Security of Cyberphysical Systems

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Securing an automated transportation system



See the video "Tackling Autonomous Vehicle Cybersecurity Issues" at https://cesg.tamu.edu/faculty/p-r-kumar/convergencelab/

Cyber-physical systems

- Next generation of engineered systems in which computing, communication, and control technologies are tightly integrated
- Many societally important future applications
 - Smart grid
 - Automated transportation
 - Unmanned Air Vehicle Transportation System
 - Water treatment facilities
 - Telesurgery systems
 - **–** ...
- Safety critical
 - Malfunctioning causes physical harm
- Critical infrastructure
 - Important to functioning of economy and society

Vulnerability of cyberphysical systems to attacks

- Hackers hitherto could tamper only with information or bits in cyber layer
- CPS tightly couples cyber and physical worlds
 - Actions in physical world taken based on information from cyber layer
- CPS, therefore, gives hacker ability to cause damage in physical world

Security of CPS

- As more systems are connected to the Internet and become more open, there are increasingly more vulnerabilities
- Can be more harmful than other violent attacks
- Next war may be "cyber" rather than "bombs"?
- Even after many decades we still cannot secure the Operating Systems
 - New patches every day
- We still cannot secure the Internet
- Interaction between bits and physical world is very complex
- How can we possibly secure CPSs?

Several attacks on critical infrastructure systems

- Several instances of attacks in the past
 - Maroochy-Shire sewage treatment plant
 - Davis-Besse nuclear power plant
 - Stuxnet
 - Ukraine power grid
 - Water filtering plant in Pennsylvania
 - Demonstrations of cyber attacks in automated cars
- Maroochy-Shire, Australia, 2003, attack on sewage treatment system, commands issued which led to a series of faults in the system
- Attack on computers controlling Davis-Besse nuclear power plant in Ohio, 2003, Slammer worm disabled the safety monitoring system
- Stuxnet worm, 2010, exploited Microsoft Windows vulnerability to subvert critical computers controlling centrifuges in Iran uranium enrichment facility
- Attacks on Supervisory Control and Data Acquisition system, natural gas pipeline systems, trams, power utilities, and water systems, etc.

Isn't network security enough for CPS security?

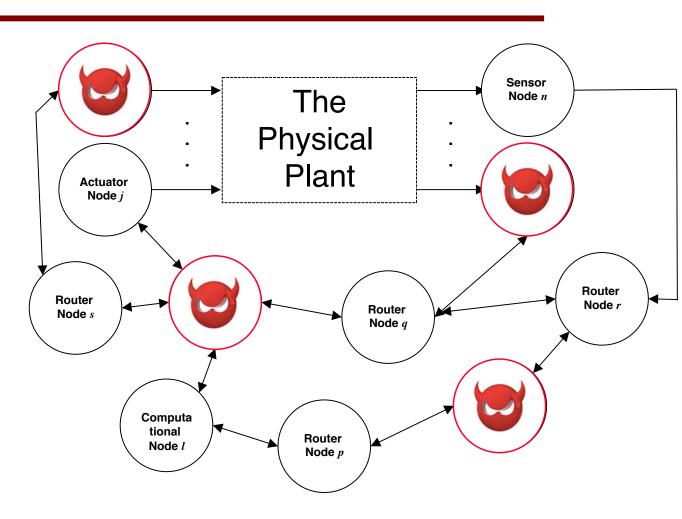
- Network and information security implemented through periodic patching.
 - CPS has a dynamic system in the loop, and may not admit controllers going online for patching
- Traditional notion of "Confidentiality, Integrity and Availability" in network and information security does not address real-time availability, which is critical for control system security
- Network or information security fundamentally cannot address physical layer attacks such as in Maroochy-Shire incident

Two-layer approach to CPS security

- Can think of CPS as consisting of two layers:
 - Cyber layer consisting routers, switches, relays, etc. providing communication backbone,
 - Physical layer consisting the plant, sensors and actuators, controllers which manipulate physical signals
- Cyber layer possibly secured using techniques such as cryptography
 - Therefore, network may possibly be abstracted as secure, reliable, delay-guaranteed bit pipes
- But how to secure the physical layer?

