Evasion attack against Multivariate Singular Spectrum Analysis based IDS

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- Industrial Control System
- Process-level IDS
- Attack Model
- MSSA based IDS
- Evasion attack
- Experiment and Result
- Conclusion

Industrial Control System (ICS)



Figure 1: High level architecture of system model.

- Cls are mostly maintained by ICS
- Consists of various workstations
- Some past incidents are:
 - Iranian nuclear plant in 2009
 - German steel mill in 2014
 - Saudi petroleum refinery in 2017
 - Indian nuclear plant (NPCIL) in 2019
 - Israel water treatment plant in 2020

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Process level IDS



Figure 2: High level architecture of system model.

- An ICS is protected by various layers of protection
- But an attacker can evade such protection
- The ultimate attacker's aim gets reflected in physical process
- A process level IDS monitors the sensor measurements to detect attack induced abnormalities

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Attack Model



Figure 3: High level architecture of attack model.

- Attacker targets to the most critical component
- But the critical components are more protected
- Attacker finds some easy manipulable components
- A MITM attacker manipulate sensor measurements [1, 2, 3]
- Proposed perturbation method to craft adversarial measurements.

- A multivariate IDS
- MSSA based IDS ^[4] is:
 - Computationally efficient
 - Captures temporal information
 - Captures mutual correlation
 - Supports noise cancellation property
 - Capable to detect even a stealthy attack
 - Suitable for large scale ICS/IIoT networks

Working of MSSA based IDS

- It works by projecting the recent sensor measurements on noise free signal subspace
- Attack is detected based on departure of projected measurement



Parameters after training:

- Projection Matrix: $U^T \in \mathbb{R}^{R \times L}$
- Centroid Vector: $\hat{c} \in \mathbb{R}^{R}$
- Classifier Threshold: θ
- Lag Parameter L

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$$\mu^{(n)}$$
, $\sigma^{(n)}$

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Consider an ICS consisting of N sensors where n^{th} sensor generates measurement $X_t^{(n)}$ at time t. The IDS performs following steps at timestamp t:

- The current evasion attack methods ^[6, 7] are against deep-learning based IDS
- Do not consider time-series based models
- Deep-learning based IDS suffers from various limitations:
 - High computation cost
 - Noise cancellation property

The low computation cost and noise cancellation property of MSSA-based IDS make it one of the most suitable in large-scale ICS/IIoT networks, which motivates us to analyse it against evasion attacks.

Evasion Attack: Capabilities





- High availability supports non-encrypted communication ^[3, 5]
- An attacker already present in the network can eavesdrop

Manipulation capability

- A rootkit can be deployed to PLC to manipulate the sensor measurements ^[1]
- A victimized employee can be used for exploitation ^[2]
- \blacksquare Various vulnerabilities are reported in OT networks, ${\approx}83\%$ violates communication $^{[3]}$

Evasion Attack: Greedy Approach

Objective Function:

$$\hat{m}_{t+1} = rg\min_{m'_{t+1}} \left(D'_{t+1}
ight)$$

Simplify D'_{t+1} for unknown:

$$D'_{t+1} = ||\hat{c} - U^T \cdot w'_{t+1}||^2$$

= $||\hat{c} - (U[1:L-1]^T \cdot w'_{t+1}[1:L-1] + U[L] * w'_{t+1}[L])||^2$
= $||y - U[L] * m'_{t+1}||^2$
= $||U[L]||^2 m'^2_{t+1} - 2(y^T \cdot U[L])m'_{t+1} + ||y||^2$

where,

$$y = \hat{c} - U[1:L-1]^T \cdot w'_{t+1}[1:L-1]$$

Minima:

$$\hat{m}_{t+1} = \frac{y^T \cdot U[L]}{||U[L]||^2}$$

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Estimate the adversarial measurements

- Break $\hat{m}_{t+1} = ||X_{t+1}||$ into adversarial and non-adversarial: $||X_{t+1}[adv]||^2 + ||X_{t+1}[adv]||^2 = \hat{m}_{t+1}^2$
- Assumption $X_{t+1} \approx X_t$:

$$|X_{t+1}[adv]||^2 = \hat{m}_{t+1}^2 - ||X_t[\overline{adv}]||^2$$

 Manipulation is performed only if departure score is above an estimated threshold

- Accessibility constraints
 - White Box attacker: $(\mathcal{D}, \mathcal{X}, f, \phi)$
 - Gray Box attacker: $(\mathcal{D}, \mathcal{X}, f, \not{X})$
- Manipulation constraints
 - Manipulated measurement must be within the normal range

- TE-process simulator ^[8] is used
- Generated normal measurements
- Generated stealthy attack dataset
- Generated Direct damage attack dataset
- Ensure main sensors impacted by attack

Stealthy attack scenarios



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Direct damage attack scenarios



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White box and Gray box



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- Time Complexity: O(RL)
- Space Complexity: *O*(*RL*)
- Which is equal to the IDS
- Experimental: 53.7 µ-Sec

- We discussed vulnerability of MSSA based IDS
- A practicality of attack model
- Evasion attack against a time-series based IDS
- A novel perturbation method
- Demonstrated on SA and DDA attack scenarios



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Thank You!!

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