Progress

Mentor

1 point

## Course outline

How does an NPTEL online course work?

Week-0

Week-1

Week-3

Week-4

Week-2

- Lec 17- Positive Semi-definite (PSD) Matrices: Definition, Properties, Eigenvalue Decomposition
- Lec 18- Positive Semidefinite Matrix: Example and Illustration of Eigenvalue
- Decomposition Lec 19- Machine Learning
- Application: Principle Component Analysis (PCA)
- Lec 20- Computer Vision
- Application: Face Recognition, Eigenfaces
- Quiz : Assignment-4
- Feedback for Week 4 Solution-4

Week-5

Week-7

Week-8

Week-9

Week-11

Week-10

Week-12

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## **Assignment-4**

The due date for submitting this assignment has passed.

Due on 2021-02-17, 23:59 IST.

As per our records you have not submitted this assignment.

1) Consider i.i.d. zero-mean Gaussian random variables  $x_1, x_2, \dots, x_n$  of variance  $\sigma^2$ . Let  $\bar{\mathbf{x}} = [x_1, x_2, \dots, x_n]^T$ . The variance of the quantity **1 point**  $\bar{\mathbf{a}}^T\bar{\mathbf{x}}$  is

 $\sigma^2 \|\bar{\boldsymbol{a}}\|$ 

 $\sigma^2(|a_1| + |a_2| + \cdots + |a_n|)$ 

 $\sigma^2 \|\bar{\mathbf{a}}\|^2$  $\sigma^2(a_1 + a_2 + \dots + a_n)$ 

Score: 0 Accepted Answers:

No, the answer is incorrect.

 $\sigma^2 \|\bar{\mathbf{a}}\|^2$ 

The eigenvalue decomposition of a real symmetric matrix A is of the form

 $U\Lambda U^{-1}$ , where the  $\Lambda$  is an arbitrary diagonal matrix of eigenvalues and U is an arbitrary matrix of eigenvectors

 $\mathbf{U}\Lambda\mathbf{U}^T$ , where the  $\Lambda$  is a diagonal matrix of real eigenvalues and  $\mathbf{U}$  is a matrix of orthonormal eigenvectors

 $\mathbf{U} \mathbf{\Lambda} \mathbf{U}^T$ , where the  $\mathbf{\Lambda}$  is a diagonal matrix of non-negative eigenvalues and  $\mathbf{U}$  is a matrix of orthonormal eigenvectors

 $\mathbf{U}\Lambda\mathbf{U}^T$ , where the  $\Lambda$  is a diagonal matrix of eigenvalues that have magnitude unity and  $\mathbf{U}$  is a matrix of orthonormal eigenvectors No, the answer is incorrect.

Score: 0 Accepted Answers:

 $\mathbf{U}\Lambda\mathbf{U}^T$ , where the  $\Lambda$  is a diagonal matrix of real eigenvalues and  $\mathbf{U}$  is a matrix of orthonormal eigenvectors

 The eigenvectors of a positive semi-definite matrix corresponding to distinct eigenvalues are 1 point Orthogonal and have unit-norm

Orthogonal but not necessarily unit-norm Neither orthogonal nor unit-norm

 Non-orthogonal and have unit-norm No, the answer is incorrect. Score: 0

Accepted Answers: Orthogonal but not necessarily unit-norm

4) The eigenvalues of the matrix below are  $\mathbf{A} = \begin{bmatrix} 2 & 3 \\ 2 & 1 \end{bmatrix}$ 

Score: 0

4, -1

Accepted Answers:

5) Consider the Gaussian random vector  $\bar{\mathbf{x}}$  with mean  $\bar{\boldsymbol{\mu}}_{x}$  and covariance matrix  $\mathbf{R}_{x}$ . Consider now  $\bar{\mathbf{y}} = \mathbf{A}\bar{\mathbf{x}} + \bar{\mathbf{b}}$ . Its mean and covariance matrix 1 point are given as

$$\mathbf{A}\bar{\boldsymbol{\mu}}_{x}, \mathbf{A}\mathbf{R}_{x}\mathbf{A}^{T}$$
 $\mathbf{A}\bar{\boldsymbol{\mu}}_{x}, \mathbf{A}\mathbf{R}_{x}\mathbf{A}^{T} + \bar{\mathbf{b}}\bar{\mathbf{b}}^{T}$ 
 $\mathbf{A}\bar{\boldsymbol{\mu}}_{x} + \bar{\mathbf{b}}, \mathbf{A}\mathbf{R}_{x}\mathbf{A}^{T} + \bar{\mathbf{b}}\bar{\mathbf{b}}^{T}$ 
 $\mathbf{A}\bar{\boldsymbol{\mu}}_{x} + \bar{\mathbf{b}}, \mathbf{A}\mathbf{R}_{x}\mathbf{A}^{T} + \bar{\mathbf{b}}\bar{\mathbf{b}}^{T}$ 
 $\mathbf{A}\bar{\boldsymbol{\mu}}_{x} + \bar{\mathbf{b}}, \mathbf{A}\mathbf{R}_{x}\mathbf{A}^{T}$ 
No, the answer is incorrect. Score: 0
Accepted Answers:

 $A\bar{\mu}_x + \bar{\mathbf{b}}, A\mathbf{R}_x \mathbf{A}^T$ 

Consider a 2 dimensional random column vector  $\bar{\mathbf{x}}$  that has the multi-variate Gaussian distribution with mean  $\bar{\boldsymbol{\mu}}_x = \begin{bmatrix} -2 \\ 4 \end{bmatrix}$  and covariance matrix  $\Sigma_x = \begin{bmatrix} 6 & 0 \\ 0 & 2 \end{bmatrix}$ . Consider the vector  $\bar{\mathbf{y}}$  defined as

