

# Using Structured System Theory to Identify Malicious Behavior in Distributed Systems

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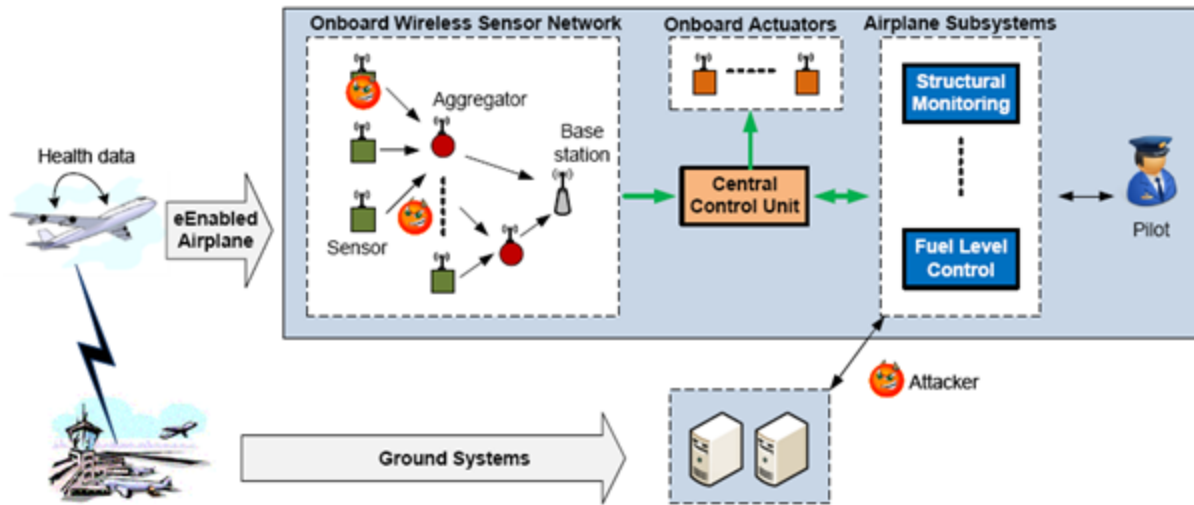
Electrical and Computer Engineering  
University of Cyprus

# Outline

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- ▶ Introduction
- ▶ Background on Linear and Structured System Theory
- ▶ Resilient Information Dissemination in Multi-Agent Networks
- ▶ Designing an Intrusion Detection System for a Wireless Control Network

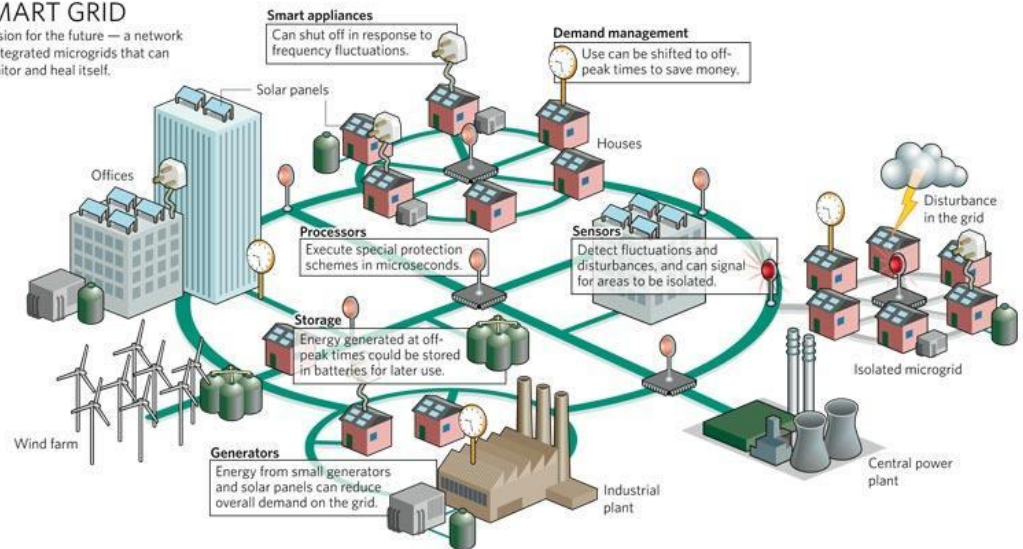
# Distributed Systems and Networks in Safety-Critical Applications



**Airplane Health Monitoring and Management System**

**Electrical Power Grid**

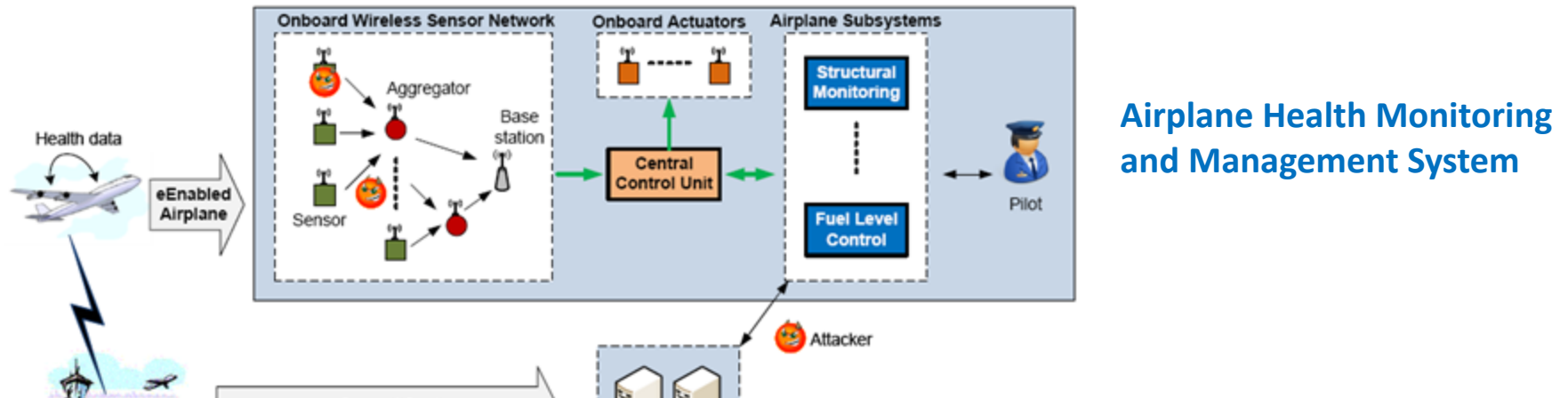
**SMART GRID**  
 A vision for the future — a network of integrated microgrids that can monitor and heal itself.



