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# **EE 521: Analysis of Power Systems**

## Lecture 6 State Estimation Concepts

Fall 2009

Mondays & Wednesdays 5:45-7:00

August 24 – December 18

**Test 216** 

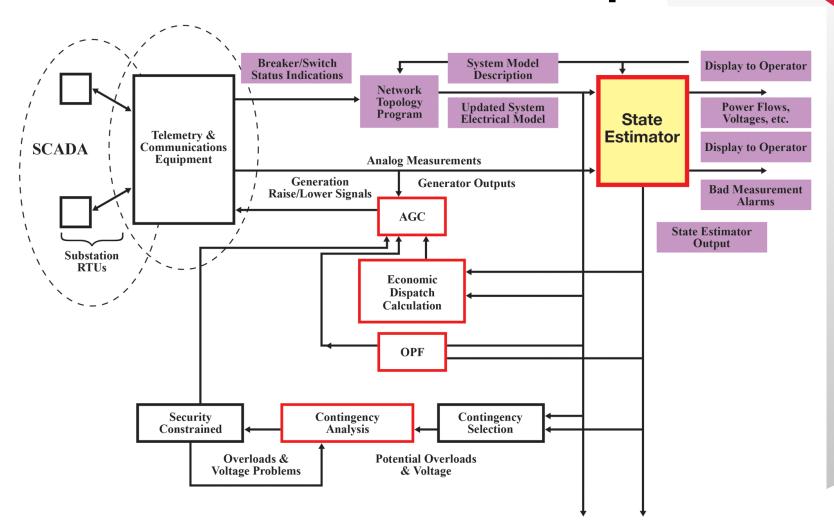


### **Topics**

- Overview of Real-Time Power Grid Operation
- Why We Need State Estimation
- Formulation of State Estimation
  - Weighted Least Square
- Solution Methods for State Estimation
  - Newton-Raphson



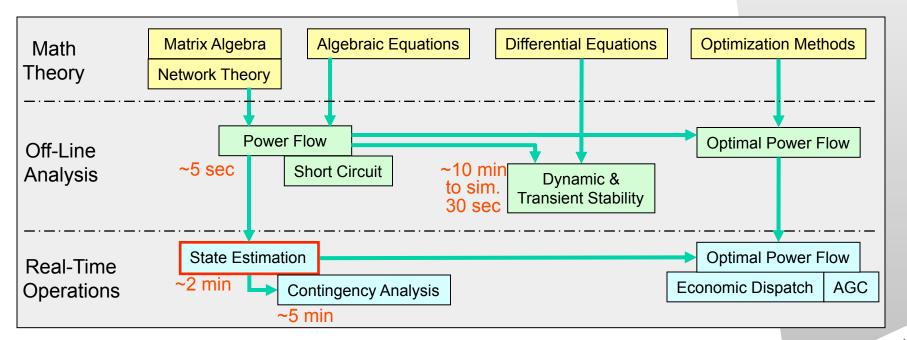
### Overview of Power Grid Operation





# Mathematical Basis for Power Grid Operation

- Based on steady-state modeling (power flow model)
- Formulated using algebraic equations in matrix form





## SCADA Systems (Supervisory Control And Data Acquisition)

#### Many uses:

- Manufacturing, production, and fabrication processes
- City water systems, oil and gas pipelines, electrical power grids, and large communication systems.
- Facilities such as buildings, airports, and space stations.

