

(I) We know that the variational form of $\lambda_{\min}^{(M)} = \sup_{x: \|x\|=1} \{x^T M x\}$. : 10^{1.2}
 If $\forall i \in [n]: u(i) \geq 0$. Let $\exists j: u(j) \neq 0, u(i) = 0: (M_{ik} \geq 0, u(k) \geq 0)$

$$(Mu)_{(i)} = \mu u(i) \rightarrow \sum_{k=1}^n M_{ik} u(k) = \mu \cdot u(i) = 0 \quad \cdot \times$$

Since $\forall k: M_{ik} \geq 0, u(k) \geq 0$ & $\exists j: u(j) \neq 0$ the running sum would not have the chance to get zero which is a paradox. Thus:

if $\forall i \in [n]: u(i) \geq 0 \Rightarrow \exists j: u(j) \neq 0$ Q.E.D. ■

which suggests that: $\boxed{\forall i \in [n]: u(i) \geq 0 \leftarrow \forall i \in [n]: u(i) \geq 0}$

(II) Since $\mu_1 = u_1^T M u_1 \geq y^T M y$. Let $y_j = |u_1^{(j)}| \cdot \forall j$: Thus:

$$y^T M y = \sum_{i,j} y_i y_j M_{ij} = \sum_{i,j} |u_1^{(i)}| \cdot |u_1^{(j)}| M_{ij} \geq \sum_{i,j} u_1^{(i)} u_1^{(j)} M_{ij} = u_1^T M u_1 = \mu_1$$

which suggests $(y^T M y \leq \mu_1) \wedge (y^T M y \geq \mu_1) \rightarrow y^T M y = \mu_1$

thus: $y^T M y = u_1^T M u_1 = \mu_1$ & since $\mu_1 = \sup_{\|z\|=1} \{z^T M z\} = u_1^T M u_1$

therefore $y = |u_1|_+ = u_1$, which proves that: $\forall j: u_1(j) = |u_1(j)|$

$\rightarrow \boxed{\forall j: u_1(j) \geq 0}$ Q.E.D. ■ from (I) $\rightarrow \boxed{\forall j: u_1(j) \geq 0}$

(III) let $y(i) = |u_n(i)| \forall i \in [n]$. Now we have:

$$\mu_1 = \sup_{\|z\|=1} \{z^T M z\} \geq y^T M y = \sum_{i,j} |u_n^{(i)}| |u_n^{(j)}| M_{ij} \geq \left| \sum_{i,j} u_n^{(i)} u_n^{(j)} M_{ij} \right| = |\mu_n|$$

Triangle Inequality

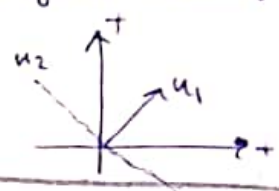
Thus: $\boxed{\mu_1 \geq |\mu_n| \geq -\mu_n}$ Q.E.D. ■

(IV) We know that for symmetric matrices $M = M^T$, if $\mu_i \neq \mu_j \rightarrow u_i \perp u_j$.
 Even if $\mu_i = \mu_j \rightarrow u_i \perp u_j$.

$u_1^T u_2 = 0 \rightarrow \sum_{i=1}^n u_1^{(i)} u_2^{(i)} = 0$. Now, since $\forall i: u_1(i) \geq 0$. Therefore, u_2 must have positive & negative elements:

$\boxed{\exists i, j \in [n]: (u_2(i) > 0) \wedge (u_2(j) < 0)}$ Q.E.D. ■

We can see this in other words as well. Since all elements of the first eigenvector are positive, and the fact that $u_2 \perp u_1$, u_2 must have negative elements as it's impossible for u_2 to have positive elements and be perpendicular to u_1 .



Ⓜ Again let $\forall i \in [n]: y(i) = |u_2(i)|$, then $\{M_{ij} \geq 0\}$:

$$\mu_2 = u_2^T M u_2 = \sum_{i,j} u_2^{(i)} u_2^{(j)} M_{ij} \leq \sum_{i,j} \underbrace{|u_2^{(i)}|}_{y(i)} \underbrace{|u_2^{(j)}|}_{y(j)} M_{ij} = y^T M y \leq u_1^T M u_1 = \mu_1.$$

Now since u_2 has negative elements as well we can deduce that:

$$u_2^T M u_2 < y^T M y \text{ which suggests } \mu_2 = u_2^T M u_2 < y^T M y \leq u_1^T M u_1 = \mu_1.$$

Thus: $\mu_2 < \mu_1$ Q.E.D. ■

Ⓜ G is connected and $\mu_1 = -\mu_n$, then we have: (let $\forall i: |u_1(i)| = y(i)$)

$$\mu_1 = -\mu_n = u_1^T M u_1 = -u_n^T M u_n \xrightarrow{\text{claim}} \sum_{i,j} M_{ij} \left(u_n^{(i)} u_n^{(j)} + |u_n^{(i)}| |u_n^{(j)}| \right) = 0 \quad (*)$$

Proof: $\rightarrow \mu_1 = u_1^T M u_1 \geq y^T M y \geq -u_n^T M u_n = -\mu_n$. So all the inequalities must be the equal case.

