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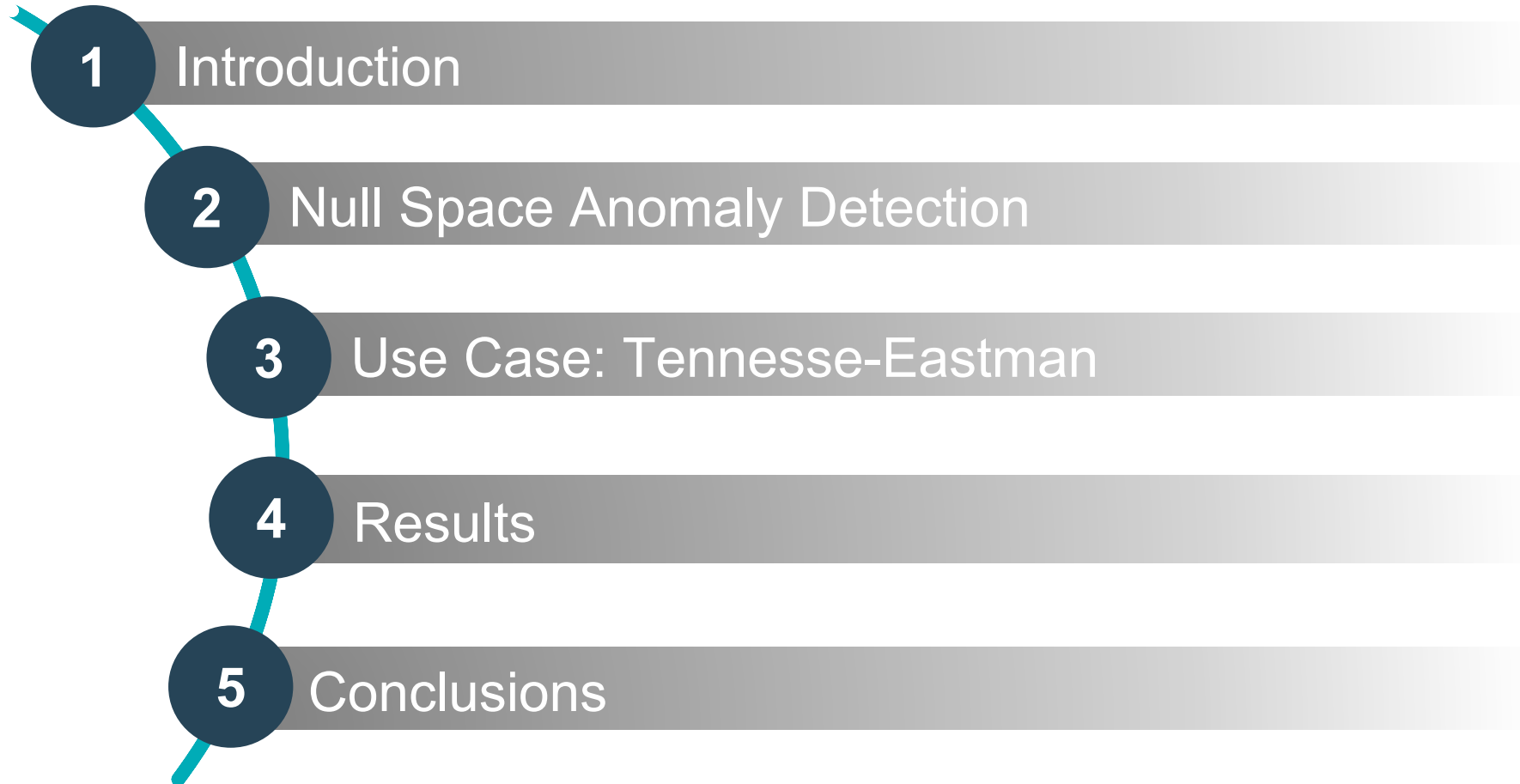
Null is Not Always Empty:

Monitoring the Null Space for Field-Level
Anomaly Detection in Industrial IoT
Environments

