



Project Goals and Approach

- Identify how a cyber-physical system may be manipulated through the environment (dynamics aware attacks)
 - Develop methodologies and metrics for assessing "how close" to unstable/unsafe/insecure states a CPS might be pushed
- General Approach
 - Develop hybrid (continuous/discrete) models of CPS
 - Develop bounds on trajectories of those systems
 - Measure "how close" a bound may approach undesirable states



Specific Technical Goals

Project aims to develop

- Security metrics (ε-stability, availability)
- Adversary classes that capture different attacks on CPS
- Annotations and algorithms that bound state trajectory
- Evaluation in context of CPS benchmark examples

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Builds on previous project

- Two problems addressed in the past year:
 - (a) Define different threat classes for CPS & analysis of CPS under these classes of attacks
 - (b) Study the cost of privacy in distributed control
- More details in the following two slides:
- Highlights:
 - Proofs from Simulations and Modular Annotations, presented by Zhenqi Huang at the 17th International Conference on Hybrid Systems: nominated for best student paper award.
 - Invariant Verification of Nonlinear Hybrid Automata Networks of Cardiac Cells, Z. Huang, C. Fan, A. Mereacre, S. Mitra, and M. Kwiatkowska. To appear in Computer Aided Verification (CAV), LNCS, 2014.
 - Control of Linear Switched Systems with Receding Horizon Modal Information. , R. Essick, J.-W. Lee, and G.E. Dullerud. To appear in *IEEE Transactions on Automatic Control*, 2014.
- Training and HR development
 - Zhenqi Huang completed MS thesis in Summer 2013 and is continuing PhD; made several conference presentations
 - Yu Wang finishing MS thesis this summer (2014); will continue to PhD.



Problem (a): CPS adversaries and their analysis

• For analyzing a CPS model under attack we need to be able to compute or approximate the behavior of the system under attack

adversaries

Our work has focused develops and implements algorithms for these approximations

Interacting network of dynamical systems

 Proofs
 from
 Simulations
 and
 Modular

 Annotation
 (HSCC 2014), gives algorithm for compositionally analyzing dynamical system models using numerical simulations. It is sound (always gives right answer) and relatively complete (i.e., terminates whenever system satisfies property robustly).
 It is sound

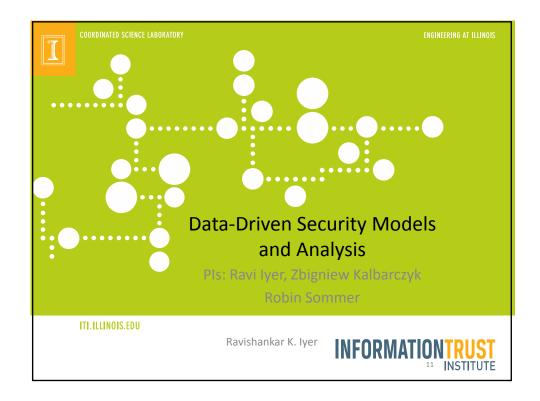
Invariant Verification of Nonlinear Hybrid Automata Networks of Cardiac Cells (CAV 2014), extends HSCC to handle deterministic hybrid models and shows that it is effective for a very challenging benchmark (a network or cardiac cells stimulated by a pacemaker)

Verification of Nonlinear Hybrid
Systems with Simulation Traces
and Compositional Reasoning.
(Zhenqi Huang's MS thesis)
develops algorithms for reasoning
about nondeterministic hybrid
models---which are essential for
capturing large classes of

Information (IEEE TAC 2014), provides <u>exact</u> and automatic analysis and synthesis algorithms for <u>worst-case</u> evaluation of adversary ability to degrade and interrupt metric-based system performance. Adversary can switch or manipulate CPS system parameters and structure; can also inject stochastic and non-stochastic disturbances. Next steps: more detailed statistical adversary models; game-oriented analysis of CPS switching.

