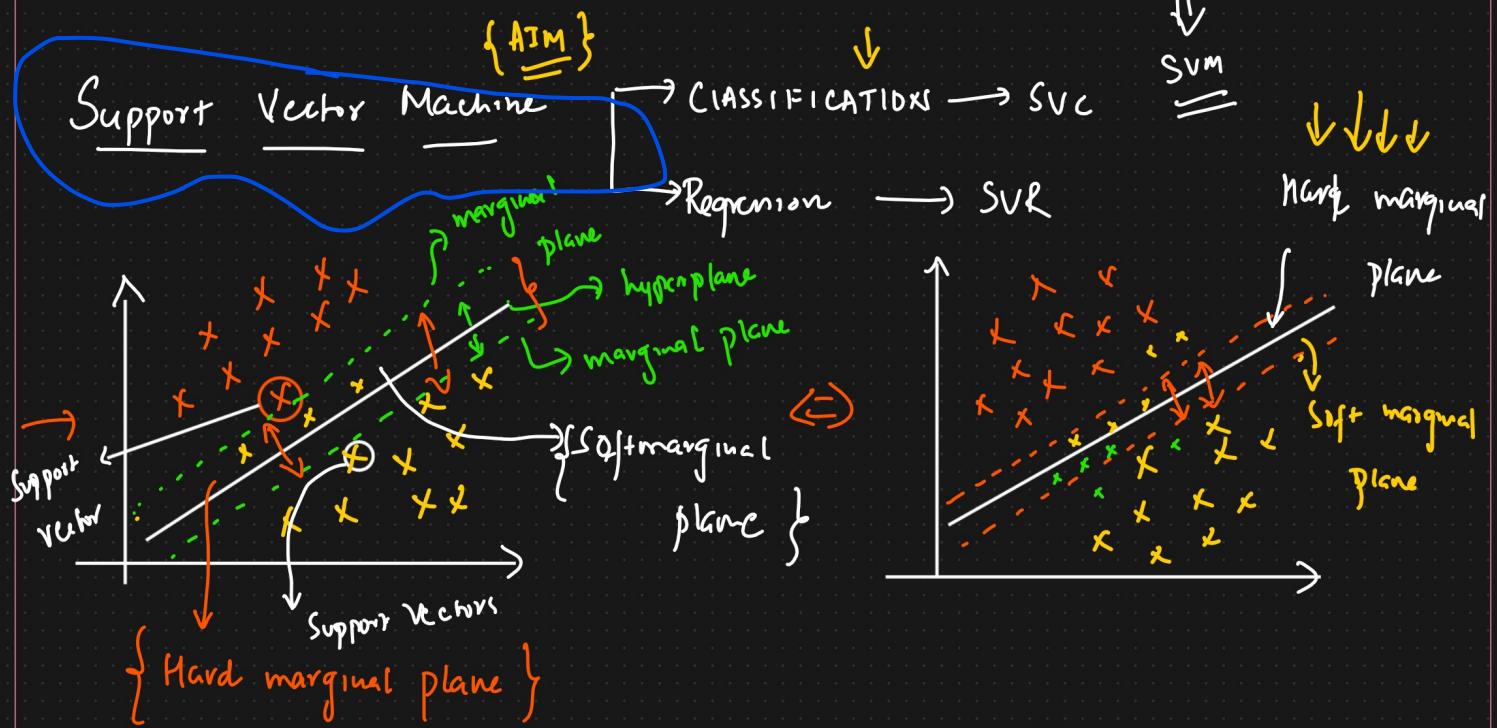


# SVM AND SVR Algorithms:-

## ① CLASSIFICATION AND REGRESSION PROBLEMS .



$$y = mx + c$$

$$y = \beta_0 + \beta_1 x$$

$$y = w^T x + b$$

$ax + by + c = 0$  → Equation of a straight line