**UAV Teams**

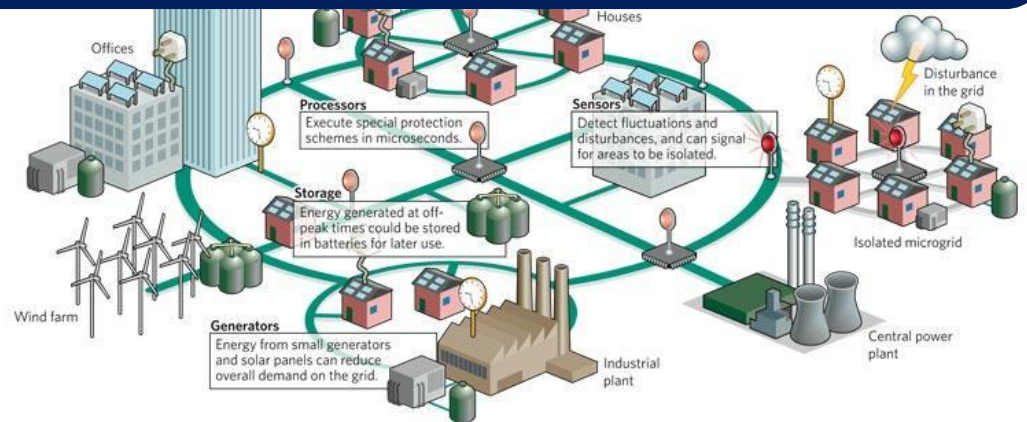


# Distributed Systems and Networks in Safety-Critical Applications



**Core requirement:** disseminate information through network

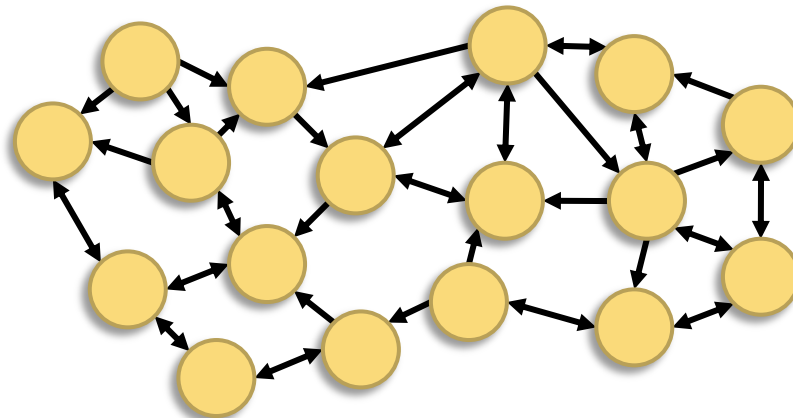
- quickly, efficiently, reliably, securely, ...



# Problem Formulation

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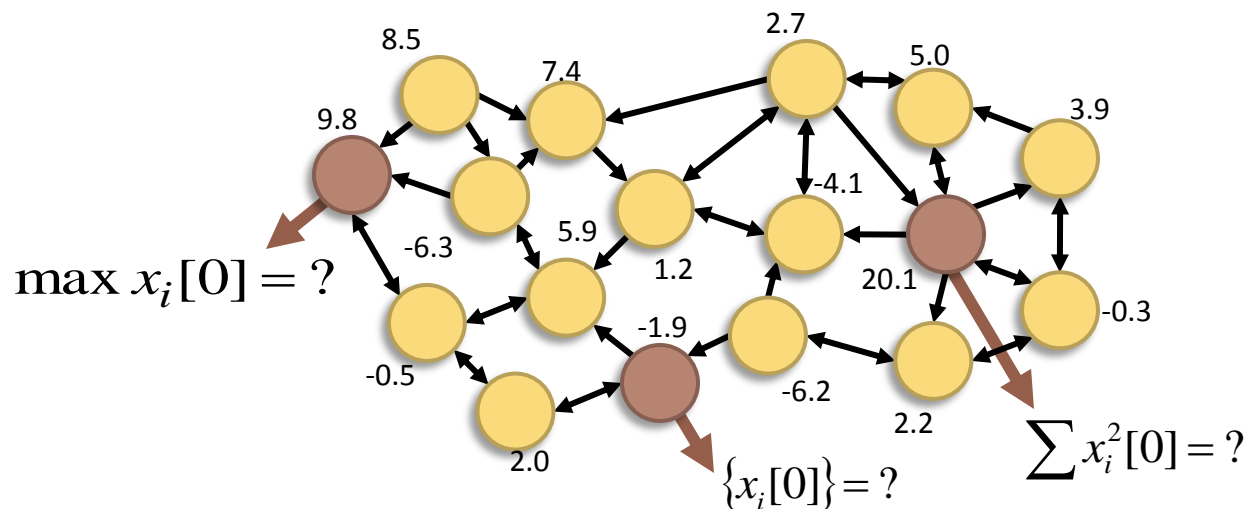
- ▶ Consider a network with nodes  $\{x_1, x_2, \dots, x_N\}$ 
  - ▶ e.g., sensors, computers, robots, “agents”, ...



- ▶ Each node can only interact with certain other nodes
  - ▶ Assume fixed interaction topology in this talk

# Function Calculation

- ▶ Each node  $x_i$  has some initial value  $x_i[0]$ 
  - ▶ e.g., temperature measurement, position, vote, ...
- ▶ Some nodes must **calculate certain functions** of these values
- ▶ Special cases
  - ▶ **Data accumulation**: some nodes gather all values
  - ▶ **Distributed consensus**: all nodes calculate same function



# Existing Work on Distributed Function Calculation

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- ▶ Distributed function calculation schemes well studied over past few decades
  - ▶ Flooding, tree-based schemes, gossip, token passing, ...
  - ▶ Communication complexity, computational complexity, time complexity, fault tolerance, ...
- ▶ Many excellent books on this topic
  - ▶ **Dissemination of Information in Communication Networks**, Hromkovic et. al., 2005
  - ▶ **Communication Complexity**, Kushilevitz and Nisan, 1997
  - ▶ **Distributed Algorithms**, Lynch, 1997
  - ▶ **Elements of Distributed Computing**, Garg, 2002
  - ▶ **Parallel and Distributed Computation**, Bertsekas and Tsitsiklis, 1997
  - ▶ ...

