

# $L_1$ -Induced Norm Analysis of Positive Systems and Its Application to Stabilization of Large Scale Interconnected Positive Systems

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# Outline

- Analysis by Copositive Lyapunov Functions

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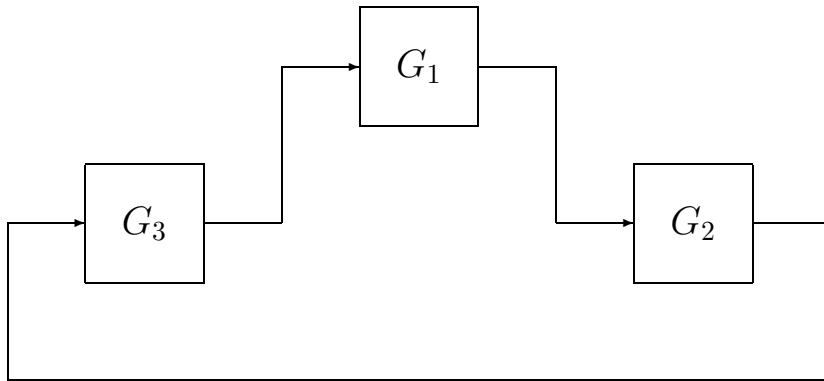
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- Analysis by Copositive Lyapunov Functions
- Weighted  $L_1$ -induced Norm
- Stability of Interconnected Positive Systems

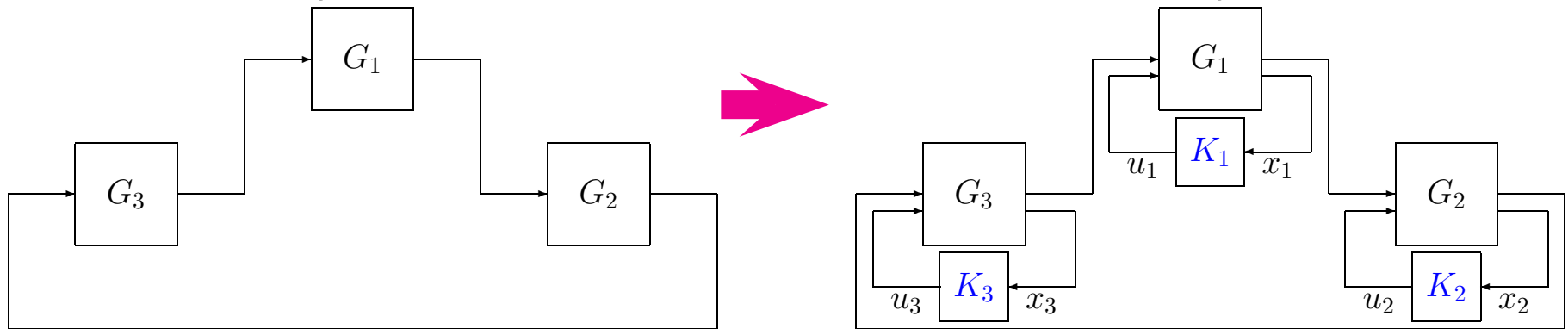
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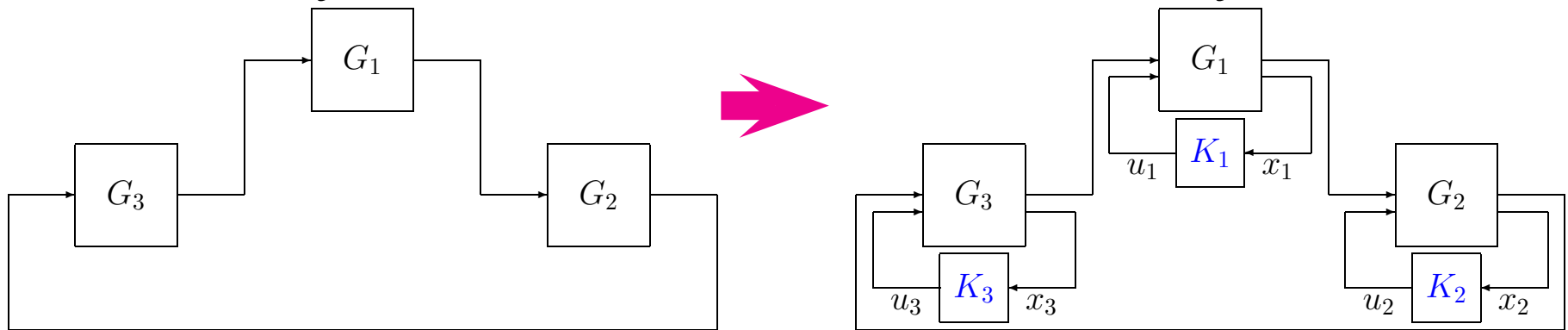
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- Decentralized Stabilizing SF Synthesis
  - purely locally by solving LP
  - optimality: minimizing real dominant pole

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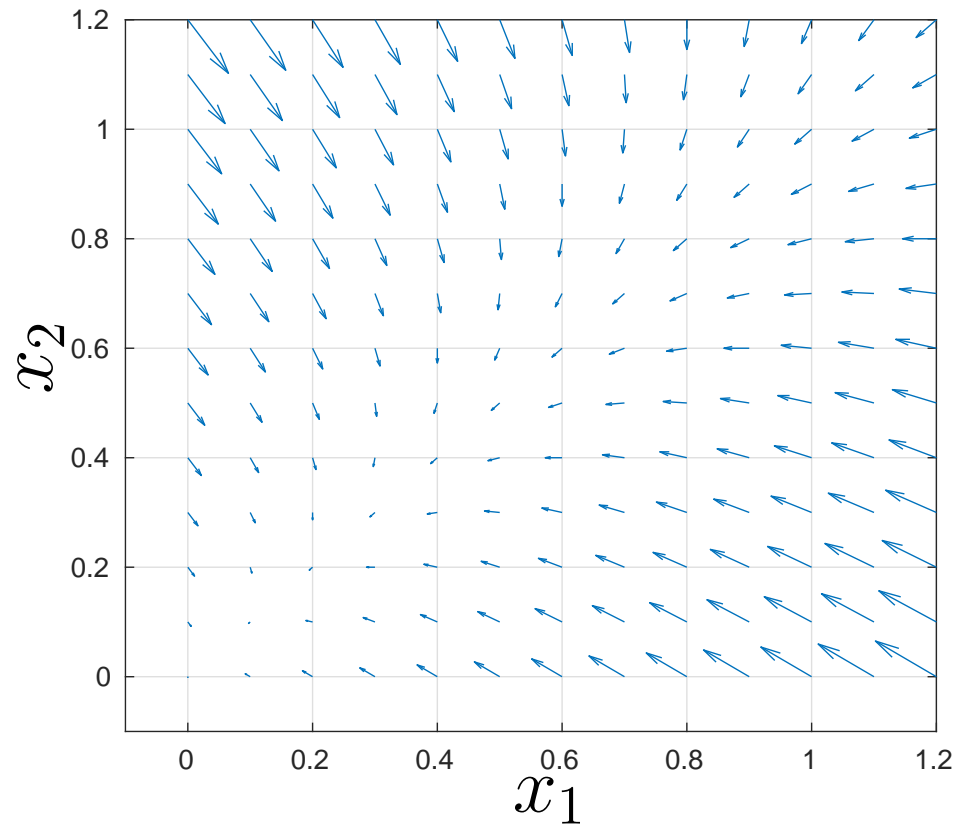
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- Decentralized Stabilizing SF Synthesis
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  - optimality: minimizing real dominant pole
- Conclusion

# Stability Analysis by Copositive Lyapunov Functions

Positive System  $\dot{x}(t) = Ax(t)$  ( $A \in \mathbb{M}^n$ )



$$x(0) \in \mathbb{R}_+^n \Rightarrow x(t) \in \mathbb{R}_+^n \quad (\forall t \in \mathbb{R}_+)$$



# Stability Analysis by Copositive Lyapunov Functions

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Requirements for LFs

$$V(x(t)) > 0, \quad -\dot{V}(x(t)) > 0 \quad (\forall x \in \mathbb{R}_+^n \setminus \{0\})$$

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- **linear:**  $V(x) = h^T x$ ,  $h \in \mathbb{R}_{++}^n$ 
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- **quadratic:**  $V(x) = x^T P x$ ,  $P \in \mathbb{S}_{++}^n$   
➤  $P \succeq I_n$ ,  $-(PA + A^T P) \succeq I_n$  (SDP)

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➤  $P \succeq_{\text{COP}} I_n$ ,  $-(PA + A^T P) \succeq_{\text{COP}} I_n$  (**COP**)

$$P \succeq_{\text{COP}} 0 \Leftrightarrow x^T P x \geq 0 \quad (\forall x \in \mathbb{R}_+^n)$$

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**COP is hard (tractable if  $n \leq 4$ )**

# Stabilizing SF Synthesis by LP

**Problem** For given  $A \in \mathbb{R}^{n \times n}$  and  $B \in \mathbb{R}^{n \times m}$ ,  
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Find  $g \in \mathbb{R}_{++}^n$ ,  $K \in \mathbb{R}^{m \times n}$  s.t.  $A + BK \in \mathbb{M}^n$ ,  $(A + BK)g \ll 0$ .

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- stabilizing SF gain:  $K = YG^{-1}$
- synthesis under structural constraint

$$K = \begin{bmatrix} K_1 & 0 \\ 0 & K_2 \end{bmatrix} \quad \longleftrightarrow \quad Y = \begin{bmatrix} Y_1 & 0 \\ 0 & Y_2 \end{bmatrix}$$

# Weighted $L_1$ -induced Norm Analysis

## Positive System

$$G : \begin{cases} \dot{x} = Ax + Bw, & x(0) = 0, \\ z = Cx + Dw \end{cases}$$

$$A \in \mathbb{M}^n, B \in \mathbb{R}_+^{n \times n_w}, C \in \mathbb{R}_+^{n_z \times n}, D \in \mathbb{R}_+^{n_z \times n_w}$$

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For given  $q_z \in \mathbb{R}_{++}^{n_z}$  and  $q_w \in \mathbb{R}_{++}^{n_w}$ , we define

$$\|G_{q_z, q_w}\|_{1+} := \sup_{\|q_w^T w\|_1=1, w \in L_{1+}} \|q_z^T z\|_1.$$

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Here,

$$\|v\|_1 := \int_0^\infty \|v(t)\|_1 dt,$$

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$L_{1+} := \{v(t) : \|v\|_1 < \infty, \quad v(t) \geq 0 \quad \forall t \in [0, \infty)\}.$   
the set of  $L_1$ -bounded and positive signals

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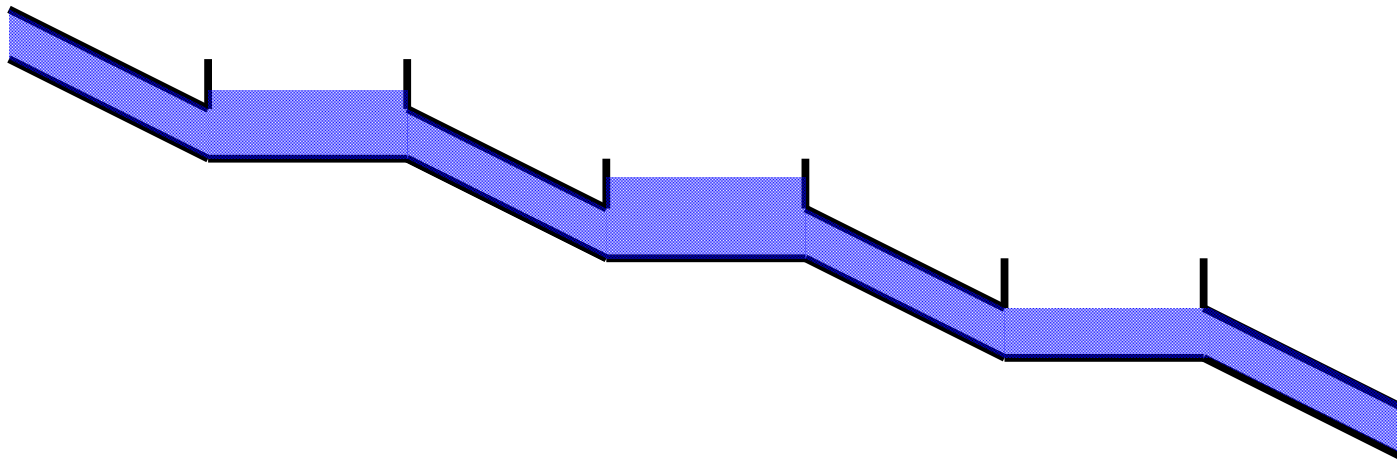


