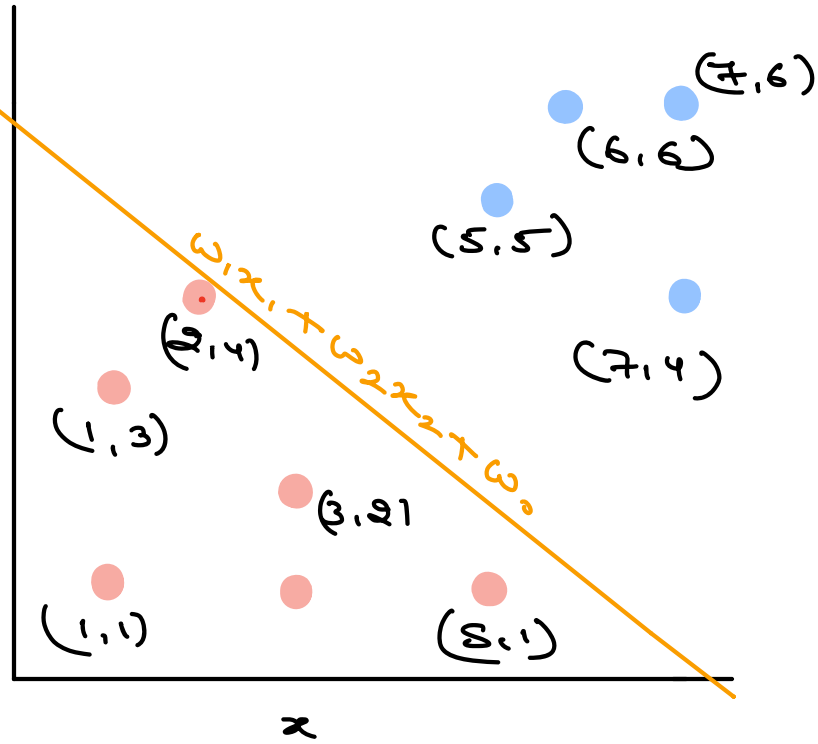


Agenda

- Revision
- Goal
- Vectors
 - ⇒ intro to Vectors
 - ⇒ Representation of Vectors
 - ⇒ Visualization of Vectors
 - ⇒ Magnitude of Vectors
 - ⇒ Norms: $L1$ and $L2$
 - ⇒ Dot product of Vectors
 - ⇒ Matrix Multiplication
 - ⇒ Angle b/w 2 vectors
 - ⇒ Connection b/w geometry and LA
 - ⇒ Unit Vector
 - ⇒ Vector projection

