

Ans 1  $G = (V, E)$ ,  $n = \# \text{vertices}$ ,  $m = \# \text{edges}$

$D = \text{degree matrix}$

$M = D^{-1} A$ ,  $A = \text{adjacency matrix}$  ( $M$  is not symmetric since  $m_{ij} = \frac{a_{ij}}{\deg(i)}$ ,  $m_{ji} = \frac{a_{ji}}{\deg(j)}$ )

$\bar{M} = D^{-1/2} A D^{-1/2}$  is symmetric

① Claim: eigenvalues of  $M$  are equal with multiplicities to those of  $\bar{M}$ .

Proof: Let eigenbasis of  $\bar{M} = v_1, v_2, \dots, v_n$ . Then,

$$\text{let } w_i = D^{-1/2} v_i$$

$$\Rightarrow M w_i = D^{-1} A D^{-1/2} v_i = D^{-1/2} (\bar{M} v_i) = D^{-1/2} \cdot \lambda_i v_i = \lambda_i w_i$$

Now, to show that  $w_i$  are linearly independent.

$$W = [w_1 \dots w_n] = D^{-1/2} [v_1 \dots v_n]$$

$$\det W = (\det D) \det V \neq 0$$

② Claim: eigenvalues of  $\bar{M}$  lie between  $-1$  and  $1$ .

Proof:  $I + \bar{M}$  and  $I - \bar{M}$  being p.s.d. will finish the proof.

$$\text{case 1: } I - \bar{M} = I - D^{-1/2} A D^{-1/2} = D^{-1/2} (D - A) D^{-1/2}$$

$$v^T D^{-1/2} (D - A) D^{-1/2} v = w^T (D - A) w$$

$$\text{where } D - A = \sum_{e \in E} h_e, \quad h_e = \begin{pmatrix} 1 & \dots & -1 \\ & & \\ -1 & \dots & 1 \end{pmatrix}$$

$$\Rightarrow w^T (D - A) w = \sum_{e = (u,v) \in E} w^T h_e w = \sum_{(u,v) \in E} (w_u - w_v)^2 \geq 0. \text{ Hence, } I - \bar{M} \text{ is p.s.d.}$$

$$\text{Similarly, } I + \bar{M} \text{ is p.s.d. as } v^T (I + \bar{M}) v = \sum_{(u,v) \in E} (w_u + w_v)^2 \geq 0.$$

Ans 2

① Claim:  $v$  is an eigenvector of  $M$  then  $D^{1/2} v$  is an eigenvector of  $\bar{M}$

$$\text{Proof: } \bar{M} D^{1/2} v = D^{-1/2} A v = D^{-1/2} (M v) = \lambda D^{1/2} v$$

② Claim:  $\forall i \in V(G)$ ,  $\lim_{t \rightarrow \infty} \|\bar{M}^t \mathbf{1}_i - x_i\|_2 = 0$  where  $x_i(j) = \sqrt{d_i d_j} / 2m$

Proof: let  $\mathbf{1}_i = \sum_{j=1}^n \alpha_j v_j$  where  $v_j$  is orthonormal eigenbasis of  $\bar{M}$ .

$$\bar{M}^t \mathbf{1}_i = \sum_j \alpha_j \bar{M}^t v_j = \sum_j \alpha_j d_i^{t/2} v_j. \text{ Hence, as } t \rightarrow \infty \text{ only } |d_i| = 1 \text{ survives.}$$

$\mathbf{1}$  is eigenvector of  $M$  then,  $\frac{D^{1/2} \mathbf{1}}{\|D^{1/2} \mathbf{1}\|}$  is eigenvector with  $\lambda = 1$  for  $\bar{M}$

As given in question,  $\lambda = 1$  is a unique eigenvalue ( $\max(|\lambda_2|, |\lambda_n|) \leq 1 - \epsilon$ )

$$\text{Hence, } \lim_{t \rightarrow \infty} M^t \mathbf{1}_i = \lim_{t \rightarrow \infty} \alpha_i \mathbf{v}_1 = \frac{\langle \mathbf{1}_i, D^{1/2} \mathbf{1} \rangle D^{1/2} \mathbf{1}}{\|D^{1/2} \mathbf{1}\|_2^2} = \frac{\sqrt{d_i} \cdot \{\sqrt{d_j}\}_{j=1}^n}{\left(\sum_{i=1}^n d_i\right)}$$

$$\text{Hence, } \alpha_i(j) = \sqrt{d_i d_j} / 2m.$$

Claim: Stationary distribution for random walks on  $M$  is  $\pi(u) = du/2m$

Proof: Interpretation: Stationary dist  $\equiv (\pi M = M)$

$$\text{Let } \pi(u) = \frac{du}{2m}.$$

$$\begin{aligned} (\pi^t M)_i &= \sum_{j=1}^n \pi_j M_{ji} = \sum_{j=1}^n \frac{\deg(j)}{2m} \cdot \frac{1}{\deg(j)} a_{ji} = \sum_{j=1}^n \frac{a_{ji}}{2m} \\ &= \frac{\deg(i)}{2m} \end{aligned}$$