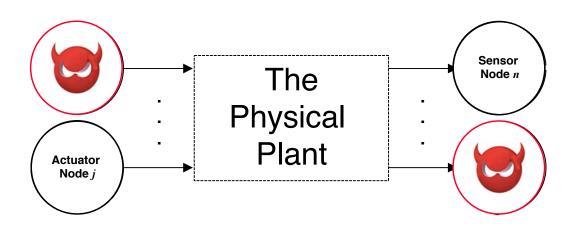
Abstraction of cyberphysical systems

- Overall system has
 - Physical plant
 - Actuators
 - Sensors
 - Routers
 - Computational nodes
 - Network
- But some of the routers, computation nodes, sensors, actuators may be compromised
- How do we secure the overall cyberphysical system?



Abstraction of security problem

- Some sensors, actuators may be compromised
- If information from a sensor is compromised, we say sensor is compromised
- It does not matter whether sensor is compromised or its information is compromised downstream

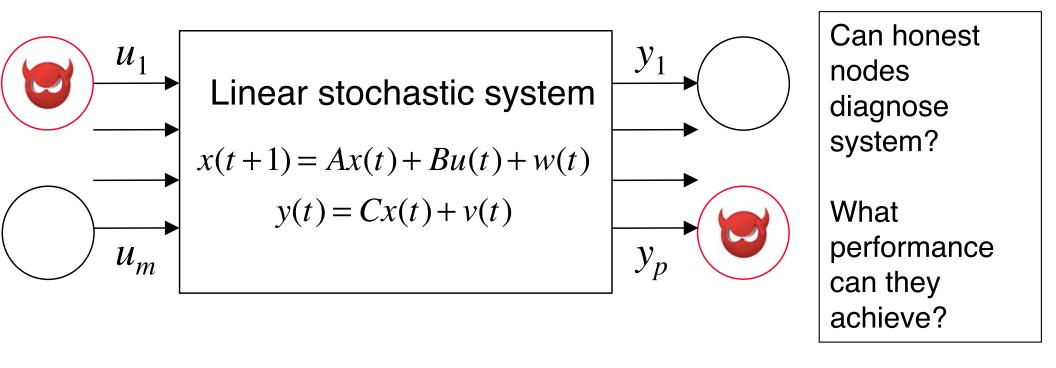


How do we secure the overall cyberphysical system when some sensors and actuators may be compromised?



Towards a paranoid theory of linear stochastic systems

Let's start with linear stochastic systems



- Physical plant modeled as linear stochastic system
 - Most common practical design
- Some actuators/sensors malicious
- Malicious actuators/sensors can collude
- Honest actuators/sensors don't know which nodes malicious

Linear systems theory in a more innocent age

- ♦ Linear system x(t+1) = Ax(t) + Bu(t)
- When is system controllable (Kalman)?

$$x(n) = A^{n}x(0) + Bu(n-1) + ABu(n-2) + \dots + A^{n-1}Bu(0)$$

$$x(n) - A^{n}x(0) = [B, AB, A^{2}B, ..., A^{n-1}B]$$

$$\vdots$$

$$u(n-1)$$

$$\vdots$$

$$u(0)$$

- ♦ Controllable subspace = Span[$B,AB,...,A^{n-1}B$]
- System is stabilizable if unstable modes of A are in controllable subspace

Linear systems theory in a more innocent age

- ♦ Linear system x(t+1) = Ax(t)y(t) = Cx(t)
- When is system state observable from outputs?

$$\begin{bmatrix} y(0) \\ y(1) \\ \vdots \\ y(n-1) \end{bmatrix} = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix} x(0)$$

$$\bullet \text{ Unobservable subspace} = \text{Null Space of } \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}$$

- System is detectable if unstable modes are observable

But what if some actuators or sensors are malicious?

$$\begin{bmatrix} x_{1}(t+1) \\ x_{2}(t+1) \\ \vdots \\ x_{n}(t+1) \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \\ \vdots \\ x_{n}(t) \end{bmatrix} + \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1m} \\ b_{21} & b_{22} & \dots & b_{2m} \\ \dots & \dots & \dots & \dots \\ b_{n1} & b_{n2} & \dots & b_{nm} \end{bmatrix} \begin{bmatrix} u_{1}(t) \\ u_{2}(t) \\ \vdots \\ u_{m}(t) \end{bmatrix}$$

$$\begin{bmatrix} y_{1}(t) \\ y_{2}(t) \\ \vdots \\ y_{p}(t+1) \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} & \dots & c_{1n} \\ c_{21} & c_{22} & \dots & c_{2n} \\ \dots & \dots & \dots & \dots \\ c_{p1} & c_{p2} & \dots & c_{pn} \end{bmatrix} \begin{bmatrix} x_{1}(t) \\ x_{2}(t) \\ \vdots \\ x_{n}(t) \end{bmatrix}$$

- Some of the u_i 's and y_i 's may be malicious
- What harm can malicious sensors/actuators cause without the honest sensors/actuators knowledge?

