# Phase Retrieval via Model-Free Power Flow Jacobian Recovery

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- 1. What is the problem?
- 2. How can we solve it?
- 3. Analytical Results
- 4. Computational Results
- 5. Outlook

# What is the problem?

Problem 1: The case of the missing power network model

#### An electric power network is a graph

$$\mathcal{G} = (\mathcal{N}, \mathcal{V}).$$

#### Every node $i \in \mathcal{N}$ has a complex **voltage phasor** and **power injection**:

$$v_i \angle \theta_i \in \mathbb{C}, \qquad p_i + jq_i \in \mathbb{C}.$$

#### The network topology is encoded in the *admittance* matrix:

$$\mathbf{Y} = \mathbf{G} + j\mathbf{B} \in \mathbb{C}^{n \times n}.$$

It is **hard to find** in the real world.

The non-linear power flow equations govern the power injections at every node  $i \in \mathcal{N}$ :

$$p_i = v_i \sum_{\substack{k=1 \\ n}}^n v_k (G_{ik} \cos \theta_{ik} + B_{ik} \sin \theta_{ik}), \qquad (1a)$$

$$q_i = v_i \sum_{k=1} v_k (G_{ik} \sin \theta_{ik} - B_{ik} \cos \theta_{ik}).$$
(1b)

The non-linear power flow equations govern the power injections at every node  $i \in \mathcal{N}$ :

$$p_{i} = v_{i} \sum_{k=1}^{n} v_{k} (\mathbf{G}_{ik} \cos \theta_{ik} + \mathbf{B}_{ik} \sin \theta_{ik}), \qquad (2a)$$
$$q_{i} = v_{i} \sum_{k=1}^{n} v_{k} (\mathbf{G}_{ik} \sin \theta_{ik} - \mathbf{B}_{ik} \cos \theta_{ik}). \qquad (2b)$$

They are functions of the network topology.

# Problem 2: The case of the missing voltage phasor

#### Every node $i \in \mathcal{N}$ has a complex voltage phasor and power injection:

$$v_i \angle \theta_i \in \mathbb{C}, \quad p_i + jq_i \in \mathbb{C}.$$

It's hard to find the voltage phase angles in the real world.

Only  $\approx$  3000 phasor measurement units (PMUs) in North America [8]. Rare in:

- 1. Distribution systems
- 2. Transmission system boundaries
- 3. Rural transmission systems
- 4. Underserved areas

Example: PJM, circa 2022

- 1. 400 PMUs throughout PJM territory<sup>1</sup>
- 2. Only required on substations

<sup>&</sup>lt;sup>1</sup>PJM, "Synchrophasor Technology Roadmap", 2022.

#### The case of the missing voltage phasor



#### Figure 1: circa 2022 PMU deployment in PJM<sup>2</sup>

<sup>&</sup>lt;sup>2</sup>Credit PJM, "Synchrophasor Technology Roadmap", 2022. Source:

https://www.pjm.com/markets-and-operations/ops-analysis/synchrophasor-technology

# OK, so who cares?

# It's pretty hard to be a power engineer without this information!

#### How to solve the non-linear power flow equations?

Classic approach: Newton-Raphson power flow.

Iteratively solve a linear system of equations of the form

$$\begin{bmatrix} \Delta p \\ \Delta q \end{bmatrix} = \begin{bmatrix} \frac{\partial p}{\partial \theta}(x) & \frac{\partial p}{\partial v}(x) \\ \frac{\partial q}{\partial \theta}(x) & \frac{\partial q}{\partial v}(x) \end{bmatrix} \begin{bmatrix} \Delta \theta \\ \Delta v \end{bmatrix} = J(x)\Delta x.$$
(3)

- 1.  $\Delta p, \Delta q \in \mathbb{R}^n$  are small perturbations in the active and reactive power injections
- 2.  $\Delta \theta, \Delta v \in \mathbb{R}^n$  are small perturbations in the voltage phase angles and magnitudes

$$\begin{bmatrix} \Delta p \\ \Delta q \end{bmatrix} = \underbrace{\begin{bmatrix} \frac{\partial p}{\partial \theta}(x) & \frac{\partial p}{\partial v}(x) \\ \frac{\partial q}{\partial \theta}(x) & \frac{\partial q}{\partial v}(x) \end{bmatrix}}_{:=J(x)} \begin{bmatrix} \Delta \theta \\ \Delta v \end{bmatrix} = J(x)\Delta x, \tag{4}$$