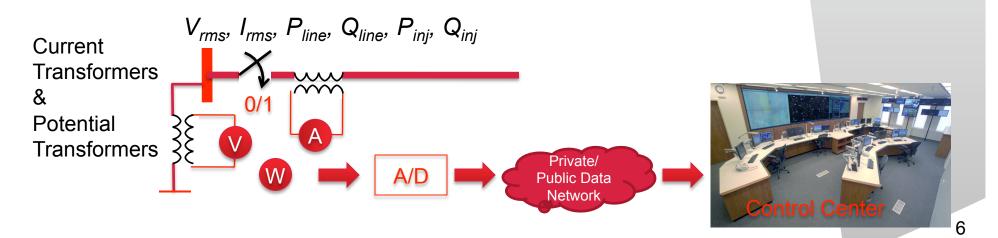
#### • Components:

- A human-Machine Interface
- A computer system
- Remote Terminal Units (RTUs) connecting to sensors
- Communication infrastructure



### **Power Grid SCADA Systems**

- Quantities Measured:
  - Status,  $V_{rms}$ ,  $I_{rms}$ ,  $P_{line}$ ,  $Q_{line}$ ,  $P_{inj}$ ,  $Q_{inj}$
- Issues with Measurements
  - Measurement Redundancy
  - Measurement Accuracy
  - Measurement Reliability





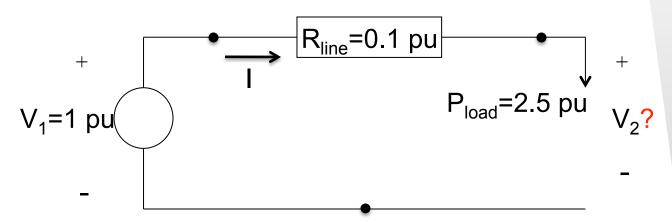
#### **Definition of State Estimation**

- Power System States:
  - V,  $\theta$  at buses, same as those in the power flow problem
- State Estimation:
  - Estimates states from measured quantities:
    - Status,  $V_{rms}$ ,  $I_{rms}$ ,  $P_{line}$ ,  $Q_{line}$ ,  $P_{inj}$ ,  $Q_{inj}$
  - Fits measurements to a model by minimizing errors
  - Objective:
    - Filter noise
    - Identify bad data and missing data
    - Estimate unmeasured quantities such as  $\theta$



#### **Power Flow Problem**

Given:  $V_1 = 1 \text{ pu}, P_{load} = 2.5 \text{ pu}. \text{ Find } V_2.$ 



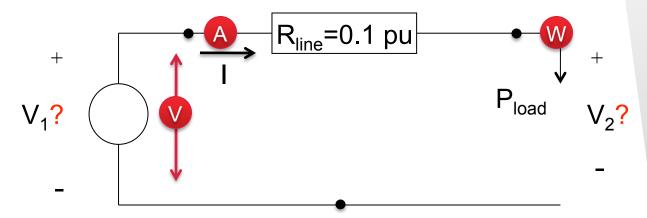
$$\begin{cases} I = \frac{V_1 - V_2}{R_{line}} \\ P_{load} = V_2 I \end{cases} \qquad P_{load} = V_2 I = V_2 \frac{V_1 - V_2}{R_{line}} \qquad V_2^2 - V_1 V_2 + P_{load} R_{line} = 0$$

$$V_2^2 - V_2 + 0.25 = 0$$
  $V_2 = 0.5$  (feasible) or  $-0.5pu$  (non-feasible)  $I = 5pu$ 



#### **State Estimation Problem**

Given Measurements:  $V_m = 0.9$  pu,  $P_m = 2.6$  pu,  $I_m = 4.5$  pu. Find  $V_1$ ,  $V_2$ .



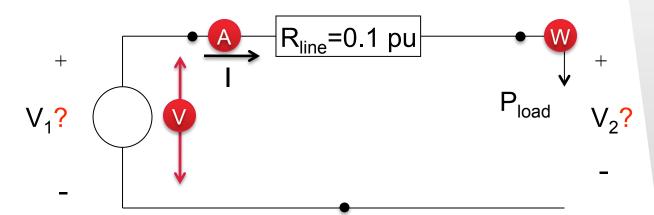
#### **Observations:**

- 1. Direct use of measurements results conflicting answers.
- 2. More measurements than necessary to solve the equations.
- 3. Measurements contain errors.
- 4. Physical laws have to be satisfied.



