

Unclassified



LDRD

Laboratory Directed Research and Development

(U)SECURE: Science and Engineering of Cyber security by Uncertainty quantification and Rigorous Experimentation

*Ali Pinar
Sandia National Laboratories
Livermore, CA*



Sandia National Laboratories is a multimission laboratory managed and operated by National Technology and Engineering Solutions of Sandia LLC, a wholly owned subsidiary of Honeywell International Inc. for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.

Sandia National Laboratories is a multimission laboratory managed and operated by National Technology & Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.

What is the return of investment for cyber?



Credit: Staff Sergeant Jason Gamble, United States Air Force



We cannot improve,
what we cannot measure



(U) SECURE's story in a nutshell

(U) Cyber experimentation should be a pillar of science of cyber security, just as computational Science and Engineering (CSE) is a pillar of science.

- (U) Cyber experimentation is commonly used to answer questions about cyber systems
 - (U) but lack of rigor limits its use in high-consequence systems
- (U) To study complex cyber systems, we need to
 - (U) answer “what if questions” with high-confidence **Emulytics**
 - (U) assess confidence in our results under uncertainty **Uncertainty Quantification**
 - (U) make robust decisions under uncertainty in an adversarial environment

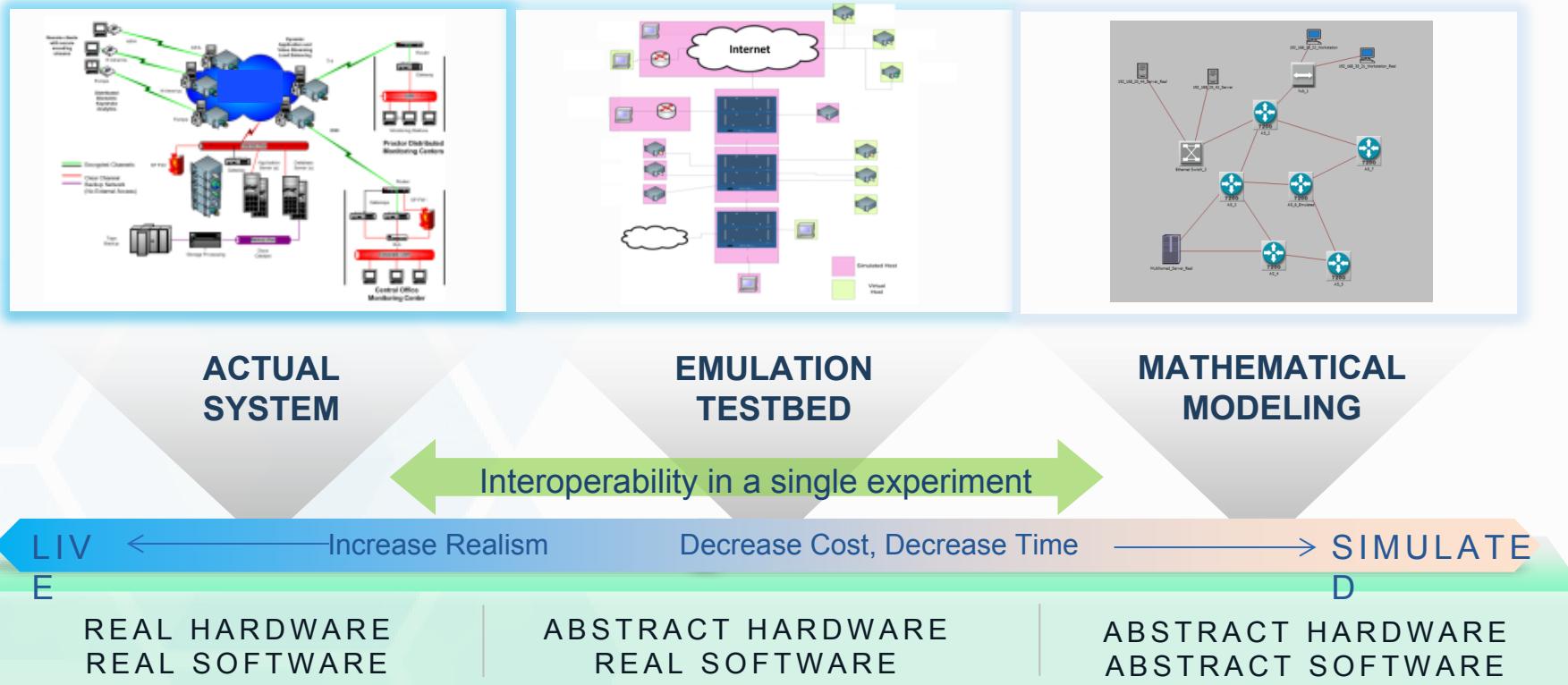
Adversarial Optimization

- (U) Inspiration: Sandia's know-how and capabilities from our nuclear stockpile stewardship
- (U) Challenge: Cyber systems are different than physics-based systems

