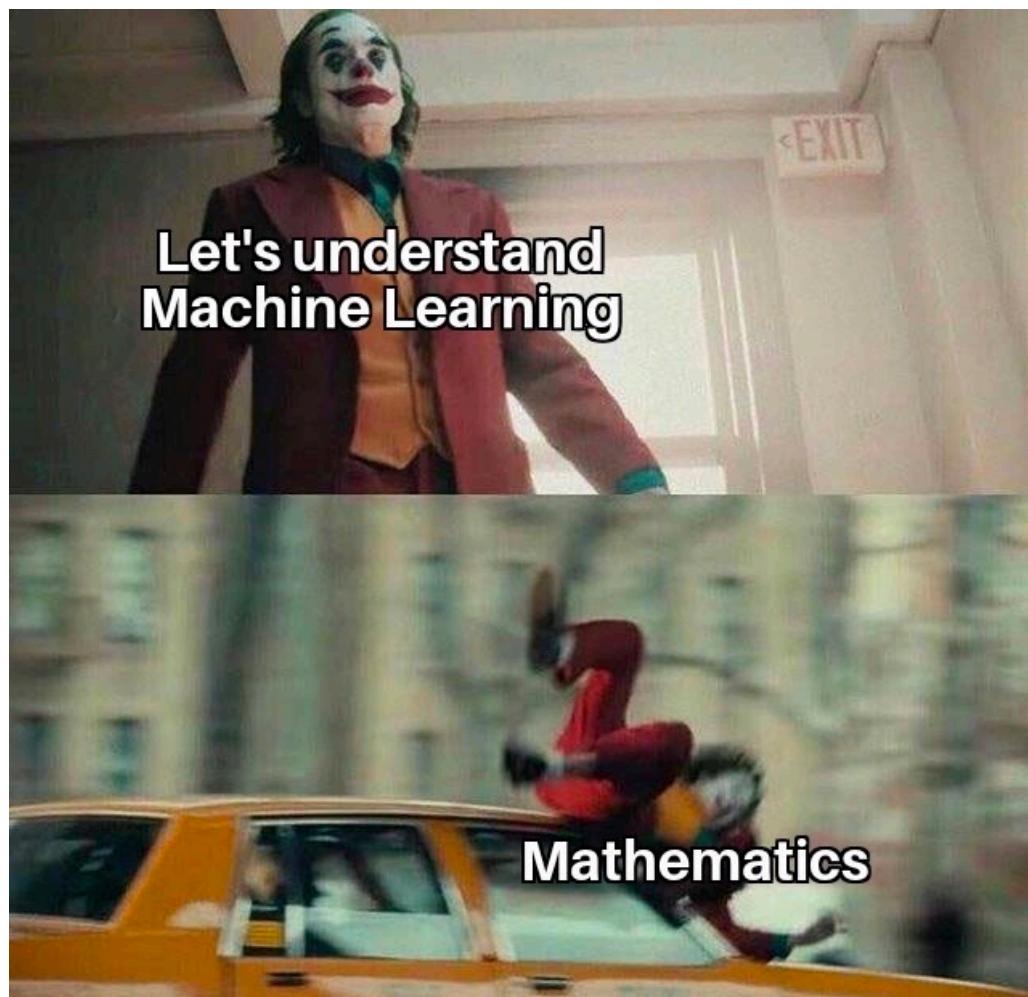


February 18, 2023

DSML : Math for ML.

Linear Algebra : Dot product and Hyperplanes.



## Recap:

\* The Machine learning Context : training classifiers.

\* New terms :

- (a) Feature
- (b) Label
- (c) Datapoint
- (d) Dataset
- (e) Classifier.

\* How to train a classifier?

Optimization !!



↑  
5 classes to study  
this!

dines. { \*  $y = mx + c \rightarrow$  <sup>slope</sup>  <sub>$x$</sub>  <sup>y-intercept</sup> .

\*  $w_1 x_1 + w_2 x_2 + w_0 = 0.$  ←

↑      ↑      ↑      ↓      ↓      ↓  
parameters.      features.

Recap: let's optimize a classifier by hand.

$x + y = 98 \rightarrow$  the line which.

$$w_1 x_1 + w_2 x_2 + w_0 = 0$$

$$\rightarrow w_1 = 1, w_2 = 1, w_0 = -98.$$

width of fish  
length of fish.

