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Substance Abuse in States and Metropolitan Areas: Model Based Estimates from the 1991-1993 National Household Surveys on Drug Abuse: Methodology Report

by:

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1. Introduction

In 1996, the Substance Abuse and Mental Health Services Administration (SAMHSA) published a report on substance abuse for 26 States and 25 metropolitan statistical areas (MSAs). That report did not provide much detail on the methodology used to produce those estimates. The report you are now reading has been written to serve two purposes. The first is to give policy analysts a deeper understanding of the properties of the estimates and of the confidence intervals on the estimates. The second is to allow other researchers to replicate the approach and to apply the method to other tasks.

The impetus for developing the local area statistics was based on the increasing demand for such estimates by State and Federal substance abuse planners. While there are existing data systems that provide small area indicators of drug abuse for selected MSAs, like the emergency room drug-related episode reports from the Drug Abuse Warning Network (DAWN) and the voluntary urine test reports among arrestees (DUF), these data systems do not provide estimates of the desired total population statistics on use, dependence, treatment and the need for treatment. Local area surveys designed to provide reliable direct estimates tend to be costly and generally do not provide comparable estimates between areas due to different sampling and data collection methodologies.

The National Household Survey on Drug Abuse (NHSDA) has been the federal government's chief source of information on the magnitude of substance use and abuse in the United States and has been used by policy makers to monitor trends in drug use and to identify problem areas. In addition to assessing prevalence and trends in drug use, SAMHSA uses the NHSDA data to investigate many special topics that are important for understanding the nature of substance abuse, and numerous researchers from around the nation conduct special analyses using the public use files that are produced from the survey. Reports based on the NHSDA are highly valued due to the use in the survey of a rigorous sampling and data collection methodology that has been developed and tested over a number of years.

Despite the value of the NHSDA for national statistics, it is not by itself an adequate source of information about drug abuse at the state and MSA levels. The local area statistics that were published in 1996 rely upon a complex algorithm for combining data from the NHSDA with local data from the Decennial Census and other sources. The algorithm is justified

¹SAMHSA, 1996.

through the use of models that relate the available local data to drug abuse as measured in the NHSDA. The use of models was required since national surveys like the NHSDA do not provide large enough samples for direct State level estimation, a limitation that persists even after pooling three years of NHSDA data together. The model-based approach to deriving the small area estimates is similar to approaches that have been used by other government agencies as the demand for local area data has increased and the required statistical and computational tools have improved.

The algorithm that was used to combine NHSDA and local data to produce small area estimates is new. It was developed by a contractor, the Research Triangle Institute. The earlier summary report that provided the estimates also had a simplified description of the algorithm and a report on an evaluation of the algorithm, but that discussion was not detailed enough to allow other researchers to replicate the method. This new report provides a detailed explanation of the methods. Although most of this report is quite technical, the introduction was written for a broad audience of social science researchers and policy analysts.

This report has four main chapters and several appendices. The balance of the introduction provides background on the NHSDA, a description of the variables and areas for which estimates were formed, a review of the need for special methodology different from that used to produce national estimates for the NHSDA and most other federal demographic surveys, a review of other methodologies that have been developed in the past for small areas such as states and MSAs and the need for a new methodology, a discussion of evaluation strategies, and guidance on interpretation of the state and MSA estimates.

Chapter 2 explains the methodology in detail. The material in this chapter is quite technical. Readers will need to be comfortable with calculus, linear algebra, and statistics. However, readers do not need to be national experts on the subject of small area estimation to appreciate this chapter. Specifically, it does not assume past experience on small area estimation.

Chapter 3 discusses the evaluation strategy developed for this project and presents the results. This chapter largely repeats the material in the summary report (SAMHSA, 1996), but there is some new material.

Chapter 4 discusses the lessons learned over the course of this project and ideas that the researchers will pursue in future applications of the methodology.

The appendices provide detail on the classification of drug and alcohol dependence, the definitions of the MSAs involved in the study, more technical information on the software used to create the estimates, more information on geographic predictors of drug and alcohol abuse, tables of actual estimates for the states and MSAs, and tables of confidence intervals on the estimates.

1.1 Background on the NHSDA

In this section, the basic design of the NHSDA is reviewed: the sample design, the questionnaire, interviewing procedures, and the data collection schedule.

1.1.1 Sample Design

The respondent universe for the NHSDA is the civilian noninstitutionalized population 12 and older within the United States. Since 1991, the NHSDA has included residents of noninstitutional group quarters (e.g., shelters, rooming houses, dormitories), residents of Alaska and Hawaii, and civilians living on military bases. Persons excluded from the universe include those with no fixed household address, e.g., homeless transients not in shelters, and residents of institutional group quarters such as jails and hospitals.

The sample design is a stratified, multistage probability design. The first stage of sampling involves selection of primary sampling units (PSUs) located across the United States. These PSUs are either very large MSAs or individual counties. Then, within each PSU, smaller land areas, or segments, are selected. Once the segments are selected, professional field staff list all of the dwelling units in each selected segment. From the resultant listings, specific dwelling units are selected for screening and/or interviewing. The NHSDA uses an area sampling frame based on block-level geographical units defined by the Decennial Census. Dwelling unit undercoverage bias is largely eliminated by using the half-open interval technique, which asks questions of screened dwelling units to identify unlisted units.

The NHSDA oversamples blacks and Hispanics to ensure an adequate representation of these minority populations in the sample. Young people are also oversampled. From 1991 to 1993, six major metropolitan areas—New York City, Washington DC, Miami, Chicago, Denver and Los Angeles—were oversampled to provide prevalence estimates for those areas. Due to the association between smoking and illicit drug use, oversampling of current smokers aged 18 to 34 was initiated for the 1993 NHSDA.

Exhibit 1.1 summarizes the sample design/size and response rate data for the NHSDA between 1991 and 1993, the years upon which the small area estimates were based.

Exhibit 1.1 NHSDA Sample Size and Response Rate by Year

	1991	1992	1993
# PSUs	125	118	117
# Segments	3,509	3,218	3,124
# DUs screened	105,311	69,996	100,340
Screener response rate	95.0%	93.6%	94.0%
# Interviews	32,594	28,832	26,489
Interview response rate among screened eligibles	84.3%	82.5%	79.5%

1.1.2 Interviewing Procedures

Prior to the interviewer's arrival at the sample dwelling unit (SDU) a letter was mailed to the resident briefly explaining the survey and requesting their cooperation. Upon arrival at the SDU a few days later, the interviewer referred the respondent to this letter and answered any questions. If the respondent had no knowledge of the lead letter, the interviewer provided another copy, explained that one was previously sent, and then answered any questions. If no one was at home during the initial call at the SDU, the interviewer left a "Sorry I Missed You" card alerting the SDU that the interviewer planned to make another callback at a later date/time. Interviewers made at least four callbacks (in addition to the initial call) to each SDU to complete the screening process and possibly obtain an interview. As necessary and appropriate, the interviewer could make use of the Appointment Card for scheduled return visits with the respondent. When an in-person contact was made with an adult member of the SDU and introductory procedures were completed, the interviewer presented a "Statement of Confidentiality" and answered questions if required. Assuming respondent cooperation, a screening of the SDU was then initiated through administration of the Housing Unit Screening Form for housing units, or the Group Quarters Unit Screening Form for group quarters units.

If a potential respondent refused to be screened, the interviewer was trained to accept the refusal in a positive manner, thereby avoiding the possibility of creating an adversarial relationship and precluding future opportunities for conversion. A conversion was usually attempted by supervisory field staff or specially selected veteran interviewers with established conversion records. If the respondent proceeded with the screening process, the interviewer answered any questions that the screening respondent may have had concerning the study. A number of informational handouts also may have been given to the respondent at this time.

Interviewers screened each sample dwelling unit within a sample segment. The interviewer listed all dwelling unit members on the screening form in order of age, beginning with the oldest. The dwelling unit type was determined based upon the race/ethnicity of the head of household, and the age groups represented in the household. Once the dwelling unit type was determined, the interviewer referred to a selection table that indicates which (if any) age groups were to be interviewed. None, one or two persons may have been selected for interview. The decision was made to limit the number of interviewed persons per household in order to avoid household fatigue, thereby keeping household response rates high.

Assuming the within dwelling unit sampling process selected a member for participation in the study, and if the selected individual was 18 or older and currently available to do the specified study questionnaire, the interviewer moved immediately to begin administering the questionnaire in a private setting within the dwelling unit. If the selected individual was 12 to 17 years of age, parental consent was obtained from the selected individual's parent or legal guardian; then, the minor was asked to participate. Once consent was obtained, the interviewer began the interview process.

The questionnaire and interview methods were designed to retain respondent interest, ensure confidentiality, and maximize the validity of response. The questionnaire was administered in such a way that interviewers would not know respondents' answers to the questions on substance use. Except for tobacco, questions for all substances were self-administered, as respondents recorded their answers on answer sheets. If a respondent was unable to read the questions or preferred not to read them, the interviewer read them out loud. However, only under circumstances of the respondent's physical inability could the interviewer record the answers on the answer sheets. Separate English and Spanish version of the questionnaire were available for each round of the NHSDA.

As the respondent recorded his/her answer choices and completed each answer sheet, they were placed in an envelope. At the end of the interview process, all materials (screening form,

study questionnaire, and answer sheets) were enclosed in this envelope, sealed, and mailed to the data processing site.

For verification purposes, respondents were asked to complete a Verification Form that requested their address and telephone number for possible follow-up to ensure that the interviewer did his/her job appropriately. Respondents were apprised that completion of the Verification Form was voluntary, and they were given the opportunity to decline to complete the verification form. These forms were then placed in a separate envelope and mailed to the data processing site. Respondents were invited to travel with the interviewer to the nearest mailbox to verify that the envelopes were immediately mailed.

A random sample of those who completed verification forms received a telephone call expressing appreciation for their participation in the study. Each respondent also was asked to answer a few questions verifying that the interview took place, whether the answer sheets were used properly, and the amount of time required to administer the interviews. Mail-out of verification letters was used when telephone numbers were unavailable.

1.1.3 Data Collection Schedule

Data collection for the 1991 NHSDA took place over a 6 month period from January to June of that year. In 1992, the NHSDA began to operate on a quarterly data collection schedule. That is, data collection extends over the entire calendar year, and is divided into four quarterly components based on the four calendar quarters. This quarterly data collection schedule allows for the realization of several associated benefits and efficiencies, such as increased flexibility in the analysis of trends and in the ability to produce data suitable for public release. Furthermore, such a schedule reduces the burden on the field staff by spreading the data collection over a longer period of time, thus allowing procedures from year to year to be implemented on more of an on-going rather than "start-stop" basis.

1.2 Selection of Variables and Areas for Estimation

It was not feasible to develop estimates of all substance abuse rates. It was necessary to pick a few behaviors. Also, there was a reluctance to prepare estimates for areas with very little NHSDA sample in them. In this section, the selected variables and areas are documented, and the choices are explained.

1.2.1 Selection of Behavior Measures

The NHSDA collects information on many aspects of substance usage, dependence, and treatment along with background demographic variables that can help provide context. Due to the cost of fitting the area estimation models, it was not feasible to produce estimates for all behaviors of interest. Accordingly, SAMHSA chose 11 critical behavior measures for which state and MSA estimates would be produced.

Alcohol and cigarette use were chosen because of their importance for public health and because independent estimates of the use of these substances are available at the state level for comparison. Three measures of usage of illicit substances were chosen, one broad measure that covers all illicit substances, a narrower measure that excludes marijuana, and a measure that focuses exclusively on cocaine.

Two measures of dependence were chosen: a measure of dependence on illicit drugs with or without dependence on alcohol and a measure of dependence only on alcohol without any dependence on illicit drugs. Persons classified as dependent are among the most severely affected substance abusers who are likely to require some kind of treatment or intervention. The dependence measure developed for the NHSDA is based on an algorithm that approximates the DSM-III-R criteria (Appendix A).

Three measures related to treatment were chosen. Two are based on questions in the NHSDA questionnaire that ask about the receipt of treatment. One covers treatment for illicit drugs (with or without treatment for alcohol abuse), and the other covers treatment for alcohol abuse (without treatment for illicit drugs). A measure of need for treatment for illicit drugs was also chosen that is based on an algorithm developed by SAMHSA's Office of Applied Studies. By subtracting the count of those actually reporting receipt of treatment from those classified as needing treatment, it is possible to measure unmet need. Note, however, that unmet treatment need was not explicitly estimated.

Finally, arrest during the past year for any reason was also chosen. This behavior measure was chosen because of its importance and its correlation with substance abuse. Note that data on arrests are also available from the FBI Uniform Crime Reports, but that the UCR

data are not directly comparable with NHSDA data since the UCR counts all arrests while the NHSDA counts persons with one or more arrests.

The 11 chosen behaviors are:

Use of legal (licit) substances:

- Cigarette Use In Past Month Smoked cigarettes at least once within the past month.
- **2. Alcohol Use In Past Month** Had at least one drink of an alcoholic beverage, that is, beer, wine or liquor or a mixed alcohol drink within the past month.

Use of illicit substances:

- 3. Any Illicit Drug Use In Past Month Use within the past month of hallucinogens, heroin, marijuana, cocaine, inhalants, or the nonmedical use of sedatives, tranquilizers, stimulants or analgesics.
- 4. Any Illicit Drug Use Other Than Marijuana In Past Month Past month use of any illicit drug excluding those whose only illicit drug use was marijuana.
- 5. Cocaine Use In Past Month Use within the past month of cocaine in any form, including crack.

Drug or alcohol dependence:²

- 6. Dependent On Illicit Drugs In Past Year Dependent on marijuana, inhalants, cocaine, hallucinogens, heroin, opiates or nonmedical use of sedatives, tranquilizers, analgesics, or stimulants. Those who are dependent on both alcohol and another illicit substance are included, but those who are dependent on alcohol only are not.
- Dependent On Alcohol But Not On Illicit Drugs In Past Year -Dependent on alcohol but not dependent on any illicit drugs.

Treatment for drug and alcohol problems:

8. Received Treatment For Illicit Drugs In Past Year - Received treatment in the past 12 months at any location (including hospitals, clinics, self-help groups, doctors) for any illicit drug. These estimates include those who received treatment in the past 12 months for both dinking and illicit drugs.

²Detailed definitions of these dependence measures can be found in Appendix A.

- 9. Received Treatment For Alcohol Use, But Not Illicit Drugs, In Past Year Received treatment in the past 12 months for drinking (including hospitals, clinics, self-help groups, doctors). These estimates exclude those who received treatment in the past 12 months for both drinking and illicit drugs.
- 10. Needed Treatment For Illicit Drug Use In Past Year (whether or not treatment received) Persons who either: were dependent on illicit drugs in the past year; were a past year heroin user; received treatment in the past 12 months for any illicit drugs; were a needle user of heroin, stimulants or cocaine in the past 12 months; were a daily marijuana user during the past 12 months; or in the past 12 months were weekly users of hallucinogens, cocaine, inhalants or had weekly nonmedical use of stimulants, sedatives, tranquilizers or analgesics.

Past year arrest:³

11. Arrested For Any Crime In Past Year - Arrested and booked at least once for breaking a law in the past 12 months.

1.2.2 States and MSAs for Which Estimates Were Produced

Estimates were only produced for those States and MSAs which had some minimum NHSDA sample in them. Although estimates could have been produced even for States with no NHSDA data, it was thought that these estimates would not have been reliable enough to meet program needs. The States and MSAs selected for small area estimation are presented in *Exhibit 1.2*. This exhibit shows the number of people who responded to the combined 1991-1993 NHSDA surveys plus information on the NHSDA sample including numbers of the sample MSA/County units, sample block groups, and the estimated 1992 population.

The States and MSAs presented in *Exhibit 1.2* were chosen for small area estimation because:

The NHSDA sample size was close to 400 persons or more, a number selected so that at least some instances of even the rarest behaviors of interest such as cocaine abuse would be observed at least once in the state sample,

³Adjustments of NHSDA substance abuse statistics based on ratios of administrative record arrest counts divided by NHSDA survey estimated arrest reports have been proposed by D. Wright and J. Gfroerer (1994). "The use of external data sources and ratio estimation to improve estimates to hard-core drug use from the NHSDA," in Harrison, L. and Hughes, A., eds. *The Validity of Self-Reported Drug Use: Improving the Accuracy of Survey Estimates.* NIDA Research Monograph 167, NIH Pub. No. 96-4147, Washington, DC, Supt. Of Docs., U.S. Government Printing Office. However, these adjustments were not done for this report.

- The number of distinct sample PSUs was greater than or equal to 4 units, and
- The number of distinct sample segments was greater than 40 segments.

Exhibit 1.2 State and MSA Small Areas Selected for Inclusion in the Study Population Size and NHSDA Sample Characteristics

STATE	SAMPLE MSA/ COUNTIES ¹	SAMPLE BLOCK GROUPS ²	SAMPLE RESPONDING PERSONS ³	1992 POPULATION PROJECTION ⁴
TOTAL UNITED STATES	213	8,942	84,974	205,945
NORTH EAST REGION New Jersey New York Pennsylvania	34 0 /	1,489 107 843 209	13,681 1,525 8,505 2,155	42,236 0,443 14,892 9,943
SOUTH REGION FIOTIGA GEOFGIA KENTÜCKY LOMISIANA NOTTH CATOMINA UKIANOMA SOUTH CATOMINA I ETMESSEE 1 EXAS VITGIMIA WEST VITGIMIA	85 13 0 0 14 4 4 13 9	3,271 904 118 112 153 222 05 48 92 503 502 444	32,346 10,000 1,001 1,001 1,001 1,002 494 3,30 780 3,082 3,338 3,34	71,396 11,205 5,442 5,055 5,387 5,061 2,063 2,759 4,117 15,751 5,227
NORTH CENTRAL REGION IIIIIIOIS IIIIIIIIIIIIIIIIIIIIIIIIII	49 0 0 4 3 0 12 5	1723 /yy 114 00 1/5 /0 120 204 55	15,456 8,088 9/18 515 1,187 084 1,039 2,021 4/13	48,968 9,378 4,381 2,010 7,013 3,370 4,223 8,940 4,021
WEST REGION California New Mexico Oregon Washington	45 22 5 4 3	2,459 1,320 /4 39 /3	23,491 12,304 6/6 412 690	43,346 24,342 1,199 2,397 4,094

¹MSA/Counties refers to geographic entities formed to estimate random effect terms in the logistic model and which are generally analogous to NHSDA primary sampling units (PSUs). The exceptions are the distinct MSA constituents of PSUs that crossed State boundaries or combined more than one MSA.

Exhibit 1.2 State and MSA Small Areas Selected for Inclusion in the Study Population Size and NHSDA Sample Characteristics (continued)

 MSA	SAMPLE BLOCK GROUPS ²	SAMPLE RESPONDING PERSONS ³	1992 POPULATION PROJECTION ⁴
Anaheim-Santa Ana, CA	*	*	1.996

²Block groups refers to the sample segments that were selected at the second stage of selection in the 1991-1993 NHSDA.

³87,915 people responded to the 1991-1993 NHSDA, however 2,941 people were omitted from the small area estimation research because of missing local area indicator variables that were used as potential predictors in the models.

⁴Population projections presented in 1000s.

Atlanta, GA	*	*	2,425
Baltimore, MD	*	*	1,996
Boston, MA	*	*	3,145
Chicago, IL	735	7,537	4,981
Dallas, TX	*	*	2,120
Denver, CO	719	7,585	1,346
Detroit, MI	*	*	3,593
El Paso, TX	*	*	456
Houston, TX	*	*	2,661
Los Angeles, CA	768	7,533	7,127
Miami-Hialeah, FL	725	8,142	1,600
Minneapolis-St. Paul, MN	*	*	2,035
Nassau-Suffolk, NY	*	*	2,178
New York, NY	730	7,676	7,086
Newark, NJ	*	*	1,500
Oakland, CA	*	*	1,727
Philadelphia, PA-NJ	*	*	4,037
Phoenix, AZ	*	*	1,770
San Antonio, TX	*	*	1,040
San Bernardino, CA	*	*	2,122
San Diego, CA	*	*	2,089
St. Louis, MO-IL	*	*	2,006
Tampa-St. Petersburg, FL	*	*	1,822
Washington, DC	725	7,795	3,345

^{*}Number of sample block groups ranged from 40 to 110; number of respondents ranged from 400 to 1200 in these MSAs.

See Appendix B for a list of counties included in each MSA.

 $^{^2}$ Block groups refers to the sample segments that were selected at the second stage of selection in the 1991-1993 NHSDA.

 $^{^3}$ 87,915 people responded to the 1991-1993 NHSDA, however 2,941 people were omitted from the small area estimation research because of missing local area indicator variables which were used as potential predictors in the models.

⁴Population projections presented in 1000s.

Four exceptions to the four or more PSU unit rule were allowed for States that met the sample person and area segment minimums.

1.3 Limits of Design-Based Estimates

Design-based estimators are the standard approach used in most national surveys to describe the population. Under the designed-based approach, one treats the substance use status of each person in the population at a particular time as an intrinsic characteristic of the person. A fixed measurement protocol, consisting of a sampling design, a sample size, a questionnaire, a set of procedures for interviewers, a weighting procedure, and an estimator is selected to describe the fixed national population. The goal is to find estimators for various parameters of the national population such as the total number of people using a particular substance, the prevalence of substance abuse in the population, and so on. Estimates of these parameters are calculated based on the sampling design used for the survey, and their expected values and variances are derived by averaging across all possible samples that could be generated using the particular sampling design. (Särndal, Swensson, and Wretman, 1992). The variance across all possible samples can be estimated based upon the data and used to form "randomization-based" or "design-consistent" confidence intervals. With these confidence intervals, statements are possible along the lines of, "If the survey were independently repeated under the same general conditions and a large number of times and a point estimate and a confidence interval were constructed for each replicate, then approximately 95 percent of the confidence intervals formed would include the average of the point estimates." This approach is commonly known as design-based estimation and statisticians who use it can be called randomization-based frequentists or design-based statisticians. This is the procedure used in almost all demographic surveys in the U. S. to produce national estimates.

A key feature of the sampling designs that are used in national surveys is that each member of the population has a nonzero probability of being in the survey. Thus, a national probability sample is also a probability sample of any subgroup within the nation, and one can, in principle, make estimates for any small area within the nation. However, because of the small sample sizes that are obtained for these areas (including some zero sample sizes), the design-consistent confidence intervals can be extremely wide and the normal approximation required for the confidence intervals to be valid may not hold.

The NHSDA sample design was developed with the aim of producing highly precise estimates for the nation, for the four Census regions, and for various demographic subgroups

defined in terms of age, race/ethnicity, and gender. It was not designed to produce state estimates. Thus, although the NHSDA is a valid probability sample of the population of each state, design-consistent confidence intervals for state estimates are likely to be very wide because the portion of the national sample that will happen to fall in a state is small or even zero, particularly for small states. In addition, there are other features of the design that make it less suitable for state estimates. A multi-stage clustered design was used. There was no attempt to construct state level strata to ensure a minimum representation within each state. Rather, the stratification focused on controlling the representation of the important demographic subgroups and employed extensive oversampling of young people, blacks, and Hispanics. Clusters at the first stage (the primary sampling units, PSUs) were large MSAs, individual counties in medium and small MSAs, individual nonmetropolitan counties, and small groups of nonmetropolitan counties. For the 1991-1993 samples, a total of 2,939 PSUs were formed. The largest 45 were selected with certainty. The remainder were then sorted by several variables such as percent minority. A sampling algorithm was used to select an additional 72-80 sample PSUs in a manner that reduced the variance for variables correlated with the sort variables. The set of sort variables did not include a marker for state.

Because there was no attempt to control the distribution within state boundaries, state estimates are extremely unstable. Furthermore, there was 1) extensive oversampling of important demographic domains, 2) clustering at the block level during the second stage of selection to reduce the cost of data collection, and 3) limits placed on the number of persons per sample household to increase the privacy of the interview resulting in people in large households having smaller probabilities of selection. The second-stage clustering and the differential sampling tend to make the estimates even more unstable. Thus, while these procedures are near optimal for national and regional estimates, they are not at all conducive to state estimation.

For example, there were just 330 people in 4 PSUs in the NHSDA in South Carolina from 1991 through 1993. Suppose that the design-based estimated rate of cocaine abuse during this period was 0.66 percent (the same as the estimate provided by the small area estimation project). Using a logit transformation and assuming that the logit of the estimated rate is normally distributed and that the 330 people had come from a *simple random sample*, the standard design-based 95 percent confidence interval would range from 0.26 percent to 1.65 percent. (Using the logit transform avoids the embarrassment of a negative lower bound on the confidence interval.) The design-based variance based on the NHSDA was not directly estimated for this project but is probably on the order of 2 to 6 times larger than the design-based variance from a simple random sample. Assuming that the factor is 4, the 95 percent confidence

interval would run from 0.10 percent to 4.05 percent. Such an estimate and confidence interval would be of little use to South Carolina officials.

In addition to the problems that the instability of the design-consistent estimators pose for state planning purposes, this instability also leads to problems for the comparisons and ranking of states. Let π_i^D be the design-consistent estimate of the drug abuse rate for the i-th state and π_i be the limit of that estimate as the state sample size approaches the state population. The "D" stands for design consistent. The parameter π_i is defined as the limit of the design-consistent estimator rather than the "true" rate for the state due to the limitations of the measurement protocol. It can be fairly easily demonstrated that the small state sample sizes and large design effects cause

$$\mathbb{E}\begin{bmatrix} \text{Range}_{i} \{ \pi_{i}^{D} \} \end{bmatrix} \times \underset{i}{\text{Range}} \{ \pi_{i} \}$$

and that

$$\mathbb{E}\left[\sum_{i} \left(\pi_{i}^{D} - \pi\right)^{2}\right] \ \ \text{\times} \ \ \ \frac{\sum_{i} \left(\pi_{i} - \pi\right)^{2},$}$$

meaning that the range among the states and the variance among the states is exaggerated by the design-consistent estimates. This exaggeration of differences could have important consequences on public perceptions and official policy unless corrected. Thus, one of the properties that a sound alternative approach should possess is a capability to shrink the state estimates together toward the national average.

1.4 Review of Alternative Approaches to Small Area Estimation

The best current review of alternative approaches to small area estimation can be found in Ghosh and Rao (1994). In order to place the SAMSHA small area estimation project in context with other work on small area estimation, it is recommended that the reader consult this article. A book of case studies edited by Schaible (1996) is also interesting and useful for providing a background on approaches to small area estimation. The article by Ghosh and Rao is terse, requiring a fairly high level of statistical expertise on the part of the reader. For this report, the goal is to extract the kernels of the ideas presented by Ghosh and Rao and to present these ideas at a level accessible to the majority of social policy researchers. It is hoped that this material will leave such readers with an intuitive impression of the shortcomings of the other methods and the benefits of the new method presented here. The discussion in this chapter is at a fundamental level. Chapter 2 and Appendices C and G contain heavier mathematics.

The alternative approaches do not treat the substance abuse status of a person at a particular point in time as an intrinsic characteristic of the person. Instead, these approaches assume that it is a matter of chance for each person but that there are underlying factors that make this chance larger for some people than for others. Before discussing the details of these alternative approaches, it will be useful to review their theoretical underpinnings. After reviewing these underpinnings, each of the most popular alternative approaches is reviewed including fixed effect models, composite estimators, and mixed effect models. The discussion in Section 1.4 closes with reasons for believing that the new approach is superior to those other alternatives.

1.4.1 Theoretical Underpinnings: Conceptions of Reality and Probability

In order to use national data to estimate state and local substance abuse rates, it is necessary to develop a theoretical framework for applying the observed patterns of substance use to groups of persons who were not interviewed. This can be done by considering each person's substance abuse status to be a random variable rather than a fixed characteristic of the person. The actual process leading to an individual's use of a particular substance may, in fact, not have a random component; it could be a complex deterministic function of genetics, family environment, local environment, government activities, and so on. However, since this process is too complex for us to understand, it is very useful to think of the process as being random within groups of people.

Under this approach, the statistician's goal is to identify groups of people within each of which there are no discernible differences in their patterns of substance abuse. It is then assumed that the probability of substance abuse is homogenous within each group. This probability can be thought of in two different ways.

One approach is to assume that there is a fixed but unknown probability for a group. With this approach, one thinks of the collection of persons in a homogenous group as an outcome

⁴Whether or not a person is using a certain drug at a certain time T can be thought of as a random variable until time T, at which point it becomes a fixed characteristic of the person. The choice of a distribution for that random variable depends upon prior information. Examining one extreme, if the person's drug using habits are known for all t<T as well as the habits of all close friends, associates and family, and if it is known whether the person was involved in some sort of treatment plan in the time leading up to time T, then the propensity is close to either one or zero. Examining the other extreme, if nothing is known about the person other than that they live in the U. S. and are at least 12 years old, then the propensity will be some sort of national average. The more we know about a person's prior history and present situation, the more we tend to view current drug use as a fixed state of existence rather than as a random event.

of an underlying random process--as just one possible manifestation of an underlying superpopulation. The statistician defines the fixed but unknown probability of drug abuse for the group as the long-run average that would be obtained if this process were repeated independently a large number of times. The collection of possible manifestations of the underlying process is called the superpopulation. This approach to inference is termed the *model-based frequentist* approach. Applying this approach, the statistician uses the data to estimate the fixed but unknown probability and also to provide a confidence region about his estimate. The interpretation of such a confidence interval (e.g., a 95 percent interval) is that over all possible manifestations of the superpopulation, 95 percent of all confidence intervals constructed in this manner from samples of the given size from the group will include the fixed but unknown true probability for members of this subgroup. A statement can thus be along the lines of, "Given a sufficiently large number of observed manifestations, 95 percent of confidence intervals constructed in this manner will include the true probability of substance abuse for people in this group."

A second way to think of this probability is to consider it as a random variable as well. This assumes that there is a random process behind the random process theorized by model-based frequentists. In other words, the probability of substance abuse in a certain group at a certain time isn't simply some fixed p; instead, there is a random process that determines whether the probability of substance abuse for the group at a certain time is above or below p. This is called the *Bayesian* approach to inference after an 18th century theologian/mathematician who discovered a theorem that demonstrates how to use observed data to update subjective estimates of probabilities of various events. The Bayesian starts out the inference task with a prior distribution on p. This means that he has a function F (the prior distribution) such that $P\{p \le \xi\} = F(\xi)$ for $0 \le \xi \le 1$. This prior is based upon past research, intuition, and personal judgment. He then uses the sample data and Bayes theorem to update his belief about the distribution of p and to derive a posterior distribution on p. This posterior distribution will depend to some extent on the prior distribution, but, if the sample size is very large, then this dependence will be slight. The Bayesian can use the posterior distribution to make point estimates and intervals (called prediction intervals) similar to the confidence intervals of the parametric frequentists. The point estimate is taken to be the mean or modal value of the posterior distribution (the expected and the most likely value of p, respectively) and the prediction interval is defined by the percentiles of the posterior distribution. For example, if a 95 percent prediction interval is desired, the lower point will be the 2.5th percentile of the posterior distribution and the upper point will be the 97.5th percentile of the same. Prediction intervals are interpreted along the lines of, "After considering the observed data, I now believe

that there is a 95 percent probability that the probability of people in this group abusing the substance in question is in the stated range."

1.4.2 Using the Sampling Weights

Having now briefly presented the model-based frequentist and Bayesian perspectives, it is useful to contrast the treatment of sampling weights in these approaches with the treatment of sampling weights in the randomization-based approach.⁵ This is critically important for an understanding of the small area estimation approach adopted for this study since both sampling weights and elements of the model-based frequentist's and Bayesian approaches have been used in developing the method described herein.

Neither the model-based frequentist nor the Bayesian uses sampling weights. They discard information on probabilities of selection and on joint probabilities of selection. They are, in fact, often just as happy to work with a purposely drawn sample as with a random sample. The design-based statisticians, of course, insist on a random sample with known marginal probabilities of selection and prefer having information on the joint probabilities of selection as well.

These differences in the use of sampling weights arise from the different conceptions discussed above. The design-based statistician is interested in estimating the drug abuse rate amongst a certain real population at a particular point in time. The model-based frequentist and the Bayesian on the other hand are not as interested in the real population at a point in time. They are more interested in the process that gave rise to that population. The latter two therefore construct models of the process and are interested in obtaining the most efficient estimates of the parameters of those models. Research has shown that using the sampling weights when estimating the parameters of a *correct* model only increases the variance. However, research has also shown that if the model is wrong, then using the weights could have reduced or eliminated the bias in the estimates due to model error (Holt, Smith and Winter, 1980). Since the design-based statistician believes that every model is wrong, he would rather have the less efficient but more robust estimates that are obtained by using the sampling weights. The other two groups reply that if the model is wrong, then there is no point in estimating the parameters at all; that all one can do is to use the best models available and then estimate the parameters of those models with as little variance as possible.

 $^{^{5}}$ For more information on the contrasts between the different schools of statisticians on this issue, see Smith (1994).

The practical importance of these differences in inferential approaches can be illustrated with a simple example. Suppose that there is a group that the modeler (whether frequentist or Bayesian) believes is homogenous, but that there is, in fact, a small difference between two subgroups that is undetected due to small sample sizes and that the sampling weights are different for the two subgroups. Let subgroup A have a process mean rate of substance abuse of 0.3, a sample size of 5 from a population of 10,000 (resulting in a sampling weight for all 5 sample persons of 2000). Let subgroup B have a process mean rate of substance abuse of 0.35, a sample size of 25 from a population of 12,500 (resulting in a sampling weight for all 25 persons of 500). The model-based frequentist will tend to estimate a substance abuse rate for the entire group of 0.3417 with an expected standard error of 0.0867 while the designed-based statistician will tend to estimate a substance abuse rate of 0.3278 with an expected standard error of 0.1054. Note that the expected estimate of 0.3278 is correct for the combined group but that it does not accurately describe the process for either subgroup and that it has substantially higher variance than that obtained by the model-based frequentist. With this small sample size, the estimate and standard error obtained by the Bayesian would depend fairly strongly on the prior distribution assumed.

For national estimates, it is more important to estimate the substance abuse rate for the population unbiasedly rather than to get more efficient estimates by assuming models either with or without prior distributions on the model parameters. The NHSDA sample was designed to do just that. However, as discussed above in Section 1.3, this approach is not feasible at the state or MSA level using current NHSDA data; so an alternative was needed.

In developing small area estimates of substance abuse rates, the various approaches were blended. Models have to be assumed to construct estimators for substance abuse at the state and local level, but it is convenient that the state estimates sum to national estimates that are close to the national estimates obtained using design-based techniques without the models. Furthermore, it was desired that state and MSA estimators, although model-based, would converge to design-consistent estimators where state/MSA sample sizes were sufficiently large.

1.4.3 Fixed Effect Models for Small Area Estimation

Fixed effect models include synthetic and regression estimators (domain-indirect estimators in the language of Schaible, 1996). Versions exist for design-based inference,

model-based frequentist inference and Bayesian inference, but the development is perhaps simplest within the model-based frequentist approach.

The model-based frequentist basically assumes that one can model the substance abuse rate in a small area as a function of other information that exists for the small area. For example, suppose that a number of domains have been identified such as demographic subsets of the population for which current estimates are maintained for small geographic areas. The Census Bureau has divided the U.S. land area into a set of nonoverlapping very small areas called block groups and has published considerable information about every block group. Furthermore, information is available at the county level on fairly current arrest rates for illegal drug possession and for illegal drug sales or manufacture, deaths related to alcohol abuse, and alcohol and substance treatment rates. All of this information can be useful in developing models that predict local substance abuse rates.

Let π^d_{ib} be the true substance abuse rate for persons in domain d who live in block group b of state i. Let N^d_{ib} be the number of people in domain d who live in block group b of state i. Let X^d_{ib} be a row vector of characteristics for people in domain d who live in block group b of state i. Examples of entries could be binary indicator flags for membership in domain d and lack of membership in other domains and median income for the domain within the block group, the percent below poverty, the percent with a high school degree, and so on. A common model would be something like

$$\operatorname{logit}\left(\pi_{ib}^{d}\right) = \ln\left(\frac{\pi_{ib}^{d}}{1 - \pi_{ib}^{d}}\right) = X_{ib}^{d}\beta \tag{1}$$

where β is a vector of unknown parameters. Maximum likelihood methods would be used to obtain an estimate β for β without paying any attention to the sample design. The substance abuse rate for the state would then be estimated as

⁶Recall that this is interpreted not as the actual rate at any particular point in time but as the mean of the underlying process. Also note that the word "true" must be interpreted in light of the inherent limitations of the NHSDA measurement protocol, meaning that only substance abuse that is voluntarily reported to interviewers is of interpret.

$$\pi_{i}^{F} = \frac{\sum_{b} \sum_{ib} N_{ib}^{d} \pi_{ib}^{d}}{\sum_{ib} \sum_{b} N_{ib}^{d}} = \frac{\sum_{d} \sum_{b} N_{ib}^{d} \frac{1}{1 + e^{-X_{ib}^{d}} \beta}}{\sum_{d} \sum_{b} N_{ib}^{d}}$$
(2)

where the "F" stands for fixed-effect model.

As an example of how to set up such a model, suppose that there were three domains, two states, two block groups in each state, and no prediction variables other than domain flags. Then the *X* matrix would be as shown on the left below.

Applies	to	Peopl	le in
---------	----	-------	-------

					Block
			Domain	State	Group
X =	1	0	1	1	1
	1	0	1	1	2
	1	0	1	2	1
	1	0	1	2	2
	0	1	2	1	1
	0	1	2	1	2
	0	1	2	2	1
	0	1	2	2	2
	0	0	3	1	1
	0	0	3	1	2
	0	0	3	2	1
	0	0	3	2	2

Also, β would consist of just two parameters, β_1 and β_2 . Comparing a person in domain 1 with a person in domain 3, the difference between the two domains in the logit of the propensity to engage in the behavior of interest is β_1 . This translates into odds ratio of e_1^{β} . For example, if β_1 =1, then the odds ratio is 2.7, meaning that someone in the first domain has odds of engaging in the behavior 2.7 time higher than for someone in the third domain. For more background on logistic regression, see Chapter 6 of Breslow and Day (1980), Chapter 4 of McCullagh and Nelder (1989) or all of Cox and Snell (1989).

In presenting such an estimate to an analyst, one could say that the national rate of substance abuse was estimated for combinations of a number of demographic domains in certain types of block groups ("neighborhoods"). One could also say that data from the Decennial Census were used to estimate how many inhabitants the state has in each combination of demographic group and block group. The national rates were applied to these state-specific population counts to estimate expected substance abuse rates for the state. Since there are many

variables that influence the likelihood that a person will abuse substances other than those measured in the decennial census, the estimates would generally be different from those that would be obtained if a large state-specific sample were interviewed using the national procedures. The analyst could then use this estimate for planning purposes, along the lines of, "If there are this many people in need of assistance, then we need to do ... to meet those needs." However, one should be cautious about using the estimates to correlate state action plans with estimated differences across the states. The estimates will not reflect differences across the states due to differences in laws, enforcement activities, advertising campaigns, outreach activities and so on, and will tend, rather, to be quite close to each other since the mix of demographic groups and block groups does not vary dramatically across the states. In essence, the fixed modeling approach shrinks the state estimates too close together. Also, if the estimates were repeated every few years, it would be entirely incorrect to try to correlate changes in the state estimates with any changes in policies or other state activities. The only changes would be due to changes at the national level.

An additional difficulty of using fixed effect models is that the estimated variances on these estimates would be highly misleading. The estimated variances would reflect sampling variability on the national rates only. Since the national rates are estimated with small variance, these estimates would appear to be quite accurate, even though they are not -- at least not in the broader sense of being close to what a large sample for the state would have found. Approaches have been suggested for replacing the variance with a mean squared error across all the states, but such approaches are not very satisfactory, as is discussed further in Section 1.5, Chapter 3, and Appendix H

Thus, it is apparent that there are several important shortcomings with this approach. Most seriously, analysts who used them to assess the impact of state interventions would end up with seriously flawed analyses. Nonetheless, if there were no state specific data in the NHSDA, this would be the best that could be done. For example, there are no NHSDA sample data for Hawaii for the period of 1991 through 1993. This approach is therefore the best that could be done for Hawaii. However, there are state-specific data for all the states for which substance abuse rates were estimated in this study.

The fact that state-specific data do exist for each of the states and MSAs that were used in this analysis allows consideration of a model along the lines of

$$\operatorname{logit}\left(\pi_{ib}^{d}\right) = \ln\left(\frac{\pi_{ib}^{d}}{1 - \pi_{ib}^{d}}\right) = X_{ib}^{d}\beta + \eta_{i}, \tag{3}$$

where η_i is a fixed effect for the *i*-th state. However, whereas model (1) will tend to make the states look too much like each other, model (3) will tend to overstate the differences among the states, much as the design-based estimates do. Such a model would produce estimates of π_i that would be only marginally more precise than the design-consistent estimates. Like the design-consistent estimates, the range and variance of substance abuse rates would be exaggerated. What is needed is a method that shrinks the estimates closer together than the design-consistent estimators but not as close together as the fixed-effect model given in equation (1).

1.4.4 Composite Estimators

When at least some survey information is available for a small area, it is possible to use composite estimators that combine information from the survey and from models. The randomization-based approach can be used to create a design-consistent estimator, π_i^D , for the substance abuse rate in each state. As discussed previously, this estimator will be subject to intolerably high variance for most states due to small state sample sizes and the lack of state stratification in the first stage of sample selection. However, this estimator is unbiased (or at least design consistent) for actual substance abuse rates in the state at the point in time of interest. If, as is probable, the model alone does not adequately describe the propensity to abuse substances, then π_i^D contains information about whether the state was higher or lower than projected from a model based only on demographic distributions across block groups. A common approach is to average π_i^D with π_i^F . Let Γ_i be the averaging factor for the i-th state. Then the composite estimator of π_i is

$$\boldsymbol{\pi}_{i}^{C} = \boldsymbol{\Gamma}_{i} \boldsymbol{\pi}_{i}^{F} + (1 - \boldsymbol{\Gamma}_{i}) \boldsymbol{\pi}_{i}^{D} \tag{4}$$

The variance of π_i^C is larger than the variance of π_i^F , but π_i^C is closer to what would have been obtained with a large sample in the state.

This approach has considerable merit, but the problem is how to decide on values for the Γ_i . Additionally, it becomes more complicated to define what is meant by variance since the randomization-based approach has been mixed with the model-based frequentist approach. Various approaches have been suggested for choosing the Γ_i , but none are very satisfying. It is desired to pick Γ_i close to 1 when the model applies well for the state and/or the sample size is small for a state and to pick Γ_i close to 0 when the model doesn't fit well for the state and/or the sample size is large for the state. It is easy enough to decide when the state has a large or small sample size, but deciding when the model fits well in a particular state is impossible unless the state sample size is very large, in which case, it is known that Γ_i should be selected close to zero without even examining its fit for the state. Ad hoc rules can be developed, but better approaches to the entire problem have recently been developed.

1.4.5 Mixed Effect Models for Small Area Estimation

With a mixed effect model, one endeavors to produce estimates for each small area that reflect not only the level of substance abuse expected for an area given the national process means for different domains and types of block groups but to also reflect the impact of unmeasured variables, such as family environments, on the local substance abuse rates. The goal is thus quite similar to the goal for composite estimation -- smaller variance than the design-based estimates but expected values closer to what a large sample in each state would yield. The difference between using a mixed effect model and composite estimation is that fitting a mixed effect model is a somewhat more objective process than picking the Γ_i in composite estimation.

A one-stage mixed effect model looks like:

$$\operatorname{logit}\left(\pi_{ib}^{d}\right) = \ln\left(\frac{\pi_{ib}^{d}}{1 - \pi_{ib}^{d}}\right) = X_{ib}^{d} \beta + U_{i}, \tag{5}$$

where U_i is a random variable with zero expected value and positive variance that is the same for all states and it is assumed that the variables are independent across states; i.e., $E(U_i) = 0$, $Var(U_i) = \sigma_u^2$, and $U_i U_{i'}$ are independent for every $i \neq i'$..

Note the difference between models (3) and (5). In model (3), each state has a fixed deviation on the logit scale from the baseline propensity expected based just on demographic

group and prediction variables. In model (5), each state has normal random variation on the logit scale from the baseline propensity. Over all possible manifestations of the state process, it is expected that each U_i will have an average value of zero. Nonetheless, after the period in question, it is possible to talk about the realized value of the random variation. In a particular past year, the deviation for each state must have been positive or negative. Model (5) can be used to get estimates, U_i , of the value of U_i for a particular manifestation of the state process. For example, if a state had a social climate that effectively suppressed substance abuse, one would expect to see a value of $U_i < 0$. Conversely, a climate that was conducive to substance abuse would lead to a value of $U_i > 0$.

Unless state sample sizes are large, the U_i from model (5) will be closer to zero than the η_i from model (3). If all the state sample sizes are large, then the models yield similar estimates. This regression to the mean with small state sample sizes is desirable since each of the state processes is being measured with error, thereby exaggerating the differences among the states, and so there is a need to "shrink" the state estimates somewhat toward the national average. This is demonstrated in Section 3.2.2. For a more thorough review of the benefits of making the state effect a random effect rather than a fixed effect, see Robinson (1991).

With this approach, the overall estimated rate of substance abuse for the state is

$$\pi_{i}^{M} = \frac{\sum_{d} \sum_{b} N_{ib}^{d} \pi_{ib}^{d}}{\sum_{b} \sum_{ib} N_{ib}^{d}} = \frac{\sum_{d} \sum_{b} N_{ib}^{d} \frac{1}{1 + e^{\left(X_{ib}^{d} \beta \ 55^{\wedge} + U_{i}\right)}}}{\sum_{d} \sum_{b} N_{ib}^{d}}$$
(6)

where the "M" stands for mixed effect. From the standpoint of being able to compare the states to determine what sorts of social climates are more resistant to substance abuse than others, it is desired that π_i^M is close to $\lim_{n_i \to N_i} \left(\pi_i^D \right)$, the limit of the randomization-based estimate for the state as the state sample size, n_i , approaches the total state population, N_i . Note that this is not the same as desiring that π_i^M is close to π_i^D given the actual state sample size from a national sample. Since π_i^D is very unstable with the small state sample sizes that are available from a national sample, the π_i^M should vary substantially from the π_i^D .

Prior efforts to fit this sort of model did not have the same goal of attempting to approximate the asymptotic values of the design-based estimators. Model-based frequentists have developed techniques to estimate β , U_i , and σ_u^2 using maximum likelihood techniques. These techniques require that the U_i be normally distributed and run into both memory space and time problems on even the largest computers available. Bayesians place prior distributions on β and σ_u^2 but are free to use distributions other than the normal distribution for U_i . Bayesians use new techniques called Monte Carlo Markov Chain (MCMC) methods (of which the most well known example is Gibbs sampling) to obtain posterior distributions for β , U_i , and σ_u^2 . These techniques use huge amounts of computer time but are conceptually easy to program. A third approach is called empirical Bayesian, in which approximations are used to obtain posterior distributions for β , U_i , and σ_u^2 with less computer time than a full Bayesian approach and without assuming prior distributions. For a review of these methods, see Chapter 9 of Diggle, Liang and Zeger (1994). The Bayesian approach is also discussed in Chapter 8 of Schaible (1996).

All of these techniques assume that the basic structure of the model given in (5) is "true." This means that the propensity to abuse substances can be totally explained by knowing the demographic group that a person belongs to, the type of block group that they live in (as characterized by a few variables from the prior Decennial Census), characteristics of the county, and the state that they live in. Even more strongly, all these techniques assume that the logit transform of this propensity is a linear function of fixed effects for the demographic group and type of block group and of a random effect for the state. The actual propensity to abuse substances may, of course, be considerably more complex reflecting things such as the quality of the local transportation network, the quality of the local police department, the proximity to major drug transportation routes, and a score of other fixed and random characteristics of the milieu in which each person lives. Since these techniques assume that the model is true, they make no use of the sampling weights. If the model is not true, they thus cannot be expected to converge to the substance abuse rates that would be estimated by very large state substance abuse surveys. To deal with this lack of convergence, a survey weighted approach was developed for this study.

1.4.6 The Survey-Weighted Empirical Bayesian Approach

This is the name given to the approach developed for this study although it could also be called a "survey weighted Penalized Quasi-likelihood" approach since the estimators can be

motivated from either an empirical Bayesian viewpoint or from a frequentist view point. The goal of this study was to be able to fit a model like (5) in such a manner that estimates would be close to design-based estimates for states with large NHSDA samples and for national domains with large sample sizes such as men and women. In terms of equations, it was desired that

$$\lim_{n_i \to N_i} \frac{\pi_i^W}{\pi_i^D} = 1, \text{ and}$$
 (7)

$$\boldsymbol{\pi}_{d}^{W} \boldsymbol{\pi}_{d}^{D}, \tag{8}$$

where π_i^W is the survey-weighted empirical Bayesian estimate of the substance abuse rate for the **state** and π_d^W is the survey-weighted empirical Bayesian estimate of the substance abuse rate for the **demographic group** across all states.⁸

The method was based upon the empirical Bayesian approach because it was thought to be the least expensive in terms of computer time. Even with this approach, large quantities of computer time were consumed in fitting all the models. Equation (8) is useful in terms of the face validity of the estimates. It is certainly awkward to explain why small area estimates do not aggregate up to national estimates. Equation (7) is somewhat useful in states with large sample sizes such as California, but questions do remain about how close π_i^W and π_i^D are to each other when the state sample size is not large. If the model is not true (as it almost certainly is not), the authors believe that their estimates will be better than those from model-based frequentist, Bayesian or Empirical Bayesian approaches, but it is not claimed that π_i^W will be extremely close to $\lim_{n_i \to N_i} \left(\pi_i^D \right)$ when the state sample size is small. It is not realistic to think that modeling exercises of this type can provide as much information about complex social phenomena as large local surveys. This question of evaluation is discussed further in Section 1.5. Guidance on interpreting the estimates is given in Section 1.6. It is not entirely straightforward to say what

⁷The approach follows the approach of Breslow and Clayton (1993) quite closely. The phrase penalized quasi-likelihood appears due to Green (1987).

⁸Within each of the four census regions, those states for which small area estimates were not produced were grouped together in a regional residual. These four regional residuals were treated the same as states in all the procedures.

is meant by $Var\left(\pi_i^W\right)$. More information on the mechanics of the survey-weighted empirical Bayesian approach is given in Chapter 2.

1.5 Evaluation Strategies

In any small area estimation project, questions arise as to the quality of the estimates. The purpose of using a model-based approach that combined NHSDA data with local data was to produce estimates of substance usage behaviors for States and MSAs that would have smaller variance than the direct estimates based solely upon the NHSDA. As discussed in Section 1.3, the direct design-consistent estimates provided confidence intervals for most States and MSAs that were too wide to be useful to researchers, policy analysts, and public health officials. It was recognized that in order to achieve the desired variance reduction, it would be necessary to let the validity of the resulting estimates depend on the models that were created to relate local data to NHSDA data. To the extent that the local data have good predictive power for the substance abuse behaviors of interest and that the models accurately reflect those relationships, the resulting model-based estimates for States and MSAs should be better than either the design-consistent estimates or simply assuming that the national rates apply to every State and MSA. Several meanings can be given to the word "better" in this context, but, regardless of the definition, actually demonstrating that the model-based estimates are better is a challenging task.

This issue is also being confronted by other government agencies. It is, in fact, at the heart of the debate about whether the Decennial Census should be adjusted for undercount. One of the chief methods for estimating the undercount is a national survey taken shortly after the census. Although there is some controversy about the adequacy of the undercount estimates at the national level, the most vigorous debate is whether adequate models can be developed to apply the national information about undercount to small areas in a way that makes the small area estimates better than the raw counts. This remains an open question despite years of major efforts by some of the best statisticians in the country.

This study had a fairly intensive evaluation component in it, but the results were more suggestive than conclusive. In Chapter 3, there is a thorough discussion of the various approaches that were used and the limitations of each. Results of the various approaches are also presented and discussed. More conclusive results might be achieved with a large scale simulation or perhaps the application of the methodology to variables that are measured well on the Decennial Population Census.

1.6 Guidance on Interpretation of Results

As mentioned earlier, the interpretation of confidence intervals provided for the state and MSA estimates of substance abuse is not entirely straightforward. First, it may be better to call these intervals prediction intervals since they are based more on Bayesian concepts than on frequentist concepts. The intervals are based on a model characterized by a national fixed pattern and a series of three random events that jointly determine whether a person abuses substances or engages in other behavior of interest. The national fixed pattern says that a person of a particular age, race and sex living in particular type of block group is going to have a certain baseline propensity to engage in the behavior of interest. The first random event is at the state level. It is assumed to affect everyone in the state in exactly the same manner by adding a small offset to the logit of the propensity to engage in the behavior of interest. The second random event is at the PSU level. It is assumed to affect everyone in the PSU in exactly the same manner by adding a second small offset to the logit of the propensity to engage in the behavior of interest. The third random event is at the person level. Given that the person has a propensity determined by his or her race, sex, age, type of block group of residence, specific PSU and specific state, it is assumed that his or her actual behavior is random with the probability of engaging in the behavior equal to the person's propensity. There was no attempt to explain a mechanism by which the random decision is reached. Statistically, it was treated as if a random number between 0 and 1 was drawn and if the random number was less than the propensity, then the person engaged in the behavior.

One can imagine averaging up the true propensities for some State, MSA or other group of interest and calling the average π . That hypothetical average π is a random variable subject to variance because of the random events at the state and PSU levels. The model states that π is normally distributed.

If all the estimated person-level propensities are averaged up for the same group, then another random variable results that is approximately normally distributed. This average is π , the estimated propensity for the group. The random variable π is subject to variance not just because of the random events at the state and PSU levels but also because of uncertainty in the estimated parameters of the fixed national model. The variable π is used as an estimate of π , but it is known that given π , the conditional expectation of π is not π . Since the conditional variance of π is a measure of how close it is to its own conditional expected value, interest does not focus on that conditional variance. There is even less interest in the unconditional variance of π since that measures the deviance of the π from its expected value over all possible state

outcomes, PSU outcomes, and person outcomes -- rather interest focuses on measuring the deviance of the π from π given the state and PSU random events that actually occurred.

According with this interest, the "variance" is defined as the expected value of $(\pi - \pi)^2$ with respect to the distribution of the state and PSU level random effects and the distribution of the estimated parameters of the fixed national model. This might be called a mean square error, but that would lead to confusion with the design-based concept of mean square error. It would probably be most accurate to call it a mean squared prediction error, but that becomes cumbersome, and so it is referred to in this report as simply the variance.

The validity of the prediction intervals depends largely but not entirely on the correctness of the model. The model can be wrong in various ways. For example, the variance of the state level random effects could be larger in one region than in another; the variance of the PSU level random effects could be larger for larger metropolitan areas than for small rural counties or vice versa; the random effects might follow a different distribution from the normal; there could be correlations between state and PSU-level random effects; there might be additional random events at the county, tract, block group, block or household levels; there might be additional random effects associated with schools and places of work; it might be possible that the behavior of interest has a strong genetic component; everyone may have their own unique propensity that they share with no one else; or everyone's behavior may be predestined with no chance variability at all.

Under some of the milder alternative models, the computed prediction intervals still have meaning, but their coverage properties may be lower than claimed. The use of weights partially protects against model failure since it is known that for large enough samples of large enough populations, the inference procedures come close to design-based inference. However, for predictions for small groups where only small sample sizes are available, the validity of the prediction intervals depends fairly strongly on the validity of the model.

For more information about the procedures that were used to estimate the variances, see Section 2.6 and Appendix G. These sections have some additional information on some acknowledged problems in the variance estimation. Those problems have probably led the prediction intervals to be somewhat too narrow. The degree of the inappropriate shortening is unknown but thought to be small. The estimated prediction intervals are shown in Appendix I.

2. The Survey-Weighted Empirical Bayesian Method

This method requires that a model be developed for predicting the propensity to engage in the behavior of interest. As a result of the modeling, it is possible to estimate the propensity to engage in the behavior of interest for every person in a state, whether or not they are in the sample. In Section 2.1, the structure of the models is described. In Section 2.2, the procedures for selecting predictor variables for the models are described. In Section 2.3, there is a sketch of the procedures used to fit the models once the variables had been selected — more details are given in Appendix C. In Section 2.4, there is a description of how separate models were fit for different segments of the population due to problems that prevented the fitting of a single model for the entire U. S. population. In Section 2.5, there is a description of how the final model was used to develop an estimate of the prevalence of the behavior of interest in targeted states and metropolitan areas. In Section 2.6, there is a discussion of how the variance of the estimator was approximated -- more details are given in Appendix G.

2.1 Structure of the Model

The full model used for most of the nation is specified here. In part of the nation, it was necessary to use a simpler model. The simpler model is described in Section 2.4.

Let the states be indexed by the subscript i, the sample PSUs within the states by the subscript j, and the sample persons within the sample PSUs by the subscript k. Then

$$y_{ijk} = \begin{cases} 1 & \text{if the } kt\text{h person in the jth PSU of the } i\text{th state reported engaging in the behavior of interest and} \\ 0 & \text{otherwise} \end{cases}$$

Let $\pi_{ijk} = (y_{ijk}) = y_{ijk} = 1$. This π_{ijk} is the propensity for the indicated person to engage in the behavior of interest. A value of π_{ijk} close to 0 indicates that the person is very unlikely to engage in the behavior of interest, while a value of π_{ijk} close to 1 indicates that the person is nearly certain to engage in that behavior.

Let U_i be a normal random variable associated with the ith state and U_{ij} be another normal random variable associated with the j-th PSU of the ith state. These variables cannot be directly observed. They represent the impact of unmeasured conditions in the state and PSU on the propensity to engage in the behavior of interest, as will be explained more fully below. It is

assumed that $E(U_i)=0$, $Var(U_i)=\sigma_1^2$, $E(U_{ij})=0$, $Var(U_{ij})=\sigma_2^2$, and that all the U_i and U_{ij} are mutually independent. The expectations and variances here are with respect to a superpopulation as was discussed in Section 1.4.

Let X_{ijkt} be the value of the t-th background variable for the indicated sample person. This can be the average income for the block group that the person lives in, the arrest rate for the county, an indicator variable for a particular type of block group or county, or an indicator variable for a particular type of person (i.e., an indicator for a particular demographic domain). By indicator variable, it is meant that

$$X_{ijkt} = \begin{cases} 1 & \text{if the person lives in the type of block group or county of interest or belongs to the demographic group or interest and } \\ 0 & \text{otherwise} \end{cases}$$

Let
$$\lambda_{ijk} = logit \ \pi_{ijk} = ln \left(\frac{\pi_{ijk}}{1 - \pi_{ijk}} \right)$$
. Note that λ_{ijk} is referred to as the logit propensity for

the individual to engage in the behavior of interest. Also note that the propensity, π_{ijk} , must be strictly between 0 and 1 for λ_{ijk} to be defined.

With this notation, the full model can be written as

$$\lambda_{ijk} = \sum_{t=1}^{p} X_{ijkt} \beta_t + U_i + U_{ij} ,$$

where β_t is an unknown parameter that specifies the impact of the t-th background variable on the logit propensity. Also note that p is simply the number of background variables in the model.

The model assumes that over all possible manifestations of states and PSUs, for people with characteristics X_{ijkl} , ..., X_{ijkp} , there is a central tendency for the logit

propensity to engage in the behavior of interest. That central tendency is $\sum_{t=1}^{p} X_{ijkt} \beta_t$.

However, the model assumes that this logit propensity varies across the possible manifestations of states and PSUs. This variation is due to unmeasured variables such as family background and peer group influences, efficacy of state and local programs to influence behavior,

cultural differences and random chance. It is assumed that the variation follows a normal law, that the process causing the state variation is independent across the states, and that the process causing the PSU variation is independent across the PSUs. The use of the normal law seems reasonable since so many natural phenomena appear to obey it such as human height, weight, and IQ. However, it is not possible to directly test the reasonableness of this normal assumption. It is possible that these state and PSU random disturbances follow nonnormal distributions.

Some additional aspects of the model are important to note. The model assumes that everyone with the same values for the p background variables has the same central tendency in the logit of the propensity to engage in the behavior of interest. This is somewhat difficult to believe. Variables about the county and about the block group may have a very strong effect on this central tendency, but it seems likely that variables on family background would exert even stronger effects. This disbelief in the model led to the decision to include the sampling weights in the model-fitting procedure as was mentioned in Section 1.4.

This project was probably the first small area estimation project outside of the Census Bureau to use block group level data. Using county level data is a more common approach that is far less expensive to implement, but then the assumption of homogeneity in the central tendency of the logit propensity would have been even harder to believe. Block group-level variables were included in the modeling process in order to make the homogeneity assumption more credible.

The objective of the modeling process was to obtain posterior Bayes predictions of β_1 , ..., β_p ; all the U_i ; and all the U_{ij} . (Note that the β_t are called fixed effects and that the U_i and U_{ij} are called random effects.) Variances and covariances on these predictions were also required in order to be able to form prediction intervals around the small area estimates. The parameters σ_1^2 and σ_2^2 were not directly of interest, but had to be estimated in order to obtain the desired predictions of the fixed and random effects.

Section 2.3 discusses how the long list of potential background variables was narrowed down to the final list of predictors. A sketch of the model fitting procedures is given in Section 2.4, and a detailed description of the model-fitting procedure is given in Appendix C.

This section concludes with an interpretation of the final model for cigarette smoking during the previous month among 18-25 year olds outside of the big cities. There were 117 fixed effects in this model. Their predicted values are given in Appendix E. The first effect is

an intercept term entitled "DUMMY". This means that $X_{ijkl} = 1$ for everyone in the sample. The predicted value of β_1 is = -0.560. This means that for a person for whom all the other background variables are equal to zero, the predicted central tendency of the logit of the propensity to have used cigarettes in the last month is -0.560. Taking the inverse of the logit transform, this means that the propensity is

$$\pi = \frac{e^{-.560}}{1 + e^{-.560}} = \frac{1}{1 + e^{.560}} = 0.364$$
. On average then, 36.4% of persons with zero values of the

other background variables are expected to report smoking in the past month.

The second effect is for being female. This means that

odds

$$X_{ijk2} = \begin{cases} 1 \\ 1 \end{cases}$$

The predicted value of β_2 is β 55 $^{\land}_2$ =0.083. This means that females with zero values on all the remaining background variables have a predicted logit propensity of β 55 $^{\land}_1$ + β 55 $^{\land}_2$ =-0.477. Again taking the inverse of the logit transforms, this means that the propensity is

$$\pi = \frac{1}{1 + e^{-377}} = 0.383$$
. On average then, 38.3 percent of females with zero values on the other

background variables are expected to report smoking in the past month. A property of the logit transform is that the ratio of the odds for males and females is constant under this model for males and females with identical remaining background variables. The ratio is

$$\left(\frac{\pi_F}{1-\pi_F}\right) / \left(\frac{subM}{\pi_1-\pi_M}\right) = \frac{e^{-.477}}{e^{-.560}} = 1.087,$$
 meaning that females are estimated to have an

for smoking in the past month that is 8.7% higher than the odds for similar males. Note that this result does not seem intuitive. It was expected that males would have a higher propensity to smoke at this age. When the significance probabilities in Appendix F are examined one sees that the *p*-value for the female term is a rather large 0.1047, meaning that standard procedures of hypothesis testing with a maximum tolerance for false positive rates of 0.05 or even 0.10 would accept the null hypothesis of no effect of sex on the propensity to smoke in this age range.

Continuing with the third through sixth effects, the modelling yielded a predicted value of β 55 $^{\land}_3$ = -.273 for black females and β 55 $^{\land}_4$ = -1.145 for other females, β 55 $^{\land}_5$ = -.903 for blacks, and -.008 for others. The central tendencies of the logit propensities for all combinations of race and sex are:

Logit (π)
---------	----

	Male	Female
White	560	477
Black	-1.463	-1.653
Hispanic	560	477
Other	568	-1.630

and the corresponding propensities are:

	Male	Female
White	.364	.383
Black	.188	.161
Hispanic	.364	.383
Other	.362	.164

It is important to bear in mind that these are not population estimates. These propensities apply only to persons with zero values for all the other background variables. Since those remaining variables include block-group-level variables for which no one in the sample has a value of zero, these absolute propensities are actually not very interesting. The odds ratios are more interesting. These are shown below with white males as the reference group.

Odds Ratios for Smoking in Last Month

	Male	Female	
White	1.000	1.087	
Black	0.405	0.335	
Hispanic	1.000	1.087	
Other	0.992	0.343	

The next three effects are for regions. The omitted Midwest region has a zero value for all three of these indicator variables. The next four effects are for type of setting the person lives in: large MSA, medium MSA, small MSA, or urban nonmetropolitan (small town distant from any major city). The omitted category here is rural nonmetropolitan (on a farm or out in the wilderness).

The next three effects are for the prevalence of children in the tract that the person lives in. Tracts are roughly the size of ZIP codes areas. There does not appear to be large variation in the percent of the population that is 0 to 18 years of age at such a high geographic level, but some tracts do have unusually high or low proportions of children. In this case, $X_{ijk, 14} =$ for a linear trend effect across five categorized levels of percent of population in tract of person ijk that was 0-18 years of age at the time of the last decennial census (April 1, 1990). Also, $X_{ijk,15} =$ a quadratic trend effect $X_{ijk,14}^2$ and $X_{ijk,16} = X_{ijk,14}^3$. So $\beta \ 55^{\wedge}_{14} = 0.174, \ \beta \ 55^{\wedge}_{15} = -0.020$, and $\beta \ 55^{\wedge}_{16} = .094$. Clearly, these parameters are of almost no interest to policy analysts. The most important point of this discussion is to illustrate the mechanisms of using the parameter estimates. The reason for estimating the parameters is only to be able to estimate the average propensity in every block group in the states for which small area estimates were produced.

The random effects are perhaps of somewhat greater interest. They were intended to measure the cumulative impact at the state and PSU levels of all the variables that could not be included in the model, such as official policies and intangible variables such as cultural patterns. In essence, the fixed effects tell us about relationships between known environmental factors and the behavior of interest and the random effects tell us about what is still unexplained. However, this interpretation is muddied somewhat by having county-level arrest and substance treatment data included amongst the background variables, since these variables are more naturally viewed as the results of substance abuse rather than as causes. Instead of representing shifts in the logit propensity to engage in the behavior of interest after controlling for age, race, sex, and local environmental variables, the random effects come to represent shifts after controlling for all those variables <u>and</u> county-level arrest and treatment data on substance abuse. Shifts of this type appear to be difficult to interpret. Although the random effects are not being published, they are available from SAMHSA.

2.2 Procedures for Selecting Fixed Effects

A wide range of possible predictor variables were considered. This section presents a description of how the variables were created and a discussion of how the list of actual fixed effects in the model was determined.

NHSDA Data: Although there are many useful background variables collected through the NHSDA survey instrument, most of these cannot be used in the modeling because it is impossible to get reliable estimates at the block group level of how many people possess these characteristics. Accordingly, the only predictor variables that could be used from the NHSDA interview were age, sex, race, and Hispanic origin. From geographic information from the sampling system, additional available variables included: region, size of metropolitan area (if metropolitan), and whether the block group was urban or rural. The four age groups were 12-17, 18-25, 26-34, and over 34. As explained in Section 2.4, separate models were fit for each age group, so age and all interactions of age with other predictor variables were forced into all the models. Sex, race, Hispanic origin, and geography were recoded into the 11 main-effect variables listed in *Exhibit 2.1*. Additionally, the interactions of sex with race/ethnicity, sex with region, and race/ethnicity with region were tested.

Exhibit 2.1 NHSDA Predictor Variables (fixed effects)

Female
Hispanic (of any race)
Black but not Hispanic
Neither white, black nor Hispanic
Northeast
South
West
Large MSA (metropolitan statistical area)
Medium MSA
Small MSA
Urban block group outside of any metropolitan area

Note that one category has been excluded from each of the classifications that the variables induce (i.e., male, white but not Hispanic, Midwest, and rural nonmetropolitan). This was done to get unique parameter estimates. It was possible to have set extra constraints to make the parameters unique, but it was easier to exclude categories.

Census data: *Exhibit 2.2* lists Census data elements that were considered for use in the modeling. There were eighteen groups of these variables. All of the attributes come from the 1990 U.S. Census long form sample. About 8 percent of the population was asked to complete a long form in 1990. For most of the variable groups, the possibilities were considered of using average responses both at the block group level and at the tract level.

Note that all of the census variables are approximately continuous at the tract level and block group level. There was concern that the relationship of the logit propensity to the predictor variables might be nonlinear. Frequently, for example, the upper and lower ends of the income distribution have more in common with respect to certain behaviors than the middle of the income distribution. Consideration was thus given to using powers of the continuous variables as predictor variables, such as the square of the poverty rate, the cube of the poverty rate, and so on. However, using these powers of continuous variables as predictor variables probably would have led to outliers when the model was used to predict block-group level propensities. The reason for this is that when a continuous variable is used as a predictor, any outlying values of the predictor variable can lead to outliers in the predicted propensities.

In order to avoid the problems with outliers while simultaneously being able to estimate linear trends, quadratic relationships, and higher order relationships, some special recodes were created. These recodes were done in the same manner for each of the variables. The procedure was as follows:

Sort the units (counties, tracts or block groups, depending on the variable being recoded) by the variable. Using the poverty rate for tracts as example, that means that all the tracts in the nation were sorted by the tract-level poverty rate. The sorted list was then divided into 5

⁹Further information on these long form Census data can be found in the technical documentation for Summary Tape File #3 of the 1990 Census of Population and Housing as well as in the booklet "1990 Census of Population and Housing Tabulation and Publication Program" prepared by the U.S. Bureau of Census.

¹⁰A well known example of this concerns annual doctor visits. Since the poor have access to Medicaid and the upper end of the income distribution has excellent access to health care, the graph of doctor visits against income is U-shaped.

Exhibit 2.2 1990 Census Variables Used to Model Prevalence of Substance Abuse¹¹

Race x Hispanic

Percent:

White nonhispanic Black nonhispanic

Hispanic Other

2. Education for persons 18 or older

Percent with:

0-8 years 9-12 years and no H.S. diploma

H.S. graduate

some college and no degree

associate degree

bachelors, graduate, or professional degree

Percent aged:

0-18 years

19-24 years

25-34 years

35-44 years 45-54 years

55-64 years

65 and over

4. Poverty

families below poverty level

5. Public Assistance

households with public assistance income

6. Disability

persons 16-64 with a work disability

7. Household composition

one-person households

of households with female heads (no spouse present) with children under 18

8. Employment

Percent:

of men 16 years and older in the labor force of women 16 years and older in the labor

9.<u>Housing value</u> - owner occupied units Median value of owner occupied housing units

10. Housing rent - rental units

Median rents for rental units

11. Sex by marital status (persons 16 years and

Percent of:

Females currently married and not

Females separated, divorced, or widowed

Females never married

Males currently married and not separated Males separated, divorced, or widowed

Males never married

12. Income

Median Household Income

13. Urbanicity

of persons residing in an urban place

14. Urbanized Area

of persons in an MSA urbanized area

15. Age of Housing

of HUs built before 1939 of HUs built from 1940 to 1949

16. High School Dropout Rate (Tract level only)

of high school age children who have

dropped out

17. Hispanic Subpopulations

Percent:

of Hispanics that are Cuban

of Hispanics that are Puerto Rican

18. Other Race Subpopulations

Percent of:

Population that is Asian and Pacific Islander Population that is Native American, Alaskan, or Aleut

¹¹From Summary Tape File 3, 1990 Census of Population and Housing. Summaries were used at the tract level and at the block group level. Where a segment consisted of blocks from multiple block groups or even tracts, the summary variables were a weighted average of the statistics from the block groups or tracts that intersected the

categories, each containing 20 percent of the tracts. Such groups are referred to as quintiles. The 20 percent of tracts with the lowest poverty rates are referred to as the first quintile of the tract-level poverty rate distribution. The fifth quintile consists then of the 20 percent of tracts with the highest poverty rates. The trend variables were then defined as shown in *Exhibit 2.3*.

Exhibit 2.3 Definitions of Trend Variables

Linear	Quadratic	Cubic	Quartic	
-2	2	-1	8	if the tract was in the first quintile
-1	-1	2	-5	if the tract was in the second quintile
0	-2	0	-6	if the tract was in the third quintile
1	-1	-2	-5	if the tract was in the fourth quintile
2	2	1	8	if the tract was in the fifth quintile

Creating four variables for every one of the variables in the list of 18 variable groups would produce more variables than could possibly be simultaneously fit. In addition, there were problems with collinearity between variables, ¹² particularly if both the tract-level variables and the block-group level variables corresponding to the same census question (such as average poverty rate for the tract and for the block group) were used. To reduce the collinearity problem, all variables that could be logically determined given other variables were dropped. As an example of this, consider the first group of variables concerning race and ethnicity. If the percent of the tract that is white but not Hispanic, the percent that is black but not Hispanic and the percent that is Hispanic are all known, then one can easily deduce the percent that is neither white nor black nor Hispanic.

To further reduce collinearity problems, block-group level variables were discarded if the correlation between the tract-level version and the block-group level version (across all the block groups in the nation) was above 0.80. Retention of the tract-level variables was favored over the block-group level variables because of the smaller sampling error on the tract level variables. However, this was a difficult decision. If all work had been done only at the tract level, the tasks of assembling the predictor variables and of obtaining population projections would have been much less expensive. There was substantial initial interest in the block group level data

 $^{^{12}}$ Two variables are collinear if the correlation between them is equal to one. The model fitting procedures would fail if any pair of variables was collinear or even approximately collinear.

because it was thought that they would have greater predictor power than tract level data. Of course, for those variables where the tract-level and block-group level variables are nearly collinear, it didn't matter much which were used since both versions contained roughly equal information in them.

There was also a desire to include some interactions of the Decennial Census variables with the NHSDA variables. For all Census variables that were kept after the preceding step, two-way interactions of the variables with sex and separately with race/ethnicity were generated. No other interactions were tested.

County level (social indicator) correlates of substance abuse: In addition to the Census variables, recoded county level 'social indicators' of substance abuse were also considered. These county level variables were obtained from three sources: The first of these sources was the FBI's Uniform Crime Reports database for 1991. This source yielded data on arrest rates per 10,000 persons for illegal drug possession, drug sales/manufacture by several reported drug categories, and total violent crime arrest rates. The second source combined data from the 1991 and 1992 National Drug and Alcoholism Treatment Unit Survey (NDATUS) conducted by the Substance Abuse and Mental Health Services Administration. From this source, data were obtained on the 1991 and 1992 average treatment rates per 1,000 county residents for (1) alcohol treatment alone and (2) for illicit drug treatment (including treatment for both drug and alcohol use). Thirdly, 1990 alcohol-related death rates per 10,000 county residents were obtained from the National Center for Health Statistics national death certificate registry. Two such rates were considered in this research: (a) the "any related rate" which includes all ICD-8 cause-of-death codes that are deemed to have a significant link to alcohol abuse, and (b) a more restrictive rate which requires explicit mention of alcohol on the death certificate. Appendix D presents the 1990 census block group, census tract, and County level variables that were considered as possible fixed effects. In appendix D, the defining variable labels of block group variables are preceded by the letter B. The tract and County variable labels are preceded by letters T and C respectively.

Underclass Indicator: Finally, there was a variable at the tract level from the Urban Institute that flagged tracts in distressed inner city areas. This was a binary recode of several tract-level census summary variables.

Further reducing list of possible predictor variables

Despite the work to eliminate collinear census variables, there were still too many variables and some of the variables had structural problems. A variable has such a problem if there exists a level of the variable such that no one in the entire NHSDA sample had the characteristic and also engaged in the behavior of interest. A logistic model always predicts a propensity strictly between 0 and 1 -- never exactly 0 or 1. If a predictor variable has a value that is associated with an observed propensity of 0 or 1, then the model fitting will fail. Accordingly, all such variables were dropped. Note that this was done separately for each of the 88 models, as were all the remaining steps in this section.

The next step was to fit a simple **linear** regression model using all the remaining possible fixed effects. The form of this simple linear regression model was

$$Y_{ijk} = \sum_{t=1}^{P} X_{ijkt} \beta_t + e_{ijk},$$

where it was assumed that all the e_{ijk} are independently and identically distributed as N(0, σ^2). Since Y_{ijk} can be only 0 or 1, while the right hand side can predict any real number, this model is not at all satisfactory for most inferences. Nonetheless, models of this type do have several important advantages over logistic models when the analysis is in the exploratory stage. First, the linear models do still produce parameter estimates that are approximately correct. Second, they do not fail when some interaction term is associated with an observed propensity of 0 or 1. Third, linear models are far faster to run than logistic models since there is no iteration required. Fourth, the software for implementing variable elimination procedures is better written for linear models than for logistic models 13. The idea was to quickly cull out poor predictors and predictors that were approximately collinear with other predictors. ¹⁴ To be on the safe side, the selection criteria were set loosely for these runs at α =0.1. If any interaction terms associated with a main effect were significant, then the interaction and the associated main effects were kept even if the main effects were not significant themselves. (This makes the model somewhat easier to interpret.) Similarly, if a higher order term was significant for one of the recodes for polynomial trends, then all the recodes for lower order trends were kept even if those lower order trends were not significant themselves.

 $^{^{13}}$ Elimination procedures are procedures that involve fitting a series of models, where the (n+1)-th model is obtained either adding the most important previously omitted variable from the n-th model or deleting the least important currently included variable in the n-th model.

¹⁴Collinear predictor variables also cause linear models to fail because the matrix X'X becomes ill-conditioned, making it difficult or impossible to invert it.

To improve the accuracy of the p-values, a modified version of the REGRESS procedure from SUDAAN was used to carry out the linear regression runs. SUDAAN increases the variance estimates to reflect the effects of clustering and unequal weighting in the sample design. The stripped-down version was a custom version made to run faster than the more flexible commercial version. SUDAAN requires the input of cluster and stratification structure for the sample design. The same structure information was used as would be used for national models from the NHSDA. This structure is different from the structure implicit in the mixed model, but this was not very important since the program was only being used to further cull out weak predictor variables.

The next step was to run fixed-effect logistic models with the retained variables. This was done using a stripped-down version of the procedure LOGISTIC from SUDAAN. The algorithm was the same as would have been used to fit a fixed-effect logistic model to any dataset from a complex sample survey, but some of the optional features were removed to make the program very fast. The same sample structure information that would have been used for national models from the NHSDA was again used. In these logistic fits, any variables that failed to pass the significance tests outlined above with α reduced to 0.05 were deleted. The fixed effect parameters obtained from these survey weighted LOGISTIC procedure runs were used as the starting β values in the mixed logistic model runs.

2.3 Fitting the Models

The full model described in Section 2.2 can be written in matrix form as $\lambda = X\beta + ZU$,

where λ , X, β , Z, and U are all matrixes. The structure of these matrices is a bit tedious to describe and understand, but the effort is well worthwhile in terms of the simplicity that they allow in describing the extremely complex model fitting procedure.

Let m be the number of states involved in the model fitting. The term "state" is used loosely here. A state-like entity in the model could be either a state, the balance of a state after taking out an MSA, or the balance of a region after taking out all targeted states. Let r_i be the number of sample PSUs in the i-th state. Let n_{ij} be the number of sample persons in the j-th PSU

of the *i*-th state. Let $n = \sum_{i=1}^{m} \sum_{j=1}^{r_i} n_{ij}$ be the total number of sample persons used in the model

fitting. The matrix λ is an $n \times 1$ matrix (i.e., n rows and 1 column). It is formed by listing all the λ_{ijk} with the first subscript varying the slowest, the middle subscript the next slowest, and the last subscript the fastest. (Recall that λ_{ijk} is the logit propensity to engage in the behavior of interest for the k-th person in the j-th PSU of the i-th state.) This is the same as listing all the logit propensities for the first sample PSU of the first state, then those of the second sample PSU and so on until all have been listed for the first state, and the proceeding to the second state and so on. The last entry is for the last person in the last sample PSU of the last state.

The matrix X is an $n \times p$ matrix (i.e., n rows and p columns), where p is the number of fixed effects in the model. The element in the j-th column of the i-th row is denoted X_{ij} and signifies the value of the j-th background variables for i-th sample person where the sample persons are ordered by state, PSU, and person, as in the description of λ above. The matrix β is a $p \times 1$ matrix with elements β_1 through β_p .

Let

$$r = \sum_{i=1}^{m} r_i$$

be the number of PSUs in sample. Let q = m+r. This is the number of random effects in the model. Then U is a qx1 matrix where U_1 through U_m are listed first and then U_{II} through U_{mr_m} are listed next with all the PSU effects for the same state listed next to each other. The matrix Z is an nxq matrix consisting of 0s and 1s. The jth element of the ith row of Z is denoted Z_{ij} and signifies whether the i-th sample person is in the state or PSU associated with the j-th random effect. Exactly two elements of each row of Z are 1's since a particular sample person can only belong to one state and one PSU. The other elements of each row are all 0s. Each column of Z corresponds to either a state random effect or a PSU random effect. If a column corresponds to a state random effect then the number of 1s in the column is equal to the state sample size. Similarly, if a column corresponds to a PSU effect, then the number of 1s in the column is equal to the PSU sample size. The distribution assumptions about the random effects can be written compactly as

$$U \sim N_q (0.55 \sim ,G)$$
 ,

where N_q denotes a q-variate normal distribution, 0 60~? is a qx1 matrix of zeroes and G is the variance-covariance matrix for U.

Given the assumptions about independence between the random effects, a uniform variance for the state-level effects, and another uniform variance for the PSU-level effects, the matrix G has a particularly simple form. Let I_n be an nxn identity matrix. (A matrix with 1s on the main diagonal and 0s off the main diagonal. Then

$$G = \sqrt{\sigma_1^2 I_m} \frac{0}{\sigma_2^2 I_r}$$

where, as before, m is the number of states and r is the number of sample PSUs. Note that G is $q \times q$.

The objective of the model fitting was to find the best possible estimates of β and U. Once these are obtained, they are used to predict a propensity for every person in the population to engage in the behavior of interest. In other words, obtaining best estimates of β and U would lead to best estimates of λ , which would in turn lead to best estimates of π for the entire population. If π is known for the entire population, then it is easy to estimate the average propensities for small areas. Complicating this simple objective is the fact that statisticians do not agree on the way to obtain best estimates of β and U. This is an area of very active current research. There is a maximum likelihood approach favored by model-based frequentists, another approach favored by Bayesians, and an approach favored by empirical Bayesians. These various approaches are discussed below. That discussion is prefaced by a quick review of probability distributions and a general discussion of the different approaches to estimation.

By way of quick review, the distribution function for a random variable is usually written F(y) and is defined as $F(y)=\Pr\{Y\leq y\}$. If F is differentiable at y, then one may define f(y)=F' (y) and call f the probability density function (pdf). If F is not differentiable at y, then f is defined as

$$f\left(y\right)=\lim_{\epsilon\to0^{+}}120\epsilon\to0^{+}\left(F(y+\epsilon)\right)-\lim_{\epsilon\to0^{-}}120\epsilon\to0^{-}\left(F(y+\epsilon)\right)$$
 and it is then called a probability function (pf) since

$$f(y) = PrY = y$$
.

The modeler usually has a family of possible distributions in mind when making inferences based on a sample. That means that he believes the distribution of Y is known except for the values of a few parameters, labeled by the vector θ . Writing out f(y) will involve the use of θ in some way. Thus, f(y) is often written as $f(y|\theta)$, drawing attention to the fact that the pdf or pf of y depends on θ . Instead of viewing it primarily as a function of y that depends on θ , the

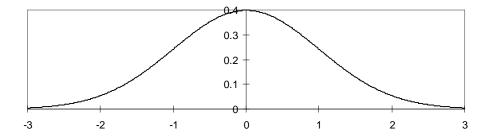
pdf or pf can be viewed as a function of θ that depends on y. When it is viewed in this manner, it is called a likelihood function instead of a pdf or a pf and it is usually written as

 $L(\theta|y)$.

Note, however, that $L(\theta|y)=f(y|\theta)$. Both functions appear identical. The difference is only one of emphasis. When $f(y|\theta)$ is used, the writer is, thinking about ways to use knowledge about θ to make predictions about y; when $L(\theta|y)$ is used, the writer is thinking about ways to use knowledge about θ to make predictions about θ .

Exhibit 2.4 shows the pdf for a standard normal distribution. The pdf reaches its maximum value at the center of the distribution at Y=0. The point that maximizes the pf or pdf is called the mode of the distribution. If one considers trying to balance the graph at a single point along the horizontal axis, the center of gravity is also found at Y=0. The center of gravity is called the mean of the distribution. If one considers trying to find the value of Y such that half the area under the graph is on the left and half the area is on the right, that point is also found at Y=0. The point with half the area on either side is called the median of the distribution. For the normal distribution, all the measures are the same; i.e., the mode, mean and median are all equal to each other. This is not true for all random variables, but since the central limit theorem

Exhibit 2.4 Standard Normal Probability Density Function



states that the mean of a large random sample of values of a random variable will be approximately normally distributed, the mode of the distribution is close to the mean in many applications.

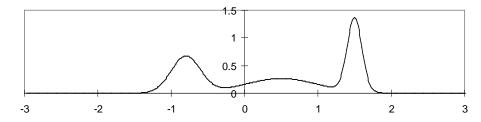
The shape of the curve in *Exhibit 2.4* is similar to the shape of the curve of the likelihood function for the unknown mean of a normal distribution given a simple random sample from that distribution, except for the fact that the likelihood function will be centered at the sample mean.

The method of maximum likelihood is to estimate a parameter by finding the mode of its likelihood function given the sample data. Suppose for example, that it is believed that Y is normally distributed with known variance but unknown mean and that a sample of 30 observations of Y yielded an average value of -34. The value that maximizes the likelihood function for the unknown mean given that particular random sample is -34.

Although this method works fine for many applications, there can be problems when the likelihood function is not symmetric or has multiple local maxima as in Exhibit 2.5. The likelihood function can look like this when the variable being measured (the *Y* variable) is not normal and the sample size is small. Note that for this random variable, the mean, median, and mode are all different. The median is 0.5, the mean is 0.4, and the mode is 1.5. For such a random variable, maximum likelihood estimation would not be a good method for finding the mean based on a small sample.

Despite the dangers of using maximum likelihood estimators for small samples, they have excellent properties when based on large samples. It has been demonstrated that for many classes of random variables, the maximum likelihood estimators of their parameters are asymptotically unbiased and exhibit minimum possible variance. In addition, the MLE

Exhibit 2.5. Probability Density Function for Mixture of Normal Variables



estimators are asymptotically normally distributed which is useful for forming confidence intervals.

The Bayesian approach to estimation would be to place a prior distribution on the parameter θ , then derive the posterior distribution of θ given the sample data, then find the mean of the posterior distribution. If $p(\theta)$ is the prior distribution on θ , then the posterior distribution of θ given Y is

$$p(\theta | y) = \frac{f(y | \theta)p(\theta)}{\int f(y | \theta)p(\theta)d\theta}$$

Note that if $p(\theta)=1$ for all θ and if $\int f(y|\theta)d\theta$ exists, then $p(\theta|y)=L(\theta|y)$, so the two methods are very similar for a flat $p(\theta)$. Such a prior distribution is called an uninformative prior because it states that all real numbers are equally likely to be the correct value of θ . Such a prior is an improper prior because a proper probability distribution has finite area underneath its graph. For some applications, use of an uninformative prior will result in the denominator of the posterior being infinite so that the posterior is not well defined. For other applications, the use of an uninformative prior is adequate. In general, use of uninformative priors is the biggest difference between the empirical Bayes approach used in this study and a fully Bayesian approach. With a fully Bayesian approach, only proper priors would be used.

If there are two random variables of interest, Y and Z, then the joint distribution of the two together is frequently of interest, as well as the marginal distributions of each by itself, and the conditional distributions of each given the other. These various distributions are related to each other as follows. Let the marginal probability functions or probability density functions for Y and Z be f(y) and g(z), let the joint pf or pdf of Y and Z be h(x,y), and let the conditional pf or pdf of Y given Y be g(z|y). Then

$$f(y|z) = \frac{h(y,z)}{g(z)}$$
,

$$g(z\big|\underline{y}) = \frac{h(y,z)}{f(y)} \quad ,$$

$$h(y,z)=f(y|z)g(z)=g(z|y)f(y)$$

$$f(y) = \int h(y,z) dz = \int f(y|z)g(z)dz$$

$$g(z|y) = \int h(y,z) dz = \int g(z|y) f(y) dy$$

where the integrals are Lebesgue integrals that correspond to summation when h is a pf instead of a pdf.

If there is a sample of n random observations of Y, then the joint pdf or pf for the entire set of observations is

$$\textbf{\textit{f}}_{n}(\textbf{\textit{y}}_{1}\textbf{\textit{,}DOTSAXIS}\textbf{\textit{,}}\textbf{\textit{y}}_{n}\,|\,\theta) = \prod_{i=1}^{n}\textbf{\textit{f}}(\textbf{\textit{y}}\,|\,\theta)$$
 where the Π notation means to multiply all n factors together.

Returning from general review to this specific study, the pf for a single observation of substance abuse or other behavior is

$$f(y_i | \beta, U, G) = [\pi_i^{y_i} (1 - \pi_i)^{1 - y_i}],$$

for $y_i = 0$ or 1. (This is a standard result for Bernoulli random variables.) Using the general rule given above, the pf for the entire y vector of behavior observations given β , U, and G is

$$f(y \mid \beta, U, G) = \prod_{i=1}^{n} [\pi_{i}^{y_{i}} (1-\pi_{i})^{1-y_{i}}].$$

The pdf for the vector of random effects given the assumption of a q-variate normal distribution is

$$f(U \mid G) = \frac{1}{\left(\sqrt{2\pi}\right)^q \sqrt{|G|}} \exp\left\{-\frac{1}{2}U^t G^{-1}U\right\} \ ,$$

where |G| denotes the determinate of G, U^{t} denotes the transpose of U, and G^{-1} indicates the inverse of G (a matrix such that $GG^{-1} = G^{-1}G = I$, the identity matrix).

Using the rules given above for conditional and joint distributions, the joint pf/pdf for y and U is

$$f(y,U|\beta,G)=f(y|\beta,U,G)f(U|\beta,G)$$

Since the distribution of U does not depend on β , this can be simplified to

$$f(y,U|\beta,G)=f(y|\beta,U,G)f(U|G)$$

$$= \left\{ \prod_{i=1}^{n} \left[\pi_{i}^{y_{i}} (1-\pi_{i})^{1-y_{i}} \right] \right\} \frac{\exp \sqrt{\frac{1}{2}} U^{t} G^{-1} U^{t}}{(\sqrt{2\pi})^{q} \sqrt{|G|}}$$

If this function is now viewed as a function of β and U given fixed y and G instead of a function of y and U given fixed β and G, something very similar to a likelihood function is obtained:

$$L(\beta, U | y, G) = \left\{ \prod_{i=1}^{n} [\pi_{i}^{y_{i}} (1 - \pi_{i})^{1 - y_{i}}] \right\} \frac{\exp \sqrt{\frac{1}{2}} U^{t} G^{-1} U^{t}}{(\sqrt{2\pi})^{q} \sqrt{|G|}}$$

Finding β and U that maximize this penalized quasi-likelihood has substantial intuitive appeal as a means for estimating β and U. However, there are two problems. The first is that G is unknown. The second is that $L(\beta, U|y, G)$ is not a true likelihood function since U is a latent variable rather than a parameter. A latent random variable is a random variable that, by definition, can not be observed. When the existence of a latent random variable is theorized, it is standard practice to also speculate on the nature of the joint distribution of the observable and the latent random variables. However, maximum likelihood estimation requires that the marginal distribution of the observable random variables given the distribution parameters be found, and that this marginal distribution be maximized to find the maximum likelihood estimates of the parameters. Using a procedure that is similar to maximum likelihood estimation but not exactly the same raises some questions about the properties of the estimators. As was mentioned above, maximum likelihood estimators are known to have good properties Little is known about the properties of estimates obtained by when sample sizes are large. maximizing $L(\beta, U|y, G)$ even when G is known.

