#### **SVM**

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Sumit Kumar Yaday

Department of Management Studies Indian Institute of Technology, Roorkee

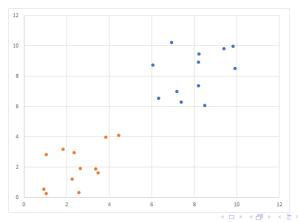


# Recap and Today

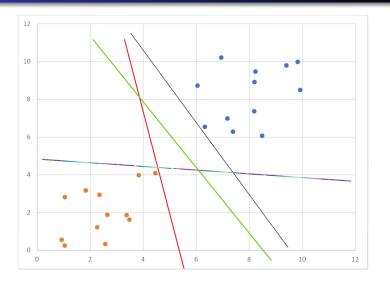
- SVM
- More of SVM

# Support Vector Machine - Introduction

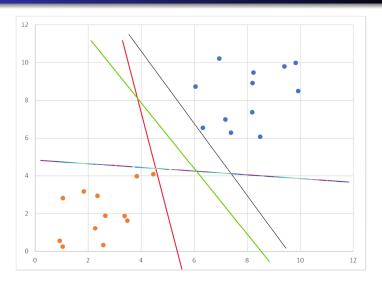
- Supervised Learning Algorithm for Classification
- Given N training points in which  $n_1$  are of type A,  $n_2$  are of type B, draw the **best** line(plane)
- To begin with, assume that the training points are linearly separable



#### Which is the best line?



#### Which is the best line?



What makes us think it is the green line? Can we make the ideas a bit more precise?

#### **Notations**

Let the data-set be denoted as -

S.No	$X_1$	$X_2$	Y (+1 or -1)
1	<i>x</i> <sub>11</sub>	<i>X</i> <sub>12</sub>	+1
2	<i>x</i> <sub>21</sub>	X22	-1
3	X31	X32	-1
N	X <sub>N1</sub>	X <sub>N2</sub>	+1

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Let the equation of the line be  $w_1x_1 + w_2x_2 + b = 0$ We need to determine  $w_1$ ,  $w_2$  and b

#### Obtaining $w_1$ , $w_2$ and b

- Arbitratily choose  $w_1$ ,  $w_2$  and b such that  $w_1x_{i1} + w_2x_{i2} + b > 0$  whenever  $y_i = +1$   $w_1x_{i1} + w_2x_{i2} + b < 0$  whenever  $y_i = -1$
- Simply put, for all points,  $y_i(w_1x_{i1} + w_2x_{i2} + b) > 0$
- Criteria Consider a line. Find the distance of the line from all the training examples (or points). Look at the minimum of all these distances.

We are interested in the line for which this minimum distance is as large as possible.

$$\bullet \max_{(w_1, w_2, b)} \left( \min_{i = \{1, 2, \dots, N\}} \frac{|w_1 x_{i1} + w_2 x_{i2} + b|}{\sqrt{w_1^2 + w_2^2}} \right)$$

 The max in the equation in maximize and min in the equation is minimum



The optimization problem thus becomes -

$$\max_{(w_1,w_2,b)} \left( \min_{i=\{1,2,...,N\}} \frac{|w_1 x_{i1} + w_2 x_{i2} + b|}{\sqrt{w_1^2 + w_2^2}} \right)$$

The max in the equation in maximize and min in the equation is minimum

subject to the following **N** constraints -  $y_i(w_1x_{i1} + w_2x_{i2} + b) > 0$ 

The optimization problem thus becomes -

$$\max_{(w_1,w_2,b)} \left( \frac{1}{\sqrt{w_1^2 + w_2^2}} \left[ \min_{i=\{1,2,\dots,N\}} (|w_1 x_{i1} + w_2 x_{i2} + b|) \right] \right)$$

subject to the following  ${f N}$  constraints -

$$y_i(w_1x_{i1}+w_2x_{i2}+b)>0$$

The optimization problem thus becomes -

$$\max_{(w_1,w_2,b)} \left( \frac{1}{\sqrt{w_1^2 + w_2^2}} \left[ \min_{i=\{1,2,....,N\}} (|w_1 x_{i1} + w_2 x_{i2} + b|) \right] \right)$$

subject to the following  ${f N}$  constraints -

$$y_i(w_1x_{i1}+w_2x_{i2}+b)>0$$

As scaling all  $w_1$ ,  $w_2$  and b by the same factor (non-zero) doesn't change the line (or hyperplane), we will choose  $w_1$ ,  $w_2$  and b such that -

