

Unclassified



**LDRD**

Laboratory Directed Research and Development

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

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# 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

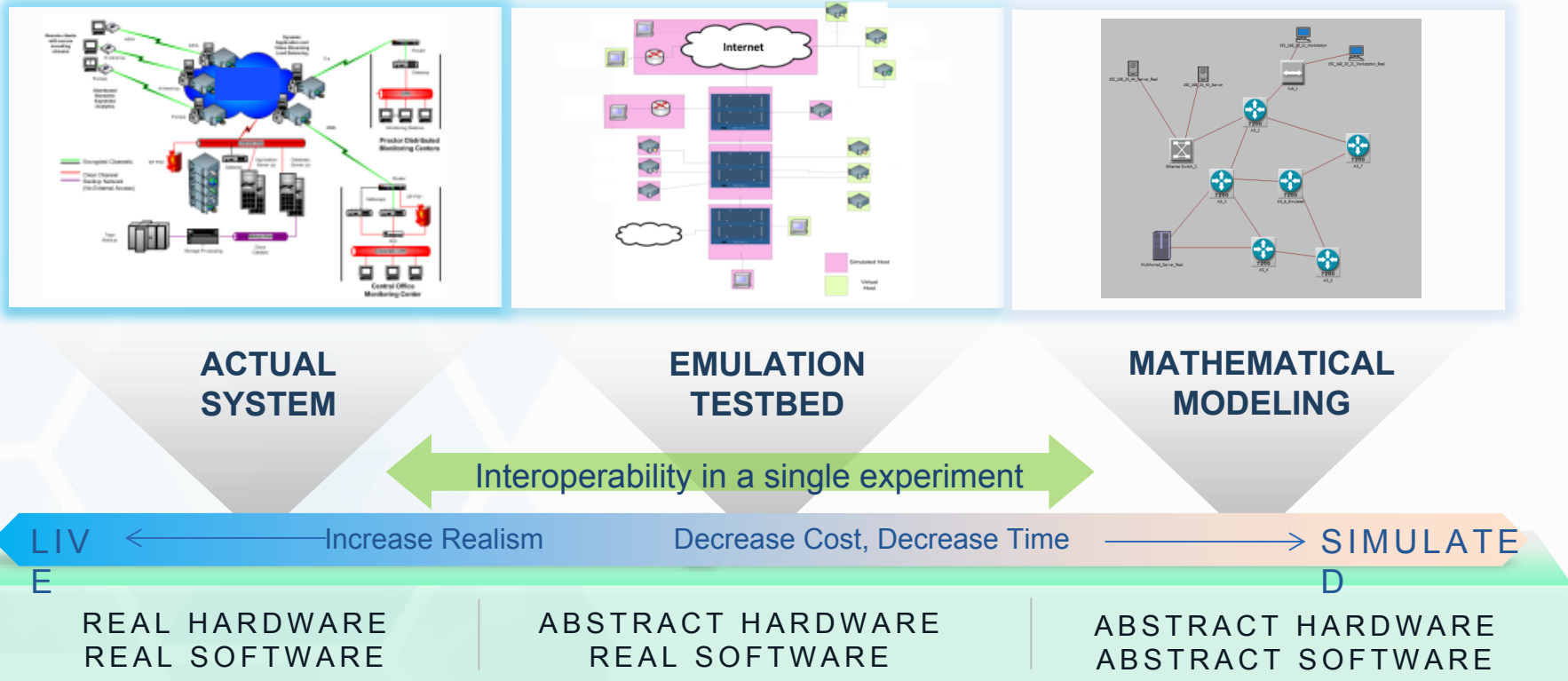
(U)Cyber experimentation should be a pillar of science of cyber security, just as computational Science and Engineering (CSE) is a pillar of science.

