

## 7. Linear Discriminant Analysis

### Discriminant functions

Consider a classifier built using functions

$$g_i(\bar{x}), \quad i = 1, 2, \dots, L$$

such that the input vector  $\bar{x}$  is assigned to class  $\lambda$  if

$$g_\lambda(\bar{x}) = \max_{1 \leq i \leq L} g_i(\bar{x})$$

These  $g_i(\bar{x})$  are termed discriminant functions.

### Gaussian density

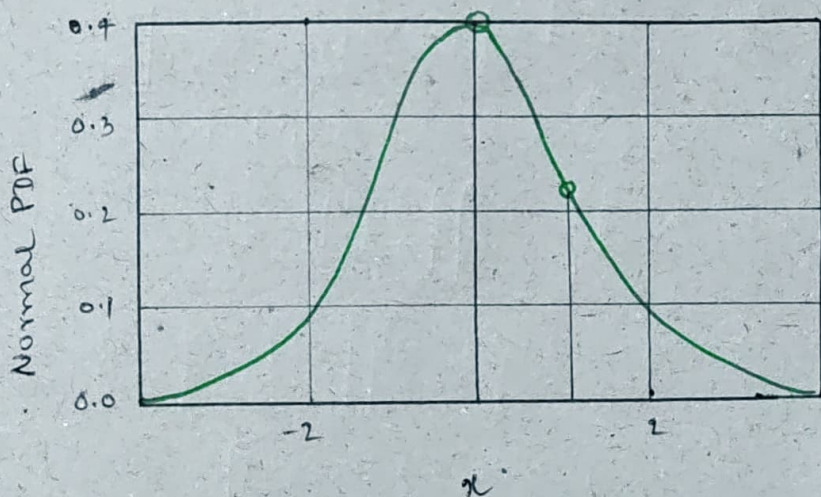
Recall, the expression for the Gaussian PDF is

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

The Mean and Variance of the Gaussian RV are

$$E\{X\} = \mu$$

$$E\{(X-\mu)^2\} = \sigma^2$$





## Multivariate Gaussian

The PDF of a Gaussian random vector is given as

$$f_{\vec{x}}(\vec{x}) = \frac{1}{\sqrt{(2\pi)^n |R|}} e^{-\frac{1}{2}(\vec{x}-\vec{\mu})^T R^{-1} (\vec{x}-\vec{\mu})}$$

The Mean and Covariance Matrix are defined as

$$E\{\vec{x}\} = \vec{\mu}$$

$$E\{(\vec{x}-\vec{\mu})(\vec{x}-\vec{\mu})^T\} = R$$

outer product

### Problem 1

Given  $R$  below, find  $|R|$  and  $R^{-1}$ .

$$R = \begin{bmatrix} 7 & 2 \\ 2 & 1 \end{bmatrix}$$

Covariance Matrix,  $R = \begin{bmatrix} 7 & 2 \\ 2 & 1 \end{bmatrix}$

$$|R| = (7)(1) - (2)(2) = 3$$

$$R^{-1} = \frac{1}{|R|} \text{adj}(R) = \frac{1}{3} \begin{bmatrix} 1 & -2 \\ -2 & 7 \end{bmatrix}$$

### Problem 2

Find  $(\vec{x}-\vec{\mu})^T R^{-1} (\vec{x}-\vec{\mu})$ , given  $\vec{\mu} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

The quantity in exponent can be simplified as

$$\begin{aligned} & \frac{1}{3} \begin{bmatrix} x_1-1 & x_2-2 \end{bmatrix} \begin{bmatrix} 1 & -2 \\ -2 & 7 \end{bmatrix} \begin{bmatrix} x_1-1 \\ x_2-2 \end{bmatrix} \\ &= \frac{1}{3} \left[ (x_1-1)^2 + 7(x_2-2)^2 - 2(x_1-1)(x_2-2) - 2(x_1-1)(x_2-2) \right] \\ &= \frac{1}{3} \left[ x_1^2 + 7x_2^2 + 6x_1 - 24x_2 - 4x_1x_2 + 21 \right] \end{aligned}$$



### Problem 3

Find the multivariate Gaussian PDF. ( $n=2$ )

$$f_{\vec{x}}(\vec{x}) = \frac{1}{\sqrt{(2\pi)^n |R|}} e^{-\frac{1}{2}(\vec{x}-\vec{\mu})^T R^{-1}(\vec{x}-\vec{\mu})}$$
$$= \frac{1}{\sqrt{12\pi^2}} e^{-\frac{1}{6}(x_1^2 + 7x_2^2 + 6x_1 - 24x_2 - 4x_1x_2 + 21)}$$

### Gaussian discriminant

\* Consider the input vectors  $\vec{x}$  drawn from two Gaussian classes

$C_0$  : Mean  $\vec{\mu}_0$  and Covariance  $R$

$C_1$  : Mean  $\vec{\mu}_1$  and Covariance  $R$

\* Also termed Gaussian Discriminant Analysis.