$by = -ax - c$  → coefficient

$y = -\frac{a}{b}x - \frac{c}{b}$  → intercept

$$y = w_1 x_1 + w_2 x_2 + w_3 x_3 + b$$

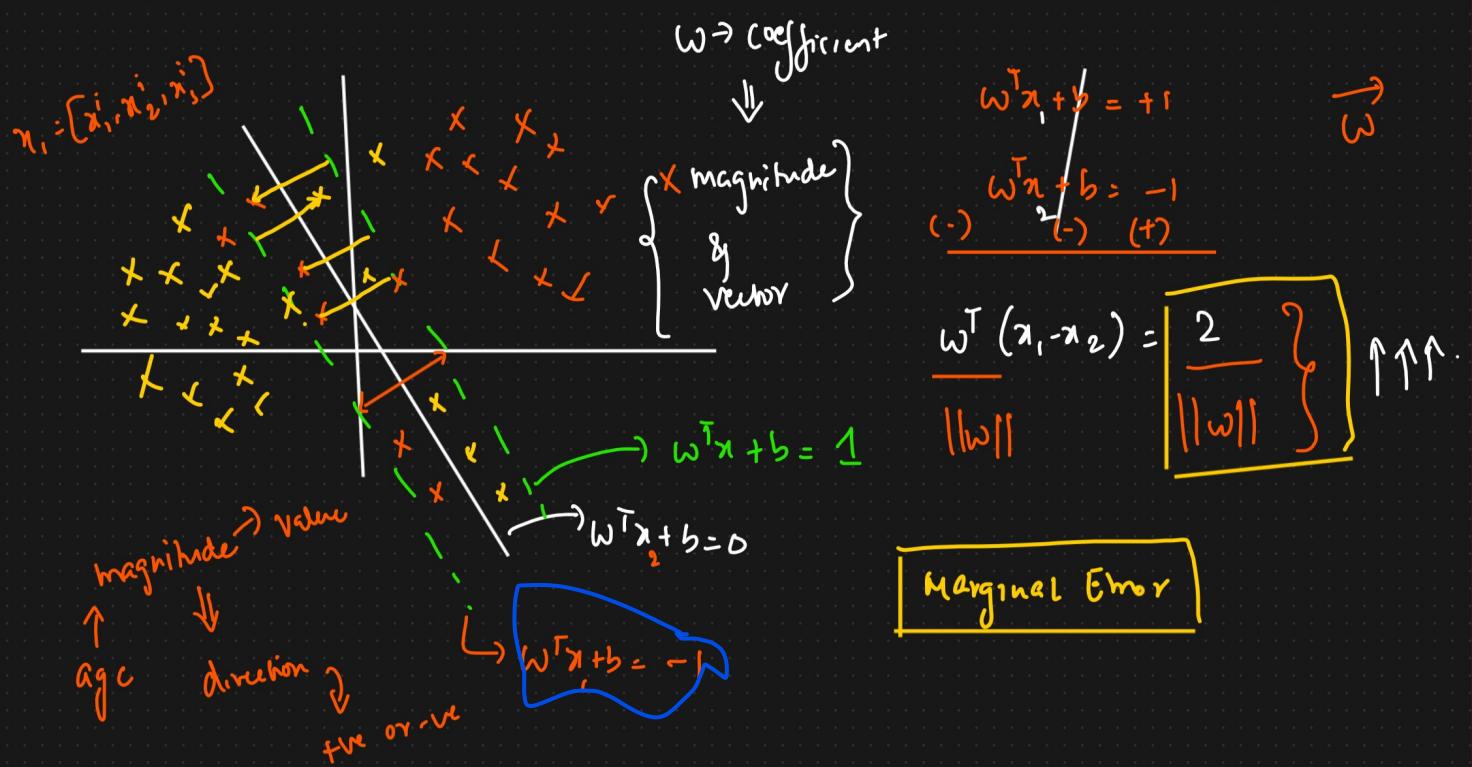
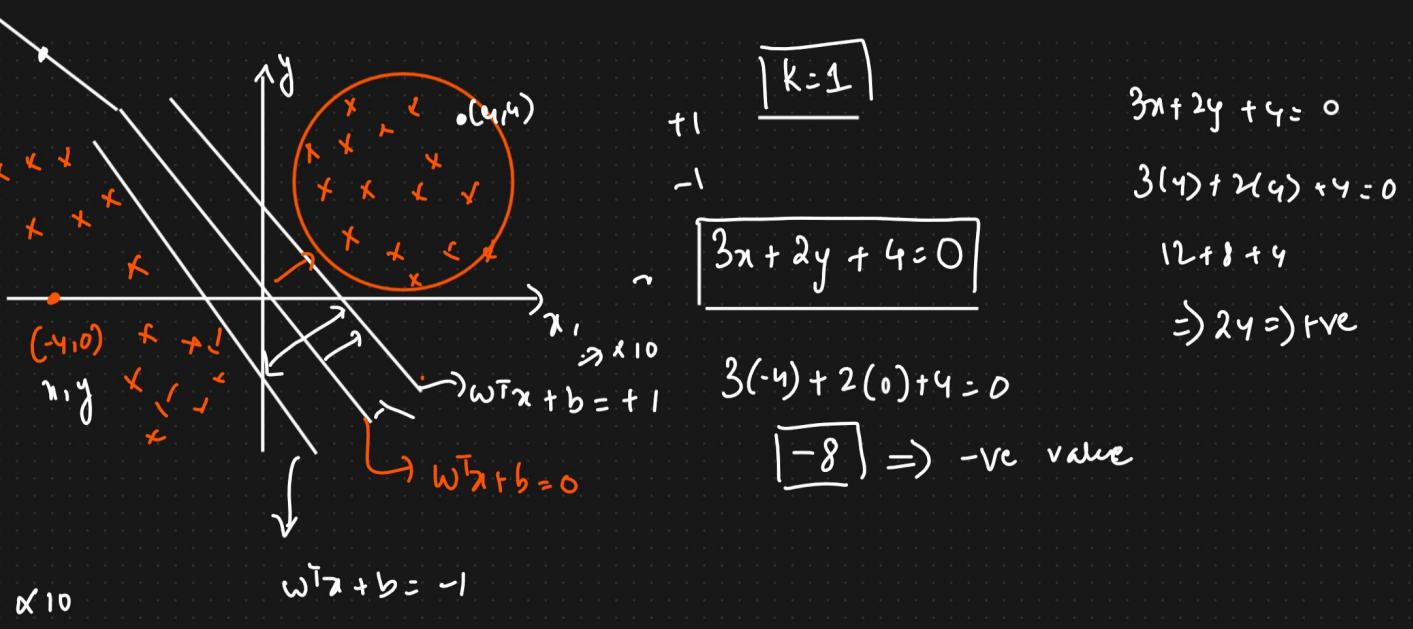
$$y = w^T x + b$$

