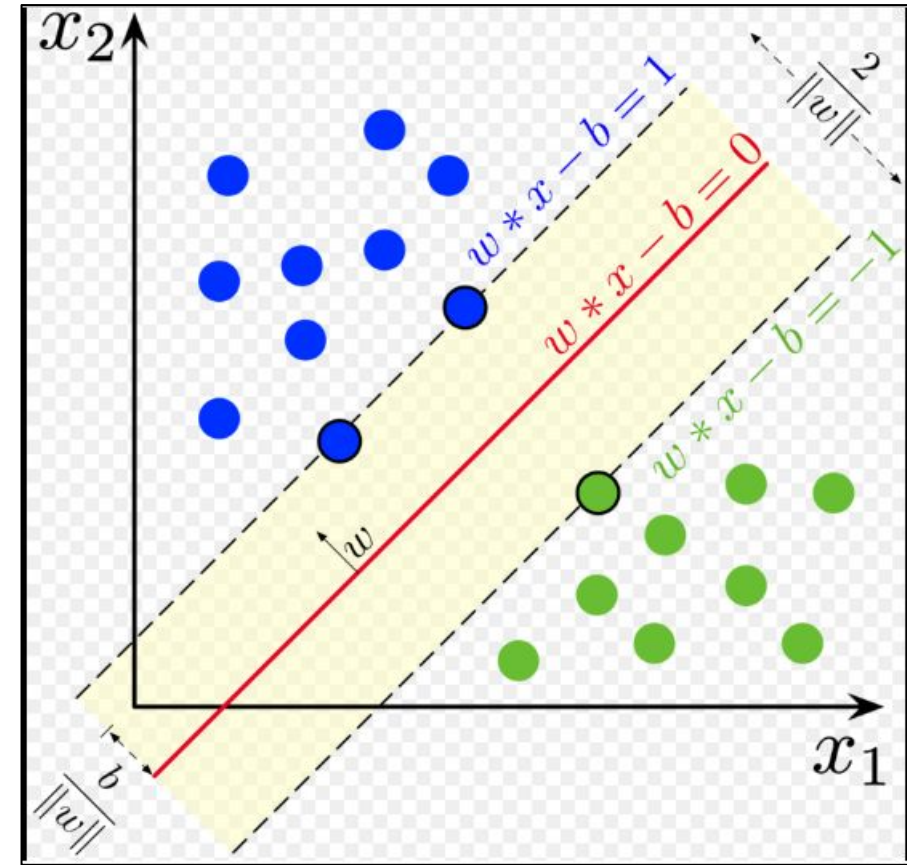


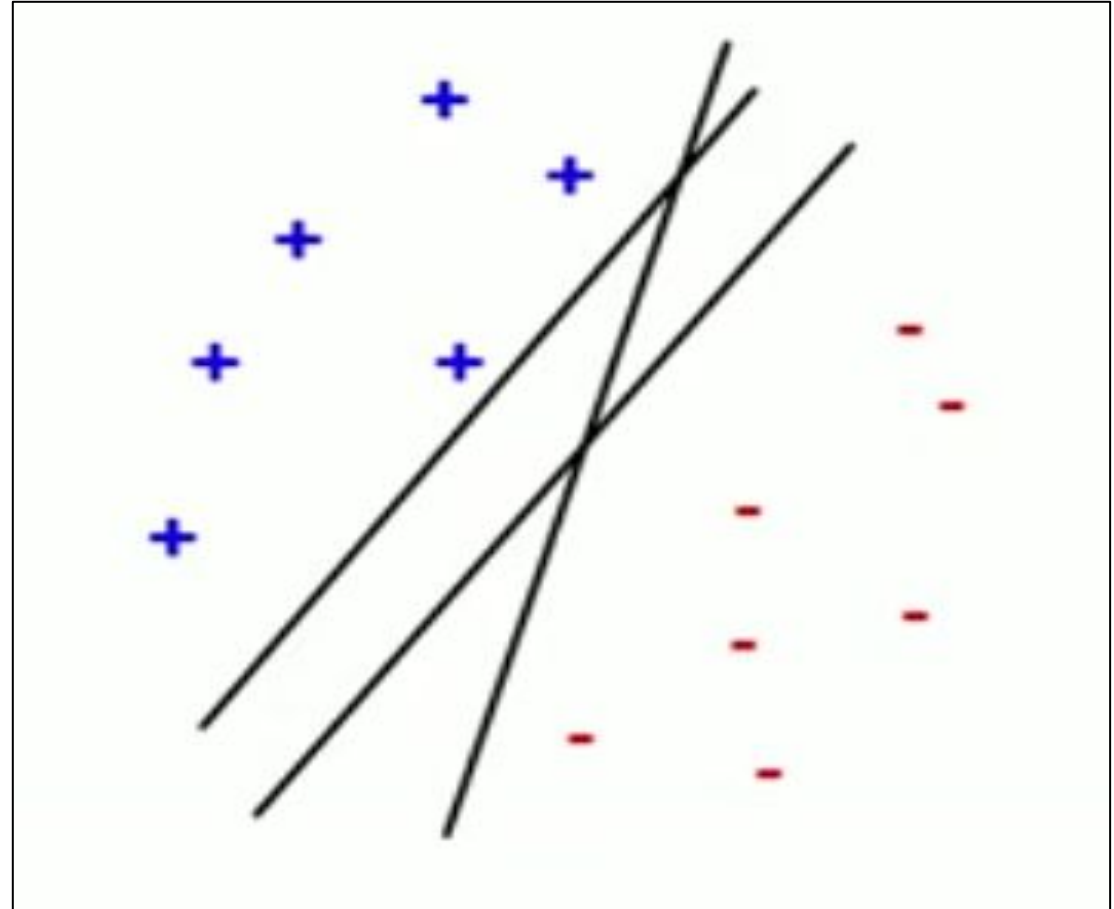
Classifier: Support Vector machine (SVM)

