

Static State Estimation in Power Systems

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Energy Management System (EMS)



Measurements

- Obtained from supervisory control and data acquisition (SCADA)
- New data every 4 seconds
- O Static state estimator
 - Uses a model of the system that includes topology and parameter information
 - Estimates real-time values for the whole system
 - Signals to operators any lines close to overload or already overloaded
 - MISO runs this every 5 minutes

Static State Estimation



- Process to assign values to unknown system state variables based on imperfect measurements obtained from the system
- State variables:
 - Voltage magnitudes
 - Q Relative phase angles at buses
- Measurements:
 - Voltage magnitudes
 - 2 Real and reactive transmission line flows
 - Ourrent through transmission lines
- Objective: produce the "best estimate" of the system voltage and phase angles, recognizing that there are errors in the measured quantities and that there may be redundant measurements.
- Basic underlying assumption: power system is in quasi-steady-state condition

Static State Estimation Criteria



• Common criteria for "best" estimate

- The maximum likelihood criterion: maximize probability that estimate of the state variable is its true value
- O The weighted least-squares criterion: minimize sum of squares of the weighted deviations of estimated measurements from actual measurements
- The minimum variance criterion: minimize expected value of sum of squares of the the deviations of estimated components of the state vector from the corresponding true values
- Assume meter error distributions are normally distributed and unbiased
- Then the three criteria above lead to identical estimators

Maximum Likelihood Least-Squares Estimation



- Let x denote vector of state variables, i.e., voltage magnitudes and relative phase angles
- Let η_i represent uncertainty in each measured value z_i^{meas} . Then,

$$z_i^{meas} = z_i^{true} + \eta_i = h_i(x) + \eta_i$$

- Model $\eta_i \sim \mathcal{N}(0, \sigma_i)$
- Assumptions:
 - **Q** Zero-mean: $\mathbb{E}[\eta_i] = 0$, for all i
 - Independent measurements:

$$\mathbb{E}[\eta_i \eta_j] = \begin{cases} \sigma_j^2 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

Maximum Likelihood Least-Squares Estimation



• The p.d.f. of z_i^{meas} is

$$f(z_i^{meas}) = \frac{1}{\sigma_i \sqrt{2\pi}} \exp\left\{\frac{-\left[z_i^{meas} - h_i(x)\right]^2}{2\sigma_i^2}\right\}$$

- Find estimate \hat{x} that maximizes the probability that the observed measurement z_i^{meas} would occur
 - \implies Find \hat{x} that maximizes p.d.f. $f(z_i^{meas})$

$$\implies$$
 Find \hat{x} that minimizes $\frac{\left[z_i^{meas} - h_i(x)\right]^2}{2\sigma_i^2}$

- Suppose there are *m* measurements
- Leads to the following objective:

$$\min J(x) = \sum_{i=1}^{m} \frac{[z_i^{meas} - h_i(x)]^2}{2\sigma_i^2}$$

Matrix Form



Let

$$z = \begin{bmatrix} z_1^{meas} \\ \vdots \\ z_m^{meas} \end{bmatrix}, \quad h(x) = \begin{bmatrix} h_1(x) \\ \vdots \\ h_m(x) \end{bmatrix}, \quad R = \begin{bmatrix} \sigma_1^2 & & \\ & \ddots & \\ & & \sigma_m^2 \end{bmatrix}$$

• Weighted least-squares (WLS) estimation problem with R^{-1} as weighting matrix

$$\min J(x) = \frac{1}{2} [z - h(x)]^T R^{-1} [z - h(x)] = J(\hat{x})$$

• Necessary conditions of optimality are

$$\nabla_x J \Big|_{x=\hat{x}} = - [z - h(\hat{x})]^T R^{-1} \underbrace{[\nabla_x h]_{x=\hat{x}}}_{H(\hat{x})} = 0$$

• So \hat{x} is the root of the set of nonlinear equations

$$H^{T}(x)R^{-1}[z-h(x)] = 0$$

Weighted Least-Squares Estimation



$$H^{T}(x)R^{-1}[z - h(x)] = 0$$

- Can solve iteratively. Or...
- If each h_i is a linear function^{*} of x, i.e., h(x) = Hx, then

$$H^T R^{-1} z - H^T R^{-1} H x = 0$$

Closed-form solution exists:

$$\hat{x} = \left[H^T R^{-1} H \right]^{-1} H^T R^{-1} z$$

- * DC power flow model assumptions
 - For each of the transmission lines, $x_i >> r_i$.
 - ► All bus voltages are approximately 1 p.u.
 - All bus angles are nearly 0. Therefore, $\cos(\theta_i \theta_j) \cong 1$ and $\sin(\theta_i \theta_j) \cong \theta_i \theta_j$.

Example—Two-Bus System





4 measurements

$$z = \begin{bmatrix} v_1 \\ v_2 \\ p_{12} \\ q_{12} \end{bmatrix} = \begin{bmatrix} 1.02 \\ 1.0 \\ 2.0 \\ 0.2 \end{bmatrix}, \quad R = \begin{bmatrix} 0.05^2 & 0 & 0 & 0 \\ 0 & 0.05^2 & 0 & 0 \\ 0 & 0 & 0.1^2 & 0 \\ 0 & 0 & 0 & 0.1^2 \end{bmatrix}$$

3 states

$$x = \begin{bmatrix} V_1 \\ V_2 \\ \theta_2 \end{bmatrix}$$

Example—Two-Bus System





Measured values

$$z = h(x) = \begin{bmatrix} v_1 \\ v_2 \\ p_{12} \\ q_{12} \end{bmatrix} = \begin{bmatrix} V_1 \\ V_2 \\ -10V_1V_2\sin\theta_2 \\ 10V_1^2 - 10V_1V_2\cos\theta_2 \end{bmatrix} + \begin{bmatrix} \eta_1 \\ \eta_2 \\ \eta_3 \\ \eta_4 \end{bmatrix}$$

Jacobian

$$H(x) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -10V_2 \sin \theta_2 & -10V_1 \sin \theta_2 & -10V_1 V_2 \cos \theta_2 \\ 20V_1 - 10V_2 \cos \theta_2 & -10V_1 \cos \theta_2 & 10V_1 V_2 \sin \theta_2 \end{bmatrix}$$

• Iterative solution. At the k^{th} iteration, solve for Δx_k in

$$[H(x_{k})]^{T} R^{-1} H(x_{k}) \Delta x_{k} = [H(x_{k})]^{T} R^{-1} [z - h(x_{k})]$$

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