

Robust PMU Placement

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Optimization Challenges in the evolution of energy networks to smart grids
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- PMU Placement Problem (PPP)
- Literature Review

2. Models for PPP

- Iterative Model
- Bilevel Model
- Computational Results

3. Robust PMU Placement

- Robust vs Redundancy
- Computational Complexity
- Models
- Computational Results

4. Conclusions and Open Questions

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PMU Placement Problem

Preliminaries

Motivation

- ▶ Need to monitor the power system network;
- ▶ Guarantee full observability;
- ▶ Minimize costs.

Preliminaries

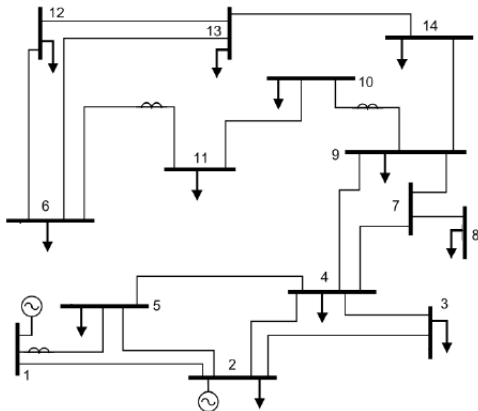
Motivation

- ▶ Need to monitor the power system network;
- ▶ Guarantee full observability;
- ▶ Minimize costs.

Phasor Measurement Units (PMUs)

- ▶ Measuring device;
- ▶ Provides time synchronized phasor measurements;
- ▶ Very expensive.

Example: IEEE 14 Bus



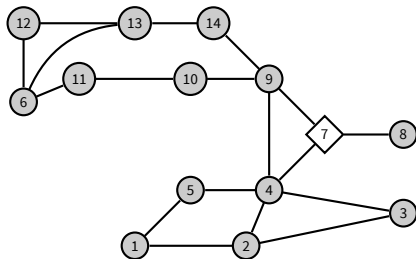
Example: IEEE 14 Bus

Power System Graph (PSG)

$G = (V, E)$: $V = \{\text{set of buses (nodes)}\}$
 $E = \{\text{set of transmission lines (edges)}\}.$

$N(i)$: is the neighborhood of node $i \in V$. $N[i] = N(i) \cup \{i\}.$

S : set of zero-injection nodes.



$$S = \{7\}$$

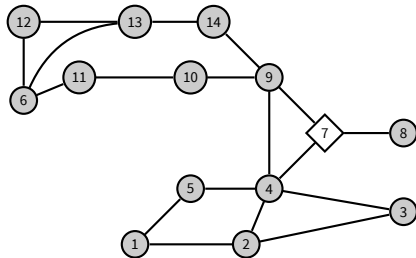
Example: IEEE 14 Bus

A node is **observed** if its voltage is known.

An edge is **observed** if its current is known.

A PSG is **fully observed** if the voltage and current in all its buses and transmission lines are known.

The **PMU placement problem** (PPP) consists in deciding the localization of PMUs such that the PSG is fully observed and the number (or cost) of PMUs is minimized.



$$S = \{7\}$$

Properties to reduce the number of PMUs

Laws

Ohm's law $I_{ij} = \frac{V_i - V_j}{R_{ij}}$, where R_{ij} is the resistance.

Kirchhoff's law $\sum_{j \in N(v_i)} I_{ij} = 0$ for a zero-injection bus i .

Rule 1: A PMU in a bus i measures its voltage V_i and the current I_{ij} for all $j \in N(i)$. By Ohm's law, the voltage for bus j can be computed.

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Rule 1: A PMU in a bus i measures its voltage V_i and the current I_{ij} for all $j \in N(i)$. By Ohm's law, the voltage for bus j can be computed.

Rule 2: A zero-injection node i is observed if all of its neighbors are observed.

Rule 3: If zero-injection node j and all its neighbors, except node i , are observed, then i is observed.