$$\begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix}$$

$$y = m_1 x_1 + m_2 x_2 + m_3 x_3 + c$$

$$y = m^T x + c$$

$$\begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix} \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix}$$



Maximize  $(w, b)$   
Constraint  
 $\frac{2}{||w||} \Rightarrow \text{Marginal plane } \checkmark$   
 Such that  
 $y_j \begin{cases} +1 \\ = \\ -1 \\ z \end{cases}$   
 $w^T x_i + b \geq 1 \Rightarrow \text{correct}$   
 $w^T x_i + b = \Rightarrow \text{=} =$   
 $w^T x_i + b \leq -1 \Rightarrow \text{fwe} =$

For all correct points

Constraint  $\rightarrow$

$$y_i * (\mathbf{w}^T \mathbf{x}_i + b) \geq 1$$

$$\text{Maximize}_{(\mathbf{w}, b)} \frac{2}{\|\mathbf{w}\|}$$

$\Leftrightarrow$

$$\text{Min}_{(\mathbf{w}, b)} \frac{\|\mathbf{w}\|}{2}$$

Linear Regression

Cost function:

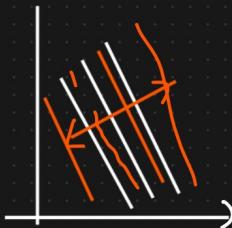
$$\text{Min}_{(\mathbf{w}, b)} \frac{\|\mathbf{w}\|}{2}$$

