Power-flow Methods

Fast Decoupled Power-Flow Algorithm

1

Power System Performance University **→** Viability/Acceptability – Voltages/currents within limits? → Small-signal stability - Can damp out small disturbances Transient stability – Can recover after large disturbances? **⇒**Economy - Cost minimal?

Washington State University

Powerflow Analysis

- Need to calculate voltages and currents and ensure they are all within limits
- \circ Represent transmission by π circuits
- o Formulate power-flow equations
- o Solve for voltage phasors at all buses
- Currents by circuit analysis

o Power quantities specified

Washington State University

3

DC Method Summary

- → Assume system lightly loaded
- ▶ Equivalent DC circuit
- → Simplify powerflow equations to be linear
- Direct solution. Easy to solve.
- →Only calculates phase angles
- → Cannot check voltage magnitudes
- Not valid for systems with heavy loads

Newton-Raphson Summary

- Can find accurate solutions
- → Good initial condition needed
- → Very fast convergence
- Poor initial conditions ⇒ Fast divergence
- ▶ Each iteration is time-consuming $(J^k)\Delta x^k = b - h(x^k)$
- · How to speed-up? Fast Decoupled Power-

Flow.

5

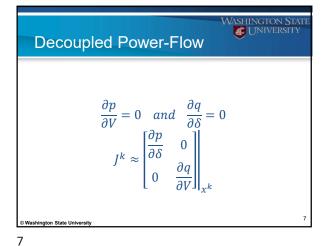
Fast Decoupled Derivation

 $P \leftrightarrow \delta$ (DC power-flow method)

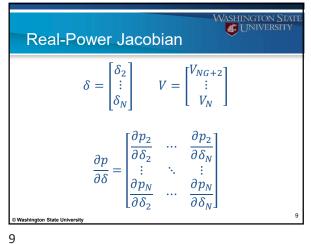
 $Q \leftrightarrow V$

$$x = \begin{bmatrix} \delta \\ V \end{bmatrix} \implies h(x) = \begin{bmatrix} p(x) \\ q(x) \end{bmatrix}$$
$$\implies \frac{\partial h}{\partial x} = \begin{bmatrix} \frac{\partial p}{\partial \delta} & \frac{\partial p}{\partial V} \\ \frac{\partial q}{\partial \delta} & \frac{\partial q}{\partial V} \end{bmatrix}$$

6



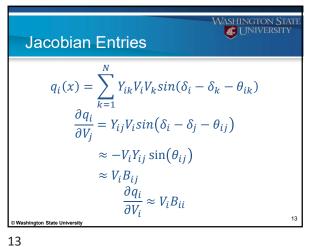
8



Jacobian Entries $p_{i}(x) = \sum_{k=1}^{N} Y_{ik} V_{i} V_{k} cos(\delta_{i} - \delta_{k} - \theta_{ik})$ $\frac{\partial p_{i}}{\partial \delta_{j}} = Y_{ij} V_{i} V_{j} sin(\delta_{i} - \delta_{j} - \theta_{ij})$ $\approx Y_{ij} V_{i} V_{j} \left[sin(\delta_{i} - \delta_{j}) \cos(\theta_{ij}) - cos(\delta_{t} - \delta_{j}) \sin(\theta_{ij}) \right]$ $\approx -Y_{ij} V_{i} V_{j} \sin(\theta_{ij})$ $\approx B_{ij} V_{i} V_{j} \approx B_{ij} V_{i} (since)$ O Washington State University

Jacobian Entry $\frac{\partial p_i}{\partial \delta_j} \approx B_{ij} V_i$ where: $B_{ij} = -Imag(\overline{Y_{ij}}) = -Y_{ij} \sin(\theta_{ij})$ $\frac{\partial p_i}{\partial \delta_j} \approx B_{ij} V_i$ $\frac{\partial p_i}{\partial \delta_i} \approx B_{ij} V_i$ $\frac{\partial p_i}{\partial \delta_i} \approx B_{ii} V_i$

