

# Dimensionality Reduction of the Pinning Control Problem for Network Synchronization

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- Dynamical Network
  - Definition
  - Representation
- Collective behavior
  - Synchronization
- Control
  - Pinning control

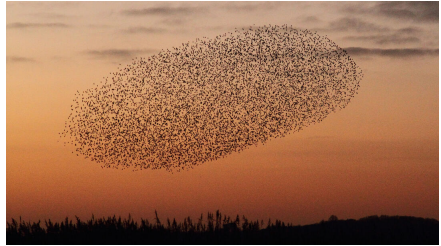


Figure: Starling murmuration <sup>a</sup>

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<sup>a</sup>Photo: Donald Macauley

# Dynamical Network

## Overview

- Dynamical Network
  - Definition: Interactive dynamical units
    - ▶ Social
    - ▶ Economical
    - ▶ Biological
    - ▶ etc.
  - Representation: Graph Theory
    - ▶ Nodes
    - ▶ Links

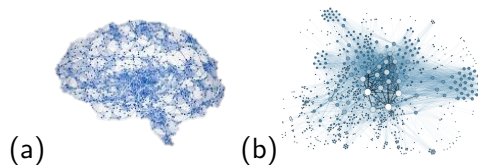


Figure: (a) Neural Network. (b) Graph

- Homogeneous,
  - Same dynamical system in each node, AND
  - Same interaction function between the nodes
- Heterogeneous
  - 1 Different Dynamical System in each nodes, OR
  - 2 Different interaction function between the nodes

- A general equation for a homogeneous network of coupled dynamical systems

$$\dot{\mathbf{x}}_i(t) = \mathbf{F}(\mathbf{x}_i(t)) + \sigma \sum_{j=1}^N a_{ij} \mathbf{H}(\mathbf{x}_j(t)) \quad i = 1, \dots, N \quad (1)$$

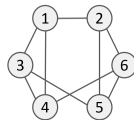
- $\mathbf{x}_i$ : the m-dimensional state vector of the  $i^{th}$  dynamical system (oscillator),
- $\mathbf{F}$ : the dynamics of each oscillator,
- $A$ : the connectivity matrix (topology of the network connections)
- $\sigma$ : the overall coupling strength
- $\mathbf{H}$ : the output function

# Synchronization

$$\dot{\mathbf{x}}_i(t) = \mathbf{F}(\mathbf{x}_i(t)) + \sigma \sum_{j=1}^N a_{ij} \mathbf{H}(\mathbf{x}_j(t)) \quad i = 1, \dots, N$$

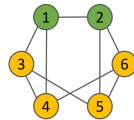
## (a) Complete synchronization in homogeneous networks

- Matrix  $A$ : Constant row sum (Usually Laplacian matrix  $L$ )
- Synchronization manifold:  $\mathbf{s}(t) = \mathbf{x}_1 = \mathbf{x}_2 = \dots = \mathbf{x}_N$



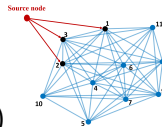
## (b) Cluster synchronization

- Cluster Synchronization of Networks via a Canonical Transformation for Simultaneous Block Diagonalization of Matrices



## (c) Complete synchronization in heterogeneous networks (links of different types)

- Pinning control of networks: dimensionality reduction through simultaneous block-diagonalization of matrices



# Complete Synchronization

## Homogeneous network

$$\dot{\mathbf{x}}_i(t) = \mathbf{F}(\mathbf{x}_i(t)) + \sigma \sum_{j=1}^N l_{ij} \mathbf{H}(\mathbf{x}_j(t)) \quad i = 1, \dots, N \quad (2)$$

- Synchronization,
  - Complete Synchronization,
  - Matrix  $L$ : Laplacian matrix (zero-row sum)
  - Synchronization manifold:  $\mathbf{s}(t) = \mathbf{x}_1 = \mathbf{x}_2 = \dots = \mathbf{x}_N$
- Stability analysis
  - Dynamics on synchronous manifold:  $\dot{\mathbf{s}}(t) = \mathbf{F}(\mathbf{s}(t))$ .
- Is this manifold stable?

