

## Notes 10: Conductance, Expansion, Normalized Laplacians

### 1. CONDUCTANCE AND EXPANSION

Last lecture we saw that a graph is connected if and only if the second smallest eigenvalue  $\lambda_2$  of its Laplacian  $L_G$  is strictly larger than the smallest eigenvalue  $\lambda_1$  (which is zero). Today we will show a robust version of this result: a graph is “well-connected” if and only if  $\lambda_2$  is much bigger than  $\lambda_1$ .

One way to measure how well a graph  $G = (V, E)$  is connected is expansion.

**Definition 1.1.** Given a graph  $G$  with positive edge weights  $w \in \mathbb{R}_+^E$ , the degree of vertex  $i$  is  $d(i) := \sum_{j:(i,j) \in E} w_{ij}$  and the total degree of a vertex subset  $S \subseteq V$  is  $d(S) := \sum_{i \in S} d(i)$ .

The conductance of a vertex subset  $S \subseteq V$  is

$$\varphi(S) = w(S, \bar{S}) / d(S),$$

where  $w(S, \bar{S}) = \sum_{i \in S, j \notin S, (i,j) \in E} w_{ij}$  is the total edge weight across the cut from  $S$  to  $\bar{S}$ .

The expansion of a graph is

$$\varphi(G) = \min_{\substack{S \subseteq V, S \neq \emptyset \\ d(S) \leq d(V)/2}} \varphi(S).$$

The condition  $d(S) \leq d(V)/2$  in expansion is equivalent to  $d(S) \leq d(\bar{S})$ .

The total degree of a subset  $S$  measures the size of a subset, weighted according to degrees. If the graph is regular (all vertices have the same degree), then  $\text{deg}(S)$  is proportional to  $|S|$ .

The conductance of a subset or the expansion a graph is always between 0 and 1. A graph is disconnected if and only if  $\varphi(G) = 0$ .

**1.1. Complete graph.** What is the expansion of the complete graph, the most well-connected graph? Given a subset  $S \subseteq V$  of size  $k$ , its conductance is  $\frac{k(n-k)}{k(n-1)} = \frac{n-k}{n-1}$ . So for any subset of size at most  $n/2$ , its conductance is at least  $\frac{n}{2(n-1)} \approx \frac{1}{2}$ . Complete graph therefore has expansion roughly  $1/2$ .

**1.2. Barbell graph.** The barbell graph on  $2n$  vertices consists of two disjoint complete subgraphs, each of size  $n$ , that are joined by a single edge. This graph is connected (every vertex has a path to any other vertex), but intuitively not well-connected, since removing the extra edge disconnects the two complete subgraphs.

What is the expansion of this graph? Consider  $S$  to be the vertex set of one of the complete subgraphs. Then  $S$  has conductance  $\frac{1}{1+n(n-1)} = O\left(\frac{1}{n^2}\right)$ . Hence the expansion of the barbell graph is also  $O(1/n^2)$ .

### 2. NORMALIZED LAPLACIANS

We are going to compare graph expansion to Laplacian eigenvalues. We will assume the graph has no isolated vertices (of degree 0).

Recall that  $L_G = \sum_{(i,j) \in E} w_{ij} (\mathbb{1}_i - \mathbb{1}_j)(\mathbb{1}_i - \mathbb{1}_j)^\top = D - A$ , where  $D$  is the diagonal matrix with  $D_{ii} = d(i)$  and  $A$  is the adjacency matrix. (We will drop subscript  $G$  and write  $L = L_G$ .)

All eigenvalues of  $L$  lie in the range  $[0, 2\Delta]$ , where  $\Delta = \max_{i \in V} d(i)$  is the maximum degree:

**Proposition 2.1.**  $-D \preceq A \preceq D$  and  $0 \preceq L \preceq 2D$ . In particular, eigenvalues of  $A$  are between  $-\Delta$  and  $\Delta$ , and those of  $L$  are between 0 and  $2\Delta$ .

