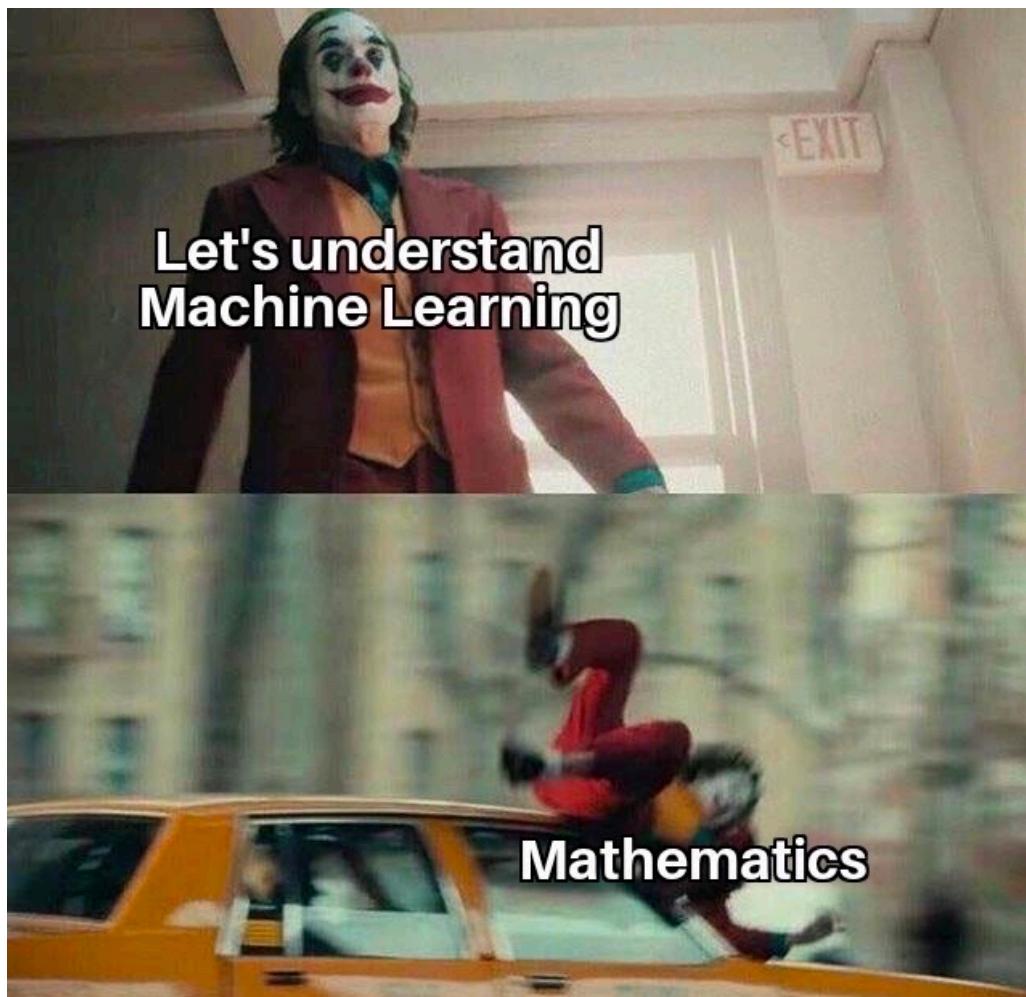


February 23, 2023

DSML : Math for ML

Linear Algebra: Loss minimization in Classification



Recap:

(a) Vectors: $\bar{x} \in \mathbb{R}^d$, $\bar{x} = [x_1 \ x_2 \ \dots \ x_d]$

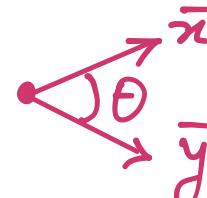
(b) Norm: $\|\bar{x}\| = \sqrt{x_1^2 + x_2^2 + \dots + x_d^2}$

(c) Inner product: $\bar{x}, \bar{y} \in \mathbb{R}^d$

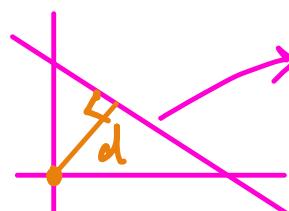
(Dot) $\bar{x} \cdot \bar{y} = \bar{x}^T \bar{y} = \sum_{i=1}^d x_i \cdot y_i$