Summary of the Rules

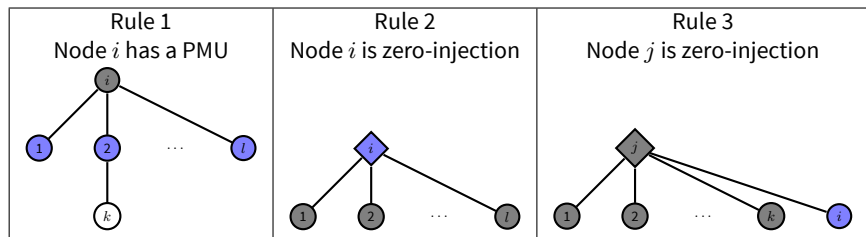


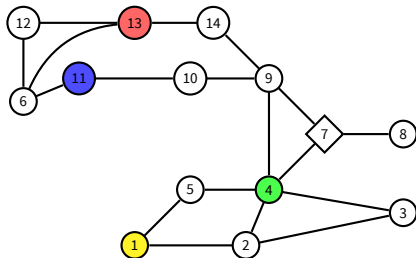
Figure 1: The grey nodes are observed. The blue nodes become observed.

Lemma

If the buses of a PSG are all observed, then the PSG is fully observed.

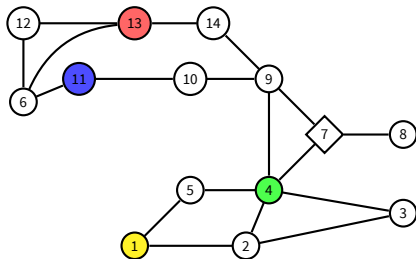
Example

Colored nodes have a PMU.



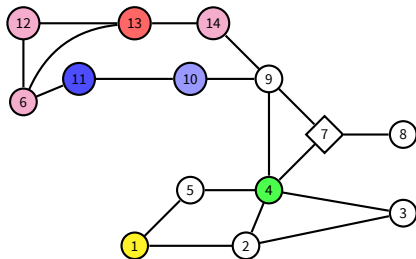
Example

Apply Rule 1.



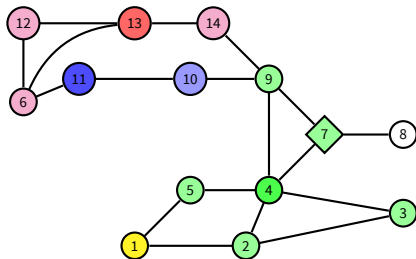
Example

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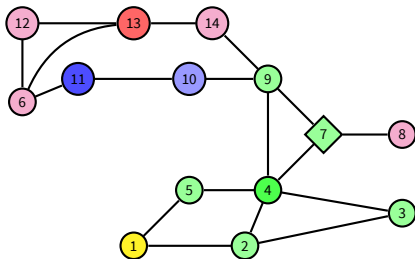
Example

Apply Rule 1.



Example

Apply Rule 1. Apply Rule 3.



Literature Review

State-of-the-art: Solution Techniques

D. Dua, S. Dambhare, R. K. Gajbhiye and S. A. Soman. *Optimal multistage scheduling PMU placement: An ILP approach*. IEEE Transactions on Power Delivery, 2008.

Model the problem with binary linear programming formulation. Consider the possibility of redundancy. **Limited propagation depth of the rules.**

N. Fan and J-P. Watson. *On integer programming models for the multi-channel PMU placement problem and their solution*. Energy Systems, 2006.

Propose a more general model with L-channel PMU's. The model is exponential. **Limited propagation depth of the rules.**

State-of-the-art: Solution Techniques

P-L. Poirion, S. Toubaline, C. D'Ambrosio and L. Liberti. *Observing the State of a Smart Grid Using Bilevel Programming*. Networks, 2016.

Model the 1-channel PMU problem with all nodes being zero-injection. First formulation properly modeling the propagation rules.

N. M. Manousakis, G. N. Korres and P. S. Georgilakis. *Taxonomy of PMU placement methodologies*. IEEE Transactions on Power Systems, 2012.

