

# Graph-theoretic Algorithm for Nonlinear Power Optimization Problems

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# Optimization for Power Networks

## □ Optimizations:

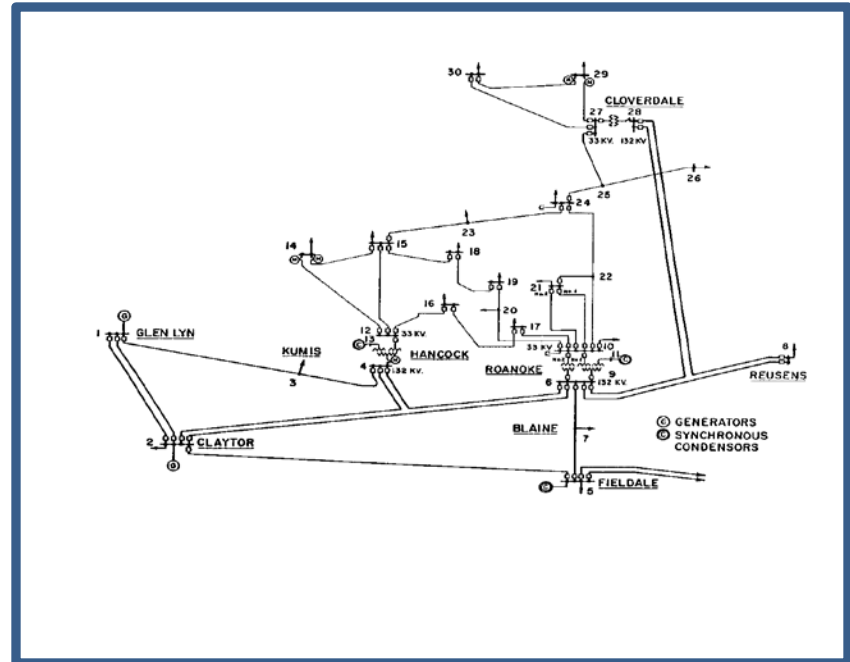
- Optimal power flow (OPF)
- Security-constrained OPF
- State estimation
- Network reconfiguration
- Unit commitment
- Dynamic energy management

## □ Issue of non-convexity:

- Discrete parameters
- Nonlinearity in continuous variables

## □ Transition from traditional grid to smart grid:

- More variables (10X)
- Time constraints (100X)



# Broad Interest in Power Optimization

- ❑ OPF-based problems solved on different time scales:
  - Electricity market
  - Real-time operation
  - Security assessment
  - Transmission planning
  
- ❑ Existing methods based on linearization or local search
  
- ❑ **Question:** How to find the best solution using a scalable robust algorithm?
  
- ❑ Huge literature since 1962 by power, OR and Econ people

# Rank-Constrained Optimization

- Consider a polynomial optimization

$$\begin{aligned} \min_{x \in \mathbb{R}^{n-1}} \quad & f_0(x) \\ \text{s.t.} \quad & f_k(x) \leq 0 \quad \text{for } k = 1, \dots, p \end{aligned}$$

- With no loss of generality, assume that  $f_k$ 's are quadratic.

- Consider the transformation:  $\mathbf{X} \triangleq \begin{bmatrix} 1 & x^T \end{bmatrix}^T \begin{bmatrix} 1 & x^T \end{bmatrix}$

- Rank-constrained reformulation:

$$\begin{aligned} \min_{\mathbf{X} \in \mathbb{S}^n} \quad & \text{trace}\{\mathbf{F}_0 \mathbf{X}\} \\ \text{s.t.} \quad & \text{trace}\{\mathbf{F}_k \mathbf{X}\} \leq 0 \quad \text{for } k = 1, \dots, p \\ & X_{11} = 1 \\ & \mathbf{X} \succeq 0 \\ & \text{rank}\{\mathbf{X}\} = 1. \end{aligned} \quad \left. \vphantom{\begin{aligned} \min_{\mathbf{X} \in \mathbb{S}^n} \quad & \text{trace}\{\mathbf{F}_0 \mathbf{X}\} \\ \text{s.t.} \quad & \text{trace}\{\mathbf{F}_k \mathbf{X}\} \leq 0 \quad \text{for } k = 1, \dots, p \\ & X_{11} = 1 \\ & \mathbf{X} \succeq 0 \\ & \text{rank}\{\mathbf{X}\} = 1. \end{aligned}} \right\} \longrightarrow \text{SDP relaxation}$$