(U) Cyber experimentation approaches



Figure unclassified



(U) SECURE's position:

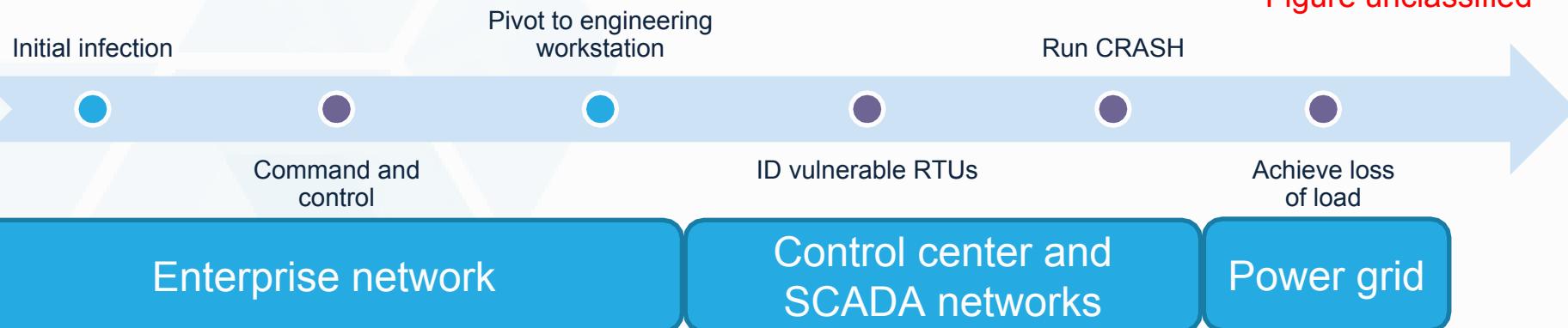
- (U) Results should be independent of the platform and the tools used for the experiment

(U) Exemplar: How vulnerable is the power grid against a cyber attack?



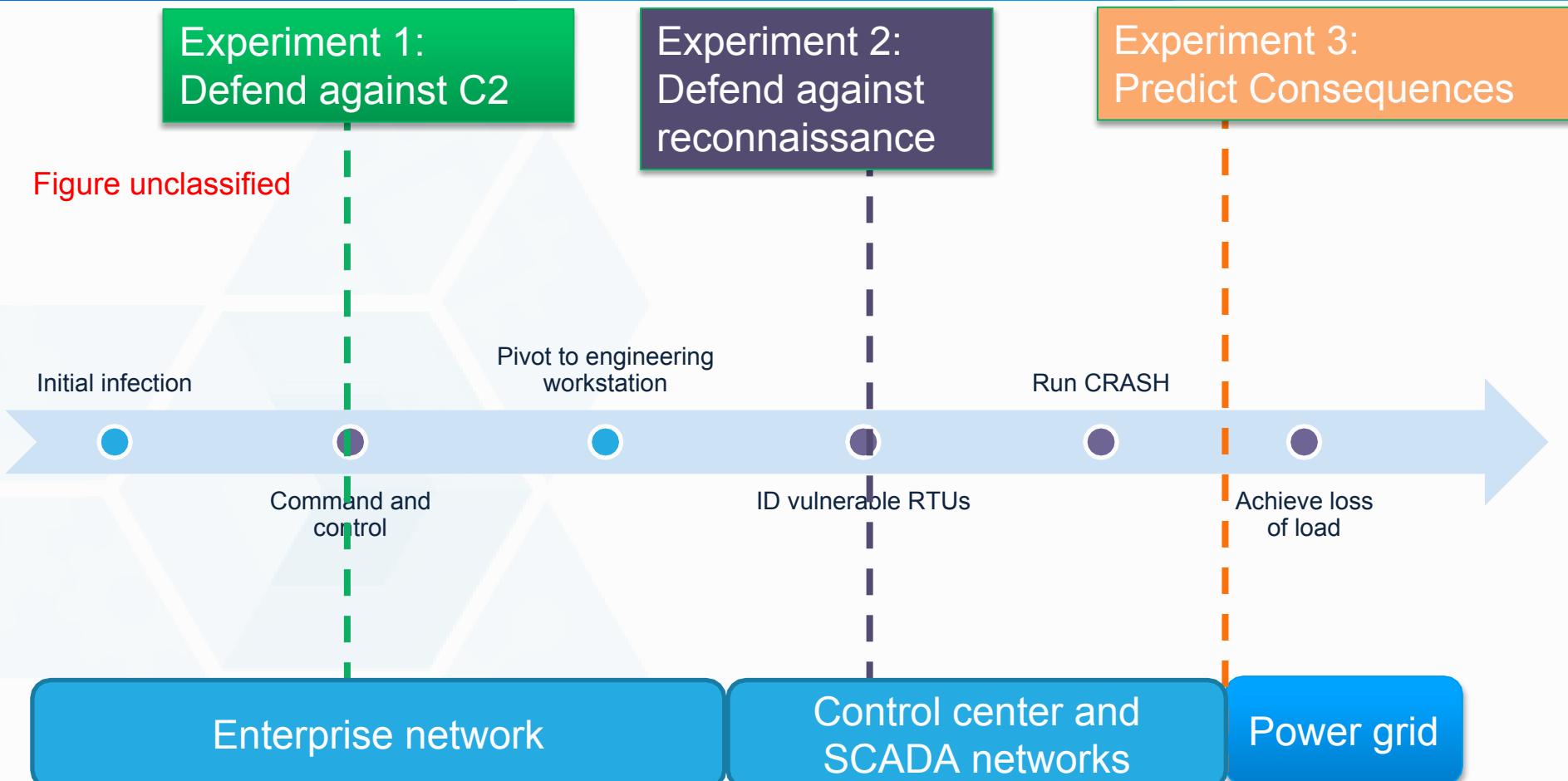
- (U) Ukraine attack was based on Crash Override Malware
- (U) The attacker gains remote access to power grid components to turn them on and off.