$x_1$	$y_1$	$z_1$
62	71	31
66	88	36
80	55	35
99	75	.
98	81	.
79	94	.
51	61	.
57	90	.
82	67	.
64	93	.

- ① Will the line equation be able to separate these out? No!
- ② What can do this? A plane.
- ③ What is the equation of a plane?

(2 D) :  $w_1 x_1 + w_2 x_2 + w_0 = 0 \rightarrow$  general "line"  
equation.  
A 2-D Hyperplane.

(3 D) :  $w_1 x_1 + w_2 x_2 + w_3 x_3 + w_0 = 0 .$   
↓  
general "plane"  
equation  
OR.  
A 3-D hyperplane.

(d D) :  $w_1 x_1 + w_2 x_2 + \dots + w_d x_d + w_0 = 0 .$   
↓  
general equation for a  
d-dimensional hyperplane!

## Linear Algebra: Vectors:

list, tuple, numpy arrays etc.

What is a vector?

It is a collection of numbers.  
array.

How to represent them? Coding

$\bar{x}$  ← the bar denotes a vector.

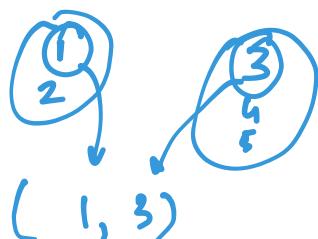
if  $\bar{x}$  has d dimensions, we can write  $\bar{x}$  as:

$$\bar{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_d \end{bmatrix}$$

$\boxed{\bar{x} \in \mathbb{R}^d}$  → it means "belongs to"

$\mathbb{R} \rightarrow$  The set of all real numbers.

$$\bar{x} = [x_1 \ x_2 \ x_3 \dots \ x_d] \rightarrow \text{row vector.}$$



## Linear Algebra: Vectors

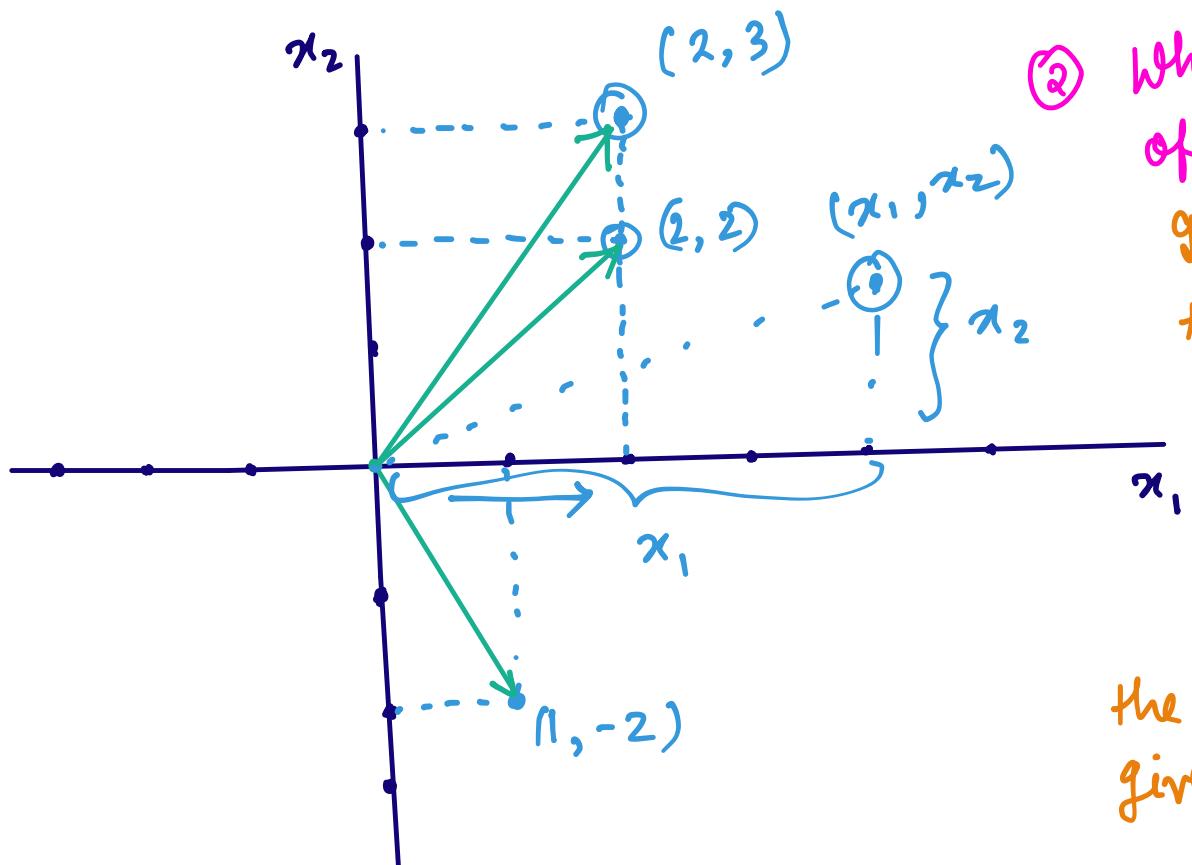
$$\bar{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