- (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

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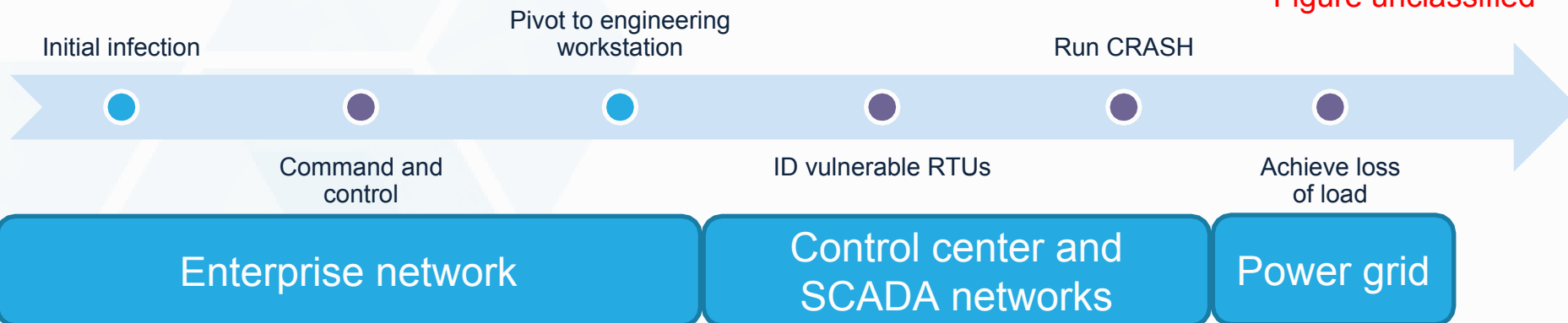


# (U) Exemplar: How vulnerable is the power grid against a cyber attack?



- (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.

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- (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