Figure unclassified



- (U) Goal: characterize loss of load resulting from malware infection in enterprise network
 - (U) Account for uncertainties in threat, network conditions
- (U) Approach: Piecewise studies to inform Markov transition probabilities and uncertainties

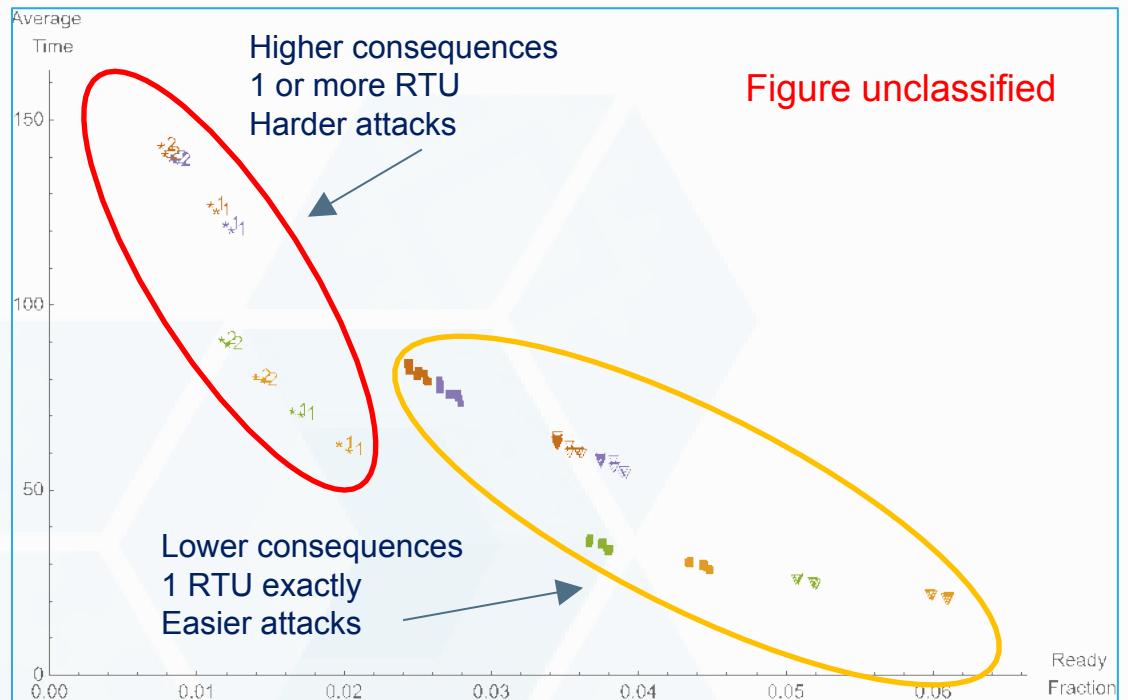
(U) We design an experiment for each step, and aggregate results with a Markov model



Overarching Themes: verification and validation, extreme events, scalable algorithms



(U) Aggregated results



(U) Defender goal: push attacker toward top-left of the plot (e.g. through better IDS)

(U) Each dot on the chart above represents a combination of C2 data, scanning/detection data, and attacker/defender strategy

- (U) Plotting attack success metrics from Markov analysis: mean time to attack success vs. fraction of time in the “READY” state.
- (U) Extended our analysis framework to support **UQ in transition probabilities**, and **variations in each step’s inherent timestep**.
- (U) Experiments provide range of transition probabilities (depending on scenario, attacker strategy, etc.)

