

Homework 7

● Graded

Student

Jaiden Rain Atterbury

Total Points

39 / 40 pts

Question 1

[Question 1](#)

10 / 10 pts

✓ - 0 pts correct

Question 2

[Question 2](#)

9 / 10 pts

✗ - 1 pt some part of answer (usually (a) or (d)) treats derivative as a map $\mathbb{R}_{\leq d} \rightarrow \mathbb{R}_{\leq d}$

Question 3

[Question 3](#)

10 / 10 pts

✓ - 0 pts correct

Question 4

[Question 4](#)

10 / 10 pts

✓ - 0 pts Complete

Question assigned to the following page: [1](#)

Jaeden Atterbury

05/17/2022

Homework #7

(1) Graph Connectivity

$$(a) D_G = \begin{bmatrix} 1 & & & & & & \\ & 1 & & & & & \\ & & 2 & & & & \\ & & & 2 & & & \\ & & & & 1 & & \\ & & & & & 1 & \\ & & & & & & 1 \end{bmatrix}$$

$$A_G = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$L_G = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & -1 & -1 & 0 & 0 \\ 0 & 0 & -1 & 2 & -1 & 0 & 0 \\ 0 & 0 & -1 & -1 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & -1 & 1 \end{bmatrix} = D_G - A_G$$

$$(b) x^T L_H x = [x_1 \ x_2 \ x_3 \ x_4 \ x_5 \ x_6 \ x_7] L_H$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \end{bmatrix}$$

$$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} +1 & -1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\begin{bmatrix} x_3 & x_4 & x_5 \end{bmatrix} \begin{bmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{bmatrix} \begin{bmatrix} x_3 \\ x_4 \\ x_5 \end{bmatrix}$$

$$\begin{bmatrix} x_6 & x_7 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} x_6 \\ x_7 \end{bmatrix}$$

$$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} x_1 & -x_2 \\ -x_1 & +x_2 \end{bmatrix}$$

$$x_1^2 - 2x_1x_2 + x_2^2$$

$$\begin{bmatrix} x_3 & x_4 & x_5 \end{bmatrix} \begin{bmatrix} 2x_3 - x_4 - x_5 \\ -x_3 + 2x_4 - x_5 \\ -x_3 - x_4 + 2x_5 \end{bmatrix}$$

$$\begin{bmatrix} x_6 & x_7 \end{bmatrix} \begin{bmatrix} x_6 - x_7 \\ -x_6 + x_7 \end{bmatrix}$$

$$(x_1 - x_2)^2$$

$$2x_3^2 + 2x_4^2 + 2x_5^2 - 2x_3x_4 - 2x_3x_5 - 2x_4x_5$$

$$x_6^2 - 2x_6x_7 + x_7^2$$

$$(x_3 - x_4)^2 + (x_3 - x_5)^2 + (x_4 - x_5)^2$$

$$(x_6 - x_7)^2$$

$$\therefore x^T L_H x =$$

$$\boxed{(x_1 - x_2)^2 + (x_3 - x_4)^2 + (x_3 - x_5)^2 + (x_4 - x_5)^2 + (x_6 - x_7)^2}$$

Question assigned to the following page: [1](#)

(c) From a theorem in the lecture notes (lesson 17), if \mathbf{L} has K connected components $\lambda_n = 0$ has multiplicity K in \mathbf{L}_G . Since \mathbf{L}_G is symmetric $\text{GM}(0) = \text{AM}(0)$.

$\therefore \mathbf{L}_G$ has 3 connected components so

$$\mathbf{E}_0 = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \right\}$$

(d) Based on the above, the first eigenvalue ~~that is~~ going to be positive is eigenvalue λ_4 . This comes from theorem 6.3.6 which states the eigenvalues of \mathbf{L}_G are $0 = \lambda_1 \leq \lambda_2 \leq \lambda_3 \leq \dots \leq \lambda_K$, since $\lambda_1, \lambda_2, \lambda_3 = 0$ represent each connected part of G , λ_4 will be the first non-zero (positive) eigenvalue since it doesn't correspond to a new connected piece and must be $\geq \lambda_3$, but since $\lambda_4 \neq 0$ it shows that $\lambda_4 > \lambda_3 = 0$.