① How to visualize?

$$\bar{x}_1 = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

$$\bar{x}_2 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$\bar{x}_3 = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$



② What is the magnitude of the vector?

It is the length of the line joining the origin to the point.

For 2-D vectors, the magnitude can be given by :  $\sqrt{x_1^2 + x_2^2}$

## Linear Algebra: Vectors

③ How to get magnitude of the vector in d-dimension?

For 2-D,  $\bar{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

magnitude ( $\bar{x}$ ) =  $\sqrt{x_1^2 + x_2^2}$ .

For 3-D,  $\bar{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$

magnitude ( $\bar{x}$ ) =  $\sqrt{x_1^2 + x_2^2 + x_3^2}$

For d-D,  $\bar{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_d \end{bmatrix}$

magnitude ( $\bar{x}$ ) =  $\sqrt{x_1^2 + x_2^2 + x_3^2 + \dots + x_d^2}$

## Linear Algebra: Norm of a vector

- \* The norm of a vector is its length!!  
↓  
magnitude.
- \* We write the norm of a vector using the following symbols:  
norm of  $\bar{x}$   $\rightarrow \|\bar{x}\|$
- \* The formula for the  $L_2$  norm of a vector is:

$$\|\bar{x}\| = \sqrt{x_1^2 + x_2^2 + \dots + x_d^2}$$

we will mostly use this:

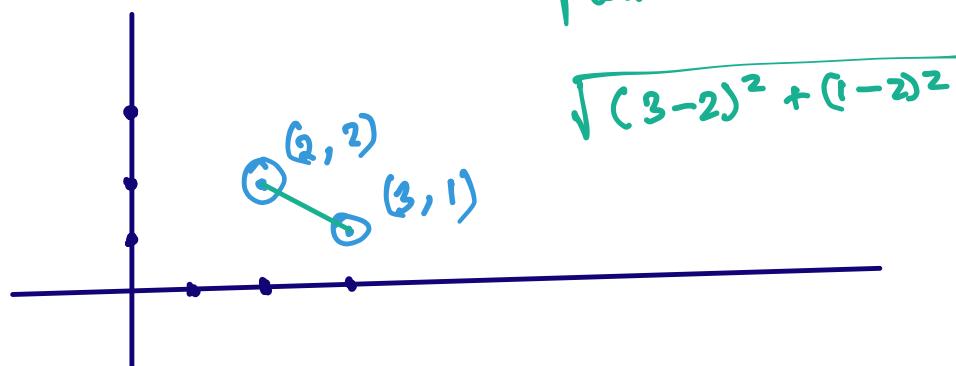
$$\|\bar{x}\| = |x_1| + |x_2| + \dots + |x_d|$$

↑  
 $L_1$  norm, similar to manhattan.

