图谱论组会讲稿 ||

We shall discuss some subtle connections of spectral graph theory and classical analysis, including

- the Laplacian operator of a (simple) graph
- eigenvalue of weighted graphs
- application in theories of random walks

The Laplacian

On the threshold of spectral graph analysis.

Laplatian operator of a graph

To begin with that, we shall review the definition of *adjacency matrix* A(G) and *degree matrix* D(G) for a simple graph G. The **Laplacian matrix**

$$L := D - A = egin{cases} \deg v & u = v, \ -1 & u \sim v, \ 0 & ext{else}. \end{cases}$$

A brandnew concept named Laplacian operator (Laplacian for short) is defined as

$$\mathcal{L} := egin{cases} 1 & u = v, \ -\sqrt{\deg v \cdot \deg u}^{-1} & u \sim v, \ 0 & ext{else}. \end{cases}$$

Here $\mathcal{L} = D^{-1/2} L D^{-1/2}$, where D^{lpha} is defined as

$$(\lambda_1^{\alpha}, \lambda_2^{\alpha}, \dots, \lambda_k^{\alpha}, 0, \dots, 0).$$

Consider the function $f:V(G)\to\mathbb{R}$ (or \mathbb{C}). Here we shall define the *inner product* (\leftrightarrow what we done in *discrete Fourier analysis*).

$$\langle f,g
angle := \sum_{v\in V} f(v) \overline{g(v)}$$

For simplicity, we only discuss the real functions below.

Laplacian of a function can be also defined via the theorem of *unit decomposition*. Let δ_v be *characteristic function* w.r.t. v, then

$$\begin{split} \mathcal{L}f &= \sum_{v \in V} \left\langle \mathcal{L}f, \delta_v \right\rangle \cdot \delta_v \\ &= \sum_{v \in V} \sum_{u \in V} \left(f(u) - \sum_{u' \sim u} \frac{f(u')}{\sqrt{\deg u \cdot \deg u'}} \right) \delta_v(u) \cdot \delta_v \\ &= \sum_{v \in V} \frac{1}{\sqrt{\deg v}} \sum_{u \sim v} \left(\frac{f(v)}{\sqrt{\deg v}} - \frac{f(u)}{\sqrt{\deg u}} \right) \cdot \delta_v \end{split}$$

Hence $\forall v \in V$, we have

$$(\mathcal{L}f)(v) = rac{1}{\sqrt{\deg v}} \sum_{u \sim v} \left(rac{f(v)}{\sqrt{\deg v}} - rac{f(u)}{\sqrt{\deg u}}
ight).$$

Notice that $\mathcal{L} \sim \sum_i \partial_{x_i}(\partial x_i)$ in the continuous case in a *uniform-densitied* and *nongradient* medium.

Basic facts about eigenvalues

Suppose that λ is a eigenvalue of \mathcal{L} w.r.t. some eigenfunction g, then

$$rac{\langle g, \mathcal{L}g
angle}{\langle g, g
angle} = \lambda.$$

The L.H.S. outcomes

$$rac{\langle g, \mathcal{L}g
angle}{\langle g, g
angle} = rac{\sum_v rac{g(v)}{\sqrt{\deg v}} \sum_{u \sim v} \left(rac{g(v)}{\sqrt{\deg v}} - rac{g(u)}{\sqrt{\deg u}}
ight)}{\sum_v g(v)^2} =?$$

Whoops, how so? The dawn of victory might occur whence you think back

$$\langle f, Lf
angle = \sum_{u \sim v} (f(v) - f(u))^2.$$

It suggests that once we substitude g with $T^{1/2}f$, we have

$$rac{\langle g, \mathcal{L}g
angle}{\langle g, g
angle} = rac{\sum_{u \sim v} (f(u) - f(v))^2}{\sum_v f(v)^2 \deg v}.$$

Furthermore, the laplacian is semi-positive since 0 is the smallest eigenvalue of L with corresponding eigenvector $\mathbf{1}_n$. Hence $T^{1/2}\mathbf{1}$ be an eigenfunction s.t.

$$\mathcal{L}(T^{1/2}\mathbf{1}) = T^{-1/2}L\mathbf{1}_n = 0.$$

 $\mathbf{1}:V o\{1\}$ is constant function.

We call eigenvalue λ a main eigenvalue if and only if $\mathcal{E}_{\lambda} \nsubseteq \mathbf{1}^{\perp}$. The **spectra** of \mathcal{L} are $0 \leq \lambda_0 \leq \lambda_1 \leq \cdots \leq \lambda_{n-1}$. We notice that the multiplicity of main eigenvalue 0 equals the number of *components* in G (trivial, but why?).

Hence λ_1 is defined as (notice that $\left\langle T^{1/2}f,T^{1/2}\mathbf{1}\right
angle=0$ while T is *self-adjoint*.)

$$\lambda_1 = \inf_{f \perp T1} rac{\sum_{u \sim v} (f(u) - f(v))^2}{\sum_v f(v)^2 \deg v}.$$

In continuous case... maybe we could steal a glance at the *Lichnerowicz-Obata theorem*. Let M denotes a compact Riemannian manifold without boundary. Consider the eigenfunction satisfying the equation $-\Delta u=\lambda u$, we obtain

$$\lambda \int_M u^2 = -\int_M (\Delta u) u = \int_M (
abla u)^2.$$

Since the constant function is a trivial solution, the non-trivial solutions should be subject to $\int_M u=0$. The function f s.t. $\left\langle T^{1/2}f,T^{1/2}\mathbf{1}\right\rangle=0$ is also known as the **harmonic eigenfunction**

