

MAT 280: Laplacian Eigenfunctions: Theory,
Applications, and Computations
Lecture 20: Introduction to Spectral Graph
Theory–III. Graph Cut and Cheeger Constants,
Isospectral Graphs, and Discrete Laplacian
Eigenvalue Problems

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1 The Cheeger Ratio and the Cheeger Constants of a Graph

This lecture is based primarily on material in [17] and [3, Sec. 2.1-2.3].

Definition 1.1. Given a graph $G = (V, E)$. Let $S \subset V$ be a subset of vertices. Then $\partial S \triangleq \{e = (x, y) \in E \mid x \in S, y \notin S\}$,

Example 1.2. In Figure 1, S is the set of dark nodes, and the dotted lines form ∂S . $\bar{S} \triangleq V \setminus S$. Also, $\text{vol}(S) = m(S) \triangleq \sum_{x \in S} m(x)$.

Definition 1.3. The **Cheeger ratio** for $S \subset V$ is defined as

$$h(S) \triangleq \frac{\#(\partial S)}{\min(\text{vol}(S), \text{vol}(\bar{S}))} = \frac{|\partial S|}{\min(\text{vol}(S), \text{vol}(\bar{S}))},$$

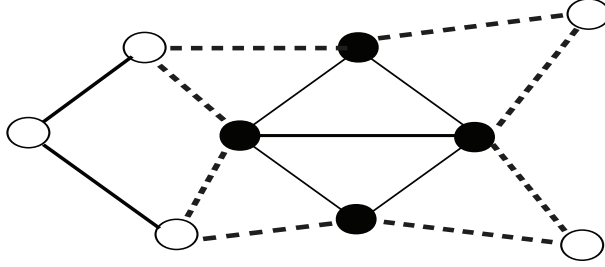


Figure 1: S consists of the dark nodes. The dotted lines form ∂S .

Example 1.4. In Figure 1, $\text{vol}(S) = 4 + 5 + 4 + 5 = 18$ and $\text{vol}(\bar{S}) = 3 + 2 + 3 + 2 + 2 = 12$. $\#(\partial S) = 8$. So the Cheeger ratio is $h(S) = 8/12 = 2/3$.

The Cheeger ratio tells about the quality of the **cut** of V into $S \cup \bar{S} = V$. In Figure 2, a graph that nearly separates into two separate graphs is shown. Note that this graph is well balanced, i.e., $\text{vol}(S) \approx \text{vol}(\bar{S})$; and there exist few connections between S and \bar{S} , i.e., $|\partial S|$ is small. In such case, this cut will give us small $h(S)$.

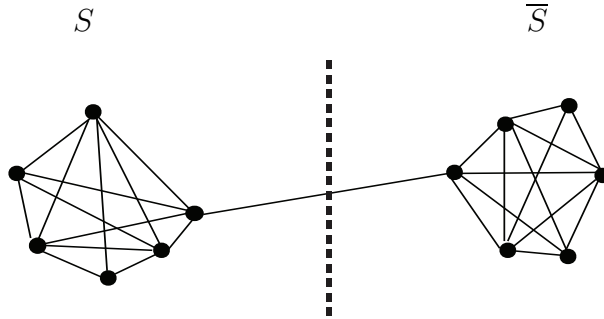


Figure 2: The cut into S and \bar{S} generates a small Cheeger ratio.

Definition 1.5. The Cheeger constants $h_P(G)$ and $h_A(G)$ are defined as

$$h_P(G) \triangleq \inf_{S \subset V, S \neq \emptyset} h(S),$$

$$h_A(G) \triangleq \inf_{S \subset V, S \neq \emptyset} \frac{\#(\partial S)}{\min\{\#(S), \#(\bar{S})\}} = \inf_{S \subset V, S \neq \emptyset} \frac{|\partial S|}{\min(|S|, |\bar{S}|)}.$$

The following theorem provides the upper bounds for $h_P(G)$ and $h_A(G)$.

Theorem 1.6.

1. *Dodzink and Kendall (1986) [5]:* $h_P(G) \leq \sqrt{2\lambda_2^{(P)}}$, i.e., $\lambda_2^{(P)} \geq \frac{h_P^2(G)}{2}$;
2. *Mohar (1987) [12]:* $h_A(G) \leq \sqrt{\lambda_2^{(A)}(2m_\infty(G) - \lambda_2^{(A)})}$ for $p = \#(V) \geq 4$;
3. *Tan (2003) [15]:* $h_P(G) \leq \sqrt{\lambda_2^{(P)}(2 - \lambda_2^{(P)})}$, for $p \geq 4$. Equality holds if and only if $G = K_{1,p-1}$.

