

Recitation Solutions - Week 6

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Stochastic matrices

1. Prove that the product of two stochastic matrices is a stochastic matrix
2. Let $A \in \mathbb{R}^{n \times n}$ be a stochastic matrix. True or False:
 1. A is always invertible
 2. The eigenvector corresponding to the largest eigenvalue of A is unique
 3. A cannot have zero as its eigenvalue
3. Let $A \in \mathbb{R}^{n \times n}$ be a stochastic matrix. Prove that the absolute value of any eigenvalue of A is ≤ 1

Stochastic matrices

1. Prove that the product of two stochastic matrices is a stochastic matrix

Solution: Let P_1 and P_2 be the two stochastic matrices

1. $P_1 P_2[i, j] = P_1[i, :] P_2[:, j] \geq 0$

2. $[1, \dots, 1] P_1 P_2 = [1, \dots, 1] P_2 = [1, \dots, 1]$ proving that sum of entries in all columns of $P_1 P_2$ is 1

Stochastic matrices

2. Let $A \in \mathbb{R}^{n \times n}$ be a stochastic matrix. True or False:

1. A is always invertible
2. The eigenvector corresponding to the largest eigenvalue of A is unique
3. A cannot have zero as its eigenvalue

Solution:

i. False. $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$

ii. False, A=Identity matrix

iii. False, $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$

Stochastic matrices

3. Let $A \in \mathbb{R}^{n \times n}$ be a stochastic matrix. Prove that the absolute value of any eigenvalue of A is ≤ 1

Solution: For any x such that $Ax = \lambda x$

$$\|Ax\|_1 = \sum_i \left| \sum_j A_{ij} x_j \right| \leq \sum_i \sum_j |A_{ij} x_j|$$

A is a stochastic matrix

$$\begin{aligned} \Rightarrow \|Ax\|_1 &\leq \sum_i \sum_j A_{ij} |x_j| = \sum_j |x_j| \sum_i A_{ij} = \sum_j |x_j| = \|x\|_1 \\ &\Rightarrow |\lambda| \|x\|_1 \leq \|x\|_1 \\ &\Rightarrow |\lambda| \leq 1 \end{aligned}$$

Spectral decomposition

1. Let $A \in \mathbb{R}^{n \times n}$ be a symmetric matrix. Give a vector v with $\|v\| = 1$ such that $\|Av\|$ is maximized
2. A symmetric matrix $M \in \mathbb{R}^{n \times n}$ is said to be positive definite if all of its eigenvalues are greater than zero. Prove that $x^T M x > 0$ for a symmetric positive definite matrix $M \in \mathbb{R}^{n \times n}$ and $x \in \mathbb{R}^n$
3. Suppose S and T are symmetric and positive definite (all eigenvalues greater than zero)
 1. True or False: ST will always be symmetric
 2. Prove that all eigenvalues of ST are still positive

Spectral decomposition

1. Let $A \in \mathbb{R}^{n \times n}$ be a symmetric matrix. Give a vector v with $\|v\| = 1$ such that $\|Av\|$ is maximized

Solution: By spectral theorem, I can write A as $A = UDU^T$ where the columns of U (namely u_1, \dots, u_n) form an orthonormal basis of A and let the diagonal entries in D be d_1, \dots, d_n

Let $v = \alpha_1 u_1 + \dots + \alpha_n u_n$ such that $\sqrt{\alpha_1^2 + \dots + \alpha_n^2} = 1$

$$Av = UDU^T v = \alpha_1 d_1 u_1 + \dots + \alpha_n d_n u_n$$

$$\Rightarrow \|Av\| = \sqrt{\alpha_1^2 d_1^2 + \dots + \alpha_n^2 d_n^2}$$

Let $d_k^2 = \max\{d_1^2, \dots, d_n^2\}$

$$\Rightarrow \|Av\| = |d_k| \sqrt{\alpha_1^2 \left(\frac{d_1}{d_k}\right)^2 + \dots + \alpha_n^2 \left(\frac{d_n}{d_k}\right)^2} \leq |d_k| \sqrt{\alpha_1^2 + \dots + \alpha_n^2} = |d_k|$$

$\Rightarrow \|Av\| \leq |d_k|$ and the maximum is achieved for $v = u_k$

Spectral decomposition

2. A symmetric matrix $A \in \mathbb{R}^{n \times n}$ is said to be positive definite if all of its eigenvalues are greater than zero. Prove that $x^T Ax > 0$ for a symmetric positive definite matrix $A \in \mathbb{R}^{n \times n}$ and $x \in \mathbb{R}^n$

Solution: Following the notation from previous question,

$$\begin{aligned} Ax &= \alpha_1 d_1 u_1 + \cdots + \alpha_n d_n u_n \\ \Rightarrow x^T Ax &= \alpha_1^2 d_1 + \cdots + \alpha_n^2 d_n > 0 \text{ since } d_1, \dots, d_n > 0 \end{aligned}$$

Spectral decomposition

3. Suppose S and T are symmetric and positive definite (all eigenvalues greater than zero)

1. True or False: Product of two symmetric matrices is always symmetric
2. Prove that all eigenvalues of ST are still positive

Solution:

1. $S_1 = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}, S_2 = \begin{bmatrix} 1 & 3 \\ 3 & 2 \end{bmatrix} \Rightarrow S_1 S_2 = \begin{bmatrix} 7 & 7 \\ 5 & 8 \end{bmatrix}$ (Not symmetric)

2. Let x be an eigenvector of ST corresponding to eigenvalue $\lambda \Rightarrow STx = \lambda x$

$$\Rightarrow (Tx)^T STx = \lambda (Tx)^T x$$

$$\Rightarrow x^T T^T STx = \lambda x^T Tx$$

$$\Rightarrow \lambda = v^T S v / x^T T x \text{ where } v = Tx$$

Since S and T are positive definite, $x^T T x > 0$ and $v^T S v > 0$

$$\Rightarrow \lambda > 0$$