Mitigating Global Failure Regimes in Large Distributed Systems (Ongoing Research)

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No effective methods exist to predict failure regimes in large distributed systems—the search space is large and causality is difficult to establish. Our research goals are to: (1) develop design-time methods that system engineers can use to detect existence and causes of costly failure regimes prior to system deployment and (2) develop run-time methods that system managers can use to detect onset of costly failure regimes in deployed systems, prior to collapse.

Complex information systems encompass an

Determining causality is difficult given only



infeasible search space.

Hard

Problem



System Response Space

System Parameter Space

System Parameter Space = k^x , e.g., for k of 2^{32} and x of 1000, = $(2^{32})^{1000} = O(10^{9633})$

Atoms in the visible universe = about 10⁸⁰

patterns of global system behavior.



Second

Hard

Problem

Mitigation pdf

Approach One: Combine Markov Models, Graph Analysis and Perturbation Analysis

Granted

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of



Innovations in Measurement Science

Using simulated failure scenarios in a

Markov chain model to predict failures in a Cloud

Example: Markov simulation and perturbation of <u>a minimal s-t cut set</u> of a Markov chain graph:

Increase in Probability of Transition from Allocating Maximum state (9) to Allocating_Partial state (11)

Approach Two: Combine Anti-Optimization and Genetic Algorithms

М	ULTIDIMENSIONAL AN	ALYSIS TECHNIOUES			
	Principal Componer	nts Analysis,	Gro	Growing Collection of Tuples:	
	GENETIC A	LGORITHM	{Generation, I {Generation, I {Generation, I {Generation, I {Generation, I	{Generation, Individual, Fitness, Parameter 1 value,Parameter N value} {Generation, Individual, Fitness, Parameter 1 value,Parameter N value}	
ſ	Recombination	Selection based on	{Generation, I {Generation, I {Generation, I	Individual, Fitness, Parameter 1 value,Parameter N value} Individual, Fitness, Parameter 1 value,Parameter N value} Individual, Fitness, Parameter 1 value,Parameter N value}	

- **Corresponds to software failure** scenario involving multiple faults/attacks.
- Simulation identifies threshold beyond which increased failure incidence causes drastic performance collapse

 \rightarrow Verified in target system being modeled (i.e., Koala, a large-scale <u>simulation of a Cloud)</u>





Approach Three: Measuring Key System Properties such as Critical Slowing Down

> <u>A simple univariate example predicting power</u> grid blackout in a human engineered system*

How might your organization benefit from collaborating with us?

IF your organization designs and deploys Clouds and other large distributed systems AND



* From P. Hines, E. Cotilla-Sanchez, and S. Blumsack. Topological Models and Critical Slowing Down: Two Approaches to Power System Risk Analysis. Proceedings of the 44th Hawaii Conference on System Sciences. IEEE Computer Society, Washington, DC, USA, pp. 1-10.

- You wish to improve the reliability of your system AND
- You have a model of your system OR
- You are willing to share sufficient information for us to construct a model AND you are willing to help us ensure our model suitability represents your system
- **THEN** working together we could help you improve the reliability of your system (or specific aspects of your system) by:
 - Applying our design-time methods to search the design space for potential collapse scenarios (and iterating on any proposed design revisions you create to mitigate collapse scenarios) AND/OR
 - Exploring run-time monitoring and measurement approaches that could signal incipient onset of collapse scenarios that were not detected using our designtime methods
- WIN-WIN: we would gain additional evaluation and refinement of our methods and $\circ \circ$ you could gain a transfer of our technology to enhance your design process.

http://www.nist.gov/itl/antd/emergent_behavior.cfm