- An SVM model is a representation of the examples as points in space, mapped so that the examples of the separate categories are divided by a clear gap that is as wide as possible (Maximum Margin Classifier).
- Supervised Algorithm

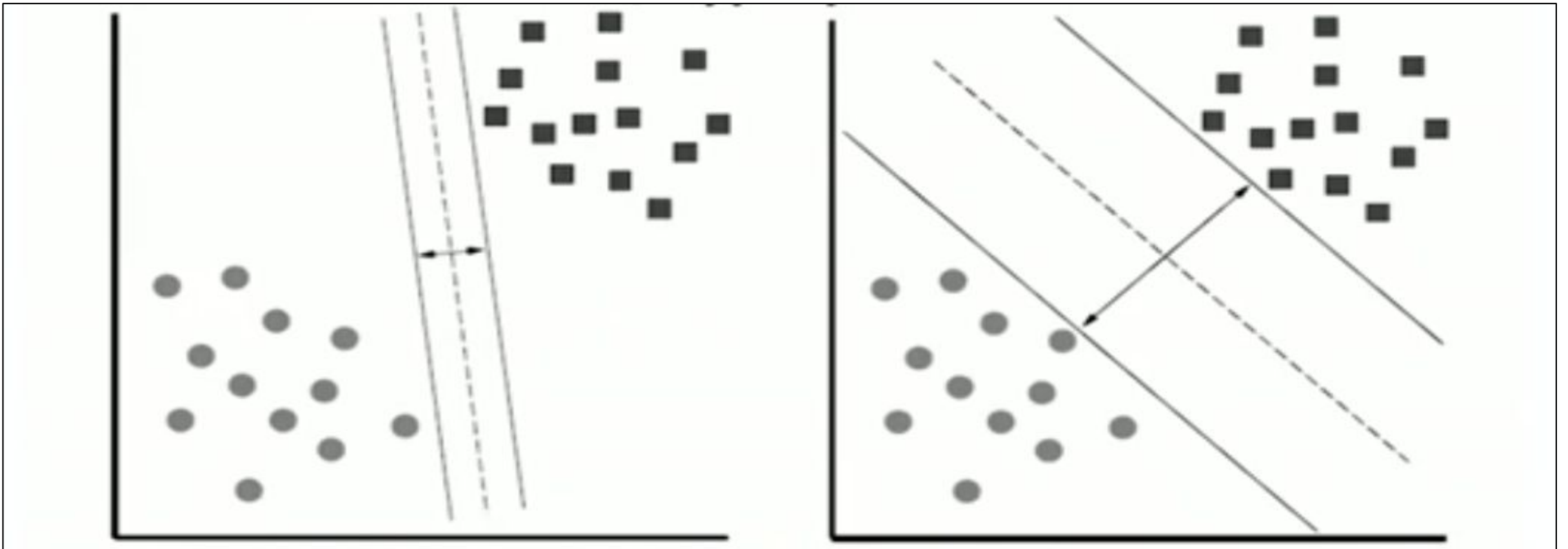


What's the Goal?

1. Want to separate '+' from '-' using a line in a 2 class problem.
2. The feature vector may be n-dimensional.
3. Multi Classes SVM works for more than 2 classes.

