

# **Power Systems Analysis**

## **Chapter 6 Branch flow models**

# Outline

1. General network
2. Radial network
3. Equivalence
4. Backward forward sweep
5. Linearized model

# Outline

1. General network
  - Complex form power flow equations
  - Real form power flow equations
2. Radial network
3. Equivalence
4. Backward forward sweep
5. Linearized model

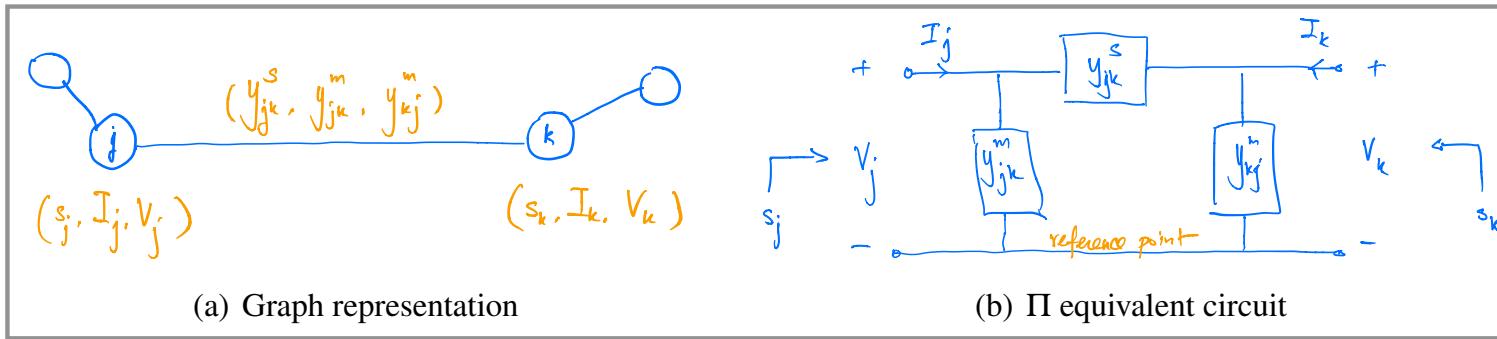
# General network

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

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

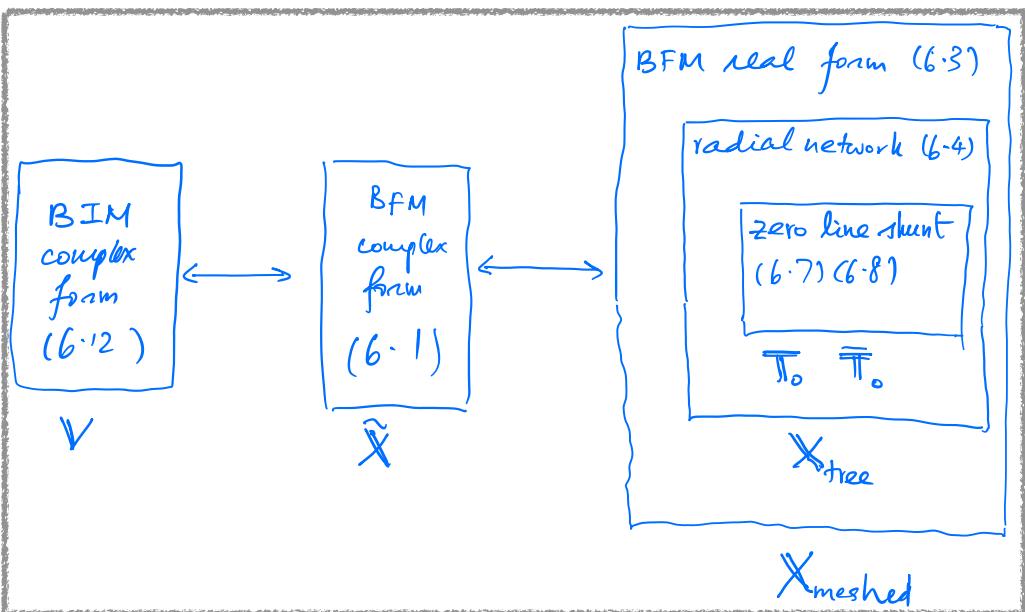
## 2. Each line $(j, k)$ is parameterized by $(y_{jk}^s, y_{jk}^m, y_{kj}^m)$

- $y_{jk}^s$  : series admittance
- $y_{jk}^m, y_{kj}^m$  : shunt admittances, generally different



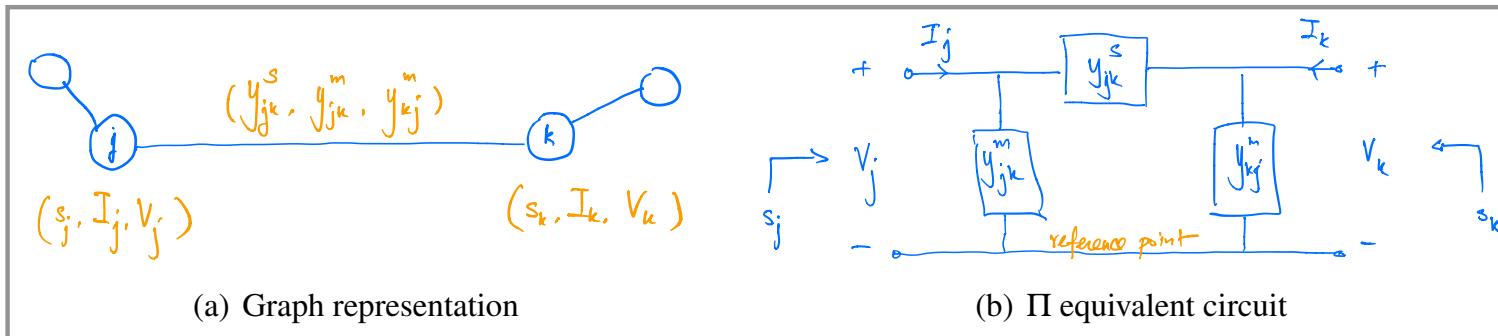
# General network

1. We will introduce several BFM models
  - General: complex form, real form
  - Radial: with/without shunt admits.
2. Each model defined by
  - Set of variables
  - Set of power flow equations relating these variables
3. These models are equivalent to each other, and to BIM



