Synchronization of heterogeneous oscillators under network modifications: Perturbation and optimization of the synchrony alignment function

Dane Taylor, Per Sebastian Skardal, Jie Sun SIAM Journal on Applied Dynamical Systems 76(5), 1984-2008 (2016)

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## Synchronization

#### Engineering applications:

- Power grids, smart grids
  - Josephson junctions
- Synthetic cell engineering

#### Examples in biology:

- Neuronal activity
- Cardiac pacemaker cells
  - Circadian rhythms



## Kuramoto Phase-Oscillator Model

-Kuramoto, Chemical Oscillations, Waves, and Turbulence (1984)

$$\dot{\theta}_n = \omega_n + K \sum_m A_{nm} H(\theta_m - \theta_n)$$

#### Kuramoto Order Parameter

$$re^{i\psi} = N^{-1} \sum_{n} e^{i\theta_n}$$

Variance Order Parameter

$$R = 1 - \sigma_{\theta}^2 / 2$$
$$\sigma_{\theta} = N^{-1} \sum_{n} (\theta_n - \overline{\theta})^2$$

• Order parameters are similar for strong phase synchronization:

$$R \le r \le R + \frac{\sum_{n} (\theta_n - \psi)^4}{24N}$$



## Strong Phase-Locked Synchronization

Focusing on the phase-locked state, we study the linear approximation

$$\dot{\theta}_n = \omega_n + K \sum_m A_{nm} H(\theta_m - \theta_n)$$
  
$$\approx \omega_n + K H(0) d_n - K H'(0) \sum_m L_{nm} \theta_n$$
  
$$= \hat{\omega}_n - \hat{K} \sum_m L_{nm} \theta_n$$

m



strong phase synchronization

where  $L_{nm} = \delta_{nm} d_n - A_{nm}$  is the unnormalized Laplacian matrix

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We call this system <u>heterogeneous Laplacian dynamics</u>, and it has the steady-state solution

$$\theta^* = \hat{K}^{-1} L^{\dagger} \hat{\omega} + \overline{\theta}$$

where  $L^{\dagger} = \sum_{m=2}^{N} \lambda_m^{-1} v^{(m)} (v^{(m)})^T$  is the Moore-Penrose pseudoinverse

## Synchrony Alignment Function (SAF)

-Skardal, Taylor, Sun. PRL 113, 144101 2014

Describes the phases' variance for heterogeneous Laplacian dynamics

$$J(\omega, L) = \frac{1}{N} ||L^{\dagger}\omega||^{2}$$
$$= \frac{1}{N} \sum_{n=2}^{N} \frac{\langle \omega, v^{(n)} \rangle^{2}}{\lambda_{n}^{2}}$$
$$= K^{2} \frac{||\theta^{*} - \overline{\theta}||^{2}}{N}$$

- Measures alignment of oscillator frequencies  $\{\omega_n\}$  with the network structure
  - Uses full set of eigenvalues  $\{\lambda_n\}$  and eigenvectors  $\{v^{(n)}\}$  of the network Laplacian matrix L

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- Measures alignment of oscillator frequencies  $\{\omega_n\}$  with the network structure
  - Uses full set of eigenvalues  $\{\lambda_n\}$  and eigenvectors  $\{v^{(n)}\}$  of the network Laplacian matrix L
- The order parameters are given by:  $R = 1 J(\omega, L)/2K^2 \approx r$
- These approximate the order parameters of the <u>Kuramoto system</u> in the regime of strong phase synchronization

### Alignment of Frequencies with Network Dramatically Effects Synchronization

 Degree-frequency correlations can affect the very nature of phase locking, influencing whether high-degree or low-degree nodes first phase lock



 Effects of degree-frequency correlations on network synchronization: Universality and full phase-locking. Skardal, Sun, Taylor & Restrepo. EPL (Europhysics Letters) 101, 20001 (2013).