End Goal

x_1	x_2
5	5
2	4
6	6
3	2
7	4
5	1



Find out the classifier

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

$$w_1, w_2, w_0 = ?$$

using x_1 's and x_2 's

③ w_1, w_2, w_0 ③ weights / parameters

③ x_1 's and x_2 's ③ features / data points

Vectors

① magnitude

② Direction

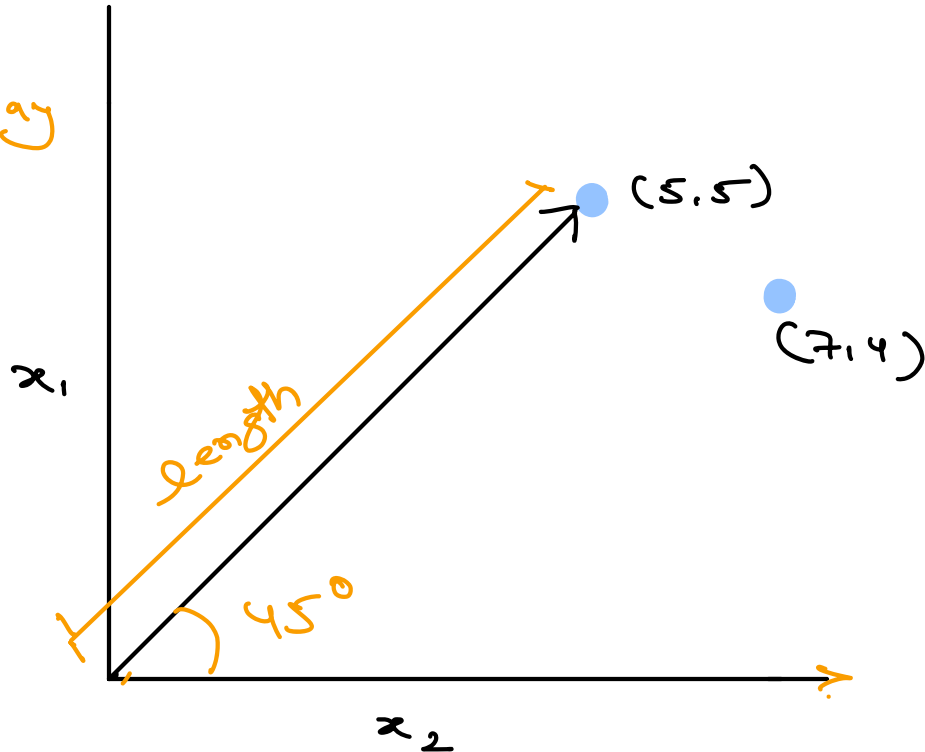
\vec{x} or \bar{x}

$(5, 5)$
 \vec{x}

$\begin{bmatrix} 5 \\ 5 \end{bmatrix}$

③ list, tuple, np-array

$\vec{x} = (5, 5)$



$\bar{x} \in 2, 6, 10, 12$

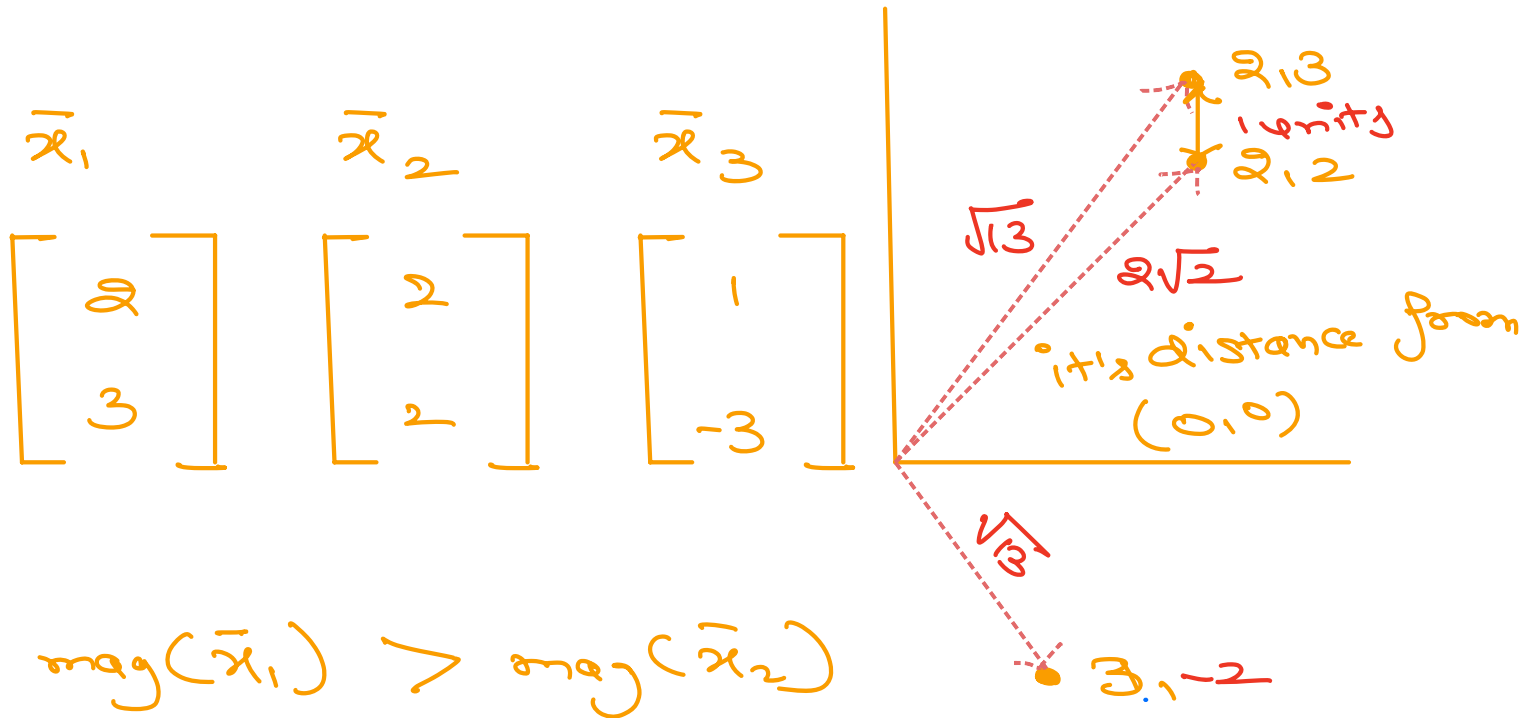
$\begin{bmatrix} 2 \\ 6 \\ 10 \\ 12 \end{bmatrix}$

