

Physics-Based Methods for Distinguishing Attacks from Faults

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Outline

- 1 Introduction
- 2 Approach
- 3 Three Tanks system example
- 4 Fault or Attack
- 5 Experimental Results
- 6 Summary and Conclusions

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Motivation

- ① Cyber-Physical Systems (CPSs) are of great interest due to the wide application area where their model can be used.
- ② System security and attacks detection can be studied through CPS models.
- ③ **Goal:** Detect and distinguish attacks from faults on a complex system using CPS models.

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Contributions

- ① Method for distinguishing attacks from faults in an observed-based framework.
- ② Physics-based methods can be effective, but they cannot deal with every kind of attack.
- ③ Demonstrate approach on hydraulic benchmark system.

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Preliminaries

- CPS model is an instance of a hybrid system, which can operate in different behaviours, called modes.

$$Modes : \begin{cases} y_{m_1} = g_1(x) \\ \dots \\ y_{m_i} = g_i(x) \end{cases}$$

- e.g. a drone has many operating modes
 - take-off, landing, wandering, surface mapping, ...
- The set of modes include also faults/attacks behaviour.

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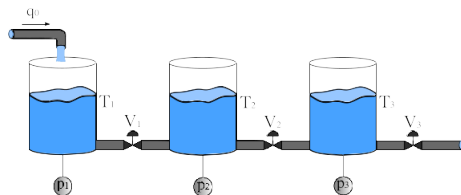
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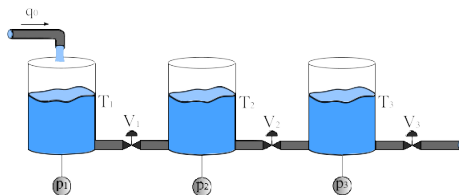
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Nominal Model



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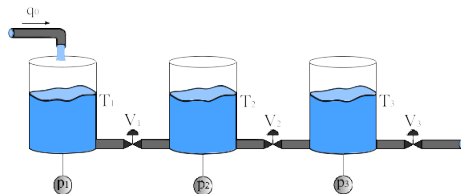


$$\frac{\delta h_1}{\delta t} = q_0 - q_1 = \frac{q_0 - k_1 \text{sign}(h_1, h_2) \sqrt{|h_1 - h_2|}}{A_1}$$

$$\frac{\delta h_2}{\delta t} = \frac{k_1 \text{sign}(h_1, h_2) \sqrt{|h_1 - h_2|} - k_2 \sqrt{h_2}}{A_2}$$

$$\frac{\delta h_3}{\delta t} = \frac{k_2 \text{sign}(h_2, h_3) \sqrt{|h_2 - h_3|} - k_3 \sqrt{h_3}}{A_3}$$

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Input: $u = \{q_0, v_1, v_2, v_3\}$ Output: $y = \{p_1, p_2, p_3\}$

Control Model

- Nominal system model:

$$\begin{aligned}x_{k+1} &= A_{\gamma}x_k + B_{\gamma}u_k + w_k \\ y_k &= C_{\gamma}x_k + v_k\end{aligned}$$

- Observer model:

$$\begin{aligned}\hat{x}_{k+1} &= A_{\gamma}\hat{x}_k + B_{\gamma}u_k + L_{\gamma}(y_k - C_{\gamma}\hat{x}_k) \\ \hat{y}_k &= C_{\gamma}\hat{x}_k + v_k \\ r_k &= y_k - C_{\gamma}\hat{x}_k \\ u_k &= -K_{\gamma}\hat{x}_k\end{aligned}$$

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External perturbation

- Faults influence:

$$\begin{aligned}x_{k+1} &= A_{\gamma}x_k + B_{\gamma}u_k + \textcolor{red}{B}_f f_k + w_k \\y_k &= C_{\gamma}x_k + \textcolor{red}{C}_f f_k + v_k\end{aligned}$$

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$$\begin{aligned}x_{k+1} &= A_{\gamma}x_k + B_{\gamma}u_k + \textcolor{red}{B}_a a_k + w_k \\y_k &= C_{\gamma}x_k + \textcolor{red}{D}_a a_k + v_k\end{aligned}$$

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f_k and a_k are the fault and attack vector respectively.