Switched Systems with Receding Horizon Modal

Proving Abstractions of Dynamical Systems through Numerical Simulations (HOTSOS 2014) shows that the HSCC ideas can be used to compute the distance between an ideal system and the same system under





Goal and Challenges

Goals

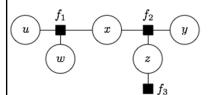
- Represent security incidents using factor graphs
- Create scientifically sound methods for preemptive detection of attacks, i.e., before the system misuse

Challenges

- Operate on a partial knowledge on the attack
- Semantics of event logs difficult to correlate with attacker's actions
 - examining an event in isolation may not be sufficient
- Attackers may enter the target system using stolen credentials

12

An Overview of Factor Graphs



A variable node represents a random variable



 \top A factor node is a function that links random variables and define their relationships. E.g., f2(x,y,z)

A factor graph of five variables (x,y,z,w,u) and three factor functions (f1,f2,f3)

 $P(X) = \frac{1}{Z} \prod_{v \in V} f(v)$

X Random variables

Z Normalization factor

V Cliques (group of variables)

A factor graph is an undirected, bipartite graph representing functional relations between random variables

- Factor graph is an effective representation of complex dependencies between random variables, including both causal and non-causal dependencies
- The joint probability distribution of random variables can be factorized into factor functions

- Factorization simplifies calculation (such as inferring value of hidden variables) on factor graph

Application of Factor Graphs to Attack Detection

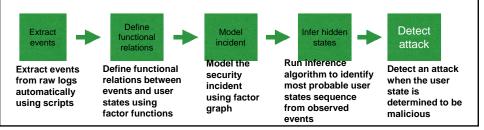
Input

- Set of observable events X, (generated by system monitoring tools, e.g., IDS alerts, netflows, syslogs)
- User profile u (e.g., places user usually logs from, applications user usually runs)
- Set of possible system/user states Y = {benign, suspicious, malicious}

Output

Given an event sequence E infer security state S of the target system/user

Use factor graphs to represent functional relation between the **observed evidence** (the event sequence) and the **hidden system/user states** to determine the most probable sequence state transitions



Factor Graph Representation of an Example Incident at NCSA

Raw log (syslog, IDS alerts, netflows)	Event	User State	
sshd: Accepted <user> from <remote- host> (a user logs into a computing node using ssh)</remote- </user>	login remotely (e ⁰)	s ^o	
HTTP GET vm.c (server6.bad-domain.com) (a source file (vm.c) downloaded from a server)	download sensitive (e ¹)	s¹	
sshd: Received SIGHUP; restarting (the Secure Shell server (SSHd) restarted)	restart system service (e ²)	s²	

Post-incident analysis of attacker actions:

- a) compromise a user account and log in from a remote location,
- b) download, compile, and execute a privilege escalation exploit (CVE-2008-0600)
- inject credential collecting code (to harvest user credentials) into the node's SSHd server,
- d) restart s the SSHd server

A Factor Graph representation of the incident

Variable nodes

(defined based on the data from security/system logs)

- e1: download sensitive
- e2: restart system service
- s¹: user state when observing e¹ s²: user state when observing e²

Factor functions/nodes

(defined based on the data from security/system, knowledge of the system, security experts opinion)

The factor function f₂ can improve detection accuracy by incorporating prior

$$f_1 = \left\{ egin{array}{ll} 1 & \mbox{if} \ e^1 = \mbox{download sensitive} \\ & \& \ s^1 = \mbox{suspicious} \\ 0 & \mbox{otherwise} \end{array}
ight.$$

$$f_2 = \left\{ egin{array}{ll} 1 & ext{if} & e^2 = restart \ service} \& \ s^1 = suspicious} \& \ s^2 = malicious} \ 0 & otherwise \end{array}
ight.$$

 $f_3 = \left\{ egin{array}{ll} 1 & ext{if} \ e^2 = ext{restart sys service} \\ & \& \ s^2 = benign \\ 0 & otherwise \end{array}
ight.$