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## Example: Tank System

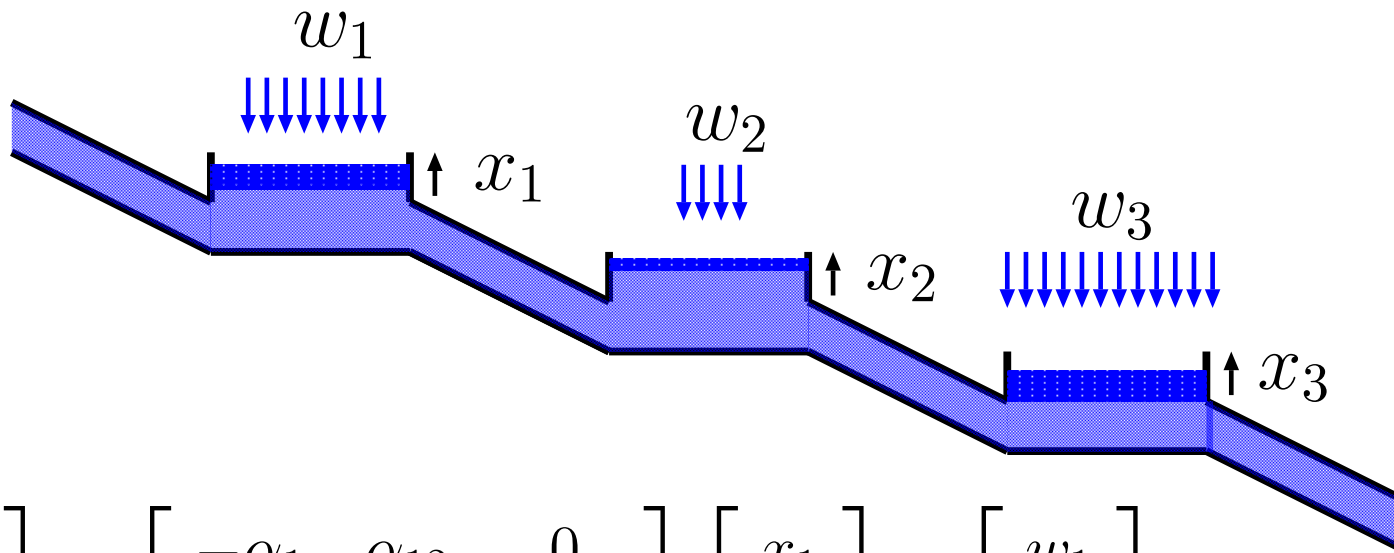


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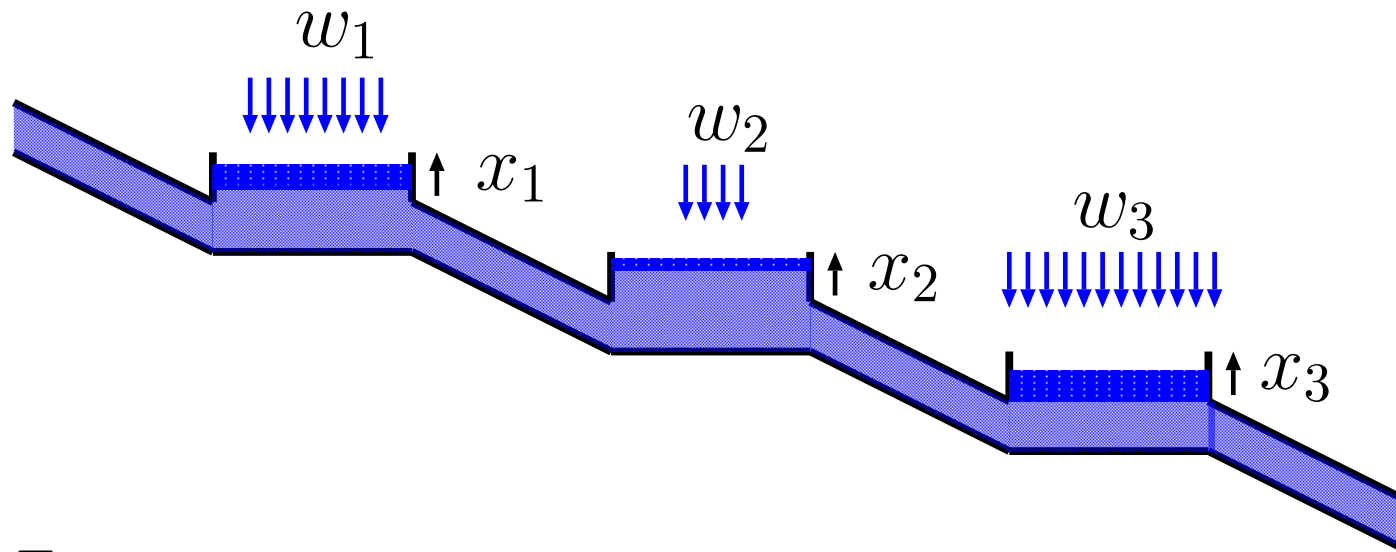
$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{bmatrix} = \begin{bmatrix} -\alpha_1 & \alpha_{12} & 0 \\ \alpha_{21} & -\alpha_2 & \alpha_{23} \\ 0 & \alpha_{32} & -\alpha_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix}.$$

# Weighted $L_1$ -induced Norm Analysis

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## Example: Tank System



- $q_z^T = [1 \ 1 \ 1]$  ➔  $z = x_1 + x_2 + x_3$  (total fluctuation)
- $q_{z,i}$ : weights for  $x_i$ ,  $q_{w,i}$ : rain strength

# Weighted $L_1$ -induced Norm Analysis

## Definition

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## Facts

$$\|G_{q_z, q_w}\|_{1+} = \|Q_z G Q_w^{-1}\|_1,$$
$$Q_z = \text{diag}(q_z), \quad Q_w = \text{diag}(q_w).$$

$L_1$ -induced norm with weightings  
on the input and output signals

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How to compute?

# Weighted $L_1$ -induced Norm Analysis

## Theorem (Ebihara et al., CDC11,TAC17)

For given  $q_z \in \mathbb{R}_{++}^{n_z}$ ,  $q_w \in \mathbb{R}_{++}^{n_w}$  and  $\gamma > 0$ , the followings are equivalent:

- (i)  $A \in \mathbb{H}^n$  and  $\|G_{q_z, q_w}\|_{1+} < \gamma$ .
- (ii)  $\exists h \in \mathbb{R}_{++}^n$  such that
$$\begin{bmatrix} h^T A + q_z^T C & h^T B + q_z^T D - \gamma q_w^T \end{bmatrix} \ll 0.$$

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## Characterization by Linear Programming

$$\inf_{h \in \mathbb{R}_{++}^n, \gamma} \gamma \text{ s. t. } \begin{bmatrix} h^T A + q_z^T C & h^T B + q_z^T D - \gamma q_w^T \end{bmatrix} \ll 0.$$

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## LP-based Positive System Analysis

- M. Rami and F. Tadeo, MTNS 2006
- C. Briat, Int. J. Robust Nonlin. Cont. 2013
- J. Shen and J. Lam, Automatica 2014
- A. Rantzer, EJC 2015



C. Briat



J. Lam



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### Application?

- stability analysis of interconnected systems
- decentralized controller synthesis

# Stability Analysis of Interconnected Positive Systems

Subsystem  $G_i$  (positive and stable)

$$G_i : \begin{cases} \dot{x}_i = A_i x_i + \sum_{j=1, j \neq i}^N B_{ij} w_{ij}, \\ z_{ji} = C_{ji} x_i \quad (j \neq i) \end{cases} \quad (i = 1, \dots, N)$$

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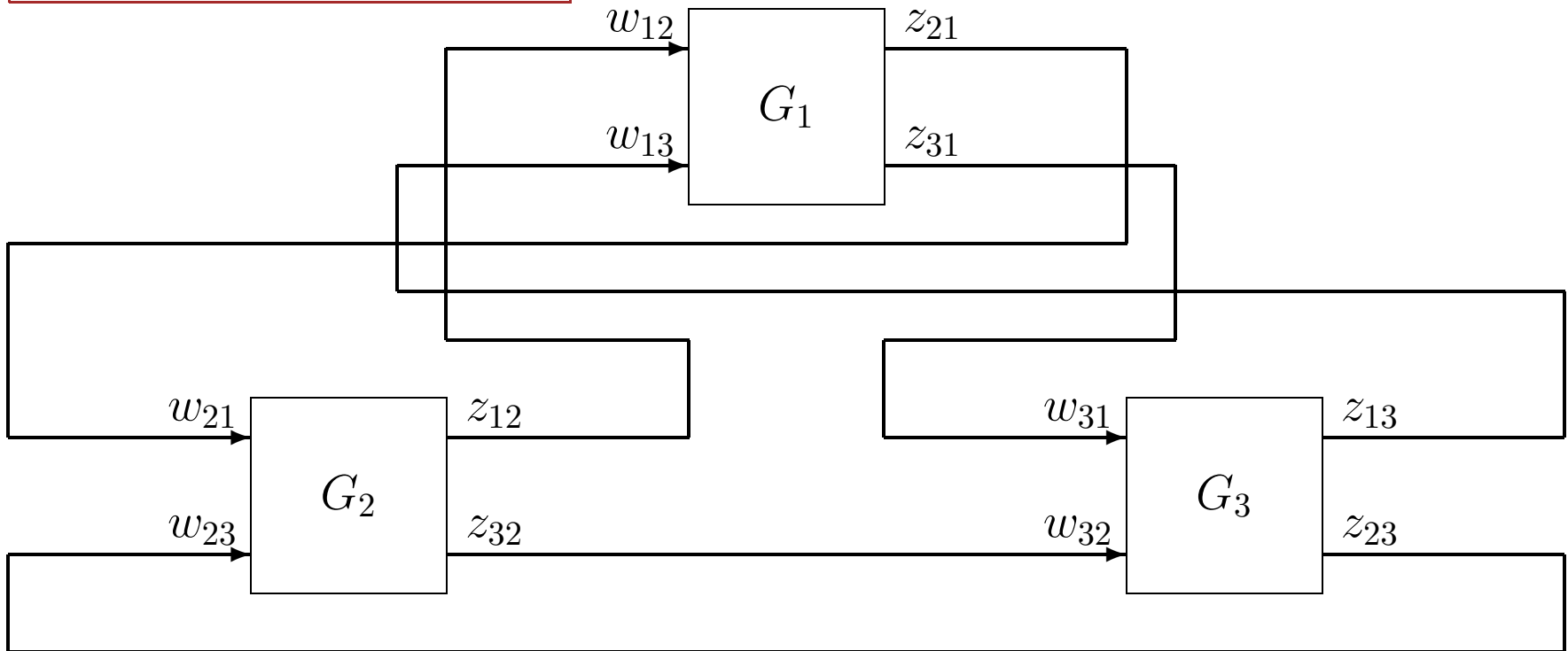
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Interconnection

$$w_{ij} = z_{ij} \quad (i, j = 1, \dots, N, \quad i \neq j)$$

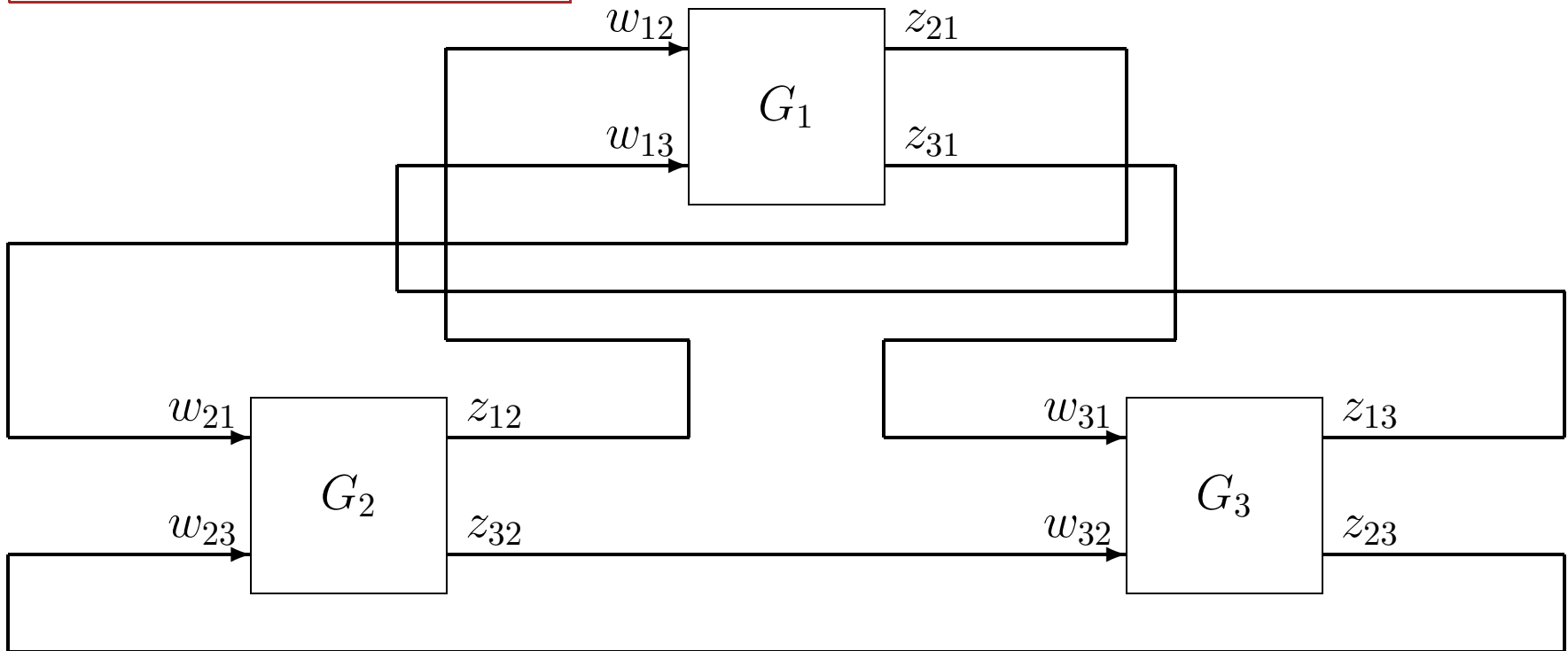
# Stability Analysis of Interconnected Positive Systems