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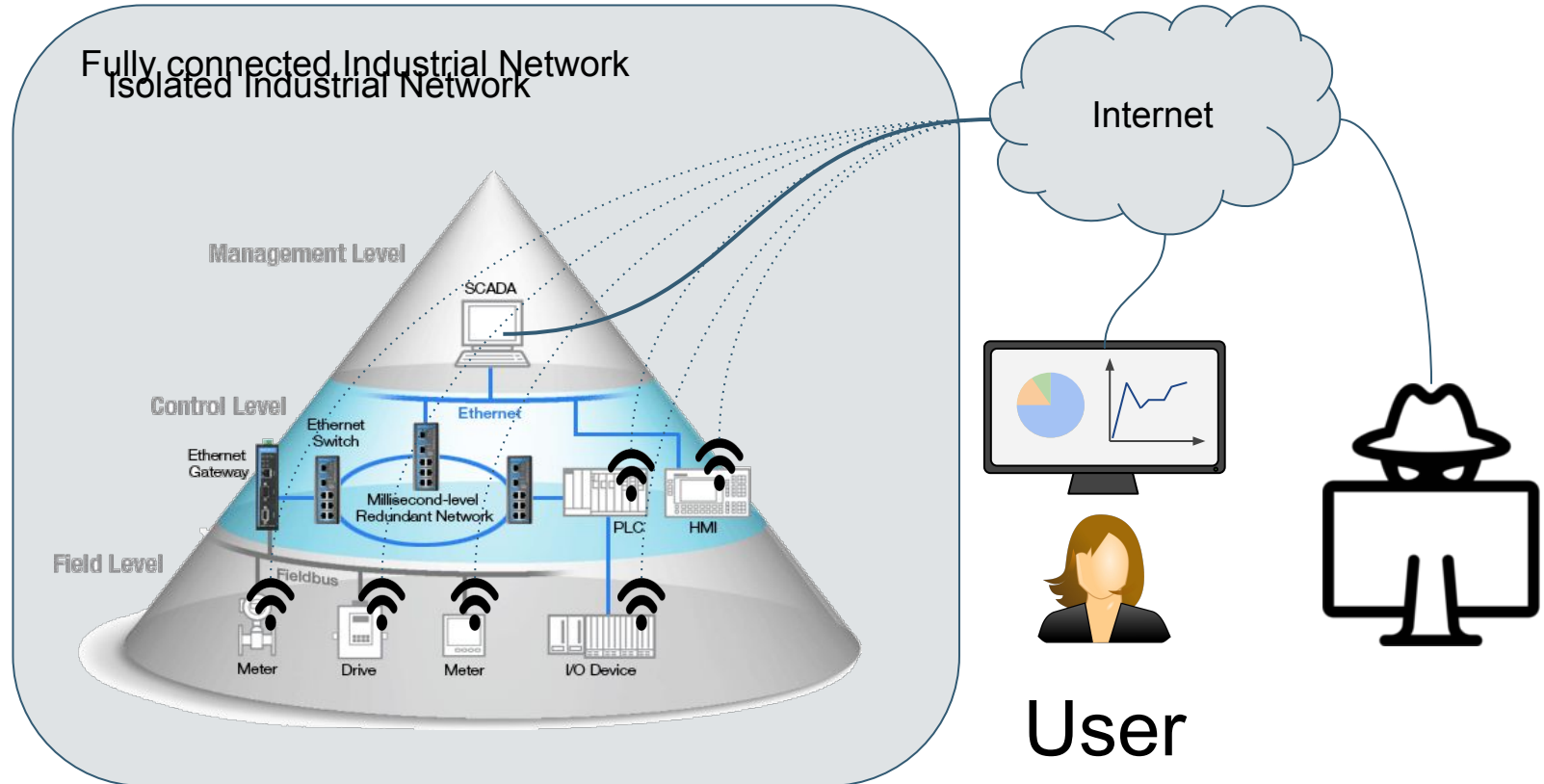
Agenda



