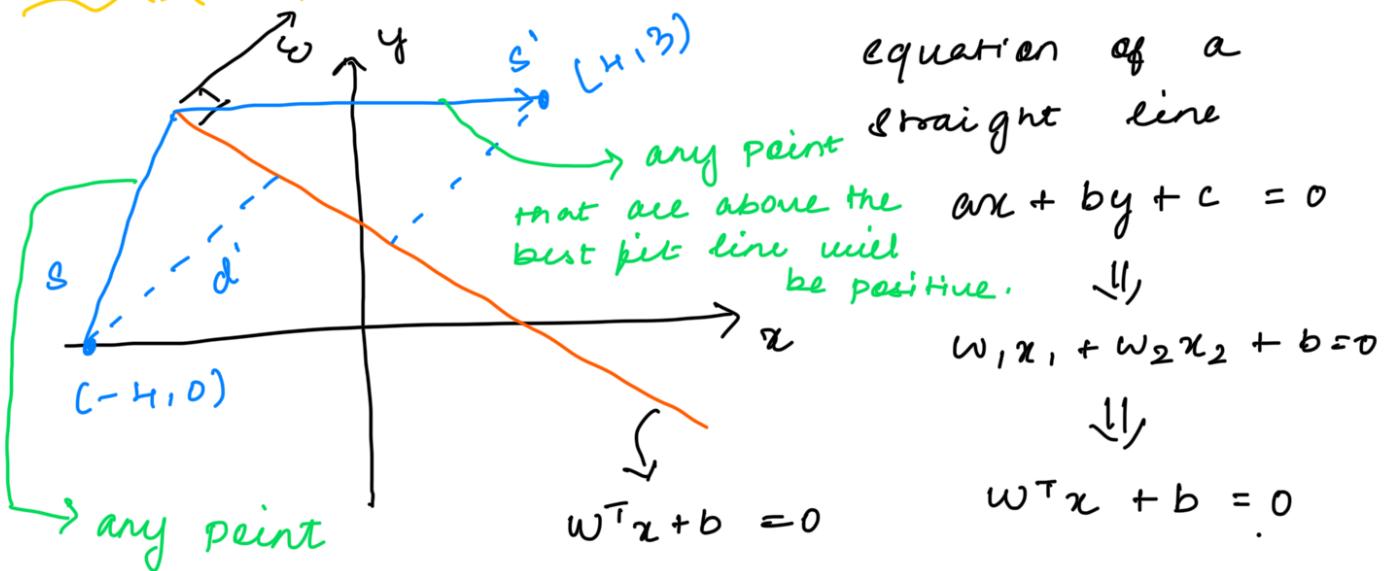


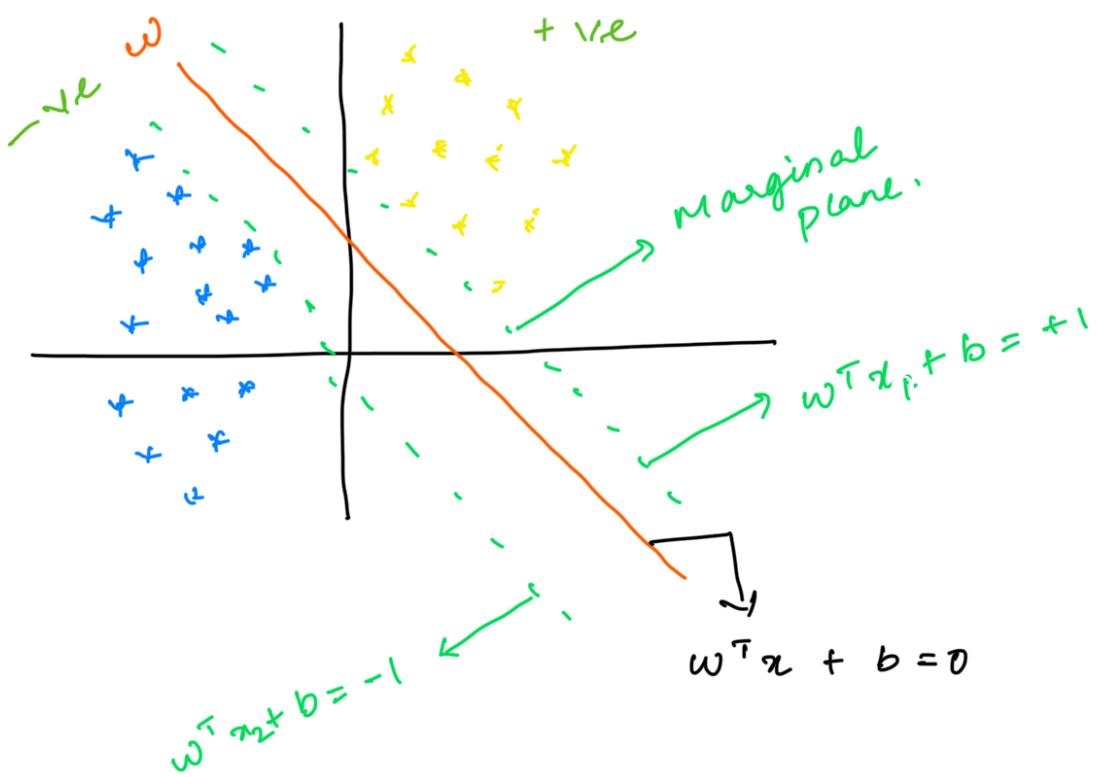
Support Vector Machine (SVC) Math Intuition



any point below the best fit line will be negative.

* distance will be -ve below the plane

* distance will be +ve above the plane.



therefore,

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

Unit Vector

Magnitude of vector is 1

cost function

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

dealing with
 \uparrow +ve points

constraint

such that

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

\downarrow dealing with
 \downarrow -ve points

These are applicable only
 for correctly classified points

For all correct points

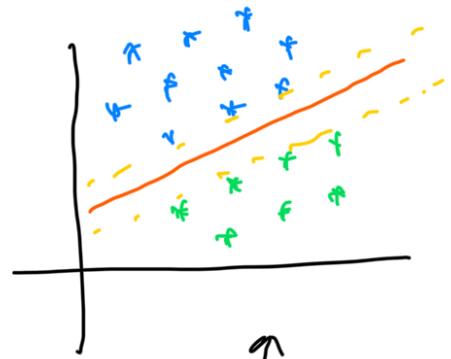
constraint \rightarrow
$$y_i (w^T x + b) \geq 1$$

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

$$\min_{w, b} \|w\|$$

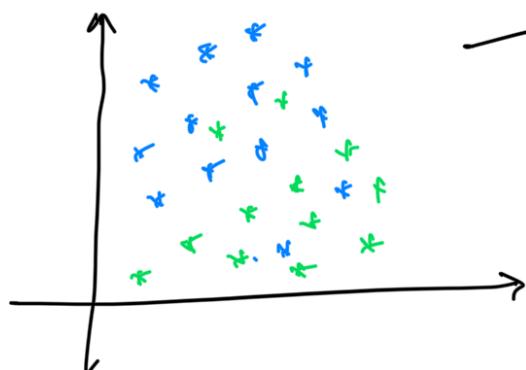
cost function of SVM (SVC)

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



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. here we will have tune it with hyper parameter.

In this case my cost function will be,

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

$C_i \Rightarrow$ how many point we want to avoid misclassification.
hyper parameter \rightarrow

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

- ^{up} ---
the marginal plane.
- * This cost function is related to soft margin.