(d) angle between vectors:

$$\cos(\theta) = \frac{\bar{x}^T \bar{y}}{\|\bar{x}\| \cdot \|\bar{y}\|}$$



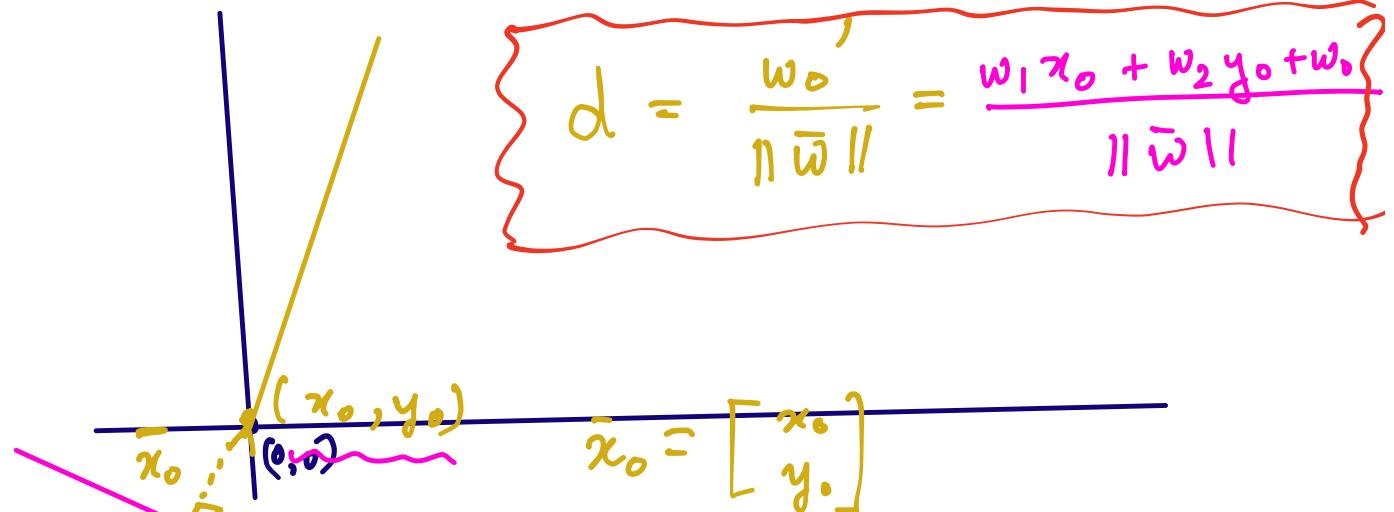
(e) Distance between origin and a line:



line equation: $\bar{w}^T \bar{x} + w_0 = 0$

$$d = \frac{|w_0|}{\|\bar{w}\|}$$

Distance between a point and a line



To make the shift,

we add x_0 to x_1 , y_0 to x_2

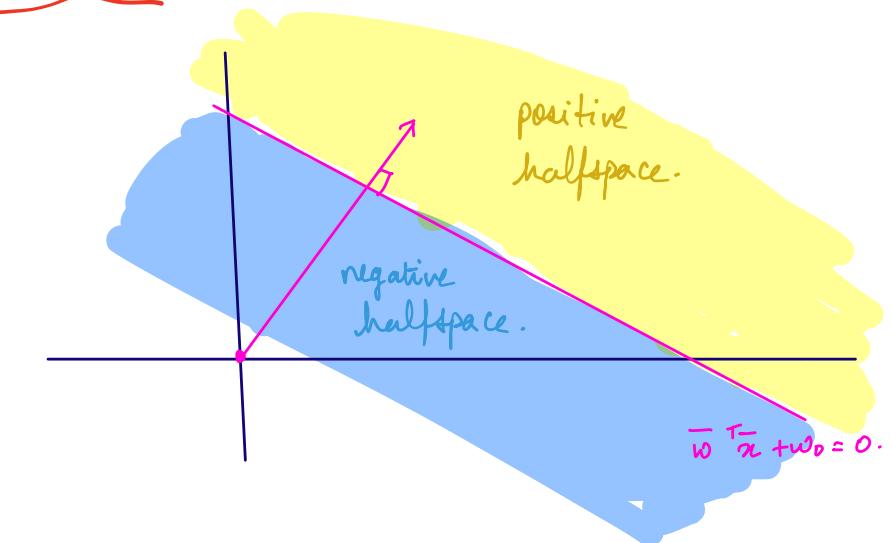
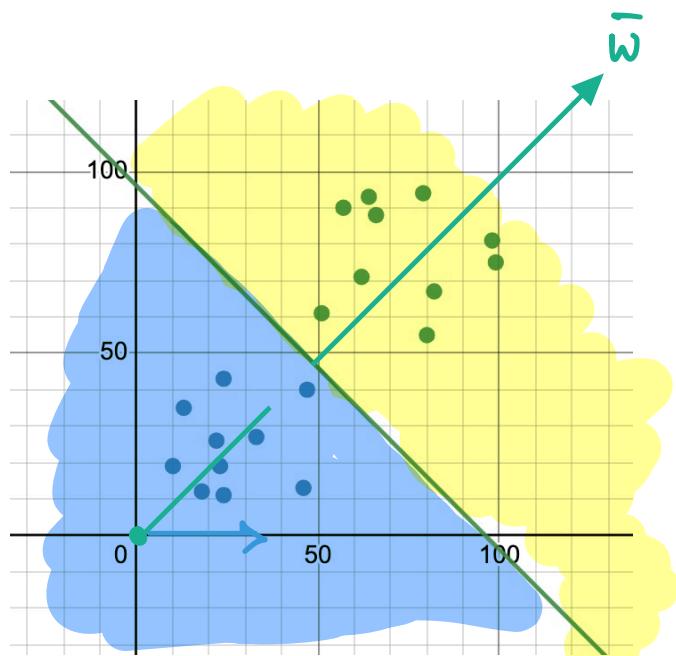
$$w_1(x_1 + x_0) + w_2(x_2 + y_0) + w_0 = 0$$

$$\bar{w}^T \bar{x} + w_0 = 0.$$

$$w_1 x_1 + w_1 x_0 + w_2 x_2 + w_2 y_0 + w_0 = 0.$$

$$w_1 x_1 + w_2 x_2 + \underbrace{(w_1 x_0 + w_2 y_0 + w_0)}_{w_0} = 0.$$

How to find the halfspace a point belongs to.



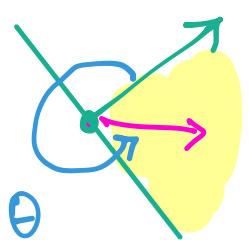
$$\bar{w} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$x + y - 98 = 0.$$