Interpretation: stationary distribution = converging distribution

$$\text{Let } p = \sum_{i=1}^n \alpha_i \mathbf{1}_i^T, \text{ where } \sum \alpha_i = 1.$$

$$\begin{aligned} \text{Then, } \tilde{p} &= \lim_{t \rightarrow \infty} p M^t = \lim_{t \rightarrow \infty} p D^{-1/2} \bar{M}^t D^{1/2} \\ &= \lim_{t \rightarrow \infty} \sum_{i=1}^n \alpha_i \mathbf{1}_i^T D^{-1/2} \bar{M}^t D^{1/2} \\ &= \lim_{t \rightarrow \infty} \left( \sum_{i=1}^n \alpha_i D^{+1/2} \bar{M}^t D^{-1/2} \mathbf{1}_i \right)^T \\ &= \left( \sum_{i=1}^n \alpha_i D^{+1/2} \left( \lim_{t \rightarrow \infty} \bar{M}^t \frac{1}{\sqrt{d_i}} \mathbf{1}_i \right) \right)^T \\ &= \sum_{i=1}^n \alpha_i D^{+1/2} \frac{1}{\sqrt{d_i}} \alpha_i \quad \text{where } \alpha_i(j) = \frac{\sqrt{d_i d_j}}{2m} \end{aligned}$$

$$\begin{aligned} \text{Hence, } \tilde{p}(j) &= \sum_{i=1}^n \alpha_i \frac{1}{\sqrt{d_i}} (D^{+1/2} \alpha_i)(j) = \sum_{i=1}^n \frac{\sqrt{d_j}}{\sqrt{d_i}} \alpha_i(j) \alpha_i \\ &= \sum_{i=1}^n \frac{d_j}{2m} \alpha_i = \frac{\deg(j)}{2m} \end{aligned}$$

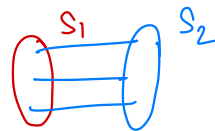
Ans 2  $G = (V, E)$  is  $d$ -regular,  $L = \mathbb{I} - \frac{A}{d}$ , eigenvalues of  $L$  are

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

Claim:  $G$  is bipartite iff  $d_n = 2$

( $\Rightarrow$ ) let  $G$  be bipartite.

Let  $S_1 = \{1, 2, \dots, l\}$ ,  $S_2 = [n] \setminus S_1$ . (wlog)



We just need to show existence of  $-1$  eigenvalue for  $\frac{A}{d}$ .

$$P(i) = \begin{cases} \frac{p d_i}{\sum_{i \in S_1} d_i} & , i \in S_1 \\ -\frac{p d_i}{\sum_{i \in S_2} d_i} & , i \in S_2 \end{cases} \quad \left| \quad \begin{array}{l} \text{Then, after one step i.e. } \tilde{P} = \\ \tilde{P}_i = \sum_{j \in S_2} \frac{-p d_j}{\sum_{i \in S_2} d_i} \cdot \frac{1}{d_j} \quad i \in S_1 \\ = \frac{-p d_i}{\sum_{i \in S_1} d_i} = -\frac{p d_i}{D} \end{array} \right.$$

where  $\sum_{i \in S_1} d_i = \sum_{i \in S_2} d_i = D$

Similarly for  $S_2$ , we have

$$\tilde{P}_i = \sum_{j \in S_1} \frac{p d_j}{\sum_{i \in S_1} d_i} \cdot \frac{1}{d_j} \quad , i \in S_2 \\ = +\frac{p d_i}{D}$$

( $\Leftarrow$ ) Now, given  $d_n = 2$  i.e.  $\frac{A}{d}$  has eigenvalue  $-1$ .

To show  $G$  is bipartite.

$$\text{let } v^T \frac{A}{d} = -v^T, \left( v^T \frac{A}{d} \right)_i = \sum_{j=1}^n v_j \frac{a_{ji}}{d} = -v_i$$

$$\Rightarrow \sum_{j \in N(i)} [v_j] = -v_i, \text{ let } S_1 = \{i \in [n] \mid v_i \geq 0\} \\ S_2 = \{i \in [n] \mid v_i < 0\}$$

claim: if  $\arg \max_i |v_i| = i^*$ ,  $|v_{i^*}| = k$ , then  $|v_i| = k \forall i$

let  $\arg \max_i |v_i| = i^*$ . Then,  $\sum_{j \in N(i^*)} [v_j] = -v_{i^*} = -k$ .

Hence  $\forall j \in N(i^*)$ ,  $v_j = -v_{i^*}$  otherwise at least one  $i$  has  $|v_i| > k$ .

Hence,  $\forall j \in N(i^*)$ , if  $i^* \in S_1 \Leftrightarrow j \in S_2$  and  $i \in S_2 \Leftrightarrow j \in S_1$ .

Repeating this argument, the absolute value of  $v_i = k$  for all  $i$  in  $G$  since all nodes are in same connected component as  $i^*$ , and we get that

$$\text{if } i \in S_1 \Rightarrow \forall j \in N(i), j \in S_2, v_j = -v_i \text{ (if not, } |\sum_{j \in N(i)} v_j| < \sum |v_j| = k)$$

Hence,  $S_1, S_2$  give partitions of the bipartite graph.

