

ECE 5314: Power System Operation & Control

Lecture 7: Power Flow Problem

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- R2 A. Gomez-Exposito, A. J. Conejo, C. Canizares, *Electric Energy Systems: Analysis and Operation*, Chapter 3.
- R1 A. J. Wood, B. F. Wollenberg, and G. B. Sheble, *Power Generation, Operation, and Control*, Wiley, 2014, Chapter 6.

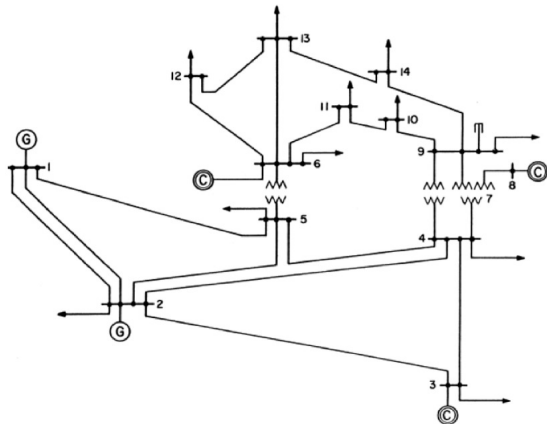
Power transmission network as an electric circuit

- N nodes (generator/load buses) and L edges (lines, transformers)

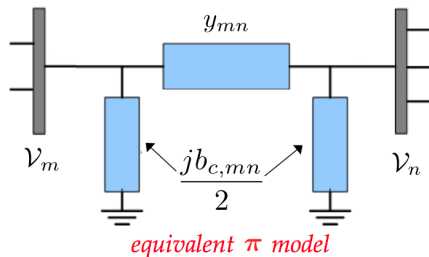
- AC voltages and currents as phasors (at nominal frequency)

$$\mathcal{V} = V e^{j\theta} = V_r + jV_i$$

- From scalar to multivariate Ohm's law: $\mathcal{V} = Z\mathcal{I} \rightarrow \mathbf{v} = \mathbf{Z}\mathbf{i}$



Transmission lines

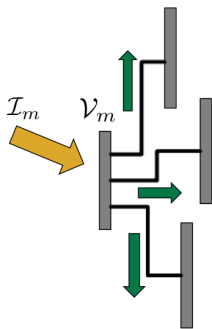
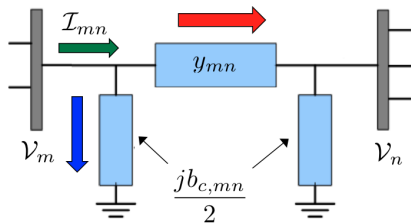


- Line series impedance: $z_{mn} = r_{mn} + jx_{mn}$ ($x_{mn} > 0$)
- Line series admittance: $y_{mn} = \frac{1}{z_{mn}} = g_{mn} - jb_{mn}$
- Line series conductance: $g_{mn} = \frac{r_{mn}}{r_{mn}^2 + x_{mn}^2}$
- Line series susceptance: $b_{mn} = \frac{x_{mn}}{r_{mn}^2 + x_{mn}^2} > 0$
- Total charging susceptance: $b_{mn}^c > 0$

Line currents

$$\mathcal{I}_{mn} = y_{mn}(\mathcal{V}_m - \mathcal{V}_n) + j\frac{b_{c,mn}^c}{2}\mathcal{V}_m$$

$$\mathcal{I}_{mn} = \left(y_{mn} + j\frac{b_{c,mn}^c}{2}\right)\mathcal{V}_m - y_{mn}\mathcal{V}_n$$



Kirchoff's current law:

$$\mathcal{I}_m = \left(\sum_{n \neq m} y_{mn} + j\frac{b_{c,mn}^c}{2}\right)\mathcal{V}_m - \sum_{n \neq m} y_{mn}\mathcal{V}_n$$

Collect currents and voltages $\{\mathcal{I}_m, \mathcal{V}_m\}_{m=1}^N$ into $\mathbf{i}, \mathbf{v} \in \mathbb{C}^{N \times 1}$