(e) The relationship between the $\text{AM}(0)$ and the number of connected components is simple: If a graph G has K different connected components then $\text{AM}(0) = K$. This means that for each connected piece of a graph there is an eigenvalue of 0 that corresponds to it.

Question assigned to the following page: [2](#)

(2) Differentiation is a linear transformation

$$f(x) = a_d x^d + a_{d-1} x^{d-1} + \dots + a_1 x + a_0$$

$$\vec{f} = [a_0, a_1, \dots, a_{d-1}, a_d]^T$$

(a) $f'(x) = d a_d x^{d-1} + (d-1) a_{d-1} x^{d-2} + \dots + a_1 + 0$

$$\vec{f}' = [0, a_1, \dots, (d-1)a_{d-1}, da_d]^T$$

(b) $D_d : \mathbb{R}[x]_{\leq d} \rightarrow \mathbb{R}[x]_{\leq d-1}, f(x) \mapsto f'(x)$

1. $D_d(kf(x)) = k D_d(f(x))$?

$$(kf(x))' = \frac{d}{dx}(kf(x)) = k \frac{d}{dx}(f(x)) = kf'(x) \quad \checkmark \text{ yes from calculus.}$$

2. $D_d(f(x) + g(x)) = D_d(f(x)) + D_d(g(x))$

$$(f(x) + g(x))' = \frac{d}{dx}(f(x) + g(x)) = \frac{d}{dx}(f(x)) + \frac{d}{dx}(g(x)) = f'(x) + g'(x) \quad \checkmark \text{ yes from calculus.}$$

$\therefore D_d$ is a linear transformation.

(c) monomial basis of $\mathbb{R}[x]_{\leq d}$

$$B_d = \{1, x, x^2, \dots, x^d\}$$

$$D_d(1) = \frac{d}{dx}(1) = 0 \rightarrow \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$D_d(x) = \frac{d}{dx}(x) = 1 \rightarrow \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

$$D_d(x^2) = \frac{d}{dx}(x^2) = 2x \rightarrow \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 2 \end{bmatrix}$$

$$D_d(x^d) = \frac{d}{dx}(x^d) = dx^{d-1} \rightarrow \begin{bmatrix} 0 \\ 0 \\ \vdots \\ d \end{bmatrix}$$

Question assigned to the following page: [2](#)

$$(d) \mathbb{R}_n^{d+1} \rightarrow \mathbb{R}_m^d$$

$$M = \begin{bmatrix} 0 & 1 & 0 & 0 & \cdots & 0 \\ 0 & 0 & 2 & 0 & \cdots & 0 \\ 0 & 0 & 0 & 3 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & d \end{bmatrix}$$

$$(e) M_5 = \begin{bmatrix} 6 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 4 & 0 \\ 0 & 0 & 0 & 0 & 0 & 5 \end{bmatrix} \quad \mathbb{R}^6 \rightarrow \mathbb{R}^5 \quad 5 \times 6$$

$$(f) S(x) = 5x^5 - 19x^3 + 24x - 3 \in \mathbb{R}[x]_5$$

$$S'(x) = 25x^4 - 57x^2 + 24 \in \mathbb{R}[x]_4$$

$$5 \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 4 & 0 \\ 0 & 0 & 0 & 0 & 0 & 5 \end{bmatrix} \begin{bmatrix} -3 \\ 24 \\ 0 \\ -19 \\ 0 \\ 5 \end{bmatrix} = \begin{bmatrix} 24 \\ 0 \\ 0 \\ -57 \\ 0 \\ 25 \end{bmatrix}$$

$$\mathbb{R}^6 \rightarrow \mathbb{R}^5$$

$$\mathbb{R}[x]_{\leq 5} \rightarrow \mathbb{R}[x]_{\leq 4}$$

Question assigned to the following page: [3](#)