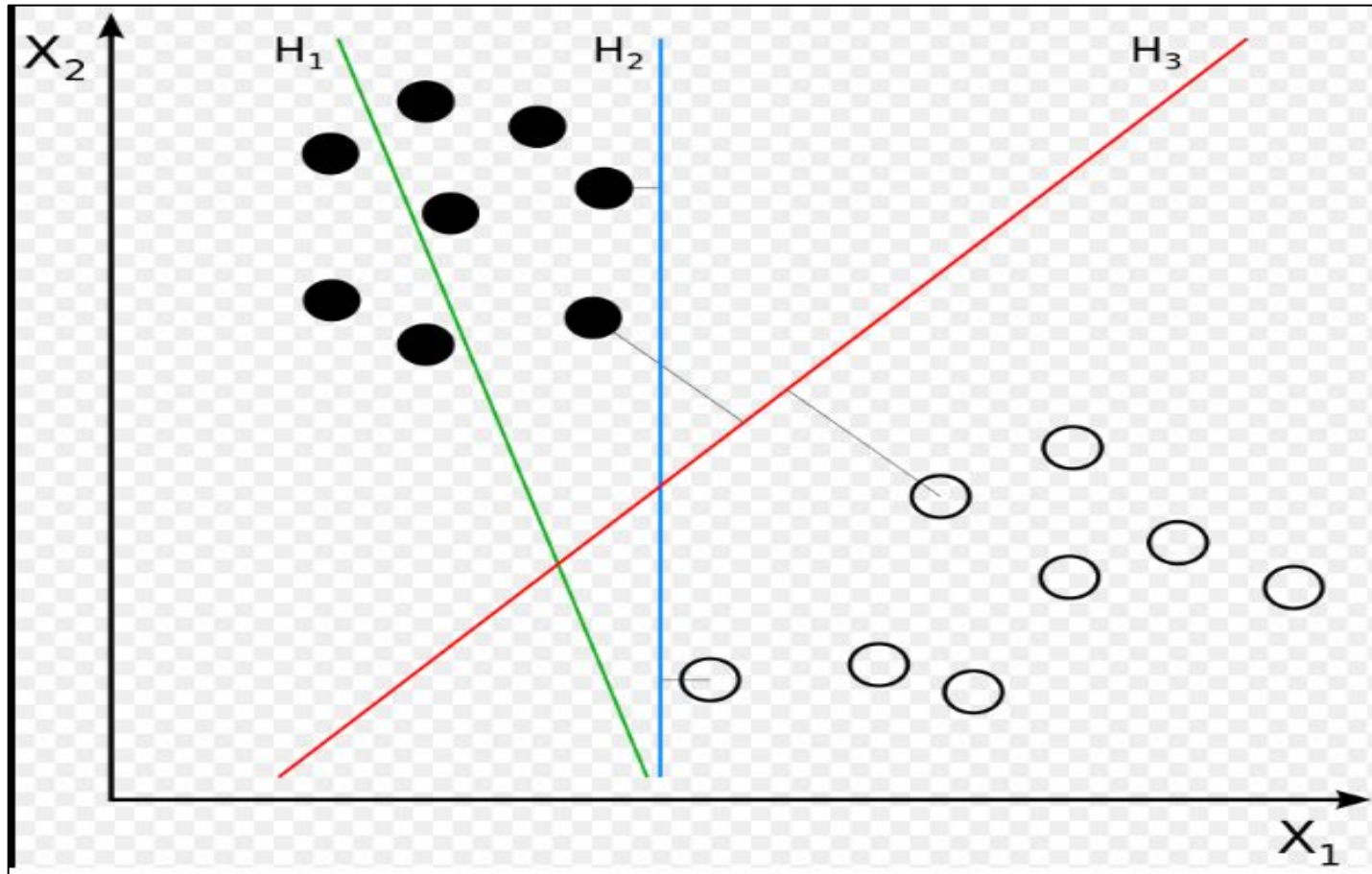


Maximizing the margin (Y)



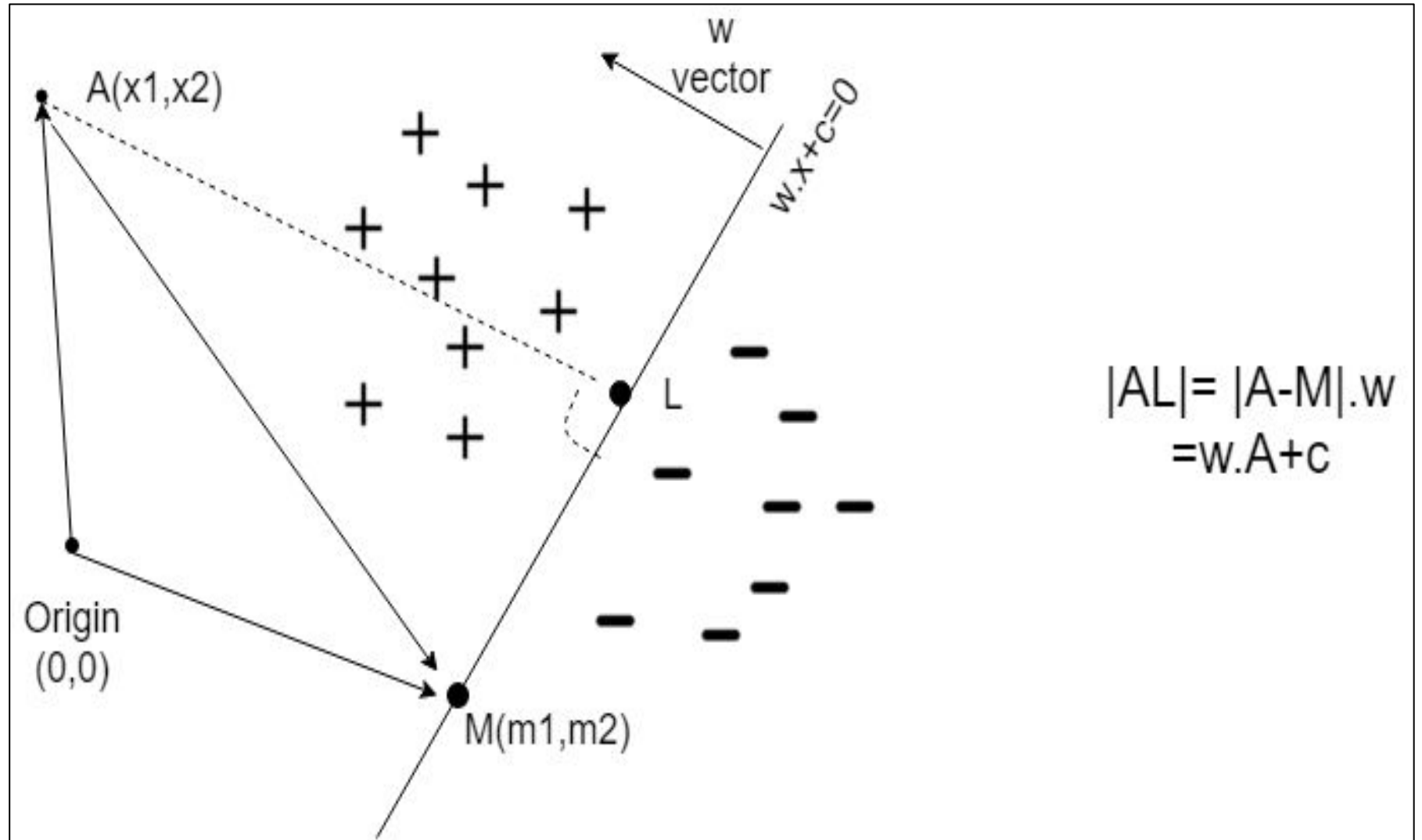
Source: <https://www.youtube.com/watch?v=ax8LxRZCORU>