$$+ C_j \sum_{i=1}^n \xi_i$$

Hinge Loss

Summation of the  
distance of the

wrong data points  
from the marginal  
plane



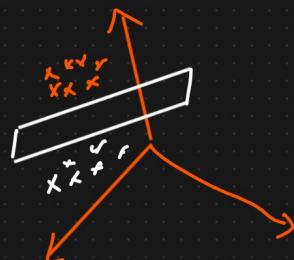
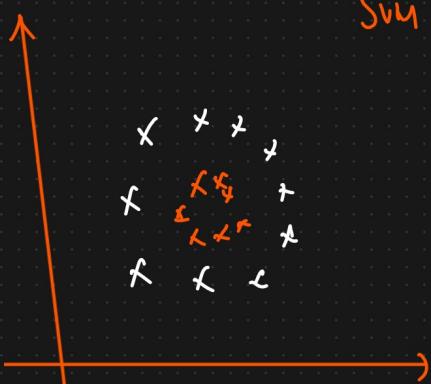
Hard marginal  
plane

How many  
points we  
want to avoid  
misclassification.

SVM

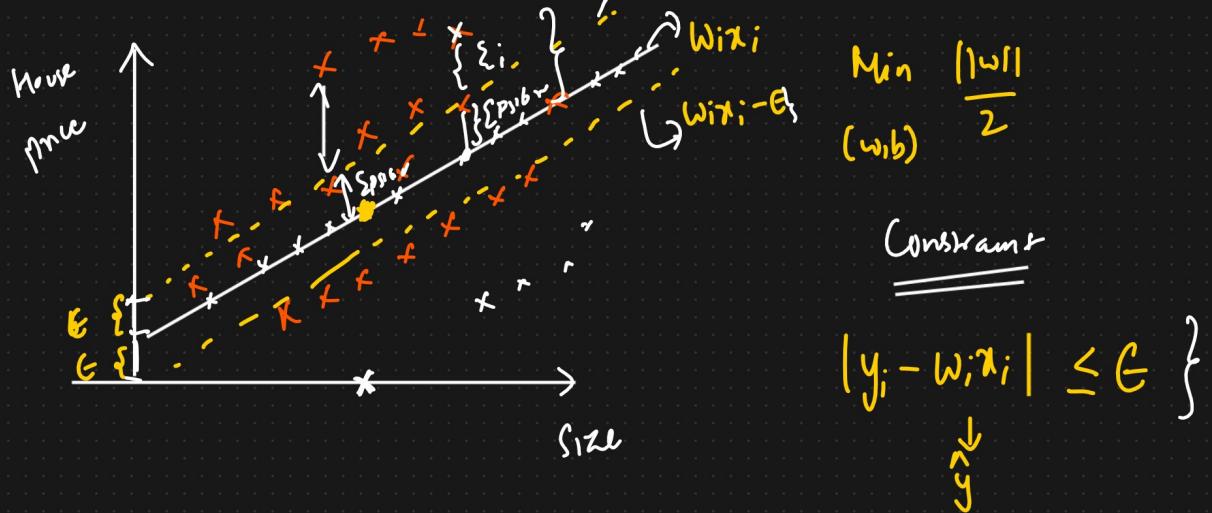
SVM Kernels

(kernel Trick)



## ④ Support Vector Regression

$$\begin{aligned} w_i x_i + b &= +1 \\ w_i x_i + b &= -1 \\ \Rightarrow w_i x_i + \epsilon \end{aligned}$$



