

Integrative Multi-Scale Biomedical Informatics

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INTEGRATIVE BIOMEDICAL INFORMATICS ANALYSIS

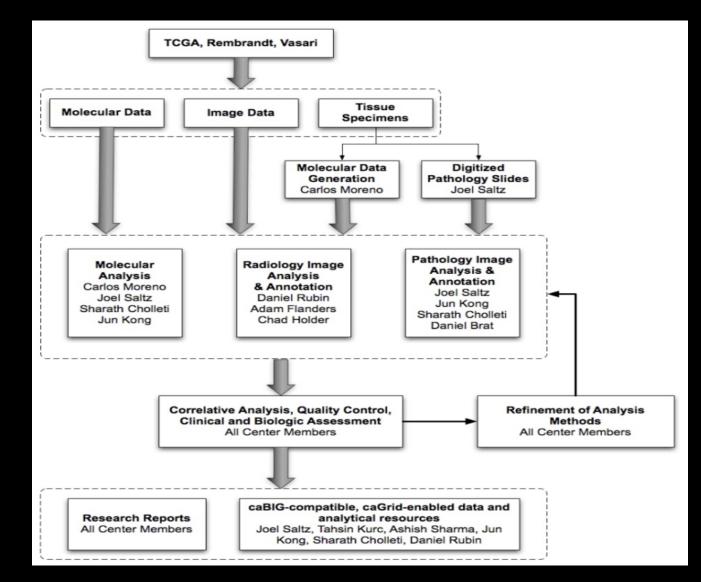
Reproducible anatomic/functional characterization at gross level (Radiology) and fine level (Pathology)

Integration of anatomic/functional characterization with multiple types of "omic" information

Create categories of jointly classified data to describe pathophysiology, predict prognosis, response to treatment

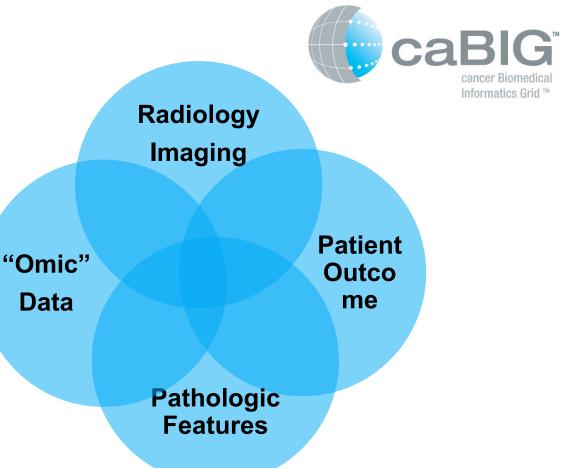


caBIG In Silico Center for Brain Tumor Research



Integration of heterogeneous multiscale information

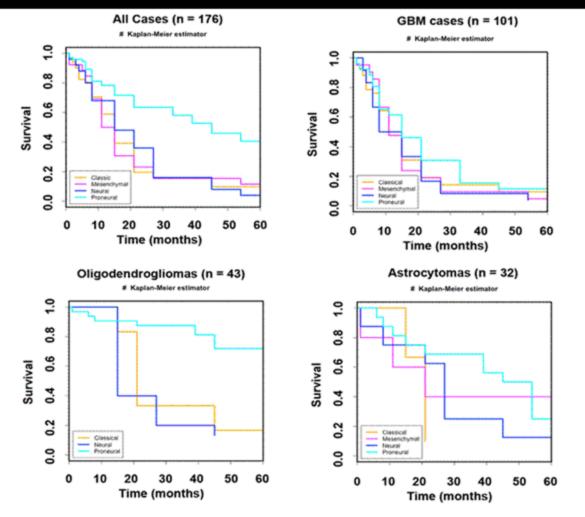
- •Coordinated initiatives Pathology, Radiology, "omics"
- •Exploit synergies between all initiatives to improve ability to forecast survival & response.





