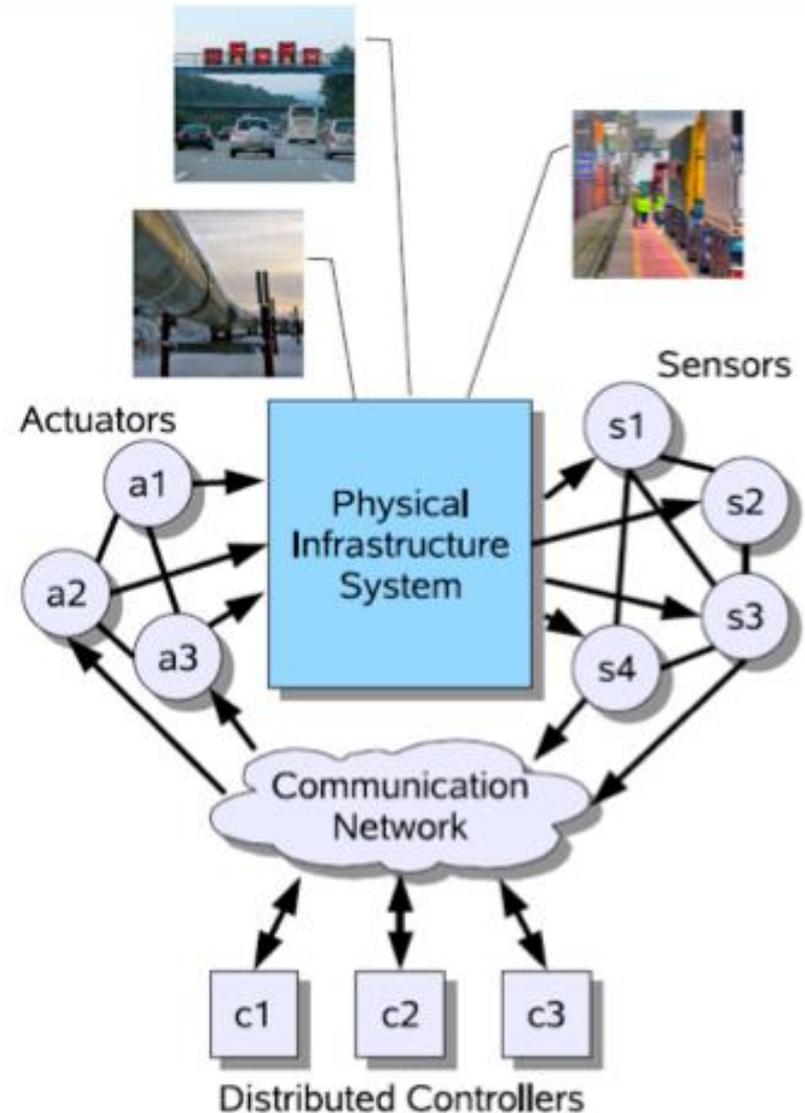


# Securing Cyber-Physical Systems

Alvaro Cárdenas  
Fujitsu Laboratories

Ricardo Moreno  
Universidad de los Andes

- Control
- Computation
- Communication
  
- Interdisciplinary Research!
  
- Example: Smart Grid



## ■ Attacks

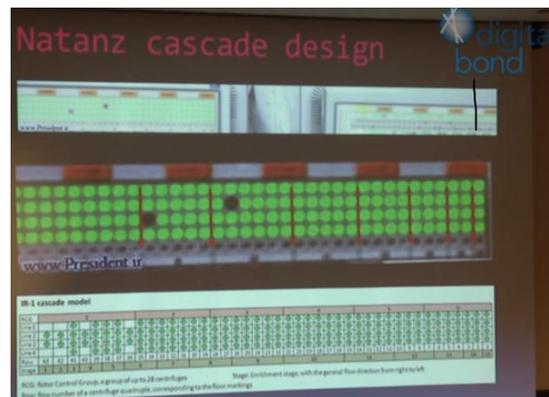
### ■ Maroochy Shire 00



### ■ HVAC 12



### ■ Stuxnet 10



## ■ Threats

Obama Adm Demonstrates In Feb. 2012 attack to power Grid

### US Video Shows Hacker Hit on Power Grid

US Video Shows Potential Destruction Caused by Hackers Seizing Control of Electrical Grid



In this image from video released by the Department of Homeland Security, smoke pours from an expensive electrical turbine during a March 4, 2007, demonstration by the Idaho National Laboratory, which was simulating a hacker attack against the U.S. electrical grid. (AP Photo/Dept. of Homeland Security)

By **TED BRIDIS** and **EILEEN SULLIVAN**  
Associated Press Writers  
WASHINGTON Sep 27, 2007 (AP)

The Associated Press

A government video shows the potential destruction caused by hackers seizing control of a crucial part of the U.S. electrical grid: an industrial turbine spinning wildly out of control until it becomes a smoking hulk and power shuts down.

Font Size  
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### Your Opinion

#### Comment & Contribute

WHAT OTHERS ARE SAYING 44 Comments

I agree with the person(s) about the old...

locoyoco2 Sep-27

'seems for every smart program there is...

OFEARTHLYGOOD Sep-27

scare tactics,Someone already pointed out,...

tmarabu Sep-27

COMMENT

#### Post Video

TALK Connect with Newsmakers

- Vulnerabilities are increasing
  - Sensors/Controllers are now computers (can be programmed for general purposes)
  - Networked (remotely accessible)
  - By necessity, billions of low-cost embedded devices
  - Physically insecure locations
- Attacks will continue to happen
  - Devices deployed for ~ 20-30 years



ABOUT THIS BLOG

## FBI: Smart Meter Hacks Likely to Spread

39  
tweets  
retweet

A series of hacks perpetrated against so-called “smart meter” installations over the past several years may have cost a single U.S. electric utility hundreds of millions of dollars annually, the **FBI** said in a cyber intelligence bulletin obtained by

Advertisement



## ■ Short Term

- Incentives
- Software reliability
- Solve basic vulnerabilities

## ■ Medium Term

- Leverage Big Data for Situational Awareness

## ■ Long Term Research

- Resilient estimation and control algorithms

- *“Making a strong business case for cybersecurity investment is complicated by the **difficulty of quantifying risk in an environment of rapidly changing, unpredictable threats with consequences that are hard to demonstrate**”*
  - DoE Roadmap
- Governments are responsible for Homeland Security, and critical infrastructure security
  - Utilities are not (outside their budget/scope?)
  - Problem:
    - Interdependencies (e.g., cascading failures)
    - It doesn't matter if one utility sets an example because this is a weakest security game
  - Nations have much more to lose from an attack than utilities

- Vendors of equipment for managing control systems have few incentives for secure development programs because customers are not requesting them
- Asset owners need to request vendors secure coding practices, hardened systems, and quick response when new vulnerabilities and attack vectors are identified
- American Law Institute (ALI)
  - Principles of the Law of Software Contracts (2009)
  - Vendors liable for knowingly shipping buggy software
  - Implied warranty of no material hidden defects (non-disclaimable)
  - Software for CIP can be first use case
- Currently congress is debating how to give incentives for asset owners to invest in security
  - Cybersecurity Act 2012 (increase regulation)
  - SECURE-IT Act 2012 (increase data sharing)

## ■ Short Term

- Incentives
- Software reliability
- Solve basic vulnerabilities

## ■ **Medium Term**

- Leverage Big Data for Situational Awareness

## ■ Long Term Research

- Resilient estimation and control algorithms

- Push back in prices
  - Billions of low-cost embedded devices
  - Can't have fancy tamper protection
- Security is hard to see
  - Hard to see advantages of hardening devices
- But, Situational Awareness is Fun to see
  - Understand the health of the system
    - Routing protocol, health of the system
  - Identify anomalies
- Big Data is new in Smart Grid
  - Redundancy
  - Diversity
  - Data Analytics to identify suspicious behavior

# Big Data Analytics in Smart Grid



EVENT

HOME APPLE BROADBAND CLEANTECH CLOUD COLLABORATION MEDIA



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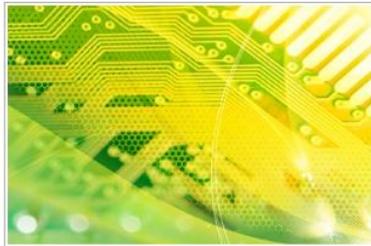
## When big IT goes after big data on the smart grid

By Adam Lesser | Mar. 20, 2012, 10:49am PT | No Comments

Tweet 136 | Share 37 | Like 7 | +1 6

This article originally appeared on *GigaOM Pro*, our premium research service (subscription required).

With many utilities facing the task of storing **petabytes** of smart meter data for as long as seven years in order to satisfy regulatory requirements, the ability to house and leverage the massive load of



## Big Data Offers Big Value for Utilities

03 Apr 2012 | United States

Share this

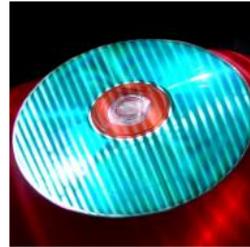
Smart meters produce data – it takes work to make the data ‘smart.’

### What happened

[Adam Lesser of Gigaom](#) wrote about the difficulties faced by utilities when dealing with “big data” and the opportunities that this offers to IT companies.

According to Lesser, utilities face petabytes of data that needs to be stored for up to seven years to comply with regulation. Not only that, these utilities also need to “mine” this data and be able to pull out useful information, in a usable format, to allow them to save the time and money promised when deploying smart meters. In other words, make data ‘smart.’ This poses a “significant IT challenge,” one that is new to utilities.

In his report, “[Smart Grid Billing Outlook 2012-2016](#),” author Danny Dicks says, “While smart meter deployments have been growing steadily over the last 3-4 years, utilities’ IT system priorities have been focused on preparing for how to deal with large volumes of smart meter data. This year we expect to see the emphasis change towards making use of that data – to develop innovative tariffs and new services ... All this will require changes to traditional billing systems and CISs.”



No end of possibilities for the fearless, forward thinking and imaginative.