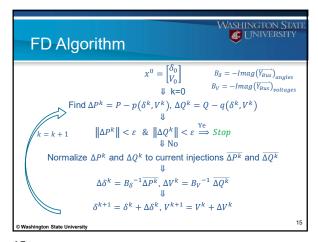
Fast Decoupled Algorithm $\frac{\partial p}{\partial \delta} \approx \begin{bmatrix} B_{22}V_2 & \cdots & B_{2N}V_2 \\ \vdots & \ddots & \vdots \\ B_{N2}V_N & \cdots & B_{NN}V_N \end{bmatrix}$ $\begin{bmatrix} B_{22}V_2^k & \cdots & B_{2N}V_2^k \\ \vdots & \ddots & \vdots \\ B_{N2}V_N^k & \cdots & B_{NN}V_N^k \end{bmatrix} \begin{bmatrix} \Delta \delta_2^k \\ \vdots \\ \Delta \delta_N^k \end{bmatrix} = \begin{bmatrix} \Delta P_2^k \\ \vdots \\ \Delta P_N^k \end{bmatrix}$ $B_{\delta} \begin{bmatrix} \Delta \delta_2^k \\ \vdots \\ \Delta P_N^k \end{bmatrix} = \begin{bmatrix} \Delta P_2^k / V_2^k \\ \vdots \\ \Delta P_N^k / V_N^k \end{bmatrix}$ $B_{\delta} [\Delta \delta] = [\overline{\Delta P}]$



ASHINGTON STA Fast Decoupled Algorithm Similarly: $B_{V} \begin{bmatrix} \Delta V_{NG+2}^{k} \\ \vdots \\ \Delta V_{N}^{k} \end{bmatrix} = \begin{bmatrix} \Delta Q_{NG+2}^{k} / V_{NG+2}^{k} \\ \vdots \\ \Delta Q_{N}^{k} / V_{N}^{k} \end{bmatrix}$ $B_V = -\operatorname{Imag}(\overrightarrow{Y_{Bus}})_{NG+2,...,N}$ $B_V[\Delta V] = [\overline{\Delta Q}]$ © Washington State University

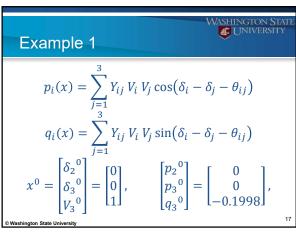
14

16



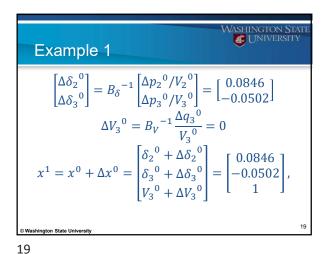
Example 1 (Same system as for NR) $V_1 = 1$ $\delta_1 = 0$ (y = -j2.5)z = j0.3z = j0.2 $P_2 = 0.7$ (y = -j5)(y=-j3.33) ε = 0.01 0.7 + j0.2 3 $\overrightarrow{Y_{Bus}} = \begin{bmatrix} -j7.5 & j2.5 & j5 \\ j2.5 & -j5.83 & j3.33 \\ j5 & j3.33 & -i8.33 \end{bmatrix}$

15



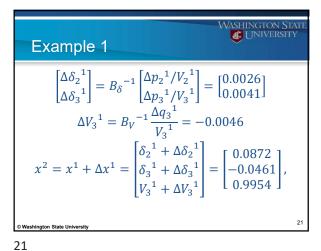
Example 1 $\begin{bmatrix} \Delta p_2^0 \\ \Delta p_3^0 \\ \Delta q_3^0 \end{bmatrix} = \begin{bmatrix} 0.7000 \\ -0.7000 \\ -0.0002 \end{bmatrix}$ $B_{\delta} = -\text{Imag}(\overrightarrow{Y_{Bus}})_{\substack{2,3\\2,3}} = \begin{bmatrix} 5.83 & -3.33\\ -3.33 & 8.33 \end{bmatrix}$ $B_{V} = -\text{Imag}(\overrightarrow{Y_{Bus}})_{\substack{3\\2,3}} = 8.33$

17 18



Example 1 $\begin{bmatrix} p_2^1 \\ p_3^1 \\ q_3^1 \end{bmatrix} = \begin{bmatrix} 0.6983 \\ -0.7254 \\ -0.1617 \end{bmatrix}$ © Washington State University

20



Example 1 $\begin{bmatrix} p_2^2 \\ p_3^2 \\ q_3^2 \end{bmatrix} = \begin{bmatrix} 0.6983 \\ -0.6966 \\ -0.1999 \end{bmatrix}$ $|max(|\Delta p^2|, |\Delta q^2|)| < 0.01 \implies \text{Stop}$ Converged in two iterations!

Fast Decoupled Summary Summary - Fast computation per iteration - More number of iterations (slower convergence) - Better tolerance of initial conditions