# Stability analysis

## Master Stability Approach

- $(\mathbf{x}_i(t) = \mathbf{s}(t) + \delta\mathbf{x}_i(t))$
- $\lim_{t \rightarrow \infty} \delta\mathbf{x}_i(t) = 0$
- Linearized system of equations

$$\delta\dot{\mathbf{x}}_i(t) = D\mathbf{F}(\mathbf{s}(t))\delta\mathbf{x}_i(t) + \sigma \sum_{j=1}^N l_{ij} \mathbf{H}(\mathbf{s}(t))\delta\mathbf{x}_j(t) \quad i = 1, \dots, N \quad (3)$$

- Master Stability approach (Pecora & Carroll (1998))
  - Diagonalizing the Laplacian matrix  $L$  using eigenvalue decomposition
  - $\boldsymbol{\eta}(t) = (V \otimes I_m)\delta\mathbf{x}(t)$



# Master Stability Function (MSF)

## Diagonalizing

- Master stability equations: a set of independently evolving perturbations:

$$\dot{\boldsymbol{\eta}}_j(t) = [D\mathbf{F}(\mathbf{s}(t)) - \sigma\lambda_j D\mathbf{H}(\mathbf{s}(t))]\boldsymbol{\eta}_j(t), \quad (4)$$

where

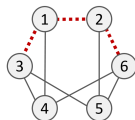
- $$\begin{cases} \lambda_1 = 0 & \text{parallel perturbation} \\ \lambda_j \quad j = 2, \dots, N & \text{transverse perturbation} \end{cases}$$

- MSF  $\mathcal{M}(\sigma\lambda_i)$ : The largest transverse Lyapunov exponent as a function of the parameter  $\sigma\lambda_i$ .

# Stability analysis

## Heterogeneous Network: Networks with different types of interactions

- The case of a heterogeneous networks with different types of interactions was first considered by F. Sorrentino in this paper: "*Synchronization of hypernetworks of coupled dynamical systems*, (2012)"



$$\dot{\mathbf{x}}_i(t) = \mathbf{F}(\mathbf{x}_i(t)) + \sigma_A \sum_{j=1}^N l_{ij}^A \mathbf{H}_A(\mathbf{x}_j(t)) + \sigma_B \sum_{j=1}^N l_{ij}^B \mathbf{H}_B(\mathbf{x}_j(t))$$

$$\delta \dot{\mathbf{x}}_i(t) = D\mathbf{F}(\mathbf{s}(t))\delta \mathbf{x}_i + \sigma_A \sum_{j=1}^N l_{ij}^A D\mathbf{H}_A(\mathbf{s}(t))\delta \mathbf{x}_j + \sigma_B \sum_{j=1}^N l_{ij}^B D\mathbf{H}_B(\mathbf{s}(t))\delta \mathbf{x}_j \quad (5)$$

- The two Laplacian matrices  $L^A$  and  $L^B$  commute
- One of the networks (either  $L^A$  or  $L^B$ ) is unweighted and fully connected
- One of the two networks (say, e.g.,  $L^A$ ) is such that  $L_{ij}^A = a_j$  for  $i, j = 1, \dots, N$

# Simultaneous Block Diagonalization

Given a set of  $n \times n$  real matrices  $\{A_1, A_2, \dots, A_N\}$ , find an  $n \times n$  orthogonal matrix  $P$  such that  $P^T A_1 P, \dots, P^T A_N P$  are in a common block-diagonal form.

- Maehara, T., Murota, K. (2011)

$\mathcal{M}_n :=$  Set of  $n \times n$  matrices  $(\{A_1, A_2, \dots, A_N\})$

**Definition:** A matrix  $*$ -algebra (subalgebra)  $T$ :

Subset of  $\mathcal{M}_n$  such that if

- 1)  $I_n \in T$  AND
- 2)  $A, B \in T$  AND
- 3)  $\alpha, \beta \in \mathbb{R}$

then  $\Rightarrow \alpha A + \beta B, AB, A^T \in T$

**Definition:**  $T'$ : Commutant algebra of  $T$ :

- 1) forms a matrix  $*$ -algebra
- 2) contains set of all matrices  $X$  that commute with every member of  $T$

# Simultaneous Block Diagonalization

- **Theorem** There exist an orthogonal matrix  $Q$  such that:

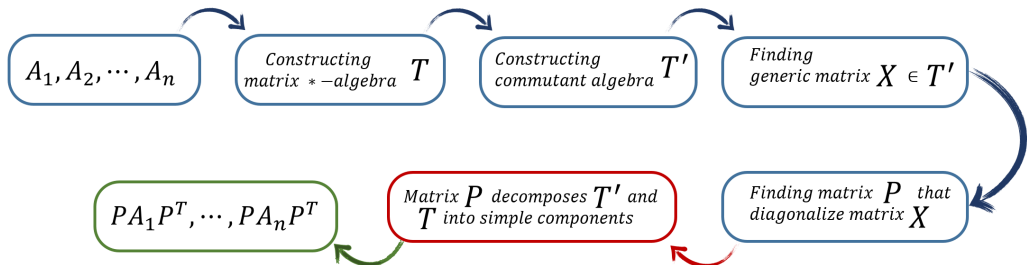
$$Q^T T Q = \oplus_I^N T_j, \quad (6)$$

where each  $T_j$  is not further reducible.