## Big Data Management for Energy and Smart Grid - Creating the Real-Time Utility Enterprise

Thursday, April 5, 2012 from 5:00 PM to 9:00 PM (PT)  
Mountain View, United States



Carnegie Mellon University  
Silicon Valley

Ticket Information				
TYPE	REMAINING	END		QUANTITY
General	Sold Out	Ended	Free	Sold Out

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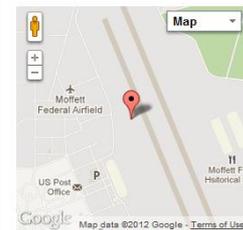
#### Producer



#### Main Sponsor



### When & Where



Building 152  
Moffett Field, Hwy 101  
Mountain View, 94035

Thursday, April 5, 2012 from 5:00 PM to 9:00 PM (PT)

Add to my calendar

- Fujitsu is chairing the working group
- Please consider contributing



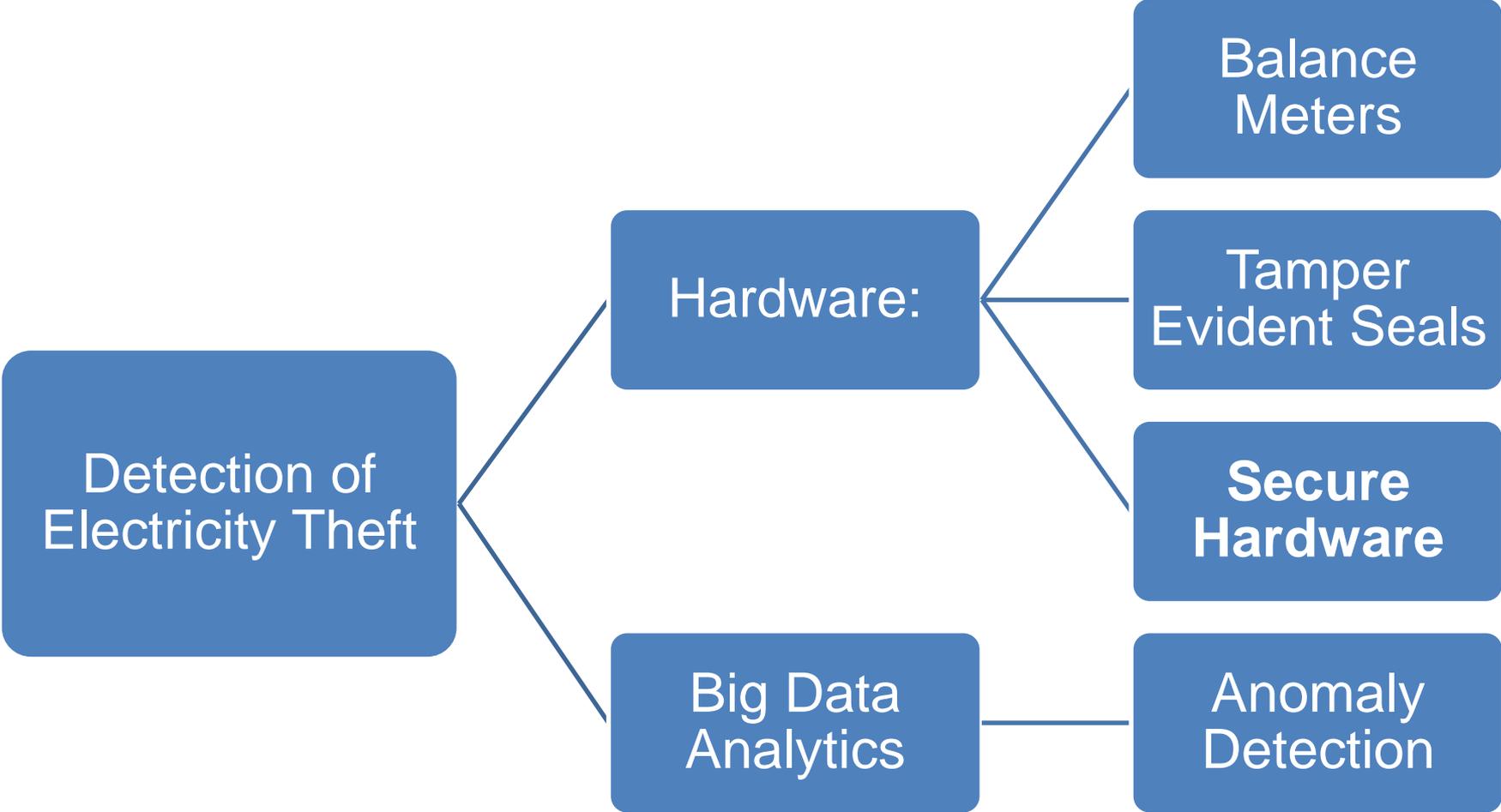
## Big Data Working Group

Proposed Charter

---

April 2012

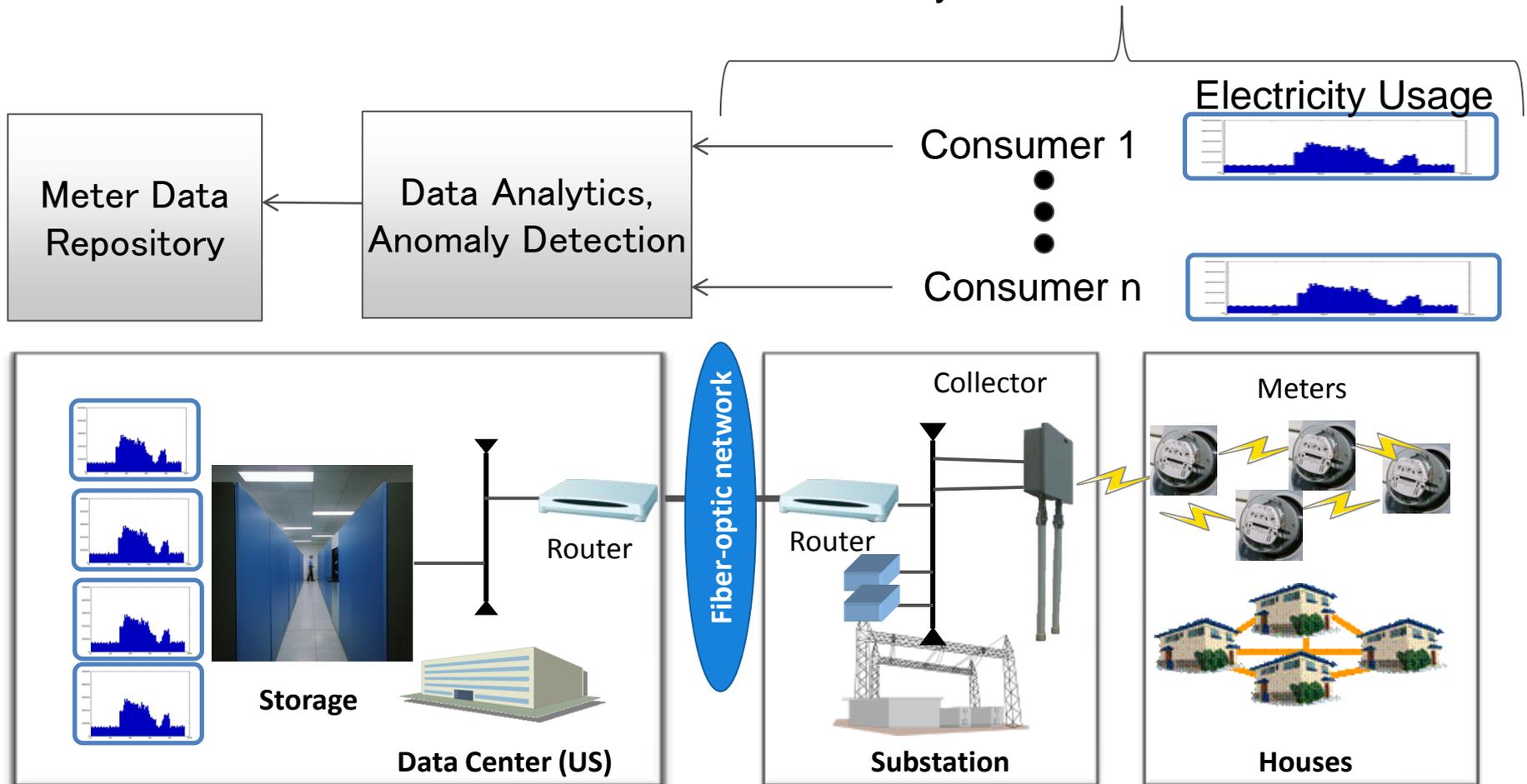
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[Mashima, Cardenas. Submitted to RAID, 2012]

# Big Data Analytics to Identify Fraud

**AMI:** Advanced Metering Infrastructure. Smart Meters send consumption data frequently (e.g., every 15 minutes) to the utility

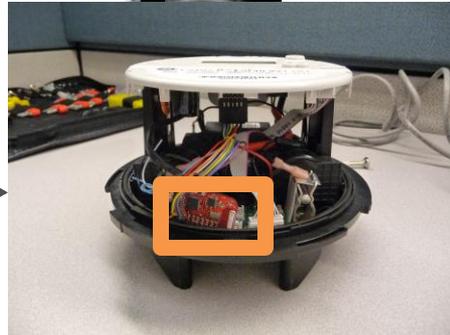
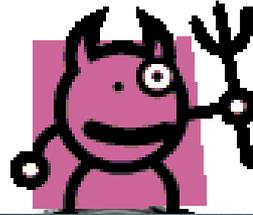


