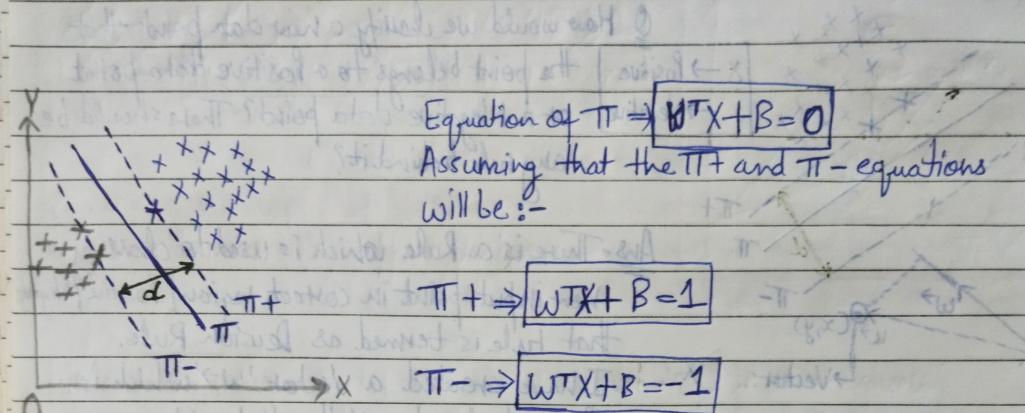


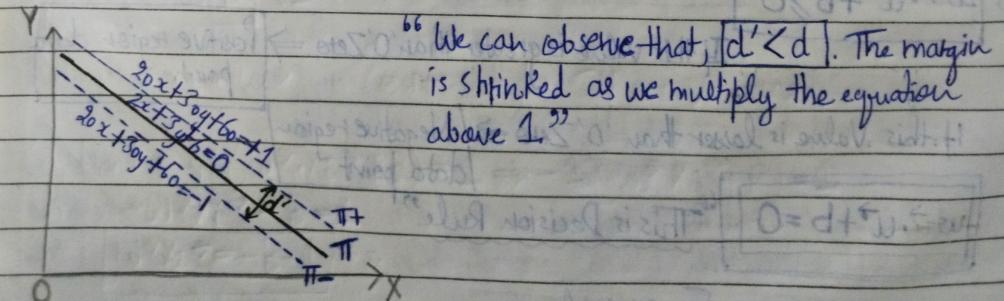
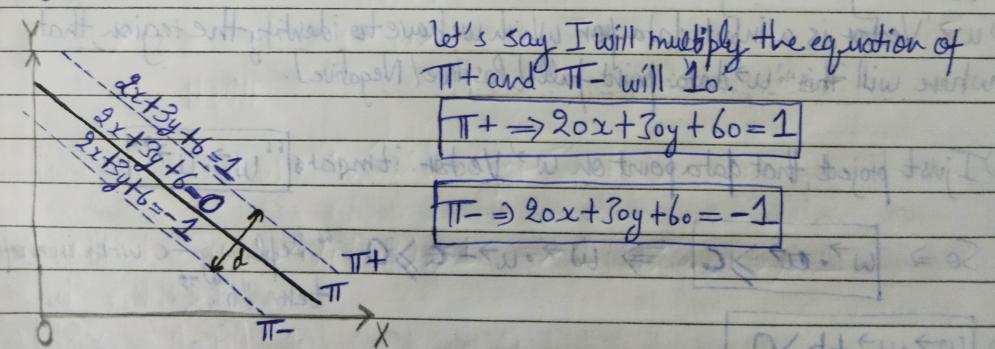
Date.....

Hard Margin SVM....

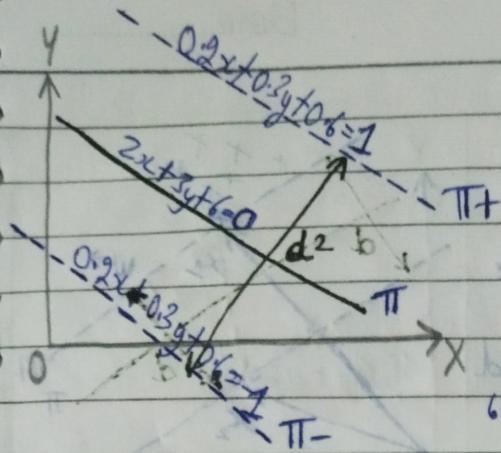


* Q Why the magnitude between the $\Pi+$ and $\Pi-$ are equal?

Ans The Decision Hyperplane is designed in such a way that the magnitude between the $\Pi-$ and the magnitude between the $\Pi+$ are always same.



