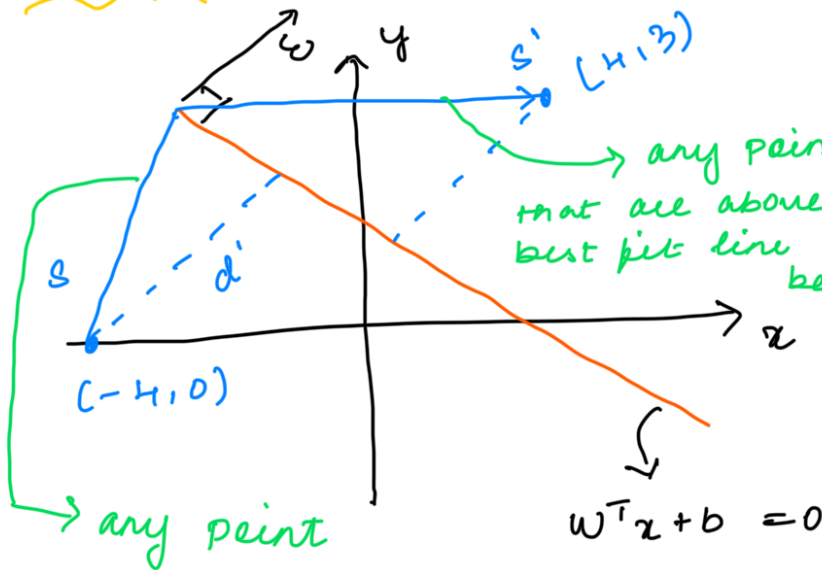


Support Vector Machine (SVC) Math Intuition



equation of a straight line

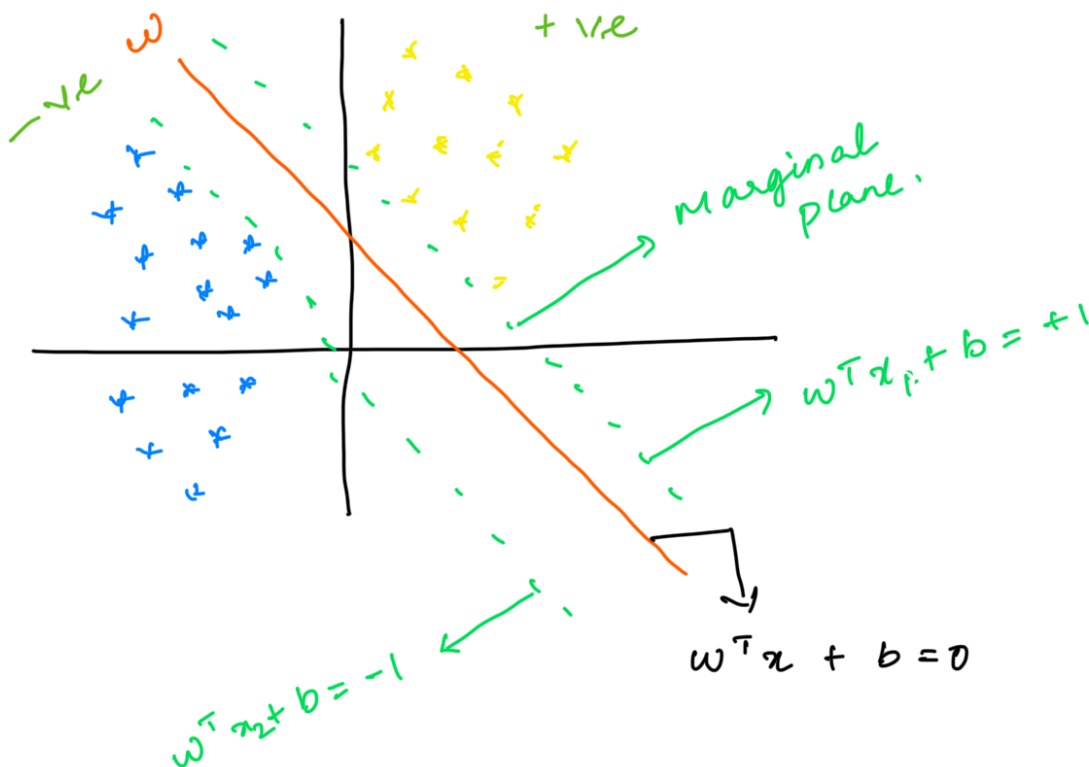
$$ax + by + c = 0$$

$$w_1 x_1 + w_2 x_2 + b = 0$$

$$w^T x + b = 0$$

* distance will be -ve below the plane

* distance will be +ve above the plane.



therefore,

$$\begin{array}{rcl} w^T x_1 + b & = & 1 \\ w^T x_2 + b & = & -1 \\ \hline (-) & & (+) \\ w^T (x_1 - x_2) & = & +2 \end{array}$$

(solving linear equation)

Unit Vector

Magnitude of vector is 1

Cost function

Maximize $\frac{2}{||w||}$ \Rightarrow Distance between marginal plane

dealing with +ve points
↑

constraint

such that

$$y_i \begin{cases} +1 & \text{if } \underline{w^T x + b} \geq 1 \\ & \text{truth value} \\ -1 & \text{if } w^T x + b \leq -1 \end{cases}$$

↓
dealing with -ve points

⇕

These are applicable only for correctly classified points

For all correct points

constraint \rightarrow

$$y_i (w^T x + b) \geq 1$$

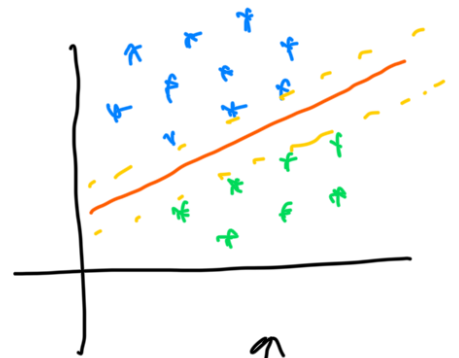
$$\text{maximize } \frac{2}{||w||} = \text{minimize } \frac{||w||}{2}$$

w, b $\|w\|$

w, b

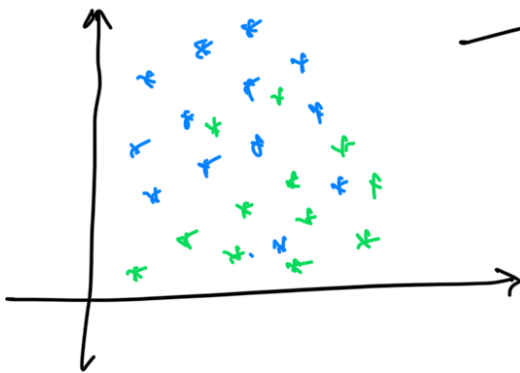
cost function of SVM (SVC)

$$\text{minimize}_{w, b} \frac{\|w\|}{2}$$



I can use this as my cost function, if my data points are clearly separable

however, in the real world that is not the case,



→ there will be data overlap, hence we will have to use it with hyper parameters.

In this case my cost function will be,

$$\text{minimize}_{w, b} \frac{\|w\|}{2} + \boxed{C_i \sum_{i=1}^n h_i} \Rightarrow \text{Hinge Loss}$$

hyper parameters → $C_i \Rightarrow$ how many point we want to avoid misclassification.

$h_i \Rightarrow$ summation of the distance of the incorrect data points from

of the marginal plane.

* This cost function is related to left margin.