$$\Rightarrow \begin{aligned} x &= a \\ x - a & \end{aligned}$$

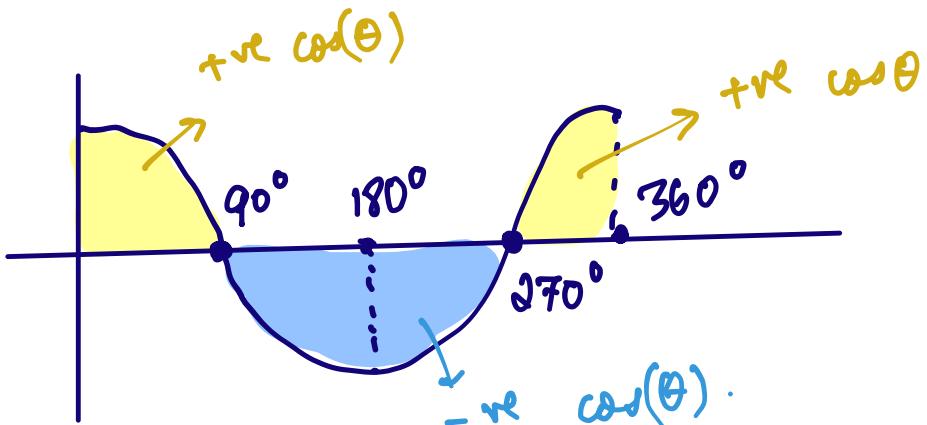
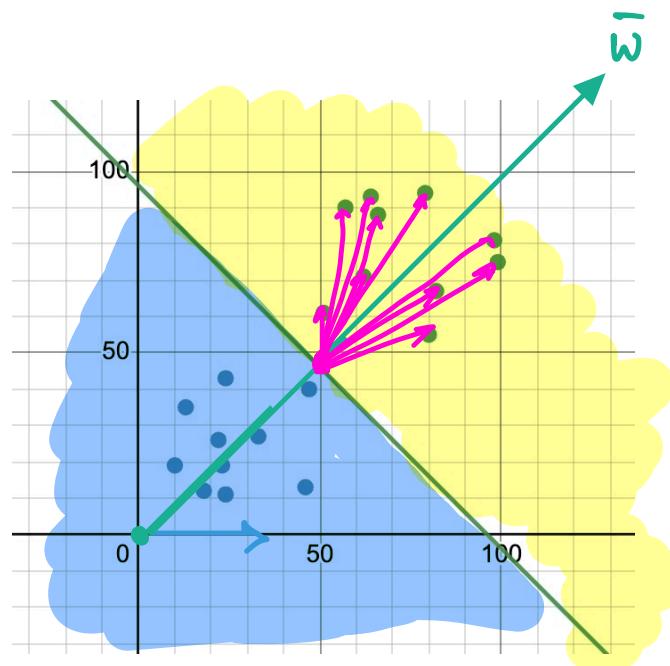
$$\begin{aligned} 1x + 0y - a &= 0 \\ [1, 0] & \end{aligned}$$

4]



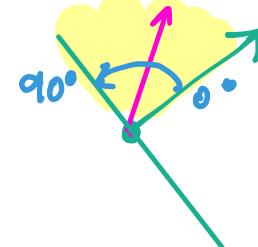
$$270^\circ \leq \theta \leq 360^\circ$$

$\cos \theta \rightarrow +ve$



Counterclockwise angles.

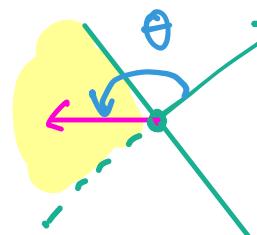
1]



$$0^\circ \leq \theta \leq 90^\circ$$

$\cos \theta \rightarrow +ve$

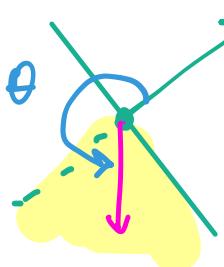
2]