- 1. The matrix  $J(x) \in \mathbb{R}^{2n \times 2n}$  is the power flow Jacobian matrix.
- 2. Derivatives of the power flow equations (2) with respect to the voltage magnitudes v and phase angles  $\theta$ .

$$\begin{bmatrix} \Delta p \\ \Delta q \end{bmatrix} = \underbrace{\begin{bmatrix} \frac{\partial p}{\partial \theta}(x) & \frac{\partial p}{\partial v}(x) \\ \frac{\partial q}{\partial \theta}(x) & \frac{\partial q}{\partial v}(x) \end{bmatrix}}_{:=J(x)} \begin{bmatrix} \Delta \theta \\ \Delta v \end{bmatrix} = J(x)\Delta x, \quad (5)$$

- 1. Can we learn this matrix as a proxy model?
- 2. Can we exploit the structure of this matrix?

#### What do we actually know in practice?

Commonly, we receive measurements of the form [9]

$$v_i \in \mathbb{R}, \quad p_i + jq_i \in \mathbb{C}.$$

- 1. v<sub>i</sub> —voltage *magnitude*
- 2.  $p_i$  —active (real) power
- 3.  $q_i$  —reactive (imaginary) power<sup>3</sup>

#### No network model, either!

<sup>&</sup>lt;sup>3</sup>e.g., from a historical or chosen power factor [9]

- 1. How can we recover voltage phasors from their magnitudes?
- 2. How can we recover a phaseless model of the voltage phasors?
- 3. How can we do this provably?



Figure 2: Voltage phasor recovery via power flow Jacobian recovery

# How can we solve it?

#### Given a matrix $\mathbf{A} \in \mathbb{C}^{m \times n}$ , we want to

# find $\mathbf{x} \in \mathbb{C}^n$ s.t. $|\mathbf{A}\mathbf{x}| = \mathbf{b}$ .

where  $\boldsymbol{b} \in \mathbb{R}^m$  are real-valued magnitude measurements.

Example applications:

- 1. X-ray crystallography [5]
- 2. Electron microscopy [4]
- 3. Optical and medical imaging [2]
- 4. Numerous tasks in experimental physics [10, 7]
- 5. ... Power systems?

For any voltage magnitudes v, we can represent the **phasor voltages**  $\overline{v} \in \mathbb{C}^n$  using a **phase vector**  $u \in \mathbb{C}^n$  in the **unit complex ball** 

$$|u_i|=1, \quad \text{atan2}(\text{Im}\{u_i\}, \text{Re}\{u_i\})=\theta_i, \tag{6}$$

where we require

$$\overline{\mathbf{v}} = \operatorname{diag}(\mathbf{v})\mathbf{u} = \begin{bmatrix} v_1 & \dots & 0\\ \vdots & \ddots & \vdots\\ 0 & \dots & v_n \end{bmatrix} \mathbf{u}.$$
 (7)

#### Apply the phase retrieval framework to recover the voltage phase

find 
$$\boldsymbol{\theta} \in (-\pi, \pi]^n$$
 st:  $|\operatorname{diag}(\boldsymbol{v})\boldsymbol{u}(\boldsymbol{\theta})| = \boldsymbol{v}.$  (8a)

# Analytical Results

**Classical power network physics:** well known structural symmetries in the power flow Jacobian matrix.

$$\frac{\partial p_{i}}{\partial \theta_{k}} = \begin{cases}
v_{k} \frac{\partial q_{i}}{\partial v_{k}} & i \neq k, \\
-v_{i} \frac{\partial q_{i}}{\partial v_{i}} - 2v_{i}^{2}B_{ii} & i = k.
\end{cases}$$

$$\frac{\partial q_{i}}{\partial \theta_{k}} = \begin{cases}
-v_{k} \frac{\partial p_{i}}{\partial v_{k}} & i \neq k, \\
v_{i} \frac{\partial p_{i}}{\partial v_{i}} - 2v_{i}^{2}G_{ii} & i = k
\end{cases}$$
(9a)
$$(9b)$$