# Linear Iterative Strategies

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- ▶ Investigate **linear iterative strategies** for distributed function calculation
- ▶ At each time-step  $k$ , every node updates its value as

$$x_i[k+1] = w_{ii}x_i[k] + \sum_{j \in nbr(i)} w_{ij}x_j[k]$$

- ▶ Extensively studied for **asymptotic consensus**
  - ▶ For some vector  $\mathbf{a}$ ,

$$\lim_{k \rightarrow \infty} x_i[k] = \mathbf{a}^T \mathbf{x}[0], \quad \forall i \in \{1, 2, \dots, N\}$$

- ▶ Also studied by communications community: **network coding**
- ▶ Survey papers:
  - ▶ [Olfati-Saber, Fax & Murray, *Proc. IEEE*, 2007], [Ren, Beard & Atkins, *Proc. ACC*, 2005], [Yeung, Li, Cai, Zhang, 2006]



# Potential for Incorrect Behavior

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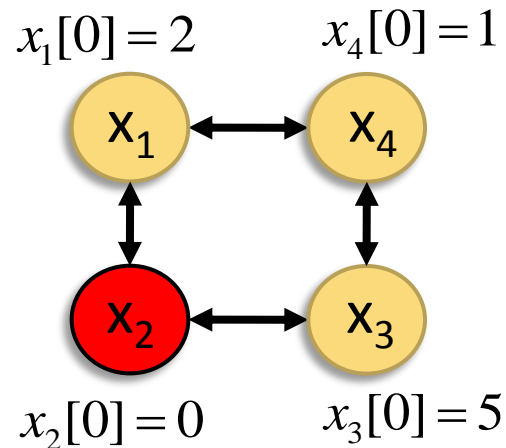
- ▶ Here, we ask:

“What happens if some nodes don’t follow the linear iterative strategy?”

- ▶ **Faulty nodes:** update their values incorrectly due to hardware faults, or stop working altogether
- ▶ **Malicious nodes:** willfully update their values incorrectly to prevent other nodes from calculating functions
- ▶ Related works:
  - ▶ [Jadbabaie, Lin & Morse '03], [Gupta, Langbort & Murray '06], [Pasqualetti, Bicchi & Bullo '07], [Sundaram & Hadjicostis '08], [Teixeira, Sandberg & Johansson '10], [Chapman & Mesbahi '10]

# Building Intuition

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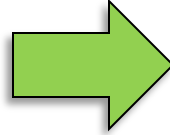
- ▶ Node  $x_1$  wants to obtain all initial values via some algorithm
- ▶ Node  $x_2$  is malicious and pretends  $x_3[0] = 9$
- ▶ Node  $x_4$  behaves correctly and uses  $x_3[0] = 5$
- ▶ Node  $x_1$  doesn't know who to believe
  - ▶ i.e., is node  $x_3$ 's value equal to 9 or 5?
- ▶ Node  $x_1$  needs another node to act as tie-breaker

# Main Results

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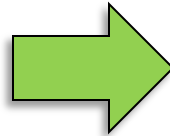
- ▶ If network has up to **b malicious nodes**, we show:

Node  $x_i$  has **2b or fewer** node-disjoint paths from some node  $x_j$



Malicious nodes can update their values in such a way that  $x_i$  **cannot calculate any function** of  $x_j$ 's initial value

Node  $x_i$  has **2b+1 or more** node-disjoint paths from every other node



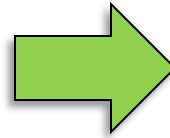
$x_i$  can obtain **all initial values** after running linear strategy for **at most N time-steps** with **almost any weights**

# Main Results

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- ▶ If network has up to **b malicious nodes**, we show:

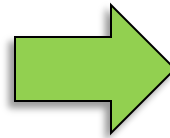
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“Easy”

Malicious nodes can update their values in such a way that  $x_i$  **cannot calculate any function** of  $x_j$ 's initial value

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“Tricky”

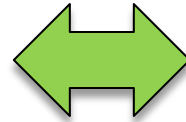
$x_i$  can obtain **all initial values** after running linear strategy for **at most N time-steps** with **almost any weights**

# Using Structured System Theory to Analyze Linear Iterations

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- ▶ To prove resilience, we use the following approach:

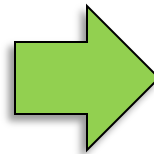
Linear strategy is resilient to  $b$  malicious nodes in a given network



A linear system has some property “**P**”



Original network has connectivity  $2b+1$



A graph has some property “**Q**”



(Structured System Theory)

# Background on Linear and Structured System Theory

# Properties of Linear Systems

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$$x[k+1] = Ax[k] + Bu[k]$$

$$y[k] = Cx[k]$$

State:  $x \in \mathfrak{R}^n$ ,

Output:  $y \in \mathfrak{R}^p$ ,

Input:  $u \in \mathfrak{R}^m$

- ▶ **Controllability:** drive state to desired value using input
- ▶ **Observability:** determine state from output, with **input known** (or zero)
- ▶ **Strong Observability:** determine state from output, with **input unknown**
- ▶ **Invertibility:** determine input from output, with **state known**

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# Properties of Linear Systems

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- ▶ **Controllability:** drive state to desired value using input

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## Standard Approach:

Use algebraic tests to determine if properties hold

- ▶ **Strong Observability:** determine state from output, with **input unknown**
- ▶ **Invertibility:** determine input from output, with **state known**

# Linear Structured Systems

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$$\begin{aligned}x[k+1] &= Ax[k] + Bu[k] \\ y[k] &= Cx[k]\end{aligned}\quad x \in \mathfrak{R}^n, y \in \mathfrak{R}^p, u \in \mathfrak{R}^m$$

- ▶ System is **structured** if every entry of the matrices (A,B,C) is either zero, or an independent free parameter
- ▶ Used to represent and analyze dynamical systems with unknown/uncertain parameters [Lin '74, Dion et al., '03]
- ▶ **Structured system theory:** determines properties of systems based on the zero/nonzero structure of matrices

# Structural Properties

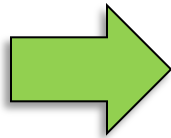
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**“Structured system has property P”:**

Property P holds for **at least one choice** of free parameters in the matrices (A, B, C)

- ▶ Structural properties are **generic!**

Structured system has property P



Structured system will have property P for **almost any choice** of free parameters