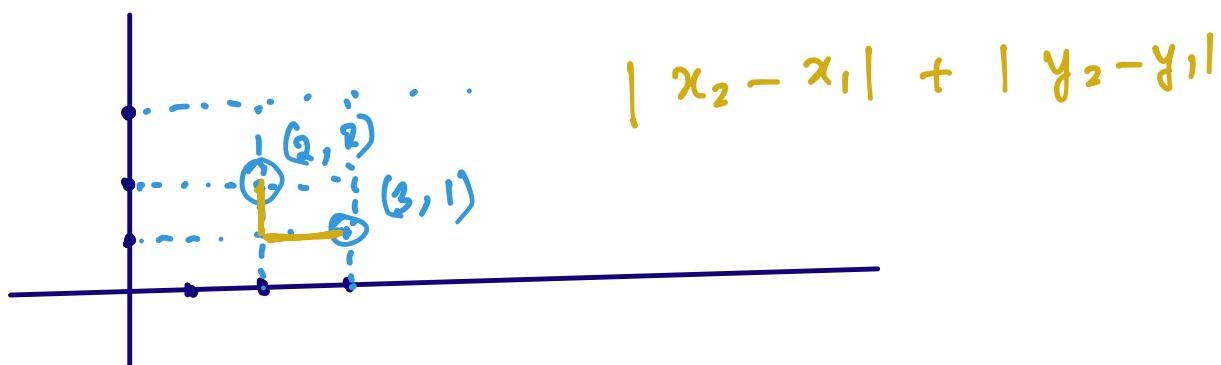
$\rightarrow L_2$  norm  
similar to euclidean.

## Different types of distances:

- ① Euclidean distance: Shortest path between two points.



- ② Manhattan distance: The distance by grid-shaped roads on the 2-D plane.



## Linear Algebra: Matrix multiplication.

$$\cdot \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}_{2 \times 2} \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}_{2 \times 2} = \begin{bmatrix} (1 \times 5 + 2 \times 7) & (1 \times 6 + 2 \times 8) \\ (3 \times 5 + 4 \times 7) & (3 \times 6 + 4 \times 8) \end{bmatrix} = \begin{bmatrix} 19 & 22 \\ 43 & 50 \end{bmatrix}.$$

Dimensions of a matrix: number of rows  $\times$  number of columns.

- \* To do matrix multiplication, we have to check if:  $n = a \cdot$  ← this holds.

$A_{m \times n}$

$B_{a \times b}$

Matrix multiplication is possible only if

Transpose:

$$\begin{bmatrix} \bar{x} \\ 1 \\ 2 \end{bmatrix}_{2 \times 1}$$

$$\begin{bmatrix} \bar{y} \\ 3 \\ 4 \end{bmatrix}_{2 \times 1}$$

If  $\bar{x}$  is a vector, its transpose is written as  
 $\bar{x}^T$

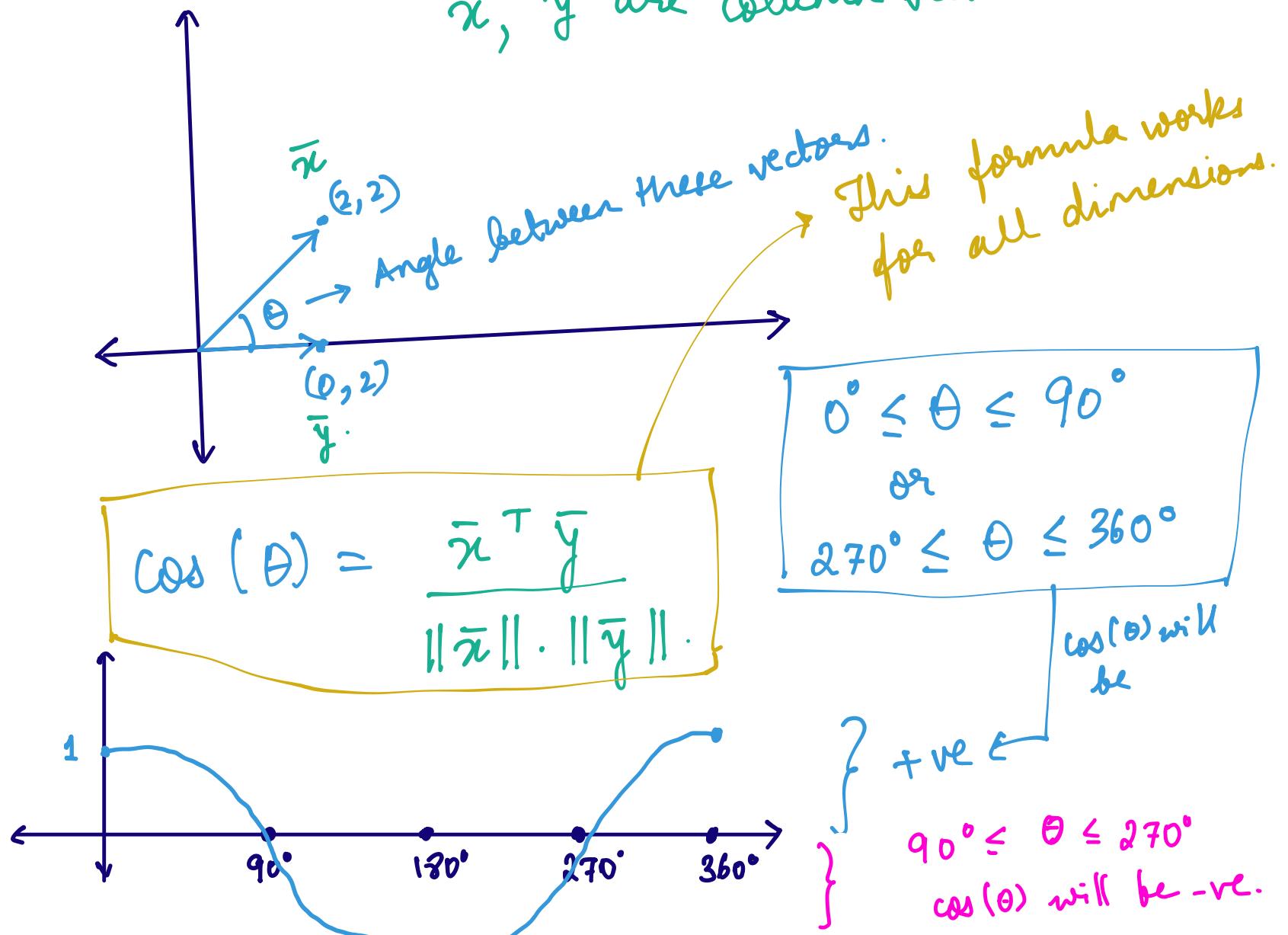
$$[ \textcircled{1} \quad \textcircled{2} ]_{1 \times 2} \begin{bmatrix} \textcircled{3} \\ \textcircled{4} \end{bmatrix}_{2 \times 1} = [ 1 \times 3 + 2 \times 4 ] \\ = 3 + 8 \\ = 11$$