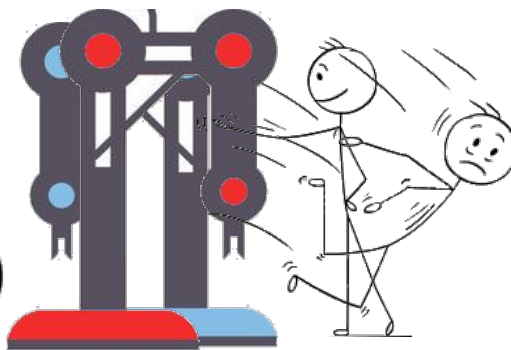
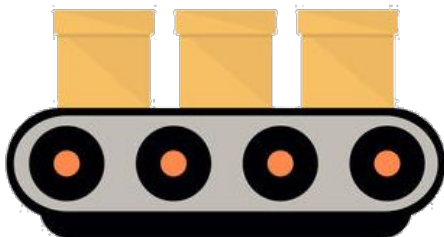
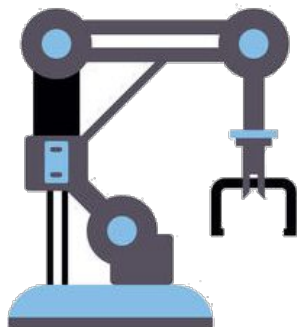
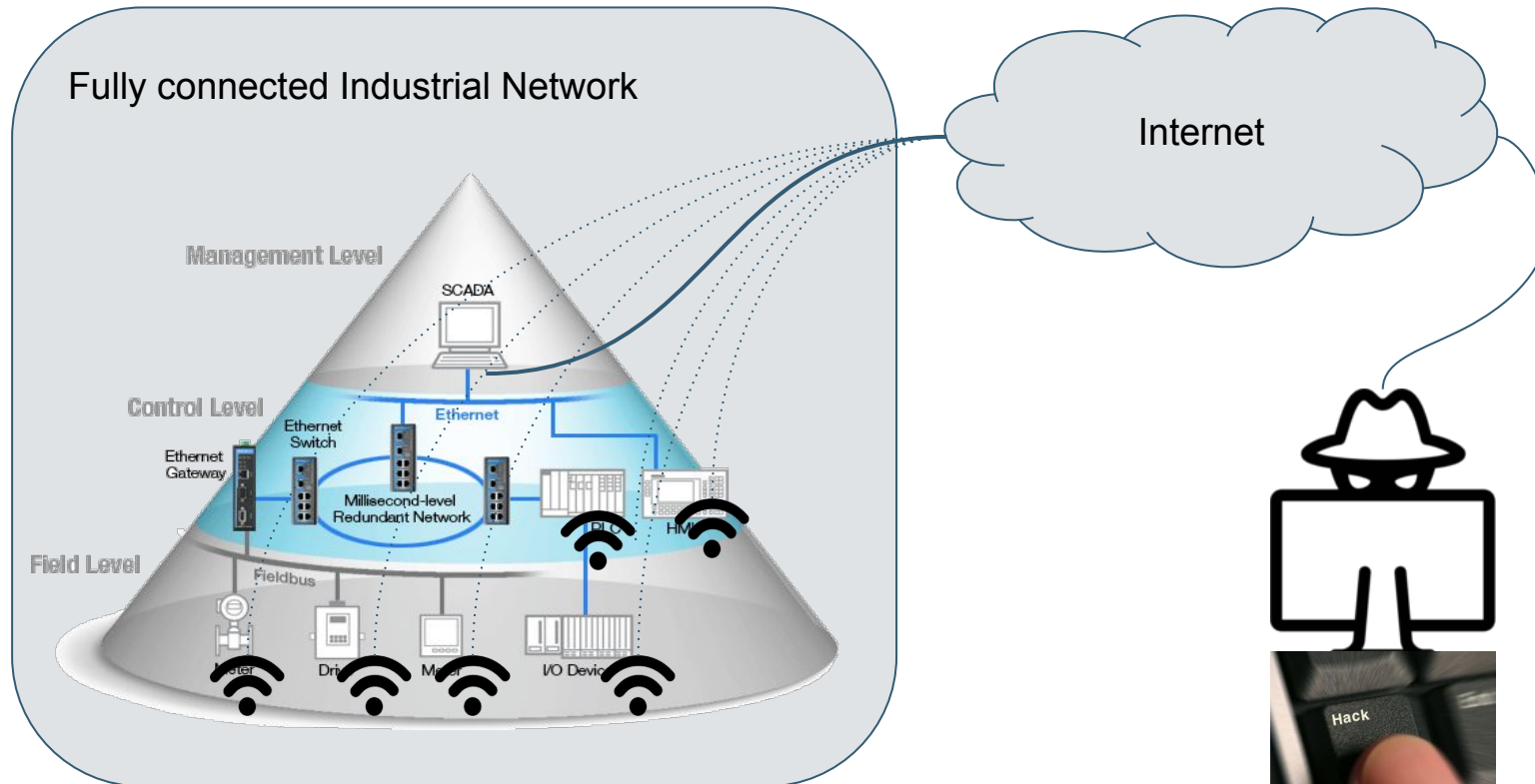
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Introduction