# Outline

- Review of four previous projects on SDP relaxation
  
- New results:
  - ❖ Notion of complexity for power networks (treewidth)
  - ❖ Connection between rank and treewidth
  - ❖ Case studies:
    1. **IEEE 300 bus:** Rank for security constrained OPF  $\leq 7$
    2. **Polish 3120 bus:** Rank for security constrained OPF  $\leq 27$

# Projects 1-3

## **Project 1: How to solve a given OPF in polynomial time?** (joint work with Steven Low)

- SDP relaxation works for various systems (IEEE systems, etc.).
- This is due to passivity.

## **Project 2: Find network topologies over which optimization is easy?** (joint work with Somayeh Sojoudi, David Tse and Baosen Zhang)

- Distribution (acyclic) networks are fine (under certain assumptions).
- Transmission networks may need phase shifters.

## **Project 3: How to design a distributed algorithm for solving OPF?** (joint work with Stephen Boyd, Eric Chu and Matt Kranning)

- A practical (infinitely) parallelizable algorithm using ADMM.
- It solves 10,000-bus OPF in 0.85 seconds on a single core machine.

# Project 4

**Project 4: How to do optimization for mesh networks?** (joint work with Ramtin Madani and Somayeh Sojoudi)

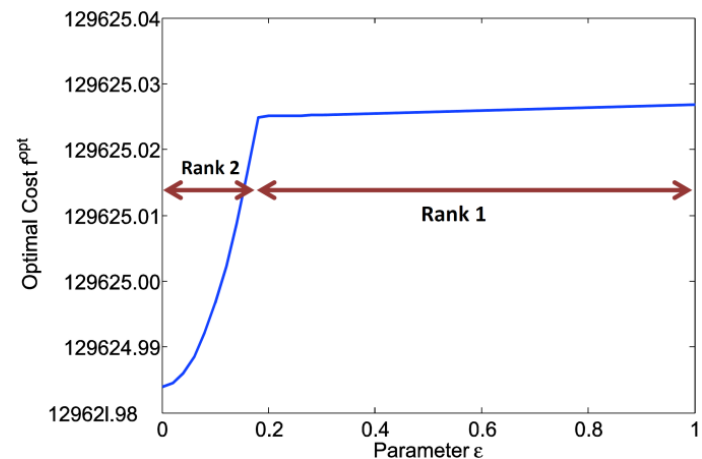
❑ Penalized SDP relaxation:

$$\sum_{k \in \mathcal{G}} f_k(P_{G_k}) \quad \longrightarrow \quad \sum_{k \in \mathcal{G}} f_k(P_{G_k}) + \varepsilon \sum_{k \in \mathcal{G}} Q_{G_k}$$

❑ Example borrowed from Bukhsh et al.:

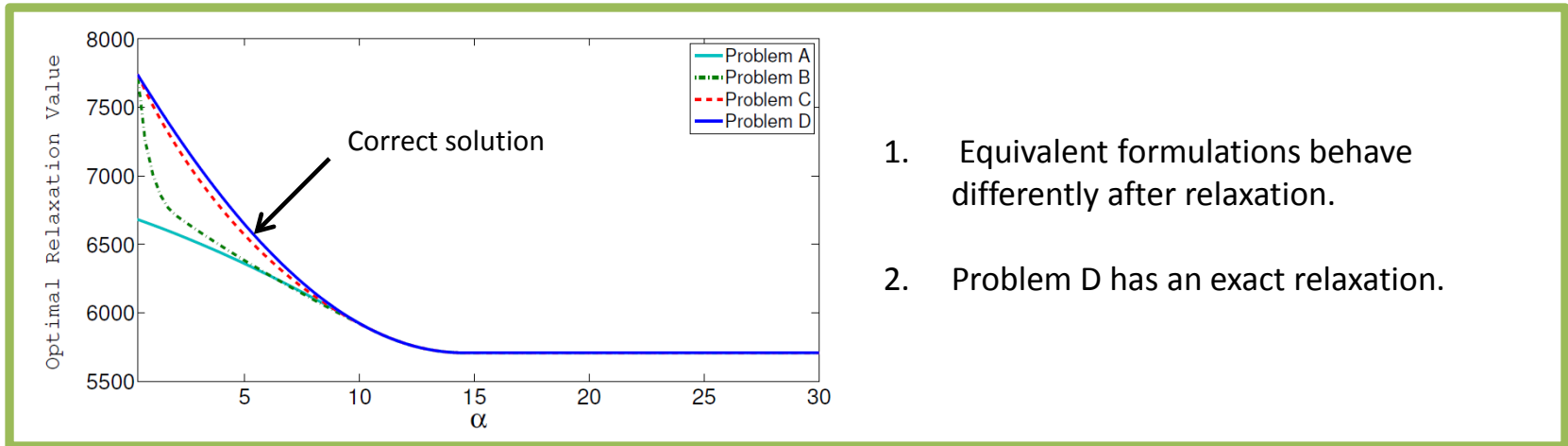
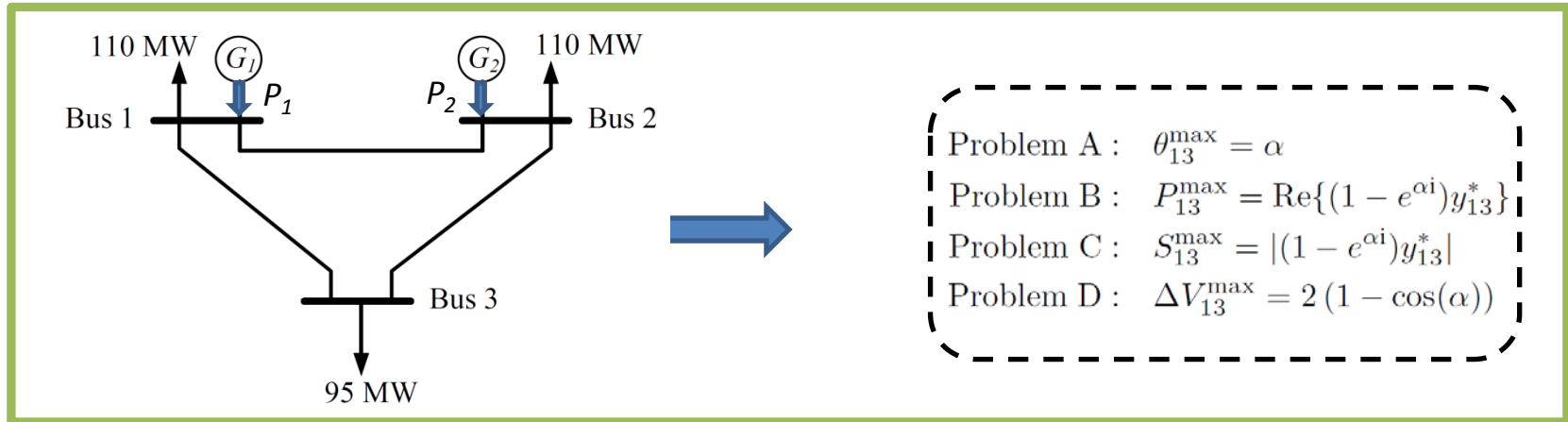
❑ Modify IEEE 118-bus system has 3 local solutions with the optimal costs 129625.03, 177984.32 and 195695.54.

❑ Our method finds the best one.



# Response of SDP to Equivalent Formulations

□ **Capacity constraint:** active power, apparent power, angle difference, voltage difference, current?