# Adversary Model

$f(t)$

Real Consumption

$$Y_1, \dots, Y_n$$

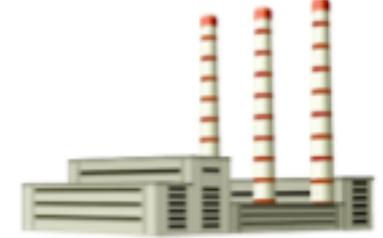


$a(t)$

Fake Meter Readings

$$\hat{Y}_1, \dots, \hat{Y}_n$$

Utility



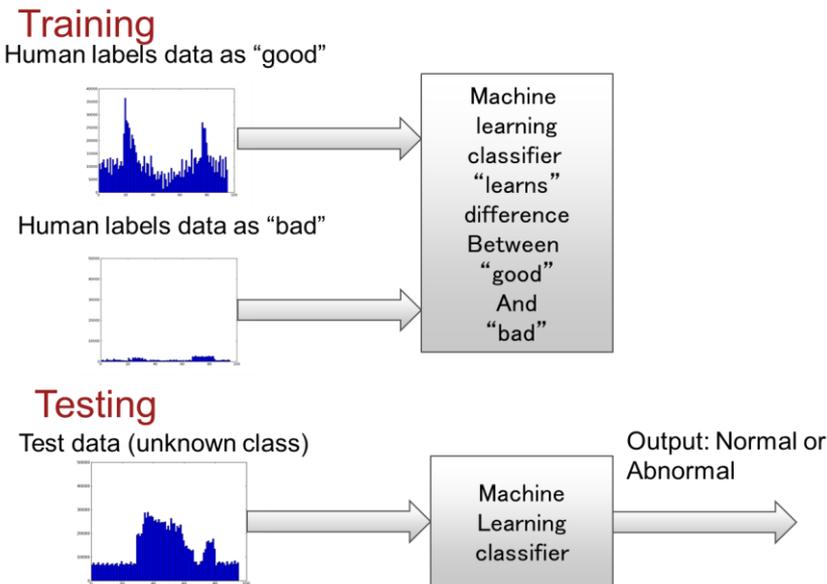
Goal of attacker: Minimize Energy Bill:

$$\min_{\hat{Y}_1, \dots, \hat{Y}_n} \sum_{i=1}^n \hat{Y}_i$$

Goal of Attacker: Not being detected by classifier "C":

$$C(\hat{Y}_1, \dots, \hat{Y}_n) = \text{normal}$$

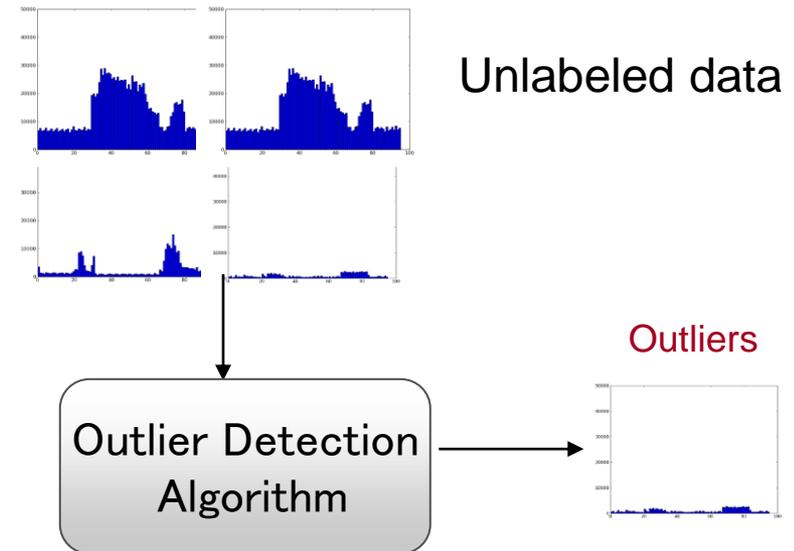
## Supervised Learning



### ■ Problems

- It is not easy to get "Attack" data
- A classifier trained with attack data might not be able to generalize to new "smart" attacks

## Unsupervised Learning



### ■ Problems

- Easier to attack
- More false positives
- E.g. Local Outlier Factor (LOF) did poorly in our tests

- We only have “good” data
  - Do not assume we have access to “attack” data
  - Train only one class (“good” class)
- We have prior knowledge of **attack invariant**
  - We know attackers want to lower energy consumption
  - Include this information for the “bad” class
- **Composite Hypothesis Testing** formulation:

$$H_0 : P_0$$

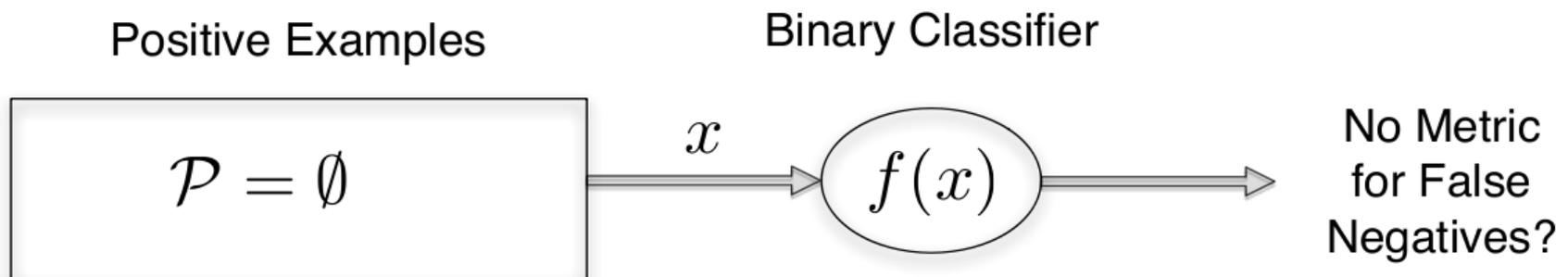
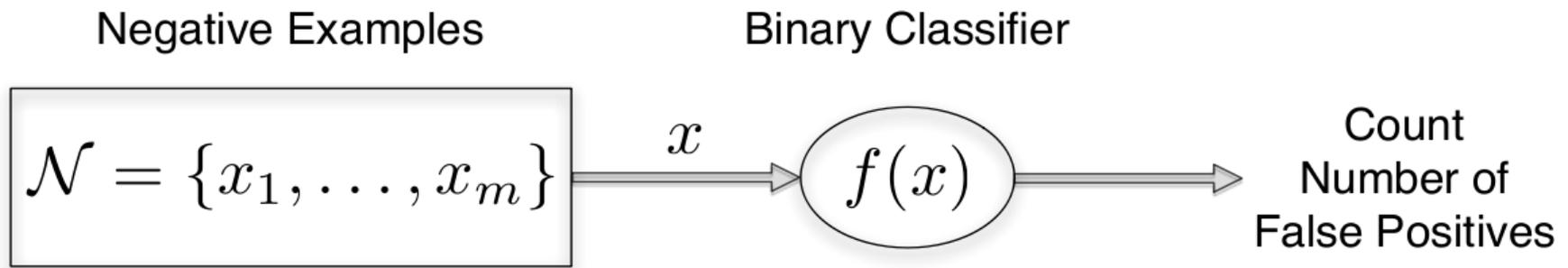
$$H_1 : P_\gamma \text{ s.t. } \mathbb{E}_\gamma[Y] < \mathbb{E}_0[Y]$$

$$Y_{k+1} = \sum_{i=1}^p A_i Y_{k-i} + \sum_{j=0}^q B_j (V_{k-j} + \theta)$$

under  $H_0 : \theta = 0$  and under  $H_1 : \theta = -\gamma, \gamma > 0$ .

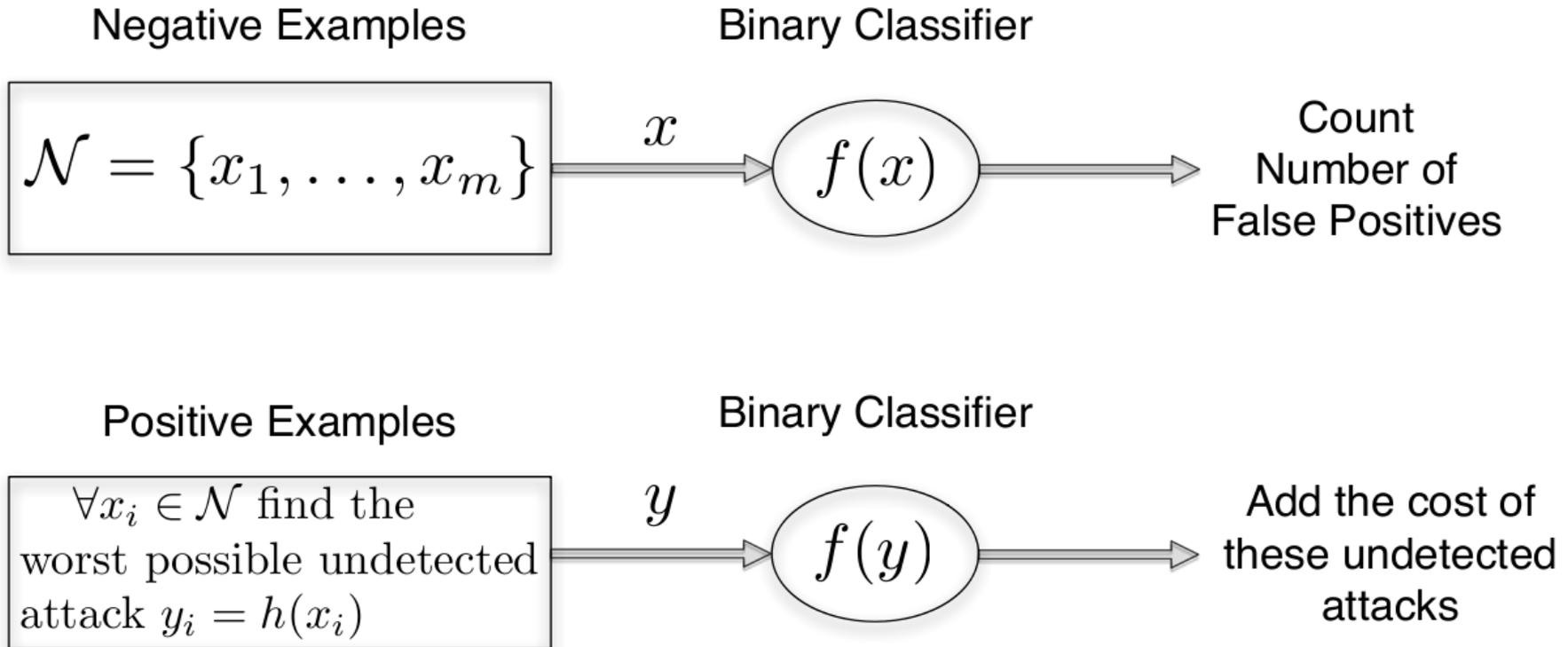
# Problem: We Do Not Have Positive Examples

- Because meters were just deployed, we do not have examples of “attacks”



# Our Proposal:

- Find the worst possible undetected attack for each classifier, and then find the cost (kWh Lost) of these attacks