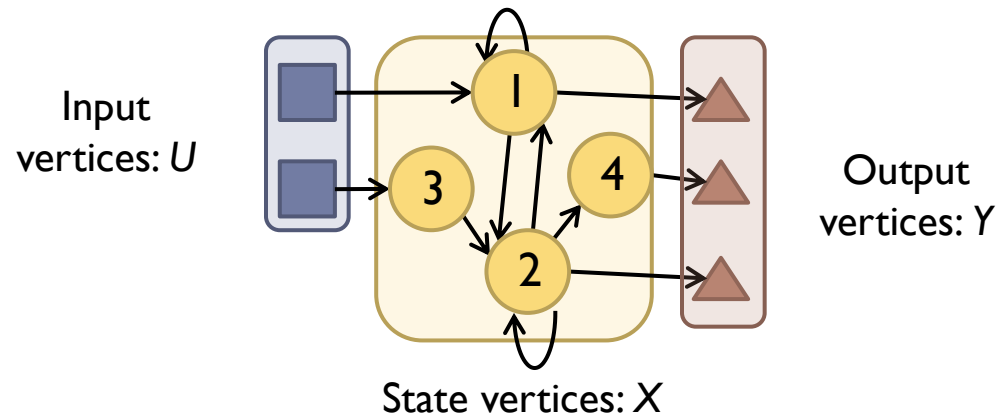
- ▶ Use graph based techniques to determine if structural properties hold

# Example of Structured System and Associated Graph

- ▶ Structured system can be represented as a graph
- ▶ Structured system:

$$x[k+1] = \begin{bmatrix} \lambda_1 & \lambda_2 & 0 & 0 \\ \lambda_3 & \lambda_4 & \lambda_5 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & \lambda_6 & 0 & 0 \end{bmatrix} x[k] + \begin{bmatrix} \lambda_7 & 0 \\ 0 & 0 \\ 0 & \lambda_8 \\ 0 & 0 \end{bmatrix} u[k], \quad y[k] = \begin{bmatrix} \lambda_9 & 0 & 0 & 0 \\ 0 & \lambda_{10} & 0 & 0 \\ 0 & 0 & 0 & \lambda_{11} \end{bmatrix} x[k]$$

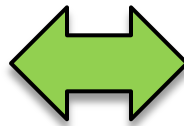
- ▶ Associated graph  $H$ :



# Example: Test for Structural Invertibility

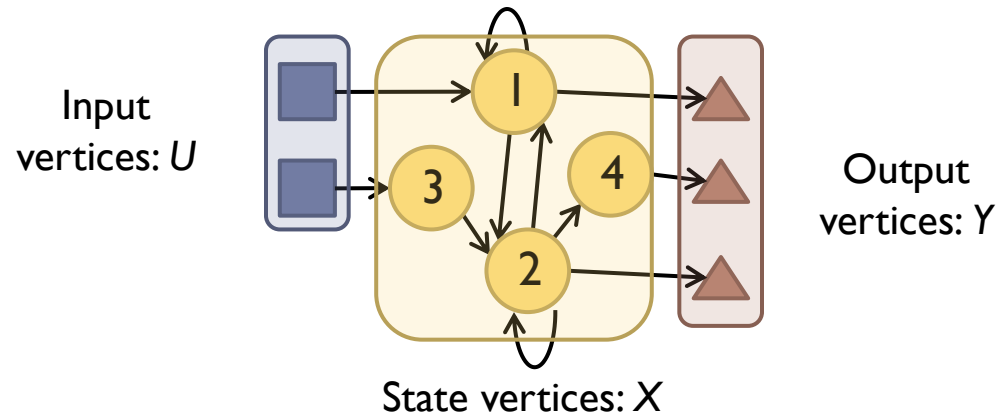
**Theorem** [*van der Woude, '91*]:

System is structurally invertible



Graph  $H$  has  **$m$  vertex-disjoint paths** from inputs to outputs

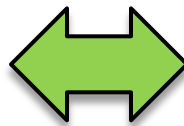
▶ e.g.,



# Example: Test for Structural Invertibility

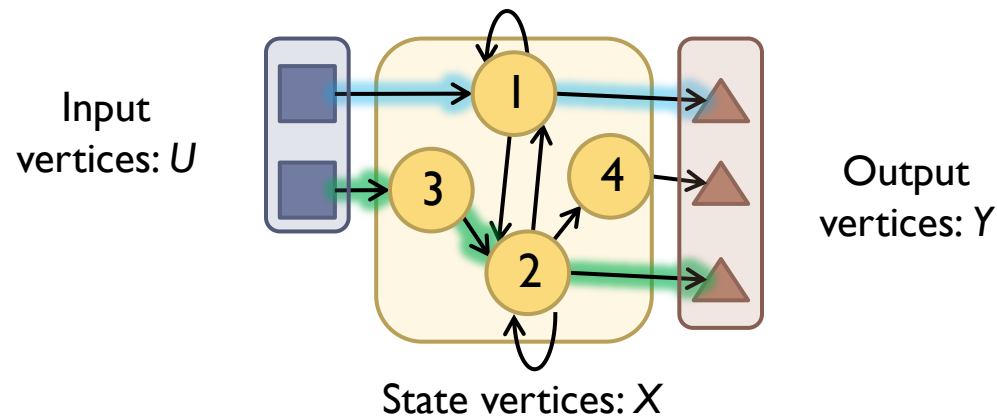
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▶ e.g.,



**Structurally Invertible**

# References on Structured Systems

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- ▶ C. T. Lin, “*Structural Controllability*”, IEEE TAC, 1974
- ▶ K. J. Reinschke, *Multivariable Control: A Graph-Theoretic Approach*, 1988
- ▶ J-M. Dion, C. Commault and J. van der Woude, “*Generic Properties and Control of Linear Structured Systems: A Survey*”, Automatica, 2003
- ▶ D. D. Siljak, *Decentralized Control of Complex Systems*, 1991
- ▶ Sundaram & Hadjicostis, CDC 2009, ACC 2010 (*Structural properties over finite fields, upper bound on generic controllability/observability indices*)

# Application to Resilient Information Dissemination



# Modeling Faulty/Malicious Behavior in Linear Iterative Strategies

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- ▶ Linear iterative strategy for information dissemination:

- ▶ **Correct** update equation for node  $x_i$ :

$$x_i[k+1] = w_{ii}x_i[k] + \sum_{j \in nbr(i)} w_{ij}x_j[k]$$

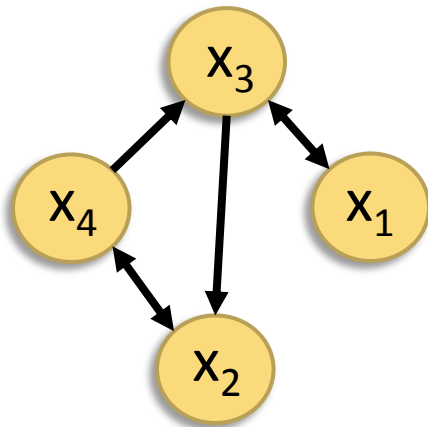
- ▶ **Faulty or malicious** update by node  $x_i$ :

$$x_i[k+1] = w_{ii}x_i[k] + \sum_{j \in nbr(i)} w_{ij}x_j[k] + f_i[k]$$

- ▶  $f_i[k]$  is an additive error at time-step  $k$
- ▶ **Note:** this model allows node  $x_i$  to update its value in a **completely arbitrary** manner

# Modeling the Values Seen by Each Node

- ▶ Each node obtains neighbors' values at each time-step
- ▶ Let  $\mathbf{y}_i[k] = \mathbf{C}_i \mathbf{x}[k]$  denote values seen by  $x_i$  at time-step  $k$ 
  - ▶ Rows of  $\mathbf{C}_i$  index portions of  $\mathbf{x}[k]$  available to  $x_i$



For node  $x_3$ :

$$\mathbf{y}_3[k] = \begin{bmatrix} x_1[k] \\ x_3[k] \\ x_4[k] \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}}_{\mathbf{C}_3} \mathbf{x}[k]$$

# Linear Iteration with Faulty/Malicious Nodes

- ▶ Let  $S = \{x_{i_1}, x_{i_2}, \dots, x_{i_b}\}$  be set of faulty/malicious nodes
  - ▶ **Unknown** a priori, but bounded by  $b$
- ▶ Update equation for entire **system**:

$$\underbrace{\begin{bmatrix} x_1[k+1] \\ \vdots \\ x_N[k+1] \end{bmatrix}}_{\mathbf{x}[k+1]} = \underbrace{\begin{bmatrix} w_{11} & \cdots & w_{1N} \\ \vdots & \ddots & \vdots \\ w_{N1} & \cdots & w_{NN} \end{bmatrix}}_{\mathbf{W}} \underbrace{\begin{bmatrix} x_1[k] \\ \vdots \\ x_N[k] \end{bmatrix}}_{\mathbf{x}[k]} + \underbrace{\begin{bmatrix} \mathbf{e}_{i_1} & \mathbf{e}_{i_2} & \cdots & \mathbf{e}_{i_b} \end{bmatrix}}_{\mathbf{B}_S} \underbrace{\begin{bmatrix} f_{i_1}[k] \\ f_{i_2}[k] \\ \vdots \\ f_{i_b}[k] \end{bmatrix}}_{\mathbf{f}_S[k]}$$

$$\mathbf{y}_i[k] = \mathbf{C}_i \mathbf{x}[k]$$

**Constraint:** weight  $w_{ij} = 0$  if node  $x_j$  is not a neighbor of node  $x_i$

# Recovering the Initial State

---

- ▶ System model for linear iteration with malicious nodes

$$\mathbf{x}[k + 1] = \mathbf{W}\mathbf{x}[k] + \mathbf{B}_S \mathbf{f}_S[k]$$

$$\mathbf{y}_i[k] = \mathbf{C}_i \mathbf{x}[k]$$

- ▶ **Objective:** Recover initial state  $\mathbf{x}[0]$  from outputs of the system, without knowing  $\mathbf{f}_S[k]$
- ▶ *Almost* equivalent to **strong observability** of system
  - ▶ The set  $S$  is also unknown here
  - ▶ Only know it has at most  $b$  elements

# Recovering the Initial State

---

- ▶ Want to ensure that the output trajectory uniquely specifies the initial state
  - ▶ Same output trajectory must not be generated by two different initial states and two (possibly) different sets of  $b$  malicious nodes
- ▶ By linearity, can show:

Can recover initial state in system

$$\mathbf{x}[k+1] = \mathbf{W}\mathbf{x}[k] + \mathbf{B}_S \mathbf{f}_S[k]$$

$$\mathbf{y}_i[k] = \mathbf{C}_i \mathbf{x}[k]$$

for **any unknown set**  $S$  of  $b$  nodes



Linear system

$$\mathbf{x}[k+1] = \mathbf{W}\mathbf{x}[k] + \mathbf{B}_Q \mathbf{f}_Q[k]$$

$$\mathbf{y}_i[k] = \mathbf{C}_i \mathbf{x}[k]$$

is strongly observable for  
**any known set**  $Q$  of  $2b$  nodes

# Structural Strong Observability

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- ▶ For any set  $Q$  of  $2b$  nodes, strong observability of

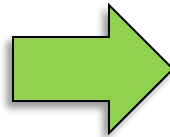
$$\mathbf{x}[k+1] = \mathbf{W}\mathbf{x}[k] + \mathbf{B}_Q \mathbf{f}_Q[k]$$

$$\mathbf{y}_i[k] = \mathbf{C}_i \mathbf{x}[k]$$

is a **structural property**

- ▶ Graph of system is given by graph of original network, with additional inputs and outputs
- ▶ Using tests for structural strong observability, we show:

$x_i$  has  **$2b+1$  node-disjoint paths** from every other node



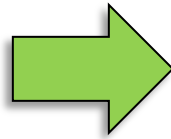
Above system will be strongly observable for any set  $Q$  of  $2b$  nodes

# Robustness of the Linear Iterative Scheme

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- ▶ By generic nature of structural properties:

Network is  
 **$2b+1$  connected**



For **almost any  $W$** , any node can recover all initial values despite actions of  $b$  malicious nodes

- ▶ How long will each node need to wait before the values that it receives uniquely specifies the initial state?