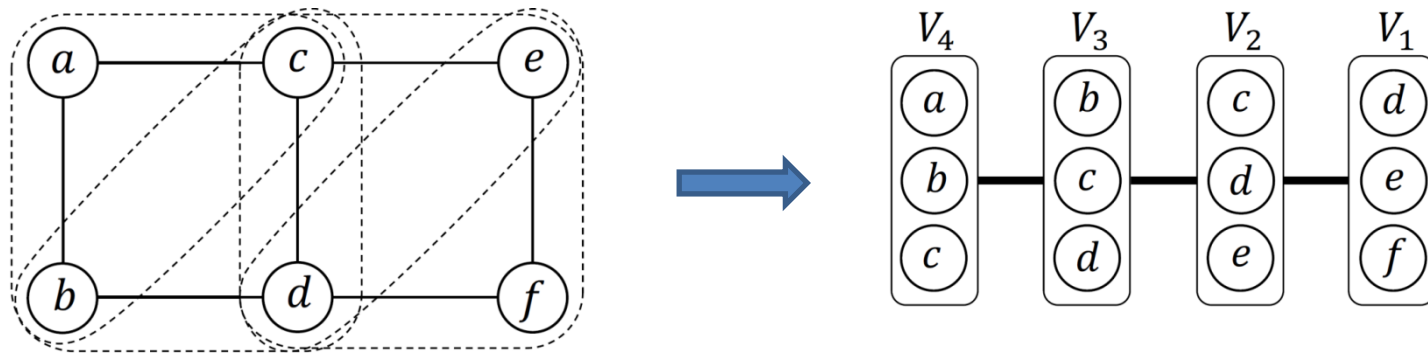


1. Equivalent formulations behave differently after relaxation.
2. Problem D has an exact relaxation.



# Treewidth

## □ Tree decomposition:



□ We map a given graph  $G$  into a tree  $T$  such that:

- ❖ Each node of  $T$  is a collection of vertices of  $G$
- ❖ Each edge of  $G$  appears in one node of  $T$
- ❖ If a vertex shows up in multiple nodes of  $T$ , those nodes should form a subtree

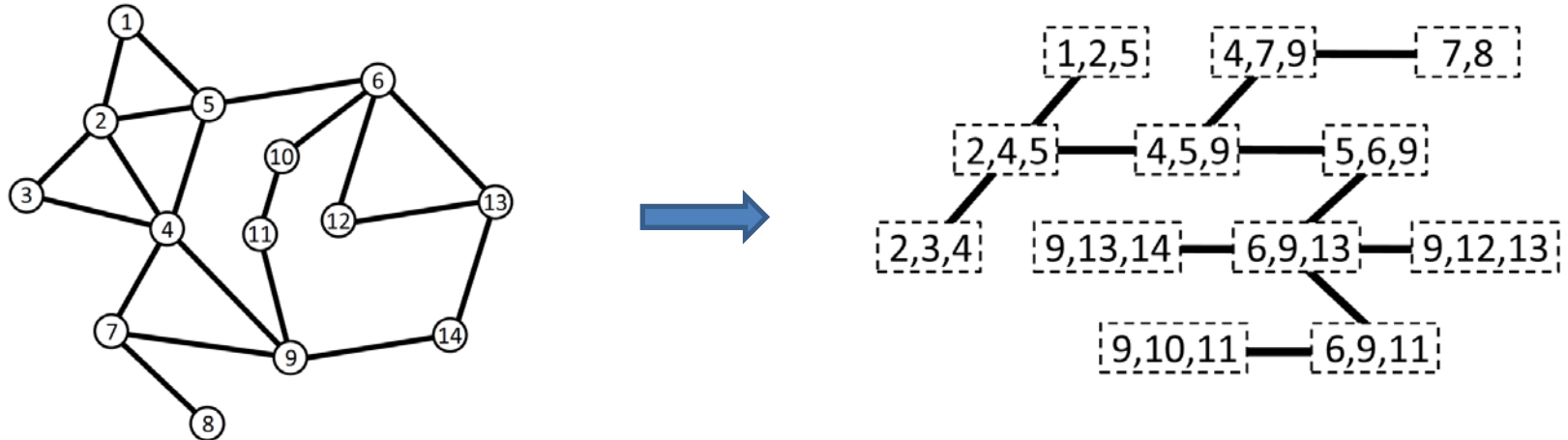
□ **Width of a tree decomposition:** The cardinality of largest node minus one

□ **Treewidth of graph:** The smallest width of all tree decompositions

# Power Networks

□ Treewidth of a tree: 1

□ How about the treewidth of IEEE 14-bus system with multiple cycles? 2



□ How to compute the treewidth of a large graph?

