

# Lecture 14 - Graph Laplacians

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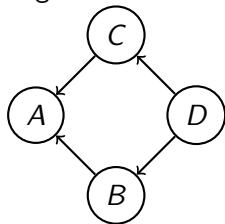
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# Graph Laplacians

Definition:

Let  $G$  be a graph. Let  $i$  be a vertex of  $G$ , the degree of  $i$  is the number of edges incident to vertex  $i$ , it is denoted  $d_i$



# Graph Laplacians

Definition:

Let  $D_G \in \mathbb{R}^{n \times n}$  be the diagonal matrix whose  $i^{th}$  diagonal entry is  $d_i$ .

Recall the adjacency matrix  $A_G = (a_{ij}) \in \mathbb{R}^{n \times n}$  is defined by:

$$a_{ij} = \begin{cases} 1, & \text{if } [i, j] \text{ is an edge in } G \\ 0, & \text{if not} \end{cases}$$

Definition:

The Laplacian of  $G$  is:

$$L_G = D_G - A_G = \begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 3 & -1 & -1 \\ 0 & -1 & 1 & 0 \\ -1 & -1 & 0 & 2 \end{bmatrix}$$

# Graph Laplacians

Facts:

- $L_G$  is PSD ( Positive Semi-Definite ).
- $\lambda = 0$  is always an eigenvalue of  $L_G$  with eigenvector  $\vec{1}$ .

$$\begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 3 & -1 & -1 \\ 0 & -1 & 1 & 0 \\ -1 & -1 & 0 & 2 \end{bmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}$$

Every row of  $L_G$  sums to 0 by construction.

$$L_G \vec{1} = \begin{pmatrix} \text{sum of row 1} \\ \cdots \\ \cdots \\ \text{sum of row n} \end{pmatrix} = \vec{0}$$

Why is  $L_G$  PSD ?

We will show that  $L_G = B^T B$  for some matrix  $B$ .

# Positive Semi-Definite Characterization

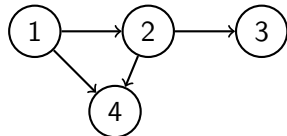
Definition:

A symmetric matrix  $A \in \mathbb{R}^{n \times n}$  is positive semi-definite, denoted  $A \geq 0$ , if any of the following equivalent conditions hold:

- 1  $\vec{x}^T A \vec{x} \geq 0, \forall \vec{x} \in \mathbb{R}^n$ . (the graph of  $g_A(\vec{x}) = \vec{x}^T A \vec{x}$  is on or above the domain,  $\mathbb{R}^n$ ).
- 2 All eigenvalues of  $A$  are non-negative.
- 3 There exists some matrix  $B$  s.t.  $A = B^T B$ .
- 4 All principal minors of  $A$  are non-negative.

# Define BG

Define the matrix  $B_G$ , the "directed" node-edge incidence matrix of G.



pick a direction for each edge

rows indexed by nodes

columns indexed by edges of G

$$B_G = \begin{matrix} & \begin{matrix} 12 & 14 & 23 & 24 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 1 \\ 0 & 0 & -1 & 0 \\ 0 & -1 & 0 & -1 \end{bmatrix} \end{matrix}$$

In the column indexed by the directed edge  $ij$ , we put 1 in row  $i$  and -1 in row  $j$ .

$$L_G = B^T B$$

# Define $B_G$

$$L_G = B_G \cdot B_G^T$$

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 1 \\ 0 & 0 & -1 & 0 \\ 0 & -1 & 0 & -1 \end{bmatrix} \cdot \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 0 & 0 & -1 \\ 0 & 1 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix} = \begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 3 & -1 & -1 \\ 0 & -1 & 1 & 0 \\ -1 & -1 & 0 & 2 \end{bmatrix}$$

If you get  $i^{th}$  row and  $i^{th}$  column,  $1^2$  or  $(-1)^2$  for every edge incident to  $i$ .  
So the sum  $d_i$ .

$i^{th}$  row and  $j^{th}$  column (if  $i \neq j$ ) if  $ij$  is on edge, you will get  $1(-1) = -1$

Therefore,  $L_G$  is PSD.

# Lemma

Lemma:

If  $G$  is a graph, the Laplacian  $L_G$  is PSD and its eigenvalues are:

$$0 = \lambda_1 \leq \lambda_2 \leq \lambda_3 \leq \cdots \leq \lambda_n$$

and  $\vec{1}$  is an eigenvector of  $L_G$  with eigenvalue 0.