#### Alignment of Frequencies with Network Dramatically Effects Synchronization

- Degree-frequency correlations can affect the very nature of phase locking, influencing whether high-degree or low-degree nodes first phase lock
- Alignment also influences level of synchronization
  - Frequencies are either aligned with dominant eigenvector,  $\omega = \mathbf{v}^{(N)}$ , or given by  $\omega = \text{perm}(\mathbf{v}^{(N)})$



### Example: SAF-Optimized Systems Exhibit Matched Heterogeneity

-Skardal, Taylor, Sun. PRL 113, 144101 2014



homogeneous network



<sup>-</sup>color indicates frequency

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## Motivation

- These network optimization experiments were based on Markov Chain Monte Carlo (MCMC) methods in which thousands of perturbations were proposed as possible improvements:
  - perturbations that decreased the SAF were implemented
  - perturbations that increased the SAF were rejected
- MCMC is inefficient since perturbations are <u>chosen at random</u>
- We will develop efficient algorithms by approximating how network modifications affect the SAF
  - Efficient in two ways:
    - (1) small number of rewires needed to enhance synchronization
    - (2) small computational complexity of rewiring algorithms

## Outline

- We develop a first-order spectral perturbation analysis for how network modifications such as rewiring affect the SAF
- We use these perturbations to rank network edges (and potential new edges) according to their importance to synchronization
  - Rankings take into account the oscillators' heterogeneous frequencies!
- Based on these rankings, we develop rewiring algorithms to efficiency enhance a systems ability to synchronize

Synchronization of heterogeneous oscillators under network modifications: Perturbation and optimization of the synchrony alignment function

Dane Taylor, Per Sebastian Skardal, Jie Sun SIAM Journal on Applied Dynamical Systems 76(5), 1984-2008 (2016)

## Perturbation Analysis of the SAF

- Let symmetric matrix  $\epsilon \Delta L$  denote a change to the Laplacian matrix encoding a modification to an undirected network and consider the SAF

$$J(\omega, L) = \frac{1}{N} \sum_{n=2}^{N} \frac{\langle \omega, v^{(n)} \rangle^2}{\lambda_n^2}$$

• The perturbed SAF is given by

$$J(\omega, L + \epsilon \Delta L) = J(\omega, L) + \epsilon J'(0) + \mathcal{O}(\epsilon^2)$$

where

$$J'(0) = \frac{2}{N} \sum_{n=2}^{N} \left( \frac{\boldsymbol{\omega}^T \boldsymbol{v}^{(n)}}{\lambda_n^3} \right) \left( \sum_{m=2}^{N} \frac{[\boldsymbol{\omega}^T \boldsymbol{v}^{(m)}][(\boldsymbol{v}^{(m)})^T \Delta L \boldsymbol{v}^{(n)}]}{(1 - \lambda_m / \lambda_n) - \delta_{nm}} \right)$$

• This result uses the classical spectral perturbation results:

## Perturbation Analysis of the SAF

- The addition (+) and removal (-) of an edge  $\left(p,q\right)$  yields the perturbation matrix

$$\Delta L_{ij}^{(pq)} = \left\{ egin{array}{cc} \pm 1, & (i,j) \in \{(p,p),(q,q)\} \ \mp 1, & (i,j) \in \{(p,q),(q,p)\} \ 0, & otherwise, \end{array} 
ight.$$

• Using  $(\boldsymbol{v}^{(m)})^T \Delta L^{(pq)} \boldsymbol{v}^{(n)} = (\boldsymbol{v}_p^{(m)} - \boldsymbol{v}_q^{(m)})(\boldsymbol{v}_p^{(n)} - \boldsymbol{v}_q^{(n)})$ , we obtain the following first-order approximation for adding a set  $\mathcal{E}^{(+)}$  of edges and removing a set  $\mathcal{E}^{(-)}$  of edges

$$J(oldsymbol{\omega},L(\epsilon)) = J(oldsymbol{\omega},L) + \sum_{(p,q)\in\mathcal{E}^{(+)}} \epsilon Q_{pq} - \sum_{(p,q)\in\mathcal{E}^{(-)}} \epsilon Q_{pq} + \mathcal{O}(\epsilon^2)$$

where

$$Q_{pq} = rac{2}{N} \sum_{n=2}^{N} \left( rac{oldsymbol{\omega}^T oldsymbol{v}^{(n)}}{\lambda_n^3} 
ight) \left( \sum_{m=1}^{N} rac{[oldsymbol{\omega}^T oldsymbol{v}^{(m)}][(oldsymbol{v}_p^{(m)} - oldsymbol{v}_q^{(m)})(oldsymbol{v}_p^{(n)} - oldsymbol{v}_q^{(n)})]}{(1 - \lambda_m / \lambda_n) - \delta_{nm}} 
ight)$$