❖ NP-hard problem

❖ We used graph reduction techniques for sparse power networks

# Power Networks

□ Upper bound on the treewidth of sample power networks:

System $\mathcal{G}$	$\text{tw}\{\mathcal{G}\}$	System $\mathcal{G}$	Bound on $\text{tw}\{\mathcal{G}\}$
IEEE 14-bus	2	Polish 2383-bus	26
IEEE 30-bus	3	Polish 2736-bus	38
New England 39-bus	3	Polish 2746-bus	40
IEEE 57-bus	5	Polish 3012-bus	28
IEEE 118-bus	4	Polish 3120-bus	26
IEEE 300-bus	6	Polish 3375-bus	28

$$\min \quad \text{trace}\{\mathbf{F}_0 \mathbf{X}\}$$

$$\text{s.t.} \quad \text{trace}\{\mathbf{F}_k \mathbf{X}\} \leq 0 \quad \text{for } k = 1, \dots, p$$

$$X_{11} = 1$$

$$\mathbf{X} \succeq 0$$

Real/complex  
optimization



□ Define  $G$  as the sparsity graph

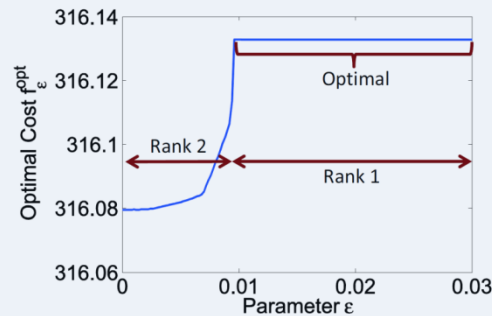
□ **Theorem:** There exists a solution with rank at most treewidth of  $G + 1$

□ We proposed infinitely many optimizations to find that solution.

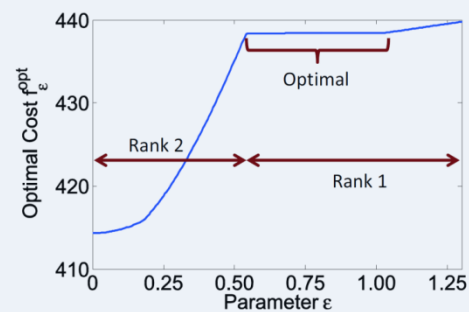
# Examples

- ❑ **Example:** Consider the security-constrained unit-commitment OPF problem.
- ❑ Use SDP relaxation for this mixed-integer nonlinear program.
- ❑ What is the rank of  $X^{opt}$ ?
  1. **IEEE 300-bus system:** rank  $\leq 7$
  2. **Polish 3120-bus system:** Rank  $\leq 27$

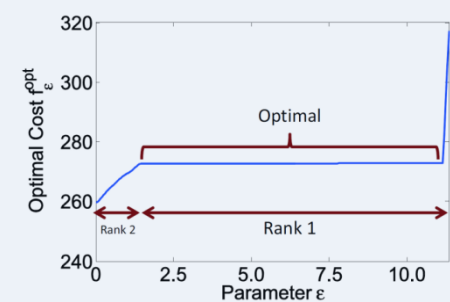
- ❑ How to go from low-rank to rank-1? Penalization (tested on 7000 examples)