Relates the Power Flow Jacobian Blocks! 😀

**Classical power network physics:** well known structural symmetries in the power flow Jacobian matrix [6].

$$\frac{\partial p_{i}}{\partial \theta_{k}} = \begin{cases}
v_{k} \frac{\partial q_{i}}{\partial v_{k}} & i \neq k, \\
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\end{cases}$$

$$\frac{\partial q_{i}}{\partial \theta_{k}} = \begin{cases}
-v_{k} \frac{\partial p_{i}}{\partial v_{k}} & i \neq k, \\
v_{i} \frac{\partial p_{i}}{\partial v_{i}} - 2v_{i}^{2}G_{ii} & i = k
\end{cases}$$
(10a)
(10b)

Need the network model and parameters 😕

Solution 1: A phaseless representation of the Power Flow Jacobian matrix.

#### Lemma 1: Phaseless power flow Jacobian structure

We can write the partial derivatives of power injections with respect to phase angles without the phase angles and without the grid model.

$$\frac{\partial p_{i}}{\partial \theta_{k}} = \begin{cases} v_{k} \frac{\partial q_{i}}{\partial v_{k}} & i \neq k \\ v_{i} \frac{\partial q_{i}}{\partial v_{i}} - 2q_{i} & i = k, \end{cases}$$
(11a)  
$$\frac{\partial q_{i}}{\partial \theta_{k}} = \begin{cases} -v_{k} \frac{\partial p_{i}}{\partial v_{k}} & i \neq k \\ -v_{i} \frac{\partial p_{i}}{\partial v_{i}} + 2p_{i} & i = k. \end{cases}$$
(11b)

These are functions of the **power injection measurements!** 

The power-voltage phase angle sensitivity matrices can be expressed as functions  $\frac{\partial p}{\partial \theta}, \frac{\partial q}{\partial \theta} : \mathbb{R}^n \times \mathbb{R}^n \mapsto \mathbb{R}^{n \times n}$  of the form

$$\frac{\partial p}{\partial \theta}(\mathbf{v}, q) = \operatorname{diag}(\mathbf{v}) \frac{\partial q}{\partial \mathbf{v}} - 2 \operatorname{diag}(q),$$
(12a)  
$$\frac{\partial q}{\partial \theta}(\mathbf{v}, p) = -\operatorname{diag}(\mathbf{v}) \frac{\partial p}{\partial \mathbf{v}} + 2 \operatorname{diag}(p),$$
(12b)

which are implicitly parameterized by  $\frac{\partial q}{\partial v}$  and  $\frac{\partial p}{\partial v}$ .

#### New characterization of the structure of the Power Flow Jacobian



**Figure 3:** Equivalence of the Newton-Raphson iterations using the  $\theta$ -free expressions and standard expressions for  $\frac{\partial p}{\partial \theta}$ ,  $\frac{\partial q}{\partial \theta}$  (right) for a simple two bus test case (left).

Solution 2: Use the power flow Jacobian structure for voltage phase retrieval Apply a classic phase retrieval algorithm [11] to solve the power flow equations using learned voltage magnitude blocks

minimize 
$$\left\| \begin{bmatrix} \Delta \boldsymbol{p}_t \\ \Delta \boldsymbol{q}_t \end{bmatrix} - \begin{bmatrix} (?) & \frac{\partial \boldsymbol{p}}{\partial \boldsymbol{v}} \\ (?) & \frac{\partial \boldsymbol{q}}{\partial \boldsymbol{v}} \end{bmatrix} \begin{bmatrix} \Delta \boldsymbol{\theta}_t \\ \Delta \boldsymbol{v}_t \end{bmatrix} \right\|_2^2$$

Learning the Jacobian blocks is well studied [3].

### Problem! we do not know $\frac{\partial q}{\partial \theta}, \frac{\partial p}{\partial \theta}$

We have to add them as decision variables.

