



ROYAL INSTITUTE
OF TECHNOLOGY

Quantifying Cyber Security for Networked Control Systems

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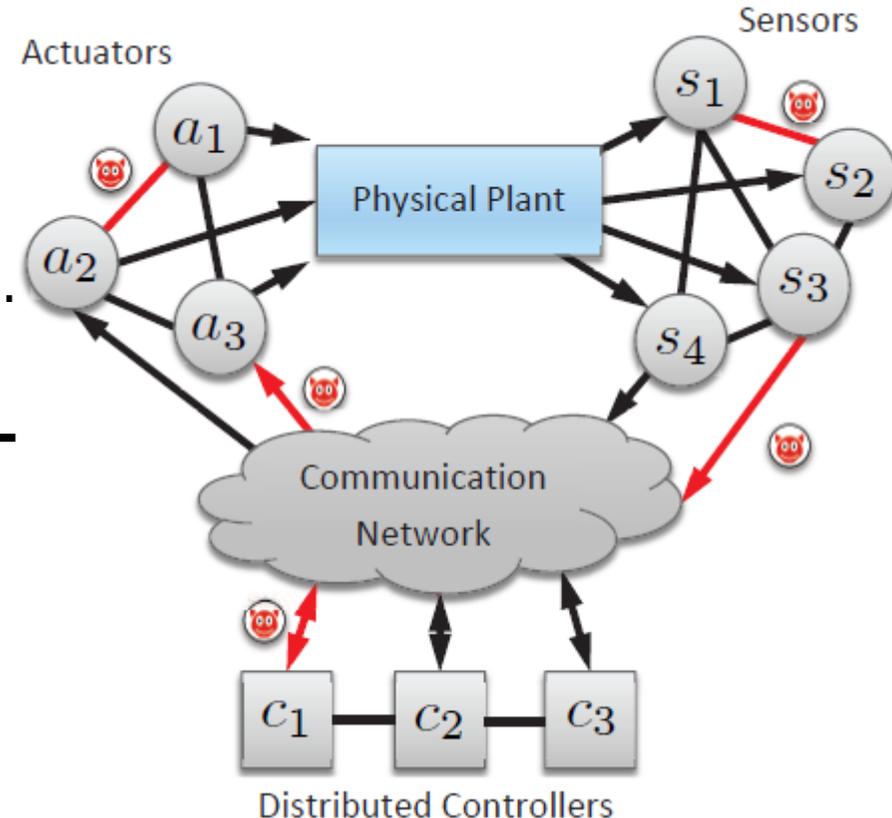


UC Berkeley
May 17th, 2013



Motivation

- Networked control systems are to a growing extent
 - based on commercial off-the-shelf components
 - integrated with data analytics environments etc.
- Leads to **increasing vulnerability to cyber-physical threats** with many potential points of attacks
- Need for tools and strategies to understand and mitigate attacks in networked control systems:
 - Which threats should we care about?
 - What impact can we expect from attacks?
 - Which resources should we protect (more)?



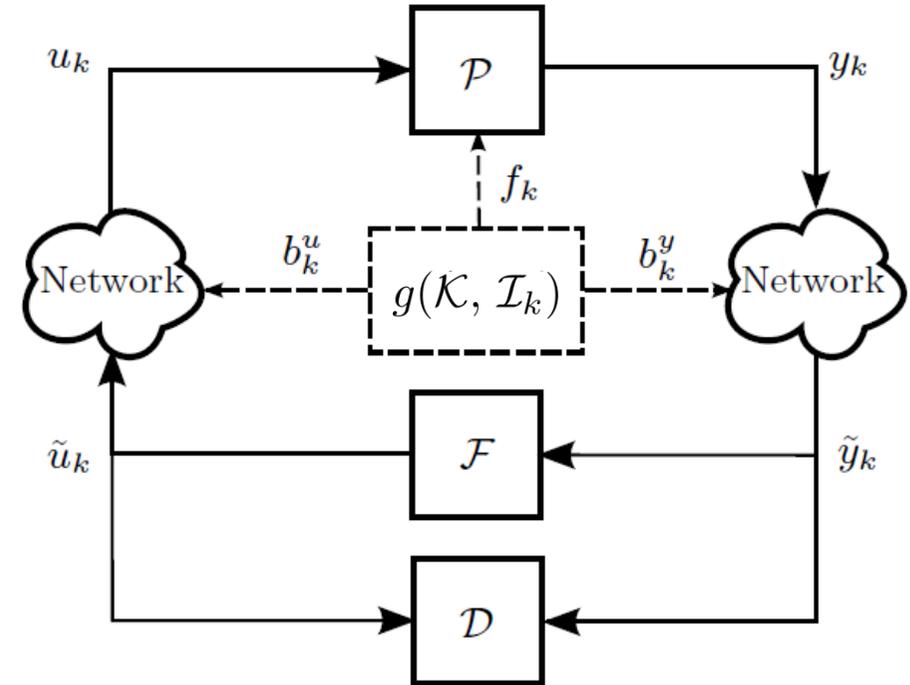
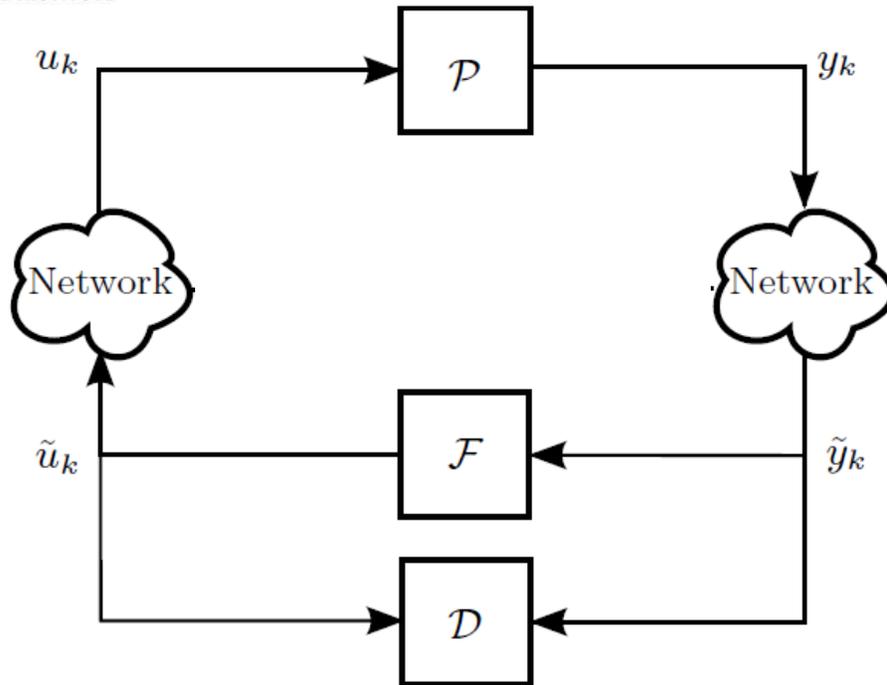
Contributions

- Adversary models for networked control systems
- Optimization tools for quantization of cyber security
 - Trade-off between protection resources and level of security
 - Trade-off between attack resources and attack impact
- Security metric for power network state estimators. Efficient computation using graph Min Cut relaxations
- Security metric for wireless LQG-controlled quadruple tank. Computation using mixed integer linear programs

Outline

- **Adversary models for networked control systems**
- Application 1: Power network state estimation
- Application 2: Wireless LQG-controlled quadruple tank

Networked Control System under Attack



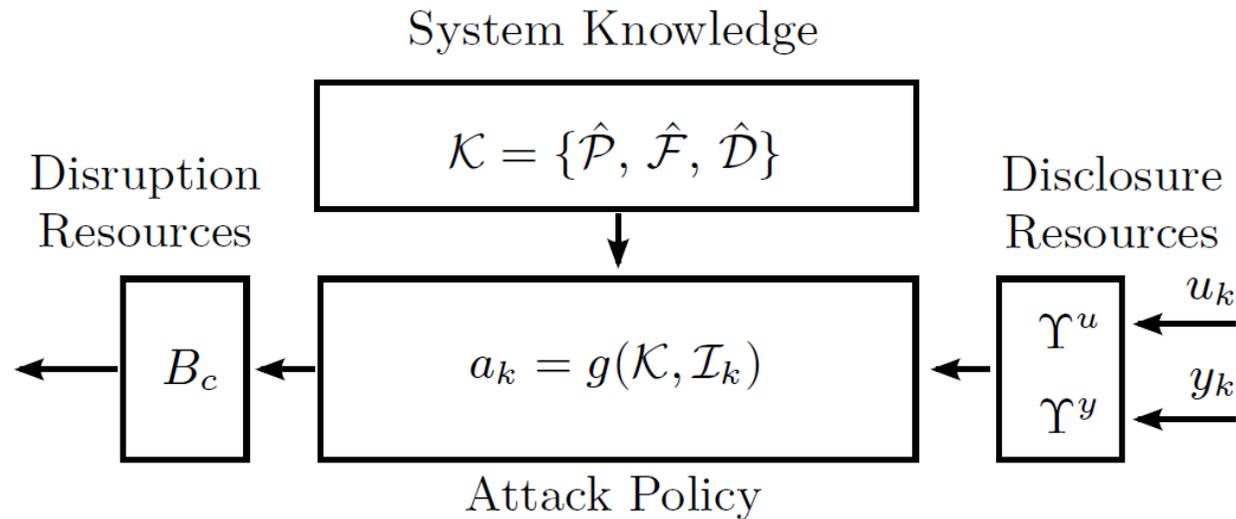
- Physical plant (\mathcal{P})
- Feedback controller (\mathcal{F})
- Anomaly detector (\mathcal{D})
- Disclosure Attacks

- Physical Attacks f_k
- Deception Attacks

$$\tilde{u}_k = u_k + \Gamma^u b_k^u$$

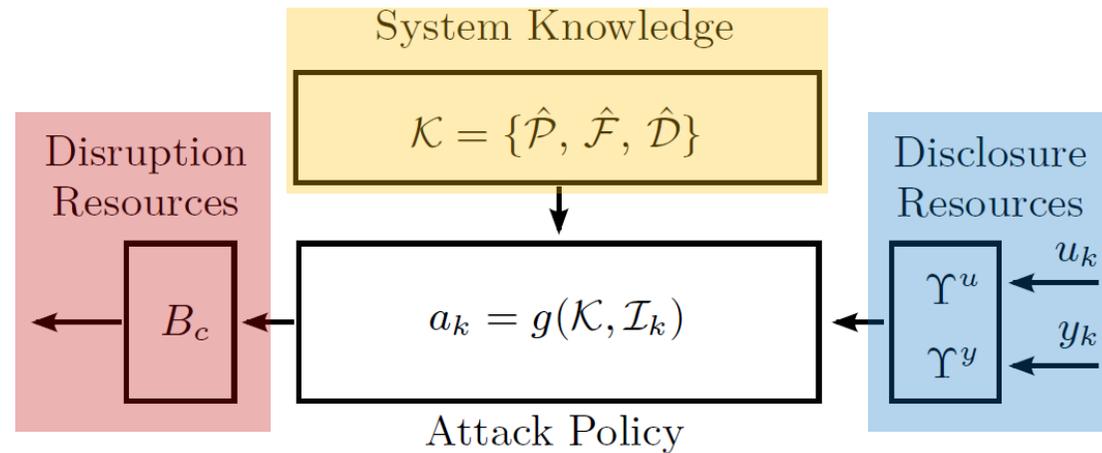
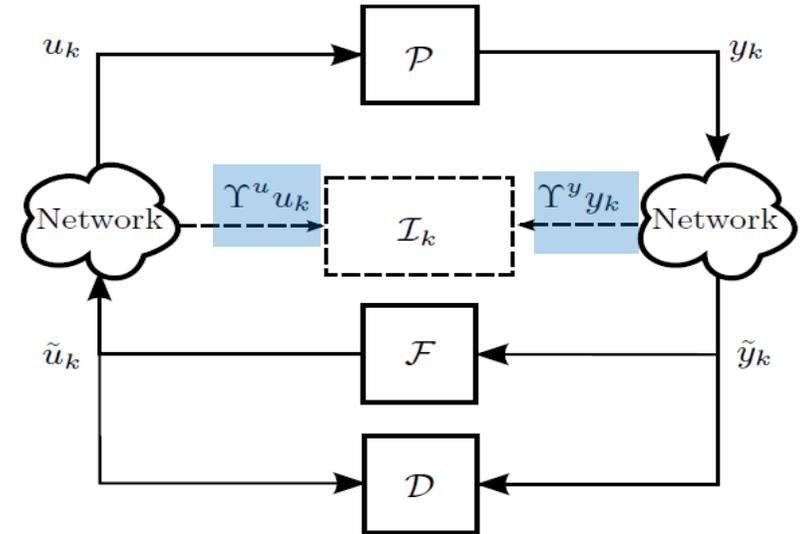
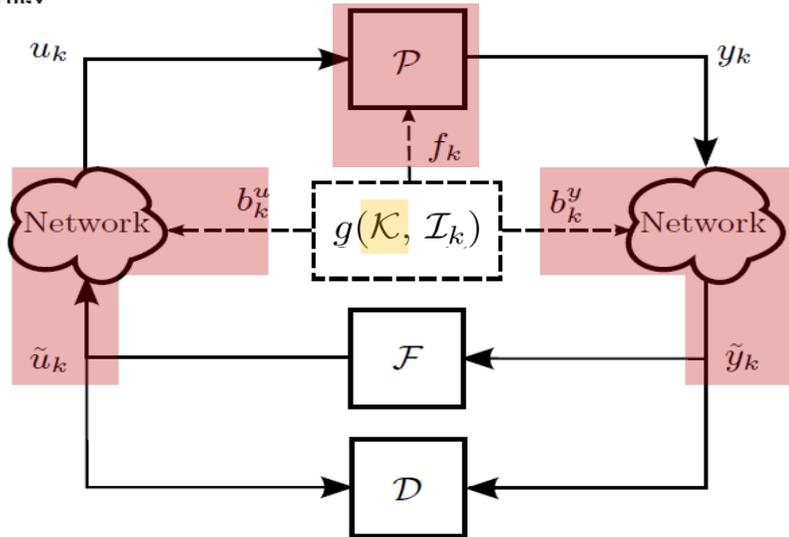
$$\tilde{y}_k = y_k + \Gamma^y b_k^y$$