Industrial Networks

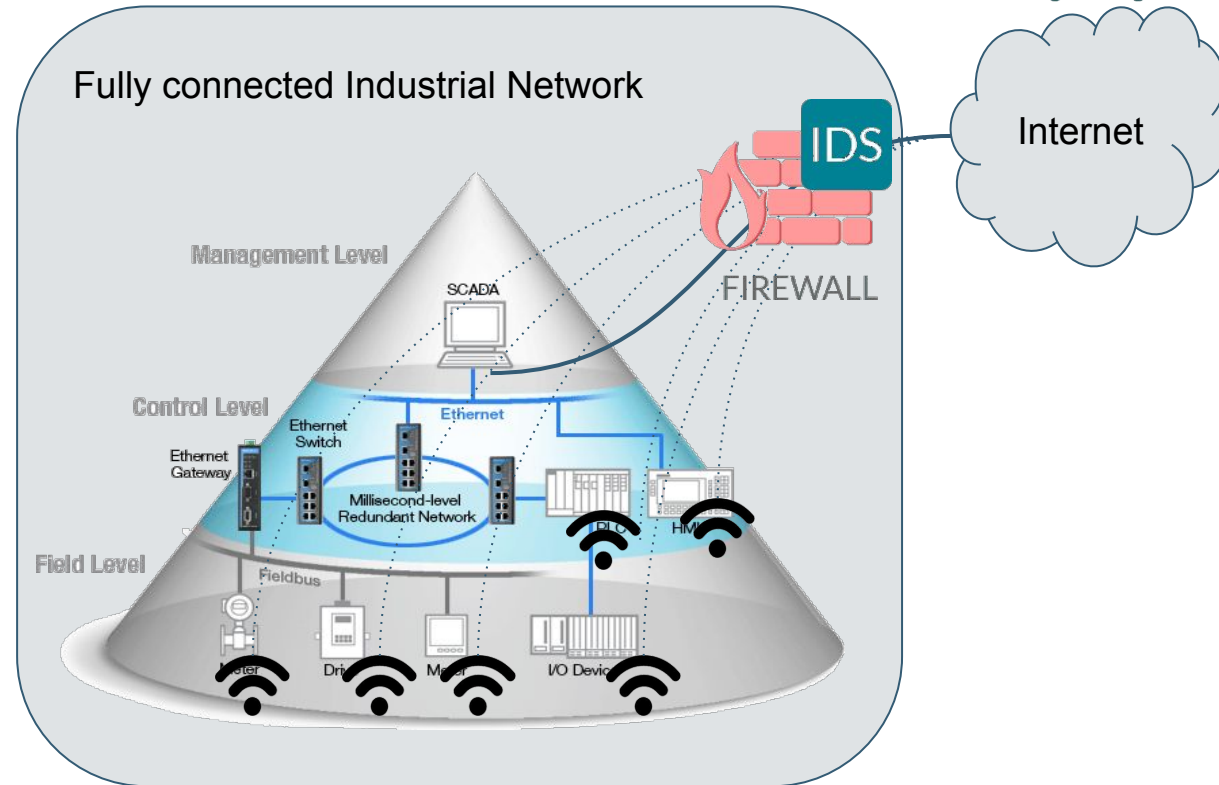


Process Control



Intrusion Detection System

1. Signature Based IDSs
2. Anomaly Detection Systems (ADS)



We present an **Anomaly Detection System** that **monitors physical quantities** of the process itself **to detect intrusions at field-level** that can lead to a unwanted activity within the monitored process

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Null Space Anomaly Detection

Null Space Anomaly Detection

- Multivariate anomaly detection system
- Validated in fields like *Structural Health Monitoring*
- Based in Stochastic Subspace Identification¹
- Uses time series measured in the process as input

$$\mathbf{Y} = [\mathbf{y}_1, \mathbf{y}_2, \dots, \mathbf{y}_m]$$

- Covariance Driven Hankel Matrix transform

$$\mathbf{H}_{p,q} = \begin{bmatrix} \Lambda_1 & \Lambda_2 & \Lambda_2 & \dots & \Lambda_q \\ \Lambda_2 & \Lambda_3 & \dots & \dots & \vdots \\ \Lambda_3 & \dots & \dots & \dots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ \Lambda_{p+1} & \dots & \dots & \dots & \Lambda_{p+q} \end{bmatrix} \quad \Lambda_i = \left(\frac{1}{N-i-1} \right) \sum_{k=1}^{N-i} \mathbf{y}_{k+i} \mathbf{y}_k^t$$

¹ P. Van Overschee and B. De Moor, *Subspace identification for linear systems: Theory–Implementation–Applications*. Springer Science & Business Media, 1996.

