



Attack Path Reconstruction from Consequence on Power Grids *with a focus on Monitoring Layer attacks*

J.K. Wang

Chris Moya

Wang.6536@osu.edu

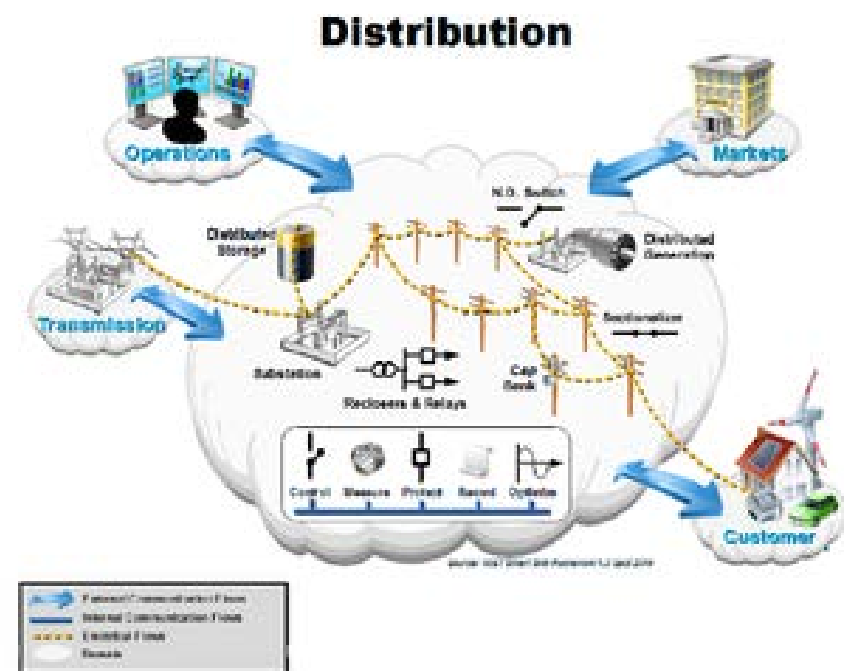
Dept. of Electrical and Computer Engineering

The Ohio State University

Columbus, Ohio, USA

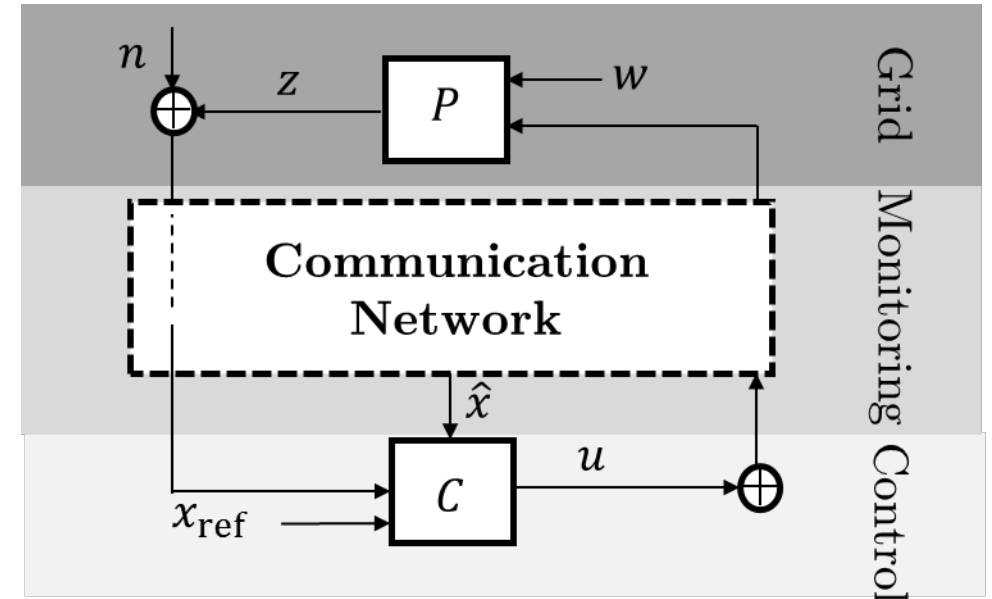
Motivation

- CPS structure of modern power grid.
 - **(Physical) grid**
 - including generators, power delivery equipment and lines, loads and protective devices;
 - **Control layer**
 - including central and local decision making units
 - **Monitoring layer**
 - including meters, actuators and hardware to store and transmit measurements