Innocent age concerns vs New age concerns

- Nature causes stability/instability
- Malicious agents cause harm
- Stability of benign systems
- Security of malicious systems
- Stabilizability/Detectability of benign systems
- Securability of malicious systems

Passive guarantees based on system structure

The securable and unsecurable subspaces for deterministic systems

 What states can the malicious sensors/actuators drive the system to without the honest sensors/actuators finding out?

◆ The unsecurable subspace V is the set of states that the malicious sensors and actuators can keep indistinguishable from the 0 state

lacktriangle The securable subspace is V^{\perp}

The unsecurable states of deterministic systems

• Suppose $x(t+1) = Ax(t) + B_m u_m(t)$

$$y_h(t) = \begin{bmatrix} x_1(t) \\ \vdots \\ x_H(t) \end{bmatrix} = C_h x(t)$$

• Then x(0) can be made indistinguishable from 0 if for some $u_m(0), u_m(1), \dots, u_m(t), \dots$

$$C_h x(0) = 0$$

 $C_h (Ax(0) + B_m u_m(0)) = 0$
 \vdots
 $C_h (A^t x(0) + A^{t-1} B_m u_m(0) + \dots B_m u_m(t-1)) = 0$

Characterization of Unsecurable and Securable subspaces

Unsecurable subspace is the maximal subspace V
such that for all v in V

$$C_h v = 0$$

There exists u such that $Av + B_m u \in V$

ullet Securable subspace is V^{\perp}

Stochastic systems

Malicious sensors and actuators in linear stochastic system

Consider a linear stochastic system

$$x(t+1) = Ax(t) + Bu^{g}(z^{t}) + B_{m}u_{m}(t) + w(t+1)$$
$$y(t) = x(t)$$

- w is white noise of variance Σ
- Honest sensors measure y_1, y_2, \dots, y_H
- Malicious sensors measure $y_{H+1}, y_{H+2}, \dots, y_p$
- Sensor measurements reported are z(t), where $z_i(t)=x_i(t)$ for $i=0,1,\ldots,H$
- But for the malicious sensor's $z_i(t)$ need not equal $x_i(t)$ for i = H+1, H+2, ..., p
- And malicious actuators may apply $u_m(t)$ different from 0

What performance can be guaranteed for a linear stochastic system?

Honest sensors conduct Test to detect if there is any malicious activity:

$$\lim_{T \to 0} \frac{1}{T} \sum_{t=0}^{T-1} \left(z(t+1) - Az(t) + Bu^{g}(z^{t}) \right) \left(z(t+1) - Az(t) + Bu^{g}(z^{t}) \right)^{T} = \Sigma$$

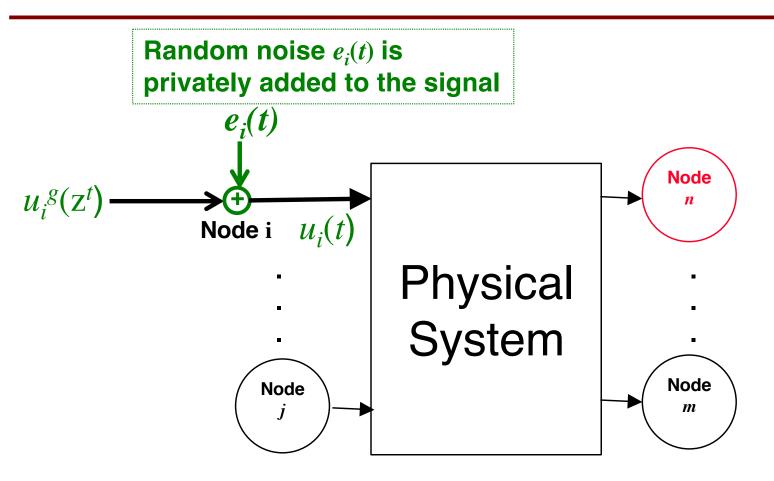
- To remain undetected malicious sensors/actuators must pass Test
- **Theorem**: Then the error in the reported state error in the securable subspace V^{\perp} is guaranteed to be of zero power

$$\lim_{T \to 0} \frac{1}{T} \sum_{0}^{T} \left\| \tilde{x}(t)_{V^{\perp}} \right\|^{2} = 0$$

where
$$\tilde{x}(t)_{V^{\perp}} = \text{Projection of } (z(t) - x(t)) \text{ on } V^{\perp}$$

Can we do better?