Adversary Model

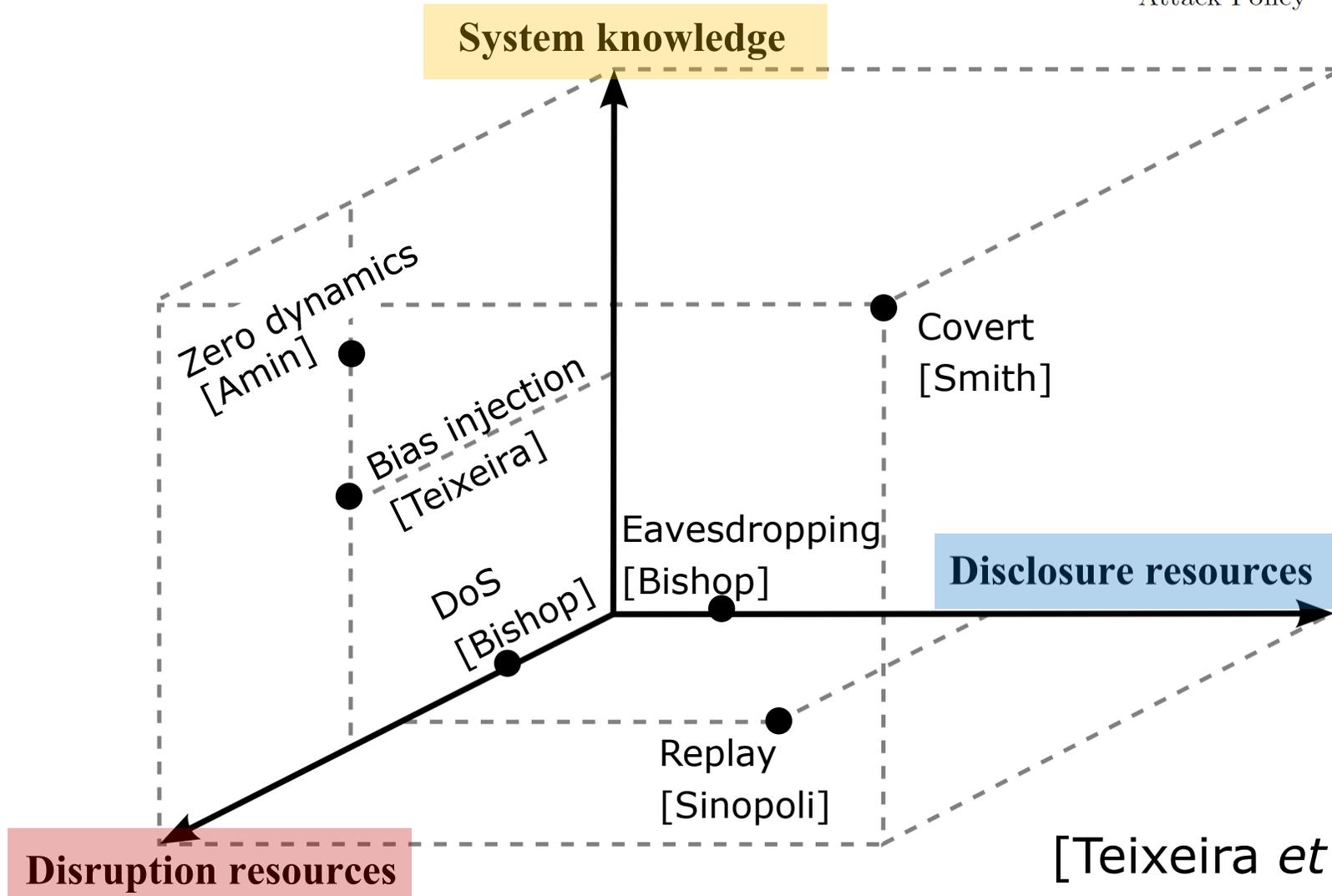
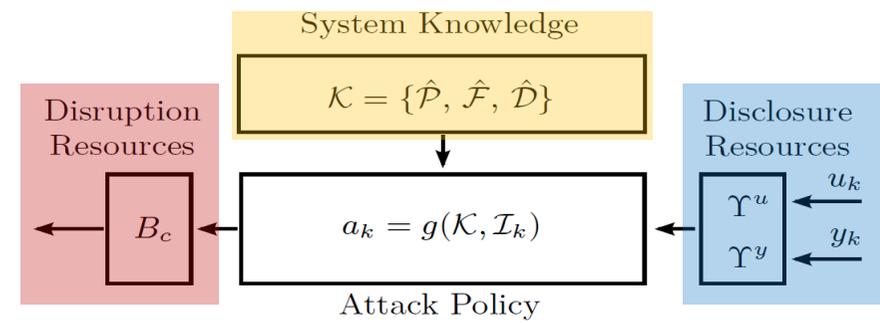


- Adversary's goal to force the process state into an unsafe region
- Attack should be stealthy, i.e., no alarms (at least until it is too late)
- Adversary constrained by limited resources

Networked Control System with Adversary Model



Attack Space

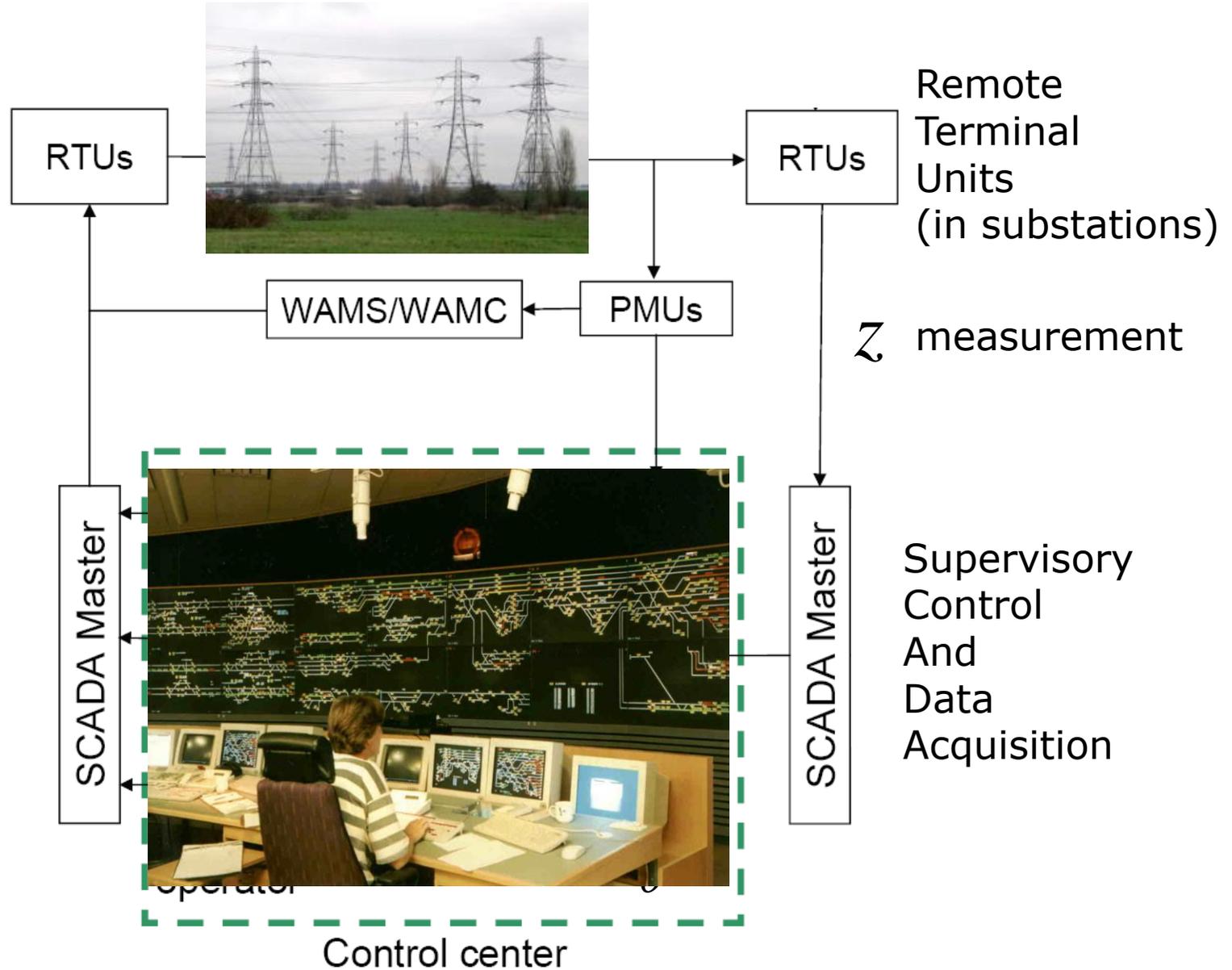


[Teixeira et al., 2012]

Outline

- Adversary models for networked control systems
- **Application 1: Power network state estimation**
 - **Security index definition**
 - **Computation with LASSO/graph Min Cut relaxations**
- Application 2: Wireless LQG-controlled quadruple tank

Power Network Control System



Model-Based State Estimation

Given redundant measurement z , find state estimate $\hat{\theta}$ based on **steady-state** model

state
 θ



measurement
 z

$$z = h(\theta)$$

Power Network State Estimation Model

States (θ)
= bus voltage **phase angles**

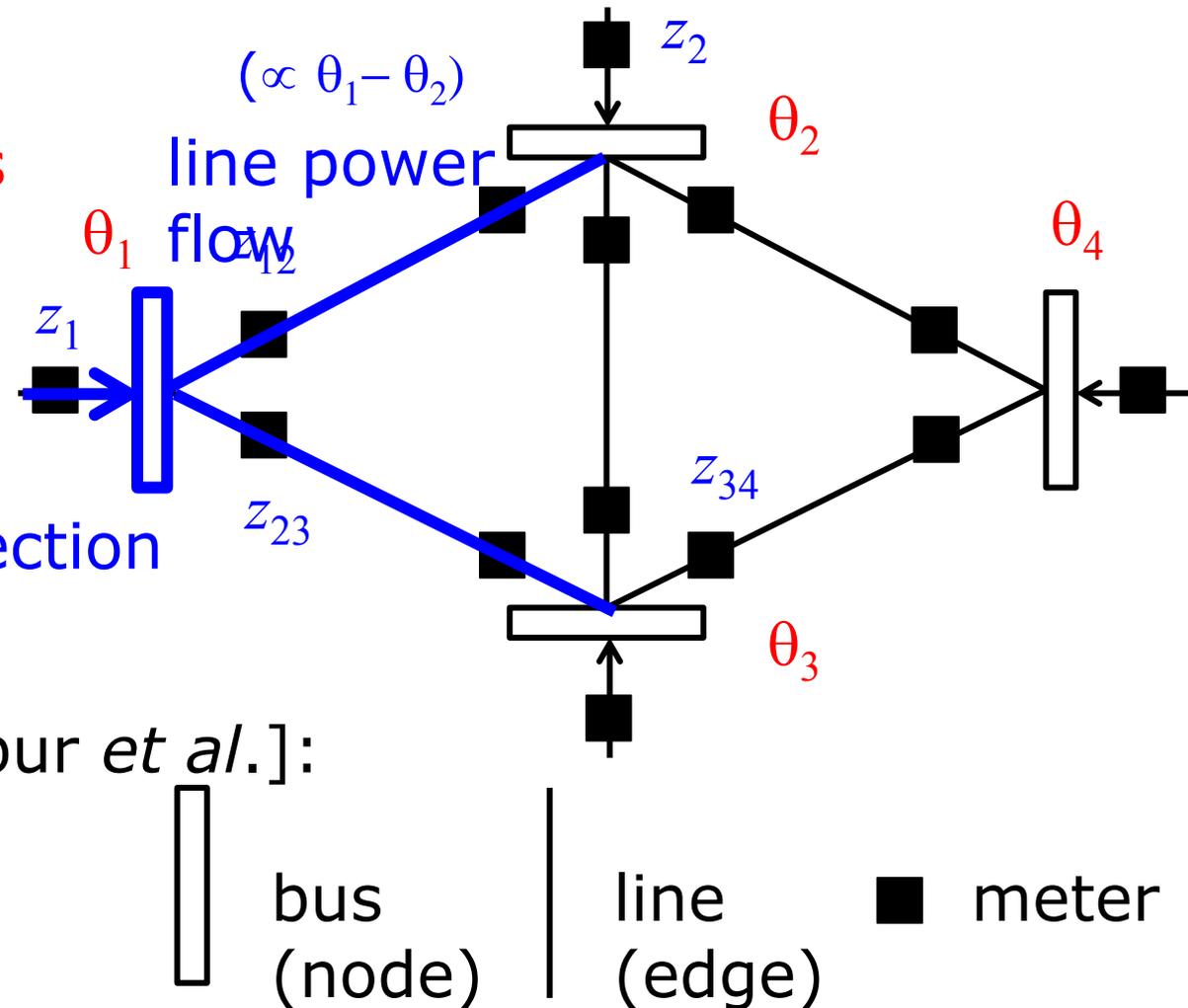
(flow conservation)
bus injection

Measurements (z)
= **line power flow & bus injection**

“DC power flow model” [Abur *et al.*]:

$$z = H \theta$$

←
measurement matrix



State Estimation by Least Squares

State estimator (LS)

$$z = H \theta$$

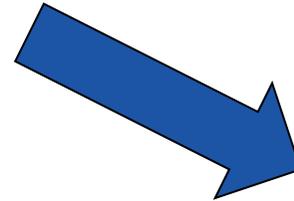
$$\Rightarrow \hat{\theta} = (H^T H)^{-1} H^T z$$

wrong



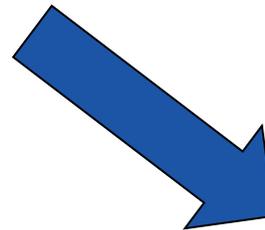
wrong

Contingency
analysis



wrong

OPF
calculations



What if the measurements were **wrong**?

$$z \leftarrow z + \Delta z \quad \text{random measurement noise}$$

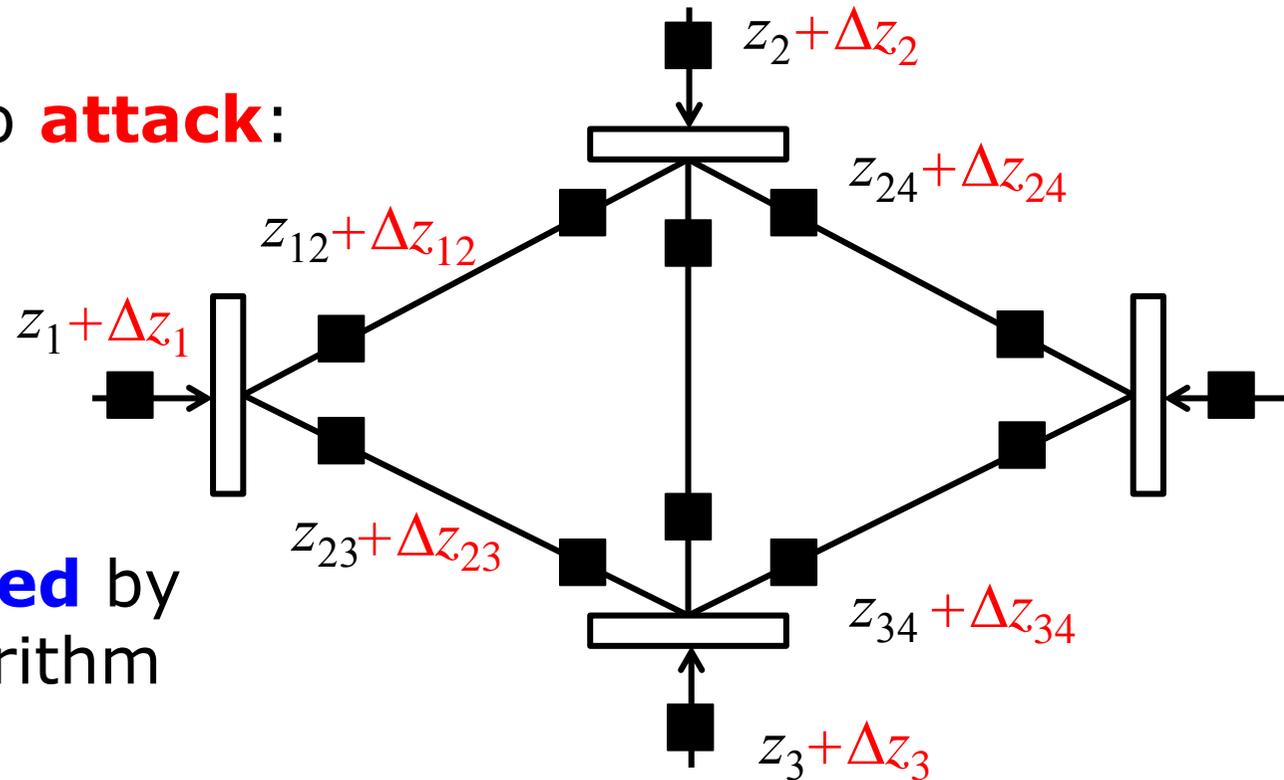
intentional data attack

Stealth Additive False-Data Attack

Measurements subject to **attack**:

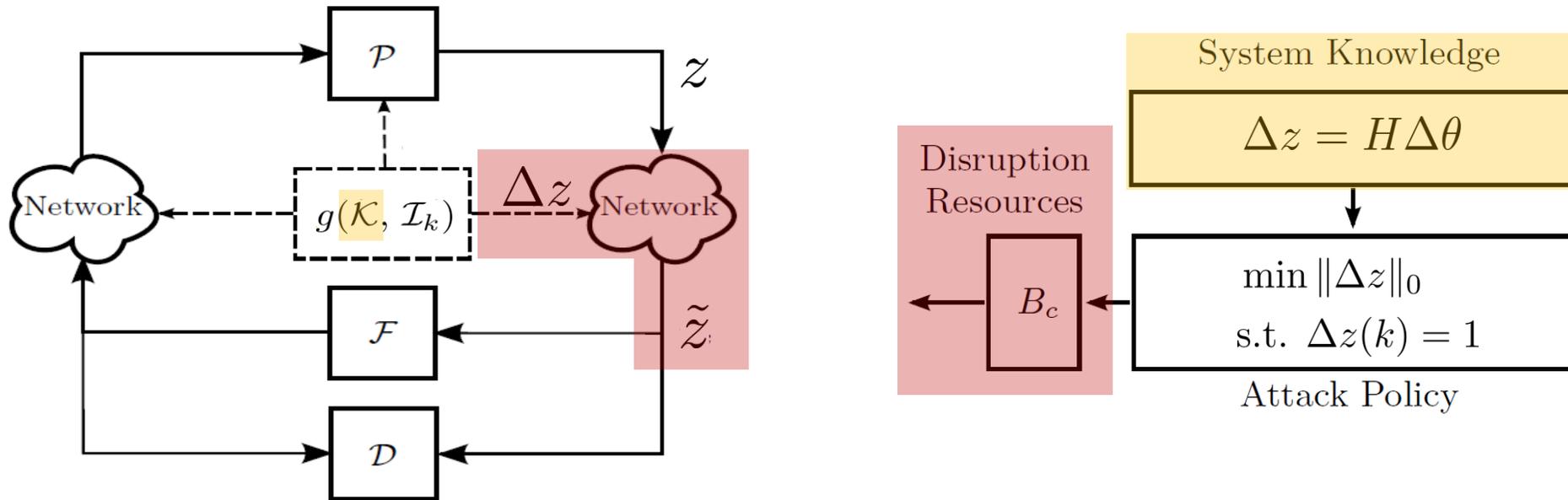
$$z \leftarrow z + \Delta z$$

Attack is **constrained**;
otherwise will be **detected** by
Bad Data Detection algorithm



$$\text{Stealth attack [Liu et al., Giani et al.]: } \Delta z = H \Delta \theta$$

Adversary Model



- Adversary's goal to induce a bias in measurement channel k
- Attack should be stealthy, i.e., no alarms
- Adversary should use minimal disruption resources

Security Index

Stealth attack $\Delta z = H\Delta\theta$

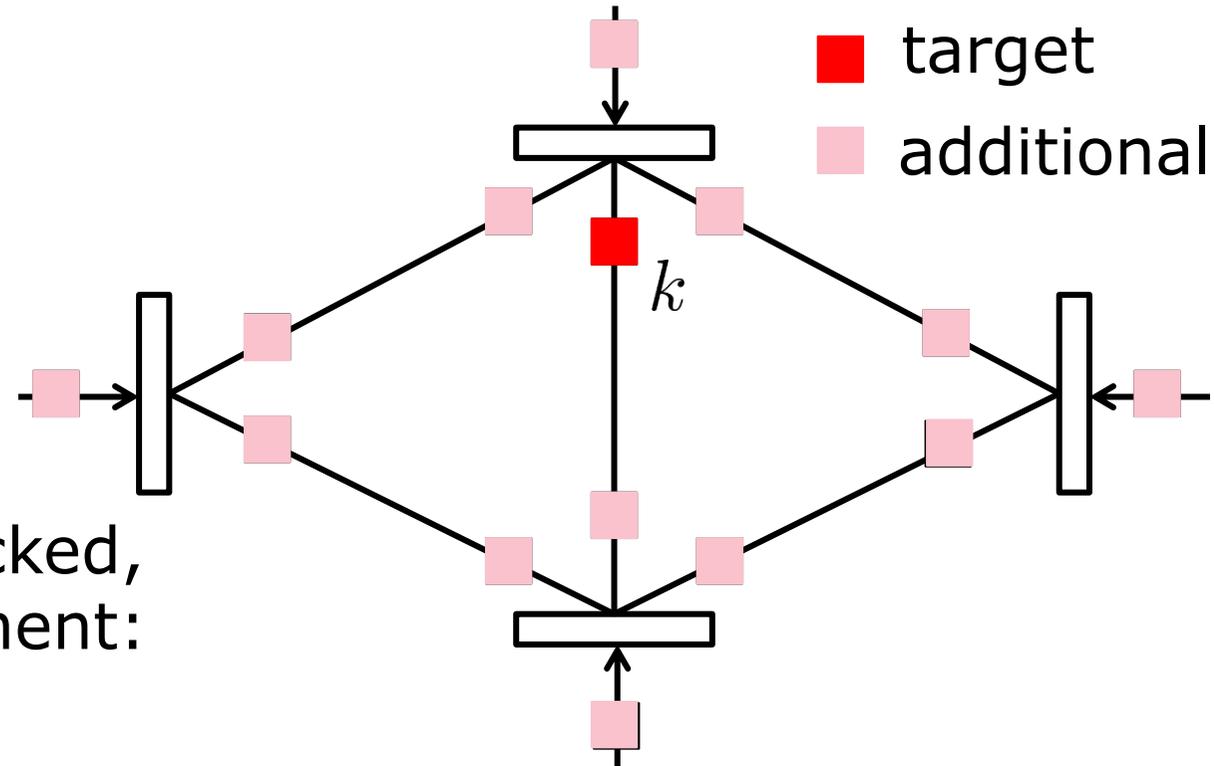
In general, $e_k \notin \text{span}(H)$

Minimum # of meters attacked,
targeting the k^{th} measurement:

$$\min_{\Delta\theta} \|H\Delta\theta\|_0$$

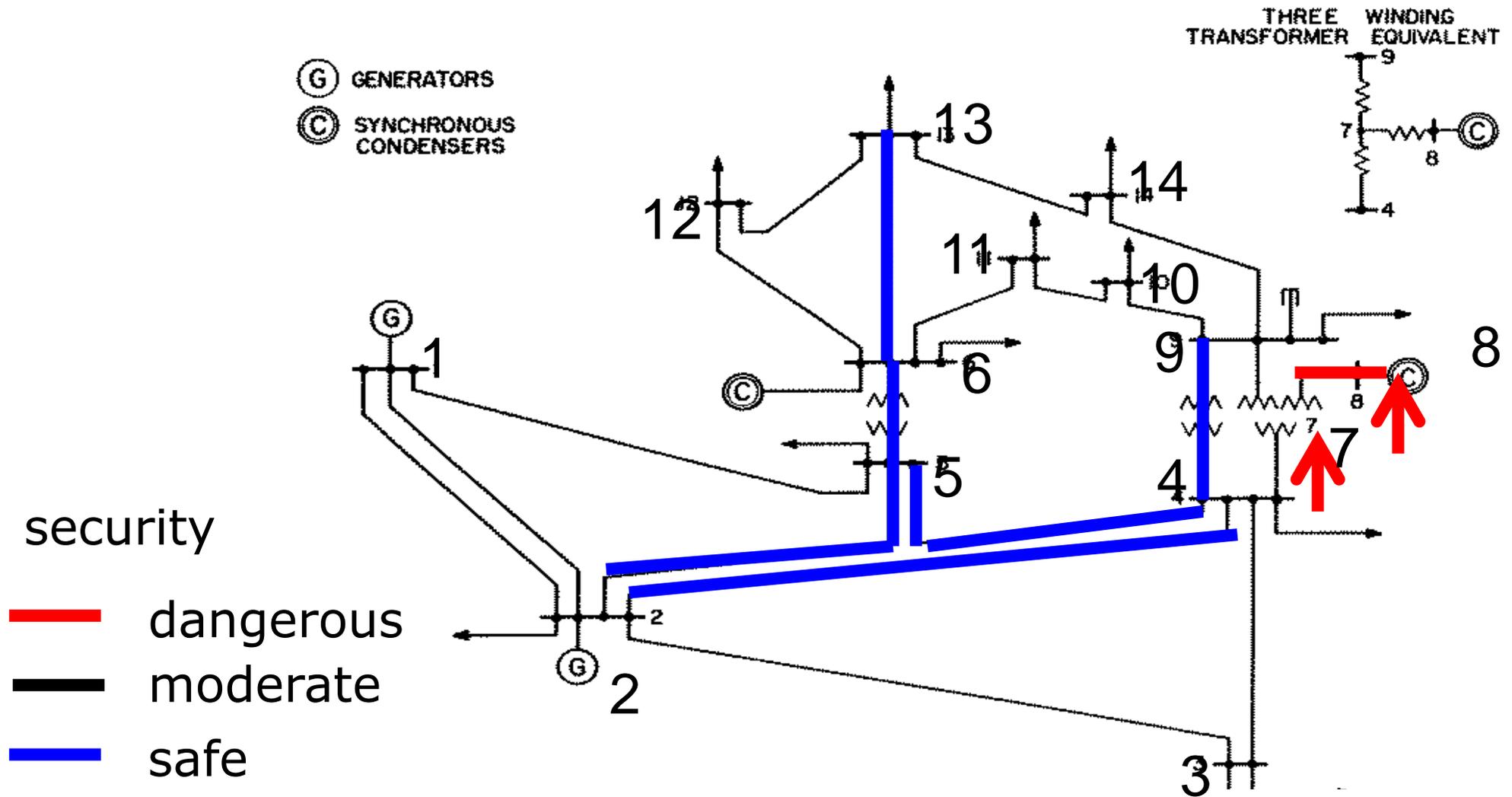
$$\text{s.t. } H(k, :) \Delta\theta = 1$$

Minimum objective value =
security index [Sandberg *et al.*]



Security index identifies network vulnerabilities

The Goal: Quantify Security

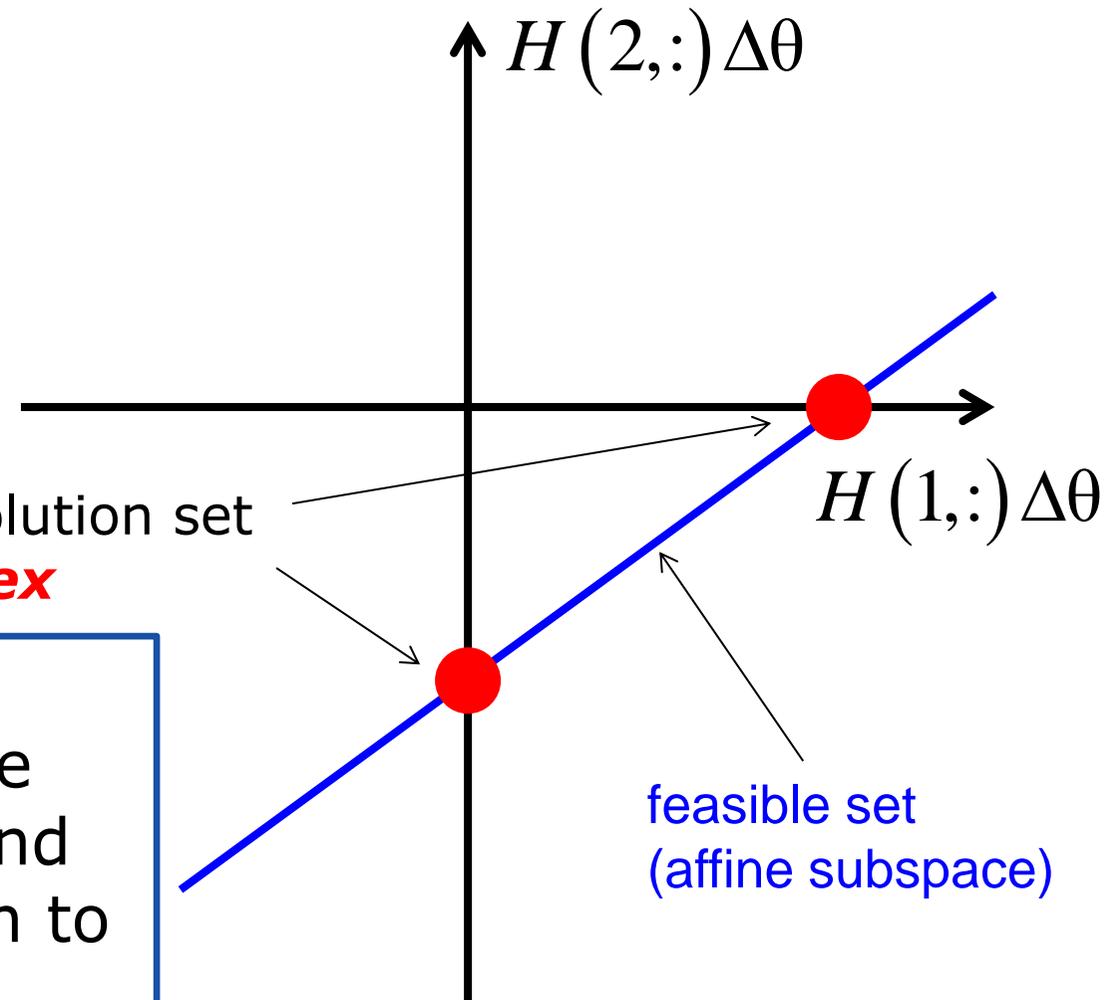


Security Index – Cardinality Minimization

Security index problem

$$\min_{\Delta\theta} \quad \|H\Delta\theta\|_0$$

$$\text{s.t.} \quad H(k,:) \Delta\theta = 1$$



Closely related to compressed sensing and computation of the **cospark** of H , see [Tillmann and Pfetsch, 2005]. Problem known to be **NP-hard** in general.