# General network

## Branch currents



## Sending-end currents

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

Bus injection model: relate nodal variables  $s$  and  $V$

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

# General network

## Complex form

Branch flow model: includes branch vars as well

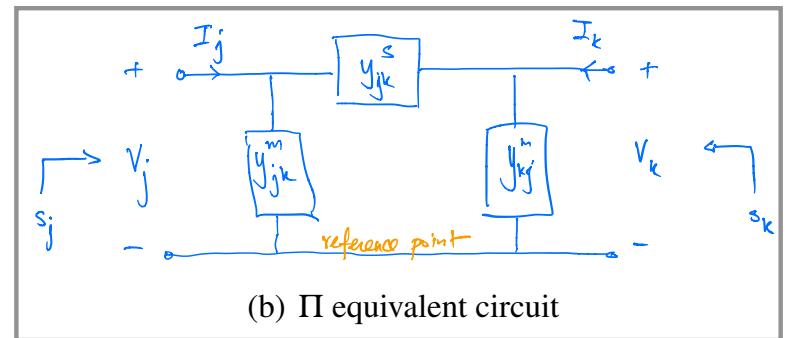
- Branch currents  $(I_{jk}, I_{kj})$
- Branch power  $(S_{jk}, S_{kj})$

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

$$S_{jk} = V_j I_{jk}^H, \quad S_{kj} = V_k I_{kj}^H$$

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

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



This model is equivalent to BIM (later)

- Serves as a bridge to BIM

# General network

## Real form

Key feature of original Dist Flow equations (branch flow model) of Baran-Wu1989

- No voltage/current phase angles
- Suitable for radial networks (tree topology)
- We generalize to meshed networks

For each bus  $j$

- $s_j := (p_j, q_j)$  or  $s_j := p_j + iq_j$  : power injection
- $v_j$  : squared voltage magnitude

For each branch  $(j, k)$

- $S_{jk} := (P_{jk}, Q_{jk})$  or  $S_{jk} := P_{jk} + iQ_{jk}$  : **sending-end** power  $j \rightarrow k$ ; also  $S_{kj}$  from  $k \rightarrow j$
- $(\ell_{jk}, \ell_{kj})$  : squared magnitude of **sending-end** current  $j \rightarrow k$ , and  $k \rightarrow j$

# General network

## Real form

The variables  $v_i$  and  $(\ell_{jk}, \ell_{kj})$  contain no angle information

Angle info must be recoverable from a power flow solution  $x := (s, v, \ell, S) \in \mathbb{R}^{3(N+1)+6M}$

- This is easy for radial networks
- Trickier for meshed networks

# General network

## Real form

For each line  $(j, k)$  let:

$$z_{jk}^s := \left( y_{jk}^s \right)^{-1} =: z_{kj}^s$$

$$\alpha_{jk} := 1 + z_{jk}^s y_{jk}^m, \quad \alpha_{kj} := 1 + z_{kj}^s y_{kj}^m$$

$\alpha_{jk} = \alpha_{kj}$  if and only if  $y_{jk}^m = y_{kj}^m$

$\alpha_{jk} = \alpha_{kj} = 1$  if and only if  $y_{jk}^m = y_{kj}^m = 0$

# General network

## Real form

For each line  $(j, k)$  let:

$$z_{jk}^s := \left( y_{jk}^s \right)^{-1} =: z_{kj}^s$$

$$\alpha_{jk} := 1 + z_{jk}^s y_{jk}^m, \quad \alpha_{kj} := 1 + z_{kj}^s y_{kj}^m$$

Given  $x := (s, v, \ell, S) \in \mathbb{R}^{3(N+1)+6M}$  define nonlinear functions:

$$\beta_{jk}(x) := \angle \left( \alpha_{jk}^H v_j - \left( z_{jk}^s \right)^H S_{jk} \right)$$

$$\beta_{kj}(x) := \angle \left( \alpha_{kj}^H v_k - \left( z_{kj}^s \right)^H S_{kj} \right)$$

If  $x$  is a power flow solution, then  $(\beta_{jk}(x), \beta_{kj}(x))$  are angle differences across  $(j, k)$

# General network

## Real form

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

power balance

# General network

## Real form

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

power balance

$$\left| S_{jk} \right|^2 = v_j \ell_{jk}, \quad \left| S_{kj} \right|^2 = v_k \ell_{kj}$$

branch power magnitude

The complex notation is only shorthand for real equations

$$p_j = \sum_k P_{jk}, \quad q_j = \sum_k Q_{jk}$$

$$v_j \ell_{jk} = P_{jk}^2 + Q_{jk}^2, \quad v_k \ell_{kj} = P_{kj}^2 + Q_{kj}^2$$

# General network

## Real form

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

power balance

$$\left| S_{jk} \right|^2 = v_j \ell_{jk}, \quad \left| S_{kj} \right|^2 = v_k \ell_{kj}$$

branch power magnitude

$$\left| \alpha_{jk} \right|^2 v_j - v_k = 2 \operatorname{Re} \left( \alpha_{jk} \left( z_{jk}^s \right)^H S_{jk} \right) - \left| z_{jk}^s \right|^2 \ell_{jk}$$

Ohm's law, KCL (magnitude)

$$\left| \alpha_{kj} \right|^2 v_k - v_j = 2 \operatorname{Re} \left( \alpha_{kj} \left( z_{kj}^s \right)^H S_{kj} \right) - \left| z_{kj}^s \right|^2 \ell_{kj}$$

# General network

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Ohm's law, KCL (magnitude)

$$\text{there exists } \theta \in \mathbb{R}^{N+1} \text{ s.t. } \beta_{jk}(x) = \theta_j - \theta_k$$

cycle condition

$$\beta_{kj}(x) = \theta_k - \theta_j$$

# General network

## Real form

Cycle condition on  $x$  is highly nonlinear

$$\beta_{jk}(x) := \angle \left( \alpha_{jk}^H v_j - \left( z_{jk}^s \right)^H S_{jk} \right)$$

$$\beta_{kj}(x) := \angle \left( \alpha_{kj}^H v_k - \left( z_{jk}^s \right)^H S_{kj} \right)$$

$$\beta(x) = \begin{bmatrix} C^T \\ -C^T \end{bmatrix} \theta \text{ for some } \theta \in \mathbb{R}^{N+1}$$