$$90^\circ \leq \theta \leq 180^\circ$$

$\cos \theta \rightarrow -ve$

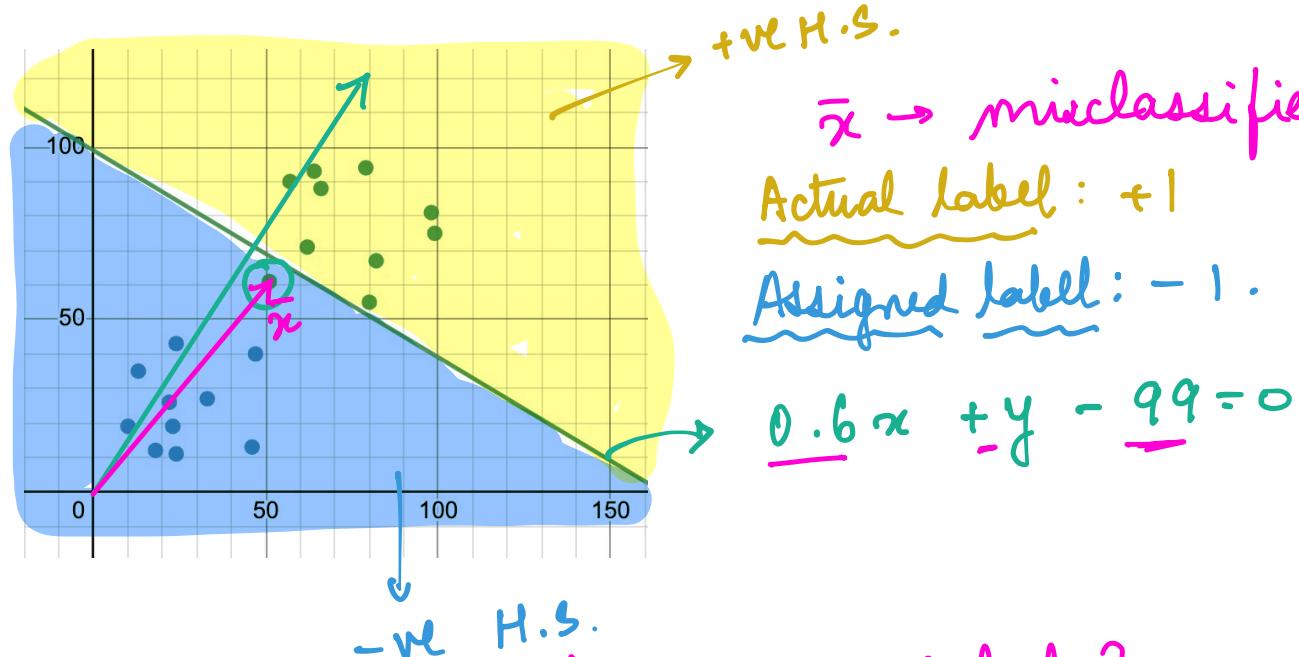
3]



$$180^\circ \leq \theta \leq 270^\circ$$

$\cos \theta \rightarrow -ve$

Putting it all together: loss function.



Q] How does a classifier assign labels?
x-bar → what is the label of this?

Ans. We first calculate $d = \frac{\bar{w}^T \bar{x} + w_0}{\|\bar{w}\|}$

If $d > 0 \rightarrow \text{label} = +1$

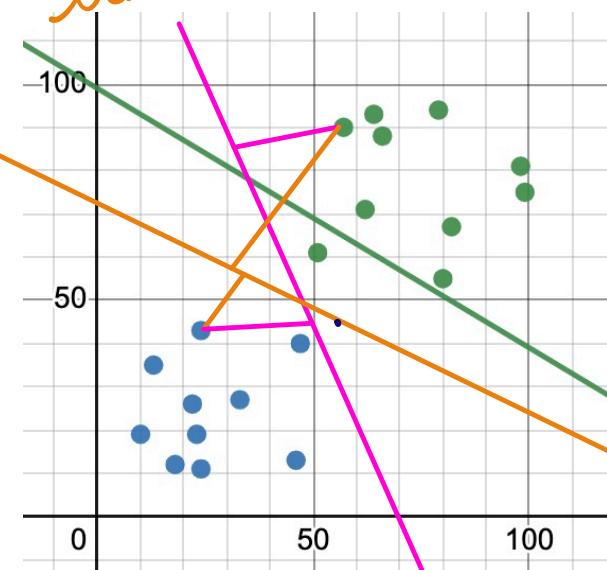
If $d < 0 \rightarrow \text{label} = -1$

Q2] Which line in the figure below is a better classifier?

Let's take the distances between our feature vectors and the line.



We want to minimize the distances between our feature vectors and the line.



option 2
↳ 100% ✓

option 3
↳ 100% ✓

In Math:

① Dataset: $X = \{(\bar{x}_i, y_i)\}_{i=1}^n$ labels.
 ↓ feature vectors

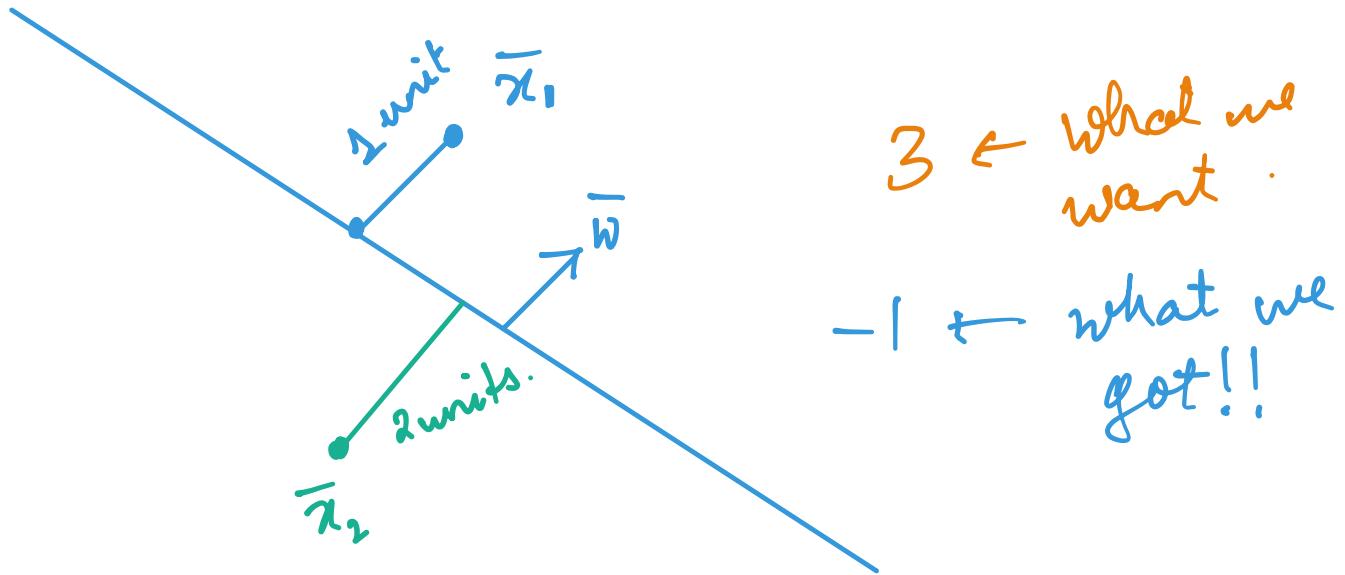
$$y_i = \begin{cases} +1 & - \text{fve H.S.} \\ -1 & - \text{ve H.S.} \end{cases}$$

② Classifier: $\bar{w}^\top \bar{x} + w_0 = 0$.

③ Sum of all distances between classifier and feature vectors.

$$L(X; \bar{w}, w_0) = \sum_{i=1}^n \left(\frac{\bar{w}^\top \bar{x}_i + w_0}{\|\bar{w}\|} \right) y_i$$

→ -ve → -ve



$$y(x; \bar{w}, w_0) = \cancel{3} + 1 + -2.$$

$$= \underline{-1}$$

not differentiable

- ① Take abs. value of distances. $f(x) = |x|$.
- ② Take square \leftarrow Taking derivative of this is more complex.
- ③ Multiply by the labels -

To automate the process of finding the best classifier, we must solve the following problem:

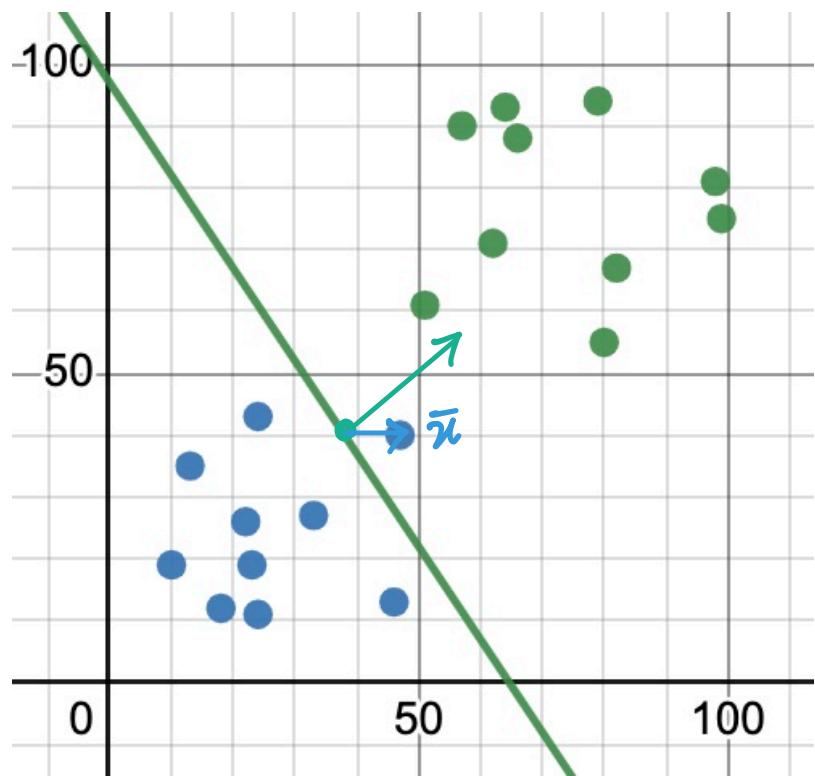
$$\max_{\bar{w}, w_0} g(X; \bar{w}, w_0)$$