Null Space Anomaly Detection

- Hankel Matrix → System identification (*Control Theory*)
- For **ADS**, we do not need to identify the system
- We use **Singular Value Decomposition** on Hankel Matrix
- and find the **Null Space** (U_{H0})

SVD decomposition of H

$$H_{p,q} = U_H S_H V_H^t$$

U_{H0} property

$$U_{H0}^t H_{p,q} = 0$$

- Null hypothesis & Residual:

NullSpace Residual

The Residual Matrix is defined:

$$R_{i,j} = U_{H0}^t H_{i,j}$$

- $R_{i,j} = 0$, Healthy State
- $R_{i,j} \neq 0$, Abnormal State

Null Space Anomaly Detection

- Algorithm
 - Learning phase: (NOC datasets)
 - extract Null Space
 - Calculate Residual values for NOC datasets
 - Threshold Calculation
 - Detection phase:
 - Calculate Residuals
 - check whether they are still under the threshold
- Residuals \approx Anomaly Indicators (AI)¹

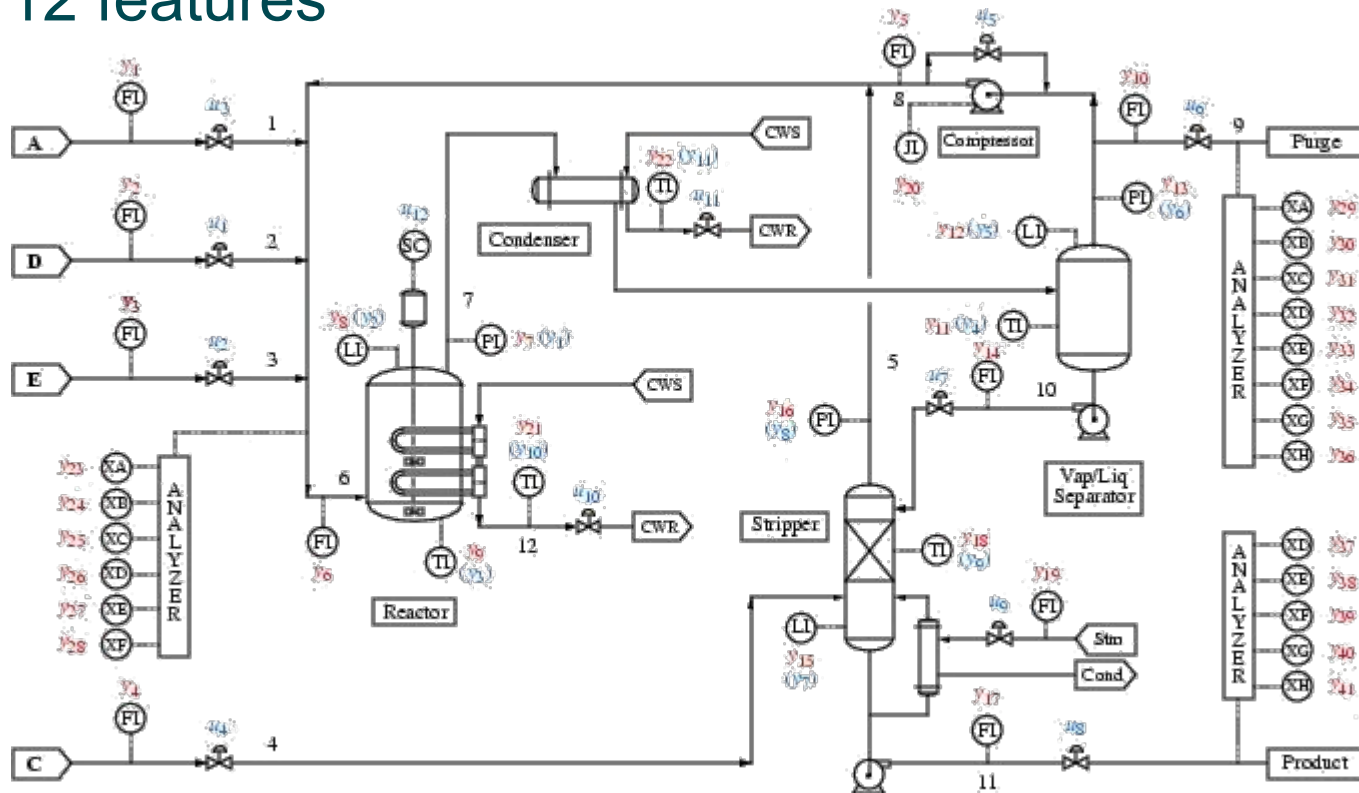
¹ E. Zugasti, A. G. González, J. Anduaga, M. A. Arregui, and F. Martínez, "Nullspace and autoregressive damage detection: a comparative study," Smart Materials and Structures, vol. 21, no. 8, p. 085010, 2012.

