

# **Power System Analysis**

## **Chapter 4 Bus injection models**

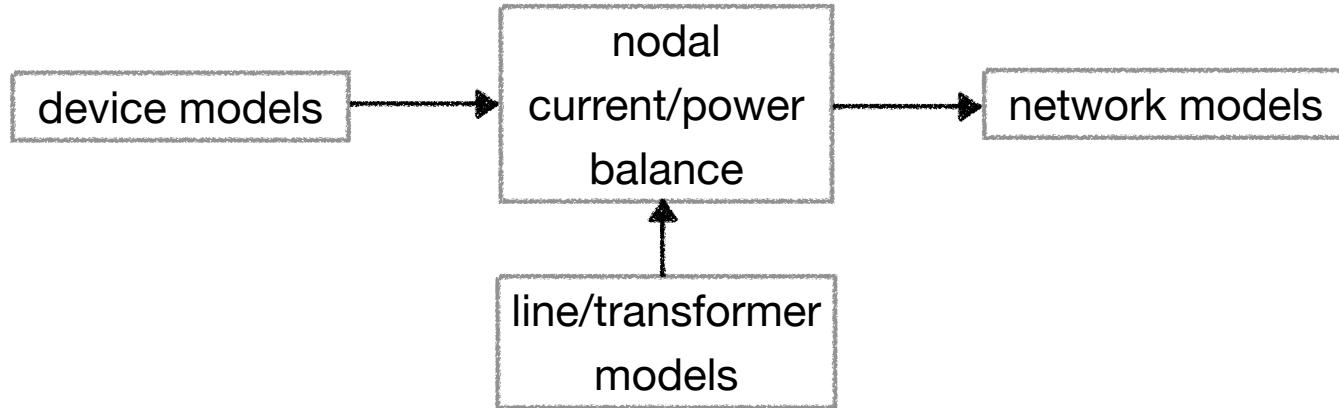
# Outline

1. Component models
2. Network model:  $VI$  relation
3. Network model:  $Vs$  relation
4. Computation methods

# Outline

1. Component models
  - Sources, impedance
  - Transmission or distribution line
  - Transformer
2. Network model:  $VI$  relation
3. Network model:  $Vs$  relation
4. Computation methods

# Overview



single-phase or 3-phase

# Single-phase devices

## 1. Single-terminal device $j$

- Voltage source  $(E_j, z_j)$ , current source  $(J_j, y_j)$ , power source  $(\sigma_j, z_j)$ , impedance  $z_j$
- Terminal variables  $(V_j, I_j, s_j)$
- External model: relation between  $(V_j, I_j)$  or  $(V_j, s_j)$

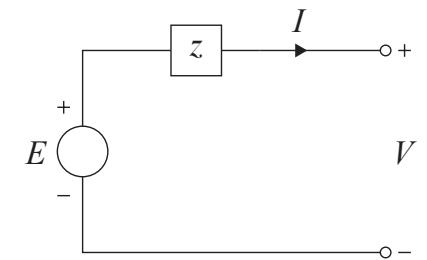
## 2. Two-terminal device $(j, k)$

- Line  $(y_{jk}^s, y_{jk}^m, y_{kj}^m)$ , transformer  $(K_{jk}(n), \tilde{y}_{jk}^s, \tilde{y}_{jk}^m)$
- Terminal variables  $(V_j, I_{jk}, S_{jk})$  and  $(V_k, I_{kj}, S_{kj})$
- External model: relation between  $(V_j, V_k, I_{jk}, I_{kj})$  or  $(V_j, V_k, S_{jk}, S_{kj})$

# Single-phase devices

## 1. Voltage source $(E_j, z_j)$

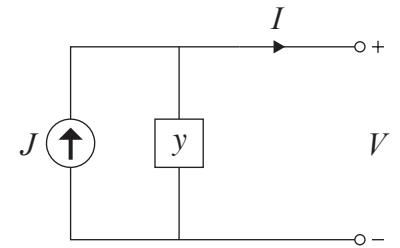
- Constant internal voltage  $E_j$  with series impedance  $z_j$
- Models for Thevenin equivalent circuit of a balanced synchronous machine, secondary side of transformer, grid-forming inverter
- External model:  $V_j = E_j - z_j I_j$
- External model:  $s_j = V_j I_j^H = y_j^H V_j (E_j - V_j)^H$



# Single-phase devices

## 2. Current source $(J_j, y_j)$

- Constant internal current  $J_j$  with shunt admittance  $y_j$
- Models for Norton equivalent circuit of a synchronous generator, load (e.g. electric vehicle charger), grid-following inverter
- External model:  $I_j = J_j - y_j V_j$
- External model:  $s_j = V_j I_j^H = V_j (J_j - y_j V_j)^H$



# Single-phase devices

## 3. Power source $(\sigma_j, z_j)$

- Constant internal power  $\sigma_j$  in series with impedance  $z_j$
- Models for load, generator, secondary side of transformer
- External model:  $\sigma_j = (V_j - z_j I_j) I_j^H$
- External model:  $s_j = V_j I_j^H = \sigma_j + z_j I_j I_j^H$

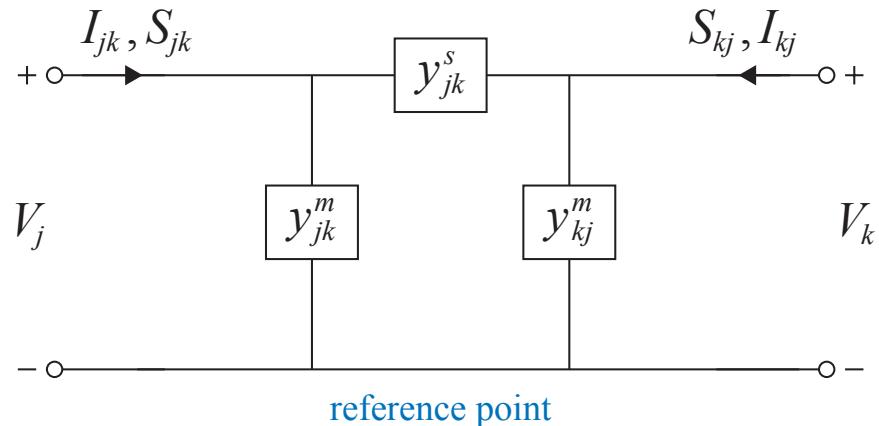
# Single-phase devices

## 4. Impedance $z_j$

- Constant impedance  $z$
- Models for load
- External model:  $V_j = z_j I_j$
- External model:  $s_j = V_j I_j^H = \frac{|V_j|^2}{z_j^H}$

# Single-phase line $(y_{jk}^s, y_{jk}^m, y_{kj}^m)$

**VI relation:  $\Pi$  circuit and admittance matrix  $Y_{\text{line}}$**



$$\begin{bmatrix} I_{jk} \\ I_{kj} \end{bmatrix} = \underbrace{\begin{bmatrix} y_{jk}^s + y_{jk}^m & -y_{jk}^s \\ -y_{jk}^s & y_{jk}^s + y_{kj}^m \end{bmatrix}}_{Y_{\text{line}}} \begin{bmatrix} V_j \\ V_k \end{bmatrix}$$

$$I_{jk} = y_{jk}^s(V_j - V_k) + y_{jk}^m V_j,$$

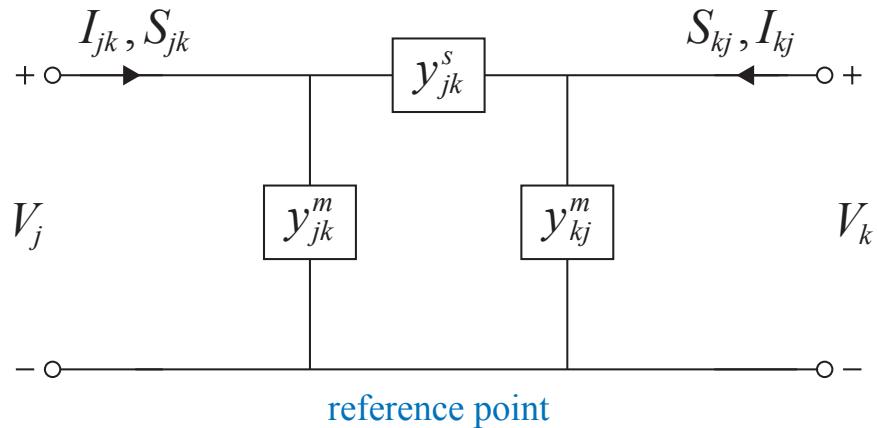
$$I_{kj} = y_{jk}^s(V_k - V_j) + y_{kj}^m V_k$$

admittance matrix  $Y_{\text{line}}$  :

- complex symmetric
- $[Y]_{jk} = -$  series admittance

# Single-phase line $(y_{jk}^s, y_{jk}^m, y_{kj}^m)$

**VI relation:  $\Pi$  circuit and admittance matrix  $Y_{\text{line}}$**



$$I_{jk} = y_{jk}^s(V_j - V_k) + y_{jk}^m V_j,$$

$$I_{kj} = y_{jk}^s(V_k - V_j) + y_{kj}^m V_k$$

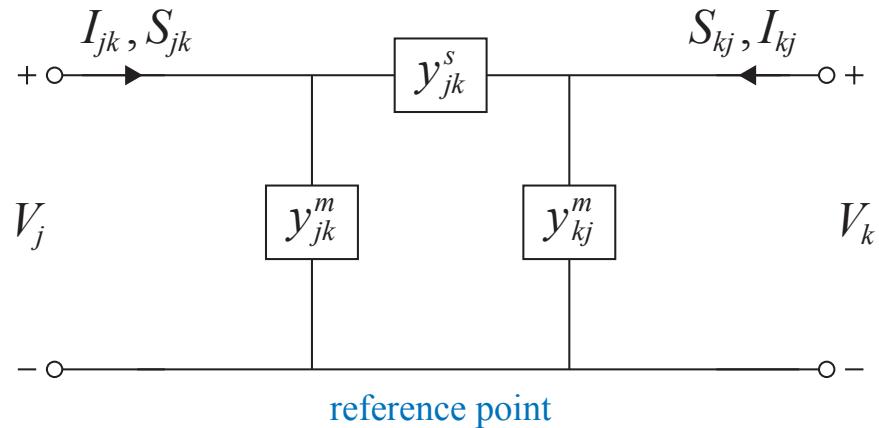
Their sum is total line current loss

$$I_{jk} + I_{kj} = y_{jk}^m V_j + y_{kj}^m V_k \neq 0$$

If  $y_{jk}^m = y_{kj}^m = 0$ , then  $I_{jk} = -I_{kj}$

# Single-phase line $(y_{jk}^s, y_{jk}^m, y_{kj}^m)$

$V_S$  relation



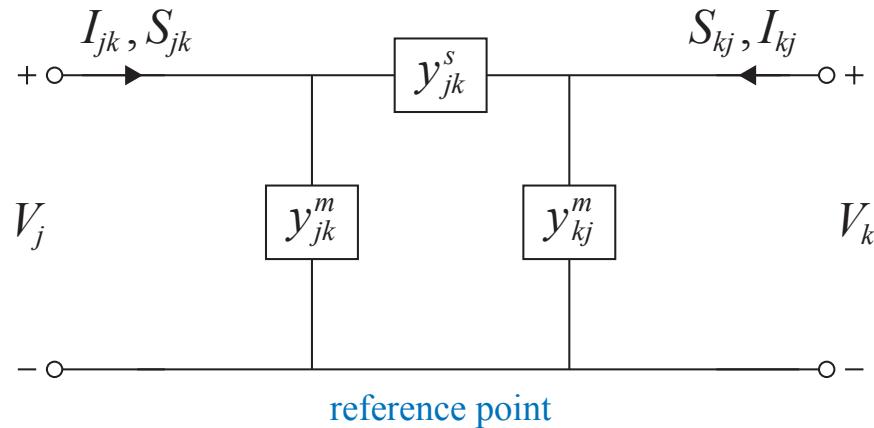
$$S_{jk} := V_j I_{jk}^H = \left( y_{jk}^s \right)^H \left( |V_j|^2 - V_j V_k^H \right) + \left( y_{jk}^m \right)^H |V_j|^2$$

$$S_{kj} := V_k I_{kj}^H = \left( y_{jk}^s \right)^H \left( |V_k|^2 - V_k V_j^H \right) + \left( y_{kj}^m \right)^H |V_k|^2$$