*Proof.*  $D - A = \sum_{(i,j) \in E} w_{ij} (\mathbb{1}_i - \mathbb{1}_j)(\mathbb{1}_i - \mathbb{1}_j)^\top \succeq 0$ .

Similarly  $D + A = \sum_{(i,j) \in E} w_{ij} (\mathbb{1}_i + \mathbb{1}_j)(\mathbb{1}_i + \mathbb{1}_j)^\top \succeq 0$ .

Inequalities for  $L$  follow from those of  $A = D - L$ . □

We want to remove the dependence on degree and normalize the Laplacian, so that its eigenvalues are between  $[0, 2]$ . To this end, we “divide”  $L$  by the positive definite matrix  $D$  — or rather, multiply by  $D^{-1/2}$  on both left and right, so that the resulting matrix is still symmetric.

**Definition 2.2.** The normalized adjacency matrix is  $\mathcal{A} = D^{-1/2}AD^{-1/2}$ .

The normalized Laplacian is  $\mathcal{L} = D^{-1/2}LD^{-1/2} = D^{-1/2}(D - A)D^{-1/2} = I - \mathcal{A}$ .

**Claim 2.3.** If  $n$ -by- $n$  real symmetric matrix  $X$  is positive semidefinite, then so is  $Y^\top XY$  for any  $n$ -by- $m$  real matrix  $Y$ . (simple proof omitted)

**Proposition 2.4.** Eigenvalues of  $\mathcal{A}$  are between  $-1$  and  $1$ . Eigenvalues of  $\mathcal{L}$  are between  $0$  and  $2$ .

*Proof.* The above proposition showed that  $D - A \succcurlyeq 0$ . Therefore  $I - \mathcal{A} = D^{-1/2}(D - A)D^{-1/2} \succcurlyeq 0$  by the above claim (with  $X = D - A, Y = D^{-1/2}$ ). Equivalently, all eigenvalues of  $\mathcal{A}$  are at most  $1$ .

Similarly  $D + A \succcurlyeq 0$ . Therefore  $I + \mathcal{A} = D^{-1/2}(D + A)D^{-1/2} \succcurlyeq 0$  by the above claim. Equivalently, all eigenvalues of  $\mathcal{A}$  are at least  $-1$ .

Eigenvalue bounds for  $\mathcal{L}$  follows from eigenvalue bounds for  $\mathcal{A} = I - \mathcal{L}$ .  $\square$

In fact  $0$  is always an eigenvalue of  $\mathcal{L}$ , with eigenvector  $v_1 = D^{1/2}\mathbf{1}$ , because

$$\mathcal{L}v_1 = D^{-1/2}LD^{-1/2}D^{1/2}\mathbf{1} = D^{-1/2}L\mathbf{1} = 0.$$

One can show that  $L$  and  $\mathcal{L}$  have the same zero eigenspace, via the invertible map  $D^{-1/2}$ .

### 3. CHEEGER–ALON–MILMAN INEQUALITY

Let  $0 = \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n \leq 2$  be the eigenvalues of the normalized Laplacian matrix  $\mathcal{L}$  of a graph  $G$ .

In the previous lecture, we showed that  $\lambda_2 = 0$  ( $= \lambda_1$ ) if and only if the graph is connected.

We now quantify well-connectedness of a graph via  $\lambda_2$  (the gap between the two smallest eigenvalues).

**Theorem 3.1** (Cheeger–Alon–Milman).

$$\frac{\lambda_2}{2} \leq \varphi(G) \leq \sqrt{2\lambda_2}.$$

We first prove the easy direction (left inequality).

By Courant–Fischer, taking  $v_1 = D^{1/2}\mathbf{1}$  to be an eigenvector of  $\mathcal{L}$  with eigenvalue  $0$ ,

$$\lambda_2 = \min_{x \perp v_1} \frac{x^\top \mathcal{L}x}{x^\top x} = \min_{x \perp v_1} \frac{x^\top D^{-1/2}LD^{-1/2}x}{x^\top x} = \min_{D^{1/2}y \perp v_1} \frac{y^\top Ly}{y^\top Dy},$$

where  $y = D^{-1/2}x$ .