← Column Vector

$\|\vec{x}\|$

$[2 \ 6 \ 10 \ 12]$ ← row Vector

Norm of Vector



* Distance b/w two points

Euclidean Distance Formula

x_1, y_1 and x_2, y_2

$$\sqrt{(x_2 - \underbrace{x_1}_0)^2 + (y_2 - \underbrace{y_1}_0)^2}$$

$(2, 2)$
 $(2, 3)$

$$\sqrt{(2-2)^2 + (3-2)^2} = \sqrt{1^2} = 1$$

magnitude of $\sqrt{x_1^2 + y_1^2}$

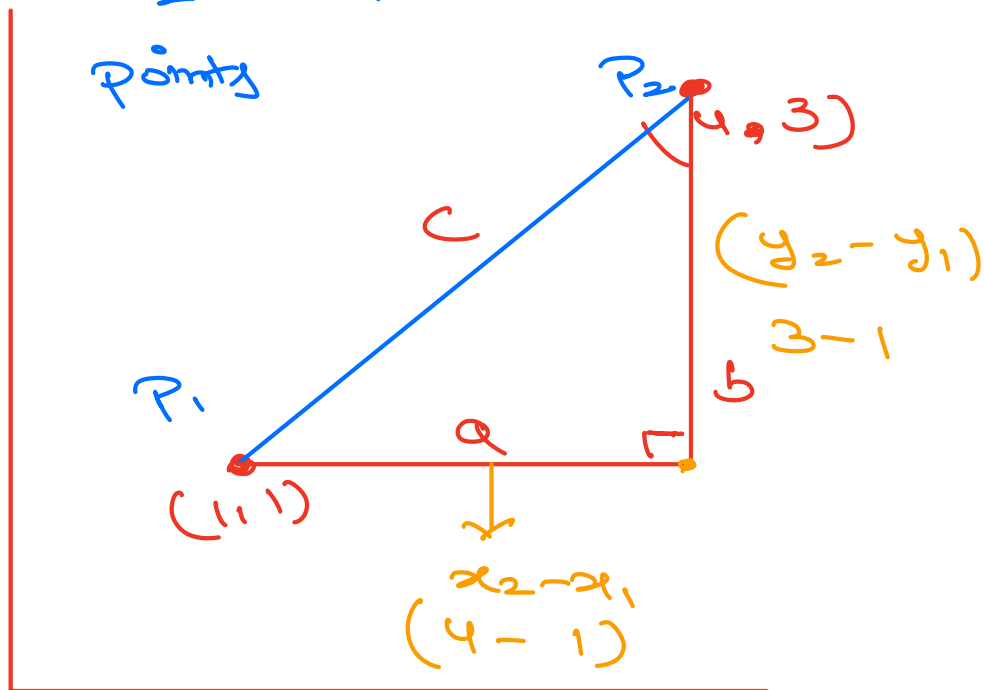
magnitude of Vector is also known as **Norm** of Vector

$$L_2 \text{ Norm} = \sqrt{x_1^2 + y_1^2} \quad (\text{Euclidean}) \\ \text{Norm}$$

$$L_1 \text{ Norm} \equiv |x_1| + |x_2| \quad (\text{Manhattan}) \\ \text{Norm}$$

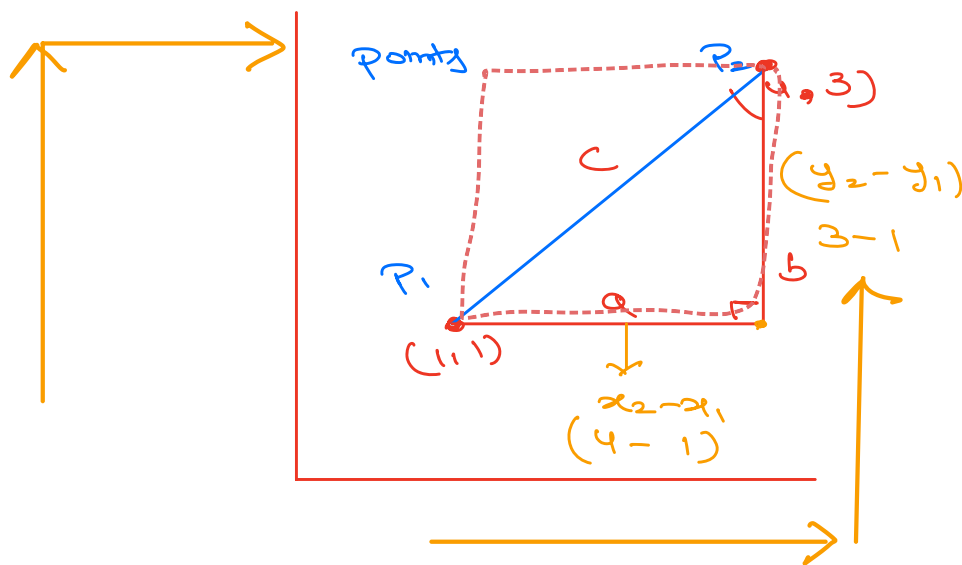
Q $(2, 4, 1, 3) \Rightarrow L_1 \equiv 10$
 $L_2 = \sqrt{30}$