N. M. Manousakis, G. N. Korres and P. S. Georgilakis. *Optimal Placement of phasor measurement units: A literature review*. In 16th Int. Conference Intelligent System Application to Power Syst., 2011.

Provide a literature review of the formulations and solution techniques that have been presented in the literature.

State-of-the-art: computational complexity

D. J. Brueni and L. S. Heath. *The PMU placement problem*. SIAM Journal on Discrete Mathematics, 2005.

Prove that the PPP without zero-injection nodes is NP-hard even for planar bipartite graphs.

D. Gyllstrom, E. Rosensweig and J. Kurose . *On the impact of PMU placement on observability and cross-validation*. In Proceedings of the 3rd International Conference on Future Energy Systems, 2012.

The PPP is NP-hard as well as some variants of it. The variants are: 1) maximize the number of observed buses for a fixed number of PMUs, 2) minimize the number of PMUs such that full observability is ensured as well as cross validation, and 3) maximize the number of observed buses for a fixed number of PMUs and ensure cross-validation.

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Decision variables

$$\forall i \in V \quad x_i = \begin{cases} 1 & \text{if a PMU is installed in node } i \\ 0 & \text{otherwise} \end{cases}$$

Iterative Model

$$\forall i \in V, \forall d = 0, \dots, T - 1 \quad w_{i,d} = \begin{cases} 1 & \text{if node } i \text{ is observed in iteration } d \\ 0 & \text{otherwise} \end{cases}$$

Iterative Model

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$$\forall i \in S, \forall d = 0, \dots, T - 1 \quad y_{i,d}^2 = \begin{cases} 1 & \text{if Rule 2 is applied in iteration } d \\ & \text{to observe node } i \\ 0 & \text{otherwise} \end{cases}$$

Iterative Model

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$$\forall i \in V, \forall j \in N(i) \cap S, \forall d = 0, \dots, T-1 \quad y_{i,j,d}^3 = \begin{cases} 1 & \text{if Rule 3 is applied to} \\ & \text{node } j \text{ in iteration } d \\ & \text{to observe node } i \\ 0 & \text{otherwise} \end{cases}$$

Iterative Model

$$\min_{x, w, y^2, y^3} \text{binary}$$

s. t

$$\sum_{i \in V} x_i$$

$$w_{i, T-1} = 1, \quad \forall i \in V$$

$$x_i + \sum_{j \in N(i)} x_j \geq w_{i, 0}, \quad \forall i \in V$$

Full observability

Rule 1

Iterative Model

$$\min_{x, w, y^2, y^3} \text{binary}$$

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$$x_i + \sum_{j \in N(i)} x_j \geq w_{i, 0}, \quad \forall i \in V$$

$$w_{i, d+1} \geq w_{i, d}, \quad \forall i \in V, d = 0, \dots, T-2$$

Full observability

Rule 1

Monotone

Iterative Model

$$\begin{array}{ll}
 \min_{x, w, y^2, y^3} & \sum_{i \in V} x_i \\
 \text{s. t.} & w_{i, T-1} = 1, \quad \forall i \in V \quad \text{Full observability} \\
 & x_i + \sum_{j \in N(i)} x_j \geq w_{i, 0}, \quad \forall i \in V \quad \text{Rule 1} \\
 & w_{i, d+1} \geq w_{i, d}, \quad \forall i \in V, d = 0, \dots, T-2 \quad \text{Monotone} \\
 & w_{i, d+1} - w_{i, d} \leq y_{i, d}^2 + \sum_{j \in N(i) \cap S} y_{i, j, d}^3, \quad \forall i \in S, d = 0, \dots, T-2 \quad \text{Active rule}
 \end{array}$$

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 & w_{i, d+1} - w_{i, d} \leq \sum_{j \in N(i) \cap S} y_{i, j, d}^3, \quad \forall i \in V \setminus S, d = 0, \dots, T-2 \quad \text{Active rule}
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 \end{array}$$

Iterative Model

Lemma

The binary requirement on the decision variables w can be relaxed to $[0, 1]$.