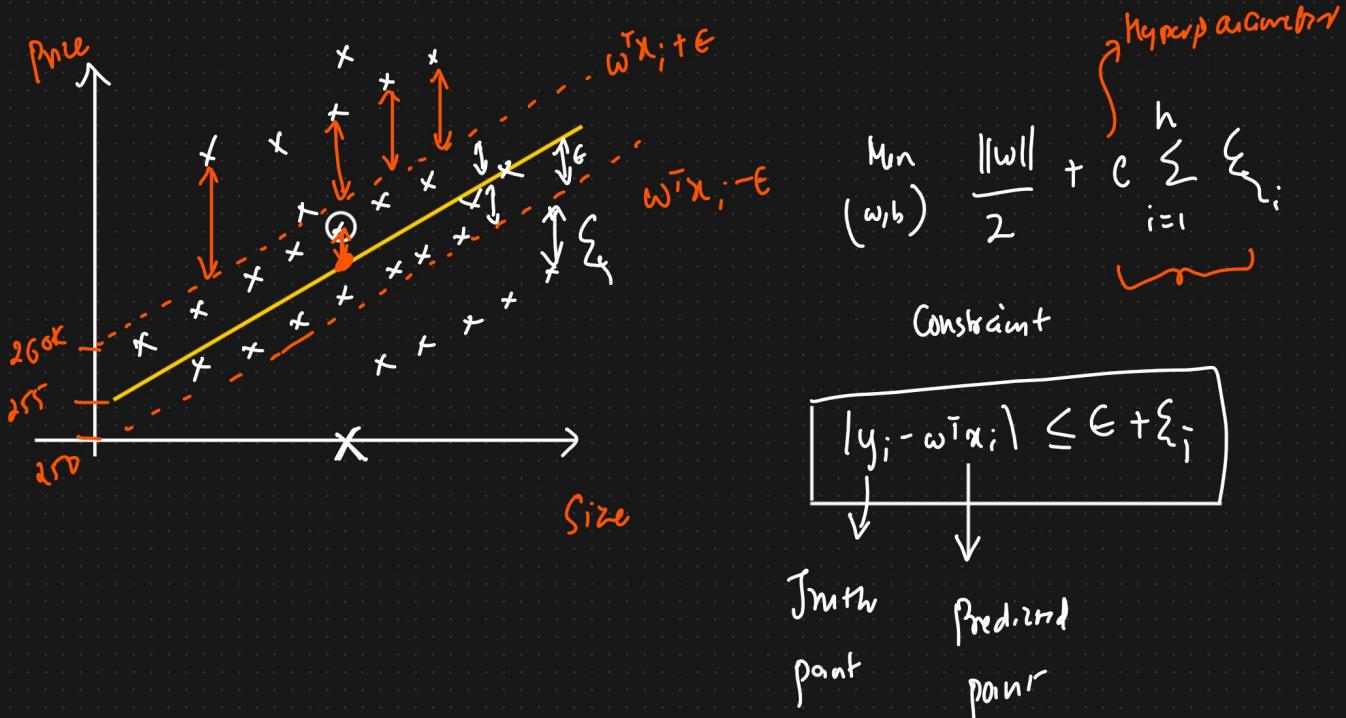
Cost function

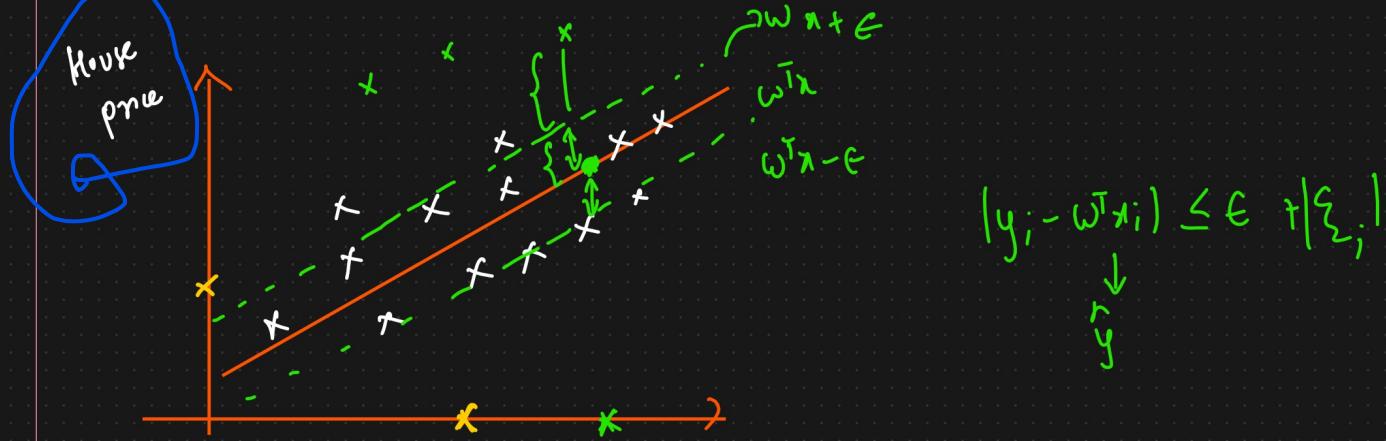
$$\underset{(w,b)}{\text{Min}} \frac{\|w\|}{2} + C \sum_{i=1}^n |\xi_i|$$