Note: Unlike Bayesian Networks which require pairwise conditional dependency, factor graphs can combine a group of two or more variables and has lower computational complexity, e.g., function f2)

Output of Factor Graph Based Analysis for the Example Incident at NCSA

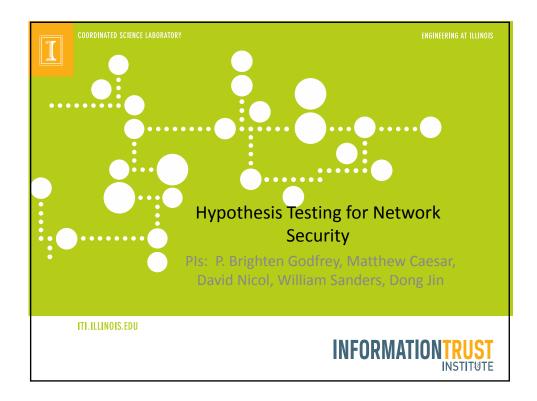
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sshd: Received SIGHUP; restarting (the Secure Shell server (SSEd) restarted)	restart system service (e²)	s ² = malicious	$\left(e^{0} ight)$	$\begin{pmatrix} e^1 \end{pmatrix}$	$\begin{pmatrix} e^2 \end{pmatrix}$		
Most probable sequence of system/user state transitions determined based on the Factor Graph							

- · Benefits of the proposed approach
 - Early detection and prevention of attacks
 - · when a user is suspicious, additional monitors can be enabled for monitoring
 - Timely identification of users' security state in a large enterprise network
 - based on output user state sequence, a security analyst can identify most suspicious users and progress of potential attacks without having to examine a large amount of raw logs

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Impact on Science of Security

- A new trend of thinking: building a common sensing infrastructure for reliability and security monitoring
- A new type of intrusion detection system that detects attacks from initial stage, using only partially observed traces.
- A theoretical foundation of orthogonal monitoring framework, where monitoring data from all layers of a system can be combined to make more accurate attack detections.





Project Goals

- Increasingly complex, large scale nature of networks makes it difficult to <u>understand</u> network's behavior
- Even very simple behaviors (e.g., packet reachability) are difficult for operators to test
 - Synthesizing simple behaviors into high-level qualitative understanding has been beyond reach
- Our work: <u>Network Hypothesis Testing Methodology</u>
 - Analysis methodology needed to support scientific reasoning about the security of networks
 - Focus on information and data flow security
 - Adds to both theoretical underpinnings and practical realization of Science of Security



Project Goals

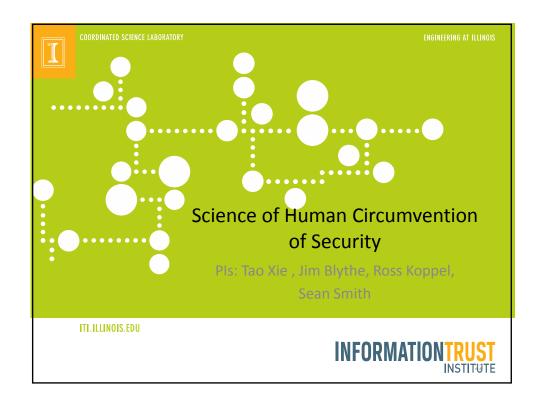
- Example: App-oriented access control policies in enterprise or mission-oriented network. Network designer needs to test various hypotheses:
 - All network paths traverse a firewall
 - Drone Operator will with 99% confidence be assured at least
 50% of bandwidth for connection with drone
 - Fewer than 5% of stepping stone attacks that use no more than 3 compromised hosts could allow for data leakage from database DB
- Key challenges:
 - Scalability: large amounts of fast-changing network data
 - Modeling: policies' cross-layer interactions, possible subversion