Example ( $N = 3$ )



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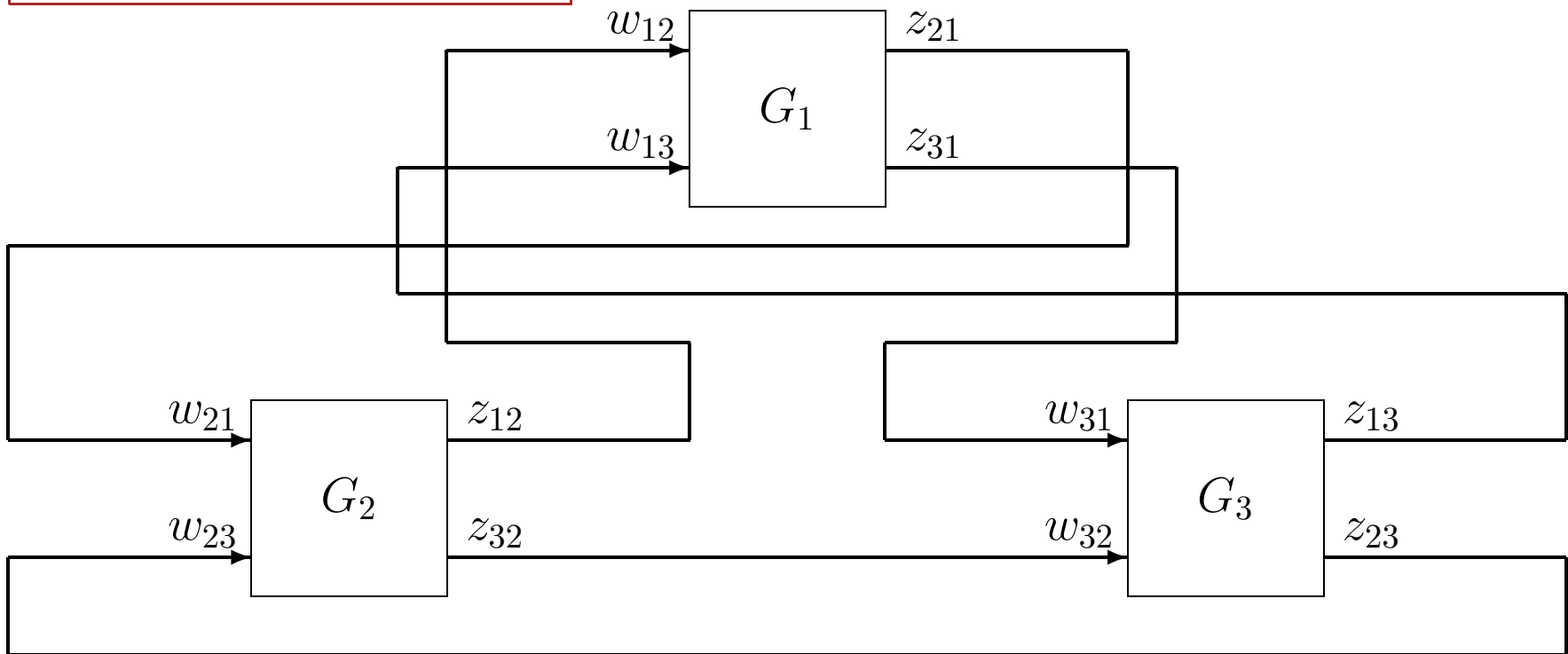
Example ( $N = 3$ )



most general interconnection structure

# Stability Analysis of Interconnected Positive Systems

Example ( $N = 3$ )



stability characterization via  
weighted  $L_1$ -induced norm?

# Stability Analysis of Interconnected Positive Systems

## Theorem (Ebihara et al., CDC2011, TAC2017)

The interconnected system is stable iff there exists  $q_{ij} \in \mathbb{R}_{++}^{n_{wij}}$  ( $i, j = 1, \dots, N, i \neq j$ ) such that  $\|G_{i,q_{i,z},q_{i,w}}\|_{1+} < 1$  ( $i = 1, \dots, N$ ) where

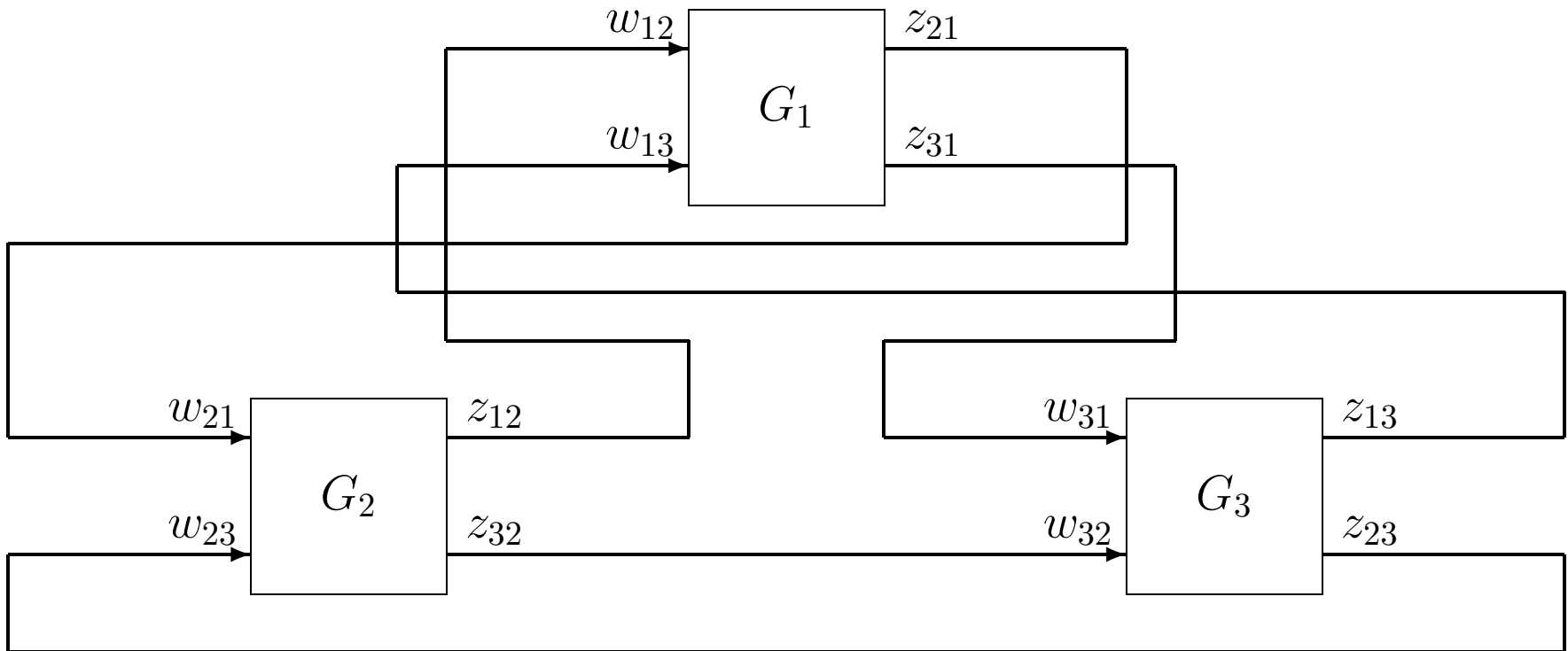
$$q_{i,z} = \begin{bmatrix} q_{1i}^T & \cdots & q_{i-1,i}^T & q_{i+1,i}^T & \cdots & q_{Ni}^T \end{bmatrix}^T,$$
$$q_{i,w} = \begin{bmatrix} q_{i1}^T & \cdots & q_{i,i-1}^T & q_{i,i+1}^T & \cdots & q_{iN}^T \end{bmatrix}^T.$$

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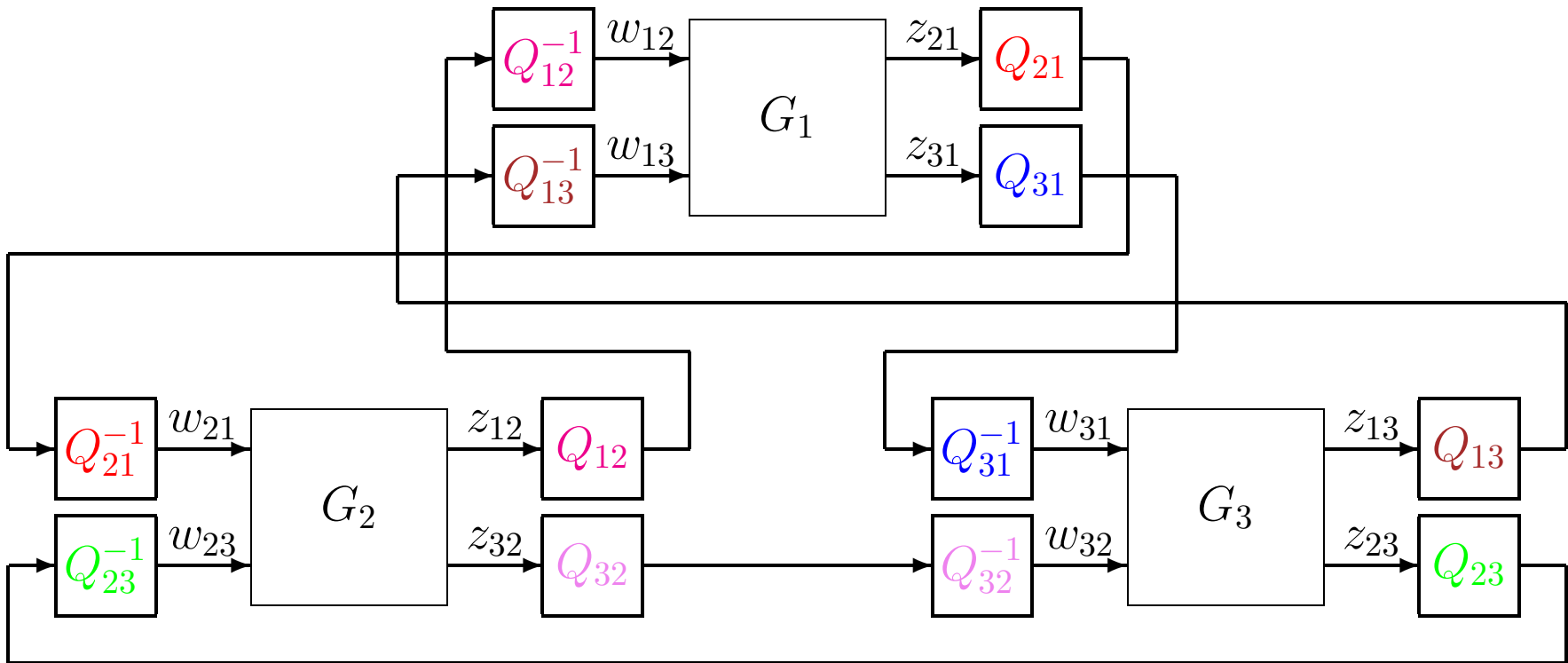
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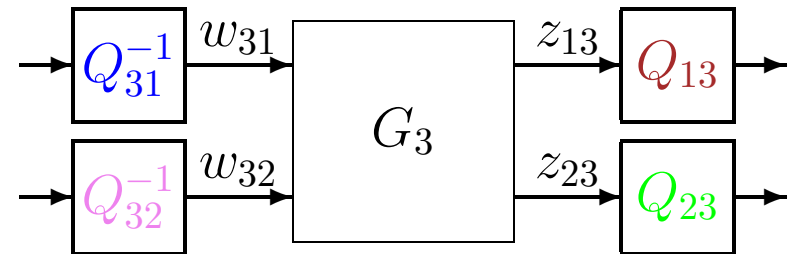
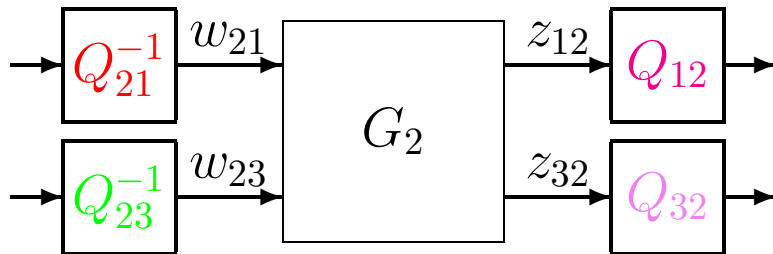
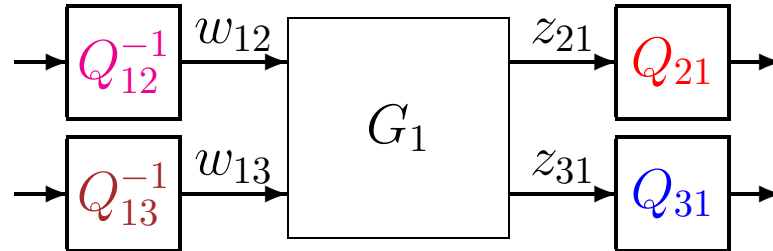
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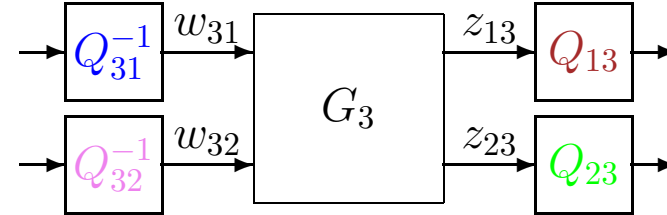
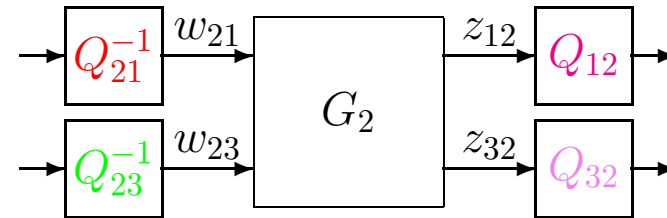
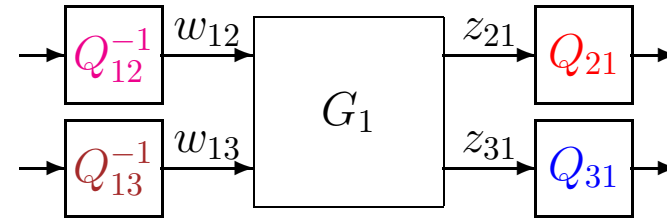
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## Example ( $N = 3$ )