(U) Markov analysis allows:

- Estimates of how secure the system is under attack
- Ranking of attacker/defender strategies



(U) So What?

- (U) What changed?
 - (U) We produced an **objective** process that can **quantify** security.
 - (U) All assumptions are listed; all processes are repeatable;
 - (U) All experiments are verified; all models are validated;
 - (U) We have a scientific processes that can, and will be improved.
 - (U) No more disagreeing with expert opinions.
 - (U) Instead challenge assumptions; propose better algorithms/metrics.
- (U) What can we do now? **Quantifiable Security**
 - (U) Quantify return on investment for cyber security
 - (U) Identify critical components both for improving security and model fidelity
 - (U) Quantify attack consequences and enable mission-driven cyber security

(U) Optimal Segmentation – rigorous comparison of two solutions



Figure unclassified

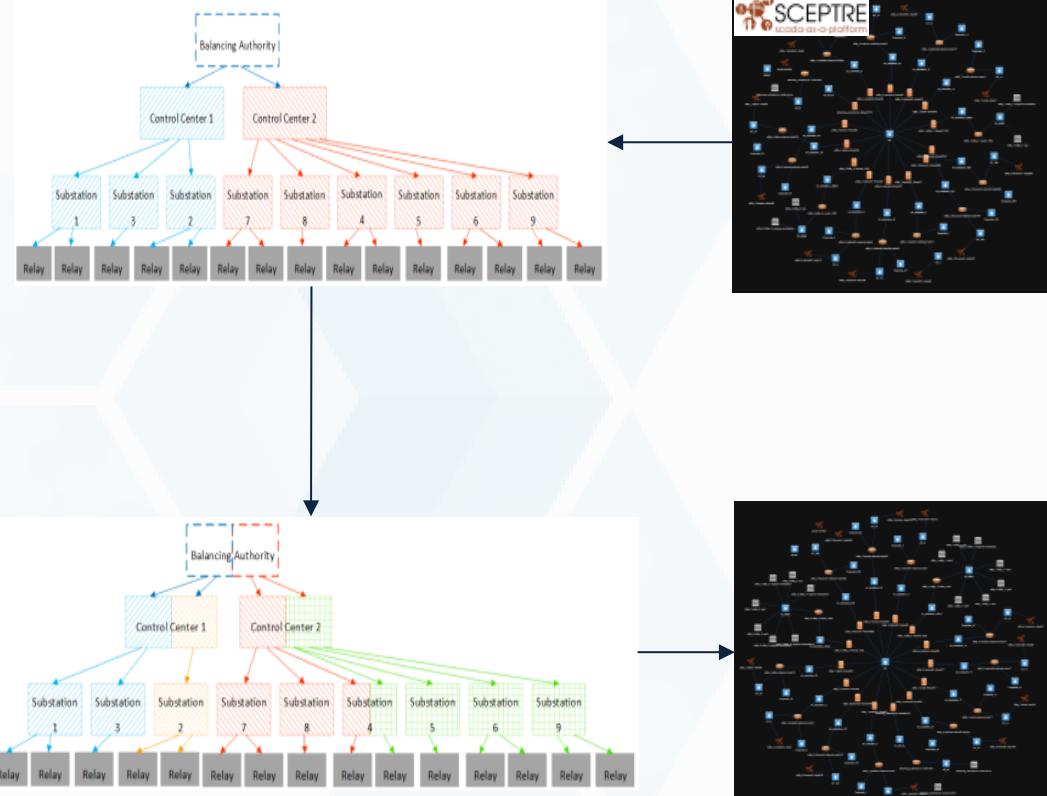
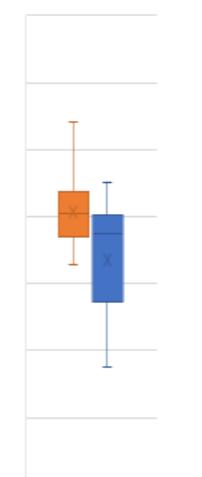
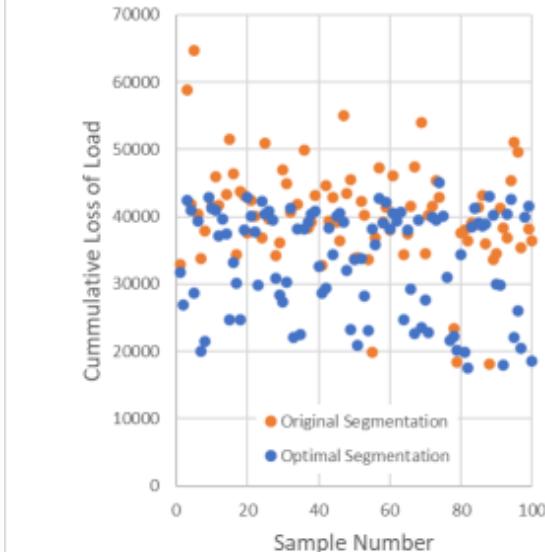
Optimization/Emulation WorkflowResults