One More Example(H3 is best line separating the classes with maximum margin)



Problem Formulation

1. A = data point from dataset
2. M = point on line
3. Aim is to find perpendicular distance $|AL|$.



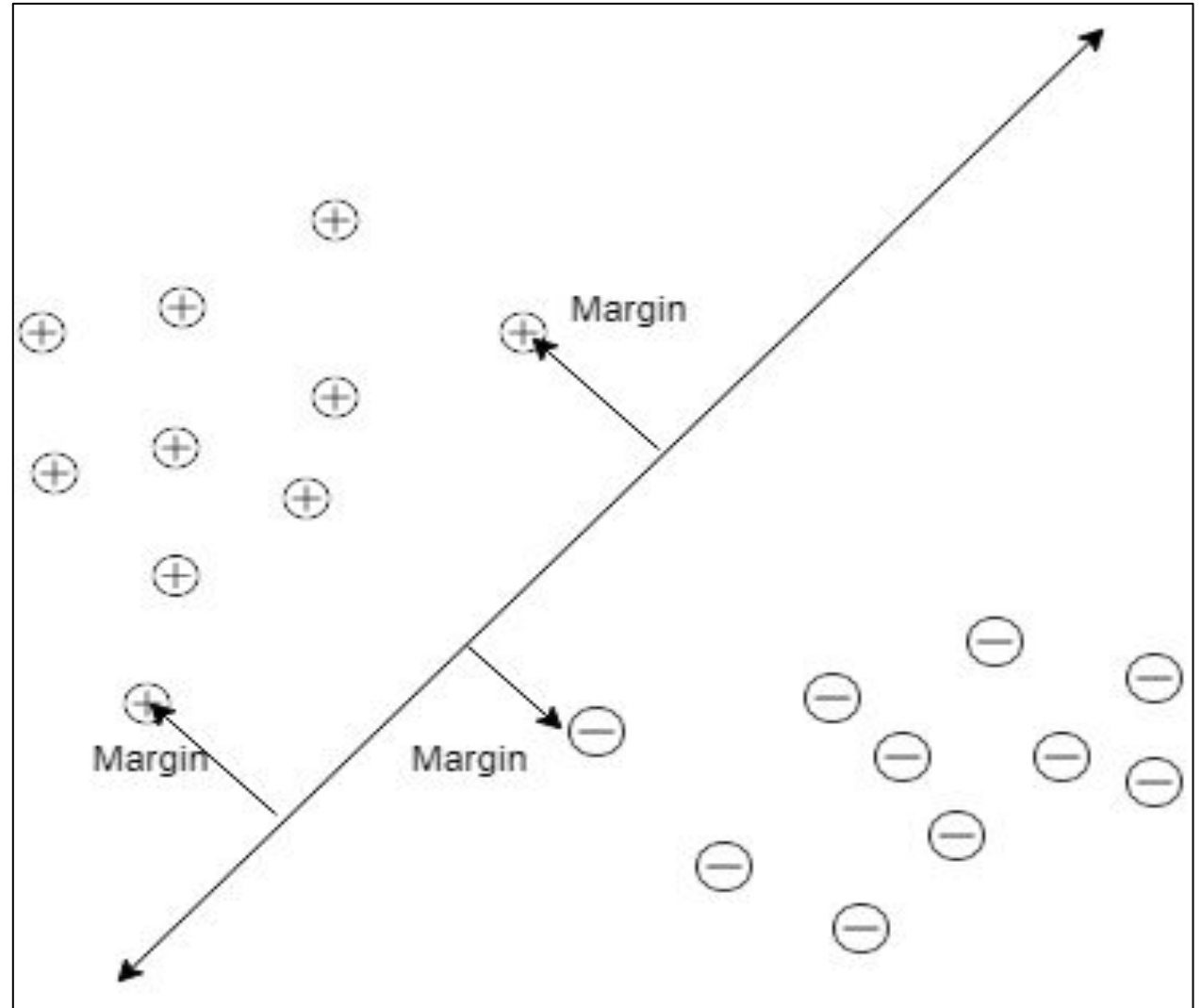
Problem formulation Continued:

1. Dataset looks like :
2. Now we define:
“Prediction” = $\text{sign}(w.x+c)$
“Confidence” = $(w.x+c).y$
 1. If $w.x+c=0$, point is on line and class can't be decided.
 2. If $w.x+c>0$, point belongs to +ve class
 3. Else, it belongs to -ve class.