The phase retrieval program for samples  $t = 1, \ldots$ , can then be written as

$$\begin{array}{c} \underset{\Delta \theta_{t}, \frac{\partial p}{\partial \theta}, \frac{\partial q}{\partial \theta}}{\min i ze} & \left\| \begin{bmatrix} \Delta p_{t} \\ \Delta q_{t} \end{bmatrix} - \begin{bmatrix} \frac{\partial p}{\partial \theta} (\mathbf{v}_{t}, q_{t}) & \frac{\partial p}{\partial \mathbf{v}} \\ \frac{\partial q}{\partial \theta} (\mathbf{v}_{t}, p_{t}) & \frac{\partial q}{\partial \mathbf{v}} \end{bmatrix} \begin{bmatrix} \Delta \theta_{t} \\ \Delta \mathbf{v}_{t} \end{bmatrix} \right\|_{2}^{2} \\ \text{subject to: power flow Jacobian structure!} \end{array}$$

Solution 3: Use the power flow Jacobian structure to guarantee voltage phase retrieval

#### Long story short:

- 1. Use the Jacobian structure to show when the voltage phase can be uniquely recovered.
- 2. Do some linear algebra
- 3. ???
- 4. Profit!

Take a look at the paper for more details

#### Theorem (Phase retrieval from active power injections)

For a set of buses in a network  $\mathcal{B} \subset \{1, ..., n\}$ , if for every bus  $i \in \mathcal{B}$ , the reactive power differential inequality

$$|q_{i}| > \frac{1}{2} v_{i} \left( \sum_{k \in \mathcal{B} \setminus \{i\}} \left| \frac{\partial q_{k}}{\partial v_{i}} \right| - \left| \frac{\partial q_{i}}{\partial v_{i}} \right| \right)$$
(13a)  
or  $|q_{i}| > \frac{1}{2} \left( \sum_{k \in \mathcal{B} \setminus \{i\}} v_{k} \left| \frac{\partial q_{i}}{\partial v_{k}} \right| - v_{i} \left| \frac{\partial q_{i}}{\partial v_{i}} \right| \right),$  (13b)

holds, then the voltage phase angles can be uniquely recovered from solely the active power (real) injections **p**.

**Theorem (Phase retrieval from reactive power injections)** Analogously, if for every bus  $i \in B$ , the active power differential inequality

$$|p_{i}| > \frac{1}{2} v_{i} \left( \sum_{k \in \mathcal{B} \setminus \{i\}} \left| \frac{\partial p_{k}}{\partial v_{i}} \right| - \left| \frac{\partial p_{i}}{\partial v_{i}} \right| \right)$$
(14a)  
or 
$$|p_{i}| > \frac{1}{2} \left( \sum_{k \in \mathcal{B} \setminus \{i\}} v_{k} \left| \frac{\partial p_{i}}{\partial v_{k}} \right| - v_{i} \left| \frac{\partial p_{i}}{\partial v_{i}} \right| \right),$$
(14b)

holds, then the voltage phase angles can be uniquely recovered from solely the reactive power (imaginary) injections **q**.

Side note: can also guarantee Jacobian invertibility Jacobian invertibility  $\iff$  No voltage collapse  $\implies$  Phase retrieval Take a look at the paper for more details

# **Computational Results**

- 1. RTS-GMLC network model [1]
- 2. Open-source
- 3. Real-world data

How does this compare to classical model-based state estimation?

#### Comparison with classical state estimation



**Figure 4:** Impact of (*RTS\_GMLC*) model uncertainty on recovered phase angle relative error vs. measurement noise level. Shaded regions indicate  $\pm 1$  standard deviation of the relative errors computed over 20 bootstraps.

# Voltage phasor recovery performance

#### Voltage phasor recovery performance



Figure 5: Voltage phasor recovery by measurement noise level.

How about the matrices?

#### Power flow Jacobian recovery



**Figure 6:** Recovery of the power-phase angle submatrices  $\frac{\partial p}{\partial \theta}$ ,  $\frac{\partial q}{\partial \theta}$  of the power flow Jacobian for the *RTS\_GMLC* network via the phase retrieval program.

# How about real-time performance?

#### Real-time voltage phasor recovery performance



Figure 7: Ground truth (blue) and estimated (orange dashed) voltage phase angles at 15 min. granularity, juxtaposed with ground truth 5 min. granularity voltage phase angles (black dots).

# Outlook

- 1. We can recover voltage phasors from their magnitudes
- 2. We can recover models of voltage phase angles from their magnitudes
- 3. This can save a lot of money<sup>4</sup>

<sup>&</sup>lt;sup>4</sup>A PMU installation costs \$40,000-\$180,000 each.

<sup>&</sup>quot;Factors affecting PMU installation costs", US Department of Energy, 2014.