Figure unclassified



(U) Takeaway: Designed a workflow that interfaces emulation with mathematical optimization to investigate network segmentation

Unclassified

(U) Takeaway: Mathematical optimization identifies a segmentation policy that is more robust under a CrashOverride attack



(U) Identifying extreme events is crucial

- (U) We need to identify events with low-likelihood yet high-consequence
 - (U) Solution: Multi-fidelity sampling for tail events; optimization for extreme points

Figure
unclassified

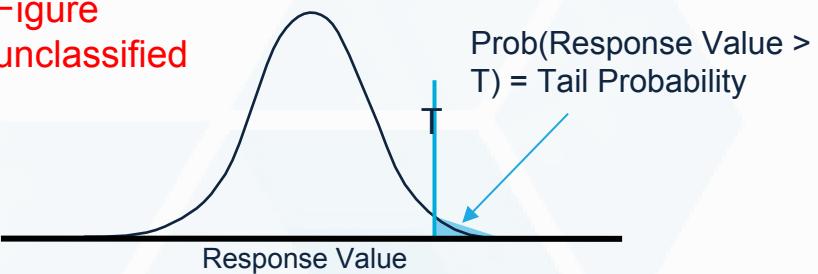
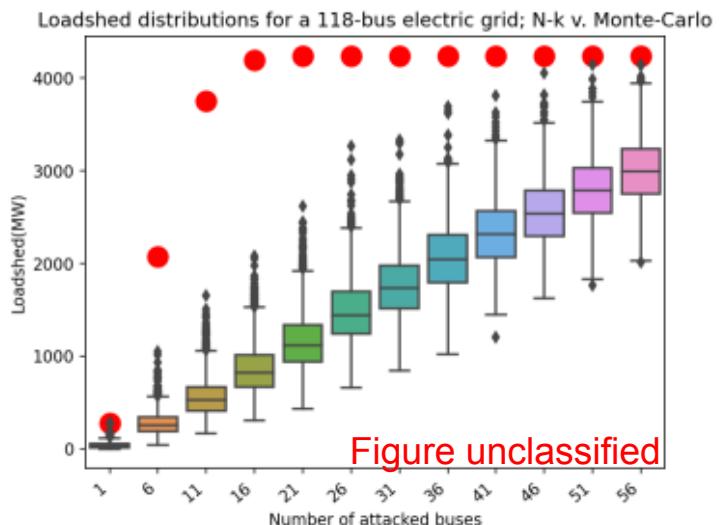
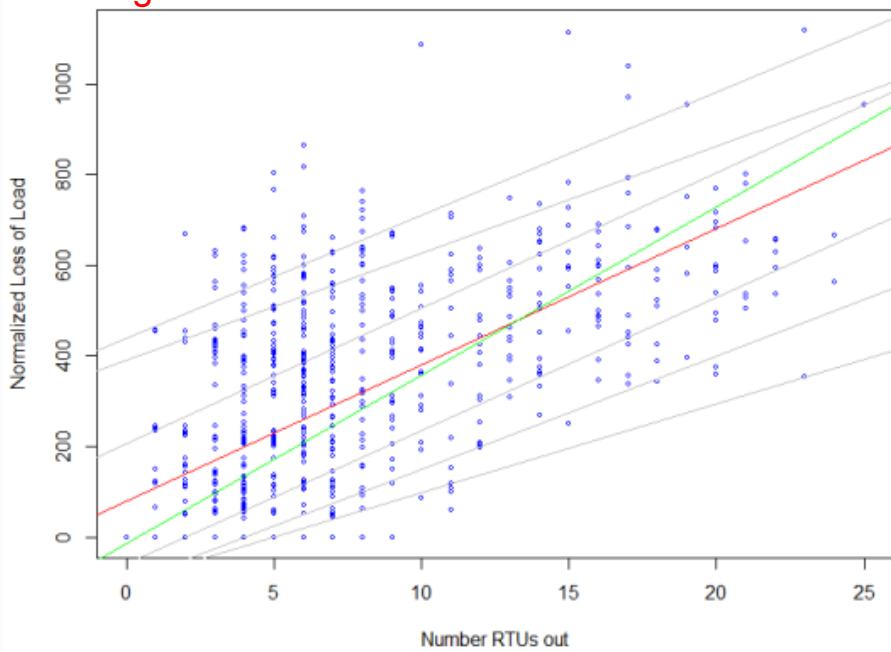


