





**Course Name: Deep Learning** 

**Faculty Name: Prof. P. K. Biswas** 

**Department: E & ECE, IIT Kharagpur** 

#### **Topic**

**Lecture 06: Discriminant Function and Decision Surface** 

#### **CONCEPTS COVERED**

**Concepts Covered:** 

☐ Bayes Minimum Error Classifier

☐ Bayes Minimum Risk Classifier

**□** Discriminant Function

Decision Boundary





## **Discriminant Function**

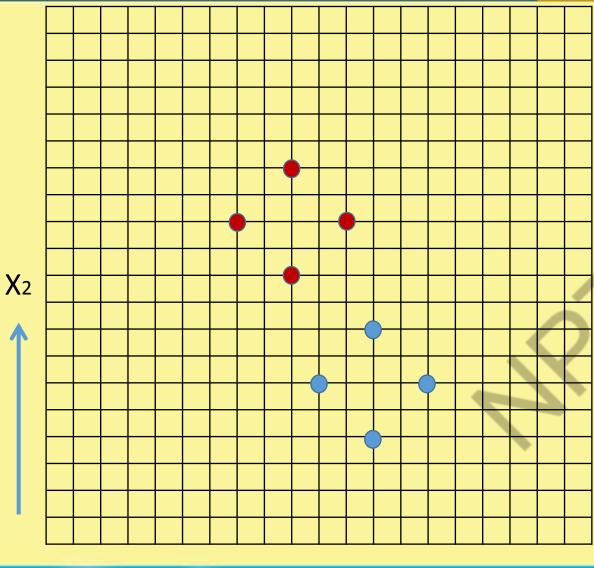




# Discriminant Function under Multivariate Normal Distribution







$$\begin{bmatrix} 12 \\ 4 \end{bmatrix} \begin{bmatrix} 12 \\ 8 \end{bmatrix} \begin{bmatrix} 10 \\ 6 \end{bmatrix} \begin{bmatrix} 14 \\ 6 \end{bmatrix} \implies \omega_1$$

$$\begin{bmatrix} 9 \\ 10 \end{bmatrix} \begin{bmatrix} 9 \\ 14 \end{bmatrix} \begin{bmatrix} 7 \\ 12 \end{bmatrix} \begin{bmatrix} 11 \\ 12 \end{bmatrix} \implies \omega_2$$

**X**1





$$\begin{bmatrix} 12 \\ 4 \end{bmatrix} \begin{bmatrix} 12 \\ 8 \end{bmatrix} \begin{bmatrix} 10 \\ 6 \end{bmatrix} \begin{bmatrix} 14 \\ 6 \end{bmatrix} \implies \omega_1 \qquad \mu_1 = \frac{1}{4} \begin{bmatrix} 12 \\ 4 \end{bmatrix} + \begin{bmatrix} 12 \\ 8 \end{bmatrix} + \begin{bmatrix} 10 \\ 6 \end{bmatrix} + \begin{bmatrix} 14 \\ 6 \end{bmatrix} = \begin{bmatrix} 12 \\ 6 \end{bmatrix}$$

$$[X_1 - \mu_1][X_1 - \mu_1]^t = \begin{bmatrix} 0 \\ -2 \end{bmatrix}[0 \quad -2] = \begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix} = M_1$$

$$[X_{2} - \mu_{1}][X_{2} - \mu_{1}]^{t} = \begin{bmatrix} 0 \\ 2 \end{bmatrix} [0 \quad 2] = \begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix} = M_{2}$$

$$[X_3 - \mu_1][X_3 - \mu_1]^t = \begin{bmatrix} -2 \\ 0 \end{bmatrix} [-2 \quad 0] = \begin{bmatrix} 4 & 0 \\ 0 & 0 \end{bmatrix} = M_3$$

$$[X_4 - \mu_1][X_4 - \mu_1]^t = \begin{bmatrix} 2 \\ 0 \end{bmatrix} [2 \quad 0] = \begin{bmatrix} 4 & 0 \\ 0 & 0 \end{bmatrix} = M_4$$

$$\Sigma_1 = \frac{1}{4} [M_1 + M_2 + M_3 + M_4]$$

$$= \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} = 2 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = 2I$$



$$\begin{bmatrix} 9 \\ 10 \end{bmatrix} \begin{bmatrix} 9 \\ 14 \end{bmatrix} \begin{bmatrix} 7 \\ 12 \end{bmatrix} \begin{bmatrix} 11 \\ 12 \end{bmatrix} \implies \omega_2 \qquad \mu_2 = \frac{1}{4} \begin{bmatrix} 9 \\ 10 \end{bmatrix} + \begin{bmatrix} 9 \\ 14 \end{bmatrix} + \begin{bmatrix} 7 \\ 12 \end{bmatrix} + \begin{bmatrix} 11 \\ 12 \end{bmatrix} = \begin{bmatrix} 9 \\ 12 \end{bmatrix}$$

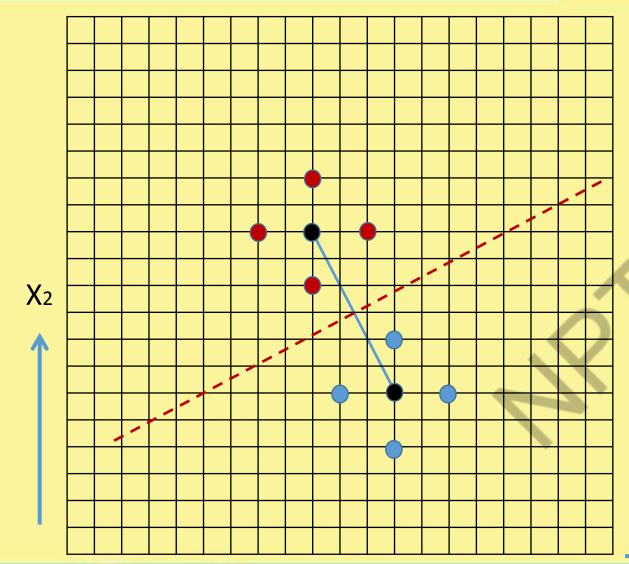
$$\Sigma_2 = 2I$$



$$\Sigma_1 = \Sigma_2 = 2I \approx \sigma^2 I$$
 Where

$$\sigma = \sqrt{2}$$





$$W^t(X - X_0) = 0$$

$$W = \mu_2 - \mu_1$$

$$X_0 = \frac{1}{2}(\mu_1 + \mu_2) - \frac{\sigma^2}{\|\mu_1 - \mu_2\|^2} \ln \frac{P(\omega_1)}{P(\omega_2)}(\mu_1 - \mu_2)$$