$$= \max_{\bar{w}, w_0} \sum_{i=1}^n \left(\frac{\bar{w}^\top \bar{x}_i + w_0}{\|\bar{w}\|} \right) y_i$$

How to get loss function from gain function?

$$l(x; \bar{w}, w_0) = -g(x; \bar{w}, w_0)$$

Perceptron learning algorithm:



Stop when
 \bar{w} is not
updated.

$$x + y - 98 = 0$$

\bar{x} is misclassified
and

\bar{x} is misclassified if \bar{x} 's actual label: -1
and assigned label: +1

Q] How to detect if \bar{x} is misclassified?

if $y_i = \text{sign}(\frac{\bar{w}^T \bar{x} + w_0}{\|\bar{w}\|})$:
print("correct")

else:
print("incorrect").

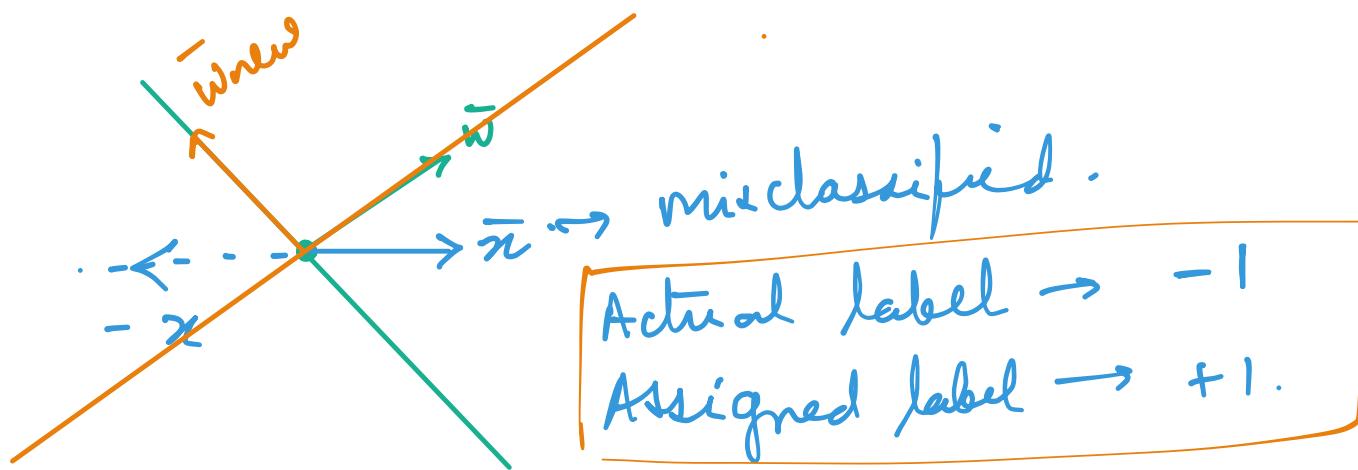
Q] How to update \bar{w} to fix this?