Security index problem

$$\begin{aligned} \min_{\Delta\theta} \quad & \|H\Delta\theta\|_0 \\ \text{s.t.} \quad & H(k, :) \Delta\theta = 1 \end{aligned}$$

How to solve?

Security Index Computation – MILP

$$\begin{aligned} \min_{\Delta\theta} \quad & \|H\Delta\theta\|_0 \\ \text{s.t.} \quad & H(k,:) \Delta\theta = 1 \end{aligned}$$

$$\begin{aligned} \min_{\Delta\theta, y} \quad & \sum_i y(i) \\ & -My \leq H\Delta\theta \leq My \\ \text{s.t.} \quad & H(k,:) \Delta\theta = 1 \\ & y(i) \in \{0,1\} \quad \forall i \end{aligned}$$

- ❑ Cardinality minimization problem
- ❑ Mixed integer linear program (MILP)
- ❑ **Exact** solution (solver: CPLEX)
- ❑ Solution algorithm **not scalable**

MILP formulation

Security Index Computation – LASSO

$$\begin{aligned} \min_{\Delta\theta} \quad & \|H\Delta\theta\|_1 \\ \text{s.t.} \quad & H(k,:) \Delta\theta = 1 \end{aligned}$$

$$\begin{aligned} \min_{\Delta\theta, y} \quad & \sum_i y(i) \\ \text{s.t.} \quad & -y \leq H\Delta\theta \leq y \\ & H(k,:) \Delta\theta = 1 \\ & y(i) \in \mathfrak{R} \quad \forall i \end{aligned}$$

LP formulation

- Convex linear program (LP)
- Known as LASSO
- Approximate** solution
- Less expensive to solve

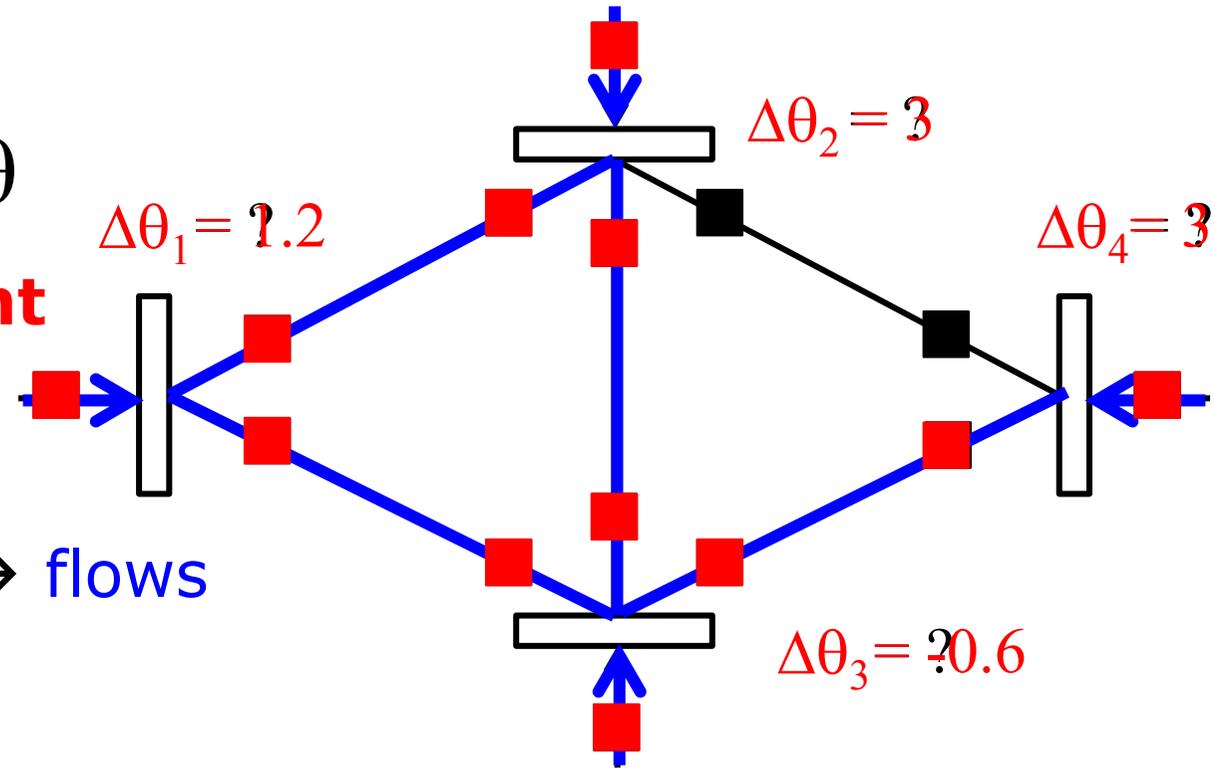
The Challenge

- Can we find solutions as **accurately** as MILP, and **faster** than LASSO?
- For general H , the answer is no (problem NP-hard)
- Let us exploit DC-power flow structure of H and make a **full measurement assumption**
- Specialize into graph problems with accurate and efficient algorithms

Graph Interpretation of Stealth Attack

Stealth attack $\Delta z = H\Delta\theta$
 = **phase angle assignment**

Phase angle differences \rightarrow flows



attack cost $\|H\Delta\theta\|_0 =$ **# of meters with nonzero flows**

Binary Phases Assignment is Optimal

No phase angle difference \rightarrow No flows  Attack cost = 0
 No attack...

Next guess: (0,1) phase angle assignment?

Theorem: Optimal $\Delta\theta_i$ can be restricted to 0 or 1, for all i

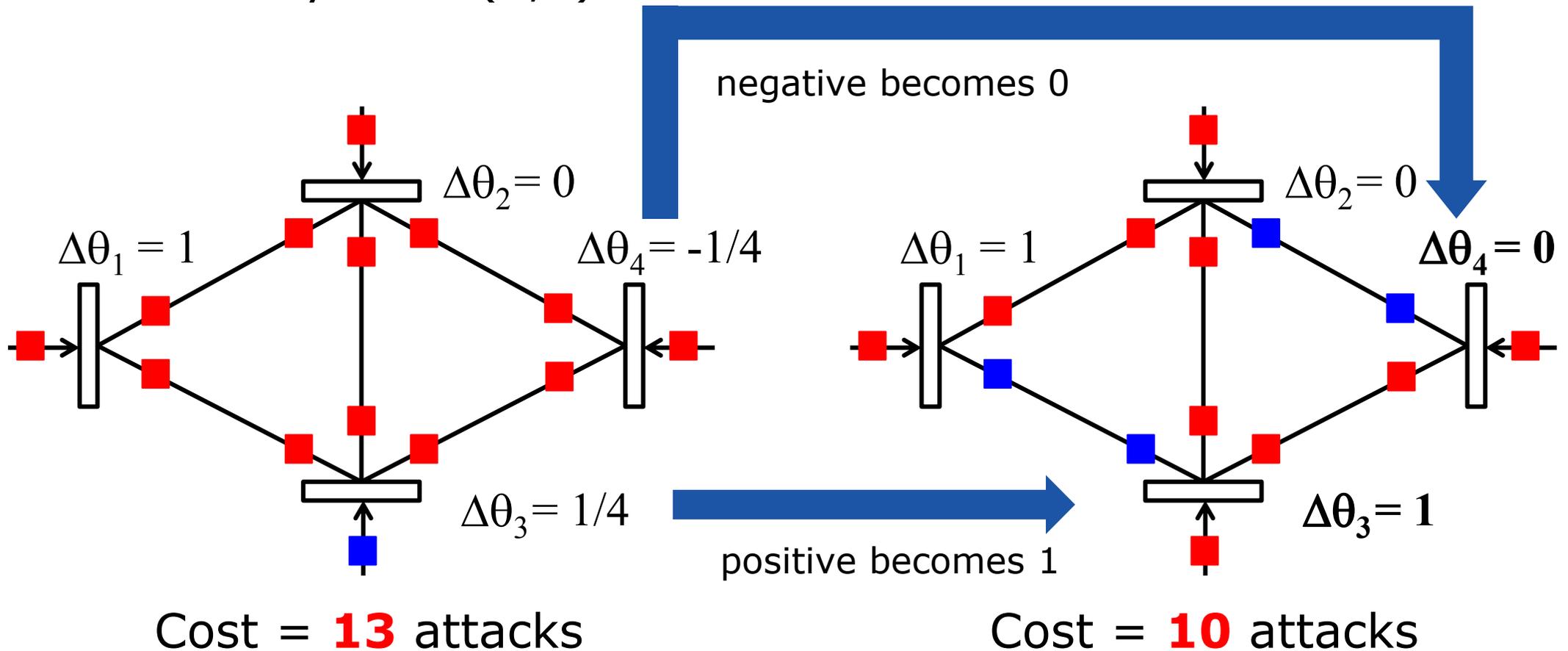
$$\min_{\Delta\theta} \|H\Delta\theta\|_0 \quad \longleftrightarrow \quad \min_{\Delta\theta} \|H\Delta\theta\|_0$$

$$\text{s.t.} \quad H(k, :) \Delta\theta = 1 \quad \text{s.t.} \quad H(k, :) \Delta\theta = 1$$

$$\Delta\theta_i \in \{0, 1\}$$

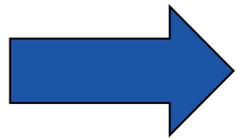
Binary Optimal Solution Justification

Can always find (0,1) feasible solution with no worst cost

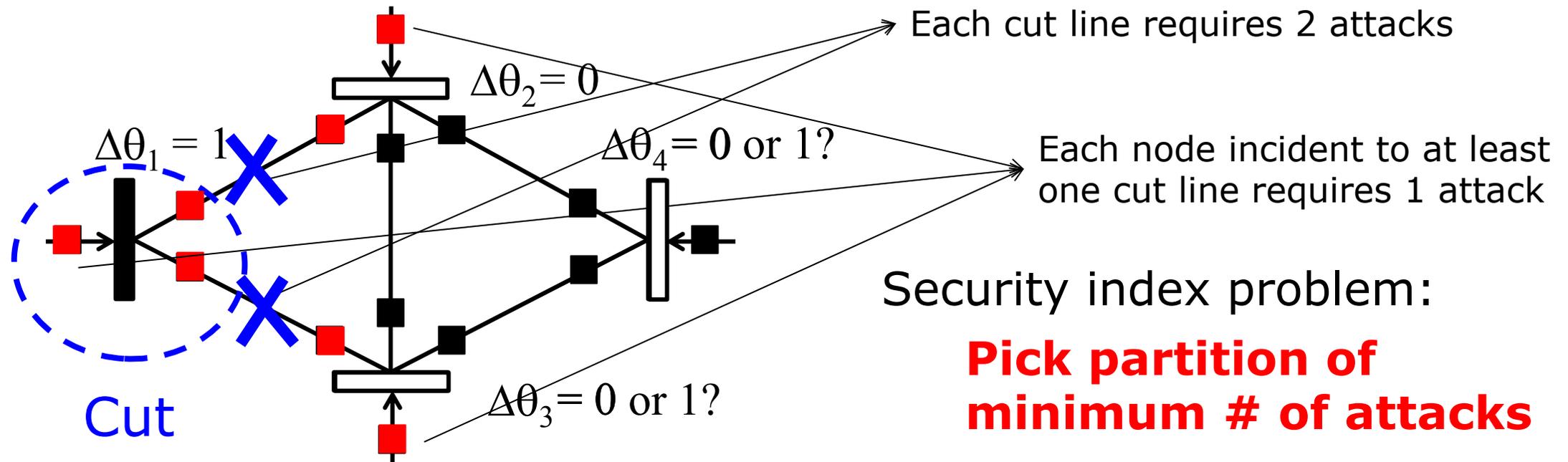


Reformulation as Node Partitioning

Optimal $\Delta\theta_i$ can be restricted to 0 or 1, for all i



Phase angle assignment becomes **node partitioning**

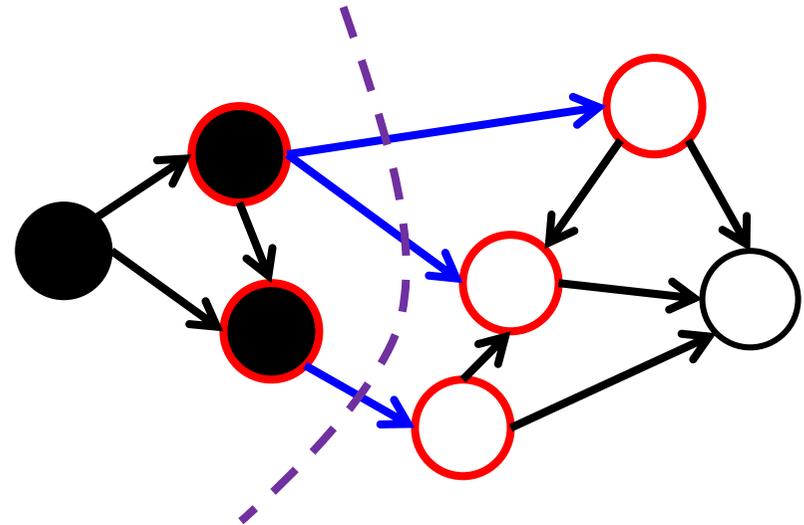


Security index problem

$$\begin{aligned} \min_{\Delta\theta} \quad & \|H\Delta\theta\|_0 \\ \text{s.t.} \quad & H(k, :) \Delta\theta = 1 \end{aligned}$$