(*) \rightarrow Since $M_{ij} \geq 0$, there are two cases:

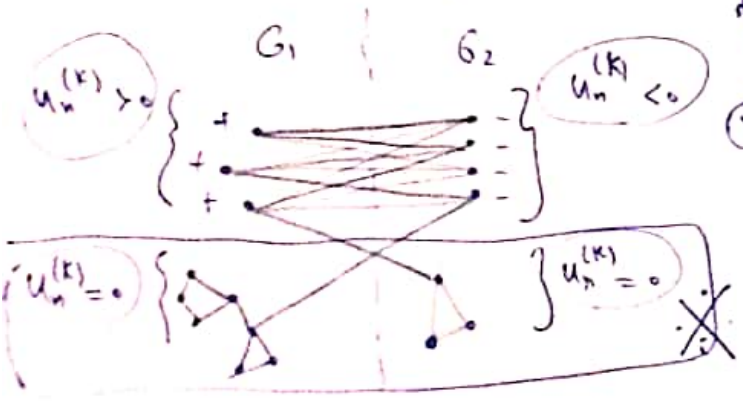
if $M_{ij} \neq 0$, then $u_n^{(i)} u_n^{(j)} + |u_n^{(i)}| |u_n^{(j)}| = 0$, otherwise; we don't know anything about $u_n^{(i)}, u_n^{(j)}$.

$$\rightarrow u_n^{(i)} u_n^{(j)} < 0$$

thus, if $\forall i,j: u_n^{(i)} u_n^{(j)} < 0$, then (i,j) are disconnected. otherwise, they could be connected or not. So the graph G, is bipartite. Since we can label each node k with $u_n^{(k)} > 0$ to subgraph G_1 and with $u_n^{(k)} < 0$ to the other subgraph G_2 .

$$(M_{ij} = 1) \Leftrightarrow u_n^{(i)} u_n^{(j)} < 0$$

Q.E.D. ■



Ⓜ $u_n(k) = 0$ does not happen since for $\mu_1 = -\mu_n$ we must have: $\forall i: u_1(i) = |u_n(i)|$ which proves $u_n(k) = 0$ does not happen.

① $K_n \rightarrow L = \begin{bmatrix} n-1 & -1 & -1 & \dots & -1 \\ -1 & n-1 & -1 & \dots & -1 \\ -1 & -1 & n-1 & \dots & -1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -1 & -1 & -1 & \dots & n-1 \end{bmatrix} = nI - \mathbb{1}_{n \times n}$: 20/2

$\det(\lambda I - L) = \det(-nI + \mathbb{1}_{n \times n} + \lambda I) = \det((\lambda - n)I + \mathbb{1}_{n \times n})$. Let λ_L be the set of eigenvalues of L , $-\mathbb{1}_{n \times n}$, respectively.

$$\lambda_L = \{ \lambda : \det(-L + \lambda I) = 0 \} = \{ \lambda : \det((\lambda - n)I - (-\mathbb{1}_{n \times n})) = 0 \} = \{ \lambda + n : \det(\lambda I - (-\mathbb{1}_{n \times n})) = 0 \}$$

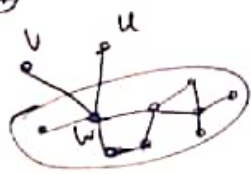
$$\rightarrow \lambda_L = n + \{ \lambda : \det(\lambda I - (-\mathbb{1}_{n \times n})) = 0 \} = n + \lambda_H.$$

However, the matrix $-\mathbb{1}_{n \times n}$ has $(n-1)$ zero eigenvalues since $\text{rank}(-\mathbb{1}_{n \times n}) = 1$.

Let $v = \begin{bmatrix} +1 \\ +1 \\ \vdots \\ +1 \end{bmatrix} \rightarrow (-\mathbb{1}_{n \times n})v = \begin{bmatrix} -n \\ -n \\ \vdots \\ -n \end{bmatrix} = -nv \rightarrow \lambda = n \rightarrow \lambda_H = \{ \underbrace{0, 0, \dots, 0}_{n-1}, n \}$

which suggest $\lambda_L = \{ \underbrace{n, n, \dots, n}_{n-1}, 0 \}$.