Motivation

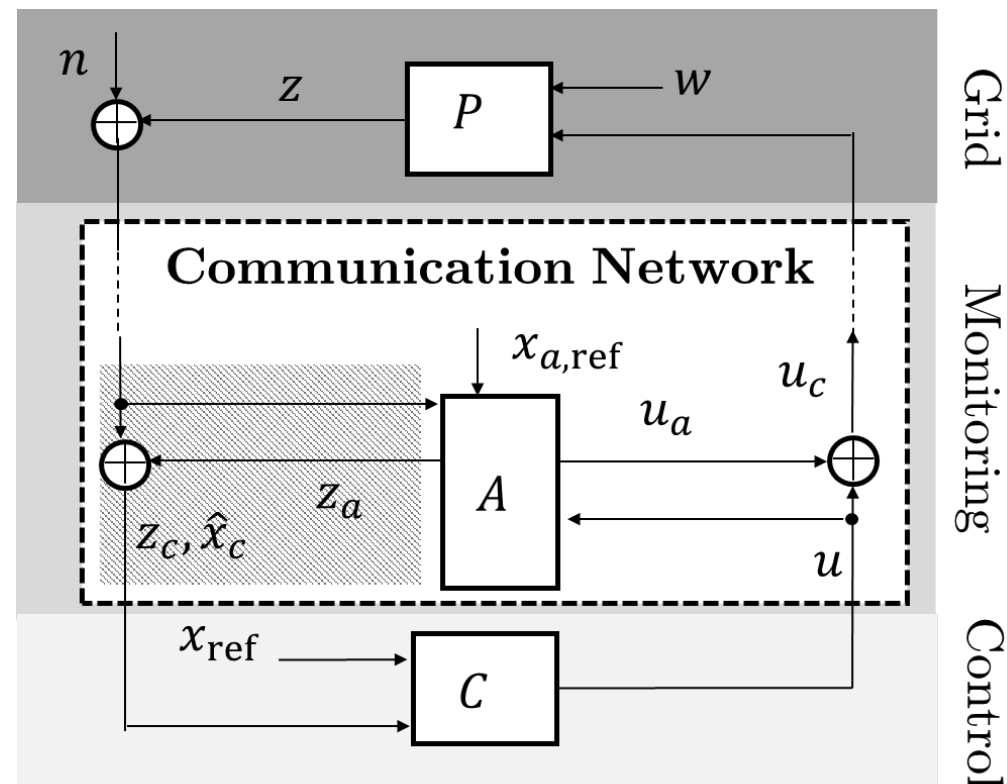
- Monitoring layer
 - **High impact**
 - Connecting a large number of devices over a wide area
 - Hovering over the control layer and the grid
 - **High vulnerability**
 - dedicated protection to individual devices practically impossible
 - nonproprietary cyber-space is resorted to important tasks, including electricity markets.



w system disturbance	z measurement
n measurement noise	\hat{x} estimated state
u operation command	x_{ref} desired state

Motivation

- Measurement attacks
 - Model I: Covert control actions
 - False measurements are injected in order to cover the malicious control actions, while real attacks are launched on the control layer
 - replay attacks,
 - stealth attacks [Amin, 2012],
 - general covert attacks [Smith, 2011], [Teixeira, 2015]...

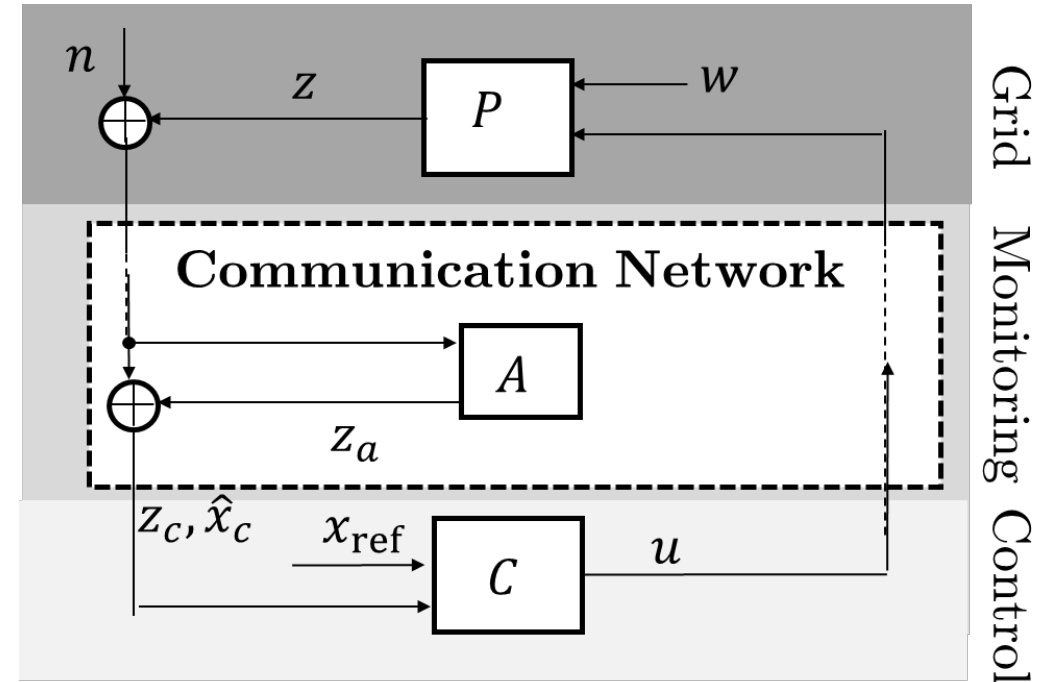


w system disturbance
 n measurement noise
 u operation command
 z true measurement

z_a attacking measurement
 z_c corrupted measurement
 \hat{x}_c corrupted estimated state
 x_{ref} operator's desired state
 $x_{a,ref}$ attacker's desired state

Motivation

- Measurement attacks
 - Model I: Covert control actions
 - Model II: Disabled alarms
 - The attacks that destroy the alarm function of the monitoring layer.
 - The formation of an attacker's action depends on the status of the physical grid.
 - Effect similar to causing "hidden failure."