Dynamic watermarking



 Actuator node superimposes a private excitation whose realization is unknown to other nodes

Why does it help?

- Private excitation $e_i(t)$ appears in transformed returned signals from sensors at time t+1
- Measurement reported by sensor at time t+1 has to contain suitably transformed contribution of e_i(t)
- So actuator can check if private excitation comes back properly from sensors
- Checks if the reported measurements have the appropriately correlations with $e_i(t)$ reported
- This provides powerful guarantees against general attacks on sensors – not just replay attack

Illustration on simple first order SISO system

- SISO system: x(t+1) = ax(t) + bu(t) + w(t+1) $w(t) \sim N(0, \sigma_w^2)$, i.i.d.
- Dynamic watermarking $u(t) = u^g(t) + e(t)$ with $e(t) \sim N(0, \sigma_e^2)$, i.i.d.
- Two tests are conducted by actuator

$$\lim_{t \to 0} \frac{1}{T} \sum_{t=0}^{T-1} (z(t+1) - az(t) - bu^{g}(t) - be(t))^{2} = \sigma_{w}^{2}$$

$$\lim_{t \to 0} \frac{1}{T} \sum_{t=0}^{T-1} (z(t+1) - az(t) - bu^{g}(t))^{\frac{?}{2}} b^{2} \sigma_{e}^{2} + \sigma_{w}^{2}$$

- If either test fails, then there is malicious sensor information
 - System goes into safety mode
 - Halted, checked, rebooted, manual operation, etc

Guarantee provided by Dynamic Watermarking

Theorem

$$\lim_{T \to 0} \frac{1}{T} \sum_{t=0}^{T-1} v^2(t) = 0$$

• Where $v(t+1) = z(t+1) - az(t) - bu^g(t) - be(t) = w(t+1)$

Interpretation:

$$z(t+1)-az(t)-bu^{g}(t)-be(t)=w(t+1)+v(t+1)$$

• So reported sensor measurements can distort actual noise w(t) only by zero power signal v(t)

- From the first test $\lim_{T\to\infty}\frac{1}{T}\sum_{k=1}^{T}(v[k]+w[k])^2=\sigma_w^2$
- Hence $\lim_{T \to \infty} \frac{1}{T} \sum_{k=1}^{T} v^2[k] + 2v[k]w[k] + w^2[k] = \sigma_w^2$
- So $\lim_{T \to \infty} \frac{1}{T} \sum_{k=1}^{T} v^2[k] + \lim_{T \to \infty} \frac{1}{T} \sum_{k=1}^{T} 2v[k]w[k] = 0$
- From the second test $\lim_{T\to\infty}\frac{1}{T}\sum_{k=1}^{T}(v[k]+be[k-1]+w[k])^2=b^2\sigma_e^2+\sigma_w^2$
- So $\lim_{T \to \infty} \frac{1}{T} \sum_{k=1}^{T} (v[k] + w[k])^2 + \lim_{T \to \infty} \frac{1}{T} \sum_{k=1}^{T} b^2 e^2 [k-1]$

$$+\lim_{T\to\infty}\frac{1}{T}\sum_{k=1}^{T}2be[k-1](v[k]+w[k])=b^{2}\sigma_{e}^{2}+\sigma_{w}^{2}$$

- Since e has variance σ^2 : $\lim_{T\to\infty}\frac{1}{T}\sum_{k=1}^{T}e[k-1](v[k]+w[k])=0$
- Since e(t-1) and w(t) are independent: $\lim_{T\to\infty}\frac{1}{T}\sum_{k=1}^{T}e[k-1]v[k]=0$
- Define the σ -algebra $S_k := \sigma(x^k, z^k, e^{k-2})$
- And $\hat{w}[k] := E[w[k] | S_k]$
- Since $w[k] = x[k] ax[k-1] bg_{k-1}(z^{k-1}) be[k-1]$
- We have the Markov Chain: $(x^{k-2}, e^{k-2}) \rightarrow (x[k-1], x[k], z^k) \rightarrow w[k]$