These inequalities are used to evaluate the lower bounds of $\lambda_2^{(P)}$, $\lambda_2^{(A)}$.

For the upper bounds of the eigenvalues, we have

Theorem 1.7 (Urakawa 1999 [16]).

For $2 \leq j \leq \lfloor \frac{\text{diam}(G)}{2} \rfloor$,

$$\lambda_j^{(A)} \leq m_\infty(G) - 2\sqrt{m_\infty(G) - 1} \cos\left(\frac{\pi}{\frac{\text{diam}(G)}{2j} + 1}\right);$$

$$\lambda_j^{(P)} \leq 1 - \frac{2\sqrt{m_\infty(G) - 1}}{m_\infty(G)} \cos\left(\frac{\pi}{\frac{\text{diam}(G)}{2j} + 1}\right).$$

Note that there also exists the simpler upper bound:

$$\lambda_2^{(P)} \leq 2h_P(G).$$

Let us prove this inequality.

Proof. Using the minimum principle (MP₂), we have

$$\begin{aligned}
\lambda_2^{(P)} &= \inf_{f \in C_0(V), \langle f, 1 \rangle = 0, f \neq 0} \frac{\langle df, df \rangle_0}{\|f\|^2} \\
&= \inf_{f \in C_0(V), \langle f, 1 \rangle = 0, f \neq 0} \frac{\langle \Delta_A f, df \rangle_0}{\|f\|^2} \\
&= \inf_{f \in C_0(V), \langle f, 1 \rangle = 0, f \neq 0} \frac{\sum_{x \sim y} (f(x) - f(y))^2}{\sum_{x \in V} m(x) f^2(x)},
\end{aligned}$$

where we have used the following definitions:

$$\begin{aligned}
\langle f, g \rangle_0 &\triangleq \sum_{u \in V} f(u)g(u), \text{ and} \\
\|f\|^2 &\triangleq \sum_{u \in V} m(u)f^2(u).
\end{aligned}$$

Now suppose S achieves the Cheeger constant $h_P(G)$. Set

$$f(x) \triangleq \begin{cases} \frac{1}{\text{vol}(S)} & \text{if } x \in S, \\ -\frac{1}{\text{vol}(\bar{S})} & \text{if } x \notin S. \end{cases}$$

Then

$$\begin{aligned}
\langle f, 1 \rangle &= \sum_{x \in V} f(x) \cdot 1 \cdot m(x) \\
&= \sum_{x \in S} \frac{1}{\text{vol}(S)} m(x) + \sum_{x \in \bar{S}} \left(-\frac{1}{\text{vol}(\bar{S})} \right) m(x) \\
&= \frac{\text{vol}(S)}{\text{vol}(S)} - \frac{\text{vol}(\bar{S})}{\text{vol}(\bar{S})} \\
&= 0.
\end{aligned}$$

We also have

$$\begin{aligned}
\|f\|^2 &= \sum_{x \in V} f^2(x)m(x) \\
&= \sum_{x \in S} \frac{m(x)}{\text{vol}(S)^2} + \sum_{x \in \bar{S}} \frac{m(x)}{\text{vol}(\bar{S})^2} \\
&= \frac{1}{\text{vol}(S)} + \frac{1}{\text{vol}(\bar{S})}.
\end{aligned}$$

Moreover,

$$\begin{aligned}
\langle df, df \rangle_0 &= \langle \Delta_A f, f \rangle_0 \\
&= \sum_{x \sim y} (f(x) - f(y))^2 \\
&= \sum_{\substack{x \sim y \\ x, y \in S}} (f(x) - f(y))^2 + \sum_{\substack{x \sim y \\ x \in S, y \in \bar{S}}} (f(x) - f(y))^2 + \sum_{\substack{x \sim y \\ x, y \in \bar{S}}} (f(x) - f(y))^2 \\
&= 0 + |\partial S| \left(\frac{1}{\text{vol}(S)} + \frac{1}{\text{vol}(\bar{S})} \right)^2 + 0.
\end{aligned}$$

Now, using the inequality $\frac{a+b}{ab} \leq \frac{2}{\min(a,b)}$ for $a, b > 0$, we find

$$\begin{aligned}
\lambda_2^{(P)} &\leq \frac{|\partial S| \cdot \left(\frac{1}{\text{vol}(S)} + \frac{1}{\text{vol}(\bar{S})} \right)^2}{\frac{1}{\text{vol}(S)} + \frac{1}{\text{vol}(\bar{S})}} \\
&= |\partial S| \left(\frac{1}{\text{vol}(S)} + \frac{1}{\text{vol}(\bar{S})} \right) \\
&\leq \frac{2|\partial S|}{\min(\text{vol}(S), \text{vol}(\bar{S}))} \\
&= 2h_P(G).
\end{aligned}$$

The last equality is derived by the assumption that S achieves $h_P(G)$.