quadratic equations

# Single-phase line $(y_{jk}^s, y_{jk}^m, y_{kj}^m)$

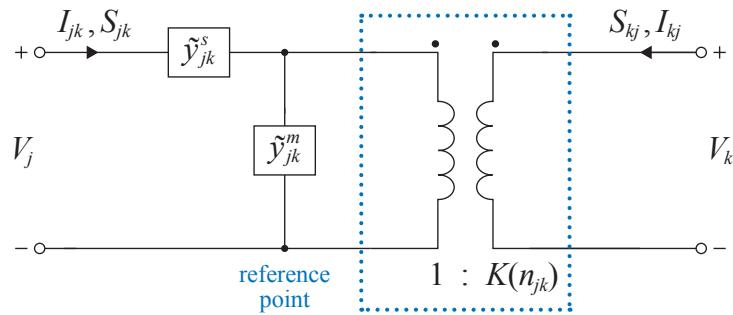
## ***Vs* relation**



## Line loss

# Single-phase transformer $\left( K\left(n_{jk}\right), \tilde{y}_{jk}^s, \tilde{y}_{jk}^m \right)$

## Complex $K\left(n_{jk}\right)$

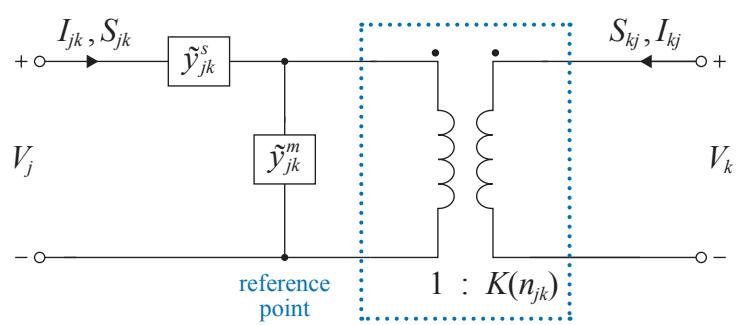


$$\begin{bmatrix} I_{jk} \\ I_{kj} \end{bmatrix} = \underbrace{\begin{bmatrix} y_{jk}^s & -y_{jk}^s/K_{jk}(n) \\ -y_{jk}^s/\bar{K}_{jk}(n) & (y_{jk}^s + y_{jk}^m)/|K_{jk}(n)|^2 \end{bmatrix}}_{Y_{\text{transformer}}} \begin{bmatrix} V_j \\ V_k \end{bmatrix}$$

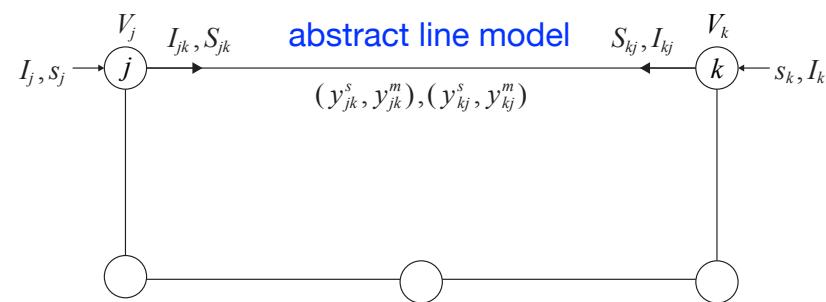
- $Y_{\text{transformer}}$  : *not symmetric*
- Has no equivalent  $\Pi$  circuit
- Use admittance or transmission matrix for analysis

# Single-phase transformer $\left(K\left(n_{jk}\right), \tilde{y}_{jk}^s, \tilde{y}_{jk}^m\right)$

## Complex $K\left(n_{jk}\right)$



$$\begin{bmatrix} I_{jk} \\ I_{kj} \end{bmatrix} = \underbrace{\begin{bmatrix} y_{jk}^s & -y_{jk}^s/K_{jk}(n) \\ -y_{jk}^s/\bar{K}_{jk}(n) & (y_{jk}^s + y_{jk}^m)/|K_{jk}(n)|^2 \end{bmatrix}}_{Y_{\text{transformer}}} \begin{bmatrix} V_j \\ V_k \end{bmatrix}$$

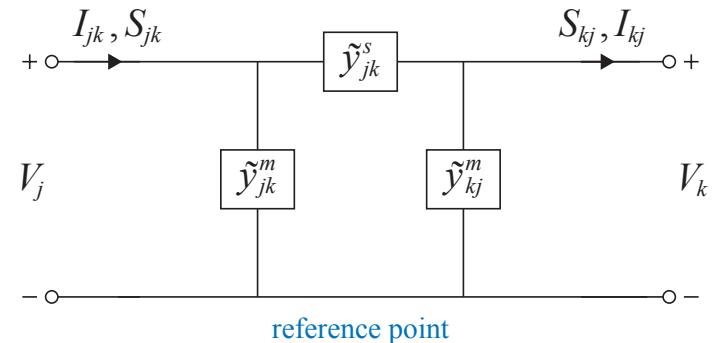
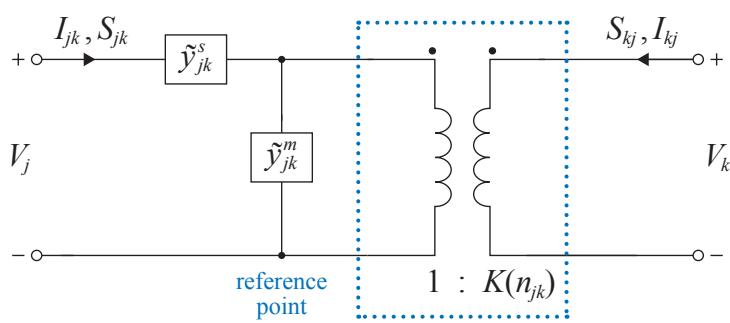


$$y_{jk}^s := \frac{\tilde{y}_{jk}^s}{K_{jk}(n)}, \quad y_{jk}^m := \left(1 - \frac{1}{K_{jk}(n)}\right)\tilde{y}_{jk}^s$$

$$y_{kj}^s := \frac{\tilde{y}_{jk}^s}{\bar{K}_{jk}(n)}, \quad y_{kj}^m := \frac{1 - K_{jk}(n)}{|K_{jk}(n)|^2}\tilde{y}_{jk}^s + \frac{1}{|K_{jk}(n)|^2}\tilde{y}_{jk}^m$$

# Single-phase transformer $\left( K \left( n_{jk} \right), \tilde{y}_{jk}^s, \tilde{y}_{jk}^m \right)$

**Real**  $K \left( n_{jk} \right) = n_{jk}$



$$I_{jk} = y_{jk}^s (V_j - a_{jk} V_k)$$

$$I_{jk} = y_{jk}^m a_{jk} V_k + n_{jk} (-I_{kj})$$

$$y_{jk}^s := a_{jk} \tilde{y}_{jk}^s = y_{kj}^s$$

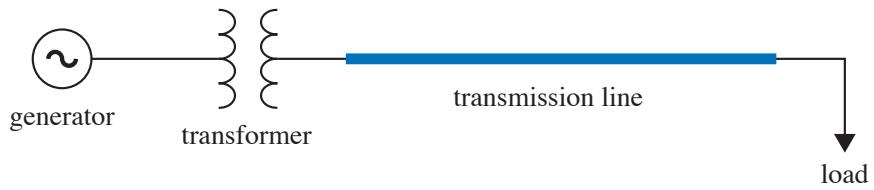
$$y_{jk}^m := (1 - a_{jk}) \tilde{y}_{jk}^s \quad \tilde{y}_{jk}^m \neq \tilde{y}_{kj}^m$$

$$y_{kj}^m := a_{jk} (a_{jk} - 1) \tilde{y}_{jk}^s + a_{jk}^2 \tilde{y}_{jk}^m$$

# Outline

1. Component models
2. Network model:  $VI$  relation
  - Example and network model
  - Admittance matrix  $Y$  and properties
  - Kron reduction  $Y/Y_{22}$  and properties
  - Radial network
3. Network model:  $Vs$  relation
4. Computation methods

# Example



## System

- Generator: current source  $(I_1, y_1)$
- Transformer  $(n, \tilde{y}^s, \tilde{y}^m)$
- Transmission line with series admittance  $y$
- Load: current source  $(I_2, y_2)$

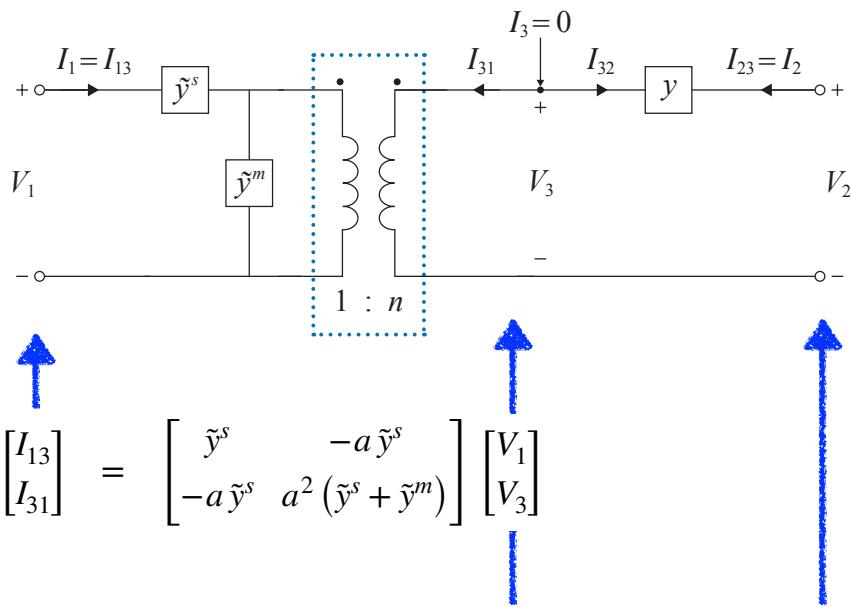
## Derive

- Derive network model (admittance matrix  $Y$ )

Derive  $Y$  in 2 steps

# Example

## Step 1: transformer + line



relate branch currents with  
nodal voltages

$$\begin{bmatrix} I_{32} \\ I_{23} \end{bmatrix} = \begin{bmatrix} y & -y \\ -y & y \end{bmatrix} \begin{bmatrix} V_3 \\ V_2 \end{bmatrix}$$

Nodal current balance (KCL):

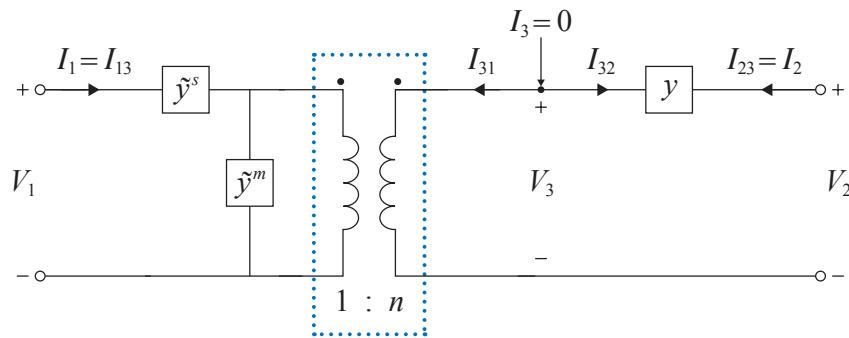
$$I_1 = I_{13}$$

$$I_3 = I_{31} + I_{32} = 0$$

$$I_2 = I_{23}$$

# Example

## Step 1: transformer + line



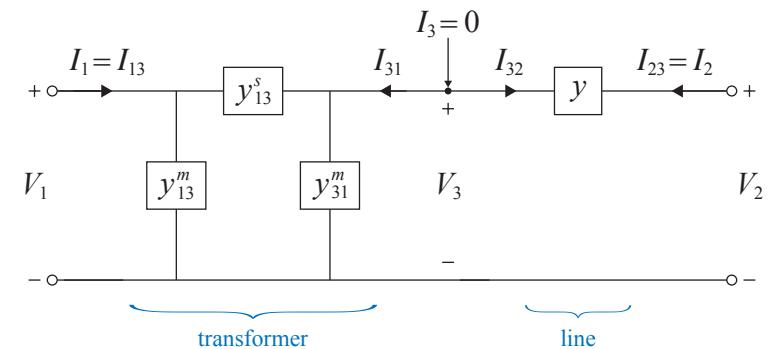
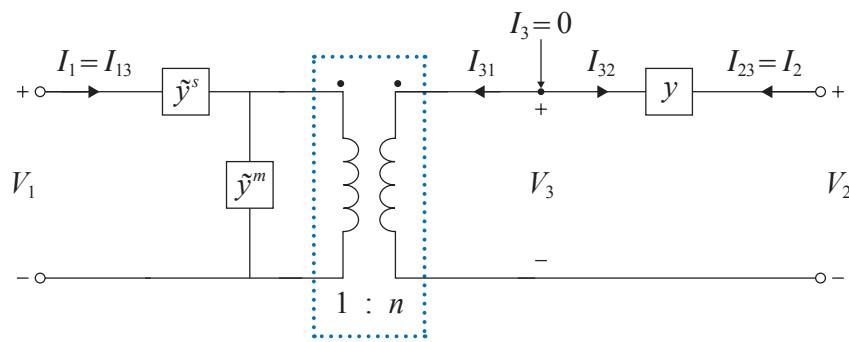
Eliminate branch currents:

$$\begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix} = \underbrace{\begin{bmatrix} \tilde{y}^s & 0 & -a\tilde{y}^s \\ 0 & y & -y \\ -a\tilde{y}^s & -y & y + a^2(\tilde{y}^s + \tilde{y}^m) \end{bmatrix}}_{Y_1} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix}$$