# Discriminant Function under Multivariate Normal Distribution











Thank you







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#### **Topic**

**Lecture 07: Discriminant Function and Decision Surface - II** 

#### **CONCEPTS COVERED**

#### Concepts Covered:

□ Discriminant Function under Multivariate

**Normal Distribution** 

☐ Decision Boundary under Various Cases of

**Covariance Matrices** 

**□** Examples

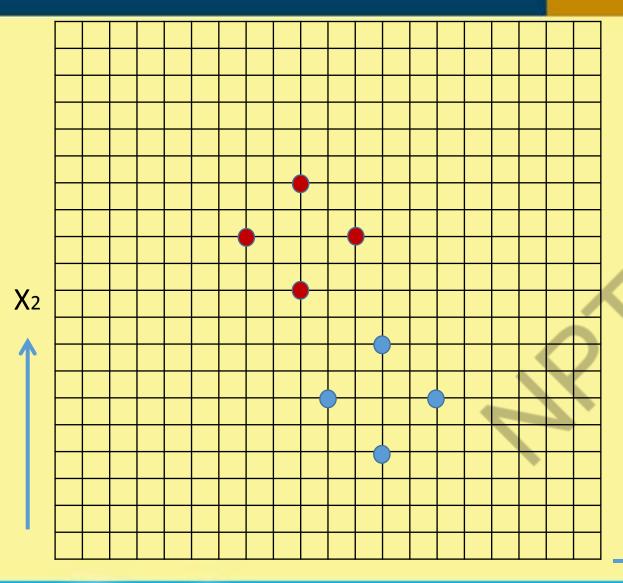




# Discriminant Function under Multivariate Normal Distribution







$$\begin{bmatrix} 12 \\ 4 \end{bmatrix} \begin{bmatrix} 12 \\ 8 \end{bmatrix} \begin{bmatrix} 10 \\ 6 \end{bmatrix} \begin{bmatrix} 14 \\ 6 \end{bmatrix} \implies \omega_1$$

$$\begin{bmatrix} 9 \\ 10 \end{bmatrix} \begin{bmatrix} 9 \\ 14 \end{bmatrix} \begin{bmatrix} 7 \\ 12 \end{bmatrix} \begin{bmatrix} 11 \\ 12 \end{bmatrix} \implies \omega_2$$



$$\begin{bmatrix} 12 \\ 4 \end{bmatrix} \begin{bmatrix} 12 \\ 8 \end{bmatrix} \begin{bmatrix} 10 \\ 6 \end{bmatrix} \begin{bmatrix} 14 \\ 6 \end{bmatrix} \implies \omega_1 \qquad \mu_1 = \frac{1}{4} \begin{bmatrix} 12 \\ 4 \end{bmatrix} + \begin{bmatrix} 12 \\ 8 \end{bmatrix} + \begin{bmatrix} 10 \\ 6 \end{bmatrix} + \begin{bmatrix} 14 \\ 6 \end{bmatrix} = \begin{bmatrix} 12 \\ 6 \end{bmatrix}$$

$$[X_1 - \mu_1][X_1 - \mu_1]^t = \begin{bmatrix} 0 \\ -2 \end{bmatrix}[0 \quad -2] = \begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix} = M_1$$

$$[X_{2} - \mu_{1}][X_{2} - \mu_{1}]^{t} = \begin{bmatrix} 0 \\ 2 \end{bmatrix} [0 \quad 2] = \begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix} = M_{2}$$

$$[X_3 - \mu_1][X_3 - \mu_1]^t = \begin{bmatrix} -2 \\ 0 \end{bmatrix} [-2 \quad 0] = \begin{bmatrix} 4 & 0 \\ 0 & 0 \end{bmatrix} = M_3$$

$$[X_4 - \mu_1][X_4 - \mu_1]^t = \begin{bmatrix} 2 \\ 0 \end{bmatrix} [2 \quad 0] = \begin{bmatrix} 4 & 0 \\ 0 & 0 \end{bmatrix} = M_4$$

$$\Sigma_1 = \frac{1}{4} [M_1 + M_2 + M_3 + M_4]$$

$$= \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} = 2 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = 2I$$



$$\begin{bmatrix} 9 \\ 10 \end{bmatrix} \begin{bmatrix} 9 \\ 14 \end{bmatrix} \begin{bmatrix} 7 \\ 12 \end{bmatrix} \begin{bmatrix} 11 \\ 12 \end{bmatrix} \implies \omega_2 \qquad \mu_2 = \frac{1}{4} \begin{bmatrix} 9 \\ 10 \end{bmatrix} + \begin{bmatrix} 9 \\ 14 \end{bmatrix} + \begin{bmatrix} 7 \\ 12 \end{bmatrix} + \begin{bmatrix} 11 \\ 12 \end{bmatrix} = \begin{bmatrix} 9 \\ 12 \end{bmatrix}$$

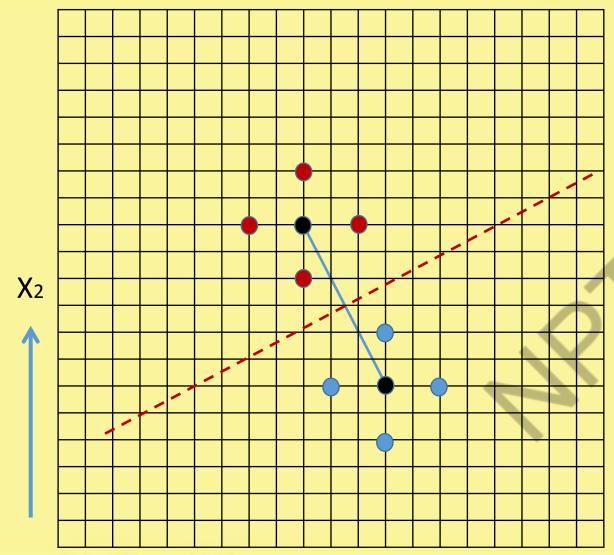
$$\Sigma_2 = 2I$$



$$\Sigma_1 = \Sigma_2 = 2I \approx \sigma^2 I$$
 Where

$$\sigma = \sqrt{2}$$





$$W^t(X - X_0) = 0$$

$$W = \mu_2 - \mu_1$$

$$X_0 = \frac{1}{2}(\mu_1 - \mu_2) - \frac{\sigma^2}{\|\mu_1 - \mu_2\|^2} \ln \frac{P(\omega_1)}{P(\omega_2)}(\mu_1 - \mu_2)$$



# Discriminant Function under Multivariate Normal Distribution











Thank you







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#### **Topic**

**Lecture 08: Discriminant Function and Decision Surface - III** 

#### **CONCEPTS COVERED**

**Concepts Covered:** 

☐ Decision Boundary under Various Cases of

**Covariance Matrices** 

**□**Examples

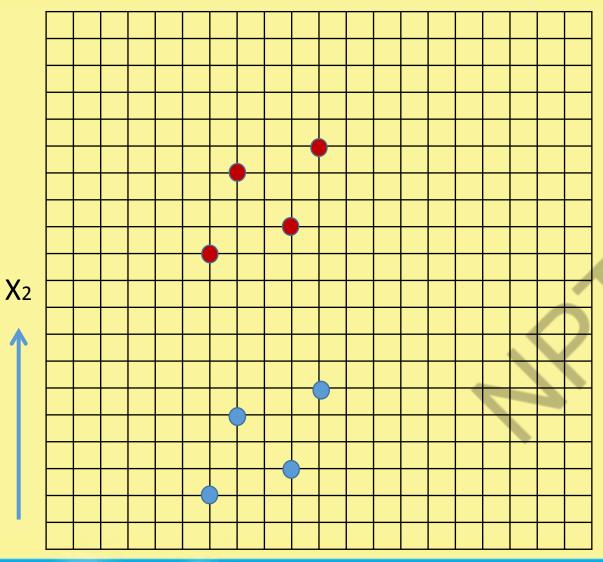




# Discriminant Function under Multivariate Normal Distribution



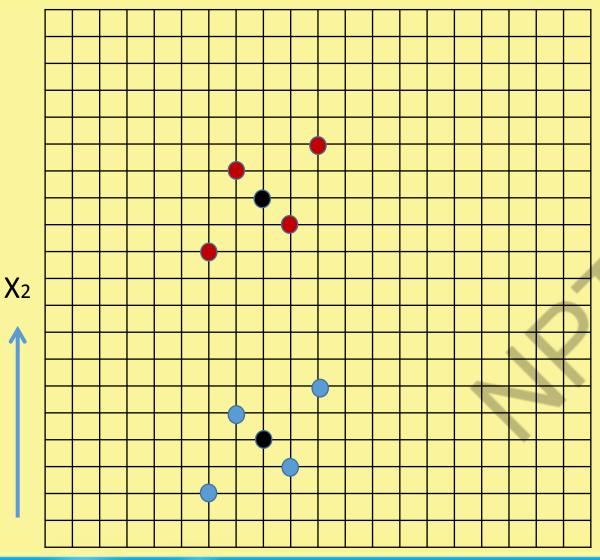




$$\begin{bmatrix} 6 \\ 2 \end{bmatrix} \begin{bmatrix} 9 \\ 3 \end{bmatrix} \begin{bmatrix} 7 \\ 5 \end{bmatrix} \begin{bmatrix} 10 \\ 6 \end{bmatrix} \Rightarrow \omega_1$$

