# Laplacians of graphs and Cheeger inequalities

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# Outline

- Review on Graph Analysis
  - Combinatorial Laplacian
  - Normalized Laplacian
  - Cut
- Cheeger Constant
- Cheeger Inequality for Edge Expansion
  - Proof
  - Examples
- Isoperimetric Inequality for Vertex Expansion

• A graph is an ordered pair G = (V, E)

• Adjacency matrix 
$$A_{uv} = \begin{cases} 1 & \text{if } (u,v) \in E(G) \\ 0 & \text{otherwise} \end{cases}$$

Undirected graph  $\Leftarrow \Rightarrow$  symmetric adjacency matrix

- $d_v$ : degree of the vertex v
- D = diag(d) = diagonal matrix of degrees

$$A = \begin{pmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}, \ d = \begin{pmatrix} 3 \\ 2 \\ 2 \\ 1 \end{pmatrix}, \ D = \begin{pmatrix} 3 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

### **Combinatorial Laplacian**

$$L(u, v) = \begin{cases} d_v & \text{if } u = v \\ -1 & \text{if } u \text{ and } v \text{ are adjacent} \\ 0 & \text{otherwise} \end{cases}$$



Normalized Laplacian

$$\mathcal{L} = SLS = D^{-\frac{1}{2}}LD^{-\frac{1}{2}}$$
  
$$S = D^{-\frac{1}{2}}$$
: diagonal matrix with the  $(v, v)$ -th entry having value  $\frac{1}{\sqrt{d_v}}$ 

$$\mathcal{L}(u,v) = \begin{cases} 1 & \text{if } u = v \\ -\frac{1}{\sqrt{d_u d_v}} & \text{if } u \text{ and } v \text{ are adjacent} \\ 0 & \text{otherwise} \end{cases}$$

with eigenvalues

$$0 = \lambda_1 \le \lambda_2 \le \dots \le \lambda_n \le 2$$

The smallest non-zero eigenvalue of  $\mathcal{L}$  is related to best edge-relative cut.

# Cut: results in disconnected graph

- Vertex cut
- Edge cut

# How easy is it to cut a graph?

Efficient Cut:

- Should be small
- Pieces left to be large





 $E(X, \overline{X})$ : denotes the set of edges with one vertex in X and one vertex in  $\overline{X}$ .

$$\sum_{x \in X} d_x = vol(X)$$

If G is disconnected,  $h_G$ =0 If G is connected,  $h_G > 0$  Cheeger Inequalities:  $2h_G \ge \lambda_2 \ge \frac{h_G^2}{2}$ 

1.  $\lambda_2 \leq 2h_G$ 

Let *h* denote a function which assigns to each vertex *v* of *G* a complex value h(v). Then  $\sum (f(u) - f(v))^2$ 

where 
$$h = S^{-1}f$$
.  
 $\lambda_2 = \min_{f \perp S^{-2}1} \frac{u \sim v}{\sum_v d_v f(v)^2}$ 

We choose f based on an optimum cut C which achieves  $h_G$  and separates the graph G into two parts, A and B:

$$f(v) = \begin{cases} \frac{1}{vol(A)} & \text{if } v \text{ is in } A\\ -\frac{1}{vol(B)} & \text{if } v \text{ is in } B \end{cases}$$

By substituting f into  $\lambda_2$ , we have the following:

$$\lambda_2 \leq |C|(1/vol(A) + 1/vol(B))$$
  
$$\leq \frac{2|C|}{\min(vol(A), vol(B))}$$
  
$$= 2h_G$$



PROOF. We consider the harmonic eigenfunction f of  $\mathcal{L}$  with eigenvalue  $\lambda_2$ We order vertices of G according to f. That is, relabel the vertices so that  $f(v_i) \ge f(v_{i+1})$ , for  $1 \le i \le n-1$ . Let  $S_i = \{v_1, \ldots, v_i\}$  and define

$$\alpha_G = \min_i h_{S_i}.$$

Let r denote the largest integer such that  $\operatorname{vol}(S_r) \leq \operatorname{vol}(G)/2$ . Since  $\sum_v g(v)d_v = 0$ ,

$$\sum_{v} g(v)^2 d_v = \min_{c} \sum_{v} (g(v) - c)^2 d_v \le \sum_{v} (g(v) - g(v_r))^2 d_v.$$