Iterative Model

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Lemma

It is enough to consider $T = |S| + 1$ for the binary programming formulation to be correct.

Bilevel Model

$$\begin{aligned} & \min_{x \in \{0,1\}^{|V|}} \sum_{i \in V} x_i \\ & \text{subject to } \sum_{i \in V} w_i \geq |V| \end{aligned}$$

Full observability

Bilevel Model

$$\min_{x \in \{0,1\}^{|V|}} \sum_{i \in V} x_i$$

subject to $\sum_{i \in V} w_i \geq |V|$

Full observability

$$\min_{w \in \{0,1\}^{|V|}} \sum_{i \in V} w_i$$

$$\text{s.t. } w_i \geq x_i, \quad \forall i \in V$$

Rule 1

$$w_i \geq x_j, \quad \forall i \in V, \forall j \in N(i)$$

Rule 1

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Bilevel Model

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Full observability

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Bilevel Model: solution approach

Possible directions to solve the bilevel model:

1. There is a MILP which is equivalent to the bilevel model. However, it demands the use of big M values.
2. Solve

$$\min_{x \in \{0,1\}^{|V|}} \sum_{i \in V} x_i \quad (1)$$

$$\text{s.t.} \quad \sum_{i \in V: \bar{x}_i = 0} x_i \geq 1, \quad \forall \bar{x} \text{ infeasible.} \quad (2)$$

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Improvements: 1) \bar{x} must be maximal.

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Improvements: 1) \bar{x} must be maximal.

$$2) \quad \sum_{i \in V: \bar{x}_i=0 \wedge \exists j \in N[i]: \bar{w}_j=0} x_i \geq 1$$

Computational Results

IEEE System		Dua et al.			It. Model			Bi. Model			Imp. Bi. Model		
	$ S $	# PMUs	# PMUs	time	# PMUs	iter	time	# PMUs	iter	time	# PMUs	iter	time
14 Bus	1	3	3	0.01	3	216	7.08	3	11	0.01	3	11	0.01
30 Bus	0	-	10	0.00			tl	10	19	0.02			
57 Bus	15	14	11	0.97			tl	11	41	0.08			
118 Bus	10	29	29	0.24			tl	29	86	0.30			

Table 1: Comparison with Dua et al. of the iterative and bilevel models.

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Robust PMU Placement Problem

Robust and Redundancy

The **redundancy** of a node i reflects the number of PMUs that observe i .

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The **robustness** of a node i measures the number of PMUs that can be attacked and i is still observed.

Robust and Redundancy

The **redundancy** of a node i reflects the number of PMUs that observe i .

The **robustness** of a node i measures the number of PMUs that can be attacked and i is still observed.

Motivation: PMUs can fail and full observability must still be guaranteed.

Robust Vs Redundancy

Redundancy = 1, i.e., each node must be observed twice.

Robust Vs Redundancy

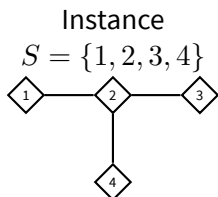
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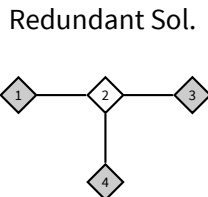
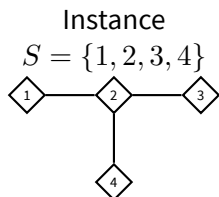
Redundant Sol.

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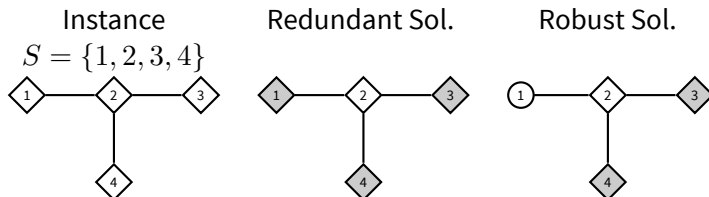


Robust Sol.