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Example: Pathology and Gene Expression Joint Predictors of Recurrence/Survival



Lee Cooper Carlos Moreno

In Silico Center for Brain Tumor Research

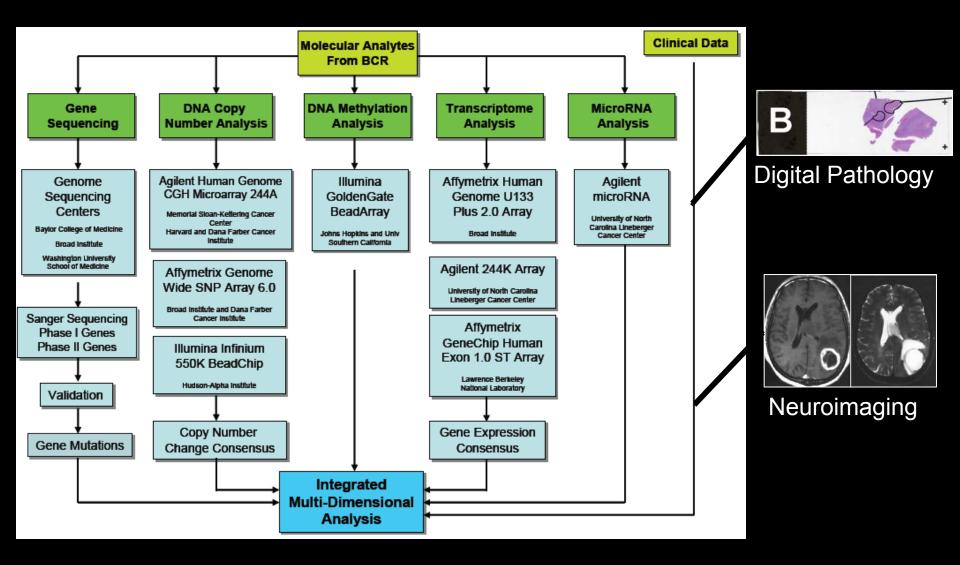
Key Data Sets

REMBRANDT: Gene expression and genomics data set of all glioma subtypes

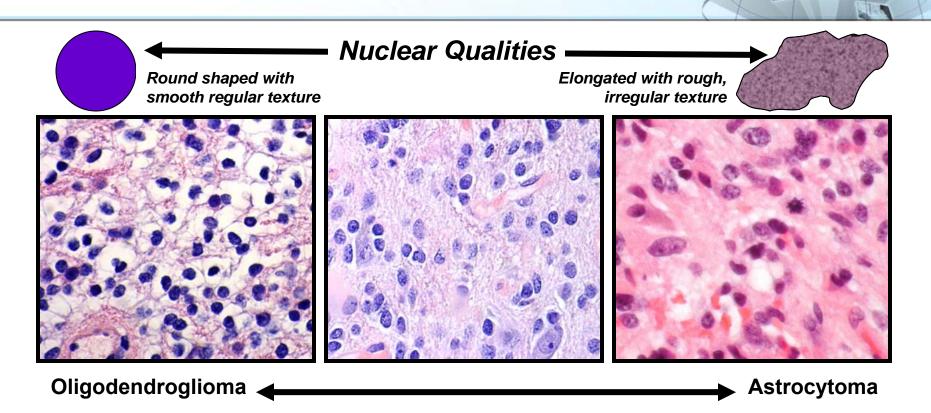
The Cancer Genome Atlas (TCGA): Rich "omics" set of GBM, digitized Pathology and Radiology

Pathology and Radiology Images from Henry Ford Hospital, Emory, Thomas Jefferson U, MD Anderson and others

TCGA Research Network



Characterizing Pathology of Gliomas



 Pathologists and image analysis algorithms describe, segment and classify microanatomic features in whole slide images



TCGA Brain Pathology Criteria Attributes that Relate to Entire Specimen Roughly 120 TCGA patients; Three Reviewers with Dan Brat adjudicating

Not Present: Not detected on any block Present: detected on any block Abundant: present in ≥ 50% of 10x fields in ≥ 50% blocks

- Microvascular hyperplasia elements (1,2)
- Complex/glomeruloid
- Circumferential endothelial hyperplasia
- Necrosis elements (3,4)
- Multiple serpentine pseduopalisading pattern
- Zonal necrosis