Transformers and phase shifters are ignored in our analysis

Multivariate Ohm's law

Currents are linearly related to voltages: $\mathbf{i} = \mathbf{Y}\mathbf{v}$

Bus admittance matrix: fundamental in power systems operations

$$Y_{mn} = \begin{cases} \sum_{k \neq m} y_{mk} + j \frac{b_{mk}^c}{2} & , m = n \\ -y_{mn} & , \exists \text{ line } (m, n) \\ 0 & , \text{ otherwise} \end{cases}$$

- symmetric ($Y_{mn} = Y_{nm}$); non-Hermitian ($Y_{mn} \neq Y_{nm}^*$)
- sparse: efficient computations and storage
- invertible if $b_{mn}^c \neq 0$ for at least one line; otherwise $\mathbf{Y}\mathbf{1} = \mathbf{0}$

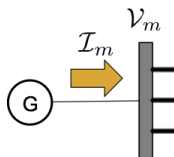
Bus impedance matrix: $\mathbf{Z} := \mathbf{Y}^{-1}$ ($\mathbf{v} = \mathbf{Z}\mathbf{i}$)

- non-sparse
- *not* the matrix of line impedances, i.e., $Z_{mn} \neq z_{mn} = \frac{1}{y_{mn}}$

Complex power

- Power $\mathcal{S}_m = \mathcal{S}_m^g - \mathcal{S}_m^d$ generated/consumed at bus m

$$\{\mathcal{S}_m = P_m + jQ_m = \mathcal{V}_m \mathcal{I}_m^*\}_{m=1}^N, \text{ and } \mathbf{i} = \mathbf{Y}\mathbf{v}$$



- Eliminate currents to get the multivariate power model

$$\mathbf{s} = \text{diag}(\mathbf{v})\mathbf{i}^* = \text{diag}(\mathbf{v})\mathbf{Y}^*\mathbf{v}^*$$

N complex equations in $2N$ complex unknowns

- Bus admittance matrix in rectangular coordinates $\mathbf{Y} = \mathbf{G} + j\mathbf{B}$
- Similar expressions for power flow on line (m, n) : $\mathcal{S}_{mn} = \mathcal{V}_m \mathcal{I}_{mn}^*$

Power flow equations

Voltages in **polar coordinates** ($\theta_{mn} = \theta_m - \theta_n$)

$$P_m = V_m \sum_{n=1}^N V_n (G_{mn} \cos \theta_{mn} + B_{mn} \sin \theta_{mn})$$

$$Q_m = V_m \sum_{n=1}^N V_n (G_{mn} \sin \theta_{mn} - B_{mn} \cos \theta_{mn})$$

dependence on phase differences only; reference bus $\theta_N = 0$

Voltages in **rectangular coordinates** (quadratic equations!)

$$P_m = V_{m,r} \sum_{n=1}^N (V_{n,r} G_{mn} - V_{n,i} B_{mn}) + V_{m,i} \sum_n (V_{n,i} G_{mn} + V_{n,r} B_{mn})$$

$$Q_m = V_{m,i} \sum_{n=1}^N (V_{n,r} G_{mn} - V_{n,i} B_{mn}) - V_{m,r} \sum_n (V_{n,i} G_{mn} + V_{n,r} B_{mn})$$

Power flow problem

There are $2N$ equations and $4N$ variables $\{(P_m, Q_m, V_m, \theta_m)\}_{m=1}^N$

Problem statement: Fixing the values of $2N$ variables, find the values of the rest $2N$ unknowns that satisfy the nonlinear power flow (PF) equations

Given values typically come from

- First N_d load buses (PQ buses) (P_m, Q_m)
- Next N_g generator buses (PV buses) (P_m, V_m)
- Reference bus $(V_N, \theta_N = 0)$