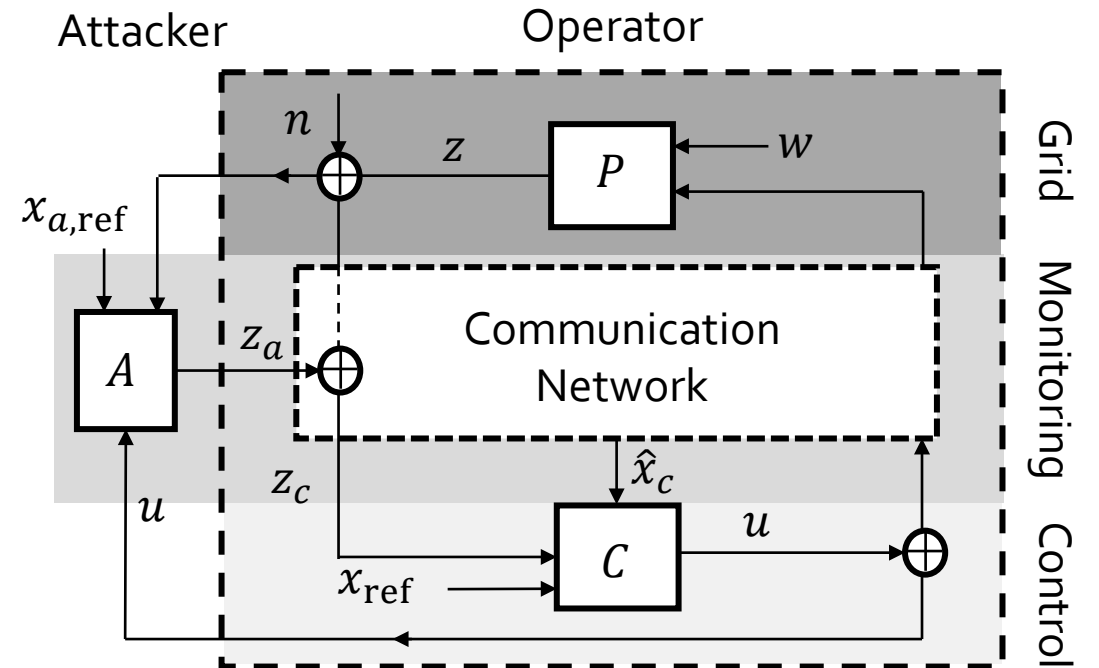


w system disturbance
 n measurement noise
 u operation command
 z true measurement

z_a attacking measurement
 z_c corrupted measurement
 \hat{x}_c corrupted estimated state
 x_{ref} operator's desired state
 $x_{a,\text{ref}}$ attacker's desired state

Motivation

- Measurement attacks
 - Model I: Covert control actions
 - Model II: Disabled alarms
 - Model III: State estimator cheater
 - inject false measurements that can bypass the detection of state estimators [Liu & Ning, 2011], [Kim & Tong, 2014], [Li & Scaglione, 2013]