Since  $L_G$  is PSD, the  $\vec{x}^T L_G \vec{x} \geq 0$ ,  $\forall \vec{x} \in \mathbb{R}^n$



# Lemma

PwP:

Let  $G = ([n], E)$  be the graph with vertices  $[n]$  and edge set  $E$ . If  $\vec{x} = (x_1, \dots, x_n)^T$ .

$$\vec{x}^T L_G \vec{x} = \sum_{\{i,j\} \in E} (x_i - x_j)^2$$

Proof

We know that  $L_G = B_G B_G^T$ . So  $\vec{x}^T L_G \vec{x} = \vec{x}^T B_G B_G^T \vec{x}$ .

Lets understand  $\vec{x}^T B_G$ .

So the entries of  $\vec{x}^T B_G$  are all  $x_i - x_j$  when  $ij$  is an edge of  $G$ .

# Lemma

PwP:

$$\text{So } \vec{x}^T L_G \vec{x} = (\vec{x}^T B_G) \cdot (B_G^T \vec{x}) = \sum_{\{i,j\} \in E} (x_i - x_j)^2.$$

Do this for the example:

# Connectivity

Definition:

A Graph is connected if there is a way to walk from any vertex to any other vertex along the edges of  $G$ .

Theorem:

$G$  is connected if and only if  $\lambda_2 > 0$ .

Proof:

If  $G$  is not connected the Graph Laplacian will look like this:

# Connectivity

We sort of proved if  $G$  is not connected, then  $\lambda_2 = 0$ . (if  $G$  has  $k$  components,  $\lambda_1 = \lambda_2 = \dots \lambda_k = 0$ )

We still need to show that if  $G$  is connected,  $\lambda_2 > 0$ .

For a symmetric matrix,  $AM(\lambda) = GM(\lambda)$  for an eigenvalues.

So we need to show  $GM(0) = 1$ . Let  $\vec{u}$  be an eigenvector with eigenvalue of 0. Then  $L_G \vec{u} = \vec{0}$ .

$$0 = \vec{u}^T L_G \vec{u} = \sum_{\{i,j\} \in E} (u_i - u_j)^2$$

If  $u_1 = c$  and  $1,k$  is an edge, the  $u_k = c$ . for all edge  $ij \in E$ .  
Similarly if  $kj$  is an edge  $u_j = c$ . Since  $G$  is connected  $\vec{u} = c\vec{1}$ .

AM, algebraic multiplicity.

GM, geometric multiplicity.

# Conclusion

Slight Generalization:

The number of times 0 is an eigenvalue of  $L_G$  counts the number of connected components of  $G$ .

Fact:

The second eigenvalue  $\lambda_2$  is called the spectral gap or the Fiedler value of  $G$ .

# Testing Latex stuff

$$\alpha = \begin{matrix} & \bar{f}_1 & \bar{f}_2 & \bar{f}_3 \\ \begin{matrix} k_1 \\ k_2 \\ k_3 \\ k_4 \end{matrix} & \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \end{matrix}$$
$$\begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 3 & -1 & -1 \\ 0 & -1 & 1 & 0 \\ -1 & -1 & 0 & 2 \end{bmatrix}$$

# Testing Latex stuff

$$\beta = \begin{matrix} & \bar{f}_1 & \bar{f}_2 & \bar{f}_3 \\ \begin{matrix} f_1 \\ f_2 \\ f_3 \end{matrix} & \begin{pmatrix} 3 & 2 & 0 \\ 2 & 4 & 2 \\ 5 & 3 & 1 \end{pmatrix} \end{matrix}$$

$$\gamma = \begin{matrix} & C_1 & C_2 & \dots & C_n \\ \begin{matrix} N_1 \\ N_2 \\ \\ N_n \end{matrix} & \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \\ & \bar{f}_1 & \bar{f}_2 & \dots & \bar{f}_n \\ \begin{matrix} k_1 \\ k_2 \\ \\ k_n \end{matrix} & \begin{bmatrix} 0 & 1 & \dots & 0 \\ 1 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix} & \begin{matrix} = 1 \\ = 1 \\ \\ = 1 \end{matrix} \end{matrix}$$