(3) Polynomial Interpolation

(a) $g(-2) = -3 \quad g(0) = 1 \quad g(1) = 0 \quad g(3) = 22$

$$\begin{bmatrix} 1 & -2 & 4 & -8 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 3 & 9 & 27 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -3 \\ 1 \\ 0 \\ 22 \end{bmatrix}$$

$$\begin{array}{c|ccccc} 1 & -2 & 4 & -8 & -3 \\ 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 3 & 9 & 27 & 22 \end{array} \xrightarrow{\substack{-R_1+R_2+R_3 \\ -R_1+R_3+R_4}} \begin{array}{c|ccccc} 1 & -2 & 4 & -8 & -3 \\ 0 & 2 & -4 & 8 & 4 \\ 0 & 3 & -3 & 9 & 3 \\ 0 & 5 & 5 & 35 & 25 \end{array} \xrightarrow{\substack{\frac{3}{2}R_2+R_3+R_4 \\ -\frac{5}{2}R_2+R_4+R_3}} \begin{array}{c|ccccc} 1 & -2 & 4 & -8 & -3 \\ 0 & 2 & -4 & 8 & 4 \\ 0 & 0 & 3 & -3 & -3 \\ 0 & 0 & 15 & 15 & 15 \end{array}$$

$$\sim \begin{array}{c|ccccc} 1 & -2 & 4 & -8 & -3 \\ 0 & 2 & -4 & 8 & 4 \\ 0 & 0 & 3 & -3 & -3 \\ 0 & 0 & 0 & 30 & 30 \end{array} \begin{array}{l} x_4 = 1 \\ x_3 = 0 \\ x_2 = -2 \\ x_1 = 1 \end{array}$$

$\therefore g(x) = x^3 - 2x + 1$

(b) $M = \begin{bmatrix} 1 & t_1 & t_1^2 & \dots & t_1^d \\ 1 & t_2 & t_2^2 & \dots & t_2^d \\ 1 & t_3 & t_3^2 & \dots & t_3^d \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & t_m & t_m^2 & \dots & t_m^d \end{bmatrix} \quad f = \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_d \end{bmatrix} \quad g = \begin{bmatrix} b_0 \\ b_1 \\ b_2 \\ \vdots \\ b_d \end{bmatrix}$

$$S(x) = a_d x^d + \dots + a_2 x^2 + a_1 x + a_0$$

$$g(x) = b_d x^d + \dots + b_2 x^2 + b_1 x + b_0$$

(i) $f(x), g(x) \in \mathbb{R}[x]_{\leq d}$

$$\text{eval}(S(x)) = Mf$$

$$\text{eval}(g(x)) = Mg$$

$$\text{eval}(S(x) + g(x)) = M(f + g) = Mf + Mg = \text{eval}(S(x)) + \text{eval}(g(x)) \checkmark$$

Question assigned to the following page: [3](#)

(ii) $S(x) \in \mathbb{R}[x] \leq d$

$\gamma \in \mathbb{R}$

$$\text{eval}(S(x)) = M_S$$

$$\text{eval}(\gamma S(x)) = M(\gamma S) = \gamma(M_S) = \gamma \text{eval}(S(x)) \checkmark$$

∴ eval is a linear transformation.

(c) 3×3 Vandermonde with t_1, t_2, t_3

$$A = \begin{bmatrix} 1 & t_1 & t_1^2 \\ 1 & t_2 & t_2^2 \\ 1 & t_3 & t_3^2 \end{bmatrix}$$

$$\det(A) = 1 \cdot \det \begin{bmatrix} t_2 & t_2^2 \\ t_3 & t_3^2 \end{bmatrix} - 1 \cdot \det \begin{bmatrix} t_1 & t_1^2 \\ t_3 & t_3^2 \end{bmatrix} + 1 \cdot \det \begin{bmatrix} t_1 & t_1^2 \\ t_2 & t_2^2 \end{bmatrix}$$

$$t_3^2 t_2 - t_2^2 t_3 - (t_3^2 t_1 - t_1^2 t_3) + t_2^2 t_1 - t_1^2 t_2 =$$

$$t_3^2 t_2 - t_2^2 t_3 - t_3^2 t_1 + t_1^2 t_3 + t_2^2 t_1 - t_1^2 t_2 =$$

$$(t_3 - t_2)(t_3 - t_1)(t_2 - t_1) =$$

$$\boxed{(-1)^3 (t_1 - t_2)(t_1 - t_3)(t_2 - t_3)}$$

Question assigned to the following page: [4](#)