- $Y_1$  : complex symmetric
- Hence: admittance matrix with  $\Pi$  circuit
- Unequal shunt elements (even if  $\tilde{y}^m = 0$ )

# Example

## Step 1: transformer + line



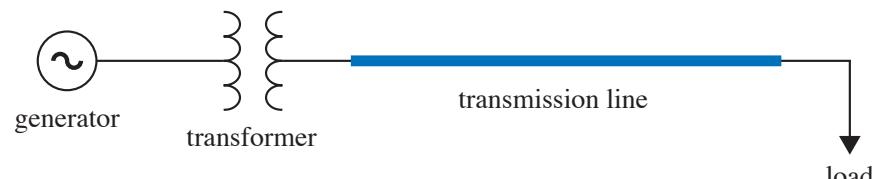
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$$\begin{aligned} y_{13}^s &:= a\tilde{y}^s \\ y_{13}^m &:= (1-a)\tilde{y}^s \\ y_{31}^m &:= a(a-1)\tilde{y}^s + a^2\tilde{y}^m \end{aligned}$$

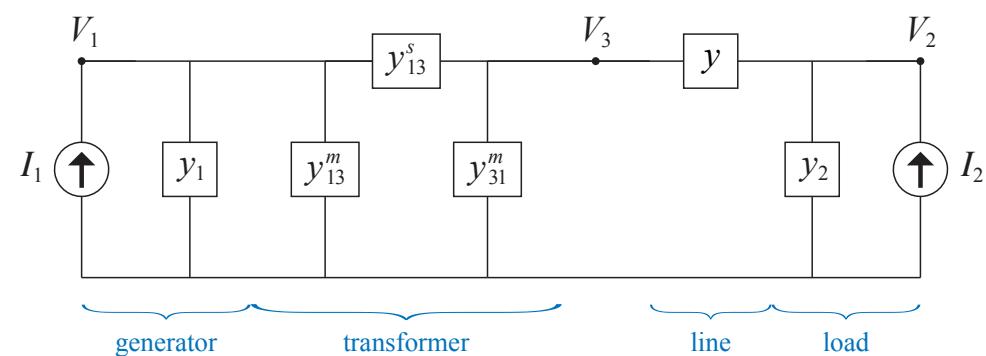
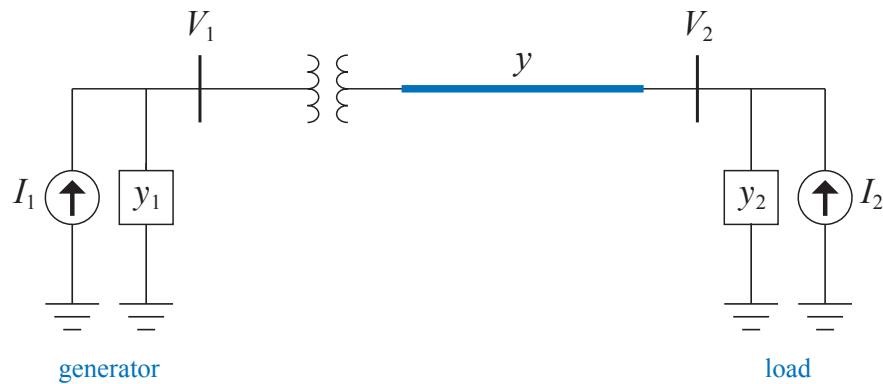
# Example

## Step 2: overall system



$$\begin{bmatrix} I_1 \\ I_2 \\ 0 \end{bmatrix} = \begin{bmatrix} \tilde{y}^s + y_1 & 0 & -a\tilde{y}^s \\ 0 & y + y_2 & -y \\ -a\tilde{y}^s & -y & y + a^2(\tilde{y}^s + \tilde{y}^m) \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix}$$

generator/load admittances



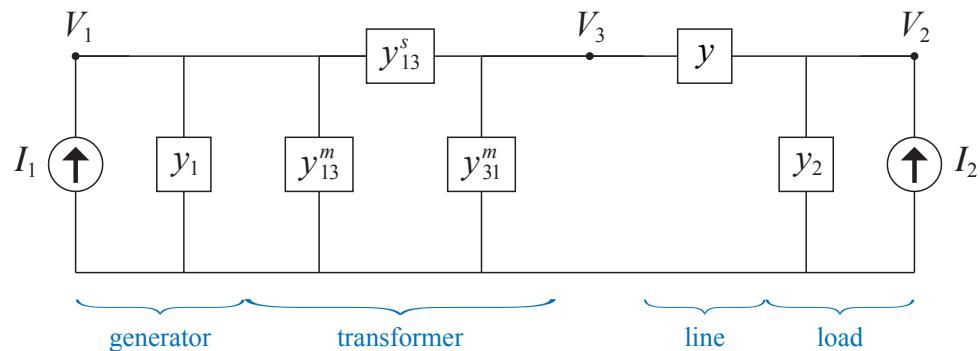
# Example

## Step 2: overall system



$$\begin{bmatrix} I_1 \\ I_2 \\ 0 \end{bmatrix} = \begin{bmatrix} \tilde{y}^s + y_1 & 0 & -a\tilde{y}^s \\ 0 & y + y_2 & -y \\ -a\tilde{y}^s & -y & y + a^2(\tilde{y}^s + \tilde{y}^m) \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix}$$

generator/load admittances



- Overall network model: **ideal** current sources connected by network
- Network: admittance matrix  $Y$
- $Y$  includes admittances of non-ideal current sources

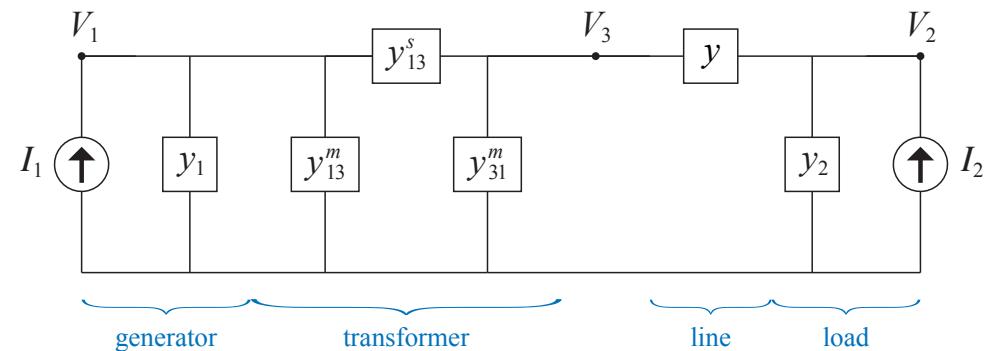
# Example

## Step 2: overall system



$$\begin{bmatrix} I_1 \\ I_2 \\ 0 \end{bmatrix} = \begin{bmatrix} \tilde{y}^s + y_1 & 0 & -a\tilde{y}^s \\ 0 & y + y_2 & -y \\ -a\tilde{y}^s & -y & y + a^2(\tilde{y}^s + \tilde{y}^m) \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix}$$

generator/load admittances



### Kron reduction (see below)

- Internal bus has zero injection  $I_3 = 0$
- Can eliminate  $(V_3, I_3)$
- External behavior: relation between  $(I_1, I_2)$  and  $(V_1, V_2)$

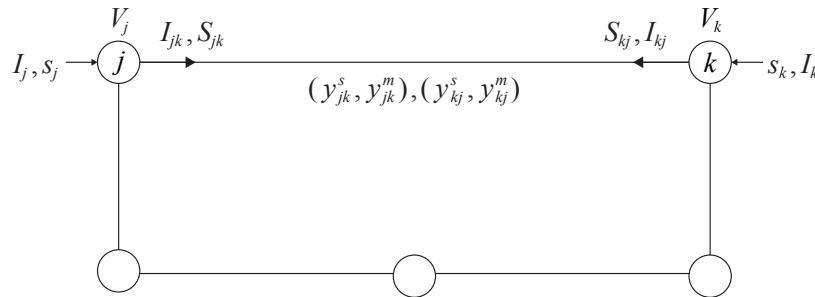
# Line model

1. Network  $G := (\bar{N}, E)$

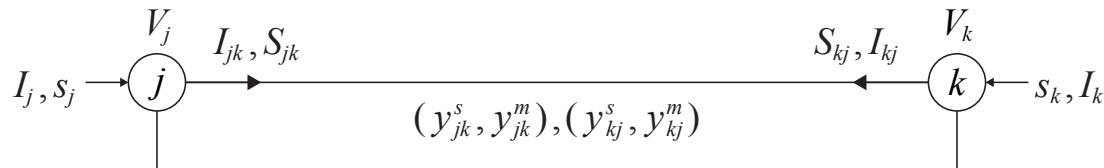
- $\bar{N} := \{0\} \cup N := \{0\} \cup \{1, \dots, N\}$  : buses/nodes/terminals
- $E \subseteq \bar{N} \times \bar{N}$  : lines/branches/links/edges

2. Each line  $(j, k)$  is parameterized by  $(y_{jk}^s, y_{jk}^m)$  and  $(y_{kj}^s, y_{kj}^m)$

- $(y_{jk}^s, y_{jk}^m)$  : series and shunt admittances from  $j$  to  $k$
- $(y_{kj}^s, y_{kj}^m)$  : series and shunt admittances from  $k$  to  $j$
- Models transmission or distribution lines, single-phase transformers



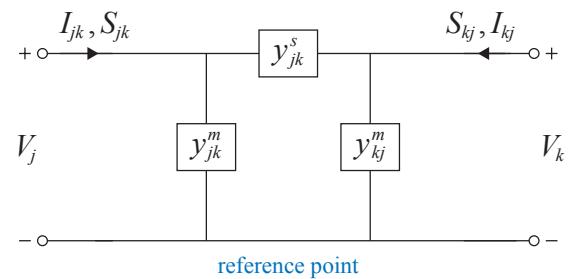
# Line model



Sending-end currents

$$I_{jk} = y_{jk}^s(V_j - V_k) + y_{jk}^m V_j, \quad I_{kj} = y_{kj}^s(V_k - V_j) + y_{kj}^m V_k,$$

If  $y_{jk}^s = y_{kj}^s$ : same relation but equivalent to  $\Pi$  circuit:



# **Network model**

## **Nodal current balance**

$$I_j = \sum_{k:j \sim k} I_{jk}$$

# Network model

## Nodal current balance

$$I_j = \sum_{k:j \sim k} I_{jk} = \left( \sum_{k:j \sim k} y_{jk}^s + y_{jj}^m \right) V_j - \sum_{k:j \sim k} y_{jk}^s V_k$$

↑  
total shunt admittance:  $y_{jj}^m := \sum_{k:j \sim k} y_{jk}^m$

# Network model

Admittance matrix  $Y$

$$I_j = \sum_{k:j \sim k} I_{jk} = \left( \sum_{k:j \sim k} y_{jk}^s + y_{jj}^m \right) V_j - \sum_{k:j \sim k} y_{jk}^s V_k$$

In vector form:

$$\textcolor{blue}{I} = \textcolor{blue}{YV} \text{ where } Y_{jk} = \begin{cases} -y_{jk}^s, & j \sim k \ (j \neq k) \\ \sum_{l:j \sim l} y_{jl}^s + y_{jj}^m, & j = k \\ 0 & \text{otherwise} \end{cases}$$

# Network model

## Admittance matrix $Y$

$Y$  can be written down by inspection of network graph

- Off-diagonal entry: – series admittance
- Diagonal entry:  $\sum$ series admittances + total shunt admittance

In vector form:

$$I = YV \text{ where } Y_{jk} = \begin{cases} -y_{jk}^s, & j \sim k \quad (j \neq k) \\ \sum_{l:j \sim l} y_{jl}^s + y_{jj}^m, & j = k \\ 0 & \text{otherwise} \end{cases}$$