If a linear system is strongly observable, outputs of system over  $N$  time-steps are sufficient to determine initial state

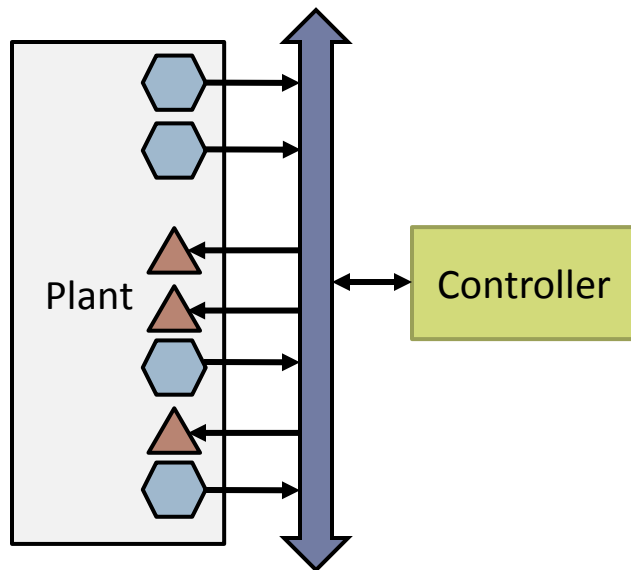
- ▶ Any node can obtain  $\mathbf{x}[0]$  after **at most  $N$  time-steps**

Application to Feedback Control:  
Monitoring for Malicious Behavior in a  
Wireless Control Network

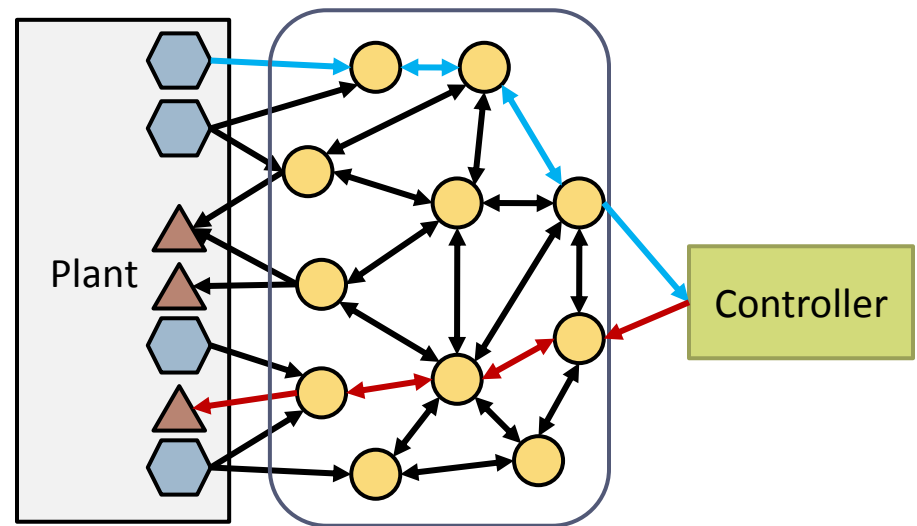


# Control Over Networks

- ▶ Sensors (⬡) and Actuators (▲) are installed on a *plant*
- ▶ Communicate with controllers (■) over a network
- ▶ Standard architectures:



Wired Control System



Wireless Control System

**Network primarily used as a *channel***

# Much Work on Networked Control

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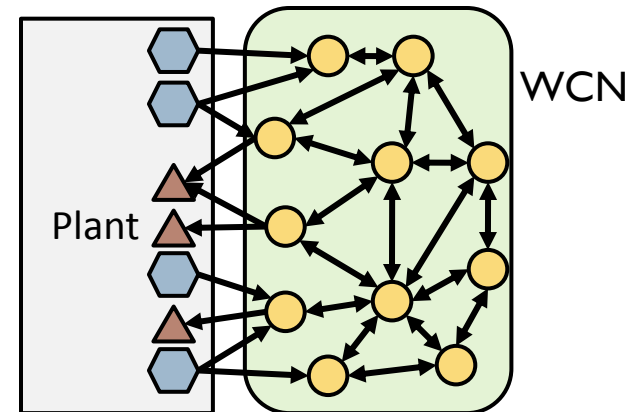
- ▶ Studied by many researchers
  - ▶ Branicky et al., Elia et al., Gupta et al., Hespanha et al., Martins et al., Murray et al., Nair et al., Seiler et al., Sinopoli et al., Walsh et al., ...
- ▶ Robustness to “network” issues:
  - ▶ Packet dropouts, Bernoulli channel failures
  - ▶ Delays in the feedback loop
- ▶ Survey paper: *Hespanha et al., Proc IEEE, 2007*

# The Wireless Control Network (WCN)

- ▶ Idea: Can we use the wireless network *itself* as a controller?
  - ▶ Leverage the computational capability *in the network*

- ▶ Desired properties:

- ✓ Lightweight computation - design for resource constrained nodes
- ✓ Static transmission schedules
- ✓ Compositionality
- ✓ Handles multiple actuation and sensing points



- ▶ Utility: Network can be used as a backup control mechanism in case primary controller fails
  - ▶ Uses existing infrastructure, *does not require additional hardware*

# Wireless Control Network

- ▶ Plant:  $\mathbf{x}[k + 1] = \mathbf{A}\mathbf{x}[k] + \mathbf{B}\mathbf{u}[k]$

$$\mathbf{y}[k] = \mathbf{C}\mathbf{x}[k]$$

- ▶ State update procedure, node  $z_i$ :

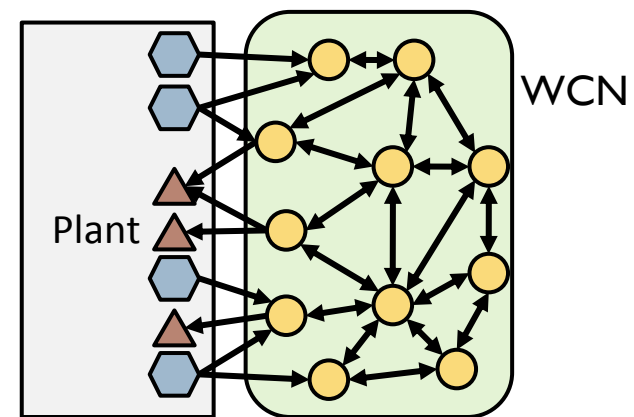
$$z_i[k + 1] = w_{ii}z_i[k] + \underbrace{\sum_{z_j \in \mathcal{N}_{z_i}} w_{ij}z_j[k]}_{\text{From neighbors}} + \underbrace{\sum_{s_j \in \mathcal{N}_{z_i}} h_{ij}y_j[k]}_{\text{From sensors}}$$

- ▶ Plant update procedure, input  $i$ :

$$u_i[k] = \sum_{z_j \in \mathcal{N}_{a_i}} g_{ij}z_j[k] \quad \text{From actuator's neighbors}$$