We define the positive and negative part of  $g - g(v_r)$ , denoted by  $g_+$  and  $g_-$ , respectively, as follows:

$$g_{+}(v) = \begin{cases} g(v) - g(v_r) & \text{if } g(v) \ge g(v_r), \\ 0 & \text{otherwise,} \end{cases}$$
$$g_{-}(v) = \begin{cases} |g(v) - g(v_r)| & \text{if } g(v) \le g(v_r), \\ 0 & \text{otherwise.} \end{cases}$$

We consider

$$\lambda_{G} = \frac{\sum_{u \sim v} (g(u) - g(v))^{2}}{\sum_{v} g(v)^{2} d_{v}}$$

$$\geq \frac{\sum_{u \sim v} (g(u) - g(v))^{2}}{\sum_{v} (g(v) - g(v_{r}))^{2} d_{v}}$$

$$\geq \frac{\sum_{u \sim v} \left( (g_{+}(u) - g_{+}(v))^{2} + (g_{-}(u) - g_{-}(v))^{2} \right)}{\sum_{v} \left( g_{+}(v)^{2} + g_{-}(v)^{2} \right) d_{u}}.$$

Without loss of generality, we assume  $R(g_+) \leq R(g_-)$  and therefore we have  $\lambda_G \geq R(g_+)$  since

$$\frac{a+b}{c+d} \ge \min\{\frac{a}{c}, \frac{b}{d}\}.$$

We here use the notation

$$\tilde{\operatorname{vol}}(S) = \min\{\operatorname{vol}(S), \operatorname{vol}(G) - \operatorname{vol}(S)\}$$

so that

 $|\partial(S_i)| \ge \alpha_G \tilde{\operatorname{vol}}(S_i).$ 

Then we have

$$\begin{split} \lambda_{G} &\geq R(g_{+}) \\ &= \frac{\sum_{u \sim v} (g_{+}(u) - g_{+}(v))^{2}}{\sum_{u} g_{+}^{2}(u) d_{u}} \\ &= \frac{\left(\sum_{u \sim v} (g_{+}(u) - g_{+}(v))^{2}\right) \left(\sum_{u \sim v} (g_{+}(u) + g_{+}(v))^{2}\right)}{\sum_{u} g_{+}^{2}(u) d_{u} \sum_{u \sim v} (g_{+}(u) + g_{+}(v))^{2}} \\ &\geq \frac{\left(\sum_{i} |g_{+}(v_{i})^{2} - g_{+}(v_{i+1})^{2} |\alpha_{G}| \tilde{\text{vol}}(S_{i})|\right)^{2}}{2\left(\sum_{u} g_{+}^{2}(u) d_{u}\right)^{2}} \\ &= \frac{\alpha_{G}^{2}}{2} \frac{\left(\sum_{i} g_{+}(v_{i})^{2} (|\tilde{\text{vol}}(S_{i}) - \tilde{\text{vol}}(S_{i+1})|\right)\right)^{2}}{\left(\sum_{u} g_{+}^{2}(u) d_{u}\right)^{2}} \\ &= \frac{\alpha_{G}^{2}}{2} \frac{\left(\sum_{i} g_{+}(v_{i})^{2} d_{v_{i}}\right)^{2}}{\left(\sum_{u} g_{+}^{2}(u) d_{u}\right)^{2}} \\ &= \frac{\alpha_{G}^{2}}{2}. \end{split}$$

### Examples

• Hypercube graph  $Q_n$ : the graph formed from the vertices and edges of an n-dimensional hypercube

$$\lambda = \frac{2k}{n}$$
, for k = 0,1, ..., n  
 $h_{Q_n} = \frac{1}{n}$ 



• Path graph  $P_n$ 

$$\lambda = 1 - \cos \frac{k\pi}{n-1}, \text{ for } k = 0, 1, \dots, n-1$$

$$h_{P_n} = \frac{1}{n-1}$$
A path graph on 6 vertices

#### Isoperimetric Inequality for Vertex Expansion

For a subset X of vertices of V, we consider

 $N(X) = \{ v \not\in X : v \sim u \in X \}.$ 

We define

$$g_G(X) = \frac{vol(N(X))}{\min(vol(X), vol(\bar{X}))}$$

and

$$g_G = \min_X g_G(X)$$

For a graph G,

$$\lambda_G \ge \frac{g_G^2}{2d(2+2g_G+g_G^2)}$$

where d denotes the maximum degree of G.

### References

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# Thank You!