# Network model

## Admittance matrix $Y$

A matrix  $Y$  has a  $\Pi$  circuit representation

- if it is complex symmetric  $(y_{jk}^s = y_{kj}^s)$

In vector form:

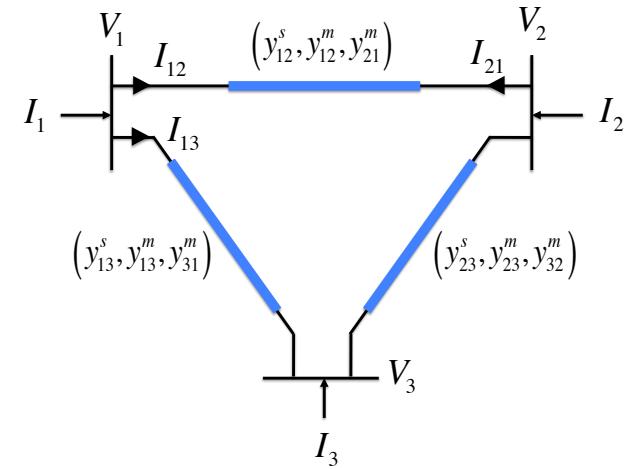
$$I = YV \text{ where } Y_{jk} = \begin{cases} -y_{jk}^s, & j \sim k \quad (j \neq k) \\ \sum_{l:j \sim l} y_{jl}^s + y_{jj}^m, & j = k \\ 0 & \text{otherwise} \end{cases}$$

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# Network model

## Example



$$\begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix} = \begin{bmatrix} y_{12}^s + y_{13}^s + y_{11}^m & -y_{12}^s & -y_{13}^s \\ -y_{12}^s & y_{12}^s + y_{23}^s + y_{22}^m & -y_{23}^s \\ -y_{13}^s & -y_{23}^s & y_{13}^s + y_{23}^s + y_{33}^m \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix}$$

total shunt admittance:  $y_{jj}^m := \sum_{k:j \sim k} y_{jk}^m$

# Admittance matrix $Y$

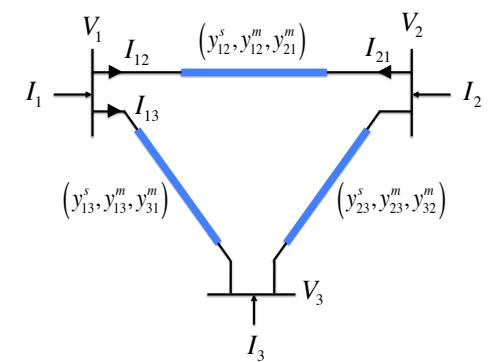
In terms of incidence matrix  $C$

bus-by-line incidence matrix

$$C_{jl} = \begin{cases} 1 & \text{if } l = j \rightarrow k \text{ for some bus } k \\ -1 & \text{if } l = i \rightarrow j \text{ for some bus } i \\ 0 & \text{otherwise} \end{cases}$$

example:

$$C = \begin{bmatrix} 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix}$$



# Admittance matrix $Y$

In terms of incidence matrix  $C$

bus-by-line incidence matrix

$$C_{jl} = \begin{cases} 1 & \text{if } l = j \rightarrow k \text{ for some bus } k \\ -1 & \text{if } l = i \rightarrow j \text{ for some bus } i \\ 0 & \text{otherwise} \end{cases}$$

$$Y = CD_y^s C^\top + D_y^m$$

where  $D_y^s := \text{diag}(y_l^s, l \in E)$ ,  $D_y^m := \text{diag}(y_{jj}^m, j \in \bar{N})$

$Y$  is a complex Laplacian matrix when  $Y^m = 0$

# Properties of $Y$

1. The inverse  $Z := Y^{-1}$ , if exists, is called a **bus impedance matrix** or an **impedance matrix**
  - Useful for fault analysis
  - Solving  $I = YV$  for  $V$
  - Advantages of  $Y$ :  $Y$  can be constructed by inspection of one-line diagram and inherits sparsity structure of  $G$ .  $Z$  can/does not.
2. Next: study existence of  $Z$ 
  - Derive (Schur complement) expressions for  $Z$ , when  $Y$  is nonsingular
  - 4 sufficient conditions for  $Y$  to be nonsingular based on the expressions for  $Z$

# Inverse of $Y$

## If exists

Let  $Y := G + iB$ ,  $Z := R + iX$

$Y$  nonsingular  $\iff \exists(R, X)$  s.t.  $YZ = ZY = \mathbb{I}$

$$\iff YZ = (GR - BX) + i(GX + BR) = \mathbb{I}$$

$$\iff \underbrace{\begin{bmatrix} G & -B \\ B & G \end{bmatrix}}_M \begin{bmatrix} R \\ X \end{bmatrix} = \begin{bmatrix} I \\ 0 \end{bmatrix}$$

Suppose  $G$  is nonsingular. Then  $Y$  nonsingular  $\iff$  Schur complement  $M/G := G + BG^{-1}B$  nonsingular

$$\text{Then } M^{-1} = \begin{bmatrix} (M/G)^{-1} & (M/G)^{-1}BG^{-1} \\ -G^{-1}B(M/G)^{-1} & G^{-1} - G^{-1}B(M/G)^{-1}BG^{-1} \end{bmatrix} \text{ and hence } \begin{bmatrix} R \\ X \end{bmatrix} = \begin{bmatrix} (M/G)^{-1} \\ -G^{-1}B(M/G)^{-1} \end{bmatrix}$$

# Invertibility of $Y$

## Theorem 1

Suppose  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ).

If  $\text{Re}(Y) > 0$ , then  $Y^{-1}$  exists, is symmetric, and  $\text{Re}(Y^{-1}) > 0$

## Proof

Let  $Y = G + iB$  with  $G > 0$ . Then  $M/G := G + BG^{-1}B > 0$  because  $G, G^{-1} > 0$  and  $B = B^T$ .

Therefore both  $G$  and  $M/G$  are nonsingular, which implies that  $Y$  is nonsingular (from previous slide).

Moreover  $\begin{bmatrix} R \\ X \end{bmatrix} = \begin{bmatrix} (M/G)^{-1} \\ -G^{-1}B(M/G)^{-1} \end{bmatrix}$  implies  $\text{Re}(Y^{-1}) = (M/G)^{-1} > 0$  since  $M/G > 0$ .

Finally, to prove  $Z := Y^{-1}$  is symmetric: substitute  $Z^T Y^T = Z^T Y$  and  $Y^T Z^T = Y Z^T$  into (transpose of)  $ZY = YZ = \mathbb{I}$  to get:

$$Z^T Y = Z^T Y^T = Y^T Z^T = Y Z^T = \mathbb{I} \quad \text{i.e., } Z^T = Y^{-1} = Z$$

# Invertibility of $Y$

Let  $y_{jk}^s =: g_{jk}^s + ib_{jk}^s$ ,  $y_{jk}^m =: g_{jk}^m + ib_{jk}^m$ ,  $y_{kj}^m =: g_{kj}^m + ib_{kj}^m$

## Conditions

1.  $g_{jk}^s, g_{jk}^m, g_{kj}^m \geq 0$  for all lines  $(j, k) \in E$ , i.e., nonnegative conductances
2.  $\sum_{k:k \sim j} g_{jk}^m \neq 0$  for all buses  $j \in \bar{N}$ , i.e., there is a shunt conductance incident on every bus
3.  $g_{jk}^s \neq 0$  for all lines  $(j, k) \in E$ , and  $\exists (j', k') \in E$  s.t.  $g_{j'k'}^m \neq 0$ , i.e., all series conductances are nonzero and there is at least one nonzero shunt conductance

## Theorem 2

Suppose  $G$  is connected and  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ). If conditions 1 and either 2 or 3 are satisfied, then

1.  $\text{Re}(Y) > 0$
2.  $Y^{-1}$  exists, is symmetric, and  $\text{Re}(Y^{-1}) > 0$

# Invertibility of $Y$

## Theorem 2

Suppose  $G$  is connected and  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ). If conditions 1 and either 2 or 3 are satisfied, then

1.  $\operatorname{Re}(Y) > 0$
2.  $Y^{-1}$  exists, is symmetric, and  $\operatorname{Re}(Y^{-1}) > 0$

## **Proof**

For any nonzero  $\rho \in \mathbb{R}^{N+1}$ , these conditions imply

$$\begin{aligned}\rho^T G \rho &= \sum_j \sum_k \rho_j \rho_k G_{jk} = \sum_j \left( \sum_{k:j \sim k} -\rho_j \rho_k g_{jk}^s + \rho_j^2 \sum_{i:j \sim i} (g_{ji}^s + g_{ji}^m) \right) \\ &= \sum_{(j,k) \in E} (\rho_j^2 - 2\rho_j \rho_k + \rho_k^2) g_{jk}^s + \sum_{j \in \bar{N}} \rho_j^2 \sum_{i:j \sim i} g_{ji}^m \\ &= \sum_{(j,k) \in E} (\rho_j - \rho_k)^2 g_{jk}^s + \sum_{j \in \bar{N}} \rho_j^2 \sum_{i:j \sim i} g_{ji}^m > 0\end{aligned}$$

# Inverse of $Y$

## If exists

Let  $Y := G + iB$ ,  $Z := R + iX$

$$Y \text{ nonsingular} \iff \underbrace{\begin{bmatrix} G & -B \\ B & G \end{bmatrix}}_M \begin{bmatrix} R \\ X \end{bmatrix} = \begin{bmatrix} I \\ 0 \end{bmatrix} \text{ which is the same as: } \underbrace{\begin{bmatrix} B & G \\ G & -B \end{bmatrix}}_{M'} \begin{bmatrix} R \\ X \end{bmatrix} = \begin{bmatrix} 0 \\ I \end{bmatrix}$$

Suppose  $B$  is nonsingular. Then  $Y$  nonsingular  $\iff$  Schur complement  $M/B := -(B + GB^{-1}G)$  nonsingular

$$\text{Then } M'^{-1} = \begin{bmatrix} B^{-1} + B^{-1}G(M'/B)^{-1}GB^{-1} & -B^{-1}G(M'/B)^{-1} \\ -(M'/B)^{-1}GB^{-1} & (M'/B)^{-1} \end{bmatrix} \text{ and hence } \begin{bmatrix} R \\ X \end{bmatrix} = \begin{bmatrix} -B^{-1}G(M'/B)^{-1} \\ (M'/B)^{-1} \end{bmatrix}$$

This leads to 2 analogous sufficient conditions in terms of  $\text{Im}(Y)$  and  $(b_{jk}^s, b_{jk}^m, b_{kj}^m)$  with similar proofs.

# Invertibility of $Y$

## Theorem 3

Suppose  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ).

If  $\text{Im}(Y) < 0$ , then  $Y^{-1}$  exists, is symmetric, and  $\text{Im}(Y^{-1}) > 0$

# Invertibility of $Y$

Let  $y_{jk}^s =: g_{jk}^s + ib_{jk}^s$ ,  $y_{jk}^m =: g_{jk}^m + ib_{jk}^m$ ,  $y_{kj}^m =: g_{kj}^m + ib_{kj}^m$

## Conditions

1.  $b_{jk}^s, b_{jk}^m, b_{kj}^m \leq 0$  for all lines  $(j, k) \in E$ , i.e., nonpositive susceptances
2.  $\sum_{k:k \sim j} b_{jk}^m \neq 0$  for all buses  $j \in \bar{N}$ , i.e., there is a shunt susceptances incident on every bus
3.  $b_{jk}^s \neq 0$  for all lines  $(j, k) \in E$ , and  $\exists (j', k') \in E$  s.t.  $b_{j'k'}^m \neq 0$ , i.e., all series susceptances are nonzero and there is at least one nonzero shunt susceptance

## Theorem 4

Suppose  $G$  is connected and  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ). If conditions 1 and either 2 or 3 are satisfied, then

1.  $\text{Im}(Y) < 0$
2.  $Y^{-1}$  exists, is symmetric, and  $\text{Im}(Y^{-1}) > 0$

# Invertibility of $Y$

## Sufficiency only

These conditions on are sufficient only

- Conditions  $(g_{jk}^s, g_{jk}^m, g_{kj}^m)$  in Theorem 2 are usually satisfied by transmission/distribution lines
- ... but not by transformers

### Example:

Example 1 with node 3 at the **primary** side of the **ideal** transformer has an admittance matrix

$$Y = \begin{bmatrix} \tilde{y}^s & 0 & -\tilde{y}^s \\ 0 & y & -ny \\ -\tilde{y}^s & -ny & \tilde{y}^s + \tilde{y}^m + n^2y \end{bmatrix}$$