It ensures angle consistency of a power flow solution  $x$

# General network

## Real form

Any  $x := (s, v, \ell, S) \in \mathbb{R}^{3(N+1)+6M}$  that satisfies power flow equations with

$$v \geq 0, \quad \ell \geq 0$$

is a power flow solution

Branch flow models have been most useful for radial networks

All BFM<sub>s</sub> for radial networks are special cases of this model

- Tree topology
- Tree topology with zero shunt admittances  $y_{jk}^m = y_{kj}^m = 0$
- Tree topology with linear approximations

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  - Without shunt admittances
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5. Linearized model

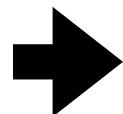
# Radial network

## Cycle condition

Major simplification for radial network: nonlinear cycle condition becomes linear in  $x$

$$\begin{aligned}\beta_{jk}(x) &:= \angle\left(\alpha_{jk}^H v_j - \left(z_{jk}^s\right)^H S_{jk}\right) \\ \beta_{kj}(x) &:= \angle\left(\alpha_{kj}^H v_k - \left(z_{kj}^s\right)^H S_{kj}\right) \\ \beta(x) &= \begin{bmatrix} C^T \\ -C^T \end{bmatrix} \theta \text{ for some } \theta \in \mathbb{R}^{N+1}\end{aligned}$$

general network



$$\alpha_{jk}^H v_j - \left(z_{jk}^s\right)^H S_{jk} = \left(\alpha_{kj}^H v_k - \left(z_{kj}^s\right)^H S_{kj}\right)^H$$

radial network

# Radial network

## With shunt admittances

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

power balance

$$\left| S_{jk} \right|^2 = v_j \ell_{jk}, \quad \left| S_{kj} \right|^2 = v_k \ell_{kj}$$

branch power magnitude

$$\left| \alpha_{jk} \right|^2 v_j - v_k = 2 \operatorname{Re} \left( \alpha_{jk} \left( z_{jk}^s \right)^H S_{jk} \right) - \left| z_{jk}^s \right|^2 \ell_{jk}$$

Ohm's law, KCL (magnitude)

$$\left| \alpha_{kj} \right|^2 v_k - v_j = 2 \operatorname{Re} \left( \alpha_{kj} \left( z_{kj}^s \right)^H S_{kj} \right) - \left| z_{kj}^s \right|^2 \ell_{kj}$$

cycle condition

$$\alpha_{jk}^H v_j - \left( z_{jk}^s \right)^H S_{jk} = \left( \alpha_{kj}^H v_k - \left( z_{kj}^s \right)^H S_{kj} \right)^H$$

$2(N+1) + 6M$  real equations in  $3(N+1) + 6M$  real vars ( $M = N$ )

# Radial network

## With shunt admittances

Any  $x := (s, v, \ell, S) \in \mathbb{R}^{3(N+1)+6M}$  that satisfies power flow equations with

$$v \geq 0, \quad \ell \geq 0$$

is a power flow solution

All equations are **linear** in  $x$ , except the **quadratic** equalities

$$\left| S_{jk} \right|^2 = v_j \ell_{jk}, \quad \left| S_{kj} \right|^2 = v_k \ell_{kj}$$

This can be relaxed to second-order cone constraint in OPF (later)

# General network

## Angle recovery

Treat network  $G := (\bar{N}, E)$  as directed graph with arbitrary orientation

- (Re-)Define branch variables  $(S_{jk}, \ell_{jk})$  only in direction of lines  $(j, k)$
- (Re-)Define  $\beta(x) := (\beta_{jk}(x), (j, k) \in E)$  where

$$\beta_{jk}(x) := \angle \left( \alpha_{jk}^H v_j - \left( z_{jk}^s \right)^H S_{jk} \right)$$

Any power flow solution  $x := (s, v, \ell, S) \in \mathbb{R}^{3(N+1)+6M}$  implies

$$\beta(x) = C^T \theta \text{ for some } \theta \in \mathbb{R}^{N+1}$$

### Angle recovery:

$$V_j = \sqrt{v_j} e^{i\theta_j}, \quad I_{jk} = \sqrt{\ell_{jk}} e^{i(\theta_j - \angle S_{jk})}$$

# Radial network

## Without shunt admittances

When shunt admittances  $y_{jk}^m = y_{kj}^m = 0$

- $\alpha_{jk} = \alpha_{kj} = 1$
- $\ell_{kj} = \ell_{jk}$  and  $S_{kj} + S_{jk} = z_{jk}^s \ell_{jk}$

Can use **directed** graph with vars  $(\ell_{jk}, S_{jk})$  defined **only** in direction of lines  $(j, k)$

Substitute  $(\ell_{kj}, S_{kj})$  in terms of  $(\ell_{jk}, S_{jk})$  into previous power flow equations yields original

DistFlow equations of [Baran-Wu 1989]

# Radial network

## Without shunt admittances

DistFlow equations [Baran-Wu 1989] (down direction):

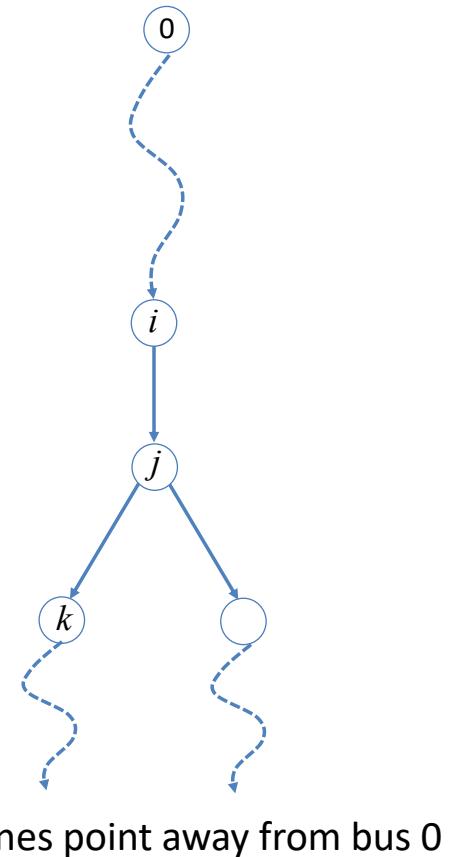
$$\sum_{k:j \rightarrow k} S_{jk} = S_{ij} - z_{ij}\ell_{ij} + s_j \quad \text{power balance}$$

$$v_j - v_k = 2 \operatorname{Re} \left( z_{jk}^H S_{jk} \right) - |z_{jk}|^2 \ell_{jk} \quad \text{Ohm's law, KCL (magnitude)}$$

$$v_j \ell_{jk} = |S_{jk}|^2 \quad \text{branch power magnitude}$$

$2(N + 1) + 2M$  real equations in  $3(N + 1) + 3M$  real vars ( $M = N$ )