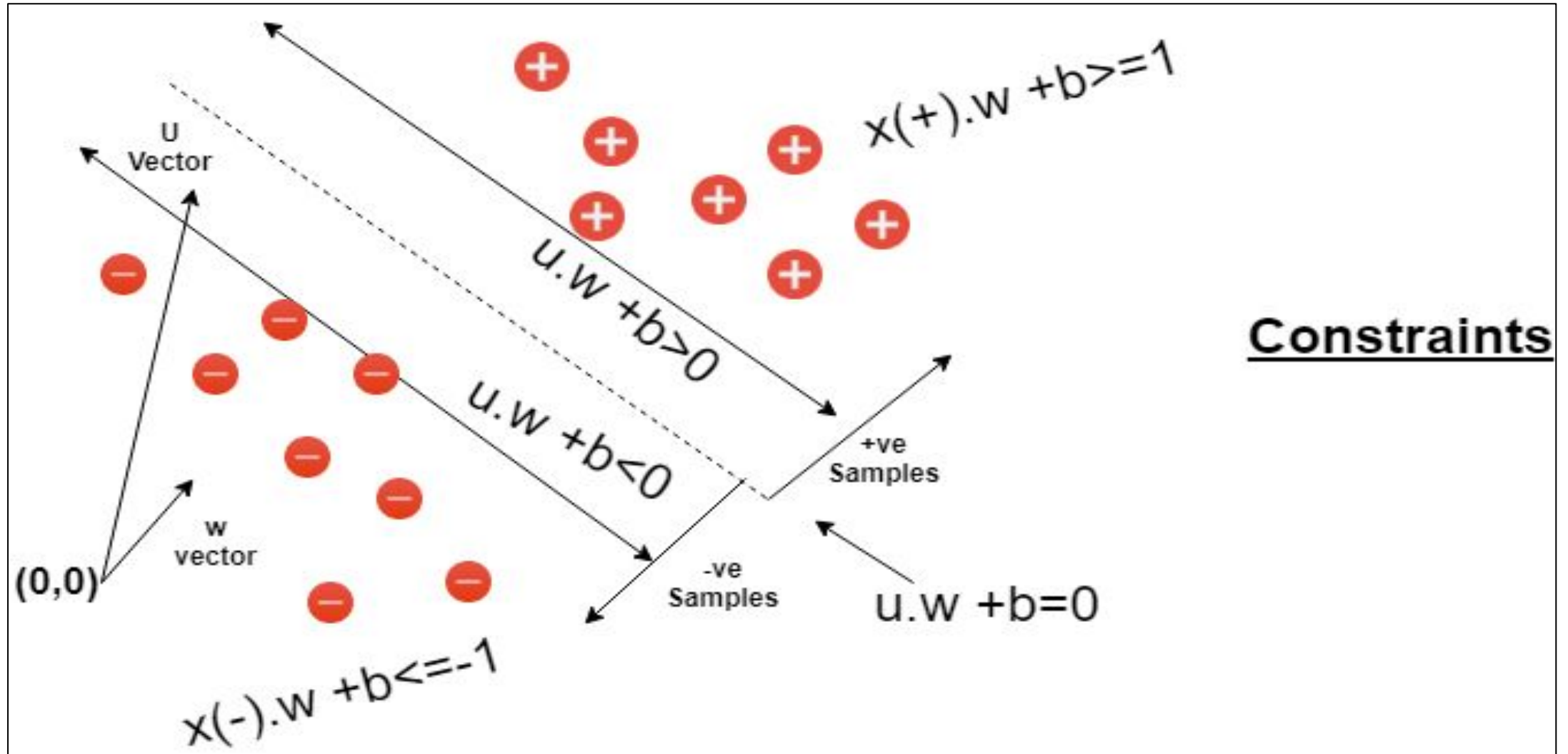
X1	X2	Class- Y
1	2	+1
2.4	4.5	-1
2.5	8	-1
0	2.1	+1

Problem Statement

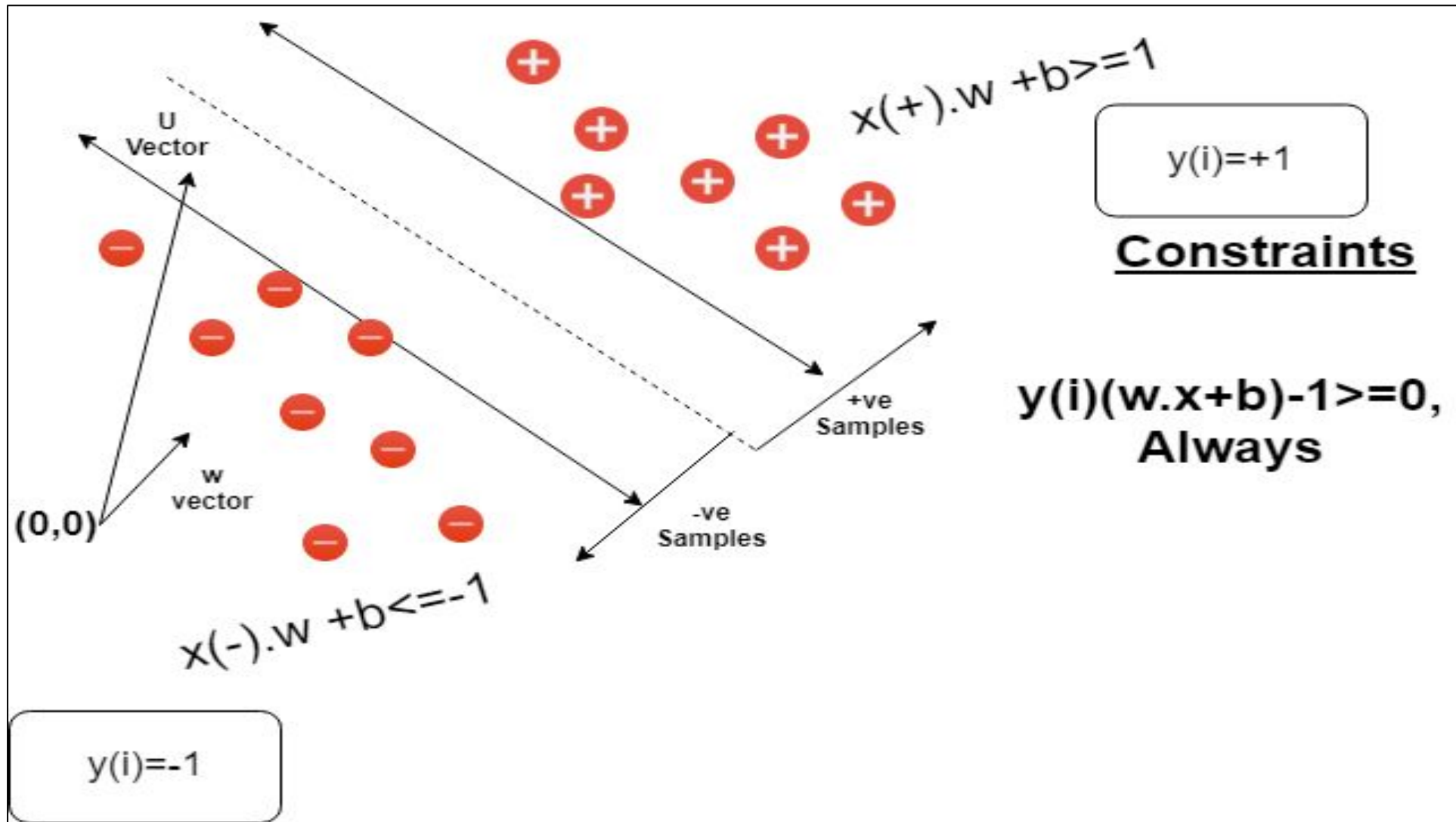
1. For i th datapoint:
 $\text{Margin}(i) = (w \cdot x(i) + c) * y(i)$
2. So we want to solve:
 $\max_w \min_i (\text{Margin}(i))$
3. Hence a Constrained Optimization Problem.