### **Measurement Equations**

Given Measurements:  $V_m = 0.9$  pu,  $P_m = 2.6$  pu,  $I_m = 4.5$  pu. Find  $V_1$ ,  $V_2$ .



Define measurement variables and state variables:

$$z = \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} V_m \\ P_m \\ I_m \end{bmatrix}, \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}$$

Include error terms and express using state variables:

$$z = z_{true} + e = \begin{bmatrix} V_1 \\ V_2 \frac{V_1 - V_2}{R_{line}} \\ \frac{V_1 - V_2}{R_{line}} \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix} = h(x) + e$$



## Weighted Least Square Formulation

Error equations: Estimated errors:

$$e = z - h(x)$$
 
$$\hat{e} = z - h(\hat{x})$$

Formulate an optimization problem using weighted least square methods: Weights are used to indicate different levels of measurement accuracy

$$\min_{x_1, x_2} f(x_1, x_2) = \min_{V_1, V_2} f(V_1, V_2) = w_1 e_1^2 + w_2 e_2^2 + w_3 e_3^2$$

The problem becomes solving the following two conditions:

$$\frac{\partial f(x_1, x_2)}{\partial x_1} = 0, \qquad \frac{\partial f(x_1, x_2)}{\partial x_2} = 0$$



#### **Solution Process**

#### Expand the derivative terms:

$$\left. \frac{\partial f(x_1, x_2)}{\partial x_1} \right|_{\hat{x}} = 2 \left( w_1 e_1 \frac{\partial e_1}{\partial x_1} + w_2 e_2 \frac{\partial e_2}{\partial x_1} + w_3 e_3 \frac{\partial e_3}{\partial x_1} \right) \right|_{\hat{x}} = 0$$

$$\frac{\partial f(x_1, x_2)}{\partial x_2}\bigg|_{\hat{x}} = 2\bigg(w_1 e_1 \frac{\partial e_1}{\partial x_2} + w_2 e_2 \frac{\partial e_2}{\partial x_2} + w_3 e_3 \frac{\partial e_3}{\partial x_2}\bigg)\bigg|_{\hat{x}} = 0$$

#### Rewrite in matrix form:

$$\begin{bmatrix} \frac{\partial e_1}{\partial x_1} & \frac{\partial e_2}{\partial x_1} & \frac{\partial e_3}{\partial x_1} \\ \frac{\partial e_1}{\partial x_2} & \frac{\partial e_2}{\partial x_2} & \frac{\partial e_3}{\partial x_2} \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} \begin{bmatrix} \hat{e}_1 \\ \hat{e}_2 \\ \hat{e}_3 \end{bmatrix} = 0$$

$$H^T W \hat{e} = 0$$



$$H^TW[z-h(\hat{x})]=0$$
  $H^TWh(\hat{x})=H^TWz$  Solve for x hat?



$$H^T W h(\hat{x}) = H^T W z$$



### **Newton-Raphson Method**

Linearize at an initial guess  $x_0$  hat:

$$h(\hat{x}) = h(\hat{x}^{(0)}) + \frac{\partial h(x)}{\partial x} \bigg|_{\hat{x}^{(0)}} (\hat{x}^{(1)} - \hat{x}^{(0)}) = h(\hat{x}^{(0)}) + H(\hat{x}^{(1)} - \hat{x}^{(0)})$$

Solve for x hat iteratively:

$$H^{T}W[h(\hat{x}^{(0)}) + H(\hat{x}^{(1)} - \hat{x}^{(0)})] = H^{T}Wz$$

$$\hat{x}^{(1)} = \hat{x}^{(0)} + (H^T W H)^{-1} H^T W \left[ z - h(\hat{x}^{(0)}) \right]$$



#### **State Estimation Procedure**

- Identify measurement variables and state variables (input and output)
  - *z* and *x*
- Formulate measurement equations

$$\bullet z = h(x) + e$$

- Derive Jacobian Matrix *H*:  $H = \frac{\partial h(x)}{\partial x}$
- Solve for estimated states using Newton-Raphson method (*H* needs to be updated at every step)