- **Definition::Generic matrix** If we sample a symmetric matrix  $A \in T$  randomly, then  $A$  is generic with probability one.
- **Proposition.** Let  $A \in T$  be a generic symmetric matrix. Then an orthogonal matrix that diagonalizes  $A$  decomposes  $T$  into simple components.

- **Proposition.** If an orthogonal matrix  $Q$  decomposes a matrix  $*$ -algebra  $T$  into the simple components then  $Q$  also decomposes the commutant algebra  $T'$  into the simple components.
- \* To find  $Q$  we need to sample a generic matrix  $X$  in  $T'$

# SBD Overview



## Finding matrix $P$

- A matrix  $X$  needs to simultaneously commute with each one of the matrices  $A_k$  which means  $[A_k, X] = A_k X - X A_k = O_n$ ,  $k = 1, \dots, M$ ,
- Define the vectorizing function  $\text{vec} : \mathbb{R}^{n \times m} \mapsto \mathbb{R}^{nm}$ , we have:

$$\text{vec}(A_k X - X A_k) = \text{vec}(A_k X) - \text{vec}(X A_k) = \text{vec}(O_n), \quad k = 1, 2, \dots, M \quad (7)$$

- To find the matrix  $X$ :
  - Define matrix  $Y_k = (I_n \otimes A_k - (A_k)^T \otimes I_n)$
  - Look for a vector  $\text{vec}(X) \in \bigcap_{k=1}^M \mathcal{N}(Y_k)$
- By calculating the matrix  $X$ , the matrix  $P$  is constructed as the eigenvectors of the matrix  $X$ .

# Application of SBD to Synchronization of Networks

- Irving, D., Sorrentino, F. (**2012**)
  - Complete Synchronization of Heterogeneous Networks
- Zhang, Y., Motter, A. E. (**2020**)
  - Cluster Synchronization
- Panahi, S., Klickstein, I., Sorrentino, F. (**2021**).
  - Canonical Transformation
- Panahi, S., Lodi, M., Storace, M., Klickstein, I., Novello, G., Sorrentino, F.
  - Pinning control problem



# SBD & Cluster Synchronization

"Cluster Synchronization of Networks via a Canonical Transformation for Simultaneous Block Diagonalization of Matrices, Chaos, 2021"

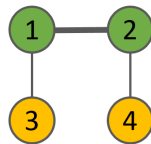
- **Goal:** Proposing a canonical transformation for simultaneous block diagonalization of matrices, applicable for investigating the cluster synchronization of networks
- **Advantages:**
  - 1) Decouple the stability problem into subproblems of minimal dimensionality while preserving physically meaningful information
  - 2) Applicable for both orbital and equitable partitions of the network nodes
  - 3) Parametrization of the problem in a small number of parameters

# Cluster synchronization

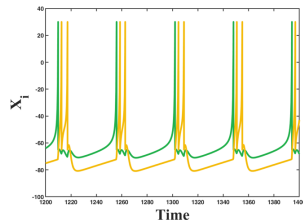
Given a network with adjacency matrix  $A$ :

$$\dot{\mathbf{x}}_i(t) = \mathbf{F}(\mathbf{x}_i(t)) + \sum_{j=1}^N a_{ij} \mathbf{H}(\mathbf{x}_j(t)) \quad i = 1, \dots, N$$

- **Clusters:** The network nodes ( $\mathcal{V}$ ) can be partitioned into  $C$  numbers of clusters  $\mathcal{C}_1, \mathcal{C}_2, \dots, \mathcal{C}_C$ ,  $\cup_{k=1}^C \mathcal{C}_k = \mathcal{V}$ ,  $\mathcal{C}_k \cap \mathcal{C}_\ell = \emptyset$  for  $k \neq \ell$ ,
- **Coloring of the Nodes:** Partitioning of the nodes of the network induces a colored network, where each node  $i$  is assigned a color  $k$  if node  $i$  is in cluster  $\mathcal{C}_k$ .
- **Cluster Synchronization:** Nodes in each cluster synchronize on the same time evolution but these time evolutions are different for nodes in different clusters

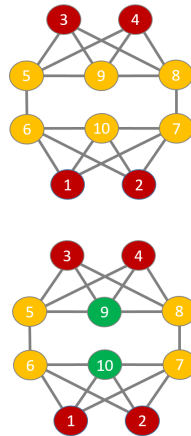


$$A = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$



# Cluster synchronization

- **Equitable Clusters:** all the nodes in one cluster has exactly the same number of neighbors in the other clusters regardless of the choice of the nodes
- **Orbital Clusters:** Nodes in the same orbital cluster can be swapped with each other without changing the overall network topology.