$$\begin{bmatrix} 6 \\ 11 \end{bmatrix} \begin{bmatrix} 9 \\ 12 \end{bmatrix} \begin{bmatrix} 7 \\ 14 \end{bmatrix} \begin{bmatrix} 10 \\ 15 \end{bmatrix} \implies \omega_2$$





$$\begin{bmatrix} 6 \\ 2 \end{bmatrix} \begin{bmatrix} 9 \\ 3 \end{bmatrix} \begin{bmatrix} 7 \\ 5 \end{bmatrix} \begin{bmatrix} 10 \\ 6 \end{bmatrix} \Rightarrow \omega_1$$

$$\begin{bmatrix} 6 \\ 11 \end{bmatrix} \begin{bmatrix} 9 \\ 12 \end{bmatrix} \begin{bmatrix} 7 \\ 14 \end{bmatrix} \begin{bmatrix} 10 \\ 15 \end{bmatrix} \implies \omega_2$$

$$\mu_1 = \begin{bmatrix} 8 \\ 4 \end{bmatrix} \qquad \mu_1 = \begin{bmatrix} 8 \\ 13 \end{bmatrix}$$

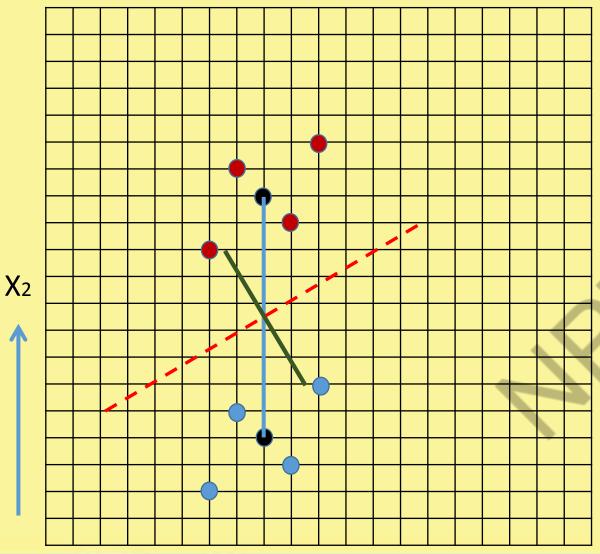


$$\begin{bmatrix} 6 \\ 2 \end{bmatrix} \begin{bmatrix} 9 \\ 3 \end{bmatrix} \begin{bmatrix} 7 \\ 5 \end{bmatrix} \begin{bmatrix} 10 \\ 6 \end{bmatrix} \Rightarrow \omega_1 \quad \begin{bmatrix} 6 \\ 11 \end{bmatrix} \begin{bmatrix} 9 \\ 12 \end{bmatrix} \begin{bmatrix} 7 \\ 14 \end{bmatrix} \begin{bmatrix} 10 \\ 15 \end{bmatrix} \Rightarrow \omega_2$$

$$\mu_1 = \begin{bmatrix} 8 \\ 4 \end{bmatrix} \qquad \mu_2 = \begin{bmatrix} 8 \\ 13 \end{bmatrix}$$

$$\Sigma = \frac{1}{2} \begin{bmatrix} 5 & 3 \\ 3 & 5 \end{bmatrix} \qquad \Sigma^{-1} = \frac{1}{8} \begin{bmatrix} 5 & -3 \\ -3 & 5 \end{bmatrix}$$





$$W^t(X - X_0) = 0$$

$$W = \Sigma^{-1}[\mu_2 - \mu_1] = \frac{1}{8} \begin{bmatrix} 5 & -3 \\ -3 & 5 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} -3 \\ 5 \end{bmatrix}$$

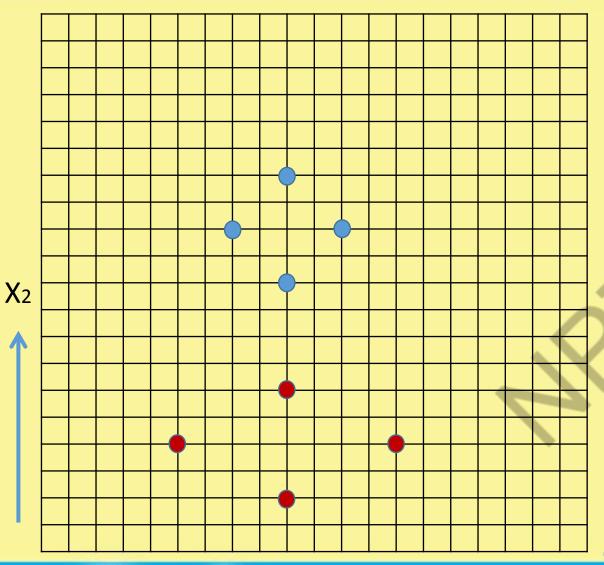
$$X_0 = \frac{1}{2}(\mu_1 + \mu_2) - \frac{1}{(\mu_1 - \mu_2)^t \Sigma^{-1}(\mu_1 - \mu_2)} \ln \frac{P(\omega_1)}{P(\omega_2)}(\mu_1 - \mu_2)$$



# Discriminant Function under Multivariate Normal Distribution



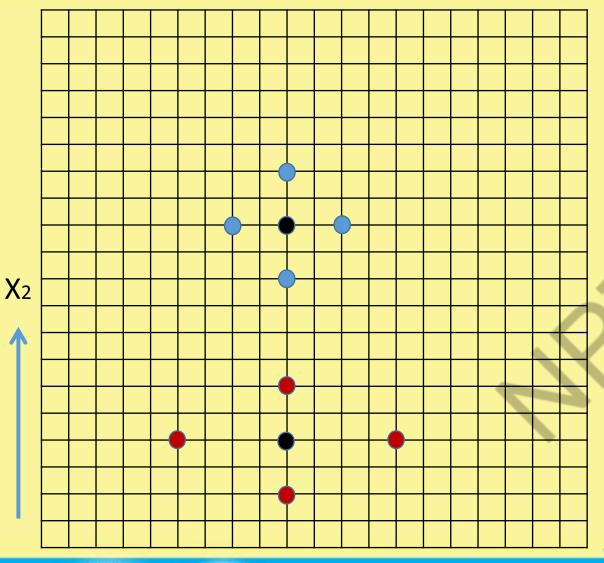




$$\begin{bmatrix} 5 \\ 4 \end{bmatrix} \begin{bmatrix} 9 \\ 2 \end{bmatrix} \begin{bmatrix} 9 \\ 6 \end{bmatrix} \begin{bmatrix} 13 \\ 6 \end{bmatrix} \Rightarrow \omega_1$$

$$\begin{bmatrix} 9 \\ 10 \end{bmatrix} \begin{bmatrix} 9 \\ 14 \end{bmatrix} \begin{bmatrix} 7 \\ 12 \end{bmatrix} \begin{bmatrix} 11 \\ 12 \end{bmatrix} \implies \omega_2$$