Experiment 1:  
Defend against C2

Experiment 2:  
Defend against  
reconnaissance

Experiment 3:  
Predict Consequences

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Initial infection

Pivot to engineering  
workstation

Run CRASH

Command and  
control

ID vulnerable RTUs

Achieve loss  
of load

Enterprise network

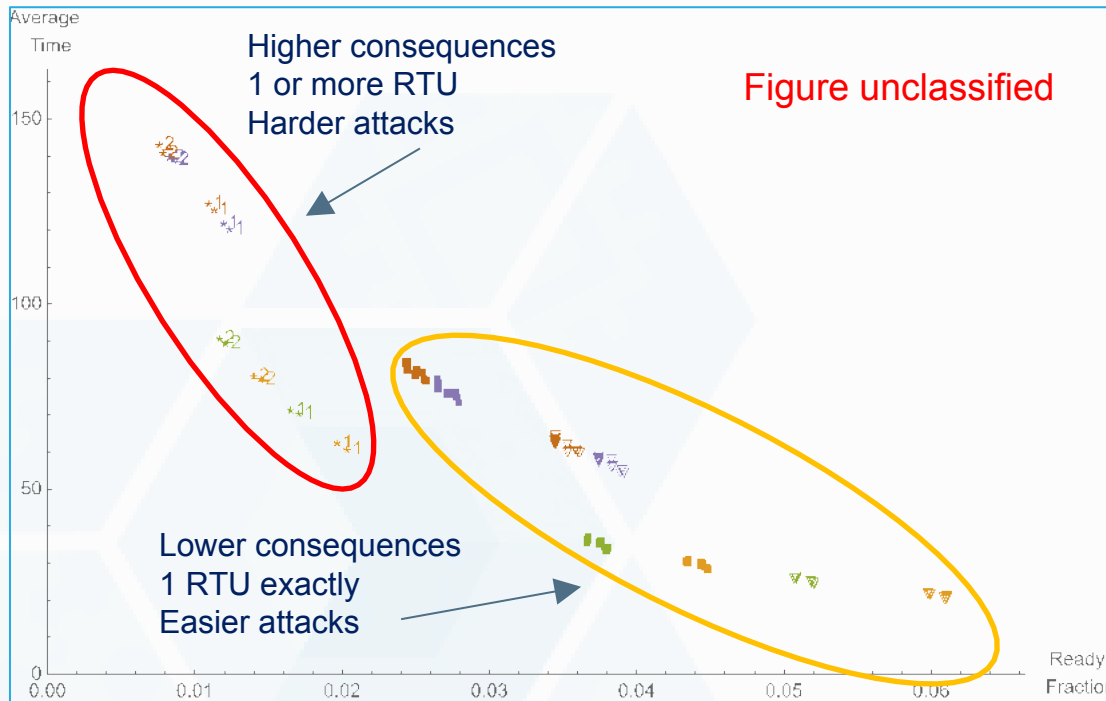
Control center and  
SCADA networks

Power grid

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

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## Optimization/Emulation Workflow

## Results

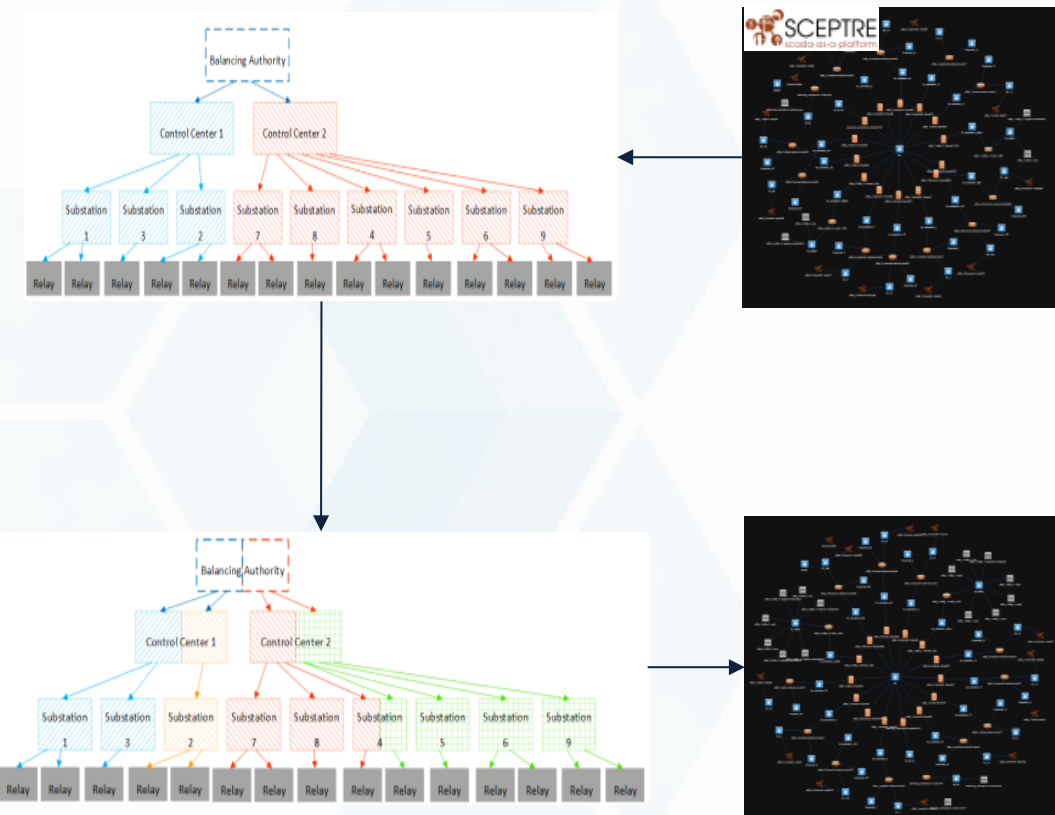
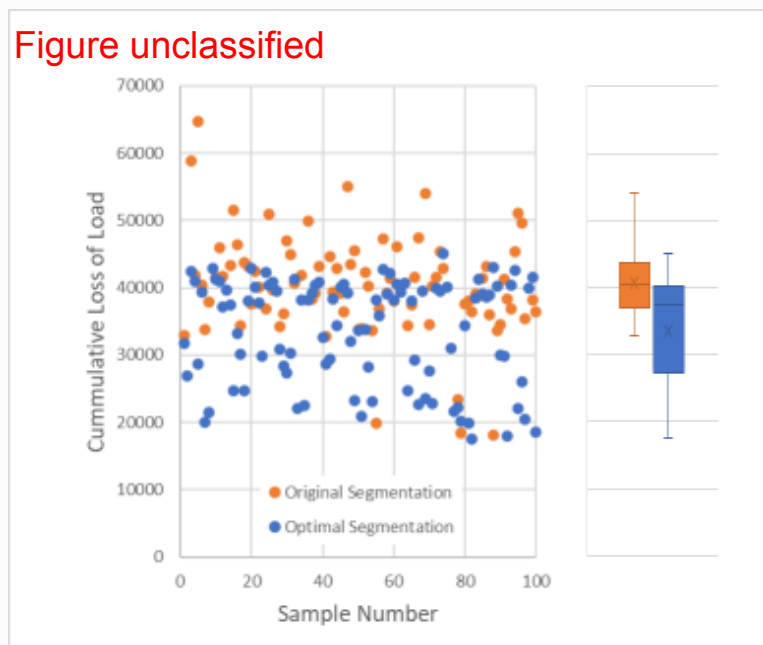


