

# Question 4

Sunday, December 3, 2023

4:35 PM

4. Let  $A \in M_n(\mathbb{R})$  be a symmetric matrix, with eigenvalues  $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_p$ . Show that

$$A = Q \Lambda Q^T = \sum \lambda_i \ell_i \ell_i^T$$

$$\max_{\|\vec{x}\| \neq 0} \frac{\vec{x}^T A \vec{x}}{\vec{x}^T \vec{x}} = \lambda_1$$

$$\min_{\|\vec{x}\| \neq 0} \frac{\vec{x}^T A \vec{x}}{\vec{x}^T \vec{x}} = \lambda_p$$

Comment at what vectors  $x$  the max and min values are attained.

$$Q(x) = \vec{x}^T A \vec{x}, \text{ Since } A \text{ symmetric} \Rightarrow A = Q \Lambda Q^T$$

$$\Rightarrow \max \frac{\vec{x}^T (Q \Lambda Q^T) \vec{x}}{\vec{x}^T \vec{x}} = \lambda_1$$

$$\text{we say } Q^T \vec{x} \Rightarrow Q^T \vec{x} = [x]_Q \text{ + likewise}$$

$$\text{for } \vec{x}^T Q = [x]_Q^T$$

$$\Rightarrow \max \frac{[x]_Q^T \Lambda [x]_Q}{\vec{x}^T \vec{x}} = \frac{\sum_{i=1}^n \lambda_i [x]_{Q,i}^2}{\vec{x}^T \vec{x}} = \lambda_1$$

$$\Rightarrow \frac{\sum_{i=1}^n \lambda_i x_i^2}{\sum_{i=1}^n x_i^2} \text{ is maximized @ } \lambda_1$$

$$\text{we know } \lambda_1 \geq \lambda_i \geq \lambda_n$$

$$\Rightarrow x_i^2 \lambda_1 \geq x_i^2 \lambda_i \geq x_i^2 \lambda_n$$

$$\Rightarrow \sum_{i=1}^n x_i^2 \lambda_1 \geq \sum_{i=1}^n x_i^2 \lambda_i \geq \sum_{i=1}^n x_i^2 \lambda_n$$

→ Divide out for indices  $i$

$$\lambda_1 \geq \frac{\sum_{i=1}^n x_i^2 \lambda_i}{\sum_i x_i} \geq \lambda_p$$

max when  $x = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$  Since  $\lambda_1$  is at  $A_{11}$

and min when  $x = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$

in other words the first vector

in  $U$  when  $A$  is orthogonally  
Diagonalized

is max & last vector in  $U$  is  
minimized