# Cluster synchronization

- **Quotient network:** Describes a network for which all the nodes in each cluster collapse to a single quotient node.



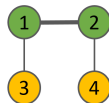
- **Quotient matrix:**  $Q_{kl} = \sum_{j \in \mathcal{C}_l} A_{ij} \quad i \in \mathcal{C}_k.$

- **Cluster synchronization dynamics:**

$$\dot{s}_k(t) = F(s_k(t)) + \sum_{l=1}^C Q_{kl} H(s_l(t)), \quad k, l = 1, 2, \dots, C,$$

- **Cluster Indicator**

**matrix:**  $E_k = \{e_{kij}\}_{N \times N}$ ,  $e_{kij} = 1$  if node  $i \in \mathcal{C}_k$ , otherwise 0.



$$E_1 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad E_2 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

# Stability Analysis

- $\delta \mathbf{x}_i = (\mathbf{x}_i - \mathbf{s}_k), i \in \mathcal{C}_k$

$$\delta \dot{\mathbf{x}}(t) = \left[ \sum_{c=1}^C E_c \otimes D\mathbf{F}(\mathbf{s}_c(t)) + \sum_{c=1}^C A E_c \otimes D\mathbf{H}(\mathbf{s}_c(t)) \right] \delta \mathbf{x}(t), \quad (8)$$

- Is it possible to reduce the  $mN$ -dimensional stability problem into a set of independent lower-dimensional equations?
- It is needed to simultaneously block diagonalize the set of  $C + 1$  matrices  $\{A, E_1, E_2, \dots, E_C\}$ 
  - IRR transformation (Pecora, L. M., Sorrentino, F., Hagerstrom, A. M., Murphy, T. E., & Roy, R. (2014))  $\Rightarrow$  [Orbital clusters](#)
  - SBD technique (Zhang, Y., & Motter, A. E. (2020))

# Canonical SBD

## Finding matrix $P$

- **Assumption:** Without loss of generality  $\Rightarrow$  we can order the network nodes so that:

$$E_1 = \begin{pmatrix} I_{n_1} & 0 \\ 0 & 0_{N-n_1} \end{pmatrix} \quad E_2 = \begin{pmatrix} 0_{n_1} & 0 & 0 \\ 0 & I_{n_2} & 0 \\ 0 & 0 & 0_{N-(n_1+n_2)} \end{pmatrix} \quad \cdots \quad E_C = \begin{pmatrix} 0_{N-n_C} & 0 \\ 0 & I_{n_C} \end{pmatrix} \quad (9)$$

- Lemma 1. Any matrix  $P$  that commutes with the set of matrices  $E_i$  has the following block-diagonal structure,

$$P = \begin{pmatrix} P_1 & 0 & \cdots & 0 \\ 0 & P_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & P_C \end{pmatrix} \quad (10)$$

- The transformation matrix  $T$ :
  - the eigenvectors of the matrix  $P$  for its columns
  - must also have the same block-diagonal structure as the matrix  $P$

$$T = \begin{pmatrix} T_1 & 0 & \cdots & 0 \\ 0 & T_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & T_C \end{pmatrix} \quad (11)$$

- $T^T = T^{-1} \Rightarrow T_k^T = T_k^{-1}$
- Each block  $T_k$  corresponds to each cluster  $\mathcal{C}_k \Rightarrow$  each block describes stability of either one cluster or a set of intertwined clusters
- After applying  $T$ ,  $\exists$  one block corresponds to the quotient dynamics.
- $J_i = TE_iT^T = E_i$  for  $i = 1, 2, \dots, C \Rightarrow T$  is 'Canonical'