- Small cell component
- Gemistocytes
- "Oligodendroglioma-like" component with perinuclear cytoplasmic halos
- Perineuronal and/or perivascular satellitosis
- Multi-nucleated/giant cells
- Epithelial metaplasia
- Mesenchymal metaplasia
- Entrapped gray matter
- Entrapped white matter
- Micro-mineralization
- •
- Inflammation
- Macrophage/histiocytic infiltrates
- Lymphocytic infiltrates
- Polymorphonuclear leukocytic infiltrates



TCGA Whole Slide Images

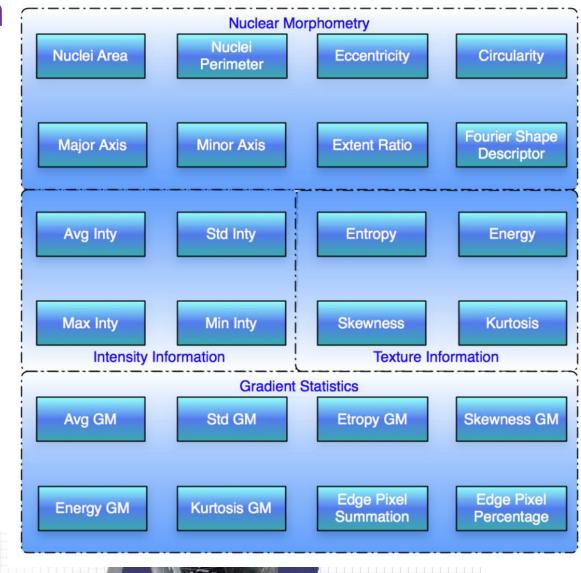
Feature Extraction

Jun Kong

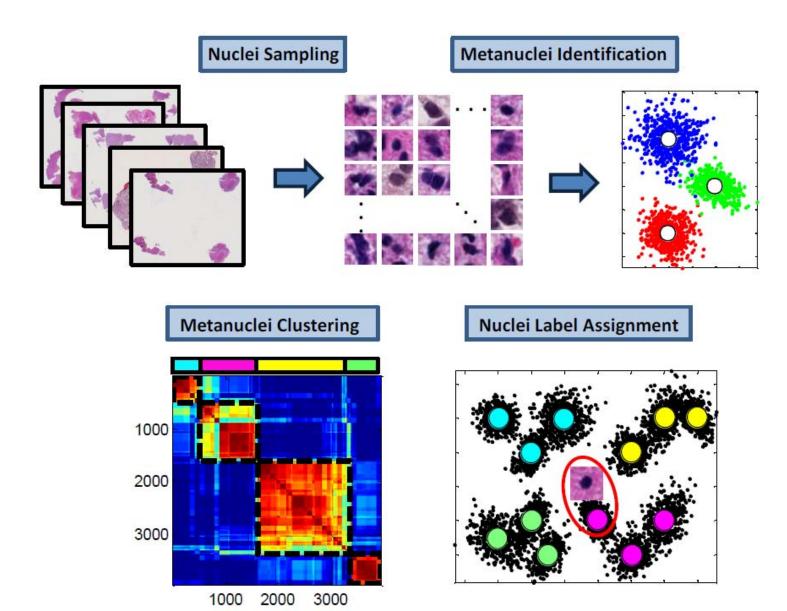
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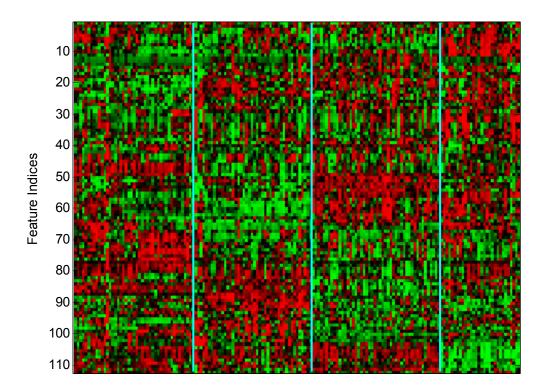
Informatics Grid "