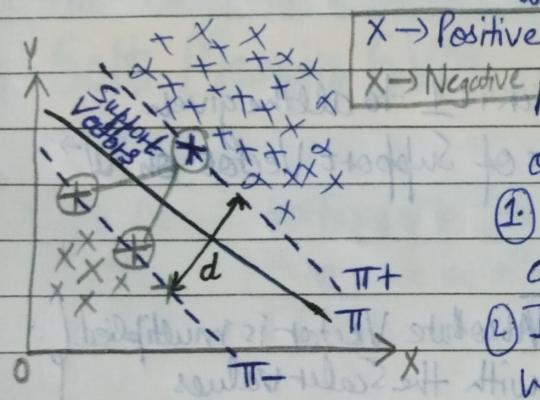
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Let's say I will multiply the $\Pi+$ and $\Pi-$ which is less than 1, say '0.1'.

$$\begin{aligned}\Pi+ &\Rightarrow 0.2x + 0.3y + 0.6 = 1 \\ \Pi- &\Rightarrow 0.2x + 0.3y + 0.6 = -1\end{aligned}$$

"We can observe that $d_2 > d$. It means when we multiply the equation of $\Pi+$ and $\Pi-$ with less than '1', the margin will expand."



As per our SVM, we have some conditions only which this 'd' will be calculated :-

- ① All the data should be properly classified and too much linear.
 - ② If any Positive data point lies inside the negative region or vice versa and if any data point is lies beyond the " $\Pi+$ or $\Pi-$ ".
- (based on this)* then also our complete SVM will fail.

Constraints in Hard Margin SVM

① Π Equation $\Rightarrow \vec{W} \cdot \vec{x}_i + b = 0$, Similarly for $\Pi+$

$\vec{W} \cdot \vec{x}_i + b = 1$, and for $\Pi- = \vec{W} \cdot \vec{x}_i + b = -1$

For Data points lies in the Positive region.

Assuming that:-

$y_i = 1$ [All $y+$ outputs are labelled as 1]

$y_i (\vec{W} \cdot \vec{x}_i + b) \geq 1$

① Constraint

For Data points lies in the Negative Region

$y_i (\vec{W} \cdot \vec{x}_i + b) \leq -1$ [All $y-$ outputs are labelled as -1]

$y_i (\vec{W} \cdot \vec{x}_i + b) \leq 1$

② Constraint

Spiral

For Data Points which lies at hyper Planes or Support Vectors

$$Y_i(\vec{w} \cdot \vec{x}_i + b) = 1$$

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Combining (I) and (II) Constraints as:-

When $y_i=1$

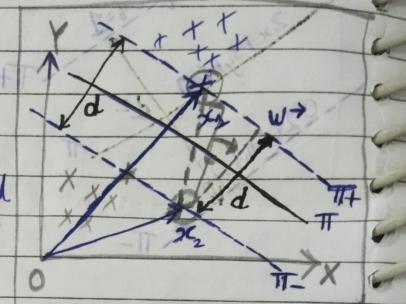
$$Y_i(\vec{w} \cdot \vec{x}_i + b) \geq 1 \Rightarrow 1(\vec{w} \cdot \vec{x}_i + b) \geq 1$$

When $y_i=-1 \Rightarrow Y_i(\vec{w} \cdot \vec{x}_i + b) \leq 1$ Both are same...

So, as according to the Graph, we have taken '2' Support Vectors and named them as ' x_1 ' and ' x_2 '.

distance between the Support Vectors is :-

$$\|x_2 - x_1\|$$



We already have a vector named " w " which is \perp to all the given hyper parameters. If we project the difference of Support Vectors on " w " Vector, we will get the margin 'd'.

$$\therefore d = (x_2 - x_1) \cdot \frac{(w)}{\|w\|}$$

Absolute Vector is multiplied
with the Scalar values

$$\Rightarrow \frac{x_2 \cdot w - x_1 \cdot w}{\|w\|}$$

$$\|w\|$$

Putting the " $x_2 \cdot w$ " in Constraint for Support Vectors:-

$$\text{Constraint} = Y_i(\vec{w} \cdot \vec{x}_i + b) \geq 1 \Rightarrow 1(\vec{w} \cdot \vec{x}_i + b) \geq 1$$

$$\Rightarrow \vec{w} \cdot \vec{x}_i + b \geq 1$$

$$\Rightarrow [\vec{w} \cdot \vec{x}_i = 1 - b] \geq 1$$

Putting the " $x_1 \cdot w$ " in Constraint for Support Vectors:-

$$\text{Constraint} = Y_i(\vec{w} \cdot \vec{x}_i + b) = 1 \Rightarrow -1(\vec{w} \cdot \vec{x}_i + b) = 1$$

$$\Rightarrow -\vec{w} \cdot \vec{x}_i - b = 1$$

$$[\vec{w} \cdot \vec{x}_i = -b - 1] \Rightarrow$$

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Putting the values in 1 and 2 in 'd' margin.

$$\frac{d = 1 - b - (-b - 1)}{\|w\|} \Rightarrow \frac{1 - b + b + 1}{\|w\|} \quad \boxed{d = 2}$$

It means $\text{argmax } (w^*, b^*)$ $\boxed{\frac{2}{\|w\|}}$ Should be maximum or highest term

given that $\boxed{y_i(w^T x_i + b) \geq 0}$