

Enhancing Observability in Power Distribution Grids

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Motivation

- Reduced observability due to limited metering infrastructure
- Leverage smart meter data [1]
- Smart PV inverters can be commanded within micro-sec.
- System state needed for grid dispatch tasks



Problem statement

- Define the following set of buses
 - Set \mathcal{M} (metered buses): $(p_n, q_n, |V_n|^2)$ specified
 - Set \mathcal{O} (non-metered buses): no information is available
 - Set \mathcal{C} of conv. PQ buses: (p_n, q_n) specified
 - Substation bus $S = \{0\}$

$$\mathcal{N} := \{1, \dots, N\} = \mathcal{C} \cup \mathcal{M} \cup \mathcal{O}$$

Given smart meter data find the system state ${f v}$, i.e. the vector of voltages and infer the unobserved loads in ${\cal O}$



Coupling flows via grid probing

- Grid transitions from state ${f v} o {f v}'$
- Grid sensing on \mathcal{M} over two consecutive time instances



Criterion for solvability

If there exists a set of vertex-disjoint paths from $\mathcal{O} \to \mathcal{M}$ in $\mathcal{T} = (\mathcal{N}, \mathcal{L})$ then the CPF equations are locally invertible in general.



Solving PF as an SDP

PF as a feasibility problem

find
$$\mathbf{v} \in \mathbb{C}^n$$

s.t. $\mathbf{v}^H \mathbf{M}_k \mathbf{v} = \hat{s}_k, \quad k = 1, \dots, m$

PF in standard form [2]

$$\min_{\mathbf{V}} \quad \operatorname{tr}(\mathbf{MV}) \\ \text{s.t.} \quad \operatorname{tr}(\mathbf{M}_{k}\mathbf{V}) = \hat{s}_{k} \quad k = 1, \dots, m \\ \mathbf{V} \succeq 0 \\ \operatorname{rank}(\mathbf{V}) = 1 \quad \mathbf{V} = \mathbf{v}\mathbf{v}^{H}$$

Coupled PSSE

 Measurements are corrupted with noise and/or unknown loads may change over time.

$$\hat{s}_k = s(\mathbf{v}) + \epsilon_k$$

 Rather than solving exactly the CPF equations, minimize a leastsquares cost

$$\sum_{k} (\hat{s}_{k} - \mathbf{v}^{H} \mathbf{M}_{k} \mathbf{v})^{2} + \sum_{k} (\hat{s}_{k} - \mathbf{v}^{\prime H} \mathbf{M}_{k} \mathbf{v}^{\prime})^{2} + \sum_{k} (\mathbf{v}^{H} \mathbf{M}_{k} \mathbf{v} - \mathbf{v}^{\prime H} \mathbf{M}_{k} \mathbf{v}^{\prime})^{2}$$

$$\underbrace{\qquad}$$
Time instance t Time instance t' Coupling

SDP-based PSSE

$$\min_{\mathbf{V},\mathbf{V}',\boldsymbol{\epsilon}} \sum_{\ell=1}^{2L+2O} f_{\ell}(\epsilon_{\ell}) + \alpha [\operatorname{tr}(\mathbf{M}\mathbf{V}) + \operatorname{tr}(\mathbf{M}\mathbf{V}')]$$

s.t. $\operatorname{tr}(\mathbf{M}_{\ell}\mathbf{V}) + \epsilon_{\ell} = \hat{s}_{\ell}, \quad \ell = 1, \dots, L$
 $\operatorname{tr}(\mathbf{M}_{\ell}\mathbf{V}') + \epsilon_{\ell} = \hat{s}_{\ell}', \quad \ell = L+1, \dots, 2L$
 $\operatorname{tr}(\mathbf{M}_{\ell}(\mathbf{V}'-\mathbf{V})) = \epsilon_{\ell}, \quad \ell = 2L+1, \dots, 2L+2O$
 $\mathbf{V} \succeq \mathbf{0}, \mathbf{V}' \succeq \mathbf{0}.$

Least-Squares (LS) [3],[4],[5],[6]

2L+2O

 $\ell = 1$

 $f(\boldsymbol{\epsilon}) = \sum \epsilon_{\ell}^2$

Least-Absolute Value (LAV)

$$f(\boldsymbol{\epsilon}) = \sum_{\ell=1}^{2L+2O} |\epsilon_{\ell}|$$

Strengthen SDP relaxation

$$\operatorname{tr}(\mathbf{M}_{pn}\mathbf{V}) < 0 \qquad \qquad \forall n \in \mathcal{O}$$

$$\operatorname{tr}(\mathbf{M}_{pn}\mathbf{V}) < 0 \qquad \qquad \forall n \in \mathcal{O}$$



• β captures the ℓ_2 -norm between the two states $(\mathbf{v}, \mathbf{v}')$ [6]

• As $||\mathbf{v} - \mathbf{v}'||_2$ increases, grid probing changes system state

Estimation error 0.12 -WLS 0.1 0.08 BMSE 0.06 0.04 0.02 0∟ 5 10 15 20 25 30 35 40 SNR

• Averaged over **20** simulations for each step **[6]**

• RMSE:
$$\sqrt{\sum_{t=1}^{N} \frac{(\hat{v}_t - v_{t,true})^2 + (\hat{v}'_t - v'_{t,true})^2}{2N}}$$

Numerical tests with real data



- Simulation over one day with real data [1],[6]
- Grid probing to change system state
- RMSE is very low

Conclusions

- Used smart inverters for grid probing
- Novel technique using time coupling
- Fewer measurements needed to estimate the system state
- Characterized local invertibility via an intuitive **criterion**
- Developed an SDP-based solver for coupled power flow (CPF) and coupled power system state estimation (CPSSE).

References

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- [3] H. Zhu and G. B. Giannakis, "Power System Nonlinear State Estimation Using Distributed Semidefinite Programming," *IEEE Journal of Selected Topics in Signal Processing*, vol. 8, no. 6, pp. 1039-1050, Dec. 2014.
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- [6] S. Bhela, V. Kekatos, and S. Veeramachaneni, "Enhancing Observability in Distribution Grids using Smart Meter Data," *IEEE Trans. on Smart Grid*, Feb. 2016 (accepted).