\* The Likelihoods of the two classes are

$$p(\vec{x}; C_0) = \frac{1}{\sqrt{(2\pi)^n |R|}} e^{-\frac{1}{2}(\vec{x}-\vec{\mu}_0)^T R^{-1}(\vec{x}-\vec{\mu}_0)}$$

$$p(\vec{x}; C_1) = \frac{1}{\sqrt{(2\pi)^n |R|}} e^{-\frac{1}{2}(\vec{x}-\vec{\mu}_1)^T R^{-1}(\vec{x}-\vec{\mu}_1)}$$

\* Choose the class that maximizes the likelihood.

Choose  $C_0$  if

$$p(\vec{x}; C_0) \geq p(\vec{x}; C_1)$$

$$\Rightarrow \frac{1}{\sqrt{(2\pi)^n |R|}} e^{-\frac{1}{2}(\vec{x}-\vec{\mu}_0)^T R^{-1}(\vec{x}-\vec{\mu}_0)} \geq \frac{1}{\sqrt{(2\pi)^n |R|}} e^{-\frac{1}{2}(\vec{x}-\vec{\mu}_1)^T R^{-1}(\vec{x}-\vec{\mu}_1)}$$

$$\Rightarrow (\vec{x}-\vec{\mu}_0)^T R^{-1}(\vec{x}-\vec{\mu}_0) \leq (\vec{x}-\vec{\mu}_1)^T R^{-1}(\vec{x}-\vec{\mu}_1)$$



This discriminant function can be simplified as

$$\text{Choose } C_0 : \bar{h}^T (\bar{x} - \bar{\mu}) \geq 0$$

$$\text{Choose } C_1 : \bar{h}^T (\bar{x} - \bar{\mu}) < 0$$

where,

$$\bar{\mu} = \frac{1}{2} (\bar{\mu}_0 + \bar{\mu}_1)$$

$$\bar{h} = R^{-1} (\bar{\mu}_0 - \bar{\mu}_1)$$

### Linear classifier

Thus, the classifier is linear, and is characterized by the hyperplane.

$$\bar{h}^T (\bar{x} - \bar{\mu}) = 0.$$

### Problem 4 :

Given the two classes  $C_0, C_1$  are distributed

$$C_0 \sim \mathcal{N} \left( \begin{bmatrix} -1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1/2 & 0 \\ 0 & 1/4 \end{bmatrix} \right)$$

$$C_1 \sim \mathcal{N} \left( \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 1/2 & 0 \\ 0 & 1/4 \end{bmatrix} \right)$$

Calculate  $\bar{h}, \bar{\mu}$ .

$$\bar{\mu} = \frac{1}{2} (\bar{\mu}_0 + \bar{\mu}_1) = \frac{1}{2} \left( \begin{bmatrix} -1 \\ 2 \end{bmatrix} + \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right) = \frac{1}{2} \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$\bar{h} = R^{-1} (\bar{\mu}_0 - \bar{\mu}_1) = \begin{bmatrix} 2 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} -3 \\ 1 \end{bmatrix} = \begin{bmatrix} -6 \\ 4 \end{bmatrix}$$



Problem 5 :

For the same classes  $C_0$  and  $C_1$ , determine the classifier.

$$\begin{aligned}\bar{h}^T (\bar{x} - \bar{\mu}) &= [-6 \ 4] \left( \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} - \begin{bmatrix} 1/2 \\ 3/2 \end{bmatrix} \right) \\ &= [-6 \ 4] \left( \bar{x} - \frac{1}{2} \begin{bmatrix} 1 \\ 3 \end{bmatrix} \right) \geq 0\end{aligned}$$

Choose $C_0$ if $-6x_1 + 4x_2 - 3 \geq 0$
Choose $C_1$ if $-6x_1 + 4x_2 - 3 < 0$

$$\max \left[ (-6x_1 + 4x_2 - 3), (6x_1 - 4x_2 + 3) \right]$$