Physical Interpretations of Image Analysis Results



Nuclear Features Used to Classify GBMs



Clustergram of selected features used in consensus clustering

an initiative of the National Cancer

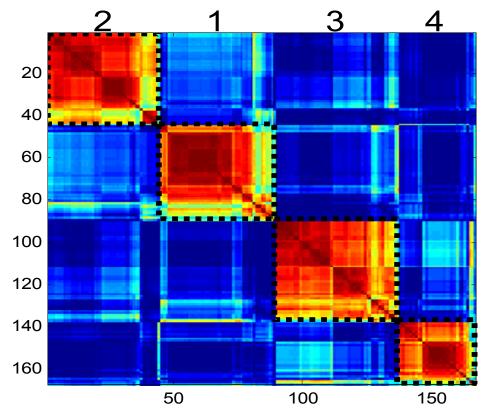


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Nuclear Features Used to Classify GBMs



Consensus clustering of morphological signatures

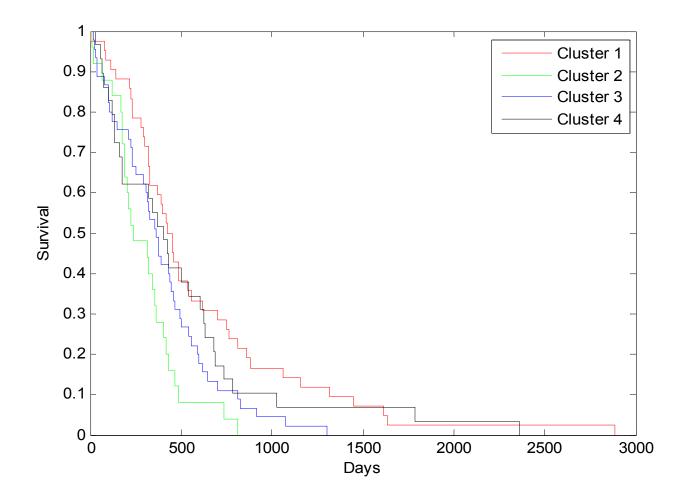
Study includes 200 million nuclei taken from 480 slides corresponding to 167 distinct patients.

an initiative of the National Cancer

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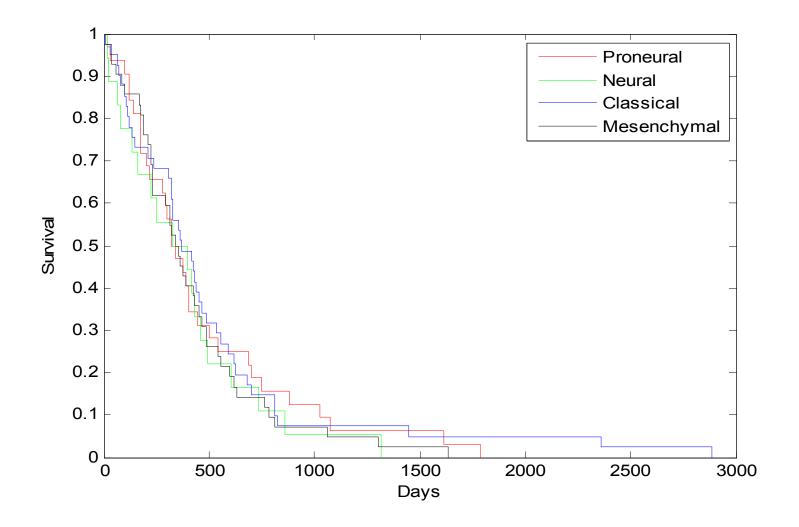
cancer Biomedical