Suppose  $g^s, \tilde{g}^s > 0, b^s, \tilde{b}^s \leq 0, \tilde{b}^m \geq 0$ . Then  $g_{23}^m := (1 - n)g^s$  and  $g_{32}^m := n(n - 1)g^s$  have opposite signs ( $n \neq 1$ )

Hence  $Y$  does not satisfy conditions in Theorem 2. But  $Y$  is nonsingular if and only if  $\tilde{b}_m > 0$

# Outline

1. Component models
2. Network model:  $VI$  relation
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  - Admittance matrix  $Y$  and properties
  - Kron reduction  $Y/Y_{22}$  and properties
  - Radial network
3. Network model:  $Vs$  relation
4. Computation methods

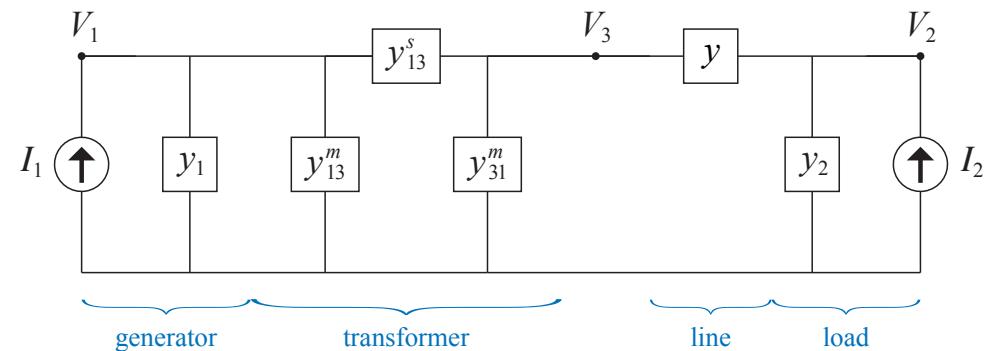
# Example

## Step 2: overall system



$$\begin{bmatrix} I_1 \\ I_2 \\ 0 \end{bmatrix} = \begin{bmatrix} \tilde{y}^s + y_1 & 0 & -a\tilde{y}^s \\ 0 & y + y_2 & -y \\ -a\tilde{y}^s & -y & y + a^2(\tilde{y}^s + \tilde{y}^m) \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix}$$

generator/load admittances



### Kron reduction (see below)

- Internal bus has zero injection  $I_3 = 0$
- Can eliminate  $(V_3, I_3)$
- External behavior: relation between  $(I_1, I_2)$  and  $(V_1, V_2)$

# Kron reduction

- $N_{\text{red}} \subseteq \bar{N}$ : buses of interest, e.g., terminal buses
- Want to relate current injections and voltages at buses in  $N_{\text{red}}$

$$\begin{bmatrix} I_1 \\ I_2 \end{bmatrix} = \underbrace{\begin{bmatrix} Y_{11} & Y_{12} \\ Y_{21} & Y_{22} \end{bmatrix}}_Y \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} \quad \begin{array}{l} \xleftarrow{N_{\text{red}}} \\ \xleftarrow{\bar{N} \setminus N_{\text{red}}} \end{array}$$

- Eliminate  $V_2 = -Y_{22}^{-1}Y_{21}V_1 + Y_{22}^{-1}I_2$
- giving  $(Y_{11} - Y_{12}Y_{22}^{-1}Y_{21})V_1 = I_1 - Y_{12}Y_{22}^{-1}I_2$   
Schur complement

# Kron reduction

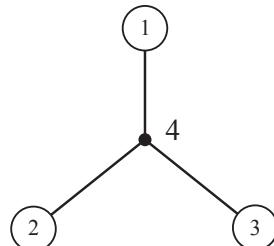
If internal injections  $I_2 = 0$  :

$$Y/Y_{22} := (Y_{11} - Y_{12}Y_{22}^{-1}Y_{21}) V_1 = I_1$$

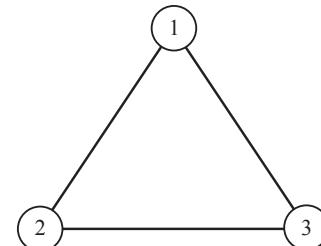
Schur complement

- Describes effective connectivity and line admittances of reduced network

Example:



(a) Original network



(b) Kron reduced network

# Existence of Kron reduction

Admittance matrix  $Y = CY^sC$  where  $Y^s := \text{diag} \left( y_{jk}^s \right)$

When  $Y$  is **real**, it is called a real Laplacian matrix

- $(N + 1) \times (N + 1)$  real symmetric matrix
- Row sum = column sum = 0
- $\text{rank}(Y) = N$ ,  $\text{null}(Y) = \text{span}(\mathbf{1})$  when all  $y_{jk}^s$  are (real &) of the **same sign** (otherwise  $\text{rank}(Y)$  can be  $< N$ )
- Any principal submatrix is invertible, i.e.,  $Y/Y_{22}$  always exists

When  $Y$  is a complex symmetric, but not Hermitian, these properties may not hold

In particular,  $Y_{22}$  may not be invertible and  $Y/Y_{22}$  may not exist

# Existence of Kron reduction

Next: Properties of  $Y_{22}$  and  $Y/Y_{22}$

- Conditions on  $(y_{jk}^s, y_{jk}^m, y_{kj}^m)$  for  $Y_{22}$  to be nonsingular, hence existence of  $Y/Y_{22}$
- Conditions on  $(y_{jk}^s, y_{jk}^m, y_{kj}^m)$  for  $Y/Y_{22}$  to be nonsingular

# Invertibility of $Y_{22}$

**When**  $y_{jk}^s = y_{kj}^s$

Recall proof of Theorem 2:

$$\rho^\top G \rho = \sum_{(j,k) \in E} \left( \rho_j - \rho_k \right)^2 g_{jk}^s + \sum_{j \in \bar{N}} \rho_j^2 \sum_{i:j \sim i} g_{ji}^m > 0$$



term associated with lines



term associated with nodes

# Invertibility of $Y_{22}$

**When**  $y_{jk}^s = y_{kj}^s$

**Recall** proof of Theorem 2:

$$\rho^\top G \rho = \sum_{(j,k) \in E} \left( \rho_j - \rho_k \right)^2 g_{jk}^s + \sum_{j \in \bar{N}} \rho_j^2 \sum_{i:j \sim i} g_{ji}^m > 0$$

Similar structure for strict principal submatrix  $Y_{22}$ :

$$\operatorname{Re} (\alpha^\top Y_{22} \alpha) = \sum_i \left( \sum_{j,k \in C_i: (j,k) \in E} g_{jk}^s \left| \alpha_j - \alpha_k \right|^2 + \sum_{j \in C_i} G_j |\alpha_j|^2 \right)$$

$$\operatorname{Im} (\alpha^\top Y_{22} \alpha) = \sum_i \left( \sum_{j,k \in C_i: (j,k) \in E} b_{jk}^s \left| \alpha_j - \alpha_k \right|^2 + \sum_{j \in C_i} B_j |\alpha_j|^2 \right)$$

# Invertibility of $Y_{22}$

## Proof

For strict principal submatrix:

$$Y_{22}[j,j] = \sum_{k \notin A:(j,k) \in E} y_{jk}^s + \sum_{k \in A:(j,k) \in E} y_{jk}^s + y_{jj}^m$$

Hence

$$\begin{aligned} \alpha^H Y_{22} \alpha &= \sum_{j \in A} \left( \left( \sum_{k \notin A:(j,k) \in E} y_{jk}^s + \sum_{k \in A:(j,k) \in E} y_{jk}^s + y_{jj}^m \right) |\alpha_j|^2 - \sum_{k \in A:(j,k) \in E} y_{jk}^s \alpha_j^H \alpha_k \right) \\ &= \sum_{j,k \in A:(j,k) \in E} \left( y_{jk}^s |\alpha_j|^2 - y_{jk}^s \alpha_j^H \alpha_k - y_{kj}^s \alpha_k^H \alpha_j + y_{kj}^s |\alpha_k|^2 \right) + \sum_{j \in A} \left( \sum_{k \notin A:(j,k) \in E} y_{jk}^s + y_{jj}^m \right) |\alpha_j|^2 \\ &= \sum_{j,k \in A:(j,k) \in E} y_{jk}^s |\alpha_j - \alpha_k|^2 + \sum_{j \in A} \left( \sum_{k \notin A:(j,k) \in E} y_{jk}^s + y_{jj}^m \right) |\alpha_j|^2 \end{aligned}$$

# Invertibility of $Y_{22}$

## Proof

For strict principal submatrix:

$$Y_{22}[j,j] = \sum_{k \notin A:(j,k) \in E} y_{jk}^s + \sum_{k \in A:(j,k) \in E} y_{jk}^s + y_{jj}^m$$

Hence

$$\operatorname{Re}(\alpha^H Y_{22} \alpha) = \sum_i \left( \sum_{j,k \in C_i:(j,k) \in E} g_{jk}^s |a_j - a_k|^2 + \sum_{j \in C_i} G_j |a_j|^2 \right)$$

$$\operatorname{Im}(\alpha^H Y_{22} \alpha) = \sum_i \left( \sum_{j,k \in C_i:(j,k) \in E} b_{jk}^s |a_j - a_k|^2 + \sum_{j \in C_i} B_j |a_j|^2 \right)$$

Similar conditions to Theorem 2:

$$\rho^T G \rho = \sum_{(j,k) \in E} (\rho_j - \rho_k)^2 g_{jk}^s + \sum_{j \in \bar{N}} \rho_j^2 \sum_{i:j \sim i} g_{ji}^m > 0$$

# Invertibility of $Y_{22}$

**When**  $y_{jk}^s = y_{kj}^s$

Let  $y_{jk}^s =: g_{jk}^s + ib_{jk}^s$ ,  $y_{jk}^m =: g_{jk}^m + ib_{jk}^m$ ,  $y_{kj}^m =: g_{kj}^m + ib_{kj}^m$

## Conditions

1. For all lines  $(j, k) \in E$ ,  $g_{jk}^s \geq 0$ ; for all buses  $j \in \bar{N}$ ,  $G_j \geq 0$
2. For all buses  $j \in \bar{N}$ ,  $G_j \neq 0$
3. For all lines  $(j, k) \in E$ ,  $g_{jk}^s \neq 0$ ; for each connected component  $C_i$ ,  $\exists j_i \in C_i$  s.t.  $G_{j_i} \neq 0$

## Theorem 5

Suppose  $G$  is connected and  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ). If conditions 1 and either 2 or 3 are satisfied, then

1.  $\text{Re}(Y_{22}) > 0$
2.  $Y_{22}^{-1}$  exists, is symmetric, and  $\text{Re}(Y_{22}^{-1}) > 0$

# Invertibility of $Y_{22}$

**When**  $y_{jk}^s = y_{kj}^s$

Let  $y_{jk}^s =: g_{jk}^s + ib_{jk}^s$ ,  $y_{jk}^m =: g_{jk}^m + ib_{jk}^m$ ,  $y_{kj}^m =: g_{kj}^m + ib_{kj}^m$

## Conditions

1. For all lines  $(j, k) \in E$ ,  $b_{jk}^s \leq 0$ ; for all buses  $j \in \bar{N}$ ,  $B_j \leq 0$
2. For all buses  $j \in \bar{N}$ ,  $B_j \neq 0$
3. For all lines  $(j, k) \in E$ ,  $b_{jk}^s \neq 0$ ; for each connected component  $C_i$ ,  $\exists j_i \in C_i$  s.t.  $B_{j_i} \neq 0$

## Theorem 6

Suppose  $G$  is connected and  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ). If conditions 1 and either 2 or 3 are satisfied, then

1.  $\text{Im}(Y_{22}) < 0$
2.  $Y_{22}^{-1}$  exists, is symmetric, and  $\text{Im}(Y_{22}^{-1}) > 0$

# Invertibility of $Y_{22}$

**When**  $y_{jk}^s = y_{kj}^s$  **and**  $y_{jk}^m = y_{kj}^m = 0$

## Corollary 7

Suppose  $G$  is connected,  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ) and  $y_{jk}^m = y_{kj}^m = 0$ .