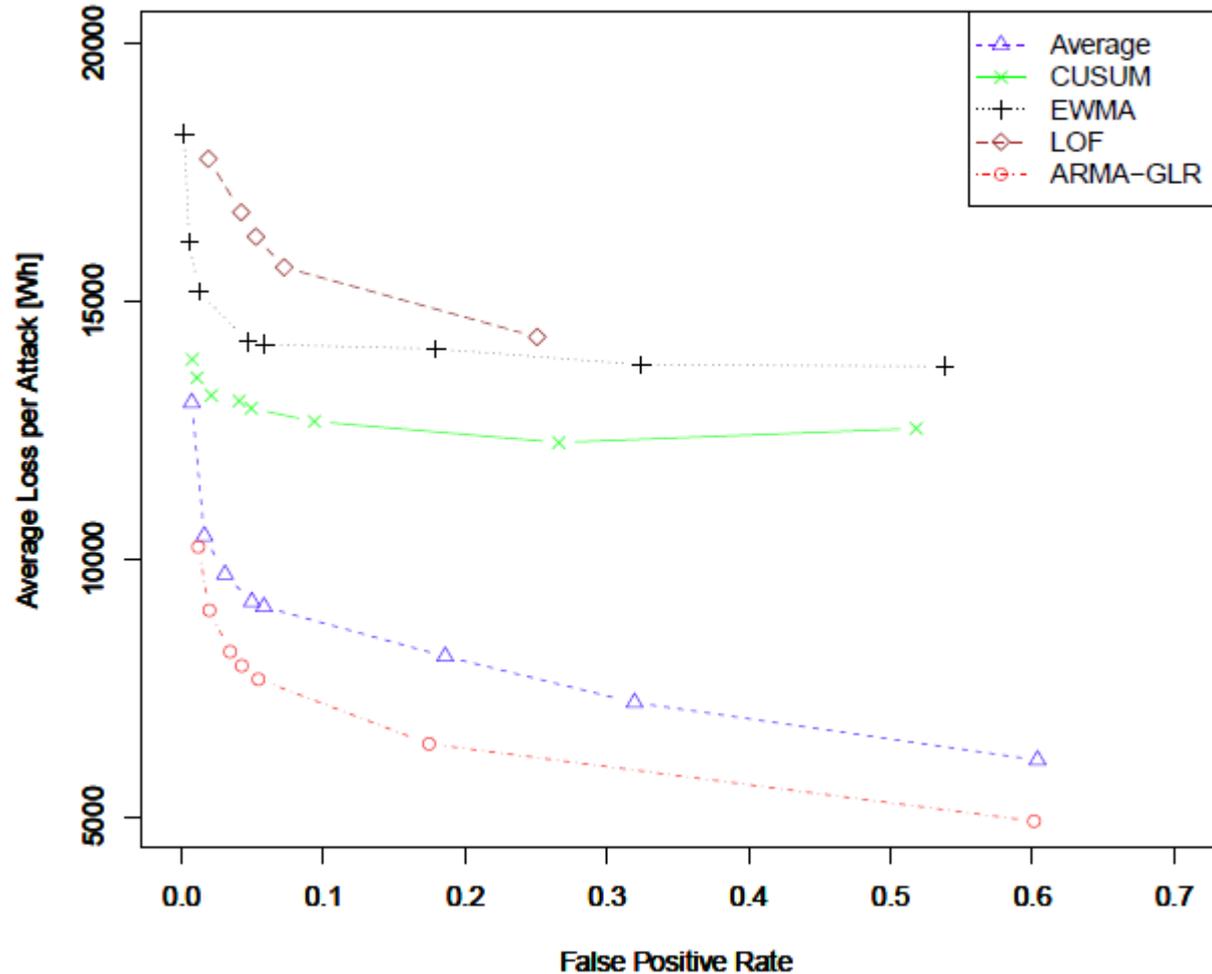


■ We tried many anomaly detectors

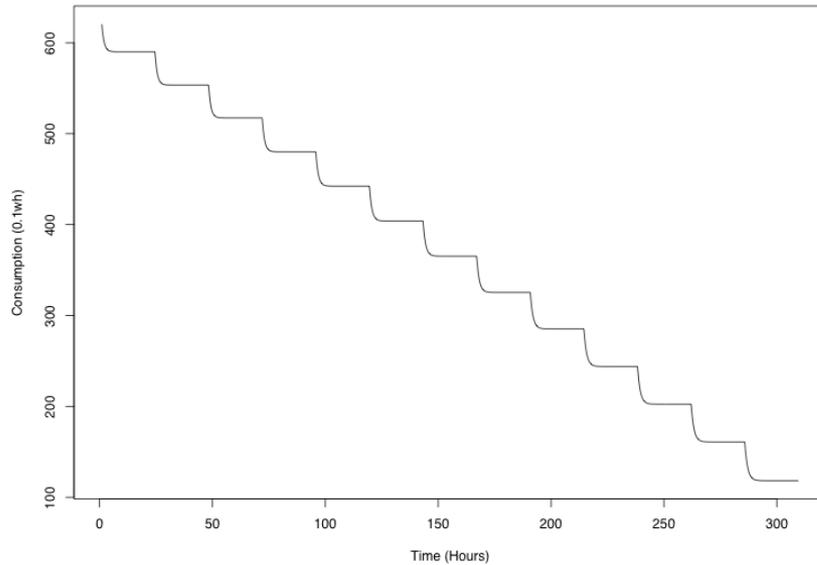
- Average
- CUSUM
- EWMA
- LOF
- ARMA-GLR

■ ARMA GLR is the best detector:

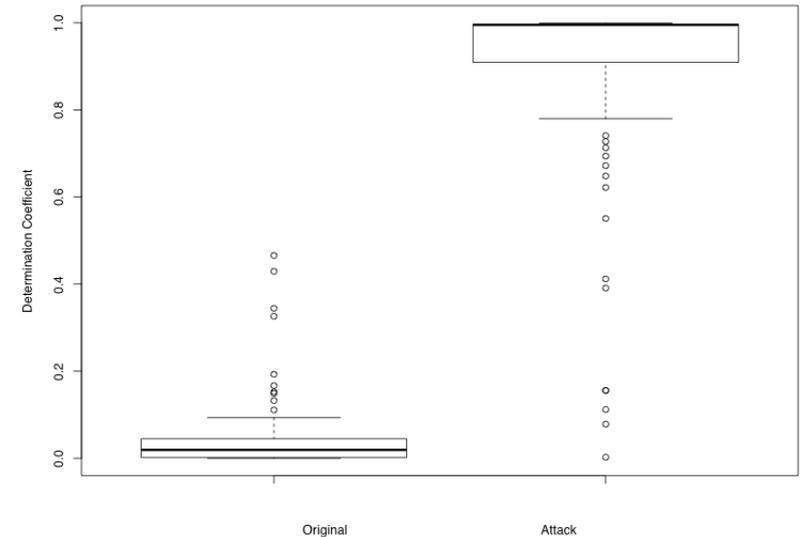
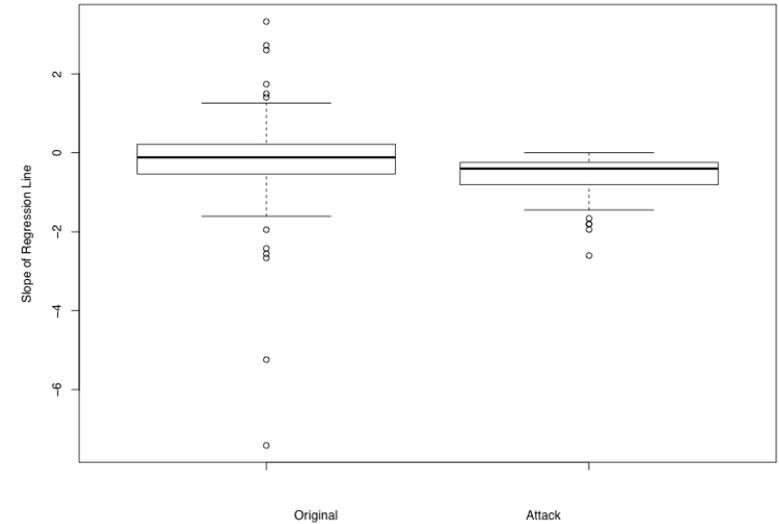
- For the same false positive rate, it minimizes the ability of an attacker to create undetected attacks



# Preventing Poisoning Attacks

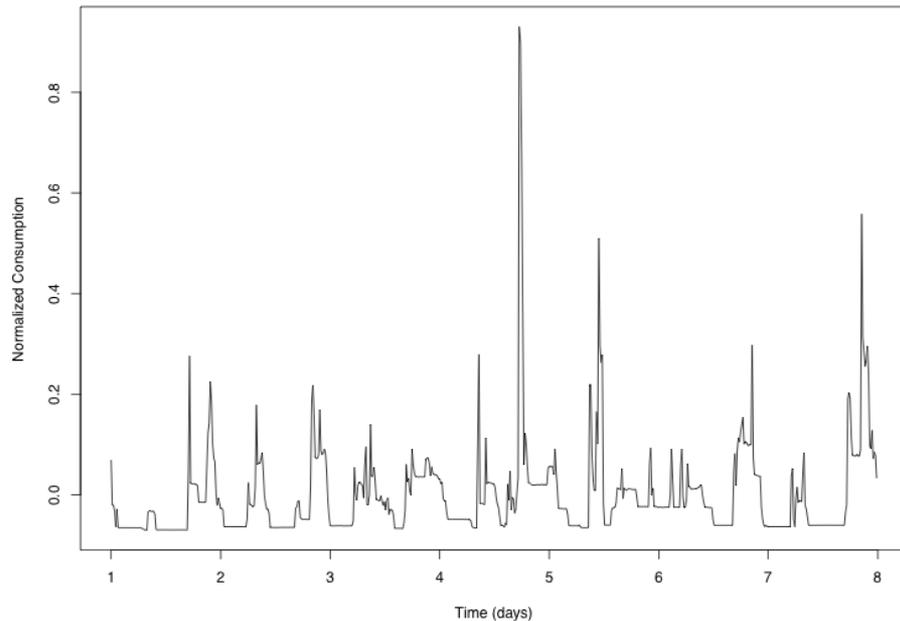


- Electricity consumption is a non-stationary distribution
- We have to “retrain” models
- Attacker might use fake data to mislead the classifier

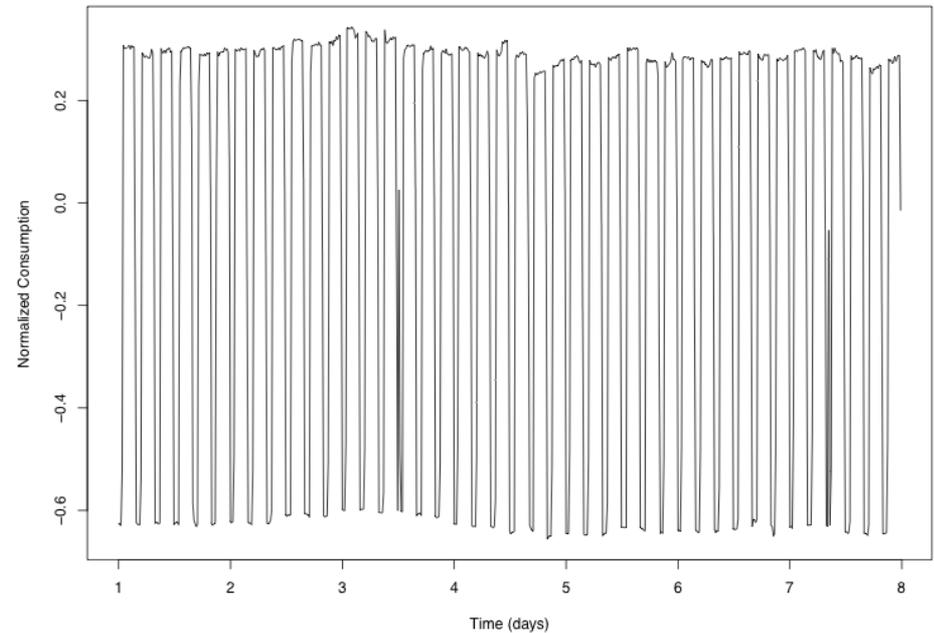


- Use in production system, experience and feedback
- Detecting other anomalies.