Generalized Min Cut problem



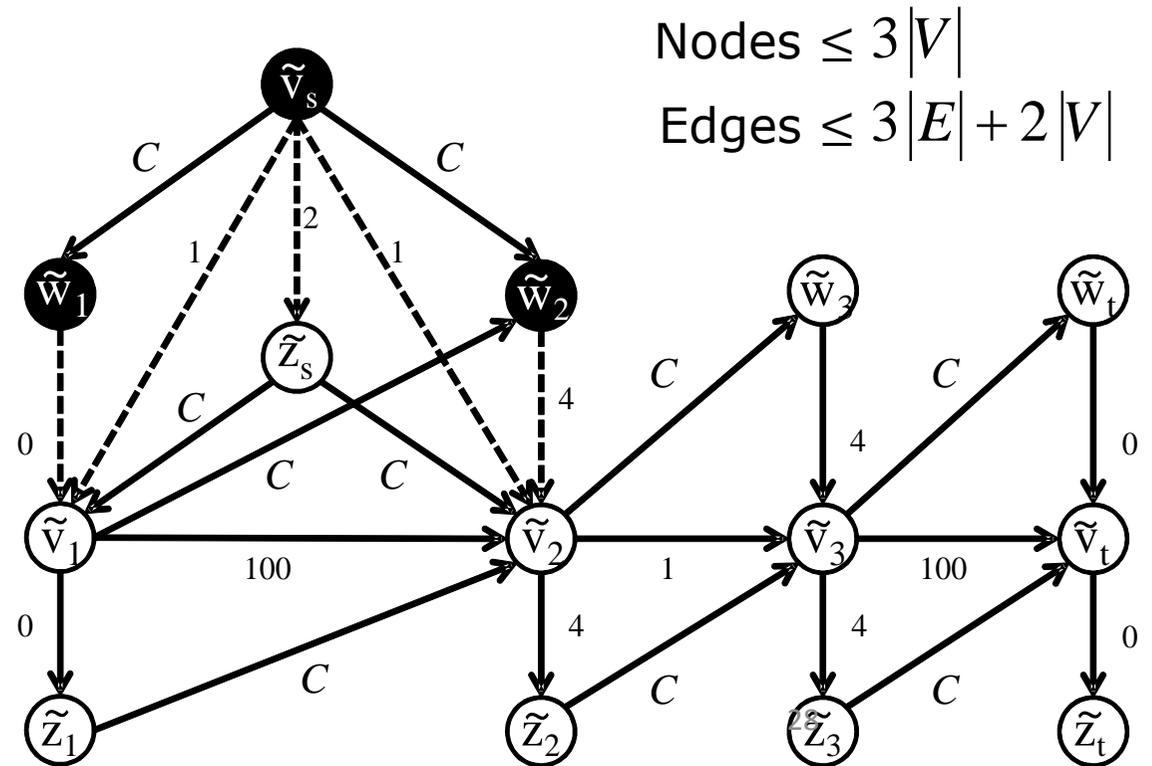
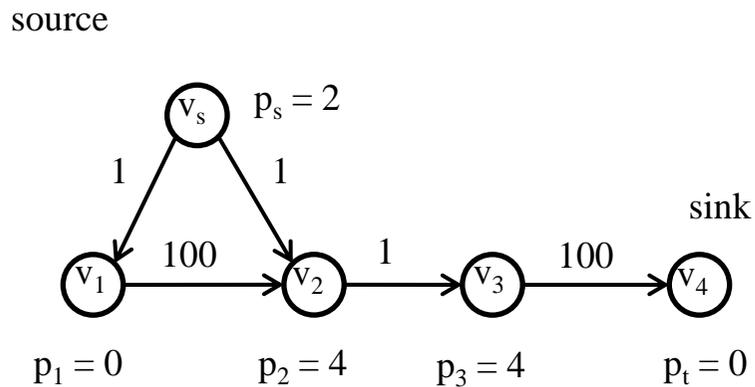
How to solve generalized Min Cut?

Standard Min Cut on Appended Graph

Generalized Min Cut = Standard Min Cut on **appended** graph

generalized min cut \longleftrightarrow standard min cut appended graph

$|V|$ nodes
 $|E|$ edges



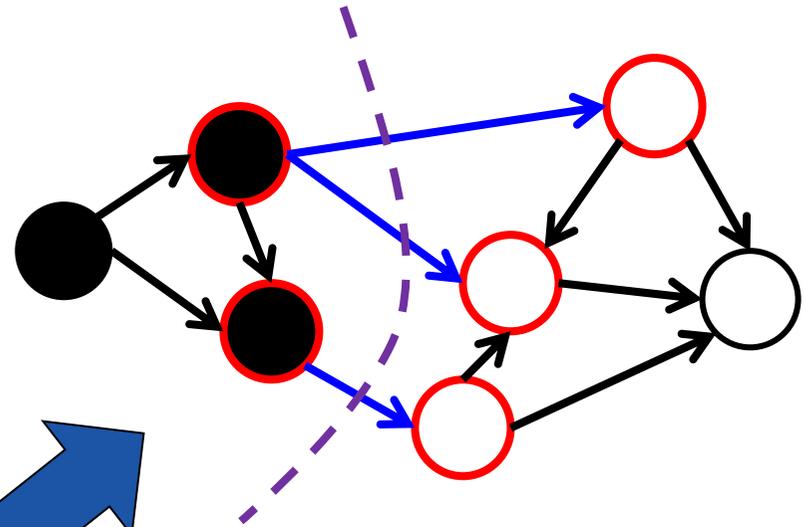
Security Index Problem – Summary

Security index problem

$$\begin{aligned} \min_{\Delta\theta} \quad & \|H\Delta\theta\|_0 \\ \text{s.t.} \quad & H(k,:) \Delta\theta = 1 \end{aligned}$$

Practical
implications?

Generalized Min Cut problem



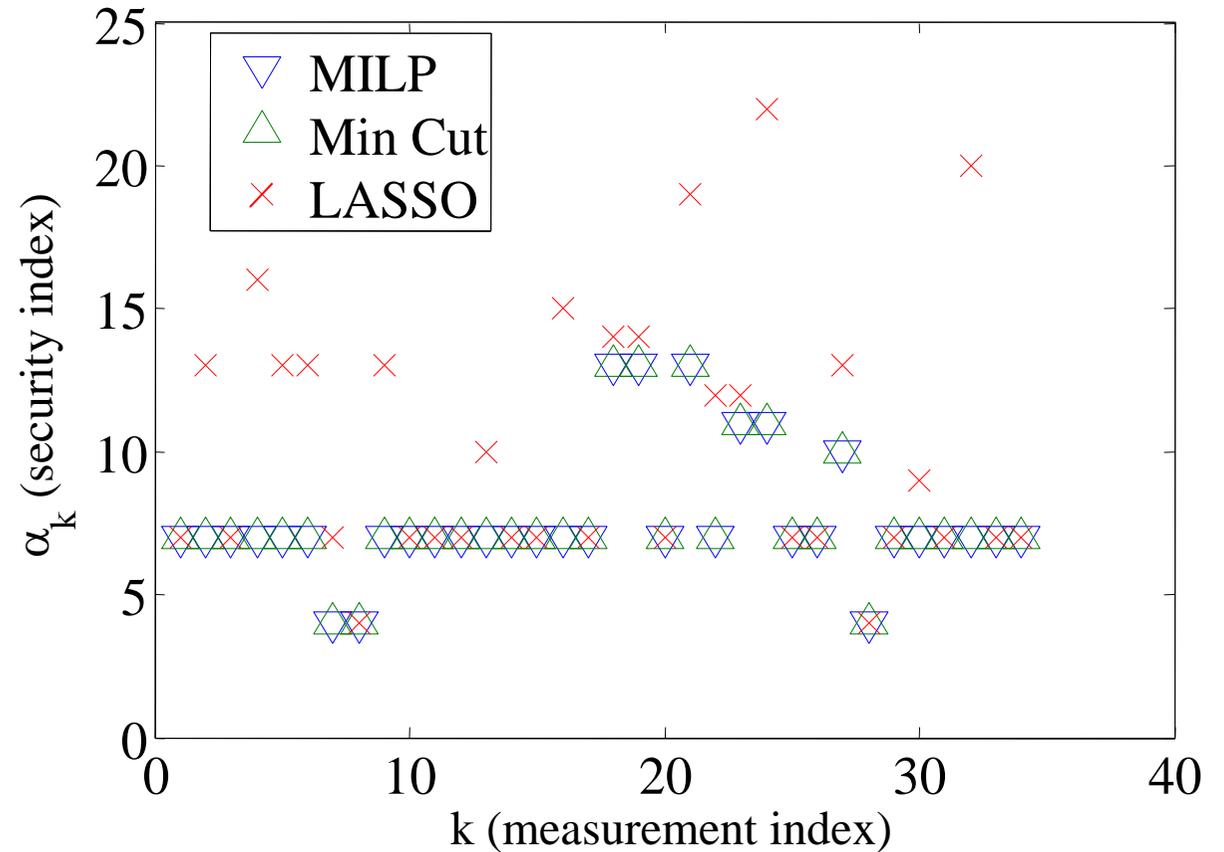
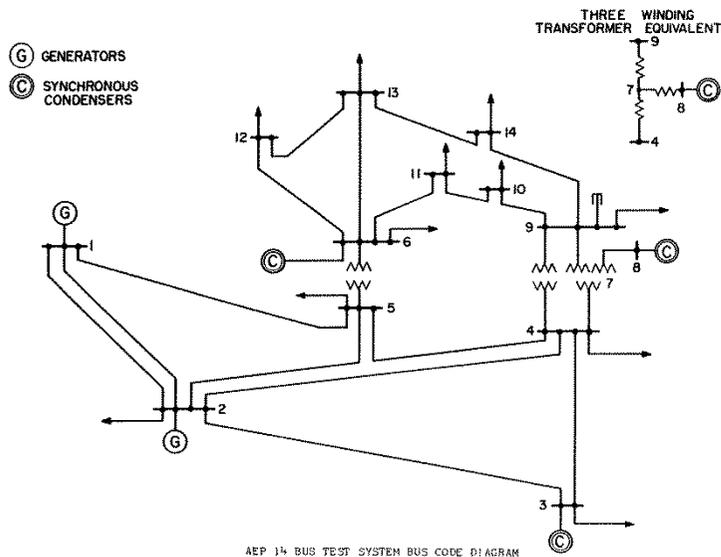
Standard Min Cut problem
on an **appended** graph

[Sou *et al.*, 2011]
[Hendrickx *et al.*,
2013]

>> [maxflow,mincut] = max_flow(A,source,sink);

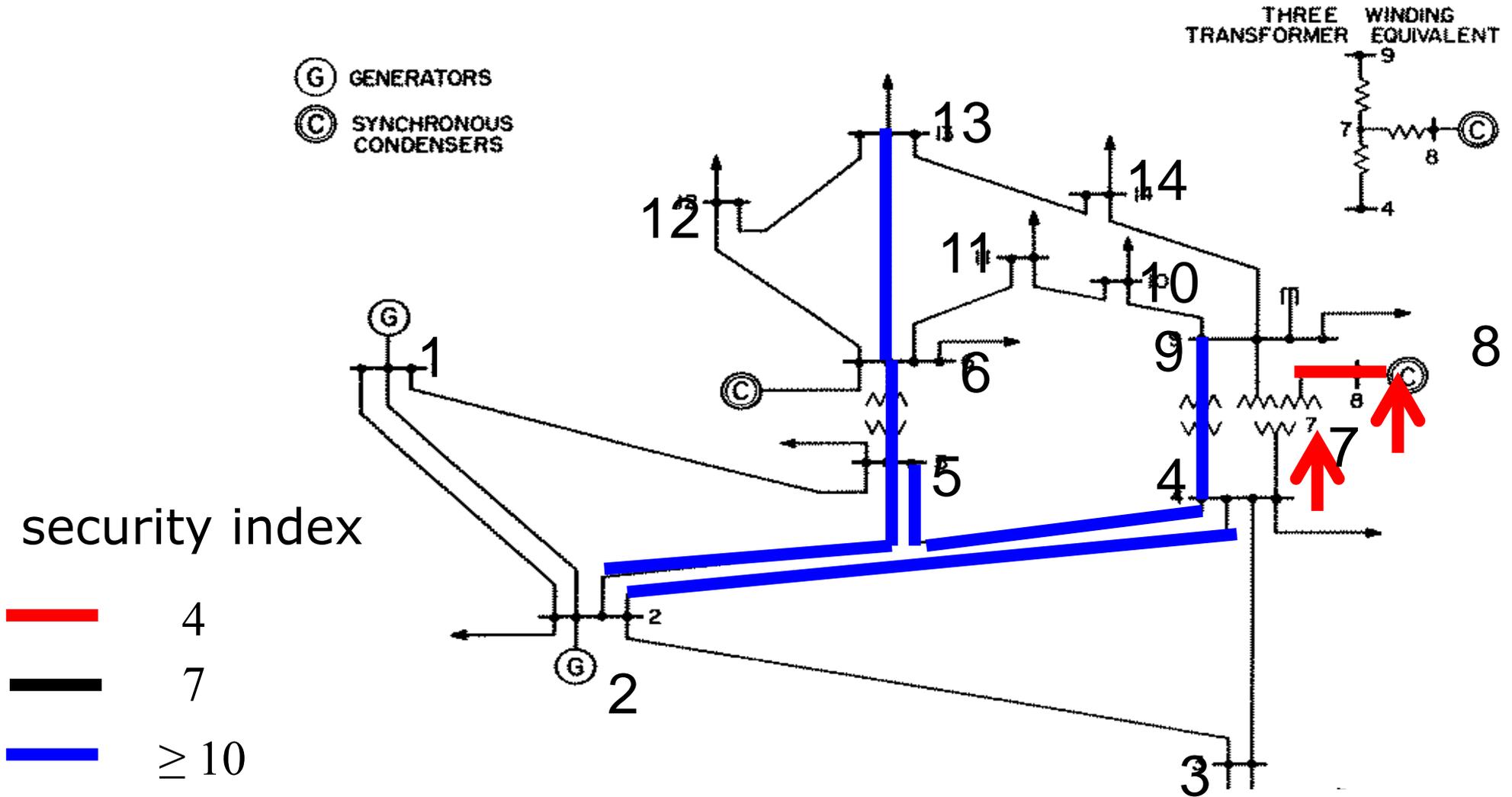
IEEE 14 Bus Benchmark Test Result

Security indices for all measurements



Solve time: MILP 1.1s; LASSO 0.6s; Min Cut 0.02s

IEEE 14 Bus Vulnerable Measurements



IEEE 118, 300, 2383 Bus Benchmarks

Min Cut solution is **exact**

Solve time comparison:

Method/Case	118 bus	300 bus	2383 bus
MILP	763 sec	6708 sec	About 5.7 days
Min Cut	0.3 sec	1 sec	31 sec

What about LASSO (1-Norm Relaxation)?

$$\begin{aligned} \min_{\Delta\theta} \quad & \|H\Delta\theta\|_1 \\ \text{s.t.} \quad & H(k,:) \Delta\theta = 1 \end{aligned}$$

We have seen LASSO relaxation in general yields non-optimal solution

Will LASSO ever work?