$$\|G_1, [q_{21}^T \ q_{31}^T]^T, [q_{12}^T \ q_{13}^T]^T\|_{1+} < 1,$$

$$\|G_2, [q_{12}^T \ q_{32}^T]^T, [q_{21}^T \ q_{23}^T]^T\|_{1+} < 1,$$

$$\|G_3, [q_{13}^T \ q_{23}^T]^T, [q_{31}^T \ q_{32}^T]^T\|_{1+} < 1.$$



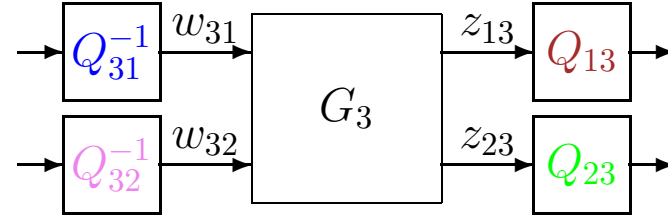
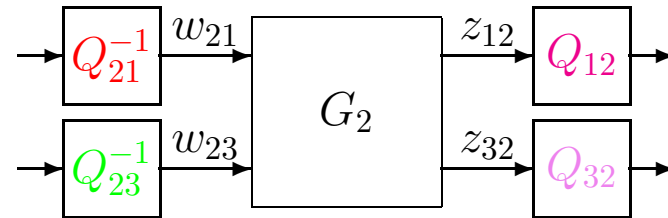
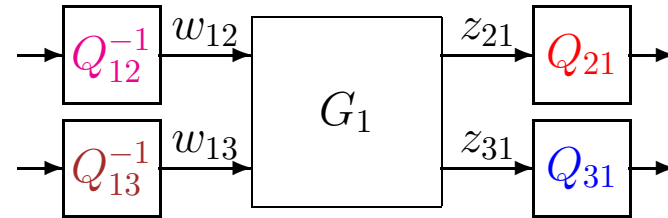
# Stability Analysis of Interconnected Positive Systems

## Example ( $N = 3$ )

$$\|G_1, [q_{21}^T \ q_{31}^T]^T, [q_{12}^T \ q_{13}^T]^T\|_{1+} < 1,$$

$$\|G_2, [q_{12}^T \ q_{32}^T]^T, [q_{21}^T \ q_{23}^T]^T\|_{1+} < 1,$$

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- necessary and sufficient condition for stability with  $L_1$ -induced norm of each subsystem
- coupled over subsystems through  $q_{ij}$

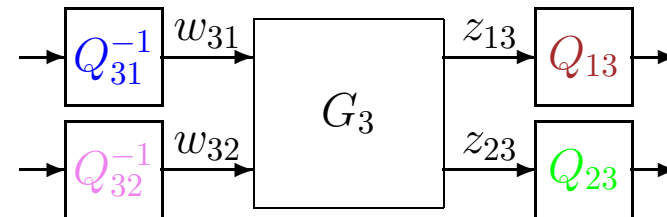
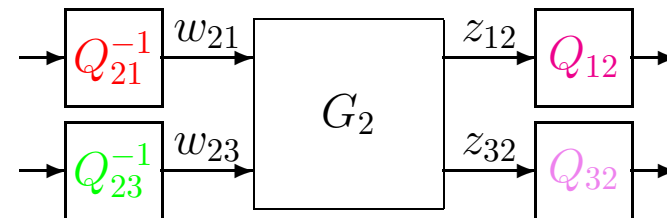
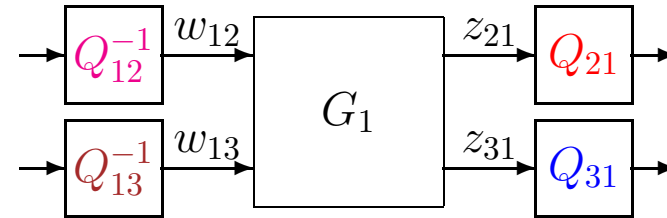
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## Decentralized Stabilizing Synthesis

# Decentralized Control of Interconnected Positive Systems

Subsystem  $G_i$  ( $i = 1, \dots, N$ )

$$G_i : \begin{cases} \dot{x}_i = A_i x_i + \sum_{j=1, j \neq i}^N B_{w_{ij}} w_{ij}, \\ z_{ji} = C_{z_{ji}} x_i \quad (j \neq i) \end{cases}$$

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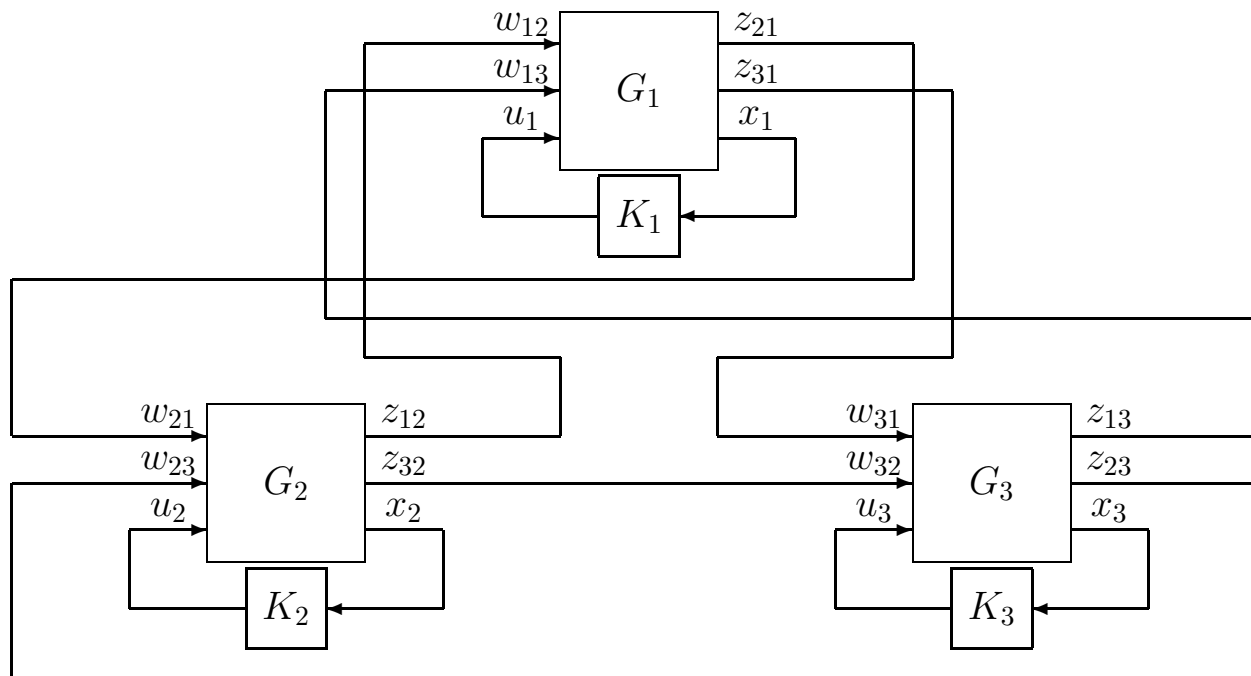
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- practical but intractable requirement

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Subsystem  $G_i$  with  $u_i = K_i x_i$

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How to solve?

# Decentralized Control of Interconnected Positive Systems

## Stability Condition of Interconnected PS's

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What is the tractable network structure?



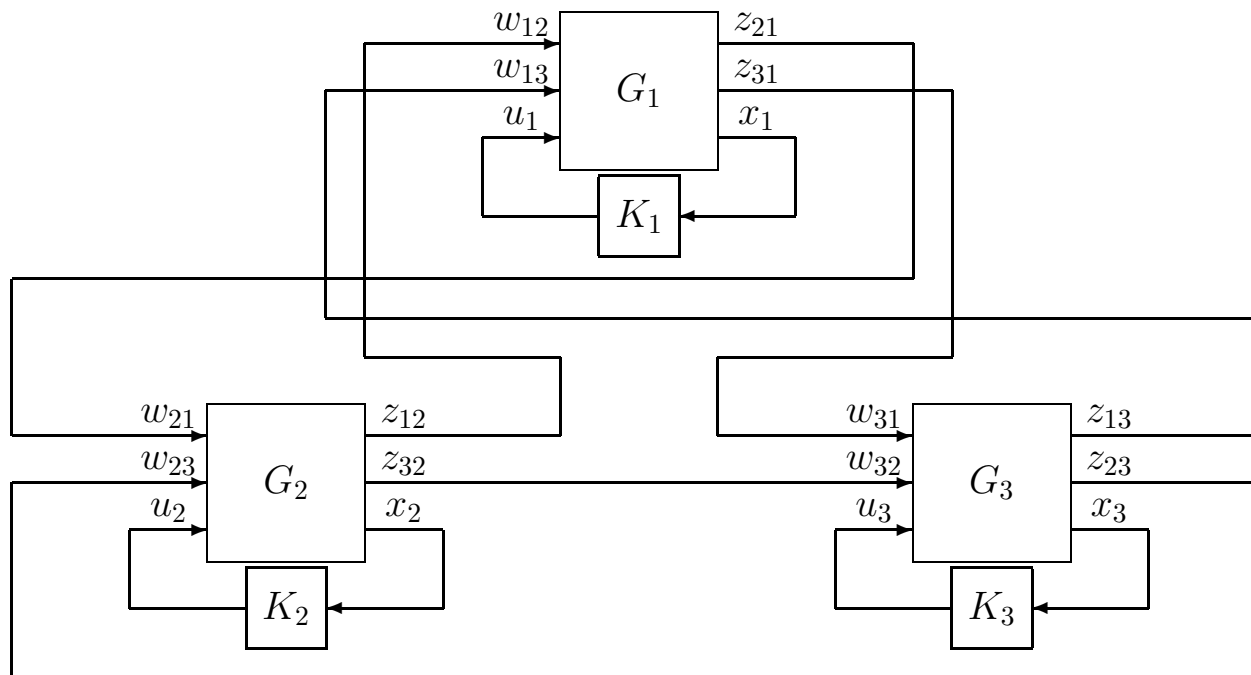
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$G_i$  provides the same scalar output to  $G_j$  ( $j \neq i$ )



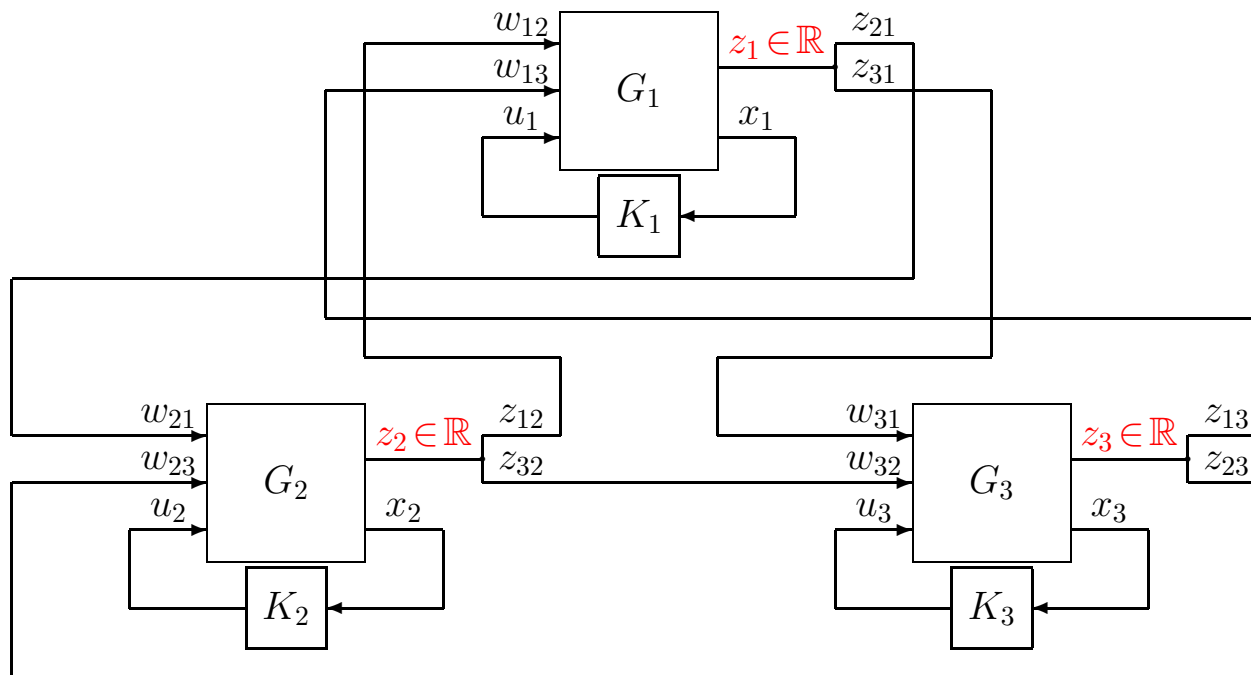
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**Simplest Case:**  $G_i$  ( $i = 1, 2, 3$ ) are all **SISO** and

$$\begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} = \Omega \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix}, \quad \Omega = \begin{bmatrix} 0 & \Omega_{12} & \Omega_{13} \\ \Omega_{21} & 0 & \Omega_{23} \\ \Omega_{31} & \Omega_{32} & 0 \end{bmatrix} \in \mathbb{R}_+^{3 \times 3}.$$