Number of buses $N = 1 + N_g + N_d$

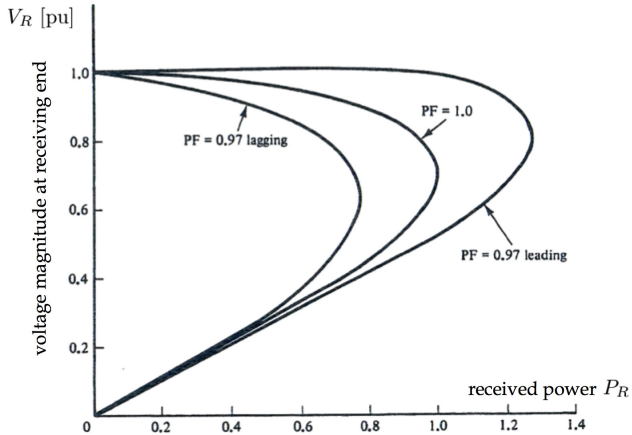
Solving the power flow equations

$$P_m = V_m \sum_n V_n (G_{mn} \cos \theta_{mn} + B_{mn} \sin \theta_{mn}), \quad m = 1, \dots, N_d + N_g = N - 1$$

$$Q_m = V_m \sum_n V_n (G_{mn} \sin \theta_{mn} - B_{mn} \cos \theta_{mn}), \quad m = 1, \dots, N_d$$

- Set of nonlinear equations in $\{(V_n, \theta_n)\}_{n=1}^N$ solved recursively
- Once voltages $\{(V_n, \theta_n)\}_{n=1}^N$ are found, any other quantity (injections, flows, currents, losses) can be calculated
- **Flat start or flat voltage profile**
voltages usually initialized at $V_n = 1$ and $\theta_n = 0$ for all n
- PF solution is not unique!

'Nose' curve



Q: How is the famous nose curve derived?

Gauss-Seidel method

$$\begin{aligned} \mathcal{S}_m &= \mathcal{V}_m \sum_{n=1}^N Y_{mn}^* \mathcal{V}_n^* \Rightarrow \mathcal{S}_m^* = \mathcal{V}_m^* \sum_{n=1}^N Y_{mn} \mathcal{V}_n \Rightarrow \\ \left(\frac{\mathcal{S}_m}{\mathcal{V}_m} \right)^* &= \sum_{n=1}^{m-1} Y_{mn} \mathcal{V}_n + Y_{mm} \mathcal{V}_m + \sum_{n=m+1}^N Y_{mn} \mathcal{V}_n \Rightarrow \\ \mathcal{V}_m &:= \frac{1}{Y_{mm}} \left[\left(\frac{\mathcal{S}_m}{\mathcal{V}_m} \right)^* - \sum_{n=1}^{m-1} Y_{mn} \mathcal{V}_n - \sum_{n=m+1}^N Y_{mn} \mathcal{V}_n \right] \end{aligned}$$

Gauss-Seidel iterations:

1. Initialize \mathbf{v}_0 at flat profile or at most recent grid state
2. Repeat until convergence $\|\mathbf{v}_{t+1} - \mathbf{v}_t\|_2 \leq \epsilon$

$$\mathcal{V}_m^{t+1} := \frac{1}{Y_{mm}} \left[\left(\frac{\mathcal{S}_m^t}{\mathcal{V}_m^t} \right)^* - \sum_{n=1}^{m-1} Y_{mn} \mathcal{V}_n^{t+1} - \sum_{n=m+1}^N Y_{mn} \mathcal{V}_n^t \right], \quad \forall m$$

where \mathcal{S}_m^t is either fixed or calculated from PF equations via \mathbf{v}_t

3. Normalize \mathcal{V}_m^{t+1} to match given magnitude for PV buses

Impedance matrix method

Power flow problem involves two equations that can be combined:

$$\left. \begin{array}{l} \mathbf{s} = \text{diag}(\mathbf{v})\mathbf{i}^* \quad \Leftrightarrow \quad \mathbf{i} = [\text{diag}(\mathbf{v}^*)]^{-1}\mathbf{s}^* \\ \mathbf{i} = \mathbf{Y}\mathbf{v} \quad \Leftrightarrow \quad \mathbf{v} = \mathbf{Z}\mathbf{i} \end{array} \right\} \Rightarrow \mathbf{v} = \mathbf{Z}[\text{diag}(\mathbf{v}^*)]^{-1}\mathbf{s}^*$$