$$\mu_1 = \begin{bmatrix} 12 \\ 6 \end{bmatrix} \qquad \mu_2 = \begin{bmatrix} 9 \\ 4 \end{bmatrix}$$

$$\Sigma_1 = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

$$\Sigma_2 = \begin{bmatrix} 8 & 0 \\ 0 & 2 \end{bmatrix}$$

 $X_1$ 



### Discriminant Function

$$g_{i}(X) = -\frac{d}{2}\ln 2\pi - \frac{1}{2}\ln \left|\Sigma_{i}\right| - \frac{1}{2}\left[(X - \mu_{i})^{t}\Sigma_{i}^{-1}(X - \mu_{i})\right] + \ln P(\omega_{i})$$

$$= X^{t}A_{i}X + B_{i}^{t}X + C_{i}$$

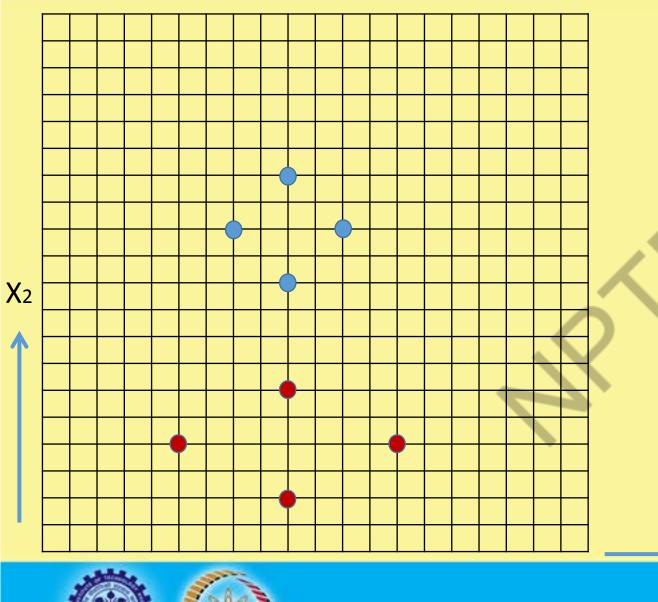
$$A_i = -\frac{1}{2} \Sigma_i^{-1}$$

$$B_i = \Sigma_i^{-1} \mu_i$$

$$C_i = -\frac{1}{2}\mu_i^t \Sigma_i^{-1} \mu_i - \frac{1}{2} \ln |\Sigma_i| + \ln P(\omega_i)$$







 $X_1$ 









Thank you







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**Topic** 

**Lecture 09: Linear Classifier** 

#### **CONCEPTS COVERED**

#### **Concepts Covered:**

- ☐ Discriminant Function and Decision Boundary
- ☐ Nearest Neighbour and k-NN Classifier
- ☐ Linear Classifier
- ■Support Vector Machine (SVM)

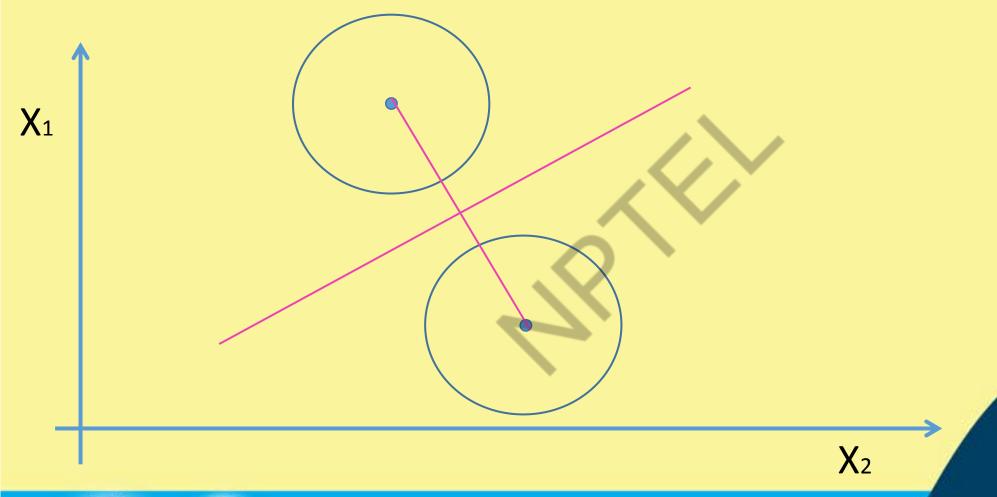




## Nearest Neighbour Rule

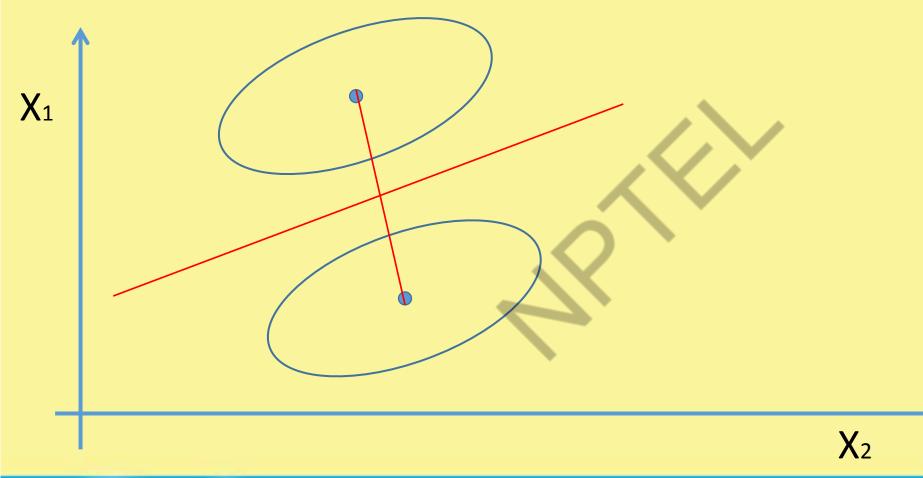


## Minimum Distance Classifier





## Minimum Distance Classifier





## Nearest Neighbour Rule

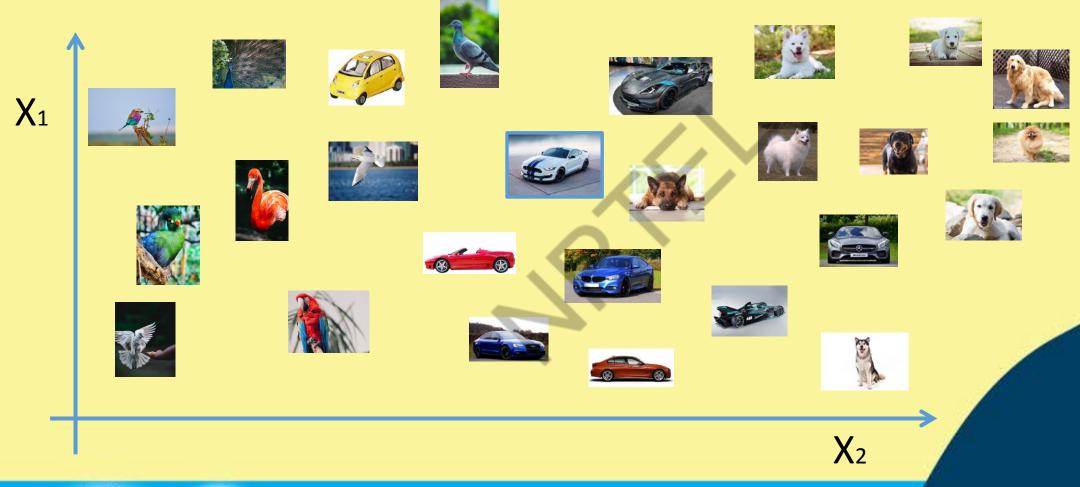




Image Source: Internet

## Nearest Neighbour Rule

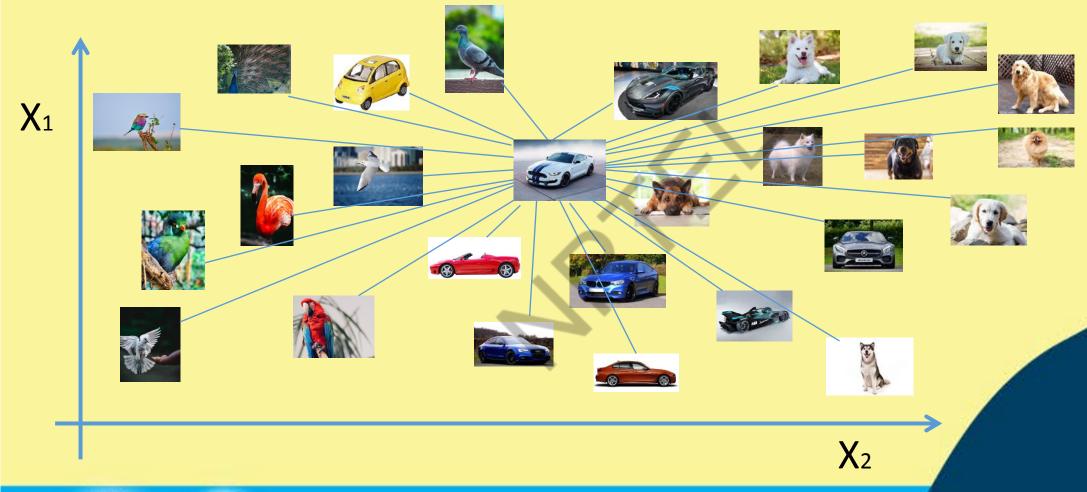




Image Source: Internet

## Nearest Neighbour Rule

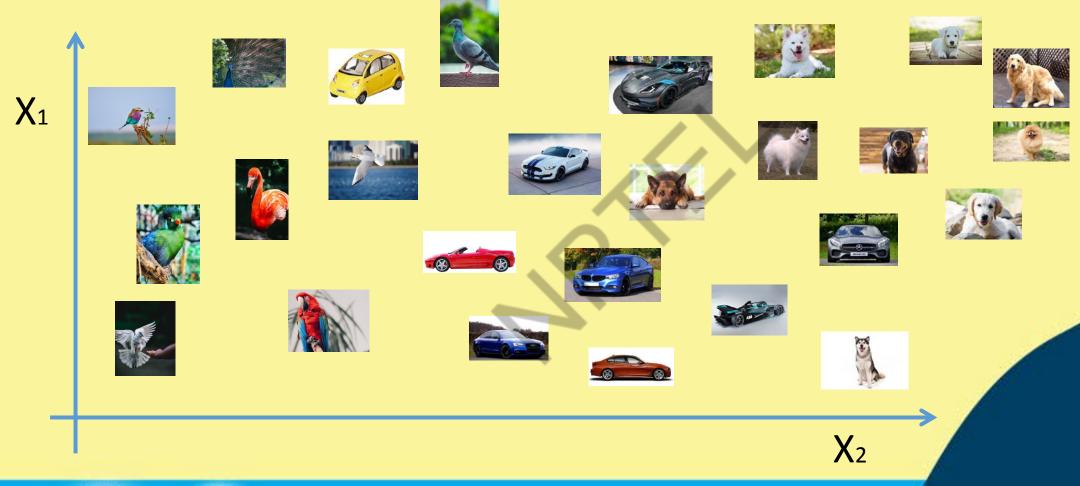
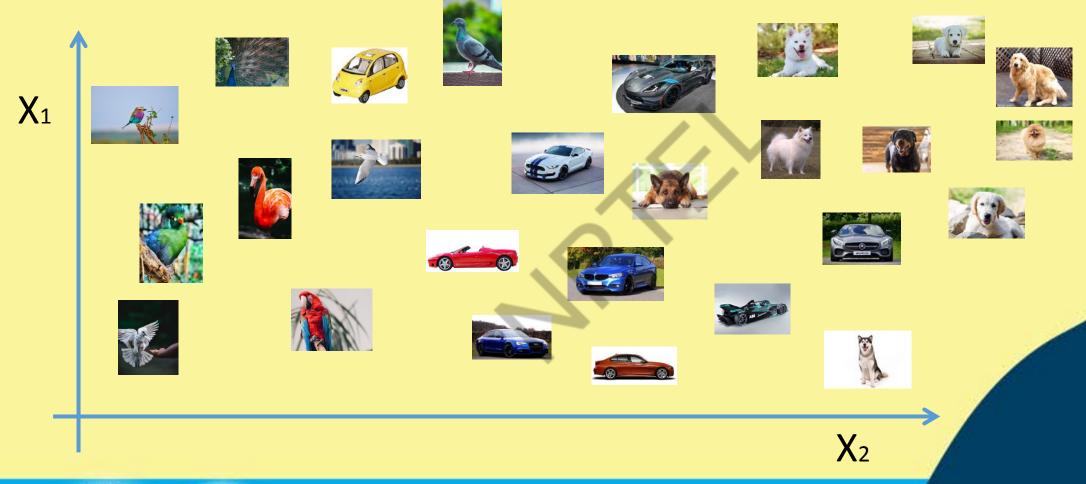




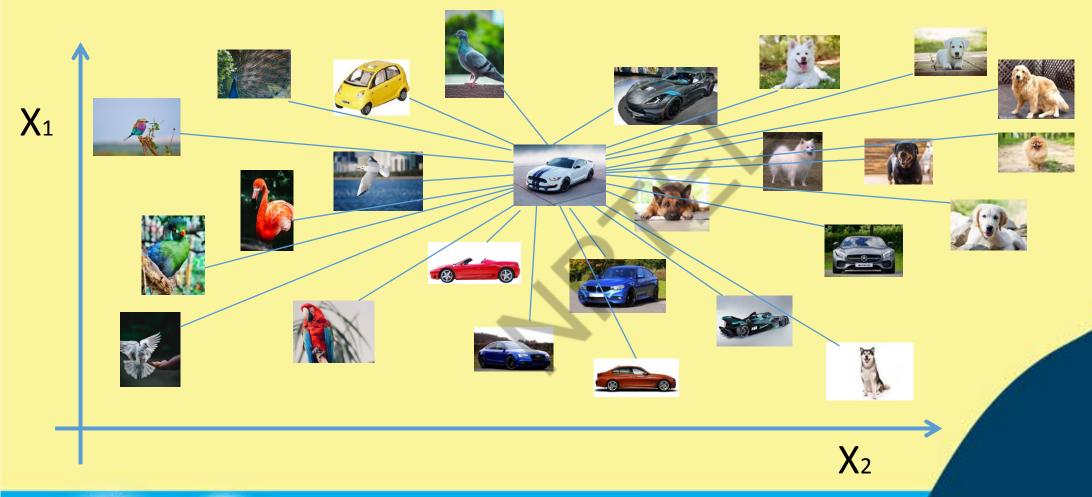
Image Source: Internet

# k-Nearest Neighbour Rule

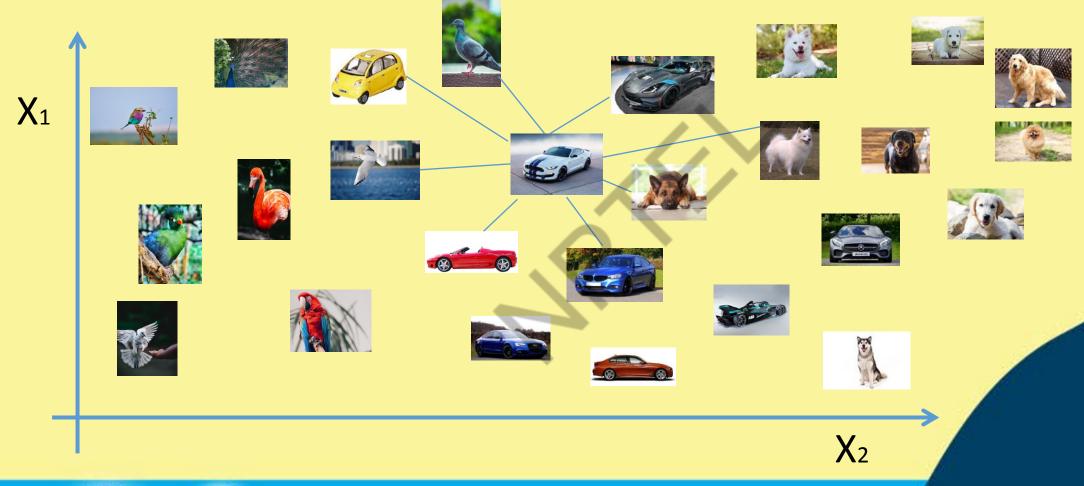




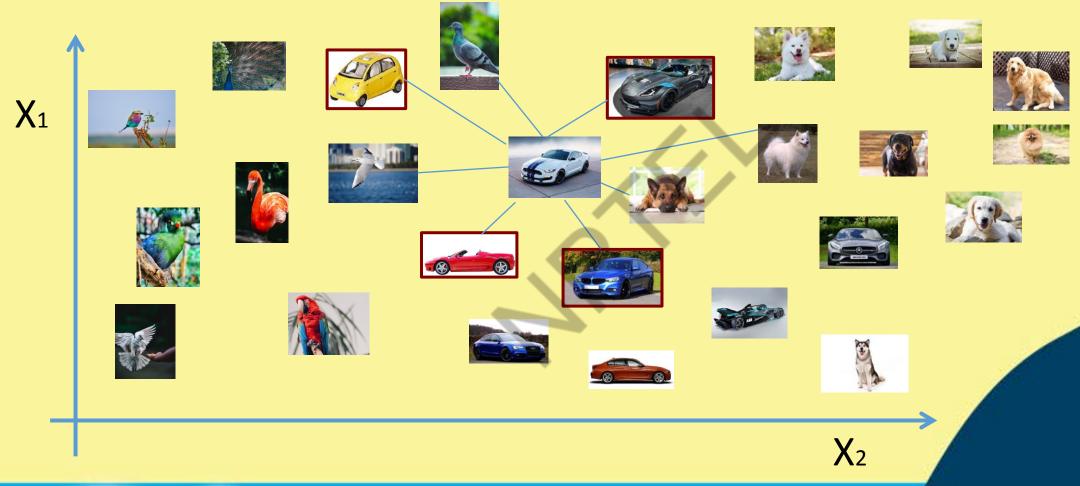




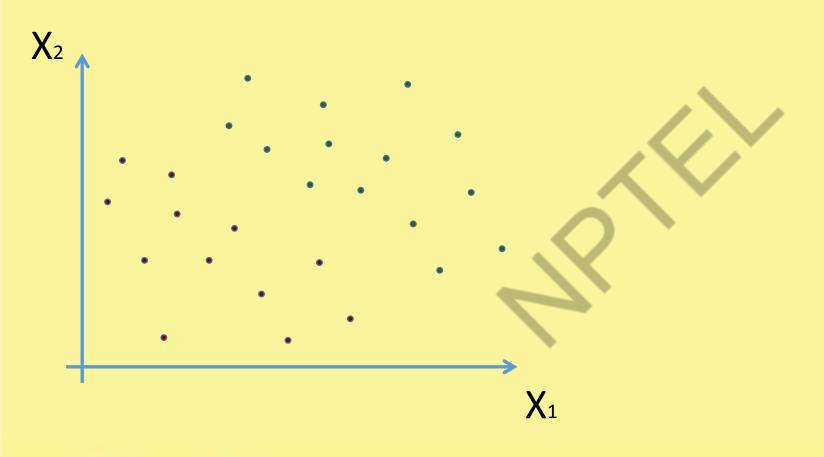




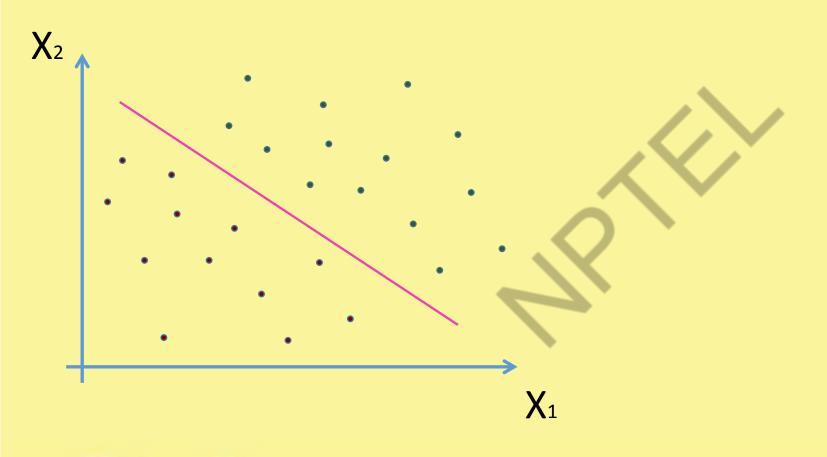








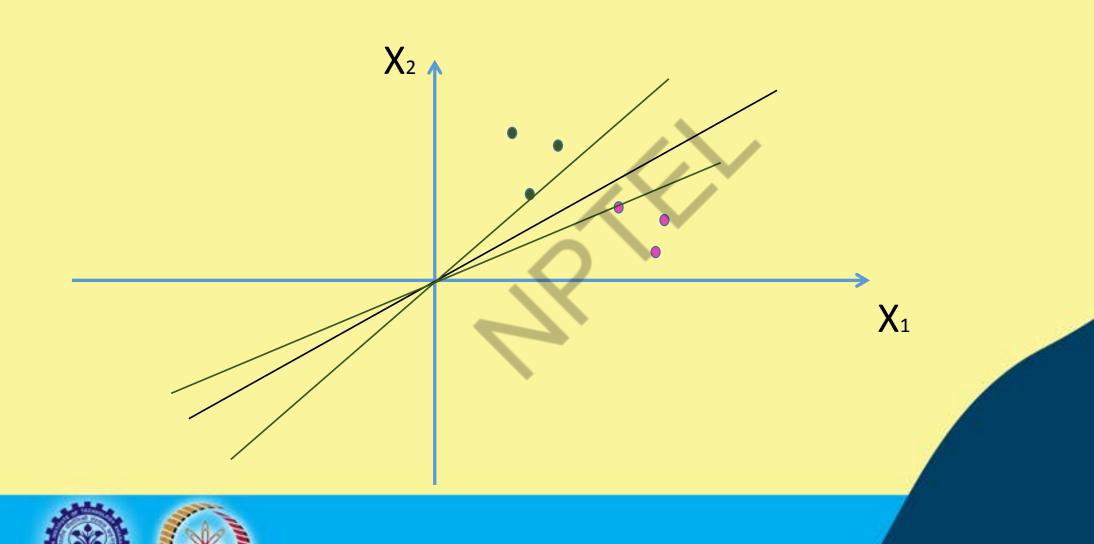


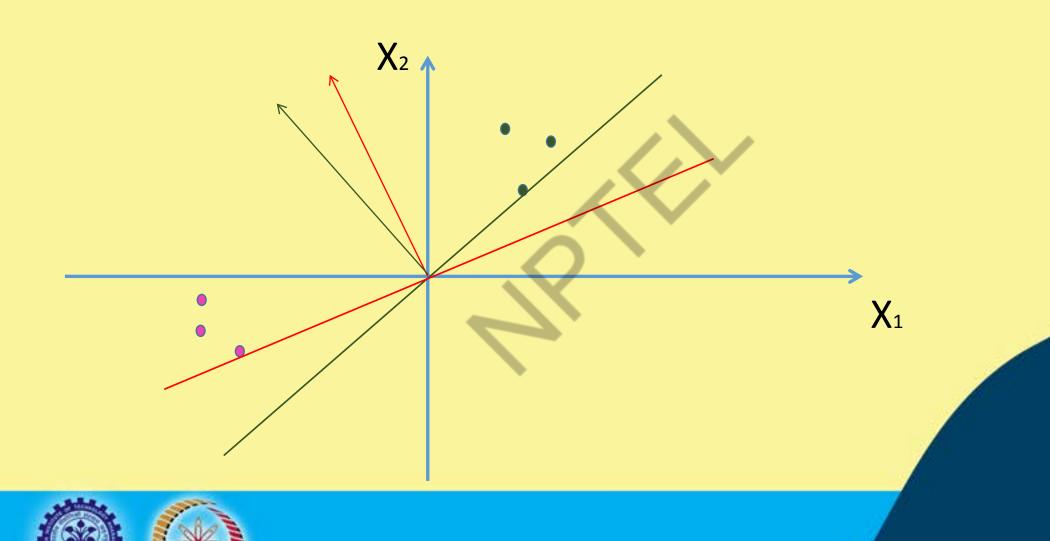






















#### NPTEL ONLINE CERTIFICATION COURSES

Thank you