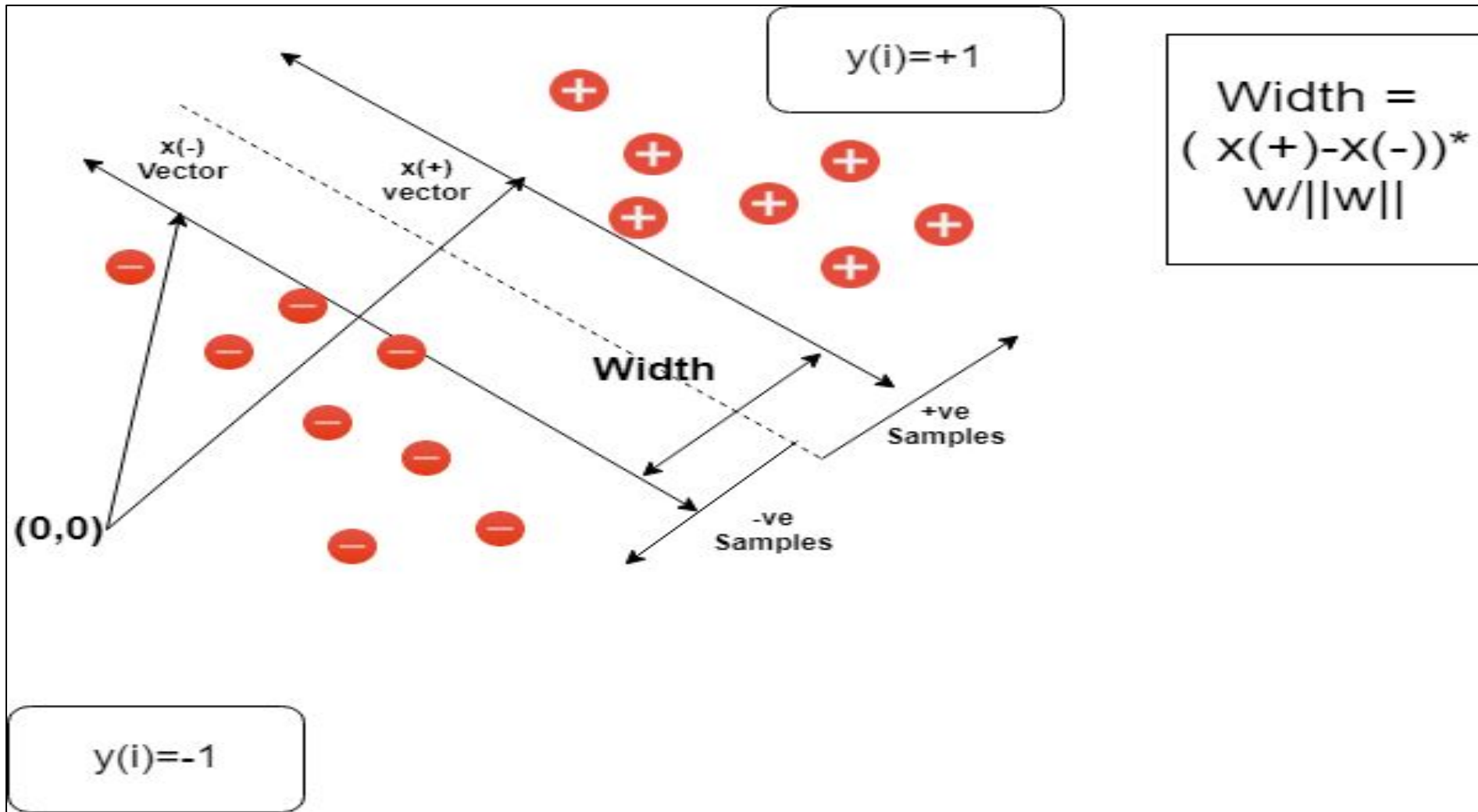
Additional Constraints



Additional Constraints



Finding the width between Two Support Vectors



Using Lagrange multiplier

(is a strategy for finding the local maxima and minima of a function subject to equality constraints)