 $\bar{\mathbf{y}} = \begin{bmatrix} 1 & -2 \\ 3 & -4 \end{bmatrix} \bar{\mathbf{x}} + \begin{bmatrix} -2 \\ 3 \end{bmatrix}.$ What is the mean  $\bar{\mu}_{\nu}$  and covariance matrix  $\Sigma_{\nu}$  of  $\bar{y}$ ?

$$\bar{\boldsymbol{\mu}}_{y} = \begin{bmatrix} -12 \\ -19 \end{bmatrix}, \boldsymbol{\Sigma}_{y} = \begin{bmatrix} 14 & 34 \\ 34 & 86 \end{bmatrix}$$

$$\bar{\boldsymbol{\mu}}_{y} = \begin{bmatrix} -12 \\ -19 \end{bmatrix}, \boldsymbol{\Sigma}_{y} = \begin{bmatrix} 12 & 18 \\ 18 & 48 \end{bmatrix}$$

$$\bar{\boldsymbol{\mu}}_{y} = \begin{bmatrix} -6 \\ -15 \end{bmatrix}, \boldsymbol{\Sigma}_{y} = \begin{bmatrix} 14 & 34 \\ 34 & 86 \end{bmatrix}$$

 $\bar{\boldsymbol{\mu}}_{y} = \begin{bmatrix} -6 \\ -15 \end{bmatrix}, \boldsymbol{\Sigma}_{y} = \begin{bmatrix} 12 & 18 \\ 18 & 48 \end{bmatrix}$ 

Accepted Answers: 
$$\bar{\boldsymbol{\mu}}_y = \begin{bmatrix} -12 \\ -19 \end{bmatrix}, \boldsymbol{\Sigma}_y = \begin{bmatrix} 14 & 34 \\ 34 & 86 \end{bmatrix}$$
7) Consider the Gaussian classification problem with the two classes distributed as

No, the answer is incorrect.

Score: 0

The classifier for this problem, to classify a new vector  $\bar{\mathbf{x}}$ , can be formulated as

 $C_1 \sim N(\bar{\boldsymbol{\mu}}_1, \boldsymbol{\Sigma})$  $C_2 \sim N(\bar{\mu}_2, \Sigma)$ 

Choose  $C_1$  if  $(\bar{\mathbf{\mu}}_1 - \bar{\mathbf{\mu}}_2)^T \mathbf{\Sigma}^{-1} \left( \bar{\mathbf{x}} - \frac{\bar{\mathbf{\mu}}_1 + \bar{\mathbf{\mu}}_2}{2} \right) \leq 0$  and  $C_2$  otherwise Choose  $C_1$  if  $(\bar{\mathbf{\mu}}_1 - \bar{\mathbf{\mu}}_2)^T \mathbf{\Sigma}^{-1} \left( \bar{\mathbf{x}} - \frac{\bar{\mathbf{\mu}}_1 + \bar{\mathbf{\mu}}_2}{2} \right) \ge 0$  and  $C_2$  otherwise

Choose  $C_1$  if  $(\bar{\mathbf{\mu}}_1 - \bar{\mathbf{\mu}}_2)^T \mathbf{\Sigma} \left(\bar{\mathbf{x}} - \frac{\bar{\mathbf{\mu}}_1 + \bar{\mathbf{\mu}}_2}{2}\right) \ge 0$  and  $C_2$  otherwise

Choose  $C_1$  if  $(\bar{\mathbf{\mu}}_1 - \bar{\mathbf{\mu}}_2)^T (\bar{\mathbf{x}} - \frac{\bar{\mathbf{\mu}}_1 + \bar{\mathbf{\mu}}_2}{2}) \ge 0$  and  $C_2$  otherwise

Accepted Answers: Choose  $C_1$  if  $(\bar{\mathbf{\mu}}_1 - \bar{\mathbf{\mu}}_2)^T \mathbf{\Sigma}^{-1} \left( \bar{\mathbf{x}} - \frac{\bar{\mathbf{\mu}}_1 + \bar{\mathbf{\mu}}_2}{2} \right) \ge 0$  and  $C_2$  otherwise

No, the answer is incorrect.

Score: 0

8) Consider the Gaussian classification problem with the two classes  $C_1$ ,  $C_2$  distributed as  $C_1 \sim N\left(\begin{bmatrix}2\\-4\end{bmatrix},\begin{bmatrix}\frac{1}{4} & 0\\0 & \frac{1}{8}\end{bmatrix}\right), C_2 \sim N\left(\begin{bmatrix}-4\\2\end{bmatrix},\begin{bmatrix}\frac{1}{4} & 0\\0 & \frac{1}{8}\end{bmatrix}\right)$ 

The classifier for this problem, to classify a new vector  $\bar{\mathbf{x}} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ , can be formulated as choose  $C_1$  if

$$4x_1 + 6x_2 \ge 2$$

$$3x_1 - 2x_2 \ge -1$$

$$-2x_1 + 4x_2 \le 3$$

$$x_1 - 2x_2 \ge 1$$

 $x_1 - 2x_2 \ge 1$ 9) Consider a zero-mean random vector to have independent random variables. Its covariance matrix is

Proportional to the identity matrix

No, the answer is incorrect.

Accepted Answers:

 A diagonal matrix but not necessarily proportional to the identity matrix An arbitrary positive semi-definite matrix, not necessarily diagonal

 Any arbitrary matrix No, the answer is incorrect.

Accepted Answers: A diagonal matrix but not necessarily proportional to the identity matrix

The eigenvectors of the matrix below are

Score: 0

Score: 0

Accepted Answers:

 $\mathbf{A} = \begin{bmatrix} 2 & 3 \\ 2 & 1 \end{bmatrix}$ 