Jacobi-type iterations:

1. Initialize \mathbf{v}_0 at flat profile or at most recent grid state
2. Repeat until convergence $\|\mathbf{v}_{t+1} - \mathbf{v}_t\|_2 \leq \epsilon$

$$\mathbf{v}_{t+1} = \mathbf{Z}[\text{diag}(\mathbf{v}_t^*)]^{-1}\mathbf{s}_t^*$$

where entries of \mathbf{s}_t are either known or calculated from PF equations via \mathbf{v}_t

Inversion of \mathbf{Y} ; (close to) singularity of \mathbf{Y} handled by eliminating the slack bus

Newton's method

- Newton-Raphson method aims at solving nonlinear equations: $\mathbf{f}(\mathbf{x}) = \mathbf{0}$
- At iteration $t + 1$, function $\mathbf{f}(\mathbf{x})$ is linearized at \mathbf{x}^t

$$\mathbf{f}(\mathbf{x}) \approx \hat{\mathbf{f}}(\mathbf{x}) = \mathbf{f}(\mathbf{x}^t) + \mathbf{J}(\mathbf{x}^t)(\mathbf{x} - \mathbf{x}^t)$$

where $\mathbf{J}(\mathbf{x}^t)$ is the Jacobian matrix of \mathbf{f}

- Variable \mathbf{x}^{t+1} is updated such that $\hat{\mathbf{f}}(\mathbf{x}^{t+1}) = \mathbf{0}$

$$\mathbf{x}^{t+1} := \mathbf{x}^t - [\mathbf{J}(\mathbf{x}^t)]^{-1}\mathbf{f}(\mathbf{x}^t)$$

- Newton's method in two steps (convergence to be studied later)

$$-\mathbf{J}(\mathbf{x}^t)\boldsymbol{\delta}^t = \mathbf{f}(\mathbf{x}^t) \quad \text{system of linear equations}$$

$$\mathbf{x}^{t+1} := \mathbf{x}^t + \boldsymbol{\delta}^t$$

Power flow via Newton's method

Equations involved in power flow problem:

$$\Delta P_m := \hat{P}_m - V_m \sum_k V_k (G_{mk} \cos \theta_{mk} + B_{mk} \sin \theta_{mk}) = 0, \quad m = 1, \dots, N_d + N_g$$

$$\Delta Q_m := \hat{Q}_m - V_m \sum_k V_k (G_{mk} \sin \theta_{mk} - B_{mk} \cos \theta_{mk}) = 0, \quad m = 1, \dots, N_d$$

or more compactly

$$\mathbf{f}(\mathbf{x}) = \begin{bmatrix} \Delta \mathbf{p}(\mathbf{x}) \\ \Delta \mathbf{q}(\mathbf{x}) \end{bmatrix} = \mathbf{0}$$

Variables involved in power flow problem

$$\mathbf{x} := \begin{bmatrix} \boldsymbol{\theta} \\ \mathbf{v} \end{bmatrix} = [\theta_1 \ \theta_2 \ \dots \ \theta_{N-1} \ V_1 \ V_2 \ \dots \ V_{N_d}]^\top$$

For Jacobian in NR, need to find: $-\frac{\partial \Delta P_m}{\partial \theta_n} = -\frac{\partial (\hat{P}_m - P_m)}{\partial \theta_n} = \frac{\partial P_m}{\partial \theta_n}$

Finding derivatives $\partial P_m / \partial \theta_n$

Repeating for convenience: $P_m = V_m \sum_k V_k (G_{mk} \cos \theta_{mk} + B_{mk} \sin \theta_{mk})$

1. For $n \neq m$, we get

$$\frac{\partial P_m}{\partial \theta_n} = V_m V_n (G_{mn} \sin \theta_{mn} - B_{mn} \cos \theta_{mn})$$