IEEE 14-bus system



IEEE 30-bus system



IEEE 57-bus system

# Polynomial Optimization

polynomial optimization  $\iff$  dense QCQP  $\iff$  sparse QCQP

- ❑ **Sparsification Technique:** distributed computation

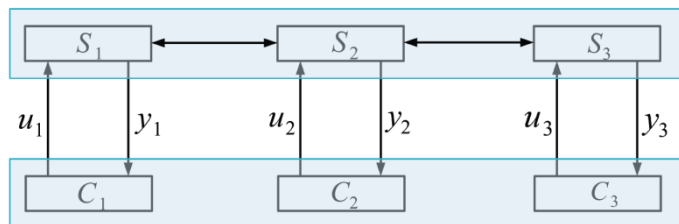
$$x_i \iff (x_{i1}, x_{i2}) \text{ s.t. } x_{i1} = x_{i2}$$

- ❑ This gives rise to a sparse QCQP with a sparse graph.
- ❑ The treewidth can be reduced to 2.

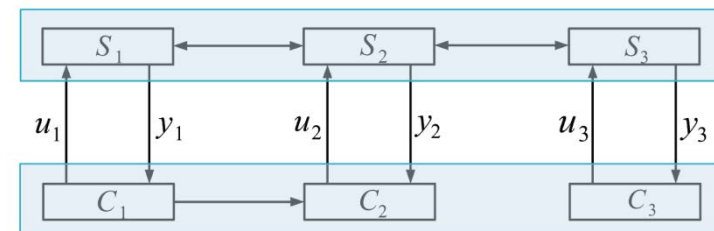
**Theorem:** Every polynomial optimization has a QCQP formulation whose SDP relaxation has a solution with rank 1, 2 or 3.

# Distributed Control

- Computational challenges arising in the control of real-world systems:
  - ❖ Communication networks
  - ❖ Electrical power systems
  - ❖ Aerospace systems
  - ❖ Large-space flexible structures
  - ❖ Traffic systems
  - ❖ Wireless sensor networks
  - ❖ Various multi-agent systems



**Decentralized control**



**Distributed control**

# Optimal Decentralized Control Problem

- ❑ **Optimal centralized control:** Easy (LQR, LQG, etc.)
- ❑ **Optimal distributed control (ODC):** NP-hard (Witsenhausen's example)

- ❑ Consider the time-varying system:

$$\begin{cases} x[\tau + 1] = A[\tau]x[\tau] + B[\tau]u[\tau] \\ y[\tau] = C[\tau]x[\tau] \end{cases}, \quad \forall \tau \in \mathbb{Z}_+$$

- ❑ The goal is to design a structured controller  $u[\tau] = Ky[\tau]$  to minimize

$$\sum_{\tau=0}^p (x[\tau]^T Q[\tau] x[\tau] + u[\tau]^T R[\tau] u[\tau]) + \mu \text{trace}\{KK^T\}$$

# Two Quadratic Formulations in Static Case

## □ Formulation in time domain:

❖ Stack the free parameters of  $K$  in a vector  $h$ .

❖ Define  $\mathbf{v}$  as:

$$\mathbf{v} = [ 1 \quad h^* \quad x[0]^* \quad x[1]^* \quad \cdots \quad x[p]^* \quad y[0]^* \quad \cdots \quad y[p]^* \quad u[0]^* \quad \cdots \quad u[p]^* ]^*$$