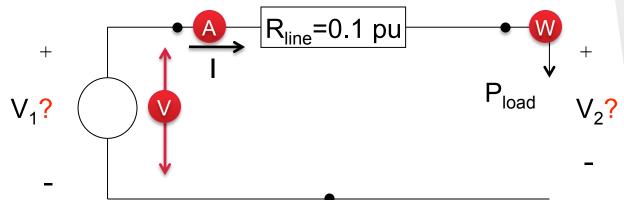
$$\hat{x}^{(k+1)} = \hat{x}^{(k)} + (H^T W H)^{-1} H^T W [z - h(\hat{x}^{(k)})]$$



### **Example – State Estimation**

#### **Problem:**

Given Measurements:  $V_m = 0.9$  pu,  $P_m = 2.6$  pu,  $I_m = 4.5$  pu, w = 100. Find  $V_1$ ,  $V_2$ .



**Solution:** 

Define measurement variables

$$z = \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} V_m \\ P_m \\ I_m \end{bmatrix}, \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}$$

Measurement equations:

and state variables: 
$$z = \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} V_m \\ P_m \\ I_m \end{bmatrix}, \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}$$
 
$$z = h(x) + e = \begin{bmatrix} V_1 \\ V_2 \frac{V_1 - V_2}{R_{line}} \\ \frac{V_1 - V_2}{R_{line}} \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix} = \begin{bmatrix} x_1 \\ 10x_1x_2 - 10x_2^2 \\ 10x_1 - 10x_2 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix}$$



### Example – State Estimation cont'd

#### Jacobian matrix:

$$h(x) = \begin{bmatrix} x_1 \\ 10x_1x_2 - 10x_2^2 \\ 10x_1 - 10x_2 \end{bmatrix} \qquad H = \frac{\partial h(x)}{\partial x} = \begin{bmatrix} 1 & 0 \\ 10x_2 & 10x_1 - 20x_2 \\ 10 & -10 \end{bmatrix}$$

#### Solve for x hat iteratively:

$$\hat{x}^{(0)} = \begin{bmatrix} 1.0 \\ 0.5 \end{bmatrix} \qquad h(\hat{x}^{(0)}) = \begin{bmatrix} 1.0 \\ 2.5 \\ 5.0 \end{bmatrix} \qquad H^{(0)} = \begin{bmatrix} 1 & 0 \\ 5 & 0 \\ 10 & -10 \end{bmatrix} \qquad z = \begin{bmatrix} 0.9 \\ 2.6 \\ 4.5 \end{bmatrix} \qquad W = \begin{bmatrix} 100 \\ 100 \\ 100 \end{bmatrix}$$

$$\hat{x}^{(1)} = \hat{x}^{(0)} + \left(H^T W H\right)^{-1} H^T W \left[z - h(\hat{x}^{(0)})\right] = \begin{bmatrix} 1.015385\\ 0.565385 \end{bmatrix}$$

$$\hat{x}^{(2)} = \hat{x}^{(1)} + \left(H^T W H\right)^{-1} H^T W \left[z - h(\hat{x}^{(1)})\right] = \begin{bmatrix} 1.021688\\0.571376 \end{bmatrix}$$

$$\hat{x}^{(3)} = \begin{bmatrix} 1.021685 \\ 0.571357 \end{bmatrix} pu$$



### Example – State Estimation cont'd

#### Objective function *f*:

$$\min_{x_1, x_2} f(x_1, x_2) = w_1 \hat{e}_1^2 + w_2 \hat{e}_2^2 + w_3 \hat{e}_3^2$$