- Given  $(v_0, s_j, j \in N)$ , there are  $4N + 2$  equations in  $4N + 2$  vars  $(s_0, v_j, j \in N, \ell, S)$



# Radial network

## Without shunt admittances

DistFlow equations [Baran-Wu 1989] (down direction):

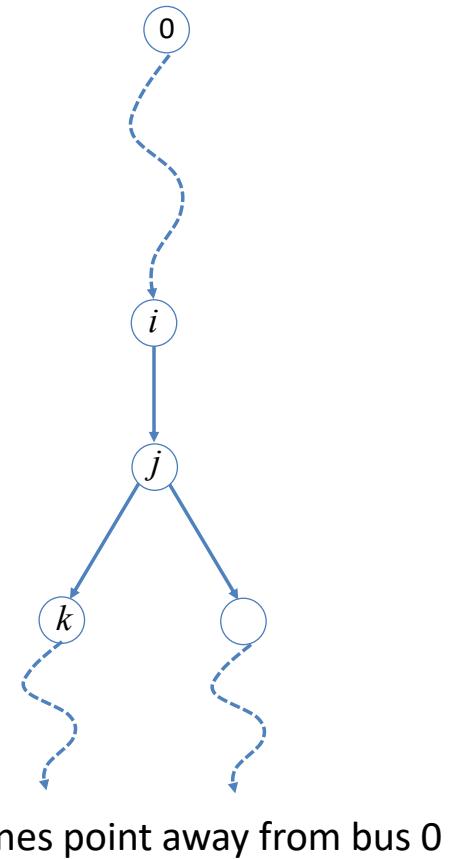
$$\sum_{k:j \rightarrow k} S_{jk} = S_{ij} - z_{ij}\ell_{ij} + s_j \quad \text{power balance}$$

$$v_j - v_k = 2 \operatorname{Re} \left( z_{jk}^H S_{jk} \right) - |z_{jk}|^2 \ell_{jk} \quad \text{Ohm's law, KCL (magnitude)}$$

$$v_j \ell_{jk} = |S_{jk}|^2 \quad \text{branch power magnitude}$$

All equations are **linear** in  $x$ , except the **quadratic** equalities

$$v_j \ell_{jk} = |S_{jk}|^2$$



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# Equivalence

## Recap

Bus injection model

- General networks: complex form, polar form, Cartesian form

Branch flow model

- General networks: complex form, real form
- Radial networks: with / without shunt admittances

All these models are **equivalent**

- In what sense?
- They consist of different equations with different variables in different domains

# Equivalence

## Solution set

Bus injection model

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

Solution set

$$\mathbb{V} := \{(s, V) \in \mathbb{C}^{2(n+1)} \mid V \text{ satisfies BIM}\}$$

# Equivalence

## Solution set

Branch flow models: solution sets

$$\tilde{\mathbb{X}} := \{\tilde{x} : (s, V, I, S) \in \mathbb{C}^{2(N+1)+4M} \mid \tilde{x} \text{ satisfies BFM complex}\}$$

$$\mathbb{X}_{\text{meshed}} := \{x : (s, v, \ell, S) \in \mathbb{R}^{3(N+1)+6M} \mid x \text{ satisfies BFM real}\}$$

$$\mathbb{X}_{\text{tree}} := \{x : (s, v, \ell, S) \in \mathbb{R}^{9N+3} \mid x \text{ satisfies BFM radial}\}$$

$$\mathbb{T}_0 := \{x : (s, v, \ell, S) \in \mathbb{R}^{6N+3} \mid x \text{ satisfies BFM radial zero } y_{jk}^m\}$$

Definition: Two sets  $A$  and  $B$  are equivalent ( $A \equiv B$ ) if there is a bijection between them

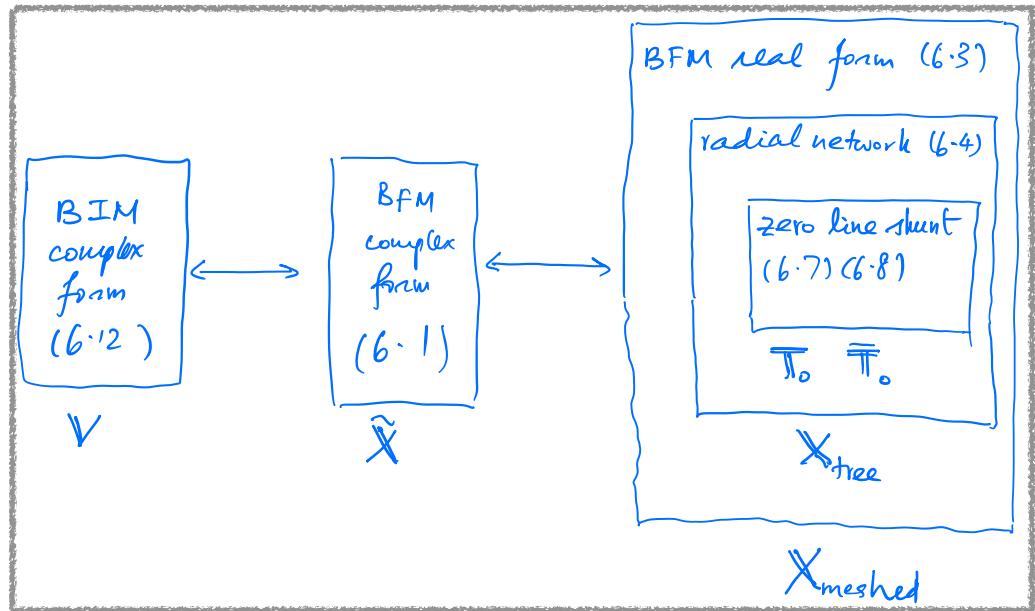
# Equivalence

## Solution set

### Theorem

Suppose  $G$  is connected

1.  $\mathbb{V} \equiv \tilde{\mathbb{X}} \equiv \mathbb{X}_{\text{meshed}}$
2. If  $G$  is a tree, then  $\mathbb{X}_{\text{meshed}} \equiv \mathbb{X}_{\text{tree}}$
3. If  $G$  is a tree and  $y_{jk}^m = y_{kj}^m = 0$ , then  $\mathbb{X}_{\text{tree}} \equiv \mathbb{T}_0 \equiv \hat{\mathbb{T}}_0$