## Normal Consumption Profile



## Abnormal Consumption Profile



## ■ Short Term

- Incentives
- Software reliability
- Solve basic vulnerabilities

## ■ Medium Term

- Leverage Big Data for Situational Awareness

## ■ **Long Term Research**

- Resilient estimation and control algorithms

# Previous Work in Security: What can Help in Securing CPS?

## ■ Prevention

- Authentication, Access Control, Message Integrity, Software Security, Sensor Networks

## ■ Detection

## ■ Resiliency

- Separation of duty, least privilege principle

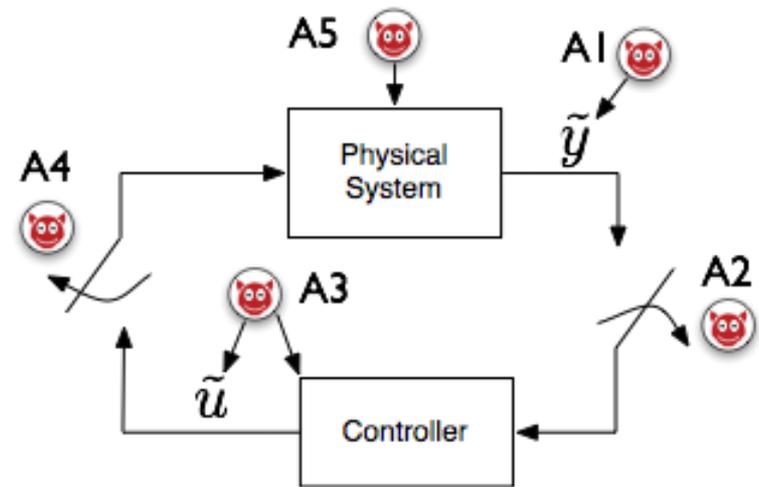
## ■ Incentives for vendors and asset owners to implement security best practices

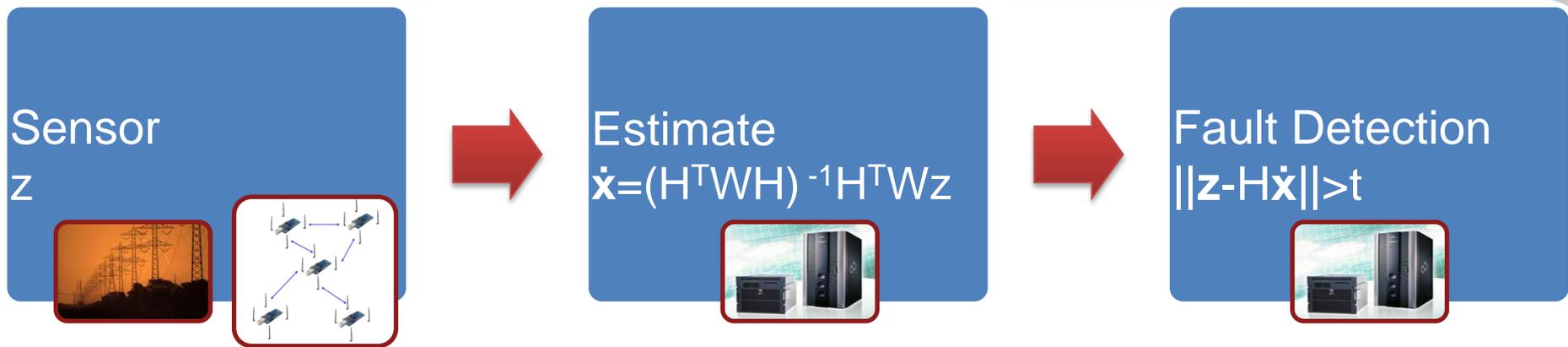
# Previous Work in Security: What is Missing for Secure CPS?

- What is new and fundamentally different in control systems security?
- Model interaction with the physical world
  - How can the attacker manipulate the physical world?

## ■ Attacks to Regulatory Control

- A1 and A3 are deception attacks: the integrity of the signal is compromised
- A2 and A4 are DoS attacks
- A5 is a physical attack to the plant





## ■ Fault-Detection Algorithms do not Work Against Attackers

- Liu, Ning, Reiter. CCS 09

## ■ Attacks are different than failures!

- Non-correlated, non-independent, etc.

## ■ Their study is missing:

- **Impact (risk assessment) of attacks?**

- **Countermeasures?**

# CPS security is different from IT and Control Systems Safety/Fault Detection

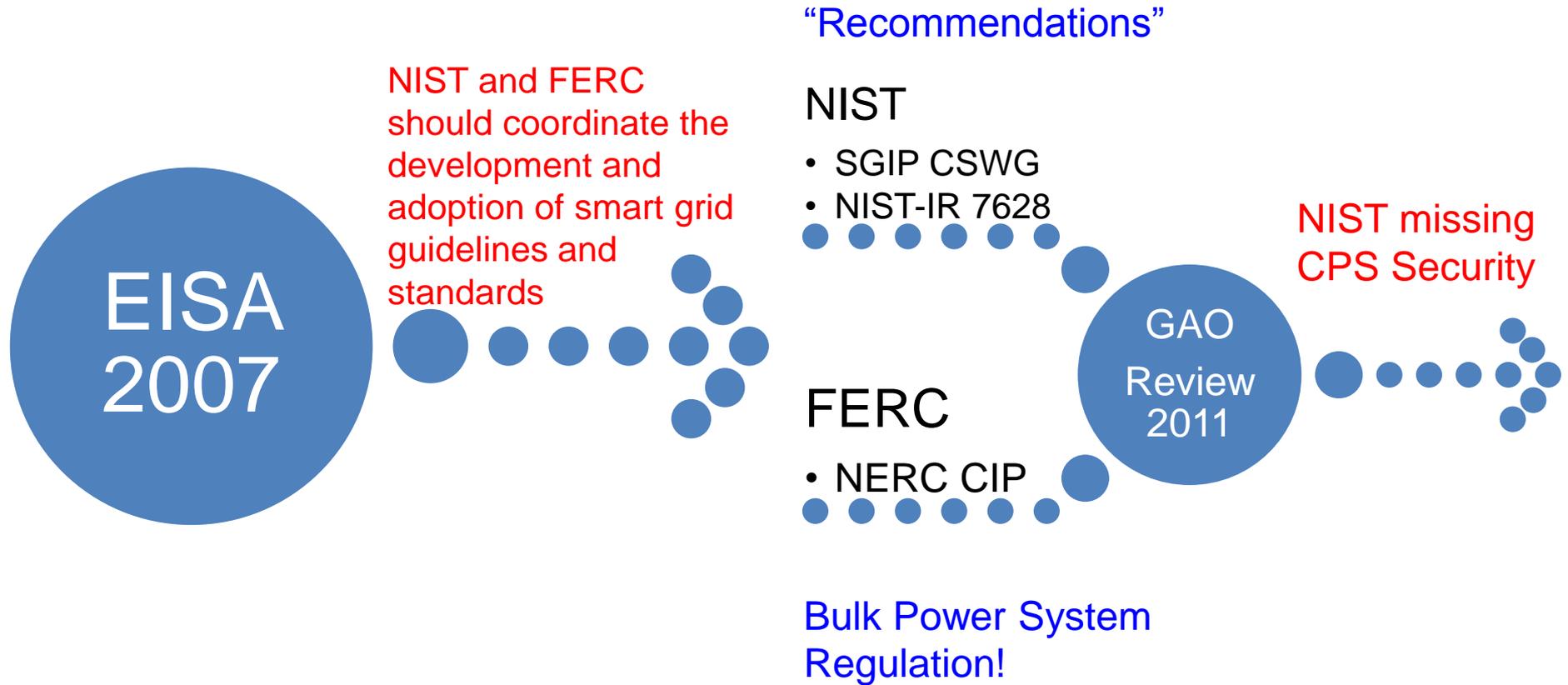
- So security is important; but are there new research problems, or can the problems be solved with
  - Traditional IT security? AC, IDS, AV, Separation of duty, least priv. etc.
  - Control Algorithms? Robust control, fault-tolerant control, safety, etc.
- Missing in IT Security
  - Understanding effects in the physical world
  - Attacker strategies
  - Attack detection algorithms based on sensor measurements
  - Attack-resilient estimation and control algorithms
- Missing in Control
  - Realistic attack models
  - Failures are different from Attacks!
    - Liu et.al. CCS 09, Maroochy, Stuxnet, etc.
- **Argument: Robust Control + IT Security => Resilient CPS**

- **Threat assessment:**
  - How to model attacker and his strategy
  - Consequences to the physical system
- **Attack-resilient** control algorithms
  - CPS systems that degrade gracefully under attacks
- **Attack-detection** by using models of the physical system
  - Study stealthy attacks (undetected attacks)
- **Big Data Analytics**
  - Situational awareness
- **Privacy**
  - Privacy-aware CPS algorithms