- After applying  $T$ , we obtain,

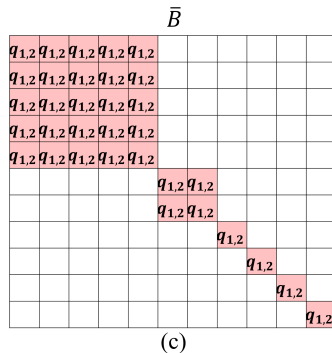
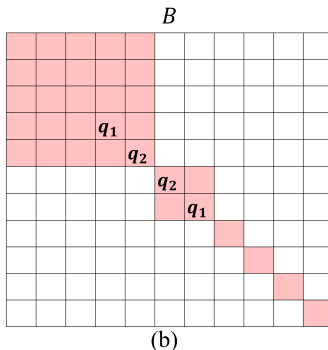
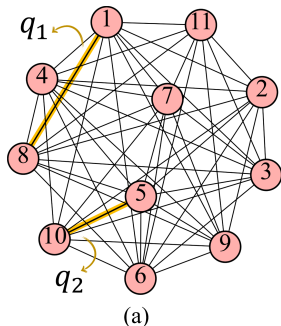
$$\dot{\eta}(t) = \left[ \sum_{k=1}^C J_k \otimes D\mathbf{F}(\mathbf{s}_k(t)) + \sum_{k=1}^C B J_k \otimes D\mathbf{H}(\mathbf{s}_k(t)) \right] \boldsymbol{\eta}(t), \quad (12)$$

$$\begin{aligned} \dot{\mathbf{Y}} = T^{-1} \delta \dot{\mathbf{X}} &= \frac{\sqrt{2}}{2} \begin{pmatrix} \delta \dot{\mathbf{x}}_1 + \delta \dot{\mathbf{x}}_2 \\ -\delta \dot{\mathbf{x}}_3 - \delta \dot{\mathbf{x}}_4 \\ \delta \dot{\mathbf{x}}_1 - \delta \dot{\mathbf{x}}_2 \\ \delta \dot{\mathbf{x}}_3 - \delta \dot{\mathbf{x}}_4 \end{pmatrix} = \begin{pmatrix} D\mathbf{F}(\mathbf{s}_1) & 0 & 0 & 0 \\ 0 & D\mathbf{F}(\mathbf{s}_2) & 0 & 0 \\ 0 & 0 & D\mathbf{F}(\mathbf{s}_1) & 0 \\ 0 & 0 & 0 & D\mathbf{F}(\mathbf{s}_2) \end{pmatrix} \mathbf{Y} \\ &+ \begin{pmatrix} 0 & -D\mathbf{H}(\mathbf{s}_2) & 0 & 0 \\ -D\mathbf{H}(\mathbf{s}_1) & D\mathbf{H}(\mathbf{s}_2) & 0 & 0 \\ 0 & 0 & 0 & D\mathbf{H}(\mathbf{s}_2) \\ 0 & 0 & D\mathbf{H}(\mathbf{s}_1) & -D\mathbf{H}(\mathbf{s}_2) \end{pmatrix} \mathbf{Y}, \end{aligned}$$



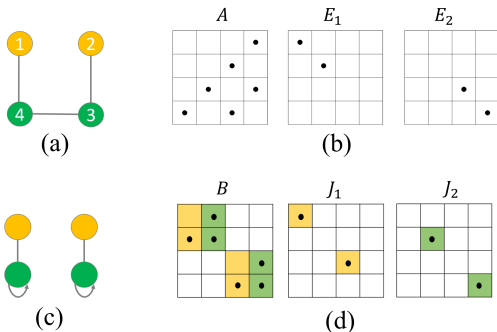
## Advantages:

- 1) Compared to previous work [Zhang and Motter (2020)] it has lower number of nonzero entries that parametrize the matrices after the transformation



## Advantages:

- 2) Each network node after the transformation belongs to one and only one cluster
  - Canonical SBD



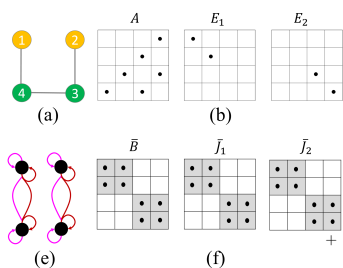
$$\dot{\mathbf{Y}} = T^{-1} \delta \dot{\mathbf{X}} = \frac{\sqrt{2}}{2} \begin{pmatrix} \delta \dot{\mathbf{x}}_1 + \delta \dot{\mathbf{x}}_2 \\ -\delta \dot{\mathbf{x}}_3 - \delta \dot{\mathbf{x}}_4 \\ \delta \dot{\mathbf{x}}_1 - \delta \dot{\mathbf{x}}_2 \\ \delta \dot{\mathbf{x}}_3 - \delta \dot{\mathbf{x}}_4 \end{pmatrix} = \begin{pmatrix} DF(s_1) & 0 & 0 & 0 \\ 0 & DF(s_2) & 0 & 0 \\ 0 & 0 & DF(s_1) & 0 \\ 0 & 0 & 0 & DF(s_2) \end{pmatrix} \mathbf{Y}$$

$$+ \begin{pmatrix} 0 & -DH(s_2) & 0 & 0 \\ -DH(s_1) & DH(s_2) & 0 & 0 \\ 0 & 0 & 0 & DH(s_2) \\ 0 & 0 & DH(s_1) & -DH(s_2) \end{pmatrix} \mathbf{Y},$$