# Equivalence

Bus injection models and branch flow models are equivalent

- Any result proved in one model holds also in another model

Some results are easier to formulate / prove in one model than the other

- BIM: semidefinite relaxation of OPF (later)
- BFM: some exact relation proofs

Should freely use whichever is more convenient for problem at hand

BFM is particularly suitable for modeling distribution systems

- Tree topology allows efficient computation of power flows (BFS)
- Seems to be much more numerically stable than BIM for large networks
- Models and relaxations extend to unbalanced  $3\phi$  networks

# Outline

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  - For radial networks
5. Linearized model

# Backward forward sweep

Efficient solution method for power flow equations

- Applicable for radial networks

Partition solution  $(x, y)$  into two groups of variables  $x$  and  $y$

Each round of spatial iteration consists of a backward sweep and a forward sweep

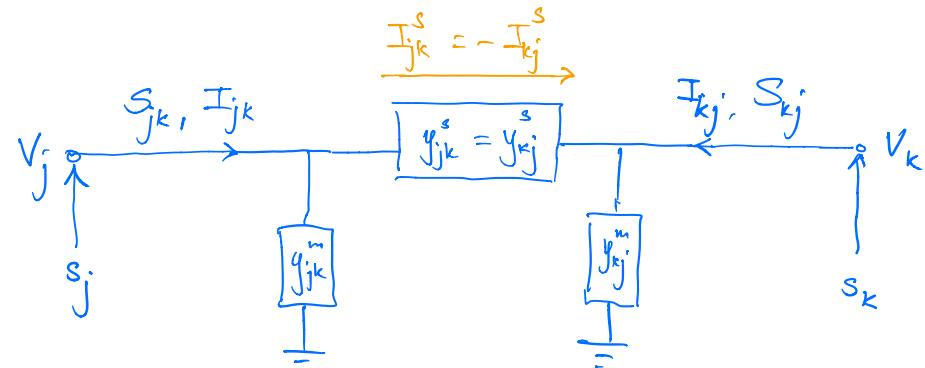
- Given  $y$ , compute each component  $x_j$  iteratively from leafs to root (backward)
- Given  $x$ , compute each component  $y_j$  iteratively from root to leaves (forward)

Iterate until stopping criterion

Different BFS methods differ in how to partition variables into  $x$  and  $y$  and the associated power flow equations

# Backward forward sweep

## Example



Use complex form BFM

Given:  $V_0$  and  $s := (s_j, j \in N)$

Compute:  $V := (V_j, j \in N)$  and currents  $I^s := (I_{jk}^s, (j, k) \in E)$  through series impedance

- All other variables  $I_{jk} = I_{jk}^s + y_{jk}^m V_j$ ,  $I_{kj}$ ,  $S_{jk}$ ,  $S_{kj}$  can then be computed
- Can also compute  $V_j$  and  $I_{jk}$  instead (exercise)
- Advantage:  $I_{jk}^s = -I_{kj}^s$

D. Shirmohammadi, H. W. Hong, A. Semlyen, and G. X. Luo. A compensation-based power flow method for weakly meshed distribution and transmission networks. *IEEE Transactions on Power Systems*, 3(2):753–762, May 1988.

# Backward forward sweep

## Example

Power flow equation

$$s_j = V_j I_{ji}^H + \sum_{k:j \rightarrow k} V_j I_{jk}^H = V_j \left( \left( I_{ji}^s + y_{ji}^m V_j \right)^H + \sum_{k:j \rightarrow k} \left( I_{jk}^s + y_{jk}^m V_j \right)^H \right)$$

Substitute  $I_{kj}^s = -I_{jk}^s$  to write all vars in direction of line  $j \rightarrow k$ :

$$\left( \frac{s_j}{V_j} \right)^H = -I_{ij}^s + y_{jj}^m V_j + \sum_{k:j \rightarrow k} I_{jk}^s$$

where  $y_{jj}^m := \sum_k y_{jk}^m$

# Backward forward sweep

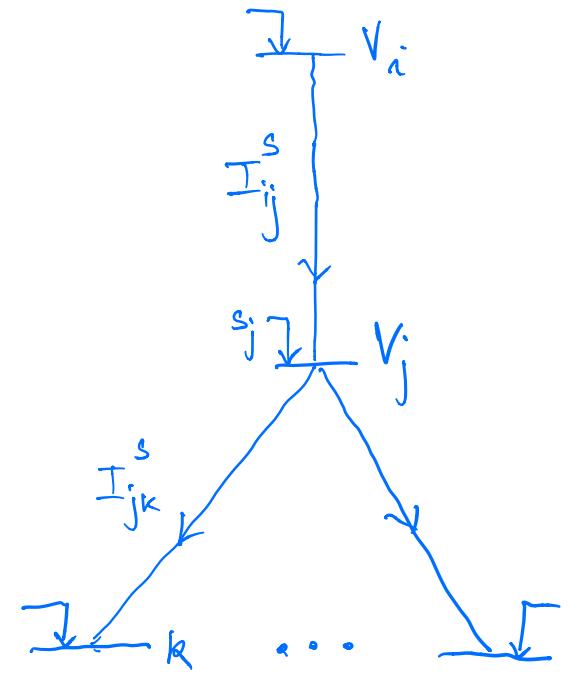
## Example

Rewrite in **spatially recursive** structure

$$I_{ij}^s = \sum_{k:j \rightarrow k} I_{jk}^s - \left( \left( \frac{s_j}{V_j} \right)^H - y_{jj}^m V_j \right)$$