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**Use Case:
Tennessee Eastman**

Tennessee Eastman Process

- Chemical Process¹
- From 4 gaseous reactants → 2 liquid products
- 41 + 12 features



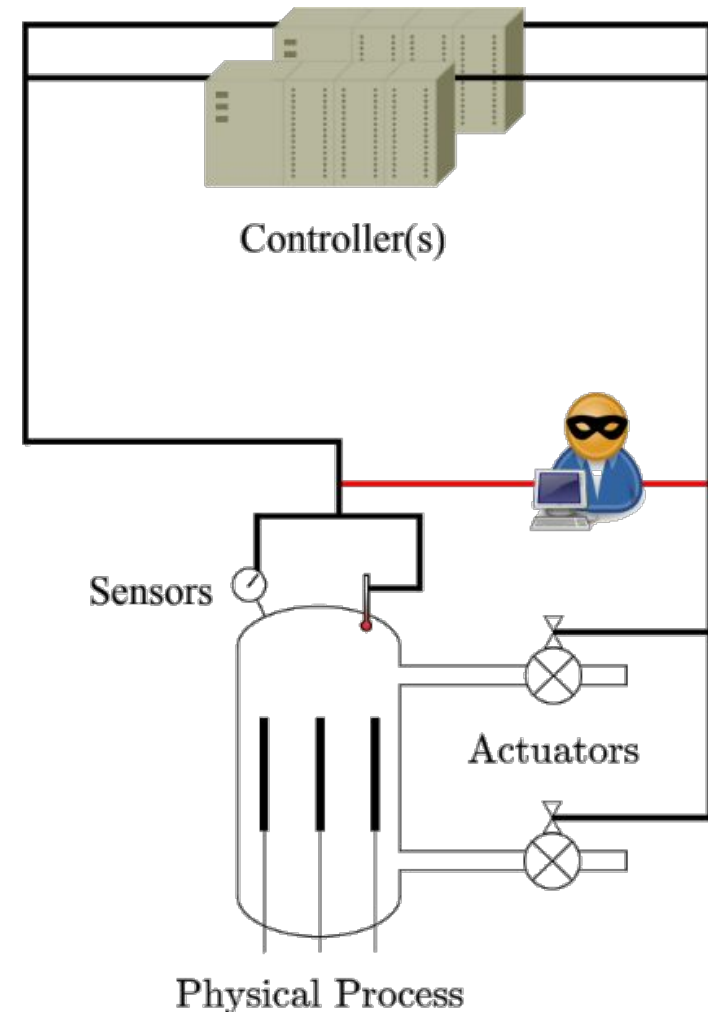
¹ J. J. Downs and E. F. Vogel, "A plant-wide industrial process control problem," Computers & Chemical Engineering, vol. 17, no. 3, pp. 245–255, 1993.