# Canonical SBD

$$\dot{\eta}(t) = \left[ \sum_{k=1}^C J_k \otimes DF(s_k(t)) + \sum_{k=1}^C BJ_k \otimes DH(s_k(t)) \right] \eta(t),$$

- $\tilde{T}$  SBD transformation

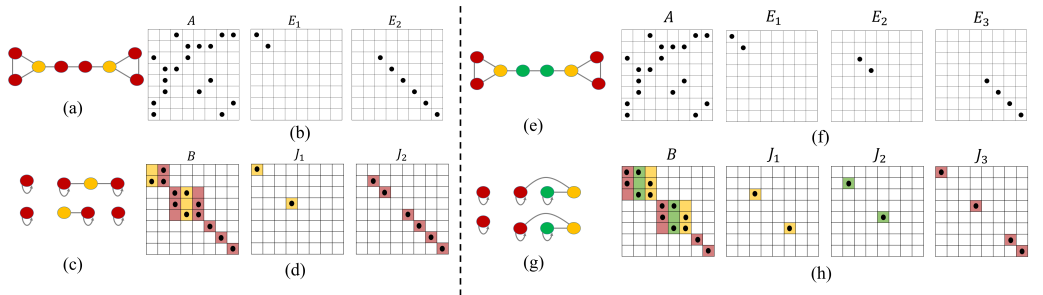


$$\dot{\mathbf{z}} = \tilde{T}^{-1} \delta \dot{\mathbf{x}} = \begin{pmatrix} -0.65\delta\dot{\mathbf{x}}_1 + 0.65\delta\dot{\mathbf{x}}_2 - 0.26\delta\dot{\mathbf{x}}_3 + 0.26\delta\dot{\mathbf{x}}_4 \\ -0.26\delta\dot{\mathbf{x}}_1 + 0.26\delta\dot{\mathbf{x}}_2 + 0.65\delta\dot{\mathbf{x}}_3 - 0.65\delta\dot{\mathbf{x}}_4 \\ 0.69\delta\dot{\mathbf{x}}_1 + 0.69\delta\dot{\mathbf{x}}_2 - 0.16\delta\dot{\mathbf{x}}_3 - 0.16\delta\dot{\mathbf{x}}_4 \\ 0.16\delta\dot{\mathbf{x}}_1 + 0.16\delta\dot{\mathbf{x}}_2 + 0.69\delta\dot{\mathbf{x}}_3 + 0.69\delta\dot{\mathbf{x}}_4 \end{pmatrix}$$

$$= \begin{pmatrix} 0.14DF(s_1) + 0.86DF(s_2) & 0.35DF(s_1) - 0.35DF(s_2) & 0 & 0 \\ 0.35DF(s_1) - 0.35DF(s_2) & 0.86DF(s_1) + 0.14DF(s_2) & 0 & 0 \\ 0 & 0 & 0.05DF(s_1) + 0.95DF(s_2) & 0.22DF(s_1) - 0.22DF(s_2) \\ 0 & 0 & 0.22DF(s_1) - 0.22DF(s_2) & 0.95DF(s_1) + 0.05DF(s_2) \end{pmatrix} \mathbf{z} + \begin{pmatrix} 0.35DH(s_1) - 0.51DH(s_2) & 0.86DH(s_1) + 0.20DH(s_2) & 0 & 0 \\ -0.14DH(s_1) + 1.2DH(s_2) & -0.35DH(s_1) - 0.49DH(s_2) & 0 & 0 \\ 0 & 0 & 0.22DH(s_1) + 1.17DH(s_2) & 0.95DH(s_1) - 0.27DH(s_2) \\ 0 & 0 & -0.05DH(s_1) + 0.72DH(s_2) & -0.22DH(s_1) - 0.17DH(s_2) \end{pmatrix} \mathbf{z}$$

## Advantages:

- 3) It is applicable for both orbital and equitable partitioning of the network nodes.