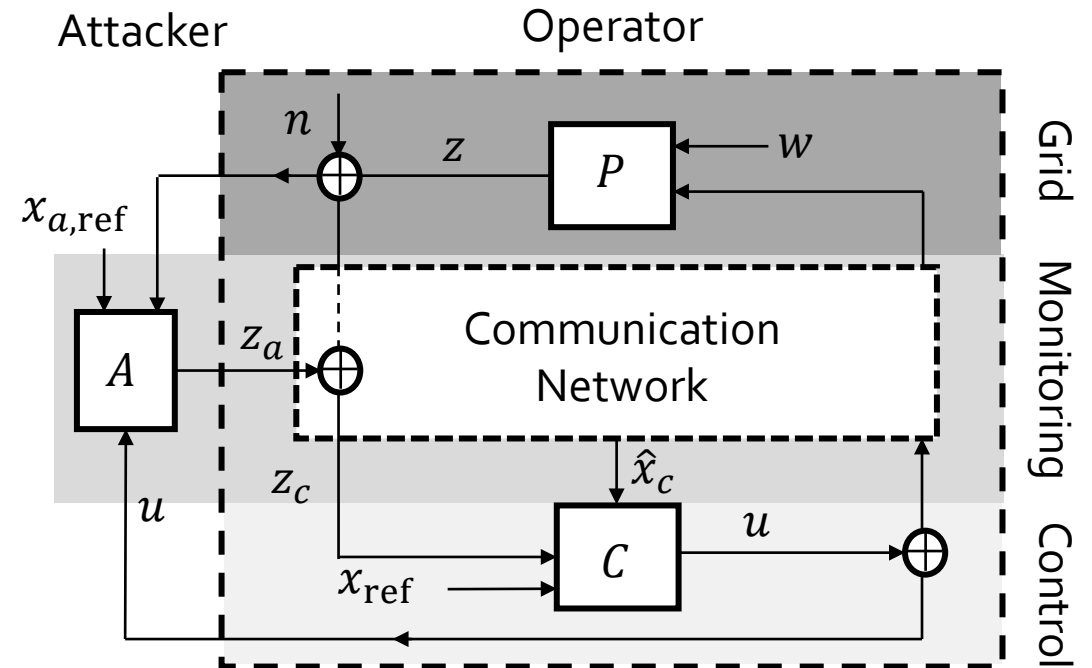


w system disturbance
 n measurement noise
 u operation command
 z true measurement

z_a attacking measurement
 z_c corrupted measurement
 \hat{x}_c corrupted estimated state
 x_{ref} operator's desired state
 $x_{a,ref}$ attacker's desired state

Motivation

- Measurement attacks
 - Model I: Covert control actions
 - Model II: Disabled alarms
 - Model III: State estimator cheater →
Monitoring Layer (ML) attack
- **(ML attacks)** inject falsified measurements \mathbf{z}_a to *manipulate* decision-making processes \mathbf{u} to accomplish the attacking goal $\mathbf{x}_{a,ref}$.

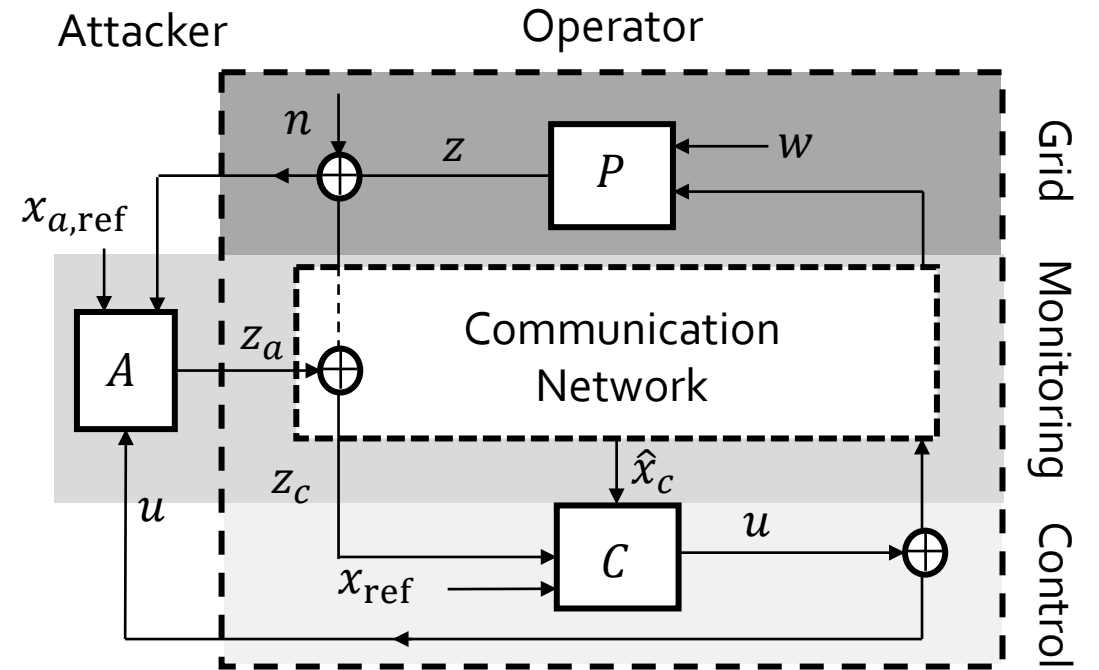


w system disturbance
 n measurement noise
 u operation command
 z true measurement

z_a attacking measurement
 z_c corrupted measurement
 \hat{x}_c corrupted estimated state
 x_{ref} operator's desired state
 $x_{a,ref}$ attacker's desired state

Motivation

- Existing studies' methodology:
 - $\max |z - z_a|_p$.
- Deficiency:
 - An attacker's goal would not be injecting maximum erroneous measurements, but causing intended physical consequences on the grid.
- Research Goal: investigating models of ML attacks based on an attacker's **motivation**



w system disturbance
 n measurement noise
 u operation command
 z true measurement

z_a attacking measurement
 z_c corrupted measurement
 \hat{x}_c corrupted estimated state
 x_{ref} operator's desired state
 $x_{a,ref}$ attacker's desired state