② we have nodes u, v with degree 1, that have a common neighbor w .



$$L = \begin{matrix} & \begin{matrix} u \\ v \\ w \end{matrix} \\ \begin{matrix} u \\ v \\ w \end{matrix} & \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 1 \\ -1 & -1 & n \end{bmatrix} \end{matrix}$$

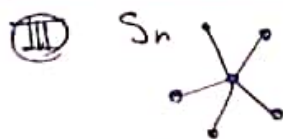
we can see that for the matrix

L , the vector: u -th v -th
 $h = [0, 0, 0, \dots, 1, 0, \dots, 0, 1, 0, \dots, 0]$

is an eigenvalue with $\lambda = 1$.

the other eigenvalues will be the basis of

$$\mathbb{R}^n \setminus \text{span}(h) \rightarrow \{ \psi_1, \dots, \psi_{n-1} \} \xrightarrow{\text{base}} \mathbb{R}^n \setminus \text{span}(h)$$



$$L = \begin{bmatrix} n-1 & -1 & -1 & \dots & -1 \\ -1 & 1 & 0 & \dots & 0 \\ -1 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -1 & 0 & 0 & \dots & 1 \end{bmatrix} = I + \overbrace{\begin{bmatrix} n-2 & -1 & \dots & -1 \\ -1 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ -1 & 0 & \dots & 0 \end{bmatrix}}^A$$

$\begin{cases} \lambda_L = 1 + \lambda_A \\ \text{Eigen}(L) = \text{Eigen}(A) \end{cases}$

since the $\text{rank}(A) = 1$, then A has $n-1$ zero eigenvalues. L has one zero eigenvalue,

which suggests $\lambda_A \ni \{-1\}$. $\text{Tr}\{A\} = \sum_{\lambda \in \lambda_L} \lambda = \underbrace{0 + 0 + \dots + 0}_{n-1} + (-1) + \lambda_{\max} = n-2$.

$\lambda_{\max} = (n-1) \rightarrow \lambda_A = \{ \underbrace{0, 0, \dots, 0}_{n-2}, -1, n-1 \} \rightarrow \lambda_L = \{ \underbrace{1, 1, \dots, 1}_{n-2}, 0, n \}$.

Now using the result of the previous section, we can let u, v as below, and have these eigenvectors:



$$u_2 = \begin{bmatrix} 1 \\ -1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, u_3 = \begin{bmatrix} 1 \\ 0 \\ -1 \\ \vdots \\ 0 \end{bmatrix}, \dots, u_{n-1} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \\ -1 \end{bmatrix}, u_n = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \\ 1 \end{bmatrix}$$

$$\lambda_2 = 1, \lambda_3 = 1, \dots, \lambda_{n-1} = 1, \lambda_n = 0$$

$$u_1 = \begin{bmatrix} \frac{n(n-1)}{n^2-n-1} \\ \frac{n^2}{n^2-n-1} \\ \vdots \\ \frac{n^2}{n^2-n-1} \end{bmatrix}$$



$$L = \begin{bmatrix} 2 & -1 & 0 & \dots & 0 \\ -1 & 2 & -1 & \dots & 0 \\ 0 & -1 & 2 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -1 & \dots & -1 & 2 \end{bmatrix}$$

size n

(iv)

$$\lambda_i \in \text{DFT}([2, -1, 0, \dots, 0]) \rightarrow \lambda_i = 2 + e^{-j\frac{2\pi}{n}i} + e^{-j\frac{2\pi}{n}i(n-1)} \quad \text{Q.E.D.}$$

$$\rightarrow \lambda_i = 2 + e^{-j\frac{2\pi}{n}i} + e^{j\frac{2\pi}{n}i} = 2 + 2\cos\left(\frac{2\pi}{n}i\right) = 2\left(1 - \cos\left(\frac{2\pi}{n}i\right)\right) \quad \forall i \in \{0, \dots, n-1\}$$

the set of eigenvectors are $\{u_k : u_k = \begin{bmatrix} e^{-j\frac{2\pi}{n}k(1-1)} \\ e^{-j\frac{2\pi}{n}k(2-1)} \\ \vdots \\ e^{-j\frac{2\pi}{n}k(n-1)} \end{bmatrix}\}$ since $e^{j\theta} = \cos\theta + j\sin\theta$

$$u_k = \text{Re}\{u_k\} + j\text{Im}\{u_k\}$$

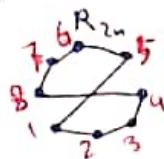
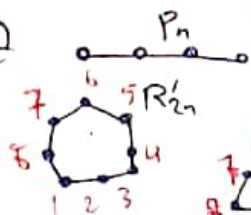
$$= \begin{bmatrix} \vdots \\ \cos\left(\frac{2\pi}{n}k(i-1)\right) \\ \vdots \end{bmatrix} + j \begin{bmatrix} \vdots \\ \sin\left(\frac{2\pi}{n}k(i-1)\right) \\ \vdots \end{bmatrix}$$

since $\lambda_i \in \mathbb{R} \rightarrow \text{Re}\{u_k\}, \text{Im}\{u_k\} \in \mathcal{E}_\lambda(\mathbb{C})$

so each of the vectors above are eigenvalues. ~~Furthermore~~ since set of

~~complex~~ thus $x_{k(i)} = \cos\left(\frac{2\pi}{n}ki\right), y_{k(i)} = \sin\left(\frac{2\pi}{n}ki\right)$ are the eigenvectors for the same λ . Q.E.D.