So

$$\hat{w}[k] := E[w[k] | \sigma(e^{k-2}, x^{k-2}, x[k-1], x[k], z^k)] = E[w[k] | \sigma(x[k-1], x[k], z^k)]$$

- Since $x[k] ax[k-1] bg_{k-1}(z^{k-1}) = be[k-1] + w[k]$ is i.i.d. Gaussian
- The estimate $\hat{w}[k] = \frac{\sigma_w^2}{b^2 \sigma_e^2 + \sigma_w^2} (be[k-1] + w[k]) = \beta (be[k-1] + w[k])$
- Note $\beta \coloneqq \frac{\sigma_w^2}{b^2 \sigma_e^2 + \sigma_w^2} < 1$
- Let $\widetilde{w}[k] = w[k] \hat{w}[k]$
- Then. $(\widetilde{w}[k-1], \mathcal{S}_k)$ Is a martingale

- ♦ Also $v[k] ∈ S_k$, in fact $v[k] ∈ σ(x^k, z^k)$
- By Martingale Stability Theorem $\sum_{k=1}^{T} v[k] \widetilde{w}[k] = o(\sum_{k=1}^{T} v^2[k]) + O(1)$
- Now $\sum_{k=1}^{T} v[k]w[k] = \sum_{k=1}^{T} v[k](\hat{w}[k] + \tilde{w}[k])$ = $\sum_{k=1}^{T} v[k]\hat{w}[k] + o(\sum_{k=1}^{T} v^{2}[k]) + O(1)$
- So $\sum_{k=1}^{T} v[k]w[k] = \beta b \sum_{k=1}^{T} v[k]e[k-1] + \beta \sum_{k=1}^{T} v[k]w[k] + o(\sum_{k=1}^{T} v^{2}[k]) + O(1)$
- Hence $\sum_{k=1}^{T} v[k]w[k] = \frac{\beta b}{1-\beta} \sum_{k=1}^{T} v[k]e[k-1] + o(\sum_{k=1}^{T} v^2[k]) + O(1)$

- $\bullet \text{ Now } \sum_{k=1}^{T} v[k]e[k-1] = o(T)$
- So it follows that $\sum_{k=1}^{T} v[k]w[k] = o(\sum_{k=1}^{T} v^2[k]) + o(T) + O(1)$
- So $\sum_{k=1}^{T} v^2[k] + \sum_{k=1}^{T} 2v[k]w[k] = (1+o(1))(\sum_{k=1}^{T} v^2[k]) + o(T)$
- ◆ Taking limits and dividing by T gives the result

Comments

Surprising complexity of proof

Difficult to handle general non-Gaussian case

Stability consequences of Dynamic Watermarking

- Theorem:
- ◆ Suppose |a| < 1, i.e., system is open-loop stable,
- Then distortion d[t] = z[t] x[t] is zero power: $\lim_{T \to \infty} \frac{1}{T} \sum_{k=0}^{T-1} d^2[k] = 0$
- Mean-square performance is same as reported performance $\lim_{T \to \infty} \frac{1}{T} \sum_{k=0}^{T-1} x^2[k] = \lim_{T \to \infty} \frac{1}{T} \sum_{k=0}^{T-1} z^2[k]$
- Suppose $u^g(t) = fx(t)$ with |a+bf| < 1
- Then mean square performance is optimal

$$\lim_{T \to \infty} \frac{1}{T} \sum_{k=0}^{T-1} x^2 [k] = \frac{\sigma_w^2 + b^2 \sigma_e^2}{1 - |a + bf|^2}$$

More general results

Results extend to

 ARMAX Systems used in process control:

$$y[t] = -\sum_{k=1}^{p} a_k y[t-k] + \sum_{k=0}^{h} b_k u[t-l-k] + \sum_{k=0}^{r} c_k w[t-k]$$

 MIMO partially observed Gaussian systems

$$\mathbf{x}[t+1] = A\mathbf{x}[t] + Bu[t] + \mathbf{w}[t+1]$$
$$y[t+1] = C\mathbf{x}[t+1] + n[t+1]$$

Some non-Gaussian systems

Example

♦ System: y(t+1)+0.7y(t)-0.2y(t-1) = u(t)+0.5u(t-1)+w(t) $w(t) \sim N(0,1)$, i.i.d.