Informatics Grid



Survival of morphological clusters



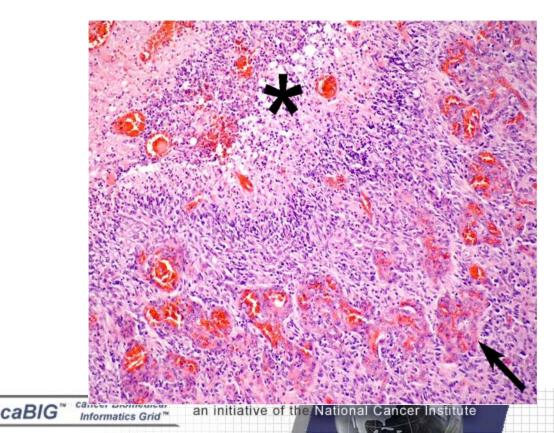


Survival of patients by molecular tumor subtype



Correlation of Necrosis, Angiogenesis, "omics" and Outcome

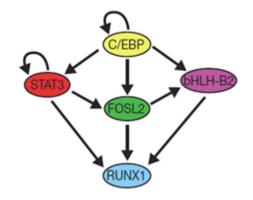
- GBMs display variable and regionally heterogeneous degrees of necrosis (asterisk) and angiogenesis
- These factors may impact gene expression profiles



Genes Correlated with Necrosis include Transcription Factors Identified as Regulators of the Mesenchymal Transition

Frozen sections from 88 GBM samples marked to identify regions of necrosis and angiogenesis
Extent of both necrosis and angiogenesis calculated as a percentage of total tissue area

Gene	SAM q-value
Symbol	(Corrected p-value)
C/EBPB	< 0.000001
C/EBPD	< 0.000001
FOSL2	< 0.000001
STAT3	0.0047
RUNX1	0.0082



Carro MS, et al. Nature 263: 318-25, 2010





Imaging Predictors of survival and molecular profiles in the TCGA Glioblastoma Data set The TCGA glioma working group

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Medicine, Boston, MA. ⁷SAIC-Frederick, Inc., Frederick, MD. ⁸University of Virginia, Charlottesville, VA. ⁹ Northwestern University Chicago, IL

<u>Emory</u>

David A Gutman¹ Lee Cooper¹ Scott N Hwang¹ Chad A Holder¹ Doris Gao¹ Carlos Moreno¹ Arun Krishnan¹ Jun Kong¹ Seena Dehkharghani¹ Joel Saltz¹ Dan Brat¹ Adam Flanders³ Eric Huang² Robert J Clifford² Dina Hammoud³ John Freymann⁷ Justin Kirby⁷ Carl Jaffe⁶

<u>TJU/CBIT/NCI</u>

UVA/Northwestern

Max Wintermark⁸ Manal Jilwan⁸ Prashant Raghavan⁸ Pat Mongkolwat⁹ Henry Ford

Lisa Scarpace⁴ Tom Mikkelsen⁴

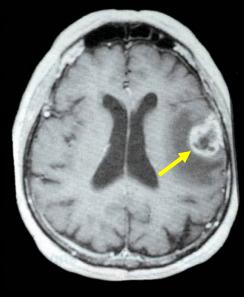
GBM Imaging Biomarkers Image Quantification Correlation with Pathology, "omics"



No enhancement Normal Vessels Stable lesion



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Rim-enhancement Vascular Changes Rapid progression Feature Sets in Neuro Radiology Adam Flanders (TJU), Daniel Rubin (Stanford), Lori Dodd (NCI), Eric Huang (NCI), many others

- Standardized validated feature sets
- Standard terminology:

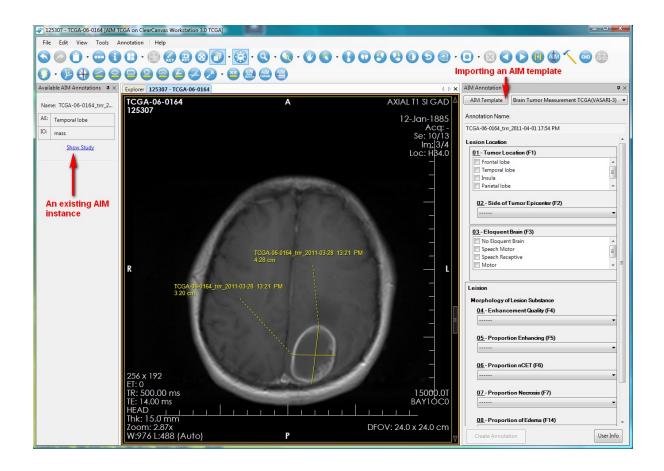
caBIG[™] cancer Biomedical

- Comprehensive, reproducible set of imaging features of cancer
- Quantitative, reproducible basis for assessing baseline disease and treatment response
- Gutman (Emory) leading effort to develop complementary algorithmic pipeline
- caBIG tools NBIA and AIME data service employed to manage images, annotations

an initiative of the National Cancer

 caBIG AIM template and ClearCanvas software supports Radiologist annotation process