1. If  $g_{jk}^s > 0$  for all  $(j, k) \in E$ , then  $Y_{22}^{-1}$  exists, is symmetric. Moreover  $\operatorname{Re}(Y_{22}) > 0$  and  $\operatorname{Re}(Y_{22}^{-1}) > 0$
2. If  $b_{jk}^s < 0$  for all  $(j, k) \in E$ , then  $Y_{22}^{-1}$  exists, is symmetric. Moreover  $\operatorname{Im}(Y_{22}) < 0$  and  $\operatorname{Im}(Y_{22}^{-1}) > 0$

## Theorem 8

Suppose  $G$  is connected,  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ) and  $y_{jk}^m = y_{kj}^m = 0$ . If  $g_{jk}^s \geq 0$  and  $b_{jk}^s \leq 0$   $\forall (j, k) \in E$  then

1.  $\operatorname{Re}(Y_{22}) \geq 0$ ,  $\operatorname{Im}(Y_{22}) \leq 0$ ,  $\operatorname{Re}(Y_{22}) - \operatorname{Im}(Y_{22}) > 0$
2.  $Y_{22}^{-1}$  exists and is symmetric

# Invertibility of $Y/Y_{22}$

**When**  $y_{jk}^s = y_{kj}^s$

## Theorem 9

Suppose  $Y_{22}$  is nonsingular.

1. If  $\text{Re}(Y) > 0$ , then  $(Y/Y_{22})^{-1}$  exists and is symmetric. Moreover  $\text{Re}(Y/Y_{22}) > 0$  and  $\text{Re}\left((Y/Y_{22})^{-1}\right) > 0$
2. If  $\text{Im}(Y) < 0$ , then  $(Y/Y_{22})^{-1}$  exists and is symmetric. Moreover  $\text{Im}(Y/Y_{22}) < 0$  and  $\text{Im}\left((Y/Y_{22})^{-1}\right) > 0$

# Outline

1. Component models
2. Network model:  $VI$  relation
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  - Radial network
3. Network model:  $Vs$  relation
4. Computation methods

# Radial networks

**When**  $y_{jk}^s = y_{kj}^s$  **and**  $y_{jk}^m = y_{kj}^m = 0$

$(N + 1) \times N$  incidence matrix  $C$ ,  $D_y^s := \text{diag}(y_l^s, l \in E)$ :

$$Y = CD_y^s C^\top \quad \text{admittance matrix}$$

$N \times N$  **reduced** incidence matrix  $\hat{C}$ ,  $D_y^s := \text{diag}(y_l^s, l \in E)$ :

$$\hat{Y} = \hat{C}D_y^s \hat{C}^\top \quad \text{reduced admittance matrix}$$

**Main property:**  $\hat{C}$  and hence  $\hat{Y}$  are always nonsingular. Moreover  $\hat{Z} := \hat{Y}^{-1}$  has a simple and useful structure

# Radial networks

**When**  $y_{jk}^s = y_{kj}^s$  **and**  $y_{jk}^m = y_{kj}^m = 0$

## Theorem 10

Suppose  $G$  is connected,  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ) and  $y_{jk}^m = y_{kj}^m = 0$ .

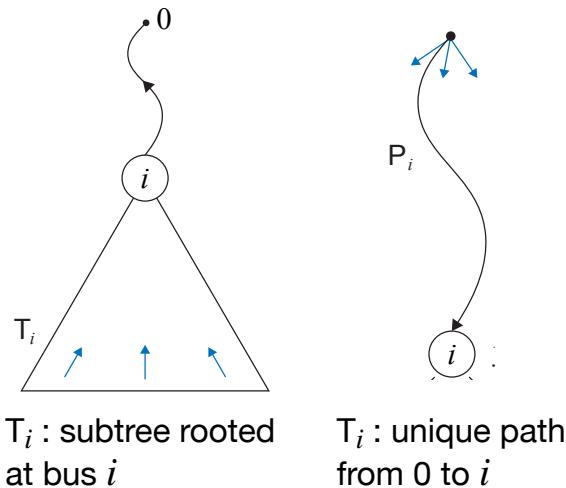
1. Reduced incidence matrix  $\hat{C}$  is nonsingular

$$[\hat{C}^{-1}]_{lj} = \begin{cases} -1 & l \in P_j \\ 1 & -l \in P_j \\ 0 & \text{otherwise} \end{cases}$$

2. Reduced admittance matrix  $\hat{Y}$  is nonsingular, and

$$\hat{Z} := \hat{Y}^{-1} = \hat{C}^{-T} D_z^s \hat{C}^{-1}$$

$$\hat{Z}_{jk} = \sum_{l \in P_j \cap P_k} z_l^s \quad \text{sum of } z_{jk}^s := 1/y_{jk}^s \text{ on common segment of paths from ref bus 0 to } j \text{ and } k$$



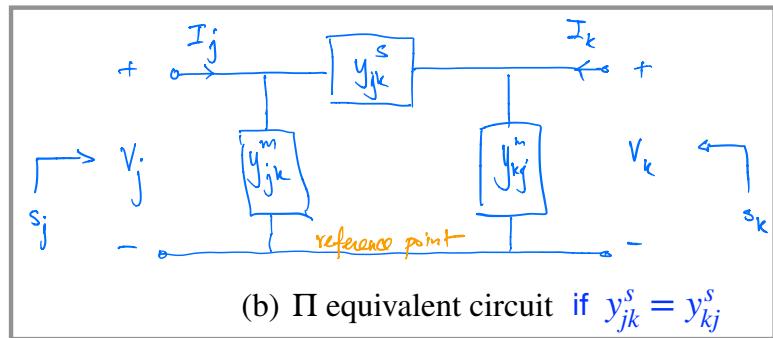
This property has been applied for topology identification, voltage control, ...

# Outline

1. Component models
2. Network model:  $VI$  relation
3. Network model:  $Vs$  relation
  - Complex form
  - Polar form
  - Cartesian form
  - Types of buses
  - Application: topology identification
4. Computation methods

# General network

## Branch currents



Sending-end currents

$$I_{jk} = y_{jk}^s(V_j - V_k) + y_{jk}^m V_j$$

$$I_{kj} = y_{kj}^s(V_k - V_j) + y_{kj}^m V_k$$

# **Power flow models**

## **Complex form**

Using  $S_{jk} := V_j I_{jk}^H$ :

$$S_{jk} = \left( y_{jk}^s \right)^H \left( |V_j|^2 - V_j V_k^H \right) + \left( y_{jk}^m \right)^H |V_j|^2$$

$$S_{kj} = \left( y_{kj}^s \right)^H \left( |V_k|^2 - V_k V_j^H \right) + \left( y_{kj}^m \right)^H |V_k|^2$$

# Power flow models

## Complex form

Bus injection model  $s_j = \sum_{k:j \sim k} S_{jk}$ :

$$s_j = \sum_{k:j \sim k} \left( y_{jk}^s \right)^H \left( |V_j|^2 - V_j V_k^H \right) + \left( y_{jj}^m \right)^H |V_j|^2$$

In terms of admittance matrix  $Y$

$$s_j = \sum_{k=1}^{N+1} Y_{jk}^H V_j V_k^H$$

N + 1 complex equations in 2(N + 1) complex variables  $(s_j, V_j, j \in \bar{N})$

# Power flow models

## Polar form

Write  $s_j = p_j + iq_j$  and  $V_j = |V_j| e^{i\theta_j}$  with  $y_{jk}^s = g_{jk}^s + ib_{jk}^s$ ,  $y_{jk}^m = g_{jk}^m + ib_{jk}^m$ :

$$p_j = \sum_{k:k \sim j} (g_{jk}^s + g_{jk}^m) |V_j|^2 - \sum_{k:k \sim j} |V_j| |V_k| (g_{jk}^s \cos \theta_{jk} + b_{jk}^s \sin \theta_{jk})$$

$$q_j = - \sum_{k:k \sim j} (b_{jk}^s + b_{jk}^m) |V_j|^2 - \sum_{k:k \sim j} |V_j| |V_k| (g_{jk}^s \sin \theta_{jk} - b_{jk}^s \cos \theta_{jk})$$

2( $N + 1$ ) real equations in 4( $N + 1$ ) real variables  $(p_j, q_j, |V_j|, \theta_j, j \in \bar{N})$

# Power flow models

## Cartesian form

Write  $s_j = p_j + iq_j$  and  $V_j = c_j + id_j$  with  $c_j = |V_j| \cos \theta_j$  and  $d_j = |V_j| \sin \theta_j$ :

$$p_j = \sum_{k:k \sim j} \left( g_{jk}^s + g_{jk}^m \right) \left( c_j^2 + d_j^2 \right) - \sum_{k:k \sim j} \left( g_{jk}^s (c_j c_k + d_j d_k) + b_{jk}^s (d_j c_k - c_j d_k) \right)$$

$$q_j = - \sum_{k:k \sim j} \left( b_{jk}^s + b_{jk}^m \right) \left( c_j^2 + d_j^2 \right) - \sum_{k:k \sim j} \left( g_{jk}^s (d_j c_k - c_j d_k) - b_{jk}^s (c_j c_k + d_j d_k) \right)$$

2( $N + 1$ ) real equations in 4( $N + 1$ ) real variables  $(p_j, q_j, c_j, d_j, j \in \bar{N})$

# Power flow models

## Types of buses

Power flow equations specify  $2(N + 1)$  real equations in  $4(N + 1)$  real variables

- Power flow (load flow) problem: given  $2(N + 1)$  values, determine remaining vars

### Types of buses

- $PV$  buses :  $(p_j, |V_j|)$  specified, determine  $(q_j, \theta_j)$ , e.g. generator
- $PQ$  buses :  $(p_j, q_j)$  specified, determine  $V_j$ , e.g. load
- Slack bus 0 :  $V_0 := 1\angle 0^\circ$  pu specified, determine  $(p_0, q_0)$

# Outline

1. Component models
2. Network model:  $VI$  relation
3. Network model:  $Vs$  relation
  - Complex form
  - Polar form
  - Cartesian form
  - Types of buses
  - Application: topology identification
4. Computation methods

# Radial networks

**When**  $y_{jk}^s = y_{kj}^s$  **and**  $y_{jk}^m = y_{kj}^m = 0$

## Theorem 10

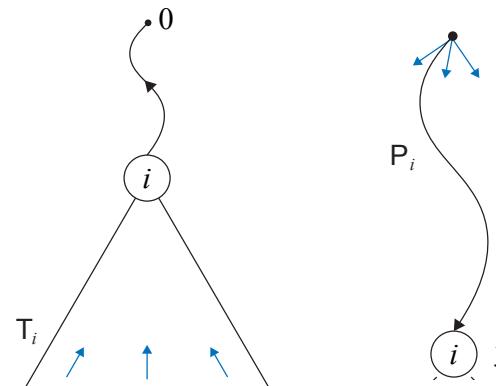
Suppose  $G$  is connected,  $Y$  is complex symmetric ( $y_{jk}^s = y_{kj}^s$ ) and  $y_{jk}^m = y_{kj}^m = 0$ .

1. Reduced incidence matrix  $\hat{C}$  is nonsingular

$$[\hat{C}^{-1}]_{lj} = \begin{cases} -1 & l \in P_j \\ 1 & -l \in P_j \\ 0 & \text{otherwise} \end{cases}$$

2. Reduced admittance matrix  $\hat{Y}$  is nonsingular, and

$$\begin{aligned} \hat{Z} &:= \hat{Y}^{-1} = \hat{C}^{-T} D_z^s \hat{C}^{-1} \\ \hat{Z}_{jk} &= \sum_{l \in P_j \cap P_k} z_l^s \end{aligned}$$



$T_i$  : subtree rooted at bus  $i$

$T_i$  : unique path from 0 to  $i$

# Topology identification

1. Distribution grid typically consists of a meshed network with sectionalizing and tie switches on some lines
2. At any time switch are configured s.t. operational network is a spanning tree (substation at its root)
3. System operator knows the meshed network, but may not always know accurately switch status and hence operational network

**Goal:** Identify operational network from measurements of voltage magnitudes

# Linearized power flow model

## Linearization of polar form

**Assumptions:** For all  $(j, k) \in E$

1.  $y_{jk}^s = y_{kj}^s = g_{jk}^s + i b_{jk}^s ; y_{jk}^m = y_{kj}^m = 0$
2.  $g_{jk}^s > 0$  and  $b_{jk}^s < 0$

Consider flat voltage profile:  $V_j^{\text{flat}} = \mu e^{i\theta} \implies (p^{\text{flat}}, q^{\text{flat}}) = (0,0)$

- All voltages have same magnitude (e.g.  $\mu = 1$  pu) and angle

Let

- $(|\hat{V}|, \hat{\theta})$  : perturbation variable around  $V^{\text{flat}}$  at non-reference buses
- $(\hat{p}, \hat{q})$  : perturbation variable around  $(p^{\text{flat}}, q^{\text{flat}}) = (0,0)$  at non-reference buses

# **Linearized power flow model**

## **Linearization of polar form**

Polar form power flow model

$$p_j = \sum_{k:k \sim j} \left( g_{jk}^s + g_{jk}^m \right) |V_j|^2 - \sum_{k:k \sim j} |V_j| |V_k| \left( g_{jk}^s \cos \theta_{jk} + b_{jk}^s \sin \theta_{jk} \right)$$