This is the situation that was faced in this project. The random Y is observable, but the random vector U is not So U is a latent random vector. The distributions of both Y and U depend on the parameter G. In addition, the distribution of Y depends of the parameter G. The parameter G is the real parameter of interest. It governs the relationship between the census background variables and the observed behaviors. The marginal conditional distribution of Y given the parameters G and G is, however, quite computer intensive to calculate. Using the relationships on marginal and conditional distributions given above, the marginal distribution of of Y given the parameters G and G is

$$f(y \mid \beta, G) = \frac{\left\{ \prod_{i=1}^{n} [\pi_{i}^{yi} (1 - \pi_{i})^{1 - y_{i}}] \right\} \frac{\exp\left[\frac{1}{2} U^{t} G^{-1} U\right]}{\left(\sqrt{2\pi}\right)^{q} \sqrt{|G|}}}{\int_{i}^{n} [\pi_{i}^{yi} (1 - \pi_{i})^{1 - y_{i}}] \right\} \frac{\exp\left[\frac{1}{2} U^{t} G^{-1} U\right]}{\left(\sqrt{2\pi}\right)^{q} \sqrt{|G|}} dU}$$

This can be turned around into a likelihood function for β and G given Y:

$$L(\beta,g \mid y) = \frac{\left\{ \prod_{i=1}^{n} \left[\pi_{i}^{yi} (1-\pi_{i})^{1-y_{i}} \right] \right\} \frac{\exp \left[\frac{1}{2} U^{t} G^{-1} U \right]}{\left(\sqrt{2\pi} \right)^{q} \sqrt{|G|}}}{\int_{i=1}^{n} \left[\pi_{i}^{yi} (1-\pi_{i})^{1-y_{i}} \right] \int_{i}^{q} \frac{\exp \left[\frac{1}{2} U^{t} G^{-1} U \right]}{\left(\sqrt{2\pi} \right)^{q} \sqrt{|G|}} dU}$$

The model-based frequentist would evaluate the integral in the denominator and then find the values β 55^ and G that jointly maximize $L(\beta,G|y)$. He would then use G to make predictions U for U, and then use β 55^ and U to make predictions for λ and hence for π .

The difficulty with this approach is that the integration to find $L(\beta,G|y)$ is q-dimensional and each of those q dimensions must be integrated with numerical techniques since closed-form integrals of $L(\beta,U|y,G)$ do not exist. Some methods exist to reduce the dimensionality of the integration to the number of levels of random effects. For just a state effect, it is possible to reformulate the task as one involving q=m single-dimensional integrations; for a model with both state and PSU effects, the task can be reformulated into some number of two-dimensional integrations. This approach was, in fact, tried for this project. However, the computer time required for the integration was found to be unacceptable even with the 2-level nested random effect model. Since at the time, there were plans to develop 3-level and even 4-level nested random effect models (with random effects for separate segments and even households), this approach was abandoned.

The estimation approach of maximizing $L(\beta, U|y, G)$ can be easily justified from an empirical Bayes point of view. (That is the reason why the method is called an empirical Bayes method.) To see this, note that if an uninformative joint prior is placed on β and U, then the joint posterior of β and U given the data y and the variance matrix G is

$$p\left(\beta,U \,\middle|\, y,G\right) = \frac{\left\{ \prod_{i=1}^{n} \left[\pi_{i}^{y_{i}}(1-\pi_{i})^{1-y_{i}}\right] \right\} \frac{\exp^{\sqrt{\frac{1}{2}}} U^{t} \, G^{-1} \, U^{\sqrt{\frac{1}{2}}}}{\left(\sqrt{2\pi}\right)^{q} \, \sqrt{|G|}}}{\int_{\sum_{i=1}^{n} \left[\pi_{i}^{y_{i}}(1-\pi_{i})^{1-y_{i}}\right]^{2}} \frac{\exp^{\sqrt{\frac{1}{2}}} U^{t} \, G^{-1} \, U^{\sqrt{\frac{1}{2}}}}{\left(\sqrt{2\pi}\right)^{q} \, \sqrt{|G|}} d\beta dU}$$

Since the denominator of this posterior is not a function of β or U (they have been integrated out), maximizing $p(\beta, U|y, G)$ is identical to maximizing $L(\beta, U|y, G)$. Thus, the method of maximizing the joint quasi-likelihood assuming known G is identical to the

empirical Bayesian method, also assuming known *G*. The fact that the two approaches agree was comforting, as was the fact that it had been previously invented, tested, generalized, and alternately motivated by Stiratelli, et al (1984), Schall (1991), and Breslow and Clayton (1993). This was the approach adopted for this study.

The problem in applying this method is that G must also be estimated. If U were known then it would be possible to estimate G by maximizing L(G|U)=f(U|G). This can be motivated through the maximum likelihood perspective or through the empirical Bayes perspective by placing an uninformative prior on G. Thus, there is a slight conundrum. Given the true value of G, it is possible to reasonably estimate G and G and G are the true value of the unobservable G, it is possible to reasonably estimate G. If one knows neither G are the it is difficult to know how to estimate either of them or G. Following, the work of the authors mentioned above, this conundrum was resolved through the use of an iterative technique.

In this iterative technique, there is an initial guess at a reasonable value for G. Using this guess, β and U are estimated by maximizing $L(\beta, U|y, G)$. The resulting estimate of U is used to estimate G through maximization of L(G|U). This procedure cycles back and forth until convergence is obtained.

A fully Bayesian approach would have been to put a joint prior distribution on β , U, and G. The Bayesian would then find the posterior means β 55 $^{\wedge}$, U, and G and use these to estimate λ and hence π . Until recently, the Bayesian approach involved even more intense integrations than did the maximum likelihood approach of the model-based frequentists. Now, with new Monte Carlo Markov Chain methods such as Gibbs Sampling, this approach may be computationally tractable. However, since these methods are new, reported to converge very slowly, and require the specification of prior distributions (a step with which SAMHSA was not very comfortable), this approach was not pursued.

As discussed in Section 1.4, the empirical Bayesian approach was modified slightly for this study to use the survey weights. The intent of this modification was to make the method more robust to model misspecification and to make the method design-consistent.

The modification to the method involves the insertion of the survey weights into the posterior distribution for y and U. Rather than maximizing

$$L\left(\beta, U | y, G\right) = \left\{ \begin{array}{l} \prod\limits_{i=1}^{n} \left[\pi_{i}^{y_{i}} (1 - \pi_{i})^{1 - y_{i}} \right] \end{array} \right\} \quad \frac{\exp\left\{ \left[-\frac{1}{2} U^{t} G^{-1} U \right] \right\}}{\sqrt{2\pi}}^{q} \quad \sqrt{|G|} \right) \ \ \, ,$$

the expression to be maximized is

$$L_{w}(\beta, U|y, G) = \left\{ \prod_{i=1}^{n} \left[\pi_{i}^{w_{i}y_{i}} (1 - \pi_{i})^{w_{i}(1 - y_{i})} \right] \right\} \frac{\exp\left\{ -\frac{1}{2}U^{t} G^{-1}U \right\}}{\sqrt{2\pi}}^{q} \sqrt{|G|} ,$$

where w_i is the sampling weight for the *i*-th person, scaled so that the sum of the weights is equal to the total sample size. The procedure for estimating G is the same although the use of the weights does change the estimate of G.

If all persons in the 1991-93 NHSDA had the same sampling weight, then the estimators resulting from maximizing L and $L_{\rm w}$ would be the same. Also, if the background variables reflected in X totally explained the variation in sampling weights, the estimators would be the same. However, the sampling weights vary according to a more complex pattern than is reflected in X. As discussed in Section 1.1, the NHSDA oversampled by age and race and undersampled persons in large households. Also the sampling rates in six large metro areas were markedly higher than the national sample rates. Finally, block groups with high percentages of Hispanics and Blacks were oversampled to reduce the cost of overrepresenting Hispanics and Blacks. There is further unplanned variation caused by undersampling of blocks that were much larger at the time of listing than had been anticipated based on Decennial Census information and by adjustments for nonresponse and undercoverage. Since most of the weight variation due to age and race is reflected in the X matrix, and since separate models were fit for the collection of 6 oversampled metro areas, the most important differences between L and $L_{\rm w}$ concern the planned oversampling of concentrated Hispanic and Black neighborhoods and the undersampling of large households and unexpectedly large blocks.

The frequentist properties of the survey-weighted empirical Bayes estimates are not well understood. Since the only prior distributions used were noninformative (also called improper or vague), it is theoretically possible to derive the frequentist properties of these estimates, but this is quite difficult and has not yet been rigorously resolved. Breslow and Clayton (1993) point out that for estimating β and predicting U, this method is approximately equivalent to a maximum likelihood approach. They also point out structural similarities to estimators proposed by Harville (1977) for normal y variables. They also report on two small simulation

studies, the results of which are moderately encouraging. As discussed in Section 1.5, larger, more varied, and more realistic simulation studies would be useful for evaluating the survey-weighted empirical Bayes model fitting procedure. This idea is discussed further in section 3.4.

2.4 Need for Multiple Models

The ideal modeling procedure would have been to fit one national model for each of the 11 behaviors of interest. For technical reasons, however, it was necessary to fit 8 models for each of the 11 behaviors for a total of 88 models. Separate models were fit for four age groups in each of two oversampling strata. The four age groups were 12-17, 18-25, 26-34, and over 34. The oversampling strata were defined by the set of metropolitan areas that were given sufficiently large sample sizes to allow design-based estimates with good precision. There were 6 such metropolitan areas. The combination of these 6 metropolitan areas constituted one of the strata for which separate models were fit. The other stratum consisted of the rest of the nation. The models for the collection of 6 targeted MSAs are referred to as the "Big City Models." The models for the balance of the country are referred to as the "Remainder Models."

There was an initial attempt to fit just one national model for each behavior for several reasons. First, a single national model would have yielded more stable estimates of the fixed effects. Second, the estimates of the fixed effects would not vary by age or stratum as they do when 8 different models are fit. Third, the estimates of mean square error of the small area estimates would have been more accurate. Lastly, it would have been somewhat easier to organize the file handling aspects of the task.

The reason for having to fit a different set of models for the oversampled MSAs was that the average sampling weight in the oversampled MSAs was much smaller than the average sampling weight in the balance of the country. The large disparity in weights caused a critical matrix in the model-fitting procedure to become ill-conditioned, meaning that it could not be inverted. (See Appendix C for a more detailed explanation of this point.) Larger and faster computers probably would not have solved this problem. It appears that although the method can make the small area estimates more robust through reflection of the sampling weights, if the variation in the sampling weights is too extreme, then the analysis must fail.

The reason for having to fit a different set of models for each age group was related to computer size and speed. Since age is the most important predictor of the behaviors of interest, it was desired to include many interactions of age with other variables. The X matrix of

predictor variables became too large to read in at acceptable speeds. It was thus decided that the best way to shorten the X matrix without giving up very much predictive power was to fit separate models by age group. With a larger and faster machine, there would not have been a need to fit separate models by age group. Another solution might have been to use fewer fixed effects. At the time, there was a reluctance to give up prediction power by using fewer fixed effects, but since the evaluation in Chapter 3 seems to indicate that the fixed part of the model was overfit (too rich, meaning too many fixed effects), this might be the best solution to the problem in future applications.

The big city models had a simpler structure than the remainder models. The structure given in Section 2.1 is for the remainder models. For the big city models, the terms for the PSU-level random effects (U_{ij}) were forced equal to zero. Although it might seem natural to fit a model with 6 state-level random effects corresponding to the 6 MSAs, it was felt that the estimates of σ_1^2 would be very unstable if they were based upon just 6 realizations of U_i . Accordingly, each of the six oversampled MSAs were divided in two. Although it would have been possible to split each MSA by county, it was felt that it would be more useful to split each MSA into two strata of block groups, one with higher SES (socio-economic status) than the other. SES was measured by rent levels and housing values. The cut points on rent and housing values were set so that about 30 percent of the population in each MSA was in the low SES stratum. Thus, for the big city models, m=q=12.

After the sample in the oversampled MSAs was dropped out from the remainder models, those models converged nicely. However, there was still too much variation in the weights within the 6 oversampled MSAs. The reduced models did not converge initially. The problem was that Denver and Miami are much smaller MSAs than New York, Los Angeles, Chicago, and Washington. In order to obtain the desired sample sizes for Denver and Miami, these two MSAs had to be oversampled at rates considerably higher than in the other four MSAs. To resolve this problem, the weights were standardized within each of the 12 random-effect groups defined by MSA and by SES stratum. This means, that within each of these groups, the sampling weights were forced to add up to the nominal sample size. This was done by simply dividing every weight by the sum of all weights for the group. This adjustment preserved the variation of weights within each group but clearly changed the variation across groups considerably.

This adjustment to the weights means that the big city estimates are no longer nearly as design consistent as the remainder estimates are, but comparisons of the model-based estimates

and the design-consistent estimates (unpublished) still show good agreement between the two, leading to the belief that the effect of these weight adjustments was fairly minor.

2.5 Form of the Estimator Given a Fitted Model

The population was divided into 32 domains by age, sex and a combination of race and Hispanic origin. The four age groups were 12-17, 18-25, 26-34, and over 34. The four race/ethnicity groups were white but not Hispanic, black but not Hispanic, other but not Hispanic, and Hispanic. (The "other but not Hispanic" group consists primarily of American Indians, Eskimos, and Aleuts; and Asians and Pacific Islanders.) For each behavior of interest, there were eight models covering different age groups and parts of the country as described in Section 2.4. After concatenating the predictions from the eight models, an estimated propensity to engage in a behavior of interest was available for each of 32 domains inside every one of the block groups in the 50 states and the District of Columbia. (There were about 230,000 block groups defined for the 1990 Decennial Census.)

In this report, the age domains are indexed by a, the sex-race/ethnicity subdomains with each age domain by d, the states and targeted metropolitan areas by i, and the block groups within each state and targeted metropolitan area by b. The list of targeted states and metropolitan areas is shown in *Exhibit 1.2*. In order to apply the fitted model, the X and Z matrices originally defined in Section 2.3 had to be redefined. When fitting the model, both of these matrices have one row per sample person. When applying the model, they both have one row for every domain defined by age, sex and race/ethnicity in every block group in every state of interest, whether or not the block group is in a sample PSU. Where the meaning is clear, X can mean either the matrix with one row per sample person or the matrix with one row per domain per block group. Where both matrices appear in the same equation or there is a desire to stress the choice of matrix, x 45*? is used for the matrix with one row per domain per block group. The same convention is followed for the matrix Z. The rows of X 45*? are denoted by X_{ib}^{ad} . Similarly, the rows of Z 45*? are denoted by Z_{ib} , dropping the a and the d subscripts since all the rows of Z 45*? in the same block group are identical. The row vector X_{ib}^{ad} consists of flags for the age group and domain, census data about the county, tract, and block group, and administrative data about the county as described in Section 2.2. For the remainder models, the row vector Z_{ib} consists mostly of zeros with two 1's in each row to indicate the state and PSU to which the block group belongs. For the big city models, the row $vector Z_{ib}$ has just one 1 to indicate the city and SES stratum of the block group.

Let β 50 $^{\land}_a$ be the column vector of estimated fixed effects for the a-th age group and let U 50 $^{\land}_a$ be the column vector of estimated state-level and PSU-level random effects for the a-th age group. Here, it is important to note that the PSU effect was taken to be zero if the PSU was not in sample. Also, to avoid conflicting predictions for the big cities, the state level random effects for states containing oversampled MSAs were actually defined as balance-of-state random effects. Finally, in order to be able to make national estimates, all the states in a region not targeted for separate estimates were combined to form a pseudo state. Each pseudo state was assigned a separate state-level random effect. Give this notation, the propensity for persons of the indicated age, sex, race and ethnicity in the indicated block group to engage in the behavior of interest was calculated as:

$$\pi \ 55^{ad}_{ib} = \frac{1}{1 + e^{-(X^{ad}_{ib}\beta \ 50^{\circ}_{a} + Z_{ib}U \ 50^{\circ}_{a})}}.$$

As an example, if π 55 $^{ad}_{ib}$ =0.1, this means that the model predicts that 10 percent of these people are likely to engage in the behavior of interest.

Using data from Claritas (a private firm specializing in current demographic data), the number of people in the specified area who have the indicated demographic characteristics was estimated. Let N_{ib}^{ad} be that estimated population size. Then the estimated number of people in that block group who engaged in the behavior of interest at the midpoint of the 1991-1993 period is $N_{ib}^{ad}\pi$ 55 N_{ib}^{ad} .

By summing this number up across all the block groups in the i-th area, a state/local estimate of the number of people in the age-sex-race-ethnicity domain ad who engaged in the behavior of interest at that time is obtained. By summing further on i, regional and national estimates are obtained for the domain ad. Summing alternatively on b, d, and a, one can obtain small area estimates for all ages, sexes, races, and ethnic groups combined.

Using formulas, the average rate of engaging in the behavior of interest in the *i*th small area is estimated to be

$$\pi \ 55^{N}_{i} = \frac{\sum_{a} \sum_{d} \sum_{b} N_{ib}^{ad} \pi \ 55^{A}_{ib}}{\sum_{a} \sum_{d} \sum_{b} N_{ib}^{ad}} \quad .$$

Similarly, the average rate of engaging in the behavior of interest in the *d*-th subdomain of the *a*-th age group is

$$\pi \ 55^{NW}_{ad} = \frac{\sum_{i} \sum_{b} N_{ib}^{ad} \pi \ 55^{N}_{ib}^{ad}}{\sum_{i} \sum_{b} N_{ib}^{ad}} \quad .$$

2.6 Estimating Variance

The estimated prevalence rate for the *i*-th state or other small area is

Error!

where a indexes the four age domains for which separate models were run, d indexes the eight race/ethnicity by gender domains which were modeled by fixed effects in β 55 $^{\wedge}_a$, b indexes block groups within the area, N^{ad}_{ib} is the domain-ad population of the indicated block group, X^{ad}_{ib} is the row vector of characteristics of the fixed predictor variables for the block group, Z_{ib} is the row vector of flag variables for the block group that indicated which random effects apply, and β 55 $^{\wedge}_a$ and U_a are the vectors of fixed and random effects peculiar to the model for age group a.

This is clearly a nonlinear estimator for which one can only hope to approximate the variance. First the types of errors that need to be reflected must be conceptualized. As discussed in Section 1.4, one could be interested in variability and bias across replications of the survey, across possible manifestations of the population, etc. In the empirical Bayes approach, there are three random events that determine each person's propensity to engage in a behavior of interest. The first is a random event at the state level that determines the U_i component of the person's propensity. The second is a random event at the PSU level that determines the U_{ij} component of the person's propensity. The third is the flip of the biased coin that decides whether or not the person will actually engage in the behavior, where the probability of "heads" on the coin is equal to the person's realized propensity given their PSU, state, and background variables.

If π_{ib}^{ad} is interpreted as the uniform propensity for persons in domain ad, block group b, and state i to engage in the behavior, interest focuses on how closely the estimate π_i^W is expected to track the average of these propensities

$$\boldsymbol{\pi}_{i} = \frac{\sum \sum \sum \sum N_{ib} n_{ib}^{ad} \boldsymbol{\pi}_{ib}^{ad}}{\sum \sum \sum \sum N_{ib} N_{ib}^{ad}}$$

across all possible outcomes of the three variables across all members of the population; that is, it was decided to estimate $E(\pi_i^W - \pi_i)^2$, where both π_i^W and π_i are random variables and the expectation is with respect to the posterior distributions of π_i^W and π_i given the observed data.

It was decided not to try to reflect any extra variation in $(\pi_i^W - \pi_i)$ across all possible samples of a fixed population caused by the segment and household-level clustering in the sample design. That is, of course, of some concern for a hybrid method that incorporates design-based weights in the Bayesian approach, but there would be severe technical difficulties in trying to reflect that extra variation.

Also for reasons of technical difficulty, it was decided not to reflect the extra variation in $(\pi_i^W - \pi_i)$ caused by the estimation of σ_1^2 and σ_2^2 , the components of variance. Instead, the mean was calculated as if $\sigma_1^2 = \sigma_1^2$ and $\sigma_2^2 = \sigma_2^2$ were fixed and known. This is not the desired procedure from a Bayesian viewpoint but seems to be unavoidable in the empirical Bayes approach. The magnitude of the error in estimating $E(\pi_i^W - \pi_i)^2$ caused by treating σ_1^2 and σ_2^2 as fixed is unknown, but some discussion in Appendix G indicates that the underestimation of $\sqrt{(\pi_i^W - \pi_i)^2}$ could be in the range of 5 to 20 percent.

Another technical difficulty prevented reflection of the fact that $Cov(\pi_{ib}^{ad}, \pi_{ib}^{a'd}) \neq 0$ for $a \neq a'$. Computing constraints required that this covariance be treated as zero. Since these covariances will generally be positive, this problem also led to the underestimation of $E(\pi_i^W - \pi_i)^2$. Again, the magnitude of the underestimation is unknown.

Perhaps most seriously, the overfitting discussed in Chapter 3 probably led to underestimation of the variance. The link between overfitting a model and underestimation of model error can perhaps best be understood with a simple extreme example. Consider the linear regression of a single continuous variable on *n* independent binary variables with a total sample size of *n* cases. Suppose further that all the predictor variables are independent of the continuous variable. In this case, the model fit will be perfect even though the predictor variables have no predictive value. Estimated model error will be zero. Actual prediction error will equal the variance of the continuous variable. Nothing that extreme was done on this project, but there were hundreds of predictor variables in some of the models. The sizes of the random effects were almost certainly underestimated due to the large number of predictor variables. This then probably led to underestimation of the error in the model.

Finally, there was an error in the variance formula that affected models with two stages. Corrected variances have been calculated for some of the estimates but are not shown in this report. The impact of the correction on standard errors was small (0-10%) for past month cigarette use, past month cocaine use, past month use of any illicit substances other than marijuana, and past year alcohol treatment. The impact of the correction on standard errors was moderate (0-30%) for past month alcohol use and past month use of any illicit substance. Appendix I shows the original confidence intervals prior to correction of the error. For a further discussion of these issues and a detailed description of the variance estimation procedure, see Appendix G.

3. Evaluation

This chapter discusses the work that was done to evaluate the quality of the small area estimates. This work can best be thought of as a partial evaluation since it is difficult to evaluate the quality of estimates based upon a mixture of models and data when the phenomenon being described is never actually measured. The purpose of small area estimation methodology is to have narrow confidence intervals without paying for extensive data collection. But it is precisely that kind of data that is needed to evaluate the methodology. If the data required for a stringent evaluation were available, then there would be no need for the small area estimation program. Despite this central dilemma, three types of partial evaluation were carried out.

The first type was to compare the model-based estimates with alternate estimates for the small areas of interest. This is a popular method with a fairly long history, but it was only marginally useful for this application. The second type of partial evaluation was to compare the model-based estimates with alternate estimates for artificial domains where theoretical considerations lead one to believe that agreement between the methods should be good. This method yielded more information, particularly for the more common behaviors of interest such as alcohol use, but is still subject to serious caveats. The third type of partial evaluation was to compare the model-based estimates with data from other federal data systems for the small areas of interest. Comparisons of this type are, of course, interesting, but methodological differences between the data systems make these comparisons difficult to interpret.

None of the three types of partial evaluation were very conclusive by themselves. The summary report for this study (SAMHSA, 1996) found that the joint preponderance of evidence favored the survey-weighted empirical Bayes approach over the alternatives and that for most of the estimates developed, there was evidence that the estimates adequately reflect the prevalence of substance abuse characteristics for States and MSAs. Upon further review of the evaluation methodology, it appears that some of the evaluation findings in the summary report should perhaps have been somewhat less definitive. The basic conclusions have not changed, but they are slightly more tentative in this report.

Another result of the methodology review has been some discussion of what could have been learned from a well crafted simulation study. In favor of the simulation study approach, it might be argued that the central reasons for using the survey-weighted empirical Bayes procedure were that it was thought to provide: (1) estimates with smaller errors than the design-based methods, (2) better error estimates than methods based only on fixed effect models, (3) better consistency with national estimates than could be achieved through the use of mixed models

without survey weights, and (4) computational savings relative to a fully Bayesian or maximum likelihood approach. The evaluation work presented in this report does not address any of these points. At this time, it is strongly suspected but unproven that the first two assertions are true. The third assertion appears to be true on theoretical grounds alone, but the magnitude of the improvement associated with using the weights is unmeasured. The fourth assertion appears to be subject to change as the methodology for each approach improves with additional research.

The central questions from the users' point of view are whether the prediction intervals really are nearly certain to include the truth and whether the intervals are short enough to be useful. Only the user can determine whether the intervals are short enough to be useful to them, but the quality of the coverage of the prediction intervals is an open question that needs additional research. A computer simulation study would address this question directly. This point is discussed further in Section 3.4.

3.1 Comparisons with Alternate Estimates for the Small Areas of Interest

Estimates based on different models were compared to see whether theories about the properties of the various models are borne out. Specifically, there was interest to see (1) if the survey-weighted empirical Bayes estimates for California would agree better with the design-based estimates for that state and (2) how the dispersion of the state estimates varied by method. California was of interest because the very large sample size of 12,000 people in California makes the design-based estimate the most believable estimate for that state.

Regarding dispersion, it was expected that estimates based upon the leanest fixed-effect model would have the most compressed dispersion. Estimates based upon a mixed model were expected to have broader dispersion but not as broad as the design-based estimates. If there had been a fixed effect model with broader dispersion of state estimates from the national average than what was observed with a mixed model or even with the design-based estimates, then this would have constituted strong evidence that the fixed effect model had been overfit.

The estimates from the different models were not compared with the idea of deciding which set was best in some sense. Other workers in the field of small area estimation have used comparisons of this type to try to decide which set is best, but there are some serious problems with that approach that are discussed in detail in Appendix H. Very briefly, the standard definition of the best small area estimator is usually stated in terms of design-based mean square

error. By definition, the design-based mean square error of an arbitrary estimator π_i is ${\rm MSE_D}\left(\pi_i\right) = {\rm Var_D}(\pi_i) + {\rm Bias_D^2}(\pi_i)$. The design-based estimates have large variance and almost no bias. Estimates based upon a lean fixed-effects model have very small variance and large bias. Composite estimates have average variance and bias. Estimates based upon mixed-effect models also have average variance and bias and can be applied to areas with no sample in them. Furthermore, procedures exist for estimating errors based upon mixed models that appear to have better validity than those based upon fixed effect models. Design-based variances can be estimated for each of the methods, but estimation of the design-based bias is not possible for composite estimates nor for estimates based upon mixed effect models. With no way to estimate the bias of these estimators, it is impossible to decide which of them is best. It was for this reason that the various estimates were compared across the small domains of interest primarily to see whether the expected dispersion patterns were realized.

Results

Exhibit 3.1 contrasts different estimators of past month alcohol use. Exhibits 3.2, 3.3 and 3.4 do the same for past month cigarette use, past year drug treatment, and past year arrest. The first column of estimates are the design-based estimates for the 26 states with nontrivial NHSDA samples in them. Recall from Section 1.3 that these estimates are unstable and vary too much across the states. The second column of estimates shows the results of using mixed-effect models with survey weighting. As desired, the mixed models compressed the dispersion of the state estimates. The range, the standard deviation and the interquartile range are all smaller for all four behaviors examined in the four exhibits.

If the random effects are removed but all the same fixed effects are kept as in the full mixed model, then the state estimates are equal to the statistics shown in the third column of estimates. For most behaviors and measures of dispersion, using a fixed effect model instead of a mixed effect model results in greater compression of the dispersion across the states, as expected.

The next three columns of estimates show the results of applying three different fixed effect models. The fixed effect model with only county and demographic effects does not contain any tract level and block group level summary variables, but does include some additional

Exhibit 3.1 Relationship of Survey-Weighted Empirical Bayes Estimates of Past Month Alcohol Consumption to Design-Based Estimates and

Estimates Based on Fixed Effect Models

			Estimators Based on Fixed Models			
		Survey-				
	Design-	Weighted			Separate	
	Based	Empirical	Same Fixed	County and	Demographic	
	NHSDA	Bayes	Effects as in	Demographic	Effects for Big	Demographic
	1991-1993	(Mixed	Mixed Model	Effects Only	City &	Effects Only
		Model)		,	Remainder	,
Total United States	53.01	53.46	53.43	53.01	53.41	53.40
California	57.69	56.67	58.07	56.05	53.08	52.55
Florida	49.67	48.45	48.52	49.75	52.43	52.78
Georgia	48.78	48.57	48.21	47.48	52.28	52.78
Illinois	55.73	54.43	55.26	55.48	55.18	53.27
Indiana	44.95	47.70	52.40	52.36	54.07	54.65
Kansas	60.82	56.51	54.33	56.40	54.10	54.60
Kentucky	32.03	41.18	44.26	40.97	54.18	54.79
Louisiana	56.62	49.40	44.13	44.66	51.53	52.02
Michigan	58.26	56.08	55.23	54.44	53.33	53.86
Minnesota	64.96	63.32	57.06	56.56	54.74	55.29
Missouri	44.10	54.04	52.22	54.71	53.49	54.07
New Jersey	61.10	59.94	62.03	60.32	52.62	52.91
New Mexico	56.21	53.98	51.57	57.40	52.41	51.88
New York	56.96	57.04	56.60	56.51	53.60	52.52
North Carolina	43.04	46.73	48.17	46.44	52.49	53.01
Ohio	50.45	52.24	53.18	55.30	53.56	54.14
Oklahoma	36.50	39.81	44.02	44.22	52.84	53.16
Oregon	59.72	55.95	54.83	59.81	53.97	54.45
Pennsylvania	52.70	55.82	53.75	55.90	53.39	53.96
South Carolina	47.03	46.84	46.67	41.34	51.83	52.36
Tennessee	35.76	40.70	45.10	39.07	53.06	53.65
Texas	55.23	52.88	48.80	50.07	53.10	53.06
Virginia	48.16	51.21	52.28	47.34	54.18	53.42
Washington	59.55	58.33	57.32	59.97	54.05	54.43
West Virginia	38.41	38.61	39.48	38.58	53.99	54.63
Wisconsin	67.92	59.15	55.95	56.31	54.38	54.94
Range	35.9	24.7	22.6	21.7	3.7	3.4
Standard Deviation	9.4	6.6	5.4	6.8	0.9	1.0
Interquartile Range	12.6	8.5	7.1	9.7	1.4	1.6
Correlation with						
Design-Based		0.917	0.769	0.769	0.172	0.020
Rank Correlation		0.917	0.709	0.709	0.172	0.020
with Design-Based		0.916	0.804	0.835	0.212	0.063
				3.300		2.302

Exhibit 3.2 Relationship of Survey-Weighted Empirical Bayes Estimates of Past Month Cigarette Use to Design-Based Estimates and Estimates Based on Fixed Effect Models

			Estimators Based on Fixed Models			
		Survey-				
	Design-	Weighted			Separate	
	Based	Empirical	Same Fixed	County and	Demographic	
	NHSDA	Bayes	Effects as in	Demographic	Effects for Big	Demographic
	1991-1993	(Mixed	Mixed Model	Effects Only	City &	Effects Only
		Model)			Remainder	
Total United States	27.66	27.16	27.68	27.73	27.49	27.43
California	25.52	24.35	24.97	25.81	26.35	26.70
Florida	26.34	25.75	27.29	26.62	27.15	27.03
Georgia	28.41	28.55	28.81	29.18	28.59	28.14
Illinois	27.40	27.87	27.01	27.15	26.15	27.54
Indiana	24.65	26.04	28.20	28.13	28.14	27.77
Kansas	24.55	25.78	27.69	27.17	27.95	27.54
Kentucky	34.97	33.74	31.17	29.55	28.15	27.78
Louisiana	24.21	27.90	30.75	29.35	28.55	28.09
Michigan	29.45	28.78	29.89	29.41	28.26	27.84
Minnesota	25.42	24.16	26.30	28.99	28.11	27.70
Missouri	27.53	26.76	28.29	28.25	28.08	27.68
New Jersey	25.43	26.08	26.24	26.37	27.75	27.25
New Mexico	28.56	30.67	27.29	28.02	26.61	26.02
New York	24.18	25.13	26.57	26.06	26.10	27.24
North Carolina	30.49	28.26	28.69	28.59	28.43	27.97
Ohio	31.80	31.18	29.26	29.10	28.16	27.77
Oklahoma	25.85	29.00	31.09	29.36	27.87	27.20
Oregon	25.20	27.11	28.59	27.42	27.65	27.20
Pennsylvania	30.06	28.56	28.47	28.24	27.92	27.53
South Carolina	30.38	31.00	30.61	29.76	28.61	28.17
Tennessee	31.39	31.44	30.33	30.44	28.22	27.82
Texas	27.51	28.38	27.16	26.74	27.54	27.11
Virginia	28.14	26.49	27.16	27.81	27.47	27.98
Washington	25.00	25.16	27.33	28.27	27.92	27.37
West Virginia	34.39	32.67	31.58	29.50	27.75	27.40
Wisconsin	27.98	24.94	27.13	28.21	28.03	27.63
Range	10.8	9.6	6.6	4.6	2.5	2.2
Standard Deviation	3.0	2.6	1.8	1.2	0.7	0.5
Interquartile Range	4.5	3.1	2.6	2.1	0.6	0.6
Completion with						
Correlation with Design-Based		0.844	0.584	0.583	0.287	0.244
Rank Correlation		0.844	0.384	0.383	0.287	0.244
with Design-Based		0.762	0.508	0.575	0.340	0.329
Design Dased		0.702	0.500	0.575	3.540	0.527

Exhibit 3.3 Relationship of Survey-Weighted Empirical Bayes Estimates of Past Year Drug Treatment to Design-Based Estimates and Estimates Based on Fixed Effect Models

				Estimators Base	ed on Fixed Models	
		Survey-				
	Design-	Weighted			Separate	
	Based	Empirical	Same Fixed	County and	Demographic	
	NHSDA	Bayes	Effects as in	Demographic	Effects for Big	Demographic
	1991-1993	(Mixed	Mixed Model	Effects Only	City &	Effects Only
		Model)			Remainder	
Total United States	0.62	0.70	0.70	0.68	0.65	0.64
California	1.04	0.97	0.92	0.77	0.65	0.65
Florida	0.47	0.69	0.73	0.69	0.60	0.60
Georgia	0.75	0.71	0.71	0.70	0.70	0.70
Illinois	0.46	0.54	0.59	0.53	0.63	0.65
Indiana	0.53	0.52	0.53	0.63	0.64	0.64
Kansas	0.48	0.66	0.66	0.63	0.63	0.63
Kentucky	0.41	0.57	0.57	0.66	0.63	0.63
Louisiana	0.35	0.64	0.66	0.61	0.70	0.70
Michigan	0.85	0.76	0.77	1.01	0.66	0.65
Minnesota	0.59	0.84	0.82	0.54	0.64	0.63
Missouri	0.78	0.70	0.64	0.79	0.63	0.63
New Jersey	0.42	0.68	0.68	0.67	0.64	0.63
New Mexico	0.25	0.77	0.80	0.76	0.64	0.61
New York	0.68	0.64	0.60	0.67	0.62	0.64
North Carolina	0.58	0.64	0.66	0.85	0.68	0.68
Ohio	0.34	0.70	0.70	0.66	0.64	0.64
Oklahoma	0.60	0.86	0.88	0.71	0.66	0.63
Oregon	0.69	0.90	0.88	0.80	0.60	0.59
Pennsylvania	0.45	0.56	0.60	0.62	0.62	0.62
South Carolina	0.39	0.53	0.54	0.79	0.69	0.70
Tennessee	0.27	0.60	0.60	0.62	0.65	0.65
Texas	0.65	0.61	0.66	0.68	0.66	0.65
Virginia	0.47	0.65	0.66	0.63	0.67	0.69
Washington	0.61	0.85	0.89	0.73	0.65	0.63
West Virginia	0.49	0.49	0.47	0.42	0.59	0.59
Wisconsin	0.19	0.61	0.63	0.50	0.63	0.62
Range	0.9	0.5	0.5	0.6	0.1	0.1
Standard Deviation	0.2	0.1	0.1	0.1	0.0	0.0
Interquartile Range	0.2	0.1	0.2	0.1	0.0	0.0
Correlation with						
Design-Based		0.536	0.446	0.509	0.065	0.070
Rank Correlation		0.330	0.440	0.509	0.003	0.070
with Design-Based		0.472	0.391	0.509	0.037	0.131
= 1811 24.004		0.172	0.571	0.507	0.057	0.151

Exhibit 3.4 Relationship of Survey-Weighted Empirical Bayes Estimates of Past Year Arrest to Design-Based Estimates and Estimates Based on Fixed Effect Models

				Estimators Base	ed on Fixed Models	
		Survey-				
	Design-	Weighted			Separate	
	Based	Empirical	Same Fixed	County and	Demographic	
	NHSDA	Bayes	Effects as in	Demographic	Effects for Big	Demographic
	1991-1993	(Mixed	Mixed Model	Effects Only	City &	Effects Only
		Model)		-	Remainder	_
Total United States	1.57	1.64	1.62	1.66	1.61	1.60
California	1.90	1.90	1.68	1.74	1.58	1.65
Florida	1.66	1.54	1.48	1.72	1.50	1.51
Georgia	2.50	2.57	2.04	1.99	1.91	1.80
Illinois	1.35	1.58	1.46	1.56	1.38	1.62
Indiana	2.68	2.29	1.88	1.73	1.64	1.57
Kansas	2.46	2.37	1.84	1.73	1.62	1.55
Kentucky	1.31	1.59	1.69	1.77	1.63	1.57
Louisiana	1.47	2.26	2.14	2.06	1.89	1.78
Michigan	1.41	1.89	1.97	1.76	1.69	1.62
Minnesota	1.75	1.50	1.70	2.01	1.58	1.52
Missouri	1.70	1.83	1.91	1.69	1.61	1.54
New Jersey	1.18	1.18	1.23	1.26	1.63	1.55
New Mexico	1.79	2.43	2.02	1.87	1.65	1.54
New York	0.66	1.18	1.16	1.05	1.40	1.59
North Carolina	1.46	1.70	1.95	2.21	1.82	1.73
Ohio	2.03	2.07	1.96	1.81	1.63	1.56
Oklahoma	2.07	1.27	1.52	1.79	1.54	1.47
Oregon	1.67	1.70	1.66	1.55	1.46	1.41
Pennsylvania	1.07	1.21	1.26	1.19	1.56	1.50
South Carolina	1.15	1.71	1.90	1.66	1.91	1.80
Tennessee	1.10	2.07	1.60	1.54	1.69	1.61
Texas	1.92	1.77	1.68	2.03	1.85	1.73
Virginia	1.31	1.40	1.60	1.77	1.67	1.73
Washington	1.21	1.51	1.77	1.60	1.56	1.50
West Virginia	1.75	1.17	1.35	1.74	1.47	1.42
Wisconsin	1.91	1.29	1.74	1.95	1.60	1.54
Range	2.0	1.4	1.0	1.2	0.5	0.4
Standard Deviation	0.5	0.4	0.3	0.3	0.1	0.1
Interquartile Range	0.6	0.6	0.4	0.2	0.1	0.1
Correlation with		0.550	0.450	0.515	0.150	0.012
Design-Based		0.560	0.468	0.545	0.178	-0.043
Rank Correlation		0.445	0.413	0.587	0.080	-0.079
with Design-Based		0.445	0.413	0.387	0.080	-0.079

county-level variables, particularly those associated with NDATUS and UCR (as described in Section 2.2). This model is interesting since it is less work to construct models with only county-level variables. Looking across the four tables, it appears that this model sometimes produced greater compression than the full fixed model and sometimes less. On balance, it does not appear that dropping tract and block group variables causes much, if any, stronger compression of the dispersion across the states than using the full fixed model (although of course the compression is stronger than using the full mixed model). This leads to the idea that perhaps for future applications, it would be satisfactory to use a mixed model where the only geographic fixed-effects were at the county level.

The fifth column shows the results from using a fixed effect model that includes only demographic effects except for one set of effects for the six oversampled metropolitan areas and a second set of effects for the remaining areas. This model resulted in very strong compression of the range and a rank correlation with the design-based estimates approaching zero. The simplest model with only demographic fixed effects was even worse.

Another estimator that would have been interesting to include in the tables is the design-consistent logistic regression estimator. The survey-weighted empirical Bayes estimates are asymptotically approximate composites of the estimates from the full fixed effect model with design-consistent logistic regression estimators. Thus, it is expected that the survey-weighted empirical Bayes estimates would lie between the other two. It is noted that while the survey-weighted empirical Bayes estimates frequently lie between the estimates from the full fixed effect model and the design consistent expansion estimators (simple Horwitz-Thompson estimates with some post-stratification), this is not always the case.

It is obvious from these tables that the survey-weighted empirical Bayes estimator performs at least some of the required compression of the dispersion across the states, but does it more lightly than is possible to do with a fixed-effect model. The correlations and rank correlations across the states indicate that using the modeling did change the ordering compared to using the design-based approach. The strength of the impact varies considerably across the behaviors examined. While it is hoped that the rank ordering was improved through the use of modeling, these tables do not contain adequate information to make such a judgment.

One could intently study the complete set of individual state statistics and try to make sense of them individually, but there is no statistical basis to support such analysis. For example, estimates of past-month alcohol use in Tennessee vary from 35.76% to 53.65%. Which is best? Even if individual estimates were determined to be bad in some sense, there

would still be the possibility that the bad value was a fluke and that on average, the method does provide better estimates than the alternatives.

The only state where it makes sense to look at the individual values is California. The California sample size was over 12,000 people, so the design-based estimates are quite reliable. *Exhibit 3.5* highlights the comparisons of the California estimates. Note that an RSE is shown only for the design-based estimates. While all the columns are subject to sampling error, and the columns are correlated with each other (making testing of differences for statistical significance difficult), the RSE at least helps place the relative deviations of the model-based estimates in perspective. The full mixed model does best for arrest and drug treatment. It also is almost the best for alcohol use. It is just as bad for cigarette use as the simple fixed model with only demographic effects. The reason for this is not clear. On average, the full mixed model provides the closest agreement with the design-based estimates for California, as expected given the large sample size in California.

Comparisons of the national estimates are highlighted in *Exhibit 3.6.* Although the agreement between the survey-weighted empirical Bayes estimates and the design-based estimates at the national level are generally satisfactory, the agreement is not as close as was hoped would be the case, particularly for drug treatment in the past year. Even though the survey-weighted mixed model gives better consistency with design-based estimates at the national level than could be obtained with an unweighted mixed model, the consistency is not as good as can be obtained with a weighted fixed model. (The fixed models were weighted, so they are design consistent.) The problem is that even though the number of sample people is large, the number of sample PSUs is fairly small.

3.2 Comparison with Alternate Estimates for Artificial Domains

The second type of partial evaluation that was done was to compare the estimates from the survey-weighted empirical Bayesian method with alternative estimates on artificial domains where good agreement was expected. The idea behind this approach was that all good methods should produce similar estimates for large homogenous domains, where a homogenous domain is defined to be a group of people all of whom have the same propensity to engage in the behavior of interest. However, the study did not attempt to identify naturally occurring large homogenous domains upon which this theory could be tested. Instead, an idea due to Lemeshow and Hosmer (1982) was adapted to this application to create artificial domains that were large and expected to be fairly homogenous.

Exhibit 3.5 California Estimates

			Relative Deviation from Design Based							
				Estimators Based on Fixed Effect Mod						
	Design- Based NHSDA 1991-1993	RSE on Design-B ased	Survey- Weighted Empirical Bayes (Mixed Model)	Same Fixed Effects as in Mixed Model	-	Separate Demographic Effects for Big City & Remainder	Demographic Effects Only			
Alcohol use	57.69	1%	-2%	1%	-3%	-8%	-9%			
Cigarette use	25.52	2%	-5%	-2%	1%	3%	5%			
Any illicit	8.43	3%	-2%	-7%	-14%	-31%	-32%			
Any illicit but marijuana	3.96	4%	-1%	-13%	-8%	-36%	-36%			
Drug treatment	1.04	9%	-7%	-12%	-26%	-38%	-38%			
Arrest	1.90	7%	0%	-12%	-8%	-17%	-13%			
	ĺ									
Average relative			-3%	-7%	-10%	-21%	-21%			
deviation										

Exhibit 3.6 U.S. Estimates

			Relative Deviation from Design Based							
				Es	timators Based	on Fixed Effect M	odels			
	NHSDA Design-B Empirical Bayes En 1991-1993 ased (Mixed Model) M		Same Fixed Effects as in Mixed Model	County and Demographic Effects Only	Separate Demographic Effects for Big City & Remainder	Demographic Effects Only				
Alcohol use	53.01	1%	1%	1%	0%	1%	1%			
Cigarette use	27.66	1%	-2%	0%	0%	-1%	-1%			
Any illicit	5.86	4%	-1%	2%	1%	0%	0%			
Any illicit but marijuana	2.49	6%	-2%	-1%	1%	-1%	-1%			
Drug treatment	0.62	11%	13%	13%	10%	5%	3%			
Arrest	1.57	7%	4%	3%	6%	3%	2%			
Average relative			2	3	3	1	1			

deviation		1		

This methodology exploits the fact that as a result of fitting logistic regressions, each sample person in the three years of the NHSDA that were used to fit the models has a predicted propensity to exhibit the behavior of interest (e.g., smoke cigarettes, need treatment for substance abuse, etc.). By sorting the sample persons by the predicted propensity as in *Exhibit 3.7*, and then cutting the list into a fairly small number of contiguous pieces of equal size, it should be possible to develop a partition that meets the desired criteria of large groups that are internally homogenous. Note that these groups don't represent any specific area or demographic domain; rather, each consists of people who, according to a model, have similar propensities to exhibit the characteristic of interest.

This methodology was used to partition the sample population into L mutually exclusive and exhaustive groups of equal size, each with a distinct average estimated propensity to exhibit the behavior of interest. According to the model used to create the partition, it is impossible to partition the sample into equal sized groups in a manner that would result in greater homogeneity within each group. The groups were ordered by increasing estimated propensity and indexed by d.

It was not expected that the various methods would agree exactly on each group due to variance on the estimates. A test was developed to determine whether the observed differences between the design-based estimates and each of the alternative estimators were significant. Two measures of agreement were also developed -- a correlation coefficient and a ratio of ranges. The construction of the test and measures of agreement is discussed in Section 3.2.1. The statistical properties of these statistics on the full sample are discussed in Section 3.2.2. Their properties on cross-validation samples are discussed in Section 3.2.3. The results of applying this method are presented in Section 3.2.4.

3.2.1 Construction of the Test and Measures of Agreement

Let π_d^F , π_d^M , π_d^W , and π_d^D be fixed-effect model estimates, mixed effect model estimates, survey-weighted empirical Bayes estimates, and design consistent estimates for the d-th large homogenous group. These were calculated as

$$\pi_d^F = \frac{\sum_{i \in d} w_i \pi_i^F}{\sum_{i \in d} w_i} \quad \pi_d^D = \frac{\sum_{i \in d} w_i y_i}{\sum_{i \in d} w_i},$$

where π_i^F is the propensity to engage in the behavior of interest predicted by a fixed model for the *i*-th sample person and y_i is a binary flag indicating whether or not the person reported engaging in the behavior of interest. The average group propensities π_d^M and π_d^W were calculated analogously \tan^F_d . Let π 50~} F be the vector of the L estimated average propensities from the fixed model. The propensity vectors π 50~} M , π 50~} M , and π 50~} D were defined similarly.

The test for agreement between the estimates from the fixed effect model with the design-based estimates was calculated as

$$T = (\pi 50 \sim)^{F} - \pi 50 \sim ^{D})^{t} \Psi^{-1} (\pi 50 \sim)^{F} - \pi 50 \sim ^{D}),$$

where

 $\Psi = \text{Var}(\pi \ 50 \sim)^F - \pi \ 50 \sim)^D = \text{E}(\pi \ 50 \sim)^F - \pi \ 50 \sim)^D t (\pi \ 50 \sim)^F - \pi \ 50 \sim)^D$.

The tests for comparing the survey-weighted empirical Bayes estimates with the design-based estimates were constructed similarly. Tests of this form are called Wald tests. The details of how the variance-covariance matrix was calculated are given in Appendix G, Section G.4. Under assumptions that appear fairly reasonable, T has a chi-square distribution with about L degrees of freedom under the null hypothesis that the two sets of estimates are in good agreement with each other. The power of this test is only fair, with about 40 degrees of freedom for alcohol use and cigarette use and about 20 degrees of freedom for the other behaviors. Also, there were many replications of the test for different behaviors and age groups. Accordingly, attention did not focus on just whether any p-values below 0.05 were found. While a p-value below 0.05 is certainly strong evidence that the model does not fit the data, a preponderance of p-values below 0.5 would also indicate problems with the model. This can be seen by noting that if the two sets of estimates were in good agreement, p-values for the test should be randomly spread across the interval from zero to one with an expected value of 0.50. Small p-values would indicate that the two estimators do not have the same expected values even for large homogenous domains, a finding that would be troubling. However, there are some difficulties in the interpretation of T as are discussed further in the next section.

The correlation of the estimates from the fixed model with the design-based estimates was calculated as

$$\rho^{F} = \frac{\sum_{d=1}^{L} (\pi_{d}^{F} - \pi_{d}^{D})^{2}}{\sqrt{\sum_{d=1}^{L} (\pi_{d}^{F} - \pi_{d}^{F})^{2}} \sqrt{\sum_{d=1}^{L} (\pi_{d}^{D} - \pi_{d}^{F} + 5\frac{D}{A})^{2}}}.$$

Similar definitions were used for ρ^M and ρ^W . The initial idea was that large correlations would indicate that the model fit the data well and that smaller correlations would indicate a less satisfactory fit. This correlation coefficient is a descriptive measure of model fit in the spirit of the multiple R^2 of linear regression. Since the design-based estimates are subject to more sampling error than the model-based alternatives, one would expect the correlations to be less than one even when the model fits. For this reason, there is uncertainty regarding how small the correlation should be to suggest serious lack of fit.

Let R^F equal the observed range of π_1^F to π_L^F . The ranges R^M , R^W , and R^D were calculated comparably. The range ratio for the fixed model relative to the design-based estimates was defined to be R^F/R^D . The range ratios for the mixed model and for the survey weighted empirical Bayes method were defined similarly. The initial idea was that range ratios close to one would indicate that the model fit the data well and that smaller or larger range ratios would indicate a less satisfactory fit. However, further review of this methodology has raised some questions about the interpretation of small range ratios when the variance of the characteristic under consideration across the domains is not large relative to the variance within the domains. (Since the domains were formed to be fairly internally homogenous, one would expect strong heterogeneity across the domains, but the degree of heterogeneity across the domains depends on the strength of the predictors in the model.) When the heterogeneity across groups is not large relative to the within-domain sampling variance, the design-based estimates will be overdispersed, resulting in a value of R^D that will tend to be larger than the true range. Since estimates based on a good model are designed to correct the overdispersion of the design-based estimates, it is not clear how much less than one the range ratio can be before serious lack of fit is suggested. These issues are discussed in the next section.

3.2.2 Statistical Properties of the Test and Measures of Agreement

In this section, there are some slight contradictions with the interpretation of evaluation results presented in the summary report (SAMHSA, 1996). Further review of the test and measures of agreement has established interpretation difficulties that can probably be ignored for common characteristics such as smoking and consumption of alcohol but that may be more serious for rare characteristics such as cocaine use. The Wald test is discussed first. The measures of agreement are then discussed under the simplifying assumption that the partition is fixed. Then there is further discussion of the consequence of the random nature of the partition when there is model overfitting.

Wald Test

The Wald test that was used in this study derives from a standard result in statistics. If y is a random vector of dimension L with a multivariate normal distribution with mean 0 and variance-covariance matrix Ψ , then

$$T=y^t\Psi^{-1}y\sim\chi^2(L)$$
.

A statistic with this form is called a Wald statistic. Wald statistics are a generalization of t-statistics to allow the simultaneous testing of whether a vector of (possibly correlated) random variables all have zero means. If the observed value of T is too large relative to what is expected for a statistic with a chi-square distribution, then it is concluded that one of the variables in the vector must have a nonzero mean. Even if y is not normally distributed and if Ψ is not known but has to be estimated, there are conditions under which T is still approximately distributed as a chi-squared variable with L degrees of freedom (Stroud, 1971). The quality of the approximation will depend on the severity of the nonnormality of the components of y, on the stability of the estimated variance-covariance matrix, and on the correlation between y and Ψ 65^? The approximation is better when y is close to normal and Ψ 65^? is stable and independent of y.

For this application, y is replaced by either π 50~ $\}^F$ - π 50~ $\}^D$ or π 50~ $\}^W$ - π 50~ $\}^D$, depending upon whether the test is being applied to estimated propensity vectors from a fixed-effect model or to survey-weighted empirical Bayes estimates. Thus, instead of testing whether Ey=0, the test is for whether E(π 50~ $\}^F$ - π 50~ $\}^D$)=0 or E(π 50~ $\}^W$ - π 50~ $\}^D$)=0.

The assumption that π 50~ $\}^F$ - π 50~ $\}^D$ and π 50~ $\}^W$ - π 50~ $\}^D$ each has a multivariate normal distribution is reasonable since the groups were formed large enough for the central limit theorem to indicate that the weighted mean of the predicted or observed propensities in the group has an approximately normal distribution. The more serious problem was how to define the

meaning of the expected value in the definition of Ψ . This is also an issue in defining what is meant by saying that the two estimators have the same expected values. There are four sets of random events with respect to which the expected value could be defined. The first is the random selection of persons for the NHSDA sample. The second consists of the set of state-level random effects. The third consists of the set of PSU-level random effects. The fourth consists of the person level random outcomes. In addition, it is possible to define the expectation as a conditional expectation, conditioning on various statistics and/or outcomes. Ideally, one should condition the expectation on as little as possible. It would have been most satisfying to have defined the expectation as over all possible samples, over the distribution of the state and PSU random effects, and over the distribution of person level random outcomes. However, the contractor was not able to derive a method of doing this within the time and budget available. Accordingly, something simpler was done. The variances were conditioned on the state and PSU random effects, on the estimated \(\beta \) vector, on the estimated set of random effects, and on the partition induced by the fitted model. This means that the expectation was just over all possible samples and over the conditional distribution of the person level random events given that state and PSU random effects are fixed and equal to the estimated state and PSU random effects and that the true β vector is equal to the estimated β vector.

A less conditional definition of the expectation in the equation for Ψ might have resulted in either larger or smaller variances. Focusing on the variance of π 50~ $\}^W$ - π 50~ $\}^D$, less conditioning would have resulted in larger variances on π 50~ $\}^W$ and on π 50~ $\}^D$, which would, in turn, have led to less significant test results since T would have tended to be smaller. On the other hand, less conditioning would also have resulted in the estimation of a positive correlation between π 50~ $\}^W$ and π 50~ $\}^D$. A positive correlation would have reduced the variance on the difference between the two, thereby leading to more significant results since T would have tended to be larger. Since the results in section 3.1.4 do show strong correlations, the p-values for the Wald tests are probably not as small as they should be. This means that there is probably stronger evidence of lack of fit than what the tests showed.

Although it would probably have been possible with more work to derive the variance not conditioning on the random effects or on the estimated parameters, there was another difficulty. Since the partition is random, the distribution of π 50~ $\}^W$ - π 50~ $\}^D$ depends on the order statistics of the entire set of person-level predicted propensities. The effects of the random partition can be quite strong, particularly for rare characteristics if there was any overfitting of the model. It is doubtful that it would be possible to derive the variance of π 50~ $\}^W$ - π 50~ $\}^D$ without conditioning on the partition induced by the model.

Measures of Agreement

To understand the statistical properties of the measures of agreement across the L artificial domains, it is useful to have a model for the domain propensities. Indeed, without a model, it is impossible to make any inferences from those measures. However, the method for constructing the domains makes it difficult to conceive of an adequate model. As a first step, a model was developed assuming that the partition was fixed prior to fitting the model or even collecting the NHSDA data.

Let the d-th domain be labeled π_d . The model is that there is an underlying random process that leads to the propensity for each domain. This process is assumed to have mean π and variance φ , where $0 < \varphi < \pi(1-\pi)$. This means that π_d is a random variable constrained to lie between 0 and 1, according to the model. Referring back to Section 1.4, this is a Bayesian interpretation of the propensity π_d . The model is rationalized by accepting that since it is not known why the propensity is different in each domain, it is best to treat the propensity in each domain as a random variable with some overall mean and variance. The parameter π can be thought of as the national average propensity to engage in the behavior of interest. The parameter φ quantifies the expected dispersion of the domain propensities from the national average. Since φ quantifies the variability in the process among the domains, it is referred to as the process variance. In a survey, there is interest in both estimating the national propensity and how much that propensity varies across domains. For example, there may be interest in identifying the domains with the highest and lowest realized propensities or in comparing the propensities for two specific domains.

The importance of the process variance depends on its size relative to the overall propensity. Clearly, a large process variance for either a rare or a nearly universal characteristic is more important than the same level of process variance for a characteristic that is neither rare nor nearly universal. This relationship is quantified through the intraclass correlation, defined as

$$\delta = \frac{\varphi}{\pi(1-\pi)}$$
.

If the intraclass correlation is large, then the underlying process leads to sharp differences in group propensities. If the intraclass correlation is small, then the underlying process leads to only mild differences in group propensities. Since the idea of the partition is that propensity is uniform within each group, if the intraclass correlation is small, that says that everyone in the nation has about the same propensity to engage in the behavior of interest.

Generally, the desirability of close agreement between model-based and design-based estimates depends strongly on the intraclass correlation and on the sample size per group. Close agreement between the two sets of estimates is desirable only when the intraclass correlation is high and the sample size per group is large. This can be demonstrated by noting that the dispersion of the design-based estimates is too large unless both conditions are met. After demonstrating this assertion, implications are drawn for the interpretation of both the range ratio and the correlation coefficient.

The expected dispersion in the design-based domain estimates can be derived fairly easily. The dispersion of π_1^D through π_L^D depends on two sources of random variability. There is the random variation that caused π_1 through π_L to be different from π and there is sampling variance on each of the estimates given the true propensity for the domain. The first type of random variation depends only on φ while the second type depends on π , φ , and n_d , the direct sample size for the domain. Let E_M denote expectation with respect to the model and Var_M denote variance with respect to the model. Similarly, let E_D denote expectation with respect to the design and Var_D denote variance with respect to the design. Then the unconditional expected squared deviation of π_d^D from π over all possible samples of a given manifestation and over all possible manifestations of the superpopulation is

$$\begin{split} \mathbf{E}_{\mathbf{M}} \mathbf{E}_{\mathbf{D}} & \frac{1}{|L|} \sum_{d=1}^{L} \left(\boldsymbol{\pi}_{d}^{D} - \boldsymbol{\pi} \right)^{2} \frac{1}{f} \\ &= \frac{1}{L} \sum_{d=1}^{L} \mathbf{Var} \left(\boldsymbol{\pi}_{d}^{D} \right) \\ &= \frac{1}{L} \sum_{d=1}^{L} \sqrt{\mathbf{E}_{\mathbf{M}}} \mathbf{Var}_{\mathbf{D}} \left(\boldsymbol{\pi}_{d}^{D} \middle| \boldsymbol{\pi}_{d} \right) + \mathbf{Var}_{\mathbf{M}} \mathbf{E}_{\mathbf{D}} \left(\boldsymbol{\pi}_{d}^{D} \middle| \boldsymbol{\pi}_{d} \right) \frac{1}{f} \end{split}$$
 by a basic statistical theorem because
$$\frac{1}{L} \sum_{d=1}^{L} \mathbf{E}_{\mathbf{M}} \left[\frac{\boldsymbol{\pi}_{d} \left(1 - \boldsymbol{\pi}_{d} \right)}{\boldsymbol{n}_{d}} \right] + \frac{1}{L} \sum_{d=1}^{L} \mathbf{Var}_{\mathbf{M}} \left(\boldsymbol{\pi}_{d} \right) \right]$$

$$= \frac{1}{L} \sum_{d=1}^{L} \left[\frac{\boldsymbol{\pi} \left(1 - \boldsymbol{\pi} \right) - \boldsymbol{\varphi}}{\boldsymbol{n}_{d}} \right] + \boldsymbol{\varphi}$$

$$= \frac{\left[\boldsymbol{\pi} \left(1 - \boldsymbol{\pi} \right) - \boldsymbol{\varphi} \right]}{\boldsymbol{n}_{d}} + \boldsymbol{\varphi}. \end{split}$$

¹⁵See for example, Hansen, Hurwitz, and Madow, Volume 2, Section 6, Theorem 15.

The left hand side of the final line is the measurement variance due to sampling, while the right hand side is the process variance. Note that the sum of the two must be larger than φ by itself. This is the same as saying that the expected dispersion of π_1^D through π_L^D is greater than the expected dispersion of π_1 through π_L . The difference between the two expected dispersions is measurement variance. In other words, it is a mistake to think that true propensities for the domains are as spread out as the design-based estimates of those propensities unless measurement variance is negligible. *Exhibit 3.8* illustrates this with some numbers. For this example, the left most column shows invented propensities for four domains. Six simple random samples of 1,062 were then drawn from each domain. (This was a common domain sample size actually used in this study.) As the table shows, the standard deviation among the estimated domain propensities can be smaller than the true standard deviation but is more often larger -- sometimes far larger. Furthermore, when the estimated standard deviation is too large, then the range is also too large. If the estimated range of the design-based estimates is too large, then the ideal model-based procedure will produce a range ratio less than one.

By means of both the derivation of the total dispersion and the illustration, it is now clear that a range ratio of one is ideal only when measurement error is negligible. This can be achieved by having a large intraclass correlation and a large sample size per domain. In light of this discussion, it is evident that the best model is not necessarily the one that most closely tracks the values of the design-based statistics across the domains. If the variation in the process is large relative to the measurement variance, then close tracking to the design-based statistics is ideal, but otherwise the goal is to obtain a tighter set of statistics with a smaller population variance across the domains and a smaller range across the domains.

Exhibit 3.8 Illustration of Extra Variation in Domain Estimates Due to Sampling Error

Actual domain propensity	1	Estima	ted domaii	n propensit samples of			random
		Sample 1	Sample 2	Sample 5	Sample 6		
	3.91	4.32	3.87	4.45	4.35	5.64	2.83
	3.54	3.89	3.72	3.82	3.21	2.77	3.97
	3.31	4.21	3.85	2.87	2.40	2.93	3.47
	3.20	3.70	2.98	3.52	2.94	2.45	3.42
Averages							
	3.49	4.03	3.61	3.67	3.22	3.44	3.42

Standard deviations						
0.31	0.29	0.42	0.66	0.82	1.47	0.46

The effect of measurement error on the correlation coefficient is more difficult to determine. More theoretical work or perhaps simulations would be useful. Note that the Wald test accounts for measurement error. Thus, the overdispersion of the design-based estimates does not mean that the ideal model should produce estimates that demonstrate lack of fit.

Given the effect of measurement variance on the range ratio, the question then becomes, what is the ideal value for the range ratio? This research did not establish the ideal value. However, *Exhibit 3.9* does show the relative overdispersion of the design-based estimates for illustrative values of the intraclass correlation and for the group sample size. For example, if the intraclass correlation is δ =0.1 (a large value) and the group sample size is 1,062, then the overdispersion in the design-based estimates is trivial. For such a situation, the ideal range ratio is equal to one, the ideal correlation coefficient is one, and the best model-based estimator would not shows signs of a lack of fit. On the other hand, when the intraclass correlation is just 0.01 and the sample size per group is just 266, then the overdispersion is 37 percent, meaning that the ideal model-based estimator will have 37 percent smaller dispersion across the domains than the design based estimator. Unfortunately, no measures of intraclass correlation were calculated for this study. Given a value for the observed variance in the design-based estimates across the groups and an overall mean, it would be fairly easy to compute the intraclass correlation, but this idea was only conceived as the final report was being written. At that point, there was inadequate time and funding to conduct more analysis.

Exhibit 3.9 Relative Overdispersion in Design Based Estimates by Intraclass Correlation and Group Sample Size

	Sample Size per Domain					
Intraclass						
correlation	1,062	531	266			
0.100	0.8%	1.7%	3.4%			
0.050	1.8%	3.6%	7.1%			
0.025	3.7%	7.3%	14.7%			
0.010	9.3%	18.6%	37.2%			
0.005	18.7%	37.5%	74.8%			
0.001	94.1%	188.1%	375.6%			

However, it is noted here that intraclass correlation where the classes are defined to be states was estimated to be 3% for alcohol use, 0.4% for cigarette smoking, and less than 0.1% for drug treatment and arrest. When the model fits, the procedure used to form the artificial domains should maximize the intraclass correlation for a given number of the domains. Also, the number of artificial domains formed for alcohol and cigarette usage were 40 each per age domain. Only 20 artificial domains were formed per age domain for the other behaviors. Since the average sample size per age domain was 21,244, the sample size per artificial domain was 531 for alcohol and cigarettes and 1,062 for the other behaviors. (The column with 266 per domain reflects the situation when 20 groups are formed per age group for a one-quarter sample as was used for the cross-validation discussed in Section 2.2.3 below.) It is hoped that the intraclass correlations for the artificial domains should be at least 5-10%, indicating that range ratios for good models should be only slightly below one.

Model Overfitting and Random Partitions

Recall that all of the discussion of the measures of agreement so far have been preconditioned on the existence of a partition that was fixed prior to analyzing the data. The interpretation becomes more difficult when one considers the random nature of the partition. For a fixed partition, it is expected that the model-based estimators will have less variation across the groups than the design-based estimator. For a model-dependent partition, the opposite can easily be true due to model overfitting.

When the model contains inappropriate (nonsignificant) predictors, or predictors with poorly estimated coefficients, the model will tend to produce predictions for individual cases that are outliers; that is, cases with inordinately big or small propensity predictions. Such models are

referred to as "overfit" since they include too many predictors. If there are only a small number of potential predictor variables, this problem does not arise since rigorous techniques exist to decide which of the predictor variables should be retained in the model. When, however, there are hundreds of potential predictor variables, as was the case on this project, those procedures tend to keep too many variables in the model.

Choosing prediction variables is an aspect of the statistician's task that was discussed in Chapter 1. Recall that the goal of any modeling exercise is to find internally homogenous groups. In this study, all the potential predictor variables were categorical variables. Picking predictor variables for the logistic regression was therefore equivalent to finding groups of sample persons that were internally homogeneous with respect to the propensity to engage in the behavior of interest. By the same logic, finding the largest set of significant predictor variables was equivalent to searching for the smallest set of groups such that it was impossible to break any of the groups into subgroups with different true propensities. However, all that was observed for each person was a yes or a no (a 1 or a 0). As different ways to break a tentative group into subgroups were examined to see if their propensities were different, their observed rates of engaging in the behavior of interest were calculated. If the observed rates were different enough, then it was concluded that the true propensities for the subgroups must be different. Each time that this test is conducted, there is only a small probability that two subgroups with identical propensities will be mistakenly classified as having truly different propensities. However, when the test is repeated hundreds of times, the number of errors that are made of this type can become nontrivial. This problem can probably be ameliorated by having more stringent variable selection rules that minimize overfitting. For this study, predictors were kept in the models if their p-values were at least 0.05 using SUDAAN as described in Section 2.2.

Recall that the artificial partition was created by sorting on the predictions. When the model is overfit, all the outliers get grouped together -- the low outliers in the first group and the high outliers in the last group. As a result, on a partition created from an overfit model, the design-based estimator will tend to vary less across the groups than the model-based estimator that was used to create the partition. Note that this is the opposite relationship from what is expected for a fixed partition. As a result, the statistical properties of the range ratio on a model-dependent partition are sharply different from those on a fixed partition.

Let the true range R of propensities for a partition be defined as the range of π_1 through π_L for a particular manifestation of the superpopulation. Let R^F be defined as the range of π_1^F through π_1^F . Let R^M , R^M , and R^D be defined similarly. As discussed above, for a fixed partition,

 $ER^D > R$ because of measurement error. For a fixed partition, a range ratio less than one $(R^F < R^D)$ could thus mean either that the model is not rich enough or that R^F is less biased than R^D . On the other hand, a range ratio greater than one $(R^F > R^D)$ is a strong sign that the model is too rich (i.e., overfit).

When the partition depends upon a model, it is expected that the positive bias in \mathbb{R}^D will increase since the same data that led to the model also get used to calculate the π_d^D . The behavior of the range of the model-based estimates is more complex. The behavior of the range will depend strongly upon whether the partition is induced by the same model that yielded the model-based estimates and upon the extent of overfitting in the model that induced the partition. If the partition is induced by the same model that is being evaluated and that model is overfit, then the range statistic is likely to be inflated due to the impact of outliers. With worse overfitting, the inflation will also become worse. It is difficult to quantify the relationship between overfitting and inflation of the range, but range ratios greater than one are likely to be signals of overfitting.

If a very rich model is used to create a partition, and the model is capturing real signal rather than just generating noise, and a lean model is used to generate the estimated propensities, then the range ratio for the lean model-based estimators will be substantially less than one.

A correlation substantially less than 1 could signal a nonlinear monotone relationship between the model-based and design-based statistics but is more likely to signal either model lack-of-fit (either underfitting or overfitting) or large measurement variance in the design-based estimators. An examination of a graph can rule out a nonlinear monotone relationship, but it will generally be impossible to tell whether a value less than one means model lack-of-fit (underfitting or overfitting) or unstable design-based estimators. At any rate, values substantially less than one are not expected for a model-dependent partition since the data that led to the model also influence the design-based estimators, tending to create relatively high correlations.

3.2.3 Cross-Validation

Methods to assess the degree of overfitting in models do exist. One way is to divide the sample into groups prior to fitting the model, develop the model based on the sample in one group, and evaluate the fit using the other group. This method is quite rigorous with properties that are well understood. However, this method does not produce the best possible models since there is less sample available for identifying significant predictors and for estimating model

parameters. For this study, it was decided to use cross-validation after the model fitting by dividing the sample into groups, refitting the model on one group and evaluating the refit model on the other group.

Choosing the group sizes was difficult since it was desired to have a large subsample both for the model refitting and for the evaluation of the refitted model. The subsample for the model refitting had to be large since the final models were too rich (i.e., contained too many variables) to fit on samples much smaller than the original sample. The subsample for the evaluation of the refitted model also had to be large because of the problems for the range ratio discussed above when the artificial domains are small and because the Wald test requires a large sample size to be valid and to have respectable power to detect lack of fit. Accordingly, it was decided to divide the sample into four equal parts, where each part contained a uniform slice of every state and sample PSU. For each model that was evaluated using this **cross-validation** procedure, the parameters were estimated for the final model using three of the four subsamples pooled together; then the fit was evaluated using the remaining subsample. This operation was repeated four times, so that all possible combinations of three out of the four subsamples were used for parameter estimation and each of the four subsamples were used for model evaluation. The four results were then averaged.

Given the already high computational burden of the model-fitting procedure, it was decided that it was not feasible to refit all 88 models. To keep the cost down, the cross-validation was done for only 12 of the models. These consisted of models for six of the eleven characteristics on the 26-34 age group, both inside and outside of the oversampled MSAs.

Note that the structure of a model was not changed when it was cross-validated, meaning that the set of fixed and random effects remained the same. Since the models were refit on a sample 25 percent smaller than the full sample, the refit parameters are less stable than the original parameters. This means that there are likely to be more outliers at the person level, meaning that range ratios will tend to be inflated. This is further exacerbated by the fact that only one-fourth of the original sample was available for the validation. Taken together, these two factors led to the expectation that the range ratios would be much larger for the cross-validated sample than for the original sample.

One of the improvements that cross validation offers when combined with the model-dependent partitions discussed in 3.2.2 is that some of the difficulties in interpreting range ratios and correlation coefficients are ameliorated. Specifically, the bias in $\mathbb{R}^{\mathbb{D}}$ when

computed on the reserved one-quarter sample is no worse than its bias on a fixed partition. It is still a biased estimator of R due to the measurement variance in the π_d^D , but because the partition is independent of the data used to compute \mathbb{R}^D , the extra bias discussed above disappears. Offsetting that to some degree is the fact that the smaller sample sizes worsen the bias due to measurement variance. Which effect will be stronger depends on the situation. The range for the model-based methods, however, does not become any better behaved since the partition is still based on the predicted propensities for the units in the quarter sample. Therefore, the range of the model-based predictions is still expected to be biased upwards by the effect of outliers. The fact that the model parameters were fit on a separate data set (the other three quarters of the full data set) does not change this fact. It is still the case that model-based estimators will look quite different on partitions induced by themselves versus other models.

Another benefit of the cross-validation is that the estimated variance-covariance matrix Ψ 65^? has better validity. Recall that for the full sample, this variance was estimated as a conditional variance given the random effects, the fixed effects, and on the partition induced by the fitted model. This conditional definition meant that the covariance between π 50~ $\}^W$ and π 50~ $\}^D$ was estimated to be zero, even though it is clear that there is a substantial unconditional covariance between them due the fact that both use information in the y vector for each domain. In the cross-validation, the y vector for the one-quarter evaluation sample was not used in the estimation of π 50~ $\}^W$, meaning that the unconditional covariance is zero. Since the lack of a positive correlation between the two will tend to increase the size of π 50~ $\}^W$ - π 50~ $\}^D$, this means that cross-validated Wald statistics are expected to be larger than the ordinary Wald statistics, thereby leading to small (more significant) p-values.

3.2.4 Results

Exhibit 3.10 shows the results of comparing the survey-weighted empirical Bayes estimates with the design-based estimates for the artificial domains described in Section 3.2.1. Each domain is predicted by the model to be fairly internally homogenous with respect to the propensity to engage in the behavior of interest. The correlations are high for most of the behaviors and age groups, as expected with a model-dependent partition of the population. A few of them are below 0.9, a possible sign of poor model fit. The p-values tend to range fairly

Exhibit 3.10 Tracking of Survey-Weighted Empirical Bayes Estimates with Design-Based Estimates Across Model-Homogenous Subgroups

			Age (Group		
Behavior						All
	Statistic	12-17	18-25	26-34	35 Plus	Ages
Licit Drugs						
Past Month Cigarette Use	Correlation χ² probability Range Ratio	0.917 0.038 0.870	0.952 0.054 0.934	0.952 0.136 1.016	0.946 0.846 1.027	0.978 0.788 0.985
Past Month Alcohol Use	Correlation χ^2 probability Range Ratio	0.871 0.056 0.878	0.960 0.066 0.872	0.972 0.049 0.894	0.981 0.984 0.968	0.990 0.759 0.948
Illicit Drugs						
Past Month Any Illicit Drug Use	Correlation χ^2 probability Range Ratio	0.939 0.340 0.905	0.973 0.639 0.827	0.962 0.384 0.880	0.942 0.474 0.881	0.990 0.450 0.868
Past Month Any Illicit But Marijuana Use	Correlation χ^2 probability Range Ratio	0.916 0.611 0.923	0.926 0.089 0.801	0.941 0.774 1.028	0.845 0.566 0.821	0.973 0.609 0.879
Past Month Cocaine Use	Correlation χ^2 probability Range Ratio	0.824 0.864 0.856	0.878 0.094 0.740	0.920 0.803 0.880	0.903 0.611 0.889	0.970 0.607 0.849
Dependence						
Past Year Dependence On Illicit Drugs	Correlation χ^2 probability Range Ratio	0.912 0.871 1.031	0.868 0.778 0.882	0.827 0.625 0.894	0.723 0.670 1.147	0.966 0.537 1.000
Past Year Dependence On Alcohol	Correlation χ^2 probability Range Ratio	0.787 0.221 0.892	0.898 0.007 0.716	0.943 0.688 0.899	0.927 0.523 0.878	0.978 0.020 0.807
Treatment						
Past Year Treatment For Illicit Drugs	Correlation χ^2 probability Range Ratio	0.908 0.489 0.879	0.772 0.437 0.920	0.918 0.539 0.923	0.884 0.354 0.984	0.962 0.184 0.944
Past Year Treatment For Alcohol	Correlation χ^2 probability Range Ratio	0.776 0.428 0.978	0.842 0.680 0.748	0.763 0.612 0.899	0.878 0.037 0.814	0.948 0.159 0.826
Needing Treatment In Past Year	Correlation χ^2 probability Range Ratio	0.899 0.112 0.892	0.910 0.255 0.851	0.923 0.355 0.937	0.934 0.859 0.823	0.980 0.403 0.872
Arrest						
Past Year Arrested *Probability of observing the calculations.	Correlation χ^2 probability Range Ratio	0.961 0.623 0.778	0.883 0.225 0.883	0.911 0.902 0.952	0.878 0.589 1.123	0.977 0.416 0.950

^{*}Probability of observing the calculated difference in the predicted and direct estimates across evaluation subgroups given that there is no difference.

uniformly over the interval (0,1), a good sign for acceptance of the model as discussed in the prior section. The actual median of the values in the table is 0.467. The range ratios are close to 1. Taken together with the high correlations, this is a sign that the two sets of estimates track closely over the artificial domains.

A somewhat different pattern emerges from a table of cross-validated statistics as discussed in Section 3.1.4. *Exhibit 3.11* shows the cross validation results for the Survey-weighted Empirical Bayes method and for several fixed effect models. Note that the correlations are sharply smaller than in *Exhibit 3.10*, the p-values are usually slightly more significant, and range ratios are larger, all of which indicates a lack of agreement between the model-based and design-based estimators.

As discussed in Section 3.2.3, the attenuation of the cross-validated correlation coefficients and the increases in the cross-validated range ratios (both relative to the ordinary versions in *Exhibit 3.10*) are expected consequences of the reduced domain sample sizes available for the evaluation. These smaller sample sizes lead to higher measurement error on the design-based estimates for the domains and also more instability in the random partition since the first and last groups will have fewer observations in them. For these descriptive statistics, there does not appear to be a good way of judging whether the cross-validated results are more dramatic than one would expect and are therefore symptomatic of lack of fit. Nonetheless, the large range ratios substantially greater than one are certainly suggestive of overfit models.

There are good theoretical reasons, on the other hand, to place more confidence in the cross-validated p-values from the Wald statistics than on the original p-values, as discussed in Section 3.2.3. Note that the p-values in *Exhibit 3.11* are mostly below the corresponding values for 26- to 34-year olds in *Exhibit 3.10* despite the lesser power of the cross-validated test given the smaller sample sizes for the cross-validation. This tendency illustrates the danger of excessive conditioning in the definition of Ψ . Since there are other random events upon which there is still conditioning and since not conditioning upon those random events would tend to increase variances, these p-values may be a little too small. Nonetheless, the fact that all the p-values in *Exhibit 3.11* are below 0.5 and some are substantially below 0.5 constitutes reasonably strong evidence of lack of fit. This is seen in all the columns, including the column for the mixed models. Evidence of lack of fit is disturbing since it impinges on the validity of the prediction intervals, but the impact on prediction interval coverage was not quantified in this study.

Exhibit 3.11 Cross-Validating the Survey-Weighted Empirical Bayes Estimates and Several Fixed Effect Models

			F	Estimators Based of	on Fixed Effect Mod	iels
Behavior	Statistic	Survey- Weighted Empirical Bayes (Mixed Model)	Same Fixed Effects as in Mixed Model	County and Demographic Effects Only	Separate Demographic Effects for Big City & Remainders	Demographic Effects Only
Past Month Cigarette Use	Correlation χ^2 probability Range Ratio	0.765 0.015 1.095	0.737 0.007 1.084	0.622 0.008 0.956	0.514 0.000 0.440	0.584 0.000 0.272
Past Month Alcohol Use	Correlation χ^2 probability Range Ratio	0.866 0.280 0.969	0.858 0.231 0.841	0.832 0.108 0.966	0.824 0.023 0.685	0.839 0.001 0.524
Past Month Any Illicit Drug Use	Correlation χ^2 probability Range Ratio	0.728 0.043 1.618	0.704 0.036 1.500	0.659 0.132 1.038	0.573 0.080 0.510	0.637 0.109 0.310
Past Month Any Illicit Drug Use But Marijuana	Correlation χ^2 probability Range Ratio	0.636 0.412 1.450	0.615 0.249 1.451	0.392 0.242 1.213	0.212 0.131 0.256	0.297 0.131 0.153
Past Year Treatment For Illicit Drugs	Correlation χ^2 probability Range Ratio	0.588 0.287 1.420	0.481 0.275 1.509	0.407 0.386 1.348	0.532 0.294 0.278	0.561 0.297 0.219
Past Year Arrested	Correlation χ^2 probability Range Ratio	0.641 0.418 1.552	0.662 0.399 1.537	0.543 0.324 1.430	0.639 0.278 0.373	0.635 0.165 0.425
MEAN	Correlation χ² probability Range Ratio	0.704 0.243 1.351	0.676 0.199 1.321	0.576 0.200 1.159	0.549 0.134 0.424	0.592 0.117 0.317

Note: These tests were restricted to the 26- to 34-year-old age group due to the cost of computations. *Probability of observing the calculated difference in the predicted and direct estimates across evaluation subgroups given that there is no difference.

An important methodological issue in the table concerns how the random partition of the data set was created for each column. For the first three columns, the same model was used to generate the partition as was being evaluated. For the last two columns, the full weighted mixed effect model was used to generate the partition. This accounts for most of the dramatic difference between the range ratios in the last two columns and the other columns. If the simple models in the last two columns had been evaluated with respect to the partitions induced by themselves, the range ratios would have been near one.

When a model is cross-validated with respect to the partition induced by itself, the range ratio is almost always above one. This is true for the full mixed effects model, for the model with the random effects removed but all the fixed effects retained, and for the model with the random effects and all of the tract and block group level fixed effects removed. This model with only demographic and county-level fixed effects is leaner than the full fixed model and yet there are still range ratios above one. This is probably good evidence that the models are overfit—even the models with just demographic and county-level variables. Note that the overfitting appears to be worse for the rarer behaviors.

Another interesting observation about the table concerns the low correlations, small p-values and small range ratios in the last two columns. Since the partition for these columns was induced by the full mixed model, this demonstrates that the full mixed model is capturing important information that the simple models cannot. The design-based estimator does vary significantly across the cells of the partition, and the simple models are not sensitive enough to mirror that variation.

3.3 Comparisons with External Data Sources

This approach is useful and important for any study where a census is occasionally conducted of the population of interest with respect to the behavior of interest. For example, this approach is often used to judge the quality of small area estimators for labor force statistics since the Decennial Census has several labor-related questions. Even in that application, however, there are difficulties since the questions are worded slightly differently than in the Current Population Survey and in other surveys that measure employment and unemployment. Systematic measurement biases tend to make evaluation of comparisons difficult.

This is all the more true with a topic as sensitive as substance abuse. It is known that reporting of substance abuse is extremely sensitive, not just to question wording, but also to the setting of the interview, the procedures that the interviewer uses, the assurances of confidentiality

and so on. Furthermore, a substantial portion of the population that engages in the behaviors of interest (such as getting arrested) may not be accessible through a household survey since they reside in prisons at the time of interview or have such a tenuous relationship with any one household that they are not included in any household roster. Administrative databases are also subject to systemic biases. Also, the data collectors for administrative systems are often quite autonomous and thus may develop reporting idiosyncracies that compromise the comparability of the data across states and MSAs.

Despite the concern for the potential for differential nonsampling biases, comparisons were made between the NHSDA estimates and estimates from other sources. Data from three external sources were used: the Behavioral Risk Factor Surveillance System (BRFSS), the National Drug and Alcoholism Treatment Unit Survey (NDATUS), and the FBI's Uniform Crime Reports (UCR). Each of these external sources are first described. Then comparisons are made.

Behavioral Risk Factor Surveillance System (BRFSS): The BRFSS is a telephone survey conducted in all 50 States under cooperative agreements with the Centers for Disease Control and Prevention with no representation of persons living in households without telephones. State sample sizes are larger for many states than in the NHSDA, but different survey data collection agents are used in every State, making comparability across the states a difficult issue. However, definitions used for alcohol consumption and cigarette use are comparable to NHSDA definitions, since the BRFSS estimates reflect past month use. BRFSS estimates of alcohol consumption and cigarette use were compared to the survey-weighted empirical Bayes estimates. Studies have shown that reporting of substance use behaviors may be lower in telephone surveys than in face-to-face surveys, particularly for illicit drugs. The BRFSS State estimates are simple averages over the three years 1991 through 1993.

National Drug and Alcoholism Treatment Unit Survey (NDATUS): NDATUS is an inventory of all specialty substance abuse treatment facilities in the U.S. Based on reporting by State substance abuse agencies, it provides estimates (including adjustments for nonresponse) of the number of clients in treatment at a given point in time. NDATUS estimates of drug treatment volumes were compared to the survey-weighted empirical Bayes estimates. To develop an estimate of persons treated during a year, the NDATUS client counts (including drug only and combined drug and alcohol clients) were multiplied by the reciprocals of average lengths of stay, and adjusted to account for multiple treatment episodes in a year by the same individual. Estimates of length of stay and multiple episodes were obtained from the Drug Services Research Survey, conducted in 1990. These calculations were done within categories of treatment modality and applied separately to each State. No adjustment was made to account for the inclusion in the NHSDA estimates of persons reporting treatment through self-help groups, private physicians, or emergency rooms, none of which are counted in NDATUS. The State estimates are averages over only 1992 and 1993 since the 1991 estimates could not be adequately adjusted for nonresponse.

Uniform Crime Reports (UCR): The UCR compiles data from local jurisdictions on the number of arrests. UCR estimates of past year arrests were compared to the survey-weighted empirical Bayes estimates. For comparison with the NHSDA estimates, an adjustment to the UCR data was made to account for persons arrested more than once during a year, so the adjusted UCR estimates reflect number of persons arrested at least once. This adjustment was made within the four Census regions, using data on multiple arrests reported by arrestees in the NHSDA sample. The State estimates are simple averages over the three years 1991 through 1993.

Exhibit 3.12 summarizes the results of the comparisons across the 26 states for which NHSDA estimates were prepared. One positive finding is that the ratio of the range of the design-based NHSDA estimates across the 26 examined states was almost always smaller than the range in the external estimates across the same states for three out of the four characteristics. Similarly, relative standard error across the states is smaller with the design based estimates than with external estimates for two out of the four characteristics. Given the earlier discussion in this chapter about the impact of measurement error on the ranges and dispersion, the expectation had been that the design-based estimates would be more dispersed than the external estimates. The fact that this is not true constitutes reasonable evidence that there are important differences in nonsampling errors between the data systems.

Turning attention to the other range ratios in *Exhibit 3.12*, one observes the same sorts of patterns as in *Exhibits 3.1* through *3.4*. The survey-weighted empirical Bayes estimates always had a tighter range across the states than the design-based NHSDA estimates, as desired. Whether the reduction in range is too much or too little is impossible to say, but it is clear that all the estimators based on fixed models almost always produce greater shrinkage than the mixed effect model, as expected. The simple demographic model, with or without a separate set of effects for oversampled metropolitan areas, seems to produce clearly undesirable levels of

Exhibit 3.12 Comparison of Alternative Small Area Estimators to Estimates from External Sources

Comment [COMMENT1]: 4/12 --- tables corrected per Mike Witt's request and faxed to client

			Weighted					
		Empirica						
		(Mixed	Models)	Estin	nators Based on	Fixed Effect M	Iodels	Design-B
State	Estimate	With Full	Big City Sub-Sam	Same Fixed	County and	Separate Demographic	Demographic	ased 91-93 NHSDA
	from External	NHSDA	pled	Effects as in	Demographic	Effects for	Effects Only	Estimates
	Source	Sample	Model	Mixed Model	Effects Only	Big City & Remainder	Zireets omy	Listination
Alcohol Use	BRFSS							
National Estimate	50.90	53.46		53.43	53.01	53.41	53.40	
			53.59					53.01
Rank Correlation ¹		0.861		0.841	0.765	0.278	0.097	
2			0.854					0.807
Range Ratio ²	1.000	0.597	0.500	0.545	0.525	0.088	0.082	0.067
RSE ³	21	12	0.598 13	10	13	2	2	0.867 18
KSE	21	12	13	10	13	2	2	18
Cigarette Use	BRFSS							
National Estimate	23.10	27.16		27.68	27.73	27.49	27.43	
			27.27					27.66
Rank Correlation		0.491		0.631	0.670	0.485	0.472	
			0.610					0.473
Range Ratio	1.000	0.930		0.642	0.449	0.244	0.208	
Dan		10	0.786	_		2		1.048
RSE	10	10	8	6	4	3	3	11
Drug Treatment	(NDATUS)							
National Estimate	0.85	0.70	0.71	0.70	0.68	0.65	0.64	0.62
Rank Correlation		0.375	0.523	0.406	0.289	-0.160	-0.240	0.063
Range Ratio	1.000	0.349	0.326	0.324	0.432	0.079	0.080	0.615
RSE	38	18	16	17	18	4	5	31
Arrest	(UCR)							
National Estimate	4.10	1.64	1.69	1.62	1.66	1.61	1.60	1.57
Rank Correlation		0.350	0.351	0.389	0.356	0.510	0.450	-0.066
Range Ratio	1.000	0.361	0.415	0.252	0.298	0.136	0.100	0.518
RSE	27	26	24	16	16	9	7	30

¹ Correlation calculated by first ranking states and then calculating the correlation of the ranks.

Note: Estimates of prevalence rates have been multiplied by 100.

Source: SAMHSA, Office of Applied Studies, National Household Survey on Drug Abuse. Model-based estimates using 1991-1993 NHSDA data.

² Ratio of the range of the predicted values to the range of data from the external source.

³ Relative standard error among state estimates as a percentage.

shrinkage. The range ratios for the county demographic model for drug treatment and arrest are higher than for the full fixed model. This is a sign that it is possible to induce as much variation in state estimates of drug treatment and arrest with just county-level variables as with a combination of county, tract, and block-group level variables. Whether the extra induced variation is good or not constitutes an open question.

The data on rank correlations are difficult to interpret. First, since it is not known which set of estimates is best, it is hard to say whether a high rank correlation is desirable. Second, the patterns are not consistent across the behaviors. The full mixed model resulted in the highest rank correlation with the external data source only for alcohol. The very lean fixed effect models (basic demographic and basic demographic with big city effects) generally produced the lowest rank correlations, but by some fluke, produced the highest rank correlations for arrests.

It is interesting that the design-based estimates for drug treatment and arrests show essentially no correlation with the NDATUS and UCR data. This would appear to indicate that the two systems are measuring different phenomena while attaching the same name to those phenomena. The fact that the models with county level predictors have substantial rank correlations with NDATUS and UCR is probably due to the fact that variables from these databases were in those models. It is not clear though how those variables got picked for inclusion in the model, when at the state level there is no correlation. It may be that these variables are part of the overfitting problem. The fact that the NDATUS and UCR statistics vary more across the states than the design-based estimates despite the fact that they are not subject to sampling error and the fact that there is no rank correlation between them and the design-based estimates raises some concerns about the appropriateness of variables from these databases as predictors in the models. Although these variables may have better face validity than census variables for some data consumers, the potential that they are not consistently collected across the states means that their presence in the models could be undermining the purpose of creating a set of state estimates that will fairly rank the states.

The column for a big city subsampled model describes the results when the survey-weighted empirical Bayes procedure is applied to a subset of the 1991-1993 NHSDA. This subsample was drawn by dropping the oversample of the six targeted metropolitan areas. Dropping the oversamples (so that the areas were sampled at national rates) allowed the fitting of

¹⁶The included predictor variables were not, however, identical to the outcome variables of interest here. Otherwise, it would have been possible to get a perfect fit and have a rank correlation of 1.

a single mixed model for each age group and behavior instead of having to fit separate models for the big cities. The fact that the two survey-weighted empirical Bayes columns are so similar indicates that the oversample in the six MSAs had little effect on most state estimates.