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## Synthesis Problem

$$\inf_{K_i \in \mathcal{K}_{p,i}} \|G_{i,K_i,1,q_{i,w}}\|_{1+} \quad (K_i \in \mathcal{K}_{p,i}: \text{positivity constraint})$$

$L_1$ -induced Norm Optimal Control Problem  
(optimal  $q_{i,w}$  is not locally available)

# Decentralized Control of Interconnected Positive Systems

## $L_1$ -induced Norm Optimal Control

$$\inf_{K_i \in \mathcal{K}_{p,i}} \|G_{i,K_i,1,q_i,w}\|_1, \quad G_{i,K_i} = \left[ \begin{array}{c|c} A_i + B_{u_i}K_i & B_{w_i} \\ \hline C_{z_i} + D_{z_i u_i}K_i & 0 \end{array} \right]$$

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## Optimal Control for Normalized System

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Preceding Results (Ebihara et al., ACC2012)

- $K_i^*$  can be computed purely locally via LP
- $K_i^*$  is robustly optimal for any  $B_{w_i} \geq 0, q_{i,w} \geq 0$



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our primary objective has been achieved!!

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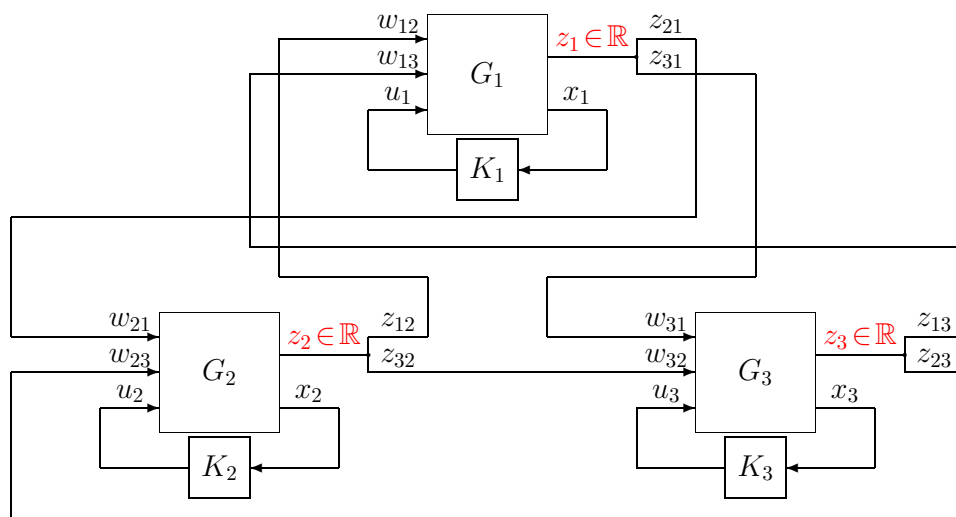
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optimality of  $K^* = \{K_1^*, \dots, K_N^*\}$  in global sense?

# Main Result



$$\dot{x}(t) = A_K x(t),$$

$$K = \{K_1, \dots, K_N\}$$

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## Performance Index

$\lambda_F(A_K)$ : maximal real part of the eigenvalues of  $A_K$

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$$\lambda_F^* := \inf_{K \in \mathcal{K}_d \cap \mathcal{K}_p} \lambda_F(A_K)$$

$\mathcal{K}_d$ : decentralized structural constraint

$\mathcal{K}_p$ : positivity constraint

# Main Result

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$$\lambda_F(A_{K^*}) = \lambda_F^*$$

$K^* = \{K_1^*, \dots, K_N^*\}$  is optimal for  $\lambda_F(A_K)$   
( $K_i^*$ : locally  $L_1$ -ind.-norm-optimal controller)

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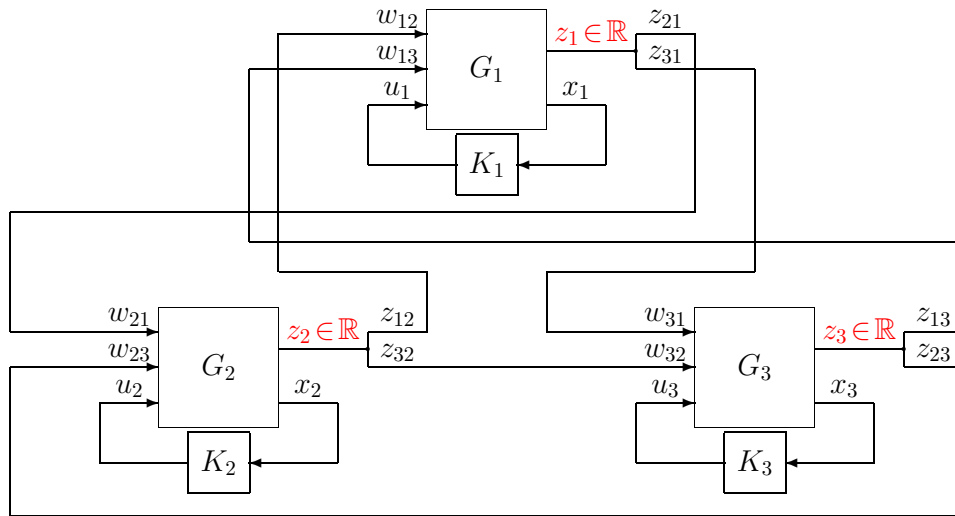
## Remark

- $K^*$  can also be designed by **centralized approach**

$$\sup_{K \in \mathcal{K}_d \cap \mathcal{K}_p, g, \alpha} \alpha \text{ s.t. } g \gg 0, (A_K + \alpha I)g \ll 0.$$

- **needs bisection search over  $\alpha \in \mathbb{R}$**
- **computationally demanding ( $g \in \mathbb{R}^{\sum n_i}$ )**

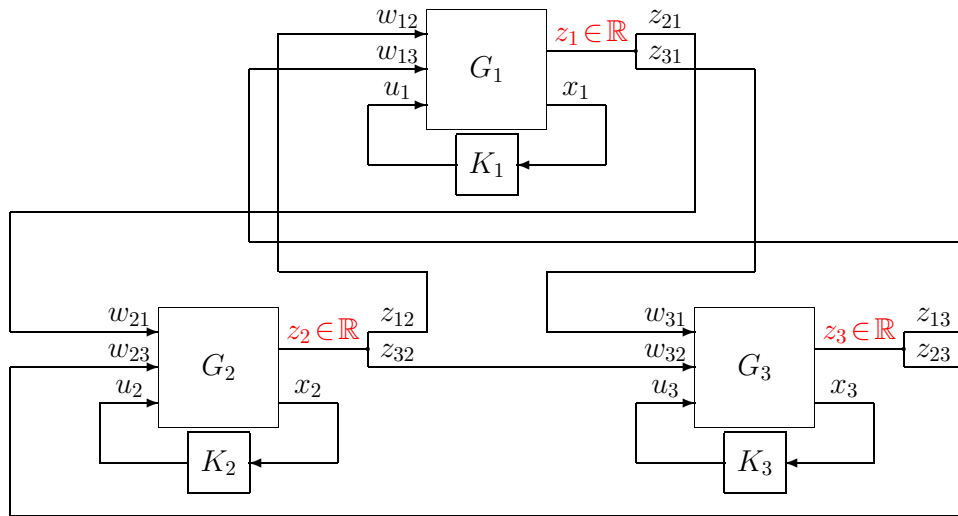
# Numerical Example 1 (Toy Example)



- $n_1 = 2, n_2 = 3, n_3 = 4$
- open-loop unstable ( $\lambda_F(A) = 1.1547$ )



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## $L_1$ -ind.-norm Optimal Gains

$$K_1^* = \begin{bmatrix} -0.4444 & -2.0000 \end{bmatrix},$$

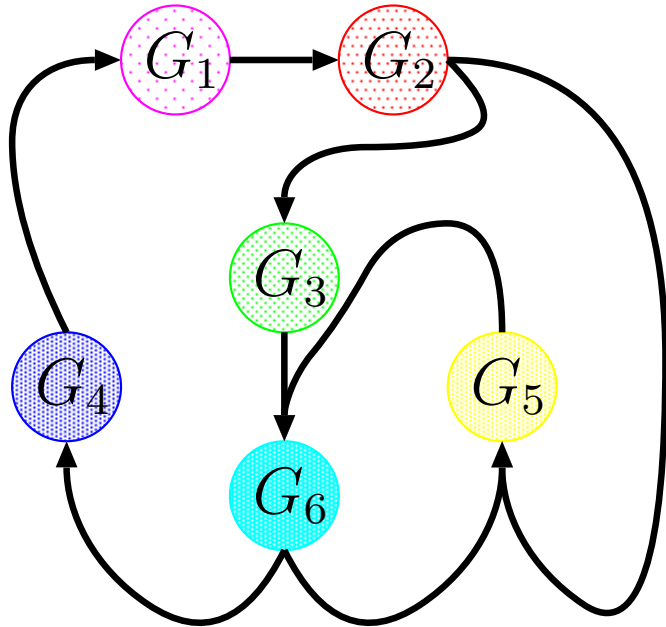
$$K_2^* = \begin{bmatrix} -0.8000 & -0.3333 & -1.0000 \end{bmatrix},$$

$$K_3^* = \begin{bmatrix} -0.5000 & -0.1111 & -0.1111 & -0.2222 \end{bmatrix}.$$

- $\lambda_F^* = \lambda_F(A_{K^*}) = -0.3056$  ( $K^* = \{K_1^*, K_2^*, K_3^*\}$ )
- coincides with centralized approach with computationally demanding bisection search

# Numerical Example 2 (Control Node Selection)

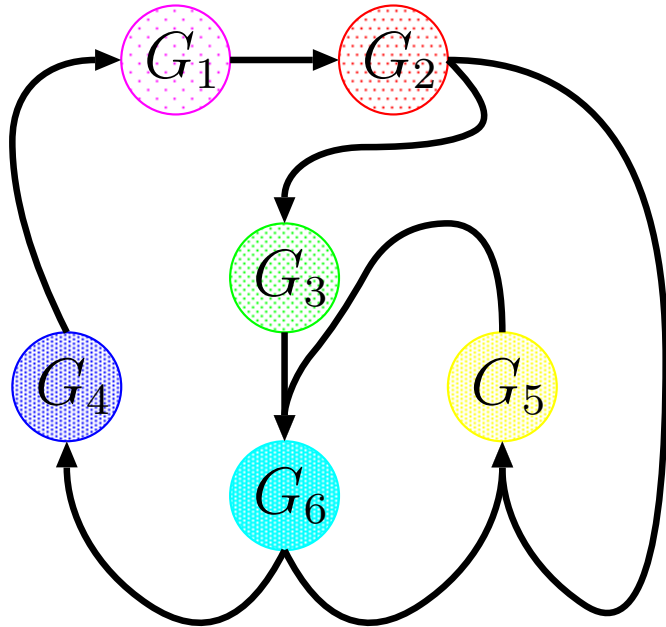
## Interconnected PS with Directed Network



- node  $i$ :  
SISO PS  $G_i$  with dim.  $n_i$
- edge  $(i, j)$ : weights  $\Omega_{i,j}$

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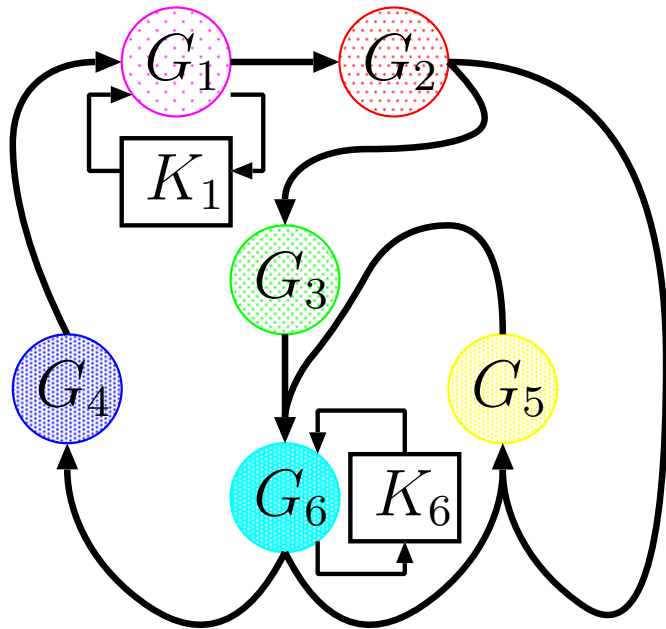