Constraint

$$|y_i - w^T x_i| \leq \epsilon + \xi_i$$

distance of the points that  
are away from the  
Epsilon line





$$(y_i - \hat{w}^T x_i) \leq \epsilon + \xi_i$$

$\downarrow$   
 $y$

$$\left\{ C \text{ & Slope} \right\}$$

$$\text{Cost function} = \frac{1}{2n} \sum_{i=1}^n (y_i - \hat{y}_i)^2 + \lambda \sum_{i=1}^n (\text{Slope})^2$$

Actual Value       $\hat{y}_i$  predicted value       $\lambda \sum_{i=1}^n (\text{Slope})^2$

$\left. \begin{array}{c} \curvearrowleft \\ \curvearrowright \end{array} \right\}$  Regularization parameters       $L_2, L_1$



$$\left\{ \alpha_2, \alpha_1, \text{Elastic Net, Hinge Loss} \right\}$$

Aim: Relationship between  $C \& \lambda$  = Logistic Sigmoid

$$\boxed{C \propto \lambda}$$

Diff

Loss & Cost function

1 - datapoint

$$\text{Loss fn} = \frac{1}{2} (y - \hat{y})^2$$

$(\hat{y} - y)^2$   
 $(y - \hat{y})^2$

Entire dataset.

$$\text{Cost funct} : \frac{1}{2n} \sum_{i=1}^n (y - \hat{y})^2$$

fit & fit-transform } transform.

Standard scalar



Transforming the data??  
↓

$f_2$

$\underline{f_1}$

$f_1$ -standard

Changing the data.

$$\begin{array}{l} \text{Train} \\ \left\{ \begin{array}{l} 2 \\ 3 \\ 4 \\ \hline 5 \end{array} \right. \\ \text{Test} \\ \left\{ \begin{array}{l} 6 \end{array} \right. \end{array}$$

$$\begin{array}{l} \left\{ \begin{array}{l} 2 \\ 3 \\ 4 \end{array} \right. \\ \hline \left\{ \begin{array}{l} 5 \\ 6 \end{array} \right. \end{array}$$

fit & transform.

fit & fit-transform

$$Z\text{-Score} = \frac{Y_i - \mu}{\sigma} = 3$$

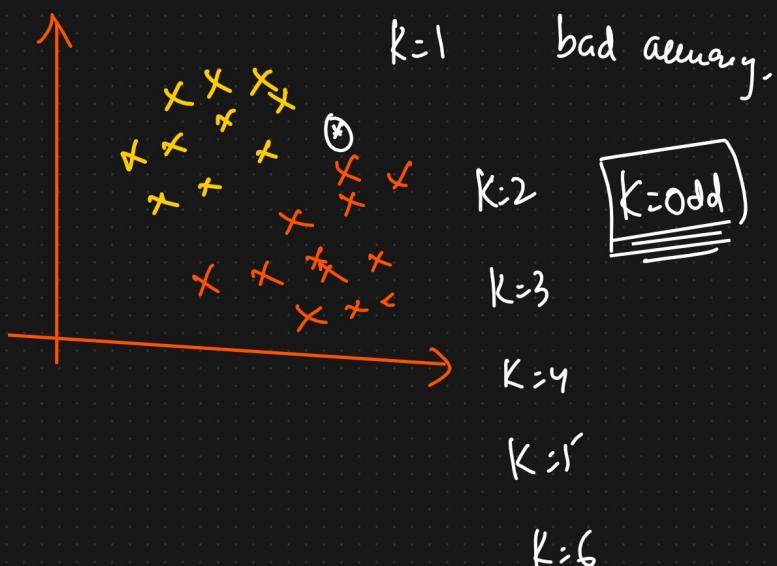
$$\frac{2+3+4}{3} = \frac{9}{3} = \sqrt{3}$$

Transform

→  
 $T_{clf}$

$$\left\{ \begin{array}{l} 5 \\ 6 \end{array} \right.$$

$$\frac{5-3}{1} = 2 \stackrel{\text{No}}{=}$$



$K \uparrow$  Accuracy ↑  
Error ↓