2. Notice similarity to summands in Q_m

$$Q_m = V_m \sum_{k \neq m} V_k (G_{mk} \sin \theta_{mk} - B_{mk} \cos \theta_{mk}) - B_{mm} V_m^2$$

3. For $n = m$, we get

$$\frac{\partial P_m}{\partial \theta_m} = -V_m \sum_{k \neq m} V_k (G_{mk} \sin \theta_{mk} - B_{mk} \cos \theta_{mk}) = -Q_m - B_{mm} V_m^2$$

Finding derivatives $\partial Q_m / \partial V_n$

Repeating for convenience: $Q_m = V_m \sum_k V_k (G_{mk} \sin \theta_{mk} - B_{mk} \cos \theta_{mk})$

1. For $n \neq m$, we get

$$\begin{aligned}\frac{\partial Q_m}{\partial V_n} &= V_m (G_{mn} \sin \theta_{mn} - B_{mn} \cos \theta_{mn}) \implies \\ V_n \frac{\partial Q_m}{\partial V_n} &= V_m V_n (G_{mn} \sin \theta_{mn} - B_{mn} \cos \theta_{mn}) = \frac{\partial P_m}{\partial \theta_n}\end{aligned}$$

2. For $n = m$, we get

$$V_m \frac{\partial Q_m}{\partial V_m} = Q_m - B_{mm} V_m^2$$

Multiplying $\partial Q_m / \partial V_n$ by V_n gives Jacobian matrix more symmetry

Finding derivatives $\partial Q_m / \partial \theta_n$

Repeating for convenience: $Q_m = V_m \sum_k V_k (G_{mk} \sin \theta_{mk} - B_{mk} \cos \theta_{mk})$

1. For $n \neq m$, we have

$$\frac{\partial Q_m}{\partial \theta_n} = -V_m V_n (G_{mn} \cos \theta_{mn} + B_{mn} \sin \theta_{mn})$$

2. Notice similarity to summands in P_m

$$P_m = V_m \sum_{k \neq m} V_k (G_{mk} \cos \theta_{mk} + B_{mk} \sin \theta_{mk}) + G_{mm} V_m^2$$

3. For $n = m$, we get

$$\frac{\partial Q_m}{\partial \theta_m} = V_m \sum_{k \neq m} V_k (G_{mk} \cos \theta_{mk} + B_{mk} \sin \theta_{mk}) = P_m - G_{mm} V_m^2$$

Finding derivatives $\partial P_m / \partial V_n$

Repeating for convenience: $P_m = V_m \sum_k V_k (G_{mk} \cos \theta_{mk} + B_{mk} \sin \theta_{mk})$

1. For $n \neq m$, we have

$$V_n \frac{\partial P_m}{\partial V_n} = V_m V_n (G_{mn} \cos \theta_{mn} + B_{mn} \sin \theta_{mn}) = -\frac{\partial Q_m}{\partial \theta_n}$$

2. For $n = m$, we get

$$\begin{aligned} V_m \frac{\partial P_m}{\partial V_m} &= V_m \sum_{k \neq m} V_k (G_{mk} \cos \theta_{mk} + B_{mk} \sin \theta_{mk}) + 2G_{mm} V_m^2 \\ &= P_m + G_{mm} V_m^2 \end{aligned}$$

Blocks of Jacobian matrix

$$\mathbf{H}_{(N-1) \times (N-1)}^t : H_{mn} = \frac{\partial P_m}{\partial \theta_n} = \begin{cases} V_m V_n (G_{mn} \sin \theta_{mn} - B_{mn} \cos \theta_{mn}), & n \neq m \\ -Q_m - B_{mm} V_m^2, & n = m \end{cases}$$

$$\mathbf{N}_{(N-1) \times N_d}^t : N_{mn} = V_n \frac{\partial P_m}{\partial V_n} = \begin{cases} -M_{mn}, & n \neq m \\ P_m + G_{mm} V_m^2, & n = m \end{cases}$$

$$\mathbf{M}_{N_d \times (N-1)}^t : M_{mn} = \frac{\partial Q_m}{\partial \theta_n} = \begin{cases} -V_m V_n (G_{mn} \cos \theta_{mn} + B_{mn} \sin \theta_{mn}), & n \neq m \\ P_m - G_{mm} V_m^2, & n = m \end{cases}$$

$$\mathbf{L}_{N_d \times N_d}^t : L_{mn} = V_n \frac{\partial Q_m}{\partial V_n} = \begin{cases} H_{mn}, & n \neq m \\ Q_m - B_{mm} V_m^2, & n = m \end{cases}$$