$$\text{width} = (\bar{x}_+ - \bar{x}_-) \cdot \frac{\bar{w}}{\|w\|}$$

$$= \frac{(\bar{x}_+ \cdot \bar{w} - \bar{x}_- \cdot \bar{w})}{\|w\|} = \frac{1-b + 1+b}{\|w\|} = \frac{2}{\|w\|}$$

$$\text{So } \text{MAX } \frac{2}{\|w\|} \Rightarrow \text{MIN } \|w\| \text{ or } \boxed{\text{MIN} \left(\frac{1}{2} \|w\|^2 \right)}$$

$$L = \frac{1}{2} \|w\|^2 - \sum d_i [y_i (\bar{w} \cdot \bar{x}_i + b) - 1]$$

$d_i = \text{multiplier}$ \uparrow
constraint

$$\text{Now, } \frac{\partial L}{\partial \bar{w}} = 0$$

$$\Rightarrow \bar{w} = \sum d_i y_i \bar{x}_i$$

$$\frac{\partial L}{\partial b} = 0$$

$$\sum d_i y_i = 0$$

Putting the terms found previous back into equation

After solving it,

$$L = \sum d_i - \frac{1}{2} \sum_i \sum_j d_i d_j y_i y_j \boxed{x_i x_j}$$

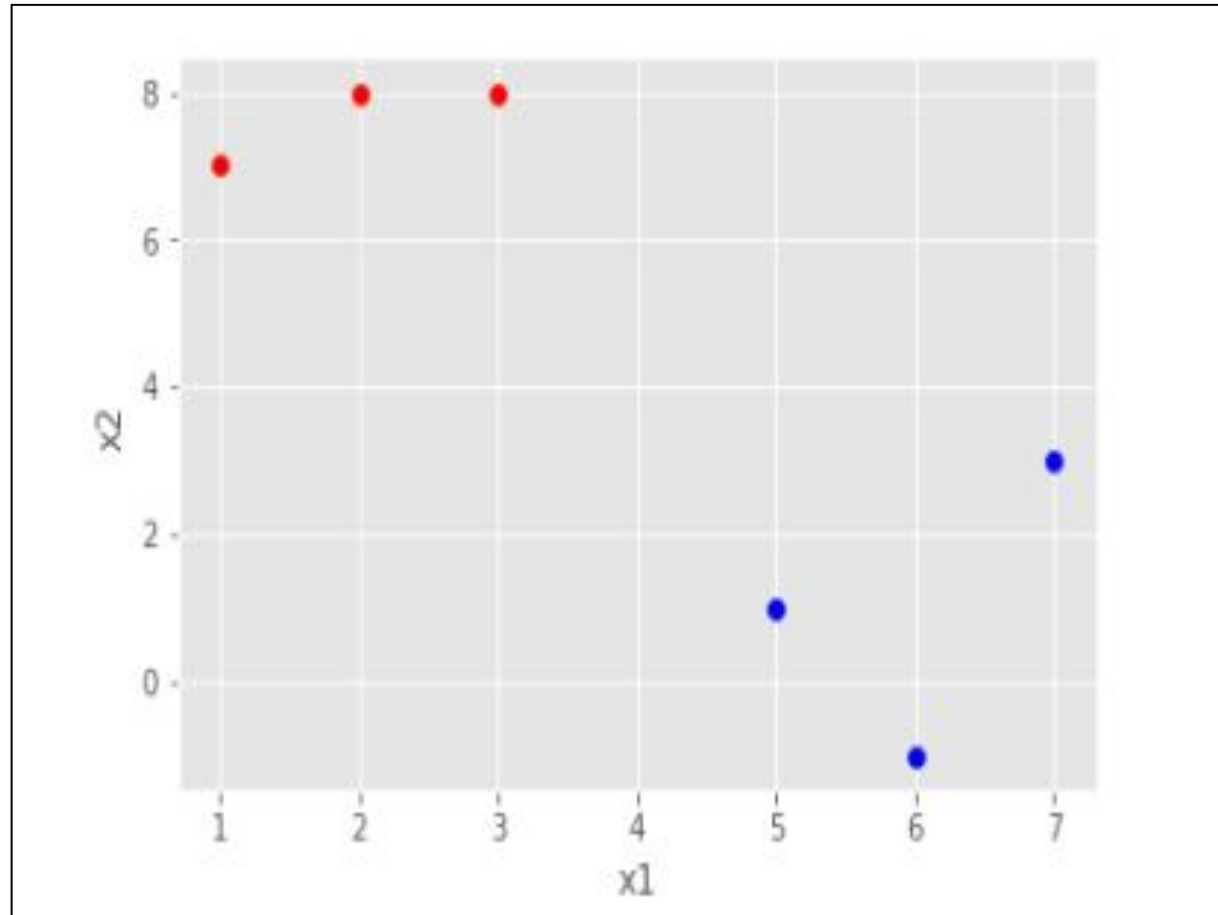
So

$$\sum d_i y_i \bar{x}_i \cdot \bar{u} + b \geq 0 \text{ Then '+'}$$

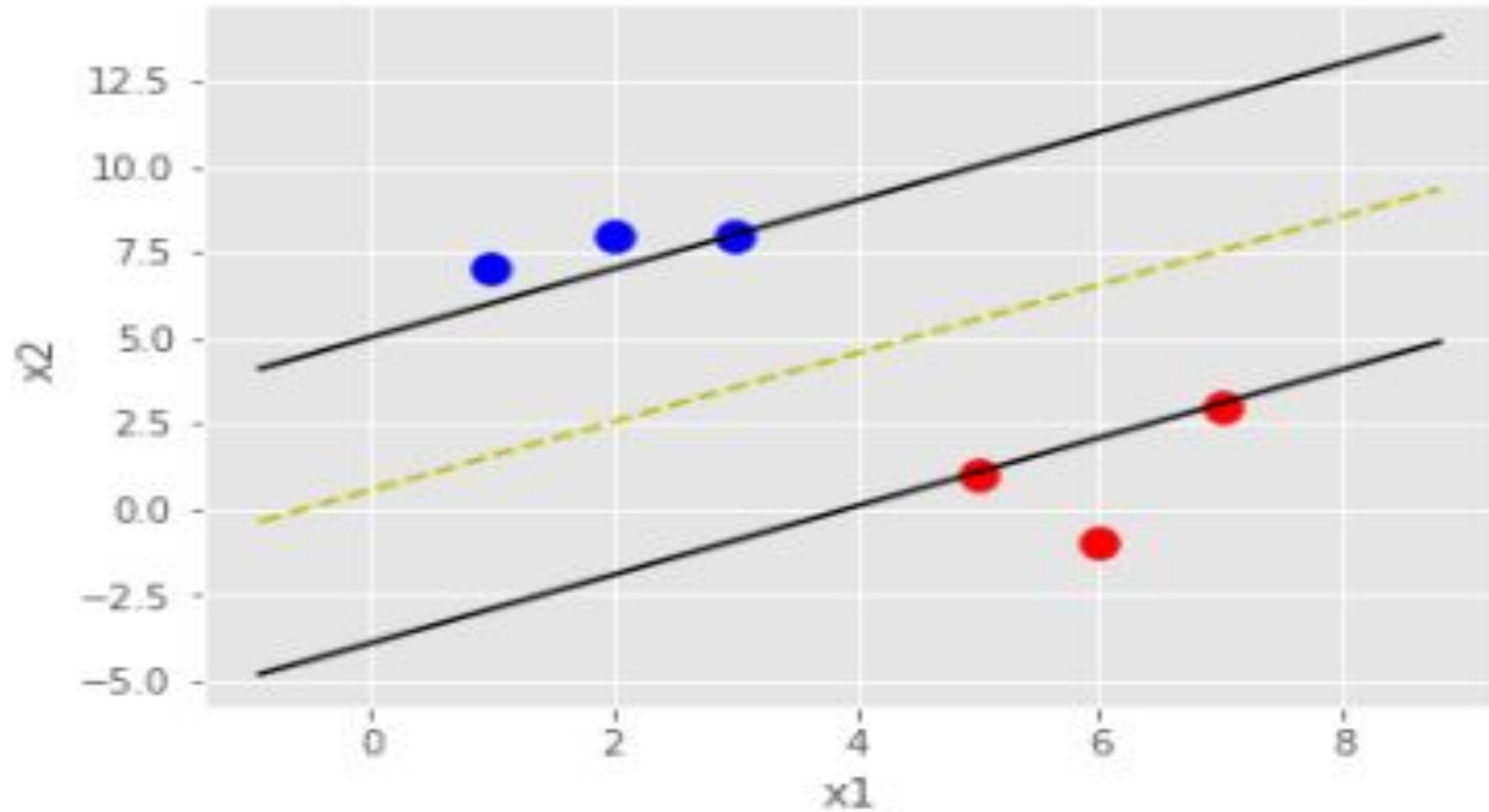
else '-'

Experiment on 2D data Points

Data Points: Red -ve samples
Blue +ve samples

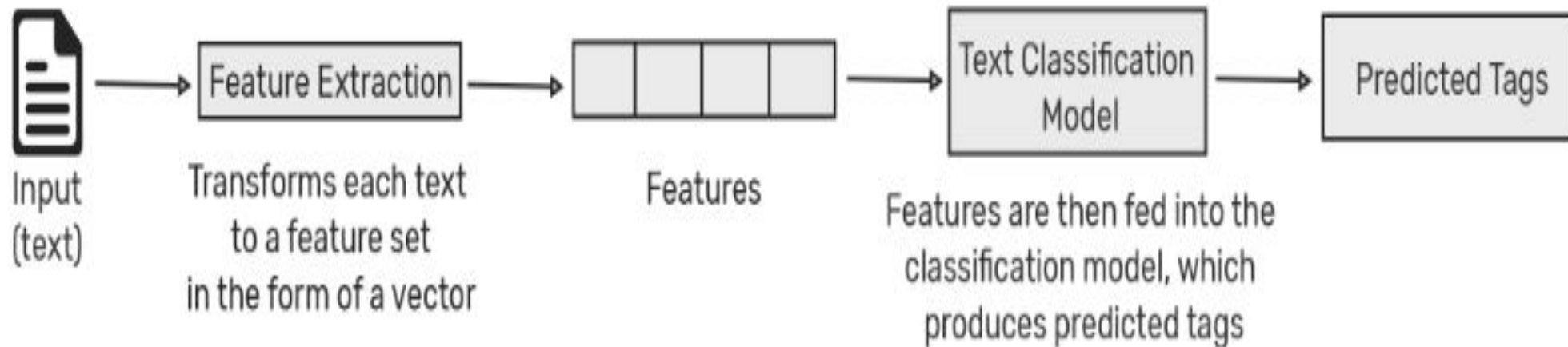


Result after applying SVM algorithm



Text Classification using SVM

- Text classification (text categorization or text tagging) is the task of assigning a set of predefined categories to free-text. Text classifiers can be used to organize, structure, and categorize pretty much anything.



Source: <https://monkeylearn.com/text-classification/>

Thanks