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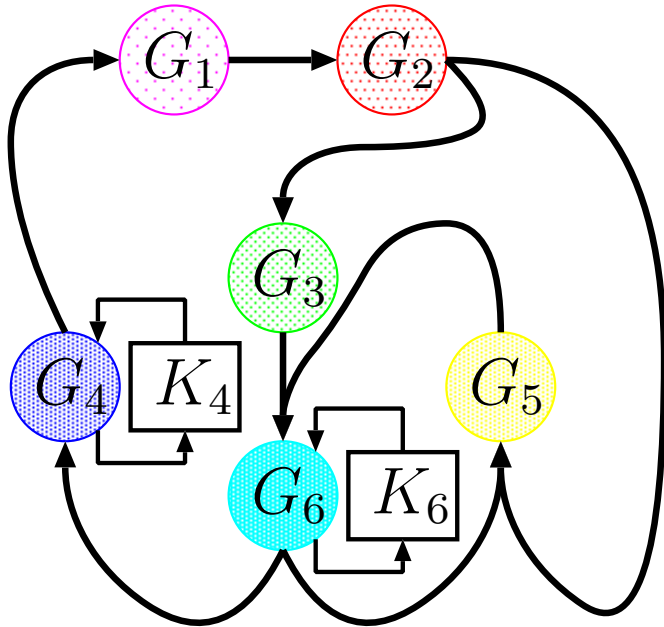


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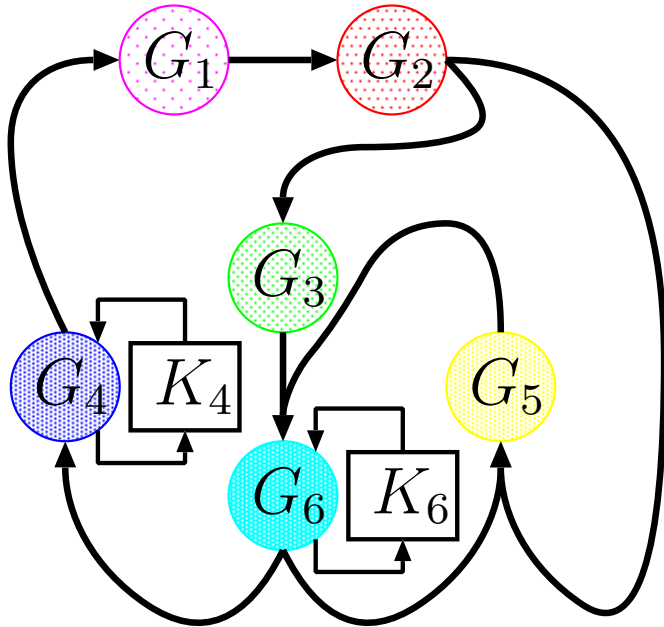


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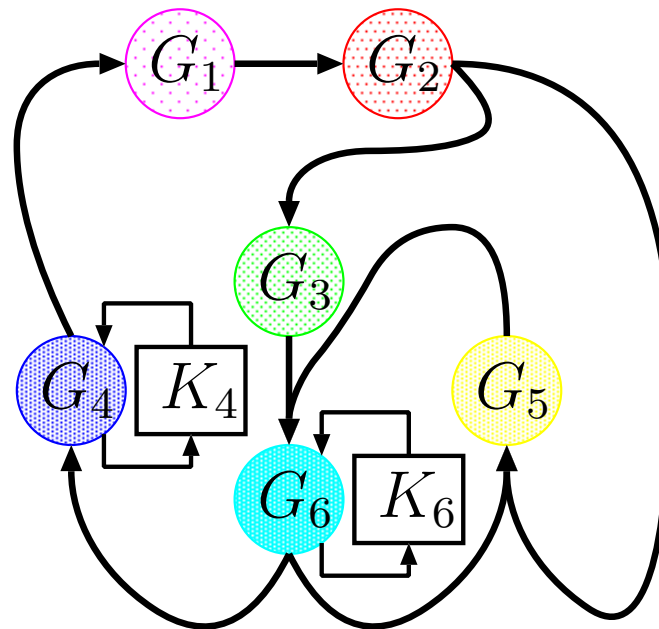
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For each selected  $N_c$  nodes  $\mathcal{N}_c$ , solve

find  $K_{\mathcal{N}_c} \in \mathcal{K}_d \cap \mathcal{K}_p$  and  $g$  s.t.  $g \gg 0$ ,  $A_{K_{\mathcal{N}_c}} g \ll 0$ .

$$K_{\mathcal{N}_c} = \begin{bmatrix} K_1 & 0 & \dots & 0 \\ 0 & K_2 & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & K_N \end{bmatrix} \quad (K_i = 0 \text{ if } i \notin \mathcal{N}_c)$$

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change of variable  $Y = K_{\mathcal{N}_c} \text{diag}(g) \rightarrow \text{LP}$

## Numerical Example 2 (Control Node Selection)

**Problem** Find  $N_c$  nodes  $\mathcal{N}_c = \{n_1, \dots, n_{N_c}\}$  and SF gains  $K_{n_i}$  ( $i = 1, \dots, N_c$ ) such that controlled interconnected system is **positive and stable**  
combinatorial problem of selecting  $N_c$  nodes from  $N$  nodes ( ${}_N C_{N_c}$ )

### Centralized Approach

For each selected  $N_c$  nodes  $\mathcal{N}_c$ , solve

find  $K_{\mathcal{N}_c} \in \mathcal{K}_d \cap \mathcal{K}_p$  and  $g$  s.t.  $g \gg 0$ ,  $A_{K_{\mathcal{N}_c}} g \ll 0$ .

LP is feasible ➡ stability and positivity achieved

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- **computationally demanding (LP with  $g \in \mathbb{R}^{\sum n_i}$ )**  
**need to solve synthesis problem  ${}_N C_{N_c}$  times**

# Numerical Example 2 (Control Node Selection)

## Proposed Approach

Step 1: Compute  $K_i^*$  ( $i = 1, \dots, N$ )

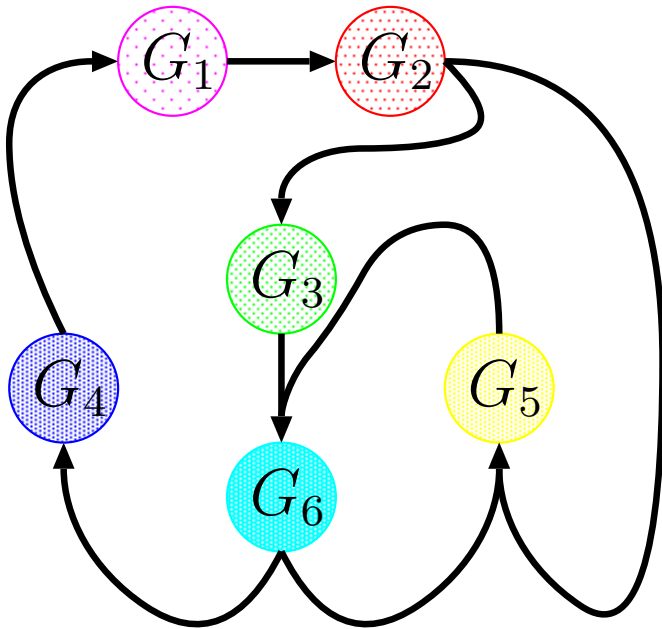
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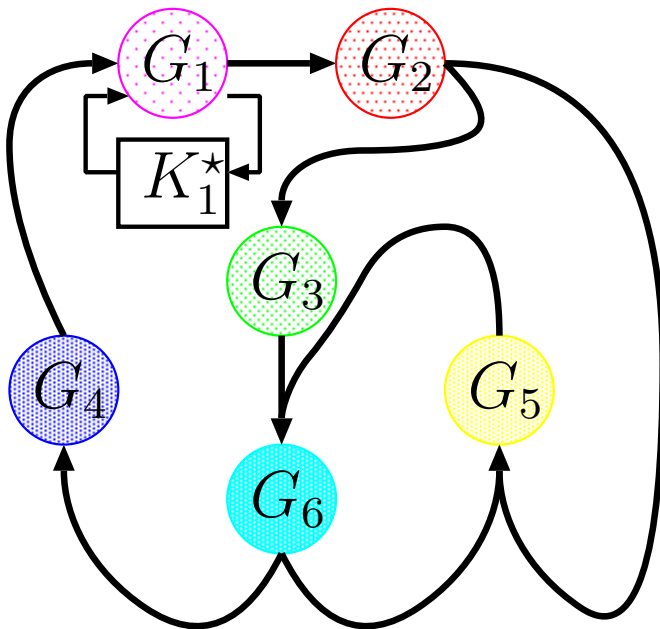


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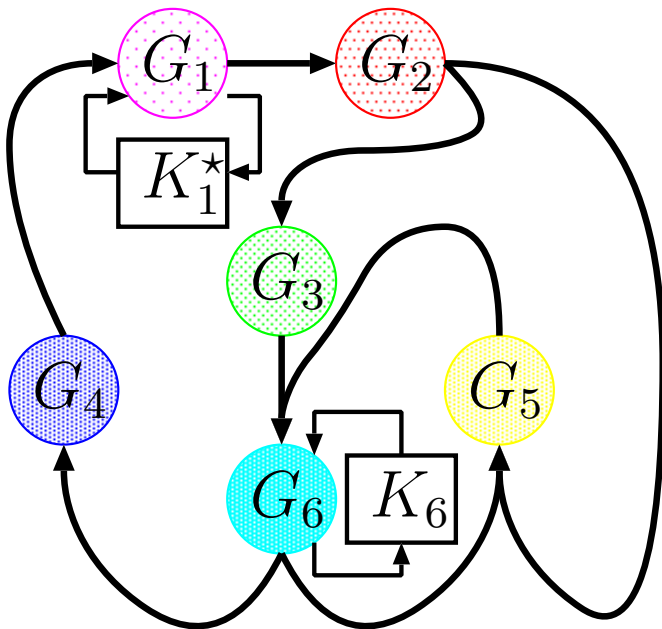


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- If node 1 is selected,  $K_1^*$  is the best for  $G_1$
- If nodes 1 and 6 are selected,  $K_1^*$  is **still the best for  $G_1$**  irrespective of  $K_6$

## Numerical Example 2 (Control Node Selection)

### Proposed Approach

Step 1: Compute  $K_i^*$  ( $i = 1, \dots, N$ )

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For each selected  $N_c$  nodes  $\mathcal{N}_c$ , the best controller is

$$K_{\mathcal{N}_c}^* = \begin{bmatrix} K_1^* & 0 & \dots & 0 \\ 0 & K_2^* & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & K_N^* \end{bmatrix} \quad (K_i^* = 0 \text{ if } i \notin \mathcal{N}_c)$$

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$\lambda_F(A_{K_{\mathcal{N}_c}^*}) < 0 \Rightarrow$  stability and positivity achieved

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- computationally efficient

suffice to solve analysis problem  $N C_{N_c}$  times

# Numerical Example 2 (Control Node Selection)

## Comparison

### Centralized Approach

- LP of each size  $\sum n_i$  for  $K_{\mathcal{N}_c}^*$ :  $N C_{N_c}$  times

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# Numerical Example 2 (Control Node Selection)

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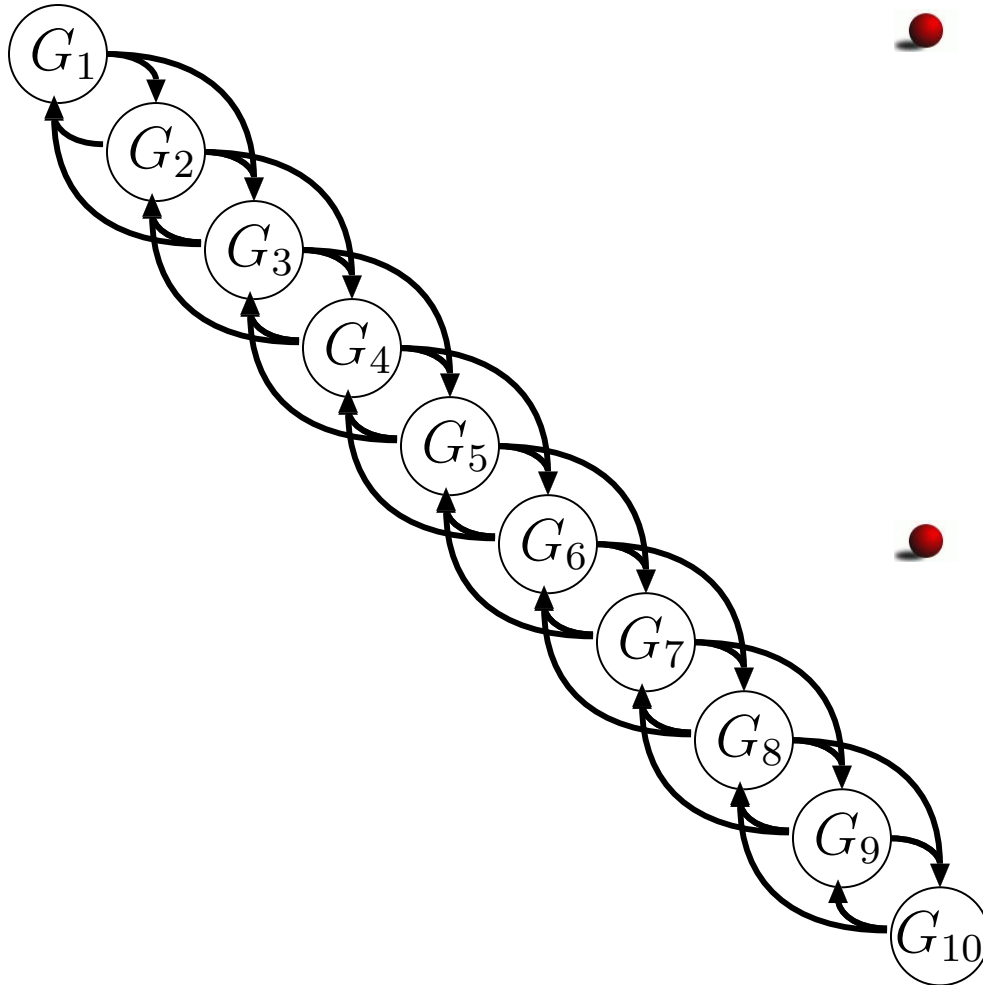
### Proposed Approach