Spatial iteration: propagating from leafs towards root (bus 0) in reverse BFS order

- Given all voltages  $V := (V_j, j \in \bar{N})$
- Given all currents  $I_{jk}^s$  in previous layer
- Compute currents  $I_{ij}^s$  in current layer



# Backward forward sweep

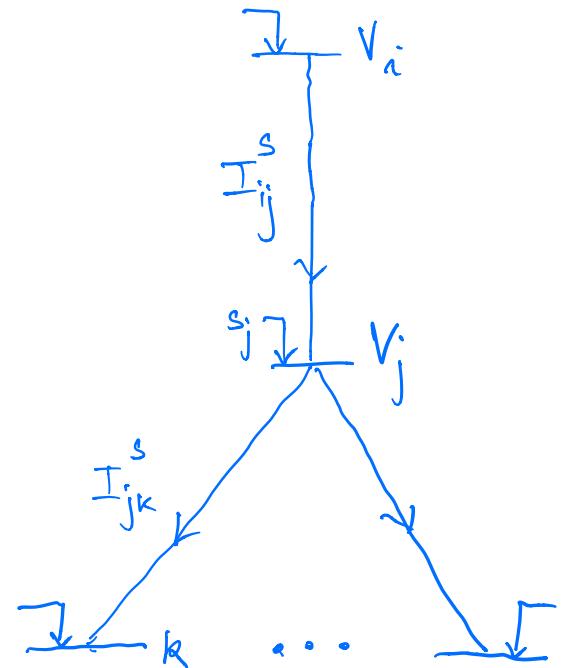
## Example

Write in **spatially recursive structure**

$$V_j = V_i - z_{ij}^s I_{ij}^s,$$

Spatial iteration: propagating from root (bus 0) towards leafs in BFS order

- Given all currents  $I^s := \left( I_{ij}^s, (i, j) \in E \right)$
- Given all voltages  $V_i$  in previous layer
- Compute voltages  $V_j$  in current layer



# Backward forward sweep

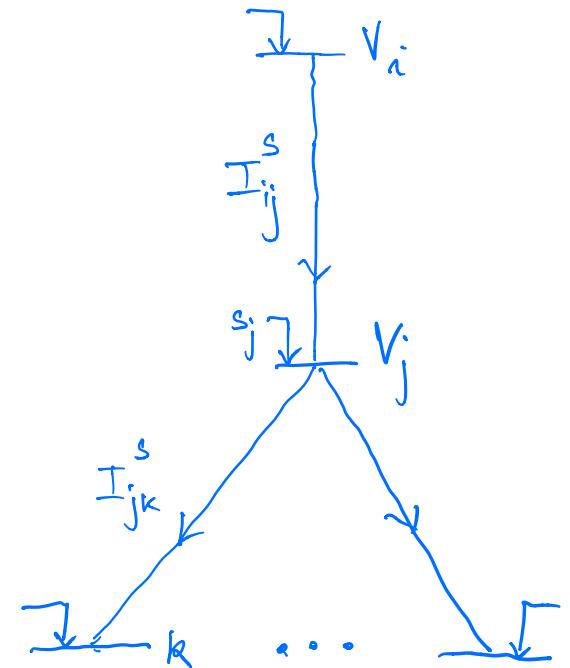
## Summary

*Input:* voltage  $V_0 = 1$  pu and injections ( $s_i, i \in N$ ).

*Output:* currents ( $I_{jk}^s, j \rightarrow k \in E$ ) and voltages ( $V_i, i \in N$ ).

### 1. Initialization.

- $V_0(t) := 1$  pu at bus  $j = 0$  for all iterations  $t = 1, 3, \dots$
- $V_j(0) := 1$  pu at all buses  $j \in N$  for iteration  $t = 0$ .



# Backward forward sweep

## Summary

2. Backward forward sweep: iterate  $t = 1, 3, 5, \dots$  till stopping criterion

(a) *Backward sweep.* Starting from the leaf nodes and working towards bus 0, compute

$$I_{ij}^s(t) \leftarrow \sum_{k:j \rightarrow k} I_{jk}^s(t) - \left( \left( \frac{s_j}{V_j(t-1)} \right)^H - y_{jj}^m V_j(t-1) \right), \quad i \rightarrow j \in E$$

(b) *Forward sweep.* Starting front bus 0 and working towards the leaf nodes, compute

$$V_j(t+1) = V_i(t+1) - z_{ij}^s I_{ij}^s(t), \quad j \in N$$

3. Output:  $I_{jk}^s := I_{jk}^s(t)$ ,  $V_i := V_i(t+1)$

# Backward forward sweep

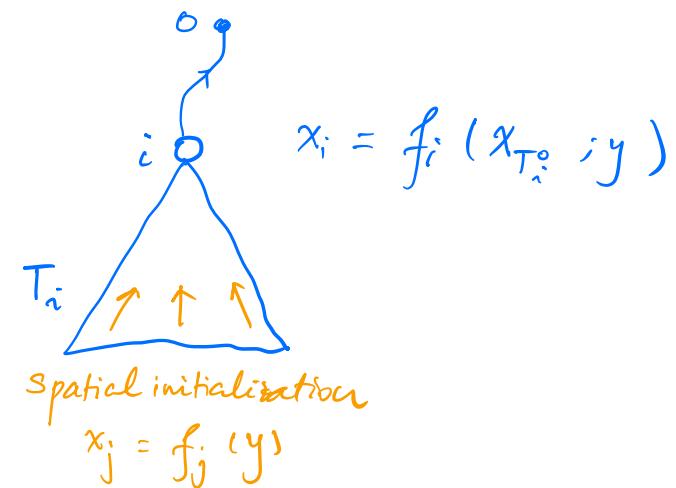
## General formulation

Backward sweep: let

- $T_i^\circ := \{\text{buses in subtree rooted at } i, \text{ excluding } i\}$
- $x_{T_i^\circ} := (x_j, j \in T_i^\circ)$