Attack model

- Integrity attack:
 - time series injection
- DoS attack
 - Communication stop
- Performed attacks

Variable number	Variable name	Attack type
XMEAS1	A feed (stream 1)	Integrity
XMEAS8	Reactor level	Integrity
XMEAS9	Reactor temperature	Denial of Service
XMEAS14	Product Separator underflow (stream 10)	Denial of Service
XMEAS17	Stripper underflow (stream 11)	Integrity

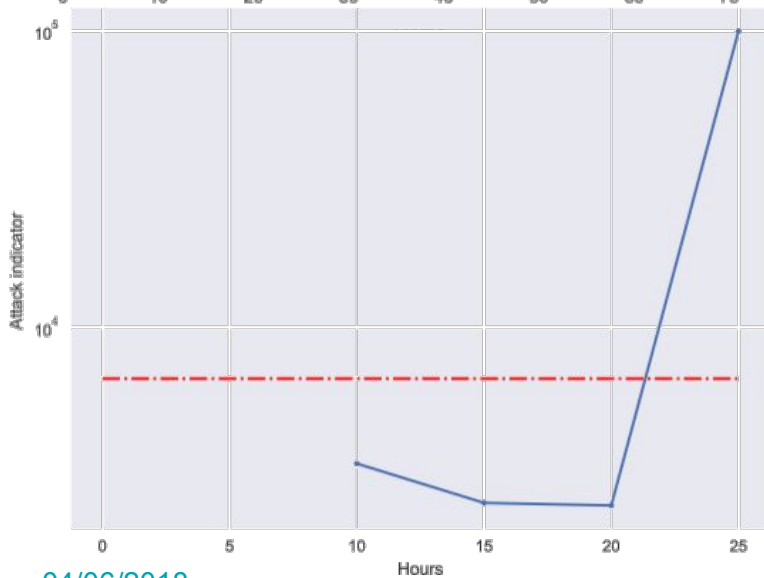
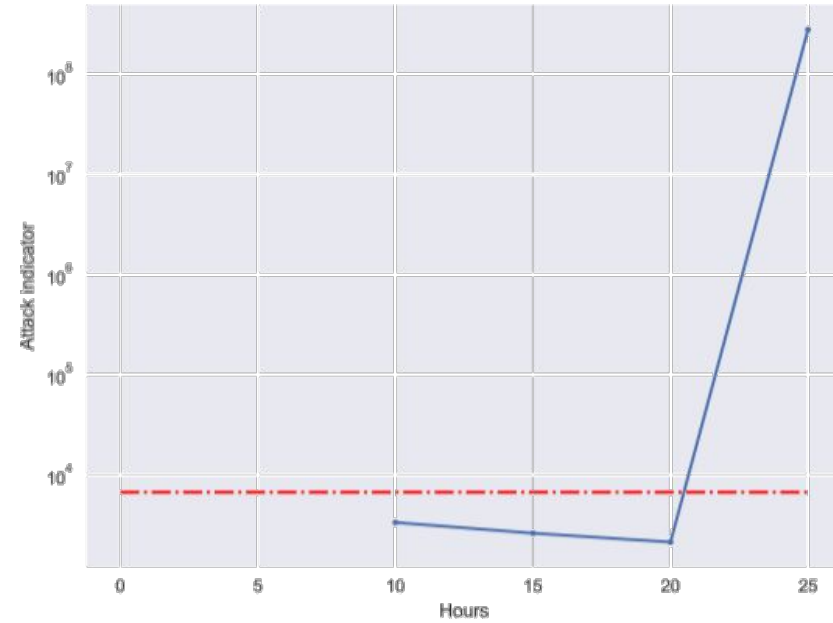
- Simulation time: 72H
 - attack starts after 24H
- $F_s=0.027$ Hz



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Results

Integrity attack results



Variable number

Variable name

XMEAS1

A feed (stream 1)

