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NPTEL (<https://swayam.gov.in/explorer?ncCode=NPTEL>) » **Pattern Recognition And Application**
(course)



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Course
outline

How does an
NPTEL
online
course
work? ()

Week 0 ()

Week 1 ()

Week 2 ()

Week 3 ()

☐ Lecture 06 :
Normal

Week 3 : Assignment 3

Assignment not submitted

Due date: 2023-08-16, 23:59 IST.

1)

2 points

Let μ is the mean vector and Σ is the covariance matrix in a d dimensional space. Which of the following is the correct expression for normal/Gaussian density function?

a) $P(X) = \frac{1}{(2\pi)^{d/2} \sqrt{|\Sigma|}} \exp\left[-\frac{1}{2}(X - \mu)^T \Sigma^{-1}(X - \mu)\right]$

b) $P(X) = \frac{1}{(2\pi)^{d/2} \Sigma} \exp\left[-\frac{1}{2}(X - \mu)^T \Sigma^{-1}(X - \mu)\right]$

c) $P(X) = \frac{1}{(2\pi)^{d/2} \sqrt{\Sigma}} \exp\left[-\frac{1}{2}(X - \mu) \Sigma^{-1}(X - \mu)^T\right]$

d) $P(X) = \frac{1}{(2\pi)^{d/2} \sqrt{\Sigma}} \exp\left[-\frac{1}{2}(X - \mu)^T \Sigma(X - \mu)\right]$

- ☐ a)
☐ b)
☐ c)
☐ d)

2)

2 points

Density and
Discriminant
Function - I
(unit?unit=26&
lesson=27)

Lecture 07 :
Normal
Density and
Discriminant
Function - II
(unit?unit=26&
lesson=28)

Lecture 08 :
Bayes
Decision
Theory -
Binary
Features
(unit?unit=26&
lesson=29)

Quiz: Week 3
: Assignment
3
(assessment?
name=111)

Feedback
Form for Week
3
(unit?unit=26&
lesson=115)

Week 4 ()

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Let $\left\{\begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 2 \\ 2 \end{pmatrix}, \begin{pmatrix} 3 \\ 1 \end{pmatrix}\right\} \in \omega_1$, $\left\{\begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ -1 \end{pmatrix}, \begin{pmatrix} 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 3 \\ 1 \end{pmatrix}\right\} \in \omega_2$. Compute the covariance matrix Σ_1 and Σ_2 .

- a) $\Sigma_1 = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$, $\Sigma_2 = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$
 b) $\Sigma_1 = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$, $\Sigma_2 = \begin{bmatrix} 2.0 & 0 \\ 0 & 0.5 \end{bmatrix}$
 c) $\Sigma_1 = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$, $\Sigma_2 = \begin{bmatrix} 0.5 & 0 \\ 0 & 2.0 \end{bmatrix}$
 d) None of these

- ☐ a)
☐ b)
☐ c)
☐ d)

3)

2 points

For a two class problem having d dimensional binary feature vectors where $a_i = P\left(x_i = 1 / \omega_1\right)$ and $b_i = P\left(x_i = 1 / \omega_2\right)$. Assume all the features are statistically independent and classes are equiprobable. Which of the following corresponds to equation of decision surface?

- a) $\sum_{i=1}^d \left[x_i \ln \left(\frac{a_i}{b_i} \right) + (1 - x_i) \ln \left(\frac{1 - a_i}{1 - b_i} \right) \right] = 0$
 b) $\sum_{i=1}^d \left[x_i \ln \left(\frac{a_i}{b_i} \right) + (1 - x_i) \ln \left(\frac{1 - a_i}{1 - b_i} \right) \right] + \ln \left(\frac{P(\omega_1)}{P(\omega_2)} \right) = 0$
 c) $\sum_{i=1}^d \left[x_i \ln \left(\frac{a_i}{b_i} \right) + (1 - x_i) \ln \left(\frac{1 - b_i}{1 - a_i} \right) \right] + \ln \left(\frac{P(\omega_1)}{P(\omega_2)} \right) = 0$
 d) Both a and b

- ☐ a)
☐ b)
☐ c)
☐ d)

4)

2 points

Let Σ_i represents the covariance matrix for i^{th} class. If the classes have the same co-variance matrix. The features are statistically independent and have same co-variance. Which of the following is true?

- a) $\Sigma_i = \Sigma$, (Σ is non - diagonal)
 b) $\Sigma_i = \Sigma$, (Σ is diagonal)
 c) $\Sigma_i = \sigma^2 I$
 d) None of these

- ☐ a)
☐ b)
☐ c)

☐ d)

5)

2 points

For a two-dimension feature vector, the data points lies on the locus of a circle as shown in Figure

1. Let the covariance matrix be $\Sigma = \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix}$. Which of the following is/are correct?

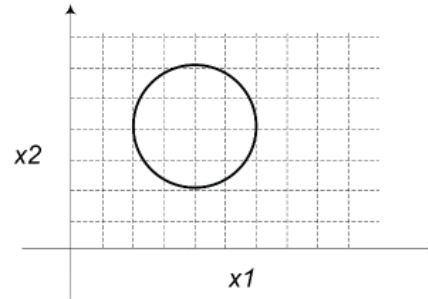


Figure 1.