Newton's iterations

1. Initialize \mathbf{v}_0 at flat profile or at a recent grid state
2. For $t = 0, 1, \dots$, until convergence $\|\mathbf{v}_{t+1} - \mathbf{v}_t\|_2 \leq \epsilon$

2.1 Evaluate Jacobian matrix at current state

2.2 Find variable update by solving the linear system

$$\begin{bmatrix} \mathbf{H}^t & \mathbf{N}^t \\ \mathbf{M}^t & \mathbf{L}^t \end{bmatrix} \begin{bmatrix} \Delta\boldsymbol{\theta}^t \\ \Delta\mathbf{v}^t/\mathbf{v}^t \end{bmatrix} = \begin{bmatrix} \Delta\mathbf{p}^t \\ \Delta\mathbf{q}^t \end{bmatrix}$$

where division by \mathbf{v}^t (known at iteration t) is for symmetry

2.3 Update the state as

$$\begin{bmatrix} \boldsymbol{\theta}^{t+1} \\ \mathbf{v}^{t+1} \end{bmatrix} := \begin{bmatrix} \boldsymbol{\theta}^t \\ \mathbf{v}^t \end{bmatrix} + \begin{bmatrix} \Delta\boldsymbol{\theta}^t \\ \Delta\mathbf{v}^t \end{bmatrix}$$

Fast decoupled power flow

Newton's iterations involve evaluating the Jacobian matrix and inverting it

Two approximations to save computations:

1. Keep the Jacobian constant by evaluating it at a specific point \mathbf{x}
2. Problem decouples by setting $\mathbf{M} = \mathbf{N} = \mathbf{0}$
3. After several approximations, matrices \mathbf{H} and \mathbf{L} simplify as

$$\mathbf{B}' \Delta \boldsymbol{\theta}^t = \Delta \mathbf{p}^t / \mathbf{v}^t$$

$$\mathbf{B}'' \Delta \mathbf{v}^t = \Delta \mathbf{q}^t / \mathbf{v}^t$$

Matrices \mathbf{B}' and \mathbf{B}'' are defined as

$$B'_{mn} = -x_{mn}^{-1}, \quad B'_{mm} = \sum_{n \neq m} x_{mn}^{-1} \quad (b_{mn} \approx x_{mn}^{-1}, \text{ no shunt, no voltage trans.})$$

$$B''_{mn} = -B_{mn}, \quad B''_{mm} = -B_{mm} \quad (b_{mn} = \frac{x_{mn}}{r_{mn}^2 + x_{mn}^2}, \text{ no phase shifters})$$

Specifications as quadratic functions

- Collect nodal voltages in rectangular coordinates in $\mathbf{v} \in \mathbb{C}^N$:

$$\mathbf{v} := [v_{1,r} + jv_{1,i} \quad \dots \quad v_{N,r} + jv_{N,i}]^T$$

- Power injections and squared voltage magn. are quadratic functions of \mathbf{v} :

$$P_m(\mathbf{v}) = \mathbf{v}^H \mathbf{M}_{P_m} \mathbf{v}$$

$$Q_m(\mathbf{v}) = \mathbf{v}^H \mathbf{M}_{Q_m} \mathbf{v}$$

$$V_m^2(\mathbf{v}) = \mathbf{v}^H \mathbf{M}_{V_m} \mathbf{v}$$

where matrices in blue are Hermitian symmetric ($\mathbf{M}_{P_m} = \mathbf{M}_{P_m}^H$)