Robust Vs Redundancy

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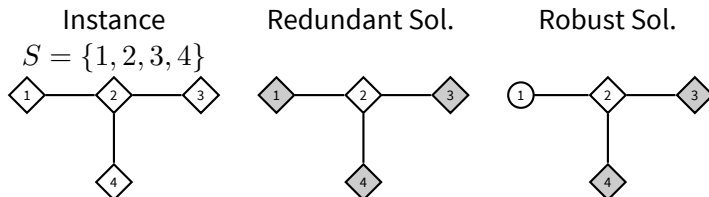
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Robust Vs Redundancy

Redundancy = 1, i.e., each node must be observed twice.

Robustness = 1, i.e., at most one PMU will be attacked.



The robust solution is less expensive.

Attacker's Problem: unable full observability

Attacker's Problem

Given a PMU placement strategy x ,

$$\min_{w, \in \{0,1\}^{|V|}} \sum_{i \in V} w_i$$

s.t.

$$w_i \geq x_i, \quad \forall i \in V$$

$$w_i \geq x_j, \quad \forall i \in V, \forall j \in N(i)$$

$$w_i \geq 1 - |N(i)| + \sum_{j \in N(i)} w_j, \quad \forall i \in S$$

$$w_i \geq 1 - |N(j)| + w_j + \sum_{k \in N(j) \setminus \{i\}} w_k, \quad \forall i \in V, \forall j \in N(i) \cap S$$

Attacker's Problem

Given a PMU placement strategy x , the attacker solves:

$$\begin{aligned}
 & \min_{w, z \in \{0,1\}^{|V|}} \sum_{i \in V} w_i \\
 & \text{s.t.} \quad \sum_{i \in V} z_i \leq \rho \\
 & \quad w_i \geq x_i(1 - z_i), \quad \forall i \in V \\
 & \quad w_i \geq x_j(1 - z_j), \quad \forall i \in V, \forall j \in N(i) \\
 & \quad w_i \geq 1 - |N(i)| + \sum_{j \in N(i)} w_j, \quad \forall i \in S \\
 & \quad w_i \geq 1 - |N(j)| + w_j + \sum_{k \in N(j) \setminus \{i\}} w_k, \quad \forall i \in V, \forall j \in N(i) \cap S
 \end{aligned}$$

Attacker's decision problem

Problem: ATTACK

Instance: A graph $G = (V, E)$, a set of zero-injection nodes $S \subseteq V$, a set of nodes $X \subseteq V$ with PMUs and two positive integers ρ and K .

Question: Does there exist a set $Z \subseteq X$ such that $|Z| \leq \rho$ and the set of observed nodes when the PMUs are in $X \setminus Z$ decreases by K ?

Attacker's problem complexity

Theorem

The ATTACK is NP-complete.

Proof: Knapsack is contained in ATTACK.

Problem: KNAPSACK

Instance: A sequence of positive integers $p_1, \dots, p_n, w_1, \dots, w_n$ and two positive integers C and P .

Question: Does there exist a set $J \subseteq \{1, \dots, n\}$ such that

$$\sum_{i \in J} w_i \leq C \text{ and } \sum_{i \in J} p_i \geq P ?$$

Attacker's problem complexity

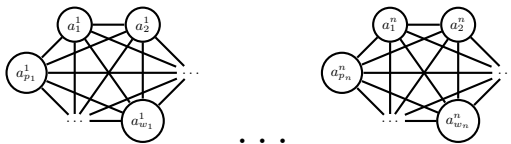


Figure 2: The blue nodes have a PMU. The blue and grey nodes are observed. The red nodes are attacked.