Yes, when H is **totally unimodular!** [Sou *et al.*, 2013]

Totally Unimodular Matrices

A matrix is totally unimodular
= determinant of all square sub-matrices are -1,0,1

network incidence matrix

$$\begin{bmatrix} 1 & 1 & 0 & 0 & -1 \\ -1 & 0 & -1 & 1 & 0 \\ 0 & -1 & 1 & -1 & 1 \end{bmatrix}$$

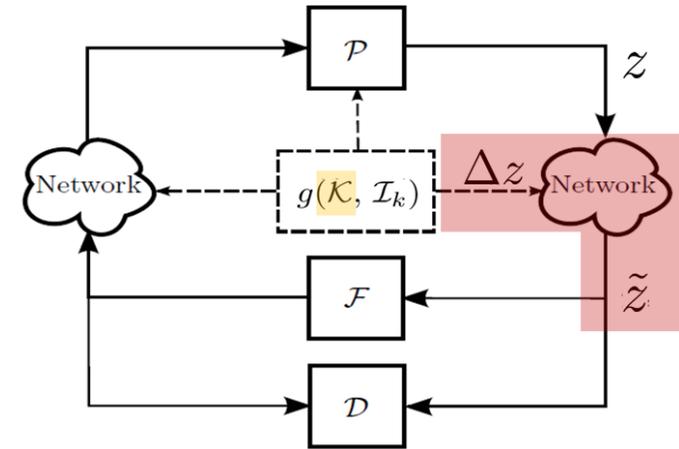
consecutive one matrix

$$\begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 \end{bmatrix}$$

Corresponds to full flow measurements
(no bus injection measurements)

Summary – Power Network State Estimation

- Adversary model
 - Induce measurement bias undetected
 - DC-power flow model known
 - Minimum disruption resources desired
- Security index problem yields lower bounds on required disruption resources. Suggests protection strategy [Vukovic *et al.*, 2012]
- Security index computation in general NP-hard. Under appropriate assumptions graph Min Cut relaxation works very well

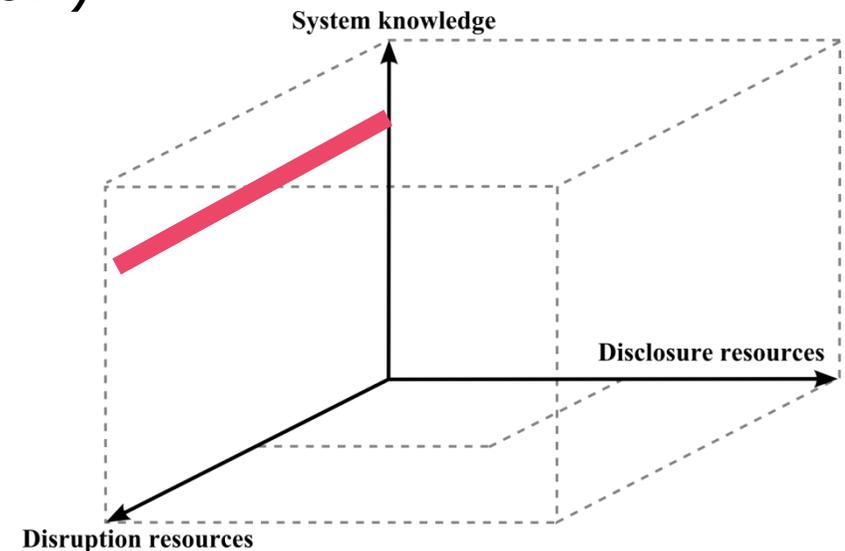


Outline

- Adversary models for networked control systems
- Application 1: Power network state estimation
- **Application 2: Wireless LQG-controlled quadruple tank**
 - **Max-impact/min-resource attacks**

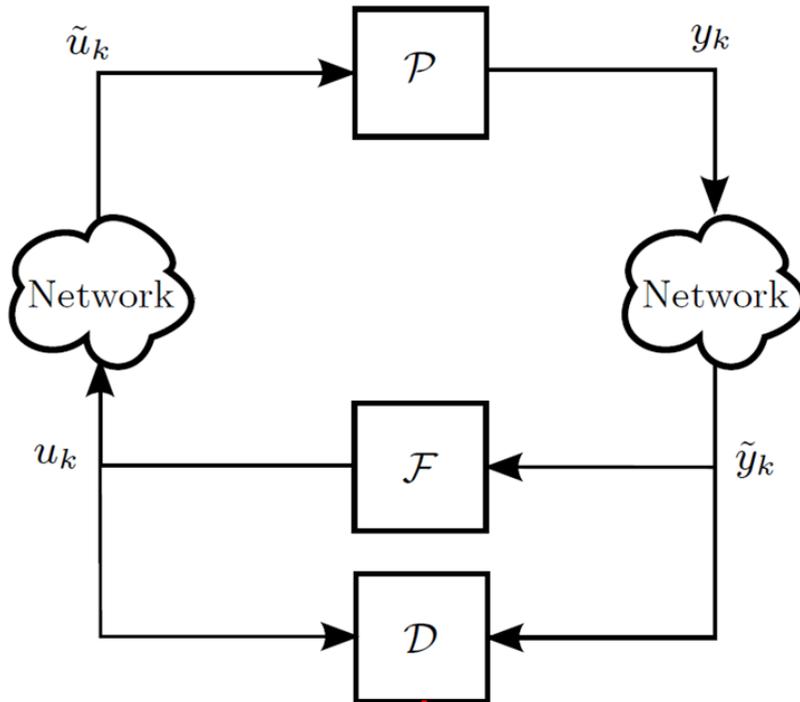
Extension to Dynamical Systems

- Attacker needs to satisfy constraints not only across channels (*spatial dimension*) but also constraints across time (*temporal dimension*)
- Cases considered:
 1. Minimum resource attacks
 2. Maximum impact attacks
 - 3. Maximum impact bounded resource attacks**



[Teixeira *et al.*, 2013]

Dynamical Networked Control System



$$\|r_k\| > \delta_r + \delta_\alpha ? \downarrow \text{Alarm}$$

- Physical Plant

$$\mathcal{P} : \begin{cases} x_{k+1} = Ax_k + B\tilde{u}_k + Gw_k \\ y_k = Cx_k + v_k \end{cases}$$

- Feedback Controller

$$\mathcal{F} : \begin{cases} z_{k+1} = A_c z_k + B_c \tilde{y}_k \\ u_k = C_c z_k + D_c \tilde{y}_k \end{cases}$$

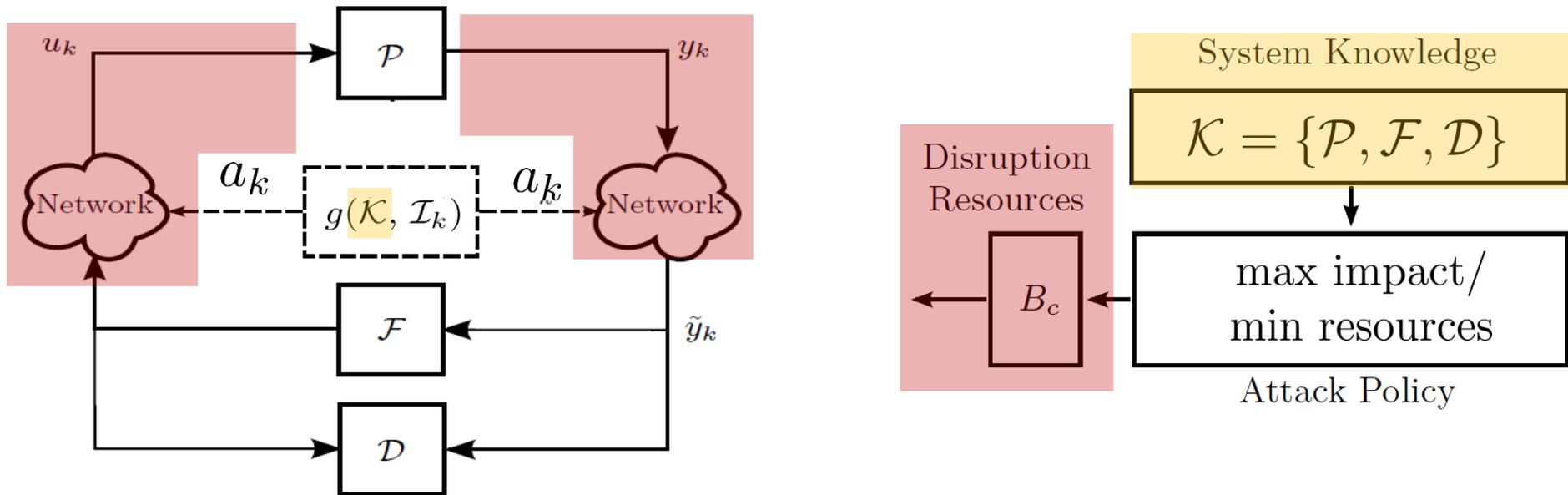
- Anomaly Detector

$$\mathcal{D} : \begin{cases} \hat{x}_{k|k} = A\hat{x}_{k-1|k-1} + Bu_{k-1} + K(\tilde{y}_k - \hat{y}_{k|k-1}) \\ r_k = V(\tilde{y}_k - \hat{y}_{k|k}) \end{cases}$$

- Alarm triggered if

$$\|r_k\| > \delta_r + \delta_\alpha$$

Adversary Model



- Adversary's goal is to force the process state into an unsafe region
- Attack should be stealthy, i.e., no alarms
- Adversary constrained by limited resources

The Dynamical Systems Case (1)

Dynamical anomaly detector for closed-loop system:

$$\begin{aligned}\xi_{k+1} &= \mathbf{A}_e \xi_k + \mathbf{B}_e a_k + \mathbf{G}_e w_k \\ r_k &= \mathbf{C}_e \xi_k + \mathbf{D}_e a_k + \mathbf{H}_e v_k\end{aligned}$$

Lift to time interval $[0, N]$

with zero-initial conditions and no noise:

$$\underbrace{\begin{bmatrix} r_0 \\ r_1 \\ r_2 \\ \vdots \\ r_N \end{bmatrix}}_{\mathbf{r}} = \underbrace{\begin{bmatrix} \mathbf{D}_e & 0 & \dots & 0 \\ \mathbf{C}_e \mathbf{B}_e & \mathbf{D}_e & \dots & 0 \\ \mathbf{C}_e \mathbf{A}_e \mathbf{B}_e & \mathbf{C}_e \mathbf{B}_e & \dots & 0 \\ \vdots & \vdots & \ddots & 0 \\ \mathbf{C}_e \mathbf{A}_e^{N-1} \mathbf{B}_e & \mathbf{C}_e \mathbf{A}_e^{N-2} \mathbf{B}_e & \dots & \mathbf{D}_e \end{bmatrix}}_{\mathcal{T}_r} \underbrace{\begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_N \end{bmatrix}}_{\mathbf{a}}$$

The Dynamical Systems Case (2)

Dynamics of plant and controller:

$$\eta_{k+1} = \mathbf{A}\eta_k + \mathbf{B}a_k + \mathbf{G}w_k$$

$$x_k = \mathbf{C}\eta_k + \mathbf{D}a_k + \mathbf{H}v_k$$

Lift to time interval $[0, N]$

with zero-initial conditions and no noise:

$$\underbrace{\begin{bmatrix} x_0 \\ x_1 \\ x_2 \\ \vdots \\ x_N \end{bmatrix}}_{\mathbf{x}} = \underbrace{\begin{bmatrix} \mathbf{D} & 0 & \dots & 0 \\ \mathbf{CB} & \mathbf{D} & \dots & 0 \\ \mathbf{CAB} & \mathbf{CB} & \dots & 0 \\ \vdots & \vdots & \ddots & 0 \\ \mathbf{CA}^{N-1}\mathbf{B} & \mathbf{CA}^{N-2}\mathbf{B} & \dots & \mathbf{D} \end{bmatrix}}_{\mathcal{T}_x} \underbrace{\begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_N \end{bmatrix}}_{\mathbf{a}}$$

Max Impact/Bounded Resource Attack

$$\begin{aligned} & \max_{\mathbf{a}} \|\mathcal{T}_x \mathbf{a}\|_{\infty} && \text{(physical impact)} \\ \text{s.t.} & && \\ & \|\mathbf{r}\|_{\infty} = \|\mathcal{T}_r \mathbf{a}\|_{\infty} \leq \delta_{\alpha} && \text{(residual in detector)} \\ & \|h_p(\mathbf{a})\|_0 \leq \epsilon && \text{(# channels attacked)} \end{aligned}$$

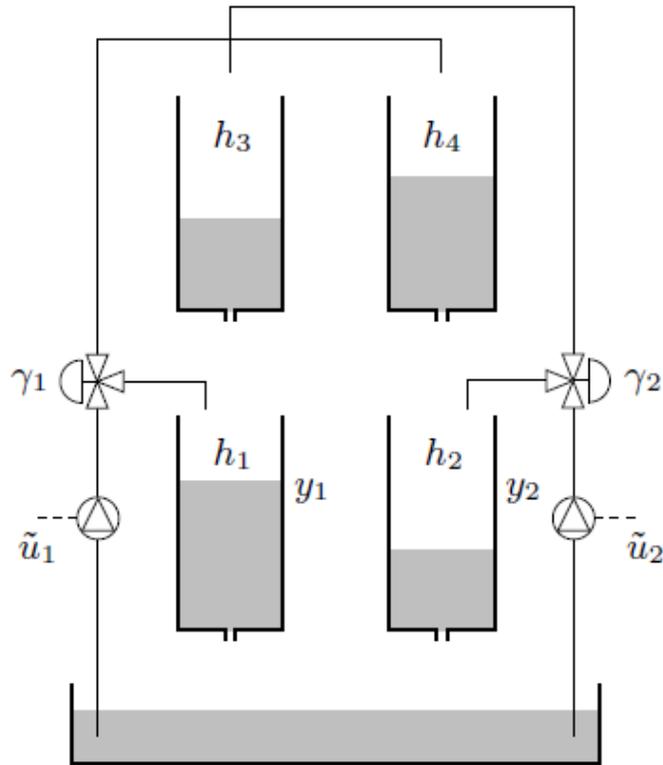
- Maximize impact (push $\|\mathbf{x}\|_{\infty}$ far away from equilibrium)
- No alarms (threshold δ_{α})
- Attack no more than ϵ channels

$$h_p(\mathbf{a}) = [\|\mathbf{a}_{(1)}\|_{\ell_p}, \dots, \|\mathbf{a}_{(i)}\|_{\ell_p}, \dots, \|\mathbf{a}_{(q_a)}\|_{\ell_p}]$$

- Mixed Integer Linear Program (MILP)

[Teixeira *et al.*, 2013]