# Application of the Canonical Transformation in Real Networks

**Table:** Real networks analysis.  $N$  is the number of nodes,  $E$  the number of edges,  $N_{ntc}$  is the number of nontrivial clusters,  $\max(|n_c|)$  is the size of the largest equitable cluster. We include the average runtime in seconds for calculation of the transformation matrix  $T$  and of the transformation matrix  $\tilde{T}$  using the code from [Zhang and Motter (2020)].

Name	$N$	$E$	$N_{ntc}$	$\max( n_c )$	Average Runtime for $\tilde{T}$	Average Runtime for $T$
ca-netscience: Scientist Collaboration Network	379	914	70	6	25.0745	6.4077
Chilean Power Grid Network	218	527	29	7	2.9198	1.9107
Power Grid Network of Western Germany	491	665	43	5	41.7946	14.4780
celegans-dir: Metabolic Network	453	2025	28	4	24.6675	12.6427
Us Airline	332	2126	31	12	11.0195	5.8412
Erdos971	429	1312	20	3	16.92	10.0446
bio-diseasome: Biological Network	516	1188	95	6	90.5763	13.9930
fb-forum: Social network	899	7036	16	5	171.0608	65.6710

# SBD and Pinning Control

Pinning control of networks:

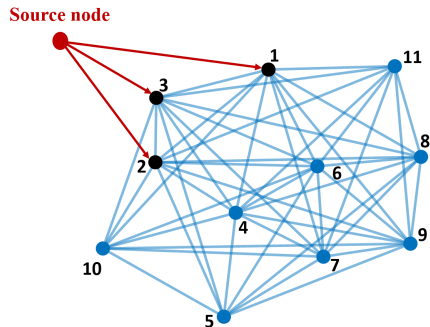
dimensionality reduction through simultaneous block-diagonalization of matrices

Considering pinning control signals :  $\mathbf{u}_i(t)$

$$\dot{\mathbf{x}}_i(t) = \mathbf{F}(\mathbf{x}_i(t)) + \sum_{j=1}^N L_{ij} \mathbf{G}(\mathbf{x}_j(t)) + \mathbf{u}_i(t), \quad (13a)$$

$$\mathbf{u}_i(t) = \gamma r_i [\mathbf{H}(\mathbf{x}_t(t)) - \mathbf{H}(\mathbf{x}_i(t))], \quad (13b)$$
$$i = 1, \dots, N$$

- Previous work (see e.g., Sorrentino, F., Di Bernardo, M., Garofalo, F., Chen, G. (2007)) has dealt with  $\mathbf{H} = \mathbf{G}$ .



# Synchronization

- Completely synchronized solution  $\mathbf{x}_1(t) = \mathbf{x}_2(t) = \dots = \mathbf{x}_n(t) = \mathbf{x}_s(t)$ , which obeys,

$$\dot{\mathbf{x}}_s(t) = \mathbf{F}(\mathbf{x}_s(t)). \quad (14)$$

- ‘Target’ synchronous solution,  $\mathbf{x}_t(t)$  produced by the initial condition  $\mathbf{x}_t^0$ ,

$$\dot{\mathbf{x}}_t(t) = \mathbf{F}(\mathbf{x}_t(t)), \quad \mathbf{x}_t(0) = \mathbf{x}_t^0. \quad (15)$$

- Considering small perturbations  $\delta\mathbf{x}_i = (\mathbf{x}_i - \mathbf{x}_t)$  and stacking them together in one vector:

$$\begin{aligned} \delta\dot{\mathbf{X}}(t) = & [I_N \otimes D\mathbf{F}(\mathbf{x}_t(t)) + L \otimes D\mathbf{G}(\mathbf{x}_t(t)) \\ & - \gamma R \otimes D\mathbf{H}(\mathbf{x}_t(t))] \delta\mathbf{X}(t), \end{aligned} \quad (16)$$

## SBD and Pining control problem

$$\delta \dot{\mathbf{X}}(t) = [I_N \otimes D\mathbf{F}(\mathbf{x}_t(t)) + L \otimes D\mathbf{G}(\mathbf{x}_t(t)) - \gamma R \otimes D\mathbf{H}(\mathbf{x}_t(t))] \delta \mathbf{X}(t).$$

We need to find transformation matrix  $T = SBD(R, L)$

- Finding matrix  $P$ :
  - matrix  $R$  can be rewritten,

$$R = \begin{pmatrix} I_s & 0 \\ 0 & \mathbf{0}_{N-s} \end{pmatrix}, \quad (17)$$