3.4 Ideas for Future Evaluations

The evaluation focused on trying to determine whether the model-based estimates were more accurate than the design-based estimates. This effort was partially successful, and it is hard to see what more could be done. While it would be interesting to compare the variances of the model-based estimates with those of the design-based estimates, the variances are almost certainly smaller with the model-based approach. The reason that this comparison was not done is that the estimates of the variances on the design-based estimates for states and MSAs would be highly unstable. It would be difficult to make the comparisons. Also, there are questions about underestimation of the variances for the model-based estimates.

One area, however, where it does appear to be possible to do more concerns the coverage of the prediction intervals, both when the model is true and when the model is not true. Two ideas are presented in this section on how the coverage could be evaluated. These methods were not used in the study since they cannot be used to evaluate the accuracy of the small area estimates themselves, but they could be useful for assessing coverage properties.

Since the labor force items are fairly consistent between the Decennial Census and the Current Population Survey, a good test of the methodology for estimating the variances for the model-based estimates would be to prepare model-based estimates and their estimated variances using variables from the CPS from April of 1990 as outcome variables and variables from the 1980 Census as predictor variables. (It is possible to link the 1990 CPS back to the 1980 Census.) By using ten-year-old predictors, it would be possible to simulate the effects of having imperfect predictors. Another advantage of such an approach is that it would have an evaluation data set completely independent of the data set used to create the predictions. Such a study would be useful for assessing the coverage properties of the prediction intervals. The actual mean squared error reductions that would be achieved for labor force statistics could, however, be quite different from the reductions expected for NHSDA statistics.

If the logistics of matching 1980 Decennial Census data to 1990 CPS data are too difficult, ¹⁷ the coverage of the prediction intervals could also be assessed through a simulation study. For such a study, it would be necessary to specify a model for the U.S. population and then to create a simulated population of very large size. Various models could be utilized to simulate the population. The methods could then be tested using either the same model structure that was used to create the simulated population or some other model. Although it might seem to be a weak test of the methodology to evaluate it assuming that the modeler knows the correct model structure, it would be of interest for the NHSDA project since approximations were used in fitting the model and in estimating the variances for the estimates. Without simulation studies, the effect of these approximations remains unknown.

Using a different model structure in the modeling than in the population simulation constitutes a much more stringent test of a methodology, but such an approach requires more judgment to design and interpret. If the model used to simulate the population is sufficiently pathological, then no reasonable method can be designed. This is even true of design-based methods since these methods assume that domain totals are normally distributed across all possible samples. It is quite difficult to decide how severe a pathology to use in testing.

One possibility to avoid having to create the population is to assemble more years of old NHSDA sample and to treat the conglomeration as a population from which samples are drawn. The problem with this approach would be the population sizes of the clusters. Fairly large clusters are required for representative testing.

A related idea would be to rerun the model on only a subsample of the 1991-1993 NHSDA, dropping out large states like California from the modeling in order to see how well the model works in predicting the behavior rates for those states when no sample was available. Although this idea has the appeal of being simple, it seems like a statistically unstable decision method. By this, it is meant that luck of the draw could easily result in the method either being accepted or rejected. It is sort of similar to making a major decision based on a single coin toss. A more complex simulation program would be more costly but would also yield much more information for the decision.

¹⁷It is likely that the Census Bureau would insist on performing the match itself and on adding noise to the prediction variables to protect confidentiality.

3.5 Summary of the Evaluations

As noted previously, all of the evaluation measures that were carried out have limitations. However, some useful findings do arise from the evaluation. First, the California estimates agree well with the design-based estimates. The method was designed to provide such agreement for domains with large sample sizes, so even though this says nothing about the quality of estimates for domains with small sample sizes, it is at least comforting that the method functioned as intended. National estimates agreed fairly well with the design-based estimates but not as well as hoped. Estimates based on simple fixed models actually fit better at the national level for several of the behaviors of interest. This is because weights were used in the fixed models, making estimates based on this design consistent, and because the number of sample PSUs was not large enough to make the national aggregates from the mixed models converge to the design-based estimates.

Second, the survey-weighted empirical Bayes estimates displayed less dispersion across the states than the design-based estimates and more dispersion than the estimates based only upon fixed effects. The method was designed to provide this sort of compression of the dispersion, so this outcome is not surprising, but it is again comforting that the method functioned as intended.

Third, there is some indication that the fixed parts of the models were overfit. Since overfitting results in negatively biased variance estimates, it might have been better to set a higher hurdle for fixed effects to enter the model. On the other hand, there is also evidence that there was genuine predictive value in at least some of the county, tract and block group level summary data, so it would have been a mistake to set the hurdle so high that only demographic variables could have entered the model. For this study, the maximum p-value for a variable to be retained was set at 0.05. It might have been better set at 0.01.

Fourth, estimates from BRFSS, NDATUS and UCR for alcohol use, cigarette use, drug treatment, and arrest vary as much or more than the NHSDA design-based estimates across the states. Since theory indicates that the NHSDA design-based estimates vary too much across the states, it might be reasonable to conclude that the estimates from the other sources also vary too much across the states. If this is true, then it might have been better not to use NDATUS and UCR county-level variables in the modeling. Variables from the decennial census may not appear as relevant and can be updated only once every ten years, but at least they are collected consistently across the states.

These findings verify that the methodology is basically functioning as intended. That is reassuring in the sense that it appears to rule out major conceptual or programming errors. The findings also indicate small ways in which the methodology can be improved. However, open questions remain about the reduction in variance achieved through the use of modeling and whether the prediction intervals have the claimed coverage levels. More research is needed to prove that these objectives were actually attained.

4. Thoughts for Future Applications

There are a number of important areas for further research. Most critically, work needs to be done on improving the variance estimates and then using simulation studies to demonstrate that the revised prediction intervals have good coverage properties, thereby establishing the validity of the methodology. Once more accurate variance estimates are obtained, it would also be important for planning purposes to determine the relationship between the variances on survey- weighted empirical Bayes estimates and design-based variances. This information could then be used to predict the width of model-based prediction intervals prior to actually preparing the estimates. Additionally, there is room for improvement in the point estimates themselves. Finally, it is important to consider the quality of the change estimates that would be obtained from model-based estimates. Each of these issues is discussed in turn below.

4.1 Improvement of Estimated Variances

There were several problems in the estimation of variance. First, a small component of variance due to sampling PSUs was inadvertently omitted. The previously published estimates reflect only the variance of the fixed part of the model in nonsample PSUs, neglecting the variance due to assuming a zero random effect for all nonsample PSU, as explained in detail in Appendix G, Section G.1. Revised variance estimates have been prepared that show a modest impact of the error on most estimates. This correction would need to be routinized for future applications.

Second, the variance due to estimating components of variance was intentionally omitted due to the lack of theory on how to properly reflect this component of variance. There are some ideas on how to proceed in this area, but progress will not be easy. As a temporary measure, one could approximate an upper bound on this effect and then use that upper bound, thereby providing prediction intervals with conservative coverage properties; i.e., one would deliberately overestimate the mean square error so that when 95% coverage is claimed, the actual coverage would be higher than 95%. For a better solution to this problem, it might be necessary to adopt a fully Bayesian approach to the small area estimation. This would require developing an entirely new software system and documentation. Since such development would be costly, it would only be recommended if simulation studies indicated that neglecting the variance on the estimated variance components was seriously affecting the coverage of the prediction intervals and that a simple upper bound did not remedy those coverage problems.

Third, estimated variances for domains that span the age groups are too low because the methodology ignored the positive covariance between the age groups across areas. There was a plan to approximate this effect, but it could not be implemented due to budget constraints. If one could fit a single model for all the age groups, this problem would disappear, although one would need to rewrite sections of the software to maintain separate random effects by age group.

Fourth, there is some evidence that the models were overfit. This means that too many predictor variables were used. If this is true, this also resulted in underestimation of the variances. The problem is that when the model is overfit, all of the estimated random effects tend to be too small. Since the variance estimate involves a term with the sum of the expected squared random effects, the overfitting thereby results in variance estimates that are too small. This problem can be fixed by keeping the models leaner.

4.2 Validation

Validation is an intrinsically difficult task with small area estimation unless there is an occasional census on the variables of interest using the same measurement techniques. Short of a census, simulation studies are the best that can be done to assess the coverage properties of the prediction intervals. The internal measures that were done for this study were innovative and interesting, but don't provide sufficiently strong evidence of validity. Comparisons with other surveys are of limited value because methodologies across surveys differ and because each has sampling error.

Various simulation studies could be designed. These can be grouped into two broad categories. The first category includes studies in which the simulated population is created according to the same model that will be used to analyze it. The second category includes studies in which pathologies are introduced into the population so that the model applies only approximately. Although the first category might appear to be an easy test of a method, it would be of interest for the NHSDA project since approximations were used in fitting the model and in estimating the variance for the estimates. Without simulation studies, the effect of these approximations is unknown.

The second category of simulation study is a much more stringent test of a methodology, but requires more judgment to design and interpret. If the pathologies are severe enough, then no reasonable method can be designed to cope with them. This is even true of design-based methods. It is quite difficult to decide how severe a pathology to use in testing.

Ideally, both sorts of simulation studies would be conducted. The second category should, of course, be delayed until the results of the first category can be studied. There is no reason to test the method on pathological pseudo populations until it has passed the test on model-conforming pseudo populations.

4.3 Reduction in Variance Due to Modeling

Before accurate predictions can be made of the variance reduction achievable through the SAE methodology, it is necessary to improve the estimates of variance as discussed above. The correction of the problem regarding a neglected bias-squared term is particularly important. After making that correction, however, more theoretical work is required since the variance of the model-based estimates is not strictly inversely proportional to the sample size. As the sample size increases, the reduction in variance due to modeling is likely to decrease. A reasonable guess is that the reduction is somewhere in the range of 10 to 75 percent. More time and study would be required to make more precise projections.

4.4 Improvement of Point Estimates

The point estimates appear to be of good quality. However, the quality could probably be improved by fitting leaner models. The overfitting mentioned earlier has several consequences for the quality of the point estimates. First, the small area estimates are not shrunk quite as close together as they should be. Another outcome of overfitting is that resulting computer time problems force one to run separate models by age. If a single model could be run including all ages, then the parameters for effects that do not interact with age would be estimated more accurately. A third outcome is that the ordering of the states may not be as stable as it could be. For all three reasons, the models should not be allowed to be as rich on a repetition of the project.

One of the ways that this could be achieved is to reverse the order of elimination for county-level summaries vis-à-vis tract or block-group level summaries. Before, the county-level summaries were only admitted to the model if they were significant after accepting the tract and block-group level summaries for the same characteristics. It may be better to let the county-level variables in first and then to accept the tract and block-group level summaries if they are marginally significant. It may also be better to require higher significance for each variable prior to admitting it to the model, perhaps $\alpha = .01$ or even smaller instead of $\alpha = .05$.

4.5 Validity of Change Estimates

Estimates of change can be made by using the survey-weighted empirical Bayes procedure to prepare estimates for each time period and then forming the differences between the two sets. Estimates of change made in this manner will be more stable than estimates of change made using purely design-based methods. However, the model-based estimates of change for particular states reflect a blend of changes at the state and national level. Particularly for states with small sample sizes, the estimates are likely to mostly reflect nation changes. The validity of these change estimates will depend on the extent to which the national changes are uniform within the cells defined by the fixed effects in the model.

4.6 Sample Design

In this study, the methodology for small area estimation was developed after the sample had already been designed and the survey had been conducted. The 1991-1993 NHSDA was designed to be nearly optimal for national estimates by race-ethnicity domain by age group. If the objective of state specific estimates is fixed in advanced, it is possible to develop a sample design that is better suited to the objective. Specifically, if state estimates were made an objective, the number of sample PSUs should be increased, the PSUs should be stratified by state, and the sample of people should be more equally allocated across the states instead of being massed in the large states. Since all of these design features would be less than optimal for national estimates by race, ethnicity and age, deciding whether to adopt these features is a difficult question.

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Appendix A: Classification of Drug and Alcohol Dependence

CLASSIFICATION OF DRUG AND ALCOHOL DEPENDENCE

The Diagnostic and Statistical Manual of Mental Disorders, Revised Third Addition (DSM-III-R), published by the American Psychiatric Association is designed to be used by clinicians as well as researchers for making diagnosis of psychiatric disorders. Substance abuse and dependence are considered psychiatric disorders under DSM-III-R. DSM-III-R defines a person as dependent on a substance if they meet 3 out of 9 criteria for that substance. The method for estimating dependence from the NHSDA is based on the NHSDA questionnaire items that are assumed to approximate five of the nine DSM-III-R criteria. These five criteria include: unable to cut down on use; reduced social, occupational, or recreational activities; continued use despite knowledge of having a problem; tolerance; withdrawal. In the NHSDA algorithm, respondents are defined as dependent on a substance if they responded affirmatively to at least two of the five NHSDA criteria for that substance.

To evaluate this method for estimating dependence, NHSDA dependence estimates were compared to estimates from the National Comorbidity Survey (NCS).^{1,2} Conducted in 1991, the NCS was designed to provide nationally representative estimates of 14 psychiatric disorders, including substance abuse and dependence, using a structured diagnostic interview. It employed a multistage area probability sample of 8,098 respondents in the household population.

Based on this analysis it was concluded that the approximation to the DSM-III-R definition of drug dependence produces dependence estimates which are comparable to the NCS. The table below indicates estimates of dependence from the NHSDA from 1991 through 1993 and estimates of dependence from the NCS.

Twelve Month Estimates of Dependence from the NCS and NHSDA (Old and New Methods) for Persons 15-54 Years of Age (1991-1994)

DRUG	NCS	NHSDA		
	1991	1991	1992	1993
Any Illicit	1.8%	1.8%	1.4%	1.5%
Marijuana	0.9	1.2	0.9	0.9
Cocaine	0.4	0.5	0.4	0.3
Alcohol	4.4	5.7	4.7	5.2

¹This is included in a paper by Joan Epstein and Joe Gfroerer, A Method for Estimating Substance Abuse Treatment Need from a National Household Survey, that was presented at the 37th International Congress on Alcohol and Drug Dependence, August 20-25, 1995.

²Lifetime and 12-Month Prevalence of DSM-III-R Psychiatric Disorders in the United States. Results from the National Comorbidity Survey. R.C. Kessler, PhD; K.A.McGonagle, PhD; S. Zhao, PhD; C.B. Nelson, MPH; M. Hughes, PhD; S. Eshleman, MA; H.U. Wittchen, PhD; K.S.Kendler, MD. Arch Gen Psychiatry, Vol.51, P.P. 8-19, January 1994.

Appendix B: Definition of 25 MSAs Selected for Small Area Estimation

Appendix B. Definition of 25 MSAs Selected for Small Area Estimation

1. Anaheim-Santa Ana, CA 1,996,343 California, Orange County 1,996,343 2. Atlanta, GA 2,424,882 Georgia, Barrow County 25,179 Georgia, Butts County 12,970 Georgia, Cherokee County 77,733 Georgia, Clayton County 151,319 Georgia, Cobb County 391,285 Georgia, Coweta County 45,782 Georgia, De Kalb County 463,515 Georgia, Douglas County 59,987 Georgia, Fayette County 55,556 Georgia, Forsyth County 38,659 Georgia, Fulton County 38,659 Georgia, Winnett County 310,368 Georgia, Wenter County 51,116 Georgia, Newton County 35,330 Georgia, Paulding County 35,385 Georgia, Paulding County 35,385 Georgia, Spalding County 46,428 Georgia, Walton County 32,395 3. Baltimore, MD 1,996,133 Maryland, Anne Arundel County 357,569 Maryland, Baltimore County 58,006 Maryland, Carroll County 104,516 Marylan
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Maryland, Howard County
Maryland, Queen Annes County
Maryland, Baltimore City
4. Boston, MA
Massachusetts, Essex County
Massachusetts, Middlesex County
Massachusetts, Norfolk County
Massachusetts, Plymouth County
Massachusetts, Suffolk County

	1992 Population <u>Projection</u>
5. Chicago, IL Illinois, Cook County Illinois, Du Page County Illinois, Mchenry County	4,175,359654,320
6. Dallas, TX	2.120.056
Texas, Collin County	
Texas, Dallas County	
Texas, Denton County	237,971
Texas, Ellis County	70,322
Texas, Kaufman County	43,729
Texas, Rockwall County	22,044
T D CO	1 245 ((0)
7. Denver, CO	
Colorado, Adams County	
Colorado, Arapahoe County	29,43/
Colorado, Douglas County	
Colorado, Jefferson County	369 394
Colorado, Jerierson County	
8. Detroit, MI	3,592,682
Michigan, Lapeer County	60,050
Michigan, Lapeer County	60,050
Michigan, Lapeer County Michigan, Livingston County Michigan, Macomb County	60,050 96,340 603,448
Michigan, Lapeer County Michigan, Livingston County Michigan, Macomb County Michigan, Monroe County	60,050 96,340 603,448 107,869
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1992

Population Projection 13. Minneapolis-St. Paul 2.035,347 Minnesota, Chisago County......24,780 Minnesota, Dakota County228,159 Minnesota, Scott County47,001 Minnesota, Wright County54,470 14. Nassau-Suffolk, NY 2,177,956 7,085,772 15. New York, NY New Jersey, Sussex County 106,854 17. Oakland, CA 1,727,459

	1992
	Population
	Projection
18. Philadelphia, PA-NJ	
New Jersey, Burlington County	
New Jersey, Camden County	407,164
New Jersey, Gloucester County	
Pennsylvania, Bucks County	456,745
Pennsylvania, Chester County	
Pennsylvania, Delaware County	
Pennsylvania, Montgomery County	
Pennsylvania, Philadelphia County	1,303,956
19. Phoenix, AZ	1,770,449
Arizona, Maricopa County	
, 1	, ,
20. San Antonio, TX	1,040,318
Texas, Bexar County	941,916
Texas, Comal County	44,601
Texas, Guadalupe County	53,801
21. San Bernardino, CA	2 122 022
California, Riverside County	
California, San Bernardino County	
Camorina, San Bornaramo County	
22. San Diego, CA	2,089,322
California, San Diego County	2,089,322
23. St. Louis, MO-IL	
Illinois, Clinton County	
Illinois, Jersey County	
Illinois, Madison County	
Illinois, Monroe County	
Illinois, St Clair County	
Missouri, Franklin County	
Missouri, Jefferson County	
Missouri, St Charles County	
Missouri, St Louis County	
Missouri, St Louis City	315,187
24. Tampa-St. Petersburg, FL	1,822,154
Florida, Hernando County	
Florida, Hillsborough County	
Florida, Pasco County	
Florida, Pinellas County	

	1992
	Population
	Projection
25. Washington, DC	
District of Columbia	513,842
Maryland, Calvert County	43,912
Maryland, Charles County	84,579
Maryland, Frederick County	127,118
Maryland, Montgomery County	651,746
Maryland, Prince Georges County	611,812
Virginia, Arlington County	
Virginia, Fairfax County	
Virginia, Loudoun County	73,193
Virginia, Prince William County	178,043
Virginia, Stafford County	
Virginia, Alexandria City	98,061
Virginia, Fairfax City	16,699
Virginia, Falls Church City	8,157
Virginia, Manassas City	23,622
Virginia, Manassas Park City	5,176

Appendix C: Fitting the Models Given Variable Selection

Appendix C. Fitting the Models Given Variable Selection

As discussed in Section 2.4, the model fitting procedure was an iterative series of iterative subprocedures. The first subprocedure is based on knowing G, the variance-covariance matrix for the random effects. The objective of this subprocedure is to find β 55[^] and U 50 [^] that maximize $L_w(\beta, U|y, G)$.

C.1 Estimating β and U Given G

Because the natural log function is monotone increasing, finding β 55^ and U 50 ^ that maximize $L_w(\beta, U/y, G)$ is identical to finding β 55^ and U 50 ^ that maximize $\ln[L_w(\beta, U/y, G)]$. Taking the log will simplify further steps considerably. Standard results from calculus tell us that if a function of one variable has a maximum at an interior point of its domain, then the first derivative will be zero at that point and the second derivative will be negative. Similar results are available for functions of many variables. Recall that if f is a real-valued function of $x_1,...,x_n$ then the transpose of the Jacobian matrix of f is an f 1 matrix defined as

$$\frac{\partial f(x)}{\partial x} = \begin{bmatrix} \frac{\partial f}{\partial x_1}(x_1, ..., x_n) \\ DOTSVERT \\ \frac{\partial f}{\partial x_n}(x_1, ..., x_n) \end{bmatrix}$$

and the Hessian of f is an n x n matrix defined as

$$\frac{\partial^{2} f(x)}{\partial x^{2}} = \begin{bmatrix} \frac{\partial^{2} f}{\partial x_{1}^{2}}(x_{1},...,x_{n}) & DOTSAXIS \frac{\partial^{2} f}{\partial x_{1}\partial x_{n}}(x_{1},...,x_{n}) \\ DOTSVERT \\ \frac{\partial^{2} f}{\partial x_{n}\partial x_{1}}(x_{1},...,x_{n}) & DOTSAXIS \frac{\partial^{2} f}{\partial x_{n}^{2}}(x_{1},...,x_{n}) \end{bmatrix}$$

For f to have a local maximum at an interior point x^* , it is necessary that

$$\frac{\partial f(x^*)}{\partial x} = \begin{bmatrix} 0 \\ DOTSVERT \\ 0 \end{bmatrix}$$

_

Sometimes the determinate of this matrix is called "the Jacobian," sometimes the matrix itself is labeled "the Jacobian."

and sufficient that this condition hold and that

$$\frac{\partial^2 f(x^*)}{\partial x^2}$$

is a negative definite matrix, by which it is meant that

$$x' \frac{\partial^2 f(x^*)}{\partial x^2} x < 0 \quad all \ x \neq 0.$$

The transpose of the Jacobian matrix of the log of the likelihood function (with respect to the parameters — not the data) is often called the efficient score of the parameters for the data or simply the score function. In our application, we write

$$S_{W}(\beta,U|y,G) = \frac{\partial \ln[L_{W}(\beta,U|y,G)]}{\partial(\beta,U)}.$$

We now derive this score function

Note that

$$ln L_w = \sum_{i=1}^{nleft} [w_i y_i ln(\pi_i) + w_i (1-y_i) ln(1-\pi_i)]$$

Recall from our model that logit $(\pi_i) = X_i \beta + Z_i U$, where X_i is the ^{i-th} row of X and Z_i is the ^{i-th} row of Z. By inverting the logit transform, we have that

$$\pi_i = \frac{e^{X_i \beta + Z_i U}}{1 + e^{X_i \beta + Z_i U}}$$

From this, it easily follows that

$$1-\pi_i=\frac{1}{1+e^{X_i\beta+Z_iU}}.$$

Thus, we may write

$$\ln L_{w} = \sum_{i=1}^{n} \left\{ w_{i} y_{i} \left[(X_{i} \beta + Z_{i} U) - \ln \left(1 + e^{X_{i} \beta + Z_{i} U} \right) \right] - w_{i} (1 - y_{i}) \ln \left(1 + e^{X_{i} \beta + Z_{i} U} \right) \right\} - \frac{1}{2} U^{t} G^{-1} U - \frac{q}{2} \ln (2\pi) - \frac{1}{2} \ln |G|$$

$$= \sum_{i=1}^{n} \left\{ w_{i} y_{i} (X_{i} \beta + Z_{i} U) - w_{i} \ln \left(1 + e^{X_{i} \beta + Z_{i} U} \right) \right\} - \frac{1}{2} U^{t} G^{-1} U - \frac{q}{2} \ln (2\pi) - \frac{1}{2} \ln |G|.$$

Taking first the derivative with respect to β_1 , we have that

$$\frac{\partial}{\partial \beta_1} \ln(L_w) = \sum_{i=1}^n \left[w_i y_i X_{iI} - \frac{w_i X_{iI} e^{X_i \beta + Z_i U}}{1 + e^{X_i \beta + Z_i U}} \right]$$

$$= \sum_{i=1}^n X_{iI} w_i (y_i - \pi_i) ,$$

where X_{i1} is the first component of the row vector X_i .

Similarly,

$$\frac{\partial \ln}{\partial \beta_p}(L_w) = \sum_{i=1}^n X_{ip} w_i (y_i - \pi_i).$$

Let W be an n x n diagonal matrix with the sampling weights on the diagonal:

$$W = \begin{cases} w_1 & O \\ DOTSDIAG \\ O & w_n \end{cases}$$

where it is important to note that the n weights were standardized to sum to n. Otherwise, the procedure behaves as if we have a sample as large as the U.S. population. Then

$$\frac{\partial \ln}{\partial \beta}(L_w) = \sqrt{\sum_{i=1}^{n} X_{iI} w_i (y_i - \pi_i)} / DOTSVERT / \sum_{i=1}^{n} X_{ip} w_i (y_i - \pi_i) / \sum_{i=1}^{n} X_{ip} w_i (y_i - \pi_i)$$

Note that $W(y-\pi)$ is an n x 1 matrix and that X^t is p x n so that $X^tW(y-\pi)$ is a p x 1 matrix, as required. Having found the Jacobian of $\ln L_w$ with respect to β , we now need to find the Jacobian with respect to U. Here it is useful to note that

$$G^{-1} = \begin{cases} \sqrt{\frac{I_m}{2}} & O \\ \sigma_1 \\ O & \frac{I_r}{\sigma_2^2} \end{cases}$$

So that $U^tG^{-1}U$ is simply

$$\frac{1}{\sigma_1^2} \sum_{i=1}^m U_i^2 + \frac{1}{\sigma_2^2} \sum_{i=m+1}^{m+r} U_i^2 ,$$

where m is the number of states and r is the number of sample PSUs. Taking first the derivative with respect to U_1 , we have that

$$\begin{split} \frac{\partial}{\partial U_{1}} \ln(L_{w}) &= \sum_{i=1}^{n} \left[w_{i} y_{i} Z_{iI} - \frac{w_{i} Z_{iI} e^{X_{i} \beta + Z_{i} U}}{1 + e^{X_{i} \beta + Z_{i} U}} \right] \frac{\partial}{\partial U_{1}} \frac{1}{2} \left[\frac{1}{\sigma_{1}^{2}} \sum_{i=1}^{m} U_{i}^{2} + \frac{1}{\sigma_{2}^{2}} \sum_{i=m+1}^{m+r} U_{i}^{2} \right] \\ &= \sum_{i=1}^{n} Z_{iI} w_{i} (y_{i} - \pi_{i}) - \frac{U_{1}}{\sigma_{1}^{2}} \end{split}$$

Similarly, for the random effect corresponding to the mth state,

$$\frac{\partial}{\partial U_m} \ln(L_w) = \sum_{i=1}^n Z_{im} w_i (y_i - \pi_i) - \frac{U_m}{\sigma_1^2}$$

For the r random effects corresponding to the PSU-level random effects, we have

$$\frac{\partial}{\partial U_j} \ln(L_w) = \sum_{i=1}^n Z_{ij} w_i (y_i - \pi_i) - \frac{U_j}{\sigma_2^2} \quad m < j \le m + r.$$

We can thus write

$$\frac{\partial}{\partial U} \ln(L_w) = \begin{pmatrix} \sum_{i=1}^{n} Z_{il} w_i (y_i - \pi_i) & \frac{1}{\sigma_1^2} U_1 \\ DOTSVERT & DOTSVERT \\ \sum_{i=1}^{n} Z_{im} w_i (y_i - \pi_i) & \frac{1}{\sigma_1^2} U_m \\ \sum_{i=1}^{n} Z_{i,m+1} w_i (y_i - \pi_i) & \frac{1}{\sigma_2^2} U_{m+1} \\ DOTSVERT & DOTSVERT \\ \sum_{i=1}^{n} Z_{i,m+r} w_i (y_i - \pi_i) & \frac{1}{\sigma_2^2} U_{m+r} \end{bmatrix}$$

By stacking $\frac{\partial}{\partial \beta} \ln(L_w)$ above $\frac{\partial}{\partial U} \ln(L_w)$, we may write

$$S_{W}(\beta, U \mid y, G) = \frac{\partial}{\partial (\beta, U)} \ln(L_{W}) = \left[X^{t} W(y - \pi) - G^{-1} U \right]$$

Note that this an (p + q) x 1 matrix as required. In order to maximize L_w for fixed y and G, we need to find β 55^and U 50 ^such that all p + g rows of S_w are identically equal to zero. Since this system of equations is nonlinear in β and U, we need to use numerical methods to solve it. The method we used is called Fisher's method of scoring (or simply "the method of scoring").

To describe this method, we need to define another term. The "Fisher information matrix" (or simply "the information matrix") is the expected value of the product of the score function with its transpose with respect to the distribution of *y*. We write this as

$$J_{W}(\beta,U|y,G)=E(S_{W}S_{W}^{t}),$$

where the expected value is with respect to the modeled distribution for y and U, ignoring the sample design and treating G and β as fixed.

In our application, note that J_w is a (p + q) x (p + q) matrix. Deriving it for our application is made simpler if we write S_w in terms of a transform of y- π .

$$\varepsilon_i = \frac{y_i - \pi_i}{\pi_i (1 - \pi_i)}$$

Note that the expected value of ε_i given fixed π_i is ε ($\varepsilon_i | \pi_i$) = 0since ε ($y_i | \pi_i$) = π_i . Also note that

$$Var\left(\varepsilon_{i}|\pi_{i}\right) = \frac{\pi_{i}(1-\pi_{i})}{[\pi_{i}(1-\pi_{i})]^{2}}$$

$$=\frac{1}{\pi_i(1-\pi_i)}$$

and that $Cov(\varepsilon_i, \varepsilon_j | \pi_i, \pi_j) = 0$ since the individual outcomes are assumed to be independent once the random effects have been fixed. If we use C to denote the n x n diagonal matrix with $\pi_i(1-\pi_i)$ sequenced down the main diagonal as

$$C = \left\langle \begin{array}{c} \pi_{11}(1-\pi_1) & O \\ DOTSDIAG \\ O & \pi_n(1-\pi_n) \end{array} \right\rangle \quad \varepsilon = \left\langle \begin{array}{c} \varepsilon_1 \\ DOTSVERT \\ \varepsilon_n \end{array} \right\rangle$$

then we may write

$$\varepsilon = C^{-1}(y-\pi)$$

the covariance matrix for ε as $Cov(\varepsilon) = C^{-1}$, and the score function as

$$S_{W}(\beta, U|\epsilon, G) = \sqrt{\frac{X^{t}WC\epsilon}{Z^{t}WC\epsilon - G^{-1}U}}.$$

The Fisher information matrix for β and U given ε and G is then

$$J_{W}(\beta, U|\epsilon, G) = E \left\{ \begin{array}{ll} X^{t}WC\epsilon \\ Z^{t}WC\epsilon - G^{-1}U \end{array} \middle| \left[\epsilon^{t}CWX - \epsilon^{t}CWZ - U^{t}G^{-1}\right] \end{array} \right\},$$

where we note that C, W, and G^{-1} are all symmetric. Expanding the product, we have

Error!

We now note that

$$E\left(\varepsilon_{i}U_{i}\right)=E_{U_{i}}\left[E(\varepsilon_{i}|U_{i})\right]=E_{U_{i}}\left[E(\varepsilon_{i}|\pi_{i})\right]=E_{U_{i}}(0)=0$$

so that $E(\varepsilon U^t) = E(U\varepsilon^t) = 0$. We also note that by definition, $E(\varepsilon \varepsilon^t) = C^{-1}$ and $E(UU^t) = G$. Furthermore, since we are treating G as fixed and known, we can move the expected value operator inside to just the factors involving U and ε . We thus have

$$J_{_{W}} = \begin{bmatrix} X^{t}WCWX & X^{t}WCWZ \\ Z^{t}WCWX & Z^{t}WCWZ + G^{-1} \end{bmatrix}.$$

Returning to the definition of Fisher's method of scoring, this is an iterative method that starts with a guess at a solution and then improves upon the guess. The improvement equation in our application is

$$\frac{\sqrt{\beta}}{U} \frac{55^{\wedge(a+1)}}{50^{\wedge(a+1)}} = \sqrt{\beta} \frac{55^{\wedge(a)}}{U} + [J_w^{(a)}]^{-1} S_w^{(a)},$$

where the a superscript indexes iterations of the cycle.

As the algorithm converges, β 55 $^{\wedge(a+1)}$ - β 55 $^{\wedge(a)}$ \rightarrow 0, U 50 $^{\wedge(a=1)}$ – U 50 $^{\wedge(a)}$ \rightarrow 0, $S_w^{(a)}$ \rightarrow 0, and $J_w^{(a)}$ \rightarrow $J_w^{(\infty)}$. We note that the matrices X, W, Z and G^{-1} are all constant during the convergence. Only the matrix C changes in J_w . Note that if any of the π_i are close to 0 or to 1, then C will have a diagonal entry close to zero, thereby making J_w ill-conditioned (i.e., difficult to invert). In such a case, convergence may be a problem. Otherwise, it may be proven that this process will converge. It is our understanding, however, that it has not been proven that there is a single solution to S_w = 0. If there are multiple solutions, it may be best to start the system with several different initial guesses. See Section C.3 for more discussion of this point.

Although we have completely specified the procedure for finding β 55^and U 50 ^ given G, it is useful for discussion purposes to describe the algorithm in an equivalent but different form. To this end, we note that the score function can be rewritten as where

$$S_{w} = \left[X^{t}W(y-\pi) \right] = \left[X^{t}WCW(\zeta-X\beta-ZU) \right],$$

$$Z^{t}W(y-\pi)-G^{-1}U \right] = \left[X^{t}WCW(\zeta-X\beta-ZU) - G^{-1}U \right],$$

$$\zeta = \sqrt{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_1 U + \frac{y_1 - \pi_1}{w_1 \pi_1 (1 - \pi_1)}} \int_{X_1 \beta + Z_$$

Continuing to transform S_w, we have

$$S_{w} = \begin{bmatrix} X^{t}WCW\zeta \\ Z^{t}WCW\zeta \end{bmatrix} - \begin{bmatrix} X^{t}WCWX\beta \\ Z^{t}WCWX\beta \end{bmatrix} - \begin{bmatrix} X^{t}WCWZU \\ Z^{t}WCWZU + G^{-1}U \end{bmatrix}$$

$$= \begin{bmatrix} X^{t}WCW\zeta \\ Z^{t}WCW\zeta \end{bmatrix} - \begin{bmatrix} X^{t}WCWX \\ Z^{t}WCWX \end{bmatrix} \beta - \begin{bmatrix} X^{t}WCWZ \\ Z^{t}WCWZ + G^{-1} \end{bmatrix} U$$

$$= \begin{bmatrix} X^{t}WCW\zeta \\ X^{t}WCW\zeta \end{bmatrix} - \begin{bmatrix} X^{t}WCWX & X^{t}WCWZ \\ Z^{t}WCWX & Z^{t}WCWZ + G^{-1} \end{bmatrix} \begin{bmatrix} \beta \\ U \end{bmatrix}$$

$$= \begin{bmatrix} X^{t}WCW\zeta \\ Z^{t}WCW\zeta \end{bmatrix} - J_{w} \begin{bmatrix} \beta \\ U \end{bmatrix}$$

So if we had the final version of C available, we could solve S_w =0 by just setting This is a nice form since it corresponds to the form that would be used if ζ were the observed

$$\begin{bmatrix} \beta \\ U \end{bmatrix} = J_w^1 \begin{bmatrix} X^t W C W \zeta \\ Z^t W C W \zeta \end{bmatrix} .$$

random variable instead of y and if ζ were a normal random variable. For this reason, we have sometimes referred to ζ as the working linear variable since it allows us to express the solution for β and U as a linear function of the "data." Of course, iteration is still required for the solution in this form since one must have preliminary values of β and U in order to calculate ζ , C, and J_w .

We give the form of \mathcal{J}_W^{-1} below and prove that the form is correct, but some additional notation and a lemma are first required. Let $R = (WCW)^{-1}$ and $V = R + ZGZ^t$. Also, let $P = V^{-1} - V^{-1}X(X^tV^{-1}X)^{-1}X^tV^{-1}$. This matrix has several useful properties that we prove as needed throughout the appendix. The properties that we need to find \mathcal{J}_W^{-1} are that P is symmetric and that P is orthogonal to X, meaning that PX = 0.

Lemma C.1: *PX*=0

Proof:

$$PX = [V^{1} - V^{1}X(X^{t}V^{1}X)^{-1}X^{t}V^{1}]X$$

$$= V^{1}[X - X(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}X]$$

$$= V^{1}[X - XI]$$

$$= V^{1}0$$

$$= 0.$$

We are now ready to prove that

$$J_{w}^{-1} = \left[(X^{t}V^{-1}X)^{-1} - (X^{t}V^{-1}X)^{-1} (X^{t}V^{-1}Z) G \right]$$

$$-G(Z^{t}V^{-1}X) (X^{t}V^{-1}X)^{-1} G - GZ^{t}PZG$$

We will prove this by demonstrating that $J_w J_w^{-1} = I$. A similar argument would demonstrate that $J_w^{-1} J_w = I$, as required. There are four steps to the proof. Using the new notation, we have that

$$J_{w} = \left[\begin{array}{ccc} X^{t}R^{-1}X & X^{t}R^{-1}Z \\ Z^{t}R^{-1}X & Z^{t}R^{-1}Z + G^{-1} \end{array} \right]$$

We must show that

$$\begin{split} X^t R^{-1} X \left(X^t V^{-1} X \right)^{-1} - X^t R^{-1} Z G \left(Z^t V^{-1} X \right) & \left(X^t V^{-1} X \right)^{-1} &= \ I_{pxp} \text{,} \\ - X^t R^{-1} X \left(X^t V^{-1} X \right)^{-1} & \left(X^t V^{-1} Z \right) & G + X^t R^{-1} Z \left(G - G Z^t P Z G \right) &= \ 0_{pxq} \text{,} \\ Z^t R^{-1} X \left(X^t V^{-1} X \right)^{-1} - & \left(Z^t R^{-1} Z + G^{-1} \right) & G \left(Z^t V^{-1} X \right) & \left(X^t V^{-1} X \right)^{-1} &= \ 0_{qxp} \text{,} \\ - Z^t R^{-1} X \left(X^t V^{-1} X \right)^{-1} & \left(X^t V^{-1} Z \right) & G + \left(Z^t R^{-1} Z + G^{-1} \right) & \left(G - G Z^t P Z G \right) &= \ I_{qxq} \text{.} \end{split}$$

Proving the first condition, we have that

$$\begin{split} X^{t}R^{-1}X\left(X^{t}V^{-1}X\right)^{-1} - X^{t}R^{-1}ZG\left(Z^{t}V^{-1}X\right) & (X^{t}V^{-1}X)^{-1} \\ &= X^{t}R^{-1}\left[I - ZGZ^{t}V^{-1}\right]X\left(X^{t}V^{-1}X\right)^{-1} \\ &= X^{t}R^{-1}\left[V - ZGZ^{t}\right]V^{-1}X\left(X^{t}V^{-1}X\right)^{-1} \\ &= X^{t}R^{-1}RV^{-1}X\left(X^{t}V^{-1}X\right)^{-1} \\ &= X^{t}V^{-1}X\left(X^{t}V^{-1}X\right)^{-1} \\ &= I_{pxp}, \end{split}$$

as required. Proving the second condition, we have that

```
 \begin{array}{l} -X^{t}R^{-1}X(X^{t}V^{-1}X)^{-1}(X^{t}V^{-1}Z)\,G + (X^{t}R^{-1}Z)\,\left(G - GZ^{t}PZG\right) \\ &= X^{t}R^{-1}\left[ -X(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}Z + Z\left(I - GZ^{t}PZ\right)\right]G \\ &= X^{t}R^{-1}\left[ -X(X^{t}V^{-1}X)^{-1}X^{t}V^{-1} + \left(I - ZGZ^{t}P\right)\right]ZG \\ &= X^{t}R^{-1}V[-V^{-1}X(X^{t}V^{-1}X)^{-1}X^{t}V^{-1} + V^{-1}\left(I - ZGZ^{t}P\right)\right]ZG \\ &= X^{t}R^{-1}V[V^{-1} - V^{-1}X(X^{t}V^{-1}X)^{-1}X^{t}V^{-1} - V^{-1}ZGZ^{t}P]ZG \\ &= X^{t}R^{-1}V[P - V^{-1}ZGZ^{t}P]ZG \qquad \qquad by \ definition \ of \ P \\ &= X^{t}R^{-1}V[I - V^{-1}ZGZ^{t}]PZG \\ &= X^{t}R^{-1}[V - ZGZ^{t}]PZG \\ &= X^{t}R^{-1}RPZG \\ &= (X^{t}P)ZG \\ &= 0ZG \qquad \qquad since \ PX = 0 \ X^{t}P = (PX)^{t} = 0^{t} = 0 \\ &= 0_{pxq} \end{array}
```

as required.

Proving the third condition, we have that

$$\begin{split} Z^t R^{-1} X \left(X^t V^{-1} X \right)^{-1} &- \left(Z^t R^{-1} Z + G^{-1} \right) G \left(Z^t V^{-1} X \right) \left(X^t V^{-1} X \right)^{-1} \\ &= \left[Z^t R^{-1} X - \left(Z^t R^{-1} Z + G^{-1} \right) G Z^t V^{-1} X \right] \left(X^t V^{-1} X \right)^{-1} \\ &= \left[Z^t R^{-1} - \left(Z^t R^{-1} Z + G^{-1} \right) G Z^t V^{-1} \right] X \left(X^t V^{-1} X \right)^{-1} \\ &= \left[Z^t R^{-1} V - \left(Z^t R^{-1} Z + G^{-1} \right) G Z^t \right] V^{-1} X \left(X^t V^{-1} X \right)^{-1} \\ &= \left[Z^t R^{-1} V - \left(Z^t R^{-1} Z G Z^t + Z^t \right) \right] V^{-1} X \left(X^t V^{-1} X \right)^{-1} \\ &= \left[Z^t R^{-1} V - Z^t R^{-1} \left(Z G Z^t + R \right) \right] V^{-1} X \left(X^t V^{-1} X \right)^{-1} \\ &= \left[Z^t R^{-1} V - Z^t R^{-1} V \right] V^{-1} X \left(X^t V^{-1} X \right)^{-1} \\ &= 0 V^{-1} X \left(X^t V^{-1} X \right)^{-1} \\ &= 0 Q_{XXY} \end{split}$$

as required.

Proving the fourth condition, we have that

```
 -Z^{t}R^{-1}X(X^{t}V^{-1}X)^{-1}(X^{t}V^{-1}Z)G + (Z^{t}R^{-1}Z + G^{-1})(G - GZ^{t}PZG) 
 = [-Z^{t}R^{-1}VV^{-1}X(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}Z + (Z^{t}R^{-1}Z + G^{-1})(I - GZ^{t}PZ)]G 
 = [-Z^{t}R^{-1}V(V^{-1} - P)Z + (Z^{t}R^{-1}Z + G^{-1})(I - GZ^{t}PZ)]G \qquad \text{by definition of } P 
 = [-Z^{t}R^{-1}Z + Z^{t}R^{-1}VPZ + Z^{t}R^{-1}Z - Z^{t}R^{-1}ZGZ^{t}PZ + G^{-1} - Z^{t}PZ]G 
 = [Z^{t}R^{-1}VPZ - Z^{t}R^{-1}(V - R)PZ + G^{-1} - Z^{t}PZ]G \qquad \text{by definition of } V 
 = [Z^{t}R^{-1}VPZ - Z^{t}R^{-1}VPZ + Z^{t}PZ + G^{-1} - Z^{t}PZ]G 
 = G^{-1}G 
 = I_{qxq},
```

as required. This completes the proof that J_w^{-1} has the form claimed. Going back to the solution for $(\beta 55^{\circ}, U 50^{\circ})$ in terms of J_w^{-1} and ζ , we have that

$$\int_{W}^{|V|} Z^{i}WCW\zeta = \int_{W}^{|V|} X^{i}R^{-1}\zeta = \int_{W}^{|V|} Z^{i}WCW\zeta = \int_{W}^{|V|} X^{i}R^{-1}\zeta = \int_{W}^{|V|} Z^{i}R^{-1}\zeta = \int_{W}^{|V|} Z^{i}W^{-1}X = \int_{W}^{|V|} Z^{i}R^{-1}\zeta = \int_{W}^{|V|} Z^{i}X^{-1}X = \int_{W}^{|V|} Z^{i}X = \int_{$$

In other words, given preliminary estimates of π , we may compute ζ , C, R, V, and P and then obtained improved estimates of β and U as

$$\beta 55^{\circ} = (X^t V^{-1} X)^{-1} X^t V^{-1} \zeta$$

and

$$U 50^ = GZ^t P\zeta.$$

Given improved estimates of β and U, we can get an improved estimate of π by

$$\pi 50^{\hat{}} = \frac{e^{X\beta} 50^{\hat{}} + ZU 50^{\hat{}}}{1 + e^{X\beta} 50^{\hat{}} + ZU 50^{\hat{}}}.$$

We iterate back and forth, first improving β 55^and U 50^ and then improving π 50^ until the process converges.

Recall, however, that this is all conditional on knowing G, which we don't. In the next section, we describe the estimation process for G.

C.2 Estimating G Given β U

Staying within the Bayes approach, the ideal procedure would be to derive the posterior distribution of G given y, β , and U, and then to find the mean of that posterior distribution or perhaps its mode. Unfortunately, this approach is not tractable. This would require integration of L_W not just with respect to U, but also with respect to β . As discussed in Section C.1, the integration of L_W with respect to just U was deemed too numerically intensive. Integrating with respect to the p components of β would raise the amount of required computer time to even higher levels.

In order to get some sort of estimate of G, we followed the approach used by Stiratelli et al (1984), Schall (1991), and Breslow and Clayton (1993). This approach works by adapting a method that has favorable properties when the outcome variable y has a multivariate normal distribution.

Specifically, consider the linear model

$$\zeta = X\beta + ZU + e$$

where X, β , Z, and U are defined the same as in our model, Cov(U,e) = 0, and $e \sim N_n(0,R)$. For this linear model, there are several alternate methods for estimating G. The maximum likelihood method finds the maximum of the likelihood of G given ζ . This means finding the value of G that maximizes

$$InL_{1}^{*}(G | \beta, \zeta) = -\frac{1}{2}ln|R + ZGZ^{t}| -\frac{1}{2}(\zeta - X\beta)^{t}(R + ZGZ^{t})^{-1}(\zeta - X\beta).$$

This method is favored by some statisticians, but others are displeased by the fact that G obtained by this method is different from the classical estimates for balanced ANOVAs and that the maximum likelihood estimate of G is seriously negatively biased when the number of fixed effects is large relative to the number of observations.

Because of these concerns, an alternative method of estimating G in linear models was developed. This method is called restricted maximum likelihood (REML). For this method, G is estimated by finding the value of G that maximizes

$$\ln L_{1} \quad (G \mid \beta, \zeta) = \frac{1}{2} \ln |R + ZGZ^{t}| - \frac{1}{2} \ln |X^{T}(R + ZGZ^{t})^{-1}X| + \frac{1}{2} (\zeta - X\beta)^{t} (R + ZGZ^{t})^{-1} (\zeta - X\beta),$$

where it is assumed that X^tX is of rank p so that the determinate in the second term is nonzero. The REML estimate G does agree with the classical G for balanced designs and it is not as biased when p is large relative to n. Further arguing in favor of the REML estimate of G, Harville (1977) noted that in Harville (1974) he had demonstrated that with a noninformative prior on (β, G) , the REML estimate of G maximizes the marginal posterior distribution of G, while the ML estimate of G is merely the value of G that together with some value of G maximizes the joint posterior distribution of G and G. For more information on REML methods, see Patterson and Thompson (1971) and Cox and Reid (1987).

We chose the REML estimator for this project because we believed that it might be more robust to the nonnormality of ζ in our application and because Breslow and Clayton conducted some simulations of it which appeared encouraging.

Further justifying the use of this approach, we note that although ζ as defined in Section C.1 is far from multivariate normal, it does have the correct mean and covariance matrix. Furthermore, if we equate $W^{-1}\varepsilon$ from our model with e in the linear model, we have that $Cov(U, W^{-1}\varepsilon) = 0$ as required and that $Cov(W^{-1}\varepsilon) = R = (WCW)^{-1}$.

Without further reference to nonnormality of $W^{-1}\epsilon$, we describe the algorithm we used to obtain the REML estimate of G. The method was iterative, as for the estimation of (β, V) . In fact, we again used Fisher's method of scoring as defined in Section C.1.

The score function for G using the REML approach is

$$S_1 (G | \beta, \zeta) = \frac{\partial}{\partial G} \ln L_1(G | \beta, \zeta).$$

Here we note that G has a very simple structure as specified in Section C.1. So we will define the score function as the derivative of $\ln L_1$, with respect to (σ_1, σ_2) . By finding the values of σ_1 and σ_2 (positive or negative) that maximize L_I , we will be able to estimate σ_1^2 by $(\sigma_1)^2$ and σ_2^2 by $(\sigma_2)^2$, thereby avoiding difficulties with negative estimates of variance components.

Starting with the first partial derivative with respect to σ_1 , we have

$$S(\sigma_{1}) = \frac{\partial}{\partial \sigma_{1}} \ln L_{1}$$

$$= \frac{\partial}{\partial \sigma_{1}} \left\{ \frac{1}{2} \ln |R + ZGZ^{t}| - \frac{1}{2} \ln |X^{t}(R + ZGZ^{t})^{-1}X| - \frac{1}{2} (\zeta - X\beta)^{t} (R + ZGZ^{t}) (\zeta - X\beta) \right\}$$

$$= \frac{\partial}{\partial \sigma_{1}} \left\{ \frac{1}{2} \ln |V| - \frac{1}{2} \ln |X^{t} V^{-1}X| - \frac{1}{2} (\zeta - X\beta)^{t} V^{-1} (\zeta - X\beta) \right\}$$

To derive this partial derivative, we need some results from stochastic linear algebra.

Lemma C.2: Let *A* be a symmetric nonsingular $n \times n$ matrix whose elements are functions of variables x_1, \ldots, x_t . Then

$$\frac{\partial}{\partial x_i} \ln |A| = \operatorname{trace}\left(A^{-1} \frac{\partial A}{\partial x_i}\right),\,$$

where

$$\frac{\partial A}{\partial x_i} = \begin{pmatrix} \frac{\partial a_{11}}{\partial x_i} & DOTSLOW & \frac{\partial a_{1n}}{\partial x_i} \\ DOTSVERT & \\ \frac{a_{n1}}{\partial x_i} & DOTSLOW & \frac{\partial a_{nn}}{\partial x_i} \end{pmatrix} ,$$

recalling that the trace of a matrix is the sum of the elements on its main diagonal. For the proof of Lemma C.2, see Graybill (1969), section 10.8.

Lemma C.3:

Let A and B be conformable matrices so that AB is defined. Then

$$\frac{\partial AB}{\partial \theta} = \frac{\partial A}{\partial \theta}B + A\frac{\partial B}{\partial \theta}.$$

Again, see Graybill for a proof.

Lemma C.4:

Let A be a square nonsingular matrix. Then $\frac{\partial A^{-1}}{\partial \theta} = -A^{-1} \frac{\partial A}{\partial \theta} A^{-1}$.

Proof:

Since
$$AA^{-1} = I$$
 and $\frac{\partial I}{\partial \theta} = 0$, we have

$$0 = \frac{\partial A}{\partial \theta} A^{-1} + A \frac{\partial A^{-1}}{\partial \theta}, \text{ by Lemma C.4}$$

leading to

$$-\frac{\partial A}{\partial \theta}A^{-1} = A\frac{\partial A^{-1}}{\partial \theta} \qquad \text{and} \qquad$$

$$-A^{-1}\frac{\partial A}{\partial \theta}A^{-1}=\frac{\partial A^{-1}}{\partial \theta},$$

as required.

Lemma C.5:

Let *A* and *B* be conformable matrices so that *AB* and *BA* are defined. Then tr(AB) = tr(BA). (See any linear algebra textbook for proof.)

This means that even though matrix multiplication is not commutative, it is possible to commute the order of multiplication inside the trace operator.

Lemma C.6:

Let x be a matrix such that $x^t A x$ is defined and x does not depend on θ . Then

$$\frac{\partial (x^t A x)}{\partial \theta} = x^t \frac{\partial A}{\partial \theta} x.$$

(Again, see Graybill for proof).

Lemma C.7:

$$V^{-1}(\zeta - X\beta) = P\zeta$$

Proof:

$$\begin{split} V^{-1} \; (\; \zeta - X \, \beta \;) \; &= \; \; V^{-1} \; (\; \zeta - X \, (\; X^t \, V^{-1} \, X \,)^{-1} \, X^t \, V^{-1} \;) \; \zeta) \\ &= \; \; V^{-1} \; (\; \mathcal{I} - X \; (\; X^t \, V^{-1} \, X \,)^{\; -1} \, X^t \, V^{-1} \;) \; \zeta \\ &= \; \; P \, \zeta \end{split}$$

Lemma C.8: If y is a random vector such that Ey = 0 with covariance matrix V and A is a fixed square conformable matrix, then

$$E(y^t A y) = trace(AV)$$

Since this lemma is simple to prove, the proof is not given here.

Lemma C.9:

$$\frac{\partial P}{\partial \sigma_i} = -P \frac{\partial V}{\partial \sigma_i} P$$

Proof:

$$\frac{\partial P}{\partial \sigma_i} = \frac{\partial}{\partial \sigma_i} (V^{-1} - V^{-1} X (X^t V^{-1} X)^{-1} X^t V^{-1})$$

$$= \frac{\partial V^{-1}}{\partial \sigma_i} (I - X(X^t V^{-1} X)^{-1} X^t V^{-1}) + V^{-1} \frac{\partial}{\partial \sigma_i} (I - X(X^t V^{-1} X)^{-1} X^t V^{-1})$$

$$= -V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) V^{-1} \left(I - X (X^t V^{-1} X)^{-1} X^t V^{-1} \right) - V^{-1} \frac{\partial}{\partial \sigma_i} \left(X (X^t V^{-1} X)^{-1} X^t V^{-1} \right)$$

$$= -V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) P - V^{-1} \left\{ X \left[\frac{\partial}{\partial \sigma_i} (X^t V^{-1} X)^{-1} \right] X^t V^{-1} + X (X^t V^{-1} X)^{-1} X^t \left(\frac{\partial V^{-1}}{\partial \sigma_i} \right) \right]$$

$$= -V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) P + \\ -V^{-1} \left\{ -X \left(X^t V^{-1} X \right)^{-1} \frac{\partial \left(X^t V^{-1} X \right)}{\partial \sigma_i} \left(X^t V^{-1} X \right)^{-1} X^t V^{-1} - X \left(X^t V^{-1} X \right)^{-1} X^t V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) V^{-1} \right\} \right\}$$

$$= -V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) P + \frac{1}{V} \left(X^t V^{-1} X \right)^{-1} X^t V^{-1} \left(\frac{\partial V^{-1}}{\partial \sigma_i} \right) V^{-1} X \left(X^t V^{-1} X \right)^{-1} X^t V^{-1} - X \left(X^t V^{-1} X \right)^{-1} X^t V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) V^{-1} \right) V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) V^{-1} \left(\frac{\partial V}{\partial \sigma$$

$$= -V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) P - V^{-1} X (X^t V^{-1} X)^{-1} X^t V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) \left[V^{-1} X (X^t V^{-1} X)^{-1} X^t V^{-1} - V^{-1} \right]$$

$$= -V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) P + (P - V^{-1}) \left(\frac{\partial V}{\partial \sigma_i} \right) (-P)$$

$$= -V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) P - P \left(\frac{\partial V}{\partial \sigma_i} \right) P + V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) P$$

$$= -P\left(\frac{\partial V}{\partial \sigma_i}\right)P$$

Returning now to the derivation of $S(\sigma_1)$, we have by Lemma C.2 that

$$\frac{\partial}{\partial \sigma_1} \ln |V| = \operatorname{trace} \left[V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right].$$

For the second term of $S(\sigma_1)$, we have that

$$\frac{\partial \ln}{\partial \sigma_1} |X^t V^{-1} X|$$

= trace [
$$(X^t V^{-1}X)^{-1} \frac{\partial (X^t V^{-1}X)}{\partial \sigma_1}$$
] by Lemma C.2

= trace [
$$(X^t \ V^{-1} X)^{-1} \ X^t \ \frac{\partial V^{-1}}{\partial \sigma_1} X$$
] by Lemma C.6

= -trace [
$$(X^t \ V^{-1} X)^{-1} \ X^t \ V^{-1}$$
 $\frac{\partial V}{\partial \sigma_1} \ V^{-1}X$] by Lemma C.4

= -trace [
$$V^1 X (X^t V^1 X)^{-1} X^t V^1 \frac{\partial V}{\partial \sigma_1}$$
] by Lemma C.5

Adding this result to the result for $\frac{\partial}{\partial \sigma_1} \ln |V|$, we have that

$$\frac{\partial}{\partial \sigma_1} \left[\, \ln |\, V| + \ln |\, X^t \, V^{\text{-}1} \, X \, | \, \right] = \, \, \text{trace} \left[\, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] - \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \frac{\partial V}{\partial \sigma_1} \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \, X^t \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X^t \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X \, V^{\text{-}1} \, X \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V^{\text{-}1} \, X\,)^{\text{-}1} \, X \, V^{\text{-}1} \, X \, V^{\text{-}1} \, X \right] + \, \text{trace} \left[\, V^{\text{-}1} \, X \, (\, X^t \, V$$

$$= \operatorname{trace} \left[V^{-1} \frac{\partial V}{\partial \sigma_1} - V^{-1} X (X^t V^{-1} X)^{-1} X^t V^{-1} \frac{\partial V}{\partial \sigma_1} \right]$$

$$= \operatorname{trace} \left\{ \left[V^{1} - V^{1}X \left(X^{t}V^{1}X \right)^{-1}X^{t}V^{1} \right] \frac{\partial V}{\partial \sigma_{1}} \right\}$$

$$= trace \left(P \frac{\partial V}{\partial \sigma_1} \right)$$

This is the general result given by Harville. Applying this general result to our particular application results in further simplification:

= trace
$$\left[P \frac{\partial}{\partial \sigma_1} (R + ZGZ^t)\right]$$
 by definition of V

$$= \operatorname{trace} \left[P Z \frac{\partial G}{\partial \sigma_1} Z^t \right]$$

Error!

= trace
$$\begin{cases} PZ & \begin{bmatrix} 2\sigma_1 0 \\ 0 \end{bmatrix} Z^t \end{cases}$$

$$= \operatorname{trace} \left\{ P \begin{bmatrix} Z & 50 \sim_1 \\ Z & 50 \sim_2 \end{bmatrix} \begin{bmatrix} 2\sigma_1 I_m 0 \\ 0 \end{bmatrix} \begin{bmatrix} Z_1 & 80 \sim^t \\ Z_2 & 80 \sim^t \end{bmatrix} \right\}$$

=
$$2\sigma_1 \operatorname{trace} \left\{ P[Z \ 50\sim_1 Z \ 50\sim_2] \begin{bmatrix} Z \ 50\sim_1^t \\ 0 \end{bmatrix} \right\}$$

=
$$2\sigma_1 \operatorname{trace} \left(PZ 55 \sim_1 Z 55 \sim_1^t \right)$$

=
$$2\sigma_1 \operatorname{trace} \left(Z 50 \sim_1^t PZ 50 \sim_1 \right)$$

where $Z 55\sim_1$ consists of the first *m* columns of *Z* and $Z 55\sim_2$ consists of the remaining *r* columns of *Z*.

Working on the third term of $S(\sigma_1)$, we have that

$$\frac{\partial}{\partial \sigma_1} \left[(\zeta - X\beta)^t V^{-1} (\zeta - Xbetahat) \right]$$

$$= \frac{\partial}{\partial \sigma_1} [\zeta - X\beta]^t P\zeta$$
 by lemma C.7

=
$$\frac{\partial}{\partial \sigma_1} [\zeta^t P \zeta]$$
 since $X^t P = 0$ by lemma C.1

$$= \zeta^t \left(\frac{\partial P}{\partial \sigma_1} \right) \zeta$$

$$= -\zeta^t P \left(\frac{\partial V}{\partial \sigma_1} \right) P \zeta \quad \text{by lemma C.9.}$$

=
$$-(\zeta - X\beta)^t V^{-1} \left(\frac{\partial V}{\partial \sigma_1}\right) V^{-1} (\zeta - X\beta)$$
 by lemma C.7

This is the general result given by Harville. Applying this general result to our particular application results in further simplification.

As demonstrated in the derivation of the second term of $S(\sigma_1)$, we know that in this application,

$$\frac{\partial V}{\partial \sigma_1} = 2\sigma_1 Z 55 \sim_1 Z 55 \sim_1^t .$$

So

$$\begin{split} &\frac{\partial}{\partial \sigma_{1}} \big[(\zeta - X\beta)^{t} V^{-1} (\zeta - X\beta) \big] \\ &= -2\sigma_{1} (\zeta - X\beta)^{t} V^{-1} Z \ 55 \sim_{1}^{t} Z \ 55 \sim_{1}^{t} V^{-1} (\zeta - X\beta) \\ &= -2\sigma_{1} \Big[Z \ 55 \sim_{1}^{t} V^{-1} (\zeta - X\beta) \Big]^{t} \Big[Z \ 55 \sim_{1}^{t} V^{-1} (\zeta - X\beta) \Big] \\ &= -2\sigma_{1} \xi_{1}^{t} \xi_{1} \quad , \\ &\text{where } \xi_{1} = Z \ 55 \sim_{1}^{t} V^{-1} (\zeta - X\beta) \end{split}$$

Finally, we can write out a complete expression for the score function for σ_1 using Harville's general form as

$$\begin{split} \mathcal{S}\left(\sigma_{1}\right) &= \frac{\partial \ln L_{1}}{\partial \sigma_{1}} &= \frac{\partial}{\partial \sigma_{1}} \sqrt[f]{-\frac{1}{2} \ln |V| - \frac{1}{2} \ln |X^{t} V^{-1} X|} - \frac{1}{2} (\zeta - X \beta)^{t} V^{-1} (\zeta - X \beta) \sqrt[f]{-\frac{1}{2} \ln |V| - \frac{1}{2} \ln |X^{t} V^{-1} X|} \\ &= -\frac{1}{2} \operatorname{trace} \left(\mathcal{P} \frac{\partial V}{\partial \sigma_{1}} \right) + \frac{1}{2} (\zeta - X \beta)^{t} V^{-1} \frac{\partial V}{\partial \sigma_{1}} V^{-1} (\zeta - X \beta) \end{split}$$

Equivalently for our application, we may write

$$S(\sigma_1) = -\sigma_1 \operatorname{trace} (Z 55 \sim_1^t PZ 55 \sim_1) + \sigma_1 \xi 50 \wedge_1^t \xi 50 \wedge_1^t$$

Going through similar algebra, we get the score function for σ_2 as

$$S(\sigma_2) = \frac{\partial \ln L_1}{\partial \sigma_2} = -\sigma_2 \operatorname{trace}(Z 55 \sim_2^t PZ 55 \sim sub2) + \sigma_2 \xi 50 \sim_2^t \xi 50 \sim_2^t$$

Just as in the estimation of (β, U) , we want to find (σ_1, σ_2) such that simultaneously solves the system $S(\sigma_1) = 0$ and $S(\sigma_2) = 0$. Clearly, $(\sigma 50^{\land}_1, \sigma 50^{\land}_2) = (0,0)$ is a root of the score function that does not interest us. It is interesting to note the lemmas C.5, C.7 and C.8 can be used to demonstrate that $E(\xi 50^{\land t}_i \xi 50^{\land t}_i) = (Z 50^{\land t}_i PZ 50^{\land t}_i)$ for σ_1 and σ_2 that solve the system. This means that finding the REML estimates of σ_1 and σ_2 is equivalent to setting $\xi 50^{\land t}_i \xi 50^{\land t}_i$ equal to its own expected value.

Solving the system requires an iterative approach. We did this using Fisher's method of scoring again. However, rather than computing the Fisher information matrix as

$$E \bigvee_{l} \begin{bmatrix} S(\sigma_{1}) \\ S(\sigma_{2}) \end{bmatrix} \begin{bmatrix} S(,\sigma_{1},) \\ S(,\sigma_{2},) \end{bmatrix} \bigvee_{l} \text{ we used a theorem from advanced probability theory.}$$

This theorem states that given suitable regularity conditions, $^2 \mathbb{E}(SS^t) = -\mathbb{E}\left[\frac{\partial S}{\partial \theta}\right]$, where

$$\frac{\partial S}{\partial \theta} = \begin{pmatrix} \frac{\partial^2 \ln f}{\partial \theta_1^2} & DOTSLOW & \frac{\partial^2 \ln f}{\partial \theta_1} \partial \theta_n \\ DOTSVERT & DOTSVERT \\ \frac{\partial^2 \ln f}{\partial \theta_1 \partial \theta_n} & DOTSLOW & \frac{\partial^2 \ln f}{\partial \theta_n^2} \end{pmatrix}$$

² See Cramer (1946) for a set of conditions and a proof.

is the Hessian of $\ln f$ with respect to θ .

For this particular application, it is easier to find the second partial derivatives of $\ln L_I$ and then to find the expected value of those second partial derivatives than to try to find the expected value of the product of the score function with its transpose. The reason for this is the difficulty of integrating products of traces. To find the second partial derivatives of $\ln L_I$, we go back to the equations for $S(\sigma_i)$ in terms of V to allow easier comparison with Harville (1977).

Focusing first on the first two components of lnL_I , we have that

$$\frac{\partial^{2}}{\partial \sigma_{i} \partial \sigma_{1}} \left[\ln |V| + \ln |X^{t} V^{-1} X| \right] = \frac{\partial}{\partial \sigma_{i}} \operatorname{trace} \left(P \frac{\partial V}{\partial \sigma_{i}} \right)$$

= trace $\left[\frac{\partial}{\partial \sigma_i} \left(P\left(\frac{\partial V}{\partial \sigma_j}\right)\right)\right]$ since the trace is a linear function of the elements of a matrix

$$= \operatorname{trace} \left[\left(\frac{\partial P}{\partial \sigma_i} \right) \! \left(\frac{\partial V}{\partial \sigma_j} \right) \! + P \! \left(\frac{\partial^2 V}{\partial \sigma_i \partial \sigma_j} \right) \right]$$

$$= \operatorname{trace} \left[-P \left(\frac{\partial V}{\partial \sigma_i} \right) P \left(\frac{\partial V}{\partial \sigma_j} \right) + P \left(\frac{\partial^2 V}{\partial \sigma_i \partial \sigma_j} \right) \right]$$

$$= \operatorname{trace} \left[P \left(\frac{\partial^2 V}{\partial \sigma_i \partial \sigma_j} \right) - P \left(\frac{\partial V}{\partial \sigma_i} \right) P \left(\frac{\partial V}{\partial \sigma_j} \right) \right]$$

$$= \operatorname{trace} \left[PZ \left(\frac{\partial^2 G}{\partial \sigma_i \partial \sigma_j} \right) Z^t - PZ \left(\frac{\partial G}{\partial \sigma_i} \right) Z^t PZ \left(\frac{\partial G}{\partial \sigma_j} \right) Z^t \right]$$

For i = j, this expression will equal

$$\mathsf{trace} \ \left[\ 2 \ PZ \ 55 \underset{i}{\sim}_{i} \ Z \ 55 \underset{i}{\sim}_{i}^{t} - 4 \ sigmasubi^{2} \ PZ \ 55 \underset{i}{\sim}_{i} \ Z \ 55 \underset{i}{\sim}_{i}^{t} \ PZ \ 55 \underset{i}{\sim}_{i} \ Z \ 55 \underset{i}{\sim}_{i}^{t} \ DZ \ 55 \underset{i}{\sim}_{i} \ Z \ 55 \underset{i}{\sim}_{i}^{t} \ DZ \ 55 \underset{i}{\sim}_{i} \ Z \ 55 \underset{i}{\sim}_{i}^{t} \ DZ \ 55 \underset{i}{\sim}_{i} \ Z \$$

For $i \neq j$, this expression will equal

-trace
$$\left[4\sigma_i\sigma_jPZ55\sim_iZ55\sim_i^tPZ55\sim_jZ55\sim_j^t\right]$$

since
$$\frac{\partial^2 G}{\partial \sigma_i \partial \sigma_j} = 0$$
 for $i \neq j$.

Focusing now on the third component of lnL_1 , we have that

$$\frac{\partial^2}{\partial \sigma_i \, \partial \sigma_j} (\zeta - X \beta)^t V^{-1} (\zeta - X \beta)$$

$$= -\frac{\partial}{\partial \sigma_i} \left\{ (\zeta - X \beta)^t V^{-1} \left(\frac{\partial V}{\partial \sigma_i} \right) V^{-1} (\zeta - X \beta) \right\}$$

$$= -\frac{\partial}{\partial \sigma_i} \left[\zeta^t P \left(\frac{\partial V}{\partial sigmasubi} \right) P \zeta \right]$$

$$= -\zeta^t \left[\left(\frac{\partial P}{\partial \sigma_i} \right) \left(\frac{\partial V}{\partial \sigma_j} \right) P + P \left[\left(\frac{\partial^2 V}{\partial \sigma_i \partial \sigma_j} \right) P + \left(\frac{\partial V}{\partial \sigma_j} \right) \left(\frac{\partial P}{\partial \sigma_i} \right) \right] \right] \zeta$$

$$= -\zeta^t \left[-P \left(\frac{\partial V}{\partial \sigma_i} \right) P \left(\frac{\partial V}{\partial \sigma_j} \right) P + P \left(\frac{\partial^2 V}{\partial \sigma_i \partial \sigma_j} \right) P - P \left(\frac{\partial V}{\partial \sigma_j} \right) P \left(\frac{\partial V}{\partial \sigma_i} \right) P \right] \zeta$$

$$= -\zeta^{t} P \left[-\left(\frac{\partial V}{\partial \sigma_{i}}\right) P\left(\frac{\partial V}{\partial \sigma_{j}}\right) + \left(\frac{\partial^{2} V}{\partial \sigma_{i} \partial \sigma_{j}}\right) - \left(\frac{\partial V}{\partial \sigma_{j}}\right) P\left(\frac{\partial V}{\partial \sigma_{i}}\right) \right] P \zeta$$

$$= -(\zeta^{t} - X\beta) V^{-1} \left[\frac{\partial^{2} V}{\partial \sigma_{i} \partial \sigma_{j}} - 2 \left(\frac{\partial V}{\partial \sigma_{j}} \right) P \left(\frac{\partial V}{\partial \sigma_{i}} \right) \right] V^{-1} (\zeta - X\beta)$$

(The last equality holds since $\frac{\partial^2 A}{\partial \sigma_i \partial \sigma_i} = \frac{\partial^2 A}{\partial \sigma_j \partial \sigma_i}$ for any twice differential matrix A.)

We have now reconfirmed the formula $\frac{\partial^2 lnL_1}{\partial \sigma_i \partial \sigma_i}$ given by Harville (1977):

Error!

Having found the Hessian of $\ln L_l$, we now need to find the expected value of this Hessian with respect to the distribution of ζ in order to get the Fisher information matrix for G. We first prove several lemmas.

Lemma C.10: $Cov(\zeta)=V$

Proof:

Lemma C.11: V is a generalized inverse of P, meaning that PVP=P

$$C \circ V \zeta = Cov(X\beta + ZU + W^{-1}C^{-1}(y - \pi))$$

$$= Cov(ZU) + Cov(W^{-1}\varepsilon) \qquad U \varepsilon$$

$$= Z(Cov U)Z^{t} + W^{-1}1(Cov \varepsilon)W^{-1} \quad (W is diagonal \ hence \ symmetric)$$

$$= ZGZ^{t} + W^{-1}C^{-1}W^{-1}$$

$$= ZGZ^{t} + (WCW)^{-1}$$

$$= ZGZ^{t} + R$$

$$= V.$$

Proof:

$$\begin{split} PVP &= \left[V^{-1} - V^{-1}X \left(X^{t}V^{-1}X \right)^{-1}V^{-1} \right] V \left[V^{-1} - V^{-1}X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1} \right] \\ &= V^{-1} \left[I - X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1} \right] \left[I - X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1} \right] \\ &= V^{-1} \left[I - 2X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1} + X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1}X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1} \right] \\ &= V^{-1} \left[I - 2X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1} + X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1} \right] \\ &= V^{-1} - V^{-1}X \left(X^{t}V^{-1}X \right)^{-1}X^{t}V^{-1} \\ &= P \,. \end{split}$$

Lemma C.12: $Cov(Ay) = A Cov(y)A^t$

The proof for Lemma C.12 is not given here but is elementary.

Continuing with the task of finding the expected value of the Hessian of lnL_l , we note that the

first term of $\frac{\partial^2 \ln L_1}{\partial \sigma_i \partial \sigma_i}$ is a constant and that the second term may be written as $\frac{1}{2} b^t A b$,

where
$$b = P\zeta = V^{-1} (\zeta - X\beta)$$
 is a

random variable with expected value 0, and

$$A = \frac{\partial^2 V}{\partial \sigma_i \partial \sigma_i} - 2 \left(\frac{\partial V}{\partial \sigma_i} \right) P \left(\frac{\partial V}{\partial \sigma_i} \right).$$

Using lemma C.8, we have that

$$E(b^t Ab) = trace(ACov(b)) = trace((Cov(b))A)$$

The covariance matrix for b is

Cov
$$b = \text{Cov}(P\zeta)$$

= $P \text{Cov}(\zeta) P$ by lemma C.12 since $P^t = P$
= PVP by lemma C.10
= P by lemma C.11.

Thus

$$\begin{split} & \mathbf{E}_{\zeta} \Bigg(\frac{\partial^{2} L_{1}}{\partial \sigma_{i} \partial \sigma_{j}} \Bigg) \\ &= -\frac{1}{2} \operatorname{trace} \left\{ \!\! / \!\! P \Bigg[\Bigg(\frac{\partial^{2} V}{\partial \sigma_{i} \partial \sigma_{j}} \Bigg) \!\! - \! \Bigg(\frac{\partial V}{\partial \sigma_{i}} \Bigg) \!\! P \Bigg(\frac{\partial V}{\partial \sigma_{j}} \Bigg) \right] \right\} \\ & + \frac{1}{2} \operatorname{trace} \left\{ \!\! / \!\! P \Bigg[\Bigg(\frac{\partial^{2} V}{\partial \sigma_{i} \partial \sigma_{j}} \Bigg) \!\! - \! 2 \Bigg(\frac{\partial V}{\partial \sigma_{i}} \Bigg) \!\! P \Bigg(\frac{\partial V}{\partial \sigma_{j}} \Bigg) \right] \right\} \!\! / \\ & = -\frac{1}{2} \mathrm{trace} \Bigg[P \Bigg(\frac{\partial V}{\partial \sigma_{i}} \Bigg) P \Bigg(\frac{\partial V}{\partial \sigma_{i}} \Bigg) \Bigg], \end{split}$$

as claimed by Harville.

Applying this general formula to our application, we have that

$$\begin{split} E_{\zeta} \bigg(\frac{\partial^2 L_1}{\partial \sigma_i \partial \sigma_j} \bigg) &= -\frac{1}{2} \operatorname{trace} \bigg[P Z \frac{\partial G}{\partial \sigma_i} Z^t P Z \frac{\partial G}{\partial \sigma_j} Z^t \bigg] \\ &= -2 \sigma_i \sigma_j \operatorname{trace} \bigg[P Z 55 \sim_i Z 55 \sim_i^t P Z 55 \sim_j Z 55 \sim_j^t \bigg] \end{split}$$

Thus, Fisher's information matrix is

Error!

Alternatively using lemma C.5, we can specify the (i,j)-th element of J_G as

$$J_G(i,j) = 2\sigma_i\sigma_j \left[(Z 55\sim_i^t PZ 55\sim_j)(Z 55\sim_i^t PZ 55\sim_j)^t \right]$$

This alternative formula is computationally useful since for an arbitrary matrix A, the trace of AA^{t} can be computed as the sum of the squared elements of A.

We now have all the information needed to write out our iterative cycle for estimating (σ_1, σ_2) using Fisher's method of scoring and the REML form of the log likelihood function.

Given initial estimates $\sigma_1^{(0)}$ $\sigma_2^{(0)}$, we calculated an initial value for G and hence V and P. We already had values for β to calculate ξ 50 $^{\circ}$. We then used these to obtain an initial value for the score function

Error!

and an initial value for Fisher's information matrix

Error!

Improved estimates of σ_1 and σ_2 were then obtained as

$$\begin{pmatrix} \sigma_1^{(1)} \\ \sigma_2^{(1)} \end{pmatrix} = \begin{pmatrix} \sigma_1^{(0)} \\ \sigma_2^{(0)} \end{pmatrix} + \begin{pmatrix} J_G^{(0)} \end{pmatrix}^{-1} S_G^{(0)}$$

This process is then allowed to iterate until convergence is obtained.

C.3 Joint Estimation of β , U, G

In Section C.1, an iterative method was described for estimating β and U, given G. In section C.2, another iterative method was described for estimating G given β U. These algorithms were run simultaneously with interchange of estimates back and forth to jointly estimate β , U, G.

We experimented with different ways of running the algorithms. The method that we chose as providing the fastest convergence was to iterate the first algorithm a single step, then to iterate the second algorithm a single step, then return to the first for one more step, then return to the second for one more step, and so on.

The main alternative we considered was to allow the first algorithm to converge, then to allow the second to converge, return to the first for another full convergence, and so on. This alternative proved to be slower. Other alternatives are possible such as allowing the second algorithm two steps for every one step of the first, but interchange after a single step of each algorithm appeared to be a good choice.

Even so, the computational burden was such that the full model (with state and PSU random effects) could not be run on a Pentium in a reasonable time. We ran the reduced models (with only state effects) on several Pentiums and the full models on a separate VAX server. For future applications, we would recommend the use of a dedicated VAX server or one or more fast UNIX workstations. For some states and variables, we could not get the full model to converge. When this occurred, we dropped back to the reduced model.

With an iterative procedure of this complexity, there is always the possibility of convergence to local maxima rather than global maxima. There is also the possibility of convergence to saddlepoints (where even though the score function is zero, the derivative of the score function is indeterminate, meaning that there are directions for change in the log likelihood that are both positive and negative). If the method were strict maximum likelihood, then there are theorems to indicate when the score function has a unique solution and that thus there were no saddlepoints and the only local maximum was also a global maximum. However, our method only approximates maximum likelihood and the method used to force $\sigma_1^2 > 0$ $\sigma_2^2 > 0$ does create saddlepoints since $S(\sigma_1, \sigma_2) = 0$ for $(\sigma_1^2, \sigma_1^2) = (0,0)$, which is clearly not an optimum estimate. Thus, the ideal procedure would have been to choose multiple starting points for the algorithm and to compare the results of starting it from those various points. However, the CPU time and VAX charges were higher per run than our time schedule and budget could support for multiple runs. We, therefore, ran each model only once, choosing the starting values carefully.

The starting values for β were taken from the ordinary logistic regressions for y or X without random effects. The starting values for U were taken to be zero, a reasonable procedure since the unconditional expectation of U is zero. The starting values for G were taken from by guessing at how much shrinkage should occur. See Section C.5 for a description of how shrinkage relates to the components of variance.

C.4 Asymptotic Properties of the Estimators

A sequence of random variables $\{Y_i\}$ is said to be consistent for a parameter θ if

$$\lim_{i\to\infty} \Pr\left(|Y_i - \theta| > \varepsilon\right) = 0 \text{ for every } \varepsilon > 0.$$

Similarly, two sequences $\{Y_i\}$ and $\{X_i\}$ are said to be asymptotically equivalent if $\lim_{i\to\infty} \Pr\left(|Y_i-X_i|>\epsilon\right)=0$ for every $\epsilon>0$. We claimed in Section 1.4.6 and again in Section 2.1.2 that

$$\lim_{n_i \to N_i} \frac{\pi \ 50^{N_i}}{\pi \ 50^{N_i}} = 1,$$

where *i* indexes the target small areas. By this rather loose statement, we meant that π 50 N_i and π 50 N_i are asymptotically equivalent as defined above. This claim has actually not yet been rigorously proven, but we do discuss our reasoning in this section for believing this assertion to be true. A rigorous proof might also require that N_i tend to infinity. Letting both n_i and N_i tend to infinity requires some care in specifying the stratification and clustering of the population as it grows larger that we do not attempt to address here.

We base our belief that π 50 $_i^w$ and π 50 $_i^D$ are asymptotically equivalent upon the observation that if the scaled residual error ε was normally distributed such that its covariance matrix did not depend on β , then the desired asymptotic equivalency can be proven. Recall from Section C.1 that we defined $\varepsilon = C^{-1}(y-\pi)$ and that we defined $\zeta = X\beta + ZU + W^{-1}\varepsilon$ with Cov(U) = G. If ε were multivariate with covariance matrix $C^{-1} = \sigma_e^2 I$, then our score function for β and U would have the same form as in Section C.1, but since the conditional covariance matrix for ζ given U would not depend on β or U, it would be possible to get an explicit formula for the root of the score function instead of having to use an iterative method to find the root. Note that these conditions hold for the standard linear mixed model where $y = X\beta + ZU + \varepsilon$, Cov(U) = G, $\varepsilon \sim N_n(O, I\sigma_e^2)$, U and U are independent and U is still defined as U and U are independent and U is still defined as U and U are independent and U is still defined as U and U are independent and U is still defined as U and U are independent and U is still defined as U and U are independent and U is still defined as U and U are independent and U are independent and U and U are independent and U and U are independent and U and U are independent and U and U are independent and U are independent and U are independent and U and U are independent and U and U are independent and U are indep

Recall that the score function for β and U given C, G and ζ is

$$S\left(\beta,U|C,G,\zeta\right) = \left(\begin{matrix} X^{t}WCW(\zeta-X\beta-ZU) \\ Z^{t}WCW(\zeta-X\beta-ZU)-G^{-1}U \end{matrix}\right).$$

Given a large enough sample size, we can assume that the method used to estimate C and G is accurate enough to treat C = C and G = G as fixed and known.

To keep discussion simple assume that there is only one level of random effects so that $G = \sigma_u^2 I_q$. (I_q is the qxq identity matrix, where m=q is the number of states in the model.)

Since $S(\beta 55^{\circ}, U 50^{\circ}|C, G, \zeta) = 0$, we know from looking at the bottom half of the score function that $Z^{t}WCW(\zeta-X\beta 55^{\circ}-ZU 50^{\circ}) = G^{-1}U 50^{\circ}$ so

$$Z^{\pm} \frac{WW}{\sigma_e^2} (\zeta - X\beta 55^{\wedge} - ZU 55^{\wedge}) = \frac{U 55^{\wedge}}{\sigma_u^2}.$$

By the definitions of y, ζ and ε , we know that $\varepsilon = W(\zeta - X\beta 55^{\circ} - ZU 50^{\circ})$ and that $\varepsilon = y - X\beta 55^{\circ} - ZU 50^{\circ}$. By equating these two expressions for ε , we have that

$$W(\zeta - X\beta 55^{\circ} - ZU 50^{\circ}) = v - X\beta 55^{\circ} - ZU 50^{\circ}.$$

Substituting this result into the earlier equation, we have that

$$\frac{Z^{t}W(y-X\beta 55^{\wedge}-ZU50^{\wedge})}{\sigma_{e}^{2}} = \frac{U50^{\wedge}}{\sigma_{u}^{2}}.$$

Collecting all the terms involving $U 50^{\circ}$, we have that

$$\frac{\sqrt{\frac{I}{\sigma_u^2}} + \frac{Z^t WZ}{\sigma_e^2}}{\sqrt{U}} = \frac{Z^t W(y - X\beta 55^{\wedge})}{\sigma_e^2}$$

So

$$U 55^{\circ} = \sqrt{\frac{\sigma_e^2}{\sigma_e^2}} I + Z^t W Z right]^{-1} Z^t W (y - X\beta 55^{\circ})$$

Note further that

where n_i is the number of sample persons in the i-th state and w_i is the average sampling weight in the i-th state.

Thus,

$$\int_{\frac{\sigma_{e}^{2}}{\sigma_{u}^{2}}} \frac{1}{n_{1}w_{1}} dt \int_{\frac{\sigma_{e}^{2}}{\sigma_{u}^{2}}} \frac{1}{n_{1}w_{1}} dt \int_{\frac{\sigma_{e}^{2}}{\sigma_{u}^{2}}}$$

where

$$\gamma_i = \frac{\sigma_u^2}{\sigma_u^2 + \frac{\sigma_e^2}{n_i w_i}}$$

is the shrinkage factor for the i-th state.

Since the weights were scaled, $w_i \ge 1$. So γ_i is the proportion of the total variance on the logit scale that is due to true differences among the states rather than measurement variance. For N_i sufficiently large,

$$\lim_{n_{i}\to N_{i}} \gamma_{i}-1.$$

Focusing on $U \ 50^{\circ}_{i}$ for an individual state, we note that the i-th row of $Z^{t}W(y-X\beta)$ has the form

$$\sum_{j=1}^{n_{i}} w_{ij} (y_{ij} - X_{ij} \beta 55^{\wedge}) = n_{i} w_{i} (y_{Di} - X_{Di} \beta 50^{\wedge}) ,$$

where $y_{Di} = \sum_{j=1}^{n_i} \frac{w_{ij}}{n_i w_i} y_{ij}$ is the design-consistent estimator of the mean of y in the i-th state, and

 X_{Di} is defined similarly. Combining these results leads to the explicit solution for $U \ 50^{\circ}_{i}$ of $U \ 50^{\circ}_{i} = \gamma_{i}(y_{Di}-X_{Di}\beta \ 50^{\circ})$.

Given this solution for $U \ 50^{\circ}_{i}$, the empirical Bayes estimator π_{i}^{w} for small area-i becomes

$$\pi 50^{\hat{}}_{i}^{w} = \frac{1}{N_{i}} \sum_{j=1}^{N_{i}} (X_{ij} \beta 50^{\hat{}} + U 50^{\hat{}})_{i}$$

$$= X_{0i}\beta 50^{\circ}_{i} + U 50^{\circ}_{i}$$

where $X_{\Omega i}$ depicts the population mean of the X_{ij} vectors. Substituting in the explicit solution for U_i leads to

$$\pi_i^{W} = X_{O_i} \beta + \gamma_i (y_{D_i} - X_{D_i} \beta)$$

$$= (1-\gamma_i)X_{\Omega i}\beta + \gamma_i[y_{Di}-(X_{Di}-X_{\Omega i}\beta)]$$

Note that y_{Di} is the traditional Horvitz-Thompsen estimator and $y_{Di} - (X_{DI} - X_{\Omega i})$ β is design consistent for the mean of y in state i since X_{Di} is a design consistent estimator of $X_{\Omega i}$. Therefore with $\pi_i^D = y_{Di}$ for this linear mixed model case

$$\lim_{n_i \to N_i} (\pi_i^w \div \pi_i^D) = 1$$

since X_{Ω_i} is a design consistent estimator of X_{Ω_i} . This establishes the desired result.

As mentioned earlier, the corresponding result for applying the method to binary variables has not yet been proven. The discreteness of the distribution for binary variables is not the primary difficulty in attempting a proof since with large enough n, the binary variables could be grouped into sets of binomial variables which would have approximately normal distributions by the Central Limit Theorem. The main difficulty in attempting a proof is that the matrix R is a function of β . The variance of a normal variable is usually not related to its mean. When there is a relationship, the method of proof given here fails.

Model Consistency

In addition to the question of design consistency, there is the issue of model consistency. Under fairly general conditions, maximum likelihood estimates are consistent for the parameters they are estimating. Less is known about the asymptotic properties of REML estimates, even for normally distributed errors. The model-based asymptotic characteristics of the approximations used here to estimate G are unknown, even when β is known. Given that in the procedure used

here, there is a pattern of alternating iterations between an approximate maximum likelihood estimate for (β, U) and an approximate restricted maximum likelihood estimate for G, the asymptotic frequentist properties of the method will be difficult to ever determine.

In a sense, however, the asymptotic design-based and model-based frequentist properties of the estimator are of only limited interest since the sample sizes available for a small area are small. Mean square error of the point estimates of π is of greater interest, along with the coverage properties of the confidence intervals. Simulation studies will probably be the only fruitful approach to studying these issues. Breslow and Clayton (1993) had some small simulation studies, but more are needed.

C.5 Computational Tricks and the Relationship of the Survey-Weighted Empirical Bayes Estimate to Composite Estimation

In this section, we show versions of the equations from Section C.1 that are faster to calculate than the forms given in that section. First, a useful lemma is reproduced from Robinson (1991).

Lemma C.13: Let A, U, B, and V be matrices such that A + UBV is well defined and invertible and such that A^{-1} exists. Then $(A + UBV)^{-1} = A^{-1} - A^{-1}U(I + BVA^{-1}U)^{-1}BVA^{-1}.$

Proof: We demonstrate below that the right-hand form of the equation is the inverse of A+UBV by multiplying it by A+UBV showing that the product is the identity matrix. Since matrix multiplication is not commutative, a full-proof would require evaluating the results of left-hand and right-hand multiplication, but we leave the other side of the proof to the interested reader.

```
 (A+UBV) [A^{-1}-A^{-1}U(I+BVA^{-1}U)^{-1}BVA^{-1}] \\ = (A+UBV) A^{-1} [I-U(I+BVA^{-1}U)^{-1}BVA^{-1}] \\ = (I+UBVA^{-1}) [I-U(I+BVA^{-1}U)^{-1}BVA^{-1}] \\ = I-U(I+BVA^{-1}U)^{-1}BVA^{-1} + UBVA^{-1} - UBVA^{-1}U(I+BVA^{-1}U)^{-1}BVA^{-1} \\ = I-U[(I+BVA^{-1}U)^{-1}-I+BVA^{-1}U(I+BVA^{-1}U)^{-1}]BVA^{-1} \\ = I-U[-I+(I+BVA^{-1}U)(I+BVA^{-1}U)^{-1}]BVA^{-1} \\ = I-U[-I+I]BVA^{-1} \\ = I-U[-I+I]BVA^{-1}
```

Applying lemma C.13 to V, we have that since $V=R+ZGZ^t$, its inverse may be written as

$$V^{1} = R^{-1} - R^{-1} Z(I + GZ^{t} R^{-1} Z)^{-1} GZ^{t} R^{-1}$$
.

Since R is a diagonal matrix, it is quite easy to find its inverse. Also $I+GZ^tR^{-1}Z$ is of dimension qxq, so it is much easier to find its inverse than to find the inverse of V directly since V is nxn. Because the matrix $I+GZ^tR^{-1}Z$ needs to be inverted very often in the iterative procedure, we developed a closed-form expression for the inverse rather than relying on numerical methods to invert the matrix repeatedly.

Once this matrix has been inverted, simplifications in the formulas for the random effects can be achieved that also help the intuitive understanding of the nature of these effects. First, we use lemma C.13 to derive an expression for estimating the random effects that is fast to calculate. From Section C.1, we have that

Error!