- ◆ Actuator applies u(t) = -0.7z(t) 0.2z(t-1) 0.5u(t-1) + e(t) $e(t) \sim N(0,1)$, i.i.d.
- Closed-loop system:

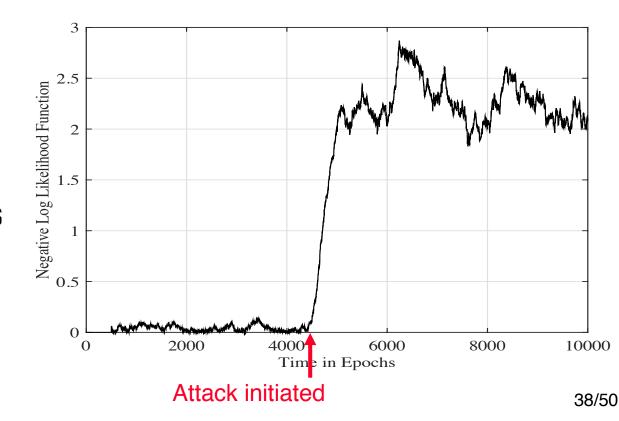
$$y[t+1] = 0.7(y[t] - z[t]) + 0.3(y[t-1] - z[t-1]) + e[t] + w[t+1]$$

Sensor estimates process noise by

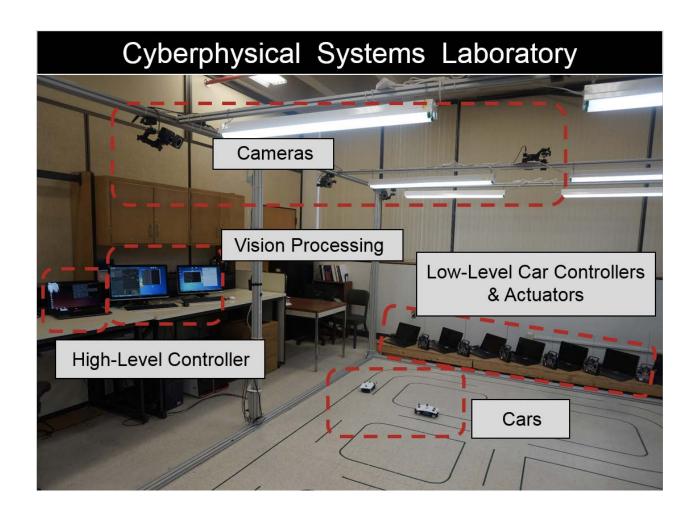
$$\widehat{w}[t+1] := \frac{1}{2} (y[t+1] - 0.7(y[t] - z[t]) - 0.3(y[t-1] - z[t-1])$$

Example

- Simulates a fake system with a fake noise $n(t) \hat{w}(t)$ $n(t) \sim N(0,1)$, i.i.d.
- Reports output of fake simulated system
- In absence of watermarking, actuator would not suspect any malicious measurements
- Sensor attack begins at time 4500



Test of autonomous transportation system in CPS lab



Automated vehicles are vulnerable to cyber attacks

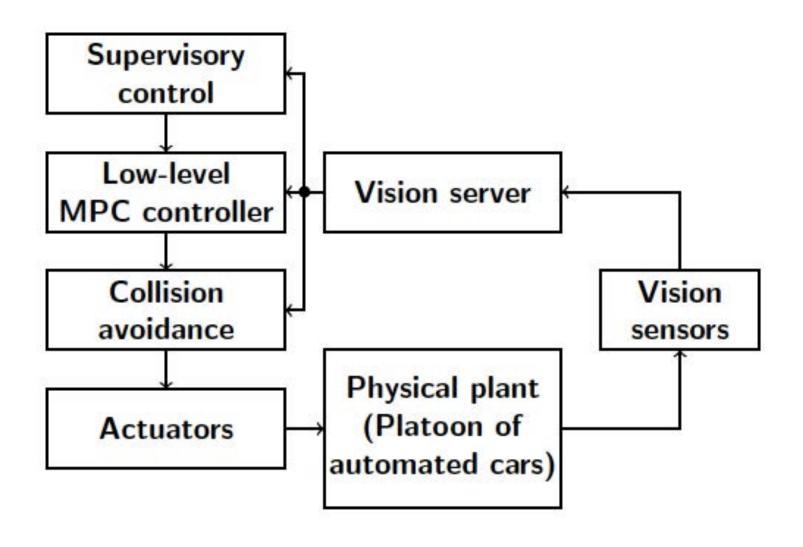
- Hackers have demonstrated remote hijack of a Jeep's digital systems over the Internet
 - Resulted in the car manufacturer recalling over a million units to patch identified security vulnerabilities
- Automated cars use various sensors
 - Ultrasound sensor to determine distance of close objects
 - mm-wave radar to map road immediately ahead
- These sensors can be jammed.
 Researchers from Zhejiang University have demonstrated such sensor attacks
- Several other demonstrations reported recently