□

2 Isospectrality

Definition 2.1. Let $G = (V, E)$. Define $\text{Spec}(G, A)$, $\text{Spec}(G, \Delta_A)$ and $\text{Spec}(G, \Delta_P)$ as the sets of spectra (eigenvalues) of A , Δ_A and Δ_P of G , respectively. For given G_1 and G_2 , if $\text{Spec}(G_1, A) = \text{Spec}(G_2, A)$, then G_1 and G_2 are said to be **cospectral**. If $\text{Spec}(G_1, \Delta_A) = \text{Spec}(G_2, \Delta_A)$ or $\text{Spec}(G_1, \Delta_P) = \text{Spec}(G_2, \Delta_P)$, then G_1 and G_2 are said to be **isospectral**.

For cospectrality and isospectrality of graphs, see [6], [1], [14] and [7]. For more on the isospectral inequalities, see [9], [10] and chapter 2. of [3, Chap. 2].

Example 2.2. (Fisher 1966 [6], Baker 1966 [1]). Figure 3 shows two cospectral graphs, however they are not isomorphic.

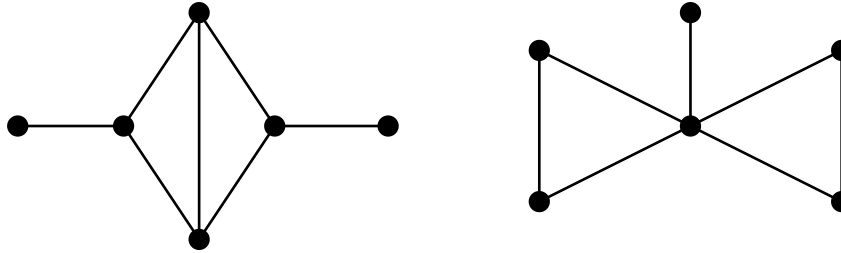


Figure 3: Cospectral but not isomorphic graphs

Example 2.3. (Fuji-Katsuda 1999 [7], Tan 1998 [14]). Figure 4 is an example of two Δ_A -isospectral graphs.

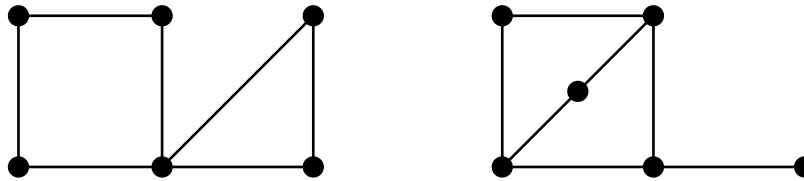


Figure 4: Δ_A -isospectral graphs.

Example 2.4. (Fuji-Katsuda 1999 [7], Tan 1998 [14]). Figure 5 shows an example of two Δ_P -isospectral graphs.

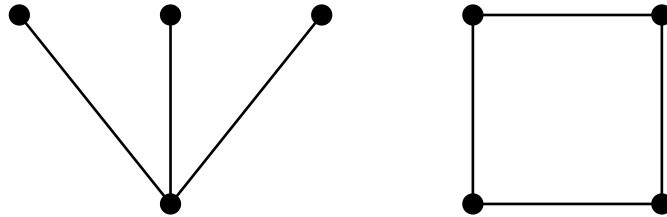


Figure 5: Δ_P -isospectral graphs.