AIM and AIM Template on ClearCanvas Workstation





caBIG"

Examples of Recent Findings Relating Radiology, Pathology and "Omics"

- Fisher's exact test demonstrated an association between the presence of CDKN2A homozygous deletion and the identification of an *ill-defined nonenhancing tumor margin* (p=0.007).
- TP53 mutant tumors had a *decreased mean tumor size* (p=0.002), measured as the maximum tumor dimension in the T2-weighted or FLAIR images
- Significant association between *minimal enhancing tumor* (≤5% proportion of the overall tumor) and Proneural classification (p=0.0006).
- Significant association between a >5% proportion of necrosis and the presence of microvascular hyperplasia in pathology slides (p=0.008).



Data Models to Represent Feature Sets and Experimental Metadata

PAIS |pās| : <u>Pathology Analytical Imaging Standards</u>

- Provide semantically enabled data model to support pathology analytical imaging
- Data objects, comprehensive data types, and flexible relationships
- Object-oriented design, easily extensible
- Reuse existing standards
 - Reuse relevant classes already defined in AIM
 - Follow DICOM WG 26 metadata specifications on WSI reference
 - Specimen information in DICOM Supplement 122 and caTissue
 - Use caDSR for CDE and NCI Thesaurus for ontology concepts



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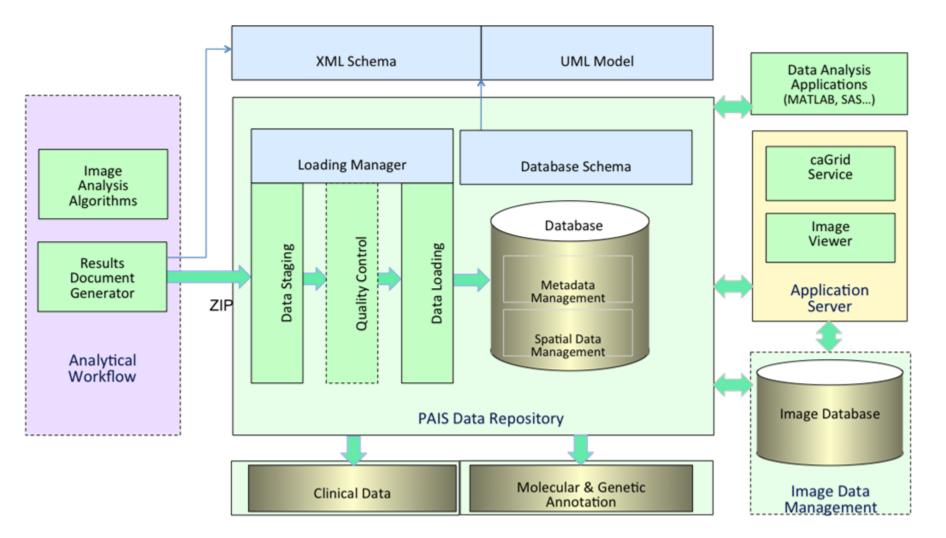
- Implemented with IBM DB2 for large scale pathology image metadata (~million markups per slide)
- Represented by a complex data model capturing multifaceted information including markups, annotations, algorithm provenance, specimen, etc.
- Support for complex relationships and spatial query: multilevel granularities, relationships between markups and annotations, spatial and nested relationships