Attacker's problem complexity

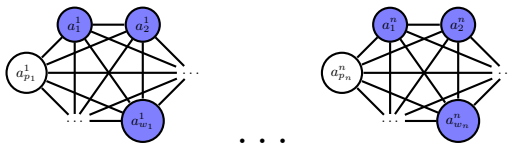


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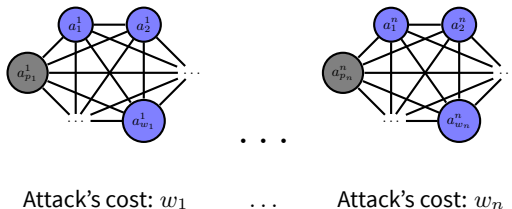


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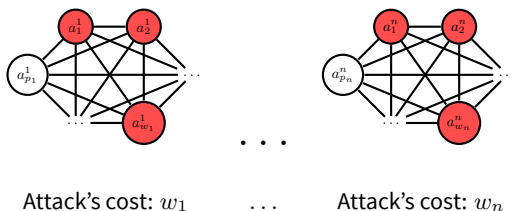


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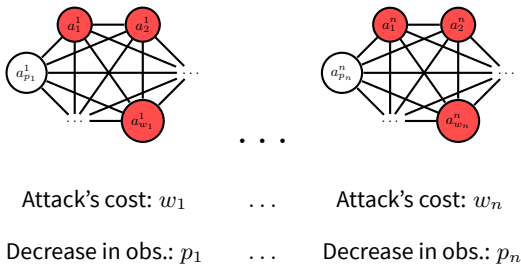


Figure 2: The blue nodes have a PMU. The blue and grey nodes are observed. The red nodes are attacked.

Solving the Robust PPP: Iterative and Bilevel Models

Iterative Model

$$\begin{aligned}
 & \min_{x, w, y^2, y^3} \sum_{i \in V} x_i \\
 & \text{subject to} \\
 & w_{i, T-1} = 1, \quad \forall i \in V, \\
 & x_i + \sum_{j \in N(i)} x_j \geq w_{i, 0}, \quad \forall i \in V, \\
 & w_{i, d+1} \geq w_{i, d}, \quad \forall i \in V, d = 0, \dots, T-2, \\
 & w_{i, d+1} - w_{i, d} \leq y_{i, d}^2 + \sum_{j \in N(i) \cap S} y_{i, j, d}^3, \quad \forall i \in S, d = 0, \dots, T-2, \\
 & w_{i, d+1} - w_{i, d} \leq \sum_{j \in N(i) \cap S} y_{i, j, d}^3, \quad \forall i \in V \setminus S, d = 0, \dots, T-2, \\
 & y_{i, j, d}^3 \leq w_{j, d}, \quad \forall i \notin V, \forall j \in N(i) \cap S, d = 0, \dots, T-2, \\
 & y_{i, j, d}^3 \leq w_{k, d}, \quad \forall i \in V, \forall j \in N(i) \cap S, \forall k \in N(j) \setminus \{i\}, d = 0, \dots, T-2, \\
 & y_{i, d}^2 \leq w_{j, d}, \quad \forall i \in S, \forall j \in N(i), d = 0, \dots, T-2,
 \end{aligned}$$

Iterative Model

$$\begin{aligned}
 & \min_{x, w, y^2, y^3 \text{ binary}} \sum_{i \in V} x_i \\
 & \text{subject to} \quad w_{i, T-1}^u = 1, \quad \forall i \in V, \forall u \in \mathcal{U} \\
 & \quad x_i + \sum_{j \in N(i)} x_j \geq w_{i, 0}, \quad \forall i \in V, \\
 & \quad w_{i, d+1} \geq w_{i, d}, \quad \forall i \in V, d = 0, \dots, T-2, \\
 & \quad w_{i, d+1} - w_{i, d} \leq y_{i, d}^2 + \sum_{j \in N(i) \cap S} y_{i, j, d}^3, \quad \forall i \in S, d = 0, \dots, T-2, \\
 & \quad w_{i, d+1} - w_{i, d} \leq \sum_{j \in N(i) \cap S} y_{i, j, d}^3, \quad \forall i \in V \setminus S, d = 0, \dots, T-2, \\
 & \quad y_{i, j, d}^3 \leq w_{j, d}, \quad \forall i \notin V, \forall j \in N(i) \cap S, d = 0, \dots, T-2, \\
 & \quad y_{i, j, d}^3 \leq w_{k, d}, \quad \forall i \in V, \forall j \in N(i) \cap S, \forall k \in N(j) \setminus \{i\}, d = 0, \dots, T-2, \\
 & \quad y_{i, d}^2 \leq w_{j, d}, \quad \forall i \in S, \forall j \in N(i), d = 0, \dots, T-2,
 \end{aligned}$$

