



# Optimizing Dynamic Power Flow

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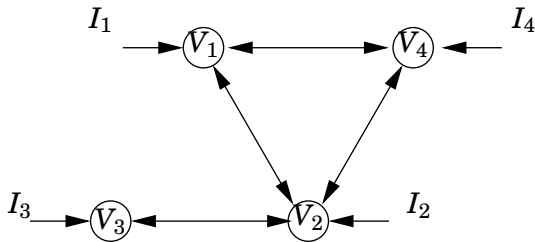


- Combine with water power reservoirs in northern Sweden
- Use wind farms to stabilize network
- AEOLUS project: Distributed coordination of wind turbines

# Outline

- **Problem Statements**
  - Positive Quadratic Programming
  - Optimizing Static Power Flow
  - Dynamic Positive Programming
  - Optimizing Dynamic Power Flow

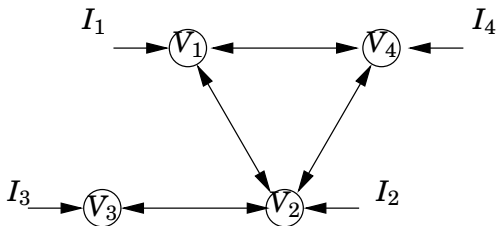
# A Power Transmission Network



$$\underbrace{\begin{bmatrix} I_1(s) \\ I_2(s) \\ I_3(s) \\ I_4(s) \end{bmatrix}}_{I(s)} = \underbrace{\begin{bmatrix} Y_{12} + Y_{14} & -Y_{12} & 0 & -Y_{14} \\ -Y_{21} & Y_{21} + Y_{23} + Y_{24} & -Y_{23} & -Y_{24} \\ 0 & -Y_{32} & Y_{32} & 0 \\ -Y_{41} & -Y_{42} & 0 & Y_{41} + Y_{42} \end{bmatrix}}_{Y(s)} \underbrace{\begin{bmatrix} V_1(s) \\ V_2(s) \\ V_3(s) \\ V_4(s) \end{bmatrix}}_{V(s)}$$

Potential differences drive currents (voltage\*current = power)  
Price differences drive commodity flows (price\*amount = value)

# An Optimal Flow Problem for AC Power



$$I_k \in \mathbf{C}$$

$$V_k \in \mathbf{C}$$

Minimize

$$\operatorname{Re} \sum_k I_k^* V_k$$

subject to  $I = YV$  and  $\underline{P}_k \leq \operatorname{Re} (I_k^* V_k) \leq \overline{P}_k$

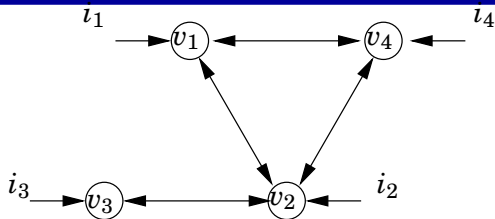
$$\underline{Q}_k \leq \operatorname{Im} (I_k^* V_k) \leq \overline{Q}_k$$

$$\underline{v}_k \leq |V_k| \leq \overline{v}_k$$

for  $k = 1, \dots, 4$

(Convex relaxation by Lavaei/Low inspired this talk.)

# Problem I: Optimizing Static Power Flow



$$i_k \in \mathbf{R}$$

$$v_k \in \mathbf{R}$$

Minimize

$$\sum_k i_k v_k$$

subject to  $i = Yv$  and

$$i_k v_k \leq \bar{p}_k$$

$$(v_k - v_j)^2 \leq c_{kj}$$

$$\underline{v}_k \leq v_k \leq \bar{v}_k \quad \text{for all } k, j$$

Notice:  $\bar{p}_k$  negative at loads, positive at generators.

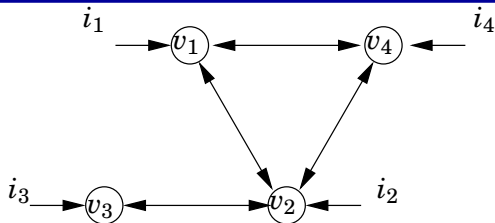
Motivation: 1) Real DC networks. 2) Approximation of AC.