- 1. Bottle necked by measurement frequency (PMUs, milliseconds, other devices, minutes-hours)
- 2. Bottle necked by measurement type (Reactive power assumptions needed)

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# Thanks for listening

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Backup slides

#### Where are the eigenvalues of a square matrix?

For any matrix  $A \in \mathbb{C}^{n \times n}$ , by the *Gershgorin Circle Theorem* the eigenvalues of A are guaranteed to lie in the union of the i = 1, ..., n Gershgorin discs  $\mathcal{G}_i(A)$  of the matrix, i.e.,

$$\lambda_i(\mathbf{A}) \in \bigcup_{i=1}^n \mathcal{G}_i(\mathbf{A}), \quad i = 1, \dots, n,$$
 (15)

where

$$\mathcal{G}_{i}(\mathbf{A}) \triangleq \{ w \in \mathbb{C} : |w - A_{ii}| \leq \sum_{k: k \neq i} |A_{ik}| \} \subseteq \mathbb{C}.$$
 (16)

Given: |Ax| = b,  $b \in \mathbb{R}^m$ , what is  $x \in \mathbb{C}^n$ ?

Classical phase retrieval problem:

$$\min_{\boldsymbol{x}\in\mathbb{C}^{n},\boldsymbol{y}\in\mathbb{C}^{m}} ||\boldsymbol{A}\boldsymbol{x}-\boldsymbol{y}||_{2}^{2} \quad \text{s.t.} \quad |\boldsymbol{y}| = \boldsymbol{b},$$
(17a)  
$$\iff \min_{\boldsymbol{x}\in\mathbb{C}^{n},\boldsymbol{u}\in\mathbb{C}^{m}} ||\boldsymbol{A}\boldsymbol{x}-\operatorname{diag}(\boldsymbol{b})\boldsymbol{u}|| \quad \text{s.t.} \quad |\boldsymbol{u}| = \mathbb{1},$$
(17b)  
$$\iff \min_{\boldsymbol{u}\in\mathbb{C}^{m}:|\boldsymbol{u}_{i}|=1 \; \forall i\in[1,m]} \boldsymbol{u}^{*}\boldsymbol{M}\boldsymbol{u}, \quad \text{s.t.} \quad \boldsymbol{M} = \operatorname{diag}(\boldsymbol{b}-\boldsymbol{A}\boldsymbol{A}^{\dagger}) \succ 0.$$
(17c)

for any candidate phase u', note that

$$\hat{x} = A^{\dagger} \operatorname{diag}(b) u' = (A^* A)^{-1} A^* \operatorname{diag}(b) u',$$
 (18)

Case	# PQ Buses	% Satisfying Thm. 1	r <sub>worst</sub>
14	9	100.0%	—
24_ieee_rts	13	100.0%	—
ieee30	24	95.83%	$1.4 \times 10^{-14}$
RTS_GMLC	40	100.0%	—
118	64	100.0%	—
89pegase	77	94.81%	8.32
ACTIVSg200	162	96.91%	0.088
ACTIVSg500	444	94.37%	3.014
ACTIVSg2000	1608	84.83%	28.31

Table 1: Analysis of Theorem 1 for PQ buses of various test cases

Case	# PQ Buses	% Satisfying Thm. ??	$\sigma_{ m max}$
14	9	100.0%	0.876
24_ieee_rts	13	100.0%	0.401
ieee30	24	95.83%	1.437
RTS_GMLC	40	100.0%	0.444
118	64	100.0%	0.473
89pegase	77	100%	0.954
ACTIVSg200	162	100%	0.698
ACTIVSg500	444	99.77%	1.090
ACTIVSg2000	1608	99.69%	1.180

Table 2: Analysis of Theorem ?? for PQ buses of various test cases

Quantity	Value
$\left\  \frac{\partial p}{\partial \theta} - \frac{\partial p}{\partial \theta} (\mathbf{V}, \mathbf{q}) \right\ _{F} \left\  \frac{\partial p}{\partial \theta} \right\ _{F}$	$2.510 \times 10^{-8}$
$\left\  \frac{\partial q}{\partial \theta} - \frac{\partial q}{\partial \theta} (\mathbf{v}, \mathbf{p}) \right\ _{F} / \left\  \frac{\partial q}{\partial \theta} \right\ _{F}$	$1.725 \times 10^{-7}$

**Table 3:** Verification that the structure expressions of the Lemma hold formultiphase unbalanced networks.