Fusheng Wang, Emory



Feature Derivation, Analysis and Management Framework



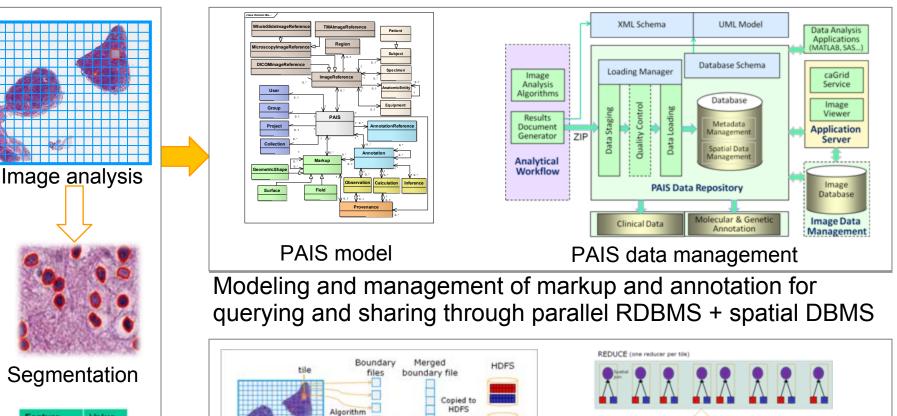






Fusheng Wang

Pathology Imaging GIS





Feature extraction

HDFS data staging MapRe

MapReduce based gueries

Data Scan

Data Scan

On the fly data processing for algorithm validation/algorithm sensitivity studies, or discovery of preliminary results

PAIS Database and Analysis Pipeline

- Suite of analysis algorithms and pipelines that carry out the following tasks:
- 1. segmentation of cells and nuclei;
- 2. characterization of shape and texture features of segmented nuclei;
- 3. storage of nuclei meta-data in relational database;
- 4. mechanism supporting spatial queries for humanannotated nuclei;
- 5. machine learning methods that integrate information from features to accomplish classification tasks.



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Pipeline for Whole Slide Feature Characterization



- 10¹⁰ pixels for each whole slide image
- 10 whole slide images per patient
- 10⁸ image features per whole slide image
- 10,000 brain tumor patients
- 10¹⁵ pixels
- 10¹³ features
- Hundreds of algorithms
- Annotations and markups from dozens of humans



Integrative Research Design Pattern: Other Emory Examples

- Minority Health Genomics and Translational Research Bio-Repository Database (MH-GRID)
- ACTSI Cardiovascular, Diabetes, Brain Tumor Registry
- Early Hospital Readmission
- CFAR (Center for AIDS Research) HIV/Cancer Project
- Radiation Therapy and Quantitative Imaging
- Integrative Analysis of Text and Discrete Data Related to Smoking Cessation and Asthma
- Semantic Query and Analysis of Integrative Datasets in Renal Transplant Clinical Studies (CTOT-C)

caBIG' une bureta

Thanks to:

- In silico center team: Dan Brat (Science PI), Tahsin Kurc, Ashish Sharma, Tony Pan, David Gutman, Jun Kong, Sharath Cholleti, Carlos Moreno, Chad Holder, Erwin Van Meir, Daniel Rubin, Tom Mikkelsen, Adam Flanders, Joel Saltz (Director)
- caGrid Knowledge Center: Joel Saltz, Mike Caliguiri, Steve Langella co-Directors; Tahsin Kurc, Himanshu Rathod Emory leads
- caBIG In vivo imaging team: Eliot Siegel, Paul Mulhern, Adam Flanders, David Channon, Daniel Rubin, Fred Prior, Larry Tarbox and many others
- In vivo imaging Emory team: Tony Pan, Ashish Sharma, Joel Saltz
- Emory ATC Supplement team: Tim Fox, Ashish Sharma, Tony Pan, Edi Schreibmann, Paul Pantalone
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- NIH/in silico TCGA Imaging Group: Scott Hwang, Bob Clifford, Erich Huang, Dima Hammoud, Manal Jilwan, Prashant Raghavan, Max Wintermark, David Gutman, Carlos Moreno, Lee Cooper, John Freymann, Justin Kirby, Arun Krishnan, Seena Dehkharghani, Carl Jaffe
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Thanks!