3) Water tanks. 4) Supply chains

# Further questions regarding Problem I

- Are there distributed solution algorithms?
- Will market mechanisms find the optimum?
- Optimize transition when demand changes! (Problem II)

## Problem II: Optimizing Dynamic Power Flow



$$i_k(t) \in \mathbf{R}$$

$$v_k(t) \in \mathbf{R}$$

Minimize

subject to  $I(s) = Y(s)V(s)$  and

$$\sum_k \int_0^\infty i_k(t)v_k(t)dt$$

$$i_k(t)v_k(t) \leq \bar{p}_k$$

$$|v_k(t) - v_j(t)|^2 \leq c_{kj}$$

$$\underline{v}_k \leq v_k(t) \leq \bar{v}_k \quad \text{for all } k, j$$

Convexly solvable when off-diagonal elements of  $Y(s)$  have non-negative impulse response! (e.g. ramp dynamics)



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- **Positive Quadratic Programming**
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- Dynamic Positive Programming
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# Positive Quadratic Programming

Given  $A_0, \dots, A_K \in \mathbf{R}^{n \times n}$  with nonnegative off-diagonal entries and  $b_1, \dots, b_K \in \mathbf{R}$ , the following equality holds:

$$\begin{array}{ll} \max & x^T A_0 x \\ \text{subject to} & x \in \mathbf{R}_+^n \\ & x^T A_k x \geq b_k \\ & k = 1, \dots, K \end{array} = \begin{array}{ll} \max & \text{trace}(A_0 X) \\ \text{subject to} & X \succeq 0 \\ & \text{trace}(A_k X) \geq b_k \\ & k = 1, \dots, K \end{array}$$

Proof

If  $X = \begin{bmatrix} |x_1|^2 & & * \\ & \ddots & \\ * & & |x_n|^2 \end{bmatrix}$  maximizes the right hand side,

then  $x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$  maximizes the left.

[Kim/Kojima, 2003]

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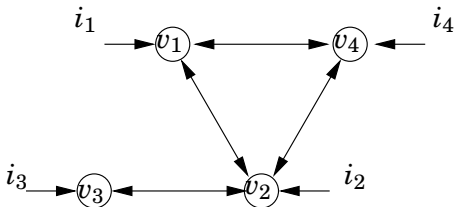
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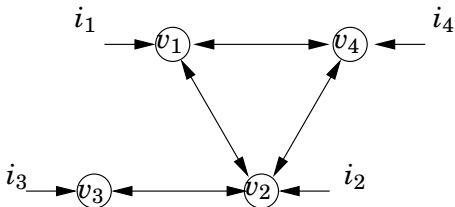
# An Optimal Flow Problem for DC Power



Minimize  $i_3 v_3 + i_4 v_4$

subject to  $i = Yv$  and  $i_1 v_1 \leq \bar{p}_1$   
 $i_2 v_2 \leq \bar{p}_2$   
 $\underline{v}_k \leq v_k \leq \bar{v}_k$  for  $k = 1, \dots, 4$

# An Optimal Flow Problem for DC Power



Minimize  $(-y_{32}v_2 + y_{32}v_3)v_3 + (-y_{41}v_1 - y_{42}v_2 + y_{41}v_4 + y_{42}v_4)v_4$

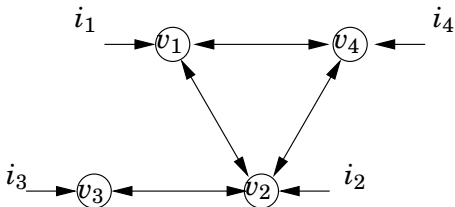
subject to  $(y_{12}v_1 + y_{14}v_1 - y_{12}v_2 - y_{14}v_4)v_1 \leq \bar{p}_1$

$(-y_{21}v_1 + y_{21}v_2 + y_{23}v_2 + y_{24}v_2 - y_{23}v_3 - y_{24}v_4)v_2 \leq \bar{p}_2$

$|\underline{v}_k|^2 \leq |v_k|^2 \leq |\bar{v}_k|^2$

**Note:** The problem is convex in  $|v_1|^2, \dots, |v_4|^2$ !

# An Optimal Flow Problem for DC Power



$$\begin{aligned} \text{Minimize} \quad & (-y_{32}v_2 + y_{32}v_3)v_3 + (-y_{41}v_1 - y_{42}v_2 + y_{41}v_4 + y_{42}v_4)v_4 \\ \text{subject to} \quad & (y_{12}v_1 + y_{14}v_1 - y_{12}v_2 - y_{14}v_4)v_1 \leq \bar{p}_1 \\ & (-y_{21}v_1 + y_{21}v_2 + y_{23}v_2 + y_{24}v_2 - y_{23}v_3 - y_{24}v_4)v_2 \leq \bar{p}_2 \\ & |\underline{v}_k|^2 \leq |v_k|^2 \leq |\bar{v}_k|^2 \end{aligned}$$

