

**Problem 1** (20%). Prove that, for any vector  $v \in \mathbb{R}^n$ ,

$$\frac{|v|_1}{\sqrt{n}} \leq \|v\|_2 \leq |v|_1,$$

where  $|v|_1 := \sum_i |v_i|$  is the  $L_1$ -norm and  $\|v\|_2 := (\sum_i v_i^2)^{1/2}$  is the  $L_2$ -norm of  $v$ .

*Hint:* Use the Cauchy-Schwarz inequality, i.e.,  $|u \cdot v| \leq \|u\|_2 \|v\|_2$  for any  $u, v \in \mathbb{R}^n$ .

**Problem 2** (20%). Let  $A$  be a square symmetric matrix and  $\lambda$  be an eigenvalue of  $A$ . Prove that, for any  $k \in \mathbb{N}$ ,  $\lambda^k$  is an eigenvalue of  $A^k$ .

**Problem 3** (20%). Let  $G$  be an  $n$ -vertex  $d$ -regular bipartite graph and  $A$  be the normalized adjacency matrix of  $G$ . Prove that, there exists a vector  $v \in \mathbb{R}^n$  such that

$$Av = -v.$$

Generalize the construction to non-regular bipartite graphs, i.e., for any bipartite graph  $G'$  with column-normalized adjacency matrix  $A'$ , prove that  $A'$  has an eigenvalue  $-1$ .

*Note:*  $A'$  is also called the *random-walk* matrix of  $G'$ .

**Problem 4** (20%). Let  $G = (V, E)$  be a  $d$ -regular graph and  $P$  be a random walk of length  $t$  in  $G$ . Prove that, for any edge  $e \in E$  and any  $1 \leq i \leq t$ ,

$$\Pr [ \text{ } e \text{ is the } i^{\text{th}} \text{-edge of } P ] = \frac{1}{|E|}.$$

*Hint:* Prove by induction on  $i$ .

**Problem 5** (20%). Let  $G = (V, E)$  be an  $(n, d, \lambda)$ -expander and  $S \subseteq V$  be a vertex subset. Prove that,

$$\Pr_{(u,v) \in E} [ u, v \in S ] \leq \frac{|S|}{n} \left( \frac{|S|}{n} + \lambda \right),$$

i.e., for any  $(u, v) \in E$ , the probability that both  $u, v$  are in  $S$  is bounded by  $\frac{|S|}{n} \left( \frac{|S|}{n} + \lambda \right)$ .

*Hint:* Use the fact that  $|E(S, S)| = (d|S| - |E(S, T)|)/2$ . Apply the crossing lemma.

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*Hint:* Use the Cauchy-Schwarz inequality, i.e.,  $|u \cdot v| \leq \|u\|_2 \|v\|_2$  for any  $u, v \in \mathbb{R}^n$ .

we want to proof  $\|v\|_2 \leq |v|_1$  and  $\frac{|v|_1}{\sqrt{n}} \leq \|v\|_2$

①  $\|v\|_2 \leq |v|_1$

$$|v|_1 = \sum_{i=1}^n v_i \quad \& \quad \|v\|_2 = \left( \sum_{i=1}^n v_i^2 \right)^{\frac{1}{2}}$$

$$|v|_1^2 = \left( \sum_{i=1}^n v_i \right)^2 = \sum_{i=1}^n v_i^2 + 2 \cdot \sum_{1 \leq i < j \leq n} v_i v_j \geq \left( \left( \sum_{i=1}^n v_i^2 \right)^{\frac{1}{2}} \right)^2 = \sum_{i=1}^n v_i^2 = (\|v\|_2)^2$$