Outline

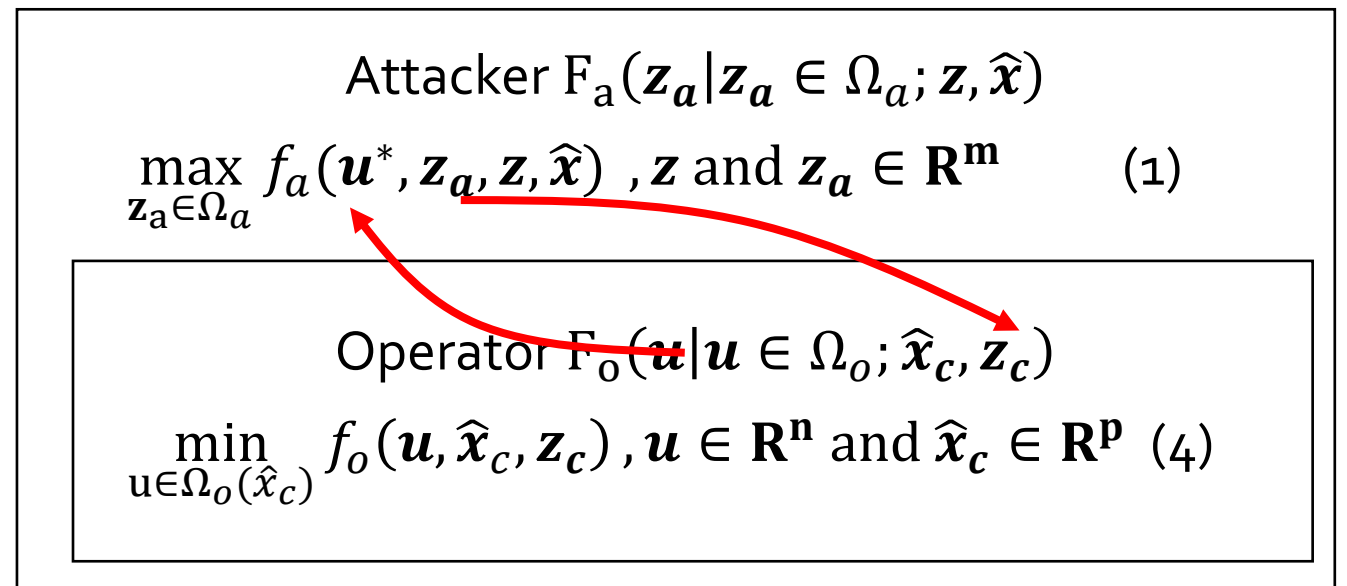
- ~~Motivation~~
- Methodology
- Numerical examples
- Implications

Methodology

- Attacking principle:
 - (**ML attacks**) inject falsified measurements \mathbf{z}_a to *manipulate* decision-making processes \mathbf{u} to accomplish the attacking goal $\mathbf{x}_{a,\text{ref}}$.
- Problem statement:
 - Find \mathbf{z}_a (on monitoring layer) that induces \mathbf{u} (on control layer) that results $\mathbf{x}_{a,\text{ref}}$ (on grid).
- Challenge: complex system
 - Large set of variables
 - Grid dynamics

Methodology

- Assumption:
 - one-step false measurement injection (zero-day attack).
 - Operation objective under manipulation is known.
- Problem form:
 - Bi-level optimization
 - corrupted measurement \mathbf{z}_c
 - $\mathbf{z}_c = \mathbf{z} + \mathbf{z}_a$
 - Manipulated decision \mathbf{u}^*
 - $\mathbf{u}^* = \arg \left\{ \min_{\mathbf{u}} f_o(\mathbf{u}, \hat{\mathbf{x}}_c) \right\}$



Methodology

- Problem structure
 - Attacker's feasible set $\Omega_a(\mathbf{z})$
 - attacker's accessibility to the network and condition to bypass bad data detection in SCADA and EMS
 - Conditions defined in prior works [Liu and Ning, 2011]:
 - Scenario I- Limited access to meters.
 - Scenario II-Limited resources available to compromise meters.

$$\text{Attacker } F_a(\mathbf{z}_a | \mathbf{z}_a \in \Omega_a; \mathbf{z}, \hat{\mathbf{x}})$$
$$\max_{\mathbf{z}_a \in \Omega_a} f_a(\mathbf{u}^*, \mathbf{z}_a, \mathbf{z}, \hat{\mathbf{x}}), \mathbf{z} \text{ and } \mathbf{z}_a \in \mathbf{R}^m \quad (1)$$

$$\text{Operator } F_o(\mathbf{u} | \mathbf{u} \in \Omega_o; \hat{\mathbf{x}}_c, \mathbf{z}_c)$$
$$\min_{\mathbf{u} \in \Omega_o(\hat{\mathbf{x}}_c)} f_o(\mathbf{u}, \hat{\mathbf{x}}_c, \mathbf{z}_c), \mathbf{u} \in \mathbf{R}^n \text{ and } \hat{\mathbf{x}}_c \in \mathbf{R}^p \quad (4)$$

Methodology

- Problem structure
 - Operator's feasible set $\Omega_o(\hat{\mathbf{x}}_c, \mathbf{z}_c)$
 - operation feasible region defined by system states
 - E.g. security constrained optimal power flow.