27.000000000000000

1.642381656804740

1.554807088533027

1.554804815717737

1.554804815535279

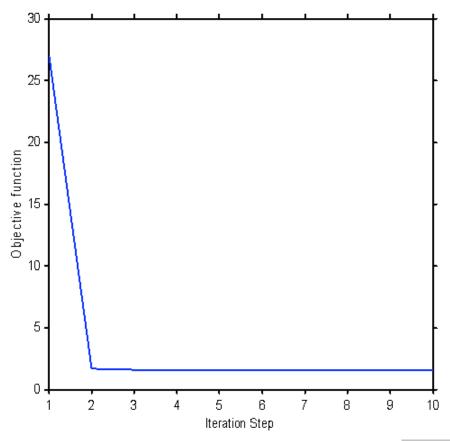
1.554804815535258

1.554804815535259

1.554804815535259

1.554804815535262

1.554804815535262





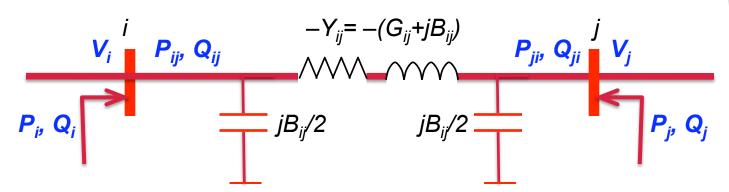
#### State Estimation vs. Power Flow

	Power Flow	State Estimation
Input	Given PV, PQ, Vθ	Measured $z = V_{rms}$ , $I_{rms}$ , $P_{line}$ , $Q_{line}$ , $P_{inj}$ , $Q_{inj}$
Output	V and θ	$x = V$ and $\theta$
Formulation	$P - P(V, \theta) = 0$ $Q - Q(V, \theta) = 0$	z - h(x) = e
Objective	Drive $\Delta P$ , $\Delta Q$ towards 0.	Drive Δz towards a minimum.
Solution Method	$\begin{bmatrix} \theta^{n+1} \\ V^{n+1} \end{bmatrix} = \begin{bmatrix} \theta^n \\ V^n \end{bmatrix} - \left[ J(x^n) \right]^{-1} \begin{bmatrix} \Delta P(x^n) \\ \Delta Q(x^n) \end{bmatrix}$	$\hat{x}^{(k+1)} = \hat{x}^{(k)} + (H^T W H)^{-1} H^T W \left[ z - h(\hat{x}^{(k)}) \right]$
Jacobian Matrix	$J(x) = \begin{bmatrix} \frac{\partial P}{\partial \theta} & \frac{\partial P}{\partial V} \\ \frac{\partial Q}{\partial \theta} & \frac{\partial Q}{\partial V} \end{bmatrix}$	$H = \left[\frac{\partial h(x)}{\partial x}\right] $ J is part of H.



#### Jacobian Matrix H

- Measurement Types
  - V<sub>i</sub>: Voltage magnitude at bus i
  - P<sub>i</sub>: Real power injection at bus i
  - Q<sub>i</sub>: Reactive power injection at bus i
  - P<sub>ii</sub>: Real power flow at bus i in line ij
  - Q<sub>ij</sub>: Reactive power flow at bus i in line ij





#### **Dimension of Jacobian Matrix**

- For a *N*-bus-*B*-line power system, maximum measurements:
  - 3N + 4B.
- State variables

• 
$$2N - 1$$
.

Redundancy Factor

• 
$$(3N + 4B)/(2N - 1)$$

Jacobian Matrix

• 
$$(3N + 4B) \times (2N - 1)$$

$$\begin{bmatrix} V_{i} \\ P_{i} \\ Q_{i} \\ P_{ji} \\ Q_{ji} \\ Q_{ji} \end{bmatrix} + e \quad H = \begin{bmatrix} //\partial \theta_{k} & //\partial V_{k} \\ \partial P_{i} / \partial \theta_{k} & /\partial V_{k} \\ \partial Q_{i} / \partial \theta_{k} & /\partial V_{k} \\ \partial P_{ij} / \partial \theta_{k} & /\partial V_{k} \\ \partial P_{ji} / \partial \theta_{k} & /\partial V_{k} \\ \partial P_{ji} / \partial \theta_{k} & /\partial V_{k} \\ \partial Q_{ij} / \partial \theta_{k} & /\partial V_{k} \\ \partial Q_{ji} / \partial Q_{ji}$$