# Linearized power flow model

## Linearization of polar form

Polar form power flow model

$$p_j = \sum_{k:k \sim j} \left( g_{jk}^s + g_{jk}^m \right) |V_j|^2 - \sum_{k:k \sim j} |V_j| |V_k| \left( g_{jk}^s \cos \theta_{jk} + b_{jk}^s \sin \theta_{jk} \right)$$

Linearize around  $(V^{\text{flat}}, p^{\text{flat}}, q^{\text{flat}})$  yields a linear model from  $|\hat{V}|$  to  $(\hat{p}, \hat{q})$  at non-reference buses:

$$|\hat{V}| = \hat{R}\hat{p} + \hat{X}\hat{q} + \hat{v}_0$$

where

$$\hat{R} := \hat{C}^{-T} D_1 \hat{C}^{-1} > 0, \quad \hat{X} := -\hat{C}^{-T} D_2 \hat{C}^{-1} < 0$$

$\hat{C}$  is reduced incidence matrix and

$$D_g := \text{diag}(g_l^s) > 0, \quad D_b := \text{diag}(b_l^s) < 0$$

$$D_1 := \left( D_g + D_b D_g^{-1} D_b \right)^{-1} > 0, \quad D_2 := \left( D_b + D_g D_b^{-1} D_g \right)^{-1} < 0$$

# Covariance of voltages and powers

Define covariance and cross-covariance matrices

$$\Sigma_v := E[|\hat{V}| - E(|\hat{V}|)][|V| - E(|V|)]^\top$$

$$\Sigma_p := E[\hat{p} - E\hat{p}][\hat{p} - E\hat{p}]^\top, \quad \Sigma_q := E[\hat{q} - E\hat{q}][\hat{q} - E\hat{q}]^\top$$

$$\Sigma_{pq} := E[\hat{p} - E\hat{p}][\hat{q} - E\hat{q}]^\top, \quad \Sigma_{qp} := E[\hat{q} - E\hat{q}][\hat{p} - E\hat{p}]^\top$$

Then

$$\Sigma_v = \hat{R}\Sigma_p\hat{R}^\top + \hat{X}\Sigma_q\hat{X}^\top + \hat{R}\Sigma_{pq}\hat{X}^\top + \hat{X}\Sigma_{qp}\hat{R}^\top$$

# Covariance of voltages and powers

**Assumptions:** power injections at same bus are positively correlated, those at different buses are uncorrelated

3. For all  $j \in N$ :  $\Sigma_p[j,j] > 0$ ,  $\Sigma_q[j,j] > 0$ ,  $\Sigma_{pq}[j,j] = \Sigma_{qp}[j,j] > 0$ ;  $y_{jk}^m = y_{kj}^m = 0$
4. For all  $j \neq k$ :  $\Sigma_p[j,k] = \Sigma_q[j,k] = \Sigma_{pq}[j,k] = \Sigma_{qp}[j,k] = 0$

# Covariance of voltages and powers

**Assumptions:** power injections at same bus are positively correlated, those at different buses are uncorrelated

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## Theorem

Under assumptions 1-4:

1. If a non-reference bus  $j \in N$  is a descendant of bus  $i$ , then  $\text{var}(|V_j|) > \text{var}(|V_i|)$
2. If bus  $i$  is a parent of bus  $j$  then the variance of  $|V_i| - |V_j|$  is given by:

$$E\left((|V_i| - |V_j|) - E(|V_i| - |V_j|)\right)^2 = \sum_{k \in T_j} \left( r_{ij}^2 \text{var}(p_k) + x_{ij}^2 \text{var}(q_k) + 2r_{ij}x_{ij}\text{cov}(p_k, q_k) \right)$$

# Covariance of voltages and powers

## Proof: part 1

Theorem 10 implies

$$\hat{R}_{jk} = \sum_{l \in P_j \cap P_k} r_l > 0, \quad \hat{X}_{jk} = \sum_{l \in P_j \cap P_k} x_l > 0$$

Hence

$$\begin{aligned} \hat{R}_{jk} &= \hat{R}_{ik} + r_{ij}, & \hat{R}_{ik} &= \sum_{l \in P_i} r_l, & \text{if } k \in T_j \\ \hat{R}_{ik} &= \hat{R}_{jk}, & & & \text{if } k \notin T_j \end{aligned}$$

Use these to evaluate the diagonal entries of  $\text{var}(|V_j|) - \text{var}(|V_i|) = \Sigma_v[j, j] - \Sigma_v[i, i]$ , for each of the four terms in

$$\Sigma_v = \hat{R}\Sigma_p\hat{R}^\top + \hat{X}\Sigma_q\hat{X}^\top + \hat{R}\Sigma_{pq}\hat{X}^\top + \hat{X}\Sigma_{qp}\hat{R}^\top$$

# Covariance of voltages and powers

Due to covariances  $\Sigma_p, \Sigma_q$ :

$$(\hat{R}\Sigma_p\hat{R}^\top)[j,j] - (\hat{R}\Sigma_p\hat{R}^\top)[i,i] = \sum_{k \in T_j} \Sigma_p[k,k] \left( 2 \sum_{l \in P_i} r_l + r_{ij} \right) r_{ij} > 0$$

similarly:  $(\hat{X}\Sigma_q\hat{X}^\top)[j,j] > (\hat{X}\Sigma_q\hat{X}^\top)[i,i]$

Due to cross-covariances  $\Sigma_{pq}, \Sigma_{qp}$ :

$$(\hat{R}\Sigma_{pq}\hat{X}^\top)[j,j] - (\hat{R}\Sigma_{pq}\hat{X}^\top)[i,i] = \sum_k \Sigma_{pq}[k,k] (\hat{R}_{jk}\hat{X}_{jk} - \hat{R}_{ik}\hat{X}_{ik}) > 0$$

similarly:  $(\hat{X}\Sigma_{qp}\hat{R}^\top)[j,j] > (\hat{X}\Sigma_{qp}\hat{R}^\top)[i,i]$

yielding:  $\Sigma_v[j,j] > \Sigma_v[i,i]$

# Covariance of voltages and powers

## Proof: part 2

If bus  $i$  is a parent of bus  $j$ , then variance of  $|V_i| - |V_j|$  is:

$$E \left( (|V_i| - E|V_i|) - (|V_j| - E|V_j|) \right)^2 = \Sigma_v[i, i] + \Sigma_v[j, j] - 2\Sigma_v[i, j]$$

Again use

$$\hat{R}_{jk} = \hat{R}_{ik} + r_{ij}, \quad \hat{R}_{ik} = \sum_{l \in P_i} r_l, \quad \text{if } k \in T_j$$

$$\hat{R}_{ik} = \hat{R}_{jk}, \quad \text{if } k \notin T_j$$

to show that the first term of

$$\Sigma_v = \hat{R}\Sigma_p\hat{R}^\top + \hat{X}\Sigma_q\hat{X}^\top + \hat{R}\Sigma_{pq}\hat{X}^\top + \hat{X}\Sigma_{qp}\hat{R}^\top$$

yields a simple expression:

$$\sigma_1 := (\hat{R}\Sigma_p\hat{R}^\top)[i, i] + (\hat{R}\Sigma_p\hat{R}^\top)[j, j] - 2(\hat{R}\Sigma_p\hat{R}^\top)[i, j] = r_{ij}^2 \sum_{k \in T_j} \Sigma_p[k, k]$$

# Covariance of voltages and powers

Similarly, the other terms of

$$\Sigma_v = \hat{R}\Sigma_p\hat{R}^\top + \hat{X}\Sigma_q\hat{X}^\top + \hat{R}\Sigma_{pq}\hat{X}^\top + \hat{X}\Sigma_{qp}\hat{R}^\top$$

yield

$$\sigma_1 := (\hat{R}\Sigma_p\hat{R}^\top)[i, i] + (\hat{R}\Sigma_p\hat{R}^\top)[j, j] - 2(\hat{R}\Sigma_p\hat{R}^\top)[i, j] = r_{ij}^2 \sum_{k \in T_j} \Sigma_p[k, k]$$

$$\sigma_2 := (\hat{X}\Sigma_q\hat{X}^\top)[i, i] + (\hat{X}\Sigma_q\hat{X}^\top)[j, j] - 2(\hat{X}\Sigma_q\hat{X}^\top)[i, j] = x_{ij}^2 \sum_{k \in T_j} \Sigma_q[k, k]$$

$$\sigma_3 := (\hat{R}\Sigma_{pq}\hat{X}^\top)[i, i] + (\hat{R}\Sigma_{pq}\hat{X}^\top)[j, j] - 2(\hat{R}\Sigma_{pq}\hat{X}^\top)[i, j] = r_{ij}x_{ij} \sum_{k \in T_j} \Sigma_{pq}[k, k]$$

$$\sigma_4 := (\hat{X}\Sigma_{qp}\hat{R}^\top)[i, i] + (\hat{X}\Sigma_{qp}\hat{R}^\top)[j, j] - 2(\hat{X}\Sigma_{qp}\hat{R}^\top)[i, j] = r_{ij}x_{ij} \sum_{k \in T_j} \Sigma_{qp}[k, k]$$

# Covariance of voltages and powers

Summing:

$$\Sigma_v[i, i] - \Sigma_v[i, j] = \sum_{k=1}^4 \sigma_k = \sum_{k \in T_j} \left( r_{ij}^2 \Sigma_p[k, k] + x_{ij}^2 \Sigma_q[k, k] + 2r_{ij}x_{ij} \Sigma_{pq}[k, k] \right)$$

# Covariance of voltages and powers

## Theorem

Under assumptions 1-4:

1. If a non-reference bus  $j \in N$  is a descendant of bus  $i$ , then  $\text{var}(|V_j|) > \text{var}(|V_i|)$
2. If bus  $i$  is a parent of bus  $j$  then the variance of  $|V_i| - |V_j|$  is given by:

$$E\left((|V_i| - |V_j|) - E(|V_i| - |V_j|)\right)^2 = \sum_{k \in T_j} \left(r_{ij}^2 \text{var}(p_k) + x_{ij}^2 \text{var}(q_k) + 2r_{ij}x_{ij}\text{cov}(p_k, q_k)\right)$$

## Implications

Property 1 identifies a leaf node  $j$  as one with  $\max \text{var}(|V_j|)$

Property 2 identifies  $j$ 's parent  $i$  as one that most closely satisfies the formula

## Algorithm

1. Identify a leaf node  $j$  among unidentified nodes.
2. Identify  $j$ 's parent.     3. Remove  $j$  from set of unidentified nodes and goto 1

# Outline

1. Component models
2. Network model:  $VI$  relation
3. Network model:  $Vs$  relation
4. Computation methods
  - Gauss-Seidel algorithm
  - Newton-Raphson algorithm
  - Fast decoupled algorithm

# Computation methods

## Gauss-Seidel algorithm

Case 1: given  $V_0$  and  $(s_1, \dots, s_N)$ , determine  $s_0$  and  $(V_1, \dots, V_N)$

Power flow equations

$$s_0 = \sum_k Y_{0k}^H V_0 V_k^H$$

$$s_j = \sum_k Y_{jk}^H V_j V_k^H, \quad j \in N$$

- First compute  $(V_1, \dots, V_N)$
- Then compute  $s_0$

# Computation methods

## Gauss-Seidel algorithm

Case 1: given  $V_0$  and  $(s_1, \dots, s_N)$ , determine  $s_0$  and  $(V_1, \dots, V_N)$

Rearrange 2nd equation:

$$\frac{s_j^H}{V_j^H} = Y_{jj}V_j + \sum_{\substack{k=0 \\ k \neq j}}^N Y_{jk}V_k, \quad j \in N$$

$$V_j = \frac{1}{Y_{jj}} \left( \frac{s_j^H}{V_j^H} - \sum_{\substack{k=0 \\ k \neq j}}^N Y_{jk}V_k \right) =: f_j(V_1, \dots, V_N), \quad j \in N$$

# Computation methods

## Gauss-Seidel algorithm

Case 1: given  $V_0$  and  $(s_1, \dots, s_N)$ , determine  $s_0$  and  $(V_1, \dots, V_N)$

2nd power flow equation:

$$V = f(V)$$

where  $V := (V_j, j \in N)$ ,  $f := (f_j, j \in N)$

Gauss algorithm is the fixed point iteration

$$V(t + 1) = f(V(t))$$

# Computation methods

## Gauss-Seidel algorithm

*Case 1: given  $V_0$  and  $(s_1, \dots, s_N)$ , determine  $s_0$  and  $(V_1, \dots, V_N)$*