- Every bus contributes two quadratic constraints/specifications on \mathbf{v}

Finding \mathbf{M} 's matrices

Voltage magnitude (\mathbf{e}_m is the m -th canonical vector)

$$V_m^2(\mathbf{v}) = \mathcal{V}_m^* \mathcal{V}_m = \mathbf{v}^H \mathbf{e}_m \mathbf{e}_m^\top \mathbf{v} \quad \Rightarrow \quad \mathbf{M}_{V_m} = \mathbf{e}_m \mathbf{e}_m^\top$$

Complex power injection

$$S_m = \mathcal{V}_m \mathcal{I}_m^* = (\mathbf{v}^\top \mathbf{e}_m)(\mathbf{e}_m^\top \mathbf{i}^*) = \mathbf{v}^\top \mathbf{e}_m \mathbf{e}_m^\top \mathbf{Y}^* \mathbf{v}^* = \mathbf{v}^H \mathbf{Y}^* \mathbf{e}_m \mathbf{e}_m^\top \mathbf{v}$$

Active power

$$P_m = \frac{S_m + S_m^*}{2} = \mathbf{v}^H \mathbf{M}_{P_m} \mathbf{v} \quad \text{where} \quad \mathbf{M}_{P_m} = \frac{1}{2} \left(\mathbf{Y}^* \mathbf{e}_m \mathbf{e}_m^\top + \mathbf{e}_m \mathbf{e}_m^\top \mathbf{Y} \right)$$

Reactive power

$$Q_m = \frac{S_m - S_m^*}{2j} = \mathbf{v}^H \mathbf{M}_{Q_m} \mathbf{v} \quad \text{where} \quad \mathbf{M}_{Q_m} = \frac{1}{2j} \left(\mathbf{Y}^* \mathbf{e}_m \mathbf{e}_m^\top - \mathbf{e}_m \mathbf{e}_m^\top \mathbf{Y} \right)$$

Power flow as a feasibility problem

- System state as solution of feasibility problem

find \mathbf{v}

$$\text{s.to } \mathbf{v}^H \mathbf{M}_k \mathbf{v} = s_k, \quad k = 1 : 2N \quad \left[\text{note } \mathbf{v}^H \mathbf{M}_k \mathbf{v} = \text{Tr}(\mathbf{M}_k \mathbf{v} \mathbf{v}^H) \right]$$

- Introduce matrix variable $\mathbf{V} = \mathbf{v} \mathbf{v}^H$

find (\mathbf{v}, \mathbf{V})

$$\text{s.to } \text{Tr}(\mathbf{M}_k \mathbf{V}) = s_k, \quad k = 1 : 2N$$

$$\mathbf{V} = \mathbf{v} \mathbf{v}^H$$

- Eliminate variable \mathbf{v} ; *non-convex* problem due to rank constraint

find \mathbf{V}

$$\text{s.to } \text{Tr}(\mathbf{M}_k \mathbf{V}) = s_k, \quad k = 1 : 2N$$

$$\mathbf{V} \succeq \mathbf{0}, \quad \text{rank}(\mathbf{V}) = 1$$

Semidefinite program relaxation

- Drop rank constraint to get semidefinite program (SDP)

find \mathbf{V}

s.to $\text{Tr}(\mathbf{M}_k \mathbf{V}) = s_k, \quad k = 1, \dots, 2N$

$\mathbf{V} \succeq \mathbf{0}$

which is a **convex problem**

- If the solution \mathbf{V}_o is rank-one, the relaxation is said to be *exact*
- If exact, find \mathbf{v}_o from $\mathbf{V}_o = \mathbf{v}_o \mathbf{v}_o^H$
- Relaxation is oftentimes exact under practical system conditions!