## Ranking Edges for SAF Optimization

- The value  $Q_{pq}$  indicates the importance of edge (p,q) to the SAF and strongly-synchronized phase-locked oscillators
- Let  $\mathcal{E} \subseteq \{1, \dots, N\} \times \{1, \dots, N\}$  denote the set of network edges
- We rank edges  $\mathcal{E}$  so that the top-ranked edge (p,q) has the smallest (most negative) value  $Q_{pq}$  so that its removal most increases  $J(\omega,L)$
- We also rank potential <u>new edges</u>  $\{1, \ldots, N\} \times \{1, \ldots, N\} \setminus \mathcal{E}$  so that the top-ranked <u>new edge</u> (p, q) corresponds to the smallest value  $Q_{pq}$

Edge rankings allow us to efficiently modify networks to tune their synchronization properties. Importantly, these rankings reflect the oscillators' heterogeneous dynamics!

# Example Illustrating Importance of Oscillator Frequencies for Edge Ranking

We add edges to a chain network



• In one case, the oscillator frequencies are nonidentical, but similar

homogeneous frequencies



# Example Illustrating Importance of Oscillator Frequencies for Edge Ranking

- We add edges to a chain network
  - In one case, the oscillator frequencies are nonidentical, but similar
  - In the other, case we make oscillator 5 very fast



# Example Illustrating Importance of Oscillator Frequencies for Edge Ranking

We add edges to a chain network



- In one case, the oscillator frequencies are nonidentical, but similar
- In the other, case we make oscillator 5 very fast
- We observe two regimes:
  - When the network is the limiting factor, the edge additions aim to improve the network's general (i.e., agnostic of frequencies) synchronizability
  - When oscillator heterogeneity is the limiting factor, the edge additions aim to counter balance the oscillators' frequencies



#### Connection to Dynamical Importance

-Restrepo, Ott, Hunt. PRL 97, 094102 (2006)

• We rank edges according to strong-synchronization regime



### Connection to Dynamical Importance

-Restrepo, Ott, Hunt. PRL 97, 094102 (2006)

• We rank edges according to strong-synchronization regime



• Our work complements <u>Dynamical Importance</u>,  $\hat{I}_k = \frac{v_k u_k}{v^T u}$ , which ranks edges according their effect on the critical value where the incoherent state becomes unstable

### Algorithms for Tuning Synchronization of Phase-Oscillator Systems

- We introduce two "gradient descent" algorithms to tune synchronization
  - One that updates  $Q_{pq}$  after modifications
  - One that does not update  $Q_{pq}$
- We compare these algorithms to two other methods
  - Modifications made at random
  - Modifications chosen to maximize  $\lambda_2$

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#### Algorithms Performance for Non-ideal Scenarios

• We consider edge additions under 3 scenarios

(a) A fraction of potential new edges are "off limits" for addition



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  - (b) We introduce misinformation about the network by rewiring a fraction of the edges before computing



#### Algorithms Performance for Non-ideal Scenarios

- We consider edge additions under 3 scenarios
  - (a) A fraction of potential new edges are "off limits" for addition
  - (b) We introduce misinformation about the network by rewiring a fraction of the edges before computing
  - (c) We introduce misinformation about the frequenting by adding Gaussian noise to them



## Conclusion

- The Synchrony Alignment Function (SAF) describes the variance of phases for strongly-coupled phase-locked oscillators
- We developed a first-order spectral perturbation analysis for how network modifications affect the SAF
- These perturbations rank network edges (and potential new edges) according to their importance to synchronization
  - Rankings take into account the oscillators' heterogeneous frequencies!
  - Observed network-dominated and oscillator-dominated regimes
- Based on these rankings, we develop rewiring algorithms to efficiency enhance a systems ability to synchronize

## Main Reference

Synchronization of heterogeneous oscillators under network modifications: Perturbation and optimization of the synchrony alignment function. Taylor, Skardal & Sun. SIAM Journal of Applied Dynamical Systems 76(5), 1984-2008 (2016).