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# Dual Positive Quadratic Programming

Given  $A_0, \dots, A_K \in \mathbf{R}^{n \times n}$  with nonnegative off-diagonal entries and  $b_1, \dots, b_K \in \mathbf{R}$ , the following equality holds:

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## Interpretation:

In the power flow example,  $\lambda_k$  is the price of power at node  $k$ .



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## Distributed solution:

The agent at node  $k$  buying power over node  $j$  compares prices at both ends and adjusts for power losses in the link.

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## Positive systems have nonnegative impulse response

If the matrices  $A$ ,  $B$ ,  $C$  and  $D$  have nonnegative coefficients except for the diagonal of  $A$ , then the system

$$\begin{aligned}\frac{dx}{dt} &= Ax + Bu \\ y &= Cx + Du\end{aligned}$$

has non-negative impulse response.

**Example:**

$$\begin{aligned}L\frac{di}{dt} &= -Ri + v && \text{inductive transmission line} \\ y &= i\end{aligned}$$

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has non-negative impulse response.

**Example:**

$$\begin{aligned}\frac{dv}{dt} &= -\alpha v + u && \text{generator ramp dynamics} \\ y &= v\end{aligned}$$

## Positive systems

Suppose the matrices  $A$ ,  $B$ ,  $C$  and  $D$  have nonnegative coefficients except for the diagonal of  $A$ :

$$\begin{aligned}\frac{dx}{dt} &= Ax + Bu \\ y &= Cx + Du\end{aligned}$$

### Properties:

- Stability verified by linear or diagonal Lyapunov functions.
- Maximal gain for zero frequency:

$$\max_{\omega} \|C(i\omega I - A)^{-1}B + D\| = \|D - CA^{-1}B\|$$

# Dynamic Positive Programming

Let  $A_0(s), \dots, A_K(s)$  have off-diagonal entries with nonnegative impulse response and  $b_1, \dots, b_K \in \mathbf{R}$ . Then the following equality holds:

$$\begin{aligned} \max \quad & \int_{-\infty}^{\infty} x(i\omega)^* A_0(i\omega) x(i\omega) d\omega \\ \text{subject to} \quad & \int_{-\infty}^{\infty} x(i\omega)^* A_k(i\omega) x(i\omega) d\omega \geq b_k \\ & x \in \mathbf{H}_+^n, k = 1, \dots, K \end{aligned}$$

$$\begin{aligned} = \max \quad & \int_{-\infty}^{\infty} \text{trace}(A_0 X) d\omega \\ \text{subject to} \quad & \int_{-\infty}^{\infty} \text{trace}(A_k X) d\omega \geq b_k \\ & X(i\omega) \succeq 0, k = 1, \dots, K \end{aligned}$$

where  $\mathbf{H}_+^n$  consists of all stable transfer functions with nonnegative impulse response.

# Positive Quadratic Programming

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$$\begin{aligned} \max \quad & \int_{-\infty}^{\infty} x^* A_0 x d\omega & = & \max \quad \int_{-\infty}^{\infty} \text{trace}(A_0 X) d\omega \\ \text{subject to} \quad & x \in \mathbf{H}_+^n & & \text{subject to} \quad X \succeq 0 \\ & \int_{-\infty}^{\infty} x^* A_k x d\omega \geq b_k & & \int_{-\infty}^{\infty} \text{trace}(A_k X) d\omega \geq b_k \\ & k = 1, \dots, K & & k = 1, \dots, K \end{aligned}$$

## Proof

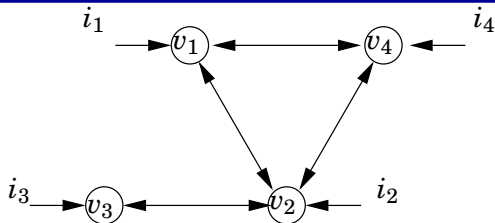
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Minimize

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Convexly solvable when off-diagonal elements of  $Y(s)$  have non-negative impulse response! (Inductive loads)

# Summary



- Positive Quadratic Programming
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