We will make this general matrix equation more specific to our application and even faster to compute by actually finding a closed form expression for the inverse of $I+GZ^tR^{-1}Z$. To find such an expression is messy, but the computing rewards are strong. We now need to consider that for some characteristics and age ranges, we were unable to get the full model with state and PSU random effects to converge. Where we were unable to obtain convergence, we simply dropped the PSU random effects from the model. The matrix inversion is considerably simpler for this simpler model and so we go through the steps for it first. If there is only a state effect, then the columns of Z are mutually orthogonal so $GZ^tR^{-1}Z$ is diagonal with I-th diagonal element

$$\sigma_{1}^{2} \sum_{i=1}^{r_{i}} \sum_{k=1}^{n_{ij}} w_{ijk}^{2} \pi_{ijk} (1 - \pi_{ijk})$$

We will simplify notation considerably by defining

$$\alpha_{ijk} = w_{ijk}^2 \pi_{ijk} (1 - \pi_{ijk})$$

and

$$\alpha_{i++} = \sum_{i=1}^{r_i} \sum_{k=1}^{n_{ij}} \alpha_{ijk}$$

and then using these α factors as weights to get weighted averages of the ζ and X variables:

$$X_{\underline{i}} = \sum_{i=1}^{r} \sum_{k=1}^{m_{ij}} \underbrace{\mathbf{Q}_{ijk}}_{ii+++} X_{ijkjk}$$

Then
$$U_{i} = \frac{1}{1 + \sigma_{1,j}^{2} w_{ij}^{2} \pi_{ij} (1 - \pi_{ij})} \sigma_{1} sup_{2} \sum_{j}^{n_{i}} w_{ij}^{2} \pi_{ij} (1 - \pi_{ij}) (\zeta_{ij} - X_{ij}) \delta_{5}^{5}$$

$$= \frac{\sigma_1^2 \alpha_{i++}}{1 + \sigma_1^2 \alpha_{i++}} \sum_{j=k}^{r_i n_{ij}} \left(\frac{\alpha_{ijk}}{\alpha_{i++}} \right) (\zeta_{ij} - X_{ij} \beta 55^{\wedge})$$

$$= \frac{\sigma_1^2}{\frac{1}{\alpha_{i+1}} + \sigma_1^2} (\zeta_i - X_i \beta 55^{\wedge})$$

$$= \gamma_i (\zeta_i - X_i \beta 55^{\wedge})$$

where

$$\gamma_i = \frac{\sigma_1^2}{\frac{1}{\alpha_{i+1}} + \sigma_1^2}$$

Note that since the weights were standardized to sum to the national sample size, α_{i++} is approximately equal to the sample size for the state multiplied by $\pi(1-\pi)$ where π is the true national average propensity. Thus,

$$\gamma_{i-} \frac{\sigma_1^2}{\frac{1}{n \cdot \pi(1-\pi)} + \sigma_1^2}$$

This form of the state random effect allows us to demonstrate a linkage between this estimation

system and composite estimation. Recall that $\lambda_{ijk} = \ln\left(\frac{\pi_{ijk}}{1-\pi_{ijk}}\right)$ is the estimated logit

propensity to engage in the behavior of interest. If we think of averaging the estimated logit propensities across the entire population in the i-th state, we obtain

Error!

This equation looks just like a classical composite estimator with the design-weighted sample data in the second term and the model-based prediction in the first term. Because of this similarity, we refer to the factor γ_i as the shrinkage factor for the *i*-th state. If γ_i is close to 1, then the sample data from the state predominate the estimated logit propensity for the state. On the other hand, if γ_i is close to zero, then national data predominate the estimated logit propensity for the state through the model. However, it is probably not appropriate to push this analogy too hard since ζ_i is a complex function of the sample data in the state and the true propensities in the state:

$$\zeta_{i} = \frac{\sum \sum_{j} \alpha_{ijk} \zeta_{ijk}}{\alpha_{i++}}$$

$$= \frac{\sum \sum_{j} \alpha_{ijk} \left[X_{ijk} + U_{i} + \frac{y_{ijk} - \pi_{ijk}}{w_{ijk} \pi_{ijk} (1 - \pi_{ijk})} \right]}{\alpha_{i++}}$$

$$= X_{i}\beta + U_{i} + \frac{\sum \sum_{j} w_{ijk} (y_{ijk} - \pi_{ijk})}{\alpha_{i++}}$$

$$= \frac{\sum \sum_{j} \alpha_{ijk} \operatorname{logit}(\pi_{ijk})}{\alpha_{i++}} + \frac{\sum \sum_{j} w_{ijk} (y_{ijk} - \pi_{ijk})}{\alpha_{i++}}$$

Also note that the composting takes place on the logit transform of the estimated propensity.

That completes the discussion of the simplified model with only state-level random effects. We now tackle the more difficult analogous equations for the full model with state-level and PSU-level random effects. With 2 levels of random effects,

$$I + GZ^{t}R^{-1}Z = \begin{bmatrix} I + \sigma_{i}^{2}Z_{1}^{t}R^{-1}Z_{1} & \sigma_{1}^{2}Z_{1}^{t}R^{-1}Z_{2} \\ \sigma_{2}^{2}Z_{2}^{t}R^{-1}Z_{1} & I + \sigma_{2}^{2}Z_{2}^{t}R^{-1}Z_{2} \end{bmatrix}$$

where Z1 and Z2 were defined in Section C.2. The columns of Z_1 are mutually orthogonal as are the columns of Z_2 , making this matrix easier to invert, but Z_1 and Z_2 , are not orthogonal to each other. We know that all columns involving different states are orthogonal. So we can re-order the random effects in the order U_1 , U_{II} , ..., U_{1r_1} ,, U_m , U_{mI} , ... U_{mr_m} and reorder the columns of Z accordingly. Let each block of columns corresponding to a state be written B_i so that $Z = \begin{bmatrix} B_1 DOTSAXISB_m \end{bmatrix}$.

Also let

$$Q_i = \begin{bmatrix} \sigma_1^2 & 0 \\ 0 & \sigma_2^2 I_{r_i} \end{bmatrix},$$

where r_i is number of sample PSUs in i-th state. Then

$$I+GZ^tR^{-1}Z = \begin{bmatrix} & I+Q_iB_1^tR^{-1}B_1 & 0\\ & DOTSDIAG\\ & 0 & I+Q_mB_m^tR^{-1}B_m \end{bmatrix},$$

where the I-th block on the main diagonal is $(1+r_i)x(1+r_i)$ in dimension. To invert this matrix, we need only invert each block. Taking the first block as an example, it has the form

$$I + \mathcal{Q}_1 B_1^t R^{-1} B_1 = \begin{pmatrix} \sqrt{1} + \sigma_1^2 \alpha_{1+} & \sigma_1^2 \alpha_{11+} & DOTSLOW & \sigma_1^2 \alpha_{1r_1+} & \sum_{r=1}^{r} \sigma_1^2 \alpha_{11+} & 1 + \sigma_2^2 \alpha_{11+} & 0 & \sum_{r=1}^{r} \sigma_1^2 \alpha_{1r_1+} & 0 & \sum_{r=1}^{r} \sigma_1^2 \alpha_{1r_1+} & 0 & 1 + \sigma_2^2 \alpha_{1r_1+} & \sum_{r=1}^{r} \sigma_1^2 \alpha_{1r_1+} & 0 & \sum_{r=1}^{r} \sigma_1^2 \alpha_{1r_1+} &$$

where $\alpha_{ijk} = w_{ijk}^2 \pi_{ijk} (1 - \pi_{ijk})$ and "+" denotes summation on a subscript.

Some grinding but routine linear algebra yields a closed-form expression for the inverse of this matrix. We don't give the derivation here, but merely the result.

Let
$$h = 1 + \sigma_1^2 \alpha_{1++} - \sum_{i=1}^{r_1} \frac{\sigma_1^2 \sigma_2^2 \alpha_{1i+}^2}{1 + \sigma_2^2 \alpha_{1i+}}$$
.

Then $(I+Q_1B_1^tR^{-1}B_1)^{-1}$ has the general form

$$\frac{1}{h} / 1 - \sigma_1^2 A^t / \left(-\sigma_2^2 A A A^t + h H \right) /$$

where

$$A^{t} = \sqrt{\frac{\alpha_{11+}}{1 + \sigma_{2}^{2} \alpha_{11+}}} DOTSLOW \frac{\alpha_{1r_{i}+}}{1 + \sigma_{2}^{2} \alpha_{1r_{i}+}} /$$

and

$$H = \begin{bmatrix} \frac{1}{1 + \sigma_2^2 \alpha_{11+}} & 0 \\ DOTSDIAG \\ 0 & \frac{1}{1 + \sigma_2^2 \alpha_{1r,+}} \end{bmatrix},$$

This means that the random effect for the first state is given by the first row of

$$(I+Q_1B_1^tR^{-1}B_1)^{-1}Q_1B_1^tR^{-1}(\zeta-X\beta 55^{\wedge})$$

The random effects for the sample PSUs in that state are given by the remaining rows. Concentrating on the first row, we have that

$$U_{1} = \frac{\sigma_{1}^{2}}{h} \sum_{j} \sum_{k} \alpha_{1jk} (\zeta_{1jk} - X_{1jk} \beta 55^{\wedge}) - \frac{\sigma_{1}^{2} \sigma_{2}^{2}}{h} \sum_{j} \frac{\alpha_{1j+}}{1 + \sigma_{2}^{2} \alpha_{1j+}^{k}} \alpha_{1jk} (\zeta_{1jk} - X_{1jk} \beta 55^{\wedge}).$$

If we let
$$\zeta_{ij} = \sum_{k} \frac{\alpha_{ijk}}{\alpha_{ij+}} \zeta_{ijk} \zeta$$
 $60 = \sum_{i} \frac{\sum_{j} \frac{\alpha_{ij+}}{\alpha_{ij+} \sigma_{2}^{2} + 1} \zeta_{ij}}{\sum_{j} \frac{\alpha_{ij+}}{\alpha_{ij+} \sigma_{2}^{2} + 1}}$

and define X_{ij} X = 60 = 100 similarly, then we have that

$$U_{1} = \frac{\sigma_{1}^{2}}{h} \sum_{j} \sum_{k} \left[\left(1 - \frac{\sigma_{2}^{2} \alpha_{1j+}}{1 + \sigma_{2}^{2} \alpha_{1j+}} \right) \alpha_{1jk} (\zeta_{1jk} - X_{1jk} \beta 55^{\wedge}) \right]$$

$$= \frac{\sigma_{1}^{2}}{h} \sum_{j} \sum_{k} \frac{1}{1 + \sigma_{2}^{2} \alpha_{1j+}} \alpha_{1jk} (\zeta_{1jk} - X_{1jk} \beta 55^{\wedge})$$

$$C - 45$$

$$= \frac{\sigma_{1}^{2}}{h} \sum_{j} \frac{\alpha_{1j+}}{1+\sigma_{2}^{2}\alpha_{1j+}} \sum_{k} \frac{\alpha_{1jk}}{\alpha_{1j+}} (\zeta_{1jk} X_{1jk} \beta 55^{\wedge})$$

$$= \frac{\sigma_{1}^{2}}{h} \sum_{j} \frac{\alpha_{1j+}}{1+\sigma_{2}^{2}\alpha_{1j+}} (\zeta_{1j} X_{1j} \beta 55^{\wedge})$$

$$= \frac{\sigma_{1}^{2}}{h} \left(\sum_{j} \frac{\alpha_{1j+}}{\sigma_{2}^{2}\alpha_{1j+}+1} \right) \left(\zeta 60 =_{1} - X 60 =_{1} \beta 55^{\wedge} \right)$$

$$= \frac{\sigma_{1}^{2} \left(\sum_{j} \frac{\alpha_{1j+}}{\sigma_{2}^{2}\alpha_{1j+}+1} \right)}{1+\sigma_{1}^{2} \sum_{j} \left[\sum_{l=j} \frac{\sigma_{lj+}^{2}}{\sigma_{2}^{2}\alpha_{lj+}+1} \right]} \left(\zeta 60 =_{1} - X 60 =_{1} \beta 55^{\wedge} \right)$$

$$= \frac{\sigma_{1}^{2}}{\sigma_{1}^{2} + \left(\sum_{j} \frac{\alpha_{1j+}^{2}}{\sigma_{2}^{2}\alpha_{1j+}+1} \right)^{-1}} \left(\zeta 60 =_{1} - X 60 =_{1} \beta 55^{\wedge} \right)$$

$$= \gamma_{11} \left(\zeta 60 =_{1} - X 60 =_{1} \beta 55^{\wedge} \right)$$

$$= \gamma_{11} \left(\zeta 60 =_{1} - X 60 =_{1} \beta 55^{\wedge} \right)$$

where

$$\gamma_{11} = \frac{\sigma_1^2}{\sigma_1^2 + \left(\frac{\sum_j \alpha_{1j+}}{\sigma_2^2 \alpha_{1j+} + 1}\right)^{-1}}$$

is the state-level shrinkage factor in the two-level model.

Looking at a PSU effect, we have that

Error!

Error!

where

$$\gamma_{21j} = \frac{\sigma_2^2}{\sigma_2^2 + \frac{1}{\alpha_{1j+}}}$$

is the PSU-level shrinkage factor for the j-th sample PSU in the first state in the 2-level model.

Similarly for the other states,

$$U_{i} = \gamma_{1i} (\zeta 60 =_{i} - X 60 =_{i} \beta 55^{\circ})$$

$$U_{ij} = \gamma_{2ij} \left[(\zeta_{ij} - X_{ij} \beta 55^{\circ}) - U_{i} \right],$$

where

$$\gamma_{1i} = \frac{\sigma_1^2}{\sigma_1^2 + \left(\sum_{j=1}^{r_i} \frac{\alpha_{ij+}}{\sigma_2^2 \alpha_{ij+} + 1}\right)^{-1}}$$

and

$$\gamma_{2ij} = \frac{\sigma_2^2}{\sigma_2^2 + \frac{1}{\alpha_{ii+}}} \quad .$$

To get a better intuitive feeling for these shrinkage factors, it is useful to introduce some additional notation. Let

$$\mathbf{v}_{ij} = Var \left[(\zeta_{ij} - X_{ij}\beta) \mid U_i \mid U_{ij} \right] .$$

This variance can be thought of as the measurement variance on the random perturbation for the j-th sample PSU in the i-th state given a particular manifestation from the superpopulation of state and PSU random effects. This variance may be derived as follows. Note that

$$\zeta_{ij} - X_{ij}\beta = U_i + U_{ij} + \sum_{k} \frac{\alpha_{ijk}}{\alpha_{ij+}} \frac{\varepsilon_{ijk}}{w_{ijk}}$$

Now, given U_i and U_{ij} , only ε_{ijk} is random. So

$$\begin{aligned} \mathbf{v}_{ij} &= \sum_{k} \left(\frac{\alpha_{ijk}}{\alpha_{ij+}}\right)^{2} \frac{Var(\varepsilon_{ijk} \mid U_{i}, U_{ij})}{w_{ijk}^{2}} \\ &= \sum_{k} \left(\frac{\alpha_{ijk}}{\alpha_{ij+}}\right)^{2} \frac{1}{w_{ijk}^{2} \pi_{ijk} (1 - \pi_{ijk})} \\ &= \sum_{k} \frac{\alpha_{ijk}}{(\alpha_{ij+})^{2}} \\ &= \frac{1}{\alpha_{ij+}} \ . \end{aligned}$$

Thus,

$$\gamma_{2ij} = \frac{\sigma_2^2}{\sigma_2^2 + V_{ii}} \quad .$$

Recall that σ_2^2 is the variance of the PSU perturbations on the logit scale. This is what was referred to in Chapters 1 and 3 as the process variation. So γ_{2ij} is the proportion of the total variance on the perturbation for the j-th PSU in the i-th state that is due to the process rather than to measurement error. This is closely analogous to the definition of the shrinkage factors for the linear mixed model discussed in Section C.4. If the measurement error in negligible, then there will be no shrinkage and so the estimated random effect for the PSU will be close to the design-based estimate for the PSU. If, on the other hand, the measurement error is very large, then there will be considerable shrinkage, meaning that the estimate for the PSU will be based largely on the fixed model and on the state perturbation.

Turning attention to the state-level shrinkage factors, note that similar algebra as used to derive v_{ij} can show that

$$Var\left[\left(\zeta_{ij}-X_{ij}\beta\right) \middle| U_{i}\right] = \sigma_{2}^{2} + v_{ij} .$$

Forming the harmonic mean of these conditional variances across the r_i sample PSUs within the i-th state yields

$$\mathbf{v}_{Hi} = \left[\frac{1}{r_i} \sum_{j=1}^{r_i} \frac{1}{\sigma_2^2 + \mathbf{v}_{ij}}\right]^{-1}$$
.

Dividing this average conditional variance by the number of sample PSUs in the states gives us a sort of measurement error on the random perturbation for the i-th state.

Now note that the state-level shrinkage factor can be rewritten as

$$\gamma_{1i} = \frac{\sigma_1^2}{\sigma_1^2 + \frac{v_{Hi}}{r_i}}$$

So it can be thought of as the proportion of the total variance on the random perturbation for the i-th state that is due to the process rather than to measurement error. Again, this is clearly analogous to the shrinkage factors for the linear mixed model. If measurement error is negligible (due to a large sample size), then there will be no shrinkage and so the estimated random effect for the state will be close to that estimated based on the design. If, on the other hand, measurement error is large (due to a small sample size), then there will be considerable shrinkage, even to the point of saying that projections for the state should be based almost entirely on the fixed model.

Appendix D: Block Group and Tract Level Variables Considered in Small Area Estimation Models

Table D.1 Block Group and Tract Level Variables Considered in Small Area Estimation Models

<u>Variable</u>	<u>Label</u>	<u>Variable</u>	<u>Label</u>
DUMMY	O: Intercept Term		
UCLASS9	T: Underclass Indicator	IBLK10 IBLK11	Black Interaction Of PDENLEV1 Black Interaction Of PDENLEV2 Black Interaction Of PDENLEV2
IFEM1	Female Interaction Of UCLASS9	IBLK12 IBLK13	Black Interaction Of PDENLEV3 Black Interaction Of PDENLEV4
IBLK1	Black Interaction Of UCLASS9	IHISP10 IHISP11	Hispanic Interaction Of PDENLEV1 Hispanic Interaction Of PDENLEV2
IHISP1	Hispanic Interaction Of UCLASS9	IHISP12 IHISP13	Hispanic Interaction Of PDENLEV3 Hispanic Interaction Of PDENLEV4
IOTH1	Other Interaction Of UCLASS9	IOTH10	Other Interaction Of PDENLEV1
PURBH	T: Percent Housing Units In Urban Areas	IOTH11 IOTH12	Other Interaction Of PDENLEV2 Other Interaction Of PDENLEV3
IFEM2	Female Interaction Of PURBH	IOTH13	Other Interaction Of PDENLEV4
IBLK2	Black Interaction Of PURBH	PHH1PLN PHH1PQU	T: Linear: Percent One Person Households T: Quadratic: Percent One Person Households
IHISP2	Hispanic Interaction Of PURBH	PHH1PCU PHH1PQR	T: Cubic: Percent One Person Households T: Quartic: Percent One Person Households
IOTH2	Other Interaction Of PURBH	IFEM15	
FEMALE	O: Female Indicator	IFEM16 IFEM17	Female Interaction Of PHH1PLN Female Interaction Of PHH1PQU Female Interaction Of PHH1PCU
FEMBLCK	O: Black Interaction Of FEMALE	IFEM18	Female Interaction Of PHH1PQR
FEMHISP	O: Hispanic Interaction Of FEMALE	IBLK15 IBLK16	Black Interaction Of PHH1PLN Black Interaction Of PHH1PQU
FEMOTHR	O: Other Interaction Of FEMALE	IBLK17 IBLK18	Black Interaction Of PHH1PCU Black Interaction Of PHH1PQR
RACEBLCK RACEHISP	O: Race/Black Indicator	IHISP15	Highania Interaction Of DIJIII DI N
RACEOTHR	O: Race/Hispanic Indicator O: Race/Other Indicator	IHISP16 IHISP17	Hispanic Interaction Of PHH1PLN Hispanic Interaction Of PHH1PQU Hispanic Interaction Of PHH1PCU
REGNOREA REGSOUTH	O: Northeast Region Indicator O: South Region Indicator	IHISP18	Hispanic Interaction Of PHH1PQR
REGWEST	O: West Region Indicator	IOTH15 IOTH16	Other Interaction Of PHH1PLN Other Interaction Of PHH1PQU
IFEM7	Female Interaction Of REGNOREA	IOTH17	Other Interaction Of PHH1PCU
IFEM8 IFEM9	Female Interaction Of REGSOUTH Female Interaction Of REGWEST	IOTH18	Other Interaction Of PHH1PQR
IBLK7	Black Interaction Of REGNOREA	POPRMLN POPRMQU	T: Linear: Average Persons Per Room T: Quadratic: Average Persons Per Room
IBLK8	Black Interaction Of REGSOUTH	POPRMCU	T: Cubic: Average Persons Per Room
IBLK9	Black Interaction Of REGWEST	POPRMQR	T: Quartic: Average Persons Per Room
IHISP7	Hispanic Interaction Of REGNOREA	IFEM20	Female Interaction Of POPRMLN
IHISP8 IHISP9	Hispanic Interaction Of REGSOUTH Hispanic Interaction Of REGWEST	IFEM21 IFEM22	Female Interaction Of POPRMQU Female Interaction Of POPRMCU
	•	IFEM23	Female Interaction Of POPRMQR
IOTH7	Other Interaction Of REGNOREA	IDI 1/20	DI LI L. C. OCDODDAMA
IOTH8 IOTH9	Other Interaction Of REGSOUTH Other Interaction Of REGWEST	IBLK20 IBLK21	Black Interaction Of POPRMLN Black Interaction Of POPRMQU
10111)	Other Interaction of REGWEST	IBLK22	Black Interaction Of POPRMCU
PDENLEV1	O: Large MSA	IBLK23	Black Interaction Of POPRMQR
PDENLEV2 PDENLEV3	O: Medium MSA O: Small MSA	IHISP20	Hispanic Interaction Of POPRMLN
PDENLEV4	O: NonMSA, Urban	IHISP21	Hispanic Interaction Of POPRMQU
IEEM10	Famala Internation Of DDENI EVI	IHISP22	Hispanic Interaction Of POPRMCU
IFEM10 IFEM11	Female Interaction Of PDENLEV1 Female Interaction Of PDENLEV2	IHISP23	Hispanic Interaction Of POPRMQR
IFEM12	Female Interaction Of PDENLEV3	IOTH20	Other Interaction Of POPRMLN
IFEM13 IOTH21	Female Interaction Of PDENLEV4 Other Interaction Of POPRMQU	IOTH23	Other Interaction Of POPRMQR
IOTH22	Other Interaction Of POPRMCU		

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

<u>Variable</u>	Label	<u>Variable</u>	<u>Label</u>
PAGE18LN	T: Linear: Percent Persons 0-18 Years	IDI V25	Digale Interaction Of DACE 241 N
PAGE18QU PAGE18CU	T: Quadratic: Percent Persons 0-18 Years T: Cubic: Percent Persons 0-18 Years	IBLK35 IBLK36	Black Interaction Of PAGE34LN Black Interaction Of PAGE34QU
PAGE18QR	T: Quartic: Percent Persons 0-18 Years	IBLK37	Black Interaction Of PAGE34QU
THOLTOOK	1. Quarte. Percent repond of 10 Tears	IBLK38	Black Interaction Of PAGE34QR
IFEM25	Female Interaction Of PAGE18LN		
IFEM26	Female Interaction Of PAGE18QU	IHISP35	Hispanic Interaction Of PAGE34LN
IFEM27	Female Interaction Of PAGE18CU	IHISP36	Hispanic Interaction Of PAGE34QU
IFEM28	Female Interaction Of PAGE18QR	IHISP37	Hispanic Interaction Of PAGE34CU
IBLK25	Black Interaction Of PAGE18LN	IHISP38	Hispanic Interaction Of PAGE34QR
IBLK26	Black Interaction Of PAGE18QU	IOTH35	Other Interaction Of PAGE34LN
IBLK27	Black Interaction Of PAGE18CU	IOTH36	Other Interaction Of PAGE34QU
IBLK28	Black Interaction Of PAGE18QR	IOTH37	Other Interaction Of PAGE34CU
		IOTH38	Other Interaction Of PAGE34QR
IHISP25	Hispanic Interaction Of PAGE18LN	DA CEAALNI	T. I. D. (D. 25.443)
IHISP26 IHISP27	Hispanic Interaction Of PAGE18QU	PAGE44LN	T: Linear: Percent Persons 35-44 Years
IHISP28	Hispanic Interaction Of PAGE18CU Hispanic Interaction Of PAGE18QR	PAGE44QU PAGE44CU	T: Quadratic: Percent Persons 35-44 Years T: Cubic: Percent Persons 35-44 Years
111131 20	Inspanie interaction of FAGE18QK	PAGE44QR	T: Ouartic: Percent Persons 35-44 Years
IOTH25	Other Interaction Of PAGE18LN	1110211211	1. Quanto Terebro Terebro 35 TT Temp
IOTH26	Other Interaction Of PAGE18QU	IFEM40	Female Interaction Of PAGE44LN
IOTH27	Other Interaction Of PAGE18CU	IFEM41	Female Interaction Of PAGE44QU
IOTH28	Other Interaction Of PAGE18QR	IFEM42	Female Interaction Of PAGE44CU
PAGE24LN	T: Linear: Percent Persons 19-24 Years	IFEM43	Female Interaction Of PAGE44QR
PAGE24QU	T: Ouadratic: Percent Persons 19-24 Years	IBLK40	Black Interaction Of PAGE44LN
PAGE24CU	T: Cubic: Percent Persons 19-24 Years	IBLK40	Black Interaction Of PAGE44QU
PAGE24QR	T: Quartic: Percent Persons 19-24 Years	IBLK42	Black Interaction Of PAGE44CU
		IBLK43	Black Interaction Of PAGE44QR
IFEM30	Female Interaction Of PAGE24LN	WWGD 40	THE STATE OF THE S
IFEM31	Female Interaction Of PAGE24QU	IHISP40	Hispanic Interaction Of PAGE44LN
IFEM32 IFEM33	Female Interaction Of PAGE24CU Female Interaction Of PAGE24QR	IHISP41 IHISP42	Hispanic Interaction Of PAGE44QU Hispanic Interaction Of PAGE44CU
II ENISS	remaie interaction of 1 AGE24QR	IHISP43	Hispanic Interaction Of PAGE44QR
IBLK30	Black Interaction Of PAGE24LN		
IBLK31	Black Interaction Of PAGE24QU	IOTH40	Other Interaction Of PAGE44LN
IBLK32	Black Interaction Of PAGE24CU	IOTH41	Other Interaction Of PAGE44QU
IBLK33	Black Interaction Of PAGE24QR	IOTH42	Other Interaction Of PAGE44CU
IHISP30	Hispanic Interaction Of PAGE24LN	IOTH43	Other Interaction Of PAGE44QR
IHISP31	Hispanic Interaction Of PAGE24EN Hispanic Interaction Of PAGE24QU	PAGE54LN	T: Linear: Percent Persons 45-54 Years
IHISP32	Hispanic Interaction Of PAGE24CU	PAGE54QU	T: Quadratic: Percent Persons 45-54 Years
IHISP33	Hispanic Interaction Of PAGE24QR	PAGE54CU	T: Cubic: Percent Persons 45-54 Years
	•	PAGE54QR	T: Quartic: Percent Persons 45-54 Years
IOTH30	Other Interaction Of PAGE24LN	TDD 645	E I I I I I I I I I I I I I I I I I I I
IOTH31	Other Interaction Of PAGE24QU	IFEM45	Female Interaction Of PAGE54LN
IOTH32 IOTH33	Other Interaction Of PAGE24CU Other Interaction Of PAGE24QR	IFEM46 IFEM47	Female Interaction Of PAGE54QU Female Interaction Of PAGE54CU
1011133	Other interaction of FAGE24QR	IFEM48	Female Interaction Of PAGE54QR
PAGE34LN	T: Linear: Percent Persons 25-34 Years		· ·
PAGE34QU	T: Quadratic: Percent Persons 25-34 Years	IBLK45	Black Interaction Of PAGE54LN
PAGE34CU	T: Cubic: Percent Persons 25-34 Years	IBLK46	Black Interaction Of PAGE54QU
PAGE34QR	T: Quartic: Percent Persons 25-34 Years	IBLK47	Black Interaction Of PAGE54CU
IFEM35	Female Interaction Of PAGE34LN	IBLK48	Black Interaction Of PAGE54QR
IFEM36	Female Interaction Of PAGE34QU	IHISP45	Hispanic Interaction Of PAGE54LN
IFEM37	Female Interaction Of PAGE34CU	IHISP46	Hispanic Interaction Of PAGE54QU
IFEM38	Female Interaction Of PAGE34QR	IHISP47	Hispanic Interaction Of PAGE54CU
IHISP48	Hispanic Interaction Of PAGE54QR	IOTH48	Other Interaction Of PAGE54QR
IOTH45	Other Interaction Of PAGE54LN	PAGE64LN	T: Linear: Percent Persons 55-64 Years
IOTH45 IOTH46	Other Interaction Of PAGE54LN Other Interaction Of PAGE54QU	PAGE64QU	T: Quadratic: Percent Persons 55-64 Years T: Ouadratic: Percent Persons 55-64 Years
IOTH47	Other Interaction Of PAGE54CU	PAGE64CU	T: Cubic: Percent Persons 55-64 Years
~			

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

Variable Label		<u>Variable</u>	<u>Label</u>
PAGE64QR	T: Quartic: Percent Persons 55-64 Years	IFEM63	Female Interaction Of PSCH12QR
IFEM50	Female Interaction Of PAGE64LN	IBLK60	Black Interaction Of PSCH12LN
IFEM51	Female Interaction Of PAGE64QU	IBLK61	Black Interaction Of PSCH12QU
IFEM52	Female Interaction Of PAGE64CU	IBLK62	Black Interaction Of PSCH12CU
IFEM53	Female Interaction Of PAGE64QR	IBLK63	Black Interaction Of PSCH12QR
IBLK50	Black Interaction Of PAGE64LN	IHISP60	Hispanic Interaction Of PSCH12LN
IBLK51	Black Interaction Of PAGE64QU	IHISP61	Hispanic Interaction Of PSCH12QU
IBLK52	Black Interaction Of PAGE64CU	IHISP62	Hispanic Interaction Of PSCH12CU
IBLK53	Black Interaction Of PAGE64QR	IHISP63	Hispanic Interaction Of PSCH12QR
IHISP50	Hispanic Interaction Of PAGE64LN	IOTH60	Other Interaction Of PSCH12LN Other Interaction Of PSCH12QU Other Interaction Of PSCH12CU Other Interaction Of PSCH12QR
IHISP51	Hispanic Interaction Of PAGE64QU	IOTH61	
IHISP52	Hispanic Interaction Of PAGE64CU	IOTH62	
IHISP53	Hispanic Interaction Of PAGE64QR	IOTH63	
IOTH50	Other Interaction Of PAGE64LN	PSCHASLN	T: Linear: Percent Associates Degree T: Quadratic: Percent Associates Degree T: Cubic: Percent Associates Degree T: Quartic: Percent Associates Degree
IOTH51	Other Interaction Of PAGE64QU	PSCHASQU	
IOTH52	Other Interaction Of PAGE64CU	PSCHASCU	
IOTH53	Other Interaction Of PAGE64QR	PSCHASQR	
PSCH8LN PSCH8QU PSCH8CU PSCH8QR	T: Linear: Percent 0-8 Years Of School T: Quadratic: Percent 0-8 Years Of School T: Cubic: Percent 0-8 Years Of School T: Quartic: Percent 0-8 Years Of School	IFEM65 IFEM66 IFEM67 IFEM68	Female Interaction Of PSCHASLN Female Interaction Of PSCHASQU Female Interaction Of PSCHASCU Female Interaction Of PSCHASQR
IFEM55	Female Interaction Of PSCH8LN Female Interaction Of PSCH8QU Female Interaction Of PSCH8CU Female Interaction Of PSCH8QR	IBLK65	Black Interaction Of PSCHASLN
IFEM56		IBLK66	Black Interaction Of PSCHASQU
IFEM57		IBLK67	Black Interaction Of PSCHASCU
IFEM58		IBLK68	Black Interaction Of PSCHASQR
IBLK55	Black Interaction Of PSCH8LN	IHISP65	Hispanic Interaction Of PSCHASLN
IBLK56	Black Interaction Of PSCH8QU	IHISP66	Hispanic Interaction Of PSCHASQU
IBLK57	Black Interaction Of PSCH8CU	IHISP67	Hispanic Interaction Of PSCHASCU
IBLK58	Black Interaction Of PSCH8QR	IHISP68	Hispanic Interaction Of PSCHASQR
IHISP55	Hispanic Interaction Of PSCH8LN	IOTH65	Other Interaction Of PSCHASLN Other Interaction Of PSCHASQU Other Interaction Of PSCHASCU Other Interaction Of PSCHASQR
IHISP56	Hispanic Interaction Of PSCH8QU	IOTH66	
IHISP57	Hispanic Interaction Of PSCH8CU	IOTH67	
IHISP58	Hispanic Interaction Of PSCH8QR	IOTH68	
IOTH55 IOTH56 IOTH57	Other Interaction Of PSCH8LN Other Interaction Of PSCH8QU Other Interaction Of PSCH8CU	PSCHCOLN PSCHCOQU Degree	T: Linear: Bachelors, Graduate, Or Professional Degree T: Quadratic: Bachelors, Graduate, Or Professional
IOTH58	Other Interaction Of PSCH8QR	PSCHCOCU PSCHCOQR	T: Cubic: Bachelors, Graduate, Or Professional Degree T: Quartic: Bachelors, Graduate, Or Professional Degree
PSCH12LN	T: Linear: Percent 9-12 Years & No HS Diploma	IFEM75	Female Interaction Of PSCHCOLN Female Interaction Of PSCHCOQU Female Interaction Of PSCHCOCU Female Interaction Of PSCHCOQR
PSCH12QU	T: Quadratic: Percent 9-12 Years & No HS Diploma	IFEM76	
PSCH12CU	T: Cubic: Percent 9-12 Years & No HS Diploma	IFEM77	
PSCH12QR	T: Quartic: Percent 9-12 Years & No HS Diploma	IFEM78	
IFEM60 IFEM61 IFEM62 IBLK78	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU Female Interaction Of PSCH12CU Black Interaction Of PSCHCOQR	IBLK75 IBLK76 IBLK77 IOTH76 IOTH77	Black Interaction Of PSCHCOLN Black Interaction Of PSCHCOQU Black Interaction Of PSCHCOCU Other Interaction Of PSCHCOQU Other Interaction Of PSCHCOQU
IHISP75 IHISP76 IHISP77	Hispanic Interaction Of PSCHCOLN Hispanic Interaction Of PSCHCOQU Hispanic Interaction Of PSCHCOCU	IOTH78 PSCHSCLN	Other Interaction Of PSCHCOQR T: Linear: Percent Some College And No Degree
IHISP78 IOTH75	Hispanic Interaction Of PSCHCOQR Other Interaction Of PSCHCOLN	PSCHSCQU PSCHSCCU PSCHSCQR	T: Quadratic: Percent Some College And No Degree T: Cubic: Percent Some College And No Degree T: Quartic: Percent Some College And No Degree

Table D.1 Block Group and Tract Level Variables Considered in Small Area Estimation Models (continued)

<u>Variable</u>	e <u>Label</u>	<u>Variable</u>	<u>Label</u>
		IFEM93	Female Interaction Of PPUBASQR
IFEM80	Female Interaction Of PSCHSCLN	IDI IZOO	DI III COCEDIUD (CINI
IFEM81	Female Interaction Of PSCHSCQU	IBLK90	Black Interaction Of PPUBASLN
IFEM82	Female Interaction Of PSCHSCCU	IBLK91 IBLK92	Black Interaction Of PPUBASQU
IFEM83	Female Interaction Of PSCHSCQR	IBLK92 IBLK93	Black Interaction Of PPUBASCU Black Interaction Of PPUBASQR
IBLK80	Black Interaction Of PSCHSCLN	IDLK93	Black Illicraction of 11 obasqk
IBLK81	Black Interaction Of PSCHSCQU	IHISP90	Hispanic Interaction Of PPUBASLN
IBLK82	Black Interaction Of PSCHSCCU	IHISP91	Hispanic Interaction Of PPUBASQU
IBLK83	Black Interaction Of PSCHSCQR	IHISP92	Hispanic Interaction Of PPUBASCU
		IHISP93	Hispanic Interaction Of PPUBASQR
IHISP80	Hispanic Interaction Of PSCHSCLN		
IHISP81	Hispanic Interaction Of PSCHSCQU	IOTH90	Other Interaction Of PPUBASLN
IHISP82	Hispanic Interaction Of PSCHSCCU	IOTH91	Other Interaction Of PPUBASQU
IHISP83	Hispanic Interaction Of PSCHSCQR	IOTH92 IOTH93	Other Interaction Of PPUBASCU
IOTH80	Other Interaction Of PSCHSCLN	1011193	Other Interaction Of PPUBASQR
IOTH81	Other Interaction Of PSCHSCQU	P64DISLN	T: Linear: Percent 16-64 With A Work Disability
IOTH82	Other Interaction Of PSCHSCCU	P64DISQU	T: Quadratic: Percent 16-64 With A Work Disability
IOTH83	Other Interaction Of PSCHSCQR	P64DISCU	T: Cubic: Percent 16-64 With A Work Disability
	· · · · · · · · · · · · · · · · · · ·	P64DISQR	T: Quartic: Percent 16-64 With A Work Disability
PPOVERLN	T: Linear: Percent Families Below Poverty Level		
PPOVERQU	T: Quadratic: Percent Families Below Poverty Level	IFEM95	Female Interaction Of P64DISLN
PPOVERCU	T: Cubic: Percent Families Below Poverty Level	IFEM96	Female Interaction Of P64DISQU
PPOVERQR	T: Quartic: Percent Families Below Poverty Level	IFEM97	Female Interaction Of P64DISCU
TEEN 60.5	E I I OCDDOVEDIN	IFEM98	Female Interaction Of P64DISQR
IFEM85	Female Interaction Of PPOVERLN	IDI 1/05	DII-I-I
IFEM86 IFEM87	Female Interaction Of PPOVERQU Female Interaction Of PPOVERCU	IBLK95 IBLK96	Black Interaction Of P64DISLN Black Interaction Of P64DISQU
IFEM88	Female Interaction Of PPOVERQR	IBLK90 IBLK97	Black Interaction Of P64DISCU
II LIVIOO	Temate interaction of FFOVERQR	IBLK98	Black Interaction Of P64DISQR
IBLK85	Black Interaction Of PPOVERLN	155150	Buth interestion of 10 (Bio Qit
IBLK86	Black Interaction Of PPOVERQU	IHISP95	Hispanic Interaction Of P64DISLN
IBLK87	Black Interaction Of PPOVERCU	IHISP96	Hispanic Interaction Of P64DISQU
IBLK88	Black Interaction Of PPOVERQR	IHISP97	Hispanic Interaction Of P64DISCU
HHADO 5	W	IHISP98	Hispanic Interaction Of P64DISQR
IHISP85	Hispanic Interaction Of PPOVERLN	IOTHO?	Od I do di OCDCADICI N
IHISP86	Hispanic Interaction Of PPOVERQU	IOTH95	Other Interaction Of P64DISLN Other Interaction Of P64DISOU
IHISP87 IHISP88	Hispanic Interaction Of PPOVERCU Hispanic Interaction Of PPOVERQR	IOTH96 IOTH97	Other Interaction Of P64DISQU Other Interaction Of P64DISCU
111131 00	Thispanic interaction of FFOVERQR	IOTH98	Other Interaction Of P64DISQR
IOTH85	Other Interaction Of PPOVERLN	1011170	Other interaction of roadisque
IOTH86	Other Interaction Of PPOVERQU	PBLACKLN	T: Linear: Percent Black Nonhispanic
IOTH87	Other Interaction Of PPOVERCU	PBLACKQU	T: Quadratic: Percent Black Nonhispanic
IOTH88	Other Interaction Of PPOVERQR	PBLACKCU	T: Cubic: Percent Black Nonhispanic
PPUBASLN	T: Linear: % HHS With Public Assist Income	PBLACKQR	T: Quartic: Percent Black Nonhispanic
PPUBASQU	T: Quadratic: % HHS With Public Assist Income	TET (100	E LA CORPO CONTA
PPUBASCU	T: Cubic: % HHS With Public Assist Income	IFEM100	Female Interaction Of PBLACKLN
PPUBASQR	T: Quartic: % HHS With Public Assist Income	IFEM101	Female Interaction Of PBLACKQU Female Interaction Of PBLACKCU
IFEM90	Female Interaction Of PPUBASLN	IFEM102 IFEM103	Female Interaction Of PBLACKQR
IFEM91	Female Interaction Of PPUBASQU	II DIVITOS	remaie interaction of i blackor
IFEM92	Female Interaction Of PPUBASCU	IBLK100	Black Interaction Of PBLACKLN
IBLK101	Black Interaction Of PBLACKOU	IOTH102	Other Interaction Of PBLACKCU
IBLK102	Black Interaction Of PBLACKCU	IOTH103	Other Interaction Of PBLACKQR
IBLK103	Black Interaction Of PBLACKQR		*
		PHISPLN	T: Linear: Percent Hispanic
IHISP100	Hispanic Interaction Of PBLACKLN	PHISPQU	T: Quadratic: Percent Hispanic
IHISP101	Hispanic Interaction Of PBLACKQU	PHISPCU	T: Cubic: Percent Hispanic
IHISP102	Hispanic Interaction Of PBLACKCU	PHISPQR	T: Quartic: Percent Hispanic
IHISP103	Hispanic Interaction Of PBLACKQR	IFEM105	Female Interaction Of PHISPLN
IOTH100	Other Interaction Of PBLACKLN	IFEM105 IFEM106	Female Interaction Of PHISPOU
IOTH101	Other Interaction of PBLACKQU	IFEM107	Female Interaction Of PHISPCU

Table D.1 Block Group and Tract Level Variables Considered in Small Area Estimation Models (continued)

<u>Variable</u>	<u>Label</u>	<u>Variable</u>	<u>Label</u>
IFEM108	Female Interaction Of PHISPQR	IFEM122 IFEM123	Female Interaction Of PHHF18CU Female Interaction Of PHHF18QR
IBLK105	Black Interaction Of PHISPLN		
IBLK106	Black Interaction Of PHISPQU	IBLK120	Black Interaction Of PHHF18LN
IBLK107 IBLK108	Black Interaction Of PHISPCU Black Interaction Of PHISPQR	IBLK121 IBLK122	Black Interaction Of PHHF18QU Black Interaction Of PHHF18CU
IDEICIOO	Black interaction of Fried Qit	IBLK123	Black Interaction Of PHHF18QR
IHISP105	Hispanic Interaction Of PHISPLN	*******	
IHISP106	Hispanic Interaction Of PHISPQU Hispanic Interaction Of PHISPCU	IHISP120	Hispanic Interaction Of PHHF18LN
IHISP107 IHISP108	Hispanic Interaction of PHISPCO	IHISP121 IHISP122	Hispanic Interaction Of PHHF18QU Hispanic Interaction Of PHHF18CU
111101 100	•	IHISP123	Hispanic Interaction Of PHHF18QR
IOTH105	Other Interaction Of PHISPLN	10771100	O.L. A OCENTRICALLY
IOTH106 IOTH107	Other Interaction Of PHISPQU Other Interaction Of PHISPCU	IOTH120 IOTH121	Other Interaction Of PHHF18LN Other Interaction Of PHHF18QU
IOTH107	Other Interaction Of PHISPQR	IOTH121 IOTH122	Other Interaction of PHHF18CU
10111100		IOTH123	Other Interaction Of PHHF18QR
POTHLN	T: Linear: Percent Other Race/Hispanicity		
POTHQU POTHCU	T: Quadratic: Percent Other Race/Hispanicity	PFLABLN	T: Linear: % Females 16+ Years Old In Labor Force
POTHOU	T: Cubic: Percent Other Race/Hispanicity T: Quartic: Percent Other Race/Hispanicity	PFLABQU PFLABCU	T: Quadratic: % Females 16+ Years Old In Labor Force T: Cubic: % Females 16+ Years Old In Labor Force
roman	1. Quarte. Percent other ruce, mapanierty	PFLABQR	T: Quartic: % Females 16+ Years Old In Labor Force
IFEM110	Female Interaction Of POTHLN		
IFEM111	Female Interaction Of POTHQU	IFEM125	Female Interaction Of PFLABLN
IFEM112 IFEM113	Female Interaction Of POTHCU Female Interaction Of POTHQR	IFEM126 IFEM127	Female Interaction Of PFLABQU Female Interaction Of PFLABCU
II LIVII I 3	Temale interaction of Format	IFEM128	Female Interaction Of PFLABQR
IBLK110	Black Interaction Of POTHLN		
IBLK111	Black Interaction Of POTHQU	IBLK125	Black Interaction Of PFLABLN
IBLK112 IBLK113	Black Interaction Of POTHCU Black Interaction Of POTHQR	IBLK126 IBLK127	Black Interaction Of PFLABQU Black Interaction Of PFLABCU
IDEKITS	Black interaction of Format	IBLK128	Black Interaction Of PFLABQR
IHISP110	Hispanic Interaction Of POTHLN	W.W.C.D.1.0.5	TY CONTRACTOR OF THE CONTRACTO
IHISP111 IHISP112	Hispanic Interaction Of POTHQU Hispanic Interaction Of POTHCU	IHISP125 IHISP126	Hispanic Interaction Of PFLABLN Hispanic Interaction Of PFLABQU
IHISP113	Hispanic Interaction Of POTHQR	IHISP127	Hispanic Interaction Of PFLABCU
		IHISP128	Hispanic Interaction Of PFLABQR
IOTH110	Other Interaction Of POTHLN	IOTILIA5	Od I a CORPE ADIN
IOTH111 IOTH112	Other Interaction Of POTHQU Other Interaction Of POTHCU	IOTH125 IOTH126	Other Interaction Of PFLABLN Other Interaction Of PFLABQU
IOTH113	Other Interaction Of POTHQR	IOTH127	Other Interaction Of PFLABCU
		IOTH128	Other Interaction Of PFLABQR
DITUETOLN	T. I : 0/ E H1-1 IIII W/NI- C 9- Child <10	DENIEWI NI	T: Linear: Percent Females Never Married
PHHF18LN PHHF18QU	T: Linear: % F-Headed HH W/No Spouse & Chld <18 T: Quad.: % F-Headed HH W/No Spouse & Chld <18	PFNEVLN PFNEVQU	T: Quadratic: Percent Females Never Married
PHHF18CU	T: Cubic: % F-Headed HH W/No Spouse & Chld <18	PFNEVCU	T: Cubic: Percent Females Never Married
PHHF18QR	T: Quartic: % F-Headed HH W/No Spouse & Chld <18	PFNEVQR	T: Quartic: Percent Females Never Married
IFEM120	Female Interaction Of PHHF18LN	IFEM130	Female Interaction Of PFNEVLN
IFEM121	Female Interaction Of PHHF18QU	IFEM131	Female Interaction Of PFNEVQU
IFEM132	Female Interaction Of PFNEVCU	IOTH131	Other Interaction Of PFNEVQU
IFEM133	Female Interaction Of PFNEVQR	IOTH132	Other Interaction Of PFNEVCU
IBLK130	Black Interaction Of PFNEVLN	IOTH133	Other Interaction Of PFNEVQR
IBLK131	Black Interaction Of PFNEVQU	PFNOTLN	T: Linear: % Females Separated, Divorced Or Widowed
IBLK132	Black Interaction Of PFNEVČU	PFNOTQU	T: Quadratic: % Fem Separated, Divorced Or Widowed
IBLK133	Black Interaction Of PFNEVQR	PFNOTCU PFNOTOR	T: Cubic: % Fem Separated, Divorced Or Widowed T: Quartic: % Fem Separated, Divorced Or Widowed
IHISP130	Hispanic Interaction Of PFNEVLN	rrnotyk	1. Quartic. 76 reiii Separated, Divorced Or Widowed
IHISP131	Hispanic Interaction Of PFNEVQU	IFEM135	Female Interaction Of PFNOTLN
IHISP132	Hispanic Interaction Of PFNEVCU	IFEM136	Female Interaction Of PFNOTQU
IHISP133	Hispanic Interaction Of PFNEVQR	IFEM137 IFEM138	Female Interaction Of PFNOTCU Female Interaction Of PFNOTOR
IOTH130	Other Interaction Of PFNEVLN	11 LIVI1 30	1 chiale interaction of 111101QR

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

<u>Variable</u>	<u>Label</u>	<u>Variable</u>	<u>Label</u>
IBLK135 IBLK136 IBLK137	Black Interaction Of PFNOTLN Black Interaction Of PFNOTQU Black Interaction Of PFNOTCU	IFEM146 IFEM147 IFEM148	Female Interaction Of PMLABQU Female Interaction Of PMLABCU Female Interaction Of PMLABQR
IBLK138 IHISP135 IHISP136 IHISP137	Black Interaction Of PFNOTQR Hispanic Interaction Of PFNOTLN Hispanic Interaction Of PFNOTQU Hispanic Interaction Of PFNOTCU	IBLK145 IBLK146 IBLK147 IBLK148	Black Interaction Of PMLABLN Black Interaction Of PMLABQU Black Interaction Of PMLABCU Black Interaction Of PMLABQR
IHISP138 IOTH135 IOTH136 IOTH137	Hispanic Interaction Of PFNOTQR Other Interaction Of PFNOTLN Other Interaction Of PFNOTQU Other Interaction Of PFNOTCU	IHISP145 IHISP146 IHISP147 IHISP148	Hispanic Interaction Of PMLABLN Hispanic Interaction Of PMLABQU Hispanic Interaction Of PMLABCU Hispanic Interaction Of PMLABQR
IOTH138 PMNEVLN PMNEVQU PMNEVCU	Other Interaction Of PFNOTQR T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married T: Cubic: Percent Males Never Married	IOTH145 IOTH146 IOTH147 IOTH148	Other Interaction Of PMLABLN Other Interaction Of PMLABQU Other Interaction Of PMLABCU Other Interaction Of PMLABQR
PMNEVQR IFEM140 IFEM141 IFEM142	T: Quartic: Percent Males Never Married Female Interaction Of PMNEVLN Female Interaction Of PMNEVQU Female Interaction Of PMNEVCU	PMNOTLN PMNOTQU PMNOTCU PMNOTQR	T: Linear: % M Separated, Divorced Or Widowed T: Quadratic: % M Separated, Divorced Or Widowed T: Cubic: % M Separated, Divorced Or Widowed T: Quartic: % M Separated, Divorced Or Widowed
IFEM143 IBLK140 IBLK141 IBLK142	Female Interaction Of PMNEVQR Black Interaction Of PMNEVLN Black Interaction Of PMNEVQU Black Interaction Of PMNEVQU	IFEM150 IFEM151 IFEM152 IFEM153	Female Interaction Of PMNOTLN Female Interaction Of PMNOTQU Female Interaction Of PMNOTCU Female Interaction Of PMNOTQR
IBLK143 IHISP140 IHISP141 IHISP142	Black Interaction Of PMNEVQR Hispanic Interaction Of PMNEVLN Hispanic Interaction Of PMNEVQU Hispanic Interaction Of PMNEVQU	IBLK150 IBLK151 IBLK152 IBLK153	Black Interaction Of PMNOTLN Black Interaction Of PMNOTQU Black Interaction Of PMNOTCU Black Interaction Of PMNOTQR
IHISP143 IOTH140 IOTH141 IOTH142	Hispanic Interaction Of PMNEVQR Other Interaction Of PMNEVLN Other Interaction Of PMNEVQU	IHISP150 IHISP151 IHISP152 IHISP153	Hispanic Interaction Of PMNOTLN Hispanic Interaction Of PMNOTQU Hispanic Interaction Of PMNOTCU
IOTH142 IOTH143 PMLABLN	Other Interaction Of PMNEVCU Other Interaction Of PMNEVQR T: Linear: % Males 16+ Years Old In Labor Force	IOTH150 IOTH151 IOTH152	Hispanic Interaction Of PMNOTQR Other Interaction Of PMNOTLN Other Interaction Of PMNOTQU Other Interaction Of PMNOTCU
PMLABQU PMLABCU PMLABQR IFEM145	T: Quadratic: % Males 16+ Years Old In Labor Force T: Cubic: % Males 16+ Years Old In Labor Force T: Quartic: % Males 16+ Years Old In Labor Force Expedicular Interaction Of PMI APLN	POLDHULN POLDHUQU POLDHUCU	Other Interaction Of PMNOTQR T: Linear: % Housing Units Built 1939 Or Earlier T: Quadratic: % Housing Units Built 1939 Or Earlier T: Cubic: % Housing Units Built 1939 Or Earlier
POLDHUQR IFEM155 IFEM156	Female Interaction Of PMLABLN T: Quartic: % Housing Units Built 1939 Or Earlier Female Interaction Of POLDHULN Female Interaction Of POLDHUQU	POLDHUCU IOTH155 IOTH156 IOTH157 IOTH158	Other Interaction Of POLDHULN Other Interaction Of POLDHUQU Other Interaction Of POLDHUCU Other Interaction Of POLDHUCU
IFEM157 IFEM158 IBLK155	Female Interaction Of POLDHUCU Female Interaction Of POLDHUQR Black Interaction Of POLDHULN	P40HULN P40HUQU P40HUCU	T: Linear: Percent Housing Units Built 1940-1949 T: Quadratic: Percent Housing Units Built 1940-1949 T: Cubic: Percent Housing Units Built 1940-1949
IBLK156 IBLK157 IBLK158 IHISP155	Black Interaction Of POLDHUQU Black Interaction Of POLDHUCU Black Interaction Of POLDHUQR Hispanic Interaction Of POLDHULN	P40HUQR IFEM160 IFEM161 IFEM162	T: Quartic: Percent Housing Units Built 1940-1949 Female Interaction Of P40HULN Female Interaction Of P40HUQU Female Interaction Of P40HUCU
IHISP156 IHISP157 IHISP158	Hispanic Interaction Of POLDHUQU Hispanic Interaction Of POLDHUCU Hispanic Interaction Of POLDHUQR	IFEM163 IBLK160 IBLK161	Female Interaction Of P40HUQR Black Interaction Of P40HULN Black Interaction Of P40HUQU

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

Variable	<u>Label</u>	<u>Variable</u>	<u>Label</u>
IBLK162 IBLK163	Black Interaction Of P40HUCU Black Interaction Of P40HUQR	IFEM171 IFEM172 IFEM173	Female Interaction Of PURBPQU Female Interaction Of PURBPCU Female Interaction Of PURBPQR
IHISP160 IHISP161 IHISP162 IHISP163	Hispanic Interaction Of P40HULN Hispanic Interaction Of P40HUQU Hispanic Interaction Of P40HUCU Hispanic Interaction Of P40HUQR	IBLK170 IBLK171 IBLK172 IBLK173	Black Interaction Of PURBPLN Black Interaction Of PURBPQU Black Interaction Of PURBPCU Black Interaction Of PURBPQR
IOTH160 IOTH161 IOTH162 IOTH163	Other Interaction Of P40HULN Other Interaction Of P40HUQU Other Interaction Of P40HUCU Other Interaction Of P40HUQR	IHISP170 IHISP171 IHISP172 IHISP173	Hispanic Interaction Of PURBPLN Hispanic Interaction Of PURBPQU Hispanic Interaction Of PURBPCU Hispanic Interaction Of PURBPQR
PRENTLN PRENTQU PRENTCU PRENTQR	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented T: Cubic: Percent Housing Rented T: Quartic: Percent Housing Rented	IOTH170 IOTH171 IOTH172 IOTH173	Other Interaction Of PURBPLN Other Interaction Of PURBPQU Other Interaction Of PURBPCU Other Interaction Of PURBPQR
IFEM165 IFEM166 IFEM167 IFEM168	Female Interaction Of PRENTLN Female Interaction Of PRENTQU Female Interaction Of PRENTCU Female Interaction Of PRENTQR	ADRATELN ADRATEQU ADRATECU ADRATEQR	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases C: Cubic: Death Rate For All Alcohol-Related Cases C: Quartic: Death Rate For All Alcohol-Related Cases
IBLK165 IBLK166 IBLK167 IBLK168 IHISP165 IHISP166	Black Interaction Of PRENTLN Black Interaction Of PRENTQU Black Interaction Of PRENTCU Black Interaction Of PRENTQR Hispanic Interaction Of PRENTLN Hispanic Interaction Of PRENTQU	IFEM175 IFEM176 IFEM177 IFEM178	Female Interaction Of ADRATELN Female Interaction Of ADRATEQU Female Interaction Of ADRATECU Female Interaction Of ADRATEQR
IHISP167 IHISP168 IOTH165	Hispanic Interaction Of PRENTCU Hispanic Interaction Of PRENTQR Other Interaction Of PRENTLN	IBLK175 IBLK176 IBLK177 IBLK178	Black Interaction Of ADRATELN Black Interaction Of ADRATEQU Black Interaction Of ADRATECU Black Interaction Of ADRATEQR
IOTH166 IOTH167 IOTH168	Other Interaction Of PRENTQU Other Interaction Of PRENTCU Other Interaction Of PRENTQR	IHISP175 IHISP176 IHISP177 IHISP178	Hispanic Interaction Of ADRATELN Hispanic Interaction Of ADRATEQU Hispanic Interaction Of ADRATECU Hispanic Interaction Of ADRATEQR
PURBPLN PURBPQU PURBPCU PURBPQR	T: Linear: Percent Population In Urban Areas T: Quadratic: Percent Population In Urban Areas T: Cubic: Percent Population In Urban Areas T: Quartic: Percent Population In Urban Areas	IOTH175 IOTH176 IOTH177	Other Interaction Of ADRATELN Other Interaction Of ADRATEQU Other Interaction Of ADRATECU
IFEM170 ADRATILN ADRATIQU ADRATICU ADRATIQR	Female Interaction Of PURBPLN C: Linr: Death Rate With Explicit Mention Of Alcohol C: Quad: Death Rate With Explicit Mention Of Alcohol C: Cubic: Death Rate With Explicit Mention Of Alcohol C: Quart: Death Rate With Explicit Mention Of Alcohol	IOTH178 IOTH180 IOTH181 IOTH182 IOTH183	Other Interaction Of ADRATEQR Other Interaction Of ADRAT1LN Other Interaction Of ADRAT1QU Other Interaction Of ADRAT1CU Other Interaction Of ADRAT1QR
IFEM180 IFEM181 IFEM182 IFEM183	Female Interaction Of ADRATILN Female Interaction Of ADRATIQU Female Interaction Of ADRATICU Female Interaction Of ADRATIQR	V18FLN V18FQU V18FCU V18FQR	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate C: Cubic: Marijuana Posession Arrest Rate C: Quartic: Marijuana Posession Arrest Rate
IBLK180 IBLK181 IBLK182 IBLK183	Black Interaction Of ADRAT1LN Black Interaction Of ADRAT1QU Black Interaction Of ADRAT1CU Black Interaction Of ADRAT1QR	IFEM185 IFEM186 IFEM187 IFEM188	Female Interaction Of V18FLN Female Interaction Of V18FQU Female Interaction Of V18FCU Female Interaction Of V18FQR
IHISP180 IHISP181 IHISP182 IHISP183	Hispanic Interaction Of ADRAT1LN Hispanic Interaction Of ADRAT1QU Hispanic Interaction Of ADRAT1CU Hispanic Interaction Of ADRAT1QR	IBLK185 IBLK186 IBLK187 IBLK188	Black Interaction Of V18FLN Black Interaction Of V18FQU Black Interaction Of V18FCU Black Interaction Of V18FQR

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

<u>Variable</u>	<u>Label</u>	<u>Variable</u>	<u>Label</u>
IHISP185	Hispanic Interaction Of V18FLN		
IHISP186	Hispanic Interaction Of V18FQU	IFEM200	Female Interaction Of V18ALN
IHISP187	Hispanic Interaction Of V18FCU	IFEM201	Female Interaction Of V18AQU
IHISP188	Hispanic Interaction Of V18FQR	IFEM202 IFEM203	Female Interaction Of V18ACU Female Interaction Of V18AOR
IOTH185	Other Interaction Of V18FLN		•
IOTH186	Other Interaction Of V18FQU	IBLK200	Black Interaction Of V18ALN
IOTH187	Other Interaction Of V18FCU	IBLK201 IBLK202	Black Interaction Of V18AQU Black Interaction Of V18ACU
IOTH188	Other Interaction Of V18FQR	IBLK202 IBLK203	Black Interaction Of V18ACO Black Interaction Of V18AQR
V18BLN	C: Linear: Marijuana Sale/Manufacture Arrest Rate		
V18BQU	C: Quadratic: Marijuana Sale/Manufacture Arrest Rate	IHISP200	Hispanic Interaction Of V18ALN
V18BCU	C: Cubic: Marijuana Sale/Manufacture Arrest Rate	IHISP201	Hispanic Interaction Of V18AQU
V18BQR	C: Quartic: Marijuana Sale/Manufacture Arrest Rate	IHISP202 IHISP203	Hispanic Interaction Of V18ACU Hispanic Interaction Of V18AQR
IFEM190	Female Interaction Of V18BLN	111131 203	mispanic interaction of viologic
IFEM191	Female Interaction Of V18BQU	IOTH200	Other Interaction Of V18ALN
IFEM192	Female Interaction Of V18BCU	IOTH201	Other Interaction Of V18AQU
IFEM193	Female Interaction Of V18BQR	IOTH202	Other Interaction Of V18ACU
IBLK190	Black Interaction Of V18BLN	IOTH203	Other Interaction Of V18AQR
IBLK191	Black Interaction Of V18BQU	V18HLN	C: Linear: Other (Non-Narcotics) Posession Arrest Rate
IBLK192	Black Interaction Of V18BCU	V18HQU	C: Quad: Other (Non-Narcotics) Posession Arst Rate
IBLK193	Black Interaction Of V18BQR	V18HCU	C: Cubic: Other (Non-Narcotics) Posession Arrest Rate
HHCD100	H I OCMIODINI	V18HQR	C: Quartic: Other (Non-Narcotics) Posession Arrest Rate
IHISP190 IHISP191	Hispanic Interaction Of V18BLN Hispanic Interaction Of V18BOU	IFEM205	Female Interaction Of V18HLN
IHISP192	Hispanic Interaction Of V18BCU	IFEM206	Female Interaction Of V18HQU
IHISP193	Hispanic Interaction Of V18BQR	IFEM207	Female Interaction Of V18HCU
	•	IFEM208	Female Interaction Of V18HQR
IOTH190	Other Interaction Of V18BLN	IDI 1/205	DI LI I COMININA
IOTH191 IOTH192	Other Interaction Of V18BQU Other Interaction Of V18BCU	IBLK205 IBLK206	Black Interaction Of V18HLN Black Interaction Of V18HQU
IOTH193	Other Interaction Of V18BQR	IBLK207	Black Interaction Of V18HCU
		IBLK208	Black Interaction Of V18HQR
		IHISP205	Hispanic Interaction Of V18HLN
V18ALN	C: Linr: Opium/Cocaine & Deriv Sale/Manuf Arst Rate	IHISP206	Hispanic Interaction Of V18HQU
V18AQU	C: Quad: Opium/Cocaine & Deriv Sale/Manuf Arst Rate	IHISP207	Hispanic Interaction Of V18HCU
V18AČU	C: Cub: Opium/Cocaine & Deriv Sale/Manuf Arst Rate	IHISP208	Hispanic Interaction Of V18HQR
V18AQR	C: Qurt: Opium/Cocaine & Deriv Sale/Manuf Arst Rate	********	
IOTH205	Other Interaction Of V18HLN	IHISP213	Hispanic Interaction Of V18DQR
IOTH206 IOTH207	Other Interaction Of V18HQU Other Interaction Of V18HCU	IOTH210	Other Interaction Of V18DLN
IOTH208	Other Interaction Of V18HQR	IOTH211	Other Interaction Of V18DQU
		IOTH212	Other Interaction Of V18DCU
V18DLN	C: Linear: Other (Non-Narcotics) Sale/Manuf Arst Rate	IOTH213	Other Interaction Of V18DQR
V18DQU	C: Quad: Other (Non-Narcotics) Sale/Manuf Arst Rate	VIOCIN	C. Lingary Symthetic Narrating Decognism Arrest Data
V18DCU V18DQR	C: Cubic: Other (Non-Narcotics) Sale/Manuf Arst Rate C: Quartic: Other (Non-Narcotics) Sale/Manuf Arst Rate	V18GLN V18GQU	C: Linear: Synthetic Narcotics Posession Arrest Rate C: Quadratic: Synthetic Narcotics Posession Arrest Rate
VIODQIC	C. Quartie. Other (1701) Praireotics) Suite/Manual First Rate	V18GCU	C: Cubic: Synthetic Narcotics Posession Arrest Rate
IFEM210	Female Interaction Of V18DLN	V18GQR	C: Quartic: Synthetic Narcotics Posession Arrest Rate
IFEM211	Female Interaction Of V18DQU	TETE 40.1.5	To the standard of the standar
IFEM212	Female Interaction Of V18DCU Female Interaction Of V18DOR	IFEM215 IFEM216	Female Interaction Of V18GLN
IFEM213	remaie interaction of viabor	IFEM217	Female Interaction Of V18GQU Female Interaction Of V18GCU
IBLK210	Black Interaction Of V18DLN	IFEM218	Female Interaction Of V18GQR
IBLK211	Black Interaction Of V18DQU		•
IBLK212	Black Interaction Of V18DCU	IBLK215	Black Interaction Of V18GLN
IBLK213	Black Interaction Of V18DQR	IBLK216	Black Interaction Of V18GQU Black Interaction Of V18GCU
IHISP210	Hispanic Interaction Of V18DLN	IBLK217 IBLK218	Black Interaction Of V18GCU Black Interaction Of V18GOR
IHISP211	Hispanic Interaction Of V18DQU	IDLIXATO	Buck interaction of Froque
IHISP212	Hispanic Interaction Of V18DCU	IHISP215	Hispanic Interaction Of V18GLN

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

<u>Variable</u>	<u>Label</u>	<u>Variable</u>	<u>Label</u>
IHISP216	Hispanic Interaction Of V18GQU		
IHISP217	Hispanic Interaction Of V18GCU	IFEM225	Female Interaction Of V18LN
IHISP218	Hispanic Interaction Of V18GQR	IFEM226	Female Interaction Of V18QU
IOTH215	Other Interaction Of V18GLN	IFEM227 IFEM228	Female Interaction Of V18CU Female Interaction Of V18OR
IOTH216	Other Interaction Of V18GQU	11 2111220	Temate interaction of Violat
IOTH217	Other Interaction Of V18GCU	IBLK225	Black Interaction Of V18LN
IOTH218	Other Interaction Of V18GQR	IBLK226	Black Interaction Of V18QU
V18CLN	C: Linear: Synthetic Narcotics Sale/Manuf Arrest Rate	IBLK227 IBLK228	Black Interaction Of V18CU Black Interaction Of V18QR
V18CQU	C: Quadratic: Synthetic Narcotics Sale/Manuf Arrest Rate	15211220	Short mitiation of Frogra
V18CCU	C: Cubic: Synthetic Narcotics Sale/Manuf Arrest Rate	IHISP225	Hispanic Interaction Of V18LN
V18CQR	C: Quartic: Synthetic Narcotics Sale/Manuf Arrest Rate	IHISP226 IHISP227	Hispanic Interaction Of V18QU Hispanic Interaction Of V18CU
IFEM220	Female Interaction Of V18CLN	IHISP228	Hispanic Interaction Of V18QR
IFEM221	Female Interaction Of V18CQU		
IFEM222	Female Interaction Of V18CCU	IOTH225	Other Interaction Of V18LN
IFEM223	Female Interaction Of V18CQR	IOTH226 IOTH227	Other Interaction Of V18QU Other Interaction Of V18CU
IBLK220	Black Interaction Of V18CLN	IOTH228	Other Interaction Of V18CO Other Interaction Of V18QR
IBLK221	Black Interaction Of V18CQU		-
IBLK222	Black Interaction Of V18CCU	VIOLLN	C: Linear: Total Violent Offenses Arrest Rate
IBLK223	Black Interaction Of V18CQR	VIOLQU VIOLCU	C: Quadratic: Total Violent Offenses Arrest Rate C: Cubic: Total Violent Offenses Arrest Rate
IHISP220	Hispanic Interaction Of V18CLN	VIOLOR	C: Quartic: Total Violent Offenses Arrest Rate
IHISP221	Hispanic Interaction Of V18CQU	-	·
IHISP222	Hispanic Interaction Of V18CCU	IFEM230	Female Interaction Of VIOLLN
IHISP223	Hispanic Interaction Of V18CQR	IFEM231 IFEM232	Female Interaction Of VIOLQU Female Interaction Of VIOLCU
IOTH220	Other Interaction Of V18CLN	IFEM233	Female Interaction Of VIOLQR
IOTH221	Other Interaction Of V18CQU	****	
IOTH222 IOTH223	Other Interaction Of V18CCU	IBLK230	Black Interaction Of VIOLEN
10111223	Other Interaction Of V18CQR	IBLK231 IBLK232	Black Interaction Of VIOLQU Black Interaction Of VIOLCU
V18LN	C: Linear: Total Drug Abuse Violations Arrest Rate	IBLK233	Black Interaction Of VIOLQR
V18QU	C: Quadratic: Total Drug Abuse Violations Arrest Rate	HHCD220	H. C. I C. COMOLIN
V18CU V18QR	C: Cubic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate	IHISP230	Hispanic Interaction Of VIOLLN
IHISP231	Hispanic Interaction Of VIOLQU	IHISP237	Hispanic Interaction Of DRATECU
IHISP232	Hispanic Interaction Of VIOLCU	IHISP238	Hispanic Interaction Of DRATEQR
IHISP233	Hispanic Interaction Of VIOLQR	IOTH235	Other Interaction Of DRATELN
IOTH230	Other Interaction Of VIOLLN	IOTH236	Other Interaction Of DRATEQU
IOTH231	Other Interaction Of VIOLQU	IOTH237	Other Interaction Of DRATECU
IOTH232	Other Interaction Of VIOLCU	IOTH238	Other Interaction Of DRATEQR
IOTH233	Other Interaction Of VIOLQR	RH43ALN	T: Linear: Recoded Median Rents For Rental Units
DRATELN	C: Linear: Mean Drug Client Treatment Rate 1991 & 92	RH43AQU	T: Quadratic: Recoded Median Rents For Rental Units
DRATEQU	C: Quad: Mean Drug Client Treatment Rate 1991 & 92	RH43ACU	T: Cubic: Recoded Median Rents For Rental Units
DRATECU DRATEOR	C: Cubic: Mean Drug Client Treatment Rate 1991 & 92 C: Quart: Mean Drug Client Treatment Rate 1991 & 92	RH43AQR	T: Quartic: Recoded Median Rents For Rental Units
DRATEQR	C. Quart. Weath Drug Chefit Treatment Rate 1991 & 92	IFEM240	Female Interaction Of RH43ALN
IFEM235	Female Interaction Of DRATELN	IFEM241	Female Interaction Of RH43AQU
IFEM236	Female Interaction Of DRATEQU	IFEM242	Female Interaction Of RH43ACU
IFEM237 IFEM238	Female Interaction Of DRATECU Female Interaction Of DRATEOR	IFEM243	Female Interaction Of RH43AQR
	· ·	IBLK240	Black Interaction Of RH43ALN
IBLK235	Black Interaction Of DRATELN	IBLK241	Black Interaction Of RH43AQU
IBLK236 IBLK237	Black Interaction Of DRATEQU Black Interaction Of DRATECU	IBLK242 IBLK243	Black Interaction Of RH43ACU Black Interaction Of RH43AQR
IBLK237 IBLK238	Black Interaction Of DRATECO	IDLIX2†J	Diack interaction of KITTJAQK
	·	IHISP240	Hispanic Interaction Of RH43ALN
IHISP235 IHISP236	Hispanic Interaction Of DRATELN Hispanic Interaction Of DRATEQU	IHISP241 IHISP242	Hispanic Interaction Of RH43AQU Hispanic Interaction Of RH43ACU
11113F 230	Thispanic interaction of DRATEQU	111131242	mspanic interaction of KH45ACU

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

<u>Variable</u>	<u>Label</u>	<u>Variable</u>	<u>Label</u>
IHISP243	Hispanic Interaction Of RH43AQR	RP80AQR	T: Quartic: Recoded Median Household Income
IOTH240	Other Interaction Of RH43ALN	IFEM250	Female Interaction Of RP80ALN
IOTH241	Other Interaction Of RH43AQU	IFEM251	Female Interaction Of RP80AQU
IOTH242	Other Interaction Of RH43ACU	IFEM252	Female Interaction Of RP80ACU
IOTH243	Other Interaction Of RH43AQR	IFEM253	Female Interaction Of RP80AQR
RH61ALN	T: Linear: Recod Median Value Of Owner Occup HUs	IBLK250	Black Interaction Of RP80ALN
RH61AQU	T: Quad: Recod Median Value Of Owner Occup HUs	IBLK251	Black Interaction Of RP80AQU
RH61ACU	T: Cubic: Recod Median Value Of Owner Occup HUs	IBLK252	Black Interaction Of RP80ACU
RH61AQR	T: Quartic: Recod Median Value Of Owner Occup HUs	IBLK253	Black Interaction Of RP80AQR
IFEM245	Female Interaction Of RH61ALN	IHISP250	Hispanic Interaction Of RP80ALN
IFEM246	Female Interaction Of RH61AQU	IHISP251	Hispanic Interaction Of RP80AQU
IFEM247	Female Interaction Of RH61ACU	IHISP252	Hispanic Interaction Of RP80ACU
IFEM248	Female Interaction Of RH61AQR	IHISP253	Hispanic Interaction Of RP80AQR
IBLK245	Black Interaction Of RH61ALN	IOTH250	Other Interaction Of RP80ALN
IBLK246	Black Interaction Of RH61AQU	IOTH251	Other Interaction Of RP80AQU
IBLK247	Black Interaction Of RH61ACU	IOTH252	Other Interaction Of RP80ACU
IBLK248	Black Interaction Of RH61AQR	IOTH253	Other Interaction Of RP80AQR
IHISP245	Hispanic Interaction Of RH61ALN	ARATELN	C: Linear: Mean A-Only Client Trmt Rate 1991 & 1992
IHISP246	Hispanic Interaction Of RH61AQU	ARATEQU	C: Quad: Mean A-Only Client Trmt Rate 1991 & 1992
IHISP247	Hispanic Interaction Of RH61ACU	ARATECU	C: Cubic: Mean A-Only Client Trmt Rate 1991 & 1992
IHISP248	Hispanic Interaction Of RH61AQR	ARATEQR	C: Quart: Mean A-Only Client Trmt Rate 1991 & 1992
IOTH245 IOTH246 IOTH247 IOTH248	Other Interaction Of RH61ALN Other Interaction Of RH61AQU Other Interaction Of RH61ACU Other Interaction Of RH61AQR	IFEM255 IFEM256 IFEM257 IFEM258	Female Interaction Of ARATELN Female Interaction Of ARATEQU Female Interaction Of ARATECU Female Interaction Of ARATEQR
RP80ALN RP80AQU RP80ACU IBLK258	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income Black Interaction Of ARATEQR	IBLK255 IBLK256 IBLK257 IHISP262 IHISP263	Black Interaction Of ARATELN Black Interaction Of ARATEQU Black Interaction Of ARATECU Hispanic Interaction Of B64DISCU Hispanic Interaction Of B64DISQR
IHISP255	Hispanic Interaction Of ARATELN	IOTH260	Other Interaction Of B64DISLN Other Interaction Of B64DISQU Other Interaction Of B64DISCU Other Interaction Of B64DISQR
IHISP256	Hispanic Interaction Of ARATEQU	IOTH261	
IHISP257	Hispanic Interaction Of ARATECU	IOTH262	
IHISP258	Hispanic Interaction Of ARATEQR	IOTH263	
IOTH255 IOTH256 IOTH257 IOTH258	Other Interaction Of ARATELN Other Interaction Of ARATEQU Other Interaction Of ARATECU Other Interaction Of ARATEQR	BAGE34LN BAGE34QU BAGE34CU BAGE34QR	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years
B64DISLN	B: Linear: % Persons 16-64 With A Work Disability	IFEM265	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU Female Interaction Of BAGE34QR
B64DISQU	B: Quadratic: % Persons 16-64 With A Work Disability	IFEM266	
B64DISCU	B: Cubic: % Persons 16-64 With A Work Disability	IFEM267	
B64DISQR	B: Quartic: % Persons 16-64 With A Work Disability	IFEM268	
IFEM260	Female Interaction Of B64DISLN	IBLK265	Black Interaction Of BAGE34LN
IFEM261	Female Interaction Of B64DISQU	IBLK266	Black Interaction Of BAGE34QU
IFEM262	Female Interaction Of B64DISCU	IBLK267	Black Interaction Of BAGE34CU
IFEM263	Female Interaction Of B64DISQR	IBLK268	Black Interaction Of BAGE34QR
IBLK260	Black Interaction Of B64DISLN	IHISP265	Hispanic Interaction Of BAGE34LN
IBLK261	Black Interaction Of B64DISQU	IHISP266	Hispanic Interaction Of BAGE34QU
IBLK262	Black Interaction Of B64DISCU	IHISP267	Hispanic Interaction Of BAGE34CU
IBLK263	Black Interaction Of B64DISQR	IHISP268	Hispanic Interaction Of BAGE34QR
IHISP260 IHISP261	Hispanic Interaction Of B64DISLN Hispanic Interaction Of B64DISQU	IOTH265	Other Interaction Of BAGE34LN

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

<u>Variable</u>	Label	<u>Variable</u>	<u>Label</u>
IOTH266	Other Interaction Of BAGE34QU		
IOTH267	Other Interaction Of BAGE34ĈU	IFEM275	Female Interaction Of BAGE54LN
IOTH268	Other Interaction Of BAGE34QR	IFEM276	Female Interaction Of BAGE54QU
BAGE44LN	B: Linear: Percent Persons 35-44 Years	IFEM277	Female Interaction Of BAGE54CU
BAGE44QU	B: Quadratic: Percent Persons 35-44 Years	IFEM278	Female Interaction Of BAGE54QR
BAGE44CU	B: Cubic: Percent Persons 35-44 Years		
BAGE44QR	B: Quartic: Percent Persons 35-44 Years	IBLK275	Black Interaction Of BAGE54LN
		IBLK276	Black Interaction Of BAGE54QU
IFEM270	Female Interaction Of BAGE44LN	IBLK277	Black Interaction Of BAGE54CU
IFEM271	Female Interaction Of BAGE44QU	IBLK278	Black Interaction Of BAGE54QR
IFEM272 IFEM273	Female Interaction Of BAGE44CU	IHISP275	Hignoria Interaction Of DACESALN
IFEIVIZ/3	Female Interaction Of BAGE44QR	IHISP276	Hispanic Interaction Of BAGE54LN Hispanic Interaction Of BAGE54QU
IBLK270	Black Interaction Of BAGE44LN	IHISP277	Hispanic Interaction Of BAGE54QU
IBLK270	Black Interaction Of BAGE44QU	IHISP278	Hispanic Interaction Of BAGE54C0
IBLK272	Black Interaction Of BAGE44CU	111101 270	Inspanie interaction of Briods (Qit
IBLK273	Black Interaction Of BAGE44QR	IOTH275	Other Interaction Of BAGE54LN
		IOTH276	Other Interaction Of BAGE54QU
IHISP270	Hispanic Interaction Of BAGE44LN	IOTH277	Other Interaction Of BAGE54CU
IHISP271	Hispanic Interaction Of BAGE44QU	IOTH278	Other Interaction Of BAGE54QR
IHISP272	Hispanic Interaction Of BAGE44CU		
IHISP273	Hispanic Interaction Of BAGE44QR	BAGE64LN	B: Linear: Percent Persons 55-64 Years
		BAGE64QU	B: Quadratic: Percent Persons 55-64 Years
IOTH270	Other Interaction Of BAGE44LN	BAGE64CU	B: Cubic: Percent Persons 55-64 Years
IOTH271	Other Interaction Of BAGE44QU	BAGE64QR	B: Quartic: Percent Persons 55-64 Years
IOTH272	Other Interaction Of BAGE44CU	IEEM200	E
IOTH273	Other Interaction Of BAGE44QR	IFEM280	Female Interaction Of BAGE64LN
BAGE54LN	B: Linear: Percent Persons 45-54 Years	IFEM281 IFEM282	Female Interaction Of BAGE64QU Female Interaction Of BAGE64CU
BAGE54QU	B: Ouadratic: Percent Persons 45-54 Years	IFEM283	Female Interaction Of BAGE64QR
BAGE54CU	B: Cubic: Percent Persons 45-54 Years	II LIVI203	Temale interaction of BAGE04QR
BAGE54QR	B: Quartic: Percent Persons 45-54 Years	IBLK280	Black Interaction Of BAGE64LN
IBLK281	Black Interaction Of BAGE64QU	IHISP288	Hispanic Interaction Of BASIANQR
IBLK282	Black Interaction Of BAGE64ČU		1
IBLK283	Black Interaction Of BAGE64QR	IOTH285	Other Interaction Of BASIANLN
		IOTH286	Other Interaction Of BASIANQU
IHISP280	Hispanic Interaction Of BAGE64LN	IOTH287	Other Interaction Of BASIANCU
IHISP281	Hispanic Interaction Of BAGE64QU	IOTH288	Other Interaction Of BASIANQR
IHISP282	Hispanic Interaction Of BAGE64CU	DOLLD LAWAY	D. T. D. L. T. L. G. L.
IHISP283	Hispanic Interaction Of BAGE64QR	BCUBANLN	B: Linear: Percent Hispanics: Cuban
IOTH280	Other Interaction Of BAGE64LN	BCUBANQU	B: Quadratic: Percent Hispanics: Cuban
IOTH281	Other Interaction Of BAGE64QU	BCUBANCU BCUBANQR	B: Cubic: Percent Hispanics: Cuban B: Quartic: Percent Hispanics: Cuban
IOTH282	Other Interaction Of BAGE64CU	DCODANQK	B. Quartic. I ciccit Hispanies. Cuban
IOTH283	Other Interaction Of BAGE64QR	IFEM290	Female Interaction Of BCUBANLN
10 111200	0 mar miletaerien er 2.1020 i Qit	IFEM291	Female Interaction Of BCUBANQU
BASIANLN	B: Linear: Percent Population: Asian, Pacific Islander	IFEM292	Female Interaction Of BCUBANCU
BASIANQU	B: Quadratic: Percent Population: Asian, Pacific Islander	IFEM293	Female Interaction Of BCUBANQR
BASIANČU	B: Cubic: Percent Population: Asian, Pacific Islander		•
BASIANQR	B: Quartic: Percent Population: Asian, Pacific Islander	IBLK290	Black Interaction Of BCUBANLN
		IBLK291	Black Interaction Of BCUBANQU
IFEM285	Female Interaction Of BASIANLN	IBLK292	Black Interaction Of BCUBANCU
IFEM286	Female Interaction Of BASIANQU	IBLK293	Black Interaction Of BCUBANQR
IFEM287	Female Interaction Of BASIANCU	HHCD200	H I OCDCHDANIA
IFEM288	Female Interaction Of BASIANQR	IHISP290	Hispanic Interaction Of BCUBANLN Hispanic Interaction Of BCUBANQU
IDI 1/205	Diagla Internation Of DACIANI N	IHISP291	
IBLK285 IBLK286	Black Interaction Of BASIANLN Black Interaction Of BASIANOU	IHISP292 IHISP293	Hispanic Interaction Of BCUBANCU Hispanic Interaction Of BCUBANQR
IBLK287	Black Interaction Of BASIANCU	111131 473	mopanic interaction of beobangs
IBLK288	Black Interaction Of BASIANCO	IOTH290	Other Interaction Of BCUBANLN
-22		IOTH291	Other Interaction Of BCUBANQU
IHISP285	Hispanic Interaction Of BASIANLN	IOTH292	Other Interaction Of BCUBANCU
IHISP286	Hispanic Interaction Of BASIANQU	IOTH293	Other Interaction Of BCUBANQR
IHISP287	Hispanic Interaction Of BASIANCU		~

Table D.1 Block Group and Tract Level Variables Considered in Small Area Estimation Models (continued)

<u>Variable</u>	<u>Label</u>	<u>Variable</u>	<u>Label</u>
BFNOTLN	B: Linear: % Fem Separated, Divorced or Widowed	IFEM300	Female Interaction Of BINDIALN Female Interaction Of BINDIAQU Female Interaction Of BINDIACU Female Interaction Of BINDIAQR
BFNOTQU	B: Quadratic: % Fem Separated, Divorced or Widowed	IFEM301	
BFNOTCU	B: Cubic: % Fem Separated, Divorced or Widowed	IFEM302	
BFNOTQR	B: Quartic: % Fem Separated, Divorced or Widowed	IFEM303	
IFEM295	Female Interaction Of BFNOTLN	IBLK300	Black Interaction Of BINDIALN Black Interaction Of BINDIAQU Black Interaction Of BINDIACU Black Interaction Of BINDIACU
IFEM296	Female Interaction Of BFNOTQU	IBLK301	
IFEM297	Female Interaction Of BFNOTCU	IBLK302	
IFEM298	Female Interaction Of BFNOTQR	IBLK303	
IBLK295	Black Interaction Of BFNOTLN	IHISP300	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU Hispanic Interaction Of BINDIACU Hispanic Interaction Of BINDIAQR
IBLK296	Black Interaction Of BFNOTQU	IHISP301	
IBLK297	Black Interaction Of BFNOTCU	IHISP302	
IBLK298	Black Interaction Of BFNOTQR	IHISP303	
IHISP295	Hispanic Interaction Of BFNOTLN	IOTH300	Other Interaction Of BINDIALN Other Interaction Of BINDIAQU Other Interaction Of BINDIACU Other Interaction Of BINDIAQR
IHISP296	Hispanic Interaction Of BFNOTQU	IOTH301	
IHISP297	Hispanic Interaction Of BFNOTCU	IOTH302	
IHISP298	Hispanic Interaction Of BFNOTQR	IOTH303	
IOTH295	Other Interaction Of BFNOTLN Other Interaction Of BFNOTQU Other Interaction Of BFNOTCU Other Interaction Of BFNOTQR	BMNOTLN	B: Linear: % Males Separated, Divorced or Widowed
IOTH296		BMNOTQU	B: Quad: % Males Separated, Divorced or Widowed
IOTH297		BMNOTCU	B: Cubic: % Males Separated, Divorced or Widowed
IOTH298		BMNOTQR	B: Quartic: % Males Separated, Divorced or Widowed
BINDIALN BINDIAQU BINDIACU BINDIAQR IFEM308	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quad: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut B: Quart: Percent Pop: American Indian, Eskimo, Aleut Female Interaction Of BMNOTQR	IFEM305 IFEM306 IFEM307 IHISP313	Female Interaction Of BMNOTLN Female Interaction Of BMNOTQU Female Interaction Of BMNOTCU Hispanic Interaction Of BPOVERQR
IBLK305	Black Interaction Of BMNOTLN	IOTH310	Other Interaction Of BPOVERLN Other Interaction Of BPOVERQU Other Interaction Of BPOVERCU Other Interaction Of BPOVERQR
IBLK306	Black Interaction Of BMNOTQU	IOTH311	
IBLK307	Black Interaction Of BMNOTCU	IOTH312	
IBLK308	Black Interaction Of BMNOTQR	IOTH313	
IHISP305	Hispanic Interaction Of BMNOTLN	BPRICALN	B: Linear: Percent Hispanics: Puerto Rican
IHISP306	Hispanic Interaction Of BMNOTQU	BPRICAQU	B: Quadratic: Percent Hispanics: Puerto Rican
IHISP307	Hispanic Interaction Of BMNOTCU	BPRICACU	B: Cubic: Percent Hispanics: Puerto Rican
IHISP308	Hispanic Interaction Of BMNOTQR	BPRICAQR	B: Quartic: Percent Hispanics: Puerto Rican
IOTH305	Other Interaction Of BMNOTLN Other Interaction Of BMNOTQU Other Interaction Of BMNOTCU Other Interaction Of BMNOTQR	IFEM315	Female Interaction Of BPRICALN
IOTH306		IFEM316	Female Interaction Of BPRICAQU
IOTH307		IFEM317	Female Interaction Of BPRICACU
IOTH308		IFEM318	Female Interaction Of BPRICAQR
BPOVERLN	B: Linear: Percent Families Below Poverty Level	IBLK315	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU Black Interaction Of BPRICACU Black Interaction Of BPRICAQR
BPOVERQU	B: Quadratic: Percent Families Below Poverty Level	IBLK316	
BPOVERCU	B: Cubic: Percent Families Below Poverty Level	IBLK317	
BPOVERQR	B: Quartic: Percent Families Below Poverty Level	IBLK318	
IFEM310 IFEM311 IFEM312 IFEM313	Female Interaction Of BPOVERLN Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU Female Interaction Of BPOVERQR	IHISP315 IHISP316 IHISP317 IHISP318 IOTH315	Hispanic Interaction Of BPRICALN Hispanic Interaction Of BPRICAQU Hispanic Interaction Of BPRICACU Hispanic Interaction Of BPRICAQR Other Interaction Of BPRICALN
IBLK310 IBLK311 IBLK312 IBLK313	Black Interaction Of BPOVERLN Black Interaction Of BPOVERQU Black Interaction Of BPOVERCU Black Interaction Of BPOVERQR	IOTH316 IOTH317 IOTH318	Other Interaction Of BPRICAQU Other Interaction Of BPRICACU Other Interaction Of BPRICAQR
IHISP310 IHISP311 IHISP312	Hispanic Interaction Of BPOVERLN Hispanic Interaction Of BPOVERQU Hispanic Interaction Of BPOVERCU	BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree

 Table D.1
 Block Group and Tract Level Variables Considered in Small Area Estimation Models

 (continued)

<u>Variable</u>	<u>Label</u>	<u>Variable</u>	<u>Label</u>
IFEM320 IFEM321 IFEM322 IFEM323	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU Female Interaction Of BSCHASCU Female Interaction Of BSCHASQR	IFEM325 IFEM326 IFEM327 IFEM328	Female Interaction Of PASIANLN Female Interaction Of PASIANQU Female Interaction Of PASIANCU Female Interaction Of PASIANQR
IBLK320 IBLK321 IBLK322 IBLK323	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU Black Interaction Of BSCHASQR	IBLK325 IBLK326 IBLK327 IBLK328	Black Interaction Of PASIANLN Black Interaction Of PASIANQU Black Interaction Of PASIANCU Black Interaction Of PASIANQR
IHISP320 IHISP321 IHISP322 IHISP323	Hispanic Interaction Of BSCHASLN Hispanic Interaction Of BSCHASQU Hispanic Interaction Of BSCHASCU Hispanic Interaction Of BSCHASQR	IHISP325 IHISP326 IHISP327 IHISP328	Hispanic Interaction Of PASIANLN Hispanic Interaction Of PASIANQU Hispanic Interaction Of PASIANCU Hispanic Interaction Of PASIANQR
IOTH320 IOTH321 IOTH322 IOTH323	Other Interaction Of BSCHASLN Other Interaction Of BSCHASQU Other Interaction Of BSCHASCU Other Interaction Of BSCHASQR	IOTH325 IOTH326 IOTH327 IOTH328	Other Interaction Of PASIANLN Other Interaction Of PASIANQU Other Interaction Of PASIANCU Other Interaction Of PASIANQR
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander	PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban
IFEM331 IFEM332 IFEM333	Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU Female Interaction Of PCUBANQR	IFEM330 IOTH335 IOTH336	Female Interaction Of PCUBANLN Other Interaction Of PINDIALN Other Interaction Of PINDIAQU
IBLK330 IBLK331 IBLK332 IBLK333	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU Black Interaction Of PCUBANQR	IOTH337 IOTH338 HSDROPLN HSDROPQU	Other Interaction Of PINDIACU Other Interaction Of PINDIAQR T: Linear: Underclass Indicator: High School Dropouts T: Quadratic: Underclass Indicator: HS Dropouts
IHISP330 IHISP331 IHISP332 IHISP333	Hispanic Interaction Of PCUBANLN Hispanic Interaction Of PCUBANQU Hispanic Interaction Of PCUBANCU Hispanic Interaction Of PCUBANQR	HSDROPCU HSDROPQR IFEM340 IFEM341	T: Cubic: Underclass Indicator: HS Dropouts T: Quartic: Underclass Indicator: HS Dropouts Female Interaction Of HSDROPLN Female Interaction Of HSDROPQU
IOTH330 IOTH331 IOTH332	Other Interaction Of PCUBANLN Other Interaction Of PCUBANQU Other Interaction Of PCUBANCU	IFEM342 IFEM343	Female Interaction Of HSDROPCU Female Interaction Of HSDROPQR
IOTH333 PINDIALN PINDIAQU PINDIACU	Other Interaction Of PCUBANQR T: Linear: % Pop: American Indian, Eskimo, Aleut T: Quadratic: % Pop: American Indian, Eskimo, Aleut T: Cubic: % Pop: American Indian, Eskimo, Aleut	IBLK340 IBLK341 IBLK342 IBLK343	Black Interaction Of HSDROPLN Black Interaction Of HSDROPQU Black Interaction Of HSDROPCU Black Interaction Of HSDROPQR
PINDIAQR IFEM335 IFEM336 IFEM337	T: Quartic: % Pop: American Indian, Eskimo, Aleut Female Interaction Of PINDIALN Female Interaction Of PINDIAQU Female Interaction Of PINDIACU	IHISP340 IHISP341 IHISP342 IHISP343	Hispanic Interaction Of HSDROPLN Hispanic Interaction Of HSDROPQU Hispanic Interaction Of HSDROPCU Hispanic Interaction Of HSDROPQR
IFEM338 IBLK335 IBLK336	Female Interaction Of PINDIAQR Black Interaction Of PINDIALN Black Interaction Of PINDIAQU	IOTH340 IOTH341 IOTH342 IOTH343	Other Interaction Of HSDROPLN Other Interaction Of HSDROPQU Other Interaction Of HSDROPCU Other Interaction Of HSDROPQR
IBLK337 IBLK338	Black Interaction Of PINDIACU Black Interaction Of PINDIAQR	HSDROP9	T: Underclass Indicator: High School Dropouts
IHISP335 IHISP336 IHISP337	Hispanic Interaction Of PINDIALN Hispanic Interaction Of PINDIAQU Hispanic Interaction Of PINDIACU	IFEM339 IBLK339	Female Interaction Of HSDROP9 Black Interaction Of HSDROP9
IHISP338	Hispanic Interaction Of PINDIAQR	IDDIK55)	Sack includion of Hobbot?