### Jacobian Matrix – V<sub>i</sub> Entries

#### Measurement equation:

$$h(x) = V_i$$

#### Jacobian entries:

$$\frac{\partial V_{i}}{\partial \theta_{k}} = 0$$

$$\frac{\partial V_{i}}{\partial V_{k}} = \begin{cases} 1, & \text{if } i = k \\ 0, & \text{if } i \neq k \end{cases}$$

$$H = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$H = \begin{bmatrix} \frac{\partial P_i}{\partial \theta_k} & \frac{\partial P_i}{\partial V_k} \\ \frac{\partial Q_i}{\partial \theta_k} & \frac{\partial Q_i}{\partial V_k} \\ \frac{\partial P_{ij}}{\partial \theta_k} & \frac{\partial P_{ij}}{\partial V_k} \\ \frac{\partial P_{ji}}{\partial \theta_k} & \frac{\partial P_{ji}}{\partial V_k} \\ \frac{\partial Q_{ij}}{\partial \theta_k} & \frac{\partial Q_{ij}}{\partial V_k} \\ \frac{\partial Q_{ji}}{\partial \theta_k} & \frac{\partial Q_{ji}}{\partial V_k} \end{bmatrix}$$



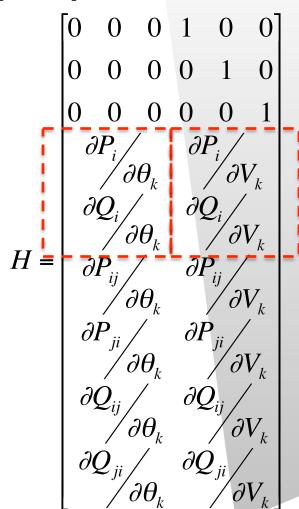
### Jacobian Matrix – $P_i$ , $Q_i$ Entries

#### Measurement equation:

$$h(x) = \begin{cases} P_i = |V_i| \sum_{j=1}^{N} |V_j| \left( G_{ij} \cos(\theta_j - \theta_i) - B_{ij} \sin(\theta_j - \theta_i) \right) \\ Q_i = -|V_i| \sum_{j=1}^{N} |V_j| \left( G_{ij} \sin(\theta_j - \theta_i) + B_{ij} \cos(\theta_j - \theta_i) \right) \end{cases}$$

Jacobian entries: (=*J* in the power flow problem)

$$\begin{cases} \frac{\partial P_{i}}{\partial \theta_{j}} = -|V_{i}||V_{j}||(G_{ij}\sin(\theta_{j} - \theta_{i}) + B_{ij}\cos(\theta_{j} - \theta_{i})) \\ \frac{\partial P_{i}}{\partial \theta_{i}} = |V_{i}|\sum_{\substack{j=1\\j \neq i}}^{N} |V_{j}||(G_{ij}\sin(\theta_{j} - \theta_{i}) + B_{ij}\cos(\theta_{j} - \theta_{i})) \\ \frac{\partial Q_{i}}{\partial |V_{j}|} = -|V_{i}||(G_{ij}\sin(\theta_{j} - \theta_{i}) + B_{ij}\cos(\theta_{j} - \theta_{i})) \\ \frac{\partial Q_{i}}{\partial |V_{i}|} = -\sum_{\substack{j=1\\i \neq i}}^{N} |V_{j}||(G_{ij}\sin(\theta_{j} - \theta_{i}) + B_{ij}\cos(\theta_{j} - \theta_{i})) - 2|V_{i}|B_{ii} \end{cases}$$