## □ Formulation in Lypunov domain:

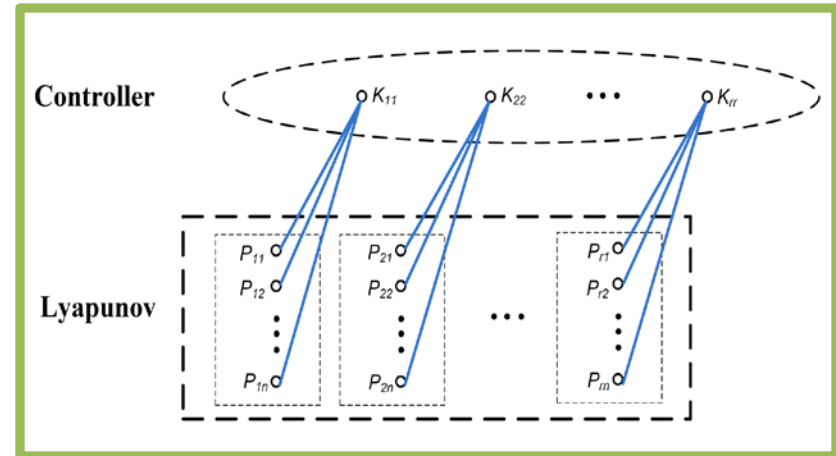
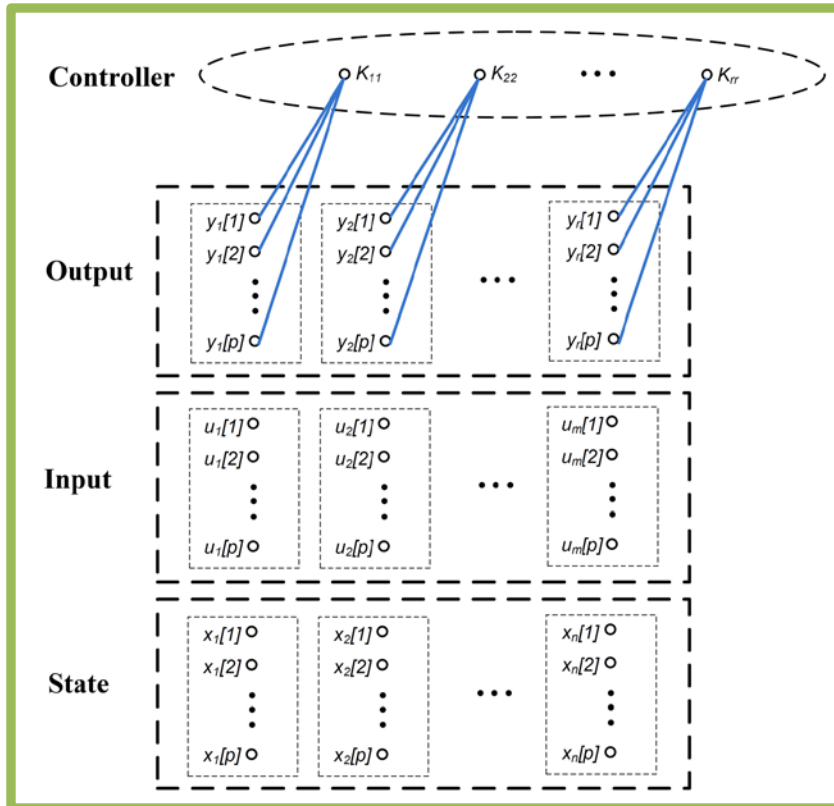
❖ Consider the BMI constraint:  $\begin{bmatrix} P & P(A + BKC)^* \\ (A + BKC)P & P \end{bmatrix} \succ 0$

❖ Define  $\mathbf{v}$  as:

$$\mathbf{v} = [ 1 \quad h^* \quad P_{11} \quad P_{12} \quad \cdots \quad P_{1n} \quad \cdots \quad P_{n1} \quad P_{n2} \quad \cdots \quad P_{nn} ]^*$$



# Graph of ODC for Time-Domain Formulation



**Theorem:** The SDP relaxation of ODC has a solution with rank 1-3 in the static/dynamic case for the time-domain or Lyapunov-domain formulation.

# Conclusions

## ❖ Optimization over power networks:

- ❑ Complexity is related to treewidth.

## ❖ Optimal decentralized control:

- ❑ NP-hard problem with a small treewidth.

## ❖ General theory for polynomial optimization:

- ❑ Every polynomial optimization has an SDP relaxation with a rank 1-3 solution.

## ❑ References:

1. R. Madani, G. Fazelnia, S. Sojoudi, J. Lavaei, "Finding Low-rank Solutions of Sparse Linear Matrix Inequalities using Convex Optimization," Preprint, 2014.
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3. G. Fazelnia, R. Madani, J. Lavaei, "Optimal Decentralized Control Problem as A Rank-Constrained Optimization," Preprint, 2014.