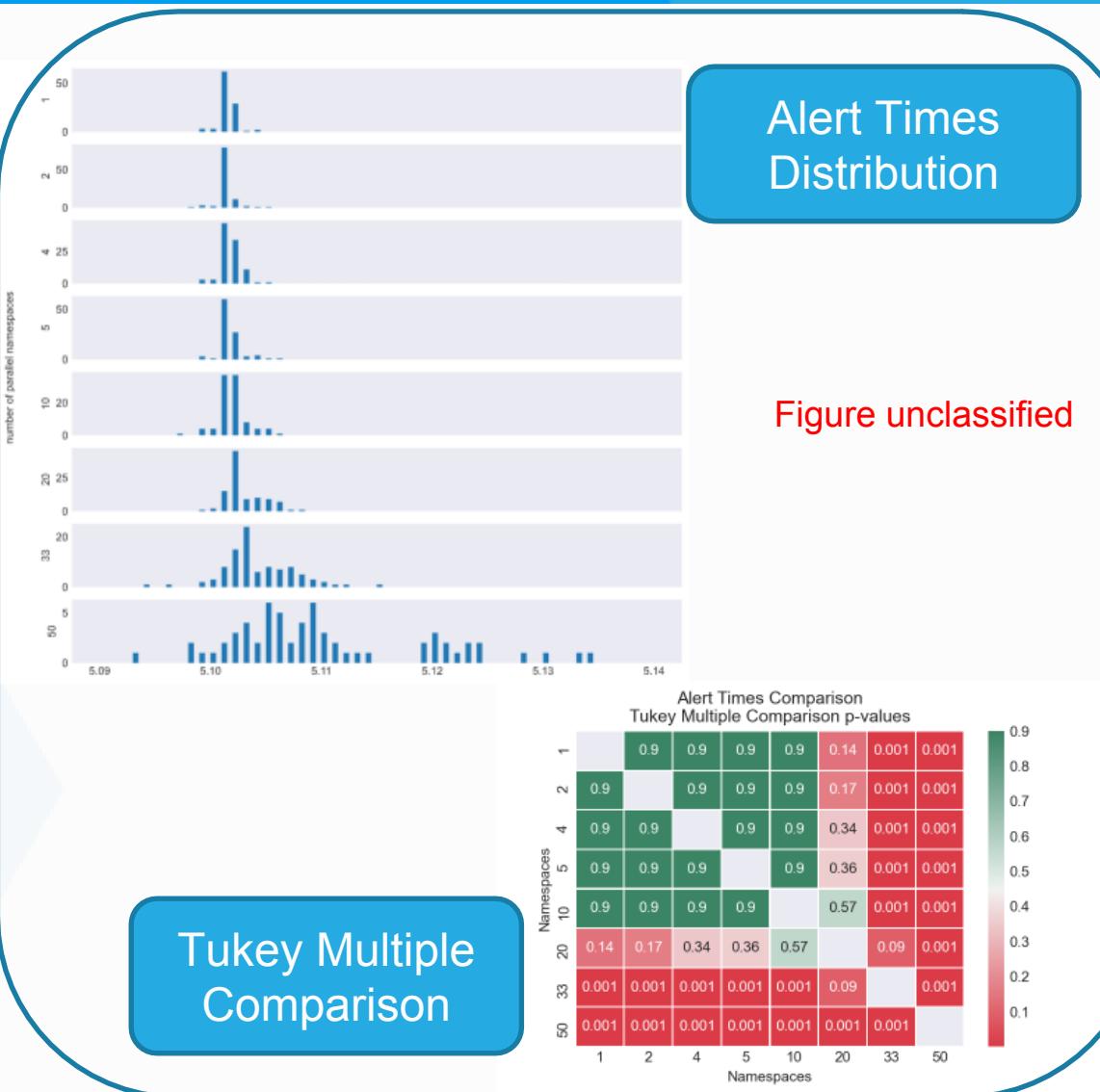
Figure unclassified





(U) Verify each experiment

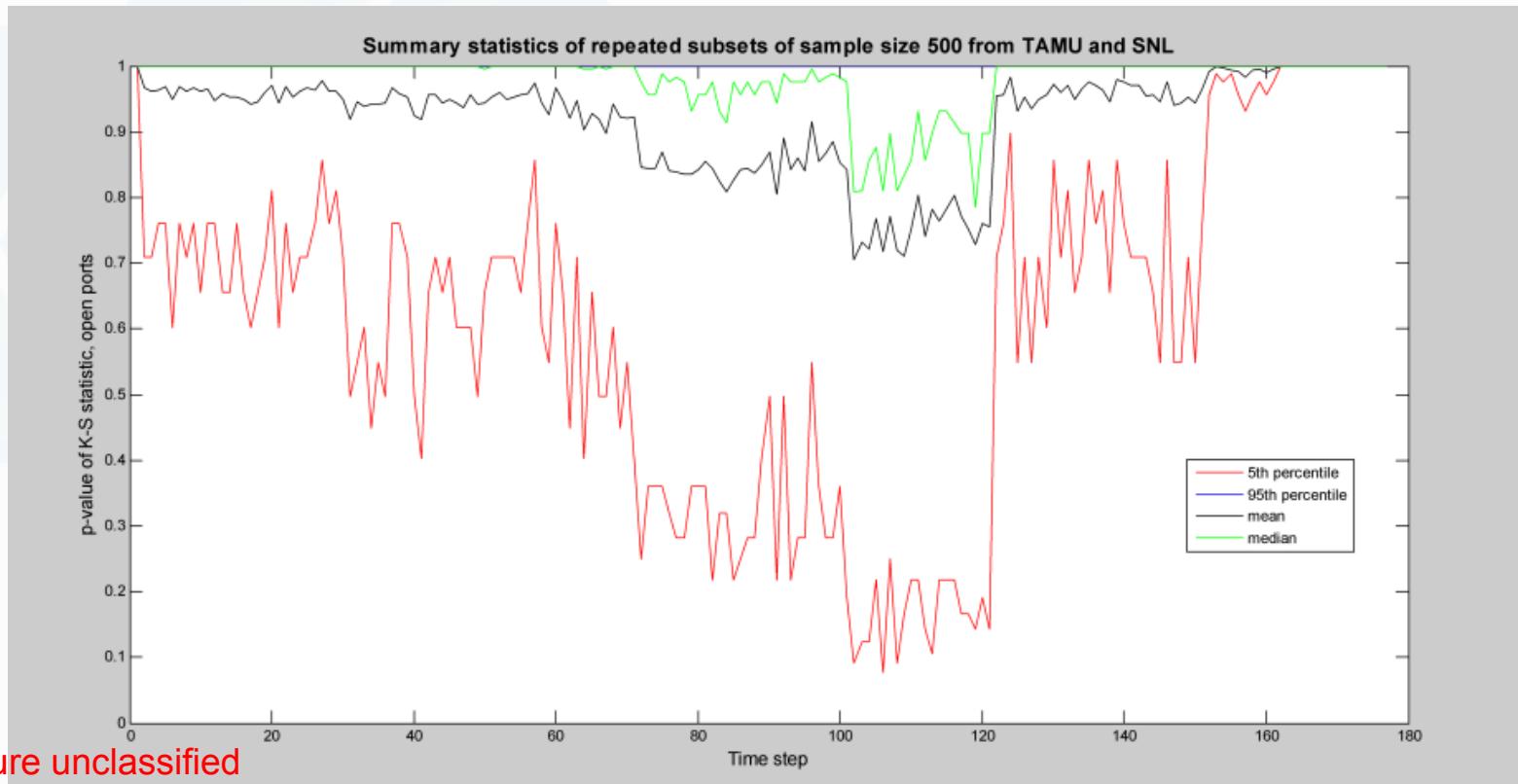
- (U) Distribution of alert times shift as namespaces are added
- (U) Quantified similarity with Tukey Multiple Comparison Test
 - (U) Shows clear drop in similarity after 10 namespaces
- (U) Large p-value indicates that the null hypothesis can't be rejected
 - (U) Larger p-value -> similar results