where \mathcal{U} are all possible attack scenarios.

Iterative Model

$$\begin{aligned}
 & \min_{x, w, y^2, y^3 \text{ binary}} \sum_{i \in V} x_i \\
 & \text{subject to} \quad w_{i, T-1}^u = 1, \quad \forall i \in V, \forall u \in \mathcal{U} \\
 & \quad x_i(1 - z_i^u) + \sum_{j \in N(i)} x_j(1 - z_j^u) \geq w_{i, 0}^u, \quad \forall i \in V, \forall u \in \mathcal{U} \\
 & \quad w_{i, d+1}^u \geq w_{i, d}^u, \quad \forall i \in V, d = 0, \dots, T-2, \forall u \in \mathcal{U} \\
 & \quad w_{i, d+1}^u - w_{i, d}^u \leq y_{i, d}^{2, u} + \sum_{j \in N(i) \cap S} y_{i, j, d}^{3, u}, \quad \forall i \in S, d = 0, \dots, T-2, \forall u \in \mathcal{U} \\
 & \quad w_{i, d+1}^u - w_{i, d}^u \leq \sum_{j \in N(i) \cap S} y_{i, j, d}^{3, u}, \quad \forall i \in V \setminus S, d = 0, \dots, T-2, \forall u \in \mathcal{U} \\
 & \quad y_{i, j, d}^{3, u} \leq w_{j, d}^u, \quad \forall i \notin V, \forall j \in N(i) \cap S, d = 0, \dots, T-2, \forall u \in \mathcal{U} \\
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 & \min_{x, w, y^2, y^3 \text{ binary}} \sum_{i \in V} x_i \\
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 & \quad x_i(1 - z_i^u) + \sum_{j \in N(i)} x_j(1 - z_j^u) \geq w_{i, 0}^u, \quad \forall i \in V, \forall u \in \mathcal{U} \\
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 \end{aligned}$$

where \mathcal{U} are all possible attack scenarios. The size of \mathcal{U} can be exponential...

Algorithm

Algorithm 1

Input: A Robust PPP instance.

Output: A robust optimal solution or proves infeasibility

1. $\bar{\mathcal{U}} \leftarrow \emptyset$
2. $x \leftarrow$ solve the Robust Iterative Model with $\bar{\mathcal{U}}$.
If infeasible, **return** infeasible.
3. $z, w \leftarrow$ solve the attacker's problem for x .
4. **If** $\sum_{i \in V} w_i = |V|$, **return** x .
Else $\bar{\mathcal{U}} \leftarrow \bar{\mathcal{U}} \cup \{z\}$, and go to step 2.

Bilevel Model

$$\begin{aligned} \min_{x \in \{0,1\}^{|V|}} \quad & \sum_{i \in V} x_i \\ \text{s. t.} \quad & \sum_{i \in V} w_i \geq |V| \end{aligned}$$

Bilevel Model

$$\begin{aligned}
 & \min_{x \in \{0,1\}^{|V|}} \sum_{i \in V} x_i \\
 & \text{s. t.} \quad \sum_{i \in V} w_i \geq |V| \\
 & \quad \min_{z, w \in \{0,1\}^{|V|}} \sum_{i \in V} w_i \\
 & \quad \text{s.t.} \quad \sum_{i \in V} z_i \leq \rho \\
 & \quad w_i \geq x_i(1 - z_i), \quad \forall i \in V \\
 & \quad w_i \geq x_j(1 - z_j), \quad \forall i \in V, \forall j \in N(i) \\
 & \quad w_i \geq 1 - |N(i)| + \sum_{j \in N(i)} w_j, \quad \forall i \in S \\
 & \quad w_i \geq 1 - |N(j)| + w_j + \sum_{k \in N(j) \setminus \{i\}} w_k, \quad \forall i \in V, \forall j \in N(i) \cap S
 \end{aligned}$$

Algorithm

Algorithm II

Input: A Robust PPP instance and a lower bound $LB \geq 1$ on the number of PMUs.