- I. $\sigma_{11} = \sigma_{22}$
- II. $\sigma_{11} \neq \sigma_{22}$
- III. $\sigma_{12} = \sigma_{21}$
- IV. $\sigma_{12} = 0$

- a) Only I and III
- b) Only II and III
- c) Only II, III and IV
- d) Only I, III and IV

☐ a)

☐ b)

☐ c)

☐ d)

6)

2 points

For a two-dimension feature vector, the data points lies on the locus of an ellipse as shown in Figure 2. Let the covariance matrix be $\Sigma = \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix}$. Which of the following is/are correct?

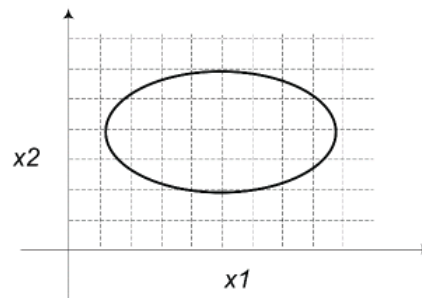


Figure 2.

- I. $\sigma_{11} = \sigma_{22}$
- II. $\sigma_{11} \neq \sigma_{22}$
- III. $\sigma_{12} = \sigma_{21}$
- IV. $\sigma_{12} = 0$

- a) Only I and III
- b) Only II and III
- c) Only II, III and IV
- d) Only I, III and IV

- ☐ a)
- ☐ b)
- ☐ c)
- ☐ d)

7)

2 points

For a two-dimension feature vector, the data points lies on the locus of an ellipse as shown in Figure 3. Let the covariance matrix be $\Sigma = \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix}$. Which of the following is/are correct?

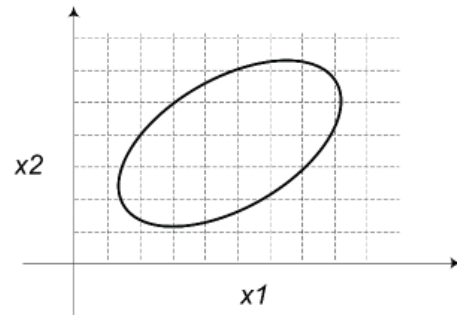


Figure 3.

- I. $\sigma_{11} = \sigma_{22}$
- II. $\sigma_{11} \neq \sigma_{22}$
- III. $\sigma_{12} = \sigma_{21}$
- IV. $\sigma_{12} = 0$

- a) Only I and III
- b) Only II and III
- c) Only II, III and IV
- d) Only I, III and IV

- ☐ a)
- ☐ b)
- ☐ c)
- ☐ d)

8)

2 points

For a two class problem, $\Sigma_i = \Sigma \neq \sigma^2 I$, where Σ_i represents the covariance matrix for i^{th} class. Both classes are normally distributed and Bayesian classifier is used. Assume $P(\omega_1) = P(\omega_2)$. Which of the following is true for decision boundary?

- a) Decision boundary passes through the midpoint of the line segment joining the means of two classes
- b) Decision boundary will be orthogonal bisector of the line joining the means of two classes.
- c) Both a and b
- d) None of the above

- ☐ a)
- ☐ b)
- ☐ c)
- ☐ d)

9)

2 points

For a two class problem, $\Sigma_i = \sigma^2 I$, where Σ_i represents the covariance matrix for i^{th} class. Both classes are normally distributed and Bayesian classifier is used. Assume $P(\omega_1) = P(\omega_2)$. Which of the following is true for decision boundary?

- a) Decision boundary passes through the midpoint of the line segment joining the means of two classes
- b) Decision boundary will be orthogonal bisector of the line joining the means of two classes.
- c) Both a and b
- d) None of the above

- ☐ a)
- ☐ b)
- ☐ c)
- ☐ d)

10)

2 points

Let $\left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 2 \\ 2 \end{pmatrix}, \begin{pmatrix} 3 \\ 1 \end{pmatrix} \right\} \in \omega_1$, $\left\{ \begin{pmatrix} 6 \\ 1 \end{pmatrix}, \begin{pmatrix} 8 \\ -1 \end{pmatrix}, \begin{pmatrix} 8 \\ 3 \end{pmatrix}, \begin{pmatrix} 10 \\ 1 \end{pmatrix} \right\} \in \omega_2$ as shown in Figure 4. For class ω_1 ,

$\mu_1 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$, $\Sigma_1 = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$. For class ω_2 , $\mu_2 = \begin{pmatrix} 8 \\ 1 \end{pmatrix}$ and $\Sigma_2 = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$. Assume classes are

normally distributed and $P(\omega_1) = P(\omega_2)$. Classify an unknown feature vector $X = \begin{pmatrix} 4.5 \\ 1 \end{pmatrix}$

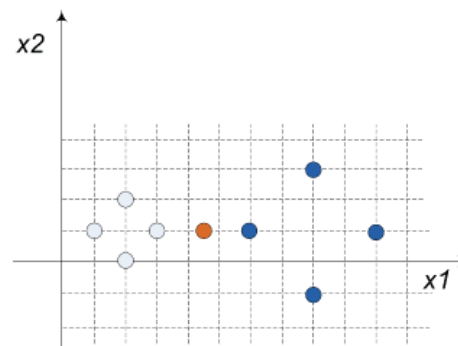


Figure 4.

- a) $X \in \omega_1$
- b) $X \in \omega_2$
- c) X lies on decision surface
- d) None of the above

- ☐ a)
- ☐ b)
- ☐ c)
- ☐ d)

You may submit any number of times before the due date. The final submission will be considered for grading.

Submit Answers