This operation is called the dot product.

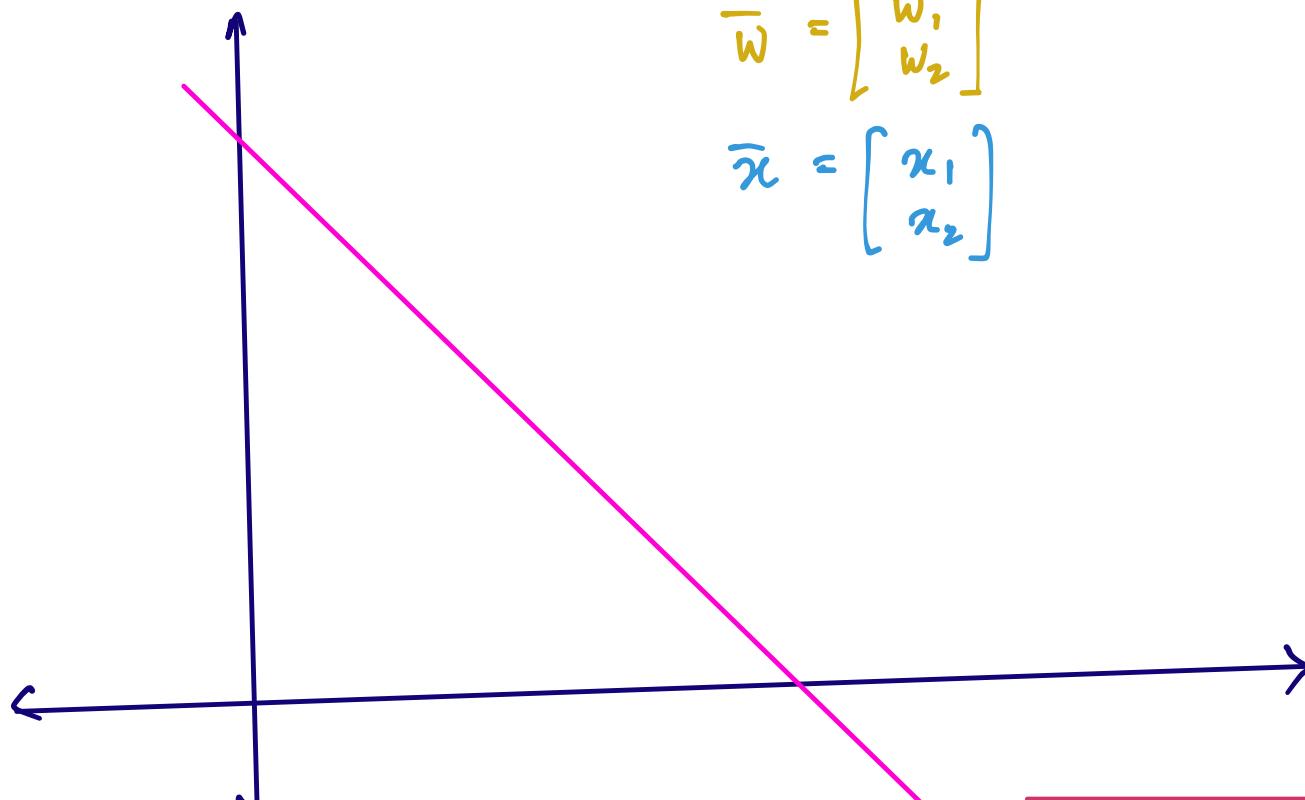
$\bar{x}^T \bar{y} \rightarrow$  notation for dot - product .

