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State Estimation

Linear State Estimation

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State Estimation

Real-time tool for estimating the voltages, phase angles of a power system

$$\vec{V}_{Bus} = \begin{bmatrix} V_1 \angle \delta_1 \\ \vdots \\ V_N \angle \delta_N \end{bmatrix}$$

$$x = \begin{bmatrix} \delta \\ V \end{bmatrix} \text{ unknown voltages and angles}$$

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State Estimation

- Reality → measure voltages, currents, P_{ij} , Q_{ij} (SCADA system)
 - Supervisory Control and Data Acquisition System
 - Remote Terminal Units
 - Communication Network
 - Phase angles cannot be measured. Lack of common phase angle reference.

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Substations

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Measurements

- Model errors
- RTU errors

Redundant measurements help get best fit

- Example

Ammeter "twice" as accurate as voltmeter

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Measurements

$$x_1 = I = \text{state, unknown}$$

$$z_1 = I = x_1 + e_1$$

e_1 = measurement error for z_1 = Random variable Gaussian (Normal)

$$z_2 = 2I + e_2 = 2x_1 + e_2$$

e_2 = measurement error for z_2

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Formulation

$$\begin{aligned}\underline{z} &= \underline{H} \underline{x} + \underline{e} \\ \underline{x} &= \text{States (known)} \\ \underline{z} &= \text{Measurements} \\ \underline{H} &= \text{Model representation} \\ \underline{e} &= \text{Measurement errors} \\ \text{Estimate: } \hat{x} &=? \\ \underline{z} &\Rightarrow \text{Estimation} \Rightarrow \hat{x} = ?\end{aligned}$$

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Least Square formulation

$$\begin{aligned}\text{Minimize } x \\ \sum_{i=1}^N \omega_i e_i^2 &\Leftarrow \text{weighted total error} \\ &= \underline{e}^T \underline{W} \underline{e} \\ \underline{e} &= \underline{z} - \underline{H} \underline{x}\end{aligned}$$

$$\begin{aligned}\text{Minimize w.r.to } x \\ (\underline{z} - \underline{H} \underline{x})^T \underline{W} (\underline{z} - \underline{H} \underline{x}) &= g(x)\end{aligned}$$

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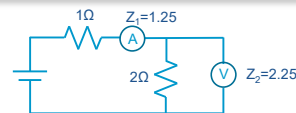
Least Square Estimate

$$\begin{aligned}\frac{\partial g}{\partial \underline{x}} &= -2 \underline{H}^T \underline{W} (\underline{z} - \underline{H} \underline{x}) = 0 \\ \underline{H}^T \underline{W} \underline{z} &= \underline{H}^T \underline{W} \underline{H} \underline{x} \\ \hat{x} &= \underline{G}^{-1} \underline{H}^T \underline{W} \underline{z} \Leftarrow \text{Least Square Estimate} \\ \underline{G} &= \underline{H}^T \underline{W} \underline{H} = \text{Gain Matrix}\end{aligned}$$

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Example

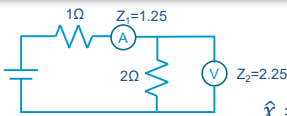


$$\begin{aligned}z_1 &= x_1 + e_1 \\ z_2 &= 2x_1 + e_2 \Rightarrow \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} x_1 + \begin{bmatrix} e_1 \\ e_2 \end{bmatrix} \\ H &= \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad W = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \\ \underline{G} &= \underline{H}^T \underline{W} \underline{H} = [1 \quad 2] \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 6\end{aligned}$$

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Estimate

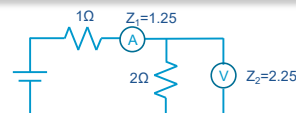


$$\begin{aligned}\hat{x} &= \underline{G}^{-1} \underline{H}^T \underline{W} \underline{z} \\ &= \frac{1}{6} [1 \quad 2] \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1.25 \\ 2.25 \end{bmatrix} \\ &= \frac{1}{6} [2 \quad 2] \begin{bmatrix} 1.25 \\ 2.25 \end{bmatrix} \\ &= \frac{7}{6} = 1.167\end{aligned}$$

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Second case



$$\hat{x} = 1.167$$

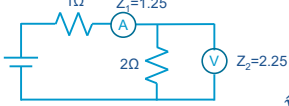
Suppose z_1 10 times as accurate as z_2

$$\begin{aligned}W &= \begin{bmatrix} 10 & 0 \\ 0 & 1 \end{bmatrix} \\ \Rightarrow \underline{G} &= \underline{H}^T \underline{W} \underline{H} = [1 \quad 2] \begin{bmatrix} 10 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 14\end{aligned}$$