- ▶ Network state vector:

$$\mathbf{z}[k] = [z_1[k] \quad z_2[k] \quad \cdots \quad z_N[k]]^T$$



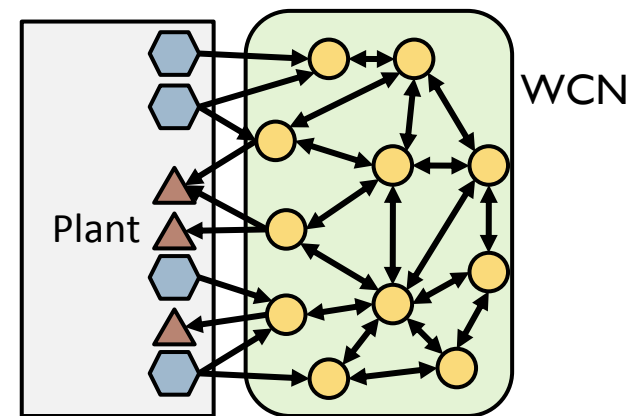
# Wireless Control Network: A Linear System

- ▶ Network update procedure:

$$\mathbf{z}[k+1] = \underbrace{\begin{bmatrix} w_{11} & w_{12} & \cdots & w_{1N} \\ w_{21} & w_{22} & \cdots & w_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ w_{N1} & w_{N2} & \cdots & w_{NN} \end{bmatrix}}_{\mathbf{W}} \mathbf{z}[k] + \underbrace{\begin{bmatrix} h_{11} & h_{12} & \cdots & h_{1p} \\ h_{21} & h_{22} & \cdots & h_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ h_{N1} & h_{N2} & \cdots & h_{Np} \end{bmatrix}}_{\mathbf{H}} \mathbf{y}[k]$$

$$\mathbf{u}[k] = \underbrace{\begin{bmatrix} g_{11} & g_{12} & \cdots & g_{1N} \\ g_{21} & g_{22} & \cdots & g_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ g_{m1} & g_{m2} & \cdots & g_{mN} \end{bmatrix}}_{\mathbf{G}} \mathbf{z}[k]$$

Only elements corresponding to existing links (link weights) are allowed to be non-zero



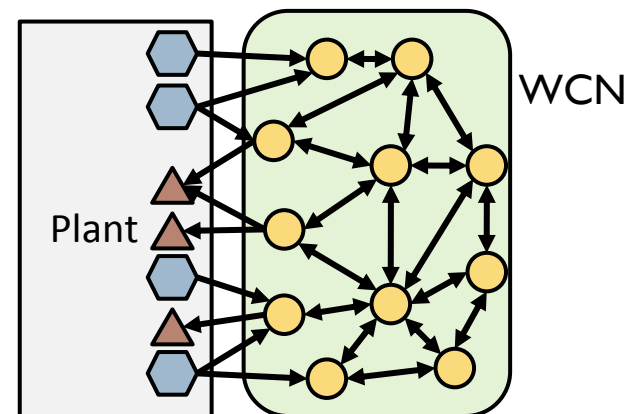
# Closed Loop System

▶ Overall system state:  $\hat{\mathbf{x}}[k] = \begin{bmatrix} \mathbf{x}[k] \\ \mathbf{z}[k] \end{bmatrix}$

▶ Closed loop system:

$$\hat{\mathbf{x}}[k + 1] = \begin{bmatrix} \mathbf{x}[k + 1] \\ \mathbf{z}[k + 1] \end{bmatrix} = \underbrace{\begin{bmatrix} \mathbf{A} & \mathbf{B}\mathbf{G} \\ \mathbf{H}\mathbf{C} & \mathbf{W} \end{bmatrix}}_{\hat{\mathbf{A}}} \begin{bmatrix} \mathbf{x}[k] \\ \mathbf{z}[k] \end{bmatrix} = \hat{\mathbf{A}}\hat{\mathbf{x}}[k]$$

- ▶ Matrices  $\mathbf{W}$ ,  $\mathbf{G}$ ,  $\mathbf{H}$  are **structured**
  - ▶ Sparsity constraints imposed by the WCN topology
  - ▶ A decentralized control problem



# Stabilizing the Closed Loop System

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- ▶ **Problem:** Find numerical matrices **W**, **H**, **G** satisfying structural constraints so that

$$\hat{\mathbf{A}} = \begin{bmatrix} \mathbf{A} & \mathbf{BG} \\ \mathbf{HC} & \mathbf{W} \end{bmatrix} \text{ is stable}$$

- ▶ Use a numerical design procedure to determine suitable **W**, **H**, **G**
  - ▶ Based on Linear Matrix Inequalities
  - ▶ Can also be made robust to Bernoulli packet drops

# Abnormal Behavior in the Network

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- ▶ What if certain nodes in the WCN become faulty or malicious?
- ▶ Security of control networks in industrial control systems is a major issue [*NIST Technical Report, 2008*]
  - ▶ **Data Historian:** Maintain and analyze logs of plant and network behavior
  - ▶ **Intrusion Detection System:** Detect and identify any abnormal activities
- ▶ Is it possible to design a Data Historian/Intrusion Detection System for the WCN?



# Intrusion Detection System

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- ▶ **Obvious Solution (?):** have IDS listen to transmissions of *every node* in WCN and double-check whether they are computing correctly
  - ▶ Obviously not satisfactory
- ▶ Can we get by with listening to the transmissions of only a **subset**  $T$  of the nodes?
  - ▶ Under what conditions?
  - ▶ Which subset should we choose?