$$\Rightarrow \|v\|_2 \leq |v|_1 \#$$

②  $\frac{|v|_1}{\sqrt{n}} \leq \|v\|_2$

$$v = (v_1, v_2, v_3, \dots, v_n), \quad \ell = (1, 1, \dots, 1)$$

$$v \cdot \ell = \sum_{i=1}^n v_i \leq \|v\|_2 \|\ell\|_2 = \left( \sum_{i=1}^n v_i^2 \right)^{\frac{1}{2}} \cdot \sqrt{n} = \|v\|_2 \sqrt{n}$$

$$\Rightarrow |v|_1 \leq \|v\|_2 \sqrt{n} \equiv \frac{|v|_1}{\sqrt{n}} \leq \|v\|_2$$

by ① ② 可推得  $\frac{|v|_1}{\sqrt{n}} \leq \|v\|_2 \leq |v|_1 \#$

**Problem 2** (20%). Let  $A$  be a square symmetric matrix and  $\lambda$  be an eigenvalue of  $A$ . Prove that, for any  $k \in \mathbb{N}$ ,  $\lambda^k$  is an eigenvalue of  $A^k$ .

1. 根據題意,  $A^k v = \lambda^k v$ , for any  $k \in \mathbb{N}$

2. By induction,  $k=1$ , we have  $\lambda$  is eigenvalue of  $A$ ,  $Av = \lambda v$

3.  $k=2$ , 左右同乘  $A$ 。

$$A \cdot A v = A \cdot \lambda v = \lambda \cdot A v = \lambda \cdot \lambda v = \lambda^2 v$$

$$\Rightarrow A^2 v = \lambda^2 v$$

4. 以此類推, for  $k+1$

$$A^{k+1} v = A \cdot A^k v = A \cdot \lambda^k v = \lambda^{k+1} v \Rightarrow A^{k+1} v = \lambda^{k+1} v$$

5. 因得證,  $\lambda^k$  is an eigenvalue of  $A^k$   $\forall k \in \mathbb{N}$  #

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Generalize the construction to non-regular bipartite graphs, i.e., for any bipartite graph  $G'$  with column-normalized adjacency matrix  $A'$ , prove that  $A'$  has an eigenvalue  $-1$ .

Note:  $A'$  is also called the *random-walk* matrix of  $G'$ .

Let  $G = (L \cup R)$ ,  $L, R$  分別是 bipartite 的兩集合

$$v \in \mathbb{R}^n, v_i = \begin{cases} +1, & i \in L \\ -1, & i \in R \end{cases}$$

$$A_{ij} = \begin{cases} \frac{1}{d}, & \text{若 } (i, j) \in E \\ 0, & \text{otherwise} \end{cases}$$

$i \in L$ , 所有 neighbor 在  $R$ , 且  $v_j = -1$ ,  $v_i = +1$

$$(Av)_i = \sum_{j:(i,j) \in E} A_{ij} v_j = \sum_{j \in N(i)} \frac{1}{d} (-1) = \frac{-1}{d} \cdot |N(i)| = \frac{-1}{d} \cdot d = -1 = -v_i$$

$i \in R$ , 所有 neighbor 在  $L$ , 且  $v_j = +1$ ,  $v_i = -1$

$$(Av)_i = \sum_{j:(i,j) \in E} \frac{1}{d} (+1) = \frac{1}{d} \cdot d = 1 = -v_i$$

$$\Rightarrow (Av) = -v$$

**Problem 4** (20%). Let  $G = (V, E)$  be a  $d$ -regular graph and  $P$  be a random walk of length  $t$  in  $G$ . Prove that, for any edge  $e \in E$  and any  $1 \leq i \leq t$ ,

$$\Pr [ e \text{ is the } i^{\text{th}}\text{-edge of } P ] = \frac{1}{|E|}.$$

*Hint:* Prove by induction on  $i$ .

I.