$$\min_{i=\{1,2,\ldots,N\}}(|w_1x_{i1}+w_2x_{i2}+b|)=1$$
  
The **min** in the above equation is **minimum**



The optimization problem thus becomes -

$$\max_{(w_1, w_2, b)} \left( \frac{1}{\sqrt{w_1^2 + w_2^2}} \right)$$

subject to the following 2N constraints -

$$y_i(w_1x_{i1}+w_2x_{i2}+b)>0$$

$$\min_{i=\{1,2,...,N\}}(|w_1x_{i1}+w_2x_{i2}+b|)=1$$

The optimization problem thus becomes -

$$\min_{\substack{(w_1,w_2,b)}} \left(\frac{w_1^2+w_2^2}{2}\right)$$
 subject to the following **2N** constraints -

$$y_i(w_1x_{i1}+w_2x_{i2}+b)>0$$

$$\min_{i=\{1,2,...,N\}} (|w_1 x_{i1} + w_2 x_{i2} + b|) = 1$$

The optimization problem thus becomes -

$$\min_{(w_1,w_2,b)} \left(\frac{w_1^2+w_2^2}{2}\right)$$
 subject to the following  ${\bf 2N}$  constraints -

$$y_i(w_1x_{i1}+w_2x_{i2}+b)>0$$

$$\min_{i=\{1,2,\ldots,N\}}(|w_1x_{i1}+w_2x_{i2}+b|)=1$$

$$\min_{i=\{1,2,...,N\}} (|w_1x_{i1} + w_2x_{i2} + b|) = 1 \text{ implies -} |w_1x_{i1} + w_2x_{i2} + b| >= 1 \quad \forall i = \{1,2,...,N\} \text{ or }$$

$$|y_i(w_1x_{i1} + w_2x_{i2} + b)| >= 1$$
  $\forall i = \{1, 2, ..., N\}$  or  $y_i(w_1x_{i1} + w_2x_{i2} + b) >= 1$   $\forall i = \{1, 2, ..., N\}$  The implies condition is not both ways, but still it can be replaced in this problem because ??

The implies condition can be replaced in this problem because ?? After some algebra, the optimization problem becomes -

$$\min_{\substack{(w_1, w_2, b) \\ \text{subject to the following}}} \left(\frac{w_1^2 + w_2^2}{2}\right)$$

subject to the following  ${\bf N}$  constraints -

$$y_i(w_1x_{i1} + w_2x_{i2} + b) >= 1$$
  $\forall i = \{1, 2, ..., N\}$ 

#### Why the name support vector?

Consider the following two optimizaion problems -

$$\min_{\substack{(w_1,w_2,b)\\\text{subject to the following N constraints -}}} \left(\frac{w_1^2+w_2^2}{2}\right) \qquad \qquad \mathbf{w_1}^*, \ \mathbf{w_2}^*, \ \mathbf{b}^*$$

$$y_i(w_1x_{i1} + w_2x_{i2} + b) >= 1$$
  $\forall i = \{1, 2, ..., N\}$ 

$$\min_{(w_1,w_2,b)} \left(\frac{w_1^2+w_2^2}{2}\right) - \sum_{i=1}^N \alpha_i (y_i(w_1x_{i1}+w_2x_{i2}+b)-1)$$
 subject to no constraints, only the fact that all  $\alpha_i$ 's are either zero or positive

Which of these two optimization problems has a lower value?



# Why the name support vector?

$$\min_{\substack{(w_1,w_2,b)}} \left(\frac{w_1^2 + w_2^2}{2}\right)$$
 subject to the following **N** constraints -

$$y_i(w_1x_{i1} + w_2x_{i2} + b) >= 1$$
  $\forall i = \{1, 2, ..., N\}$ 

$$\min_{(w_1,w_2,b)} \left( \frac{w_1^2 + w_2^2}{2} \right) - \sum_{i=1}^{N} \alpha_i (y_i (w_1 x_{i1} + w_2 x_{i2} + b) - 1)$$

subject to no constraints, only the fact that all  $\alpha_i$  's are either zero or positive

The one in the red box has a lower value. Now, let us keep playing with putting different values of  $\alpha_i$ 's.



Thank you for your attention