Ⓟ



$$\mathcal{L}(R_{2n}) = \begin{bmatrix} 2 & -1 & 0 & 0 & 0 & 0 & 0 & -1 \\ -1 & 2 & -1 & 0 & 0 & 0 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 2 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & -1 & 2 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & -1 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 & 2 & -1 \\ -1 & 0 & 0 & 0 & 0 & 0 & -1 & 2 \end{bmatrix}$$

$$\mathcal{L}(P_n) = \begin{bmatrix} 1 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -1 & 1 \end{bmatrix}$$

$$\mathcal{L}(R_{2n}) = \begin{bmatrix} 2 & -1 & 0 & 0 & -1 & 0 & 0 & 0 \\ -1 & 2 & -1 & 0 & 0 & 0 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 2 & 0 & 0 & 0 & -1 \\ -1 & 0 & 0 & 0 & 2 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & -1 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 & 2 & -1 \\ 0 & 0 & 0 & -1 & 0 & 0 & -1 & 2 \end{bmatrix}$$

$$= \mathcal{L}(P_n) \otimes \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + A \otimes \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$[I_n | I_n] \mathcal{L}(R_{2n}) \begin{bmatrix} I_n \\ I_n \end{bmatrix} = 2\mathcal{L}(P_n)$$

Q.E.D.

$$\begin{bmatrix} I_n & | & I_n \end{bmatrix} \left(\begin{array}{c|c} \mathcal{L}(P_n) - A & A \\ \hline A & \mathcal{L}(P_n) - A \end{array} \right) \begin{bmatrix} I_n \\ I_n \end{bmatrix} = \begin{bmatrix} I_n & | & I_n \end{bmatrix} \begin{bmatrix} \mathcal{L}(P_n) \\ \mathcal{L}(P_n) \end{bmatrix} = \mathcal{L}(P_n) + \mathcal{L}(P_n) = 2\mathcal{L}(P_n)$$

Let $\psi = \begin{pmatrix} \phi \\ \phi \end{pmatrix} = \phi \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ where $\phi \in \text{eigen}(\mathcal{L}(P_n)) : \mathcal{L}(P_n)\phi = \lambda\phi$:

$$\phi^T [I_n | I_n] \mathcal{L}(R_{2n}) \begin{bmatrix} I_n \\ I_n \end{bmatrix} \phi = \psi^T \mathcal{L} \psi = 2\phi^T \mathcal{L}(P_n) \phi = 2\lambda \|\phi\|^2 = \lambda \|\psi\|^2$$

$\rightarrow \psi^T \mathcal{L} \psi = \lambda \|\psi\|^2$ shows that:

Q.E.D.

$$\forall \lambda \in \Lambda_{P_n}, \phi \in \text{eigen}(P_n, \lambda) : (\psi^T \mathcal{L} \psi = \lambda \|\psi\|^2) \equiv \left\{ \lambda, \psi = \begin{pmatrix} \phi \\ \phi \end{pmatrix} \right\} \in \Lambda_{\mathcal{L}(R_{2n})}$$

(we could also do this $[I_n | I_n] \mathcal{L}(R_{2n}) \begin{bmatrix} I_n \\ I_n \end{bmatrix} \phi = \begin{bmatrix} I_n & | & I_n \end{bmatrix} \mathcal{L}(R_{2n}) \psi = \lambda \|\psi\|^2$ to make sure λ, ψ are (eigenvalue, eigenvector) pair.)

: 20th April

$$v_j(R_{2n}) = \begin{bmatrix} \omega^{0j} \\ \omega^{1j} \\ \omega^{2j} \\ \omega^{(2n-1)j} \end{bmatrix} = \begin{bmatrix} \phi_j \\ \omega^j \phi_j \end{bmatrix} = \begin{bmatrix} \phi_j \\ \phi_j \end{bmatrix} \text{ if } \omega^{nj} = 1 \rightarrow j \equiv 0$$

$\omega = e^{j \frac{2\pi}{n}}$

Let $\Phi = \begin{bmatrix} \omega \\ 1 \end{bmatrix} \rightarrow$ the eigenvectors of R_{2n} with even indices are of the form $\begin{pmatrix} \phi \\ \phi \end{pmatrix}$. For these eigenvectors $\lambda_k = 2(1 - \cos(\frac{2\pi}{n}k))$, $0 \leq k < n$