For every vertex subset  $S$ , we will construct a vector  $y$  satisfying the orthogonality constraint whose Rayleigh quotient is controlled by the conductance of  $S$ :

**Lemma 3.2.** Every nonempty subset  $S \subseteq V$  corresponds to a vector  $y$  such that  $D^{1/2}y \perp v_1$  and

$$\frac{y^\top Ly}{y^\top Dy} = \varphi(S) \frac{d(V)}{d(S)}.$$

If  $d(S) \leq d(V)/2$  (equivalently  $d(\bar{S}) \geq d(V)/2$ ), then the quotient  $\frac{y^\top Ly}{y^\top Dy} \leq 2\varphi(S)$ . Every nonempty subset  $S \subseteq V$  with at most half of the total degree therefore gives us an upperbound  $\lambda_2 \leq 2\varphi(S)$ . Minimizing over all such  $S$  yields  $\lambda_2 \leq 2\varphi(G)$ .

It remains to prove the lemma.

The condition  $D^{1/2}y \perp v_1$  means  $0 = (D^{1/2}y)^\top D^{1/2}\mathbf{1} = y^\top D\mathbf{1} = \sum_{i \in V} d(i)y_i$ .

Also, the denominator in the quotient is  $y^\top Dy = \sum_{i \in V} d(i)y_i^2$ .

In summary,

$$\lambda_2 = \min_{\sum_{i \in V} d(i)y_i = 0} \frac{\sum_{(i,j) \in E} w_{ij}(y_i - y_j)^2}{\sum_{i \in V} d(i)y_i^2}.$$

How to construct  $y$  from  $S$ ? A natural choice is  $y = \mathbb{1}_S$ , the indicator function for  $S$ , i.e.  $y_i = 1$  if  $i \in S$  and  $y_i = 0$  if  $i \notin S$ .

Then the numerator  $\sum_{(i,j) \in E} w_{ij}(y_i - y_j)^2 = w(S, \bar{S})$  and denominator  $\sum_{i \in V} d(i)y_i^2 = d(S)$ , so the quotient gives us exactly  $\varphi(S)$ .

But this  $y$  fails to satisfy the orthogonality constraint, because  $0 \neq \sum_{i \in V} d(i)y_i = d(S)$ .

Instead we pick real numbers  $a$  and  $b$  and assign  $y_i = a$  if  $i \in S$  and  $y_i = b$  if  $i \notin S$ . We want

$$0 = \sum_{i \in V} d(i)y_i = d(S)a + d(\bar{S})b .$$

Solving gives

$$a = \frac{1}{d(S)} \quad \text{and} \quad b = \frac{-1}{d(\bar{S})} .$$

For this  $y$ ,

$$\begin{aligned} \frac{\sum_{i \sim j} w_{ij}(y_i - y_j)^2}{\sum_{i \in V} d(i)y_i^2} &= \frac{w(S, \bar{S}) \left( \frac{1}{d(S)} + \frac{1}{d(\bar{S})} \right)^2}{d(S) \frac{1}{d(S)^2} + d(\bar{S}) \frac{1}{d(\bar{S})^2}} = \frac{w(S, \bar{S}) \left( \frac{1}{d(S)} + \frac{1}{d(\bar{S})} \right)^2}{\frac{1}{d(S)} + \frac{1}{d(\bar{S})}} \\ &= w(S, \bar{S}) \left( \frac{1}{d(S)} + \frac{1}{d(\bar{S})} \right) = w(S, \bar{S}) \frac{d(S) + d(\bar{S})}{d(S)d(\bar{S})} = \frac{w(S, \bar{S})d(V)}{d(S)d(\bar{S})} . \end{aligned}$$

We will prove the hard direction (right inequality) of Cheeger–Alon–Milman in the next lecture.