Fault Model

- Valve faults, leaks, sensor faults, etc..
- Valve setting: $V_i \in [0, 1]$
 - $V_i = 0$ is closed; $V_i = 1$ is open
- Additive model:

$$v_i = \begin{cases} \max\{0, v_i + \Delta_{v_i}\}, & \text{if } \Delta_{v_i} \leq 0 \\ \min\{1, v_i + \Delta_{v_i}\}, & \text{if } \Delta_{v_i} > 0 \end{cases}$$

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Attacks Model

- The attacker cannot monitor the system, only data injection.
- **Sensor:** fake sensor reading in $[0, p_i^{max}]$
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Fault or Attack

- Our system runs over different modes, each of which has a physical model ψ_i , creating the behaviour ξ_i having measurement \hat{y}_i .
- **Mode estimation:** closest mode to anomalous observation \tilde{y}_i

$$\psi^* = \arg \min_{\psi_i \in \Psi} \|\tilde{y}_i - \hat{y}_i\| = \arg \min_{\psi_i \in \Psi} r_i$$

- **Mode identifiability:**
 - distinguishable behaviour $\xi_i \forall j \neq i$
 - activated residual $r_i > \delta$ if system is in mode ψ_i

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Experiments

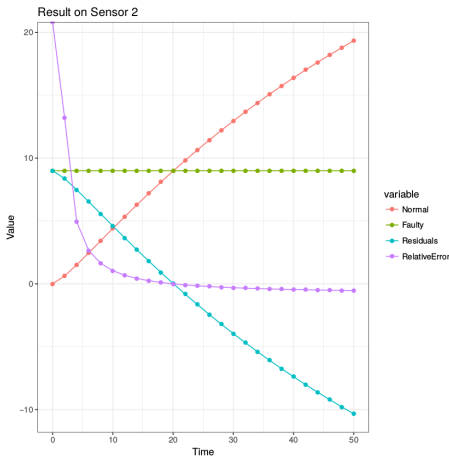
- Three types of tests:
 - Sensors attacks
 - Actuators attacks
 - Multiple components attacks
- Experimental environment:
 - Time domain: $[0, 50]$ seconds
 - Sensor data gathered every 2 seconds
 - Nominal setting: $v_1 = v_2 = v_3 = 0.5$

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Attacks on Sensors

Injected data on the second sensor of our system

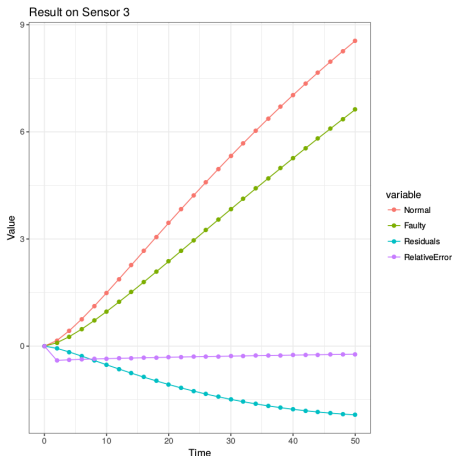


Attack identified through first derivative comparison:

$$\dot{y}_k = -\dot{r}_k$$

Attacks on Actuators

System complexity makes identifiability harder when the actuators are under attack, creating false positives.



Test	Valve 1	Valve 2	Valve 3
155	✓	X	X
355	✓	X	X
755	✓	X	X
955	✓	X	X
515	X	✓	X
535	X	✓	X
575	X	✓	X
595	X	✓	X
551			✓
553			✓
557			✓
559			✓
158	✓	X	✓
544	X	✓	✓
658	✓	X	✓
745	✓	✓	X
958	✓	X	✓
247	✓	✓	✓
638	✓	✓	✓

Multi Attacks and Results

Sensors problems correctly detected and identified.
Actuators errors detected.

Test	Valve 1	Valve 2	Valve 3	Sensor
s1_325		✓	X	1
s2_553	X		✓	2
s3_148	✓	✓		3
s12_558			✓	1-2
s23_647	✓			2-3
s31_348		✓		1-3
s123_666				1-2-3

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- Showed a security system approach based on Cyber-Physical Systems.
- Distinguishing attacks from faults is difficult when the system has few sensors.
- Future work
 - Deeper studies on the synergies of the system and between sensors' data.
 - Optimize the number of sensors in the system.

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