Furthermore: $v_j(R_{2n}) \in \text{Eigen}(R_{2n})$, $j \equiv 0$ then:

$$\begin{cases} \text{Re}\{v_j(R_{2n})\} \in \text{Eigen}(R_{2n}) \\ \text{Im}\{v_j(R_{2n})\} \in \text{Eigen}(R_{2n}) \end{cases}$$

P.E.D. ■

which suggests that for $\begin{cases} x_k(i) = \cos(\frac{2\pi}{n}ki) \\ y_k(i) = \sin(\frac{2\pi}{n}ki) \end{cases}$ we have $\begin{cases} \begin{pmatrix} x_k \\ y_k \end{pmatrix}, \begin{pmatrix} y_k \\ x_k \end{pmatrix} \\ \in \text{Eigen}(R_{2n}) \end{cases}$

Note that we only found n (eigenvector, eigenvalue) for $\mathcal{L}(R_{2n})$.

$$\forall j \equiv 0 \rightarrow \begin{cases} v_j = \begin{pmatrix} x_{j/2} \\ x_{j/2} \end{pmatrix} \\ \lambda_j = 2(1 - \cos(\frac{\pi j}{n})) \end{cases} \quad \begin{cases} v_j = \begin{pmatrix} y_{j/2} \\ y_{j/2} \end{pmatrix} \\ \lambda_j = 2(1 - \cos(\frac{\pi j}{n})) \end{cases} \quad \begin{cases} v_j = \begin{bmatrix} \omega^{0j} \\ \omega^{1j} \\ \omega^{2j} \\ \omega^{(2n-1)j} \end{bmatrix} \\ \lambda_j = 2(1 - \cos(\frac{\pi j}{n})) \end{cases}$$

① Suppose we have graphs $G(n, V_1, B_1)$ & $H(m, V_2, B_2)$. then we know: 30/12
 that for their Cartesian Product $S = G \times H$, we have $A_S = A_G \otimes I_m + I_n \otimes A_H$ and
 also $D_S = D_G \otimes I_m + I_n \otimes D_H$. therefore, we get:

$$L_S = D_S - A_S = (\underbrace{D_G - A_G}_{L_G}) \otimes I_m + I_n \otimes (\underbrace{D_H - A_H}_{L_H}) = L_G \otimes I_m + I_n \otimes L_H.$$

Now we want to show that $\beta_{ij}(a, b) = \psi_i(a) \phi_j(b)$ is an eigenvector associated with the eigenvalue $\alpha(i, j) = \lambda_i + \gamma_j$ where $1 \leq i \leq n, 1 \leq j \leq m$. We can easily see that $\forall a, b: \beta_{ij}(a, b) = \psi_i(a) \phi_j(b) \xrightarrow[\beta_{ij} \in \mathbb{R}]{\beta_{ij} = \psi_i \otimes \phi_j} \beta_{ij} = \psi_i \otimes \phi_j$. So we prove our claim now:

$$\begin{aligned} L_S \beta_{ij} &= (L_G \otimes I_m + I_n \otimes L_H)(\psi_i \otimes \phi_j) = (L_G \otimes I_m)(\psi_i \otimes \phi_j) + (I_n \otimes L_H)(\psi_i \otimes \phi_j) \\ &= (L_G \psi_i) \otimes (I_m \phi_j) + (I_n \psi_i) \otimes (L_H \phi_j) \quad \text{P.E.D.} \\ &= \lambda_i (\psi_i \otimes \phi_j) + \gamma_j (\psi_i \otimes \phi_j) = (\lambda_i + \gamma_j) (\psi_i \otimes \phi_j) = (\lambda_i + \gamma_j) \beta_{ij}. \end{aligned}$$

Now that $\forall i \in [n], j \in [m]: L_S \beta_{ij} = (\lambda_i + \gamma_j) \beta_{ij}$, we can see that the vector β_{ij} are eigenvectors of L_S with the eigenvalues $(\lambda_i + \gamma_j) \forall i, j$.

② First, we notice that: $(\hat{\alpha}_{ij} = A_H(i, j)), (e_{ij} = A_S(i, j))$

$$\{(\hat{\alpha}_{ij}=1) \wedge (\hat{\alpha}_{i'j'}=1)\} \xrightarrow{\beta_{ij} \in \mathbb{R}} \{e_{ij}=1\}, \{(\hat{\alpha}_{ij}=1) \wedge (\hat{\alpha}_{i'j'}=1)\} \leftrightarrow \{e_{ij}=-1\}$$

thus, we can see that if we decompose A_S into $A_S^{(+)}, A_S^{(-)}$ such that:

$$A_S = A_S^{(+)} - A_S^{(-)}, \text{ then we could see this equivalency: } \begin{pmatrix} \forall i, j: \\ A_S^{(+)}(i, j) \geq 0 \\ A_S^{(-)}(i, j) \geq 0 \end{pmatrix}$$

$$(\hat{\alpha}_{ij}=1) \wedge (\hat{\alpha}_{i'j'}=1) \equiv \{A_S^{(+)}(i, j)=1\} \Leftrightarrow \{e_{ij}=1\}$$

$$(\hat{\alpha}_{ij}=1) \wedge (\hat{\alpha}_{i'j'}=1) \equiv \{A_S^{(-)}(i, j)=1\} \Leftrightarrow \{e_{ij}=-1\}$$

thus, we can see A_H as below: Since $A_H(i, j) = \hat{\alpha}_{ij}$ is 1 whenever

$$A_H = \begin{bmatrix} \begin{matrix} 1 & 2 & \dots & n \\ \vdots & & & \end{matrix} & \begin{matrix} A_S^{(+)} & A_S^{(-)} \\ A_S^{(-)} & A_S^{(+)} \end{matrix} \end{bmatrix}$$

$A_S^{(+)}(i, j)$ is 1 and viceversa. As you can see, we can write A_H as block matrices of $A_S^{(+)}, A_S^{(-)}$.

Furthermore, we can obtain $A_S^{(+)}$, $A_S^{(-)}$ in terms of A_G, A_S :

$$A_S^{(+)} = \frac{1}{2} (A_S + A_G), \quad A_S^{(-)} = \frac{1}{2} (A_G - A_S). \quad (*)$$

We can also write A_H as below:

$$\begin{aligned} A_H &= \left[\begin{array}{c|c} A_S^{(+)} & A_S^{(-)} \\ \hline A_S^{(-)} & A_S^{(+)} \end{array} \right] = A_S^{(+)} \otimes \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + A_S^{(-)} \otimes \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \xrightarrow{(*)} \\ &= \frac{A_S}{2} \otimes \left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \right) + \frac{A_G}{2} \otimes \left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \right) \\ &= A_S \otimes \underbrace{\begin{bmatrix} 1/2 & -1/2 \\ -1/2 & 1/2 \end{bmatrix}}_{M_S} + A_G \otimes \underbrace{\begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}}_{M_G} = A_G \otimes M_G + A_S \otimes M_S \end{aligned}$$

$$M_G \rightarrow \lambda_G \in \{0, 1\} \rightarrow u_1 = (1, 1)^T, \quad u_2 = (1, -1)^T$$

$$M_S \rightarrow \lambda_S \in \{0, -1\} \rightarrow v_1 = (1, 1)^T, \quad v_2 = (1, -1)^T$$

Now let $\{\psi_i\}_{i=1}^n, \{\phi_i\}_{i=1}^n$ be the eigenvectors of A_S, A_G respectively. let:

(i) $\beta_{ij} = \psi_i \otimes u_j$ ~~(let $j=1$)~~ else let $j=2 \rightarrow u_j = (1, -1)^T$

$$\begin{aligned} A_H \beta_{ij} \Big|_{j=2} &= (A_S \otimes M_S)(\psi_i \otimes u_j) + (A_G \otimes M_G)(\psi_i \otimes u_j) \Big|_{j=2} \\ &= (A_S \psi_i) \otimes (M_S u_j) + (A_G \psi_i) \otimes \underbrace{(M_G u_j)}_0 \Big|_{j=2} \\ &= (A_S \psi_i) \otimes (M_S u_j) \Big|_{j=2} = (\lambda_i \psi_i) \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix} \end{aligned}$$

$$\left\{ \psi_i \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix} \in \text{Eig}(A_H) \right\}$$

So we got $A_H \beta_{i2} = A_H (\psi_i \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix}) = \lambda_i (\psi_i \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix}) \Rightarrow \lambda_i \in \Lambda_{A_H}$

(ii) on the contrary we have: let $\tilde{\beta}_{ij} = \phi_i \otimes u_j$ (let $j=1$), $u_j = (1, 1)^T$

$$\begin{aligned} A_H \tilde{\beta}_{ij} \Big|_{j=1} &= (A_S \otimes M_S)(\phi_i \otimes u_j) + (A_G \otimes M_G)(\phi_i \otimes u_j) \Big|_{j=1} \\ &= (A_S \phi_i) \otimes \underbrace{(M_S u_j)}_0 + (A_G \phi_i) \otimes (M_G u_j) \Big|_{j=1} \\ &= (A_G \phi_i) \otimes (M_G u_j) \Big|_{j=1} = (\gamma_i \phi_i) \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix} \end{aligned}$$

$$\left\{ \phi_i \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix} \in \text{Eig}(A_H) \right\}$$

which results in $A_H \tilde{\beta}_{i1} = A_H (\phi_i \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix}) = \gamma_i (\phi_i \otimes \begin{pmatrix} 1 \\ 1 \end{pmatrix}) \Rightarrow \gamma_i \in \Lambda_{A_H}$