3 Discrete Laplacian Eigenvalue Problems

Definition 3.1. $G = (V, E)$ is said to have **boundary** $\partial G = (\partial V, \partial E)$ if the following two conditions are met:

1.

$$\begin{cases} V = \overset{\circ}{V} \cup \partial V & \overset{\circ}{V} \cap \partial V = \emptyset; \\ E = \overset{\circ}{E} \cup \partial E & \overset{\circ}{E} \cap \partial E = \emptyset. \end{cases}$$

2. For each $e = (x, y) \in E, x, y \in V$,

$$\begin{aligned} e \in \overset{\circ}{E} &\iff x, y \in V; \\ e \in \partial E &\iff x \in \overset{\circ}{V}, y \in \partial V \text{ or } x \in \partial V, y \in \overset{\circ}{V}. \end{aligned}$$

Example 3.2. Note that there exists some arbitrariness of ∂G . In Figure 6, the open nodes are in $\overset{\circ}{V}$, the closed nodes in ∂V , the solid lines in E , and the dashed lines in ∂E .

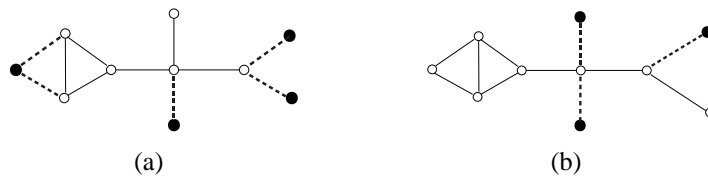


Figure 6: The arbitrariness of ∂G .

Consider the discrete Laplacian eigenvalue problem, defined as

$$\begin{cases} \Delta_P u = \mu^{(P)} u & \text{in } \overset{\circ}{V}, \\ u = 0 & \text{on } \partial V. \end{cases} \text{ Dirichlet-Laplacian (D-L}^{(P)});$$

$$\begin{cases} \Delta_P u = \nu^{(P)} u & \text{in } \overset{\circ}{V}, \\ du = 0 & \text{on } \partial E. \end{cases} \text{ Neumann-Laplacian (N-L}^{(P)}).$$

Of course, we can define the discrete Laplacian eigenvalue problem with ΔA . We can order these eigenvalues as

$$D - L^{(A)} : \mu_1^{(A)} \leq \mu_2^{(A)} \leq \dots \leq \mu_k^{(A)}, \quad k = \#(\overset{\circ}{V}).$$

$$D - L^{(P)} : \mu_1^{(P)} \leq \mu_2^{(P)} \leq \dots \leq \mu_k^{(P)}, \quad k = \#(\overset{\circ}{V}).$$

We have the following theorem.

Theorem 3.3. $\mu_1^{(P)} > 0$, and $\mu_1^{(A)} > 0$, both with multiplicity 1. And there exist $\varphi_1^{(P)}(x) > 0$ and $\varphi_1^{(A)}(x) > 0$ for all $x \in \overset{\circ}{V}$.

Proof. The proof is essentially the same as the continuum case, i.e., use the discrete Green's identity with the boundary condition. For example, let $\varphi_1^{(P)}$ be an eigenfunction for $\mu_1^{(P)}$ with $\|\varphi_1^{(P)}\| = 1$, then using the discrete Green's identity,

$$\mu_1^{(P)} = \langle \Delta \varphi_1^{(P)}, \varphi_1^{(P)} \rangle = \langle d\varphi_1^{(P)}, d\varphi_1^{(P)} \rangle = \|d\varphi_1^{(P)}\|^2 \geq 0.$$

If $d\varphi_1^{(P)} \equiv 0$, then $\varphi_1^{(P)} \equiv \text{const}$ on V . But $\varphi_1^{(P)} = 0$ on ∂V forces us to have $\varphi_1^{(P)} \equiv 0$. So $\mu_1^{(P)} \neq 0$. Therefore, $\mu_1^{(P)} > 0$. □

Theorem 3.4 (The discrete Faber-Krahn inequality). *see [9], [10].*

If $\#(\overset{\circ}{E} \cup \partial E) = n$, then

$$\mu_1^{(P)}(L_n) \leq \mu_1^{(P)}(G)$$

where equality holds if and only if $G = L_n$. L_n a graph is shown in Figure 7. There are n nodes. Only the last to the right belongs to ∂V , the rest belong to $\overset{\circ}{V}$, and the same is true for the edges.

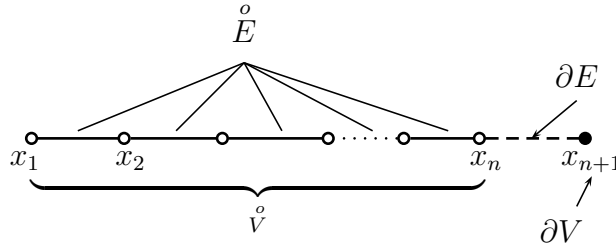


Figure 7: Graph L_n .

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