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(U) Takeaway: Designed a workflow that interfaces emulation with mathematical optimization to investigate network segmentation

(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

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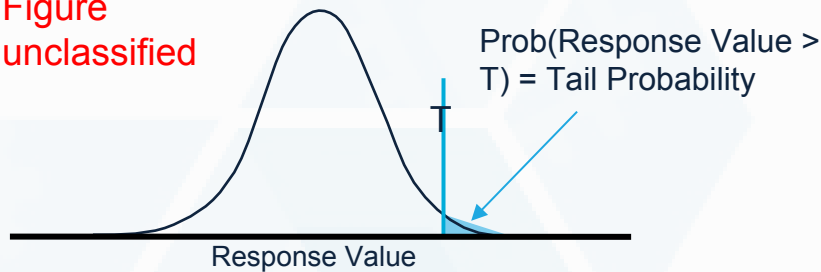
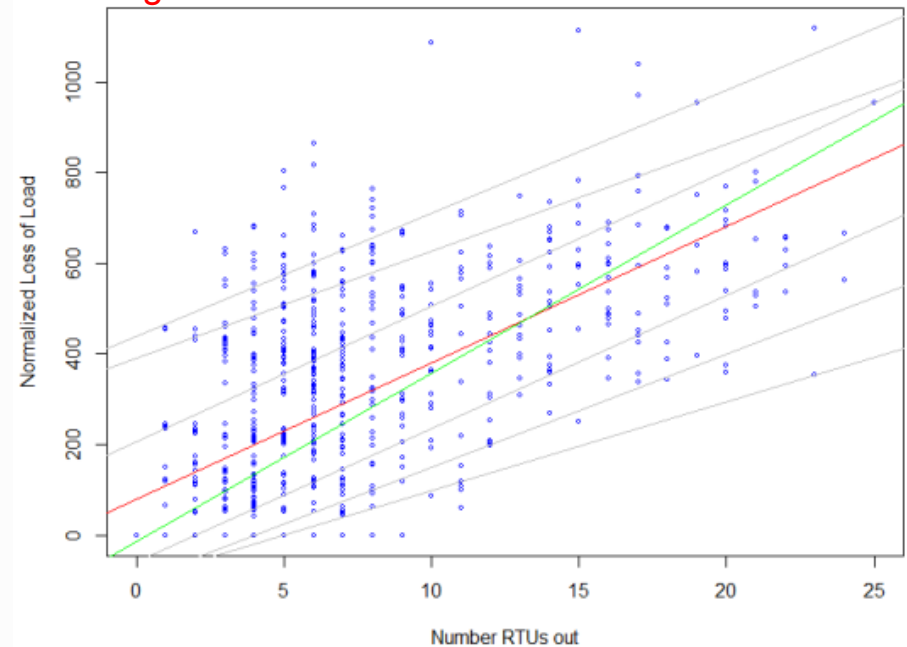


