



# Eigenspaces of Graphs and their Utility

Garrett J. Kepler  
Joint Mathematics Meeting  
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- ① Preliminaries
- ② Characterization
- ③ Classification
- ④ Application
- ⑤ Future Work

## 1 Preliminaries

## 2 Characterization

## 3 Classification

## 4 Application

## 5 Future Work

# Background

## Definition

The *Laplacian* matrix  $L = [\ell_{ij}] \in M_n$  is defined such that

$$\ell_{ij} = \begin{cases} d_i & \text{if } i = j \\ -1 & \text{if } i \text{ adjacent to } j \\ 0 & \text{otherwise} \end{cases}$$

## Definition

Let  $0 = \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$  be the eigenvalues of  $L$ . We call the second smallest eigenvalue of  $L$ ,  $\lambda_2$ , and its corresponding eigenvector the *Fiedler value* and *Fiedler vector* respectively.

# Combinatorial View of Fiedler Value

## Theorem

*Since the constant vector  $\mathbb{1}$  is an eigenvector of  $L$  corresponding to 0, by Rayleigh-Ritz, the Fiedler value of  $L$  corresponds to the following minimum:*

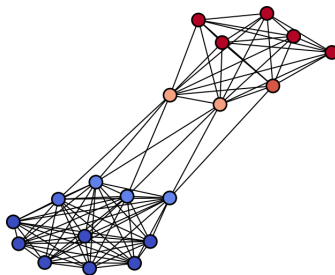
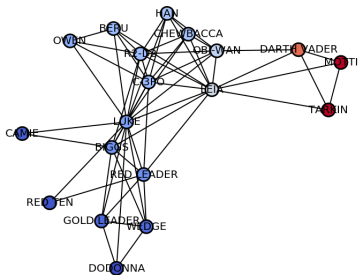
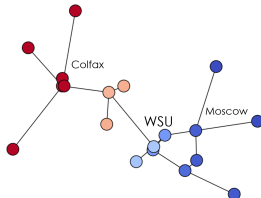
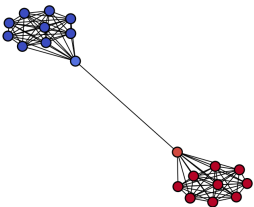
$$\lambda_2 = \min_{\mathbf{x} \perp \mathbb{1}} \frac{\mathbf{x}^T L \mathbf{x}}{\mathbf{x}^T \mathbf{x}} = \min_{\mathbf{x} \perp \mathbb{1}} \frac{\sum_{i \sim j} (x(i) - x(j))^2}{\sum_i x(i)^2}$$

# Fiedler Partitioning Theorem

## Theorem

*Let  $G = (V, E)$  be a connected graph and let  $x$  be its corresponding Fiedler vector. Then the subgraphs induced by the vertex sets  $V_1 = \{i : x_i > 0\}$  and  $V_2 = V \setminus V_1$  are connected.*

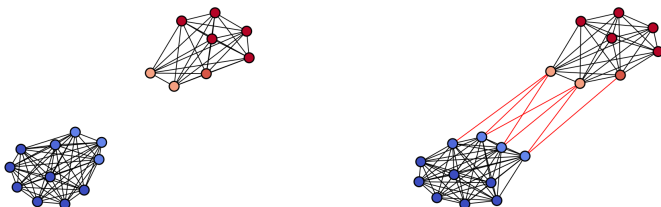
Informally: we can partition  $G$  in to two connected components using the entries of the Fiedler vector.



# Reconstruction Question

## Question

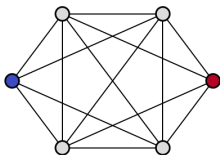
Given the two connected components of a Fiedler-partitioned graph, can we determine what set of edges are missing using the entries of the Fiedler vector?





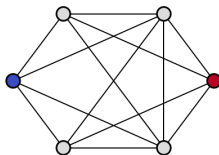
## Answer

Not always.



$$x = [1, 0, 0, -1, 0, 0]^T$$

$\lambda = 4$



$$x = [1, 0, 0, -1, 0, 0]^T$$

$\lambda = 4$

## Theorem

*Let  $x$  be an eigenvector of  $L$  corresponding to  $\lambda$ . If  $x(i) = x(j)$  for some entries of  $x$ , the inclusion or removal of edge  $(i, j)$  maintains  $x$  as an eigenvector and  $\lambda$  as its corresponding eigenvalue.*

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# Main Characterization Questions

## Question

- Can we find graphs with Fiedler vectors that reveal as little information as possible about the cut edges?
- Can we find graphs with Fiedler vectors that reveal as much as information possible?

