

# Lecture 15 - Spectral Clustering

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# Spectral Clustering

Let  $G$  be a graph,  $G(V, E)$ , where  $V = \{1, 2, 3, \dots, n\}$

The Laplacian matrix:

$$L_G = \underbrace{D_G}_{\substack{\text{Diagonal matrix} \\ \text{with degrees of} \\ \text{each vertex}}} - \underbrace{A_G}_{\text{adjacency matrix}}$$

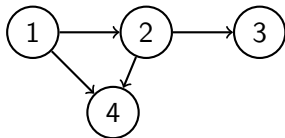
$L_G$  is a PSD Matrix whose eigenvalues are

$$0 = \lambda_1 \leq \underbrace{\lambda_2}_{\text{Fiedler value}} \leq \lambda_3 \leq \dots \leq \lambda_n$$

An eigenvector with eigenvalue  $\lambda_2$  is called the Fiedler vector of  $\vec{W}$ .

# Spectral Clustering

We will use  $\vec{W}$ ,  $L_G$  to understand clusters in a graph.



$$A = \{4, 5, 6, 7, 8\}, |A| = 5$$

Notation

If  $X$ ,  $Y$  are sets  $X \setminus Y = \{x \in X \mid x \notin Y\}$

$|x|$  = size of  $x$

# Cut

## Definition

A cut in  $G$  is a partition of  $V$  into two sets,  $A$  and  $V \setminus A$ , when  $A \neq V$ . (The cut induced by  $A$ )

## Notation

If  $X, Y \subseteq V$ , let  $E(X, Y)$  denote all edges with one vertex in  $X$  and one vertex in  $Y$ .

$$E(A, V \setminus A) = \{ \{1, 6\} \}$$

## Definition

The density of the cut induced by  $A$  is:

# Sparsest cut

Definition:

Let  $d_{\min}$  denote the smallest possible density of a cut in  $G$ . Any cut with density  $d_{\min}$  is called a sparsest cut in  $G$ .

# Example in Julia

Need to figure out how to make code look pretty in Latex

# Exploring

$$Q(\vec{x}) = n \cdot \frac{\sum_{\{i,j\} \in E(G)} (x_i - x_j)^2}{\sum_{1 \leq i < j \leq n} (x_i - x_j)^2}$$

Let  $A \subseteq V$ ,

$$A = \{4, 5, 6, 7, 8\}, \quad \vec{C}_A = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 1 \\ 1 \\ 1 \\ 0 \\ 0 \end{pmatrix}$$

Need to add a visual representation demonstrating the Graph Laplacian and the example in code. About the information we can extract from the Graph Laplacian.



# Conclusion

here we go: