

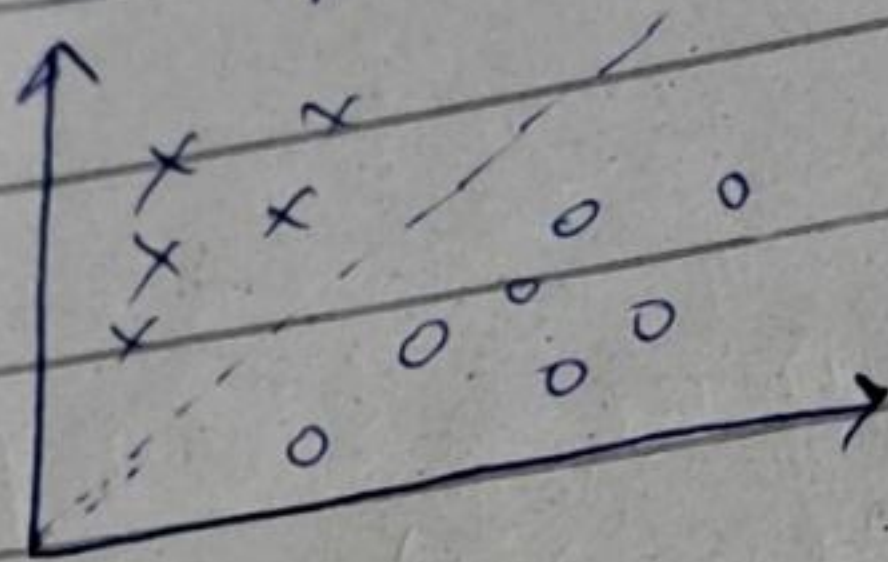
Support Vector Machine

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SVM:-

SVM is the versatile machine learning algorithm, supervised learning algorithm. It is used for both regression and classification tasks.

→ finding the optimal hyperplane.



Hyperplane:-

It is a plane that linearly divide the n -dimensional data points in two components.

Linear SVM:-

find a hyperplane defined by:-

$$w \cdot x + b = 0$$

where,

→ $w \in \mathbb{R}^d$ is the weight vector

$b \in \mathbb{R}$ is the bias

Decision Rule,

$$f(x) = \text{Sign}(w \cdot x + b)$$

Margin:-

The distance between hyperplane and the closest data points.

classified point (x_i, y_i) , margin constraint is

$$y_i(w \cdot x_i + b) \geq 1$$

The maximum margin is achieved by minimizing $\|w\|$.

Optimization Problem:-

$$\min_{w, b} \frac{1}{2} \|w\|^2 \quad \text{Subject to } y_i(w \cdot x_i + b) \geq 1$$

Non-linear SVM:-

Kernel functions,

$$k(x_i, x_j) = \phi(x_i) \cdot \phi(x_j)$$

functions,

$$\rightarrow \text{Linear } k(x_i, x_j) = x_i \cdot x_j$$

$$\rightarrow \text{Polynomial } : k(x_i, x_j) = (x_i \cdot x_j + C)^d$$

$$\rightarrow \text{Radial Basis function } : k(x_i, x_j) = \exp(-\gamma \|x_i - x_j\|)$$

$$\rightarrow f(x) = \text{Sign}\left(\sum_{i=1}^n \alpha_i y_i k(x_i, x) + b\right)$$