Attacker $F_a(\mathbf{z}_a | \mathbf{z}_a \in \Omega_a; \mathbf{z}, \hat{\mathbf{x}})$

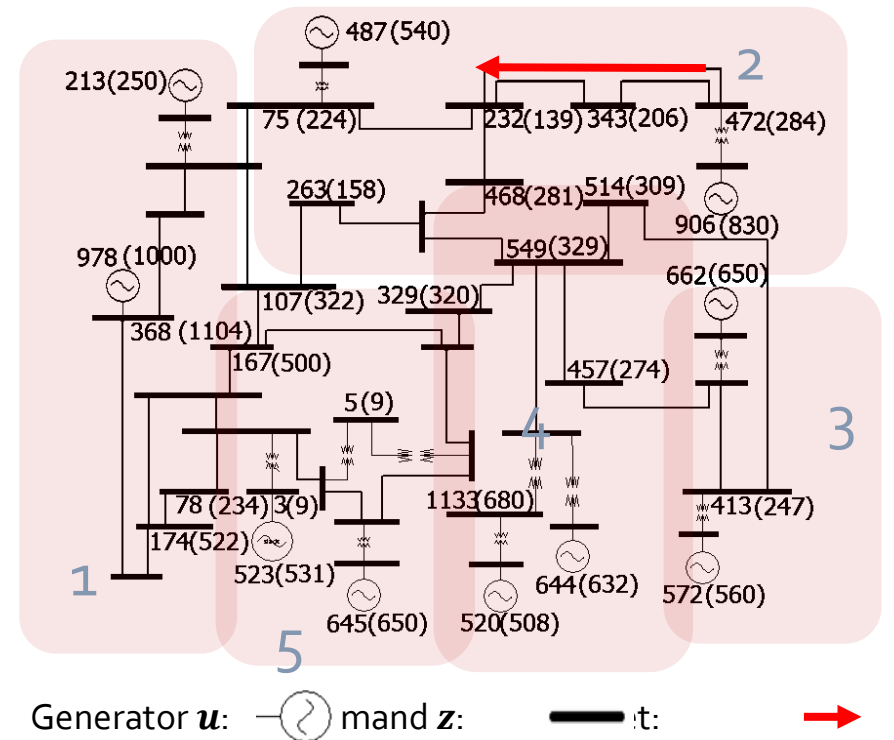
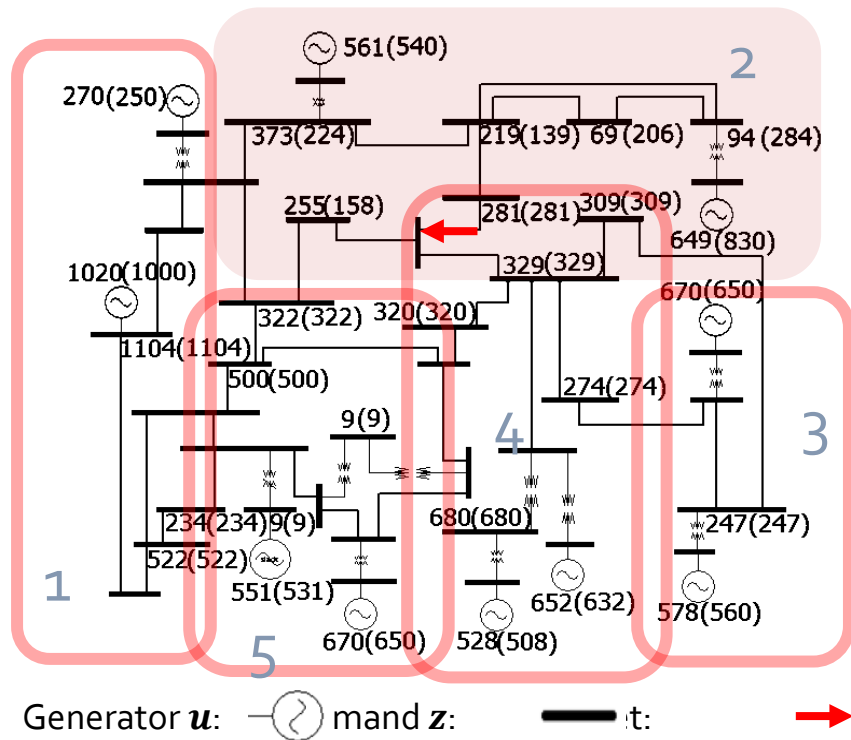
$$\max_{\mathbf{z}_a \in \Omega_a} f_a(\mathbf{u}^*, \mathbf{z}_a, \mathbf{z}, \hat{\mathbf{x}}), \mathbf{z} \text{ and } \mathbf{z}_a \in \mathbf{R}^m \quad (1)$$

Operator $F_o(\mathbf{u} | \mathbf{u} \in \Omega_o; \hat{\mathbf{x}}_c, \mathbf{z}_c)$

$$\min_{\mathbf{u} \in \Omega_o(\hat{\mathbf{x}}_c)} f_o(\mathbf{u}, \hat{\mathbf{x}}_c, \mathbf{z}_c), \mathbf{u} \in \mathbf{R}^n \text{ and } \hat{\mathbf{x}}_c \in \mathbf{R}^p \quad (4)$$