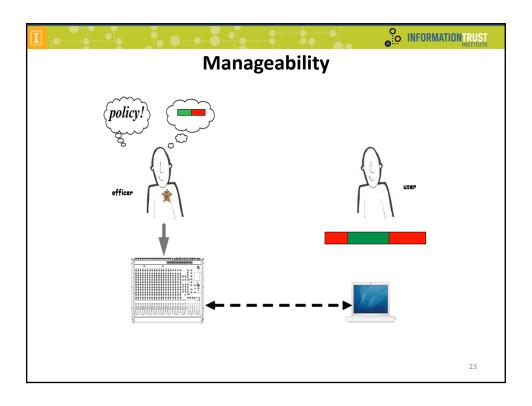
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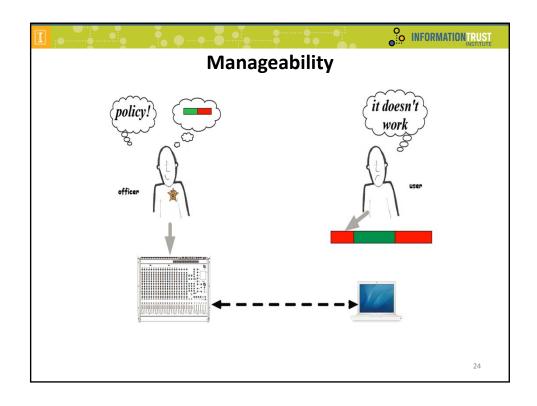


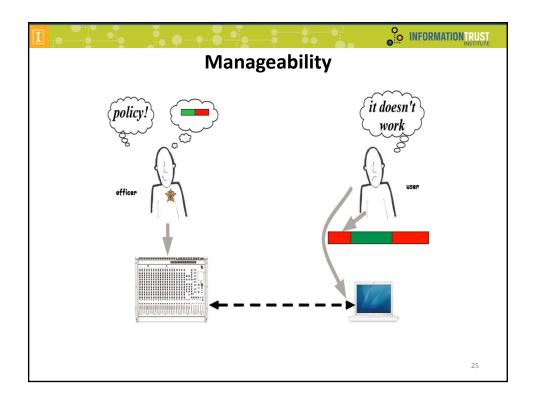
Project Approach

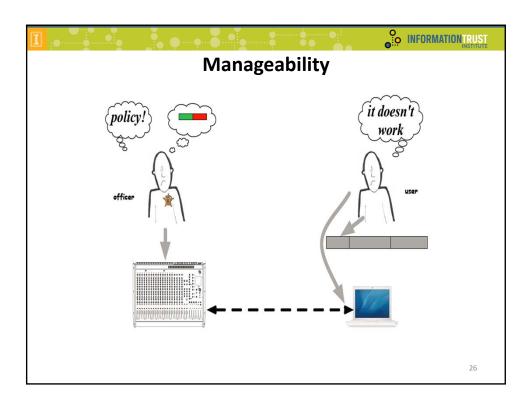
- Our approach: Network Hypothesis Testing
 - General formal model representing diversity of multi-layer network behavior and testing formal hypothesis against that model, resulting in quantitative metrics
 - Leverages results from two of our SoS lablets
 - "Towards a Science of Securing Network Forwarding" [Godfrey, Caesar] developed real-time Data Plane Verification, found real-world vulnerabilities
 - "Quantitative Assessment of Access Control in Complex Distributed Systems" [Nicol, Sanders] developed mathematical techniques to assess completeness of firewall protection
- Key tasks:
 - 1. Develop metrics for expressing quantitative hypotheses on network behavior
 - 2. Develop techniques to <u>compose and evaluate metrics</u> across system components
 - Enhance core evaluation methodologies to <u>scale to large and complex</u> systems
 - 4. Formulate a <u>hypothesis language</u> and automatically design experiments to test hypotheses within the language