#### **NPTEL ONLINE CERTIFICATION COURSES**

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**Department: E & ECE, IIT Kharagpur** 

**Topic** 

**Lecture 10: Linear Classifier** 

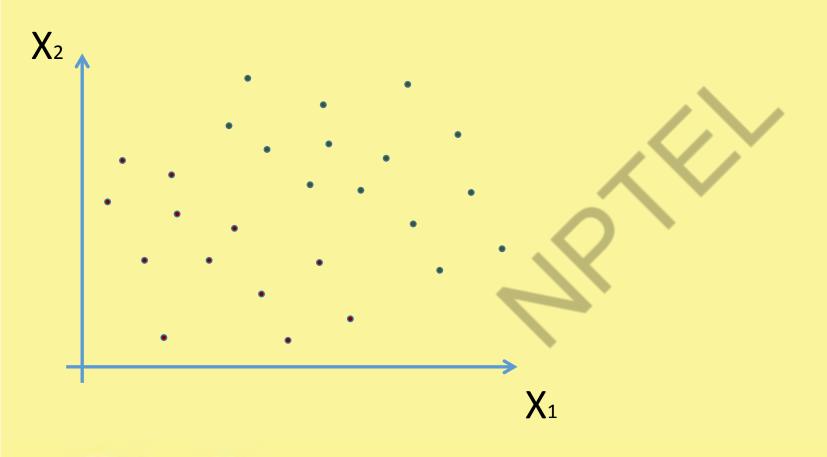
#### **CONCEPTS COVERED**

#### **Concepts Covered:**

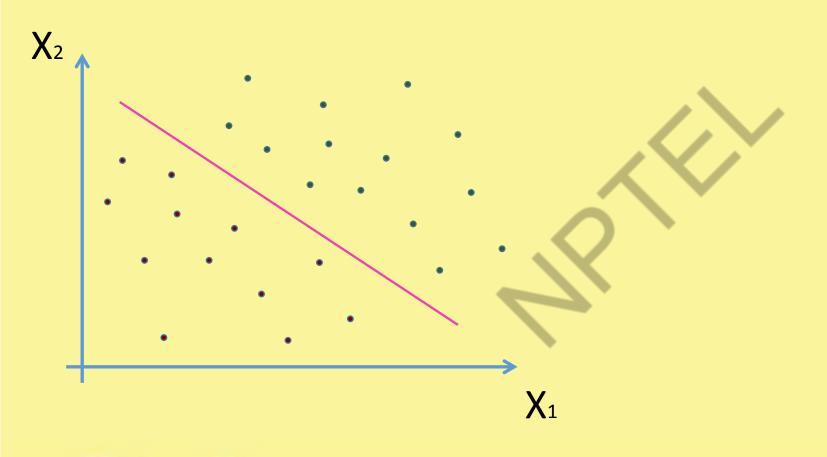
- ☐ Discriminant Function and Decision Boundary
- ☐ Nearest Neighbour and k-NN Classifier
- ☐ Linear Classifier
- ■Support Vector Machine (SVM)