## Jacobian Matrix – $P_{ij}$ , $Q_{ij}$ Entries

$$h(x) = \begin{cases} P_{ij} = -|V_{i}|^{2} G_{ij} + |V_{i}| V_{j} | (G_{ij} \cos(\theta_{j} - \theta_{i}) - B_{ij} \sin(\theta_{j} - \theta_{i})) \\ Q_{ij} = -|V_{i}|^{2} \left(\frac{B_{ij}'}{2} - B_{ij}\right) - |V_{i}| V_{j} | (G_{ij} \sin(\theta_{j} - \theta_{i}) + B_{ij} \cos(\theta_{j} - \theta_{i})) \end{cases}$$

$$\begin{bmatrix} 0 & \frac{\partial P_{ij}}{\partial \theta_i} & 0 & \frac{\partial P_{ij}}{\partial \theta_j} & 0 \end{bmatrix} \begin{bmatrix} 0 & \frac{\partial P_{ij}}{\partial V_i} & 0 & \frac{\partial P_{ij}}{\partial V_j} & 0 \end{bmatrix}$$
$$\begin{bmatrix} 0 & \frac{\partial Q_{ij}}{\partial \theta_i} & 0 & \frac{\partial Q_{ij}}{\partial \theta_j} & 0 \end{bmatrix} \begin{bmatrix} 0 & \frac{\partial Q_{ij}}{\partial V_i} & 0 & \frac{\partial Q_{ij}}{\partial V_j} & 0 \end{bmatrix}$$

$$\begin{aligned} & \textbf{Jacobian Matrix} - P_{ij}, \ Q_{ij} \ \textbf{Entries} \\ & \textbf{Measurement equation:} \\ & h(x) = \begin{cases} P_{ij} = -|V_i|^2 G_{ij} + |V_i| V_j | (G_{ij} \cos(\theta_j - \theta_i) - B_{ij} \sin(\theta_j - \theta_i)) \\ Q_{ij} = -|V_i|^2 \left( \frac{B_{ij}^{'}}{2} - B_{ij} \right) - |V_i| V_j | (G_{ij} \sin(\theta_j - \theta_i) + B_{ij} \cos(\theta_j - \theta_i)) \end{cases} \end{aligned}$$

$$\begin{aligned} & \textbf{Jacobian entries: (only positions } \textbf{ij has value}) \\ & \left[ 0 \quad \frac{\partial P_{ij}}{\partial \theta_i} \quad 0 \quad \frac{\partial P_{ij}}{\partial \theta_j} \quad 0 \right] 0 \quad \frac{\partial P_{ij}}{\partial V_i} \quad 0 \quad \frac{\partial P_{ij}}{\partial V_j} \quad 0 \\ & 0 \quad 0 \quad 0 \quad 1 \quad 0 \\ 0 \quad 0 \quad 0 \quad 0 \quad 1 \quad 0 \\ 0 \quad 0 \quad 0 \quad 0 \quad 1 \quad 0 \\ 0 \quad 0 \quad 0 \quad 0 \quad 1 \quad 0 \\ 0 \quad 0 \quad 0 \quad 0 \quad 1 \quad 0 \\ 0 \quad 0 \quad 0 \quad 0 \quad 0 \quad 1 \quad 0 \\ 0 \quad 0 \quad 0 \quad 0 \quad 0 \quad 1 \\ \partial P_{ij} / \partial \theta_k & \partial P_{ij} / \partial V_k \\ \partial Q_{ij} / \partial \theta_k & \partial P_{ij} / \partial V_k \\ \partial P_{ji} / \partial \theta_k & \partial P_{ij} / \partial V_k \\ \partial Q_{ji} / \partial \theta_k & \partial Q_{ji} / \partial V_k \\ \partial Q_{ji} / \partial \theta_k & \partial Q_{ji} / \partial V_k \\ \partial Q_{ji} / \partial \theta_k & \partial Q_{ji} / \partial V_k \end{aligned}$$



### **Summary of Jacobian Matrix Entries**

Table 15.4 Table 15.5

It is required to understand how these elements are derived. Feel free to let me know if you have any questions.



## **Assignment**

Textbook Problem 15.14 Due: Sept 23.



#### **Questions?**