Table D.1 Block Group and Tract Level Variables Considered in Small Area Estimation Models (continued)

<u>Variable</u> <u>Label</u> <u>Label</u>

IHISP339 Hispanic Interaction Of HSDROP9
IOTH339 Other Interaction Of HSDROP9



Appendix E: Fixed Effect Coefficients for Logistic Models

E1. Coefficients of Model Parameters for Past Month Alcohol Use

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	-1.371	0.862	1.003	0.392	-1.258	0.970	0.991	0.913
FEMALE	O: Female Indicator	-0.070	-0.277	-0.448	-0.570	-0.203	-0.481	-0.704	-0.797
FEMBLCK	O: Black Interaction Of FEMALE	-0.172	-0.105	-0.442	-0.340	-0.156	-0.138	0.220	-0.200
FEMHISP FEMOTHR	O: Hispanic Interaction Of FEMALE O: Other Interaction Of FEMALE	-0.309 -0.167	-0.564 -0.080	-0.823 -0.883	-0.622 -0.609	-0.005 0.742	-0.585 0.010	-0.511 0.128	-0.461 0.576
RACEBLCK	O: Other Interaction Of FEMALE O: Race/Black Indicator	-0.731	-0.589	-0.865	-0.064	-0.303	-0.407	-0.471	-0.343
RACEHISP RACEOTHR	O: Race/Hispanic Indicator O: Race/Other Indicator	0.005 -0.631	-0.208 -1.065	-0.013 -0.912	-0.005 -0.939	-0.000 -0.967	-0.105 -2.323	-0.077 -0.896	0.143 -3.210
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator						0.143 -0.262 0.097		-0.173 -0.451 -0.138
IOTH7 IOTH8 IOTH9	Other Interaction Of REGNOREA Other Interaction Of REGSOUTH Other Interaction Of REGWEST						1.864 1.558 1.154		2.478 3.241 1.051
UCLASS9	T: Underclass Indicator								0.864
PAGE18LN PAGE18QU	T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years				-0.074				-0.014 -0.038
IOTH25 IOTH26	Other Interaction Of PAGE18LN Other Interaction Of PAGE18QU								0.375 -0.503
PAGE24LN PAGE24QU	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years						-0.037 -0.087		
IFEM30 IFEM31	Female Interaction Of PAGE24LN Female Interaction Of PAGE24QU						-0.009 0.141		
PAGE34LN	T: Linear: Percent Persons 25-34 Years	-0.076							
PAGE44LN PAGE44QU PAGE44CU PAGE44QR	T: Linear: Percent Persons 35-44 Years T: Quadratic: Percent Persons 35-44 Years T: Cubic: Percent Persons 35-44 Years T: Quartic: Percent Persons 35-44 Years T: Quartic: Percent Persons 35-44 Years				0.108				0.044 0.013 0.004 0.023
PAGE64LN	T: Linear: Percent Persons 55-64 Years	0.043							
PSCH8LN	T: Linear: Percent 0-8 Years Of School		-0.097						
PSCHCOLN	T: Linear: Bachelors, Graduate, Or Professional Degree		0.155				0.165	0.153	
IBLK75	Black Interaction Of PSCHCOLN						-0.226	-0.170	
PSCHSCLN	T: Linear: Percent Some College And No Degree		-0.060			0.056			
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Quartic: Percent Families Below Poverty Level					-0.076 -0.030 -0.116 -0.023			
IBLK85 IBLK86 IBLK87 IBLK88	Black Interaction Of PPOVERLN Black Interaction Of PPOVERQU Black Interaction Of PPOVERCU Black Interaction Of PPOVERQR					0.235 0.048 0.078 0.063			
IBLK88 PBLACKLN	T: Linear: Percent Black Nonhispanic		-0.132			0.003			
POTHLN POTHQU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity						0.034 -0.052		
PFLABLN PFLABQU	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force	0.039 0.049							
PFNEVLN	T: Linear: Percent Females Never Married						0.091		
PFNOTLN PFNOTQU PFNOTCU PFNOTQR	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed T: Cubic: Percent Females Separated, Divorced Or Widowed T: Ouartic: Percent Females Separated. Divorced Or Widowed				-0.037 -0.002 -0.060 0.028				
PMNEVLN PMNEVQU PMNEVCU PMNEVQR	T: Linear: Percent Males Never Married T: Cuadratic: Percent Males Never Married T: Cubic: Percent Males Never Married T: Cuaric: Percent Males Never Married T: Cuartic: Percent Males Never Married			0.067	2.020	0.103		0.046 -0.021 0.031	0.069 -0.044 0.003 0.027
IFEM140 IFEM141	Female Interaction Of PMNEVLN Female Interaction Of PMNEVQU								0.038 0.162
IBLK140	Black Interaction Of PMNEVLN					-0.262			
IHISP140	Hispanic Interaction Of PMNEVLN			-0.152					
IOTH140 IOTH141 IOTH142	Other Interaction Of PMNEVLN Other Interaction Of PMNEVQU Other Interaction Of PMNEVCU			0.289				0.079 0.039 -0.425	
P40HULN	T: Linear: Percent Housing Units Built 1940-1949					0.124			
IFEM160	Female Interaction Of P40HULN					-0.161			
PRENTLN	T: Linear: Percent Housing Rented			-0.026	j	-0.008			
IBLK165	Black Interaction Of PRENTLN		0.002	0.059	ļ				
ADRATELN	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases		0.002 -0.077						

			BIG CITY				REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IOTH185	Other Interaction Of V18FLN								-0.514
V18ALN V18AQU V18ACU V18AQR	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Cubic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quartic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate		-0.003 -0.044		-0.027 -0.041 -0.034				-0.082 -0.026 -0.014 -0.038
IHISP200 IHISP201	Hispanic Interaction Of V18ALN Hispanic Interaction Of V18AQU		-0.164 0.117						
V18LN	C: Linear: Total Drug Abuse Violations Arrest Rate				ľ				0.174
DRATELN DRATEQU DRATECU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992 C: Cubic: Mean Drug Client Treatment Rate 1991 & 1992					0.051 -0.013 -0.024			
IOTH235 IOTH236 IOTH237	Other Interaction Of DRATELN Other Interaction Of DRATEQU Other Interaction Of DRATECU					-0.020 0.070 -0.436			
RH61ALN RH61AQU	T: Linear: Recoded Median Value Of Owner Occupied HUs T: Quadratic: Recoded Median Value Of Owner Occupied HUs		-0.003 -0.006						0.227
IHISP245 IHISP246	Hispanic Interaction Of RH61 ALN Hispanic Interaction Of RH61 AQU		-0.099 0.103		ŀ				
RP80ALN RP80AQU	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income				0.035 0.051				
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratıc: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992		0.015 0.013 -0.013						
IFEM255 IFEM256 IFEM257	Female Interaction Of ARATELN Female Interaction Of ARATEQU Female Interaction Of ARATECU		-0.073 0.035 -0.097						
B64DISLN B64DISQU B64DISCU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability				-0.009 0.040		-0.017 -0.007 0.054		-0.150
IHISP260 IHISP261 IHISP262	Hispanic Interaction Of B64DISLN Hispanic Interaction Of B64DISQU Hispanic Interaction Of B64DISCU						-0.046 0.116 -0.160		0.202
BAGE34LN IOTH265	B: Linear: Percent Persons 25-34 Years Other Interaction Of BAGE34LN				ļ			0.027 -0.437	
BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years				ļ			0.437	-0.022 -0.005
IHISP270 IHISP271	Hispanic Interaction Of BAGE44LN Hispanic Interaction Of BAGE44QU				ļ				0.130 0.182
BAGE54LN	B: Linear: Percent Persons 45-54 Years				0.057				0.038
IBLK275 BASIANLN	Black Interaction Of BAGE54LN B: Linear: Percent Population: Asian, Pacific Islander				į			0.002	-0.164 -0.040
BASIANQU BASIANCU BASIANQR	B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander							-0.007 -0.055 -0.003	-0.004 0.005
IOTH285 IOTH286 IOTH287 IOTH288	Other Interaction Of BASIANLN Other Interaction Of BASIANQU Other Interaction Of BASIANCU Other Interaction Of BASIANCU							-0.221 0.002 0.242 0.199	0.642 -0.761 0.757
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban		-0.032			0.091 0.042 0.057			
IHISP290 IOTH290	Hispanic Interaction Of BCUBANLN Other Interaction Of BCUBANLN		0.185			0.294			
IOTH291 IOTH292	Other Interaction Of BCUBANQU Other Interaction Of BCUBANQU				ŀ	0.055 -1.073			
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed		0.034 0.001 0.024 -0.010				-0.030 -0.034		
IOTH295 IOTH296 IOTH297	Other Interaction Of BFNOTLN Other Interaction Of BFNOTQU Other Interaction Of BFNOTCU		0.103 -0.001 -0.145				0.093 0.263		
IOTH298 BINDIALN BINDIAQU BINDIACU	Other Interaction Of BENOTQR B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut		0.062			-0.109	-0.030 -0.017 0.171		
IFEM300 IFEM301 IFEM302	Female Interaction Of BINDIAQU Female Interaction Of BINDIAQU Female Interaction Of BINDIAQU					0.225	-0.012 -0.008 -0.185		
IBLK300	Black Interaction Of BINDIALN					0.291			
BMNOTUN BMNOTOU BMNOTCU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed	0.045							-0.030 0.002 -0.042
IHISP305 IHISP306 IHISP307	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU Hispanic Interaction Of BMNOTCU	-0.127	·						0.086 0.012 0.171
BPOVERLN BPRICALN	B: Linear: Percent Families Below Poverty Level B: Linear: Percent Hispanics: Puerto Rican	-0.052		-0.084	į	-0.016	-0.027		
BPRICACU BPRICACU	B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican	0.010 0.075			ļ	0.000 -0.031	-0.027 -0.083 0.017		

NOTE: T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

			BIG CITY			REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BPRICAQR	B: Quartic: Percent Hispanics: Puerto Rican						0.072		
IBLK315 IBLK316	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU						-0.124 0.260		
BSCHASLN BSCHASQU BSCHASCU	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree				0.053 -0.026 0.013				0.059
IBLK320 IBLK321 IBLK322	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU				-0.072 0.083 -0.113				
PASIANLN PASIANQU	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander			0.035 -0.062					
IBLK325 IBLK326	Black Interaction Of PASIANLN Black Interaction Of PASIANQU			0.055 0.090					
PCUBANLN PCUBANQU PCUBANCU	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban				-0.025 -0.007 -0.109				
IFEM330 IFEM331 IFEM332	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU				0.013 0.019 0.129				
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Aleut			-0.040 0.028 -0.011 0.015					-0.005 -0.086 0.051 -0.039
IFEM335 IFEM336 IFEM337 IFEM338	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU Female Interaction Of PINDIACU Female Interaction Of PINDIAQR								-0.026 0.035 -0.099 0.059

E2. Coefficients of Model Parameters for Past Month Any Illicit Drug Use

Variable DUMMY O: Intercept Term FEMALE O: Female Indicator FEMBLCK O: Black Interaction O FEMHISP O: Hispanic Interaction O FEMOTHR O: Other Interaction O RACEBLCK RACEBLCK RACEBLSP O: Race/Black Indica RACEHISP O: Race/Hispanic Ind	on Of FEMALE Of FEMALE cor icator or or indicator cator	12-17 -2.903 0.049 -0.534 -0.274 0.459 0.074 0.255 -1.180	18-25 -1.573 -0.420 -0.327 -0.085 -0.055 -0.103 -0.518 -1.001	26-34 -1.738 -0.769 0.257 0.330 -0.549 -0.354 -0.878 -1.000	35+ -3.271 -0.581 0.062 -0.180 0.192 0.101	12-17 -2.807 0.088 -0.050 -0.357 0.144	18-25 -1.406 -0.718 0.203 0.282 -0.165 -1.613	26-34 -2.537 -0.353 -0.148 -0.258 1.096	35+ -3.825 -0.424 -0.277 0.116 0.600
FEMALE O: Female Indicator FEMBLCK O: Black Interaction (FEMHISP O: Hispanic Interaction (FEMOTHR O: Other Interaction (RACEBLCK O: Race/Black Indica RACEHISP O: Race/Hispanic Ind	on Of FEMALE Of FEMALE cor icator or or indicator cator	0.049 -0.534 -0.274 0.459 0.074 0.255	-0.420 -0.327 -0.085 -0.055 -0.103 -0.518	-0.769 0.257 0.330 -0.549 -0.354 -0.878	-0.581 0.062 -0.180 0.192 0.101	0.088 -0.050 -0.357	-0.718 0.203 0.282 -0.165 -1.613	-0.353 -0.148 -0.258 1.096	-0.424 -0.277 0.116
FEMBLCK O: Black Interaction (FEMHISP O: Hispanic Interaction FEMOTHR O: Other Interaction (RACEBLCK O: Race/Black Indica RACEHISP O: Race/Hispanic Ind	on Of FEMALE Of FEMALE cor icator or or indicator cator	-0.534 -0.274 0.459 0.074 0.255	-0.327 -0.085 -0.055 -0.103 -0.518	0.257 0.330 -0.549 -0.354 -0.878	0.062 -0.180 0.192 0.101	-0.050 -0.357	0.203 0.282 -0.165 -1.613	-0.148 -0.258 1.096	-0.277 0.116
FEMHISP O: Hispanic Interaction FEMOTHR O: Other Interaction O RACEBLCK O: Race/Black Indica RACEHISP O: Race/Hispanic Ind	on Of FEMALE Of FEMALE cor icator or or indicator cator	-0.274 0.459 0.074 0.255	-0.085 -0.055 -0.103 -0.518	0.330 -0.549 -0.354 -0.878	-0.180 0.192 0.101	-0.357	0.282 -0.165 -1.613	-0.258 1.096	0.116
FEMOTHR O: Other Interaction (RACEBLCK O: Race/Black Indica RACEHISP O: Race/Hispanic Ind	OFFEMALE tor icator or ndicator sator	0.459 0.074 0.255	-0.055 -0.103 -0.518	-0.549 -0.354 -0.878	0.192 0.101	-0.357 0.144	-0.165 -1.613	1.096	
RACEHISP O: Race/Hispanic Ind	icator or indicator cator	0.255	-0.518	-0.878	0.101	-0.357 0.144		0.046	
	or indicator cator		-0.518 -1.001	-0.878 -1.000		0.144		0.016	0.437
RACEOTHR O: Race/Other Indicat	cator				-1.055		-0.581 -0.541	0.614 -0.951	-0.511 -1.295
REGNOREA REGSOUTH REGWEST O: Northeast Region Indic		Ì					0.336 0.217 0.662	0.411 0.173 0.813	
PDENLEV1 O: Large MSA PDENLEV2 O: Medium MSA PDENLEV3 O: Small MSA PDENLEV4 O: NonMSA, Urban						0.073 0.201 0.297 0.121	-0.476 -0.321 -0.428 -0.347	-0.077 -0.004 0.024 -0.054	
IBLK10 Black Interaction Of I IBLK11 Black Interaction Of I IBLK12 Black Interaction Of I IBLK13 Black Interaction Of I	PDENLEV2 PDENLEV3						1.150 0.770 1.469 1.484		
IHISP10 Hispanic Interaction (IHISP11 Hispanic Interaction (IHISP12 Hispanic Interaction (IHISP13 Hispanic Interaction (Of PDENLEV2 Of PDENLEV3							-1.284 -1.705 -1.047 -0.962	
UCLASS9 T: Underclass Indicate	or						0.774		
PHH1PLN T: Linear: Percent On PHH1PQU T: Quadratic: Percent	e Person Households One Person Households						0.062 0.061		
POPRMLN T: Linear: Average Pe POPRMQU T: Quadratic: Average	ersons Per Room e Persons Per Room			-0.012				-0.196 -0.138	
IFEM20 Female Interaction Of IFEM21 Female Interaction Of IFEM21	POPRMLN							0.157 0.137	
PAGE18LN T: Linear: Percent Per PAGE18QU T: Quadratic: Percent PAGE18CU T: Cubic: Percent Per	Persons 0-18 Years						0.077 0.020 0.108		
IFEM25 Female Interaction Of IFEM26 Female Interaction Of IFEM27 Female Interaction Of IFEM27 Female Interaction Of IFEM27 Female Interaction Of IFEM27 Female Interaction Of IFEM25 Female In	PAGE18QU						-0.147 -0.001 -0.146		
PAGE24LN PAGE24QU PAGE24CU PAGE24QR T: Linear: Percent Per T: Quadratic: Percent Per T: Cubic: Percent Per T: Quartic: Percent Per	Persons 19-24 Years sons 19-24 Years							0.136 -0.032 -0.089 0.005	
IHISP30 Hispanic Interaction (IHISP31 Hispanic Interaction (IHISP32 Hispanic Interaction (IHISP33 Hispanic Interaction (Of PAGE24QU Of PAGE24CU							0.047 0.039 0.023 0.020	
PAGE34LN T: Linear: Percent Per PAGE34QU T: Quadratic: Percent						-0.002 -0.119	-0.022 -0.038		
IHISP35 Hispanic Interaction C IHISP36 Hispanic Interaction C							-0.076 0.186		
PAGE44LN T: Linear: Percent Per	rsons 35-44 Years							0.060	
PAGE54LN T: Linear: Percent Per PAGE54QU T: Quadratic: Percent PAGE54CU T: Cubic: Percent Per	Persons 45-54 Years				0.068 0.012		-0.136 -0.032 -0.057		-0.066 -0.057
IBLK45 Black Interaction Of I BLK46 Black Interaction Of I	PAGE54LN PAGE54QU				-0.133 0.061				
IHISP45 Hispanic Interaction (Of PAGE54LN				l		0.267		
PSCH8LN T: Linear: Percent 0-8 PSCH8QU T: Quadratic: Percent								-0.081 0.026	
IBLK55 Black Interaction Of I IBLK56 Black Interaction Of I	PSCH8LN							0.296 -0.228	
PSCH12LN T: Linear: Percent 9-1 PSCH12QU T: Quadratic: Percent PSCH12CU T: Cubic: Percent 9-1	2 Years & No High School Diploma 9-12 Years & No High School Diploma 2 Years & No High School Diploma 12 Years & No High School Diploma	-0.002 0.067 0.036 0.028			-0.182	0.113 0.133			0.043 0.137
IFEM60 Female Interaction Of IFEM61 Female In	PSCH12LN	0.026				-0.115 -0.181			
PSCHASLN PSCHASQU T: Linear: Percent As T: Quadratic: Percent	sociates Degree Associates Degree							0.023 -0.088	
IBLK65 Black Interaction Of I Black Interaction Of I	PSCHASLN PSCHASQU							0.100 0.158	
PSCHCOQU T: Quadratic: Bachelo	Graduate, Or Professional Degree rs, Graduate, Or Professional Degree					0.134 -0.007			-0.003 -0.042
PSCHCOCU T: Cubic: Bachelors, of IFEM75 Female Interaction Of Interaction	Graduate, Or Professional Degree PSCHCOLN PSCHCOOLI					0.082 0.024 0.024			
PPOVERLN T: Linear: Percent Fai	milies Below Poverty Level Families Below Poverty Level	-0.134 0.009	.	 -		-0.122 -0.006	-0.174	0.126 0.021	

			BIG CI	ITY			REMA	INDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
PPOVERCU PPOVERQR	1: Cubic: Percent Families Below Poverty Level T: Quartic: Percent Families Below Poverty Level	0.115				-0.052 -0.025		0.037 -0.023	
IFEM85	Female Interaction Of PPOVERLN				ļ		0.059		
PBLACKLN PBLACKQU PBLACKCU PBLACKQR	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic T: Cubic: Percent Black Nonhispanic T: Quartic: Percent Black Nonhispanic			0.076 -0.003 0.053			0.080 0.020 0.016 -0.025	0.063 -0.053 0.016	
PHISPLN PHISPQU PHISPCU PHISPQR	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic T: Cubic: Percent Hispanic T: Quartic: Percent Hispanic								0.082 -0.057 -0.067 -0.014
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity					-0.063 0.001 -0.020			
PHHF18LN PHHF18QU PHHF18CU PHHF18QR	T: Linear: % Female-Headed HH W/No Spouse & Chid Under 18 T: Quadratic: % Female-Headed HH W/No Spouse & Chid Under 18 T: Cubic: % Female-Headed HH W/No Spouse & Chid Under 18 T: Quartic: % Female-Headed HH W/No Spouse & Chid Under 18					0.045 0.020 -0.014 -0.014			0.018
IFEM120	Female Interaction Of PHHF18LN								0.073
PFNOTLN PFNOTQU	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed			0.086				0.003 0.033	
IHISP135 IHISP136	Hispanic Interaction Of PFNOTLN Hispanic Interaction Of PFNOTQU							-0.243 -0.077	
PMNEVLN	T: Linear: Percent Males Never Married				0.157				
PMNOTEN PMNOTQU	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed							0.095 -0.037	
IHISP150 IHISP151	Hispanic Interaction Of PMNOTLN Hispanic Interaction Of PMNOTQU							0.041 0.157	
POLDHULN POLDHUQU POLDHUCU	T: Linear: Percent Housing Units Built 1939 Or Earlier T: Quadratic: Percent Housing Units Built 1939 Or Earlier T: Cubic: Percent Housing Units Built 1939 Or Earlier			0.056 -0.016 -0.083			0.100 0.015	0.007 0.062	
P40HULN	T: Linear: Percent Housing Units Built 1940-1949					-0.025		0.145	
PRENTLN PRENTQU PRENTCU PRENTQR	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented T: Cubic: Percent Housing Rented T: Quartic: Percent Housing Rented					-0.120	-0.037	-0.160 0.059 0.038 0.032	0.261 0.137 0.007 -0.018
IFEM165	Female Interaction Of PRENTLN					0.280		0.222	
IBLK165 IBLK166 IBLK167	Black Interaction Of PRENTLN Black Interaction Of PRENTQU Black Interaction Of PRENTCU							0.232 -0.053 0.147	
IHISP165	Hispanic Interaction Of PRENTLN						0.086		
ADRATELN ADRATEQU ADRATECU ADRATEQR	C: Lınear: Death Rate For All Alcohol-Related Cases C: Quadratıc: Death Rate For All Alcohol-Related Cases C: Cubic: Death Rate For All Alcohol-Related Cases C: Quartic: Death Rate For All Alcohol-Related Cases		0.004 0.004 0.104			-0.028 -0.025 -0.059	-0.100 0.021 -0.004 0.054		
IBLK175 IBLK176 IBLK177 IBLK178	Black Interaction Of ADRATELN Black Interaction Of ADRATEQU Black Interaction Of ADRATECU Black Interaction Of ADRATEQR						-0.061 0.098 0.059 -0.089		
V18FLN	C: Linear: Marijuana Posession Arrest Rate				ļ	0.016			
V18ELN V18EQU V18ECU V18EQR	C: Lmear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate C: Quartic: Opium/Cocaine & Deriv Posession Arrest Rate						0.058 -0.040 0.030 -0.049	0.175	0.302 0.287 0.086 -0.002
IFEM195 IFEM196 IFEM197 IFEM198	Female Interaction Of V18ELN Female Interaction Of V18EQU Female Interaction Of V18ECU Female Interaction Of V18EQR						0.128 0.056 -0.030 0.058		
IBLK195	Black Interaction Of V18ELN						-0.282		
IHISP195 V18ALN	Hispanic Interaction Of V18ELN C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate	0.052	0.141			0.052	-0.352	-0.007	
V18AQU V18ACU	C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Cubic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate	0.091 -0.066	0.141			0.032 0.048 -0.119		-0.007	
IFEM200 IFEM201 IFEM202	Female Interaction Of V18ALN Female Interaction Of V18AQU Female Interaction Of V18ACU	-0.000 -0.047 0.136							
IBLK200 IBLK201 IBLK202 IHISP200	Black Interaction Of V18ALN Black Interaction Of V18AQU Black Interaction Of V18ACU Hispanic Interaction Of V18ALN		-0.254			-0.172 -0.065 0.211			
V18LN V18QU V18CU V18QR	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate		-0.049	0.003 0.025 0.009 0.019	0.066			-0.150 -0.070 -0.014 0.030	-0.076 -0.348
IFEM225 IFEM226 IFEM227 IFEM228	Female Interaction OFV18LN Female Interaction OFV18QU Female Interaction OFV18CU Female Interaction OFV18QR							0.042 0.061 0.017 -0.027	
IHISP225	Hispanic Interaction Of V18LN		0.007						
VIOLLN VIOLQU	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate				-0.022 -0.016				-0.297 -0.147

NOTE:

			BIG C	ITY			REMA	INDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IBLK230 IBLK231	Black Interaction Of VIOLLN Black Interaction Of VIOLQU				-0.146 0.029				
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992					0.099	0.020 -0.065	-0.021 0.080	
IFEM235 IFEM236	Female Interaction Of DRATELN Female Interaction Of DRATEQU							0.069 -0.156	
RH43ALN RH43AQU RH43ACU RH43AQR	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units T: Cubic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units	-0.024 0.016 0.020 -0.034						-0.000	-0.255 0.110
IBLK240 IBLK241	Black Interaction Of RH43ALN Black Interaction Of RH43AQU								-0.012 -0.233
RH61ALN RH61AQU RH61ACU	T: Linear: Recoded Median Value Of Owner Occupied HUs T: Quadratic: Recoded Median Value Of Owner Occupied HUs T: Cubic: Recoded Median Value Of Owner Occupied HUs						-0.019 -0.003 -0.097		
RP80ALN RP80AQU RP80ACU RP80AQR	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income T: Quartic: Recoded Median Household Income					-0.144 0.014 0.056 -0.011		-0.057	
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratue: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992	0.130				0.005 0.018 0.003			
B64DISLN B64DISQU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability					-0.003 -0.037	0.004 0.023		
IBLK260	Black Interaction Of B64DISLN						0.172 0.097		
BAGE34LN BAGE44LN	B: Linear: Percent Persons 25-34 Years B: Linear: Percent Persons 35-44 Years					-0.118	0.097		
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years						0.015 0.019 -0.065	-0.015 -0.068 0.039 0.034	0.051 -0.046 0.119
IFEM275 IFEM276 IFEM277 IFEM278	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU Female Interaction Of BAGE54QR						-0.017 -0.000 0.144	-0.003 -0.003 -0.041 -0.056	-0.023 0.120 -0.175
IHISP275 IHISP276 IHISP277	Hispanic Interaction Of BAGE54LN Hispanic Interaction Of BAGE54QU Hispanic Interaction Of BAGE54CU							-0.159 -0.028 -0.204	
BAGE64LN BAGE64QU BAGE64CU	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years	-0.032	-0.054 -0.053 0.008	0.009 -0.016 0.054	-0.101 0.048	-0.055 -0.010 -0.075	0.071 -0.034 0.048		-0.239 -0.042 0.065
IFEM280 IFEM281 IFEM282	Female Interaction Of BAGE64LN Female Interaction Of BAGE64QU Female Interaction Of BAGE64CU	0.140			0.112 -0.107	0.045 0.021 0.157			0.082 0.087 -0.020
IBLK280 IBLK281	Black Interaction Of BAGE64LN Black Interaction Of BAGE64QU								0.130 -0.030
IHISP280 IHISP281 IHISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		0.095 0.013 -0.124	-0.009 -0.032 -0.128			-0.017 -0.028 -0.092		
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander		0.005 -0.024	-0.009 0.001 -0.048 0.012			0.061	0.027 0.165 -0.004	0.051 0.026 0.048 0.025
IBLK285 IBLK286 IBLK287 IBLK288	Black Interaction Of BASIANLN Black Interaction Of BASIANQU Black Interaction Of BASIANCU Black Interaction Of BASIANCU								-0.082 -0.239 0.179 -0.132
IHISP285 IHISP286 IHISP287 IHISP288	Hispanic Interaction Of BASIANLN Hispanic Interaction Of BASIANQU Hispanic Interaction Of BASIANQU Hispanic Interaction Of BASIANQU Hispanic Interaction Of BASIANQR			0.045 0.079 0.082 -0.062				-0.032 -0.076 0.210	
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban			-0.002			0.270	-0.082	0.030 -0.113 0.330
IFEM290	Female Interaction Of BCUBANLN							0.214	
IBLK290 IBLK291 IBLK292	Black Interaction Of BCUBANLN Black Interaction Of BCUBANQU Black Interaction Of BCUBANCU								-0.017 0.347 -1.115
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed	0.074 0.025 0.033 0.025		-0.002 -0.030			-0.031 -0.007 0.070 0.018	0.128 -0.033	0.142
IFEM295 IFEM296 IFEM297	Female Interaction Of BFNOTLN Female Interaction Of BFNOTQU Female Interaction Of BFNOTCU		·	.		·	0.006 -0.050 -0.149	·	
IBLK295 IBLK296	Black Interaction Of BFNOTCU Black Interaction Of BFNOTLN Black Interaction Of BFNOTQU			-0.056 0.115			-v.147	-0.237 0.208	
IHISP295 IHISP296 IHISP297 IHISP298	Hispanic Interaction Of BFNOTUN Hispanic Interaction Of BFNOTQU Hispanic Interaction Of BFNOTCU Hispanic Interaction Of BFNOTCU Hispanic Interaction Of BFNOTQR			0.113			0.012 -0.068 -0.023 -0.039	0.200	
BINDIALN BINDIAQU BINDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut					0.108 -0.082 0.148	0.074	0.158	-0.043 -0.029

			BIG CI	ITY			REMA	INDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IBLK300	Black Interaction Of BINDIALN					0.563	0.099		
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU								0.351 -0.480
BMNOTEN BMNOTQU BMNOTCU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed				0.259 -0.007 0.082	0.064 0.019	0.019 0.015	-0.008 -0.007 0.049	
IFEM305 IFEM306 IFEM307	Female Interaction OFBMNOTLN Female Interaction OFBMNOTQU Female Interaction OFBMNOTCU				-0.276 0.000 -0.038			-0.081 -0.007 -0.108	
IHISP305 IHISP306	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU						-0.049 -0.112		
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.059 0.007 -0.012 -0.010				-0.030		-0.271 -0.035 0.043 -0.101
IFEM310 IFEM311 IFEM312 IFEM313	Female Interaction Of BPOVERLN Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU Female Interaction Of BPOVERQR						0.151		0.063 -0.029 -0.135 0.138
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican	-0.071	-0.098 -0.094			-0.168 -0.159 -0.016 0.010	0.010 -0.004 -0.078	0.046 -0.051 0.022	
IFEM315 IFEM316 IFEM317 IFEM318	Female Interaction Of BPRICALN Female Interaction Of BPRICAQU Female Interaction Of BPRICACU Female Interaction Of BPRICAQR					0.064 0.150 -0.064 -0.102			
IBLK315 IBLK316 IBLK317	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU Black Interaction Of BPRICACU		-0.035 0.200				-0.114 0.210	-0.168 0.080 0.204	
IHISP315 IHISP316 IHISP317	Hispanic Interaction Of BPRICALN Hispanic Interaction Of BPRICAQU Hispanic Interaction Of BPRICACU						-0.147 -0.019 0.303		
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree		-0.037 0.013			0.057 -0.029 -0.013 -0.026	-0.016 0.035 -0.028	0.031 0.031 0.066	0.110 0.026 -0.012
IFEM320 IFEM321 IFEM322 IFEM323	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU Female Interaction Of BSCHASCU Female Interaction Of BSCHASQR					-0.071 0.080 0.053 0.064			
IBLK320 IBLK321 IBLK322	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU					0.038 0.126 0.281		0.041 0.063 -0.087	
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander				-0.143		-0.073 -0.000 -0.067 0.023	0.002 -0.051 -0.111	
IFEM325 IFEM326 IFEM327 IFEM328	Female Interaction Of PASIANLN Female Interaction Of PASIANQU Female Interaction Of PASIANCU Female Interaction Of PASIANQR						-0.035 -0.053 0.015 -0.045		
IHISP325 IHISP326 IHISP327	Hispanic Interaction Of PASIANLN Hispanic Interaction Of PASIANQU Hispanic Interaction Of PASIANCU						0.035 0.099 0.169		
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban	-0.060 -0.114 -0.080		-0.055 0.015		-0.141 -0.167 -0.123		-0.019 0.052 0.012 0.007	
IFEM330 IFEM331 IFEM332	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU					0.137 0.155 0.356			
IBLK330 IBLK331 IBLK332	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU	0.016 0.060 0.044							
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Aleut	-0.038 0.069 -0.108		0.008 -0.035			-0.019		-0.071 0.004 -0.017 0.025
IFEM335 IFEM336 IFEM337 IFEM338	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU Female Interaction Of PINDIACU Female Interaction Of PINDIACU								0.055 0.055 0.054 -0.096
IHISP335 IHISP336	Hispanic Interaction Of PINDIALN Hispanic Interaction Of PINDIAQU			-0.083 0.152					

E3. Coefficients of Model Parameters for Past Month Cigarette Use

			BIG C	ITY			REMAIN	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35
DUMMY	O: Intercept Term	-2.096	-0.561	-0.606	-1.067	-2.495	-0.560	-0.930	-0.9
FEMALE	O: Female Indicator	0.115	-0.224	-0.098	-0.133	0.040	0.083	0.155	-0.3
FEMBLCK	O: Black Interaction Of FEMALE	-0.681	0.081	-0.271	-0.385	-0.022	-0.273	-0.376	-0.0
FEMHISP	O: Hispanic Interaction Of FEMALE	-0.186	-0.555	-0.640	-0.511	0.210	1 145	-0.947	-0.2
FEMOTHR RACEBLCK	O: Other Interaction Of FEMALE O: Race/Black Indicator	-0.212 -1.242	-0.942 -1.112	-1.580 0.049	-1.273 0.355	0.319	-1.145 -0.903	-0.980 0.075	-0.0 0.1
RACEHISP RACEOTHR	O: Race/Hispanic Indicator O: Race/Other Indicator	-0.441 -1.992	-0.251 -0.251	-0.194 0.047	0.282 -0.217	-2.685	-0.008	0.124 0.200	-0.0 0.3
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator						0.332 0.073 -0.060		
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban					0.191 0.476 0.353 0.436	-0.121 -0.085 -0.108 0.004	0.105 0.122 0.010 0.014	
IHISP10 IHISP11	Hispanic Interaction Of PDENLEV1 Hispanic Interaction Of PDENLEV2							-0.255 -0.193	
POPRMLN POPRMQU POPRMCU	T: Linear: Average Persons Per Room T: Quadratic: Average Persons Per Room T: Cubic: Average Persons Per Room			0.081		-0.023 -0.059 -0.083		-0.014	
IFEM20	Female Interaction Of POPRMLN							0.095	
IBLK20 IBLK21 IBLK22	Black Interaction Of POPRMLN Black Interaction Of POPRMQU Black Interaction Of POPRMCU					0.126 0.070 0.351			
IOTH20 IOTH21 IOTH22	Other Interaction Of POPRMLN Other Interaction Of POPRMQU Other Interaction Of POPRMCU					-0.949 -0.118 0.894			
PAGE18LN PAGE18QU PAGE18CU	T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years T: Cubic: Percent Persons 0-18 Years		0.123				0.174 -0.020 0.094		
IFEM25 IFEM26 IFEM27	Female Interaction Of PAGE18LN Female Interaction Of PAGE18QU Female Interaction Of PAGE18CU		-0.150				-0.085 0.056 -0.110		
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years T: Cubic: Percent Persons 19-24 Years T: Quartic: Percent Persons 19-24 Years T: Quartic: Percent Persons 19-24 Years					0.069 -0.074		-0.019 -0.030 -0.027 -0.007	0.0 -0.0 -0.0
IHISP30 IHISP31 IHISP32 IHISP33	Hispanic Interaction Of PAGE24LN Hispanic Interaction Of PAGE24QU Hispanic Interaction Of PAGE24CU Hispanic Interaction Of PAGE24QR							0.157 0.085 -0.010 0.021	
IOTH30 IOTH31 IOTH32 IOTH33	Other Interaction Of PAGE24LN Other Interaction Of PAGE24QU Other Interaction Of PAGE24CU Other Interaction Of PAGE24QR								0.0 0.0 0.0 0.2
PAGE44LN	T: Linear: Percent Persons 35-44 Years							-0.023	
AGE54LN AGE54QU	T: Linear: Percent Persons 45-54 Years T: Ouadratic: Percent Persons 45-54 Years				0.029 0.001	0.028 0.064			0.0 -0.0
BLK45	Black Interaction Of PAGE54LN				-0.057	0.018			
BLK46 PAGE64LN PAGE64QU PAGE64CU	Black Interaction Of PAGE54QU T: Linear: Percent Persons 55-64 Years T: Quadratic: Percent Persons 55-64 Years T: Cubic: Percent Persons 55-64 Years				0.078	-0.238 0.069 0.022 0.010			
PAGE64QR	T: Quartic: Percent Persons 55-64 Years					0.031			
PSCH8LN OTH55	T: Linear: Percent 0-8 Years Of School Other Interaction Of PSCH8LN					-0.043			
PSCH12LN PSCH12QU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma					0.036 0.033			
PSCH12ČU PSCHASLN PSCHASQU PSCHASCU PSCHASQR	T: Čubic: Percent 9-12 Years & No High School Diploma T: Linear: Percent Associates Degree T: Quadratic: Percent Associates Degree T: Cubic: Percent Associates Degree T: Quartic: Percent Associates Degree					-0.045 0.034 0.043 -0.029 -0.063		0.010 0.003	
FEM65 FEM66 FEM67	Female Interaction Of PSCHASLN Female Interaction Of PSCHASQU Female Interaction Of PSCHASCU					0.075 -0.087 0.005			
FEM68 PSCHCOLN	Female Interaction Of PSCHASQR T: Linear: Bachelors, Graduate, Or Professional Degree					0.075 -0.103			-0.
PSCHCOQU FEM75 FEM76	T: Quadratic: Bachelors, Graduate, Or Professional Degree Female Interaction Of PSCHCOLN Female Interaction Of PSCHCOQU					-0.051 0.148 0.130			-0.0
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	T: Linear: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Quaric: Percent Families Below Poverty Level					0.130 0.031 0.047 -0.072 0.002			0.0
FEM85	Female Interaction Of PPOVERLN					-0.170			-0.2

NOTE: T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

			BIG C	ITY			REMAIN	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35-
IFEM88	Female Interaction Of PPOVERQR			•		-0.069	•		
IOTH85 IOTH86	Other Interaction Of PPOVERLN Other Interaction Of PPOVERQU Other Interaction Of PPOVERCU					-0.409 -0.275			
IOTH87 PPUBASLN	T: Linear: Percent Households With Public Assist Income					0.640 -0.131			
IFEM90	Female Interaction Of PPUBASLN					0.264			
P64DISLN	T: Linear: Percent 16-64 With A Work Disability	0.086				-0.051			
IOTH95 PBLACKLN	Other Interaction Of P64DISLN T: Linear: Percent Black Nonhispanic	-0.516		-0.025		0.704		-0.033	-0.02
PBLACKQU PBLACKCU	T: Quadratic: Percent Black Nonhispanic T: Cubic: Percent Black Nonhispanic			-0.018 0.023				-0.048 -0.031	-0.08
PHISPLN PHISPQU PHISPCU PHISPQR	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic T: Cubic: Percent Hispanic T: Quartic: Percent Hispanic					-0.067 -0.042 0.042			-0.10- -0.079 -0.02- 0.000
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity					-0.198 0.008 -0.005			0.04
IOTH110	Other Interaction Of POTHLN								-0.46
PHHF18LN PHHF18QU PHHF18CU PHHF18QR	T: Linear: % Female-Headed HH W/No Spouse & Chld Under 18 T: Quadratic: % Female-Headed HH W/No Spouse & Chld Under 18 T: Cubic: % Female-Headed HH W/No Spouse & Chld Under 18 T: Quartic: % Female-Headed HH W/No Spouse & Chld Under 18					-0.039 -0.000 0.036 0.002			0.05
IFEM120	Female Interaction Of PHHF18LN								0.05
PFLABLN PFLABQU PFLABCU PFLABQR	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force T: Cubic: Percent Females 16+ Years Old In Labor Force T: Quartic: Percent Females 16+ Years Old In Labor Force						0.047 0.003 0.050 0.018		
IBLK125 IBLK126 IBLK127 IBLK128	Black Interaction Of PFLABLN Black Interaction Of PFLABQU Black Interaction Of PFLABCU Black Interaction Of PFLABQR						0.007 0.033 -0.046 -0.056		
PFNEVLN PFNEVQU PFNEVCU PFNEVQR	T: Linear: Percent Females Never Married T: Quadratic: Percent Females Never Married T: Cubic: Percent Females Never Married T: Quartic: Percent Females Never Married T: Quartic: Percent Females Never Married								0.04 0.05 0.01 -0.01
IHISP130 IHISP131	Hispanic Interaction Of PFNEVLN Hispanic Interaction Of PFNEVQU								0.00
IHISP132 IHISP133	Hispanic Interaction Of PFNEVCU Hispanic Interaction Of PFNEVQR								-0.00 0.04
PFNOTLN PFNOTQU PFNOTCU PFNOTQR	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed T: Cubic: Percent Females Separated, Divorced Or Widowed T: Quartic: Percent Females Separated, Divorced Or Widowed		0.106	0.001			0.023 -0.004 0.015 -0.009	0.039 0.014 0.035 0.014	
IHISP135 IHISP136 IHISP137 IHISP138	Hispanic Interaction Of PFNOTLN Hispanic Interaction Of PFNOTQU Hispanic Interaction Of PFNOTCU Hispanic Interaction Of PFNOTQR							0.021 -0.014 -0.031 -0.038	
IOTH135 IOTH136 IOTH137 IOTH138	Other Interaction Of PFNOTLN Other Interaction Of PFNOTQU Other Interaction Of PFNOTICU Other Interaction Of PFNOTQR						0.089 0.103 -0.259 -0.145		
PMLABLN PMLABQU	T: Linear: Percent Males 16+ Years Old In Labor Force T: Ouadratic: Percent Males 16+ Years Old In Labor Force	i !				-0.027 -0.065	*****		
IOTH145	Other Interaction Of PMLABLN					-0.283			
IOTH146 PMNOTEN	Other Interaction Of PMLABQU T: Linear: Percent Males Separated, Divorced Or Widowed					0.465	0.015	0.090	
PMNOTQU IFEM150	T: Quadratic: Percent Males Separated, Divorced Or Widowed Female Interaction Of PMNOTEN						0.012 0.084	-0.005	
IFEM151	Female Interaction Of PMNOTQU Hispanic Interaction Of PMNOTLN						-0.069	0.110	
IHISP150 IHISP151	Hispanic Interaction Of PMNOTQU							-0.118 -0.020	
POLDHULN P40HULN	T: Linear: Percent Housing Units Built 1939 Or Earlier T: Linear: Percent Housing Units Built 1940-1949			-0.013	-0.044	0.017		-0.014	0.09
P40HUQU P40HUCU P40HUQR	T: Quadratic: Percent Housing Units Built 1940-1949 T: Cubic: Percent Housing Units Built 1940-1949 T: Quartic: Percent Housing Units Built 1940-1949			-0.008 -0.000		-0.014 -0.050		-0.012 0.002 -0.009	
IBLK160 IBLK161	Black Interaction Of P40HULN Black Interaction Of P40HUOU			0.031 -0.041				0.084 0.001	
IBLK162 IBLK163	Black Interaction Of P40HUCU Black Interaction Of P40HUCU Black Interaction Of P40HUQR			0.092				-0.029 -0.044	
IOTH160 IOTH161 IOTH162	Other Interaction Of P40HULN Other Interaction Of P40HUQU Other Interaction Of P40HUCU					-0.589 0.797 0.496			
PRENTLN PRENTQU PRENTCU PRENTQR	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented T: Cubic: Percent Housing Rented T: Quartic: Percent Housing Rented						0.060 0.016 0.046	-0.091	-0.0 0.00 -0.00 0.00
IOTH165 IOTH166	Other Interaction Of PRENTLN Other Interaction Of PRENTQU						0.039 0.329		
ADRATELN	C: Linear: Death Rate For All Alcohol-Related Cases) 				0.032			

NOTE: T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

			BIG C	ITY			REMAIN	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
ADRATILN ADRATIQU ADRATICU ADRATIQR	C: Linear: Death Rate With Explicit Mention Of Alcohol C: Quadratus: Death Rate With Explicit Mention Of Alcohol C: Cubic: Death Rate With Explicit Mention Of Alcohol C: Quartic: Death Rate With Explicit Mention Of Alcohol		-0.034 0.004 -0.003 0.000				0.011 0.019 -0.006 -0.021		0.030 0.035 0.072 0.027
IBLK180 IBLK181 IBLK182 IBLK183	Black Interaction Of ADRATILN Black Interaction Of ADRATIQU Black Interaction Of ADRATICU Black Interaction Of ADRATIQR		0.050 -0.039 0.054 -0.061				-0.097 0.108 -0.050 0.070		
V18FLN V18FQU V18FCU V18FQR	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate C: Cubic: Marijuana Posession Arrest Rate C: Quartic: Marijuana Posession Arrest Rate			-0.158		-0.017		-0.048 -0.008 -0.010 -0.026	
V18BLN	C: Linear: Marijuana Sale/Manufacture Arrest Rate				0.042			-0.017	
IBLK190 V18ELN	Black Interaction Of V18BLN				0.138		-0.010	0.189	-0.111
V18ELN V18EQU V18ECU V18EQR	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate C: Quartic: Opium/Cocaine & Deriv Posession Arrest Rate						0.010 0.001 -0.056		0.002 -0.006 -0.009
V18ALN	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate		0.026				0.074	0.010	0.132
IBLK200 IHISP200	Black Interaction Of V18ALN Hispanic Interaction Of V18ALN		-0.195						-0.168
V18LN	C: Linear: Total Drug Abuse Violations Arrest Rate		-0.016	-0.041			-0.065	-0.047	
V18QU V18CU V18QR	C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate			0.029 0.097 0.010			-0.020	0.047 0.067 0.011	
IFEM225 IFEM226 IFEM227 IFEM228	Female Interaction Of V18LN Female Interaction Of V18QU Female Interaction Of V18CU Female Interaction Of V18QR							0.095 -0.044 -0.044 -0.024	
IHISP225	Hispanic Interaction Of V18LN		0.061						
VIOLLN VIOLQU	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate				-0.055 -0.001				-0.060 -0.046
IBLK230 IBLK231	Black Interaction Of VIOLLN Black Interaction Of VIOLQU				-0.141 -0.050				0.173
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992					0.042	0.061 0.002		0.009 -0.005
IFEM235 IFEM236	Female Interaction Of DRATELN Female Interaction Of DRATEQU						-0.037 -0.029		
IOTH235 IOTH236	Other Interaction Of DRATELN Other Interaction Of DRATEQU								-0.134 -0.615
RH43ALN RH43AQU RH43ACU RH43AQR	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units T: Cubic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units	-0.015 0.055 -0.001 -0.001				-0.077 -0.057 -0.126		-0.054	
RP80ALN RP80AQU RP80ACU RP80AQR	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income T: Quartic: Recoded Median Household Income		0.056	-0.086		-0.056 -0.008 0.025 -0.027	0.003 -0.035 -0.025	-0.116	-0.016 0.026
IBLK250 IBLK251 IBLK252	Black Interaction Of RP80ALN Black Interaction Of RP80AQU Black Interaction Of RP80ACU						-0.060 -0.014 0.005		
IHISP250 IHISP251	Hispanic Interaction Of RP80ALN Hispanic Interaction Of RP80AQU		0.168						-0.034 -0.157
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992	0.056				0.038 0.031 -0.040			
B64DISLN B64DISQU B64DISCU B64DISQR	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability B: Quartic: Percent Persons 16-64 With A Work Disability			0.075	0.069	0.008 -0.065 -0.026 -0.025	0.089	0.088	
IFEM260 IFEM261 IFEM262 IFEM263	Female Interaction Of B64DISLN Female Interaction Of B64DISQU Female Interaction Of B64DISCU Female Interaction Of B64DISCU Female Interaction Of B64DISQR					0.022 0.078 0.071 0.071			
BAGE34LN BAGE34QU BAGE34CU	B: Linear: Percent Persons 25-34 Years B: Ouadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years				0.086	-0.005 0.058		0.038 -0.033 -0.018	0.116 0.028 0.066
IFEM265 IFEM266 IFEM267	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU			<u> </u>			<u> </u>	<u> </u>	-0.039 -0.035 -0.112
IBLK265	Black Interaction Of BAGE34LN								-0.130
IOTH265 IOTH266 IOTH267	Other Interaction Of BAGE34LN Other Interaction Of BAGE34QU Other Interaction Of BAGE34CU					0.953 -0.677		-0.322 0.352 0.445	
BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years							-0.027 0.047	
IHISP270 IHISP271	Hispanic Interaction Of BAGE44LN Hispanic Interaction Of BAGE44QU							0.073 -0.121	
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years					-0.017 0.003 0.039	0.001 -0.049 -0.099	-0.061 -0.023 -0.019 0.007	0.012 0.017 -0.017

NOTE:

			BIG CI	ITY			REMAIN	IDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IFEM275 IFEM276 IFEM277	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU						0.024 0.051 0.135		-0.001 0.021 0.008
IBLK275 IBLK276	Black Interaction Of BAGE54LN Black Interaction Of BAGE54QU						-0.005 0.013		
IOTH275 IOTH276 IOTH277 IOTH278	Other Interaction Of BAGE54LN Other Interaction Of BAGE54QU Other Interaction Of BAGE54CU Other Interaction Of BAGE54QR					0.404 -0.054 -0.754		0.057 -0.049 0.187 0.144	
BAGE64LN BAGE64QU BAGE64CU BAGE64QR	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years B: Quartic: Percent Persons 55-64 Years		-0.010 -0.043 0.007	0.004 0.008 -0.004	-0.084 0.014	-0.097	0.018 0.006 0.025 -0.008		-0.036 0.001 -0.021
IFEM280 IFEM281 IFEM282	Female Interaction OF BAGE64LN Female Interaction OF BAGE64QU Female Interaction OF BAGE64CU				0.028 -0.004				0.056 -0.026 -0.001
IBLK280 IBLK281	Black Interaction Of BAGE64LN Black Interaction Of BAGE64QU								-0.040 0.033
IHISP280 IHISP281 IHISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		0.003 0.007 -0.042	-0.063 -0.030 0.008					
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander		-0.040 -0.014	-0.005 0.008 -0.008 0.003			-0.002 0.052 0.085 0.012	-0.079 0.001 0.035 0.030	-0.078 0.010 -0.140
IFEM285 IFEM286 IFEM287 IFEM288	Female Interaction Of BASIANLN Female Interaction Of BASIANQU Female Interaction of BASIANCU Female Interaction Of BASIANQR						-0.033 -0.028 -0.055 -0.047		0.031 0.006 0.183
IHISP285 IHISP286 IHISP287 IHISP288	Hispanic Interaction Of BASIANLN Hispanic Interaction Of BASIANQU Hispanic Interaction Of BASIANCU Hispanic Interaction Of BASIANQR			-0.003 0.058 0.003 -0.041				0.197	
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban						-0.079 0.029 0.064	0.052	
IFEM290 BFNOTLN	Female Interaction Of BCUBANLN B: Linear: Percent Females Separated, Divorced or Widowed					0.025		-0.114 0.004	
BFNOTQU BFNOTCU IFEM295	B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed Female Interaction Of BFNOTLN					0.008 0.061 0.008		-0.095 0.026	
IFEM296 IFEM297	Female Interaction Of BFNOTQU Female Interaction Of BFNOTQU Female Interaction Of BFNOTCU					-0.045 -0.176		0.085	
BINDIALN BINDIAQU BINDIACU BINDIAQR	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut B: Quartic: Percent Pop: American Indian, Eskimo, Ale		0.061	-0.003 0.024			0.057 -0.014 0.004 0.006		
IHISP300	Black Interaction Of BINDIALN Hispanic Interaction Of BINDIALN			-0.120			0.222		
IHISP301 IOTH300 IOTH301	Hispanic Interaction Of BINDIAQU Other Interaction Of BINDIALN Other Interaction Of BINDIAQU			0.168			0.440 -0.426		
BMNOTLN BMNOTQU BMNOTCU BMNOTQR	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed B: Quartic: Percent Males Separated, Divorced or Widowed			0.041 0.016 -0.014 -0.011	0.132 0.030 0.070	0.024 -0.111	0.058 0.000 0.023		0.067 -0.025
IFEM305 IFEM306 IFEM307	Female Interaction Of BMNOTLN Female Interaction Of BMNOTQU Female Interaction Of BMNOTCU			0.011	-0.146 -0.060 -0.011	-0.030 0.157			
IHISP305 IHISP306 IHISP307	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU Hispanic Interaction Of BMNOTCU			-0.011 -0.011 0.028 0.041	-0.011				-0.048 0.171
BPOVERLN BPOVERQU	Hispanic Interaction Of BMNOTQR B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level		0.105 0.025	0.041		0.006 0.015	-0.035 -0.017		0.089 -0.030
BPOVERČU BPOVERQR IFEM310	B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level Female Interaction Of BPOVERLN		0.040 0.008			-0.039	-0.045 0.004		0.054
IFEM311 IFEM312	Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU								-0.012 -0.126
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican	0.040				-0.082 0.046 -0.189		0.097 -0.009 -0.050 -0.039	
IFEM315 IBLK315 IBL K316	Female Interaction Of BPRICALN Black Interaction Of BPRICALN Black Interaction Of BPRICAQU					0.128		-0.177	
IBLK316 IBLK317 IBLK318	Black Interaction Of BPRICACU Black Interaction Of BPRICAQR							0.094 0.026 0.080	
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree		0.031 0.047			0.019 -0.051	0.048 0.004 0.016 -0.014	-0.003 -0.023 0.007	0.011 0.042 0.023
IFEM320 IFEM321	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU					-0.157 0.094	-0.072		

			BIG CI	ITY			REMAIN	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IBLK320 IBLK321 IBLK322	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU							-0.023 0.116 -0.003	
IHISP320 IHISP321	Hispanic Interaction Of BSCHASLN Hispanic Interaction Of BSCHASQU		-0.075 -0.088						
IOTH320 IOTH321	Other Interaction Of BSCHASLN Other Interaction Of BSCHASQU								-0.130 0.569
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander I: Quartic: Percent Population: Asian, Pacific Islander			-0.004 -0.016		0.102 0.006 0.005	-0.018 -0.014 -0.035 -0.019		
IBLK325 IBLK326	Black Interaction Of PASIANLN Black Interaction Of PASIANQU			0.073 0.101					
IOTH325 IOTH326 IOTH327	Other Interaction Of PASIANLN Other Interaction Of PASIANQU Other Interaction Of PASIANCU					0.286 -0.117 0.575			
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban	0.011 0.073 -0.065		0.016 0.035 0.013 0.027	-0.179	0.009 0.093 -0.099	0.224	0.051 0.010 0.024 -0.005	
IFEM330 IFEM331 IFEM332	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU	0.067 -0.219				-0.096 -0.052 0.300			
IBLK330 IBLK331 IBLK332 IBLK333	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU Black Interaction Of PCUBANQR	-0.319 -0.058 0.128					-0.278	0.043 -0.133 0.073 0.068	
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Aleut			0.049		0.018		0.025 -0.004 -0.039 0.008	-0.015 -0.013 0.005 0.026
IFEM335 IFEM336 IFEM337 IFEM338	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU Female Interaction Of PINDIACU Female Interaction Of PINDIAQR							-0.007 0.001 0.016 -0.042	
IBLK335 IBLK336	Black Interaction Of PINDIALN Black Interaction Of PINDIAQU					0.457		0.082 0.125	
IHISP335 IHISP336 IHISP337 IHISP338	Hispanic Interaction Of PINDIALN Hispanic Interaction Of PINDIAQU Hispanic Interaction Of PINDIACU Hispanic Interaction Of PINDIAQR							-0.041 0.050 -0.016 0.068	
ЮТН335	Other Interaction Of PINDIALN							0.318	

E4. Coefficients of Model Parameters for Past Month Cocaine Use

			BIG C	ITY			REMA	INDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	-6.184	-3.510	-3.721	-5.496	-6.529	-4.421	-5.037	-7.819
FEMALE	O: Female Indicator	0.941	-0.810	-0.624	-1.825	-0.262	-0.811	-0.335	-0.100
FEMBLCK FEMHISP	O: Black Interaction Of FEMALE	-1.977	0.086 0.197	0.319 -0.208	0.875 0.432	0.246 0.330	-0.386 -0.419	-0.538	-2.219
RACEBLCK RACEHISP RACEOTHR	O: Hispanic Interaction Of FEMALE O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	-0.554 0.539 0.992 -0.937	-0.182 -0.703 -2.396	-0.175 -0.488	1.287 0.850	-0.530 1.358	-0.419 -0.286 0.247 -0.246	-0.621 -0.215 0.797 -0.051	-1.322 2.995 1.442
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: Small MSA	0.55,	2.370				0.186 0.285 0.538 0.856	0.133 0.277 -0.319 -0.344	
POPRMLN	T: Linear: Average Persons Per Room			-0.141				-0.439	
IFEM20	Female Interaction Of POPRMLN							0.529	
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Lunear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years T: Cubic: Percent Persons 19-24 Years T: Quartic: Percent Persons 19-24 Years							0.091 -0.002 -0.223 0.077	
IHISP30 IHISP31 IHISP32 IHISP33	Hispanic Interaction Of PAGE24LN Hispanic Interaction Of PAGE24QU Hispanic Interaction Of PAGE24CU Hispanic Interaction Of PAGE24QR							0.114 -0.010 0.344 -0.154	
PAGE44LN	T: Linear: Percent Persons 35-44 Years							0.250	
PAGE54LN PAGE54QU	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years				-0.128 -0.160				-0.065 -0.315
IBLK45 IBLK46	Black Interaction Of PAGE54LN Black Interaction Of PAGE54QU				0.103 0.325				
PSCHASLN PSCHASQU	T: Linear: Percent Associates Degree T: Quadratic: Percent Associates Degree							-0.042 -0.185	
PSCHCOLN PSCHCOQU	T: Linear: Bachelors, Graduate, Or Professional Degree T: Quadratic: Bachelors, Graduate, Or Professional Degree					-0.169 0.356			-0.117 -0.290
IFEM75 IFEM76	Female Interaction Of PSCHCOLN Female Interaction Of PSCHCOQU					0.606 -0.710			
PBLACKLN PBLACKQU PBLACKCU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic T: Cubic: Percent Black Nonhispanic			0.155 0.035 0.107				0.116 0.052 0.189	
PHISPLN PHISPQU PHISPCU PHISPQR	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic T: Cubic: Percent Hispanic T: Quartic: Percent Hispanic								0.128 -0.112 0.063 0.140
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity					-0.046 -0.165 0.239			
PHHF18LN PHHF18QU PHHF18CU PHHF18QR	T: Lımear: % Female-Headed HH W/No Spouse & Chid Under 18 T: Quadratic: % Female-Headed HH W/No Spouse & Chid Under 18 T: Cubic: % Female-Headed HH W/No Spouse & Chid Under 18 T: Quartic: % Female-Headed HH W/No Spouse & Chid Under 18					0.252 -0.070 0.140 0.129			-0.498
IFEM120	Female Interaction Of PHHF18LN								0.979
PFNOTLN PFNOTQU	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed			0.172				0.176 0.191	
IHISP135 IHISP136	Hispanic Interaction Of PFNOTLN Hispanic Interaction Of PFNOTQU							-0.322 -0.381	
PMNOTLN PMNOTQU	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed						-0.178 0.171	-0.066 -0.121	
IFEM150 IFEM151	Female Interaction Of PMNOTLN Female Interaction Of PMNOTQU						0.550 -0.299		
IHISP150 IHISP151	Hispanic Interaction Of PMNOTLN Hispanic Interaction Of PMNOTQU							-0.053 0.285	
PRENTLN PRENTQU PRENTCU PRENTQR	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented 1: Cubic: Percent Housing Rented T: Quartic: Percent Housing Rented								0.586 0.155 0.274 -0.136
V18BLN V18BQU V18BCU	C: Linear: Marijuana Sale/Manufacture Arrest Rate C: Quadratıc: Marijuana Sale/Manufacture Arrest Rate C: Cubic: Marijuana Sale/Manufacture Arrest Rate							0.192 0.028 0.288	
V18ELN V18EQU V18ECU V18EQR	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratıc: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate C: Quartic: Opium/Cocaine & Deriv Posession Arrest Rate								0.250 0.446 0.088 0.165
V18ALN	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate		0.240					0.272	
IHISP200	Hispanic Interaction Of V18ALN		-0.361	0.073			(1) (2) (2)	0.703	
V18LN V18QU V18CU V18QR	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate		-0.279	0.072 -0.004 -0.106 0.012			0.028 -0.072	-0.691 0.018 -0.069 0.140	

			BIG C	ITY			REMA	INDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IFEM225 IFEM226 IFEM227	Female Interaction Of V18LN Female Interaction Of V18QU Female Interaction Of V18CU							0.161 -0.039 0.174	
IFEM228 IHISP225	Female Interaction Of V18QR Hispanic Interaction Of V18LN		0.432					-0.132	
VIOLLN VIOLQU	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate				0.164 -0.116				-0.310 -0.551
IBLK230 IBLK231	Black Interaction Of VIOLLN Black Interaction Of VIOLOU				-0.331 0.208				0.551
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992				0.200		0.046 -0.171		
IFEM235 IFEM236	Female Interaction Of DRATELN Female Interaction Of DRATEQU						-0.048 0.334		
RH43ALN RH43AQU RH43ACU RH43AQR	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units T: Cubic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units	-0.068 0.052 -0.178 -0.083						0.489	
RP80ALN RP80AQU RP80ACU RP80AOR	T: Linear: Recoded Median Household Income T: Quadratus: Recoded Median Household Income T: Cubis: Recoded Median Household Income T: Quartis: Recoded Median Household Income T: Quartis: Recoded Median Household Income	0.003				-0.026 0.190 0.014 -0.054	-0.027 -0.071 -0.183	-0.686	
IBLK250 IBLK251 IBLK252	Black Interaction Of RP80ALN Black Interaction Of RP80AQU Black Interaction Of RP80ACU					0.054	-0.280 0.175 0.438		
ARATELN BAGE54LN	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 B: Linear: Percent Persons 45-54 Years	0.339					-0.016		-0.270
BAGE54QU BAGE54CU	B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years						-0.194		-0.154 0.163
IFEM275 IFEM276 IFEM277	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU								0.578 0.417 -0.840
IBLK275 IBLK276	Black Interaction Of BAGE54LN Black Interaction Of BAGE54LU						0.013 0.438		-0.840
BAGE64LN BAGE64QU BAGE64CU	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years		-0.089 -0.100 0.039	-0.020 0.005 0.058	-0.156 0.195		0.176 -0.032 0.110		-0.004 -0.579 0.204
BAGE64QR IFEM280 IFEM281	B: Quartic: Percent Persons 55-64 Years Female Interaction Of BAGE64LN Female Interaction Of BAGE64QU				0.027 -0.272		-0.003		-0.343 0.491
IFEM282 IBLK280	Female Interaction Of BAGE64ČU Black Interaction Of BAGE64LN				V.=,=				-0.629 0.208
IBLK281 IHISP280	Black Interaction Of BAGE64QU Hispanic Interaction Of BAGE64LN		-0.002	-0.110					0.628
IHISP281 IHISP282	Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64QU		0.057 -0.352	0.047 -0.415					
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander		-0.037 -0.115				0.074 0.088 -0.022 -0.083		
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban						0.028 0.071 -0.348	-0.605	
IFEM290	Female Interaction Of BCUBANLN							0.686	
BMNOTLN BMNOTQU BMNOTCU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed				0.373 -0.187 0.303	0.250 0.269	0.124 -0.062 0.120		
IFEM305 IFEM306 IFEM307	Female Interaction Of BMNOTLN Female Interaction Of BMNOTQU Female Interaction Of BMNOTCU				-0.394 0.159 -0.380				
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.172 -0.050 -0.050 -0.049				0.077 0.022 -0.070 -0.055		
BPRICALN BPRICAQU BPRICACU	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican	0.165				0.072		-0.248 0.174 -0.296	
IFEM315 IBLK315	Female Interaction Of BPRICALN Black Interaction Of BPRICALN					-0.388		0.028	
IBLK316 IBLK317	Black Interaction Of BPRICAQU Black Interaction Of BPRICACU							-0.352 1.067	
BSCHASLN BSCHASQU BSCHASCU	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree		-0.083 0.106			0.181 -0.336	0.120 0.029	-0.167 -0.038 0.155	0.052 0.033 -0.370
IFEM320 IFEM321	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU					-0.243 0.552	-0.369		
IBLK320 IBLK321 IBLK322	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU							-0.174 0.164 -0.379	
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispamics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban	0.284 -0.227 -0.144		-0.147 0.102				0.143 -0.083 0.172 0.126	
IBLK330	Black Interaction Of PCUBANLN	0.256			ļ			-	

			BIG CITY 12-17 18-25 26-34 35+				REMAINDER					
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+			
IBLK331 IBLK332	Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU	-0.567 0.598										

E5. Coefficients of Model Parameters for Past Month Any Illicit Drug But Marijuana

			BIG C	ITY			REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	-3.502	-2.451	-2.988	-4.111	-3.613	-2.902	-3.810	-5.29
FEMALE	O: Female Indicator	0.173	-0.545	-0.343	-0.442	-0.199	-0.457	-0.271	0.05
FEMBLCK	O: Black Interaction Of FEMALE	-0.547	-0.150	0.136	0.267	0.696	0.263	-0.345	-0.58
FEMHISP FEMOTHR	O: Hispanic Interaction Of FEMALE O: Other Interaction Of FEMALE	0.093	0.358 1.261	0.370 -0.015	-0.239	-0.670	-0.001 -0.526	-0.035	0.08 1.07
RACEBLCK	O: Race/Black Indicator	-0.409	-0.732	-0.371	0.225	-0.914	-0.497	-0.313	1.15
RACEHISP RACEOTHR	O: Race/Hispanic Indicator O: Race/Other Indicator	-0.022 -1.119	-0.717 -2.089	-0.549 -1.395	0.615	-0.929	-0.320 -3.417	-0.227 -1.208	-0.30 -0.81
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator						0.606	0.540 0.380 1.056	
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban								0.37 -0.03 1.14 0.02
PHH1PLN	T: Linear: Percent One Person Households				İ	0.222			
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years T: Cubic: Percent Persons 19-24 Years T: Ouartic: Percent Persons 19-24 Years T: Ouartic: Percent Persons 19-24 Years							0.087 -0.021 -0.121 0.045	-0.12 0.12
PAGE34LN PAGE34QU	T: Linear: Percent Persons 25-34 Years T: Quadratic: Percent Persons 25-34 Years				į	-0.080 -0.168			
PAGE44LN PAGE44QU	T: Linear: Percent Persons 35-44 Years T: Ouddratic: Percent Persons 35-44 Years					0.037 -0.002			
PAGE54LN PAGE54QU	T: Linear: Percent Persons 45-54 Years T: Ouddratic: Percent Persons 45-54 Years					-0.002			-0.11 -0.34
IFEM45	Female Interaction Of PAGE54LN								0.01
IFEM46 PSCH8LN	Female Interaction Of PAGE54QU T: Linear: Percent 0-8 Years Of School				:			-0.171	0.32
PSCH12LN	T: Linear: Percent 9-12 Years & No High School Diploma	0.042			ŀ	0.103		0.171	
PSCH12QU PSCH12CU PSCH12QR	T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma T: Quartic: Percent 9-12 Years & No High School Diploma	0.104 0.005 0.052				0.101 0.010 0.040			
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU	-0.122 -0.183			ŀ	-0.166 -0.187			
PSCHASLN PSCHASQU	T: Linear: Percent Associates Degree T: Quadratic: Percent Associates Degree				į	0.065 -0.081			
IBLK65 IBLK66	Black Interaction Of PSCHASLN Black Interaction Of PSCHASQU				ŀ	0.228 0.370			
PSCHSCLN PSCHSCQU	T: Linear: Percent Some College And No Degree T: Quadratic: Percent Some College And No Degree					0.012 0.106			
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Quartic: Percent Families Below Poverty Level							0.074 -0.037 -0.003 -0.051	
P64DISLN P64DISQU P64DISCU	T: Linear: Percent 16-64 With A Work Disability T: Quadratic: Percent 16-64 With A Work Disability T: Cubic: Percent 16-64 With A Work Disability	0.073 0.057 0.084				0.126 0.085 0.075			
IBLK95 IBLK96 IBLK97	Black Interaction Of P64DISLN Black Interaction Of P64DISQU Black Interaction Of P64DISCU	-0.215 0.007 -0.296				0.094 0.252 -0.579			
PBLACKLN PBLACKQU PBLACKCU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic T: Cubic: Percent Black Nonhispanic			0.064 0.004 0.078				0.129 -0.015 0.125	
PFNOTLN	T: Linear: Percent Females Separated, Divorced Or Widowed			0.168	ŀ			0.228	
PMNEVLN	T: Linear: Percent Males Never Married				ŀ	-0.177			
PMNOTEN PMNOTQU PMNOTCU PMNOTQR	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed T: Cubic: Percent Males Separated, Divorced Or Widowed T: Quartic: Percent Males Separated, Divorced Or Widowed					0.015 0.004 0.035 -0.036			
P40HULN P40HUQU P40HUCU P40HUQR	T: Linear: Percent Housing Units Built 1940-1949 T: Quadratic: Percent Housing Units Built 1940-1949 T: Cubic: Percent Housing Units Built 1940-1949 T: Quartic: Percent Housing Units Built 1940-1949					-0.145		0.123 -0.039 -0.077 -0.044	
PRENTLN PRENTQU	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented							-0.084 0.111	0.35
ADRATEQR	C: Quartic: Death Rate For All Alcohol-Related Cases				[0.048		
ADRAT1LN	C: Linear: Death Rate With Explicit Mention Of Alcohol				ŀ		-0.176		
V18FLN V18FQU V18FCU V18FQR	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate C: Cubic: Marijuana Posession Arrest Rate C: Quartic: Marijuana Posession Arrest Rate					0.092 -0.081		-0.119 -0.035 -0.083 -0.053	
IHISP185	Hispanic Interaction Of V18FLN				ľ			0.246	
V18BLN	C: Linear: Marijuana Sale/Manufacture Arrest Rate			.			.		-0.0 -0.1

			BIG C	ITY			REMAI	NDER		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+	
V18ALN V18AQU V18ACU V18AQR	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quadratie: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Cubie: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quartic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate					-0.026 0.035 -0.119 -0.051				
V18CU	C: Cubic: Total Drug Abuse Violations Arrest Rate		-0.184				0.121			
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992					0.042	0.131 -0.117			
B64DISLN B64DISQU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Linear: Percent Persons 25-34 Years					-0.062 -0.078	0.150	0.101		
BAGE34LN BAGE34QU BAGE34CU	B: Linear: Percent Persons 25-34 Years B: Quadratuc: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years					0.008 -0.035 -0.055	0.158	0.191 -0.113 0.104		
IFEM265 IFEM266 IFEM267	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU							-0.142 0.112 -0.186		
IOTH265 IOTH266 IOTH267	Other Interaction Of BAGE34LN Other Interaction Of BAGE34QU Other Interaction Of BAGE34CU					-0.576 0.798 1.010		-1.421		
BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years					-0.182	-0.028 0.066	-0.004 -0.046		
IBLK270 IBLK271	Black Interaction Of BAGE44LN Black Interaction Of BAGE44QU							0.028 0.259		
IOTH270 IOTH271	Other Interaction Of BAGE44LN Other Interaction Of BAGE44QU						0.272 1.219			
BAGE54LN BAGE54QU	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years							-0.031 -0.102		
BAGE64LN BAGE64QU BAGE64CU BAGE64QR	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years B: Quartic: Percent Persons 55-64 Years		-0.098 -0.062 0.045	-0.023 0.002 0.083		-0.143	0.138 -0.055 0.107 -0.032			
IHISP280 IHISP281 IHISP282 IHISP283	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU Hispanic Interaction Of BAGE64QR		0.114 0.001 -0.233	0.041 -0.100 -0.250			0.144 0.077 0.095 0.100			
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander						0.012 0.099 -0.034 -0.046	-0.050 0.167		
IOTH285	Other Interaction Of BASIANLN						2.215			
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban				-0.071 -0.117	0.260 0.264 -0.155				
IFEM290 IFEM291 IFEM292	Female Interaction Of BCUBANLN Female Interaction Of BCUBANQU Female Interaction Of BCUBANCU					-0.070 0.008 0.408				
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed	-0.038 -0.014 -0.008 0.035				0.004 0.017 -0.056 0.024	-0.032 -0.040 0.035			
IFEM295 IFEM296 IFEM297	Female Interaction Of BFNOTLN Female Interaction Of BFNOTQU Female Interaction Of BFNOTCU						0.117 0.028 -0.190			
IBLK295 IBLK296 IBLK297 IBLK298	Black Interaction Of BFNOTLN Black Interaction Of BFNOTQU Black Interaction Of BFNOTCU Black Interaction Of BFNOTQR					0.102 -0.032 0.219 -0.118	-0.024 0.298			
IHISP295	Hispanic Interaction Of BFNOTLN					0.110	-0.244			
BINDIALN BINDIAQU BINDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut					0.095 -0.056 0.215				
BMNOTEN BMNOTQU BMNOTCU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed				0.164				-0.018 0.062 0.231	
IFEM305	Female Interaction Of BMNOTLN				-0.351				0.231	
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.092 0.043 -0.033 -0.027				-0.072 -0.132 -0.151 -0.032			
IFEM310 IFEM311 IFEM312	Female Interaction Of BPOVERLN Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU		0.061 -0.108 0.191				0.188 0.129 0.204			
IHISP310 IHISP311	Hispanic Interaction Of BPOVERLN Hispanic Interaction Of BPOVERQU		-0.280				-0.015 0.246			
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican	-0.161				-0.206 -0.246 -0.160 0.022		-0.087 0.162 -0.224		
IFEM315 IFEM316 IFEM317 IFEM318	Female Interaction OF BPRICALN Female Interaction OF BPRICAQU Female Interaction OF BPRICACU Female Interaction OF BPRICAQR					-0.027 0.444 0.118 -0.269		0.225 -0.100 0.379		
BSCHASLN BSCHASQU	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree	0.020 0.026	-0.092			-0.040 -0.137	-0.004 0.077	-0.080 0.005		

			BIG CI	TY		REMAINDER					
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+		
BSCHASCU BSCHASQR	B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree	0.072 0.041	•	•		-0.089 -0.038		0.119			
IFEM320 IFEM321 IFEM322 IFEM323	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU Female Interaction Of BSCHASCU Female Interaction Of BSCHASQR					-0.033 0.257 0.190 0.061					
PASIANLN PASIANQU PASIANCU	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander						-0.004 -0.036 -0.110	0.039 -0.046 -0.124	0.215		
IHISP325 IHISP326 IHISP327	Hispanic Interaction Of PASIANLN Hispanic Interaction Of PASIANQU Hispanic Interaction Of PASIANCU						-0.161 0.093 0.209				
IOTH325	Other Interaction Of PASIANLN						-1.705				
PINDIALN	T: Linear: Percent Pop: American Indian, Eskimo, Aleut						-0.056	0.116			
IHISP335	Hispanic Interaction Of PINDIALN							-0.325			
ЮТН335	Other Interaction Of PINDIALN				:		1.008				

E6. Coefficients of Model Parameters for Past Year Alcohol Treatment

			BIG C	ITY		REMAINDER				
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+	
DUMMY	O: Intercept Term	-6.554	-5.328	-4.383	-5.549	-6.694	-6.307	-4.376	-7.191	
FEMALE	O: Female Indicator	0.006	-1.113	-0.878	-0.744	-0.111	-1.695	-1.074	-1.606	
FEMBLCK FEMHISP	O: Black Interaction Of FEMALE O: Hispanic Interaction Of FEMALE	0.075 0.658	-1.013	0.354 -0.610	-1.245 -1.864	0.372	0.461 -1.665	-0.522 -1.319	-1.905 -1.947	
RACEBLCK RACEHISP	O: Race/Black Indicator O: Race/Hispanic Indicator	-0.946 -0.130	-1.951 -0.067	-1.432 -2.533	-1.170 -1.033	-1.783 -0.244	-1.335 -0.632	-0.602 -0.649	0.362 -3.137	
REGNOREA REGWEST	O: Northeast Region Indicator O: West Region Indicator						-0.058 0.079			
PAGE18LN PAGE18QU PAGE18CU PAGE18QR	T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years T: Cubic: Percent Persons 0-18 Years T: Quartic: Percent Persons 0-18 Years T: Quartic: Percent Persons 0-18 Years		0.513				-0.254 -0.284 0.085 -0.071			
IHISP25 IHISP26 IHISP27 IHISP28	Hispanic Interaction Of PAGE18LN Hispanic Interaction Of PAGE18QU Hispanic Interaction Of PAGE18CU Hispanic Interaction Of PAGE18QR						0.037 0.330 0.272 0.225			
PSCH12LN PSCH12QU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma		-0.486 -0.404				0.128 -0.801			
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU						-0.170 0.760			
IHISP60 IHISP61	Hispanic Interaction Of PSCH12LN Hispanic Interaction Of PSCH12QU						0.270 0.727			
PSCHCOLN PSCHCOQU PSCHCOCU PSCHCOQR	T: Linear: Bachelors, Graduate, Or Professional Degree T: Quadratic: Bachelors, Graduate, Or Professional Degree T: Cubic: Bachelors, Graduate, Or Professional Degree T: Quartic: Bachelors, Graduate, Or Professional Degree							-0.219 -0.192	0.172 -0.006 0.388 0.081	
PSCHSCLN PSCHSCQU PSCHSCCU	T: Linear: Percent Some College And No Degree T: Quadratic: Percent Some College And No Degree T: Cubic: Percent Some College And No Degree	0.548 -0.093				-0.093 0.180 -0.177				
IHISP80 IHISP81	Hispanic Interaction Of PSCHSCLN Hispanic Interaction Of PSCHSCQU	-0.393 0.491								
PPOVERLN PPOVERQU PPOVERCU	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level	-0.246 -0.410				0.255 -0.247 -0.473				
PPUBASLN PPUBASQU PPUBASCU	T: Linear: Percent Households With Public Assist Income T: Quadratic: Percent Households With Public Assist Income T: Cubic: Percent Households With Public Assist Income		-0.549 0.118 0.288				0.317 0.185			
IHISP90 IHISP91	Hispanic Interaction Of PPUBASLN Hispanic Interaction Of PPUBASQU		0.603				-0.253 -0.712			
PBLACKLN PBLACKQU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic								-0.066 -0.303	
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity						0.062 -0.087 -0.235			
IFEM110 IFEM111 IFEM112	Female Interaction Of POTHLN Female Interaction Of POTHQU Female Interaction Of POTHCU						0.249 0.146 0.689			
PFNEVLN	T: Linear: Percent Females Never Married						0.632			
PMNEVLN PMNEVQU PMNEVCU	T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married T: Cubic: Percent Males Never Married		0.407		-0.026		-0.566 -0.213 0.162		0.288 0.037 -0.329	
IBLK140 IBLK141 IBLK142	Black Interaction Of PMNEVLN Black Interaction Of PMNEVQU Black Interaction Of PMNEVCU						0.028 0.135 -0.758		0.120 -0.126 0.873	
IHISP140	Hispanic Interaction Of PMNEVLN		-0.737		-0.815		-0.633		-1.569	
PMLABLN PMLABQU PMLABCU PMLABQR	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force T: Cubic: Percent Males 16+ Years Old In Labor Force T: Quartic: Percent Males 16+ Years Old In Labor Force				-0.011 0.209	-0.003 0.053 0.038 -0.108			0.355 0.060 0.033	
IFEM145 IFEM146 IFEM147	Female Interaction Of PMLABLN Female Interaction Of PMLABQU Female Interaction Of PMLABCU				0.336 -0.559				-0.023 0.248 0.920	
POLDHULN POLDHUQU POLDHUCU	T: Linear: Percent Housing Units Built 1939 Or Earlier T: Quadratic: Percent Housing Units Built 1939 Or Earlier T: Cubic: Percent Housing Units Built 1939 Or Earlier				0.230	0.133 -0.229 -0.156				
IFEM155 IFEM156 IFEM157	Female Interaction Of POLDHULN Female Interaction Of POLDHUQU Female Interaction Of POLDHUCU				-0.578	-0.255 0.317 0.480				
P40HULN	T: Linear: Percent Housing Units Built 1940-1949								0.320	
ADRATILN ADRATIQU ADRATICU ADRATIQR	C: Linear: Death Rate With Explicit Mention Of Alcohol C: Quadratic: Death Rate With Explicit Mention Of Alcohol C: Cubic: Death Rate With Explicit Mention Of Alcohol C: Quartic: Death Rate With Explicit Mention Of Alcohol			0.026 -0.078 0.134 -0.095			-0.371 -0.360 -0.551			
V18FLN V18FQU	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate						-0.416 -0.311			
IBLK185 IBLK186	Black Interaction Of V18FLN Black Interaction Of V18FQU	1					0.381 0.519			

			BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+	
V18ELN V18EQU	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate						-0.129 -0.033			
IFEM195 IFEM196	Female Interaction Of V18ELN Female Interaction Of V18EQU						-0.959 -1.036			
IBLK195 IBLK198	Black Interaction Of V18ELN Black Interaction Of V18EQR					-0.345 0.158				
V18LN V18QU V18CU	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate					0.126 -0.006 0.379				
IFEM225 IFEM226 IFEM227	Female Interaction Of V18LN Female Interaction Of V18QU Female Interaction Of V18CU					-0.047 0.024 -0.584				
VIOLLN VIOLQU VIOLCU VIOLQR	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate C: Cubic: Total Violent Offenses Arrest Rate C: Quartic: Total Violent Offenses Arrest Rate					0.006 -0.033 0.235 0.109				
RH43ALN RH43AQU	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units				0.295 0.335				-0.162 0.288	
RH61ALN	T: Linear: Recoded Median Value Of Owner Occupied HUs				-0.451					
IBLK245 RP80ALN RP80AQU RP80ACU	Black Interaction Of RH61ALN T: Linear: Recoded Median Household Income 1: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income				1.166	0.374 0.072 -0.295	-0.221 0.206			
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992			0.095 0.168 -0.250		-0.243		-0.037 -0.097 0.184		
IHISP255 IHISP256 IHISP257	Hispanic Interaction Of ARATELN Hispanic Interaction Of ARATEQU Hispanic Interaction Of ARATECU			2.239 -1.707 1.671				0.539 -0.108 -0.715		
B64DISLN B64DISQU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability						-0.340 -0.101			
IBLK260 IBLK261	Black Interaction Of B64DISLN Black Interaction Of B64DISQU						-0.190 0.785			
BAGE34LN BAGE34QU BAGE34CU BAGE34QR	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years				0.207 0.161 -0.332	-0.086 -0.179			0.037 -0.236 0.450 0.021	
IFEM265 IFEM266 IFEM267 IFEM268	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU Female Interaction Of BAGE34QR				-0.263 -0.233 0.586				-0.568 -0.052 -0.659 -0.296	
IBLK265 IBLK266	Black Interaction Of BAGE34LN Black Interaction Of BAGE34QU					-0.087 0.759				
IHISP265 IHISP266 IHISP267	Hispanic Interaction Of BAGE34LN Hispanic Interaction Of BAGE34QU Hispanic Interaction Of BAGE34CU				-0.087 -0.729				0.142 -1.552 -0.894	
BAGE44LN BAGE44QU BAGE44CU BAGE44QR	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years B: Quartic: Percent Persons 35-44 Years						-0.175 -0.022 0.029 0.086			
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years					-0.436 -0.077 0.210 -0.066		-0.076 0.225		
IFEM275 IFEM276 IFEM277 IFEM278	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU Female Interaction Of BAGE54QR					0.439 0.334 0.001 0.219				
BAGE64LN BAGE64QU BAGE64CU BAGE64QR	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years B: Quartic: Percent Persons 55-64 Years				-0.223				0.295 -0.046 0.162 0.053	
IBLK280 IBLK282 IBLK283	Black Interaction Of BAGE64LN Black Interaction Of BAGE64CU Black Interaction Of BAGE64QR				0.817				-0.567 0.128 -0.292	
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population; Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander	0.242 0.177 -0.121 0.096				-0.029 0.252 -0.324 -0.198			0.151 0.466	
IFEM285 IFEM286 IFEM287 IFEM288	Female Interaction Of BASIANLN Female Interaction Of BASIANQU Female Interaction Of BASIANCU Female Interaction Of BASIANQR	-0.474 -0.248 -0.010 -0.210				-0.250 -0.104 0.071 0.402				
IHISP285 IHISP286	Hispanic Interaction Of BASIANLN Hispanic Interaction Of BASIANQU								-0.555 -0.735	
BCUBANLN BCUBANQU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban				0.251				-0.198 -0.443	
IHISP290 IHISP291	Hispanic Interaction Of BCUBANLN Hispanic Interaction Of BCUBANQU		·		-1.268				-0.680 1.153	
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed				0.238	-0.048 -0.054 -0.396			0.092 -0.006 0.135 0.125	

			BIG CI	TY			REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BINDIALN BINDIAQU BINDIACU BINDIAQR	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut B: Quartic: Percent Pop: American Indian, Eskimo, Aleut			0.045 -0.052	0.001 -0.167 -0.112 -0.123			-0.165 -0.155	0.624 -0.291 0.509
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU			0.070 0.395				-0.014 0.614	
BMNOTLN BMNOTQU BMNOTCU BMNOTQR	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed B: Quartic: Percent Males Separated, Divorced or Widowed	0.082 -0.209				-0.041 -0.175 0.221 -0.078			
IFEM305 IFEM306 IFEM307 IFEM308	Female Interaction Of BMNOTLN Female Interaction Of BMNOTQU Female Interaction Of BMNOTCU Female Interaction Of BMNOTQR	-0.304 0.447				0.115 0.231 -0.008 0.219			
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level	0.209 0.070 -0.022 -0.088						0.049 -0.024 -0.236	
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican							-0.128 0.098 -0.171 -0.082	0.100
IHISP315 IHISP316 IHISP317 IHISP318	Hispanic Interaction Of BPRICALN Hispanic Interaction Of BPRICAQU Hispanic Interaction Of BPRICACU Hispanic Interaction Of BPRICAQR							0.033 -0.564 0.464 0.201	-0.866
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree B: Quartic: Percent Associates Degree				0.214 -0.082 -0.092 0.090				0.046 0.324 -0.289 -0.143
IBLK320 IBLK321	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU				-1.164				-0.180 -1.163
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander				0.284			-0.276 -0.235	-0.328 -0.431 -0.076 0.140
IFEM325 IFEM326	Female Interaction Of PASIANLN Female Interaction Of PASIANQU				-0.664			0.372 0.455	0.856
IHISP325	Hispanic Interaction Of PASIANLN							0.625	
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban						0.068 0.395 -0.119 -0.102	0.426 0.195 -0.068 0.135	
IFEM330 IFEM331 IFEM332 IFEM333	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU Female Interaction Of PCUBANQR						-0.160 -0.334 -0.150 0.236		
IBLK330	Black Interaction Of PCUBANLN							-0.435	
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Ale			0.173 0.100 -0.228 0.121				0.213	