Output: A robust optimal solution or proves infeasibility

$$1. \bar{U} \leftarrow \{x \in \{0, 1\}^{|V|} : \sum_{i \in V} x_i \geq \rho + LB\}$$

$$2. \bar{x} \leftarrow \operatorname{argmin}_{x \in \bar{U}} \sum_{i \in V} x_i.$$

If infeasible, **return** infeasible.

$$3. \bar{z}, \bar{w} \leftarrow \text{solve attacker's problem for } \bar{x}.$$

If $\sum_{i \in V} w_i = |V|$, **return** \bar{x} .

4. Make \bar{x} maximal.

If \bar{x} is not capable of guaranteeing full observability, **return** infeasible.

$$\bar{U} \leftarrow \bar{U} \cap \{x : \sum_{i \in V : \bar{x}=0 \wedge \exists j \in N[i]; \bar{w}_j=0} x_i \geq 1\}$$

Go to step 2.

Computational Results

IEEE System	$\rho = 0$		Robust Solution $\rho = 1$						Robust Solution $\rho = 5$					
	num. PMUs	$ S $	num. PMUs	time	iter	UB	R1+R2 time	iter	num. PMUs	time	iter	UB	R1+R2 time	iter
14 Bus	3	1	7	0.92	12	7	0.19	12	inf	0.80	11	inf	0.22	10
30 Bus	10	0	21	0.43	24	-	-	-	inf	0.17	8	-	-	-
57 Bus	11	15		tl		27	24.12	43	inf*			inf	4.44	16
118 Bus	29	10				62	34.01	80	inf*			inf	5.68	26

Table 2: Algorithm I. The value inf* means that even if there is a PMU in all nodes, there is an attack that strictly decreases observability.

IEEE System	Our Model		Robust Solution $\rho = 1$			Robust Solution $\rho = 5$		
	num. PMUs	$ S $	num. PMUs	time	iter	num. PMUs	time	iter
14 Bus	3	1	7	0.80	12	inf	0.04	1
30 Bus	10	0	21	0.39	24	inf	0.13	6
57 Bus	11	15		tl		inf	tl	
118 Bus	29	10						

Table 3: Algorithm I with valid cuts of Algorithm II.

Computational Results

IEEE System	Our Model		Robust Solution $\rho = 1$			Robust Solution $\rho = 5$		
	num. PMUs	$ S $	num. PMUs	time	iter	num. PMUs	time	iter
14 Bus	3	1	7	0.15	25	inf	0.02	1
30 Bus	10	0	21	0.61	68	inf	0.02	2
57 Bus	11	15	23	4.20	163	inf	0.12	5
118 Bus	29	10	62	9.55	254	inf	0.13	3

Table 4: Algorithm II.

Contents

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4. Conclusions and Open Questions

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For the first time a robust approach was analyzed, instead of a redundant one.

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We proved that the attacker's problem is NP-complete.

Conclusions

For the first time a robust approach was analyzed, instead of a redundant one.

The mathematical formulations properly reflect propagation rules. For PPP and robust PPP two formulations were proposed: an iterative and a bilevel programming program.

We proved that the attacker's problem is NP-complete.

We were able to solve PPP and robust PPP for the 4 IEEE instances generally used in the literature.

On going work

More computational tests to validate our solution techniques.

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Consider the possibility of L-channel PMUs and their combination with other measuring devices.

Thank you for your attention

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