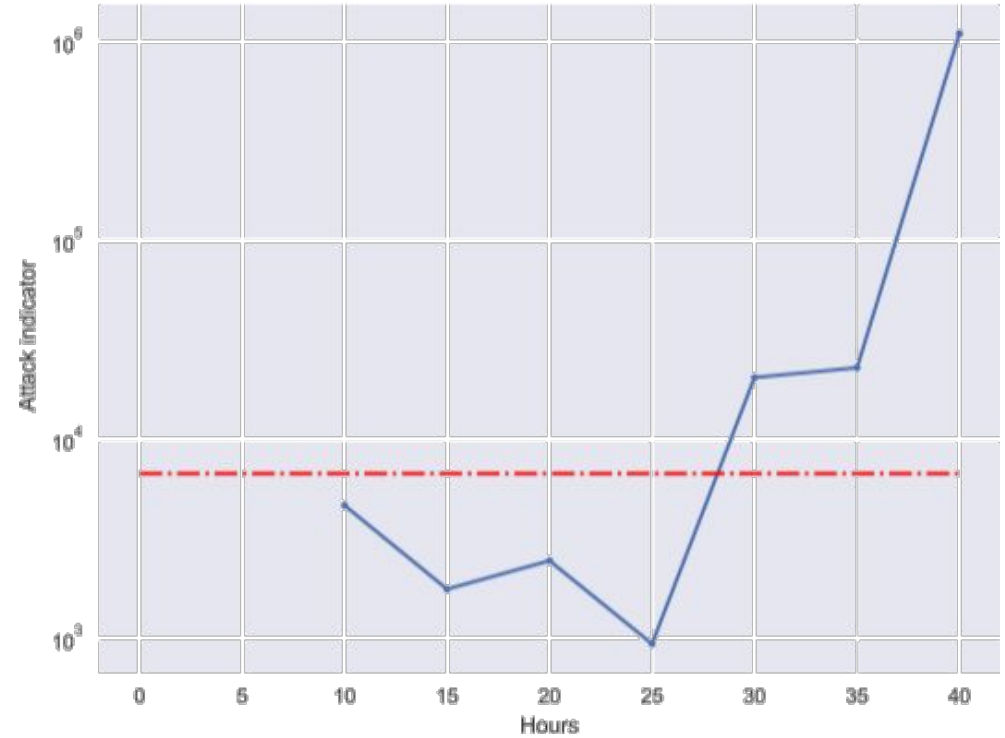
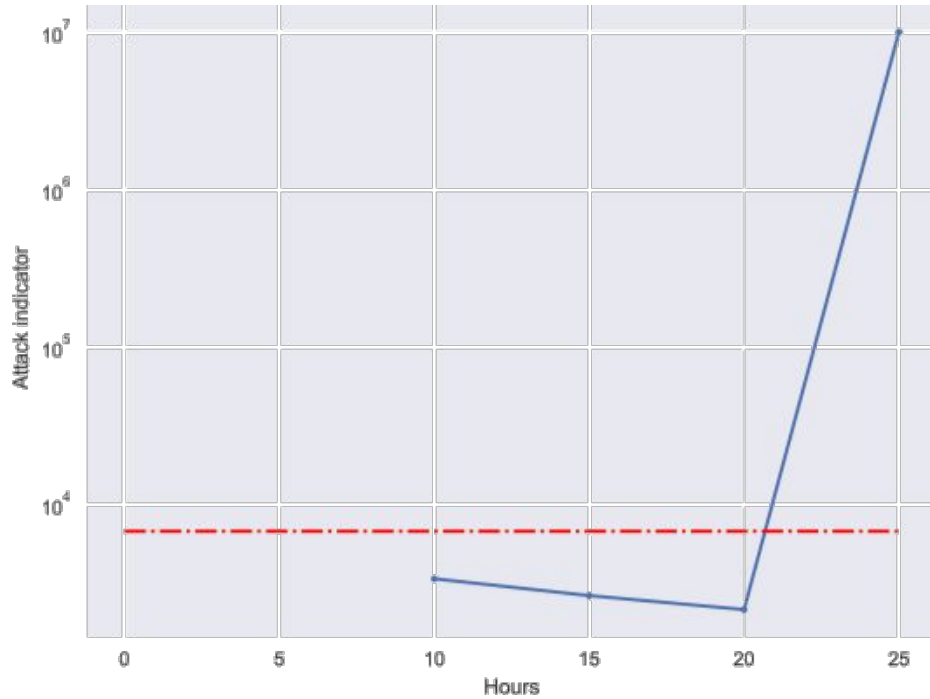
XMEAS8

Reactor level

XMEAS17

Stripper underflow (stream 11)

DoS attack results



Variable number	Variable name
XMEAS9	Reactor temperature
XMEAS14	Product Separator underflow (stream 10)

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Conclusions

Conclusions

- **Attack detection** in IIoT is still an **open challenge**
- We **present** an **ADS** that **detects field-level anomalies**
- The ADS computes an **Attack Indicator**
- **Approach validated** with Tennessee-Eastman process
 - Integrity attacks
 - DoS attacks

Future Work

- Preprocessing data to have a more sensitive method
 - Normalize the inputs
 - Feature transformation methods
- Sliding-window approach for a faster detection
- Add network-level variables to the ADS
- Use more validation scenarios

Acknowledgments

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**Eskerrik asko
Muchas gracias
Thank you**

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