E7. Coefficients of Model Parameters for Past Year Illicit Drug Use Treatment

			BIG C	ITY		REMAINDER				
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+	
DUMMY	O: Intercept Term	-5.043	-4.773	-4.581	-6.415	-5.359	-4.595	-4.979	-7.462	
FEMALE	O: Female Indicator	0.860	-0.370	-0.419	-0.020	0.745	-0.472	-0.527	-1.312	
FEMBLCK	O: Black Interaction Of FEMALE	0.387	0.089	-0.072	-1.335	-3.185	-0.594	-0.410	0.329	
FEMHISP	O: Hispanic Interaction Of FEMALE	-1.317	-1.305	-0.844	-0.362	-0.910	-0.367	-0.679	-0.798	
FEMOTHR PACEDICK	O: Other Interaction Of FEMALE	2 402	0.524	0.224	1 470	0.022	1.027	-1.065	0.076	
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	-3.492 0.091	-0.534 -0.029 -1.761	0.334 -2.206	1.470 -0.778	-0.932 -0.163 -2.205	-1.037 -1.958	0.498 -0.944 0.171	0.976 0.648 1.545	
REGNOREA REGSOUTH	O: Northeast Region Indicator O: South Region Indicator					-1.227 -0.207				
IHISP7 IHISP8 IHISP9	Hispanic Interaction Of REGNOREA Hispanic Interaction Of REGSOUTH Hispanic Interaction Of REGWEST			2.957 2.157 1.724						
PHH1PLN PHH1PQU PHH1PCU PHH1PQR	T: Linear: Percent One Person Households T: Quadratic: Percent One Person Households T: Cubic: Percent One Person Households T: Quartic: Percent One Person Households							0.262 0.149 -0.118 0.085	0.202 -0.042 -0.090 -0.098	
POPRMLN	T: Linear: Average Persons Per Room							0.347		
PAGE34LN	T: Linear: Percent Persons 25-34 Years					-0.111				
IFEM35	Female Interaction Of PAGE34LN					0.661				
PSCH12LN PSCH12QU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma				-0.670		0.452		-0.233 0.284	
PSCHSCLN PSCHSCQU	T: Linear: Percent Some College And No Degree T: Quadratic: Percent Some College And No Degree		0.294 -0.123				-0.152 -0.147		0.320 -0.381	
IFEM80 IFEM81	Female Interaction Of PSCHSCLN Female Interaction Of PSCHSCQU		-0.355 0.327						-0.277 0.521	
IHISP80 IHISP81	Hispanic Interaction Of PSCHSCLN Hispanic Interaction Of PSCHSCQU						0.381 0.658			
PPOVERLN	T: Linear: Percent Families Below Poverty Level				0.807				0.806	
PBLACKLN PBLACKQU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic							-0.214 -0.170		
PHISPLN PHISPQU PHISPCU PHISPQR	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic T: Cubic: Percent Hispanic T: Quartic: Percent Hispanic			0.003 0.116 -0.172 0.062				0.037 0.173 0.241 0.058		
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity						-0.075 -0.042 -0.165			
IHISP110 IHISP111 IHISP112	Hispanic Interaction Of POTHLN Hispanic Interaction Of POTHQU Hispanic Interaction Of POTHCU						-0.156 0.614 0.498			
PHHF18LN PHHF18QU PHHF18CU PHHF18QR	T: Linear: % Female-Headed HH W/No Spouse & Child Under 18 T: Quadratic: % Female-Headed HH W/No Spouse & Child Under 18 T: Cubic: % Female-Headed HH W/No Spouse & Child Under 18 T: Quartic: % Female-Headed HH W/No Spouse & Child Under 18					0.136 -0.035 -0.150 -0.105				
PFLABLN PFLABQU PFLABCU PFLABQR	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force T: Cubic: Percent Females 16+ Years Old In Labor Force T: Quartic: Percent Females 16+ Years Old In Labor Force	0.029 -0.204				-0.196 0.050 0.027 0.140				
PFNEVLN PFNEVQU PFNEVCU PFNEVQR	T: Linear: Percent Females Never Married 1: Quadratic: Percent Females Never Married 1: Cubic: Percent Females Never Married T: Quartic: Percent Females Never Married			0.203				-0.032 0.074 -0.052 -0.089		
PMNEVLN PMNEVQU	T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married						0.263		0.115 0.354	
IBLK140	Black Interaction Of PMNEVLN						-0.707			
PMLABLN PMLABQU PMLABCU PMLABQR	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force T: Cubic: Percent Males 16+ Years Old In Labor Force T: Quartic: Percent Males 16+ Years Old In Labor Force								0.075 0.042 -0.091 -0.136	
ADRATELN ADRATEQU	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases						-0.032 0.245			
V18FLN V18FQU V18FCU V18FQR	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate C: Cubic: Marijuana Posession Arrest Rate C: Quartic: Marijuana Posession Arrest Rate				-0.856		0.072 0.159 -0.088 -0.081		-0.829 -0.304 -0.300	
IHISP185 IHISP186 IHISP187	Hispanic Interaction Of V18FLN Hispanic Interaction Of V18FQU Hispanic Interaction Of V18FCU								0.695 0.108 0.932	
V18BLN V18BQU	C: Linear: Marijuana Sale/Manufacture Arrest Rate C: Quadratic: Marijuana Sale/Manufacture Arrest Rate		0.239				-0.484 -0.322			
IBLK190 IBLK191	Black Interaction Of V18BLN Black Interaction Of V18BQU						-0.003 0.491			
IHISP190 IHISP191	Hispanic Interaction Of V18BLN Hispanic Interaction Of V18BQU	1	-0.724				0.426 0.590			

	Label		BIG CI	ITY		REMAINDER			
Variable		12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
V18ELN	U: Linear: Opium/Cocaine & Deriv Posession Arrest Rate		•	·		•	0.380	•	
V18LN	C: Linear: Total Drug Abuse Violations Arrest Rate	0.150			0.774				0.484
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992	-0.158 0.207 -0.332							
B64DISLN B64DISQU B64DISCU B64DISQR	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability B: Quartic: Percent Persons 16-64 With A Work Disability			0.182 0.047 0.015 0.065	0.022 0.358			0.484 -0.135 0.009 -0.057	
IHISP260 IHISP261 IHISP262 IHISP263	Hispanic Interaction Of B64DISLN Hispanic Interaction Of B64DISQU Hispanic Interaction Of B64DISCU Hispanic Interaction Of B64DISQR							-0.142 0.099 0.681 0.183	
BAGE34LN BAGE34QU BAGE34CU	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years					-0.080 -0.021 -0.007			
BAGE44LN BAGE44QU BAGE44CU BAGE44QR	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years B: Quartic: Percent Persons 35-44 Years			-0.119 0.015 0.159		-0.074 -0.037 0.196	0.063 0.030 -0.085 -0.044	-0.014 -0.080 -0.109 0.062	0.135 -0.189 0.473
IFEM270 IFEM271 IFEM272 IFEM273	Female Interaction Of BAGE44LN Female Interaction Of BAGE44QU Female Interaction Of BAGE44CU Female Interaction Of BAGE44QR						-0.008 -0.115 0.381 0.117		
BAGE54LN	B: Linear: Percent Persons 45-54 Years	-0.233				0.193	0.207		
IFEM275	Female Interaction Of BAGE54LN	1.510					-0.499		
IBLK275 BASIANLN	Black Interaction Of BAGE54LN B: Linear: Percent Population: Asian, Pacific Islander	1.518	-0.269	0.288			-0.235		
BASIANQU BASIANCU BASIANQR	B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander		,	-0.202			0.039 -0.092 -0.045		
IFEM285 IFEM286 IFEM287 IFEM288	Female Interaction Of BASIANLN Female Interaction Of BASIANQU Female Interaction Of BASIANCU Female Interaction Of BASIANQR						0.226 -0.276 0.077 0.201		
IBLK285	Black Interaction Of BASIANLN		0.470				1.131		
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban	0.419 -0.071 0.479					-0.352 -0.140		
IFEM290 IFEM291	Female Interaction Of BCUBANLN Female Interaction Of BCUBANQU						-0.175 0.568		
IBLK290	Black Interaction Of BCUBANLN	-1.371							
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed					0.318	0.095 0.009 0.211 0.072	-0.187 -0.019 0.057 0.062	-0.325 0.216
IFEM295 IFEM296	Female Interaction Of BFNOTLN Female Interaction Of BFNOTQU					-0.419			0.193 -0.818
IBLK295 IBLK297 IBLK298	Black Interaction Of BFNOTLN Black Interaction Of BFNOTCU Black Interaction Of BFNOTQR						-0.107 -0.225 -0.189		
BINDIALN BINDIAQU BINDIACU BINDIAQR	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut B: Quartic: Percent Pop: American Indian, Eskimo, Ale		0.010 0.097 0.258 -0.092		-0.569 -0.540 -0.557		0.358 -0.064 0.294 -0.103		0.304 0.642 -1.237 0.308
IBLK300 IBLK301 IBLK302	Black Interaction Of BINDIALN Black Interaction Of BINDIAQU Black Interaction Of BINDIACU		0.487 -0.122 0.226		0.668 0.686 1.054		-0.004 0.419 0.085		0.253 -0.353 0.879
IBLK303 BMNOTLN	Black Interaction Of BINDIAQR B: Linear: Percent Males Separated, Divorced or Widowed		0.139				0.246		-0.373 0.460
BPOVERLN	B: Linear: Percent Families Below Poverty Level		0.053						-0.488
BPOVERQU IOTH311	B: Quadratic: Percent Families Below Poverty Level Other Interaction Of BPOVERQU		0.132 -0.470						-0.336
BPRICALN	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican		-0.354				0.251 0.104		0.530
BPRICAQU IHISP315 IHISP316	Hispanic Interaction Of BPRICALN Hispanic Interaction Of BPRICAQU						0.588 -1.177		
BSCHASLN	B: Linear: Percent Associates Degree						-1.177	0.148	
BSCHASQU BSCHASCU	B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree							-0.063 0.150	
IBLK320 IBLK321 IBLK322	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU							-0.633 -0.272 -0.648	
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander					0.122		0.016 0.131 -0.037 0.082	
PCUBANLN PCUBANQU	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban						-0.011 0.169	-	
PCUBANCU PCUBANQR	T: Quartic: Percent Hispanics. Cuban T: Quartic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban						0.024 -0.083		
IFEM330	Female Interaction Of PCUBANLN						0.682		

			BIG C	ITY		REMAINDER					
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+		
IFEM331 IFEM332 IFEM333 PINDIALN PINDIAQU PINDIACU IBLK335 IBLK336 IBLK337	Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU Female Interaction Of PCUBANCU Female Interaction Of PCUBANQR T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut Black Interaction Of PINDIALN Black Interaction Of PINDIAQU Black Interaction Of PINDIACU						-0.255 -0.096 0.242	0.254 -0.115 0.301 -0.108 0.093 -0.494			

E8. Coefficients of Model Parameters for Past Year Dependency on Alcohol Only

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	-3.963	-2.644	-3.080	-3.729	-4.195	-2.365	-2.898	-4.946
FEMALE FEMBLCK	O: Female Indicator O: Black Interaction Of FEMALE	-0.383 -1.158	-0.366 -0.018	-0.418 -0.071	-0.598 -0.648	0.113 -0.850	-0.845 -0.065	-1.140 0.021	-1.297 -0.187
FEMHISP	O: Hispanic Interaction Of FEMALE	0.022	-1.116	-0.961	-1.060	0.122	-0.003	-0.427	-0.187
FEMOTHR	O: Other Interaction Of FEMALE	0.373	-0.253	-0.200	-1.735		2.041	-0.388	
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	-0.176 -0.581 -1.575	-0.939 0.066 -0.675	-0.428 0.186 -2.097	0.542 0.017 -3.071	-2.878 0.067	-0.844 -0.229 -1.577	0.024 0.449 -0.957	-2.764 0.312 -0.134
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator					-0.118 -0.211 0.225			
IBLK7 IBLK8 IBLK9	Black Interaction Of REGNOREA Black Interaction Of REGSOUTH Black Interaction Of REGWEST					1.705 1.512 2.175			
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban					-0.530 -0.419 0.076 0.174			0.498 0.074 0.648 0.902
IBLK10 IBLK11 IBLK12 IBLK13	Black Interaction Of PDENLEV1 Black Interaction Of PDENLEV2 Black Interaction Of PDENLEV3 Black Interaction Of PDENLEV4					0.911 0.208 2.387 1.470			3.298 3.914 1.964 1.283
PHH1PLN PHH1PQU	T: Linear: Percent One Person Households T: Quadratte: Percent One Person Households								0.112 -0.153
IFEM15	Female Interaction Of PHH1PLN								0.033
POPRMLN POPRMQU POPRMCU	Female Interaction Of PHH1PQU T: Linear: Average Persons Per Room T: Quadratic: Average Persons Per Room T: Cubic: Average Persons Per Room		-0.194				-0.223 -0.141 -0.141		0.288
IBLK20 IBLK21	Black Interaction Of POPRMLN Black Interaction Of POPRMOU						-0.045 0.238		
PAGE44LN PAGE44QU	T: Linear: Percent Persons 35-44 Years T: Quadratic: Percent Persons 35-44 Years					-0.004 0.111	0.230		
PSCH12LN PSCH12QU PSCH12CU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma				0.003 0.137	0.111			-0.365 0.069 -0.138
PSCHCOLN	T: Linear: Bachelors, Graduate, Or Professional Degree								-0.436
PSCHSCLN PSCHSCQU PSCHSCCU	T: Linear: Percent Some College And No Degree T: Quadratic: Percent Some College And No Degree T: Cubic: Percent Some College And No Degree			0.048 -0.005 0.118				0.046 0.089 -0.008	
IFEM80 IFEM81 IFEM82	Female Interaction Of PSCHSCLN Female Interaction Of PSCHSCQU Female Interaction Of PSCHSCCU							0.061 -0.023 0.260	
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Quartic: Percent Families Below Poverty Level					0.013 -0.058 -0.079 0.001		0.184 -0.104 -0.027 -0.051	
IFEM85 IFEM86 IFEM87 IFEM88	Female Interaction Of PPOVERLN Female Interaction Of PPOVERQU Female Interaction Of PPOVERCU Female Interaction Of PPOVERQR					0.040 0.194 -0.072 -0.104			
IHISP85 IHISP86	Hispanic Interaction Of PPOVERLN Hispanic Interaction Of PPOVERQU					-0.398 -0.362		0.042 0.249	
P64DISLN P64DISQU P64DISCU P64DISQR	T: Linear: Percent 16-64 With A Work Disability T: Quadratic: Percent 16-64 With A Work Disability T: Cubic: Percent 16-64 With A Work Disability T: Quartic: Percent 16-64 With A Work Disability		-0.092 0.033 -0.033 -0.010				-0.161 0.051 -0.036 0.046		
IBLK95 IBLK96 IBLK97 IBLK98	Black Interaction Of P64DISLN Black Interaction Of P64DISQU Black Interaction Of P64DISCU Black Interaction Of P64DISCR		-0.122 0.272 -0.006 -0.133				0.266 0.099 -0.204 -0.146		
PHISPLN	T: Linear: Percent Hispanic		0.133				0.110		-0.233
PHISPQU IBLK105	T: Quadratic: Percent Hispanic Black Interaction Of PHISPLN Black Interaction Of PHISPLN								-0.315 0.094
IBLK106 PFNOTLN	Black Interaction Of PHISPQU T: Linear: Percent Females Separated, Divorced Or Widowed						0.033		0.356
PFNOTQU IOTH135	T: Quadratic: Percent Females Separated, Divorced Or Widowed Other Interaction Of PFNOTLN						-0.020 0.125		
IOTH136 PMNOTLN PMNOTQU	Other Interaction Of PFNOTQU T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed						0.445		-0.079 -0.072
PMNOTČU POLDHULN POLDHUQU POLDHUČU	T: Cubic: Percent Males Separated, Divorced Or Widowed T: Linear: Percent Housing Units Built 1939 Or Earlier 1: Quadratic: Percent Housing Units Built 1939 Or Earlier T: Cubic: Percent Housing Units Built 1939 Or Earlier								-0.189 0.046 0.015 0.181
ADRATTLN	C: Linear: Death Rate With Explicit Mention Of Alcohol	-0.113				-0.101			0.101
ADRATIQU ADRATICU	C: Quadratic: Death Rate With Explicit Mention Of Alcohol C: Cubic: Death Rate With Explicit Mention Of Alcohol	-0.029 -0.067				0.006 0.010			

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
ADRATIQK	C: Quartic: Death Rate With Explicit Mention Of Alcohol	-0.098				0.066			
V18FLN V18FQU	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate					-0.028 0.097			
IFEM185 IFEM186	Female Interaction Of V18FLN Female Interaction Of V18FQU			0.262	0.012	0.011 -0.283	0.117	0.100	0.024
V18ELN V18EQU V18ECU	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate			0.363 0.253	-0.013 0.192 0.218		0.117	0.108 0.112	0.034 0.131 0.327
IHISP195 IHISP196	Hispanic Interaction Of V18ELN Hispanic Interaction Of V18EQU								0.255 -0.345
V18ALN V18AQU	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate			0.121 -0.060				0.019 -0.097	0.051 -0.155
IHISP200 IHISP201	Hispanic Interaction Of V18ALN Hispanic Interaction Of V18AQU								-0.349 0.356
IOTH200 IOTH201	Other Interaction Of V18ALN Other Interaction Of V18AQU			1.082 -0.680					
V18LN V18QU V18CU V18QR	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate	0.156		-0.455 -0.125	0.027 0.006 -0.080	-0.259 -0.131 -0.114 -0.088		-0.218 -0.112	0.058 -0.156 -0.271
IBLK225 IBLK226	Black Interaction Of V18LN Black Interaction Of V18QU								-0.137 0.371
DRATELN DRATEQU DRATECU DRATEQR	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992 C: Cubic: Mean Drug Client Treatment Rate 1991 & 1992 C: Quartic: Mean Drug Client Treatment Rate 1991 & 1992	0.193 0.137 0.026 0.097				0.162 -0.133			
IBLK235 IBLK236 IBLK237 IBLK238	Black Interaction Of DRATELN Black Interaction Of DRATEQU Black Interaction Of DRATECU Black Interaction Of DRATEQR	-0.113 -0.300 0.460 -0.159							
RP80ALN RP80AQU RP80ACU RP80AQR	T: Linear: Recoded Median Household Income T: Quadratıc: Recoded Median Household Income T: Cubic: Recoded Median Household Income T: Quartic: Recoded Median Household Income		0.057 0.049 -0.028 -0.007				-0.150 -0.014 -0.010 -0.038		
IBLK250 IBLK251 IBLK252 IBLK253	Black Interaction Of RP80ALN Black Interaction Of RP80AQU Black Interaction of RP80ACU Black Interaction Of RP80AQR		-0.057 -0.137 0.134 0.085				0.056 0.298 0.393 0.087		
ARATELN B64DISLN	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 B: Linear: Percent Persons 16-64 With A Work Disability		0.094			-0.225	-0.008		
B64DISQU B64DISCU B64DISQR	B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability B: Quartic: Percent Persons 16-64 With A Work Disability		0.009 0.019 -0.005				-0.018 0.050		
IHISP260 IHISP261 IHISP262 IHISP263	Hispanic Interaction Of B64DISLN Hispanic Interaction Of B64DISQU Hispanic Interaction Of B64DISCU Hispanic Interaction Of B64DISQR		-0.097 -0.015 -0.075 0.059						
BAGE34LN BAGE34QU BAGE34CU BAGE34QR	B: Lunear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years		0.037 -0.017				0.018 -0.066 0.109		0.067 -0.064 0.050 -0.080
IFEM265 IFEM266	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU		-0.166 0.166						
BAGE44LN BAGE44QU BAGE44CU BAGE44QR	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years B: Quartic: Percent Persons 35-44 Years			0.124 -0.025 0.080		0.030 0.067 -0.140		0.045 0.054 -0.037 -0.037	
IOTH270 IOTH271	Other Interaction Of BAGE44LN Other Interaction Of BAGE44QU			-0.168 -0.286				0.511	
BAGE54LN BAGE54QU	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years						-0.023 0.053		
IOTH275 IOTH276	Other Interaction Of BAGE54LN Other Interaction Of BAGE54QU						-0.121 -0.374		
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander			-0.009	-0.064 -0.123 -0.001 -0.035			-0.031 -0.019 -0.010 -0.033	
IHISP285 IHISP286 IHISP287 IHISP288	Hispanic Interaction Of BASIANLN Hispanic Interaction Of BASIANQU Hispanic Interaction Of BASIANCU Hispanic Interaction Of BASIANQR			-0.277				-0.086 0.136 0.122 0.116	
IOTH285 IOTH286 IOTH287 IOTH288	Other Interaction Of BASIANLN Other Interaction Of BASIANQU Other Interaction Of BASIANCU Other Interaction Of BASIANQR				0.164 1.819 -3.294 0.795				
BCUBANLN BCUBANQU BCUBANCU BCUBANQR	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban B: Quartic: Percent Hispanics: Cuban		0.212 -0.153	0.171 0.041			-0.096 0.085 -0.088 0.135		
IFEM290 IFEM291 IFEM292 IFEM293	Female Interaction Of BCUBANLN Female Interaction Of BCUBANQU Female Interaction Of BCUBANCU Female Interaction Of BCUBANQR		-0.320 0.211				0.073 -0.082 0.422 -0.216		
IBLK290 IBLK291	Black Interaction Of BCUBANLN Black Interaction Of BCUBANQU			-0.104 0.252					

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BFNOTLN BFNOTQU	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed							0.074 -0.002	-0.227 -0.015
IFEM295 IFEM296	Female Interaction Of BFNOTLN Female Interaction Of BFNOTQU								0.097 0.236
IOTH295 IOTH296	Other Interaction Of BFNOTLN Other Interaction Of BFNOTQU							0.417 0.721	
BINDIALN BINDIAQU BINDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut			-0.214		-0.053		-0.093 0.048 0.087	0.069 0.220
IBLK300 IBLK301 IBLK302	Black Interaction Of BINDIALN Black Interaction Of BINDIAQU Black Interaction Of BINDIACU			0.418		0.916		0.122 -0.070 -0.442	
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU			0.309				0.160 -0.268	
BMNOTLN BMNOTQU BMNOTCU BMNOTQR	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed B: Quartic: Percent Males Separated, Divorced or Widowed			0.085	-0.075 0.097 -0.004 -0.045	-0.043 0.091 0.009 -0.050	-0.052 0.021	-0.053 -0.094	-0.059 -0.064 -0.175 0.006
IBLK305 IBLK306 IBLK307 IBLK308	Black Interaction Of BMNOTLN Black Interaction Of BMNOTQU Black Interaction Of BMNOTQU Black Interaction Of BMNOTQR				0.200 -0.200 0.308				0.644 0.040 0.312 -0.183
IHISP305 IHISP306 IHISP307 IHISP308	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU Hispanic Interaction Of BMNOTCU Hispanic Interaction Of BMNOTQU					0.299 -0.128 -0.048 0.169			
BPOVERLN BPOVERQU BPOVERCU	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level				-0.069 -0.061 0.118				0.093
IHISP310	Hispanic Interaction Of BPOVERLN		0.220	0.000			0.172	0.112	0.408
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican		-0.229	0.089 -0.040 -0.140			-0.173	-0.112 -0.060 -0.062 -0.079	
IBLK315 IBLK316 IBLK317 IBLK318	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU Black Interaction Of BPRICACU Black Interaction Of BPRICAQR							-0.265 0.090 -0.070 0.192	
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander	-0.019 0.157 -0.145				0.078 0.147 0.075 -0.041	-0.039 -0.032 -0.076		-0.144 -0.048 -0.204
IFEM325 IFEM326 IFEM327 IFEM328	Female Interaction Of PASIANLN Female Interaction Of PASIANQU Female Interaction Of PASIANCU Female Interaction Of PASIANQR	-0.009 -0.242 0.249				-0.287 -0.119 0.033 0.137			
IBLK325 IBLK326 IBLK327	Black Interaction Of PASIANLN Black Interaction Of PASIANQU Black Interaction Of PASIANCU						0.151 0.072 0.415		
PCUBANLN PCUBANQU PCUBANCU	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban	-0.057	-0.185		0.036 -0.065	0.032	0.069 0.006		-0.142 -0.037 -0.129
IBLK330 IBLK331 IBLK332	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU	0.519			0.197 0.296	0.696			0.277 -0.301 0.674
IOTH330 IOTH331	Other Interaction Of PCUBANLN Other Interaction Of PCUBANQU						0.535 0.515		
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Aleut	0.055 -0.062 0.015			-0.179 0.114 -0.057 0.042	-0.064 0.159 0.154			
IFEM335 IFEM336 IFEM337	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU Female Interaction Of PINDIACU	0.134 0.168 -0.275				0.007 -0.264 -0.319			
IHISP335 IHISP336 IHISP337 IHISP338	Hispanic Interaction Of PINDIALN Hispanic Interaction Of PINDIAQU Hispanic Interaction Of PINDIACU Hispanic Interaction Of PINDIAQU				0.282 -0.091 0.151 -0.115				
IOTH335 IOTH336	Other Interaction Of PINDIALN Other Interaction Of PINDIAQU				0.359 -0.987				

E9. Coefficients of Model Parameters for Past Year Illicit Dependency

			BIG C	ITY			REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	-4.381	-3.268	-4.208	-5.485	-4.057	-3.268	-4.618	-6.476
FEMALE	O: Female Indicator	0.449	-0.684	-0.706	-0.607	-0.320	-0.878	-0.774	-0.071
FEMBLCK FEMHISP	O: Black Interaction Of FEMALE O: Hispanic Interaction Of FEMALE	-0.149 -0.388	0.076 0.306	0.230 0.913	-0.177 -0.262	0.859	0.533 0.240	-0.178 0.094	-0.466 -0.306
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	-1.289 0.150	-0.503 -1.149 -1.719	0.181 -1.313	0.696 -0.980	-1.851 -0.302	-0.864 -0.585 -1.661	0.220 -0.778	0.884 -0.234
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator							0.349 0.286 0.840	
PHH1PLN	T: Linear: Percent One Person Households						0.258		
POPRMLN POPRMQU POPRMCU POPRMQR	T: Linear: Average Persons Per Room T: Quadratic: Average Persons Per Room T: Cubic: Average Persons Per Room T: Quartic: Average Persons Per Room			-0.156 -0.130				-0.232 0.040 0.082 0.074	
PAGE18LN PAGE18QU PAGE18CU PAGE18QR	T: Linear: Percent Persons 0-18 Years 1: Quadratic: Percent Persons 0-18 Years 1: Cubic: Percent Persons 0-18 Years T: Quartic: Percent Persons 0-18 Years		0.051 -0.067 -0.072 0.044				0.248 0.108 0.078 0.067		
IFEM25 IFEM26 IFEM27 IFEM28	Female Interaction Of PAGE18LN Female Interaction Of PAGE18QU Female Interaction Of PAGE18CU Female Interaction Of PAGE18QR		-0.201 0.084 0.158 -0.135				-0.128 0.036 -0.085 -0.096		
PAGE34LN PAGE34QU	T: Linear: Percent Persons 25-34 Years T: Quadratic: Percent Persons 25-34 Years		0.072			-0.183 -0.202		0.127 -0.187	
IBLK35	Black Interaction Of PAGE34LN							-0.691	
IHISP35	Hispanic Interaction Of PAGE34LN		-0.416						0.201
PAGE44LN PAGE44QU	T: Linear: Percent Persons 35-44 Years T: Quadratic: Percent Persons 35-44 Years				0.052	0.055			0.381 0.267
PAGE54LN PAGE54QU PAGE54CU	T': Linear: Percent Persons 45-54 Years T': Quadratic: Percent Persons 45-54 Years T': Cubic: Percent Persons 45-54 Years				-0.073 0.146 -0.299	0.057 0.040 -0.158			
IHISP45 IHISP46	Hispanic Interaction Of PAGE54LN Hispanic Interaction Of PAGE54QU				-0.761 -0.915				
PSCH12LN PSCH12QU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma	-0.109 0.111							
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU	0.011 -0.248							
PPOVERLN PPOVERQU PPOVERCU	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level				-0.112 -0.011 -0.220				
PMNEVLN PMNEVQU PMNEVCU	T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married T: Cubic: Percent Males Never Married			0.159 -0.057 0.141					
PMLABLN PMLABQU PMLABCU	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force T: Cubic: Percent Males 16+ Years Old In Labor Force		0.140 -0.073 0.145						
IHISP145 IHISP146 IHISP147	Hispanic Interaction Of PMLABLN Hispanic Interaction Of PMLABQU Hispanic Interaction Of PMLABCU		-0.275 -0.006 -0.492						
PMNOTLN PMNOTQU	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed								0.033 0.184
PRENTLN PRENTOU	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented							0.178 0.164	0.617
ADRATELN ADRATEQU	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases							0.062 0.193	
V18FLN	C: Linear: Marijuana Posession Arrest Rate			-0.374					
V18ELN V18EQU V18ECU V18EQR	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubrc: Opium/Cocaine & Deriv Posession Arrest Rate C: Quartic: Opium/Cocaine & Deriv Posession Arrest Rate			0.560 -0.143 0.113 0.059				-0.056 -0.078 -0.070 -0.068	
IHISP195 IHISP196 IHISP197 IHISP198	Hispanic Interaction Of V18ELN Hispanic Interaction Of V18EQU Hispanic Interaction Of V18ECU Hispanic Interaction Of V18EOR			-0.211 0.098 -0.420 0.187					
V18LN V18QU	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate			/			0.013 -0.118		
VIOLLN VIOLQU VIOLCU	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate C: Cubic: Total Violent Offenses Arrest Rate								-0.271 -0.238 -0.404
RP80ALN RP80AQU RP80ACU	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income					-0.124 0.259			-0.540 -0.005 -0.234
IFEM250 IFEM251	Female Interaction Of RP80ALN Female Interaction Of RP80AQU					-0.068 -0.224			
ARATELN	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992	 					0.118		

			BIG C	ITY			REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
ARATEQU B64DISLN	C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 B: Linear: Percent Persons 16-64 With A Work Disability			0.206			0.176		
BAGE34LN BAGE34QU BAGE34CU	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years			0.200				-0.087 -0.070 0.090	
IFEM265 IFEM266 IFEM267	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU							-0.066 0.169 -0.365	
BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years					-0.273		-0.023 0.111	
IHISP270 IHISP271	Hispanic Interaction Of BAGE44LN Hispanic Interaction Of BAGE44QU							-0.032 -0.374	
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years			-0.118		-0.079 -0.145 0.110	0.132 -0.083 -0.131	-0.194 -0.161 0.119	0.017 0.066 0.256 0.114
IFEM275 IFEM276 IFEM277 IFEM278	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU Female Interaction Of BAGE54QR					-0.025 0.295		0.398	-0.334 -0.072 -0.491 -0.234
IBLK275 IBLK276 IBLK277	Black Interaction Of BAGE54LN Black Interaction Of BAGE54QU Black Interaction Of BAGE54CU							-0.331 -0.018 -0.313	
BAGE64LN BAGE64QU	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years			-0.055 -0.219		-0.243	-0.009 -0.128	0.051 -0.140	
BAGE64ČU IFEM280	B: Cubic: Percent Persons 55-64 Years Female Interaction Of BAGE64LN			-0.118		0.406		-0.052	
IBLK280 IBLK281 IBLK282	Black Interaction Of BAGE64LN Black Interaction Of BAGE64QU Black Interaction Of BAGE64CU			-0.010 0.326 0.268				-0.199 0.076 0.277	
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratuc: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander			·· <u>·</u> ···				0.215	0.310 0.056 -0.270 -0.011
IBLK285 IBLK286 IBLK287 IBLK288	Black Interaction Of BASIANLN Black Interaction Of BASIANQU Black Interaction Of BASIANCU Black Interaction Of BASIANCR								0.052 0.235 0.853 -0.291
BCUBANLN BCUBANQU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban				-0.198 0.300				
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed	0.223 0.025 -0.011 0.064				0.119 -0.103 0.023 0.069	-0.162		
IFEM295	Female Interaction Of BFNOTLN					0.420	0.304		
IBLK295 IBLK296 IBLK297	Black Interaction Of BFNOTLN Black Interaction Of BFNOTQU Black Interaction Of BFNOTCU					-0.139 -0.072 -0.475			
IHISP295 IHISP296 IHISP297 IHISP298	Hispanic Interaction Of BFNOTLN Hispanic Interaction Of BFNOTQU Hispanic Interaction Of BFNOTCU Hispanic Interaction Of BFNOTQR	-0.046 -0.084 0.048 -0.089							
BINDIALN BINDIAQU BINDIACU BINDIAQR	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut B: Quartic: Percent Pop: American Indian, Eskimo, Ale			-0.153 0.071 0.230			0.082 -0.123 0.076 -0.095	0.222 -0.164 0.063 0.048	0.533
IBLK300 IBLK301 IBLK302 IBLK303	Black Interaction OFBINDIALN Black Interaction OFBINDIAQU Black Interaction OFBINDIACU Black Interaction OFBINDIAQR							-0.267 0.346 -0.032 -0.228	
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU							-0.552 0.624	
BPOVERLN BPOVERQU BPOVERCU	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level		0.145 0.064 0.017			-0.290	-0.040 0.010 -0.114		-0.662 -0.342
IOTH310 IOTH311 IOTH312	Other Interaction Of BPOVERLN Other Interaction Of BPOVERQU Other Interaction Of BPOVERCU		-0.334 0.406 0.726						
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Ouartic: Percent Hispanics: Puerto Rican					0.104 -0.306 -0.041 -0.155	0.455		
IFEM315	Female Interaction Of BPRICALN					-0.625			
BSCHASLN BSCHASQU BSCHASCU	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree				-0.083 0.004 -0.308				
IHISP320 IHISP321 IHISP322	Hispanic Interaction Of BSCHASLN Hispanic Interaction Of BSCHASQU Hispanic Interaction Of BSCHASCU				-0.232 -0.188 0.568				
PASIANLN PCUBANLN	1: Linear: Percent Population: Asian, Pacific Islander T: Linear: Percent Hispanics: Cuban	-0.068	0.133	0.057			0.241	-0.152	-0.483
PCUBANQU PCUBANCU PCUBANQR	T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban	-0.011		0.189				0.173 -0.114 0.147	0.293 0.183 -0.178

			BIG C	ITY			REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IFEM330 IBLK330 IBLK331 IBLK332 IBLK333	Female Interaction Of PCUBANLN Black Interaction Of PCUBANU Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU Black Interaction Of PCUBANQR	0.333 0.103 -0.800							-0.102 -0.366 -0.242 0.480

E10. Coefficients of Model Parameters for Past Year Arrested

		BIG CITY F			REMAINDER				
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	-3.028	-3.002	-3.690	-4.853	-2.878	-2.294	-4.243	-5.665
FEMALE	O: Female Indicator	-1.145	-1.322	-0.960	-1.385	-1.128	-1.634	-1.813	-2.582
FEMBLCK	O: Black Interaction Of FEMALE	-0.180	-0.571	-0.083	-1.057	0.256	0.361	0.296	0.692
FEMHISP RACEBLCK RACEHISP RACEOTHR	O: Hispanic Interaction Of FEMALE O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	0.260 0.228 0.185 -1.490	-1.209 0.537 0.050 -1.138	-0.578 0.299 0.249	-0.624 0.410 0.099	0.246 0.136 0.369 -0.380	-0.382 -0.746 -0.214 -0.543	-0.074 0.333 -0.315	-1.022 0.006 0.722
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator	-1.470	-1.136			-0.208 -0.434 0.193	-0.545		-1.114 -0.679 -0.372
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban						-0.214 -0.162 -0.005 -0.154	0.594 0.676 1.159 1.026	
IBLK10 IBLK11 IBLK12 IBLK13	Black Interaction Of PDENLEV1 Black Interaction Of PDENLEV2 Black Interaction Of PDENLEV3 Black Interaction Of PDENLEV4						0.515 0.547 1.799 0.850	1.020	
PHH1PLN PHH1PQU PHH1PCU	T: Linear: Percent One Person Households T: Quadratic: Percent One Person Households T: Cubic: Percent One Person Households		-0.106 0.032 -0.020				-0.088 0.049 -0.155		
PHH1PQR IBLK15 IBLK16 IBLK17	T: Quartic: Percent One Person Households Black Interaction Of PHHIPLN Black Interaction Of PHHIPQU Black Interaction OI PHHIPQU Black Interaction OI PHHIPQU		-0.034 0.016 0.009 0.110				-0.039 0.232 0.043 0.240		
IBLK18 POPRMLN	Black Interaction Of PHH1PQR T: Linear: Average Persons Per Room		0.089 -0.053				0.076 -0.143		
PAGE34LN	T: Linear: Percent Persons 25-34 Years		0.000				0.175		0.501
PAGE54LN PAGE54QU PAGE54CU	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years T: Cubic: Percent Persons 45-54 Years							-0.063 0.045 -0.257	
IHISP45 IHISP46 IHISP47	Hispanic Interaction Of PAGE54LN Hispanic Interaction Of PAGE54QU Hispanic Interaction Of PAGE54CU							-0.196 -0.221 0.396	
PAGE64LN	T: Linear: Percent Persons 55-64 Years					0.205			
PSCH8LN PSCH8QU	T: Linear: Percent 0-8 Years Of School T: Quadratic: Percent 0-8 Years Of School	-0.100 -0.017							
IOTH55 IOTH56	Other Interaction Of PSCH8LN Other Interaction Of PSCH8QU	-0.661 -0.773							
PSCH12LN PSCH12QU PSCH12CU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma		0.187				0.157 0.061 -0.092		
IHISP60	Hispanic Interaction Of PSCH12LN						-0.351		
PSCHCOLN PSCHCOQU PSCHCOCU	T: Linear: Bachelors, Graduate, Or Professional Degree T: Quadratic: Bachelors, Graduate, Or Professional Degree T: Cubic: Bachelors, Graduate, Or Professional Degree								0.222 0.027 -0.315
IBLK75 IBLK76 IBLK77	Black Interaction Of PSCHCOLN Black Interaction Of PSCHCOQU Black Interaction Of PSCHCOCU								-0.227 0.076 0.608
PPUBASLN P64DISLN	T: Linear: Percent Households With Public Assist Income T: Linear: Percent 16-64 With A Work Disability			0.200			0.199	0.261	
PBLACKLN	T: Linear: Percent Black Nonhispanic		-0.070	0.200				0.201	
PBLACKQU POTHLN	T: Quadratic: Percent Black Nonhispanic T: Linear: Percent Other Race/Hispanicity		-0.094		0.071				
IHISP110	Hispanic Interaction Of POTHLN				0.287				
PFLABLN PFLABQU	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force				0.284				0.066 -0.333
IFEM125	Female Interaction Of PFLABLN				0.444				0.555
PFNOTLN PFNOTQU	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed							0.112 0.169	
PMLABLN PMLABQU	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force				-0.038		0.137 -0.088		-0.424
IFEM145	Female Interaction Of PMLABLN				-0.636				
PMNOTEN PMNOTQU PMNOTCU PMNOTQR	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed T: Cubic: Percent Males Separated, Divorced Or Widowed T: Quartic: Percent Males Separated, Divorced Or Widowed						-0.020 -0.052 -0.037 0.040		
POLDHULN POLDHUQU POLDHUCU	T: Linear: Percent Housing Units Built 1939 Or Earlier T: Quadratic: Percent Housing Units Built 1939 Or Earlier T: Cubic: Percent Housing Units Built 1939 Or Earlier	-0.145				-0.024 -0.027 -0.119			
P40HULN P40HUQU	T: Linear: Percent Housing Units Built 1940-1949 T: Quadratic: Percent Housing Units Built 1940-1949	0.201						-0.022 -0.152	
PRENTLN	T: Linear: Percent Housing Rented	-				0.362			

		BIG CITY				REMAINDER				
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+	
ADRATILN IFEM180	C: Linear: Death Rate With Explicit Mention Of Alcohol Female Interaction Of ADRATTLN		-0.012 0.434	-0.196				•		
V18FLN V18FQU	C: Linear: Marijuana Posession Arrest Rate C: Ouadratic: Marijuana Posession Arrest Rate			-0.359				-0.201 -0.136		
V18ELN V18EQU	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate			0.414	0.052 0.218			0.383	-0.015 0.080	
IFEM195	Female Interaction Of V18ELN				0.218				-0.540	
IFEM196 V18ALN	Female Interaction Of V18EQU C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate							-0.219	-0.968	
DRATELN DRATEQU DRATECU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992 C: Cubic: Mean Drug Client Treatment Rate 1991 & 1992	-0.091 -0.005 0.146								
RH43ALN RH43AQU RH43ACU RH43AQR	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units T: Cubic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units				-0.254			-0.055 -0.031 0.053 0.056	-0.700	
RH61ALN RH61AQU RH61ACU	T: Linear: Recoded Median Value Of Owner Occupied HUs T: Quadratic: Recoded Median Value Of Owner Occupied HUs T: Cubic: Recoded Median Value Of Owner Occupied HUs		-0.007	-0.147 -0.100			0.059 -0.061 -0.099	0.029 -0.000 0.132		
IHISP245 IHISP246 IHISP247	Hispanic Interaction Of RH61ALN Hispanic Interaction Of RH61AQU Hispanic Interaction Of RH61ACU							0.155 0.166 -0.282		
ARATELN ARATEQU ARATECU ARATEQR	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quartic: Mean A-Only Client Treatment Rate 1991 & 1992								0.010 0.050 -0.175 -0.036	
IBLK255 IBLK256 IBLK257 IBLK258	Black Interaction Of ARATELN Black Interaction Of ARATEQU Black Interaction Of ARATECU Black Interaction Of ARATEQR								-0.037 -0.264 0.089 0.155	
B64DISLN B64DISQU B64DISCU B64DISQR	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability B: Quartic: Percent Persons 16-64 With A Work Disability								0.086 0.025 0.085 -0.076	
BAGE34LN BAGE34QU BAGE34CU BAGE34QR	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years				0.128 0.024		-0.012 0.046 0.098 0.012			
IFEM265 IFEM266 IFEM267 IFEM268	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU Female Interaction Of BAGE34QR						0.114 -0.132 0.001 -0.101			
IHISP265 IHISP266	Hispanic Interaction Of BAGE34LN Hispanic Interaction Of BAGE34QU				-0.008 -0.388					
BAGE44LN BAGE44QU BAGE44CU BAGE44QR	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years B: Quartic: Percent Persons 35-44 Years						-0.080 -0.086 -0.035 0.042	0.002 0.006 -0.110		
IBLK270 IBLK271 IBLK272 IBLK273	Black Interaction Of BAGE44LN Black Interaction Of BAGE44QU Black Interaction Of BAGE44CU Black Interaction Of BAGE44QR						0.094 0.093 -0.075 -0.127			
IHISP270 IHISP271	Hispanic Interaction Of BAGE44LN Hispanic Interaction Of BAGE44QU							-0.127 -0.323		
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Ouartic: Percent Persons 45-54 Years						-0.126 -0.030 -0.060 -0.040			
BAGE64LN BAGE64QU BAGE64CU	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years		-0.139			-0.186	-0.040 -0.057 0.046			
IHISP280 IHISP281 IHISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		0.298				-0.033 -0.086 -0.240			
BINDIALN BINDIAQU BINDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut		-0.005			0.170 -0.046 0.184				
IOTH300 BMNOTLN	Other Interaction Of BINDIALN B: Linear: Percent Males Separated, Divorced or Widowed	0.043	0.671							
BMNOTQU	B: Quadratic: Percent Males Separated, Divorced or Widowed	-0.139						0.057		
BPOVERLN BPOVERQU BPOVERCU	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level							0.057 0.049 -0.171		
BPRICALN BPRICAQU BPRICACU	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican E-cubic: Percent Hispanics: Puerto Rican					0.120 0.167 -0.174				
IFEM315 IFEM316 IFEM317	Female Interaction Of BPRICALN Female Interaction Of BPRICAQU Female Interaction Of BPRICACU					-0.283 -0.125 0.403				
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree			-0.168		-0.041 0.018 0.172		0.001 0.018 0.023 -0.059		
PASIANLN	T: Linear: Percent Population: Asian, Pacific Islander	-0.083				0.061				

		BIG CITY				REMAINDER	Ł		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
PASIANQU PASIANCU	1: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander					0.048 -0.152			
IFEM325 IFEM326 IFEM327	Female Interaction Of PASIANLN Female Interaction Of PASIANQU Female Interaction Of PASIANCU					0.118 -0.029 0.334			
PCUBANLN PCUBANQU PCUBANCU	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban							0.026 -0.023 -0.311	

E11. Coefficients of Model Parameters for Past Year Treatment Needed for Drug Abuse

		BIG CITY					REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	-3.441	-2.427	-2.751	-3.791	-3.445	-2.226	-3.181	-4.922
FEMALE	O: Female Indicator	0.246	-0.602	-0.654	-0.524	-0.011	-0.583	-0.279	-0.990
FEMBLCK	O: Black Interaction Of FEMALE	-0.318	-0.109	0.236	-0.776		0.303	-0.658	-0.173
FEMHISP FEMOTHR	O: Hispanic Interaction Of FEMALE O: Other Interaction Of FEMALE	-0.570 0.479	0.186 0.422	0.405 1.042	0.028		0.069 -2.145	-0.771 0.083	0.355 1.541
RACEBLCK	O: Race/Black Indicator	-0.331	-0.299	0.010	0.689		-1.157	0.615	0.939
RACEHISP RACEOTHR	O: Race/Hispanic Indicator O: Race/Other Indicator	0.264 -4.288	-0.626 -1.315	-0.642 -2.825	-0.281 -0.586		-0.463 -1.072	-0.207 -3.867	0.132 -0.274
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban						-0.235 -0.171 -0.056 -0.014		
IBLK10 IBLK11 IBLK12 IBLK13	Black Interaction Of PDENLEV1 Black Interaction Of PDENLEV2 Black Interaction Of PDENLEV3 Black Interaction Of PDENLEV4						0.319 0.508 1.306 1.185		
PHH1PLN PHH1PQU PHH1PCU PHH1PQR	T: Lmear: Percent One Person Households T: Quadratic: Percent One Person Households T: Cubic: Percent One Person Households T: Quartic: Percent One Person Households				0.210 0.071 -0.016 -0.017				-0.175 0.059 -0.088 -0.086
PAGE18LN PAGE18QU PAGE18CU	T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years T: Cubic: Percent Persons 0-18 Years						0.021 0.078 0.054		
IOTH25 IOTH26	Other Interaction Of PAGE18LN Other Interaction Of PAGE18QU						0.052 -0.794		
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years T: Cubic: Percent Persons 19-24 Years T: Quartic: Percent Persons 19-24 Years								-0.107 0.107 -0.054 0.074
PAGE44LN	T: Linear: Percent Persons 35-44 Years				0.054				
PAGE54LN PAGE54QU PAGE54CU PAGE54QR	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years T: Cubic: Percent Persons 45-54 Years T: Quartic: Percent Persons 45-54 Years				-0.023		-0.130		-0.155 -0.042 -0.091 -0.042
IFEM45	Female Interaction Of PAGE54LN				0.205				
IHISP45	Hispanic Interaction Of PAGE54LN						0.274		
PSCH12LN PSCH12QU PSCH12CU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma							0.176 -0.084 0.043	
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU							0.025 0.161	
PPOVERLN PPOVERQU PPOVERCU	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level					-0.026 0.020 -0.133			
PBLACKLN PBLACKQU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic							-0.144 -0.129	
IOTH100 IOTH101	Other Interaction Of PBLACKLN Other Interaction Of PBLACKQU							-2.087 -1.492	
PFLABLN PFLABQU PFLABCU PFLABQR	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force T: Cubic: Percent Females 16+ Years Old In Labor Force T: Quartic: Percent Females 16+ Years Old In Labor Force					-0.108 -0.096	0.055 -0.043 0.012 0.031		
IBLK125 IBLK126	Black Interaction Of PFLABLN Black Interaction Of PFLABOU						-0.006 0.203		
PFNEVLN PFNEVQU	T: Linear: Percent Females Never Married T: Quadratic: Percent Females Never Married							0.092 0.126	
IFEM130 IFEM131	Female Interaction Of PFNEVLN Female Interaction Of PFNEVQU							0.036 -0.107	
PMNEVLN PMNEVQU PMNEVCU	T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married T: Cubic: Percent Males Never Married				-0.005 -0.100				0.223 0.062 -0.268
PMLABLN PMLABQU PMLABCU PMLABQR	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force T: Cubic: Percent Males 16+ Years Old In Labor Force T: Quartic: Percent Males 16+ Years Old In Labor Force								0.032 -0.040 -0.073 -0.040
PMNOTLN	T: Linear: Percent Males Separated, Divorced Or Widowed							0.149	
POLDHULN	T: Linear: Percent Housing Units Built 1939 Or Earlier	0				-0.094			
P40HULN P40HUQU P40HUCU	T: Linear: Percent Housing Units Built 1940-1949 T: Quadratic: Percent Housing Units Built 1940-1949 T: Cubic: Percent Housing Units Built 1940-1949	0.026 0.044				-0.050 -0.023 -0.013			
IHISP160 IHISP161	Hispanic Interaction Of P40HULN Hispanic Interaction Of P40HUQU	-0.049 -0.096							
PRENTLN PRENTQU PRENTCU	Lunear: Percent Housing Rented Quadratic: Percent Housing Rented Cubic: Percent Housing Rented			0.097				0.035 0.055 0.010	0.590

			BIG C	ITY			REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
PRENTQR	T: Quartic: Percent Housing Rented		·	•			•	0.011	
IHISP165 IHISP166 IHISP167	Hispanic Interaction Of PRENTLN Hispanic Interaction Of PRENTQU Hispanic Interaction Of PRENTCU							-0.150 0.085 -0.203	
ADRATELN ADRATEQU ADRATECU ADRATEQR	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases C: Cubic: Death Rate For All Alcohol-Related Cases C: Quartic: Death Rate For All Alcohol-Related Cases						-0.023 0.117 -0.030 0.027		
IBLK175 IBLK176 IBLK177 IBLK178	Black Interaction Of ADRATELN Black Interaction Of ADRATEQU Black Interaction Of ADRATECU Black Interaction Of ADRATEQR						0.058 -0.084 0.133 -0.046		
ADRAT1LN ADRAT1QU	C: Linear: Death Rate With Explicit Mention Of Alcohol C: Quadratic: Death Rate With Explicit Mention Of Alcohol								0.249 0.083
IFEM180 IFEM181	Female Interaction Of ADRATILN Female Interaction Of ADRATIQU								-0.354 -0.570
V18FLN	C: Linear: Marijuana Posession Arrest Rate				-0.224				
V18BLN V18BQU V18BCU V18BQR	C: Linear: Marijuana Sale/Manufacture Arrest Rate C: Quadratic: Marijuana Sale/Manufacture Arrest Rate C: Cubic: Marijuana Sale/Manufacture Arrest Rate C: Quartic: Marijuana Sale/Manufacture Arrest Rate			0.089				-0.059 0.037 0.111	-0.007 0.089 0.358 0.082
IFEM190 IFEM191 IFEM192 IFEM193	Female Interaction Of V18BLN Female Interaction Of V18BQU Female Interaction Of V18BCU Female Interaction Of V18BOR								-0.051 -0.167 -0.570 -0.137
IBLK190 IBLK191 IBLK192	Black Interaction Of V18BLN Black Interaction Of V18BQU Black Interaction Of V18BCU							0.092 0.093 -0.211	
IHISP190	Hispanic Interaction Of V18BLN			-0.143					0.054
V18ALN V18AQU	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate								-0.264 -0.275
IHISP200 IHISP201	Hispanic Interaction Of V18ALN Hispanic Interaction Of V18AQU								0.135 0.432
V18LN V18QU	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate						0.133 -0.073		
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992						-0.051 -0.024		
IBLK235 IBLK236	Black Interaction Of DRATELN Black Interaction Of DRATEQU						0.066 -0.035		
RH43ALN RH43AQU	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units								0.009 0.301
IBLK240 IBLK241	Black Interaction Of RH43ALN Black Interaction Of RH43AQU								0.165 -0.379
RP80ALN	T: Linear: Recoded Median Household Income		0.132						
BLK250 B64DISLN B64DISQU B64DISCU	Black Interaction Of RP80ALN B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability		-0.368				0.073	0.189 -0.038 -0.088	
IHISP260 IHISP261 IHISP262	Hispanic Interaction Of B64DISLN Hispanic Interaction Of B64DISQU Hispanic Interaction Of B64DISCU							-0.063 -0.008 0.210	
IOTH260 IOTH261	Other Interaction Of B64DISLN Other Interaction Of B64DISQU							-0.344 0.652	
BAGE34LN BAGE34QU BAGE34CU BAGE34QR	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years						0.012		-0.054 0.008 0.050 -0.014
IBLK265 IBLK266 IBLK267 IBLK268	Black Interaction Of BAGE34LN Black Interaction Of BAGE34QU Black Interaction Of BAGE34CU Black Interaction Of BAGE34QR								-0.067 -0.040 -0.039 0.056
BAGE44LN BAGE44QU BAGE44CU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years	0.013				-0.164		-0.011	-0.012 -0.001 0.125
IOTH270	Other Interaction Of BAGE44LN	2.045				0.007	0.114	0.855	
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years					-0.003 -0.040 0.039 0.034	0.116		
IOTH275	Other Interaction Of BAGE54LN		V VIDV				-1.094		
BAGE64LN BAGE64QU BAGE64CU	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years		-0.090 -0.016 0.068				-0.025 -0.059 0.038		
IBLK280 IBLK281 IBLK282	Black Interaction Of BAGE64LN Black Interaction Of BAGE64QU Black Interaction Of BAGE64CU						0.179 0.025 -0.136		
IHISP280 IHISP281 IHISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		0.194 -0.088 -0.173						
BCUBANLN BCUBANQU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban								-0.193 -0.388

			BIG CI	TTY			REMAI	NDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed	0.033	0.076 0.010			0.044 -0.062 0.011 0.028	0.061 -0.054 0.042 0.005	0.064 -0.097 -0.051 0.007	
IBLK295 IBLK296 IBLK297 IBLK298	Black Interaction Of BFNOTLN Black Interaction Of BFNOTQU Black Interaction Of BFNOTCU Black Interaction Of BFNOTQR						-0.144 0.204 -0.085 -0.070		
IHISP295 IHISP296 IHISP297	Hispanic Interaction Of BFNOTLN Hispanic Interaction Of BFNOTQU Hispanic Interaction Of BFNOTCU		-0.123 0.118				-0.187 -0.162 -0.185		
IOTH295 IOTH296 IOTH297 IOTH298	Other Interaction Of BFNOTLN Other Interaction Of BFNOTQU Other Interaction Of BFNOTCU Other Interaction Of BFNOTQR	0.657						0.544 0.523 0.228 0.352	
BINDIALN BINDIAQU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut						0.137 -0.094		-0.024 0.253
IFEM300 IFEM301	Female Interaction Of BINDIALN Female Interaction Of BINDIAQU						-0.061 0.127		
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU								-0.160 -0.371
BMNOTLN BMNOTQU BMNOTCU BMNOTQR	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed B: Quartic: Percent Males Separated, Divorced or Widowed							-0.045 0.011 -0.021 0.021	
IHISP305 IHISP306 IHISP307 IHISP308	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU Hispanic Interaction Of BMNOTCU Hispanic Interaction Of BMNOTQR							0.081 0.009 0.236 -0.092	
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.134 0.106		0.256 0.104		-0.051		-0.195 -0.194 -0.034 -0.042
IFEM310 IFEM311 IFEM312 IFEM313	Female Interaction Of BPOVERLN Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU Female Interaction Of BPOVERQR				-0.040 -0.199		0.192		0.034 0.233 -0.165 0.102
IBLK310 IBLK311	Black Interaction Of BPOVERLN Black Interaction Of BPOVERQU		-0.299 -0.123						
BPRICALN BPRICAQU	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican						0.231 0.036		
IBLK315 IBLK316	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU						-0.332 0.479		
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree			0.029				0.069 0.016 0.074 -0.037	0.144
IFEM320 IFEM321 IFEM322 IFEM323	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU Female Interaction Of BSCHASCU Female Interaction Of BSCHASQR							-0.001 0.018 0.047 0.059	
IBLK320	Black Interaction Of BSCHASLN			-0.279					
IOTH320 IOTH321	Other Interaction Of BSCHASLN Other Interaction Of BSCHASQU							-0.410 0.971	
PASIANLN PASIANQU PASIANCU	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander				-0.032			0.101 0.034 -0.085	
IBLK325 IBLK326 IBLK327	Black Interaction Of PASIANLN Black Interaction Of PASIANQU Black Interaction Of PASIANCU							-0.083 0.104 0.321	
IOTH325	Other Interaction Of PASIANLN								-0.664
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Limear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban			-0.021 0.127 0.100				0.022 0.139 -0.093 0.037	
PINDIALN PINDIAQU PINDIACU	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut					0.064 -0.084 0.097	0.096 0.093	0.061 0.026 0.113	
IFEM335 IFEM336	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU						-0.047 -0.110		
IOTH335 IOTH336 IOTH337	Other Interaction Of PINDIALN Other Interaction Of PINDIAQU Other Interaction Of PINDIACU							-0.255 0.425 -0.934	

Appendix F: SignificanceProbabilities for Fixed Effect Coefficients

F1. Significance Probabilities for Fixed Effect Coefficients for Past Month Alcohol Use by Age

		BIG CITY				REMAINDER	l .		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	29.20	0.00	0.00	0.00	0.12	0.00	0.00	0.00
FEMBLCK	O: Black Interaction Of FEMALE	15.10	5.21	0.00	0.00	46.56	27.70	4.71	0.00
FEMHISP	O: Hispanic Interaction Of FEMALE	0.36	0.00	0.00	0.00	98.60	0.04	0.04	0.00
FEMOTHR	O: Other Interaction Of FEMALE	41.04	30.60	0.00	0.00	2.09	95.46	44.24	0.00
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	0.00 95.35 0.00	0.00 0.00 0.00	0.00 66.62 0.00	0.00 58.11 0.00	5.11 99.80 0.03	0.00 37.20 0.00	0.00 43.40 0.00	0.00 0.53 0.00
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator						32.62 2.01 50.63		30.40 0.06 39.18
IOTH7 IOTH8 IOTH9	Other Interaction Of REGNOREA Other Interaction Of REGSOUTH Other Interaction Of REGWEST						0.00 0.00 0.00		0.00 0.00 0.00
UCLASS9 PAGE18LN	T: Underclass Indicator T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years				0.00		0.00		0.00 0.19 0.00
PAGE18QU IOTH25 IOTH26	Other Interaction Of PAGE18LN Other Interaction Of PAGE18OU								0.00 0.00 0.00
PAGE24LN PAGE24QU	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years						7.83 0.00		0.00
IFEM30 IFEM31	Female Interaction Of PAGE24LN Female Interaction Of PAGE24QU						69.15 0.00		
PAGE34LN	T: Linear: Percent Persons 25-34 Years	0.01							
PAGE44LN PAGE44QU PAGE44CU PAGE44QR	T: Linear: Percent Persons 35-44 Years T: Quadratic: Percent Persons 35-44 Years T: Cubic: Percent Persons 35-44 Years T: Quartic: Percent Persons 35-44 Years				0.00				0.00 0.00 15.17 0.00
PAGE64LN	T: Linear: Percent Persons 55-64 Years	1.90							
PSCH8LN	T: Linear: Percent 0-8 Years Of School		0.00						
PSCHCOLN	T: Linear: Bachelors, Graduate, Or Professional Degree		0.00				0.00	0.00	
IBLK75	Black Interaction Of PSCHCOLN						0.00	0.01	
PSCHSCLN	T: Linear: Percent Some College And No Degree		0.00						
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Quartic: Percent Families Below Poverty Level					2.48 22.11 0.00 0.27			
IBLK85 IBLK86 IBLK87 IBLK88	Black Interaction Of PPOVERLN Black Interaction Of PPOVERQU Black Interaction Of PPOVERCU Black Interaction Of PPOVERQR					2.49 50.49 36.10 3.40			
PBLACKLN	T: Linear: Percent Black Nonhispanic		0.00						
POTHLN POTHQU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity						7.76 0.00		
PFLABLN PFLABQU	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force	6.05 0.07							
PFNEVLN	T: Linear: Percent Females Never Married						0.00		
PFNOTLN PFNOTQU PFNOTCU PFNOTQR	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed T: Cubic: Percent Females Separated, Divorced Or Widowed T: Quartic: Percent Females Separated, Divorced Or Widowed				0.00 20.07 0.00 0.00				
PMNEVLN PMNEVQU PMNEVCU PMNEVQR	T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married T: Cubic: Percent Males Never Married T: Quartic: Percent Males Never Married			0.00		0.22		0.21 6.51 0.94	0.00 0.00 54.64 0.00
IFEM140 IFEM141	Female Interaction Of PMNEVLN Female Interaction Of PMNEVQU								0.00 0.00
IBLK140	Black Interaction Of PMNEVLN					0.53			
IHISP140	Hispanic Interaction Of PMNEVLN			0.00					
IOTH140 IOTH141 IOTH142	Other Interaction Of PMNEVLN Other Interaction Of PMNEVQU Other Interaction Of PMNEVCU			0.00				24.42 42.50 0.00	
P40HULN	T: Linear: Percent Housing Units Built 1940-1949					0.01			
IFEM160	Female Interaction Of P40HULN					0.04			
PRENTLN	T: Linear: Percent Housing Rented			0.48		80.55			
IBLK165	Black Interaction Of PRENTLN			0.01					

		BIG CITY				REMAINDER	<u> </u>		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
ADRATELN	C: Linear: Death Rate For All Alcohol-Related Cases		86.86						
ADRATEQU	C: Quadratic: Death Rate For All Alcohol-Related Cases		0.00						
V18FLN IOTH185	C: Linear: Marijuana Posession Arrest Rate Other Interaction Of V18FLN								0.00
V18ALN	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate		86.81		0.00				0.00
V18AQU V18ACU V18AQR	C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Cubic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quartic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate		0.01		0.00 0.00				0.00 4.60 0.00
IHISP200 IHISP201	Hispanic Interaction Of V18ALN Hispanic Interaction Of V18AQU		0.00 0.00						
V18LN	C: Linear: Total Drug Abuse Violations Arrest Rate								0.00
DRATELN DRATEQU DRATECU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992 C: Cubic: Mean Drug Client Treatment Rate 1991 & 1992					1.22 49.97 22.78			
IOTH235 IOTH236 IOTH237	Other Interaction Of DRATELN Other Interaction Of DRATEQU Other Interaction Of DRATECU					86.06 55.24 0.05			
RH61ALN RH61AQU	T: Linear: Recoded Median Value Of Owner Occupied HUs T: Quadratic: Recoded Median Value Of Owner Occupied HUs		83.25 52.81						0.00
IHISP245 IHISP246	Hispanic Interaction Of RH61ALN Hispanic Interaction Of RH61AQU		0.01 0.00						
RP80ALN RP80AQU	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income				0.00 0.00				
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992		47.83 37.47 48.01						
IFEM255 IFEM256 IFEM257	Female Interaction Of ARATELN Female Interaction Of ARATEQU Female Interaction Of ARATECU		0.12 5.10 0.00						
B64DISLN B64DISQU B64DISCU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability				0.00 0.00		26.54 53.81 0.00		0.00
IHISP260 IHISP261 IHISP262	Hispanic Interaction Of B64DISLN Hispanic Interaction Of B64DISQU Hispanic Interaction Of B64DISCU						45.33 2.01 0.61		0.00
BAGE34LN	B: Linear: Percent Persons 25-34 Years							1.59	
IOTH265	Other Interaction Of BAGE34LN							0.00	
BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years								0.00 7.53
IHISP270 IHISP271	Hispanic Interaction Of BAGE44LN Hispanic Interaction Of BAGE44QU								0.00 0.00
BAGE54LN	B: Linear: Percent Persons 45-54 Years				0.00				0.00
IBLK275	Black Interaction Of BAGE54LN								0.00
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander							90.86 57.39 0.02 57.06	0.00 29.56 22.18
IOTH285 IOTH286	Other Interaction Of BASIANLN Other Interaction Of BASIANQU							0.05 97.29	0.00
IOTH287 IOTH288	Other Interaction Of BASIANČÚ Other Interaction Of BASIANQR							0.01 0.00	0.00
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban		7.86			27.48 49.56 39.14			
IHISP290	Hispanic Interaction Of BCUBANLN		0.00						
IOTH290 IOTH291 IOTH292	Other Interaction Of BCUBANLN Other Interaction Of BCUBANQU Other Interaction Of BCUBANCU					53.03 86.43 5.64			
BFNOTLN BFNOTQU BFNOTCU	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed		0.05 91.65 0.22				3.85 0.19		
BFNOTQR IOTH295	B: Quartic: Percent Females Separated, Divorced or Widowed Other Interaction Of BFNOTLN		0.02				22.41		
IOTH296 IOTH297 IOTH298	Other Interaction Of BFNOTQU Other Interaction Of BFNOTCU Other Interaction Of BFNOTQR		97.61 0.00 0.00				0.00		
BINDIALN BINDIAQU BINDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut					1.69	26.85 43.34 0.00		
IFEM300 IFEM301 IFEM302	Female Interaction Of BINDIALN Female Interaction Of BINDIAQU Female Interaction Of BINDIACU					0.04	/3.6/ 80.16 0.00	·	
IBLK300	Black Interaction Of BINDIALN					2.38			
BMNOTUU BMNOTUU BMNOTUU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed	4.45							0.00 39.55 0.00

		BIG CITY				REMAINDER	-		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IHISP305 IHISP306 IHISP307	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU Hispanic Interaction Of BMNOTCU	0.12							0.24 59.03 0.00
BPOVERLN	B: Linear: Percent Families Below Poverty Level			0.00					
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: L.mear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican	5.52 60.48 0.11				73.88 99.91 46.06	37.62 0.05 49.01 0.00		
IBLK315 IBLK316	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU						18.30 0.05		
BSCHASLN BSCHASQU BSCHASCU	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree				0.00 0.00 0.00				0.00
IBLK320 IBLK321 IBLK322	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU				0.00 0.00 0.00				
PASIANLN PASIANQU	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander			0.01 0.00					
IBLK325 IBLK326	Black Interaction Of PASIANLN Black Interaction Of PASIANQU			0.07 0.00					
PCUBANLN PCUBANQU PCUBANCU	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban				0.00 1.19 0.00				
IFEM330 IFEM331 IFEM332	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU				0.41 0.00 0.00				
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Aleut			0.00 0.00 9.34 0.00					30.06 0.00 0.00 0.00
IFEM335 IFEM336 IFEM337 IFEM338	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU Female Interaction Of PINDIACU Female Interaction Of PINDIAQR								0.00 0.00 0.00 0.00

F2. Significance Probabilities for Fixed Effect Coefficients for Past Month Any Illicit Drug Use by Age

		BIG CITY				REMAINDER			
Varibable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
UCLASS9	T: Underclass Indicator						0.55		
FEMALE	O: Female Indicator	64.64	0.00	0.00	0.00	48.67	0.00	0.00	0.00
FEMBLCK	O: Black Interaction Of FEMALE	0.41	0.00	0.01	9.18	88.02	31.63	43.68	1.24
FEMHISP	O: Hispanic Interaction Of FEMALE	10.40	27.91	0.00	0.00		28.86	40.07	59.64
FEMOTHR	O: Other Interaction Of FEMALE	23.94	69.98	0.09	0.00	10.51	56.72	0.00	0.01
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	59.06 4.56 0.02	4.39 0.00 0.00	0.00 0.00 0.00	0.00 0.00 0.00	18.51 48.56	0.31 0.14 0.20	90.64 24.39 0.00	0.00 0.15 0.00
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator						4.96 12.14 0.01	1.92 22.19 0.00	
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban					72.37 29.66 16.25 53.05	0.13 1.67 0.88 0.39	51.18 97.05 84.57 64.75	
IBLK10 IBLK11 IBLK12 IBLK13	Black Interaction Of PDENLEV1 Black Interaction Of PDENLEV2 Black Interaction Of PDENLEV3 Black Interaction Of PDENLEV4					55.05	4.41 17.62 1.87 0.98	01.75	
IHISP10 IHISP11 IHISP12 IHISP13	Hispanic Interaction Of PDENLEV1 Hispanic Interaction Of PDENLEV2 Hispanic Interaction Of PDENLEV3 Hispanic Interaction Of PDENLEV4							2.08 0.38 12.70 13.06	
PHH1PLN PHH1PQU	T: Linear: Percent One Person Households T: Quadratic: Percent One Person Households						9.71 0.03		
POPRMLN POPRMQU	T: Linear: Average Persons Per Room T: Quadratic: Average Persons Per Room			38.68				0.00 0.00	
IFEM20 IFEM21	Female Interaction Of POPRMLN Female Interaction Of POPRMQU							0.93 0.04	
PAGE18LN PAGE18QU PAGE18CU	T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years T: Cubic: Percent Persons 0-18 Years						4.17 37.39 0.00		
IFEM25 IFEM26 IFEM27	Female Interaction Of PAGE18LN Female Interaction Of PAGE18QU Female Interaction Of PAGE18CU						0.09 98.64 0.01		
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years T: Cubic: Percent Persons 19-24 Years T: Quartic: Percent Persons 19-24 Years							0.00 4.50 0.00 37.40	
IHISP30 IHISP31 IHISP32 IHISP33	Hispanic Interaction Of PAGE24LN Hispanic Interaction Of PAGE24QU Hispanic Interaction Of PAGE24CU Hispanic Interaction Of PAGE24QR							66.58 64.69 81.90 62.74	
PAGE34LN PAGE34QU	T: Linear: Percent Persons 25-34 Years T: Quadratic: Percent Persons 25-34 Years					96.79 0.02	41.91 2.16		
IHISP35 IHISP36	Hispanic Interaction Of PAGE34LN Hispanic Interaction Of PAGE34QU						48.67 1.50		
PAGE44LN	T: Linear: Percent Persons 35-44 Years							0.63	
PAGE54LN PAGE54QU PAGE54CU	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years T: Cubic: Percent Persons 45-54 Years				0.00 0.68		0.00 6.44 0.16		0.00 0.00
IBLK45 IBLK46	Black Interaction Of PAGE54LN Black Interaction Of PAGE54QU				$0.00 \\ 0.00$				
IHISP45	Hispanic Interaction Of PAGE54LN						1.71		
PSCH8LN PSCH8QU	T: Linear: Percent 0-8 Years Of School T: Quadratic: Percent 0-8 Years Of School							2.32 14.81	
IBLK55 IBLK56	Black Interaction Of PSCH8LN Black Interaction Of PSCH8QU							0.16 0.03	
PSCH12LN PSCH12QU PSCH12CU PSCH12QR	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma T: Quartic: Percent 9-12 Years & No High School Diploma	95.60 0.46 18.91 0.48			0.00	18.22 0.64			4.37 0.00
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU					35.76 1.18			
PSCHASLN PSCHASQU	T: Linear: Percent Associates Degree T: Quadratic: Percent Associates Degree							44.93 0.00	
IBLK65 IBLK66	Black Interaction Of PSCHASLN Black Interaction Of PSCHASQU							30.45 0.80	
PSCHCOLN	T: Linear: Bachelors, Graduate, Or Professional Degree	1				11.51			87.93

		BIG CITY				REMAINDER	t		
Varibable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
PSCHCOQU PSCHCOCU	Cubic: Bachelors, Graduate, Or Professional Degree Cubic: Bachelors, Graduate, Or Professional Degree					88.61 1.19			0.00
IFEM75 IFEM76	Female Interaction Of PSCHCOLN Female Interaction Of PSCHCOQU					83.45 73.24			
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Quartic: Percent Families Below Poverty Level	0.58 72.28 0.00				15.37 88.13 15.02 3.83	0.00	0.09 25.01 2.96 0.01	
IFEM85	Female Interaction Of PPOVERLN						28.49		
PBLACKLN PBLACKQU PBLACKCU PBLACKQR	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic T: Cubic: Percent Black Nonhispanic T: Quartic: Percent Black Nonhispanic			0.00 73.24 0.00			1.80 34.80 40.82 0.01	5.57 1.16 37.41	
PHISPLN PHISPQU PHISPCU PHISPQR	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic T: Cubic: Percent Hispanic T: Quartic: Percent Hispanic								0.66 0.23 0.00 0.05
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity					15.11 96.69 53.72			
PHHF18LN PHHF18QU PHHF18CU PHHF18QR	T: Linear: % Female-Headed HH W/No Spouse & Chid Under 18 T: Quadratic: % Female-Headed HH W/No Spouse & Chid Under 18 T: Cubic: % Female-Headed HH W/No Spouse & Chid Under 18 T: Quartic: % Female-Headed HH W/No Spouse & Chid Under 18					47.63 59.87 69.18 25.38			29.88
IFEM120	Female Interaction Of PHHF18LN								0.22
PFNOTUN PFNOTQU	1: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed			0.00				92.70 4.77	
IHISP135 IHISP136	Hispanic Interaction Of PFNOTLN Hispanic Interaction Of PFNOTQU							7.57 40.16	
PMNEVLN PMNOTLN	T: Linear: Percent Males Never Married T: Linear: Percent Males Separated, Divorced Or Widowed				0.00			0.21	
PMNOTQU	T: Quadratic: Percent Males Separated, Divorced Or Widowed							2.95	
IHISP150 IHISP151	Hispanic Interaction Of PMNOTLN Hispanic Interaction Of PMNOTQU							75.31 6.65	
POLDHULN POLDHUQU POLDHUCU	T: Linear: Percent Housing Units Built 1939 Or Earlier T: Quadratic: Percent Housing Units Built 1939 Or Earlier T: Cubic: Percent Housing Units Built 1939 Or Earlier			0.00 3.75 0.00			0.01 35.15	81.76 0.01	
P40HULN	T: Linear: Percent Housing Units Built 1940-1949					55.89		0.00	
PRENTLN PRENTQU PRENTCU PRENTQR	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented T: Cubic: Percent Housing Rented T: Quartic: Percent Housing Rented					11.66	32.47	0.00 0.13 3.96 0.00	0.00 0.00 46.39 0.00
IFEM165	Female Interaction Of PRENTLN					0.14			
IBLK165 IBLK166 IBLK167	Black Interaction Of PRENTLN Black Interaction Of PRENTQU Black Interaction Of PRENTCU							0.31 37.29 2.03	
IHISP165 ADRATELN	Hispanic Interaction Of PRENTLN C: Linear: Death Rate For All Alcohol-Related Cases		82.11			54.19	51.22 0.37		
ADRATEQU ADRATECU ADRATEQR	C: Quadratic: Death Rate For All Alcohol-Related Cases C: Cubic: Death Rate For All Alcohol-Related Cases C: Quartic: Death Rate For All Alcohol-Related Cases C: Quartic: Death Rate For All Alcohol-Related Cases		80.43 0.01			44.92 9.55	39.73 89.07 0.00		
IBLK175 IBLK176 IBLK177 IBLK178	Black Interaction Of ADRATELN Black Interaction Of ADRATEQU Black Interaction Of ADRATECU Black Interaction Of ADRATECU						47.11 11.73 43.82 0.12		
V18FLN	C: Linear: Marijuana Posession Arrest Rate					71.95			
V18ELN V18EQU V18ECU V18EQR	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate C: Quartic: Opium/Cocaine & Deriv Posession Arrest Rate						19.71 17.38 34.23 0.00	0.01	0.00 0.00 0.05 76.84
IFEM195 IFEM196 IFEM197 IFEM198	Female Interaction Of V18ELN Female Interaction Of V18EQU Female Interaction Of V18ECU Female Interaction Of V18EQR						0.80 10.92 43.77 0.00		
IBLK195	Black Interaction Of V18ELN						0.24		
IHISP195	Hispanic Interaction Of V18ELN	22.64	0.00			22.40	0.16	95.05	
V18ALN V18AQU V18ACU	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Cubic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate	32.64 1.42 19.69	0.00			33.40 17.74 0.16		85.95	
IFEM200 IFEM201 IFEM202	Female Interaction Of V18ALN Female Interaction Of V18AQU Female Interaction Of V18ACU	99.67 27.95 1.11							
IBLK200 IBLK201 IBLK202	Black Interaction Of V18ALN Black Interaction Of V18AQU Black Interaction Of V18ACU		.	<u> </u>		19.64 55.11 6.06	.	<u>. </u>	
IHISP200	Hispanic Interaction Of V18ALN		0.00						
V18LN	C: Linear: Total Drug Abuse Violations Arrest Rate		8.26	84.55	0.26			0.07	16.24

T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

		BIG CITY				REMAINDER			
Varibable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
V18QU V18CU V18QR	C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate			1.59 57.21 0.02				0.05 57.05 0.02	0.00
IFEM225 IFEM226 IFEM227 IFEM228	Female Interaction Of V18LN Female Interaction Of V18QU Female Interaction Of V18CU Female Interaction Of V18OR							30.93 4.42 64.77 3.73	
IHISP225	Hispanic Interaction Of V18LN		86.80						
VIOLLN VIOLQU	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate				40.38 1.52				0.00 0.00
IBLK230 IBLK231	Black Interaction Of VIOLLN Black Interaction Of VIOLQU				0.00 3.40				
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992					0.68	45.77 0.34	37.47 0.00	
IFEM235 IFEM236	Female Interaction Of DRATELN Female Interaction Of DRATEQU							4.44 0.00	
RH43ALN RH43AQU RH43ACU RH43AQR	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units T: Cubic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units	58.26 53.33 41.72 0.06						99.05	0.00 0.00
IBLK240 IBLK241	Black Interaction Of RH43ALN Black Interaction Of RH43AQU	0.00							80.90 0.00
RH61ALN RH61AQU RH61ACU	T: Linear: Recoded Median Value Of Owner Occupied HUs T: Quadratic: Recoded Median Value Of Owner Occupied HUs T: Cubic: Recoded Median Value Of Owner Occupied HUs						60.84 87.63 0.00		0.00
RP80ALN RP80AQU RP80ACU RP80AQR	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income T: Quatric: Recoded Median Household Income T: Quartic: Recoded Median Household Income					8.20 70.51 11.44 34.09		20.29	
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992	2.33				89.33 56.76 94.43			
B64DISLN B64DISQU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability					94.94 20.83	87.27 13.54		
IBLK260	Black Interaction Of B64DISLN						2.11		
BAGE34LN BAGE44LN	B: Linear: Percent Persons 25-34 Years B: Linear: Percent Persons 35-44 Years					0.19	0.01		
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years					0.19	59.98 32.73 0.25	51.67 0.00 4.57 0.00	0.09 0.00 0.00
IFEM275 IFEM276 IFEM277	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54QU Female Interaction Of BAGE54QR						65.53 99.78 0.00	94.23 91.20 20.49 0.00	22.39 0.00 0.00
IFEM278 IHISP275 IHISP276 IHISP277	Hispanic Interaction Of BAGE54LN Hispanic Interaction Of BAGE54QU Hispanic Interaction Of BAGE54QU							20.63 77.50 5.36	
BAGE64LN BAGE64QU BAGE64CU	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years	34.53	0.00 0.00 53.23	39.60 6.11 0.00	0.00 0.00	27.42 79.11 9.93	0.17 2.60 0.61		0.00 0.01 0.00
IFEM280 IFEM281 IFEM282	Female Interaction Of BAGE64LN Female Interaction Of BAGE64QU Female Interaction Of BAGE64CU	0.34			0.00 0.00	52.09 71.10 1.46			0.00 0.00 26.4
IBLK280 IBLK281	Black Interaction Of BAGE64LN Black Interaction Of BAGE64QU								0.06 32.25
IHISP280 IHISP281 IHISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		0.12 56.69 0.00	71.49 12.48 0.00			87.64 72.95 32.52		
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander		74.24 3.18	51.28 93.90 0.01 0.59			3.98	33.12 0.00 85.32	0.08 2.25 0.00 0.00
IBLK285 IBLK286 IBLK287 IBLK288	Black Interaction Of BASIANLN Black Interaction Of BASIANQU Black Interaction Of BASIANCU Black Interaction Of BASIANQR								26.24 0.00 0.14 0.00
IHISP285 IHISP286 IHISP287 IHISP288	Hispanic Interaction Of BASIANLN Hispanic Interaction Of BASIANQU Hispanic Interaction Of BASIANCU Hispanic Interaction Of BASIANQR		· ·	10.80 0.08 0.20 0.00			· ·	81.83 48.99 7.31	
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban			0.00			0.00	18.70	40.55 0.00 0.00
IFEM290	B: Cubic: Percent Hispanics: Cuban Female Interaction Of BCUBANLN							3.34	0.00
IBLK290 IBLK291 IBLK292	Black Interaction Of BCUBANLN Black Interaction Of BCUBANQU Black Interaction Of BCUBANCU								90.40 0.20 0.00
BFNOTLN	B: Linear: Percent Females Separated, Divorced or Widowed	1.80		90.00			28.43	0.00	0.00

T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

		BIG CITY				REMAINDER	١		
Varibable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BFNOTQU BFNOTCU BFNOTQR	B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed	23.28 19.05 0.46		0.08			73.10 0.09 0.48	3.52	
IFEM295 IFEM296 IFEM297	Female Interaction Of BFNOTLN Female Interaction Of BFNOTQU Female Interaction Of BFNOTCU						88.46 9.85 0.00		
IBLK295 IBLK296	Black Interaction Of BFNOTLN Black Interaction Of BFNOTQU			4.48 0.00				$0.08 \\ 0.02$	
IHISP295 IHISP296 IHISP297 IHISP298	Hispanic Interaction Of BFNOTLN Hispanic Interaction Of BFNOTQU Hispanic Interaction Of BFNOTCU Hispanic Interaction Of BFNOTQR						92.14 46.17 79.42 24.17		
BINDIALN BINDIAQU BINDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut					6.79 6.55 0.33	1.99	0.00	1.04 2.54
IBLK300	Black Interaction Of BINDIALN					0.24	45.47		1.67
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU				0.00	11.07	42.26	76.21	1.62 0.00
BMNOTLN BMNOTQU BMNOTCU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratuc: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed				0.00 16.19 0.00	11.87 51.44	42.26 36.06	76.21 70.91 1.10	
IFEM305 IFEM306 IFEM307	Female Interaction Of BMNOTLN Female Interaction OF BMNOTQU Female Interaction OF BMNOTCU				0.00 99.19 0.00			2.44 81.91 0.10	
IHISP305 IHISP306	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU						67.77 17.44		
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.00 48.17 26.14 0.73				36.84		0.00 0.81 0.03 0.00
IFEM310 IFEM311 IFEM312 IFEM313	Female Interaction Of BPOVERLN Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU Female Interaction Of BPOVERQR						0.51		1.45 10.93 0.00 0.00
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican	9.63	0.00 0.00			12.61 7.30 86.81 76.78	80.94 89.16 2.71	23.33 7.76 48.66	
IFEM315 IFEM316 IFEM317 IFEM318	Female Interaction Of BPRICALN Female Interaction Of BPRICAQU Female Interaction Of BPRICACU Female Interaction Of BPRICAQR					66.56 25.37 61.94 5.13			
IBLK315 IBLK316 IBLK317	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU Black Interaction Of BPRICACU		36.24 0.00				36.64 5.82	16.42 48.41 7.66	
IHISP315 IHISP316 IHISP317	Hispanic Interaction Of BPRICALN Hispanic Interaction Of BPRICAQU Hispanic Interaction Of BPRICACU						33.73 88.73 5.73		
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree		0.20 15.77			29.12 47.22 77.76 12.90	42.56 2.05 9.98	21.96 6.83 0.01	0.00 0.15 20.91
IFEM320 IFEM321 IFEM322 IFEM323	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU Female Interaction Of BSCHASCU Female Interaction Of BSCHASQR					33.96 15.65 41.80 0.79			
IBLK320 IBLK321 IBLK322	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU					73.42 19.99 2.19		64.11 27.97 15.99	
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander				0.00		3.35 99.45 0.43 0.52	94.00 0.50 0.00	
IFEM325 IFEM326 IFEM327 IFEM328	Female Interaction Of PASIANLN Female Interaction Of PASIANQU Female Interaction Of PASIANCU Female Interaction Of PASIANCU			·			42.13 11.45 67.28 0.07		
IHISP325 IHISP326 IHISP327	Hispanic Interaction Of PASIANLN Hispanic Interaction Of PASIANQU Hispanic Interaction Of PASIANCU						75.83 26.60 8.32		
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban	26.15 0.38 5.07		0.16 21.33		22.66 5.21 16.69		63.25 5.02 68.25 47.64	
IFEM330 IFEM331 IFEM332	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU					38.74 20.49 0.49			
IBLK330 IBLK331 IBLK332	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU	85.50 41.03 57.84							
PINDIALN PINDIAQU PINDIACU	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut	24.01 0.56 0.02		57.41 0.05			44.62		0.00 70.93 25.32

NOTE: T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

		BIG CITY				REMAINDER	t .		
Varibable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
PINDIAQR IFEM335 IFEM336 IFEM337 IFEM338 IHISP335 IHISP336	T: Quartic: Percent Pop: American Indian, Eskimo, Aleut Female Interaction Of PINDIAQU Female Interaction Of PINDIAQU Female Interaction Of PINDIACU Female Interaction Of PINDIAQR Hispanic Interaction Of PINDIALN Hispanic Interaction Of PINDIAQU			0.11 0.00					0.00 0.70 0.18 1.03 0.00

F3. Significance Probabilities for Fixed Effect Coefficients for Past Month Cigarette Use by Age

			BIG CI	TY			REMA	INDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	18.66	0.00	0.00	0.00	69.46	10.47	0.01	0.00
FEMBLCK	O: Black Interaction Of FEMALE	0.08	22.30	0.00	0.00	95.57	12.13	0.24	6.07
FEMHISP	O: Hispanic Interaction Of FEMALE	22.71	0.00	0.00	0.00			0.00	0.26
FEMOTHR	O: Other Interaction Of FEMALE	61.50	0.00	0.00	0.00	63.05	0.00	0.00	33.73
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	0.00 0.01 0.00	0.00 0.00 0.00	16.43 0.00 26.30	0.00 0.00 0.00	0.01 0.04	0.00 96.08	42.15 51.45 12.02	0.00 14.68 0.00
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator						0.69 45.28 62.65		
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban					27.60 0.35 4.30 0.50	18.67 32.48 26.97 95.89	10.88 4.90 88.49 83.20	
IHISP10 IHISP11	Hispanic Interaction Of PDENLEV1 Hispanic Interaction Of PDENLEV2							22.94 40.45	
POPRMLN POPRMQU POPRMCU	T: Linear: Average Persons Per Room T: Quadratic: Average Persons Per Room T: Cubic: Average Persons Per Room			0.00		79.47 26.43 3.06		48.67	
IFEM20	Female Interaction Of POPRMLN							0.09	
IBLK20 IBLK21 IBLK22	Black Interaction Of POPRMLN Black Interaction Of POPRMQU Black Interaction Of POPRMCU					58.00 65.78 2.41			
IOTH20 IOTH21 IOTH22	Other Interaction Of POPRMLN Other Interaction Of POPRMQU Other Interaction Of POPRMCU					1.82 57.88 0.33			
PAGE18LN PAGE18QU PAGE18CU	T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years T: Cubic: Percent Persons 0-18 Years		0.00				0.00 27.41 0.00		
IFEM25 IFEM26 IFEM27	Female Interaction Of PAGE18LN Female Interaction Of PAGE18QU Female Interaction Of PAGE18CU		0.00				0.88 2.96 0.00		
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years T: Cubic: Percent Persons 19-24 Years T: Quartic: Percent Persons 19-24 Years					6.32 0.62		20.06 0.15 0.95 5.18	21.56 0.00 0.17 0.00
IHISP30 IHISP31 IHISP32 IHISP33	Hispanic Interaction OT PAGE24LN Hispanic Interaction OT PAGE24QU Hispanic Interaction OT PAGE24CU Hispanic Interaction OT PAGE24QR							1.23 8.94 85.50 36.06	
IOTH30 IOTH31 IOTH32 IOTH33	Other Interaction Of PAGE24LN Other Interaction Of PAGE24QU Other Interaction Of PAGE24CU Other Interaction Of PAGE24QR								0.00 0.02 0.42 0.00
PAGE44LN	T: Linear: Percent Persons 35-44 Years							14.74	
PAGE54LN PAGE54QU	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years				0.00 72.44	51.19 1.40			0.00 0.00
IBLK45 IBLK46	Black Interaction Of PAGE54LN Black Interaction Of PAGE54QU				0.00 0.00	91.08 6.12			
PAGE64LN PAGE64QU PAGE64CU PAGE64QR	T: Linear: Percent Persons 55-64 Years T: Quadratic: Percent Persons 55-64 Years T: Cubic: Percent Persons 55-64 Years T: Quartic: Percent Persons 55-64 Years T: Quartic: Percent Persons 55-64 Years					15.22 41.91 74.00 0.18			
PSCH8LN	T: Linear: Percent 0-8 Years Of School					47.40			
IOTH55	Other Interaction Of PSCH8LN					0.24			
PSCH12LN PSCH12QU PSCH12CU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma					55.63 30.15 10.93			
PSCHASLN PSCHASQU PSCHASCU PSCHASQR	T: Linear: Percent Associates Degree T: Quadratic: Percent Associates Degree T: Cubic: Percent Associates Degree T: Quartic: Percent Associates Degree					56.16 22.97 41.91 0.00		53.02 77.20	
IFEM65 IFEM66 IFEM67 IFEM68	Female Interaction Of PSCHASLN Female Interaction Of PSCHASQU Female Interaction Of PSCHASCU Female Interaction Of PSCHASQR					35.64 8.51 91.64 0.02			
PSCHCOLN PSCHCOQU	T: Linear: Bachelors, Graduate, Or Professional Degree T: Quadratic: Bachelors, Graduate, Or Professional Degree			•		14.46 18.11		·	0.00 0.00
IFEM75 IFEM76	Female Interaction Of PSCHCOLN Female Interaction Of PSCHCOQU					6.27 0.82			

T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

			BIG C	ITY		REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	I: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level I: Cubic: Percent Families Below Poverty Level I: Quartic: Percent Families Below Poverty Level					77.08 34.98 9.53 89.49			0.00 0.00
IFEM85 IFEM86 IFEM87 IFEM88	Female Interaction Of PPOVERLN Female Interaction Of PPOVERQU Female Interaction Of PPOVERCU Female Interaction Of PPOVERQR					14.84 19.79 80.14 0.03			0.00 0.00
IOTH85 IOTH86 IOTH87	Other Interaction Of PPOVERLN Other Interaction Of PPOVERQU Other Interaction Of PPOVERCU					31.27 14.67 0.98			
PPUBASLN IFEM90	T: Linear: Percent Households With Public Assist Income Female Interaction Of PPUBASLN					11.10 1.79			
P64DISLN	T: Linear: Percent 16-64 With A Work Disability	0.68				35.32			
IOTH95	Other Interaction Of P64DISLN	0.59				6.36			
PBLACKLN PBLACKQU PBLACKCU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic T: Cubic: Percent Black Nonhispanic			0.87 0.59 0.03				9.66 0.04 0.77	0.01 0.00
PHISPLN PHISPQU PHISPCU PHISPQR	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic T: Cubic: Percent Hispanic T: Quartic: Percent Hispanic					36.24 37.78 26.20			0.00 0.00 0.00 0.00
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity					0.02 84.46 88.27			0.00
IOTH110	Other Interaction Of POTHLN								0.00
PHHF18LN PHHF18QU PHHF18CU PHHF18QR	T: Linear: % Female-Headed HH W/No Spouse & Chld Under 18 T: Quadratic: % Female-Headed HH W/No Spouse & Chld Under 18 T: Cubic: % Female-Headed HH W/No Spouse & Chld Under 18 T: Quartic: % Female-Headed HH W/No Spouse & Chld Under 18					50.39 99.10 26.23 81.69			0.00
IFEM120	Female Interaction Of PHHF18LN								0.00
PFLABLN PFLABQU PFLABCU PFLABQR	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force T: Cubic: Percent Females 16+ Years Old In Labor Force T: Quartic: Percent Females 16+ Years Old In Labor Force						2.11 78.75 0.02 0.02		
IBLK125 IBLK126 IBLK127 IBLK128	Black Interaction Of PFLABLN Black Interaction Of PFLABQU Black Interaction Of PFLABCU Black Interaction Of PFLABQR						93.27 55.82 43.02 1.47		
PFNEVLN PFNEVQU PFNEVCU PFNEVQR	T: Linear: Percent Females Never Married T: Quadratic: Percent Females Never Married T: Cubic: Percent Females Never Married T: Quartic: Percent Females Never Married								0.00 0.00 0.01 0.00
IHISP130 IHISP131 IHISP132 IHISP133	Hispanic Interaction Of PFNEVLN Hispanic Interaction Of PFNEVQU Hispanic Interaction Of PFNEVCU Hispanic Interaction Of PFNEVCR								96.53 9.53 88.95 0.00
PFNOTLN PFNOTQU PFNOTCU PFNOTQR	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed T: Cubic: Percent Females Separated, Divorced Or Widowed T: Quartic: Percent Females Separated, Divorced Or Widowed		0.00	91.65			32.64 78.97 28.17 8.59	5.49 18.06 0.07 0.03	
IHISP135 IHISP136 IHISP137 IHISP138	Hispanic Interaction Of PFNOTLN Hispanic Interaction Of PFNOTQU Hispanic Interaction Of PFNOTQU Hispanic Interaction Of PFNOTQU						0.57	79.07 79.51 58.54 7.09	
IOTH135 IOTH136 IOTH137 IOTH138	Other Interaction Of PFNOTLN Other Interaction Of PFNOTQU Other Interaction Of PFNOTCU Other Interaction Of PFNOTCU						36.03 23.14 0.10 0.00	7.07	
PMLABLN	T: Linear: Percent Males 16+ Years Old In Labor Force					56.43	0.00		
PMLABQU IOTH145 IOTH146	T: Quadratic: Percent Males 16+ Years Old In Labor Force Other Interaction Of PMLABLN Other Interaction Of PMLABQU					1.69 49.23 3.23			
PMNOTLN PMNOTQU	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed					3.23	61.54 52.04	0.00 62.39	
IFEM150 IFEM151	Female Interaction Of PMNOTLN Female Interaction Of PMNOTOU						1.01 0.47		
IHISP150	Hispanic Interaction Of PMNOTEN						V.T/	15.03	
IHISP151 POLDHULN	Hispanic Interaction Of PMNOTQU T: Linear: Percent Housing Units Built 1939 Or Earlier				0.00			70.62	0.00
P40HULN P40HUQU P40HUCU	T: Linear: Percent Housing Units Built 1940-1949 T: Quadratic: Percent Housing Units Built 1940-1949 T: Cubic: Percent Housing Units Built 1940-1949			10.21 18.70 99.98	5.00	65.83 63.79 7.17		32.36 24.99 85.11	0.00
P40HUQR IBLK160 IBLK161 IBLK162	T: Quartic: Percent Housing Units Built 1940-1949 Black Interaction Of P40HULN Black Interaction Of P40HUQU Black Interaction Of P40HUCU			5.44 0.15 0.00				0.86 6.17 98.27 50.56	
IBLK163 IOTH160 IOTH161	Black Interaction Of P40HUQR Other Interaction Of P40HULN Other Interaction Of P40HUQU					7.60 0.22		0.85	