#### Other SAF-Related Work

- Optimal synchronization of complex networks. Skardal, Taylor & Sun, *PRL* 113, 144101 (2014).
- Erosion of synchronization in networks of coupled oscillators. Skardal, Taylor, Sun & Arenas. *PRE* 91, 010802R (2015).
- Optimal synchronization of directed complex networks. Skardal, Taylor & J Sun. *Chaos* 26, 094807 (2016).
- Collective frequency variation in network synchronization and reverse PageRank. Skardal, Taylor, Sun & Arenas. *PRE* 93, 042314 (2016).

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## Accuracy of First-Order Approximation



• The error vanishes as  $\epsilon \to 0$ 

## Accuracy of First-Order Approximation

Approximation error for edge additions

$$\Delta J = J(\omega, L(0)) - J(\omega, L(\epsilon))$$
  

$$\approx \epsilon Q_{pq}$$





- The error vanishes as  $\epsilon \to 0$
- The error also vanishes with fixed  $\epsilon$  and increasing network size N and number of edges

## **Power-Grid Synchronization**

- T Nishikawa & AE Motter (2015) Comparative analysis of existing models for power-grid synchronization. New Journal of Physics, 17(1), 015012.
- M Rohden, A Sorge, M Timme & D Witthaut (2012) Self-organized synchronization in decentralized power grids. Physical Review Letters, 109(6), 064101.
- F Dorfler & F Bullo (2012) Synchronization and transient stability in power networks and nonuniform Kuramoto oscillators. SIAM Journal on Control and Optimization, 50(3), 1616-1642.



## **Spectral Perturbation Analysis**

Classical perturbation analysis of eigenvalues and eigenvectors
 [4] KE Atkinson, "An introduction to numerical analysis," John Wiley & Sons, 2008.

THEOREM 4.1 (Perturbation of Simple Eigenvalues and their Eigenvectors [4]). Let L be a symmetric matrix with simple eigenvalues  $\{\lambda_n\}$  and normalized eigenvectors  $\{v^{(n)}\}$ . Consider a fixed symmetric perturbation matrix  $\Delta L$ , and let  $L(\epsilon) = L + \epsilon \Delta L$ . Denote the eigenvalues and eigenvectors of  $L(\epsilon)$  by  $\lambda_n(\epsilon)$  and  $v^{(n)}(\epsilon)$ , respectively, for n = 1, 2, ..., N. It follows that

$$\lambda_n(\epsilon) = \lambda_n + \epsilon \lambda'(0) + \mathcal{O}(\epsilon^2),$$
  
$$\boldsymbol{v}^{(n)}(\epsilon) = \boldsymbol{v}^{(n)} + \epsilon \boldsymbol{v}^{(n)'}(0) + \mathcal{O}(\epsilon^2),$$
 (4.1)

and the derivatives with respect to  $\epsilon$  at  $\epsilon = 0$  are given by

$$\lambda'_{n}(0) = (\boldsymbol{v}^{(n)})^{T} \Delta L \boldsymbol{v}^{(n)}$$
$$\boldsymbol{v}^{(n)'}(0) = \sum_{m \neq n} \frac{(\boldsymbol{v}^{(m)})^{T} \Delta L \boldsymbol{v}^{(n)}}{\lambda_{n} - \lambda_{m}} \boldsymbol{v}^{(m)}.$$
(4.2)

### General Network-Perturbation Result

• Perturbation result for SAF: 
$$J(\omega,L)=rac{1}{N}\sum_{n=2}^{N}rac{\langle\omega,v^{(n)}
angle^2}{\lambda_n}$$