Since $T^{1/2}f\perp T^{1/2}\mathbf{1}$, we can rewrite the expression of λ_1 into

$$\lambda_1 = \inf_f \sup_{t \in \mathbb{R}} rac{\sum_{u \sim v} (f(u) - f(v))^2}{\sum_v (f(v) - t)^2 \deg v}.$$

whence the maximum holds, $2\sum_v (f(v)-t)\deg v=0$. Let

$$\overline{f} := rac{\sum_{v} f(v) \deg v}{\sum_{v} \deg v},$$

where $\sum_v \deg v = 2|E|$ is the **volume** of G. Notice that (why?)

$$\sum_v (f(v)-\overline{f})^2 \deg v = \sum_{u,v} (f(u)-f(v))^2 \deg u \deg v.$$

Hence

$$\lambda_1 = 2|E|\inf_f rac{\sum_{u\sim v}(f(u)-f(v))^2}{\sum_{u\sim v}(f(u)-f(v))^2\deg u\deg v}.$$

We can iterate the formula of Rayleigh quotient to get

$$\lambda_k = \inf_{f \perp T \cdot V_k} rac{\sum_{u \sim v} (f(u) - f(v))^2}{\sum_v f(v)^2 \deg v}.$$

Here $V_k=\operatorname{span}\{\oplus_{i=0}^{k-1}x_k\}$, where x_k is the k-th eigenvector (notice that $\mathcal L$ is diagonalisable).

Spectrum of ${\cal L}$

For $K_{m,n}$, P_n , C_n , how to evaluate their eigenvalues and corresponding eigenvectors? Hint: using *Chebyshev polynomials* to evaluate the spectra of path and circles, deducing the corresponding eigenvectors by multiplying $(\alpha, \alpha^2, \dots, \alpha^n)$.

As for graphs in general, some basic facts (bounds of eigenvalues) includes:

- $\sum_i \lambda_i \leq n$, with equality holds $\Leftrightarrow G$ has no isolated points \Leftrightarrow trace no less than n. Therefore, $\lambda_1 \leq \frac{n}{n-1} \leq \lambda_{n-1}$.
- $\lambda_1 \leq 1$ when G is not complete (why?).

$$\bullet \ \ \lambda \leq \sup_f \frac{\sum_{u \sim v} (f(u) - f(v))^2}{\sum_u f(u)^2 \deg u} \leq \frac{2 \sum_{u \sim v} (f(u)^2 + f(v)^2)}{\sum_u f(u)^2 \deg u} \leq 2.$$

The equality in the last statement holds some non-trivial case. For instance, the following statements are equivalent (Hint: proof of connected case is sufficient enough; consider the eigen function $\mathbb{1}_A \cdot f(x) - \mathbb{1}_B \cdot f(x)$ for any bipartite with consisting part A and B):

- 1. G is bipartite.
- 2. $\lambda_{n-i}=2$ if G has i+1 connected components.
- 3. For each λ_i , the value $2 \lambda_i$ is also an eigenvalue of G.

(Spectrum bounded by diameter and volume) Let D denotes the *diametre*. For $f \perp T\mathbf{1}$, let v_M be a vertex such that $|f(v_M)| = \max_v |f(v)|$. Since $\sum_v f(v) \deg v = 0$, there exists a u_0 such that $f(u_0)f(v_0) < 0$. Let P_{u_0,v_0} be the shortest path that connects u_0 and v_0 , i.e.

$$v_0-v_1-\cdots-v_k=u_0.$$

Hence,

$$egin{aligned} \lambda_1 &= rac{\sum_{u \sim v} (f(u) - f(v))^2}{\sum_u f(u)^2 \deg u} \ &\geq rac{\sum_{i=1}^k (f(v_i) - f(v_{i-1}))^2}{f(v_0)^2 \sum_u \deg u} \ &\geq rac{k \cdot \left(rac{f(v_0) - f(v_k)}{k}
ight)^2}{f(v_0)^2 \sum_u \deg u} \ &\geq rac{1}{D \cdot \sum_u \deg u} \end{aligned}$$

A pure corollary deriving from $\mathcal{L}(\sqrt{T}f) = \lambda \sqrt{T}f$ is

$$\lambda f(v) = rac{1}{\deg v} \sum_{u \sim v} (f(u) - f(v))$$

Hint: Proof via $\mathcal{L}f=\lambda\sqrt{T}f$.

Spectrum of the weighted graphs

Denote w as a function defining *edge weights*, i.e.

$$w:E o \mathbb{R}, (u,v)\mapsto w(u,v).$$

Here w(u,v)=w(v,u)=w(e) is the weight of edge e between u and v.

Thence $\deg v = \sum_u w(u,v)$. In many cases of applications, the graph may equip with *self-loops* (any examples?). The Laplacian matrix is revised to be

$$L := D - A = egin{cases} \deg v - w(v,v) & u = v, \ -w(u,v) & u \sim v, \ 0 & ext{else}. \end{cases}$$

Hence

$$\mathcal{L} := egin{cases} 1 - \dfrac{w(v,v)}{\deg v} & u = v, \ - \dfrac{w(u,v)}{\sqrt{\deg v \cdot \deg u}} & u \sim v, \ 0 & ext{else}. \end{cases}$$

Similarly, some above-mentioned theorems are re-stated as:

ullet Here $\mathcal{L}=D^{-1/2}LD^{-1/2}$, where D^lpha is defined as

$$(\lambda_1^{\alpha}, \lambda_2^{\alpha}, \ldots, \lambda_k^{\alpha}, 0, \ldots, 0).$$

ullet For $f:V o \mathbb{R}$,

$$Lf(v) = \sum_{u\sim v} (f(u) - f(v))w(u,v).$$

Hence
$$\langle f, Lf \rangle = \sum_{u \sim v} (f(u) - f(v))^2 w(u,v)$$
. λ_1

• Question: How to determine λ_k ?