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New Estimate



$$\hat{x} = \underline{G}^{-1} \underline{H}^T \underline{W} \underline{z}$$

$$= \frac{1}{14} \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} 10 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1.25 \\ 2.25 \end{bmatrix}$$

$$= \frac{1}{14} \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} 12.5 \\ 2.25 \end{bmatrix}$$

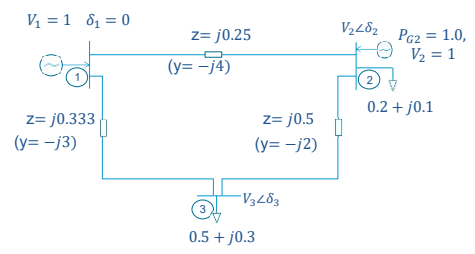
$$= \frac{17}{14} = 1.214$$

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Power System State Estimation

Power System:



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DC Power-flow solution

$$\begin{bmatrix} \delta_2 \\ \delta_3 \end{bmatrix} = \begin{bmatrix} 6 & -2 \\ -2 & 5 \end{bmatrix}^{-1} \cdot \begin{bmatrix} 0.8 \\ -0.5 \end{bmatrix}$$

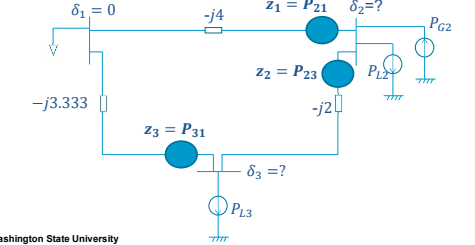
$$= \begin{bmatrix} 0.1154 \\ -0.0538 \end{bmatrix}$$

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DC powerflow:



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DC Power-flow formulation

$$\delta_2 = ? \quad \delta_3 = ?$$

$$\underline{x} = \begin{bmatrix} \delta_2 \\ \delta_3 \end{bmatrix}$$

$$z_1 = P_{21} = 4\delta_2 + e_1$$

$$z_2 = P_{23} = 2\delta_2 - 2\delta_3 + e_2$$

$$z_3 = P_{31} = 3\delta_3 + e_3$$

$$\underline{z} = \begin{bmatrix} 4 & 0 \\ 2 & -2 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} \delta_2 \\ \delta_3 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix}$$

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DC Power-flow formulation

$$\underline{z} = \begin{bmatrix} 4 & 0 \\ 2 & -2 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} \delta_2 \\ \delta_3 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix}$$

$$\begin{bmatrix} \delta_2 \\ \delta_3 \end{bmatrix} = \begin{bmatrix} 0.1154 \\ -0.0538 \end{bmatrix}$$

$$\underline{z} = \begin{bmatrix} 0.4616 \\ 0.3384 \\ -0.1614 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix}, \quad \underline{z} = \begin{bmatrix} 0.45 \\ 0.35 \\ -0.12 \end{bmatrix}$$

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Measurements

$$\begin{aligned}
 P_{21} &= 0.45 \\
 P_{23} &= 0.35 \\
 P_{31} &= -0.12 \\
 \underline{W} &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\
 \underline{z} &= \begin{bmatrix} 0.45 \\ 0.35 \\ -0.12 \end{bmatrix}
 \end{aligned}$$

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State Estimate

$$\begin{aligned}
 \underline{G} &= \underline{H}^T \underline{W} \underline{H} = \begin{bmatrix} 4 & 2 & 0 \\ 0 & -2 & 3 \end{bmatrix} \underline{W} \begin{bmatrix} 4 & 0 \\ 2 & -2 \\ 0 & 3 \end{bmatrix} \\
 &= \begin{bmatrix} 20 & -4 \\ -4 & 13 \end{bmatrix} \\
 \hat{x} &= \underline{G}^{-1} \underline{H}^T \underline{W} \underline{z} = \begin{bmatrix} 0.1158 \\ -0.0459 \end{bmatrix} = \begin{bmatrix} \hat{\delta}_2 \\ \hat{\delta}_3 \end{bmatrix}
 \end{aligned}$$

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State Estimation

- LSE Basic Theory Introduced.
- Bad Data Detection – Chi Square tests
- Nonlinear Models in State Estimation
- Topology or Model errors

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