THEOREM 4.2 (Perturbation of the SAF under a Network Modification). Let  $J(\omega, L)$  denote the SAF given by Eq. (3.3) for natural frequencies  $\omega$  and symmetric network Laplacian L, and let  $J(\omega, L(\epsilon))$  denote the SAF for the network after it undergoes a symmetric modification  $\epsilon \Delta L$ . Assume the eigenvalues of L and  $L(\epsilon) = L + \epsilon \Delta L$  are simple, and that the original and perturbed networks are both connected. Then the first-order expansion in  $\epsilon$  for the perturbed SAF is given by

$$J(\boldsymbol{\omega}, L(\epsilon)) = J(\boldsymbol{\omega}, L) + \epsilon J'(\epsilon) + \mathcal{O}(\epsilon^2), \qquad (4.3)$$

where

$$J'(\epsilon) = \frac{2}{N} \sum_{n=2}^{N} \left( \frac{\boldsymbol{\omega}^T \boldsymbol{v}^{(n)}}{\lambda_n^3} \right) \left( \sum_{m=2}^{N} \frac{[\boldsymbol{\omega}^T \boldsymbol{v}^{(m)}][(\boldsymbol{v}^{(m)})^T \Delta L \boldsymbol{v}^{(n)}]}{(1 - \lambda_m / \lambda_n) - \delta_{nm}} \right).$$
(4.4)

#### Observation 2: SAF-Optimized Systems Exhibit Correlations

Positive correlation between a node's frequency and degree

![](_page_35_Figure_2.jpeg)

![](_page_35_Figure_3.jpeg)

<sup>-</sup>color indicates frequency

#### Observation 2: SAF-Optimized Systems Exhibit Correlations

![](_page_36_Figure_1.jpeg)

# Synchrony Alignment Function (SAF)

 The SAF estimates the variance of phases for phase-locked oscillators

$$J(\omega, L) \approx \frac{||\theta^*||^2}{K^2 H'^2(0)}$$

![](_page_37_Picture_3.jpeg)

and approximates the Kuramoto order parameter

 $r \simeq 1 - J(\omega, L) / 2K^2 H^{\prime 2}(0)$ 

• We develop methodology to optimize the SAF, thereby optimizing r

- **Problem:** Given a network, allocate a set of oscillator frequencies with unit variance to maximize the order parameter  $r = 1 1/2\lambda_N^2 K^2$
- Solution: If one can choose <u>any frequencies</u>, the optimal frequency vector is , which gives  $\omega \propto v^N$

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Results shown for allocating frequencies to a scale-free network with N=1000 nodes, exponent 3, and minimum degree 2

![](_page_40_Figure_4.jpeg)

- **Problem:** Given a network, allocate a set of oscillator frequencies with unit variance to maximize the order parameter
- Solution: If one must use <u>a given set of frequencies</u>, then we approximately minimize  $J(\boldsymbol{\omega}, L)$  using an accept/reject algorithm

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Results shown for allocating frequencies drawn from a normal distribution to a scalefree network

![](_page_42_Figure_4.jpeg)

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Results shown for allocating frequencies drawn from a normal distribution to a scalefree network

As a fast approximation, one can shuffle the frequencies to best align with  $v^N$ 

![](_page_43_Figure_5.jpeg)

For small *K*, one must choose particular frequencies to achieve strong synchronization.

MAIN INSIGHT

![](_page_44_Figure_2.jpeg)

For moderate *K*, strong synchronization can be achieved just by rearranging the oscillators.

\*\*observed for our experiments

# Optimization Experiment II: design network for given frequencies

- **Problem:** Given a set of oscillator frequencies, design a network with a fixed number of edges to maximize the order parameter  $J(\omega, L)$
- Solution: We use an accept/reject algorithm to iteratively rewire an initial network so as to always decrease

# Optimization Experiment II: design network for given frequencies

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- **Solution:** We use an accept/reject algorithm to iteratively rewire an initial network so as to always decrease

Results shown for frequencies drawn from a normal distribution.

The initial network is scale free with N=1000 nodes, exponent 3, and minimum degree 2

![](_page_46_Figure_5.jpeg)

#### Optimization Requires Network Heterogeneity to Match Oscillator Heterogeneity

![](_page_47_Figure_1.jpeg)

homogeneous network

heterogeneous network

0

![](_page_47_Picture_4.jpeg)

## Further Observations for Synchrony-Optimized Systems

![](_page_48_Figure_1.jpeg)

# Beyond Maximizing the Order Parameter

- One can tune the order parameter of a system by aligning the oscillator frequencies with other eigenvectors
  - For a given scale-free network with N=1000 nodes, we consider setting the frequencies as  $\omega \propto v^{100}, v^{200}, ..., v^{1000}$

![](_page_50_Figure_3.jpeg)