Numerical Example



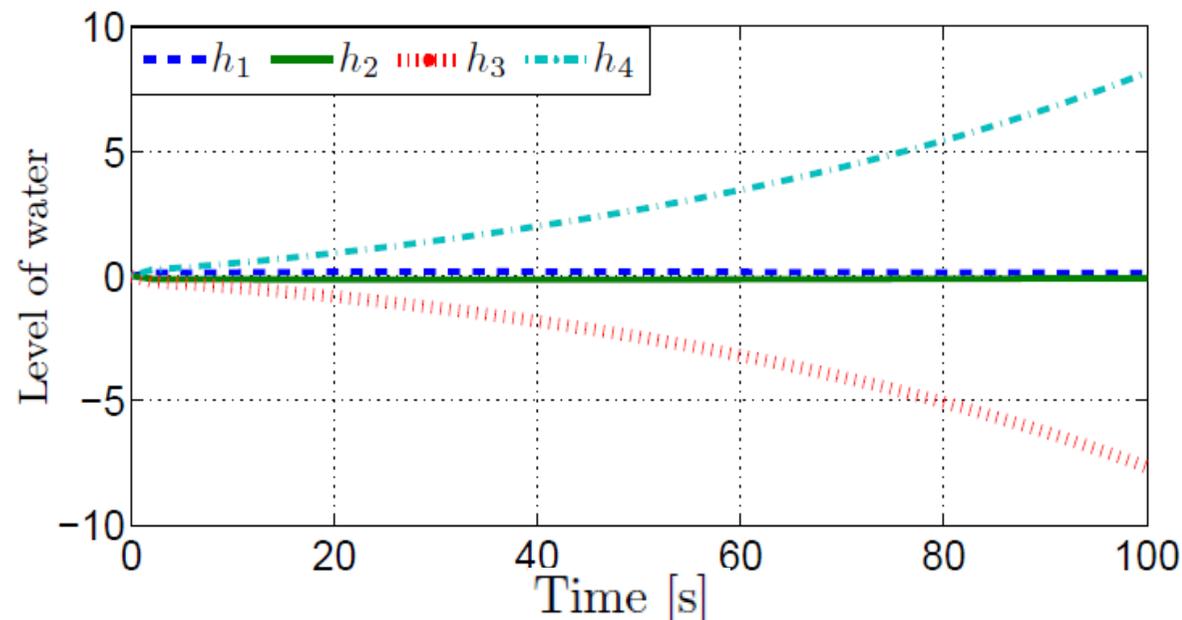
$$\begin{aligned} \dot{h}_1 &= -\frac{a_1}{A_1} \sqrt{2gh_1} + \frac{a_3}{A_1} \sqrt{2gh_3} + \frac{\gamma_1 k_1}{A_1} u_1, \\ \dot{h}_2 &= -\frac{a_2}{A_2} \sqrt{2gh_2} + \frac{a_4}{A_2} \sqrt{2gh_4} + \frac{\gamma_2 k_2}{A_2} u_2, \\ \dot{h}_3 &= -\frac{a_3}{A_3} \sqrt{2gh_3} + \frac{(1 - \gamma_2) k_2}{A_3} u_2, \\ \dot{h}_4 &= -\frac{a_4}{A_4} \sqrt{2gh_4} + \frac{(1 - \gamma_1) k_1}{A_4} u_1, \end{aligned}$$

- Wireless LQG controller
- 4 channels: 2 actuators and 2 measurements
- Minimum phase or non-minimum phase depending on γ_1, γ_2

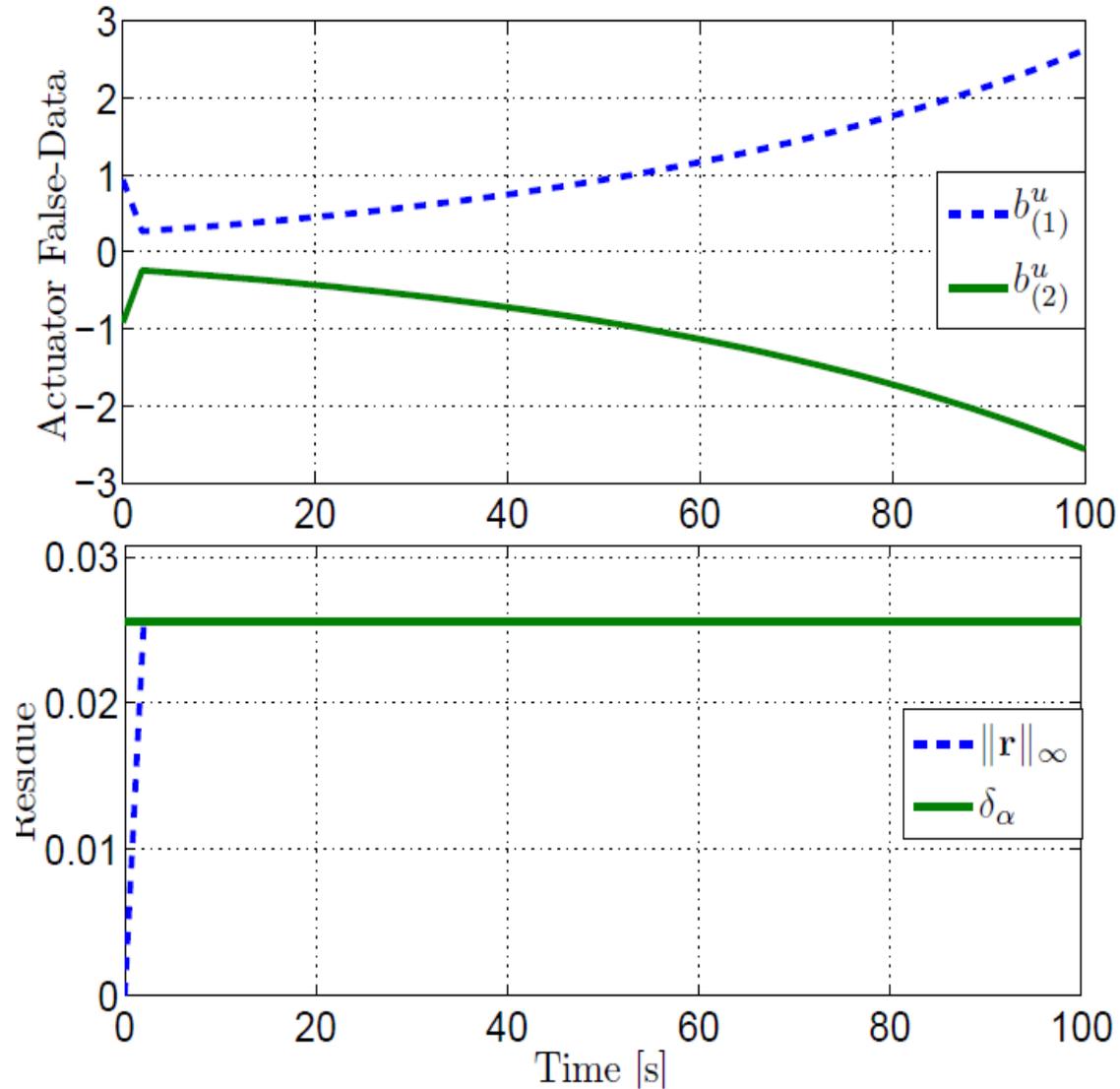
Numerical Example (Non-Min Phase)

Values of $\|\mathbf{x}\|_2$ for max impact/bounded resource attack
 $\delta_\alpha = 0.15$

	$\ h_p(\mathbf{a})\ _0$			
	1	2	3	4
Minimum phase	1.15	140.39	∞	∞
Non-minimum phase	2.80	689.43	∞	∞



Numerical Example (Non-Min Phase)



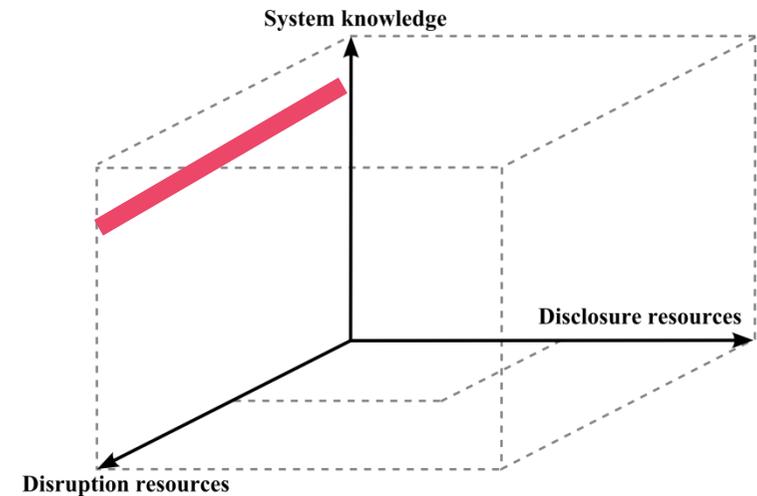
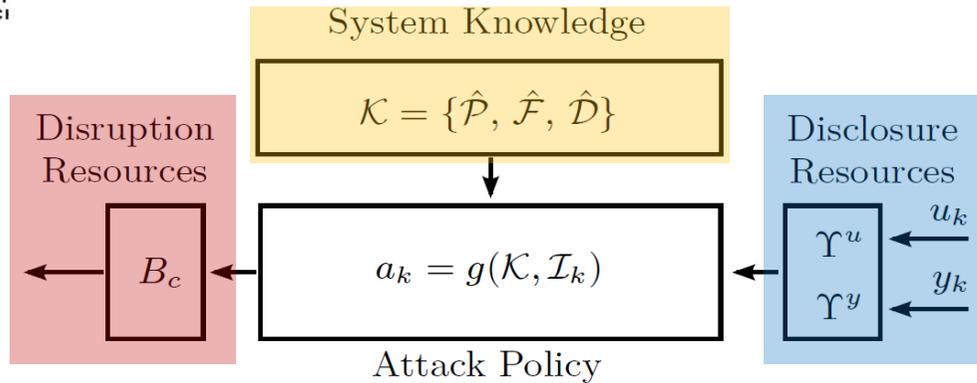
Numerical Example

- Maximum Impact/Bounded Resource attack illustrated
- 2 channels allowed: MILP selects the actuators
- 3-4 channels allowed: Unbounded impact (any attack on actuators can be hidden by corrupting 2 measurements)
- Infinity norm criteria yields more aggressive attack than 2-norm criteria (bounds get saturated)
- Not surprisingly, non-min phase plant more sensitive

Steady-State Attacks

- Consider attacks over $[0, N]$ where
 - $N \rightarrow \infty$
 - $a_k = ge^{i\omega k}$, $\omega \in \mathbb{R}$, $g \in \mathbb{C}^{q_a}$ (sinusoidal attacks)
- Similar analysis carries through but make substitutions
 - $\mathcal{T}_r \rightarrow G_r(e^{i\omega})$
 - $\mathcal{T}_x \rightarrow G_x(e^{i\omega})$
- Yields worst-case attack frequency ω etc.

Summary



- Tools for quantitative trade-off analysis between attacker's impact and resources, also important for cyber defense prioritization
- For dynamical systems there are *temporal* as well as *spatial (channel) constraints* for attacker to fulfill
 - Enforced through lifting and frequency-response models
 - Solved using MILP. No well-working relaxation known by us



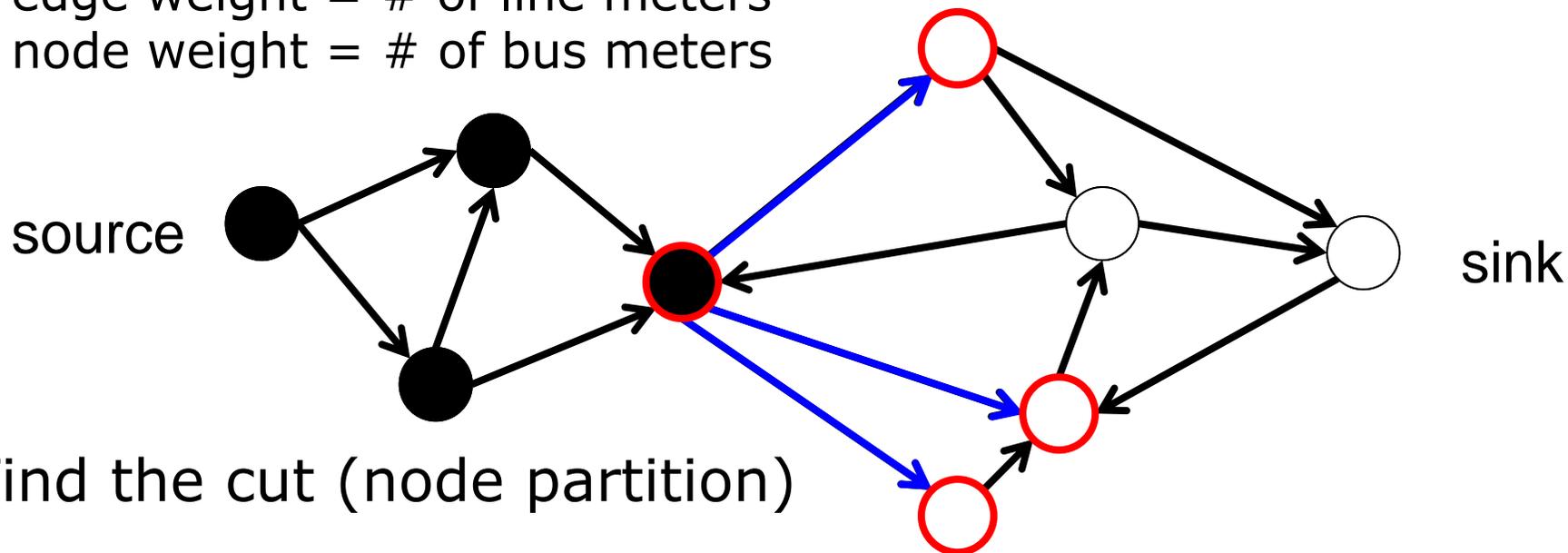
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 - Vukovic *et al.*, "Network-aware Mitigation of Data Integrity Attacks on Power System State Estimation", IEEE JSAC, 2012
- Efficient computation, and Min Cut relaxation
 - Sou *et al.*, "On the Exact Solution to a Smart Grid Cyber-Security Analysis Problem", IEEE Trans. on Smart Grid, 2013
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Generalized Min Cut with Costly Nodes