			BIG C	TY		REMAINDER				
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+	
IOTH162	Other Interaction Of P40HUCU					6.37				
PRENTLN PRENTQU PRENTCU PRENTQR	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented T: Cubic: Percent Housing Rented T: Quartic: Percent Housing Rented						1.89 24.93 0.10	0.00	9.86 50.22 7.39 48.40	
IOTH165 IOTH166	Other Interaction Of PRENTLN Other Interaction Of PRENTQU						68.28 0.01			
ADRATELN ADRATEQU ADRATECU	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases C: Cubic: Death Rate For All Alcohol-Related Cases					42.91 17.55 2.84				
ADRATILN ADRATIQU ADRATICU ADRATIQR	C: Linear: Death Rate With Explicit Mention Of Alcohol C: Quadratic: Death Rate With Explicit Mention Of Alcohol C: Cubic: Death Rate With Explicit Mention Of Alcohol C: Quartic: Death Rate With Explicit Mention Of Alcohol		1.34 78.46 82.61 97.97				67.84 34.11 72.98 0.11		0.00 0.00 0.00 0.00	
IBLK180 IBLK181 IBLK182 IBLK183	Black Interaction Of ADRATILN Black Interaction Of ADRATIQU Black Interaction Of ADRATICU Black Interaction Of ADRATIQR		10.99 12.54 3.63 0.00				19.41 6.46 44.45 0.16			
V18FLN V18FQU V18FCU	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate C: Cubic: Marijuana Posession Arrest Rate			0.00		64.41		1.52 52.39 38.64		
V18FQR V18BLN	C: Quartic: Marijuana Posession Arrest Rate C: Linear: Marijuana Sale/Manufacture Arrest Rate				0.00			0.00 40.04		
IBLK190	Black Interaction Of V18BLN				0.00			0.01		
V18ELN V18EQU V18ECU V18EQR	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate C: Quartic: Opium/Cocaine & Deriv Posession Arrest Rate						80.34 98.09 0.21		0.00 60.42 15.33 0.00	
V18ALN	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate		25.97				1.57	63.12	0.00	
IBLK200 IHISP200	Black Interaction Of V18ALN Hispanic Interaction Of V18ALN		0.00						0.00	
V18LN V18QU V18CU V18QR	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate		45.24	0.44 0.00 0.00 0.19			6.19 27.56	7.80 0.08 0.00 6.22		
IFEM225 IFEM226 IFEM227 IFEM228	Female Interaction Of V18LN Female Interaction Of V18QU Female Interaction Of V18CU Female Interaction Of V18CU			0.19				0.02 2.36 6.57 0.43		
IHISP225	Hispanic Interaction Of V18LN		5.53					0.43		
VIOLLN VIOLQU	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate				0.00 75.90				0.00 0.00	
IBLK230 IBLK231	Black Interaction Of VIOLLN Black Interaction Of VIOLQU				0.00 0.00				0.00	
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Ouadratic: Mean Drug Client Treatment Rate 1991 & 1992					19.26	0.41 90.70		0.89 6.47	
IFEM235 IFEM236	Female Interaction Of DRATELN Female Interaction Of DRATEQU						15.73 22.31			
IOTH235 IOTH236	Other Interaction Of DRATELN Other Interaction Of DRATEQU								0.00 0.00	
RH43ALN RH43AQU RH43ACU	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units T: Cubic: Recoded Median Rents For Rental Units	66.83 1.16 97.17 90.74				14.95 4.50 0.00		0.32	0.00	
RH43AQR RP80ALN	T: Quartic: Recoded Median Rents For Rental Units T: Linear: Recoded Median Household Income	90.74	0.09	0.00		47.91	91.18	0.00	2.10	
RP80AQU RP80ACU RP80AQR	T: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income T: Quartic: Recoded Median Household Income					82.14 38.84 0.45	1.25 8.12		0.00	
IBLK250 IBLK251 IBLK252	Black Interaction Of RP80ALN Black Interaction Of RP80AQU Black Interaction Of RP80ACU						56.45 84.11 93.86			
IHISP250 IHISP251	Hispanic Interaction Of RP80ALN Hispanic Interaction Of RP80AQU		0.00						28.82 0.00	
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992	14.49				29.55 24.32 32.00				
B64DISLN B64DISQU B64DISCU B64DISQR	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability B: Quartic: Percent Persons 16-64 With A Work Disability			0.00	0.00	87.78 6.06 46.04 5.35	0.00	0.00		
IFEM260 IFEM261 IFEM262 IFEM263	Female Interaction Of B64DISLN Female Interaction Of B64DISQU Female Interaction Of B64DISCU Female Interaction Of B64DISCV					75.29 9.25 16.93 0.02				
BAGE34LN BAGE34QU BAGE34CU	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years				0.00	88.83 2.08		0.28 0.03 6.87	0.00 0.00 0.00	
IFEM265 IFEM266	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU								0.00 0.00	

			BIG CI	TY			REMA	INDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IFEM267	Female Interaction Of BAGE34CU								0.00
IBLK265 IOTH265	Black Interaction Of BAGE34LN Other Interaction Of BAGE34LN					1.63		0.00	0.00
IOTH266 IOTH267	Other Interaction Of BAGE34QU Other Interaction Of BAGE34CU					0.50		0.00	
BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years							5.70 0.00	
IHISP270 IHISP271	Hispanic Interaction Of BAGE44LN Hispanic Interaction Of BAGE44QU							25.74 1.72	
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Lunear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years					64.51 90.79 14.56	95.57 0.17 0.00	0.00 0.88 5.08 7.17	1.65 0.00 0.00
IFEM275 IFEM276 IFEM277	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU						37.51 2.09 0.00		85.96 0.00 14.88
IBLK275 IBLK276	Black Interaction Of BAGE54LN Black Interaction Of BAGE54QU						94.22 80.74		
IOTH275 IOTH276 IOTH277 IOTH278	Other Interaction Of BAGE54LN Other Interaction Of BAGE54QU Other Interaction Of BAGE54CU Other Interaction Of BAGE54QR					16.67 80.34 0.77		38.98 36.63 0.10 0.00	
BAGE64LN BAGE64QU BAGE64CU BAGE64QR	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years B: Quartic: Percent Persons 55-64 Years		36.79 0.00 49.19	56.55 19.32 58.04	0.00 0.00	0.91	29.28 62.11 5.82 11.28		0.00 68.56 0.00
IFEM280 IFEM281 IFEM282	Female Interaction Of BAGE64LN Female Interaction Of BAGE64QU Female Interaction Of BAGE64CU				0.00 24.61				0.00 0.00 92.42
IBLK280 IBLK281	Black Interaction Of BAGE64LN Black Interaction Of BAGE64QU								1.21 0.76
IHISP280 IHISP281 IHISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		89.38 67.45 2.47	0.00 1.13 53.65					
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander		0.08 9.72	69.16 32.30 35.49 30.23			95.06 1.60 0.05 17.13	0.00 91.23 1.46 0.00	0.00 7.44 0.00
IFEM285 IFEM286 IFEM287 IFEM288	Female Interaction Of BASIANLN Female Interaction Of BASIANOU Female Interaction Of BASIANCU Female Interaction Of BASIANCU						34.28 33.07 12.38 0.03		0.07 45.23 0.00
IHISP285 IHISP286 IHISP287 IHISP288	Hispanic Interaction Of BASIANLN Hispanic Interaction Of BASIANQU Hispanic Interaction Of BASIANCU Hispanic Interaction Of BASIANQR			86.33 0.00 86.51 0.00				1.20	
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban						12.06 37.38 7.35	20.62	
IFEM290	Female Interaction Of BCUBANLN							7.19	
BFNOTLN BFNOTQU BFNOTCU	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed					62.65 82.60 10.60		79.84 0.00	
IFEM295 IFEM296 IFEM297	Female Interaction Of BFNOTLN Female Interaction Of BFNOTQU Female Interaction Of BFNOTCU					91.77 35.85 0.08		21.84 0.00	
BINDIALN BINDIAQU BINDIACU BINDIAQR	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut B: Quartic: Percent Pop: American Indian, Eskimo, Ale		0.00	80.41 0.77		0.08	1.26 42.08 85.33 45.55		
IBLK300	Black Interaction Of BINDIALN						5.29		
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU			0.00 0.00					
IOTH300 IOTH301	Other Interaction Of BINDIALN Other Interaction Of BINDIAQU						0.01 0.00		
BMNOTLN BMNOTQU BMNOTCU BMNOTQR	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed B: Quartic: Percent Males Separated, Divorced or Widowed			0.00 0.63 3.27 0.00	0.00 0.00 0.00	64.82 0.18	0.38 98.04 8.19		0.00 0.00
IFEM305 IFEM306 IFEM307	Female Interaction Of BMNOTLN Female Interaction Of BMNOTQU Female Interaction Of BMNOTCU				0.00 0.00 0.29	68.58 0.15			
IHISP305 IHISP306 IHISP307 IHISP308	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU Hispanic Interaction Of BMNOTCU Hispanic Interaction Of BMNOTQR			56.38 42.61 6.37 0.00					11.40 0.00
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Lunear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.00 0.15 0.00 0.98			91.59 64.98 19.25	10.32 23.28 0.13 43.51		0.00 0.00 0.00

NOTE: T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

			BIG C	TY			REMA	AINDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IFEM310 IFEM311 IFEM312	Female Interaction Of BPOVERLN Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU								0.00 9.04 0.00
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican	30.32				33.42 41.14 0.19		0.00 64.40 1.61 0.00	
IFEM315	Female Interaction Of BPRICALN					23.63			
IBLK315 IBLK316 IBLK317 IBLK318	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU Black Interaction Of BPRICACU Black Interaction Of BPRICAQR							7.33 32.76 77.74 2.99	
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree		0.49 0.00			72.59 14.45	1.36 76.34 22.74 0.84	86.58 2.27 48.17	0.14 0.00 0.00
IFEM320 IFEM321	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU					3.88 6.02	0.60		
IBLK320 IBLK321 IBLK322	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASCU							62.08 0.17 94.56	
IHISP320 IHISP321	Hispanic Interaction Of BSCHASLN Hispanic Interaction Of BSCHASQU		0.03 0.00						
IOTH320 IOTH321	Other Interaction Of BSCHASLN Other Interaction Of BSCHASQU								0.00 0.00
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander			74.91 3.31		9.33 90.98 90.90	46.98 38.53 2.13 0.03		
IBLK325 IBLK326	Black Interaction Of PASIANLN Black Interaction Of PASIANQU			0.01 0.00					
IOTH325 IOTH326 IOTH327	Other Interaction Of PASIANLN Other Interaction Of PASIANQU Other Interaction Of PASIANCU					39.46 65.04 4.16			
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban	83.27 8.02 5.83		16.22 0.00 8.52 0.00	0.00	91.85 19.63 21.13	0.00	5.44 57.08 19.76 44.74	
IFEM330 IFEM331 IFEM332	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU	30.49 0.01				44.89 59.93 0.61			
IBLK330 IBLK331 IBLK332 IBLK333	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU Black Interaction Of PCUBANQR	0.23 50.61 14.67					3.24	65.59 9.17 40.84 1.82	
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Aleut			0.00		63.26		14.08 74.48 1.46 19.00	0.01 0.00 20.53 0.00
IFEM335 IFEM336 IFEM337 IFEM338	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU Female Interaction Of PINDIACU Female Interaction Of PINDIAQR							74.73 97.48 48.90 0.00	
IBLK335 IBLK336	Black Interaction Of PINDIALN Black Interaction Of PINDIAQU					1.57		17.96 1.05	
IHISP335 IHISP336 IHISP337 IHISP338	Hispanic Interaction Of PINDIALN Hispanic Interaction Of PINDIAQU Hispanic Interaction Of PINDIACU Hispanic Interaction Of PINDIACU							52.44 35.86 82.11 0.72	
ЮТН335	Other Interaction Of PINDIALN							0.00	

F4. Significance Probabilities for Fixed Effect Coefficients for Past Month Cocaine Use by Age

		BIG CITY				REMAINDER	1		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	1.88	0.00	0.00	0.00	68.84	0.00	10.68	47.23
FEMBLCK	O: Black Interaction Of FEMALE	0.25	64.10	0.98	0.00	88.15	44.96	30.22	0.00
FEMHISP	O: Hispanic Interaction Of FEMALE	28.95	31.40	21.27	0.01	76.91	48.80	31.16	1.47
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	30.25 2.71 17.11	10.25 0.00 0.00	6.69 0.00	0.00	61.05 8.08	47.58 36.90 46.97	51.66 1.74 88.74	0.00 0.00
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban						51.01 28.49 9.48 0.14	62.66 28.80 29.55 22.58	
POPRMLN	T: Linear: Average Persons Per Room			0.00				0.00	
IFEM20	Female Interaction Of POPRMLN							0.00	
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years T: Cubic: Percent Persons 19-24 Years T: Quartic: Percent Persons 19-24 Years							6.30 95.50 0.00 0.00	
IHISP30 IHISP31 IHISP32 IHISP33	Hispanic Interaction Of PAGE24LN Hispanic Interaction Of PAGE24QU Hispanic Interaction Of PAGE24CU Hispanic Interaction Of PAGE24QR							56.68 95.46 5.38 7.99	
PAGE44LN	T: Linear: Percent Persons 35-44 Years							0.00	
PAGE54LN PAGE54QU	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years				0.00 0.00				17.83 0.00
IBLK45 IBLK46	Black Interaction Of PAGE54LN Black Interaction Of PAGE54QU				0.00 0.00				
PSCHASLN PSCHASQU	T: Linear: Percent Associates Degree T: Quadratic: Percent Associates Degree							53.87 0.00	
PSCHCOLN PSCHCOQU	T: Linear: Bachelors, Graduate, Or Professional Degree T: Quadratic: Bachelors, Graduate, Or Professional Degree					52.13 5.87			1.98 0.00
IFEM75 IFEM76	Female Interaction Of PSCHCOLN Female Interaction Of PSCHCOQU					15.29 3.00			
PBLACKLN PBLACKQU PBLACKCU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic T: Cubic: Percent Black Nonhispanic			0.00 9.18 0.00				7.45 22.48 0.00	
PHISPLN PHISPQU PHISPCU PHISPQR	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic T: Cubic: Percent Hispanic T: Quartic: Percent Hispanic								10.76 1.98 18.05 0.00
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity					81.78 27.78 13.38			
PHHF18LN PHHF18QU PHHF18CU PHHF18QR	T: Linear: % Female-Headed HH W/No Spouse & Chld Under 18 T: Quadratic: % Female-Headed HH W/No Spouse & Chld Under 18 T: Cubic: % Female-Headed HH W/No Spouse & Chld Under 18 T: Quartic: % Female-Headed HH W/No Spouse & Chld Under 18					31.22 65.13 47.40 2.90			0.00
IFEM120	Female Interaction Of PHHF18LN								0.00
PFNOTLN PFNOTQU	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed			0.00				0.74 0.00	
IHISP135 IHISP136	Hispanic Interaction Of PFNOTLN Hispanic Interaction Of PFNOTQU							21.71 3.36	
PMNOTLN PMNOTQU	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed						2.29 0.04	34.05 0.32	
IFEM150 IFEM151	Female Interaction Of PMNOTLN Female Interaction Of PMNOTQU						0.00 0.06		
IHISP150 IHISP151	Hispanic Interaction Of PMNOTUN Hispanic Interaction Of PMNOTQU							82.93 7.03	
PRENTLN PRENTQU PRENTCU PRENTQR	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented T: Cubic: Percent Housing Rented T: Quartic: Percent Housing Rented								0.00 0.00 0.00 0.00
V18BLN V18BQU V18BCU	C: Linear: Marijuana Sale/Manufacture Arrest Rate C: Quadratic: Marijuana Sale/Manufacture Arrest Rate C: Cubic: Marijuana Sale/Manufacture Arrest Rate							7.03 65.75 0.00	
V18ELN V18EQU V18ECU V18EQR	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadric: Opium/Cocaine & Deriv Posession Arrest Rate								0.50 0.00 10.98 0.00
V18ALN	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate		0.00					0.33	
IHISP200	Hispanic Interaction Of V18ALN		0.01						

		BIG CITY				REMAINDER	\		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
V18LN V18QU	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate		0.00	3.35 86.91			67.38 6.64	0.00	
V18QU V18QU V18QR	C: Quartic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate			0.10 24.78			0.04	79.65 39.64 0.00	
IFEM225 IFEM226 IFEM227 IFEM228	Female Interaction Of V18LN Female Interaction Of V18QU Female Interaction Of V18CU Female Interaction Of V18QR							18.05 67.85 21.59 0.31	
IHISP225	Hispanic Interaction Of V18LN		0.00						
VIOLLN VIOLQU	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate				0.00 0.00				0.03 0.00
IBLK230 IBLK231	Black Interaction Of VIOLLN Black Interaction Of VIOLQU				0.00 0.00				
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992						42.31 0.04		
IFEM235 IFEM236	Female Interaction Of DRATELN Female Interaction Of DRATEQU						57.92 0.01		
RH43ALN RH43AQU RH43ACU RH43AQR	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units T: Cubic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units	48.95 48.97 1.63 1.27						0.00	
RP80ALN RP80AQU RP80ACU RP80AQR	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income T: Quartic: Recoded Median Household Income					91.39 21.16 92.55 40.19	71.40 8.19 0.01	0.00	
IBLK250 IBLK251 IBLK252	Black Interaction Of RP80ALN Black Interaction Of RP80AQU Black Interaction Of RP80ACU						30.82 40.66 5.03		
ARATELN BAGE54LN	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 B: Linear: Percent Persons 45-54 Years	0.58					77.16		0.00
BAGE54QU BAGE54CU	B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years						0.00		0.00 0.00
IFEM275 IFEM276 IFEM277	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU								0.00 0.00 0.00
IBLK275 IBLK276	Black Interaction Of BAGE54LN Black Interaction Of BAGE54QU						93.61 0.34		
BAGE64LN BAGE64QU BAGE64CU BAGE64QR	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years B: Quartic: Percent Persons 55-64 Years		0.87 0.01 18.22	36.83 79.44 0.45	0.00 0.00		0.05 39.57 0.95 86.67		95.01 0.00 0.00
IFEM280 IFEM281 IFEM282	Female Interaction Of BAGE64LN Female Interaction Of BAGE64QU Female Interaction Of BAGE64CU				36.61 0.00				0.00 0.00 0.00
IBLK280 IBLK281	Black Interaction Of BAGE64LN Black Interaction Of BAGE64QU								1.49 0.00
IHISP280 IHISP281 IHISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		97.76 31.03 0.00	5.34 31.50 0.00					
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander		29.95 0.00				21.50 8.64 66.63 0.06		
BCUBANLN BCUBANQU BCUBANCU	B: Lınear: Percent Hıspanıcs: Cuban B: Quadratic: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban						83.19 48.05 0.19	0.03	
IFEM290	Female Interaction Of BCUBANLN							0.99	
BMNOTLN BMNOTQU BMNOTCU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed				0.00 0.00 0.00	12.75 4.15	4.44 11.50 0.35		
IFEM305 IFEM306 IFEM307	Female Interaction Of BMNOTLN Female Interaction Of BMNOTQU Female Interaction Of BMNOTCU				0.00 0.00 0.00				
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.00 3.93 4.10 0.00				26.03 63.23 8.48 0.03		
BPRICALN BPRICAQU BPRICACU	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican	16.63				84.59		0.41 2.28 0.07	
IFEM315	Female Interaction Of BPRICALN					51.12			
IBLK315 IBLK316 IBLK317	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU Black Interaction Of BPRICACU							94.69 34.82 0.07	
BSCHASLN BSCHASQU BSCHASCU	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree		0.16 0.00	-		51.87 10.31	3.29 44.14	0.44 34.23 0.04	17.63 22.26 0.00
IFEM320 IFEM321	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU					49.80 5.39	0.02		

T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

		BIG CITY				REMAINDER	1		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IBLK320 IBLK321 IBLK322 PCUBANLN PCUBANQU PCUBANQU PCUBANQR	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU Black Interaction Of BSCHASQU T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban	6.13 3.78 21.45		0.00 0.00				35.71 28.54 2.37 21.02 25.58 5.33 0.00	
IBLK330 IBLK331 IBLK332	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU	60.28 5.32 4.58							

F5. Significance Probabilities for Fixed Effect Coefficients for Past Month Any Illicit Drug But Marijuana by Age

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	29.69	0.00	0.00	0.00	19.09	0.00	0.05	15.97
FEMBLCK	O: Black Interaction Of FEMALE	7.24	27.46	16.55	0.00	15.03	42.92	27.75	0.00
FEMHISP	O: Hispanic Interaction Of FEMALE	71.27	0.24	0.02	0.00		99.81	92.54	77.59
FEMOTHR RACEBLCK	O: Other Interaction Of FEMALE	48.96 5.49	0.00	95.28 0.00	0.00	37.39 6.81	23.59 2.52	11.42	0.00
RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	90.88 1.31	0.00 0.00 0.00	0.00 0.00 0.00	0.00	33.65	17.13 0.00	11.62 30.45 0.04	13.65 0.00
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator						0.02	0.05 0.54 0.00	
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban								0.00 62.05 0.00 76.27
PHH1PLN	T: Linear: Percent One Person Households					0.22			
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Linear: Percent Persons 19-24 Years T: Quadratic: Percent Persons 19-24 Years T: Cubic: Percent Persons 19-24 Years T: Quartic: Percent Persons 19-24 Years							0.70 34.90 0.00 0.00	0.00 0.00
PAGE34LN PAGE34QU	T: Linear: Percent Persons 25-34 Years T: Quadratic: Percent Persons 25-34 Years					27.92 0.04			
PAGE44LN PAGE44QU	T: Linear: Percent Persons 35-44 Years T: Quadratic: Percent Persons 35-44 Years					60.65 95.50			
PAGE54LN PAGE54QU	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years								0.00 0.00
IFEM45 IFEM46	Female Interaction Of PAGE54LN Female Interaction Of PAGE54QU								70.27 0.00
PSCH8LN	T: Linear: Percent 0-8 Years Of School							0.00	
PSCH12LN PSCH12QU PSCH12CU PSCH12QR	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma T: Quartic: Percent 9-12 Years & No High School Diploma	51.88 1.30 90.69 0.02				32.65 9.49 83.93 1.53			
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU	11.21 0.10				14.46 1.89			
PSCHASLN PSCHASQU	T: Linear: Percent Associates Degree T: Quadratic: Percent Associates Degree					41.46 8.40			
IBLK65 IBLK66	Black Interaction Of PSCHASLN Black Interaction Of PSCHASQU					19.76 1.30			
PSCHSCLN PSCHSCQU	T: Linear: Percent Some College And No Degree T: Quadratic: Percent Some College And No Degree					87.11 0.96			
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Quartic: Percent Families Below Poverty Level							9.97 14.90 89.27 0.00	
P64DISLN P64DISQU P64DISCU	T: Linear: Percent 16-64 With A Work Disability T: Quadratic: Percent 16-64 With A Work Disability T: Cubic: Percent 16-64 With A Work Disability	18.53 11.22 3.37				16.27 8.68 12.12			
IBLK95 IBLK96 IBLK97	Black Interaction Of P64DISLN Black Interaction Of P64DISQU Black Interaction Of P64DISCU	2.12 92.69 0.14				68.77 18.19 1.25			
PBLACKLN PBLACKQU PBLACKCU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic T: Cubic: Percent Black Nonhispanic			0.24 78.57 0.00				0.39 60.13 0.00	
PFNOTLN	T: Linear: Percent Females Separated, Divorced Or Widowed			0.00				0.00	
PMNEVLN	T: Linear: Percent Males Never Married					1.85			
PMNOTEN PMNOTQU PMNOTCU PMNOTQR	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed T: Cubic: Percent Males Separated, Divorced Or Widowed T: Quartic: Percent Males Separated, Divorced Or Widowed					84.60 93.42 46.88 4.29			
P40HULN P40HUQU P40HUCU P40HUQR	T: Linear: Percent Housing Units Built 1940-1949 T: Quadratic: Percent Housing Units Built 1940-1949 T: Cubic: Percent Housing Units Built 1940-1949 T: Quartic: Percent Housing Units Built 1940-1949					2.32		0.00 11.39 0.07 0.00	
PRENTLN PRENTQU	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented							7.03 0.00	0.00
ADRATEQR	C: Quartic: Death Rate For All Alcohol-Related Cases						0.00		
ADRAT1LN	C: Linear: Death Rate With Explicit Mention Of Alcohol						0.00		
V18FLN	C: Linear: Marijuana Posession Arrest Rate	İ				20.21		0.04	

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
V18FQU V18FCU	C: Quadratic: Marijuana Posession Arrest Kate C: Cubic: Marijuana Posession Arrest Rate					9.03		18.82 0.10	
V18FQR IHISP185	C: Quartic: Marijuana Posession Arrest Rate Hispanic Interaction Of V18FLN							0.00 8.45	
V18BLN V18BQU	C: Linear: Marijuana Sale/Manufacture Arrest Rate C: Quadratic: Marijuana Sale/Manufacture Arrest Rate								39.46 0.00
V18ALN V18AQU V18ACU V18AQR	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Cubic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quartic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate					74.11 51.23 2.58 2.06			
V18CU	C: Cubic: Total Drug Abuse Violations Arrest Rate		0.00				0.00		
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992					20.12	0.00		
B64DISLN B64DISQU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability					39.13 8.72			
BAGE34QU BAGE34CU	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years					90.45 45.41 22.41	0.00	0.00 0.02 0.07	
IFEM265 IFEM266 IFEM267	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU							0.78 1.18 0.02	
IOTH265 IOTH266 IOTH267	Other Interaction Of BAGE34LN Other Interaction Of BAGE34QU Other Interaction Of BAGE34CU					9.69 8.78 8.97		0.00	
BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years					0.37	28.15 0.18	89.84 3.71	
IBLK270 IBLK271	Black Interaction Of BAGE44LN Black Interaction Of BAGE44QU							77.59 0.38	
IOTH270 IOTH271	Other Interaction Of BAGE44LN Other Interaction Of BAGE44QU						3.83 0.00		
BAGE54LN BAGE54QU	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years							28.21 0.00	
BAGE64LN BAGE64QU BAGE64CU BAGE64QR	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years B: Quartuc: Percent Persons 55-64 Years		0.00 0.01 1.52	15.99 85.44 0.00		1.04	0.00 1.25 0.00 0.03		
IHISP280 IHISP281 IHISP282 IHISP283	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU Hispanic Interaction Of BAGE64QR		0.95 97.41 0.00	27.69 0.08 0.00			28.65 45.45 50.76 3.55		
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander						77.41 0.15 25.67 0.01	20.12 0.00	
IOTH285	Other Interaction Of BASIANLN						0.07		
BCUBANLN BCUBANQU BCUBANCU	B: Linear: Percent Hispanics: Cuban B: Quadratıc: Percent Hıspanics: Cuban B: Cubic: Percent Hispanics: Cuban				0.00 0.00	22.52 17.04 47.13			
IFEM290 IFEM291 IFEM292	Female Interaction Of BCUBANLN Female Interaction Of BCUBANQU Female Interaction Of BCUBANCU					82.47 97.63 18.24			
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed	38.88 64.59 82.97 0.45				94.70 70.01 25.05 18.27	38.07 16.14 22.98		
IFEM295 IFEM296 IFEM297	Female Interaction Of BFNOTLN Female Interaction Of BFNOTQU Female Interaction Of BFNOTCU						3.10 53.21 0.01		
IBLK295 IBLK296 IBLK297 IBLK298	Black Interaction Of BFNOTLN Black Interaction Of BFNOTQU Black Interaction Of BFNOTQU Black Interaction Of BFNOTQR					66.87 86.29 20.55 8.56	82.80 0.27		
IHISP295	Hispanic Interaction Of BFNOTLN						7.58		
BINDIALN BINDIAQU BINDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut					22.06 35.55 0.16			
BMNOTLN BMNOTQU BMNOTCU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed				0.00				14.90 0.00 0.00
IFEM305	Female Interaction Of BMNOTLN				0.00				
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.03 1.67 9.25 0.00				12.49 0.01 0.00 0.02		
IFEM310 IFEM311 IFEM312	Female Interaction Of BPOVERLN Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU		12.14 0.04 0.00				0.22 0.72 0.00		
IHISP310 IHISP311	Hispanic Interaction Of BPOVERCN Hispanic Interaction Of BPOVERQU		0.00				91.32 2.41		

T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican	0.42				24.31 6.37 26.72 65.32		12.64 0.06 0.00	
IFEM315 IFEM316 IFEM317 IFEM318	Female Interaction Of BPRICALN Female Interaction Of BPRICAQU Female Interaction Of BPRICACU Female Interaction Of BPRICAQR					90.15 3.78 50.82 0.18		2.46 26.43 0.01	
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree	61.98 38.21 5.57 0.18	0.00			63.54 2.16 15.09 9.29	88.36 0.03	0.73 81.98 0.00	
IFEM320 IFEM321 IFEM322 IFEM323	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU Female Interaction Of BSCHASCU Female Interaction Of BSCHASQR					74.76 0.11 3.39 6.10			
PASIANLN PASIANQU PASIANCU	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander						92.01 19.27 0.01	33.52 7.88 0.00	0.00
IHISP325 IHISP326 IHISP327	Hispanic Interaction Of PASIANLN Hispanic Interaction Of PASIANQU Hispanic Interaction Of PASIANCU						24.64 42.91 8.42		
IOTH325	Other Interaction Of PASIANLN						0.01		
PINDIALN	T: Linear: Percent Pop: American Indian, Eskimo, Aleut						7.72	0.03	
IHISP335	Hispanic Interaction Of PINDIALN							1.37	
IOTH335	Other Interaction Of PINDIALN						0.00		

F6. Significance Probabilities for Fixed Effect Coefficients for Past Year Alcohol Treatment by Age

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	99.00	0.00	0.00	0.00	85.66	0.34	0.00	0.00
FEMBLCK	O: Black Interaction Of FEMALE	94.17	0.00	21.49	0.00	81.87	63.81	58.55	0.01
FEMHISP RACEBLCK	O: Hispanic Interaction Of FEMALE O: Race/Black Indicator	37.43 18.02	9.98 0.00	0.53	0.00	79.23 27.57	34.38 12.09	27.93 9.89	5.21 11.02
RACEHISP REGNOREA	O: Race/Hispanic Indicator	83.50	77.95	0.00	0.00	81.83	48.59	23.48	0.70
REGWEST	O: Northeast Region Indicator O: West Region Indicator						89.75 87.02		
PAGE18LN PAGE18QU PAGE18CU PAGE18QR	1: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years 1: Cubic: Percent Persons 0-18 Years T: Quartic: Percent Persons 0-18 Years		0.00				3.42 0.04 23.42 0.25		
IHISP25 IHISP26 IHISP27 IHISP28	Hispanic Interaction OF PAGE 18LN Hispanic Interaction Of PAGE 18QU Hispanic Interaction OF PAGE 18CU Hispanic Interaction OF PAGE 18QR						93.64 35.64 63.08 19.87		
PSCH12LN PSCH12QU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma		0.01 0.00				45.61 0.00		
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU						39.05 0.00		
IHISP60 IHISP61	Hispanic Interaction Of PSCH12LN Hispanic Interaction Of PSCH12QU						65.20 7.25		
PSCHCOLN PSCHCOQU PSCHCOCU PSCHCOQR	Linear: Bachelors, Graduate, Or Professional Degree Quadratic: Bachelors, Graduate, Or Professional Degree Cubic: Bachelors, Graduate, Or Professional Degree Quadratic: Bachelors, Graduate, Or Professional Degree							0.05 0.00	0.00 83.37 0.00 0.00
PSCHSCLN PSCHSCQU PSCHSCCU	T: Linear: Percent Some College And No Degree T: Quadratic: Percent Some College And No Degree T: Cubic: Percent Some College And No Degree	6.84 58.91				54.91 8.79 17.73			
IHISP80 IHISP81	Hispanic Interaction Of PSCHSCLN Hispanic Interaction Of PSCHSCQU	24.70 5.15							
PPOVERLN PPOVERQU PPOVERCU	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level	34.82 0.29				38.89 13.76 0.13			
PPUBASLN PPUBASQU PPUBASCU	T: Linear: Percent Households With Public Assist Income T: Quadratic: Percent Households With Public Assist Income T: Cubic: Percent Households With Public Assist Income		0.00 7.25 0.00				0.96 1.00		
IHISP90 IHISP91	Hispanic Interaction Of PPUBASLN Hispanic Interaction Of PPUBASQU		0.19				71.66 10.49		
PBLACKLN PBLACKQU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic								19.46 0.00
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity						57.98 20.25 0.18		
IFEM110 IFEM111 IFEM112	Female Interaction Of POTHLN Female Interaction Of POTHQU Female Interaction Of POTHCU						16.77 28.30 0.00		
PFNEVLN	T: Linear: Percent Females Never Married	•					0.00		
PMNEVLN PMNEVQU PMNEVCU	T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married T: Cubic: Percent Males Never Married		0.00		12.69		0.01 0.12 1.89		0.00 25.00 0.00
IBLK140 IBLK141 IBLK142	Black Interaction Of PMNEVLN Black Interaction Of PMNEVQU Black Interaction Of PMNEVCU						94.45 67.45 6.56		35.73 22.42 0.00
IHISP140	Hispanic Interaction Of PMNEVLN		0.00		0.00		20.05		0.00
PMLABLN PMLABQU PMLABCU PMLABQR	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force T: Cubic: Percent Males 16+ Years Old In Labor Force T: Quartic: Percent Males 16+ Years Old In Labor Force				53.29 0.00	98.58 68.70 72.12 3.66			0.00 3.14 26.22
IFEM145 IFEM146 IFEM147	Female Interaction Of PMLABLN Female Interaction Of PMLABQU Female Interaction Of PMLABCU				0.00 0.00				68.96 0.00 0.00
POLDHULN POLDHUQU POLDHUCU	T: Linear: Percent Housing Units Built 1939 Or Earlier T: Quadratic: Percent Housing Units Built 1939 Or Earlier T: Cubic: Percent Housing Units Built 1939 Or Earlier				0.00	59.32 16.03 34.04			
IFEM155 IFEM156 IFEM157	Female Interaction Of POLDHULN Female Interaction Of POLDHUQU Female Interaction Of POLDHUCU				0.00	40.32 15.54 4.60			
P40HULN	T: Linear: Percent Housing Units Built 1940-1949								0.00
ADRATILN ADRATIQU	C: Linear: Death Rate With Explicit Mention Of Alcohol C: Quadratic: Death Rate With Explicit Mention Of Alcohol			57.85 6.40			2.41 0.35		

		BIG CITY				REMAINDER	\		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
ADRATICU ADRATIQR	C: Cubic: Death Kate With Explicit Mention Of Alcohol C: Quartic: Death Rate With Explicit Mention Of Alcohol			0.97 0.00			0.00		
V18FLN V18FQU	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate						0.08 0.01		
IBLK185 IBLK186	Black Interaction Of V18FLN Black Interaction Of V18FQU						33.22 13.51		
V18ELN V18EQU	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate						35.71 68.20		
IFEM195 IFEM196	Female Interaction Of V18ELN Female Interaction Of V18EQU						2.75 0.00		
IBLK195 IBLK198	Black Interaction Of V18ELN Black Interaction Of V18EQR					64.77 46.33			
V18LN V18QU V18CU	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate					71.34 97.77 5.59			
IFEM225 IFEM226 IFEM227	Female Interaction Of V18LN Female Interaction Of V18QU Female Interaction Of V18CU					88.51 91.88 3.69			
VIOLLN VIOLQU VIOLCU VIOLQR	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate C: Cubic: Total Violent Offenses Arrest Rate C: Quartic: Total Violent Offenses Arrest Rate					98.51 82.82 22.64 6.80			
RH43ALN RH43AQU	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units				0.00 0.00				0.07 0.00
RH61ALN IBLK245	T: Linear: Recoded Median Value Of Owner Occupied HUs Black Interaction Of RH61 ALN				0.00				
RP80ALN RP80AQU RP80ACU	T: Linear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income					18.43 58.81 2.31	5.66 0.05		
ARATELN ARATEQU ARATECU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992			14.28 0.01 0.00		3.01		50.20 1.89 0.60	
IHISP255 IHISP256 IHISP257	Hispanic Interaction Of ARATELN Hispanic Interaction Of ARATEQU Hispanic Interaction Of ARATECU			0.00 0.00 0.00				7.02 68.23 4.47	
B64DISLN B64DISQU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability						0.01 8.40		
IBLK260 IBLK261	Black Interaction Of B64DISLN Black Interaction Of B64DISQU						51.68 3.84		
BAGE34LN BAGE34QU BAGE34CU BAGE34QR	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years				0.00 0.00 0.00	62.42 15.40			33.53 0.00 0.00 6.03
IFEM265 IFEM266 IFEM267 IFEM268	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU Female Interaction Of BAGE34QR				0.00 0.00 0.00				0.00 42.83 0.00 0.00
IBLK265 IBLK266	Black Interaction Of BAGE34LN Black Interaction Of BAGE34QU					86.13 19.44			
IHISP265 IHISP266 IHISP267	Hispanic Interaction Of BAGE34LN Hispanic Interaction Of BAGE34QU Hispanic Interaction Of BAGE34CU				20.02 0.00				93.37 0.60 31.11
BAGE44LN BAGE44QU BAGE44CU BAGE44QR	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years B: Quartic: Percent Persons 35-44 Years						2.06 69.76 69.73 0.03		
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years					4.68 64.43 19.15 25.03		6.29 0.00	
IFEM275 IFEM276 IFEM277 IFEM278	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU Female Interaction Of BAGE54QR					14.02 14.42 99.82 3.05			
BAGE64LN BAGE64QU BAGE64CU BAGE64QR	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years B: Quartic: Percent Persons 55-64 Years				0.00				0.00 8.39 0.00 0.00
IBLK280 IBLK282 IBLK283	Black Interaction Of BAGE64LN Black Interaction Of BAGE64CU Black Interaction Of BAGE64QR				0.00				0.00 22.10 0.00
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander	24.24 29.06 62.49 23.64				91.31 44.69 15.03 23.27			0.22 0.00
IFEM285 IFEM286 IFEM287 IFEM288	Female Interaction Of BASIANLN Female Interaction Of BASIANQU Female Interaction Of BASIANQU Female Interaction Of BASIANQR	11.70 32.74 97.25 5.71				53.48 79.62 87.99 4.69			
IHISP285	Hispanic Interaction Of BASIANLN	5.71							11.49

T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
1H15P286	Hispanic Interaction OI BASIANQU						•	<u> </u>	0.1/
BCUBANLN BCUBANQU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban				0.00				8.21 0.00
IHISP290 IHISP291	Hispanic Interaction Of BCUBANLN Hispanic Interaction Of BCUBANQU				0.00				16.37 1.55
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed				0.00	78.51 66.11 0.25			0.20 76.24 0.00 0.00
BINDIALN BINDIAQU BINDIACU BINDIAQR	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut B: Quartic: Percent Pop: American Indian, Eskimo, Aleut			40.17 19.35	98.08 0.00 0.00 0.00			3.98 0.56	0.00 0.00 0.00
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU			38.45 0.00				96.34 2.61	
BMNOTLN BMNOTQU BMNOTCU BMNOTQR	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed B: Quartic: Percent Males Separated, Divorced or Widowed	69.97 18.54				86.27 31.79 18.19 24.53			
IFEM305 IFEM306 IFEM307 IFEM308	Female Interaction Of BMNOTLN Female Interaction Of BMNOTQU Female Interaction Of BMNOTCU Female Interaction Of BMNOTQR	23.89 3.24				70.83 31.76 97.78 2.40			
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level	32.43 60.68 84.12 5.34						46.73 60.11 0.00	
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican							17.52 25.91 4.87 4.47	7.43
IHISP315 IHISP316 IHISP317 IHISP318	Hispanic Interaction Of BPRICALN Hispanic Interaction Of BPRICAQU Hispanic Interaction Of BPRICACU Hispanic Interaction Of BPRICAQR							94.62 12.72 30.37 14.08	0.73
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree				0.00 0.00 0.00 0.00				14.87 0.00 0.00 0.00
IBLK320 IBLK321	Black Interaction Of BSCHASLN Black Interaction Of BSCHASQU				0.00				21.52 0.00
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander I: Quartic: Percent Population: Asian, Pacific Islander				0.00			0.03 0.00	0.00 0.00 3.07 0.00
IFEM325 IFEM326	Female Interaction Of PASIANLN Female Interaction Of PASIANOU				0.00			0.37 0.00	0.00
IHISP325	Hispanic Interaction Of PASIANLN							5.15	
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban						69.17 1.46 42.87 18.70	0.00 0.56 50.48 0.00	
IFEM330 IFEM331 IFEM332 IFEM333	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU Female Interaction Of PCUBANQR						64.02 18.59 59.22 1.97		
IBLK330	Black Interaction Of PCUBANLN							28.02	
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratu: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Aleut			0.00 0.04 0.00 0.00				0.03	

F7. Significance Probabilities for Fixed Effect Coefficients for Past Year Illicit Drug Use Treatment by Age

		BIG CITY	ITY			REMAINDER	2		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	0.14	4.25	0.00	68.27	3.58	0.79	0.01	0.00
FEMBLCK	O: Black Interaction Of FEMALE	63.75	76.89	64.53	0.00	44.15	46.19	42.27	14.83
FEMHISP	O: Hispanic Interaction Of FEMALE	1.08	0.19	0.01	0.12	54.30	70.89	46.50	25.73
FEMOTHR	O: Other Interaction Of FEMALE							12.59	
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	0.17 80.64	0.84 88.44 0.02	0.16 1.06	0.00	37.33 85.61 21.58	9.46 5.98	25.89 23.33 60.38	0.00 16.68 0.00
REGNOREA REGSOUTH	O: Northeast Region Indicator O: South Region Indicator					3.15 51.98			
IHISP7 IHISP8 IHISP9	Hispanic Interaction Of REGNOREA Hispanic Interaction Of REGSOUTH Hispanic Interaction Of REGWEST			0.07 1.33 4.74					
PHH1PLN PHH1PQU PHH1PCU PHH1PQR	T: Linear: Percent One Person Households T: Quadratic: Percent One Person Households T: Cubic: Percent One Person Households T: Quartic: Percent One Person Households							0.01 0.01 1.33 0.00	0.00 11.18 0.01 0.00
POPRMLN	T: Linear: Average Persons Per Room							0.00	
PAGE34LN	T: Linear: Percent Persons 25-34 Years					61.80			
IFEM35	Female Interaction Of PAGE34LN	1				0.61			
PSCH12LN PSCH12QU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma				0.00				0.00 0.00
PSCHSCLN PSCHSCQU	T: Linear: Percent Some College And No Degree T: Quadratic: Percent Some College And No Degree	; ;	0.00 0.78				2.09 0.19		0.00 0.00
IFEM80 IFEM81	Female Interaction Of PSCHSCLN Female Interaction Of PSCHSCQU		0.02 0.00						0.00 0.00
IHISP80 IHISP81	Hispanic Interaction Of PSCHSCLN Hispanic Interaction Of PSCHSCQU						38.73 8.65		
PPOVERLN	T: Linear: Percent Families Below Poverty Level				0.00				0.00
PBLACKLN PBLACKQU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic							0.64 0.06	
PHISPLN PHISPQU PHISPCU PHISPQR	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic T: Cubic: Percent Hispanic T: Quartic: Percent Hispanic			93.38 0.00 0.00 0.00				64.08 0.03 0.00 0.16	
POTHLN POTHQU POTHCU	T: Linear: Percent Other Race/Hispanicity T: Quadratic: Percent Other Race/Hispanicity T: Cubic: Percent Other Race/Hispanicity						34.18 42.63 0.17		
IHISP110 IHISP111 IHISP112	Hispanic Interaction Of POTHLN Hispanic Interaction Of POTHQU Hispanic Interaction Of POTHCU						77.49 15.30 24.84		
PHHF18LN PHHF18QU PHHF18CU PHHF18QR	T: Linear: % Female-Headed HH W/No Spouse & Chid Under 18 T: Quadratic: % Female-Headed HH W/No Spouse & Chid Under 18 T: Cubic: % Female-Headed HH W/No Spouse & Chid Under 18 T: Quartic: % Female-Headed HH W/No Spouse & Chid Under 18					42.41 78.22 16.55 0.82			
PFLABLN PFLABQU PFLABCU PFLABQR	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force T: Cubic: Percent Females 16+ Years Old In Labor Force T: Quartic: Percent Females 16+ Years Old In Labor Force	76.03 0.29				16.18 56.85 83.02 0.05			
PFNEVLN PFNEVQU PFNEVCU PFNEVQR	T: Linear: Percent Females Never Married T: Quadratic: Percent Females Never Married T: Cubic: Percent Females Never Married T: Quartic: Percent Females Never Married			0.00				66.06 11.79 24.54 0.00	
PMNEVLN PMNEVQU	T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married						0.00		0.15 0.00
IBLK140	Black Interaction Of PMNEVLN						2.39		
PMLABLN PMLABQU PMLABCU PMLABQR	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force T: Cubic: Percent Males 16+ Years Old In Labor Force T: Quartic: Percent Males 16+ Years Old In Labor Force								2.51 9.23 0.00 0.00
ADRATELN ADRATEQU	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases						59.54 0.00		0.00
V18FLN V18FQU V18FCU V18FQR	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate C: Cubic: Marijuana Posession Arrest Rate C: Quartic: Marijuana Posession Arrest Rate				0.00		42.28 1.06 9.02 0.11		0.00 0.00 0.00
IHISP185 IHISP186 IHISP187	Hispanic Interaction Of V18FLN Hispanic Interaction Of V18FQU Hispanic Interaction Of V18FCU								9.25 74.08 0.06
V18BLN	C: Linear: Marijuana Sale/Manufacture Arrest Rate	-	0.12				0.00		

Miles 1997			BIG CITY				REMAINDER			
BILLY	Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
SELVI	v18RA∩	U: Quadratic: Marijuana Sale/Manufacture Afrest Kate					•	0.00		
HISPPIN Segue the fraction of VYHBLN 0.00 2.10 0.00 0	IBLK190	Black Interaction Of V18BLN								
Visilab C. Lamer Option Course & Derr Processin Arrest (face Visilab IHISP190	Hispanic Interaction Of V18BLN		0.00				21.06			
CHAIN C. Lames Total Dairy Awards North Area State C. C. C. C. C. C. C. C	IHISP191									
MAINENDANE C. Lemer: Mann A-Order Court Treatment Rate 1991 & 1992 0.42		·				0.00		0.00		0.00
MACHENION C. Quadratic Mean A Orbit Chief Treatment Rate 1991 & 1992 O.12	ARATELN	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992	8.52			0.00				0.00
1.0	ARATEQU ARATECU	C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992	0.43		0.00	12.00			0.00	
Histories Hispanic Intensection Of BelighSUL \$2.00	B64DISQU B64DISQU B64DISQR	B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability			1.38 56.16	0.00			0.53 84.63	
AGEFEAD B. Quadratic Percent Percents 3-4 Versis \$2.00	IHISP260 IHISP261 IHISP262 IHISP263	Hispanic Interaction Of B64DISQU Hispanic Interaction Of B64DISCU							82.66 21.84	
AGGERGE AGGE	BAGE34LN BAGE34QU BAGE34CU	B: Quadratic: Percent Persons 25-34 Years					82.90			
SAU-HAUE, S. Quartic: Percent Persons 3-4 Years S. S. S. S. S. S. S. S	BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years			46.69		55.25	59.71	3.34	0.01 0.00
FEMALY Female Interaction OF BAGEHAU	BAGE44CU BAGE44QR	B: Quartic: Percent Persons 35-44 Years			0.00		6.31	8.37		0.00
SAGE-SALN B. Linear: Percent Persons 45-54 Vears 0.36 10.36 0.25	IFEM271 IFEM272	Female Interaction Of BAGE44QU Female Interaction Of BAGE44CU						18.10 0.05		
Black Interaction Of BAGE-SIAN	BAGE54LN	· ·	0.36				10.96			
ASIANIN B. Linear Percent Population Asian, Pacific Islander 0.00 0.00 0.00 0.3 0.35 0.35 0.35 0.35 0.35 0.35 0.35 0.35 0.35 0.00 0.00 0.00 0.00 0.00 0.35	IFEM275	Female Interaction Of BAGE54LN						0.00		
ASSIANOU B. Quadratic Percent Population: Asian, Pacific Islander ASSIANOU B. Cubic Percent Population: Asian, Pacific Islander 19-20 19-2	IBLK275	Black Interaction Of BAGE54LN	0.22							
FEMDS Female Interaction Of BASIANN	BASIANLN BASIANQU BASIANCU BASIANQR	B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander		0.00				63.36 27.65		
BLK285 Black Interaction Of BASIANLN 2.09 11.54 34.75 34.7	IFEM285 IFEM286 IFEM287 IFEM288	Female Interaction Of BASIANQU Female Interaction Of BASIANCU						1.90 58.98		
SCUBANQU B. Quadratic: Percent Hispanics: Cuban 57.86 34.75	IBLK285	Black Interaction Of BASIANLN		0.03				1.72		
FEM290 Female Interaction Of BCUBANLN 1.33	BCUBANLN BCUBANQU BCUBANCU	B: Quadratic: Percent Hispanics: Cuban	57.86					11.54 34.75		
Black Interaction Of BCUBANLN Black Interaction Of BCUBANLN Black Interaction Of BCUBANLN Black Interaction Of BFNOTIOL Black Interaction Of BFNOTIO	IFEM290 IFEM291	Female Interaction Of BCUBANLN	0.03							
BROTQU B. Quadratic: Percent Females Separated, Divorced or Widowed B: Volte: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed BibLR297 Black Interaction Of BFNOTLN Black Interaction Of BFNOTCU Black Interaction	IBLK290	Black Interaction Of BCUBANLN	1.33							
FEM295 Female Interaction Of BFNOTLN SLX295 Black Interaction Of BFNOTCU SLX295 BFNOTLN BFNOTQU BFNOTCU BFNOTOR	B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed					7.18	83.72 0.03	61.42 21.01	0.00 0.00	
BLK295 Black Interaction Of BFNOTLN 36.26 Black Interaction Of BFNOTCU 36.26 Black Interaction Of BFNOTCU Black Interaction Of BFNOTQN 36.26 Black Interaction Of BINDIAQN Bi Quadratic: Percent Pop: American Indian, Eskimo, Aleut 14.74 0.00 0.00 0.01 0.00 31NDIAQN Bi Quadratic: Percent Pop: American Indian, Eskimo, Aleut 0.39 0.00 0	IFEM295	Female Interaction Of BFNOTLN					5.88	0.02	0.01	0.77
B. Linear: Percent Pop: American Indian, Eskimo, Aleut 89.05 0.00 0.03 0.00		Black Interaction Of BFNOTLN Black Interaction Of BFNOTCU						36.26		0.00
BLK300 BLK301 Black Interaction Of BINDIALN 0.68 0.00 99.37 6.75	BINDIALN BINDIAQU BINDIACU	B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut		14.74 0.00		0.00		43.39 0.01		0.00 0.00 0.00
BLK303 Black Interaction Of BINDIAQR 3.12 19.93 0.00	IBLK300 IBLK301	Black Interaction Of BINDIALN Black Interaction Of BINDIAQU		0.68 40.46		0.00		99.37 35.31		6.75 0.73
BPOVERLN Bright	IBLK303	Black Interaction Of BINDIAQR				0.00		19.93		0.00
B: Quadratic: Percent Families Below Poverty Level 0.02 0.00	BPOVERLN			29 37						0.00
BRICALN B: Linear: Percent Hispanics: Puerto Rican 0.00 2.12 0.00 B: Quadratic: Percent Hispanics: Puerto Rican 22.65 HISP315 Hispanic Interaction Of BPRICALN 1.94 HISP316 Hispanic Interaction Of BPRICAQU 1.94 B: Linear: Percent Hispanics: Puerto Rican 0.00 2.12 0.00 HISP315 Hispanic Interaction Of BPRICALN 1.94 B: Linear: Percent Associates Degree 0.063 BSCHASQU B: Quadratic: Percent Associates Degree 0.09 B: Cubic: Percent Associates Degree 0.09	BPOVERQU	B: Quadratic: Percent Families Below Poverty Level		0.02						0.00
Br Quadratic: Percent Hispanics: Puerto Rican 22.65	IOTH311	• • • • • • • • • • • • • • • • • • • •						2.12		0.00
HISP316 Hispanic Interaction Of BPRICAQU 1.94 3SCHASLN B: Linear: Percent Associates Degree 0.63 3SCHASQU B: Quadratic: Percent Associates Degree 13.65 3SCHASCU B: Cubic: Percent Associates Degree 0.099	BPRICALN BPRICAQU	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican		0.00				22.65		0.00
BSCHASQU B: Quadratic: Percent Associates Degree 13.65 B: Cubic: Percent Associates Degree 0.09	IHISP315 IHISP316	Hispanic Interaction Of BPRICALN Hispanic Interaction Of BPRICAQU						51.89 1.94		
	BSCHASLN BSCHASQU BSCHASCU	B: Quadratic: Percent Associates Degree		<u>.</u>			.		13.65	
	IBLK320	Black Interaction Of BSCHASLN							7.22	

NOTE: T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IBLK322	BIACK INTERACTION OF BSCHASCU							0.42	
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratuc: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander					34.34		76.52 0.09 48.88 0.00	
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Čubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban						94.41 14.84 81.24 4.25		
IFEM330 IFEM331 IFEM332 IFEM333	Female Interaction Of PCUBANLN Female Interaction Of PCUBANQU Female Interaction Of PCUBANCU Female Interaction Of PCUBANQR						0.58 17.81 62.35 0.01		
PINDIALN PINDIAQU PINDIACU	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut							0.01 2.07 0.00	
IBLK335 IBLK336 IBLK337	Black Interaction Of PINDIALN Black Interaction Of PINDIAQU Black Interaction Of PINDIACU							69.10 63.35 1.95	

F8. Significance Probabilities for Fixed Effect Coefficients for Past Year Dependency on Alcohol Only by Age

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	4.96	0.00	0.00	0.00	65.85	0.00	0.00	0.00
FEMBLCK	O: Black Interaction Of FEMALE	0.74	89.92	50.43	0.00	25.55	84.93	94.45	20.97
FEMHISP FEMOTHR	O: Hispanic Interaction Of FEMALE	95.05 64.23	0.00 19.71	0.00 32.50	0.00	86.83	83.81 0.00	26.07 50.36	0.52
RACEBLCK	O: Other Interaction Of FEMALE O: Race/Black Indicator	54.46	0.00	0.00	0.00	23.20	0.41	87.95	0.00
RACEHISP RACEOTHR	O: Race/Hispanic Indicator O: Race/Other Indicator	1.90 0.73	35.01 0.00	0.06 0.00	52.63 0.00	91.23	25.78 0.00	0.75 0.99	4.51 15.38
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator					70.41 45.32 51.64			
IBLK7 IBLK8 IBLK9	Black Interaction Of REGNOREA Black Interaction Of REGSOUTH Black Interaction Of REGWEST					28.15 30.03 18.21			
PDENLEV1 PDENLEV2 PDENLEV3	O: Large MSA O: Medium MSA O: Small MSA					9.37 16.12 81.48			2.73 70.98 1.63
PDENLEV4	O: NonMSA, Urban					53.42			0.00
IBLK10 IBLK11 IBLK12 IBLK13	Black Interaction Of PDENLEV1 Black Interaction Of PDENLEV2 Black Interaction Of PDENLEV3 Black Interaction Of PDENLEV4					66.79 92.18 23.52 46.49			0.00 0.00 0.54 6.10
PHH1PLN PHH1PQU	T: Linear: Percent One Person Households T: Quadratic: Percent One Person Households								0.00 0.00
IFEM15 IFEM16	Female Interaction Of PHH1PLN Female Interaction Of PHH1PQU								22.53 0.00
POPRMLN POPRMQU POPRMCU	T: Linear: Average Persons Per Room T: Quadratic: Average Persons Per Room T: Cubic: Average Persons Per Room		0.00				0.01 0.01 0.00		
IBLK20 IBLK21	Black Interaction Of POPRMLN Black Interaction Of POPRMQU						78.61 4.50		
PAGE44LN PAGE44QU	T: Linear: Percent Persons 35-44 Years T: Quadratic: Percent Persons 35-44 Years					95.95 3.94			
PSCH12LN PSCH12QU PSCH12CU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma				68.91 0.00				0.00 0.00 0.00
PSCHCOLN	T: Linear: Bachelors, Graduate, Or Professional Degree								0.00
PSCHSCLN PSCHSCQU PSCHSCCU	T: Linear: Percent Some College And No Degree T: Quadratic: Percent Some College And No Degree T: Cubic: Percent Some College And No Degree			0.32 63.96 0.00				11.60 0.01 73.54	
IFEM80 IFEM81 IFEM82	Female Interaction Of PSCHSCLN Female Interaction Of PSCHSCQU Female Interaction Of PSCHSCCU							29.01 63.88 0.00	
PPOVERLN PPOVERQU PPOVERCU PPOVERQR	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level T: Quartic: Percent Families Below Poverty Level					92.01 49.93 34.87 98.40		0.00 0.01 23.15 0.00	
IFEM85 IFEM86 IFEM87 IFEM88	Female Interaction Of PPOVERLN Female Interaction Of PPOVERQU Female Interaction Of PPOVERCU Female Interaction Of PPOVERQR					78.71 9.66 51.85 1.34			
IHISP85 IHISP86	Hispanic Interaction Of PPOVERLN Hispanic Interaction Of PPOVERQU					24.62 18.44		68.85 0.52	
P64DISLN P64DISQU P64DISCU P64DISQR	T: Linear: Percent 16-64 With A Work Disability T: Quadratic: Percent 16-64 With A Work Disability T: Cubic: Percent 16-64 With A Work Disability T: Quartic: Percent 16-64 With A Work Disability		0.35 9.42 11.69 15.40				0.01 2.99 15.47 0.00		
IBLK95 IBLK96 IBLK97 IBLK98	Black Interaction Of P64DISLN Black Interaction Of P64DISQU Black Interaction Of P64DISCU Black Interaction Of P64DISOR		3.64 0.00 88.81 0.00				12.94 43.73 16.76 2.58		
PHISPLN PHISPQU	T: Linear: Percent Hispanic T: Quadratic: Percent Hispanic								0.00 0.00
IBLK105 IBLK106	Black Interaction Of PHISPLN Black Interaction Of PHISPQU								33.38 0.00
PFNOTLN PFNOTQU	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed						36.75 40.18		
IOTH135 IOTH136	Other Interaction Of PFNOTLN Other Interaction Of PFNOTQU						41.02 0.28		
PMNOTLN PMNOTQU PMNOTCU	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed T: Cubic: Percent Males Separated, Divorced Or Widowed								0.07 0.00 0.00

NOTE:

		BIG CITY				REMAINDER	?		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
POLDHULN POLDHUQU POLDHUCU	T: Linear: Percent Housing Units Built 1939 Or Earlier T: Quadratic: Percent Housing Units Built 1939 Or Earlier T: Cubic: Percent Housing Units Built 1939 Or Earlier								0.55 16.50 0.00
ADRATILN ADRATIQU ADRATICU ADRATIQR	C: Linear: Death Rate With Explicit Mention Of Alcohol C: Quadratic: Death Rate With Explicit Mention Of Alcohol C: Cubic: Death Rate With Explicit Mention Of Alcohol C: Quartic: Death Rate With Explicit Mention Of Alcohol	30.62 77.02 59.08 0.83				34.36 93.74 90.66 1.38			
V18FLN V18FQU	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate					85.01 25.85			
IFEM185 IFEM186	Female Interaction Of V18FLN Female Interaction Of V18FQU					93.74 0.85			
V18ELN V18EQU V18ECU	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate			0.00 0.00	72.83 0.00 0.00		0.84	6.69 0.05	67.52 0.64 0.00
IHISP195 IHISP196	Hispanic Interaction Of V18ELN Hispanic Interaction Of V18EQU								13.91 0.05
V18ALN V18AQU	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate			0.64 0.42				70.37 0.02	36.12 0.00
IHISP200 IHISP201	Hispanic Interaction Of V18ALN Hispanic Interaction Of V18AQU								1.59 0.00
IOTH200 IOTH201	Other Interaction Of V18ALN Other Interaction Of V18AQU			0.12 0.00					
V18LN V18QU V18CU V18QR	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate C: Cubic: Total Drug Abuse Violations Arrest Rate C: Quartic: Total Drug Abuse Violations Arrest Rate	11.07		0.00 0.01	43.44 74.10 0.00	11.66 19.02 23.24 0.48		0.01 0.01	42.18 0.05 0.00
IBLK225 IBLK226	Black Interaction Of V18LN Black Interaction Of V18QU								0.62 0.00
DRATELN DRATEQU DRATECU DRATEQR	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992 C: Cubic: Mean Drug Client Treatment Rate 1991 & 1992 C: Quartic: Mean Drug Client Treatment Rate 1991 & 1992	7.10 6.13 85.42 3.80				2.29 2.25			
IBLK235 IBLK236 IBLK237 IBLK238	Black Interaction Of DRATELN Black Interaction Of DRATEQU Black Interaction Of DRATECU Black Interaction Of DRATECR	61.09 7.02 0.76 0.30							
RP80ALN RP80AQU RP80ACU RP80AQR	T: Lınear: Recoded Median Household Income T: Quadratic: Recoded Median Household Income T: Cubic: Recoded Median Household Income T: Quartic: Recoded Median Household Income		3.73 0.48 11.70 32.55				0.01 57.07 66.61 0.00		
IBLK250 IBLK251 IBLK252 IBLK253	Black Interaction Of RP80ALN Black Interaction Of RP80AQU Black Interaction Of RP80ACU Black Interaction Of RP80AQR		36.53 0.07 1.45 0.00				77.22 1.83 4.15 18.41		
ARATELN	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992		0.10			0.20	W2 20		
B64DISLN B64DISQU B64DISCU B64DISQR	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability B: Quartic: Percent Persons 16-64 With A Work Disability		0.12 64.85 33.49 45.70				82.29 44.26 3.11		
IHISP260 IHISP261 IHISP262 IHISP263	Hispanic Interaction Of B64DISLN Hispanic Interaction Of B64DISQU Hispanic Interaction Of B64DISCU Hispanic Interaction Of B64DISQR		1.75 64.92 4.41 0.00						
BAGE34LN BAGE34QU BAGE34CU BAGE34QR	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years		5.81 32.45				50.12 0.16 0.00		0.00 0.00 0.00 0.00
IFEM265 IFEM266	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU		0.00 0.00						
BAGE44LN BAGE44QU BAGE44CU BAGE44QR	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years B: Quartic: Percent Persons 35-44 Years			0.00 3.56 0.00		68.94 20.51 1.93		7.09 0.63 8.29 0.00	
IOTH270 IOTH271	Other Interaction Of BAGE44LN Other Interaction Of BAGE44QU			5.75 0.00				0.75	
BAGE54LN BAGE54QU	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years						39.63 0.81		
IOTH275 IOTH276	Other Interaction Of BAGE54LN Other Interaction Of BAGE54QU						47.24 0.58		
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population; Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander			66.77	0.00 0.00 89.60 0.00			35.64 48.54 74.27 0.67	
IHISP285 IHISP286 IHISP287 IHISP288	Hispanic Interaction Of BASIANLN Hispanic Interaction Of BASIANQU Hispanic Interaction Of BASIANCU Hispanic Interaction Of BASIANQR			0.00				49.43 17.81 35.93 0.94	
IOTH285 IOTH286 IOTH287 IOTH288	Other Interaction Of BASIANLN Other Interaction Of BASIANQU Other Interaction Of BASIANCU Other Interaction Of BASIANCU				37.09 0.00 0.00 0.00				

T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

NOTE:

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BCUBANLN BCUBANQU BCUBANCU BCUBANQR	B: Linear: Percent Hispanics: Cuban B: Quadratte: Percent Hispanics: Cuban B: Cubic: Percent Hispanics: Cuban B: Quartic: Percent Hispanics: Cuban		0.00 0.00	0.00 3.38			32.91 20.78 29.06 0.00		
IFEM290 IFEM291 IFEM292 IFEM293	Female Interaction Of BCUBANLN Female Interaction Of BCUBANQU Female Interaction Of BCUBANCU Female Interaction Of BCUBANQR		0.00 0.00				63.31 49.55 0.04 0.00		
IBLK290 IBLK291	Black Interaction Of BCUBANLN Black Interaction Of BCUBANQU			10.19 0.00					
BFNOTLN BFNOTQU	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed							1.10 92.94	0.00 24.93
IFEM295 IFEM296	Female Interaction Of BFNOTLN Female Interaction Of BFNOTQU								0.06 0.00
IOTH295 IOTH296	Other Interaction Of BFNOTLN Other Interaction Of BFNOTQU							0.16 0.00	
BINDIALN BINDIAQU BINDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut			0.00		60.23		0.92 10.17 1.09	0.01 0.00
IBLK300 IBLK301 IBLK302	Black Interaction Of BINDIALN Black Interaction Of BINDIAQU Black Interaction Of BINDIACU			0.00		2.59		50.83 63.91 0.47	
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU			0.00				35.26 3.21	
BMNOTEN BMNOTQU BMNOTCU BMNOTQR	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed B: Quartic: Percent Males Separated, Divorced or Widowed			0.00	0.00 0.00 46.81 0.00	55.18 11.37 88.51 4.51	9.17 31.83	8.02 0.00	0.14 0.00 0.00 17.98
IBLK305 IBLK306 IBLK307 IBLK308	Black Interaction Of BMNOTLN Black Interaction Of BMNOTQU Black Interaction Of BMNOTCU Black Interaction Of BMNOTQR				0.00 0.00 0.00				0.00 53.28 0.00 0.00
IHISP305 IHISP306 IHISP307 IHISP308	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU Hispanic Interaction Of BMNOTCU Hispanic Interaction Of BMNOTQR					31.66 57.74 87.35 7.17			
BPOVERLN BPOVERQU BPOVERCU	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level				0.00 0.00 0.00				0.00
IHISP310	Hispanic Interaction Of BPOVERLN								0.00
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican		0.00	0.02 2.63 0.00			0.13	3.05 16.50 12.62 0.00	
IBLK315 IBLK316 IBLK317 IBLK318	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU Black Interaction Of BPRICACU Black Interaction Of BPRICAQR							14.04 57.20 70.65 0.87	
PASIANLN PASIANQU PASIANCU PASIANQR	T: Linear: Percent Population: Asian, Pacific Islander I: Quadratuc: Percent Population: Asian, Pacific Islander I: Cubic: Percent Population: Asian, Pacific Islander T: Quartic: Percent Population: Asian, Pacific Islander	78.78 0.77 2.40				47.49 11.47 43.05 33.47	26.45 16.96 0.27		0.00 0.00 0.00
IFEM325 IFEM326 IFEM327 IFEM328	Female Interaction Of PASIANLN Female Interaction Of PASIANQU Female Interaction Of PASIANCU Female Interaction Of PASIANQR	93.85 1.06 1.57				6.84 37.34 83.07 1.61			
IBLK325 IBLK326 IBLK327	Black Interaction Of PASIANLN Black Interaction Of PASIANQU Black Interaction Of PASIANCU						31.62 55.41 0.22		
PCUBANLN PCUBANQU PCUBANCU	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban	56.65	0.00		0.03 0.00	82.30	26.17 89.19		0.00 6.12 0.00
IBLK330 IBLK331 IBLK332	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU	1.54			0.00 0.00	14.80			7.06 0.69 0.00
IOTH330 IOTH331	Other Interaction Of PCUBANLN Other Interaction Of PCUBANQU						8.87 3.06		
PINDIALN PINDIAQU PINDIACU PINDIAQR	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut T: Quartic: Percent Pop: American Indian, Eskimo, Aleut	50.38 32.08 82.38			0.00 0.00 0.00 0.00	50.95 5.07 14.36			
IFEM335 IFEM336 IFEM337	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU Female Interaction Of PINDIACU	26.50 10.17 1.88				95.94 1.57 2.55			
IHISP335 IHISP336 IHISP337 IHISP338	Hispanic Interaction Of PINDIALN Hispanic Interaction Of PINDIAQU Hispanic Interaction Of PINDIACU Hispanic Interaction Of PINDIAQR				0.00 0.00 0.00 0.00				
IOTH335 IOTH336	Other Interaction Of PINDIALN Other Interaction Of PINDIALN				0.00 0.00				

F9. Significance Probabilities for Fixed Effect Coefficients for Past Year Illicit Dependency by Age

		BIG CITY				REMAINDER	t		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	2.68	0.00	0.00	0.00	9.87	0.00	0.00	23.75
FEMBLCK FEMHISP	O: Black Interaction Of FEMALE	73.18 24.30	66.35 9.14	7.75 0.00	3.29 1.36	40.18 39.07	23.88	68.50 89.55	1.74 53.40
RACEBLCK RACEHISP RACEOTHR	O: Hispanic Interaction Of FEMALE O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	0.08 55.36	0.00 0.00 0.00 0.00	3.65 0.00	0.00 0.00	2.27 56.54	0.24 4.08 0.00	43.36 7.68	0.00 47.27
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator		0.00				0.00	5.24 8.23 0.00	
PHH1PLN	T: Linear: Percent One Person Households						0.00	0.00	
POPRMLN POPRMQU POPRMCU POPRMQR	T: Linear: Average Persons Per Room T: Quadratic: Average Persons Per Room T: Cubic: Average Persons Per Room T: Quartic: Average Persons Per Room			0.00 0.00				0.05 37.41 9.63 0.00	
PAGE18LN PAGE18QU PAGE18CU PAGE18QR	T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years T: Cubic: Percent Persons 0-18 Years T: Quartic: Percent Persons 0-18 Years		9.79 0.33 2.47 0.01				0.00 0.13 7.37 0.00		
IFEM25 IFEM26 IFEM27 IFEM28	Female Interaction Of PAGE18LN Female Interaction Of PAGE18QU Female Interaction Of PAGE18CU Female Interaction Of PAGE18QR		0.00 4.81 0.13 0.00				11.06 58.38 25.26 0.07		
PAGE34LN PAGE34QU	T: Linear: Percent Persons 25-34 Years T: Quadratic: Percent Persons 25-34 Years		1.33			3.30 0.06		3.46 0.00	
IBLK35	Black Interaction Of PAGE34LN					0.00		0.00	
IHISP35	Hispanic Interaction Of PAGE34LN		0.00						
PAGE44LN PAGE44QU	T: Linear: Percent Persons 35-44 Years T: Quadratic: Percent Persons 35-44 Years								0.00 0.00
PAGE54LN PAGE54QU PAGE54CU	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years T: Cubic: Percent Persons 45-54 Years				0.00 0.00 0.00	51.45 48.28 0.85			
IHISP45 IHISP46	Hispanic Interaction Of PAGE54LN Hispanic Interaction Of PAGE54QU				0.00 0.00				
PSCH12LN PSCH12QU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma	18.10 5.27							
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU	91.76 0.13							
PPOVERLN PPOVERQU PPOVERCU	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level				0.00 35.49 0.00				
PMNEVLN PMNEVQU PMNEVCU	T: Linear: Percent Males Never Married T: Quadratic: Percent Males Never Married T: Cubic: Percent Males Never Married			0.00 2.37 0.00					
PMLABLN PMLABQU PMLABCU	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force T: Cubic: Percent Males 16+ Years Old In Labor Force		0.01 0.15 0.00						
IHISP145 IHISP146 IHISP147	Hispanic Interaction Of PMLABLN Hispanic Interaction Of PMLABQU Hispanic Interaction Of PMLABCU		0.10 92.82 0.00						
PMNOTLN PMNOTQU	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed								11.82 0.00
PRENTLN PRENTQU	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented							0.08 0.00	0.00
ADRATELN ADRATEQU	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases							15.97 0.00	
V18FLN	C: Linear: Marijuana Posession Arrest Rate			0.00					
V18ELN V18EQU V18ECU V18EQR	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate C: Cubic: Opium/Cocaine & Deriv Posession Arrest Rate C: Quartic: Opium/Cocaine & Deriv Posession Arrest Rate			0.00 0.15 1.23 0.01				29.72 3.28 6.34 0.00	
IHISP195 IHISP196 IHISP197 IHISP198	Hispanic Interaction Of V18ELN Hispanic Interaction Of V18EQU Hispanic Interaction Of V18ECU Hispanic Interaction Of V18EQR			21.79 43.45 0.71 0.00					
V18LN V18QU	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate						76.05 0.01		
VIOLLN VIOLOU VIOLCU	C: Linear: Total Violent Offenses Arrest Rate C: Quadratic: Total Violent Offenses Arrest Rate C: Cubic: Total Violent Offenses Arrest Rate								0.00 0.00 0.00

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
KP8UALN RP80AQU RP80ACU	I.: Linear: Recoded Median Household Income T.: Quadratic: Recoded Median Household Income T.: Cubic: Recoded Median Household Income					21.98 0.01			0.00 76.48 0.00
IFEM250 IFEM251	Female Interaction Of RP80ALN Female Interaction Of RP80AQU					59.34 2.43			
ARATELN ARATEQU	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992						0.02 0.00		
B64DISLN	B: Linear: Percent Persons 16-64 With A Work Disability			0.00				15.10	
BAGE34LN BAGE34QU BAGE34CU	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years							15.10 10.47 3.28	
IFEM265 IFEM266 IFEM267	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU							46.07 1.93 0.00	
BAGE44LN BAGE44QU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years					0.01		58.46 0.06	
IHISP270 IHISP271	Hispanic Interaction Of BAGE44LN Hispanic Interaction Of BAGE44QU							91.16 12.09	
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years			0.00		43.40 4.50 7.08	0.05 0.25 0.00	0.07 0.00 0.15	56.68 0.21 0.00 0.00
IFEM275 IFEM276 IFEM277 IFEM278	Female Interaction Of BAGE54LN Female Interaction Of BAGE54QU Female Interaction Of BAGE54CU Female Interaction Of BAGE54QR					84.69 0.32		0.00	0.00 4.12 0.00 0.00
IBLK275 IBLK276 IBLK277	Black Interaction Of BAGE54LN Black Interaction Of BAGE54QU Black Interaction Of BAGE54CU							6.13 90.00 2.12	
BAGE64LN BAGE64QU BAGE64CU	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years			7.74 0.00 0.00		0.71	82.16 0.00	30.09 0.01 16.19	
IFEM280	Female Interaction Of BAGE64LN					0.21			
IBLK280 IBLK281 IBLK282	Black Interaction Of BAGE64LN Black Interaction Of BAGE64QU Black Interaction Of BAGE64CU			83.00 0.00 0.00				18.88 50.12 4.33	
BASIANLN BASIANQU BASIANCU BASIANQR	B: Linear: Percent Population: Asian, Pacific Islander B: Quadratic: Percent Population: Asian, Pacific Islander B: Cubic: Percent Population: Asian, Pacific Islander B: Quartic: Percent Population: Asian, Pacific Islander							0.00	0.00 1.30 0.00 26.99
IBLK285 IBLK286 IBLK287 IBLK288	Black Interaction Of BASIANLN Black Interaction Of BASIANQU Black Interaction Of BASIANCU Black Interaction Of BASIANQR								70.91 6.88 0.00 0.00
BCUBANLN BCUBANQU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban				0.00 0.00				
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed	0.08 58.29 86.25 0.26				12.10 5.56 72.29 0.12	0.03		
IFEM295 IBLK295	Female Interaction Of BFNOTLN Black Interaction Of BFNOTLN					74.70	0.00		
IBLK296 IBLK297	Black Interaction Of BFNOTQU Black Interaction Of BFNOTQU					82.01 24.76			
IHISP295 IHISP296 IHISP297 IHISP298	Hispanic Interaction Of BFNOTLN Hispanic Interaction Of BFNOTQU Hispanic Interaction Of BFNOTCU Hispanic Interaction Of BFNOTCU	71.13 37.92 64.52 1.22							
BINDIALN BINDIAQU BINDIACU BINDIAQR	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut B: Quartic: Percent Pop: American Indian, Eskimo, Aleut			0.00 0.60 0.00			13.06 0.48 8.48 0.00	0.16 0.24 29.84 1.39	0.00
IBLK300 IBLK301 IBLK302 IBLK303	Black Interaction Of BINDIALN Black Interaction Of BINDIAQU Black Interaction Of BINDIACU Black Interaction Of BINDIAQR							28.21 18.39 87.54 3.85	
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU							2.15 1.22	
BPOVERLN BPOVERQU BPOVERCU	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level		0.00 0.16 45.05			0.16	35.74 74.12 0.03		0.00 0.00
IOTH310 IOTH311 IOTH312	Other Interaction Of BPOVERLN Other Interaction Of BPOVERQU Other Interaction Of BPOVERCU		0.59 0.51 0.00						
BPRICALN BPRICAQU BPRICACU BPRICAQR	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican B: Quartic: Percent Hispanics: Puerto Rican					62.71 2.66 72.91 0.10	0.00		
IFEM315	Female Interaction Of BPRICALN P. Lingar, Paragut Associates Pagree				0.00	3.55			
BSCHASLN	B: Linear: Percent Associates Degree				0.00	l			

T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

NOTE:

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BSCHASQU BSCHASCU	B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree				/5.20 0.00				
IHISP320 IHISP321 IHISP322	Hispanic Interaction Of BSCHASLN Hispanic Interaction Of BSCHASQU Hispanic Interaction Of BSCHASCU				0.00 0.00 0.00				
PASIANLN	T: Linear: Percent Population: Asian, Pacific Islander		0.00				0.00		
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban I: Quartic: Percent Hispanics: Cuban	57.23 87.26		11.78 0.00				5.25 0.25 13.14 0.00	0.00 0.00 0.00 0.00
IFEM330	Female Interaction Of PCUBANLN	1.97							
IBLK330 IBLK331 IBLK332 IBLK333	Black Interaction Of PCUBANLN Black Interaction Of PCUBANQU Black Interaction Of PCUBANCU Black Interaction Of PCUBANQR	71.96 0.02							55.75 0.75 24.02 0.00

F10. Significance Probabilities for Fixed Effect Coefficients for Past Year Arrested by Age

			BIG CI	ITY			REMAI	INDER	
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMBLCK	O: Black Interaction Of FEMALE	48.80	0.03	56.74	0.00	57.42	22.89	41.07	0.09
FEMHISP	O: Hispanic Interaction Of FEMALE	29.73	0.00	0.05	0.00	62.42	49.28	90.88	5.05
RACEBLCK RACEHISP RACEOTHR	O: Race/Black Indicator O: Race/Hispanic Indicator O: Race/Other Indicator	12.36 22.85 0.51	0.00 55.04 0.00	0.03 0.11	0.00 5.57	58.00 20.39 22.45	23.82 32.54 2.82	4.25 32.31	97.92 0.00
REGNOREA REGSOUTH REGWEST	O: Northeast Region Indicator O: South Region Indicator O: West Region Indicator					52.14 11.27 54.06			7.59 13.43 51.68
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban						15.24 25.55 97.72 32.15	0.44 0.08 0.00 0.00	
IBLK10 IBLK11 IBLK12 IBLK13	Black Interaction Of PDENLEV1 Black Interaction Of PDENLEV2 Black Interaction Of PDENLEV3 Black Interaction Of PDENLEV4						44.04 41.26 1.16 26.12		
PHH1PLN PHH1PQU PHH1PCU PHH1PQR	T: Linear: Percent One Person Households T: Quadratic: Percent One Person Households T: Cubic: Percent One Person Households T: Quartic: Percent One Person Households		0.00 11.46 39.48 0.06				2.03 5.53 0.00 0.04		
IBLK15 IBLK16 IBLK17 IBLK18	Black Interaction Of PHH1PLN Black Interaction Of PHH1PQU Black Interaction Of PHH1PCU Black Interaction Of PHH1PCU		69.17 79.59 1.78 0.00				8.32 68.31 2.22 3.89		
POPRMLN	T: Linear: Average Persons Per Room		11.41				0.10		
PAGE34LN	T: Linear: Percent Persons 25-34 Years	:							0.00
PAGE54LN PAGE54QU PAGE54CU	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years T: Cubic: Percent Persons 45-54 Years							8.43 10.86 0.00	
IHISP45 IHISP46 IHISP47	Hispanic Interaction Of PAGE54LN Hispanic Interaction Of PAGE54QU Hispanic Interaction Of PAGE54CU							43.22 21.15 1.71	
PAGE64LN	T: Linear: Percent Persons 55-64 Years					0.14			
PSCH8LN PSCH8QU	T: Linear: Percent 0-8 Years Of School T: Quadratic: Percent 0-8 Years Of School	4.50 58.52							
IOTH55 IOTH56	Other Interaction Of PSCH8LN Other Interaction Of PSCH8QU	10.17 0.97							
PSCH12LN PSCH12QU PSCH12CU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma		0.00				0.16 2.03 0.05		
IHISP60	Hispanic Interaction Of PSCH12LN						0.92		
PSCHCOLN PSCHCOQU PSCHCOCU	T: Linear: Bachelors, Graduate, Or Professional Degree T: Quadratic: Bachelors, Graduate, Or Professional Degree T: Cubic: Bachelors, Graduate, Or Professional Degree								0.00 12.60 0.00
IBLK75 IBLK76 IBLK77	Black Interaction Of PSCHCOLN Black Interaction Of PSCHCOQU Black Interaction Of PSCHCOCU								4.05 33.42 0.00
PPUBASLN	T: Linear: Percent Households With Public Assist Income						0.00		
P64DISLN	T: Linear: Percent 16-64 With A Work Disability			0.00	l			0.00	
PBLACKLN PBLACKQU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic		1.37 0.00						
POTHLN	T: Linear: Percent Other Race/Hispanicity				0.00				
IHISP110	Hispanic Interaction Of POTHLN				0.00				
PFLABLN PFLABQU	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force				0.00				4.67 0.00
IFEM125	Female Interaction Of PFLABLN				0.00				
PFNOTLN PFNOTQU	T: Linear: Percent Females Separated, Divorced Or Widowed T: Quadratic: Percent Females Separated, Divorced Or Widowed							0.28 0.00	
PMLABLN PMLABQU	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force				3.07		0.02 0.02		0.00
IFEM145	Female Interaction Of PMLABLN				0.00				
PMNOTEN PMNOTQU PMNOTCU PMNOTQR	T: Linear: Percent Males Separated, Divorced Or Widowed T: Quadratic: Percent Males Separated, Divorced Or Widowed T: Cubic: Percent Males Separated, Divorced Or Widowed T: Quartic: Percent Males Separated, Divorced Or Widowed						63.85 5.35 20.42 0.00		
POLDHULN	T: Linear: Percent Housing Units Built 1939 Or Earlier	0.43			1	69.50			

NOTE: T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

			BIG CI	TY		REMAINDER				
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+	
POLDHUQU POLDHUCU	1: Quadratic: Percent Housing Units Built 1939 Or Earlier T: Cubic: Percent Housing Units Built 1939 Or Earlier	·				52.75 1.14	•			
P40HULN	T: Linear: Percent Housing Units Built 1940-1949	0.00				1.14		57.80		
P40HUQU PRENTLN	T: Quadratic: Percent Housing Units Built 1940-1949 T: Linear: Percent Housing Rented					0.00		0.00		
ADRATILN	C: Linear: Death Rate With Explicit Mention Of Alcohol		69.93	0.00		0.00				
IFEM180	Female Interaction Of ADRAT1LN		0.00							
V18FLN V18FQU	C: Linear: Marijuana Posession Arrest Rate C: Quadratic: Marijuana Posession Arrest Rate			0.00	ļ			0.00 0.00		
V18ELN V18EQU	C: Linear: Opium/Cocaine & Deriv Posession Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Posession Arrest Rate			0.00	1.21 0.00			0.00	61.90 0.00	
IFEM195	Female Interaction Of V18ELN Female Interaction Of V18EQU								0.00	
FEM196 V18ALN	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate							0.01	0.00	
ORATELN ORATEQU ORATECU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992 C: Cubic: Mean Drug Client Treatment Rate 1991 & 1992	10.41 88.91 2.64								
RH43ALN RH43AQU RH43ACU RH43AQR	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units 1: Cubic: Recoded Median Rents For Rental Units T: Quartic: Recoded Median Rents For Rental Units				0.00			31.59 32.80 15.38 0.00	0.00	
RH61ALN RH61AQU RH61ACU	T: Linear: Recoded Median Value Of Owner Occupied HUs T: Quadratic: Recoded Median Value Of Owner Occupied HUs T: Cubic: Recoded Median Value Of Owner Occupied HUs		85.17	0.00 0.00	ļ		12.59 1.43 0.03	58.62 99.01 0.01		
HISP245 HISP246 HISP247	Hispanic Interaction Of RH61 ALN Hispanic Interaction Of RH61 AQU Hispanic Interaction Of RH61 ACU				į į			25.33 17.71 5.11		
ARATELN ARATEQU ARATECU ARATEQR	C: Linear: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quadratic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Cubic: Mean A-Only Client Treatment Rate 1991 & 1992 C: Quartic: Mean A-Only Client Treatment Rate 1991 & 1992								74.64 2.53 0.00 0.42	
BLK255 BLK256 BLK257 BLK258	Black Interaction Of ARATELN Black Interaction Of ARATEQU Black Interaction Of ARATECU Black Interaction Of ARATEQR								80.25 1.24 55.85 0.02	
64DISLN 64DISQU 64DISCU 64DISQR	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability B: Quartic: Percent Persons 16-64 With A Work Disability								0.01 13.60 0.00 0.00	
BAGE34LN BAGE34QU BAGE34CU BAGE34QR	B: Linear: Percent Persons 25-34 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years				0.00 1.30		71.42 6.91 0.07 28.51			
FEM265 FEM266 FEM267 FEM268	Female Interaction Of BAGE34LN Female Interaction Of BAGE34QU Female Interaction Of BAGE34CU Female Interaction Of BAGE34QR						11.54 3.06 98.61 0.00			
HISP265 HISP266	Hispanic Interaction Of BAGE34LN Hispanic Interaction Of BAGE34QU				80.41 0.00					
AGE44LN AGE44QU AGE44CU AGE44QR	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years B: Quartic: Percent Persons 35-44 Years B: Quartic: Percent Persons 35-44 Years						1.16 0.02 22.01 0.00	95.30 81.49 0.02		
BLK270 BLK271 BLK272 BLK273	Black Interaction Of BAGE44LN Black Interaction Of BAGE44QU Black Interaction Of BAGE44CU Black Interaction Of BAGE44QR						39.46 32.07 41.39 0.13			
HISP270 HISP271	Hispanic Interaction Of BAGE44LN Hispanic Interaction Of BAGE44QU							49.68 2.28		
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years						0.01 21.09 1.70 0.00	_,_,		
BAGE64LN BAGE64QU BAGE64CU	B: Linear: Percent Persons 55-64 Years B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years		0.00			0.13	23.83 1.55 7.40			
HISP280 HISP281 HISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		0.00				83.83 47.75 8.08			
INDIALN INDIAQU INDIACU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut B: Cubic: Percent Pop: American Indian, Eskimo, Aleut		86.71			4.08 47.84 0.91				
OTH300	Other Interaction Of BINDIALN		0.00							
MNOTLN BMNOTQU	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed	31.53 0.00								
POVERLN BPOVERQU BPOVERCU	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level							20.47 7.60 0.00		
BPRICALN	B: Linear: Percent Hispanics: Puerto Rican				l	28.41				

NOTE: T: Indicates a tract-level variable, C: Indicates a county-level variable, B: Indicates a block-level variable, O: Other

			TY		REMAI	NDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
BPRICAQU BPRICACU IFEM315 IFEM316 IFEM317 BSCHASLN BSCHASQU BSCHASQU BSCHASQR PASIANLN PASIANQU PASIANCU IFEM325 IFEM326 IFEM326 IFEM327	B: Quadratic: Percent Hispanics: Puerto Kican B: Cubic: Percent Hispanics: Puerto Rican B: Cubic: Percent Hispanics: Puerto Rican Female Interaction Of BPRICAQU Female Interaction Of BPRICAQU B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree T: Linear: Percent Associates Degree T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander Female Interaction Of PASIANLN Female Interaction Of PASIANQU Female Interaction Of PASIANQU Female Interaction Of PASIANCU	4.73		0.00		9.86 12.75 16.41 49.28 4.19 44.93 65.53 0.03 39.47 39.05 1.23 31.71 77.29 0.43		97.97 50.87 40.06 0.00	
PCUBANLN PCUBANQU PCUBANCU	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics; Cuban							70.18 66.99 0.00	

F11. Significance Probabilities for Fixed Effect Coefficients for Past Year Treatment Needed for Drug Abuse by Age