From feasibility to minimization

- Feasibility problem can be converted to the convex minimization problem

$$\begin{aligned} \min_{\mathbf{V} \succeq \mathbf{0}} \quad & \text{Tr}(\mathbf{M}\mathbf{V}) \\ \text{s.to} \quad & \text{Tr}(\mathbf{M}_k \mathbf{V}) = s_k, \quad k = 1, \dots, 2N \end{aligned}$$

- Design matrix \mathbf{M} so that rank-one solutions are favored
 - selecting $\mathbf{M} = \mathbf{Y}^H \mathbf{Y}$ minimizes $\|\mathbf{i}\|_2^2$
 - selecting $\mathbf{M} = \mathbf{B}$ minimizes losses
 - both yield the “high-voltage solution” of the power flow equations

R. Madani, J. Lavaei, and R. Baldick, “Convexification of power flow problem over arbitrary networks,” in *Proc. IEEE Conf. on Decision and Control*, Dec. 2015, Osaka, Japan.

DC power flow model

Power flow equations

$$P_m = V_m \sum_n V_n (G_{mn} \cos \theta_{mn} + B_{mn} \sin \theta_{mn})$$

Assumptions:

A1. Low r/x ratios in transmission lines (1/5-1/10 for 220-400kV)

$$r_{mn} \ll x_{mn} \rightarrow g_{mn} \ll b_{mn} \rightarrow \mathbf{G} \simeq \mathbf{0} \quad \text{and} \quad b_{mn} = \frac{x_{mn}}{r_{mn}^2 + x_{mn}^2}$$

A2. Small angle differences $\theta_m - \theta_n \simeq 0$; $\cos \theta_{mn} \simeq 1$ and $\sin \theta_{mn} \simeq \theta_{mn}$

A3. Voltage magnitudes close to unity (pu) $V_m \simeq 1$

DC power flow model:

[why called 'DC' ?]

$$P_m \simeq \sum_{n \neq m} b_{mn} (\theta_m - \theta_n)$$

Coincides with 1st-order Taylor's series of P_m at \mathbf{v}_{flat} under A1.

B matrix

Power injections (and flows) relate linearly to phase differences

$$P_m = \sum_{n:n \sim m} P_{mn} = \sum_{n:n \sim m} b_{mn}(\theta_m - \theta_n)$$

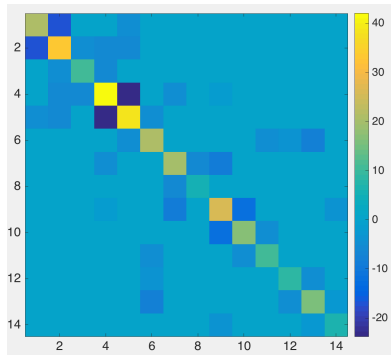
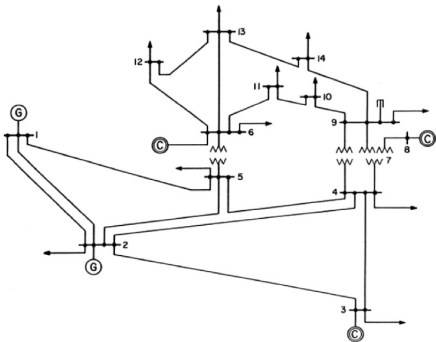
Multivariate power flow model: $\mathbf{p} = \mathbf{B}\boldsymbol{\theta}$

DC bus admittance matrix: (different from matrix \mathbf{B} in $\mathbf{Y} = \mathbf{G} + j\mathbf{B}$)

$$B_{mn} = \begin{cases} \sum_{n \neq m} b_{mn} & , m = n \\ -b_{mn} & , \exists \text{ line } (m, n) \\ 0 & , \text{ otherwise} \end{cases}$$

- Real; symmetric; sparse; and positive semidefinite [Q: Why?]
- Lossless lines: $\mathbf{B}\mathbf{1}_N = \mathbf{0}_N \Rightarrow \mathbf{p}^T \mathbf{1}_N = 0$
- Oftentimes further simplify $b_{mn} = \frac{x_{mn}}{r_{mn}^2 + x_{mn}^2} \simeq \frac{1}{x_{mn}}$

Example for the IEEE 14-bus system



```
c = loadcase('case14'); % load case file
B = makeBdc(c); % B in sparse form; use B = full(B) if full form needed
imagesc(B);
axis square;
```