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Loadshed distributions for a 118-bus electric grid; N-k v. Monte-Carlo

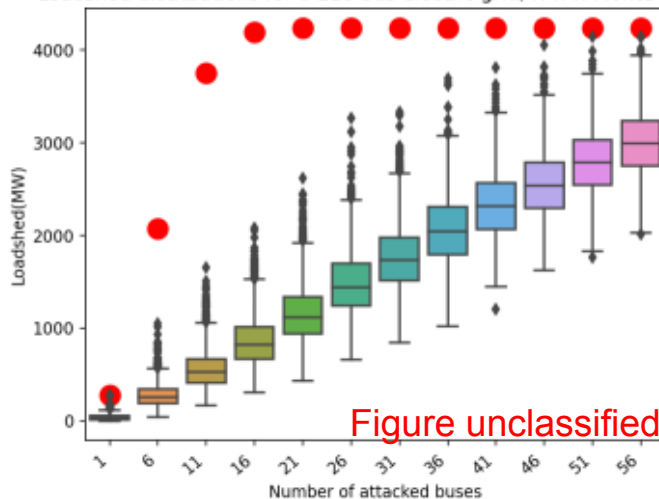
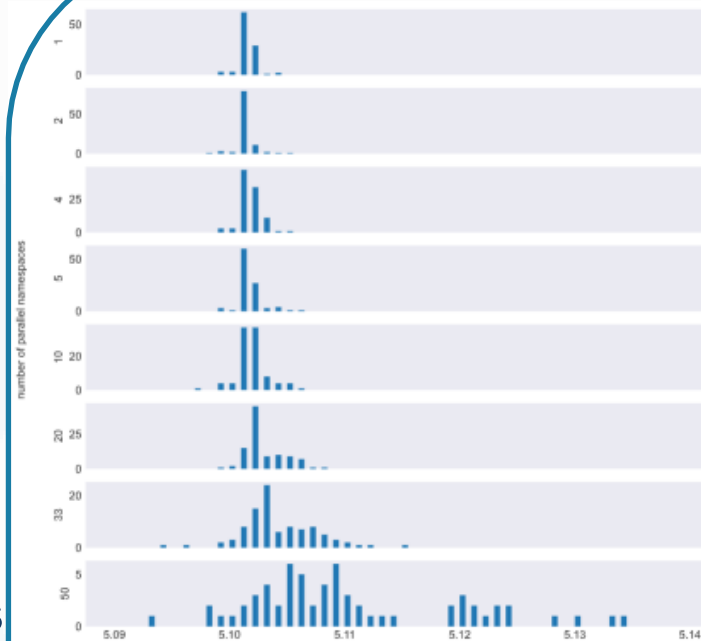


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# (U) Verify each experiment

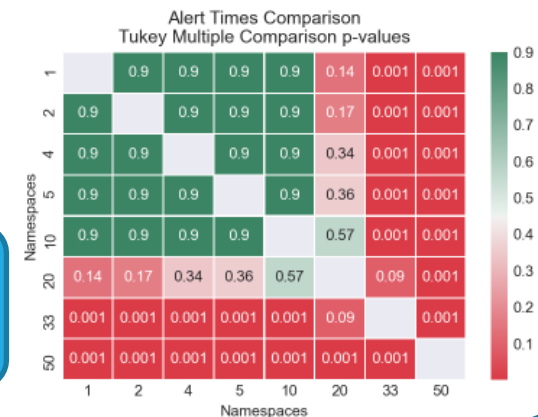
- (U) Distribution of alert times shift as namespaces are added
- (U) Quantified similarity with Tukey Multiple Comparison Test
  - (U) Shows clear drop in similarity after 10 namespaces
- (U) Large p-value indicates that the null hypothesis can't be rejected
  - **(U) Larger p-value -> similar results**



Alert Times Distribution

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Tukey Multiple Comparison





# (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 5<sup>th</sup> percentile between 100-120 seconds help identify times which have some realizations with less agreement.