Up until now, we got that $(\lambda_i, \psi_i \otimes (1)), (\gamma_i, \phi_i \otimes (1))$ are the (eigenvalues, eigenvectors) of A_{ii} . Since we have at most 2n eigenvalues for A_{ii} , and for each (λ_i, ψ_i) & (γ_i, ϕ_i) we can find exactly one (eigenvalue - eigenvector) pair for A_{ii} , we can observe that $\Lambda_{A_{ii}} = \{\lambda_i \in \text{Eigen}(A_{ii})\} \cup \{\gamma_i \in \text{Eigen}(A_{ii})\}$. which proves the theorem.

Q.E.D. ■

$$\begin{cases} \Lambda_{A_{ii}} = \{\lambda_i\}_{i=1}^n \cup \{\gamma_i\}_{i=1}^n \\ \text{Eigen}(A_{ii}) = \left\{ \psi_i \otimes (1) = \begin{pmatrix} \psi_i \\ 0 \end{pmatrix} \right\}_{i=1}^n \cup \left\{ \phi_i \otimes (1) = \begin{pmatrix} 0 \\ \phi_i \end{pmatrix} \right\}_{i=1}^n \end{cases}$$

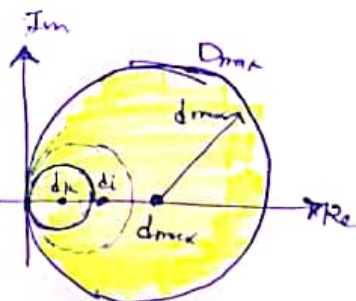
① Gershgorin theorem states that: every eigenvalue of matrix A lies on 40/2 union of disks in complex plane characterized by $D_i(a_{ii}, \sum_{j \neq i} |a_{ij}|)$

Proof: $\forall \lambda \in \Lambda_A \exists i = \arg \max_k \{ |x_k| \} \rightarrow Ax = \lambda x \rightarrow (Ax)_{ii} = \lambda x_{ii}$

$$|\lambda - a_{ii}| \leq \left| \sum_{j \neq i} a_{ij} x_j \right| \leq \sum_{j \neq i} |a_{ij}| \frac{|x_j|}{|x_i|} \leq \sum_{j \neq i} |a_{ij}| = R_i$$

Triangle inequality $|x_j| \leq |x_i|$

Now, for a Laplacian matrix, we know that $a_{ii} = d(i) = \sum_{j \neq i} |l_{ij}|$ the degree of i-th node.
 So these disks are centered at positive half of real axis.
 since $a_{ii} = d(i) \geq 0$. So each disk D_i is centered at $d(i)$ and with radius $d(i)$. So we can deduce that all of these disks D_i are covered completely in a disk with d_{\max} . So all eigenvalues $\lambda \in \Lambda_L$ lie in the circle $D(d_{\max}, d_{\max})$



$\forall \lambda \in \Lambda_L: |\lambda - d_{\max}| \leq d_{\max} \rightarrow 0 \leq \lambda \leq 2d_{\max} \rightarrow$ Now taking the biggest eigenvalue, implies that: $\lambda_1 \leq 2d_{\max}$ Q.E.D. ■

④ we know that $\lambda_1 = 0$, thus $\lambda_2 = \inf_{x \in \mathbb{R}^n \setminus \text{Span}(u_1)} \left\{ \frac{x^T L x}{x^T x} \right\} \leq \frac{y^T L y}{y^T y}$

now let $y = [0, \dots, 0, 1, 0, \dots, 0, 1, 0, \dots, 0]$, since $l_{ij} = l_{ji} = 0$ (i, j are disconnected)

$$y = (\hat{e}_i + \hat{e}_j) \rightarrow y^T L y = \hat{e}_i^T L \hat{e}_i + \hat{e}_j^T L \hat{e}_j + \hat{e}_i^T L \hat{e}_j + \hat{e}_j^T L \hat{e}_i \rightarrow y^T L y = 2$$

$$\rightarrow y^T L y = l_{ii} + l_{jj} + l_{ji} + l_{ij} = \sum_{k \neq i} -l_{ki} + \sum_{k \neq j} -l_{kj} = d(i) + d(j)$$

which results in:

■ Q.E.D. $\begin{cases} \lambda_2 \leq \frac{y^T L y}{y^T y} = \frac{d(i) + d(j)}{2} & i, j \text{ are disconnected} \\ \lambda_2 \leq \frac{d(i) + d(j)}{2} & i, j \text{ are connected} \end{cases}$