$$\textcircled{1} \quad \bar{w}_{\text{new}} = \bar{w} + \bar{x}$$

if \bar{x} 's actual label: +1
assigned label: -1

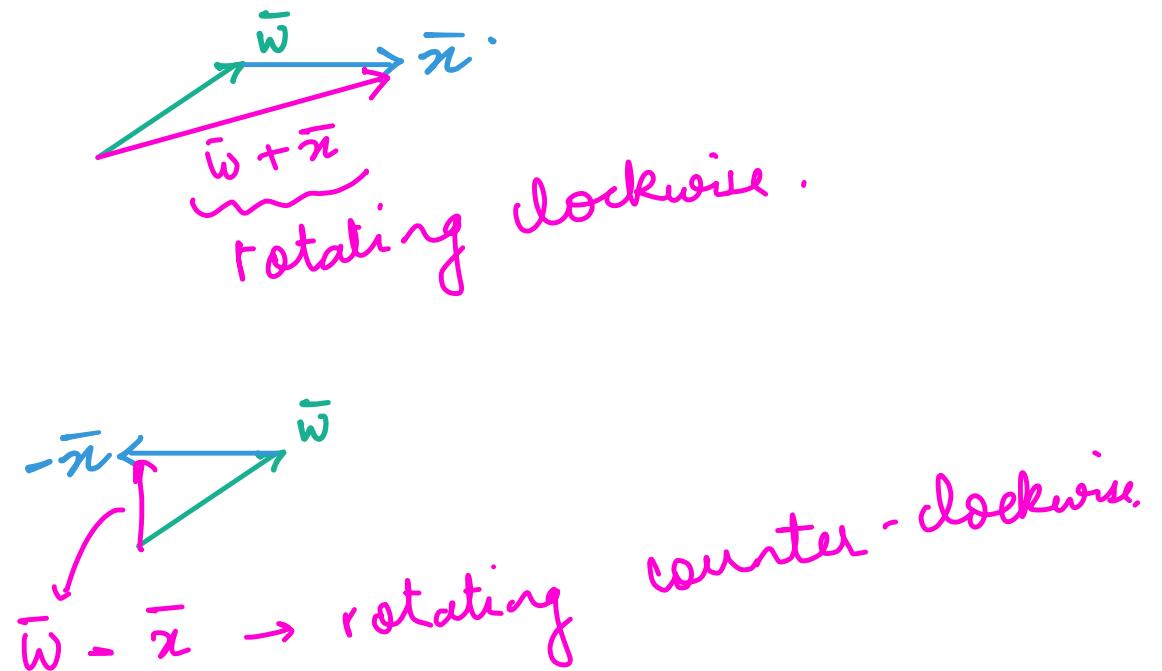
$$\textcircled{2} \quad \bar{w}_{\text{new}} = \bar{w} - \bar{x}$$

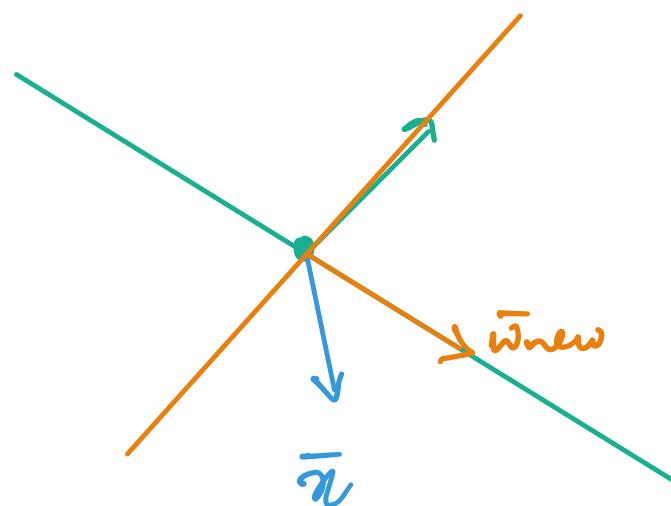
\bar{x} is misclassified if \bar{x} 's actual label: -1
and assigned label: +1



$$\textcircled{1} \quad \bar{w}_{\text{new}} = \bar{w} + \bar{x}$$

$$\textcircled{2} \quad \bar{w}_{\text{new}} = \bar{w} - \bar{x}$$





Actual label : +1

Assigned label : -1

$$\bar{w}_{\text{new}} = \bar{w} + \bar{x}$$

Prefrontal learning code (approx.)

How to calculate y :

$$(x-a)^2 + (y-b)^2 - r^2 = 0.$$