		BIG CITY	BIG CITY			REMAINDER	<u> </u>		
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
DUMMY	O: Intercept Term	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
FEMALE	O: Female Indicator	7.65	0.00	0.00	0.00	92.83	0.00	0.22	0.00
FEMBLCK	O: Black Interaction Of FEMALE	19.99	35.11	0.61	0.00		28.88	1.63	21.27
FEMHISP	O: Hispanic Interaction Of FEMALE	0.80	9.91	0.00	58.34		85.11	7.65	20.98
FEMOTHR RACEBLCK	O: Other Interaction Of FEMALE O: Race/Black Indicator	57.47 8.01	5.65 0.01	0.22 85.08	0.00		0.06 14.20	84.98 0.02	0.00
RACEHISP RACEOTHR	O: Race/Hispanic Indicator O: Race/Other Indicator	9.19 0.16	0.00 0.00	0.00 0.00	0.00 0.00 0.00		4.24 0.05	40.03 0.00	50.49 2.56
PDENLEV1 PDENLEV2 PDENLEV3 PDENLEV4	O: Large MSA O: Medium MSA O: Small MSA O: NonMSA, Urban						13.27 24.72 75.63 92.29		
IBLK10 IBLK11 IBLK12 IBLK13	Black Interaction Of PDENLEV1 Black Interaction Of PDENLEV2 Black Interaction Of PDENLEV3 Black Interaction Of PDENLEV4						69.86 54.22 14.95 14.56		
PHH1PLN PHH1PQU PHH1PCU PHH1PQR	T: Linear: Percent One Person Households T: Quadratic: Percent One Person Households T: Cubic: Percent One Person Households T: Quartic: Percent One Person Households T: Quartic: Percent One Person Households				0.00 0.00 1.43 0.00				0.00 0.00 0.00 0.00
PAGE18LN PAGE18QU PAGE18CU	T: Linear: Percent Persons 0-18 Years T: Quadratic: Percent Persons 0-18 Years T: Cubic: Percent Persons 0-18 Years						48.37 0.05 2.40		
IOTH25 IOTH26	Other Interaction Of PAGE18LN Other Interaction Of PAGE18QU						79.97 0.01		
PAGE24LN PAGE24QU PAGE24CU PAGE24QR	T: Linear: Percent Persons 19-24 Years 1: Quadratic: Percent Persons 19-24 Years 1: Cubic: Percent Persons 19-24 Years 1: Quatric: Percent Persons 19-24 Years 1: Quartic: Percent Persons 19-24 Years								0.00 0.00 0.08 0.00
PAGE44LN	T: Linear: Percent Persons 35-44 Years				0.00				
PAGE54LN PAGE54QU PAGE54CU PAGE54QR	T: Linear: Percent Persons 45-54 Years T: Quadratic: Percent Persons 45-54 Years T: Cubic: Percent Persons 45-54 Years T: Quartic: Percent Persons 45-54 Years				0.52		0.01		0.00 0.07 0.00 0.00
IFEM45	Female Interaction Of PAGE54LN				0.00				
IHISP45	Hispanic Interaction Of PAGE54LN						2.75		
PSCH12LN PSCH12QU PSCH12CU	T: Linear: Percent 9-12 Years & No High School Diploma T: Quadratic: Percent 9-12 Years & No High School Diploma T: Cubic: Percent 9-12 Years & No High School Diploma							0.00 0.31 6.24	
IFEM60 IFEM61	Female Interaction Of PSCH12LN Female Interaction Of PSCH12QU							67.76 0.04	
PPOVERLN PPOVERQU PPOVERCU	T: Linear: Percent Families Below Poverty Level T: Quadratic: Percent Families Below Poverty Level T: Cubic: Percent Families Below Poverty Level					69.30 64.31 0.24			
PBLACKLN PBLACKQU	T: Linear: Percent Black Nonhispanic T: Quadratic: Percent Black Nonhispanic							0.04 0.00	
IOTH100 IOTH101	Other Interaction Of PBLACKLN Other Interaction Of PBLACKQU							0.00 0.01	
PFLABLN PFLABQU PFLABCU PFLABQR	T: Linear: Percent Females 16+ Years Old In Labor Force T: Quadratic: Percent Females 16+ Years Old In Labor Force T: Cubic: Percent Females 16+ Years Old In Labor Force T: Quartic: Percent Females 16+ Years Old In Labor Force					7.88 2.02	9.40 4.63 58.51 0.01		
IBLK125 IBLK126	Black Interaction Of PFLABLN Black Interaction Of PFLABQU						95.71 2.17		
PFNEVLN PFNEVQU	T: Linear: Percent Females Never Married T: Quadratic: Percent Females Never Married							3.55 0.00	
IFEM130 IFEM131	Female Interaction Of PFNEVLN Female Interaction Of PFNEVQU							55.72 1.73	
PMNEVLN PMNEVQU PMNEVCU	T: Lmear: Percent Males Never Married T: Quadratic: Percent Males Never Married T: Cubic: Percent Males Never Married				61.10 0.00				0.00 0.00 0.00
PMLABLN PMLABQU PMLABCU PMLABQR	T: Linear: Percent Males 16+ Years Old In Labor Force T: Quadratic: Percent Males 16+ Years Old In Labor Force T: Cubic: Percent Males 16+ Years Old In Labor Force T: Quartic: Percent Males 16+ Years Old In Labor Force								13.88 0.26 0.00 0.00
PMNOTLN	T: Linear: Percent Males Separated, Divorced Or Widowed							0.02	
POLDHULN P40HULN	T: Linear: Percent Housing Units Built 1939 Or Earlier T: Linear: Percent Housing Units Built 1940-1949	55.11				16.13 44.36			
P40HÚQU P40HÚCU	T: Quadratic: Percent Housing Units Built 1940-1949 T: Cubic: Percent Housing Units Built 1940-1949	20.64				61.03 76.71			

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IHISP160 IHISP161	Hispanic Interaction Of P40HULN Hispanic Interaction Of P40HUQU	50.65 11.69							
PRENTLN PRENTQU PRENTCU PRENTQR	T: Linear: Percent Housing Rented T: Quadratic: Percent Housing Rented T: Cubic: Percent Housing Rented T: Quartic: Percent Housing Rented			0.00				41.92 2.13 67.57 20.14	0.00
IHISP165 IHISP166 IHISP167	Hispanic Interaction Of PRENTLN Hispanic Interaction Of PRENTQU Hispanic Interaction Of PRENTCU							26.75 43.50 12.01	
ADRATELN ADRATEQU ADRATECU ADRATEQR	C: Linear: Death Rate For All Alcohol-Related Cases C: Quadratic: Death Rate For All Alcohol-Related Cases C: Cubic: Death Rate For All Alcohol-Related Cases C: Quartic: Death Rate For All Alcohol-Related Cases						49.93 0.00 32.60 0.99		
IBLK175 IBLK176 IBLK177 IBLK178	Black Interaction Of ADRATELN Black Interaction Of ADRATEQU Black Interaction Of ADRATECU Black Interaction Of ADRATECU						66.11 37.95 23.28 26.61		
ADRAT1LN ADRAT1QU	C: Linear: Death Rate With Explicit Mention Of Alcohol C: Quadratic: Death Rate With Explicit Mention Of Alcohol								0.00 0.11
IFEM180 IFEM181	Female Interaction Of ADRATILN Female Interaction Of ADRATIQU								0.00
V18FLN	C: Linear: Marijuana Posession Arrest Rate				0.00				
V18BLN V18BQU V18BCU V18BQR	C: Linear: Marijuana Sale/Manufacture Arrest Rate C: Quadratic: Marijuana Sale/Manufacture Arrest Rate C: Cubic: Marijuana Sale/Manufacture Arrest Rate C: Quartic: Магіjuana Sale/Manufacture Arrest Rate			0.02				9.17 14.45 0.00	91.31 0.54 0.00 0.00
IFEM190 IFEM191 IFEM192 IFEM193	Female Interaction Of V18BLN Female Interaction Of V18BQU Female Interaction Of V18BCU Female Interaction Of V18BCR								29.94 0.00 0.00 0.00
IBLK190 IBLK191 IBLK192	Black Interaction Of V18BLN Black Interaction Of V18BQU Black Interaction Of V18BCU							29.26 19.27 1.36	
IHISP190	Hispanic Interaction Of V18BLN			0.00					
V18ALN V18AQU	C: Linear: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate C: Quadratic: Opium/Cocaine & Deriv Sale/Manuf Arrest Rate								0.00 0.00
IHISP200 IHISP201	Hispanic Interaction Of V18ALN Hispanic Interaction Of V18AQU								17.14 0.00
V18LN V18QU	C: Linear: Total Drug Abuse Violations Arrest Rate C: Quadratic: Total Drug Abuse Violations Arrest Rate						0.05 0.10		
DRATELN DRATEQU	C: Linear: Mean Drug Client Treatment Rate 1991 & 1992 C: Quadratic: Mean Drug Client Treatment Rate 1991 & 1992						7.48 31.24		
IBLK235 IBLK236	Black Interaction Of DRATELN Black Interaction Of DRATEQU						54.88 70.51		
RH43ALN RH43AQU	T: Linear: Recoded Median Rents For Rental Units T: Quadratic: Recoded Median Rents For Rental Units								71.29 0.00
IBLK240 IBLK241	Black Interaction Of RH43ALN Black Interaction Of RH43AQU								0.50 0.00
RP80ALN	T: Linear: Recoded Median Household Income		0.00						0.00
IBLK250	Black Interaction Of RP80ALN		0.00						
B64DISLN B64DISQU B64DISCU	B: Linear: Percent Persons 16-64 With A Work Disability B: Quadratic: Percent Persons 16-64 With A Work Disability B: Cubic: Percent Persons 16-64 With A Work Disability						0.98	0.00 9.12 0.02	
IHISP260 IHISP261	Hispanic Interaction Of B64DISLN Hispanic Interaction Of B64DISQU							72.92 95.80	
IHISP262 IOTH260	Hispanic Interaction Of B64DISCU Other Interaction Of B64DISLN							14.57	
IOTH261 BAGE34LN	Other Interaction Of B64DISQU B: Linear: Percent Persons 25-34 Years						44.43	0.00	0.12
BAGE34QU BAGE34CU BAGE34QR	B: Linear: Percent Persons 25-24 Years B: Quadratic: Percent Persons 25-34 Years B: Cubic: Percent Persons 25-34 Years B: Quartic: Percent Persons 25-34 Years						66.62		0.13 49.08 0.04 1.06
IBLK265 IBLK266 IBLK267 IBLK268	Black Interaction Of BAGE34LN Black Interaction Of BAGE34QU Black Interaction Of BAGE34CU Black Interaction Of BAGE34QR								18.20 31.12 43.05 0.06
BAGE44LN BAGE44QU BAGE44CU	B: Linear: Percent Persons 35-44 Years B: Quadratic: Percent Persons 35-44 Years B: Cubic: Percent Persons 35-44 Years	72.09				0.10		69.21	45.14 90.57 0.00
IOTH270	Other Interaction Of BAGE44LN	0.15						0.01	
BAGE54LN BAGE54QU BAGE54CU BAGE54QR	B: Linear: Percent Persons 45-54 Years B: Quadratic: Percent Persons 45-54 Years B: Cubic: Percent Persons 45-54 Years B: Quartic: Percent Persons 45-54 Years					95.24 27.71 38.39 2.77	0.03		
IOTH275 BAGE64LN	Other Interaction Of BAGE54LN B: Linear: Percent Persons 55-64 Years		0.00				0.00 40.18		
BAGE64QU BAGE64CU	B: Quadratic: Percent Persons 55-64 Years B: Cubic: Percent Persons 55-64 Years		28.61 0.01				0.39 9.70		

NOTE:

		BIG CITY				REMAINDER			
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IBLK280 IBLK281 IBLK282	Black Interaction Of BAGE644LN Black Interaction Of BAGE64QU Black Interaction Of BAGE64CU						9.66 77.29 19.48		
IHISP280 IHISP281 IHISP282	Hispanic Interaction Of BAGE64LN Hispanic Interaction Of BAGE64QU Hispanic Interaction Of BAGE64CU		0.00 0.67 0.00						
BCUBANLN BCUBANQU	B: Linear: Percent Hispanics: Cuban B: Quadratic: Percent Hispanics: Cuban								0.01 0.00
BFNOTLN BFNOTQU BFNOTCU BFNOTQR	B: Linear: Percent Females Separated, Divorced or Widowed B: Quadratic: Percent Females Separated, Divorced or Widowed B: Cubic: Percent Females Separated, Divorced or Widowed B: Quartic: Percent Females Separated, Divorced or Widowed	43.24	0.06 53.26			40.71 10.52 80.21 7.38	3.68 1.17 6.69 56.30	5.21 0.00 2.52 36.06	
IBLK295 IBLK296 IBLK297 IBLK298	Black Interaction Of BFNOTLN Black Interaction Of BFNOTQU Black Interaction Of BFNOTCU Black Interaction Of BFNOTQR						20.11 4.19 38.34 12.24		
IHISP295 IHISP296 IHISP297	Hispanic Interaction Of BFNOTLN Hispanic Interaction Of BFNOTQU Hispanic Interaction Of BFNOTCU		0.49 0.04				21.23 18.49 11.71		
IOTH295 IOTH296 IOTH297 IOTH298	Other Interaction Of BFNOTLN Other Interaction Of BFNOTOU Other Interaction Of BFNOTCU Other Interaction Of BFNOTCU	1.49						0.24 0.00 24.95 0.00	
BINDIALN BINDIAQU	B: Linear: Percent Pop: American Indian, Eskimo, Aleut B: Quadratic: Percent Pop: American Indian, Eskimo, Aleut						0.89 0.76		23.95 0.00
IFEM300 IFEM301	Female Interaction Of BINDIALN Female Interaction Of BINDIAQU						49.06 3.44		
IHISP300 IHISP301	Hispanic Interaction Of BINDIALN Hispanic Interaction Of BINDIAQU								30.92 0.45
BMNOTEN BMNOTQU BMNOTCU BMNOTQR	B: Linear: Percent Males Separated, Divorced or Widowed B: Quadratic: Percent Males Separated, Divorced or Widowed B: Cubic: Percent Males Separated, Divorced or Widowed B: Quartic: Percent Males Separated, Divorced or Widowed							18.85 60.11 37.47 1.37	
IHISP305 IHISP306 IHISP307 IHISP308	Hispanic Interaction Of BMNOTLN Hispanic Interaction Of BMNOTQU Hispanic Interaction Of BMNOTCU Hispanic Interaction Of BMNOTQR							57.75 94.14 4.89 6.21	
BPOVERLN BPOVERQU BPOVERCU BPOVERQR	B: Linear: Percent Families Below Poverty Level B: Quadratic: Percent Families Below Poverty Level B: Cubic: Percent Families Below Poverty Level B: Quartic: Percent Families Below Poverty Level		0.00 0.00		0.00 0.00		17.78		0.00 0.00 5.41 0.00
IFEM310 IFEM311 IFEM312 IFEM313	Female Interaction Of BPOVERLN Female Interaction Of BPOVERQU Female Interaction Of BPOVERCU Female Interaction Of BPOVERQR				0.95 0.00		0.04		30.50 0.00 0.00 0.00
IBLK310 IBLK311	Black Interaction Of BPOVERLN Black Interaction Of BPOVERQU		0.00 0.03						
BPRICALN BPRICAQU	B: Linear: Percent Hispanics: Puerto Rican B: Quadratic: Percent Hispanics: Puerto Rican						0.00 35.86		
IBLK315 IBLK316	Black Interaction Of BPRICALN Black Interaction Of BPRICAQU						3.36 0.15		
BSCHASLN BSCHASQU BSCHASCU BSCHASQR	B: Linear: Percent Associates Degree B: Quadratic: Percent Associates Degree B: Cubic: Percent Associates Degree B: Quartic: Percent Associates Degree			7.11				3.48 53.40 0.61 0.08	0.00
IFEM320 IFEM321 IFEM322 IFEM323	Female Interaction Of BSCHASLN Female Interaction Of BSCHASQU Female Interaction Of BSCHASCU Female Interaction Of BSCHASQR							98.07 68.29 34.09 0.19	
IBLK320	Black Interaction Of BSCHASLN			0.00					
IOTH320 IOTH321	Other Interaction Of BSCHASLN Other Interaction Of BSCHASQU							0.92 0.00	
PASIANLN PASIANQU PASIANCU	T: Linear: Percent Population: Asian, Pacific Islander T: Quadratic: Percent Population: Asian, Pacific Islander T: Cubic: Percent Population: Asian, Pacific Islander				0.01			0.12 15.38 0.14	
IBLK325 IBLK326 IBLK327	Black Interaction Of PASIANLN Black Interaction Of PASIANQU Black Interaction Of PASIANCU							41.19 23.63 0.12	
IOTH325	Other Interaction Of PASIANLN				0.00				
PCUBANLN PCUBANQU PCUBANCU PCUBANQR	T: Linear: Percent Hispanics: Cuban T: Quadratic: Percent Hispanics: Cuban T: Cubic: Percent Hispanics: Cuban T: Quartic: Percent Hispanics: Cuban		·	35.61 0.00 0.00				65.83 0.02 4.15 2.37	
PINDIALN PINDIAQU PINDIACU	T: Linear: Percent Pop: American Indian, Eskimo, Aleut T: Quadratic: Percent Pop: American Indian, Eskimo, Aleut T: Cubic: Percent Pop: American Indian, Eskimo, Aleut					24.02 4.85 5.68	0.76 0.07	3.33 25.37 0.01	
IFEM335 IFEM336	Female Interaction Of PINDIALN Female Interaction Of PINDIAQU						46.11 2.05		
IOTH335 IOTH336	Other Interaction Of PINDIALN Other Interaction Of PINDIAQU							14.52 0.12	

		BIG CITY			REMAINDER				
Variable	Label	12-17	18-25	26-34	35+	12-17	18-25	26-34	35+
IO1H35/	Other Interaction Of PINDIACO							0.00	

Appendix G: Details of Estimating Variances

Appendix G. Details of Estimating Variances

G.1 Derivation of Formulas

As discussed in Section 2.6, we want to estimate¹

$$MSE_{i} = E(\hat{\pi}_{i}^{W} - \pi_{i})^{2}$$
, (G.1)

where π_i is the sum of person-level propensities for area i. Expanding (G.1), we have that it is equal to

$$MSE_{i} = E\left[\sum_{a} \sum_{b} \sum_{b} N_{ib}^{ad} \hat{\pi}_{ib}^{ad} - \sum_{a} \sum_{d} \sum_{b} N_{ib}^{ad} \pi_{ib}^{ad}\right]^{2}$$

$$= \sum_{a} E\left[\sum_{d} \sum_{b} N_{ib}^{ad} (\hat{\pi}_{ib}^{ad} - \pi_{ib}^{ad})\right]^{2} = \sum_{a} MSE_{i}^{a}$$
(G.2)

This approximation ignores the covariances of the errors across the domains for which separate models were fit. Recall that there were four such domains, defined by age, as discussed in Section 2.5. We believe that these covariances are mostly positive since over- or underestimation of the substance abuse rate for one age domain in an area is probably accompanied by similar over- or underestimation of the rate for the other age domains in the same area, but the sizes and signs of the covariances are unknown. We were forced to ignore them by the computing constraints.

To simplify notation, we will now focus on a particular area i and age group a, dropping the a subscript since the age groups were modeled separately. We may then write the inner expectation in (G.2) as

MSE_i = E
$$\left[\sum_{d}\sum_{b}N_{ib}^{d}(\hat{\pi}_{ib}^{d} - \pi_{ib}^{d})\right]^{2} = E\left[\tilde{N}^{t}(\hat{\pi} - \pi)\right]^{2}$$
 (G.3)

where \tilde{N} is the column matrix of populations for age group a by domain d for the various block groups in the i-th area, $\hat{\pi}$ is the column matrix of estimated propensities to engage in the behavior

[&]quot;Mean square error" is used to mean the same thing that is labeled "variance" in sections 1.6 and 2.6. Neither term is wholely satisfactory. "Mean square prediction error" would probably be more appropriate but is rather long.

of interest, and π is the column matrix of propensities given the manifestation of state and PSU effects.

Since \tilde{N} is fixed, we may rewrite (G.3) as

$$MSE_{i}^{a} = \tilde{N}^{t} E \left[(\hat{\pi} - \pi)(\hat{\pi} - \pi)^{t} \right] \tilde{N}$$
 (G.4)

Since $\hat{\pi}$ and π are nonlinear functions of $(\hat{\beta}, \hat{U})$ and (β, U) , respectively, the inner expectation of $\hat{\pi}$ (G.4) is difficult to compute exactly. Following McGibbon and Tomerlin (1989), we used a first-order Taylor linearization to simplify the calculation. Focusing on the predicted propensity for a specific domain and block group, and treating it as a function of $(\hat{\beta}, \hat{U})$, we have that

$$\hat{\pi}_{ib}^{d}(\hat{\beta}, \hat{U}) = \hat{\pi}_{ib}^{d}(\beta, U) + \begin{bmatrix} \frac{\partial \hat{\pi}_{ib}^{d}}{\partial \hat{\beta}}(\beta, U) \\ \frac{\partial \hat{\alpha}}{\partial \hat{U}}(\beta, U) \end{bmatrix}^{t} \begin{bmatrix} \hat{\beta} - \beta \\ \hat{U} - U \end{bmatrix}$$
(G.5)

The first partial derivative of $\hat{\pi}_{ib}^d$ with respect to $\hat{\beta}$ is

$$\partial \frac{\hat{\pi}_{ib}^{d}}{\partial \hat{\beta}} = \frac{\partial}{\partial \hat{\beta}} \left[\frac{1}{1 + \exp[-(X_{ib}^{d} \hat{\beta} + Z_{ib} \hat{U})]} \right]$$

$$= \frac{-\frac{\partial}{\partial \hat{\beta}} \left[1 + \exp[-(X_{ib}^{d} \hat{\beta} + Z_{ib} \hat{U})] \right]}{\left\{ 1 + \exp[-(X_{ib}^{d} \hat{\beta} + Z_{ib} \hat{U})] \right\}^{2}}$$

$$= \left(X_{ib}^{d} \right)^{t} \frac{\exp[-(X_{ib}^{d} \hat{\beta} + Z_{ib} \hat{U})]}{\left\{ 1 + \exp[-(X_{ib}^{d} \hat{\beta} + Z_{ib} \hat{U})] \right\}^{2}} ,$$

where X_{ib}^d and Z_{ib} are defined as in Sections 2.5 and 2.6. Evaluating at (β, U) , we have

$$\frac{\partial \hat{\pi}_{ib}^d}{\partial \hat{\beta}} (\beta, U) = \left(X_{ib}^d \right)^t \frac{\exp[-(X_{ib}^d \beta + Z_{ib} U)]}{\left\{ 1 + \exp[-(X_{ib}^d \beta + Z_{ib} U)] \right\}^2} = \left(X_{ib}^d \right)^t \pi_{ib}^d (1 - \pi_{ib}^d)$$

Similarly,

$$\frac{\partial \hat{\pi}_{ib}^d}{\partial \hat{U}} (\beta, U) = Z_{ib}^t \pi_{ib}^d (1 - \pi_{ib}^d)$$

Substituting into (G.5) and noting that $\hat{\pi}_{ib}^d(\beta, U) = \pi_{ib}^d$, we have that

$$\hat{\pi}_{ib}^{d}(\hat{\beta}, \hat{U}) = \pi_{ib}^{d} + \pi_{ib}^{d}(1 - \pi_{ib}^{d}) \left[X_{ib}^{d} Z_{ib} \right] \begin{bmatrix} \hat{\beta} - \beta \\ \hat{U} - U \end{bmatrix}$$

$$= \pi_{ib}^{d} + \pi_{ib}^{d}(1 - \pi_{ib}^{d}) [X_{ib}^{d}(\hat{\beta} - \beta) + Z_{ib}(\hat{U} - U)] .$$

Thus $\hat{\pi}_{ib}^d - \pi_{ib}^d = \pi_{ib}^d (1 - \pi_{ib}^d) [X_{ib}^d (\hat{\beta} - \beta) + Z_{ib} (\hat{U} - U)]$ (G.6)

Substituting into (G.4), we have

$$MSE_{i}^{a} = \tilde{N}^{t} E \left\{ [\tilde{C}X^{*}(\hat{\beta} - \beta) + \tilde{C}Z^{*}(\hat{U} - U)][\tilde{C}X^{*}(\hat{\beta} - \beta) + \tilde{C}Z^{*}(\hat{U} - U)]^{t} \right\} \tilde{N}$$
 (G.7)

where an asterisk over a C, X, or Z indicates that the matrix has one row for every domain within every block group in the area as opposed to having one row per sample person as was the case for the model fitting.

We now make an additional simplifying assumption, treating C as fixed even through it must be estimated. With that additional assumption, we have that

$$MSE_{i}^{a} \doteq \tilde{N}^{t} \mathring{C} E \left\{ [\mathring{X}(\hat{\beta} - \beta) + \mathring{Z}(\hat{U} - U)] [\mathring{X}(\hat{\beta} - \beta) + \mathring{Z}(\hat{U} - U)]^{t} \right\} \mathring{C} \tilde{N}$$

$$= \tilde{N}^{t} \mathring{C} E [\mathring{X}(\hat{\beta} - \beta)(\hat{\beta} - \beta)^{t}] \mathring{C} \tilde{N} +$$

$$+ 2\tilde{N}^{t} \mathring{C} E [\mathring{X}(\hat{\beta} - \beta)(\hat{U} - U)^{t} \mathring{Z}^{t}] \mathring{C} \tilde{N} +$$

$$+ \tilde{N}^{t} \mathring{C} E [\mathring{Z}(\hat{U} - U)(\hat{U} - U)^{t} \mathring{Z}^{t}] \mathring{C} \tilde{N} .$$

$$(G.8)$$

Since $\overset{*}{X}$ and $\overset{*}{Z}$ are fixed, we may further simplify to

$$MSE_{i}^{a} \doteq \tilde{N}^{t} \mathring{C} \mathring{X} E[(\hat{\beta} - \beta)(\hat{\beta} - \beta)^{t}] \mathring{X}^{t} \mathring{C} \tilde{N}$$

$$+ 2\tilde{N}^{t} \mathring{C} \mathring{X} E[(\hat{\beta} - \beta)(\hat{U} - U)^{t}] \mathring{Z}^{t} \mathring{C} \tilde{N} + \tilde{N}^{t} \mathring{C} \mathring{Z} E[(\hat{U} - U)(\hat{U} - U)^{t}] \mathring{Z}^{t} \mathring{C} \tilde{N}$$

$$(G.9)$$

At this point, it is important to note that we estimated \hat{U}_{ij} =0 for every PSU j which was not in sample. We could have made random draws for these variables from the normal distribution with mean zero and variance $\hat{\sigma}_2^2$, but we saw no reason to introduce extra noise into the system. Random draws might have been an appropriate step if we had been interested in the distribution of propensities across all PSUs, but this was not a primary goal of our work. Rather, our primary goal was to minimize the mean square error of the predictions for the states and large MSAs. For this goal, we think that estimating a zero random effect for all nonsample counties was the best course of action. Our initial MSE estimates ignored the fact that these zero predictions have a nonzero error associated with them. We show both the original formula for estimating mean square error and the revised formula.

Let U_S denote the vector of random effects for the states and the sample PSUs. Let U_S denote the corresponding vector for the nonsample PSUs in states of interest. Let Z_S be a matrix with as many rows as $\overset{*}{Z}$ and as many columns as there are nonsample PSUs in the states of interest, where each row of Z_S is mostly zeroes. A one in the j-th column of the i-th row of Z_S indicates that the i-th domain and block group combination is in the j-th nonsample PSU. Let Z_S denote the columns of $\overset{*}{Z}$ corresponding to the random effects for states and sample PSUs. Then we may write

$$\lambda = X\beta + \begin{bmatrix} Z_S & Z_{\$} \end{bmatrix} \begin{bmatrix} U_S \\ U_{\$} \end{bmatrix}$$
 (G.10)

and

$$\hat{\pi} - \pi = C \left[X(\hat{\beta} - \beta) + Z_S(\hat{U}_S - U_S) - Z_S U_S \right]$$
(G.11)

Since all the true random effects (those for states, sample PSUs and nonsample PSUs) are assumed to be mutually independent and since the estimates of the model parameters are independent of the true random effects for areas not in sample, the mean square error can be rewritten as

$$MSE_{i}^{a} \doteq \tilde{N}^{t} \tilde{C} \overset{*}{X} E[(\hat{\beta} - \beta)(\hat{\beta} - \beta)^{t}] \overset{*}{X}^{t} \overset{*}{C} \tilde{N}$$

$$+ 2\tilde{N}^{t} \overset{*}{C} \overset{*}{X} E[(\hat{\beta} - \beta)(\hat{U}_{S} - U_{S})^{t}] \overset{*}{Z}_{S}^{t} \overset{*}{C} \tilde{N} +$$

$$+ \tilde{N}^{t} \overset{*}{C} \overset{*}{Z}_{S} E[(\hat{U}_{S} - U_{S})(\hat{U}_{S} - U_{S})^{t}] \overset{*}{Z}_{S}^{t} \overset{*}{C} \tilde{N} +$$

$$+ \tilde{N}^{t} \overset{*}{C} Z_{S} E(U_{S} U_{S}^{t}) Z_{S}^{t} \overset{*}{C} \tilde{N}$$

$$(G.12)$$

The fourth term is the term that was omitted from the original MSE calculations.

We now examine each of the terms in (G.12) in turn. Starting with the β -term, we have that

$$\beta = (X^t V^{-1} X)^{-1} (X^t V^{-1} X) \beta$$

and

$$\hat{\beta} = (X^t V^{-1} X)^{-1} X^t V^{-1} \zeta$$

So

$$\hat{\beta} - \beta = (X^{t} V^{-1} X)^{-1} X^{t} V^{-1} (\zeta - X \beta)$$
 (G.13)

We demonstrated in Section C.1, that the covariance matrix for ζ is V. Thus, if we treated V as fixed rather than as a parameter to be estimated, we have that

$$E[(\hat{\beta} - \beta)(\hat{\beta} - \beta)^{t}] = (X^{t}V^{-1}X)^{-1}X^{t}V^{-1} E[(\zeta - X\beta)(\zeta - X\beta)^{t}] V^{-1}X(X^{t}V^{-1}X)^{-1}
= (X^{t}V^{-1}X)^{-1}X^{t}V^{-1} VV^{-1}X(X^{t}V^{-1}X)^{-1}
= (X^{t}V^{-1}X)^{-1}(X^{t}V^{-1}X)(X^{t}V^{-1}X)^{-1}
= (X^{t}V^{-1}X)^{-1}$$
(G.14)

Temporarily skipping the second term of (G.12) and looking instead at the third term of (G.12), we have from Section C.1 that

$$\hat{U} = GZ^{t}P\zeta$$

where we have dropped the S subscript since it is clear that we are only talking about the random effects for states and sample PSUs in this equation. Using Lemma C.1 that PX=0, we can rewrite this as

$$\hat{U} = GZ^{t}P(\zeta - X\beta)$$

By the definition of ζ and ϵ from Section C.1, we have then that

$$\hat{U} = GZ^{t}P(ZU+W^{-1}\varepsilon)$$

Thus,

$$E(\hat{U}-U)(\hat{U}-U)^{t} = E\left\{ [GZ^{t}P(ZU+W^{-1}\epsilon)-U][U^{t}Z^{t}+\epsilon^{t}W^{-1}PZG-U^{t}] \right\}$$

$$= E(GZ^{t}PZUU^{t}Z^{t}PZG) + E(GZ^{t}PW^{-1}\epsilon U^{t}Z^{t}PZG) +$$

$$+ E(GZ^{t}PZU\epsilon^{t}W^{-1}PZG) + E(GZ^{t}PW^{-1}\epsilon\epsilon^{t}W^{-1}PZG) +$$

$$- E(GZ^{t}PZUU^{t}) - E(UU^{t}Z^{t}Z^{t}PZG)$$

$$-E(U\epsilon^{t}W^{-1}PZG) + E(UU^{t})$$

$$(G.15)$$

We now ignore the fact that G and P are both estimated from the data and instead treat them as fixed. We can then bring the expected value operator inside each term to focus just on the variability in U and in ε . Here we note that in Section C.1, we showed that U and ε are uncorrelated, the $Cov(\varepsilon)=C^{-1}$, and that PVP=P.

Using this simplifying assumption and the results from Section C.1, we have that

$$\mathbf{E}(\hat{U}-U)(\hat{U}-U)^{t} \doteq GZ^{t}PZGZ^{t}PZG + 0 + 0 + GZ^{t}PW^{-1}C^{-1}W^{-1}PZG + GZ^{t}PZG - GZ^{t}PZG - 0 + G$$

$$= GZ^{t}P(V-R)PZG + GZ^{t}PRPZG - 2GZ^{T}PZG + G$$

$$= GZ^{t}PVPZG - 2GZ^{t}PZG + G$$

$$= GZ^{t}PZG - 2GZ^{t}PZG + G$$

$$= GZ^{t}PZG - 2GZ^{t}PZG + G$$

$$= GZ^{t}PZG$$

The fourth term is quite simple. By definition, $E(U_\$U_\$^t) = \sigma_2^2 I$, where I is the identity matrix of order equal to the number of nonsample PSUs in the states of interest.

Returning now to the second term of (G.12) and using the representations developed for the first and third terms, we have that

$$E(\hat{\beta}-\beta)(\hat{U}-U)^{t} = E\{[(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}(ZU+W^{-1}\epsilon)][U^{t}Z^{t}+\epsilon^{t}W^{-1})PZG-U^{t}]\}$$

$$= E[(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZUU^{t}Z^{t}PZG] + E[(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZU\epsilon^{t}W^{-1}PZG] +$$

$$+ E[(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}W^{-1}\epsilon U^{t}Z^{t}PZG] + E[(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}W^{-1}\epsilon\epsilon^{t}W^{-1}PZG] +$$

$$- E[(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZUU^{t}] - E[(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}W^{-1}\epsilon U^{t}]$$
(G.17)

Again, we ignore the stochastic nature of V, P and G, and write

$$E(\hat{\beta}-\beta)(\hat{U}-U)^{t} \doteq (X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZGZ^{t}PZG^{t} + 0 + 0 (X^{t}V^{-1}X)^{-1}X^{t}V^{-1}W^{-1}C^{-1}W^{-1}PZG^{t} + \\ - (X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZG^{t} - 0$$

$$= (X^{t}V^{-1}X)^{-1}X^{t}V^{-1}(V-R)PZG^{t} + (X^{t}V^{-1}X)^{-1}X^{t}V^{-1}RPZG^{t} - (X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZG^{t}$$

$$= (X^{t}V^{-1}X)^{-1}X^{t}PZG^{t} - (X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZG^{t}$$

$$= (X^{t}V^{-1}X)^{-1}X^{t}(P-V^{-1})ZG^{t}$$

$$= -(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}X(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZG^{t}$$

$$= -(X^{t}V^{-1}X)^{-1}X^{t}V^{-1}ZG^{t}$$

Substituting the results from (G.14), (G.15), and (G.16) back into (G.12), we have that

$$MSE_{i}^{a} = \tilde{N}^{t} \overset{*}{C} \overset{*}{X} (X^{t}V^{-1}X)^{-1} \overset{*}{X}^{t} \overset{*}{C} \tilde{N} - 2\tilde{N}^{t} \overset{*}{C} \overset{*}{X} (X^{t}V^{-1}X)^{-1} X^{t}V^{-1}ZG\overset{*}{Z}^{t} \overset{*}{C} \tilde{N}$$

$$+ \tilde{N}^{t} \overset{*}{C} \overset{*}{Z} (G - GZ^{t}PZG) \overset{*}{Z}^{t} \overset{*}{C} \tilde{N} + \sigma_{2}^{2} \tilde{N}^{t} \overset{*}{C} Z_{S} Z_{S}^{t} \overset{*}{C} \tilde{N}$$
(G.19)

We now make a final approximation which is to substitute estimated parameters for unknown parameters. This substitution should be accurate asymptotically.

$$MSE_{i}^{a} \doteq \tilde{N}^{t} \overset{\hat{\mathcal{C}}}{C} \overset{*}{X} (X^{t} \hat{V}^{-1} X)^{-1} \overset{*}{X^{t}} \overset{\hat{\mathcal{C}}}{C} \tilde{N} - 2 \tilde{N}^{t} \overset{\hat{\mathcal{C}}}{C} \overset{*}{X} (X^{t} \hat{V}^{-1} X)^{-1} X^{t} \hat{V}^{-1} Z \mathring{G} \overset{*}{Z^{t}} \overset{\hat{\mathcal{C}}}{C} \tilde{N} + \\ + \tilde{N}^{t} \overset{\hat{\mathcal{C}}}{C} \overset{*}{Z} (\hat{G} - \hat{G} Z^{t} \hat{P} \hat{Z} \hat{G}) \overset{*}{Z^{t}} \overset{\hat{\mathcal{C}}}{C} \tilde{N} + \hat{\sigma}_{2}^{2} \tilde{N}^{t} \overset{\hat{\mathcal{C}}}{C} Z_{s} Z_{s}^{t} \overset{\hat{\mathcal{C}}}{C} \tilde{N}$$

$$(G.20)$$

The original report on this project also gave a computationally fast form of this mean square error estimator. The problems with calculating the forms given here are that V is too large to practically invert and that Z is mostly full of zeroes. By using lemma C.13 and by writing out the summations implicit in the matrix equations given here, the calculations become much more practical. The derivation of that form is, however, extremely tedious and doesn't add very much to the intuitive understanding of the mean square error estimation and was thus dropped from this report. If someone, however, is trying to replicate the results, it would be useful to have that original version of the report, keeping in mind that the fourth term was inadvertently omitted in that derivation.

G.2 Asymptotic Theory

As $n\to\infty$, we expect that most of the approximations employed to estimate $E(\hat{\pi}_i^W - \pi_i)^2$ will tend to become more and more accurate. Recall that there were five major approximations:

- (1) Assume that $E\left[(\hat{\pi}_{ib}^{ad} {\pi}_{ib}^{ad})(\hat{\pi}_{ib}^{a'd} {\pi}_{ib}^{a'd})\right] = \text{for } a \neq a'$ (2) Assume that $E\left[(\hat{\pi} \pi)(\hat{\pi} \pi)'\right] = E\left\{[CX(\hat{\beta} \beta) + CZ(\hat{U} U)][CZ(\hat{\beta} \beta) + [CX(\hat{\beta} \beta) + CZ(\hat{U} U)]'\right\}$
- (3) Assume that $Cov(\hat{C}) = 0$
- (4) Assume that $Cov(\hat{G}) = 0$
- (5) Assume that C, G, P and V may be replaced by \hat{C} , \hat{G} , \hat{P} and \hat{V} in the final expression.

All of these assumptions should be asymptotically correct with the exception of the first. The third assumption is perhaps the least troubling since there is a fairly large set of block groups represented in the sample. The fourth assumption, on the other hand, is the most troubling since the numbers of states and PSUs represented in the sample are such smaller.

G.3 Unrepresented Sources of Error

As mentioned in Sections 2.6 and G.2, the unrepresented component of most concern is the component due to estimation of G, the matrix of the between-state and between-PSU residual variances. This is a limitation of the empirical Bayes method that had been of concern since the earliest work using the method (e.g., Harville, 1977). Prasad and Rao (1990) developed an approximate method for taking the uncertainty about G into effect when estimating $\mathbf{E}\left[(\hat{U}-U)(\hat{U}-U)^t\right]$, but their method has only been demonstrated to work well for linear mixed models where G does not depend on π and where a strictly unbiased estimator for G is available. We worked out a rough application of their correction to MSE estimates when there was only one level of random effects. This rough correction had almost no impact on the MSEs for the oversampled city estimates, but did inflate the square root of the MSEs on the smaller states and on the balances of the states with oversampled cites by 2 to 20 percent. Since these inflations appeared to be unstable, we decided not to incorporate them into the official MSE estimates.

Leaving out the correlations between age groups does not have any effect on the MSEs for age group specific estimates, but does have an effect on area totals and on estimates for domains that cut across ages such as estimates by sex or race, not crossed by age. We could have incorporated these correlations with larger and faster computers.

The final source of error that was neglected was that due to the clustering of the sample by segment and by household. Other aspects of the sample design such as clustering at the county/MSA level and differential weighting were fully reflected in the estimated MSEs. We had originally intended on fitting models with additional random effects for segments and households, but the data became too thin. With a rare characteristic, the segment and household level residuals tend to be small and negative but are occasionally large and positive. This causes the empirical Bayesian procedure to become biased and unstable. This problem might be partially rectified by going to a full Bayesian approach, but even with such an approach, estimates of variance components at the segment and household levels are likely to remain unstable for rare characteristics. Another possible solution might be to use a leaner X matrix.

In the course of evaluating the effect of initially omitting the fourth term of the MSE, we noticed another possible source of underestimation of the MSE. We noted that on theoretical grounds, one would expect σ_2^2 to be larger (perhaps considerably larger) than σ_1^2 . This is because one expects the heterogeneity in the logit propensity to be greater among PSUs than among states. We, in fact, often observed the opposite pattern in our estimated components of variance. Furthermore, we often estimated σ_2^2 to be zero. We theorize that our estimated values of σ_2^2 are too small due to overfitting of the fixed part of the model. If we had not allowed so many fixed predictors into the model, our estimates of σ_2^2 would certainly have been larger and that would have increased the importance of the fourth component of the MSE.

G.4 Derivation of Goodness of Fit Statistic

As discussed in Section 3.1.3, our approach was to form a Wald statistic along the lines of $T = (\hat{\pi}^W - \hat{\pi}^D)^t \hat{\Psi}^{-1} (\hat{\pi}^W - \hat{\pi}^D)$

where the squiggly line under a symbol indicates the vector of estimated propensities for the L homogenous groups described in that section and

$$\hat{\Psi} = \mathbf{Cov}(\hat{\pi}^{W} - \hat{\pi}^{D}) = \mathbf{E}(\hat{\pi}^{W} - \hat{\pi}^{D})^{t}(\hat{\pi}^{W} - \hat{\pi}^{D}).$$

Under the null hypothesis that $E(\hat{\pi}^W - \hat{\pi}^D) = 0$ and certain regularity assumptions, T is approximately asymptotically distributed as a chi-square variable with L degrees of freedom. The model is rejected if T is too large relative to what is expected from a chi-square variable. The problem that we faced was how to define the meaning of the expected value in null hypothesis and in the definition of the variance-covariance matrix $\hat{\Psi}$. We would have found it most satisfying to have defined the expectation as over all possible samples, over the distribution of the state and PSU random effects, and over the distribution of person level random outcomes. However, we were not able to derive a method of doing this within the time and budget available. Accordingly, we did something simpler. We conditioned on the state and PSU random effects, on the estimated β vector, on the estimated set of random effects, and on the partition induced by the fitted model. This means that the expectation we used was just over all possible samples and over the conditional distribution of the person level random events given that state and PSU random effects are fixed and equal to our estimated state and PSU random effects and that the true β vector is equal to the estimated β vector. A less conditional definition of the expectation in the

equation for $\hat{\Psi}$ would have almost certainly resulted in larger variances which would, in turn, have led to less significant test results since T would have tended to be smaller. Although it would probably have been possible with more work to derive the variance not conditioning on the random effects or on the estimated parameters, it is doubtful that we could have found the variance without conditioning on the partition induced by the model. Since the partition is random, the distribution of $\hat{\pi}^W - \hat{\pi}^D$ depends on the order statistics of the entire set of personlevel predicted propensities.

Conditioning on fixed and random effects, $E(\hat{\pi}^D|S,\hat{\beta},\hat{U}) = \hat{\pi}^W$ is a constant, so the conditional variance of $\hat{\pi}^W - \hat{\pi}^D$ given the NHSDA sample (denoted by S) and the estimated model parameters is

$$\Psi_{1} = \operatorname{Cov}(\hat{\pi}^{W} - \hat{\pi}^{D} | S, \hat{\beta}, \hat{U}) = \begin{bmatrix} \operatorname{Var} \left(\frac{\sum_{i,j,k \in G_{1}} w_{ijk} y_{ijk}}{\sum_{i,j,k \in G_{1}} w_{ijk}} | S, \hat{\beta}, \hat{U} \right) & 0 \\ 0 & \operatorname{Var} \left(\frac{\sum_{i,j,k \in G_{L}} w_{ijk} y_{ijk}}{\sum_{i,j,k \in G_{L}} w_{ijk}} | S, \hat{\beta}, \hat{U} \right) \end{bmatrix} \\ = \begin{bmatrix} \hat{\pi}_{1}^{W} (1 - \hat{\pi}_{1}^{W}) Deff_{1} & 0 \\ & \ddots & \\ & 0 & \frac{\hat{\pi}_{L}^{W} (1 - \hat{\pi}_{L}^{W}) Deff_{L}}{n_{L}} \end{bmatrix},$$

 G_g is the set of sample cases in the g-th homogenous group, n_g is the sample size for the group, and $Deff_g$ is the design effect for the group due to unequal weighting. Standard sampling theory informs us that this design effect is equal to one plus the relative variation in the sampling weights within the group. That is

$$Deff_g = 1 + \frac{\left(\sum_{i,j,k \in G_g} w_{ijk} - \overline{w}_g\right)^2}{\overline{w}_g^2}.$$

We could have used Ψ_1 to form a Wald statistic $T_1 = (\hat{\mathbb{R}}^{W} - \hat{\mathbb{R}}^{D})^t \Psi_1^{-1} (\hat{\mathbb{R}}^{W} - \hat{\mathbb{R}}^{D})$, but we

were concerned that the variance was too small due to the strongly conditional nature of the definition of Ψ_1 . To remove the conditioning on the particular sample that was selected for the 1991-1993 NHSDA, we used SUDAAN. We instructed SUDAAN to get the variance covariance matrix for the L group means, where the dependent variable was defined as $W^{-1}\varepsilon$, as defined in Section 1 of Appendix C and the weight vector was defined as WCW_{1n} , where 1 is just a

column vector of n ones. Since $\varepsilon = C^{-1}(y-\pi)$, the weighted dependent variable as defined here is equal to $WCWW^{-1}C^{-1}(y-\pi)=W(y-\pi)$, as desired. If the NHSDA were a simple random sample and we instructed SUDAAN accordingly, SUDAAN would then estimate the variance covariance matrix of the L group means to be Ψ_1 . However, we instructed SUDAAN to treat the sample as the stratified two stage design that the NHSDA really is. For first stage strata, we used pairs of noncertainty PSUs and collections of 25 area segments from each certainty PSU, including the six oversampled MSAs. For the first stage clusters, we used noncertainty PSUs and half samples of 12 to 13 area segments from each of the strata defined in the certainty PSUs. We also instructed SUDAAN to treat the PSUs as drawn with unequal probability and with replacement. As a result, the variance covariance matrix estimated by SUDAAN is Ψ_2 , the variance of $\hat{\pi}^W - \hat{\pi}^D$

conditioned on the estimated parameters but not conditioned on the sample or on the person level outcomes.

We could then have calculated the Wald statistic as $T_2 = (\hat{\pi}^W - \hat{\pi}^D)^t \Psi_2^{-1} (\hat{\pi}^W - \hat{\pi}^D)$, as indeed, SUDAAN has an option to support. However, it is fairly well known that when the dimension of the $\hat{\pi}^W - \hat{\pi}^D$ vector is large, then the stability of the variance-covariance matrix becomes more critical for the chi-square approximation to the distribution of T to be reasonably good. Since we knew that in this case the vector did have high dimension (40) and that the stability of our estimated variance covariance matrix was limited by perhaps fewer than 40

degrees of freedom, we did not feel comfortable using this test. Instead, we used the Satterwaithe adjustment to T_1 suggested by Rao and Scott (1981).

Let $\Lambda = \Psi_1^{-1} \Psi_2$. This is often referred to as a design effect matrix. If both Ψ_1 and Ψ_2 were diagonal, then the matrix Λ would have the design effects for the L group means on its main diagonal. From linear algebra, we know that the eigenvalues of a diagonal matrix are equal to the elements on the main diagonal. Rao and Scott used these facts to suggest using the eigenvalues of Λ to adjust T_1 . Let $\overline{\mu}$ be the average eigenvalue of Λ and let V_{μ}^2 be the relative variance of the eigenvalues. Then Rao and Scott's Satterwaithe-adjusted Wald statistic is

$$T_3 = \frac{T_1}{\overline{\mu}(1 + V_{\mu}^2)}$$

with adjusted degrees of freedom $L' = \frac{L}{1 + V_{\mu}^2}$.

Appendix H: Inadequacy of
Traditional Measure of
Design-Based Mean
Square Error

Appendix H. Inadequacy of Traditional Measure of Design-Based Mean Square Error

The first formal approach to evaluation of small area estimates that was proposed in the literature on the subject was to examine the mean-squared difference between the model-based estimates and the design-consistent estimates across the small areas (Gonzalez, 1973). This approach has some appeal for estimators based on fixed-effect models that do not have area-specific effects, but it does not apply to composite estimators or estimators based on mixed effect models. The approach is based on the following error decomposition, where $\mathbb{E}_{\mathbb{D}}$ indicates expectation with respect to the sample design:

$$\begin{split} &\mathbb{E}_{\mathbb{D}} \left[\frac{1}{L} \sum_{i} (\pi_{i}^{F} - \pi_{i}^{D})^{2} \right] = &\mathbb{E}_{\mathbb{D}} \left[\frac{1}{L} \sum_{i} (\pi_{i}^{F} - \mathbf{E}_{D} \pi_{i}^{F} + \mathbf{E}_{D} \pi_{i}^{F} - \mathbf{E}_{D} \pi_{i}^{D} + \mathbf{E}_{D} \pi_{i}^{D} - \pi_{i}^{D})^{2} \right] \\ &= &\mathbb{E}_{\mathbb{D}} \left[\frac{1}{L} \sum_{i} (\pi_{i}^{F} - \mathbf{E}_{D} \pi_{i}^{F})^{2} \right] + \mathbb{E}_{\mathbb{D}} \left[\frac{1}{L} \sum_{i} (\mathbf{E}_{D} \pi_{i}^{F} - \mathbf{E}_{D} \pi_{i}^{D})^{2} \right] + \\ &+ \mathbb{E}_{\mathbb{D}} \left[\frac{1}{L} \sum_{i} (\pi_{i}^{D} - \mathbf{E}_{D} \pi_{i}^{D})^{2} \right] - \mathbb{E}_{\mathbb{D}} \left[\frac{1}{L} \sum_{i} (\pi_{i}^{F} - \mathbf{E}_{D} \pi_{i}^{F}) (\pi_{i}^{D} - \mathbf{E}_{D} \pi_{i}^{D}) \right] \\ &= &\frac{1}{L} \sum_{i} \mathbb{E}_{\mathbb{D}} (\pi_{i}^{F} - \mathbf{E}_{D} \pi_{i}^{F})^{2} + \frac{1}{L} \sum_{i} \mathbb{E}_{\mathbb{D}} (\mathbf{E}_{D} \pi_{i}^{F} - \mathbf{E}_{D} \pi_{i}^{D})^{2} + \\ &+ \frac{1}{L} \sum_{i} \mathbb{E}_{\mathbb{D}} (\pi_{i}^{D} - \mathbf{E}_{D} \pi_{i}^{D})^{2} - \frac{1}{L} \sum_{i} \mathbb{E}_{\mathbb{D}} (\pi_{i}^{F} - \mathbf{E}_{D} \pi_{i}^{F}) (\pi_{i}^{D} - \mathbf{E}_{D} \pi_{i}^{D}) \\ &= &\frac{1}{L} \sum_{i} \mathbb{V} \text{Var}_{\mathbb{D}} (\pi_{i}^{F}) + \frac{1}{L} \sum_{i} \mathbb{B} \text{ias}_{\mathbb{D}} (\pi_{i}^{F}) + \frac{1}{L} \sum_{i} \mathbb{V} \text{Var}_{\mathbb{D}} (\pi_{i}^{D}) - \frac{1}{L} \sum_{i} \mathbb{Cov}_{\mathbb{D}} (\pi_{i}^{F}, \pi_{i}^{D}) \end{split}$$

By definition, the design-based mean square error of π_i^F is

$$MSE_{D}(\pi_{i}^{F}) = Var_{D}(\pi_{i}^{F}) + Bias_{D}(\pi_{i}^{F}).$$

Often when unbiased estimators are impractical, one tries to obtain an estimator with minimum mean square error (MSE). To estimate the MSE separately for each small area is too unstable. By averaging across the small areas, the stability of this evaluation measure is improved. The sum of the first two terms on the right is the average of the design-based mean square error of π_i^F across all the small areas. It is straightforward to estimate the two variance terms using survey methods. If the covariance term can be assumed to be zero, then the design-based variance of π_i^D can be subtracted off the observed mean squared deviation between the two estimators across the areas so as to leave an unbiased estimate of the average mean square

error of π_i^F . As mentioned above, when a fixed effect model is used that does not contain any area-specific effects, it is reasonable to assume that π_i^F and π_i^D are independent and thus that their covariance is zero. However, when random effect models are used and the estimator is π_i^M , the covariance between π_i^M and π_i^D can become quite strong, thereby invalidating this approach. This is most easily seen by recalling that small area estimators based upon mixed effect models are closely related to composite estimators. Since $\pi_i^C = \Gamma_i \pi_i^F + (1 - \Gamma_i) \pi_i^D$, it is obvious that by choosing $\Gamma_i = 0$, we would achieve

$$\mathbb{E}_{D}\left[\frac{1}{L}\sum_{i}\left(\boldsymbol{\pi}_{i}^{C}-\boldsymbol{\pi}_{i}^{D}\right)^{2}\right]=0.$$

If one were to naively assume that the covariance between π_i^C and π_i^D were zero and then used this approach to estimate the mean square error of π_i^C , one would estimate negative mean square error. This approach is thus not appropriate for evaluating small area estimates based on compositing or upon mixed effect models. Accordingly, the approach was not used. The various estimates were compared across the small domains of interest only to see whether the expected shrinkage patterns were realized.

Appendix I: Estimates and Confidence Intervals for Estimates

Appendix I. Estimated Percentage of Population with Substance Abuse Behaviors for 26 States. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals

		Licit Substance	Use In Pas	t Month		Illicit	Substar	nce Use In Past	Month	
State	C	Cigarette Use	1	Alcohol Use	Any Il	licit Drug Use	Any I	llicit But Mrj Use	Cocaine Use	
Total United States	25.46	(25.29-25.64)	49.92	(49.32-50.51)	5.83	(5.69-5.98)	2.44	(2.37-2.52)	0.80	(0.75-0.85)
North East Region										
New Jersey	24.54	(23.72-25.37)	56.36	(54.43-58.26)	5.40	(4.85-6.02)	1.89	(1.64-2.19)	1.01	(0.80-1.26)
New York	23.73	(23.27-24.19)	53.37	(51.71-55.02)	6.44	(6.01-6.89)	2.37	(2.19-2.56)	0.82	(0.74 - 0.91)
Pennsylvania	26.87	(26.26-27.49)	52.32	(50.11-54.53)	5.07	(4.60-5.59)	2.04	(1.81-2.30)	0.59	(0.48-0.72)
South Region										
Florida	24.36	(23.82-24.90)	45.84	(43.89-47.81)	4.95	(4.52-5.41)	1.99	(1.78-2.23)	0.69	(0.58-0.83)
Georgia	26.34	(25.43-27.27)	45.10	(42.39-47.83)	5.82	(5.02-6.73)	2.04	(1.72-2.41)	0.59	(0.42 - 0.83)
Kentucky	31.44	(30.48-32.41)	38.45	(35.44-41.55)	4.53	(3.93-5.21)	1.66	(1.38-2.00)	0.44	(0.32 - 0.61)
Louisiana	25.73	(24.67-26.83)	45.79	(43.19-48.42)	4.38	(3.78-5.08)	2.01	(1.68-2.40)	0.81	(0.59-1.12)
North Carolina	26.60	(25.91-27.31)	43.76	(41.14-46.42)	5.85	(5.27-6.50)	1.79	(1.55-2.07)	0.59	(0.46-0.75)
Oklahoma	26.93	(25.92-27.96)	37.52	(34.32-40.84)	6.96	(5.99-8.06)	4.15	(3.69-4.66)	0.56	(0.42 - 0.76)
South Carolina	28.95	(27.70-30.24)	43.56	(40.04-47.15)	4.99	(4.15-5.99)	1.66	(1.35-2.05)	0.66	(0.44-0.98)
Tennessee	29.51	(28.35-30.70)	38.20	(34.90-41.62)	4.53	(3.80-5.39)	1.91	(1.60-2.28)	0.96	(0.71-1.30)
Texas	26.19	(25.47-26.92)	48.90	(47.06-50.74)	5.57	(5.05-6.15)	2.24	(1.97-2.54)	0.79	(0.63-0.98)
Virginia	25.01	(24.29-25.73)	48.09	(45.74-50.45)	5.55	(4.95-6.23)	2.71	(2.46-2.98)	1.25	(1.00-1.57)
West Virginia	30.63	(29.48-31.79)	36.40	(32.84-40.12)	4.23	(3.61-4.95)	1.41	(1.15-1.73)	0.56	(0.41-0.75)
North Central Region										
Illinois	25.96	(25.40-26.52)	50.64	(48.99-52.28)	4.64	(4.36-4.94)	1.94	(1.78-2.10)	0.69	(0.61-0.78)
Indiana	24.52	(23.71-25.34)	44.50	(41.37-47.68)	4.49	(3.95-5.11)	1.85	(1.56-2.19)	0.50	(0.38-0.67)
Kansas	24.15	(23.07-25.26)	52.56	(49.35-55.76)	5.01	(4.26-5.89)	3.19	(2.79-3.63)	0.77	(0.58-1.02)
Michigan	26.95	(26.12-27.79)	52.23	(49.38-55.07)	5.52	(4.89-6.23)	1.55	(1.29-1.85)	0.61	(0.48-0.79)
Minnesota	22.57	(21.70-23.47)	58.68	(55.77-61.54)	4.62	(4.02-5.31)	1.38	(1.14-1.68)	0.32	(0.23-0.45)
Missouri	25.22	(24.44-26.01)	50.52	(47.85-53.19)	4.98	(4.37-5.67)	2.82	(2.50-3.19)	0.89	(0.71-1.11)
Ohio	29.19	(28.53-29.86)	48.87	(46.49-51.25)	5.37	(4.87-5.92)	2.52	(2.27-2.80)	0.78	(0.64-0.94)
Wisconsin	23.51	(22.50-24.54)	55.07	(51.48-58.61)	4.07	(3.49-4.73)	1.93	(1.64-2.27)	0.84	(0.64-1.11)
West Region										
California	22.84	(22.46-23.22)	52.90	(52.15-53.66)	8.23	(7.86-8.62)	3.91	(3.69-4.14)	0.97	(0.86-1.10)
New Mexico	28.42	(26.73-30.16)	49.82	(46.71-52.92)	7.76	(6.44-9.33)	3.54	(2.90-4.32)	0.85	(0.57-1.27)
Oregon	25.53	(24.61-26.47)	52.29	(48.87-55.69)	7.08	(6.28-7.99)	3.31	(2.89-3.78)	0.38	(0.28-0.51)
Washington	23.73	(22.80-24.67)	54.48	(51.64-57.29)	6.07	(5.32-6.91)	3.20	(2.74-3.74)	0.46	(0.34-0.63)

Appendix I. Estimated Percentage of Population with Substance Abuse Behaviors for 26 States. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals (Continued)

		Past Year I	Depende	ncy			Past Y	ear Treatment				
State	Depen	dent On Illicit Drugs		oendent On Alcohol		ved Treatment icit Drug Use		ved Treatment Alcohol Use	Needed Treatment For Illicit Drug Use		Past	Year Arrest
Total United States	1.24	(1.19-1.29)	3.08	(2.97-3.20)	0.70	(0.66-0.74)	0.69	(0.65-0.74)	2.85	(2.76-2.95)	1.79	(1.72-1.86)
North East Region												
New Jersey	1.26	(1.11-1.43)	2.48	(2.14-2.88)	0.68	(0.56-0.83)	0.64	(0.49 - 0.83)	2.03	(1.77-2.34)	1.35	(1.16-1.56)
New York	1.09	(1.00-1.18)	2.02	(1.82-2.24)	0.64	(0.55-0.75)	0.41	(0.33-0.51)	2.46	(2.28-2.66)	1.32	(1.21-1.43)
Pennsylvania	0.88	(0.78 - 0.99)	2.76	(2.41-3.16)	0.56	(0.48 - 0.65)	0.69	(0.56-0.85)	2.19	(1.95-2.45)	1.28	(1.16-1.42)
South Region												
Florida	0.93	(0.84-1.04)	2.36	(2.07-2.69)	0.69	(0.60-0.80)	0.58	(0.47-0.70)	2.46	(2.21-2.74)	1.61	(1.47-1.75)
Georgia	1.21	(1.06-1.38)	3.61	(3.01-4.32)	0.71	(0.58-0.87)	0.45	(0.33-0.62)	3.78	(3.23-4.42)	2.49	(2.25-2.76)
Kentucky	0.90	(0.79-1.02)	2.06	(1.72-2.45)	0.57	(0.47 - 0.67)	0.48	(0.35-0.65)	2.27	(1.93-2.66)	1.67	(1.48-1.88)
Louisiana	0.99	(0.86-1.14)	3.93	(3.31-4.67)	0.64	(0.53-0.77)	0.59	(0.42-0.82)	2.77	(2.29-3.36)	2.32	(2.06-2.62)
North Carolina	1.14	(1.02-1.27)	2.55	(2.21-2.95)	0.64	(0.53-0.76)	0.50	(0.39 - 0.64)	2.40	(2.11-2.74)	1.68	(1.53-1.85)
Oklahoma	1.50	(1.34-1.69)	4.05	(3.32-4.93)	0.86	(0.71-1.04)	1.49	(0.97-2.26)	3.77	(3.07-4.63)	1.37	(1.19-1.58)
South Carolina	1.18	(1.02-1.36)	3.01	(2.41-3.77)	0.53	(0.42 - 0.66)	0.55	(0.38-0.81)	2.14	(1.72-2.66)	1.80	(1.55-2.08)
Tennessee	0.91	(0.80-1.04)	2.06	(1.69-2.51)	0.60	(0.50 - 0.70)	0.53	(0.38-0.76)	2.08	(1.72-2.52)	2.10	(1.83-2.41)
Texas	1.47	(1.33-1.63)	3.37	(2.95-3.84)	0.61	(0.51-0.73)	0.69	(0.55-0.87)	3.06	(2.73-3.43)	1.84	(1.66-2.05)
Virginia	1.11	(1.01-1.23)	3.03	(2.63-3.48)	0.65	(0.55-0.76)	0.61	(0.50 - 0.75)	2.91	(2.54-3.33)	1.54	(1.39-1.69)
West Virginia	0.84	(0.73-0.97)	2.12	(1.73-2.59)	0.49	(0.39-0.61)	0.73	(0.52-1.02)	2.15	(1.77-2.62)	1.29	(1.10-1.50)
North Central Region												
Illinois	0.88	(0.81 - 0.94)	3.08	(2.83-3.36)	0.54	(0.48-0.59)	0.66	(0.56-0.78)	2.32	(2.15-2.51)	1.72	(1.57-1.90)
Indiana	0.97	(0.87-1.09)	2.15	(1.80-2.55)	0.52	(0.43-0.62)	0.66	(0.49 - 0.89)	1.98	(1.70-2.30)	2.30	(2.08-2.55)
Kansas	1.13	(1.00-1.27)	3.51	(2.88-4.26)	0.66	(0.55-0.79)	0.86	(0.59-1.24)	2.73	(2.25-3.31)	2.40	(2.10-2.74)
Michigan	1.08	(0.96-1.22)	3.36	(2.82-3.99)	0.76	(0.65-0.89)	0.70	(0.52 - 0.93)	3.05	(2.63-3.53)	1.99	(1.79-2.21)
Minnesota	1.00	(0.87-1.14)	3.12	(2.58-3.77)	0.84	(0.70-1.02)	0.80	(0.59-1.10)	2.19	(1.84-2.60)	1.67	(1.45-1.91)
Missouri	1.03	(0.92-1.15)	2.16	(1.83-2.55)	0.70	(0.59 - 0.84)	0.68	(0.50-0.91)	3.26	(2.82-3.76)	2.10	(1.85-2.38)
Ohio	0.96	(0.87-1.07)	2.73	(2.36-3.15)	0.70	(0.59 - 0.83)	0.81	(0.64-1.01)	2.56	(2.28-2.87)	2.25	(2.06-2.47)
Wisconsin	0.98	(0.88-1.11)	2.95	(2.41-3.60)	0.61	(0.51-0.73)	0.59	(0.44-0.79)	2.41	(1.99-2.92)	1.53	(1.30-1.80)
West Region												
California	1.91	(1.78-2.04)	4.87	(4.62-5.14)	0.97	(0.87-1.08)	0.81	(0.71-0.94)	4.23	(3.99-4.48)	2.15	(1.99-2.33)
New Mexico	1.62	(1.41-1.87)	3.66	(2.93-4.57)	0.77	(0.61-0.96)	0.92	(0.55-1.56)	3.30	(2.61-4.17)	2.63	(2.07-3.32)
Oregon	1.99	(1.79-2.22)	2.75	(2.32-3.25)	0.90	(0.75-1.08)	0.61	(0.43-0.85)	2.22	(1.90-2.58)	1.88	(1.64-2.15)
Washington	1.96	(1.74-2.19)	3.51	(2.94-4.19)	0.85	(0.72-1.00)	0.51	(0.36-0.71)	3.55	(3.01-4.18)	1.91	(1.61-2.26)

Appendix I. Estimated Percentage of Population with Substance Abuse Behaviors for 25 MSAs. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals

]	Licit Substance U	Jse In Pas	st Month		Illicit S	ubstanc	e Use In Past M	onth	
MSA	C	igarette Use	A	lcohol Use	Any I	llicit Drug Use	Any	Illicit But Mrj Use	Cocaine Use	
Anaheim-Santa Ana, CA	21.01	(20.19-21.84)	52.25	(50.58-53.91)	8.82	(7.78-9.98)	4.85	(4.25-5.53)	1.23	(0.93-1.61)
Atlanta, GA	23.70	(22.80-24.63)	50.98	(49.52-52.45)	5.86	(5.15-6.66)	1.95	(1.62-2.36)	0.56	(0.40 - 0.79)
Baltimore, MD	24.96	(24.00-25.95)	44.84	(43.48-46.20)	5.03	(4.37-5.80)	1.80	(1.48-2.19)	0.70	(0.53-0.93)
Boston, MA	27.51	(26.70-28.33)	61.51	(60.37-62.63)	6.73	(5.98-7.55)	2.57	(2.17-3.03)	0.94	(0.74-1.20)
Chicago, IL	24.73	(24.42-25.04)	53.82	(53.49-54.16)	5.53	(5.34-5.72)	2.34	(2.23-2.45)	0.94	(0.87-1.02)
Dallas, TX	25.95	(25.06-26.85)	50.23	(48.39-52.08)	5.69	(4.95-6.53)	2.38	(2.04-2.78)	1.05	(0.80-1.37)
Denver, CO	27.13	(26.83-27.44)	58.42	(58.09-58.75)	8.30	(8.07 - 8.54)	2.94	(2.80-3.07)	1.01	(0.93-1.08)
Detroit, MI	26.52	(25.61-27.46)	54.17	(52.79-55.54)	5.50	(4.83-6.25)	1.38	(1.12-1.71)	0.60	(0.45 - 0.81)
El Paso, TX	20.41	(18.86-22.05)	45.85	(42.18-49.56)	3.58	(2.62-4.89)	1.85	(1.35-2.51)	0.53	(0.33-0.85)
Houston, TX	25.07	(24.03-26.13)	50.94	(49.37-52.50)	4.05	(3.47-4.71)	2.43	(2.07-2.86)	0.60	(0.44-0.82)
Los Angeles, CA	21.82	(21.54-22.11)	49.34	(49.01-49.68)	6.68	(6.48-6.89)	2.72	(2.60-2.84)	0.90	(0.84-0.97)
Miami-Hialeah, FL	20.88	(20.60-21.17)	44.40	(44.05-44.75)	3.75	(3.59-3.92)	2.26	(2.15-2.37)	0.89	(0.82 - 0.96)
Minneapolis-St. Paul, MN	22.64	(21.73-23.57)	65.87	(64.62-67.10)	5.19	(4.54-5.92)	1.51	(1.22-1.87)	0.36	(0.26-0.50)
Nassau-Suffolk, NY	20.22	(19.38-21.08)	59.60	(58.36-60.84)	6.57	(5.84-7.38)	1.99	(1.65-2.41)	0.37	(0.28 - 0.49)
New York, NY	23.15	(22.85-23.45)	48.84	(48.50-49.19)	5.99	(5.80-6.19)	2.40	(2.29-2.51)	1.06	(0.99-1.15)
Newark, NJ	24.10	(23.10-25.13)	62.17	(60.45-63.85)	6.20	(5.39-7.11)	1.91	(1.60-2.28)	1.37	(0.98-1.92)
Oakland, CA	23.33	(22.61-24.07)	65.02	(63.55-66.46)	11.39	(10.31-12.57)	5.11	(4.59-5.69)	1.47	(1.17-1.85)
Philadelphia, PA-NJ	25.89	(25.20-26.59)	59.13	(58.06-60.19)	5.72	(5.11-6.40)	2.22	(1.94-2.55)	0.64	(0.51-0.81)
Phoenix, AZ	24.83	(23.85-25.84)	53.69	(52.12-55.25)	6.82	(5.95-7.81)	3.99	(3.45-4.62)	1.01	(0.77-1.31)
San Antonio, TX	26.09	(24.71-27.52)	54.43	(51.55-57.27)	4.35	(3.39-5.55)	2.17	(1.67-2.81)	0.78	(0.51-1.17)
San Bernardino, CA	23.32	(22.39-24.27)	49.12	(47.64-50.60)	7.42	(6.62-8.32)	4.60	(4.07-5.20)	0.64	(0.49 - 0.84)
San Diego, CA	22.25	(21.36-23.16)	49.08	(47.60-50.55)	7.07	(6.22-8.03)	3.59	(3.09-4.17)	0.93	(0.70-1.23)
St. Louis, MO-IL	25.27	(24.52-26.03)	55.02	(53.73-56.30)	5.19	(4.56-5.90)	2.53	(2.21-2.88)	0.95	(0.75-1.21)
Tampa-St. Petersburg, FL	25.55	(24.76-26.36)	41.22	(39.75-42.70)	5.27	(4.56-6.08)	2.03	(1.66-2.47)	0.57	(0.43-0.74)
Washington, DC	22.61	(22.40-22.81)	53.82	(53.59-54.05)	5.38	(5.26-5.51)	2.58	(2.50-2.67)	1.03	(0.98-1.08)

Appendix I. Estimated Percentage of Population with Substance Abuse Behaviors for 25 MSAs. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals (Continued)

		Past Year I	Depende	ency			Past Yo	ear Treatment				
MSA	Depen	dent On Illicit Drugs		pendent On Alcohol		Received Treatment For Illicit Drug Use		Received atment For cohol Use		ed Treatment licit Drug Use	Past	Year Arrest
Anaheim-Santa Ana, CA	2.17	(1.88-2.51)	6.01	(5.37-6.72)	1.01	(0.77-1.31)	0.55	(0.39-0.78)	4.45	(3.91-5.06)	1.53	(1.20-1.94)
Atlanta, GA	1.24	(1.06-1.45)	4.47	(3.84-5.20)	0.74	(0.60-0.91)	0.38	(0.27-0.55)	4.83	(4.21-5.53)	1.74	(1.52-1.98)
Baltimore, MD	0.89	(0.75-1.05)	2.15	(1.74-2.65)	0.58	(0.45-0.75)	0.45	(0.32 - 0.65)	2.28	(1.90-2.73)	1.30	(1.11-1.52)
Boston, MA	1.68	(1.46-1.92)	4.38	(3.83-5.01)	0.97	(0.77-1.22)	2.18	(1.69-2.81)	3.88	(3.41-4.41)	1.66	(1.39-1.98)
Chicago, IL	0.89	(0.82 - 0.97)	3.46	(3.33-3.60)	0.53	(0.49 - 0.59)	0.71	(0.66-0.77)	2.75	(2.63-2.88)	1.51	(1.41-1.62)
Dallas, TX	1.63	(1.45-1.84)	3.31	(2.80-3.91)	0.65	(0.53-0.80)	0.45	(0.31-0.66)	2.47	(2.10-2.90)	1.55	(1.31-1.82)
Denver, CO	1.26	(1.16-1.37)	4.57	(4.41-4.73)	0.63	(0.57-0.70)	0.96	(0.89-1.04)	3.83	(3.68-3.99)	2.26	(2.12-2.40)
Detroit, MI	1.18	(1.02-1.37)	3.59	(3.11-4.14)	0.78	(0.66-0.92)	0.51	(0.38-0.69)	3.58	(3.15-4.06)	1.77	(1.53-2.04)
El Paso, TX	0.79	(0.57-1.08)	3.66	(2.59-5.13)	0.55	(0.36-0.85)	0.53	(0.28-1.03)	2.50	(1.92-3.25)	1.62	(1.20-2.19)
Houston, TX	2.10	(1.87-2.37)	1.96	(1.53-2.50)	0.81	(0.68-0.98)	0.71	(0.53-0.94)	3.86	(3.40-4.37)	1.56	(1.32-1.84)
Los Angeles, CA	1.50	(1.39-1.61)	7.13	(6.96-7.31)	0.75	(0.65-0.88)	0.50	(0.46-0.55)	4.04	(3.89-4.20)	1.85	(1.74-1.97)
Miami-Hialeah, FL	0.74	(0.67-0.82)	2.04	(1.94-2.14)	0.56	(0.44-0.72)	0.48	(0.44-0.52)	2.16	(2.04-2.29)	1.22	(1.13-1.32)
Minneapolis-St. Paul, MN	1.15	(0.99-1.34)	3.45	(2.93-4.06)	0.93	(0.75-1.16)	0.93	(0.65-1.34)	2.38	(2.04-2.78)	1.77	(1.50-2.08)
Nassau-Suffolk, NY	1.00	(0.85-1.17)	1.10	(0.83-1.44)	0.64	(0.49 - 0.83)	0.50	(0.36 - 0.68)	1.87	(1.56-2.24)	1.14	(0.93-1.40)
New York, NY	0.94	(0.85-1.02)	1.77	(1.68-1.87)	0.66	(0.52 - 0.84)	0.26	(0.23-0.29)	2.50	(2.38-2.63)	0.79	(0.72 - 0.87)
Newark, NJ	1.43	(1.24-1.65)	2.50	(2.00-3.12)	0.89	(0.74-1.07)	0.77	(0.59-1.01)	1.90	(1.56-2.31)	1.40	(1.10-1.77)
Oakland, CA	2.98	(2.68-3.31)	3.88	(3.24-4.65)	1.37	(1.19-1.57)	2.07	(1.69-2.53)	5.22	(4.63-5.87)	1.94	(1.64-2.28)
Philadelphia, PA-NJ	0.98	(0.85-1.13)	2.82	(2.45-3.25)	0.82	(0.71-0.95)	0.61	(0.47 - 0.78)	2.99	(2.67-3.33)	1.37	(1.20-1.57)
Phoenix, AZ	1.55	(1.37-1.75)	3.61	(3.04-4.28)	0.65	(0.53-0.80)	0.67	(0.46-0.98)	3.14	(2.68-3.68)	1.48	(1.12-1.96)
San Antonio, TX	1.38	(1.19-1.61)	2.53	(1.69-3.76)	0.53	(0.37-0.74)	0.42	(0.22 - 0.79)	2.90	(2.20-3.81)	1.60	(1.30-1.96)
San Bernardino, CA	2.09	(1.82-2.39)	3.92	(3.42-4.50)	1.15	(0.96-1.37)	1.19	(0.89-1.59)	3.84	(3.35-4.39)	2.41	(2.01-2.89)
San Diego, CA	1.38	(1.15-1.65)	2.99	(2.51-3.56)	0.74	(0.56-0.98)	0.52	(0.34-0.79)	3.19	(2.75-3.71)	2.18	(1.82-2.60)
St. Louis, MO-IL	0.96	(0.85-1.09)	2.46	(2.09-2.90)	0.74	(0.62 - 0.87)	0.62	(0.45-0.85)	2.75	(2.37-3.19)	2.09	(1.77-2.45)
Tampa-St. Petersburg, FL	0.98	(0.84-1.13)	2.10	(1.67-2.63)	0.66	(0.55-0.80)	0.41	(0.29 - 0.60)	2.08	(1.76-2.45)	1.44	(1.22-1.70)
Washington, DC	1.07	(1.02-1.13)	3.59	(3.49-3.69)	0.56	(0.51-0.60)	0.87	(0.83-0.92)	2.66	(2.58-2.75)	1.33	(1.27-1.40)

Appendix I. Estimated Number of Persons with Substance Abuse Behaviors for 25 MSAs. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals

Appendix I. Estimated Number of Persons with Substance Abuse Behaviors for 26 States. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals

		Licit Substance	Use In Past I	Month	Illicit Substance Use In Past Month							
State	Ci	garette Use	A	lcohol Use	Any I	llicit Drug Use	Any Illi	icit But Mrj Use	С	ocaine Use		
Total United States	52,444	(52,084-52,805)	102,802	(101,582-104,022)	12,015	(11,712-12,326)	5,031	(4,878-5,189)	1,647	(1,552-1,747)		
North East Region	l											
New Jersey	1,581	(1,529-1,635)	3,631	(3,507-3,754)	348	(312-388)	122	(106-141)	65	(51-81)		
New York	3,534	(3,466-3,602)	7,948	(7,701-8,193)	959	(896-1,026)	352	(326-381)	122	(110-136)		
Pennsylvania	2,672	(2,612-2,733)	5,203	(4,983-5,423)	505	(458-556)	203	(180-229)	59	(48-72)		
South Region	i											
Florida	2,744	(2,683-2,805)	5,164	(4,944-5,386)	557	(509-610)	225	(201-252)	78	(65-93)		
Georgia	1,433	(1,384-1,484)	2,454	(2,307-2,603)	317	(273-366)	111	(94-131)	32	(23-45)		
Kentucky	960	(930-989)	1,174	(1,082-1,268)	138	(120-159)	51	(42-61)	14	(10-19)		
Louisiana	872	(835-909)	1,551	(1,463-1,640)	148	(128-172)	68	(57-81)	27	(20-38)		
North Carolina	1,501	(1,461-1,541)	2,468	(2,321-2,618)	330	(297-366)	101	(87-117)	33	(26-42)		
Oklahoma	690	(664-717)	962	(880-1,047)	178	(154-207)	106	(95-119)	14	(11-19)		
South Carolina	851	(814-889)	1,280	(1,177-1,386)	147	(122-176)	49	(40-60)	19	(13-29)		
Tennessee	1,215	(1,167-1,264)	1,573	(1,437-1,713)	187	(157-222)	79	(66-94)	39	(29-53)		
Texas	3,601	(3,502-3,702)	6,724	(6,472-6,977)	767	(695-845)	308	(271-349)	108	(87-135)		
Virginia	1,307	(1,270-1,345)	2,514	(2,391-2,637)	290	(259-326)	141	(128-156)	65	(52-82)		
West Virginia	458	(441-476)	545	(492-601)	63	(54-74)	21	(17-26)	8	(6-11)		
North Central	1											
Region	•											
Illinois	2,434	(2,382-2,488)	4,749	(4,594-4,903)	435	(409-463)	182	(167-197)	65	(57-73)		
Indiana	1,123	(1,086-1,161)	2,039	(1,895-2,184)	206	(181-234)	85	(72-100)	23	(17-31)		
Kansas	485	(464-508)	1,057	(992-1,121)	101	(86-118)	64	(56-73)	15	(12-20)		
Michigan	2,052	(1,989-2,116)	3,978	(3,760-4,194)	420	(372-474)	118	(98-141)	47	(36-61)		
Minnesota	807	(776-839)	2,098	(1,994-2,200)	165	(372 + 74) $(144-190)$	49	(41-60)	12	(8-16)		
Missouri	1,065	(1,032-1,098)	2,134	(2,021-2,246)	210	(185-239)	119	(106-135)	38	(30-47)		
Ohio	2,612	(2,552-2,672)	4,372	(4,159-4,585)	481	(436-530)	226	(203-251)	70	(58-84)		
Wisconsin	945	(905-987)	2,215	(2,070-2,357)	164	(141-190)	78	(66-91)	34	(26-45)		
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West Region		(5.466.5.650)	40.000	(10.600.10.051)	• • • •	(1.010.0.000)	0.54	(007.1.007)		(210.260)		
California	5,559	(5,466-5,652)	12,877	(12,693-13,061)	2,004	(1,913-2,099)	951	(897-1,007)	237	(210-268)		
New Mexico	341	(321-362)	597	(560-635)	93	(77-112)	43	(35-52)	10	(7-15)		
Oregon	612	(590-634)	1,253	(1,171-1,335)	170	(150-191)	79	(69-91)	9	(7-12)		
Washington	971	(934-1,010)	2,230	(2,114-2,345)	248	(218-283)	131	(112-153)	19	(14-26)		

Appendix I. Estimated Number of Persons with Substance Abuse Behaviors for 25 MSAs. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals

MCA		Licit Substance U	Use In Past M	onth		Illicit S	Substance	Use In Past Mo	nth	
MSA	Cig	arette Use	Alc	cohol Use	Any Illic	eit Drug Use	Any Illic	it But Mrj Use	Cocaine Use	
Anaheim-Santa Ana,										
CA	419	(403-436)	1,043	(1,010-1,076)	176	(155-199)	97	(85-110)	25	(19-32)
Atlanta, GA	575	(553-597)	1,236	(1,201-1,272)	142	(125-161)	47	(39-57)	14	(10-19)
Baltimore, MD	498	(479-518)	895	(868-922)	100	(87-116)	36	(29-44)	14	(11-18)
Boston, MA	865	(840-891)	1,934	(1,899-1,970)	212	(188-238)	81	(68-95)	30	(23-38)
Chicago, IL	1,232	(1,217-1,247)	2,681	(2,664-2,698)	275	(266-285)	117	(111-122)	47	(44-51)
Dallas, TX	550	(531-569)	1,065	(1,026-1,104)	121	(105-139)	51	(43-59)	22	(17-29)
Denver, CO	365	(361-369)	786	(782-791)	112	(109-115)	40	(38-41)	14	(13-15)
Detroit, MI	953	(920-987)	1,946	(1,897-1,995)	197	(173-225)	50	(40-61)	22	(16-29)
El Paso, TX	93	(86-101)	209	(192-226)	16	(12-22)	8	(6-11)	2	(2-4)
Houston, TX	667	(640-695)	1,355	(1,314-1,397)	108	(92-125)	65	(55-76)	16	(12-22)
Los Angeles, CA	1,555	(1,535-1,576)	3,517	(3,493-3,541)	476	(462-491)	194	(185-203)	65	(60-69)
Miami-Hialeah, FL	334	(330-339)	710	(705-716)	60	(57-63)	36	(34-38)	14	(13-15)
Minneapolis-St. Paul,										
MN	461	(442-480)	1,341	(1,315-1,366)	106	(92-120)	31	(25-38)	7	(5-10)
Nassau-Suffolk, NY	440	(422-459)	1,298	(1,271-1,325)	143	(127-161)	43	(36-52)	8	(6-11)
New York, NY	1,640	(1,619-1,662)	3,461	(3,437-3,485)	425	(411-439)	170	(163-178)	75	(70-81)
Newark, NJ	361	(346-377)	932	(907-958)	93	(81-107)	29	(24-34)	21	(15-29)
Oakland, CA	403	(391-416)	1,123	(1,098-1,148)	197	(178-217)	88	(79-98)	25	(20-32)
Philadelphia, PA-NJ	1,045	(1,017-1,073)	2,387	(2,344-2,430)	231	(206-258)	90	(78-103)	26	(21-33)
Phoenix, AZ	440	(422-457)	951	(923-978)	121	(105-138)	71	(61-82)	18	(14-23)
San Antonio, TX	271	(257-286)	566	(536-596)	45	(35-58)	23	(17-29)	8	(5-12)
San Bernardino, CA	495	(475-515)	1,042	(1,011-1,074)	158	(140-177)	98	(86-110)	14	(10-18)
San Diego, CA	465	(446-484)	1,025	(995-1,056)	148	(130-168)	75	(65-87)	19	(15-26)
St. Louis, MO-IL	507	(492-522)	1,104	(1,078-1,130)	104	(91-118)	51	(44-58)	19	(15-24)
Tampa-St.		` '	<i>*</i>	, , , , , ,		, ,		` ′		, ,
Petersburg, FL	466	(451-480)	751	(724-778)	96	(83-111)	37	(30-45)	10	(8-13)
Washington, DC	756	(749-763)	1,800	(1,793-1,808)	180	(176-184)	86	(84-89)	34	(33-36)

Appendix I. Estimated Number of Persons with Substance Abuse Behaviors for 25 MSAs. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals

Appendix I. Estimated Number of Persons with Substance Abuse Behaviors for 26 States. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals

		Past Year D	ependen	cy			Past Yea	ar Treatment					
State		ent On Illicit Orugs			Received Treatment For Illicit Drug Use Received Treatment For Alcohol Use				Needed Treatment For Illicit Drug Use		Past Y	Past Year Arrest	
Total United States	2,549	(2,444-2,658)	6,350	(6,126-6,582)	1,447	(1,365-1,533)	1,429	(1,336-1,529	5,877	(5,690-6,071)	3,689	(3,549-3,835)	
North East Region													
New Jersey	81	(72-92)	160	(138-185)	44	(36-53)	41	(32-53)	131	(114-151)	87	(75-100)	
New York	162	(149-175)	301	(270-334)	95	(82-111)	62	(50-76)	367	(340-395)	196	(180-213)	
Pennsylvania	87	(78-98)	275	(240-315)	55	(47-65)	68	(55-85)	217	(194-243)	128	(115-142)	
South Region													
Florida	105	(94-117)	266	(233-303)	78	(67-90)	65	(53-79)	277	(248-309)	181	(166-197)	
Georgia	66	(58-75)	197	(164-235)	39	(32-47)	25	(18-34)	206	(176-240)	136	(123-150)	
Kentucky	28	(24-31)	63	(53-75)		(14-21)	15	(11-20)	69	(59-81)	51	(45-57)	
Louisiana	34	(29-39)	133	(112-158)	22	(18-26)	20	(14-28)	94	(77-114)	79	(70-89)	
North Carolina	64	(58-72)	144	(125-166)	36	(30-43)	28	(22-36)	136	(119-154)	95	(86-105)	
Oklahoma	39	(34-43)	104	(85-126)	22	(18-27)	38	(25-58)	97	(79-119)	35	(30-41)	
South Carolina	35	(30-40)	89	(71-111)	15	(12-19)	16	(11-24)	63	(50-78)	53	(46-61)	
Tennessee	38	(33-43)	85	(70-103)	25	(21-29)	22	(15-31)	86	(71-104)	87	(75-99)	
Texas	202	(182-225)	463	(405-528)	84	(71-100)	95	(75-119)	421	(376-472)	254	(228-282)	
Virginia	58	(53-64)	158	(138-182)	34	(29-40)	32	(26-39)	152	(133-174)	80	(73-89)	
West Virginia	13	(11-15)	32	(26-39)	7	(6-9)	11	(8-15)	32	(27-39)	19	(17-22)	
North Central Region													
Illinois	82	(76-89)	289	(265-315)	50	(45-55)	62	(52-73)	218	(201-235)	162	(147-178)	
Indiana	45	(40-50)	98	(83-117)	24	(20-28)	30	(23-41)	91	(78-105)	105	(95-117)	
Kansas	23	(20-25)	70	(58-86)	13	(11-16)	17	(12-25)	55	(45-67)	48	(42-55)	
Michigan	82	(73-93)	256	(215-304)	58	(50-68)	53	(40-71)	232	(200-269)	151	(136-168)	
Minnesota	36	(31-41)	112	(92-135)	30	(25-37)	29	(21-39)	78	(66-93)	60	(52-68)	
Missouri	43	(39-48)		(77-108)	30	(25-35)	29	(21-39)	138	(119-159)	89	(78-100)	
Ohio	86	(78-95)	244	(211-281)	63	(53-74)	72	(57-91)	229	(204-256)	202	(184-221)	
Wisconsin	40	(35-44)	118	(97-145)	25	(21-29)	24	(18-32)	97	(80-117)	62	(52-72)	
West Region													
California	464	(433-498)	1.186	(1.124-1.251)	236	(213-262)	198	(172-228)	1.029	(971-1.090)	524	(483-568)	
New Mexico	19	(433-498)	1,180	(35-55)	230	(7-12)	198	(7-19)	40	(31-50)	32	(25-40)	
	48	(43-53)		(56-78)		(18-26)	15	(10-20)	-	(46-62)	32 45	(39-52)	
Oregon	1 48	(43-33)	00	(30-78)	22	(16-20)	15	(10-20)	53	(40-02)	43	(39-32)	

Appendix I. Estimated Number of Persons with Substance Abuse Behaviors for 25 MSAs. Estimates Based on Data From the 1991-1993 NHSDA. 95% Confidence Intervals

Washington	80	(71-90)	144	(120-172)	35	(29-41)	21	(15-29)	145	(123-171)	78	(66-93)
	1	Past Year D	ependency	7		Past Year Treatment						
MSAs	Dependent Dru		Dependen	t On Alcohol		Treatment For Drug Use		d Treatment cohol Use		Treatment For t Drug Use	Past	Year Arrest
Anaheim-Santa Ana, CA	43	(38-50)	120	(107-134)	20	(15-26)	11	(8-16)	89	(78-101)	31	(24-39)
Atlanta, GA	30	(26-35)	108	(93-126)	18	(14-22)	9	(7-13)	117	(102-134)	42	(37-48)
Baltimore, MD	18	(15-21)	43	(35-53)	12	(9-15)	9	(6-13)	45	(38-54)	26	(22-30)
Boston, MA	53	(46-60)	138	(120-157)	30	(24-38)	69	(53-88)	122	(107-139)	52	(44-62)
Chicago, IL	45	(41-49)	172	(166-179)	27	(24-29)	36	(33-38)	137	(131-144)	75	(70-81)
Dallas, TX	35	(31-39)	70	(59-83)	14	(11-17)	10	(7-14)	52	(44-61)	33	(28-39)
Denver, CO	17	(16-18)	61	(59-64)	9	(8-9)	13	(12-14)	52	(50-54)	30	(29-32)
Detroit, MI	42	(37-49)	129	(112-149)	28	(24-33)	18	(14-25)	129	(113-146)	63	(55-73)
El Paso, TX	4	(3-5)	17	(12-23)	3	(2-4)	2	(1-5)	11	(9-15)	7	(5-10)
Houston, TX	56	(50-63)	52	(41-66)	22	(18-26)	19	(14-25)	103	(90-116)	41	(35-49)
Los Angeles, CA	107	(99-115)	508	(496-521)	54	(46-63)	36	(33-39)	288	(277-300)	132	(124-141)
Miami-Hialeah, FL	12	(11-13)	33	(31-34)	9	(7-11)	8	(7-8)	35	(33-37)	20	(18-21)
Minneapolis-St. Paul, MN	23	(20-27)	70	(60-83)	19	(15-24)	19	(13-27)	49	(42-57)	36	(31-42)
Nassau-Suffolk, NY	22	(19-25)	24	(18-31)	14	(11-18)	11	(8-15)	41	(34-49)	25	(20-30)
New York, NY	66	(61-73)	125	(119-132)	47	(37-59)	18	(16-21)	177	(168-186)	56	(51-62)
Newark, NJ	21	(19-25)	38	(30-47)	13	(11-16)	12	(9-15)	28	(23-35)	21	(17-27)
Oakland, CA	51	(46-57)	67	(56-80)	24	(21-27)	36	(29-44)	90	(80-101)	33	(28-39)
Philadelphia, PA-NJ	40	(34-46)	114	(99-131)	33	(29-38)	25	(19-32)	121	(108-135)	55	(48-63)
Phoenix, AZ	27	(24-31)	64	(54-76)	12	(9-14)	12	(8-17)	56	(48-65)	26	(20-35)
San Antonio, TX	14	(12-17)	26	(18-39)	5	(4-8)	4	(2-8)	30	(23-40)	17	(14-20)
San Bernardino, CA	44	(39-51)	83	(72-95)	24	(20-29)	25	(19-34)	81	(71-93)	51	(43-61)
San Diego, CA	29	(24-35)	62	(52-74)	15	(12-20)	11	(7-16)	67	(57-78)	45	(38-54)
St. Louis, MO-IL	19	(17-22)	49	(42-58)	15	(12-18)	12	(9-17)	55	(47-64)	42	(36-49)
Tampa-St. Petersburg, FL	18	(15-21)	38	(30-48)	12	(10-15)	8	(5-11)	38	(32-45)	26	(22-31)
Washington, DC	36	(34-38)	120	(117-124)	19	(17-20)	29	(28-31)	89	(86-92)	45	(42-47)