Papers articulating new research for CPS security

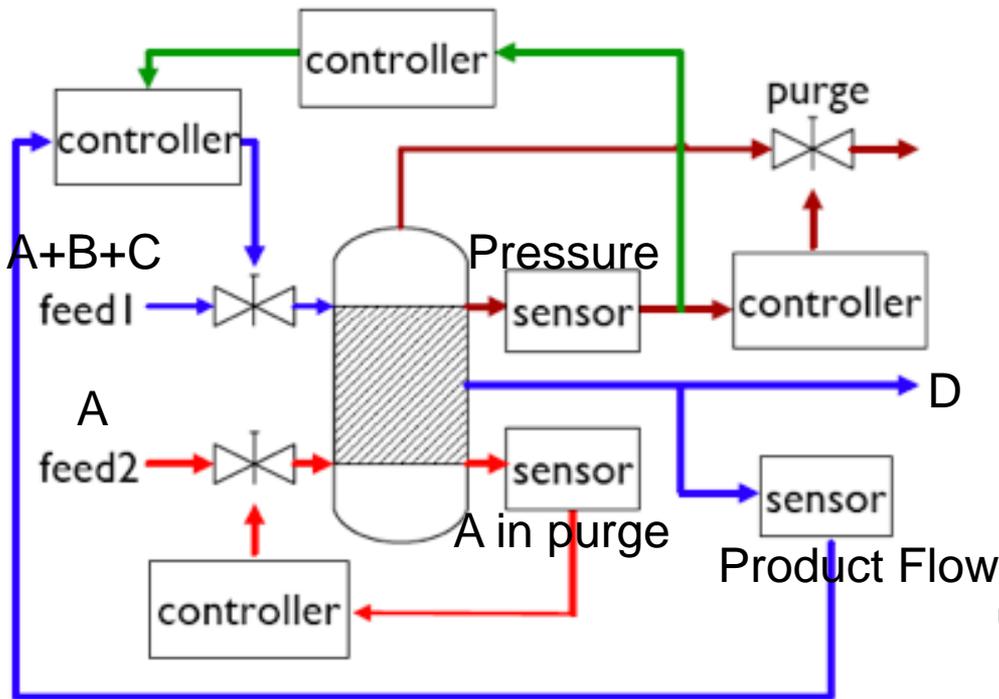
Cardenas, Amin, Sastry, HotSec 08, & ICDCS Workshop (08)

# GAO Agrees: We Need new Research for CPS Security



# Requirements for Secure Control

- Step 1: Threat Model/Assessment
  - Identify requirements
- Traditional Security Requirements: CIA (Confidentiality, Integrity, Availability)
  - What are the requirements of secure control?



## ■ Safety Constraint:

- Pressure < 3000kPa

## ■ Operational Goal:

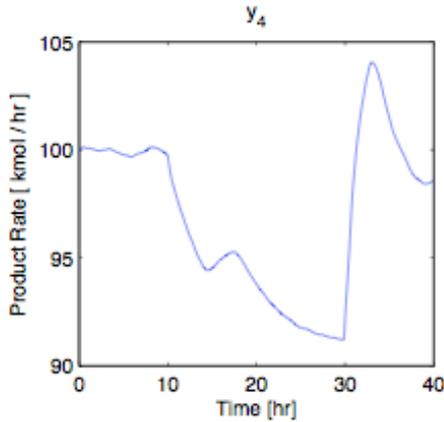
### ■ Cost:

- Proportional to the quantity of A and C in purge,
- Inversely proportional to the quantity of the final product D

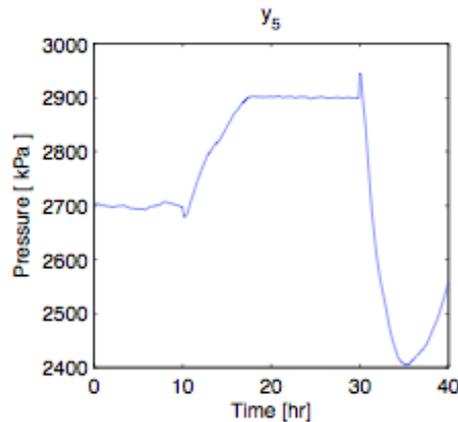
$$\text{Cost} = \frac{F_3}{F_4} (2.206y_{A3} + 6.177y_{C3})$$

# Not all Compromises affect Safety

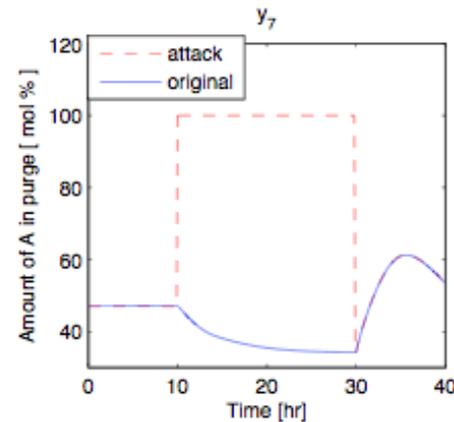
## Production



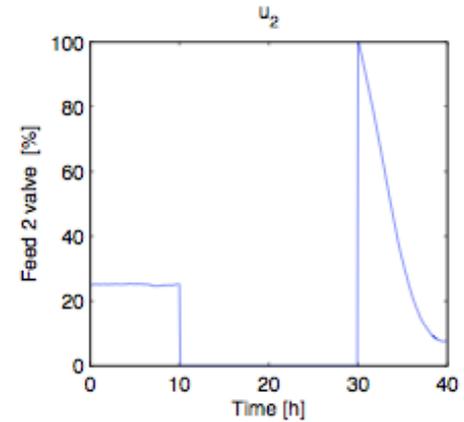
## Pressure



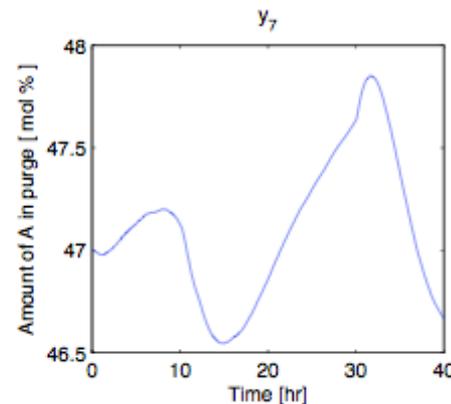
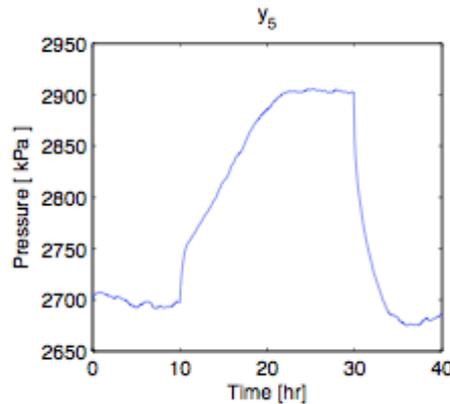
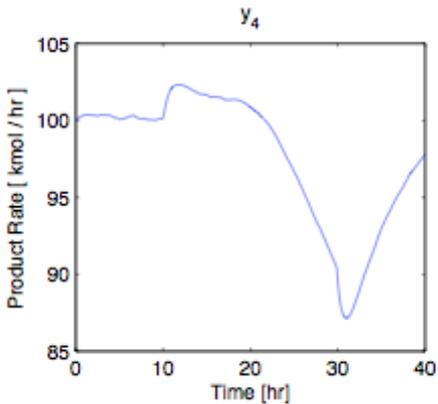
## A in Purge