Ans 4  $G = (V, E)$  is a  $d$ -regular bipartite graph.

$$L = I - \frac{A}{d} \text{ with eigs } 0 = \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n = 2$$

claim:  $\lambda_i + \lambda_{n-i+1} = 2$

Proof: It is enough to show that if  $v$  is an eigenvector with eigenvalue  $\lambda$ ,  $v' = f(v)$  with eigenvalue  $-\lambda$  where  $f$  is a bijection.

$$(f(v))_i = \begin{cases} v_i & , i \in S_1 \\ -v_i & , i \in S_2 \end{cases} \quad (\text{Clearly } f \text{ is a bijection}).$$

Now, let  $\frac{Av}{d} = \lambda v$

$$\left( \frac{A}{d} f(v) \right)_i = \sum_{j=1}^n \frac{a_{ij}}{d} \cdot v'_j = \sum_{j \in N(i)} \frac{v'_j}{d} = - \sum_{j \in N(i)} \frac{v_j}{d}$$

$$\left\{ \text{But } \left( \frac{A}{d} v \right)_i = \sum_{j \in N(i)} \frac{v_j}{d} = \lambda v_i \right\}$$

$$= -\lambda v_i \quad \square$$

Ans 5  $G = (V, E)$ ,  $|C_1| = |C_2| = n/2$ ,  $|V| = n$ ,  $(C_1 \cap C_2 = \emptyset)$   
 $G[C_1], G[C_2]$  have conductance  $\geq \phi$   
 $|E(C_1, C_2)| \leq \epsilon n d$ ,  $\epsilon \ll \phi^2$   
 $0 = \lambda_1 \leq \lambda_2 \leq \lambda_3 \leq \dots \leq \lambda_n = 2$

claim: If  $H \subseteq G$  then  $L_H \leq L_G$

Proof: Showing  $L_G - L_H$  is p.s.d. is enough.

$$x^T (L_G - L_H) x = \sum_{e \in E(G) \setminus E(H)} x^T L_e x = \sum_{(u,v) \in E(G) \setminus E(H)} (x_u - x_v)^2 \geq 0$$

Now, let's remove  $E(C_1, C_2)$  and replace the cut-edges with self-loops.

Now, in the resultant graph's Laplacian,  $\lambda_1 = \lambda_2 = 0$  with eigenspace  $\text{span} \langle v_1, v_2 \rangle$  where  $(v_1)_i = \begin{cases} 1, & i \in C_1 \\ 0, & i \in C_2 \end{cases}$ ,  $(v_2)_i = \begin{cases} 0, & i \in C_1 \\ 1, & i \in C_2 \end{cases}$

Then,  $v_3$  corresponding to  $\lambda_3$  is  $\perp$  to  $v_1, v_2$ . Hence,

$$\sum_{i \in C_1} (v_3)_i = \sum_{j \in C_2} (v_3)_j = 0$$

Consider vectors  $u \in \mathbb{R}^{|C_1|}$ ,  $v \in \mathbb{R}^{|C_2|}$ , s.t.  $\sum u_i = \sum v_i = 0$ .

Let  $L'$  denote Laplacian of new graph with  $E(C_1, C_2)$  removed.

$$\text{Then, } \lambda_3 = \min \frac{\begin{bmatrix} u^T & v^T \end{bmatrix} L' \begin{bmatrix} u \\ v \end{bmatrix}}{u^T u + v^T v}, \text{ where columns of } L' \text{ are ordered appropriately}$$

$$\lambda_3 = \min \frac{u^T L_{C_1} u + v^T L_{C_2} v}{u^T u + v^T v}, \quad \sum u_i = \sum v_i = 0.$$

Now, by Cheeger's inequality on  $C_1, C_2$  individually,

$$\lambda_2^{C_1} = \min \frac{u^T L_{C_1} u}{u^T u}, \quad \phi \leq \phi_{C_1} \leq \sqrt{2\lambda_2^{C_1}} \\ \Rightarrow \lambda_2^{C_1} \geq \frac{\phi^2}{2}.$$

$$\text{Similarly } \lambda_2^{C_2} \geq \frac{\phi^2}{2}.$$

$$\lambda_3 = \min \frac{u^T L_{C_1} u + v^T L_{C_2} v}{u^T u + v^T v} \quad \text{where} \quad \begin{aligned} u^T L_{C_1} u &\geq \frac{\phi^2}{2} u^T u \\ v^T L_{C_2} v &\geq \frac{\phi^2}{2} v^T v \end{aligned} \\ \Rightarrow \lambda_3 \geq \phi^2/2$$

Now, since  $\lambda_3$  calculated here is for  $H$ ,

$$\lambda_3 \leq (\lambda_3)^G \text{ from an earlier claim}$$

$$\Rightarrow (\lambda_3)^G \geq \frac{\phi^2}{2}.$$

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