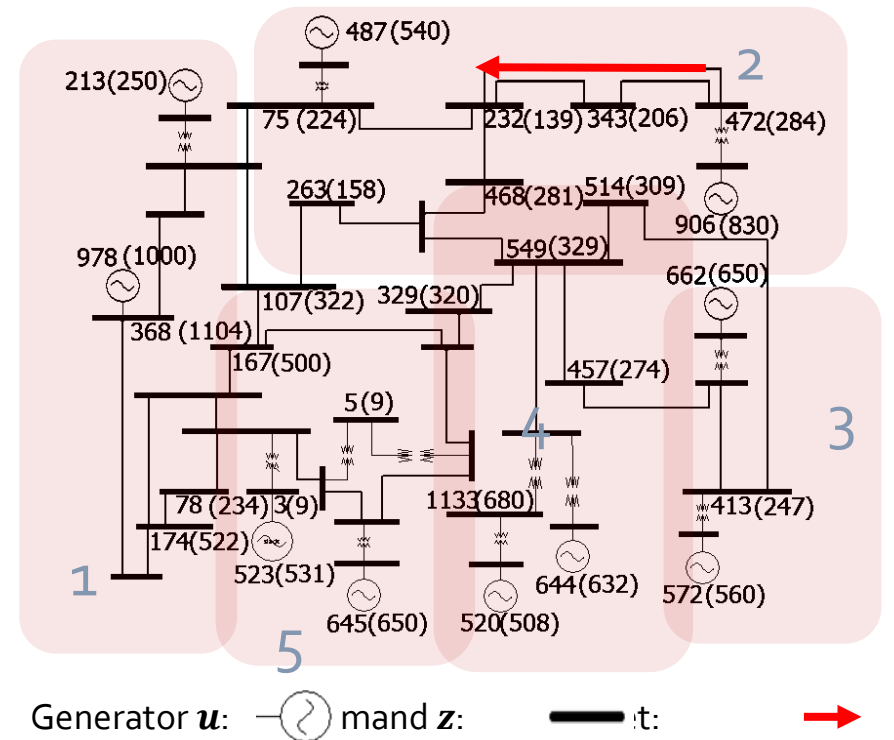
Numerical Example I

- IEEE 39-bus system, 5 substations
 - **Attacker's goal:** Congest transmission lines
 - **Manipulated operation:** security constrained OPF
 - $\mathbf{z}, \mathbf{u}(MW)$



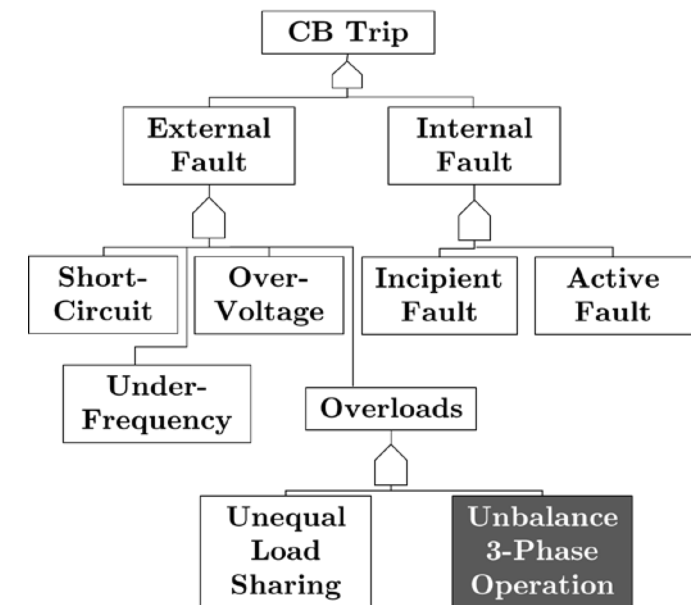
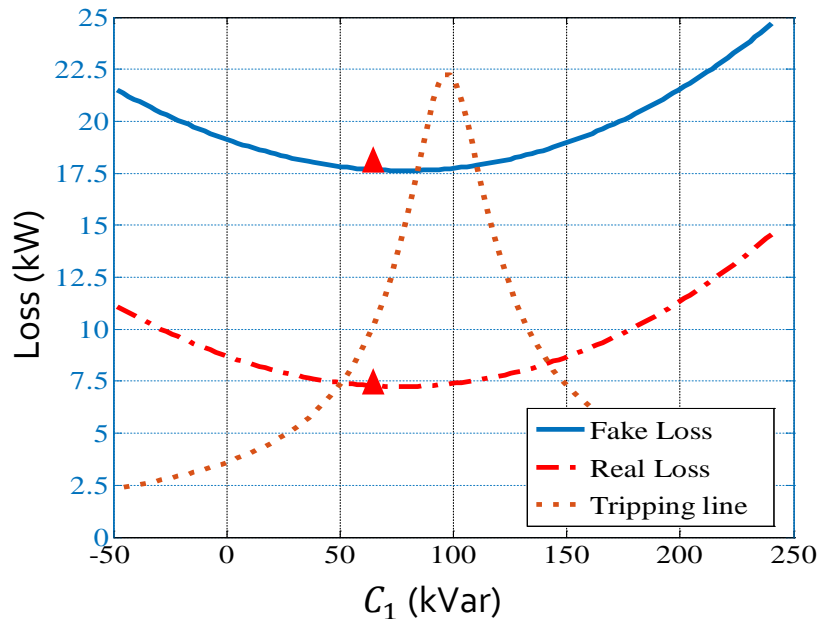
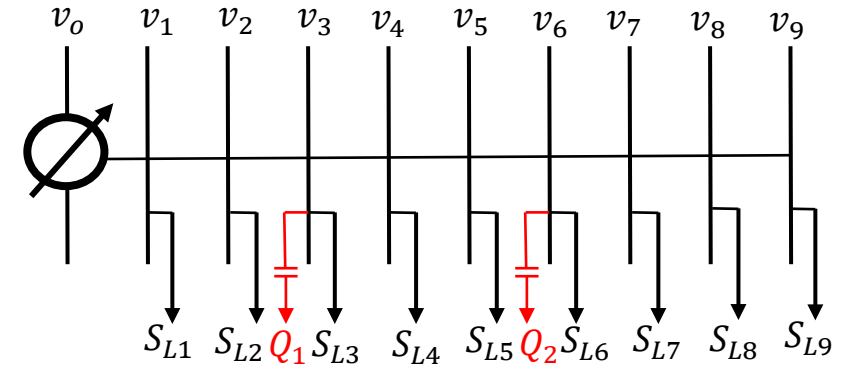
Numerical Example I