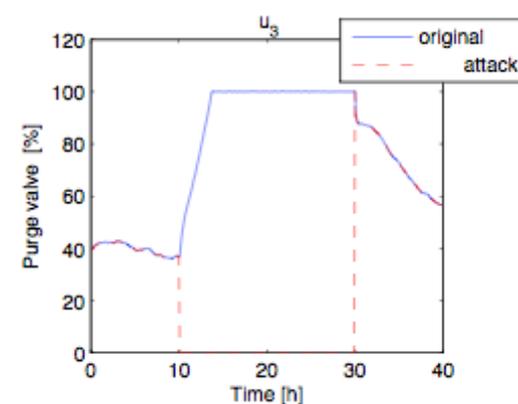
## Feed of A



## Resilient by Redundancy:

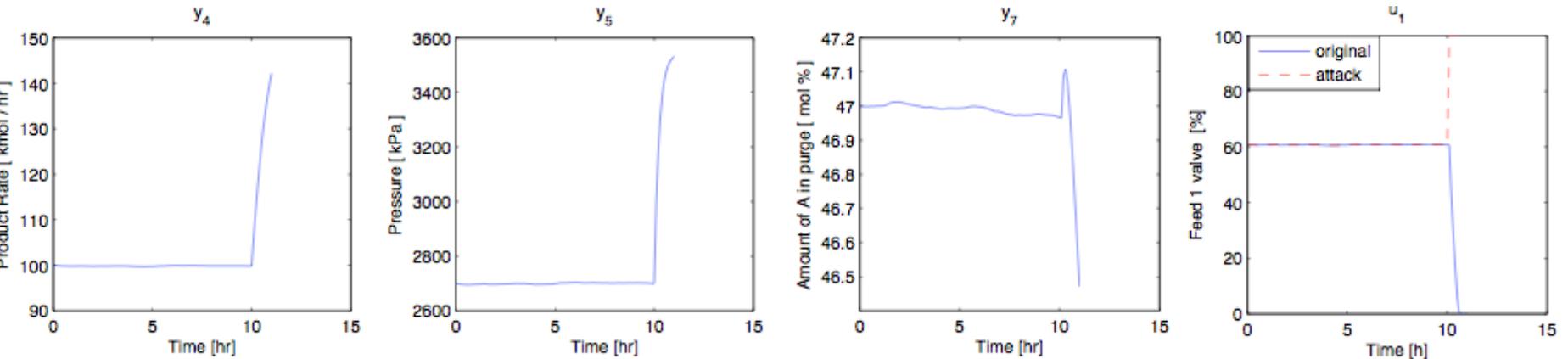


## Purge Valve

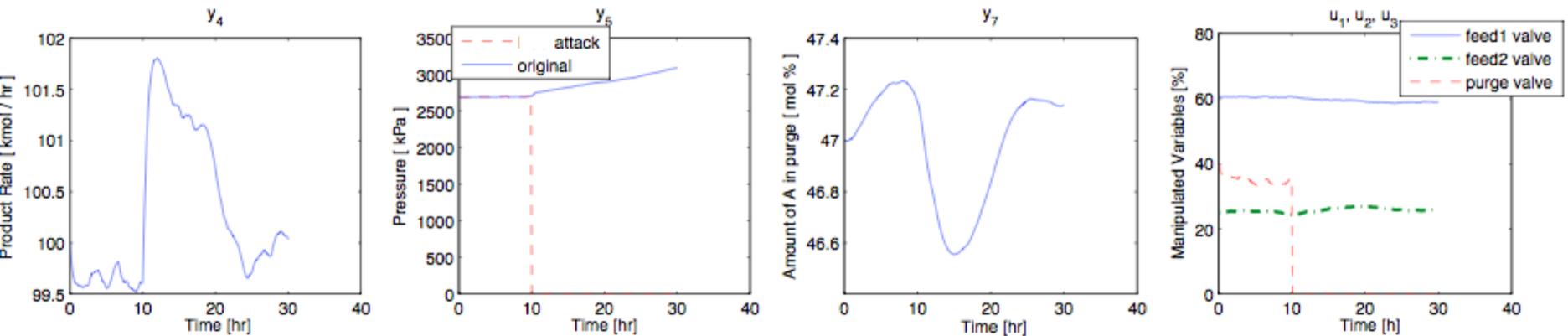


# Safety can be Compromised at Different Time Scales

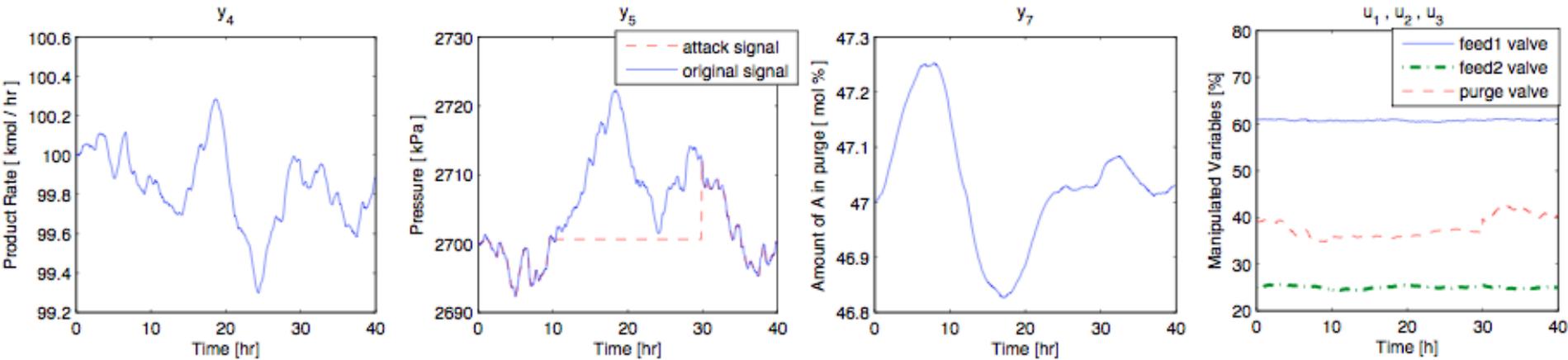
Prioritize protection of control signal for A+B+C feed



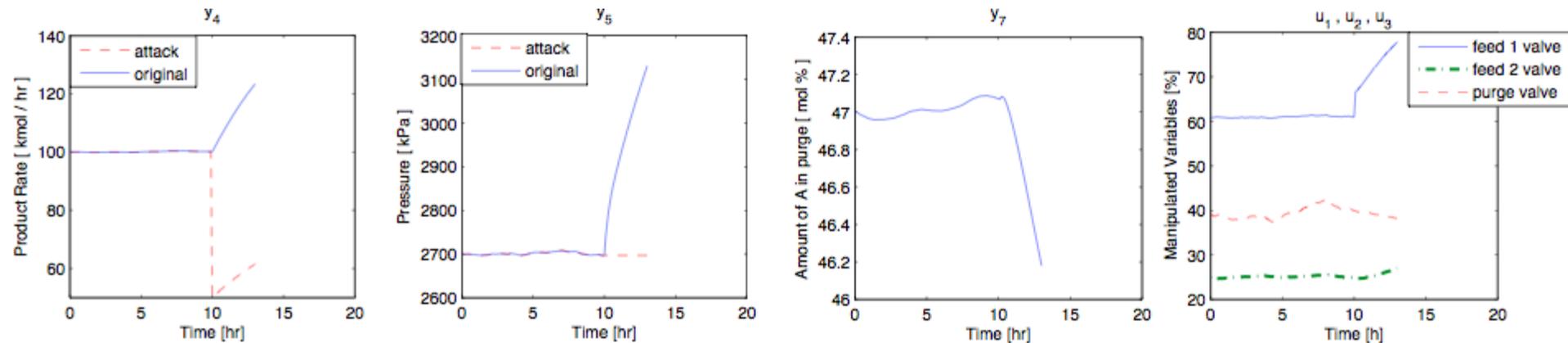
It takes 20 hours to violate Safety by compromising the pressure sensor signal (prevention vs. detection&response)



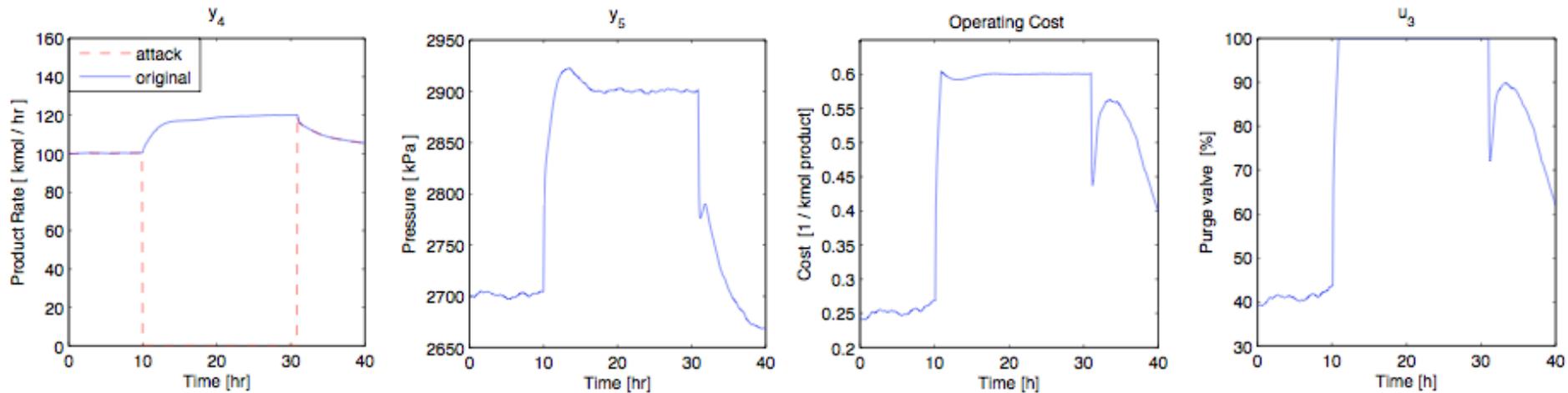
# DoS Attacks: No Impact when the System is at Steady State



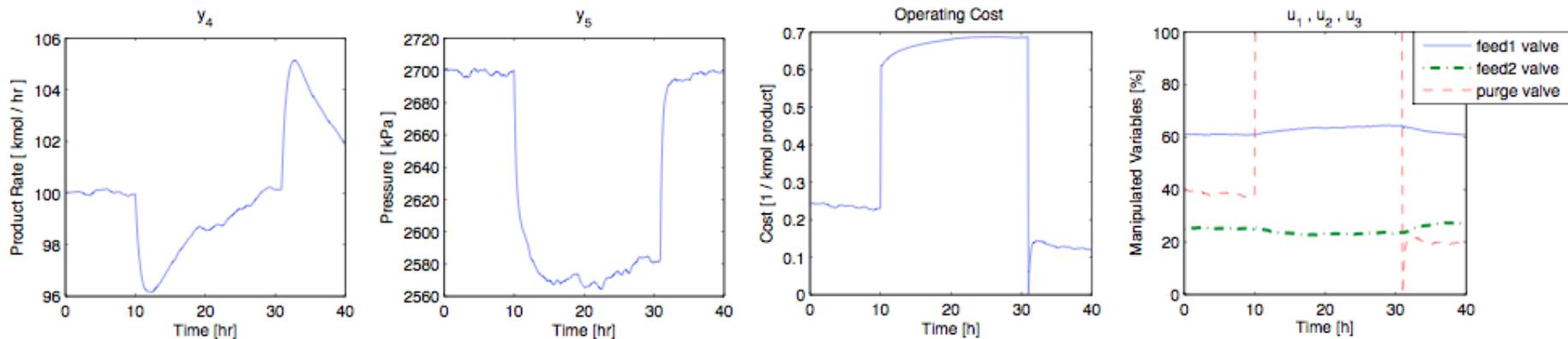
However: A previous “innocuous” integrity attack becomes significant with the help of DoS attacks



# Attacks to the Operational Cost Involve Devices that do not Matter in Safety



Attack increases safety but lowers profits

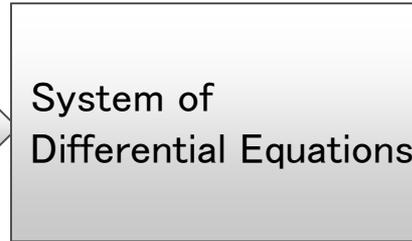


## 1<sup>st</sup> Step: Model the Physical World

Physical World

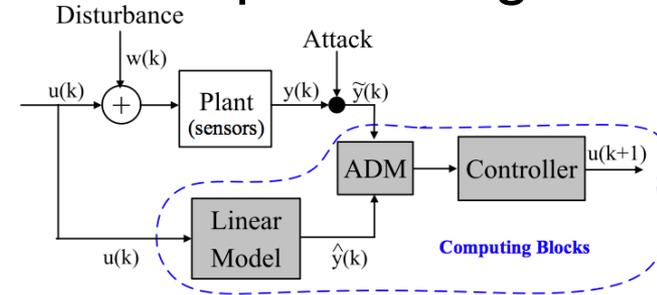


Model



## 2<sup>nd</sup> Step: Detect Attacks

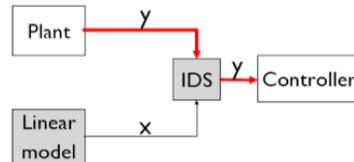
- Compare received signal from expected signal



## 3<sup>rd</sup> Step: Response to Attacks

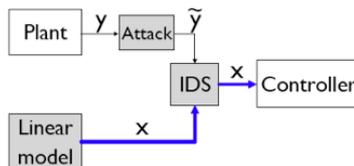
### • No attack

- Use real plant signal



### • IDS detects attack

- Switch to linear model
- Detection time
- False alarm



## 4<sup>th</sup> Step: Security Analysis

- Missed Detections
  - Study stealthy attacks
- False Positives
  - Ensure safety of automated response