- The matrix  $P$  has the following block-diagonal structure,

$$P = \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix} \quad (18)$$

- Transformation matrix  $T$

$$T = \begin{pmatrix} T_1 & 0 \\ 0 & T_2 \end{pmatrix}, \quad (19)$$



# Driven and Undriven Blocks

- By applying Transformation  $T$  to the pair  $(L, R)$ , we have:

$$\begin{aligned} T^{-1}LT &= L_T = \oplus_{j=1}^I \hat{L}_j, \\ T^{-1}RT &= R_T = \oplus_{j=1}^I \hat{R}_j, \end{aligned} \quad (20)$$

- $L_T$  and  $R_T$  are block-diagonal:

$$L_T = \begin{pmatrix} L_d & 0 \\ 0 & L_{ud} \end{pmatrix} \quad R_T = \begin{pmatrix} R_d & 0 \\ 0 & 0_{(N-c)} \end{pmatrix}, \quad (21)$$

- Equation (16) can be split into the one driven and one undriven equations,

$$\dot{\eta}_d(t) = [I_c \otimes D\mathbf{F}(\mathbf{x}_s(t)) + L_d \otimes D\mathbf{G}(\mathbf{x}_s(t)) - \gamma R_d \otimes D\mathbf{H}(\mathbf{x}_s(t))]\eta_d(t),$$

$$\dot{\eta}_{ud}(t) = [I_{N-c} \otimes D\mathbf{F}(\mathbf{x}_s(t)) + L_{ud} \otimes D\mathbf{G}(\mathbf{x}_s(t))]\eta_{ud}(t). \quad (22)$$

- Dimension of the driven pair  $(L_d, R_d)$  = the rank of the controllability matrix of the pair  $(L, R)$ .

# Transformation $\hat{T}$

- $\hat{T} = T_c T_q$ 
  - $T_c$  : Decouples the problem (16) into a controllable equation and an uncontrollable equation
  - $T_q$  : Decouples the problem (16) into a quotient equation and a redundant equation.
- By applying  $\hat{T}$  to the pair  $(L, R)$ , we have:

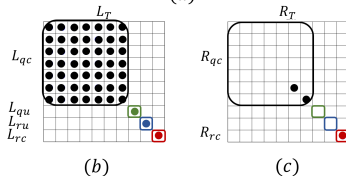
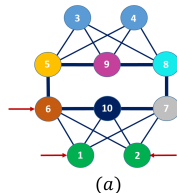
$$\hat{T}L\hat{T}^{-1} = \begin{pmatrix} L_{qc} & 0 & 0 & 0 \\ 0 & L_{rc} & 0 & 0 \\ 0 & 0 & L_{qu} & 0 \\ 0 & 0 & 0 & L_{ru} \end{pmatrix} \hat{T}R\hat{T}^{-1} = \begin{pmatrix} R_{qc} & 0 & 0 & 0 \\ 0 & R_{rc} & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}, \quad (23)$$

# Application of transformation $\hat{T}$

- The pair  $(L, R)$  is decoupled by application of the transformation  $\hat{T}$  into four blocks,

Pair	Type	Size
$(L_{qc}, R_{qc})$	Quotient controllable block (qc)	$u$
$(L_{rc}, R_{rc})$	Redundant controllable block (rc)	$c - u$
$(L_{qu}, 0)$	Quotient uncontrollable block (qu)	$N - c$
$(L_{ru}, 0)$	Redundant uncontrollable block (ru)	

- The transformation  $\hat{T}$  leads to a finest SBD of the pair  $(L, R)$ .



(c)

# Pinned node selection and multiple driven blocks

- Dimension of the driven pair is affected by
  - Topology of the network
  - Number and choice of the pinned nodes
- it is possible to have the multiple driven pairs  $(L_d^1, R_d^1), (L_d^2, R_d^2), \dots, (L_d^w, R_d^w)$ , where  $L_d \oplus_{i=1}^w L_d^i$  and  $R_d \oplus_{i=1}^w R_d^i$ .
  - A pinned node symmetry (PNS) is a symmetry of the pair  $(L, R)$  i.e. a permutation matrix  $\Pi$  that
    - ▶ commutes with both matrices  $L$  and  $R$
    - ▶ it swaps two or more pinned nodes
  - Choosing two or more pinned nodes which belong to the same PNS results into independent driven pairs for each one of these pinned nodes.

- Application of SBD to Synchronization of Dynamical Networks
  - Canonical SBD Transformation on Cluster Synchronization
    - ▶ MATLAB code to compute the canonical SBD for the cluster synchronization is available online and can be accessed from <https://github.com/SPanahi/Clustered-SBD>
  - SBD & Pinning Control Problem in Heterogeneous Networks

Thank you!