(4) Spectral Clustering

Q1 Since we are now minimizing over a larger set the minimum value of Q will become smaller than when we were minimizing over all C_A 's, therefore:

$$\mu = \min_{\substack{x \in \mathbb{R}^n \\ x \neq \mathbf{0}}} Q(x) \leq \min_{A \in \mathcal{V}} Q(C_A) \leq \gamma_G$$

thus $\mu \leq \gamma_G$

Q2 $Q(x) = Q(x + t\mathbf{1})$? $\lambda_1 = 0$ with eigenvector $\mathbf{1}$

$$\begin{aligned} Q(x + t\mathbf{1}) &= (x + t\mathbf{1})^T L_G (x + t\mathbf{1}) \rightarrow x^T L_G x + \underbrace{t x^T L_G \mathbf{1}}_0 + \underbrace{t \mathbf{1}^T L_G \mathbf{1}}_0 + t^2 \mathbf{1}^T L_G \mathbf{1} = x^T L_G x \\ &= (x + t\mathbf{1})^T L_K n (x + t\mathbf{1}) \rightarrow x^T L_K n x + \underbrace{t x^T L_K n \mathbf{1}}_0 + \underbrace{t \mathbf{1}^T L_K n x}_0 + \underbrace{t^2 \mathbf{1}^T L_K n \mathbf{1}}_0 = x^T L_K n x \\ &= \frac{x^T L_G x}{x^T L_K n x} = Q(x) \checkmark \end{aligned}$$

Q3 Let $V = \text{span}\{\mathbf{1}\}$

$$x \in \mathbb{R}^n \rightarrow x = t\mathbf{1} + e$$

$$\begin{array}{c} \xrightarrow{\text{proj}_{V^\perp}} e \in V^\perp / \text{span}\{\mathbf{1}\}^T \\ \downarrow \\ \in \text{span}(e) \end{array}$$

$$Q(x) = Q(e + t\mathbf{1})$$

$$Q(x) = Q(e)$$

$$\therefore x \in \text{span}\{\mathbf{1}\}^T$$

$$x \perp \mathbf{1} \rightarrow \mathbf{1}^T x = 0$$

$$\text{thus } \mu = \min_{\substack{x \in \mathbb{R}^n \\ x \neq \mathbf{0} \\ x \perp \mathbf{1}}} Q(x)$$

$$\|x\|_F^2 = \sqrt{\mathbf{1}^T x^T x} = \sqrt{\mathbf{1}^T x^T L_K n x}$$

Question assigned to the following page: [4](#)

$$Q_4 \text{ (i)} L_G = D_G - A_G \rightarrow L_{K_n} = D_{K_n} - A_{K_n}$$

If K_n is a complete graph of n components, the degree of vertex i , d_i , is $n-1$ for every vertex i :

$$D_{K_n} = \begin{bmatrix} n-1 & 0 & 0 & \cdots & 0 \\ 0 & n-1 & 0 & \cdots & 0 \\ 0 & 0 & n-1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & n-1 \end{bmatrix}$$

and since K_n is complete the adjacency matrix will be all ones except on the diagonal:

$$A_{K_n} = \begin{bmatrix} 0 & 1 & 1 & \cdots & 1 \\ 1 & 0 & 1 & \cdots & 1 \\ 1 & 1 & 0 & \cdots & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & 1 & \cdots & 0 \end{bmatrix}$$

and thus the Laplacian matrix of K_n is:

$$L_{K_n} = \begin{bmatrix} n-1 & -1 & -1 & \cdots & -1 \\ -1 & n-1 & -1 & \cdots & -1 \\ -1 & -1 & n-1 & \cdots & -1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -1 & -1 & -1 & \cdots & -1 \end{bmatrix}$$

$$nI_n = \begin{bmatrix} n & 0 & 0 & \cdots & 0 \\ 0 & n & 0 & \cdots & 0 \\ 0 & 0 & n & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 0 \end{bmatrix}$$

$$J_n = \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & 1 & 1 & \cdots & 1 \\ 1 & 1 & 1 & \cdots & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & 1 & \cdots & 1 \end{bmatrix}$$

$$nI_n - J_n = \begin{bmatrix} n-1 & -1 & -1 & \cdots & -1 \\ -1 & n-1 & -1 & \cdots & -1 \\ -1 & -1 & n-1 & \cdots & -1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -1 & -1 & -1 & \cdots & n-1 \end{bmatrix} = L_{K_n} \checkmark$$