# Attacker Strategy: Stealthy Attacks

## ■ Attacker

- Knows our detection model and its parameters
- Wants to be undetected for n time steps
- Wants to maximize the pressure in the tank

## ■ Surge attack

$$\tilde{y}_K = \begin{cases} y^{min} & \text{if } S_{k+1} \leq \tau \\ \hat{y}_K - |\tau + b - S_k| & \text{if } S_{k+1} > \tau \end{cases}$$

## ■ Bias attack

$$\tilde{y}_k = \hat{y}_k - (\tau/n + b)$$

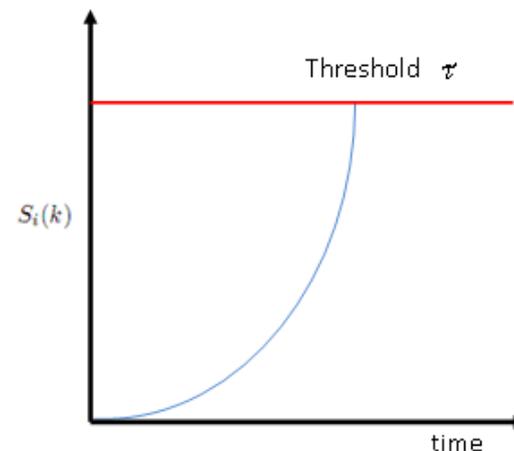
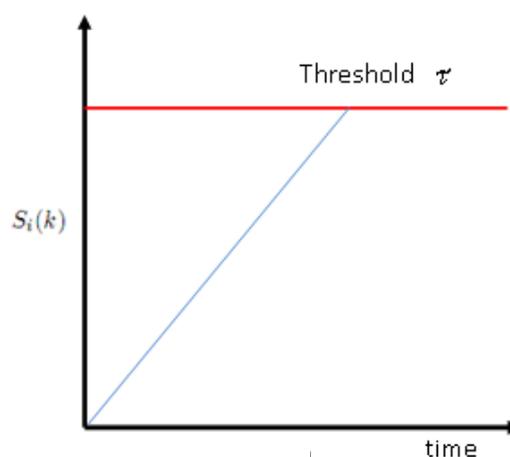
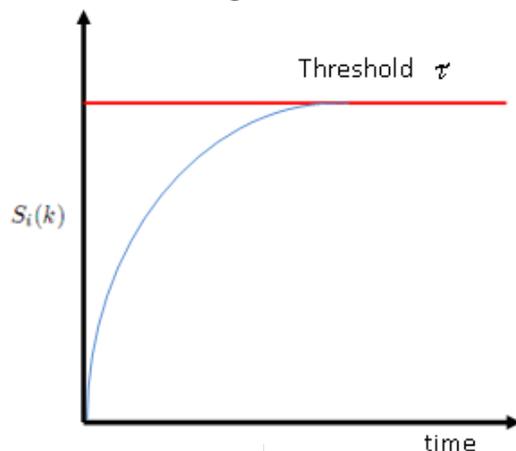
## ■ Geometric attack

$$\tilde{y}_k = \hat{y}_k - \beta\alpha^{n-k}$$

Surge Attack

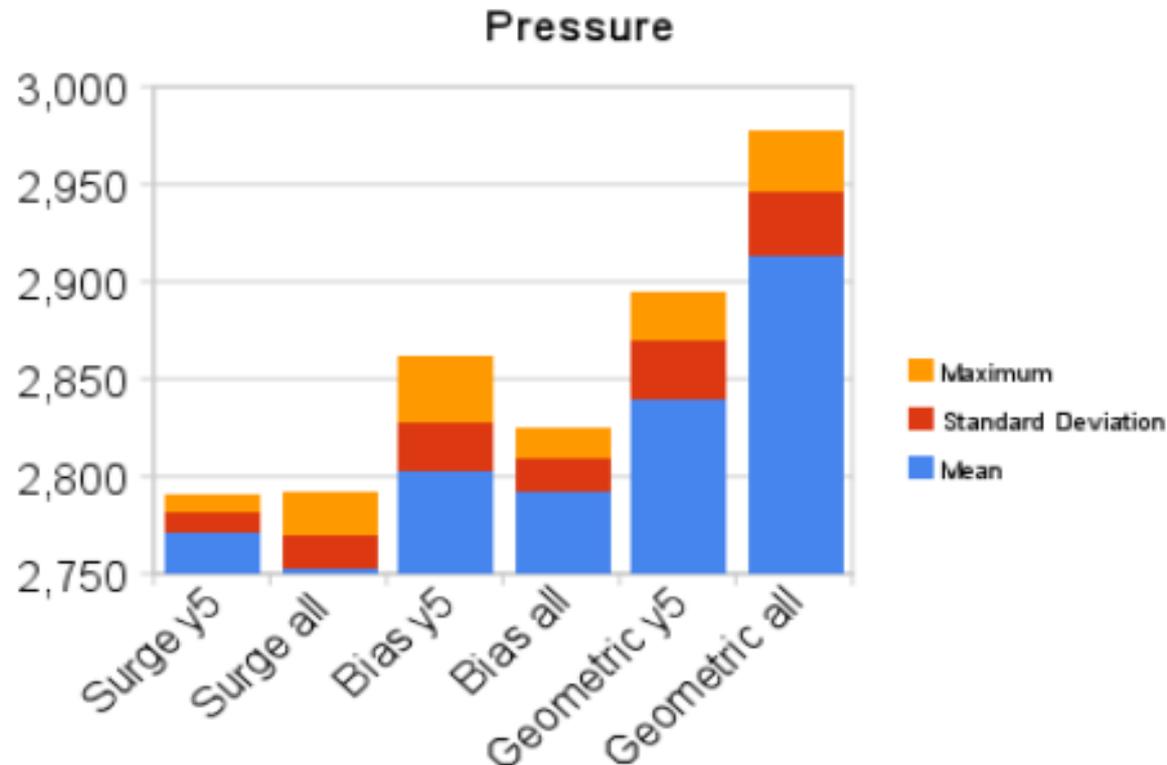
Bias Attack

Geometric Attack



# Impact of Undetected Attacks

- Even geometric attacks cannot drive the system to an unsafe state
- If an attacker wants to remain undetected, she cannot damage the system



For constrained linear systems

$$\begin{aligned}x_{k+1} &= Ax_k + Bu_k^a + w_k, & k &= 1, \dots, N-1 \\x_k^a &= \gamma_k x_k, u_k^a = \nu_k u_k, & (\gamma_k, \nu_k) &\in \{0, 1\}^2\end{aligned}$$

find **causal feedback policies**  $u_k = \mu_k(x_0^a, \dots, x_k^a)$ , that **minimize**  $J(x_0, \mathbf{u}, \mathbf{w}) = \sum_{k=1}^N x_k^\top Q^{xx} x_k + \sum_{k=1}^{N-1} \nu_k u_k^\top Q^{uu} u_k$ , subject to **power constraints**

$$\begin{pmatrix} x_k^a \\ u_k^a \end{pmatrix}^\top \begin{pmatrix} H_i^{xx} & 0 \\ 0 & H_i^{uu} \end{pmatrix} \begin{pmatrix} x_k^a \\ u_k^a \end{pmatrix} \leq \beta_i, \quad i = 1, \dots, L_1,$$

and **safety constraints**

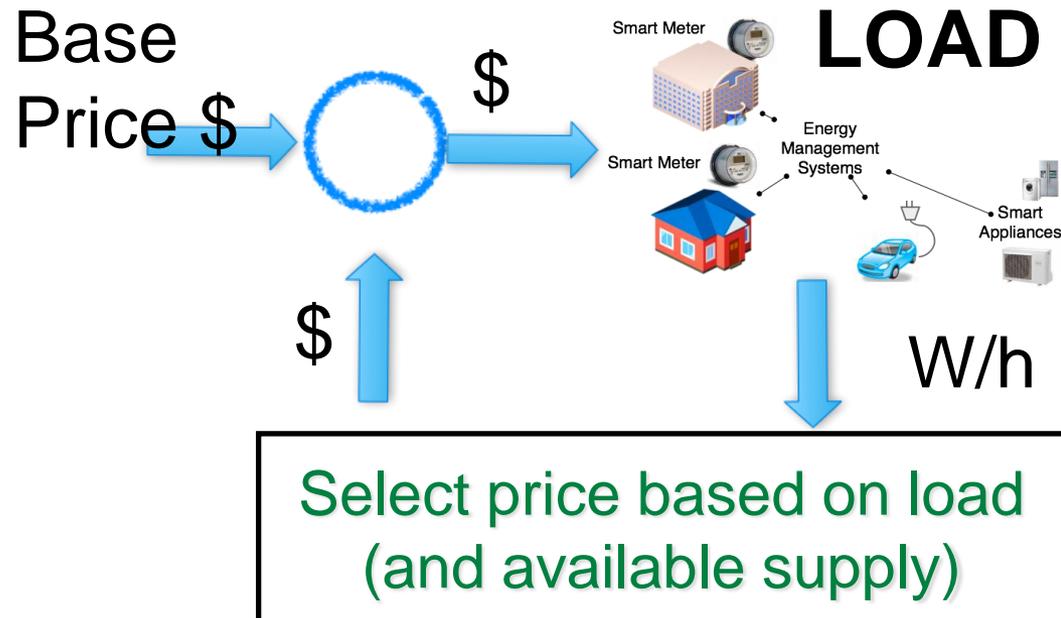
$$\begin{pmatrix} x_k^a \\ u_k^a \end{pmatrix} \in \mathcal{T}_j, \quad j = 1, \dots, L_2,$$

for all **disturbances**  $\mathbf{w} \in \mathbf{W}_\alpha$  OR  $\mathbf{w} \sim \mathcal{N}(0, W)$  and a given set of  $(\gamma_0^{N-1}, \nu_0^{N-1}) \in \mathcal{A}_{pq}$  **attack signatures**.

## ■ Data Minimization Principle

- How much data do we really need to collect for accurate estimation/control?
- Quantity: sampling
- Quality: quantization

## ■ Demand Response (DR)



## ■ DoE 2020 Vision:

- Maintain Smart Grid functions under attack

## ■ Develop resilient algorithms for:

Smart Grid Function	• Untrusted input
State Estimation	• Power flow sensors
Network Topology Processor	• Breakers
Electricity Markets	• Prices
Load balancing	• Smart meter, control data
Transmission/Distribution Automation	• Flexible Alternate Current Transmission System (FACTS)