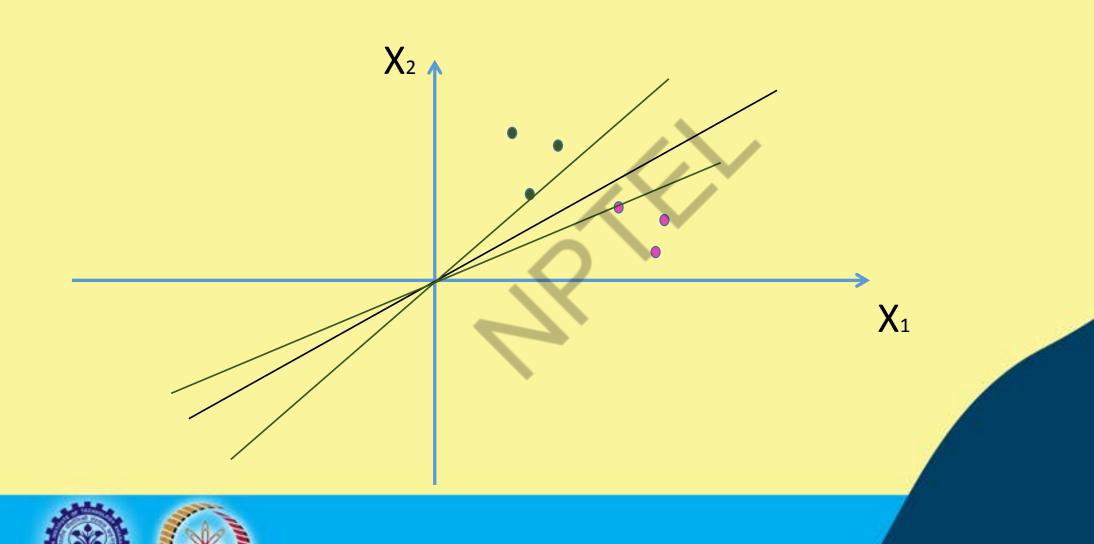


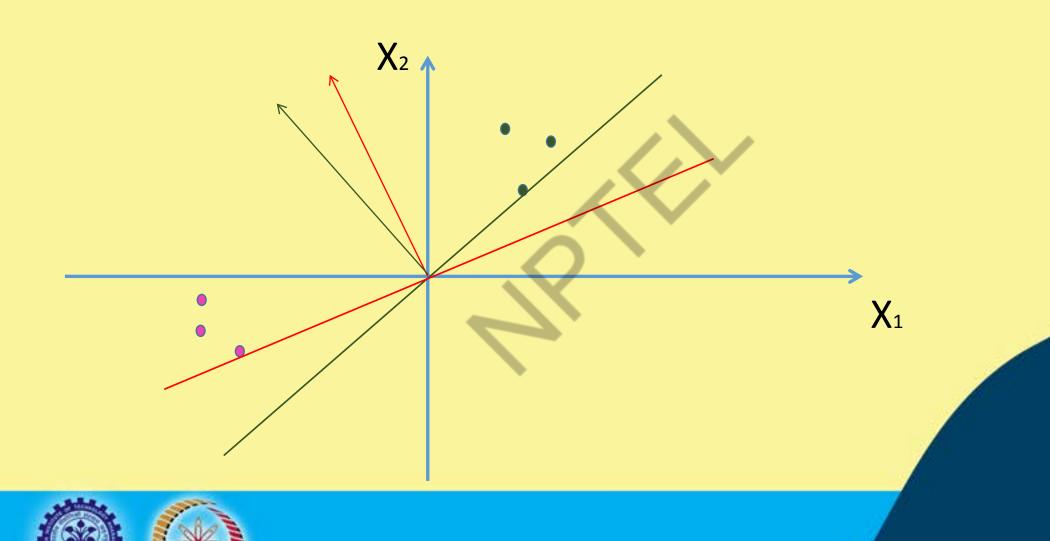






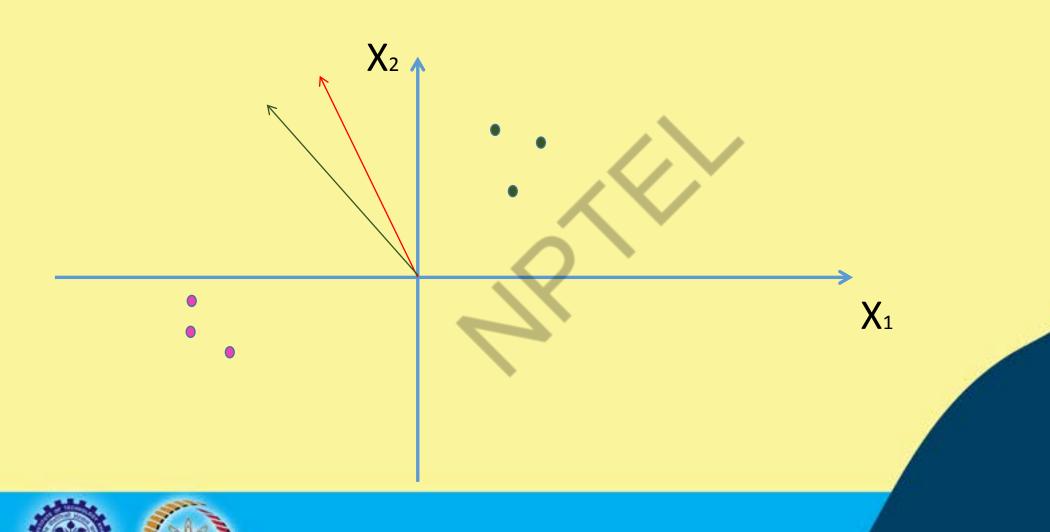


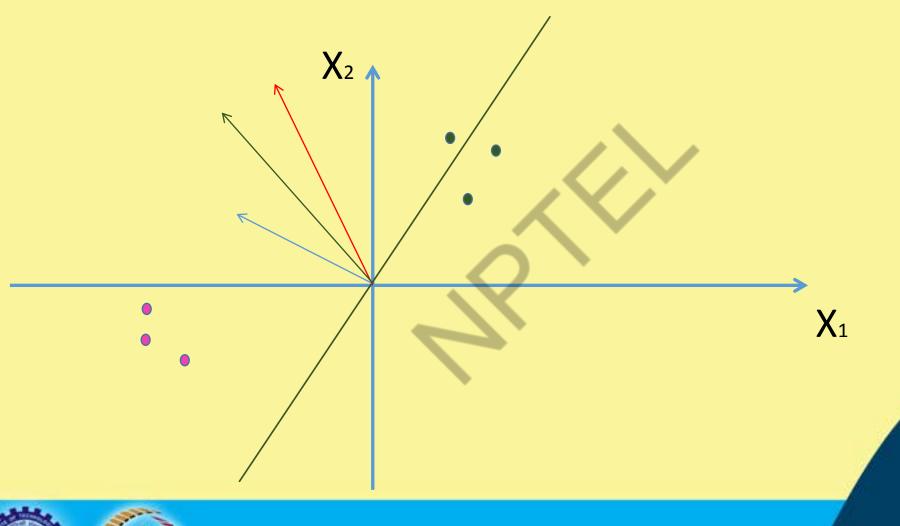




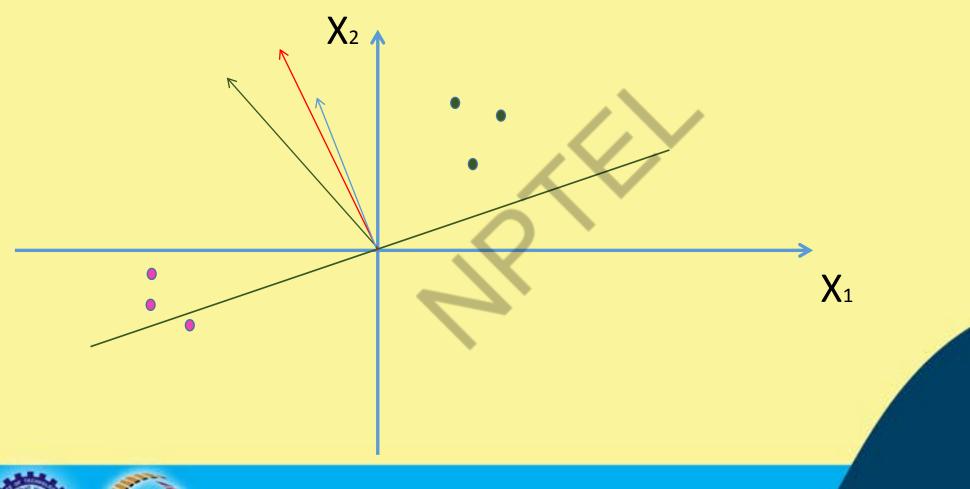




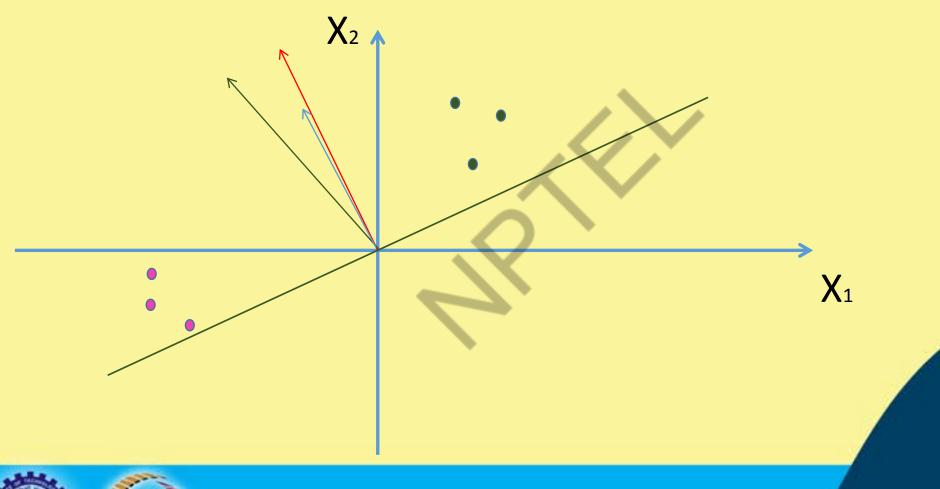




















#### NPTEL ONLINE CERTIFICATION COURSES

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