設  $X_i$  為 random walk 在第  $i$  步落在的 vertex

1.  $\forall i (1 \leq i \leq t)$ , then  $X_i \sim \text{Uniform}(V)$

令  $i=1$ , 則  $X_1$  是起點。

2. 設  $X_i = v$  為目前的 vertex,  $X_{i+1} = w$  為下一步的 vertex

By induction, 任取  $i$ , if  $X_i$  是 uniform, 求走到下一步  $w$  的  $\Pr [X_{i+1} = w]$ 。

3. 則  $\Pr [X_{i+1} = w] = \sum_{v \in N(w)} \Pr [X_i = v] \cdot \frac{1}{d}$  ( $v \rightarrow w$  的概率)

$$= \sum_{v \in N(w)} \frac{1}{n} \cdot \frac{1}{d} = \frac{|N(w)|}{nd} = \frac{d}{nd} = \frac{1}{n}$$

4.  $X_{i+1}$  仍是 uniform。

II.

1. 設  $E_i = \{X_i, X_{i+1}\}$ ,  $e = \{u, v\}$

2. 則  $\Pr (E_i = e) = \Pr (X_i = u, X_{i+1} = v) + \Pr (X_i = v, X_{i+1} = u)$

$$= \frac{1}{n} \cdot \frac{1}{d} + \frac{1}{n} \cdot \frac{1}{d} = \frac{2}{nd} = \frac{1}{|E|}$$

3.  $|E| = nd/2 \quad \forall i$

4. 推得  $e \in E, 1 \leq i \leq t, \Pr [e \text{ is the } i^{\text{th}}\text{-edge}] = \frac{1}{|E|}$  #

**Problem 5** (20%). Let  $G = (V, E)$  be an  $(n, d, \lambda)$ -expander and  $S \subseteq V$  be a vertex subset. Prove that,

$$\Pr_{(u,v) \in E} [u, v \in S] \leq \frac{|S|}{n} \left( \frac{|S|}{n} + \lambda \right),$$

i.e., for any  $(u, v) \in E$ , the probability that both  $u, v$  are in  $S$  is bounded by  $\frac{|S|}{n} \left( \frac{|S|}{n} + \lambda \right)$ .

*Hint:* Use the fact that  $|E(S, S)| = (d|S| - |E(S, T)|)/2$ . Apply the crossing lemma.

1. 由題目知  $G = (V, E)$  and  $(n, d, \lambda)$ -expander, then  $|E| = \frac{nd}{2}$

2. we want to proof " $\Pr_{(u,v) \in E} [u, v \in S] \leq \frac{|S|}{n} \left( \frac{|S|}{n} + \lambda \right)$ "

3.  $x \in \{0, 1\}^n$ ,  $x_i = \begin{cases} 1, & i \in S \\ 0, & i \notin S \end{cases}$

4.  $x = \frac{|S|}{n} \cdot l + y$ ,  $l = (1, 1, \dots, 1)^T$ ,  $l^T y = 0$

5.  $x^T P_x = \left( \frac{|S|}{n} l + y \right)^T P \left( \frac{|S|}{n} l + y \right)$

$$= \frac{|S|^2}{n^2} l^T P l + 2 \frac{|S|}{n} l^T P y + y^T P y$$

$$\times l^T P y = (P l)^T y = l^T y = 0$$

$$6. y^T P y \leq \lambda \|y\|^2 = \lambda \left( |S| - \frac{|S|^2}{n} \right)$$

$$7. \text{推得 } x^T P_x \leq \frac{|S|^2}{n} + \lambda \left( |S| - \frac{|S|^2}{n} \right) = |S| \left( \frac{|S|}{n} + \lambda \right) - \lambda \frac{|S|^2}{n} \leq |S| \left( \frac{|S|}{n} + \lambda \right)$$

8.  $\Pr_{(u,v) \in E} [u, v \in S]$

$$= \frac{1}{n} x^T P_x \leq \frac{1}{n} \left[ |S| \left( \frac{|S|}{n} + \lambda \right) \right] = \frac{|S|}{n} \left( \frac{|S|}{n} + \lambda \right) \#$$