Gauss algorithm:

$$V_1(t+1) = f_1(V_1(t), \dots, V_N(t))$$

$$V_2(t+1) = f_2(V_1(t), \dots, V_N(t))$$

⋮

$$V_N(t+1) = f_N(V_1(t), \dots, V_{N-1}(t), V_N(t))$$

# Computation methods

## Gauss-Seidel algorithm

Case 1: given  $V_0$  and  $(s_1, \dots, s_N)$ , determine  $s_0$  and  $(V_1, \dots, V_N)$

Gauss-Seidel algorithm:

$$V_1(t + 1) = f_1(V_1(t), \dots, V_N(t))$$

$$V_2(t + 1) = f_2(V_1(t + 1), \dots, V_N(t))$$

⋮

$$V_N(t + 1) = f_N(V_1(t + 1), \dots, V_{N-1}(t + 1), V_N(t))$$

# Computation methods

## Gauss-Seidel algorithm

Case 2: given  $(V_0, \dots, V_m)$  and  $(s_{m+1}, \dots, s_N)$ , determine  $(s_j, j \leq m)$  and  $(V_j, j > m)$

Power flow equations

$$s_j = \sum_k Y_{jk}^H V_j V_k^H, \quad j \leq m$$

$$s_j = \sum_k Y_{jk}^H V_j V_k^H, \quad j > m$$

- First compute  $(V_{m+1}, \dots, V_N)$  from 2nd set of equations using the same algorithm
- Then compute  $(s_j, j \leq m)$  from 1st set of equations

# Computation methods

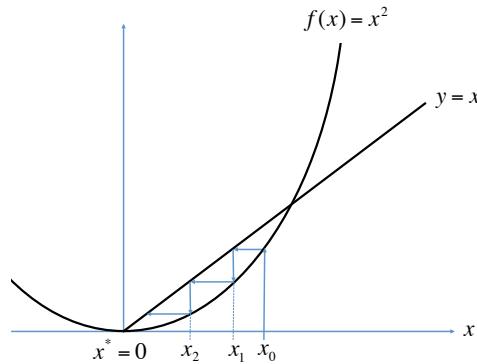
## Gauss-Seidel algorithm

If algorithm converges, the limit is a fixed point and a power flow solution

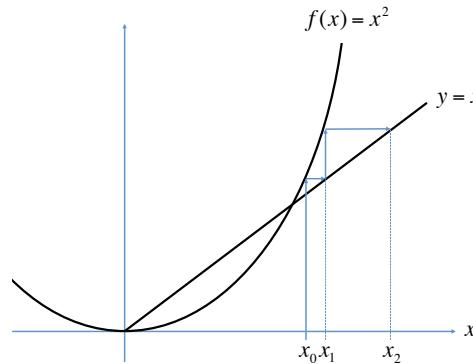
Algorithm converges linearly to unique fixed point if  $f$  is a contraction mapping

- Contraction is sufficient, but not necessary, for convergence

In general, algorithm may or may not converge depending on initial point



(a) Convergence



(b) Divergence

# Computational methods

## Newton-Raphson algorithm

To solve

$$f(x) = 0$$

where  $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ , e.g.  $\nabla F(x) = 0$  for unconstrained optimization

Idea:

- Linear approximation

$$\hat{f}(x(t+1)) = f(x(t)) + J(x(t)) \Delta x(t)$$

- Choose  $\Delta x(t)$  such that  $\hat{f}(x(t+1)) = 0$ , i.e., solve

$$J(x(t)) \Delta x(t) = -f(x(t))$$

- Next iterate  $x(t+1) := x(t) + \Delta x(t)$

$$J(x) := \frac{\partial f}{\partial x}(x) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(x) & \cdots & \frac{\partial f_1}{\partial x_n}(x) \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1}(x) & \cdots & \frac{\partial f_n}{\partial x_n}(x) \end{bmatrix}$$

# Computational methods

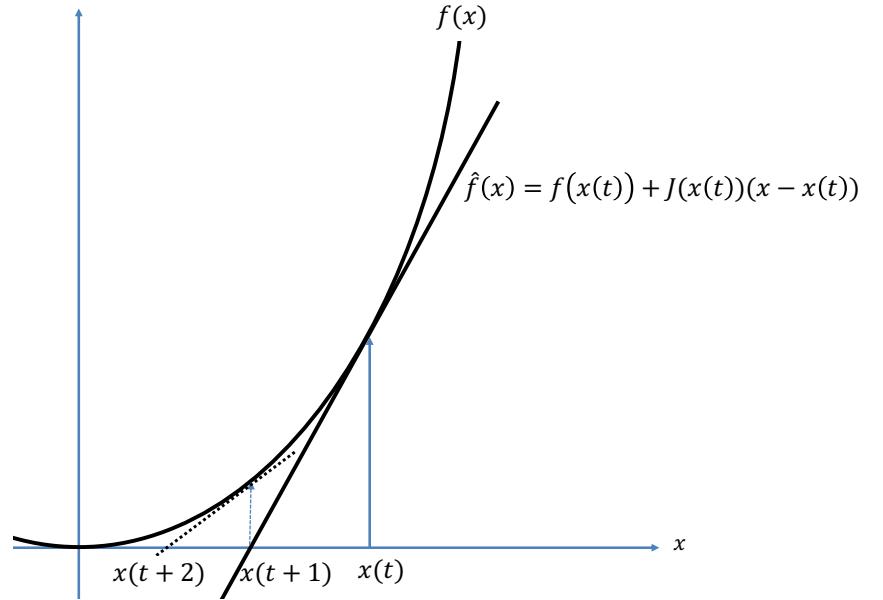
## Newton-Raphson algorithm

To solve

$$f(x) = 0$$

where  $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$ , e.g.  $\nabla F(x) = 0$  for unconstrained optimization

$$x(t+1) := x(t) - (J(x(t)))^{-1} f(x(t))$$



# Computational methods

## Newton-Raphson algorithm

### Kantorovic Theorem

Consider  $f : D \rightarrow \mathbb{R}^n$  where  $D \subseteq \mathbb{R}^n$  is an open convex set. Suppose

- $f$  is differentiable and  $\nabla f$  is Lipschitz on  $D$ , i.e.,  $\|\nabla f(y) - \nabla f(x)\| \leq L\|y - x\|$
- $x_0 \in D$  and  $\nabla f(x_0)$  is invertible

Let  $\beta \geq \left\| (\nabla f(x_0))^{-1} \right\|$ ,  $\eta \geq \left\| (\nabla f(x_0))^{-1} f(x_0) \right\|$  and

$$h := \beta\eta L, \quad r := \frac{1 - \sqrt{1 - 2h}}{h} \eta$$

# Computational methods

## Newton-Raphson algorithm

### Kantorovic Theorem

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- $f$  is differentiable and  $\nabla f$  is Lipschitz on  $D$ , i.e.,  $\|\nabla f(y) - \nabla f(x)\| \leq L\|y - x\|$
- $x_0 \in D$  and  $\nabla f(x_0)$  is invertible

If the closed ball  $B_r(x_0) \subseteq D$  and  $h \leq 1/2$ , then Newton iteration

$$x(t+1) := x(t) - (\nabla f(x(t)))^{-1} f(x(t))$$

converges to a solution  $x^* \in B_r(x_0)$  of  $f(x) = 0$

Newton-Raphson converges if it starts close to a solution, often quadratically

# Computational methods

## Newton-Raphson algorithm

Apply to power flow equations in polar form:

$$p_j(\theta, |V|) = p_j, \quad j \in N$$

$$q_j(\theta, |V|) = q_j, \quad j \in N_{pq}$$

where

$$p_j(\theta, |V|) := \left( \sum_{k=0}^N g_{jk} \right) |V_j|^2 - \sum_{k \neq j} |V_j| |V_k| \left( g_{jk} \cos \theta_{jk} + b_{jk} \sin \theta_{jk} \right)$$

$$q_j(\theta, |V|) := - \left( \sum_{k=0}^N b_{jk} \right) |V_j|^2 - \sum_{k \neq j} |V_j| |V_k| \left( g_{jk} \sin \theta_{jk} - b_{jk} \cos \theta_{jk} \right)$$

# Computational methods

## Newton-Raphson algorithm

Define  $f: \mathbb{R}^{N+N_{qp}} \rightarrow \mathbb{R}^{N+N_{qp}}$

$$f(\theta, |V|) := \begin{bmatrix} \Delta p(\theta, |V|) \\ \Delta q(\theta, |V|) \end{bmatrix} := \begin{bmatrix} p(\theta, |V|) - p \\ q(\theta, |V|) - q \end{bmatrix}$$

with

$$J(\theta, |V|) := \begin{bmatrix} \frac{\partial p}{\partial \theta} & \frac{\partial p}{\partial |V|} \\ \frac{\partial q}{\partial \theta} & \frac{\partial q}{\partial |V|} \end{bmatrix}$$

# Computational methods

## Newton-Raphson algorithm

1. Initialization: choose  $(\theta(0), |V(0)|)$

2. Iterate until stopping criteria

(a) Determine  $(\Delta\theta(t), \Delta|V|(t))$  from

$$J(\theta(t), |V|(t)) \begin{bmatrix} \Delta\theta(t) \\ \Delta|V|(t) \end{bmatrix} = - \begin{bmatrix} \Delta p(\theta(t), |V|(t)) \\ \Delta q(\theta(t), |V|(t)) \end{bmatrix}$$

(b) Set

$$\begin{bmatrix} \theta(t+1) \\ |V|(t+1) \end{bmatrix} := \begin{bmatrix} \theta(t) \\ |V|(t) \end{bmatrix} + \begin{bmatrix} \Delta\theta(t) \\ \Delta|V|(t) \end{bmatrix}$$

# Computational methods

## Fast Decoupled algorithm

Key observation: the Jacobian is roughly block-diagonal

$$J(\theta, |V|) := \begin{bmatrix} \frac{\partial p}{\partial \theta} & \frac{\partial p}{\partial |V|} \\ \frac{\partial q}{\partial \theta} & \frac{\partial q}{\partial |V|} \end{bmatrix} \approx \begin{bmatrix} \frac{\partial p}{\partial \theta} & 0 \\ 0 & \frac{\partial q}{\partial |V|} \end{bmatrix}$$

i.e., decoupling between  $p$  and  $|V|$ , and between  $q$  and  $\theta$

# Computational methods

## Fast Decoupled algorithm

Key observation: the Jacobian is roughly block-diagonal

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i.e., decoupling between  $p$  and  $|V|$ , and between  $q$  and  $\theta$

This simplifies the computation of  $\Delta x(t)$

$$\frac{\partial p}{\partial \theta}(\theta(t), |V|(t)) \Delta \theta(t) = -\Delta p(\theta(t), |V|(t))$$

$$\frac{\partial q}{\partial |V|}(\theta(t), |V|(t)) \Delta |V|(t) = -\Delta q(\theta(t), |V|(t))$$

# Computational methods

## Fast Decoupled algorithm

Decoupling assumption:  $g_{jk} = 0, \sin \theta_{jk} = 0$

$$\frac{\partial p_j}{\partial |V_k|} = \begin{cases} -|V_j| \left( g_{jk} \cos \theta_{jk} + b_{jk} \sin \theta_{jk} \right), & j \neq k \\ \frac{p_j(\theta, |V|)}{|V_j|} + \left( \sum_i g_{ji} \right) |V_j|, & j = k \end{cases}$$

$$g_{jk} = 0, \sin \theta_{jk} = 0, p_j(\theta, |V|) = 0 \Rightarrow \frac{\partial p}{\partial |V|} = 0$$

# Computational methods

## Fast Decoupled algorithm

Decoupling assumption:  $g_{jk} = 0, \sin \theta_{jk} = 0$

$$\frac{\partial q_j}{\partial \theta_k} = \begin{cases} |V_j| |V_k| \left( g_{jk} \cos \theta_{jk} + b_{jk} \sin \theta_{jk} \right), & j \neq k \\ p_j(\theta, |V|) - \left( \sum_i g_{ji} \right) |V_j|^2, & j = k \end{cases}$$

$$g_{jk} = 0, \sin \theta_{jk} = 0, p_j(\theta, |V|) = 0 \Rightarrow \frac{\partial q}{\partial \theta} = 0$$

# Summary

1. Component models
  - Single-phase devices, line, transformer
2. Network models
  - $VI$  relation (admittance matrix  $Y$ ),  $Vs$  relation (power flow equations)
  - Radial network: inverse of reduced admittance matrix has simple structure
3. Computation methods
  - Gauss-Seidel algorithm, Newton-Raphson algorithm, Fast decoupled algorithm