The same theorem in Part 4 will prove that $\lambda_{\max}^{(M)} \leq d_{\max}$:

let $\hat{D} = \bigcup_{i=1}^n \text{Disk}(0, d(i))$. Since $M_{ii} = 0 \forall i$, we can

say that all eigenvalues lie in \hat{D} since

$$\forall \lambda, \exists i \quad |\lambda - M_{ii}| \leq \sum_{j \neq i} M_{ij} = d(i) \rightarrow \forall \lambda \in \Lambda_M: |\lambda - 0| \leq d(i) \rightarrow \text{thus}$$

$$\lambda_2 \in \hat{D}. \text{ Now let } \lambda = \lambda_{\max} \rightarrow |\lambda_{\max}^{(M)}| \leq d_{\max} \rightarrow \boxed{\lambda_{\max}^{(M)} \leq d_{\max}} \quad \text{Q.E.D.}$$

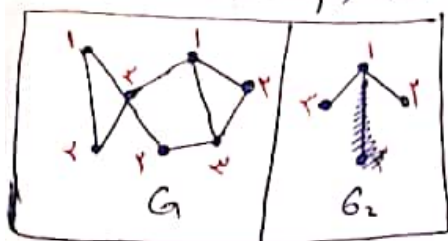
$$\hat{D} \subseteq \text{Disk}(0, d_{\max})$$

Since $\lambda_{\max}^{(M)} = \sup \frac{x^T M x}{x^T x}$, then $\forall y \in \mathbb{R}^n: \lambda_{\max}^{(M)} \geq \frac{y^T M y}{y^T y}$.

$$\text{let } y = \mathbf{1}_{n \times 1} \rightarrow \lambda_{\max}^{(M)} = \frac{1}{n} \sum_{i,j=1}^n M_{ij} = \frac{1}{n} \sum_{i=1}^n \left(\sum_{j \neq i} M_{ij} \right) = \frac{1}{n} \sum_{i=1}^n d(i) = d_{\text{avg}}$$

$$\text{which leads to } \boxed{\lambda_{\max}^{(M)} \geq d_{\text{avg}} = \frac{2e}{n}}$$

II Suppose we have a graph G_1 , and we eliminate nodes from G_1 such that $\chi(G_1) = \chi(G_2)$, and G_2 can't get smaller any further. ($G = G_1$)



then, we can see that $\forall i \in (n): d(i) \geq \chi(G_2) - 1$.
this is true since the number of colors is bounded by the degree of node i plus 1.

$$(*) \rightarrow \frac{1}{n} \sum_{i=1}^n d(i) = d_{\text{avg}} \geq \frac{1}{n} \sum_{i=1}^n (\chi(G_2) - 1) = \chi(G_2) - 1 \Rightarrow d_{\text{avg}} \geq \chi(G_2) - 1$$

$$\begin{aligned} \text{Now Since } \left\{ \begin{array}{l} \chi(G_1) = \chi(G_2) \\ \text{From II} \end{array} \right. & \rightarrow d_{\text{avg}} \leq \lambda_{\max}^{(M_1)} \rightarrow \boxed{\lambda_{\max}^{(M_1)} \geq \lambda_{\max}^{(M_2)} \geq d_{\text{avg}} \geq \chi(G_2) - 1} \\ \text{From Interlacing theorem} & \rightarrow \lambda_{\max}^{(M_1)} \geq \lambda_{\max}^{(M_2)} \rightarrow \boxed{\lambda_{\max}^{(M_1)} \geq \chi(G_1) - 1} \end{aligned}$$

$$\text{which proves: } \boxed{\chi(G_1) \leq \left\lceil \lambda_{\max}^{(G)} \right\rceil + 1} \quad \text{Q.E.D.}$$

(III) let us assume that the graph G , is k -colorable, then 50/2/21

we can write A as:

$$A = \begin{bmatrix} \phi_{1,1} & \phi_{1,2} & \dots & \phi_{1,n} \\ \phi_{2,1} & \phi_{2,2} & \dots & \phi_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ \phi_{n,1} & \phi_{n,2} & \dots & \phi_{n,n} \end{bmatrix}$$

from the last assignment we know that: (ob)

$$(k-1) \lambda_{\min}(A) + \lambda_{\max}(A) \leq \sum_{i=1}^n \lambda_{\max}(M_{ii})$$

Since $\forall i: \lambda_{\max}(M_{ii}) = \lambda_{\max}(\phi_{ii}) = 0 \rightarrow$ from (*) we get:

$$(*) \quad k \geq -\frac{\lambda_{\max}(A)}{\lambda_{\min}(A)} + 1. \text{ Since this } (*) \text{ equation holds for all } k \geq \chi(G),$$

getting min of k will result in $k \geq \chi(G) \geq -\frac{\lambda_{\max}(A)}{\lambda_{\min}(A)} + 1$ Q.E.D. ■