$x$  satisfies a **spatially recursive** structure if

$$x_i = f_i(x_{T_i^\circ}; y)$$



# Backward forward sweep

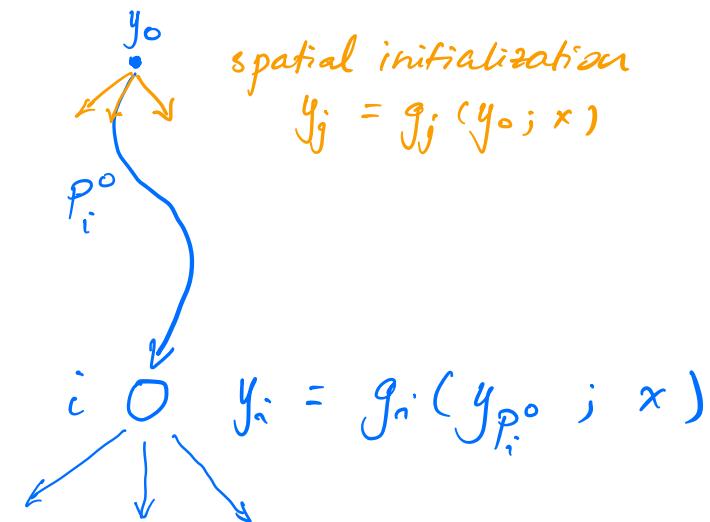
## General formulation

Forward sweep: let

- $P_i^\circ := \{\text{buses in path from root to } i, \text{ inc. 0 but exc. } i\}$
- $y_{P_i^\circ} := (y_j, j \in P_i^\circ)$

$y$  satisfies a **spatially recursive** structure if

$$y_i = g_i(y_{P_i^\circ}; x)$$



# Backward forward sweep

## General formulation

**while** stopping criterion not met **do**

(a)  $t \leftarrow t + 1; y_0(t) \leftarrow y_0;$

(b) *Backward sweep: for i starting from the leaf nodes and iterating towards bus 0 do*

$$x_i(t) \leftarrow f_i \left( x_{\mathsf{T}_i^\circ}(t); y(t-1) \right), \quad i \in \overline{N}$$

(c) *Forward sweep: for i starting front bus 0 and iterating towards the leaf nodes do*

$$y_i(t+1) \leftarrow g_i \left( y_{\mathsf{P}_i^\circ}(t+1); x(t) \right), \quad i \in N$$

# **Backward forward sweep**

**Open question**

Convergence analysis

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4. Backward forward sweep
5. Linearized model
  - Analytical solution
  - Bounds on nonlinear solutions
  - Application: decentralized volt/var control

# Linearized model

## Linear DistFlow equations

For radial networks

Set  $y_{jk}^m = y_{kj}^m = 0$ ,  $\ell_{jk} = 0$

Linear DistFlow equations [Baran-Wu 1989]

$$\sum_{k:j \rightarrow k} S_{jk} = \sum_{i:i \rightarrow j} S_{ij} + s_j$$
$$v_j - v_k = 2 \operatorname{Re} \left( z_{jk}^H S_{jk} \right)$$

# Linearized model

## Linear DistFlow equations

In vector form:

bus-by-line incidence matrix

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

Linear DistFlow equations

$$\begin{aligned} s &= CS \\ C^T v &= 2(D_r P + D_x Q) \end{aligned}$$

where  $D_r := \text{diag}(r_l, l \in E)$ ,  $D_x := \text{diag}(x_l, l \in E)$

$$\sum_{k:j \rightarrow k} S_{jk} = \sum_{i:i \rightarrow j} S_{ij} + s_j$$
$$v_j - v_k = 2 \operatorname{Re} \left( z_{jk}^H S_{jk} \right)$$

# Linearized model

## Linear DistFlow equations

Linear DistFlow can be solved explicitly

Given:  $v_0 = 1 \text{ pu}$ , injection  $\hat{s} := (s_j, j \in N)$

Determine: line power  $S := (S_{jk}, j \rightarrow k \in E)$ , voltage  $\hat{v} := (v_j, j \in N)$ , injection  $s_0$

Key observation: **Reduced** incidence matrix has full rank

$G$  connected  $\implies (N + 1) \times N$  incidence matrix  $C$  has rank  $N$

Decompose  $C =: \begin{bmatrix} -c_0^T & - \\ \hat{C} & \end{bmatrix}$

$G$  tree topology  $\implies N \times N$  **reduced** incidence matrix  $\hat{C}$  is invertible

# Linearized model

## Linear DistFlow solution

Linear DistFlow:

$$\begin{aligned}\hat{s} &= \hat{C}S \\ v_0 c_0 + \hat{C}^T \hat{v} &= 2(D_r P + D_x Q)\end{aligned}$$

and  $s_0 = c_0^T S$

$$\begin{aligned}s &= CS \\ C^T v &= 2(D_r P + D_x Q)\end{aligned}$$

Solution:

$$S = \hat{C}^{-1} \hat{s}$$

$$\hat{v} = v_0 \mathbf{1} + 2(R\hat{p} + X\hat{q}) \quad \text{voltages} = v_0 + \text{correction term } (\hat{p}, \hat{q})$$

where  $R := \hat{C}^{-T} D_r \hat{C}^{-1}$ ,  $X := \hat{C}^{-T} D_x \hat{C}^{-1}$  are positive definite matrices

# Linearized model

## Bounds on nonlinear solution

### Corollary

Fix  $v_0$  and injections  $\hat{s} \in \mathbb{R}^{2N}$  at non-reference buses. Let  $(v, \ell, S)$  and  $(v^{\text{lin}}, \ell^{\text{lin}}, S^{\text{lin}})$  be a solution of nonlinear and linear DistFlow equations respectively (in the down direction).

1.  $S_{ij} \geq S_{ij}^{\text{lin}}$
2.  $v_j \leq v_j^{\text{lin}}$

Linear DistFlow ignores line losses and underestimates required power to supply loads

# Outline

1. General network
2. Radial network
3. Equivalence
4. Backward forward sweep
5. Linearized model
  - Analytical solution
  - Bounds on nonlinear solutions
  - Application: decentralized volt/var control

# Linearized model

## Application: volt/var control

Volv/var control: control reactive power injections  $q$  to stabilize voltages  $\hat{v}$

How should  $q$  adapt as voltages fluctuate?