- LP of each size  $n_i$  for  $K_i^*$ :  $N$  times
- computation of  $\lambda_F(A_{K_{\mathcal{N}_c}^*})$ :  $N C_{N_c}$  times

proposed approach is much more efficient

# Numerical Example 2 (Control Node Selection)

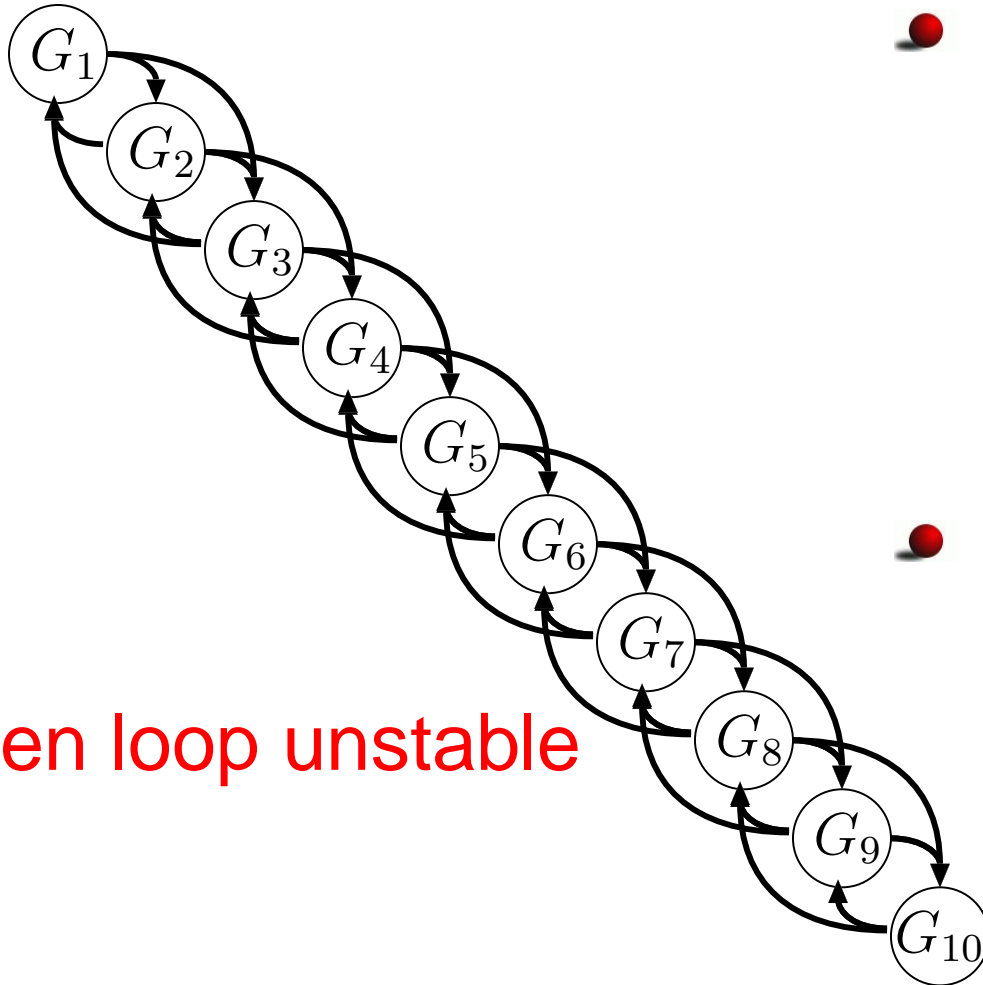
## Concrete Example



- $A_i \in \mathbb{M}^3 \cap \mathbb{H}^3$ ,  
 $B_{w_i} \in \mathbb{R}_+^{3 \times 1}$ ,  
 $B_{u_i} \in \mathbb{R}^{3 \times 1}$ ,  
 $C_{z_i} \in \mathbb{R}_+^{1 \times 3}$ ,  
 $D_{z_i u_i} = 0$
- $\Omega_{i,j} > 0 \ (|i - j| \leq 2)$   
randomly generated

# Numerical Example 2 (Control Node Selection)

## Concrete Example

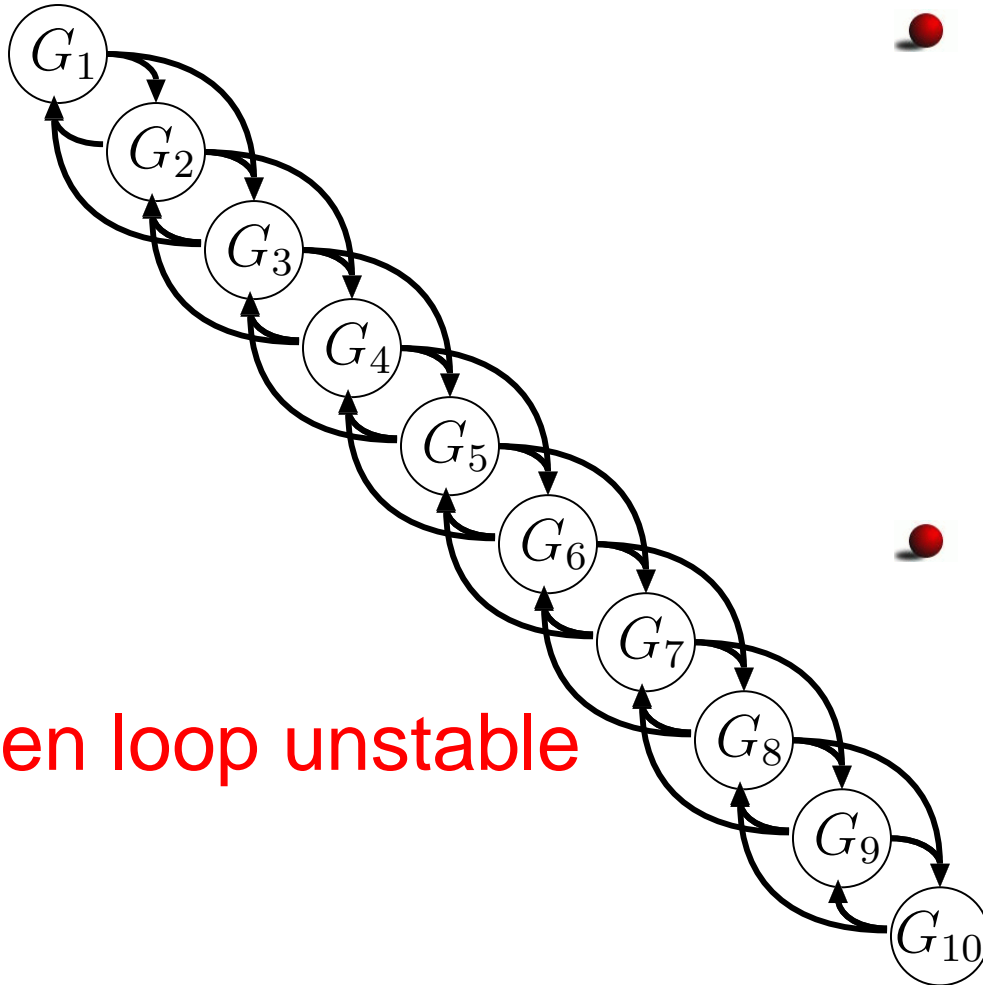


open loop unstable

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select four nodes ( $N_c = 4$ )

## Numerical Example 2 (Control Node Selection)

Results ( ${}_{10}C_4 = 210$ )

Centralized Approach: 37.80 [sec] ( $\approx 0.18 \times 210$ )

Proposed Approach: 1.43 [sec] ( $\approx 0.14 \times 10$ )

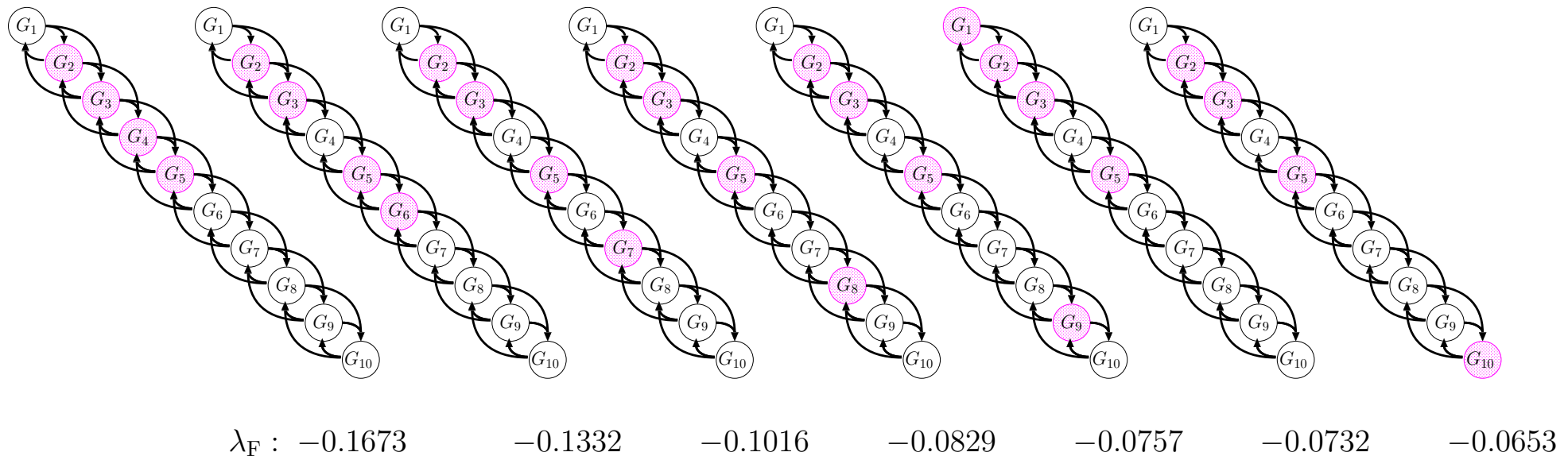
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Proposed Approach: 1.43 [sec] ( $\approx 0.14 \times 10$ )

Selected Nodes (7 patterns)



● nodes 2, 3, 5 are important

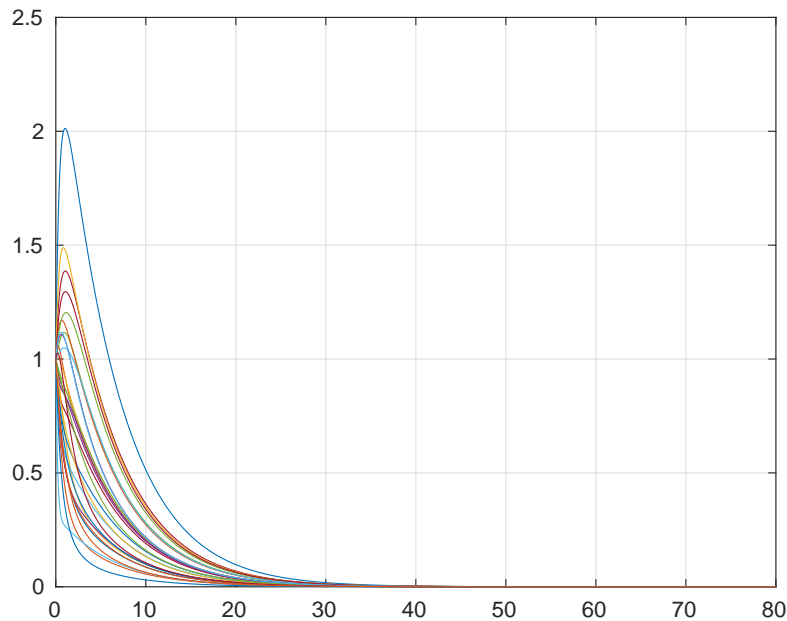
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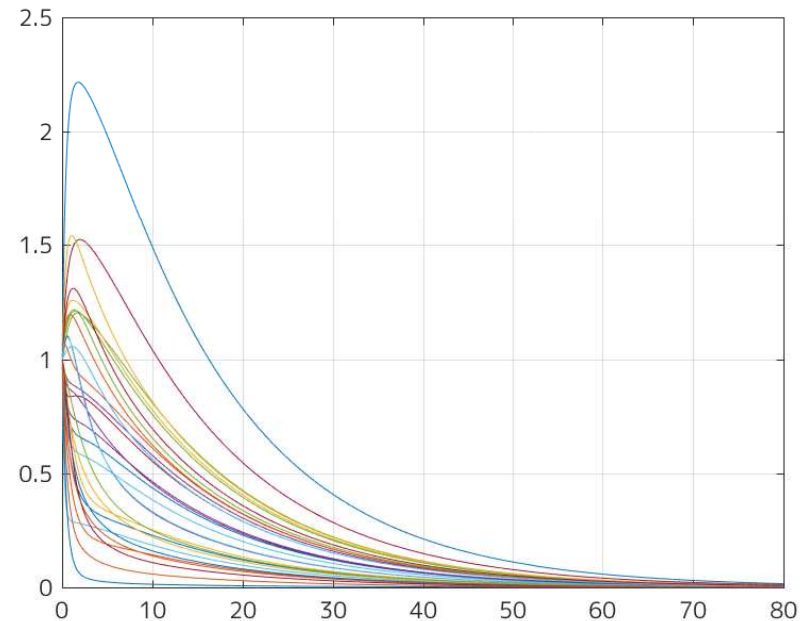
Centralized Approach: 37.80 [sec] ( $\approx 0.18 \times 210$ )