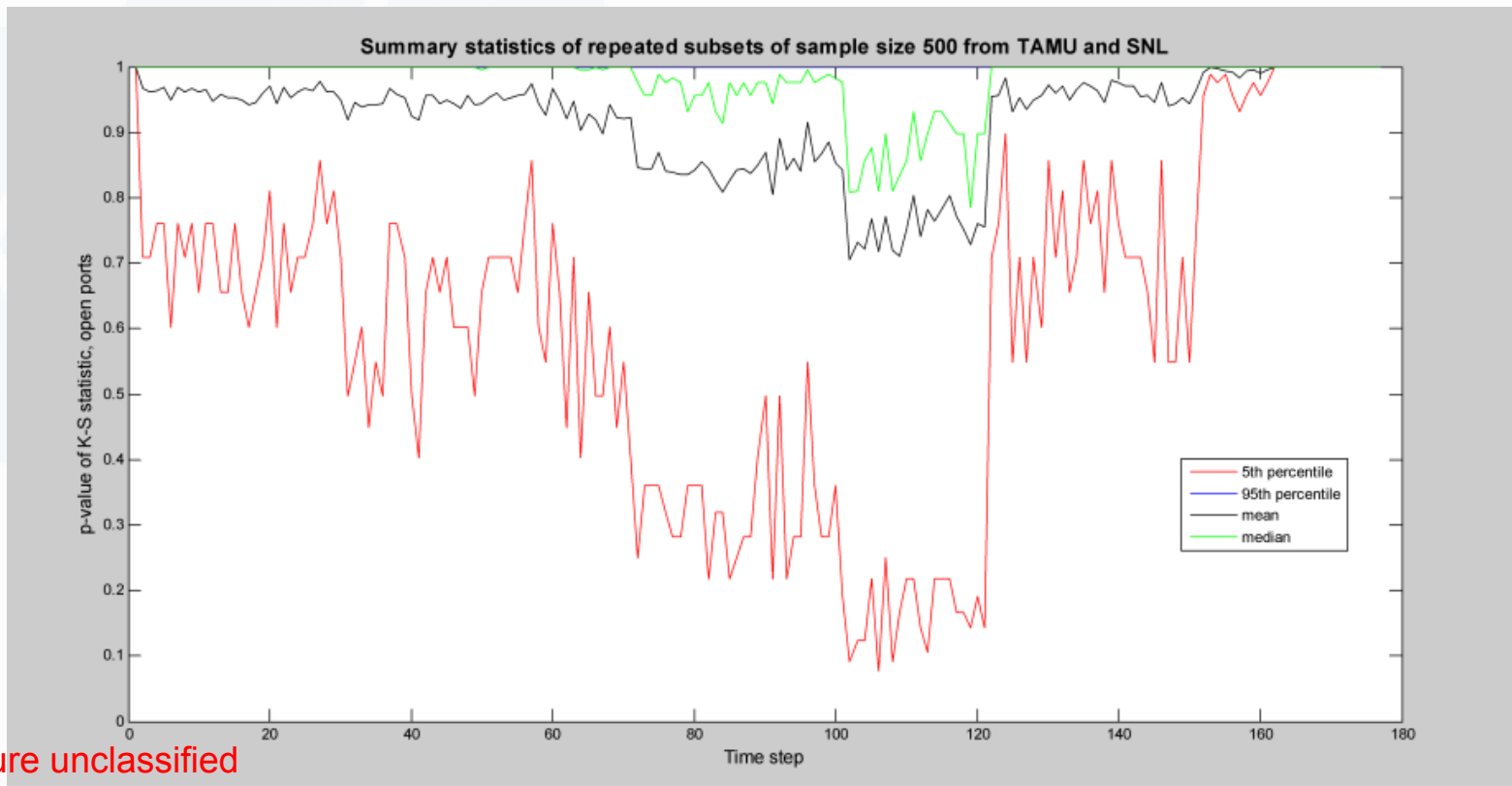
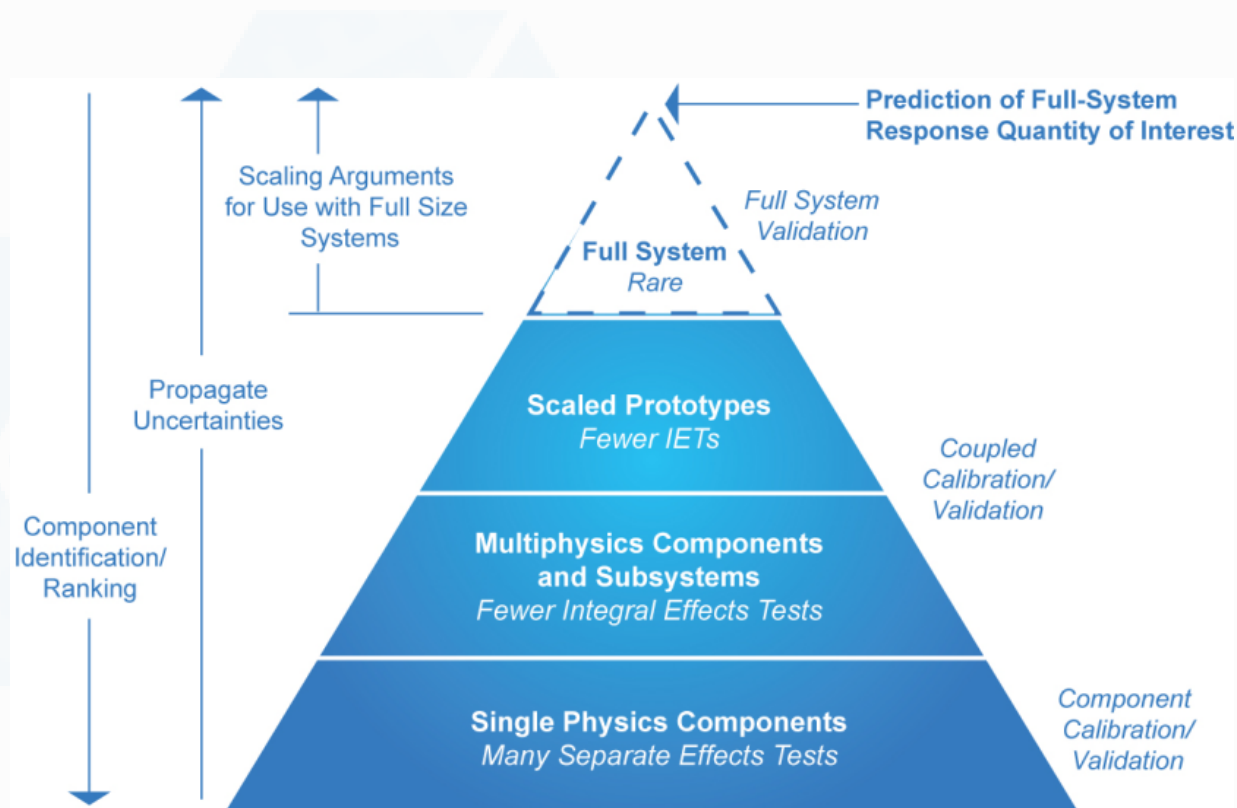


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# Validation Lessons from Weapons program: small scale tests → full system



Leverage information  
across the hierarchy

# 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. It 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