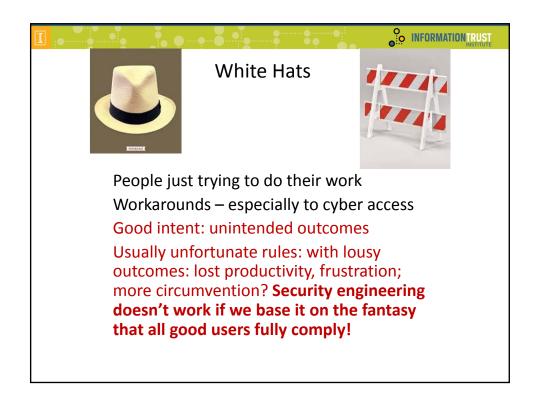


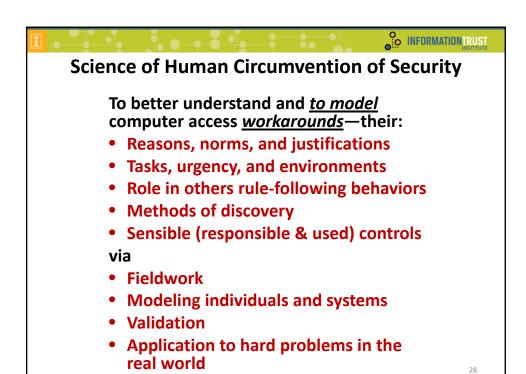


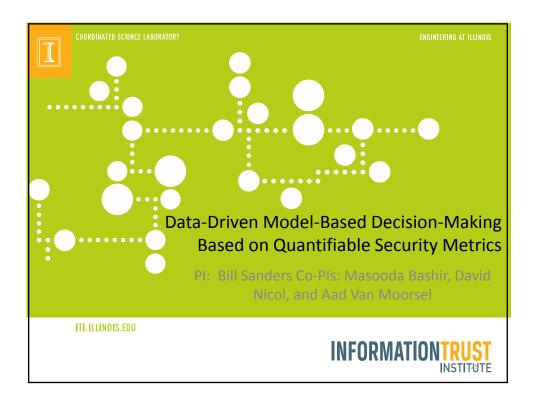














Project Goals

Hypothesis: Quantitative security models, augmented by collected data, can be used to make credible business decisions about the use of particular security technologies to protect an organization's infrastructure.

Goals:

- Develop quantitative, scientifically grounded, decisionmaking tools to guide information security investments in private or public organizations, combining human and technological concerns
- Demonstrate their use in two or more real-life case studies
- Validate the usefulness of the developed tools through analysis of the results of the completed case studies

30



Research Strands

- Stochastic modeling of human-behavioral aspects that influence the use of security policies in a private or public sector organization (led by Illinois)
- Development of a data collection strategy that is optimal with respect to a combination of factors, including sensitivity of the model to the data, statistical accuracy of the model inputs/outputs, and the cost of collecting the data (led by Newcastle Univ.)
- Validation of our project hypothesis, through a series of case studies (together)



Project Timeline

Year 1:

- A complete implementation of HITOP in Mobius. This work was started as part of a project in
 the original UIUC lablet, and a substantial portion of the development has been completed.
 The completed work from that project will be the starting point for this project, enabling us to
 focus on the construction of the quantitative models themselves, their integration with
 security data collection strategies.
- Design of a generalized implementable data collection optimization algorithm which makes
 use of statistical techniques to quantify the amount of data that needs to be collected, and
 the sensitivity of each data value on the with respect to the security measures of interest.

Year 2:

- A prototype software tool development to assist data collection strategy formulation.
- Initial case study to test efficacy of approach. (3) Refined modeling formalisms and data collection algorithms, based on results of initial case study

Year 3

- Distributable tool for use by others.
- Two or more case studies on cyber chosen infrastructure or critical infrastructure systems to validate our hypothesis.
- Tool refinements based on use by us and others