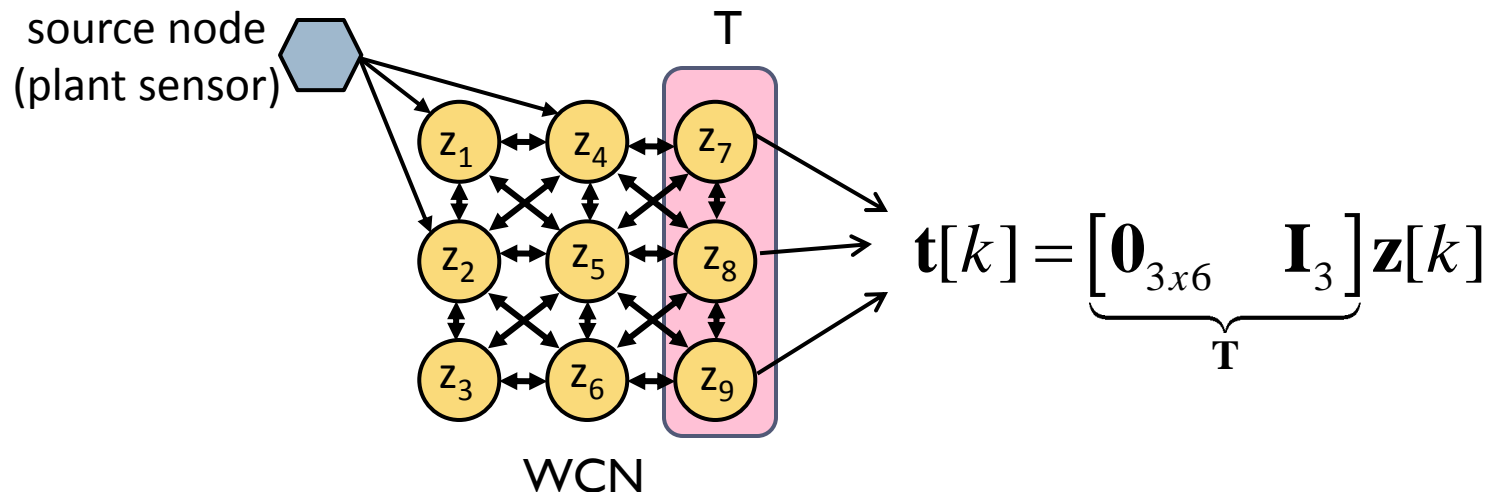
# Monitored Nodes

- ▶ Denote transmissions of any set  $T$  of monitored nodes by

$$\mathbf{t}[k] = \mathbf{Tz}[k]$$

- ▶  $\mathbf{T}$  is a matrix with a single 1 in each row, indicating which elements of  $\mathbf{z}[k]$  are being monitored

- ▶ Example:



# System Model with Malicious Nodes

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- ▶ As before, **linear system model** for WCN with set  $S$  of malicious nodes:

$$\mathbf{z}[k+1] = \mathbf{W}\mathbf{z}[k] + \mathbf{H}\mathbf{y}[k] + \mathbf{B}_S \mathbf{f}_S[k]$$

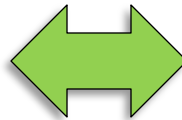
$$\mathbf{t}[k] = \mathbf{T}\mathbf{z}[k]$$

- ▶ **Objective:** Recover  $\mathbf{y}[k]$ ,  $\mathbf{f}_S[k]$  and  $S$  (initial state  $\mathbf{z}[0]$  assumed known)
  - ▶ *Almost* equivalent to **invertibility** of system
  - ▶ Problem: Don't know the set of faulty nodes  $S$ 
    - ▶ Assumption: At most  $b$  faulty nodes
  - ▶ Must ensure that output sequence cannot be generated by a different  $\mathbf{y}[k]$  and possibly different set of  $b$  malicious nodes

# Conditions for IDS/Data Historian Design

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IDS can recover  $\mathbf{y}[k]$  and identify up to  $b$  faulty nodes in the network by monitoring transmissions of set  $T$



Can recover inputs and set  $S$  in system

$$\mathbf{z}[k+1] = \mathbf{W}\mathbf{z}[k] + \begin{bmatrix} \mathbf{H} & \mathbf{B}_S \end{bmatrix} \begin{bmatrix} \mathbf{y}[k] \\ \mathbf{f}_S[k] \end{bmatrix}$$

$$\mathbf{t}[k] = \mathbf{T}\mathbf{z}[k]$$

for **any unknown set**  $S$  of  $b$  nodes



Linear system

$$\mathbf{z}[k+1] = \mathbf{W}\mathbf{z}[k] + \begin{bmatrix} \mathbf{H} & \mathbf{B}_Q \end{bmatrix} \begin{bmatrix} \mathbf{y}[k] \\ \mathbf{f}_Q[k] \end{bmatrix}$$

$$\mathbf{t}[k] = \mathbf{T}\mathbf{z}[k]$$

is invertible for

**any known set**  $Q$  of  $2b$  nodes

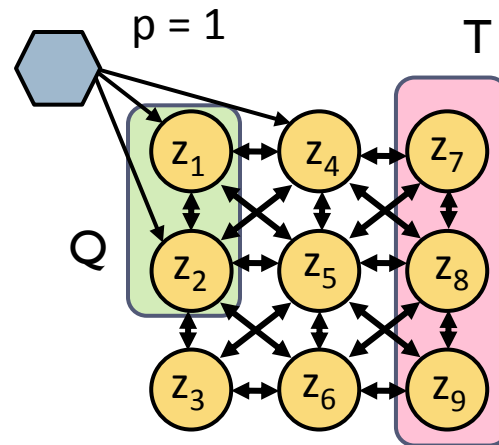
There are  $p+2b$  node-disjoint paths from sensors and any set  $Q$  of  $2b$  nodes to set  $T$



(generically)

# Example

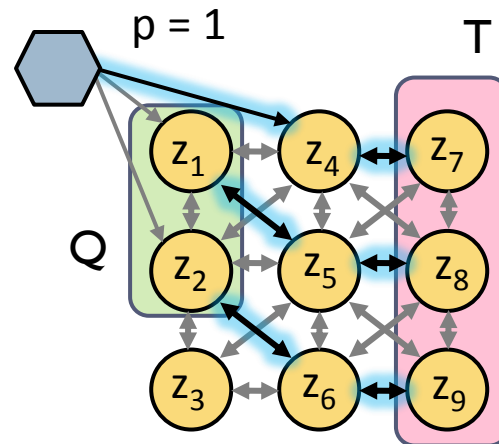
- ▶ Suppose we want to identify  $b = 1$  faulty/malicious node and recover the plant outputs in this setting:



- ▶ Consider set  $Q = \{z_1, z_2\}$

# Example

- ▶ Suppose we want to identify  $b = 1$  faulty/malicious node and recover the plant outputs in this setting:



- ▶ Consider set  $Q = \{z_1, z_2\}$ 
  - ▶  $p+2b$  vertex disjoint paths from sensor and  $Q$  to  $T$
- ▶ Can verify that this holds for any set  $Q$  of  $2b$  nodes
  - ▶ Can identify any single faulty/malicious node and recover  $y[k]$

# Summary

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- ▶ Analyzed linear iterative strategies in networks with malicious nodes
- ▶ Key tool: **structured system theory**
  - ▶ Provides graph-theoretic analysis of linear systems
- ▶ Linear iterative strategies for information dissemination:
  - ▶ Robust to  $b$  malicious nodes if connectivity of network is  $2b+1$  or higher (**strong observability**)
- ▶ Linear iterative strategies for stabilization with a wireless network
  - ▶ IDS can identify  $b$  malicious nodes by monitoring a subset of nodes if those nodes have enough disjoint paths from other nodes (**invertibility**)