- Implication
 - No guarantee that a Play Safe [in game theory] operation strategy is always available (versus [Zhu, 2011]).
 - The correlation pattern of attacks depends on the system topology and original bus power injections.



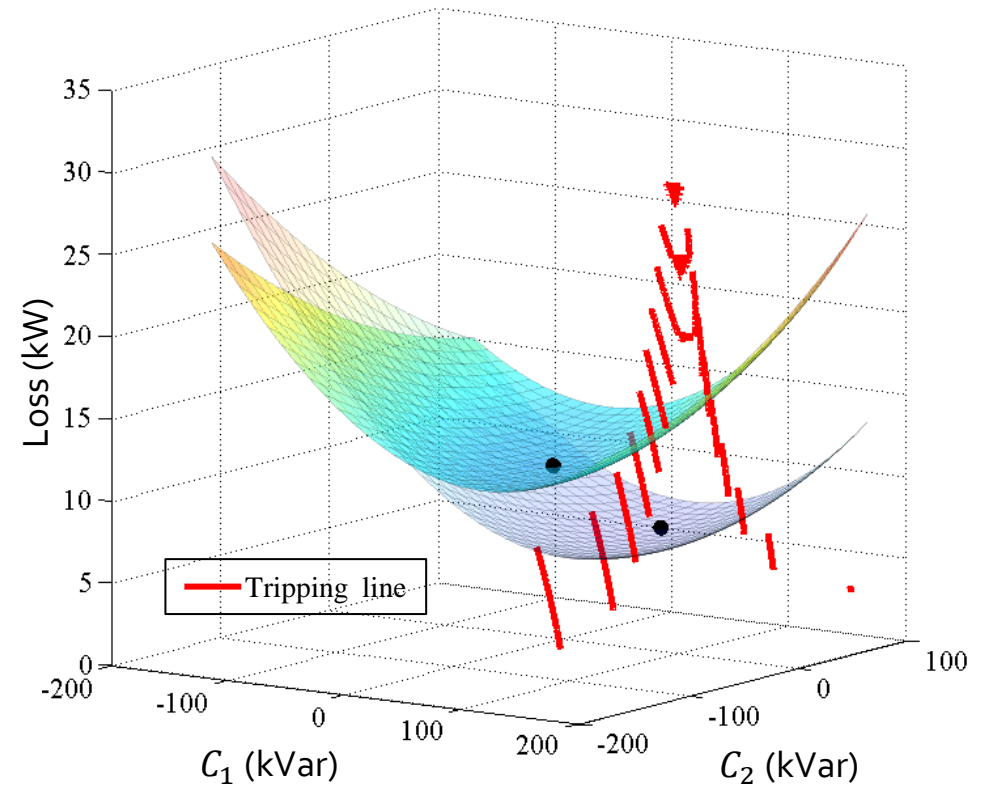
Numerical Example II

- Radial distribution system (IEEE-9 bus system)
- **Attacker's goal:** disrupting regional supply
- **Manipulated operation:** Volt/Var management on the distribution system.
- **Attacking outlet:**
 - Tripping protection on substation transformer.



Numerical Example II

- Implications
 - Attack can be defended by trading off operation goals.
 - Effective trading off can be found (indicated by dual variables in the example).





Questions?

Wang.6536@osu.edu