Focus on directed graph (undirected = bi-directed)

edge weight = # of line meters
node weight = # of bus meters

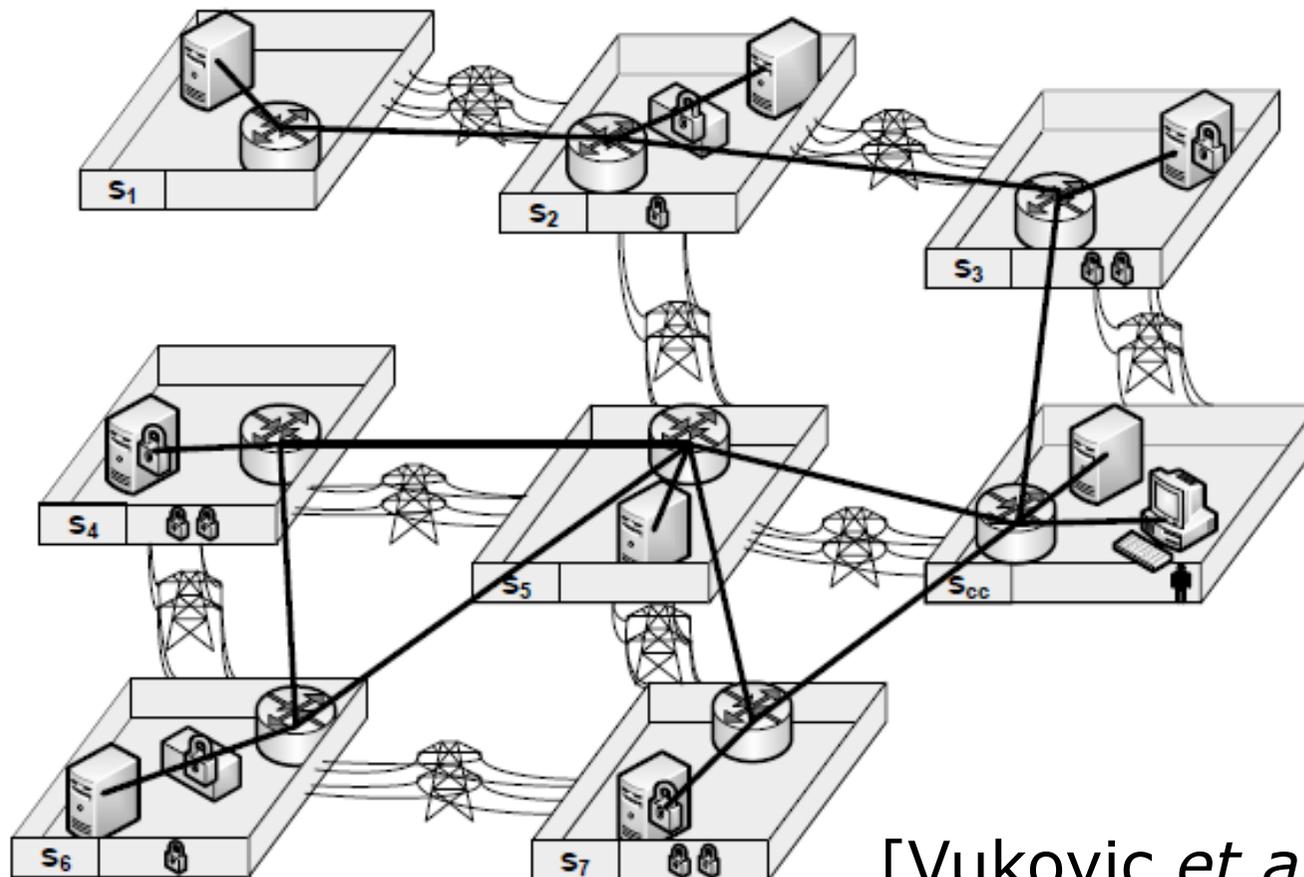
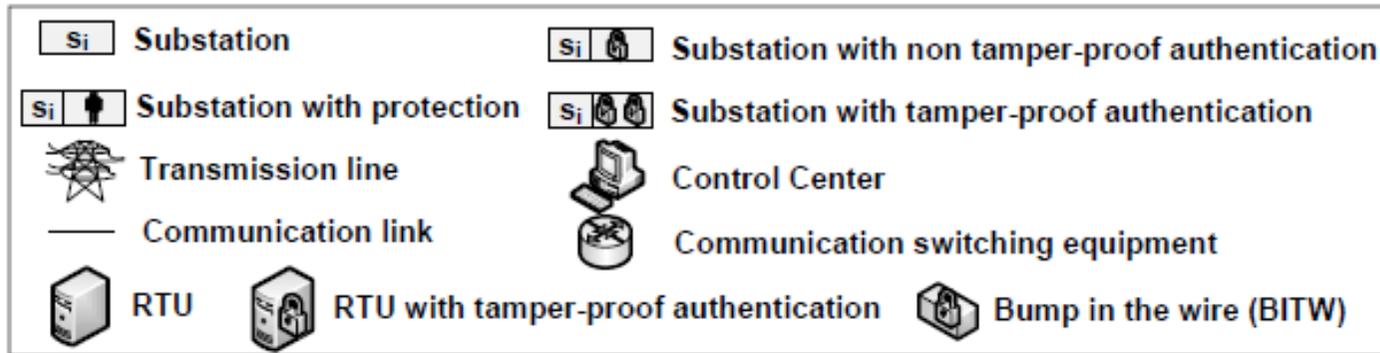


Find the cut (node partition)

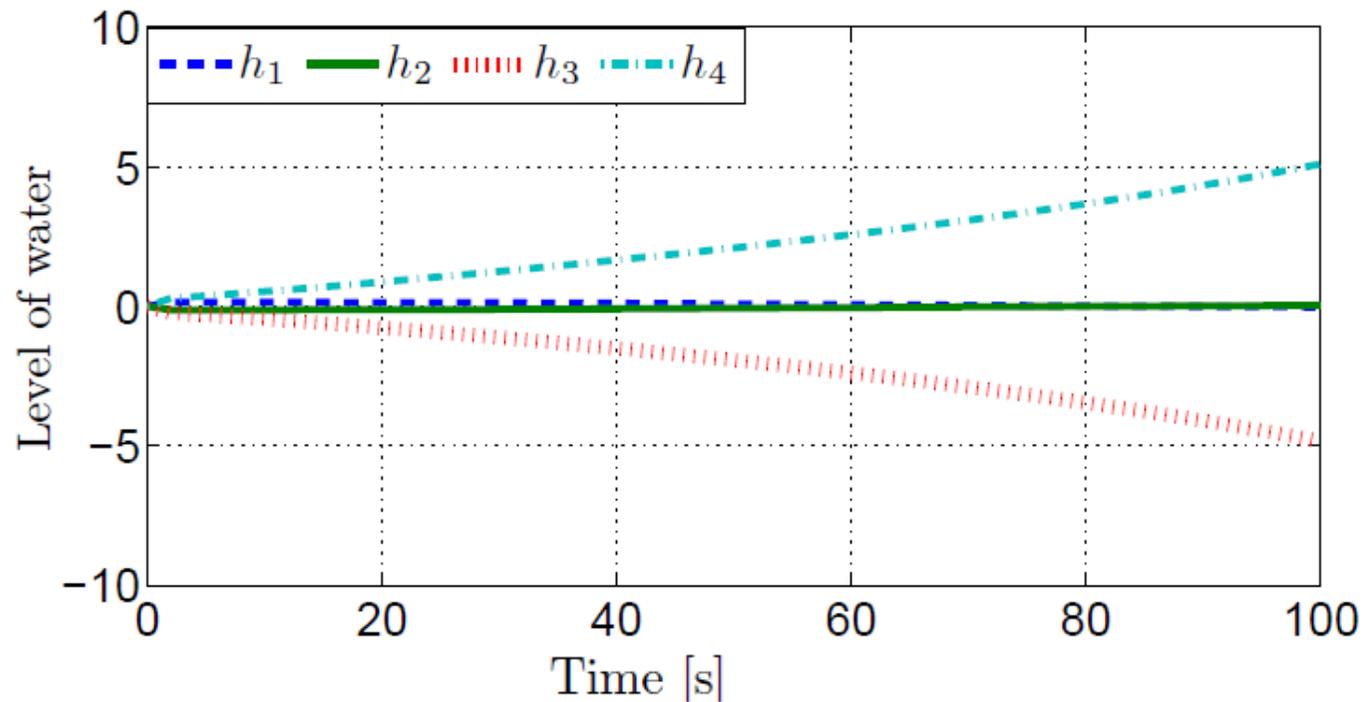
to minimize weights of cut **edge** + incident **node** weights...

Generalization of standard Min Cut!

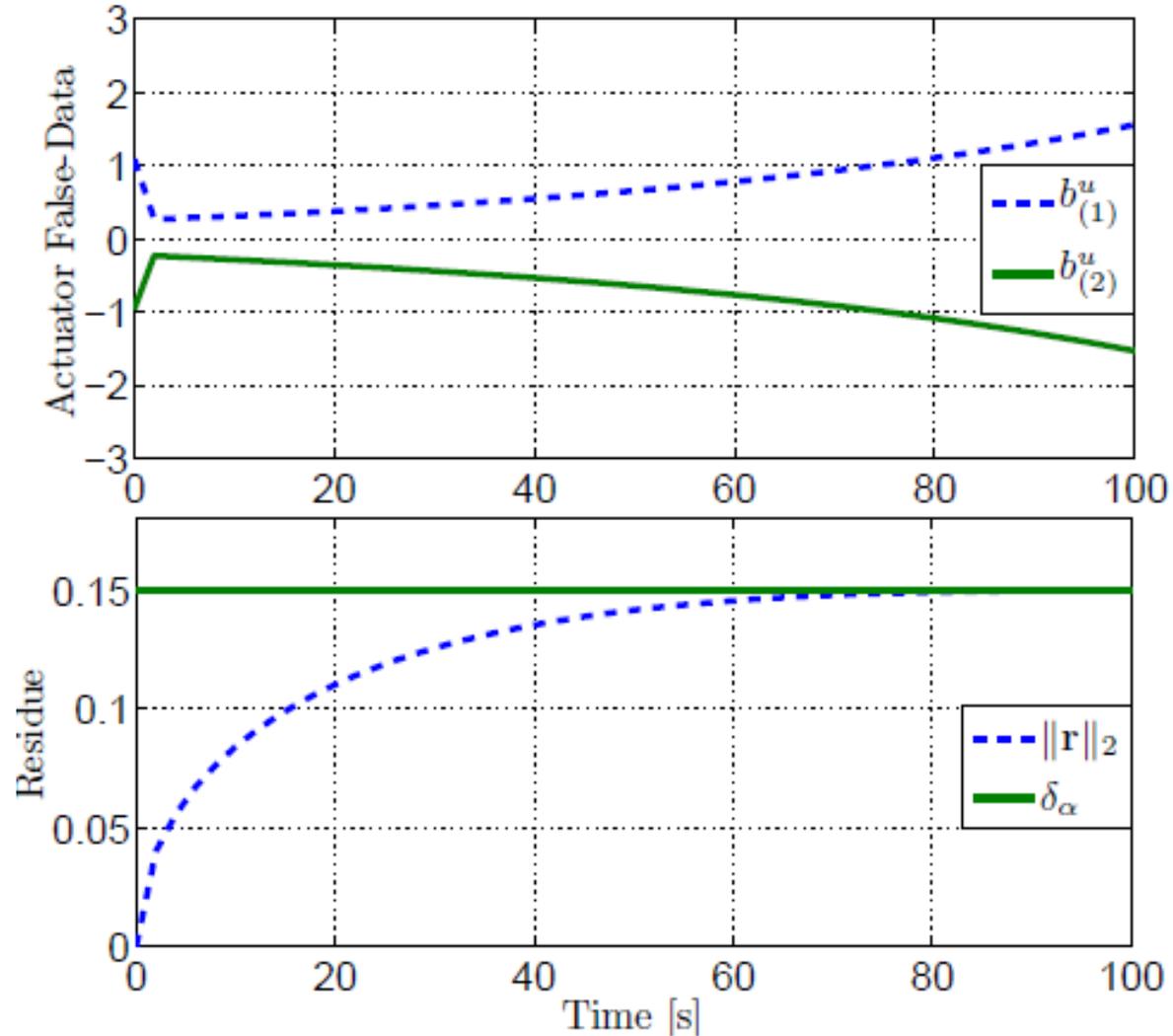
The Network: SCADA System



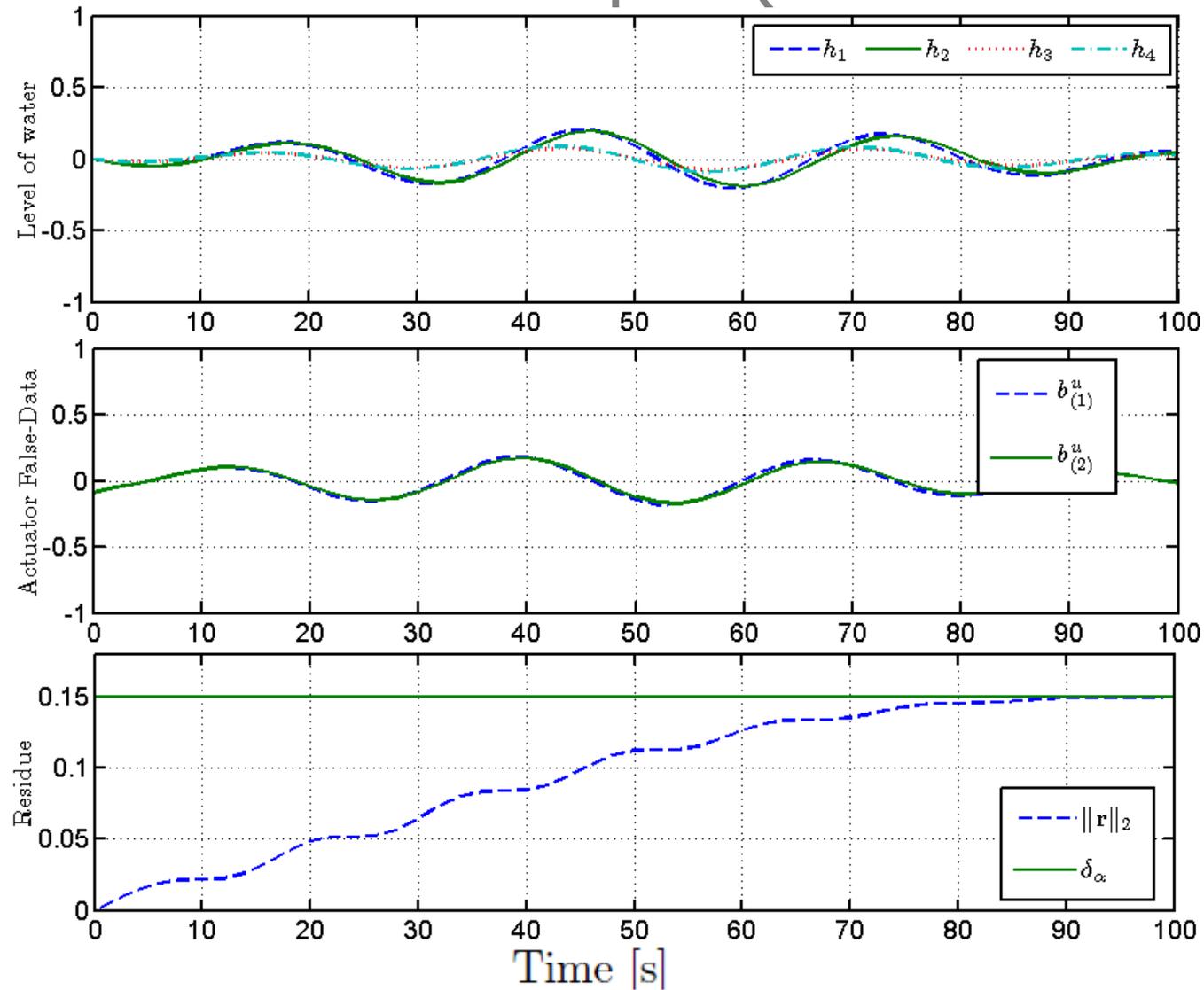
Numerical Example (2-Norm, Non-Min Phase)



Numerical Example (2-Norm, Non-Min Phase)



Numerical Example (Min Phase 2-norm)



Numerical Example (Min Phase inf-norm)