Attacks on cars

- Car hacking is the future and sooner or later you'll be hit
 - https://www.theguardian.com/technology/2016/aug/28/car-hackingfuture-self-driving-security
- Critical reasons for crashes investigated in the national motor vehicle crash causation survey
 - https://crashstats.nhtsa.dot.gov/Api/Public/ViewPublication/812115
- Hackers Remotely Kill a Jeep On the Highway- With Me in it"
 - https://www.wired.com/2015/07/hackers-remotely-kill-jeep-highway/

Feature of daily news

Testbed architecture



System model for automatic vehicles

Plant model for vehicle i given by its kinematic equations

$$x_{i}[t+1] = x_{i}[t] + h\cos(\theta_{i}[t])v_{i}[t] + h\cos(\theta_{i}[t])w_{ix}[t]$$

$$y_{i}[t+1] = y_{i}[t] + h\sin(\theta_{i}[t])v_{i}[t] + h\sin(\theta_{i}[t])w_{iy}[t]$$

$$\theta_{i}[t+1] = \theta_{i}[t] + h\omega_{i}[t] + hw_{i\theta}[t]$$

- ♦ h is the sampling period (100ms)
- $v_i[t]$ a control input, denoting speed
- $\omega_i[t]$ a control input, denoting angular
- $w_{ix}[t]$, $\underline{w}_{iy}[t]$, $w_{i\theta}[t]$ all N(0,2), i.i.d.
- Non-linear system

Watermarked system's performance in absence of attack

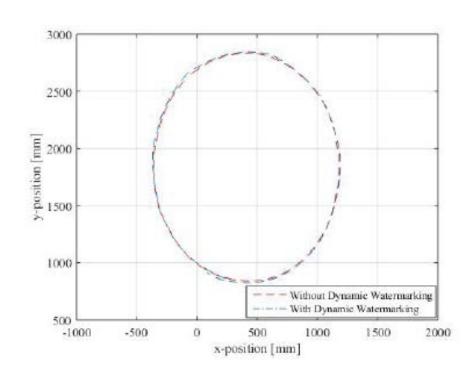
Watermarked system

$$x_{i}[t+1] = x_{i}[t] + h\cos(\theta_{i}[t])u_{i}^{g}(\mathbf{z}_{1}^{t}, \mathbf{z}_{2}^{t}) + h\cos(\theta_{i}[t])e_{iv}[t] + h\cos(\theta_{i}[t])w_{ix}[t]$$

$$y_{i}[t+1] = y_{i}[t] + h\sin(\theta_{i}[t])u_{i}^{g}(\mathbf{z}_{1}^{t}, \mathbf{z}_{2}^{t}) + h\sin(\theta_{i}[t])e_{iv}[t] + h\sin(\theta_{i}[t])w_{iy}[t]$$

$$\theta_{i}[t+1] = \theta_{i}[t] + h\omega_{i}[t] + he_{i\theta}[t] + hw_{i\theta}[t]$$

- Performance with and without watermarking
- Watermarks do not result in any added penalty on performance



Sensor attack

Sensor attack

$$z_{2x}[t_A] = x_2[t_A] + \tau, \text{ where } \tau = \text{bias}$$

$$z_{2x}[t+1] = z_{2x}[t] + h\cos(\theta_2[t])u_2^g(\mathbf{z}_1^t, \mathbf{z}_2^t) + \cos(\theta_2[t])n[t]$$

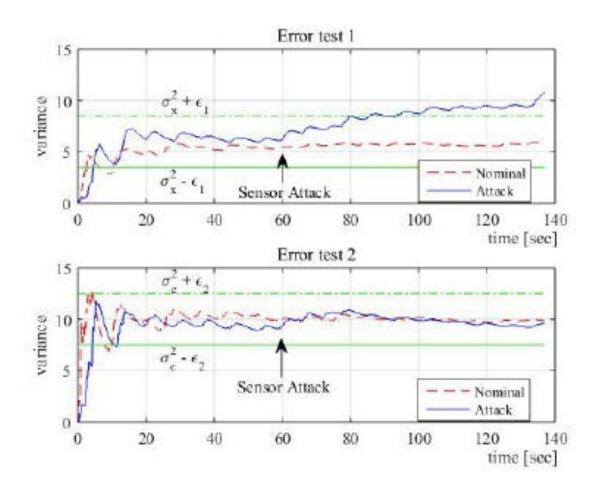
$$n[t] \sim \mathcal{N}(0, \sigma_x^2)$$