# Method

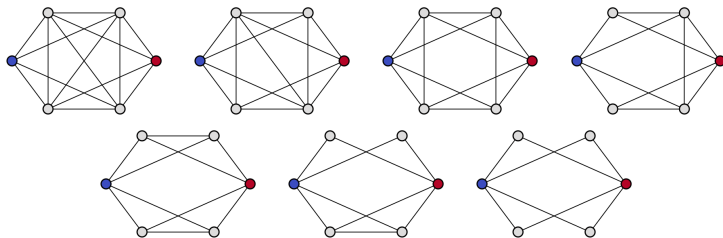
*An Approach*: Construct graph around vector

Pick  $x$  such that

- $x \perp \mathbb{1}$
- $x$  has as many/as little repeated entries as possible
- find graph that has  $x$  as Fiedler vector

# An Example

- Consider the vector from before:  $x = [1, 0, 0, -1, 0, 0]^T$
- What graphs can we make with  $x$  as its Fiedler vector?



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# Main Classification Question

## Question

Given a graph, can we determine how much information about the missing edges can be revealed by its Fiedler vector?

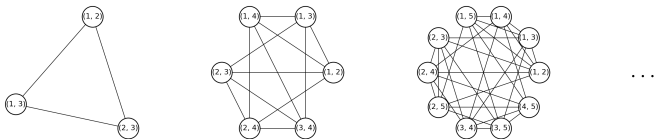
# Method

**An Approach:** Construct vector around class of graphs

- Pick a class of graphs
- Characterize their Fiedler vectors
- Observe what eigenvector preserving perturbations are permitted
- Obtain graphs that share that Fiedler vector



# An Example



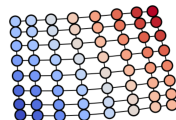
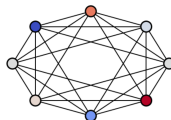
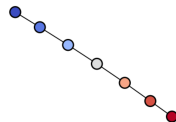
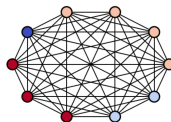
$J(2, n)$ :

- Constant vector ( $\lambda = 0$ )
- Assign 1 to node  $\{a, b\}$ , -1 to node  $\{c, d\}$ , and 0 elsewhere ( $\lambda = n$ )
- Assign 1 to pair of nodes  $\{a, b\}$  &  $\{b, c\}$ , -1 to pair of nodes  $\{a, d\}$  &  $\{c, d\}$ , and 0 elsewhere ( $\lambda = 2(n - 1)$ )

(**Bonus**: eigenvector construction above generalizes to  $J(k, n)$  by merely keeping track of intersections!)

# Already Characterized Classes

- Complete graphs ✓
- Path graphs ✓
- Grid graphs ✓
- Circulant graphs ✓
- $\vdots$



① Preliminaries

② Characterization

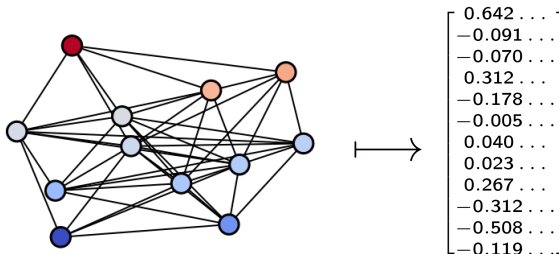
③ Classification

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⑤ Future Work

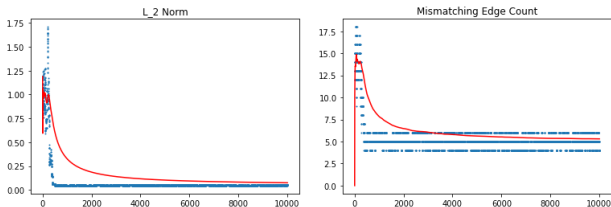
# Dimension Reduction

- In the case that we are able to reconstruct some number of missing edges, we are able to reduce the amount of required data from a graph to determine bottlenecking information



# MCMC Methods

- Some recent work has been done investigating MCMC approximation of graphs via the Fiedler vector
- Classifying the reconstructibility of graphs via the Fiedler vector can aid in the reliability of these methods



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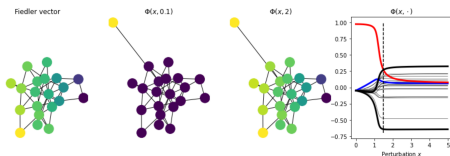
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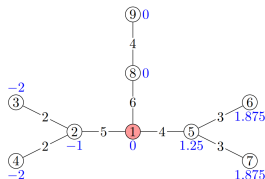
# Future Work

- In finding graphs with maximally informative Fiedler vectors, we cleverly picked vectors we knew would be an eigenvector for some graph. What are the constraints for entries that guarantee finding such a graph is possible?
- Extending to  $k$ -eigenvector partitioning, what subset of eigenvectors do we need to maximize/minimize the information we can reconstruct via their entries?

# Other Interesting Work



- "Perturbation of Fiedler vector: interest for graph measures and shape analysis" by Lefevre, Fraize, & Germaud  
url: <https://arxiv.org/abs/2306.04327>



- "Inverse Fiedler vector problem of a graph" by Lin & Shirazi  
url: <https://arxiv.org/abs/2410.09736>



Thank you!  
garrett.kepler@wsu.edu