Question assigned to the following page: [4](#)

$$(ii) \quad \mathbf{J}^T = [1 \ 1 \ 1 \ \dots \ 1] \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix}$$

$$\mathbf{J}^T \mathbf{x} = x_1 + x_2 + x_3 + \dots + x_n = 0$$

$$\mathbf{J}_n = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & 1 & 1 & \dots & 1 \\ 1 & 1 & 1 & \dots & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} + x_2 \begin{bmatrix} 1 \\ 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} + x_3 \begin{bmatrix} 1 \\ 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} + \dots + x_n \begin{bmatrix} 1 \\ 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} x_1 + x_2 + x_3 + \dots + x_n \\ x_1 + x_2 + x_3 + \dots + x_n \\ x_1 + x_2 + x_3 + \dots + x_n \\ \vdots \\ x_1 + x_2 + x_3 + \dots + x_n \end{bmatrix} = \vec{0}$$

$$(iii) \quad Q(\mathbf{x}) \text{ denominator} = \mathbf{x}^T \mathbf{L}_{K_n} \mathbf{x} \rightarrow \mathbf{x}^T (\mathbf{n} \mathbf{I}_n - \mathbf{J}_n) \mathbf{x} = \mathbf{x}^T (\mathbf{n} \mathbf{I}_n) \mathbf{x} - \mathbf{x}^T (\mathbf{J}_n) \mathbf{x} = \mathbf{n} \mathbf{x}^T \mathbf{I}_n \mathbf{x}$$

$$= \mathbf{n} \mathbf{x}^T \mathbf{x}$$

$$= \mathbf{n} \|\mathbf{x}\|^2$$

$$(iv) \quad \text{for any matrix } M \quad (\alpha \mathbf{x})^T M (\alpha \mathbf{x}) = \alpha^2 \mathbf{x}^T M \mathbf{x}$$

$$\text{thus } Q(\alpha \mathbf{x}) = \mathbf{n} \cdot \frac{(\alpha \mathbf{x})^T \mathbf{L}_G (\alpha \mathbf{x})}{(\alpha \mathbf{x})^T \mathbf{L}_{K_n} (\alpha \mathbf{x})} = \mathbf{n} \cdot \frac{\alpha^2 \mathbf{x}^T \mathbf{L}_G \mathbf{x}}{\alpha^2 \mathbf{x}^T \mathbf{L}_{K_n} \mathbf{x}} = Q(\mathbf{x})$$

$\therefore Q(\mathbf{x}) = Q(\alpha \mathbf{x}) \text{ for any non-zero } \alpha \in \mathbb{R}.$

(v) Since $Q(\mathbf{x}) = Q(\alpha \mathbf{x})$, Q is determined by direction and not length,
thus we can take \mathbf{x} as a unit vector.

$$\hookrightarrow \mathbf{n} \|\mathbf{x}\|^2 \text{ when } \|\mathbf{x}\| = 1 \text{ is } \mathbf{n}$$

$$\therefore \mathbf{x}^T \mathbf{L}_{K_n} \mathbf{x} = \mathbf{n}$$

$$\therefore Q(\mathbf{x}) = \frac{\mathbf{n}}{\mathbf{n}} \cdot \mathbf{x}^T \mathbf{L}_G \mathbf{x} = \mathbf{x}^T \mathbf{L}_G \mathbf{x}$$

$$\text{thus } \lambda_{\mathbf{n}} = \min_{\substack{\mathbf{x} \in \mathbb{R}^n \\ \mathbf{x} \neq \mathbf{0} \\ \|\mathbf{x}\|=1 \\ \mathbf{x} \perp \mathbf{J}}} \mathbf{x}^T \mathbf{L}_G \mathbf{x}$$

Question assigned to the following page: [4](#)

Q5a) L_G is a connected graph with eigenvalues:

$$0 = \lambda_1 < \lambda_2 \leq \lambda_3 \leq \dots \leq \lambda_n$$

with a set of corresponding orthonormal eigenvectors

$$\{v_1, v_2, v_3, \dots, v_n\}$$

$\vec{x} \in \text{span}\{I\}^\perp$ since $\vec{x} \perp I$

$\dim(\text{span}(I)^\perp) = n-1$ since $\dim(\text{span}(I)) = 1$ and by

proposition 4.2.6 $\dim(V) + \dim(V^\perp) = n$

since $\dim(V) = 1$

$$\dim(V^\perp) = n - \dim(V) = n-1$$

The set $\{v_2, \dots, v_n\} \subseteq \text{span}\{I\}^\perp$

since $v_i \neq I$ and eigenvectors of symmetric matrices are mutually orthogonal by proposition 5.1.6. Furthermore,