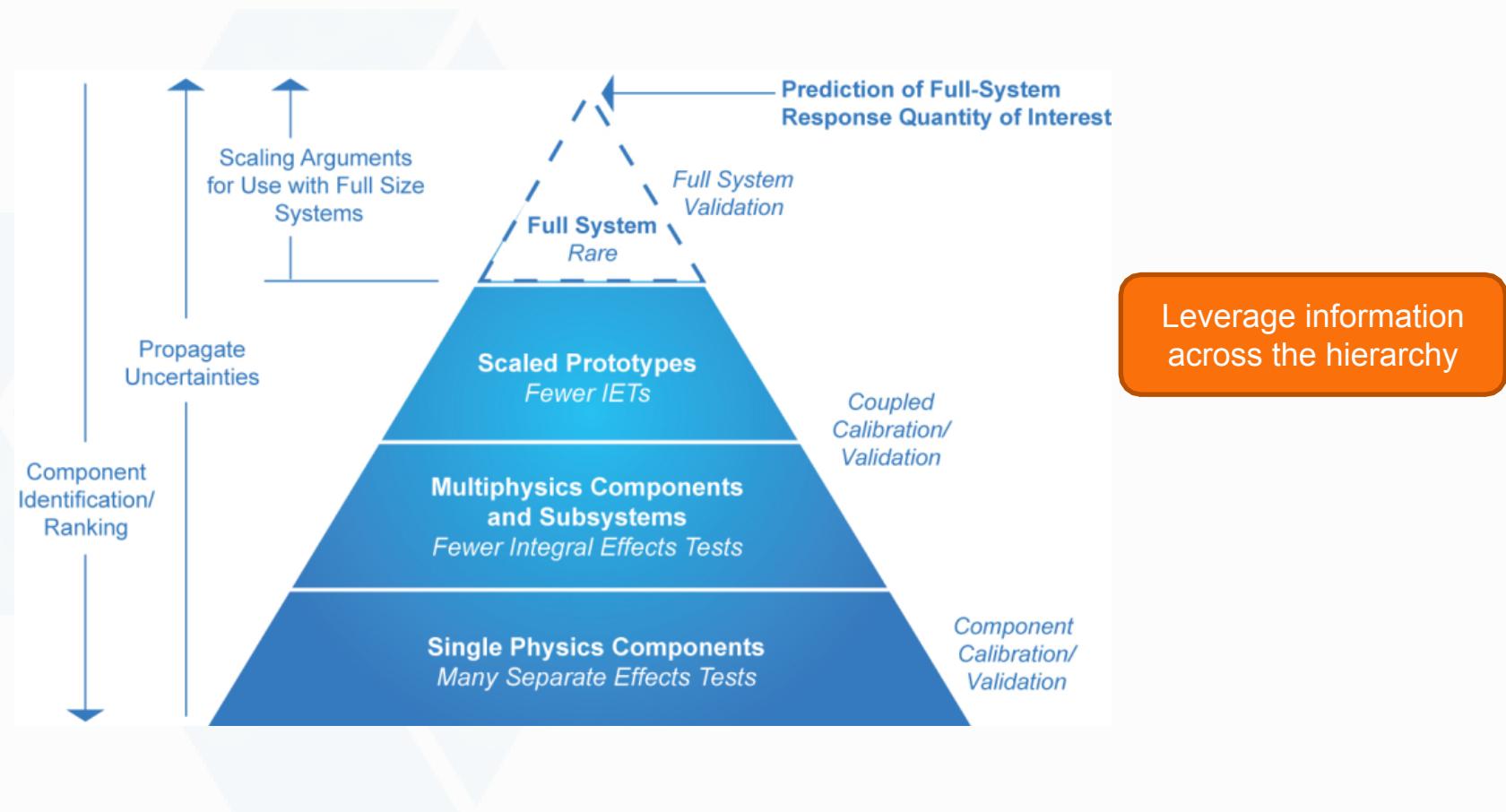


(U) Validate each model

- (U) Can we validate our models against data from real systems?
 - (U) Joint study with TAMU
- (U) Mean and median indicate good agreement. The low values of the 5th percentile between 100-120 seconds help identify times which have some realizations with less agreement.



Validation Lessons from Weapons program: small scale tests → full system



Example study for NC3 survivability/endurability



Experimental plan

- Experimental questions
- Identify inputs, outputs, topology

Topology

- Topology information from NC3
- Transfer this information to experimentation topology

Tools

- Scenario orchestration
- Fault/degradation injection (Netflix “Chaos Monkey”, but for experimental testbeds)

Efficient experiments

- Multifidelity models (emulation + math/simulation/surrogate models)
- Sampling strategies to comprehensively cover space of possible fault scenarios

Validation

- Validate low fidelity models against high fidelity models
- Start with small topology, exhaustively enumerate fault scenarios, test MF model against exhaustive results
- Where possible, compare to real world data

Rigorous Cyber Experimentation can provide NC3, what CSE provided to the nuclear weapons programs



- We cannot improve what we cannot measure
- Cyber experimentation provides measurements and is an essential tool for designing future complex systems
 - Rigor is paramount for high-consequence systems
- We need to look at the whole system and
 - build our confidence bottom up
 - tailor requirements top down
- Computation Science and Engineering (CSE) is a pillar of our nuclear weapons programs
 - Inspiration behind SECURE
- SECURE has been developing methods and tools to bring rigor into cyber experimentation. I can be used to
 - Assess a system and/or its components
 - Set justifiable requirements for components
 - Enable survivability/endurance by design
- We worked with NC3 in mind, and we are ready to face this challenge