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Initial Response with  $x_i(0) = \mathbf{1}_3$



$$\mathcal{N}_c = \{2, 3, 4, 5\}$$



$$\mathcal{N}_c = \{2, 3, 5, 10\}$$

# Conclusion

- Copositive Lyapunov Functions
- Weighted  $L_1$ -induced Norm
- Stability of Interconnected Positive Systems
  - characterized by weighted  $L_1$ -induced norm



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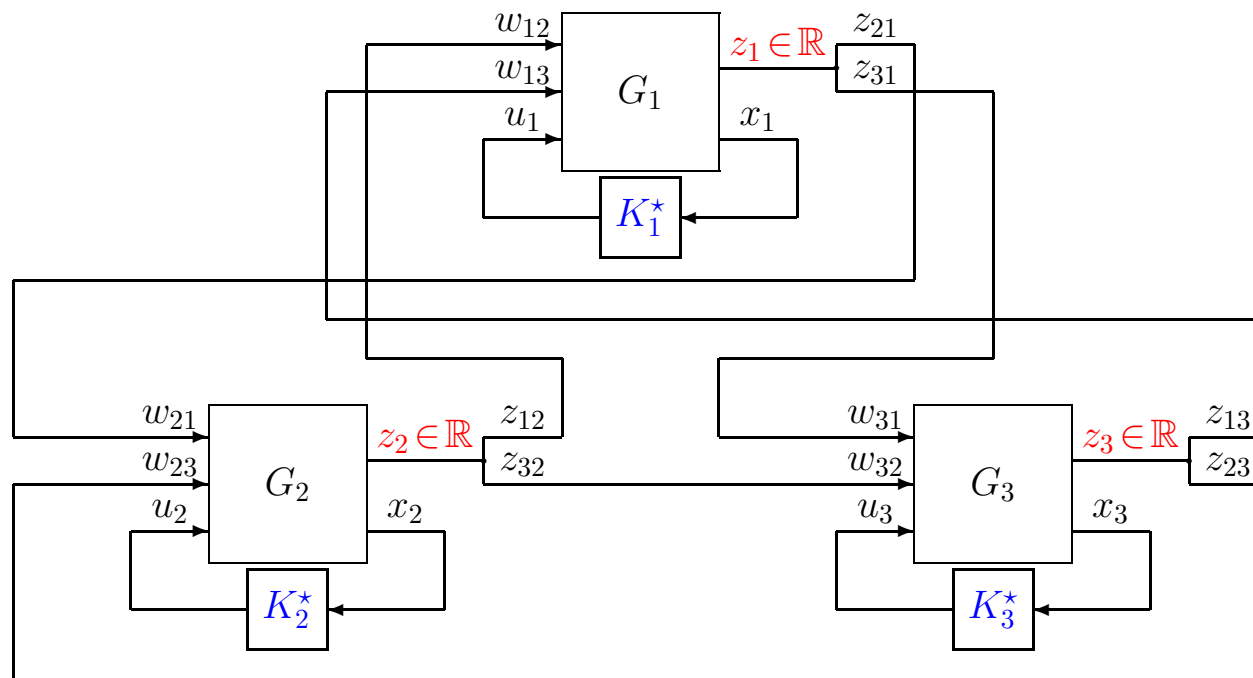
strong analysis results  
using positivity

# Conclusion

- Decentralized Stabilizing SF Synthesis

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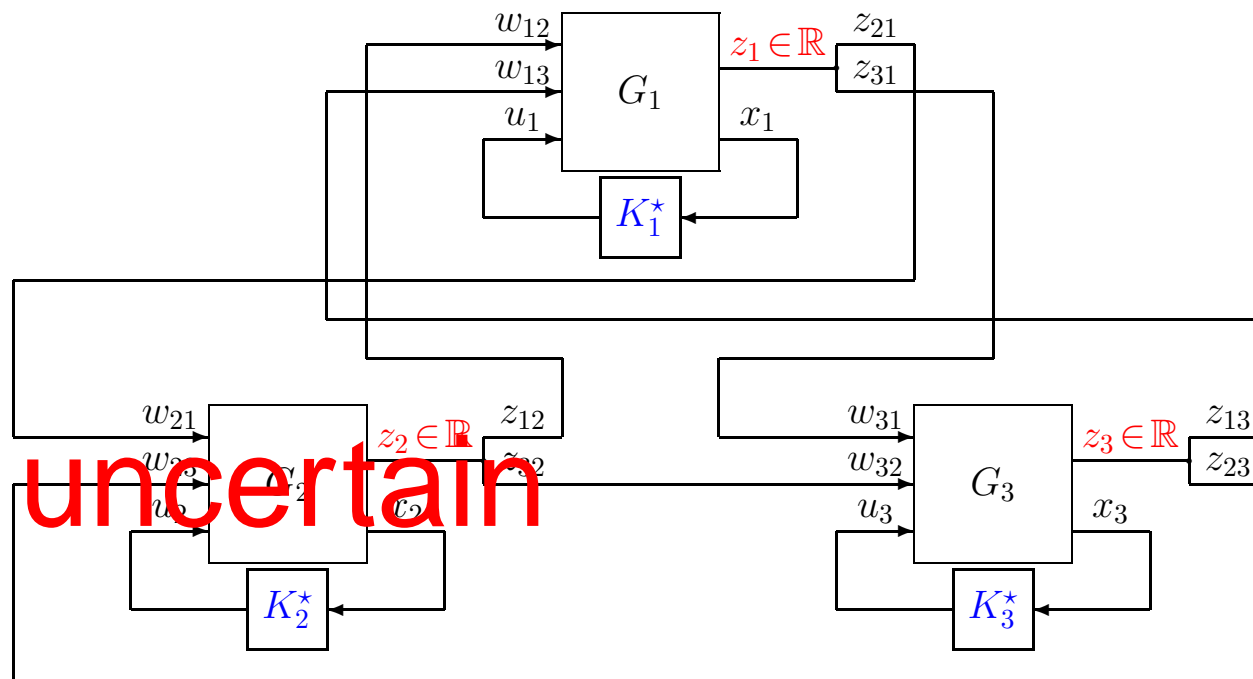
## Decentralized Stabilizing SF Synthesis



- synthesis of  $K_i^*$  purely locally by LP
- optimality property of  $K^* = \{K_1^*, \dots, K_N^*\}$ 
  - minimizes  $\lambda_F(A_K)$

# Conclusion

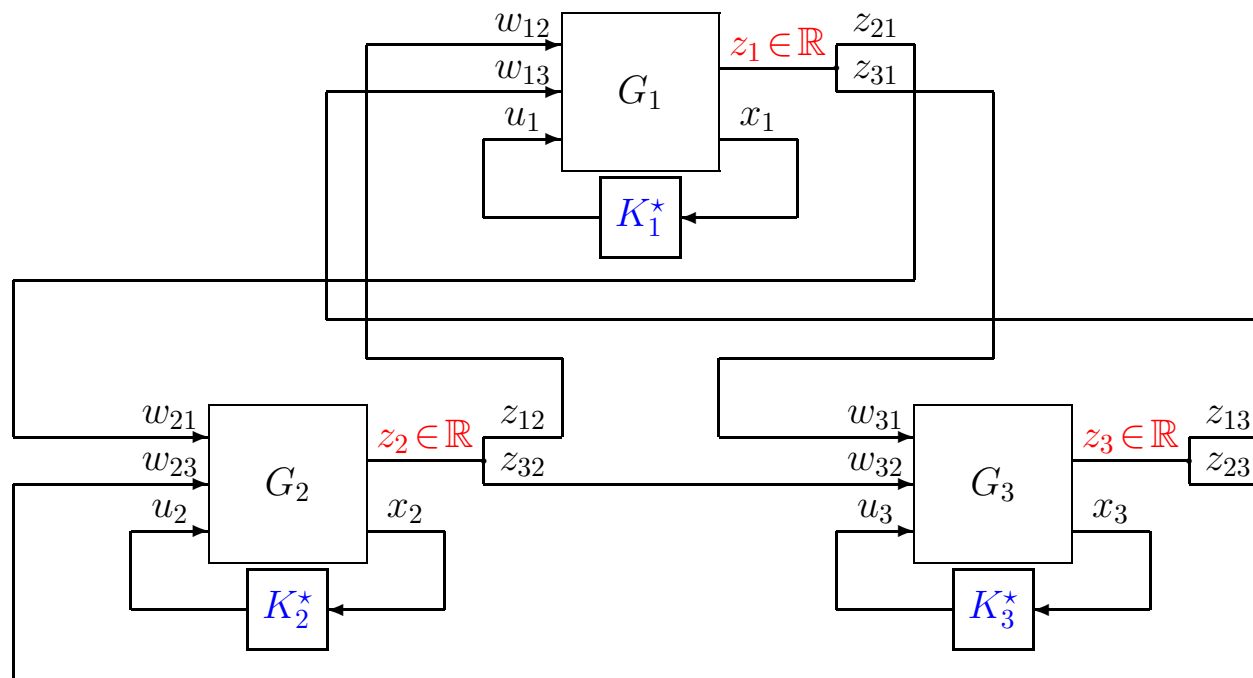
## Decentralized Stabilizing SF Synthesis



- $K_1^*$  and  $K_3^*$  remains to be optimal even if  $G_2$  is totally uncertain
- centralized approach infeasible

# Conclusion

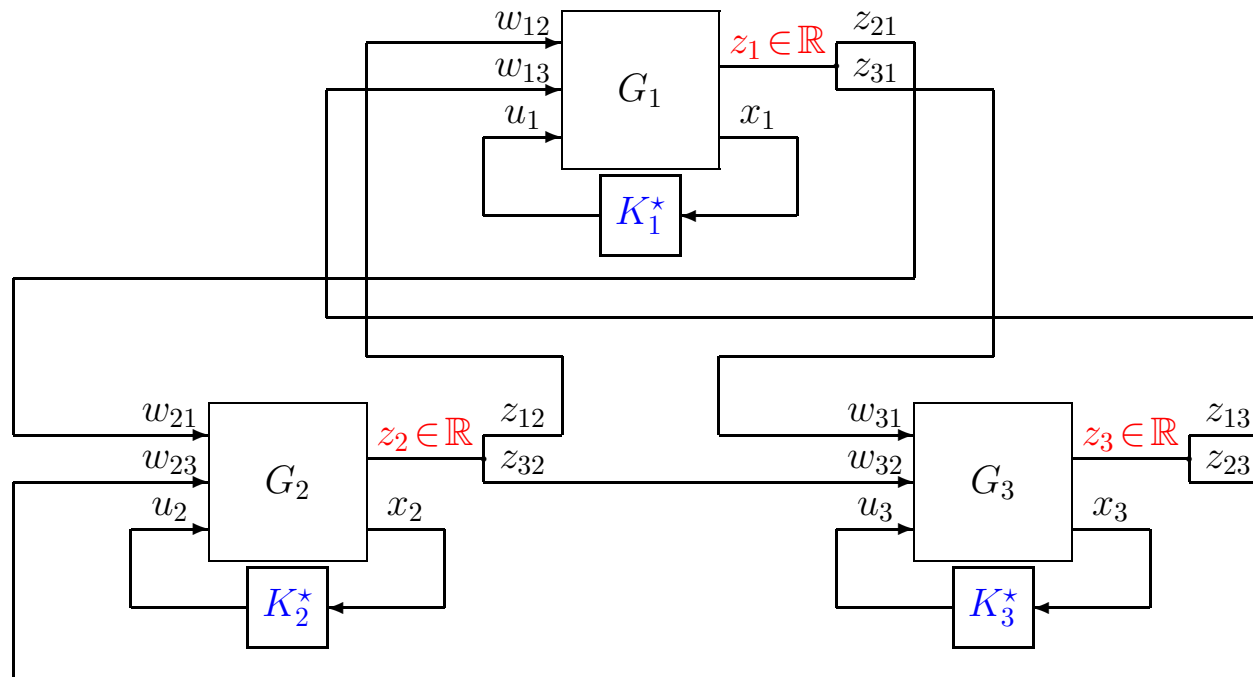
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