# Linearized model

## Application: volt/var control

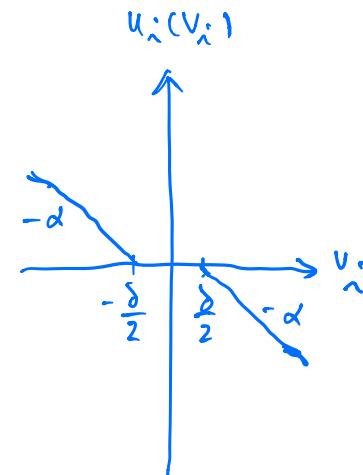
Local memoryless feedback control:

$$q_j(t+1) = \left[ u_j \left( v_j(t) - v_j^{\text{ref}} \right) \right]_{U_j} \quad U_j := \{ q_j : \underline{q}_j \leq q_j \leq \bar{q}_j \}$$

Adapt reactive power  $q_i(t)$  to drive voltage  $v_j(t)$  towards target  $v_j^{\text{ref}}$

Control  $q_j(t+1)$  depends only on:

- Feedback: measured system state  $v(t)$
- Memoryless: latest voltage  $v(t)$  at time  $t$ , not history  $v(s)$ ,  $s < t$
- Local: local voltage  $v_j(t)$  at bus  $j$ , not other voltages  $v_k(t)$



# Linearized model

## Application: volt/var control

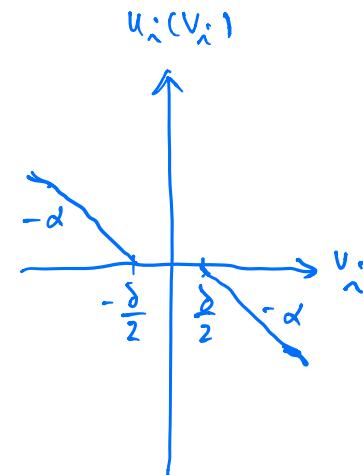
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Adapt reactive power  $q_i(t)$  to drive voltage  $v_j(t)$  towards target  $v_j^{\text{ref}}$

How does the closed-loop system behave ?

- under this simple local control
- if network is described by Linear DistFlow



# Linearized model

## Application: volt/var control

Linear DistFlow model describes how voltages (linearly) depend on control  $q$ :

$$v(q) = v_0 \mathbf{1} + 2(Rp + Xq) = 2Xq + \tilde{v}$$

where  $\tilde{v} := v_0 \mathbf{1} + 2Rp$

Since

$$\frac{\partial v_j}{\partial q_j} = 2X_{jj} = \sum_{(i,k) \in P_j} x_{ik} > 0$$

it justifies choosing  $u_j$  to be a decreasing function of  $v_j(t) - v_j^{\text{ref}}$

# Linearized model

## Application: volt/var control

Assume measured voltage is given by Linear DistFlow, i.e.,  $v_j(t) = v_j(q(t))$

Closed-loop system is discrete-time dynamical system:

$$q_j(t+1) = \left[ u_j \left( v_j(q(t)) - v_j^{\text{ref}} \right) \right]_{U_j}$$

where  $v(q) = 2Xq + \tilde{v}$

Definition:  $q^*$  is a **fixed point** or **equilibrium point** if  $q^* = \left[ u \left( v(q^*) - v^{\text{ref}} \right) \right]_{U_j}$

# Linearized model

## Application: volt/var control

Assume measured voltage is given by Linear DistFlow, i.e.,  $v_j(t) = v_j(q(t))$

Closed-loop system is discrete-time dynamical system:

$$q_j(t+1) = \left[ u_j \left( v_j(q(t)) - v_j^{\text{ref}} \right) \right]_{U_j}$$

where  $v(q) = 2Xq + \tilde{v}$

What are convergence and optimality properties of closed-loop system ?

# Linearized model

## Application: volt/var control

Closed-loop system is discrete-time dynamical system:

$$q_j(t+1) = \left[ u_j \left( v_j(q(t)) - v_j^{\text{ref}} \right) \right]_{U_j}$$

where  $v(q) = 2Xq + \tilde{v}$

### Assumptions

- $u_j$  are differentiable and  $|u'_j(v_j)| \leq \alpha_j$
- $u_j$  are strictly decreasing

# Linearized model

## Application: volt/var control

**Theorem** (Convergence)

If largest singular value  $\sigma_{\max}(AX) < 1/2$  then

$$A := \text{diag} \left( \alpha_j, j \in N \right)$$

1.  $\exists !$  equilibrium point  $q^* \in U$
2. Closed-loop system converges to  $q^*$  geometrically, i.e.,

$$\|q(t) - q^*\| \leq \beta^t \|q(0) - q^*\| \rightarrow 0$$

for some  $\beta \in (0,1)$

# Linearized model

## Application: volt/var control

**Theorem** (Optimality)

The unique equilibrium point  $q^* \in U$  solves

$$\min_{q \in U} \sum_j c_j(q_j) + q^T X q + q^T \Delta \tilde{v}$$

where  $c_j(q_j) := - \int_0^{q_j} u_j^{-1}(\hat{q}_j) d\hat{q}_j$  and  $\Delta \tilde{v} := \tilde{v} - v^{\text{ref}}$

Reverse engineering: by choosing a control function  $u_j$ , we implicitly choose a cost function  $c_j(q_j)$  that the closed-loop equilibrium optimizes

# Linearized model

## Application: volt/var control

### Theorem (Optimality)

The unique equilibrium point  $q^* \in U$  solves

$$\min_{q \in U} \sum_j c_j(q_j) + q^T X q + q^T \Delta \tilde{v}$$

where  $c_j(q_j) := - \int_0^{q_j} u_j^{-1}(\hat{q}_j) d\hat{q}_j$  and  $\Delta \tilde{v} := \tilde{v} - v^{\text{ref}}$

Forward engineering: Choose a cost function  $c_j(q_j)$  and derive control functions  $u_j$  as distributed algorithm to solve the optimization problem

# Summary

1. General network
  - Complex form, real form
2. Radial network
  - With and without shunt admittances
3. Equivalence
4. Backward forward sweep
5. Linearized model
  - Analytical solution, bounds, local volt/var control