the dimension of $\{v_2, \dots, v_n\}$ is $n-1$ so

$$\{v_2, \dots, v_n\} = \text{span}\{I\}^\perp$$

since $\vec{x} \in \text{span}\{I\}^\perp$ it can be written as a linear combination of the vectors in the set:

$$\therefore \vec{x} = \alpha_2 v_2 + \dots + \alpha_n v_n$$

(ii) $x = \alpha_2 v_2 + \alpha_3 v_3 + \dots + \alpha_n v_n = \sum_{i=2}^n \alpha_i v_i$

since $\|x\| = 1$:

$$\begin{aligned} \|x\|^2 &= \|\alpha_2 v_2 + \alpha_3 v_3 + \dots + \alpha_n v_n\|^2 \\ &= \alpha_2^2 + 0 + \dots + 0 \end{aligned}$$

$$\begin{aligned} 0 + \alpha_3^2 + \dots + 0 &\rightarrow \text{since } v_i^T v_i = 1 \quad (\|v_i\| = 1 \text{ since } \{v_1, \dots, v_n\} \text{ is an orthonormal set}) \\ 0 + 0 + \dots + \alpha_n^2 &\quad v_i^T v_j = 0 \text{ if } i \neq j \end{aligned}$$

$$= \alpha_2^2 + \alpha_3^2 + \dots + \alpha_n^2 = \sum_{i=2}^n \alpha_i^2 = 1$$

Question assigned to the following page: [4](#)

$$(iii) (\alpha_2 v_2 + \alpha_3 v_3 + \dots + \alpha_n v_n)^T L_G (\alpha_2 v_2 + \alpha_3 v_3 + \dots + \alpha_n v_n)$$

$$= \alpha_2 v_2^T L_G \alpha_2 v_2 + \alpha_2 v_2^T L_G \alpha_3 v_3 + \dots + \alpha_2 v_2^T L_G \alpha_n v_n$$

$$= \alpha_2^2 \lambda_2 \underbrace{v_2^T v_2}_0 + \alpha_2 \alpha_3 \lambda_3 \underbrace{v_2^T v_3}_0 + \dots + \alpha_2 \alpha_n \lambda_n \underbrace{v_2^T v_n}_0$$

$$= \alpha_2^2 \lambda_2 + 0 + \dots + 0$$

repeating this calculation for $\alpha_3 v_3$ and $\alpha_n v_n$ we get

$$\alpha_2^2 \lambda_2 + \alpha_3^2 \lambda_3 + \dots + \alpha_n^2 \lambda_n \checkmark$$

(iv) since $\lambda_2 \leq \lambda_3 \leq \dots \leq \lambda_n$

$$\text{and } x^T L_G x = \alpha_2^2 \lambda_2 + \alpha_3^2 \lambda_3 + \dots + \alpha_n^2 \lambda_n$$

we can say:

$$\begin{aligned} \alpha_2^2 \lambda_2 + \alpha_3^2 \lambda_3 + \dots + \alpha_n^2 \lambda_n &\geq \alpha_2^2 \lambda_2 + \alpha_3^2 \lambda_2 + \dots + \alpha_n^2 \lambda_2 \\ &\geq (\sum_{i=2}^n (\alpha_i^2)) \lambda_2 \end{aligned}$$

$$\therefore \min_{\substack{x \in \mathbb{R}^n \\ \|x\|=1 \\ x \neq 0}} x^T L_G x = \mu \geq \lambda_2$$

(v) $\|v_2\|=1$

$$L_G v_2 = \lambda_2 v_2$$

$$\therefore v_2^T L_G v_2 = v_2^T \lambda_2 v_2 = \lambda_2 \|v_2\|^2 = \lambda_2$$

$$\text{thus } \mu = \min_{\substack{x \in \mathbb{R}^n \\ \|x\|=1 \\ x \neq 0}} x^T L_G x \leq \lambda_2$$

(vi) since $\lambda_2 \leq \mu \leq \lambda_2$

$$\mu = \lambda_2$$