## Linear Algebra: Angle between two vectors.

$\bar{x}, \bar{y}$  are column vectors.



Crucial connection between co-ordinate geometry  
and linear algebra.



$$\bar{w} = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$$

$$\bar{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

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

$d_1 : w_1 x_1 + w_2 x_2 + w_0 = 0.$

$$\bar{w}^T \bar{x} = \bar{x}^T \bar{w}$$

Interpret  
this as  
follows.

Can we do this for the d-dim hyperplane?

$$(d \text{ D}) : w_1 x_1 + w_2 x_2 + \dots + w_d x_d + w_0 = 0.$$

$$\bar{w} = \begin{bmatrix} w_1 \\ w_2 \\ w_3 \\ \vdots \\ w_d \end{bmatrix}$$

$$\bar{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_d \end{bmatrix}$$

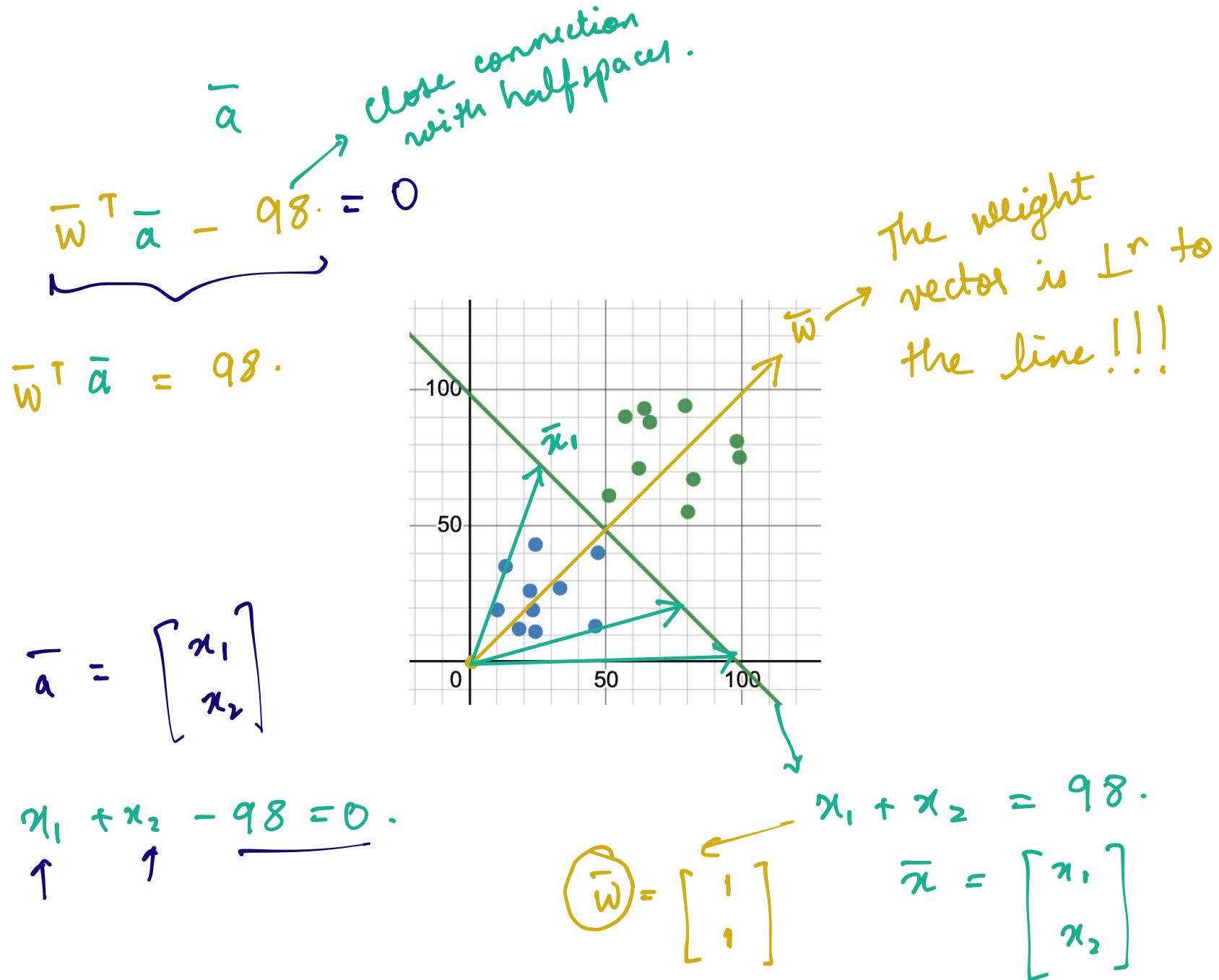
feature vector.