This attack passes Test 2, but fails Test 1

Test Statistics

Fails Test 1

PassesTest 2



Defending against arbitrary sensor attacks



See the video "Tackling Autonomous Vehicle Cybersecurity Issues" at https://cesg.tamu.edu/faculty/p-r-kumar/convergencelab/

Remarks

- CPS is important for society and economy
- Lot of future infrastructure may be CPS
- Societally and economically important
- Security of CPS is a very rapidly emerging area
- Critical for safety of future infrastructure
- Lots of attacks have already been demonstrated

References

- ♦ Yilin Mo, and Bruno Sinopoli, "Secure control against replay attacks," Allerton, pp. 911-918. IEEE, 2009.
- Sean Weerakkody, Yilin Mo, and Bruno Sinopoli, "Detecting integrity attacks on control systems using robust physical watermarking," 53rd IEEE Conference on Decision and Control, pp. 3757-3764. IEEE, 2014.
- Yilin Mo, Rohan Chabukswar, and Bruno Sinopoli. "Detecting integrity attacks on SCADA systems," IEEE Transactions on Control Systems Technology, no. 4 (2014): 1396-1407.
- Yilin Mo, Sean Weerakkody, and Bruno Sinopoli, "Physical authentication of control systems: designing watermarked control inputs to detect counterfeit sensor outputs," IEEE Control Systems 35, no. 1 (2015): 93-109.
- Bharadwaj Satchidanandan and P. R. Kumar, "Dynamic Watermarking: Active Defense of Networked Cyber-Physical Systems." Proceedings of the IEEE, vol. 105, No. 2, pp. 219-240, February 2017.
- Woo-Hyun Ko, Bharadwaj Satchidanandan and P. R. Kumar, "Theory and Implementation of Dynamic Watermarking for Cybersecurity of Advanced Transportation Systems." International Workshop on Cyber-Physical Systems Security (CPS-Sec), pp. 235-239, Philadelphia, October 17-19, 2016.
- Bharadwaj Satchidanandan and P. R. Kumar, "Secure Control of Networked Cyber-Physical Systems." Proceedings of 55th IEEE Conference on Decision and Control, pp. 283-289, December 12–14, 2016, Las Vegas.
- Bharadwaj Satchidanandan and P. R. Kumar, "On Minimal Tests of Sensor Veracity for Dynamic Watermarking-Based Defense of Cyber-Physical Systems." Proceedings of the 9th International Conference on Communication Systems & Networks (COMSNETS 2017), pp. 23-30, January 4-8, 2017, Bengaluru, India.
- Bharadwaj Satchidanandan and P. R. Kumar, "Defending Cyber-Physical Systems from Sensor At- tacks." In to appear in From 9th International Conference on Communication Systems & Networks (COMSNETS 2017), Lecture Notes in Computer Science, Springer–Verlag, Berlin.
- Bharadwaj Satchidanandan and P. R. Kumar, "Control Systems Under Attack: The Securable and Unsecurable Subspaces of a Linear Stochastic System." To appear in Emerging Applications of Control and System Theory, 2017.
- Pasqualetti, Fabio, Florian Dorfler, and Francesco Bullo. "Control-theoretic methods for cyberphysical security: Geometric principles for optimal cross-layer resilient control systems." IEEE Control Systems 35, no. 1 (2015): 110-127.
- Pasqualetti, Fabio, Florian Drfler, and Francesco Bullo. "Attack detection and identification in cyber-physical systems." IEEE Transactions on Automatic Control 58, no. 11 (2013): 2715- 2729.
- Pasqualetti, Fabio, Florian Drfler, and Francesco Bullo. "Cyber-physical security via geometric control: Distributed monitoring and malicious attacks." In Decision and Control (CDC), 2012 IEEE 51st Annual Conference on, pp. 3418-3425.
- Teixeira, A, Iman Shames, Henrik Sandberg, and Karl H. Johansson. "Revealing stealthy attacks in control systems." In Communication, Control, and Computing (Allerton), 2012 50th Annual Allerton Conference on, pp. 1806-1813. 2012.

Thank you