↳ weight vector.

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



Bias term.



## Cheat sheet:

①  $\bar{x}, \bar{y}$  are vectors.

If their dimension is  $d$ , then:

(i)  $\bar{x}, \bar{y} \in \mathbb{R}^d$ .

(ii)  $\bar{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_d \end{bmatrix}$ ,  $\bar{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_d \end{bmatrix}$

②  $\|\bar{x}\| = \sqrt{x_1^2 + x_2^2 + \dots + x_d^2}$

③  $\bar{x}^\top \bar{y} \rightarrow$  Dot product.

$$= \sum_{i=1}^d x_i \cdot y_i.$$

④ If angle between  $\bar{x}$  &  $\bar{y}$  is  $\theta$ ,

$$\cos \theta = \frac{\bar{x}^\top \bar{y}}{\|\bar{x}\| \cdot \|\bar{y}\|}$$

Let's say, we don't know  $\theta \rightarrow$  angle between  $\bar{x}$  and  $\bar{w}$ . We want to find  $\theta$ .

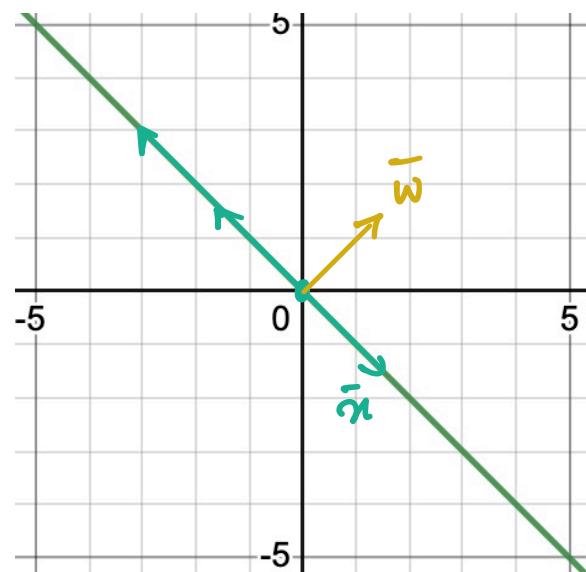
$$\boxed{\bar{w}^T \bar{x} = 0.} \quad \textcircled{1}$$

Substitute ① in

② /  
we have

$$\cos(\theta) = \frac{0}{\|\bar{x}\| \cdot \|\bar{w}\|}$$

$$\begin{aligned} \cos(\theta) &= 0. \\ \therefore \theta &= 90^\circ. \end{aligned}$$



$$\boxed{\cos(\theta) = \frac{\bar{w}^T \bar{x}}{\|\bar{x}\| \|\bar{w}\|}} \quad \textcircled{2}$$