L_2 Norm = Shortest Distance b/w points



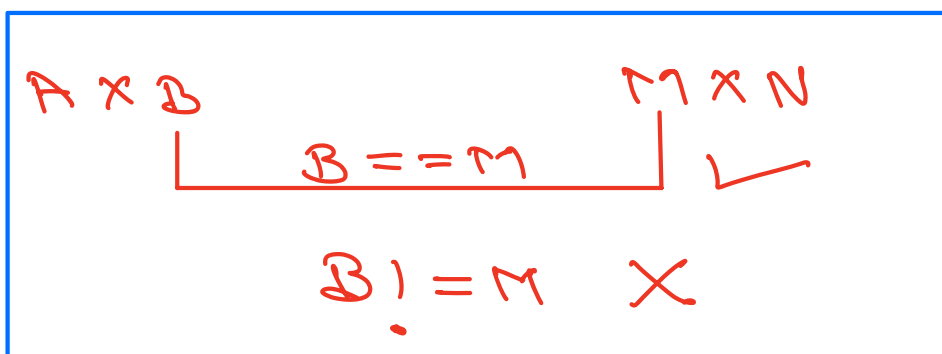
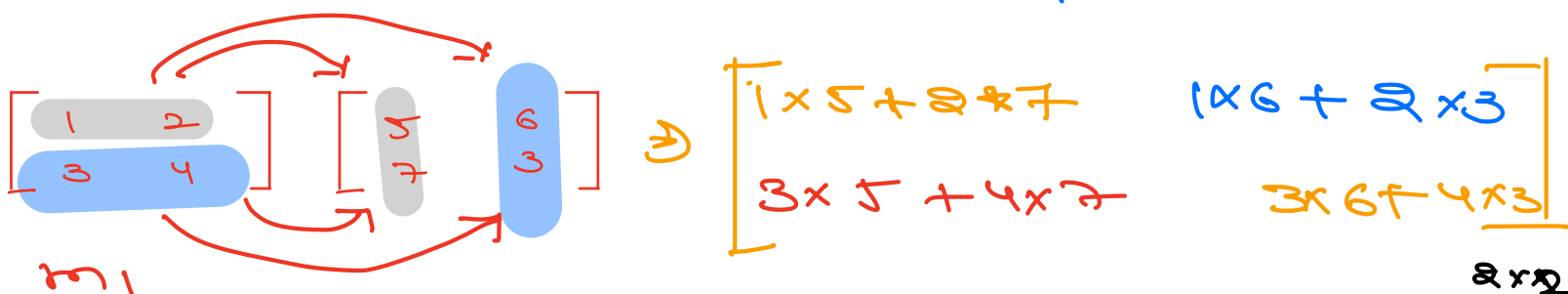
$$a^2 + b^2 = c^2$$

$$c = \sqrt{a^2 + b^2}$$



$L1 \text{ dist} \equiv a + b = \text{manhattan Distance}$

Matrix Multiplication



Validity Condition

Output $\Rightarrow A \times N$

Dot product of Vector

$$\vec{x} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

2×1

$$\vec{y} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

2×1

----- ✗ -----

① $\vec{y}^T = \begin{bmatrix} 3 & 4 \end{bmatrix}_{1 \times 2} \rightarrow \vec{x} * \vec{y}^T$
 $2 \times 1 * 1 \times 2$
 $\Rightarrow 2 \times 2$

② $\vec{x}^T = \begin{bmatrix} 1 & 2 \end{bmatrix}_{1 \times 2} \rightarrow \vec{x}^T * \vec{y}$
 $1 \times 2 * 2 \times 1 \Rightarrow 1 \times 1$



$\vec{x}^T \vec{y} \Rightarrow$ Dot Product of Vectors

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 3 \\ 4 \end{bmatrix} \Rightarrow 11$$

$1 \times 1 \quad 1 \times 1 \Rightarrow \text{Dot product} = 1 \times 1$

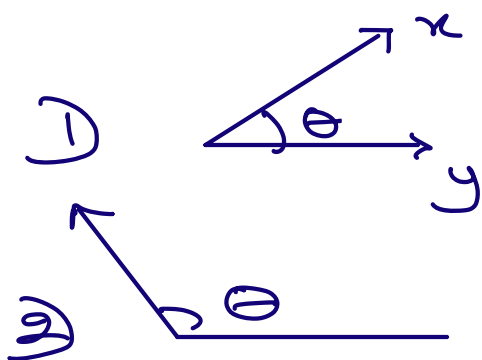
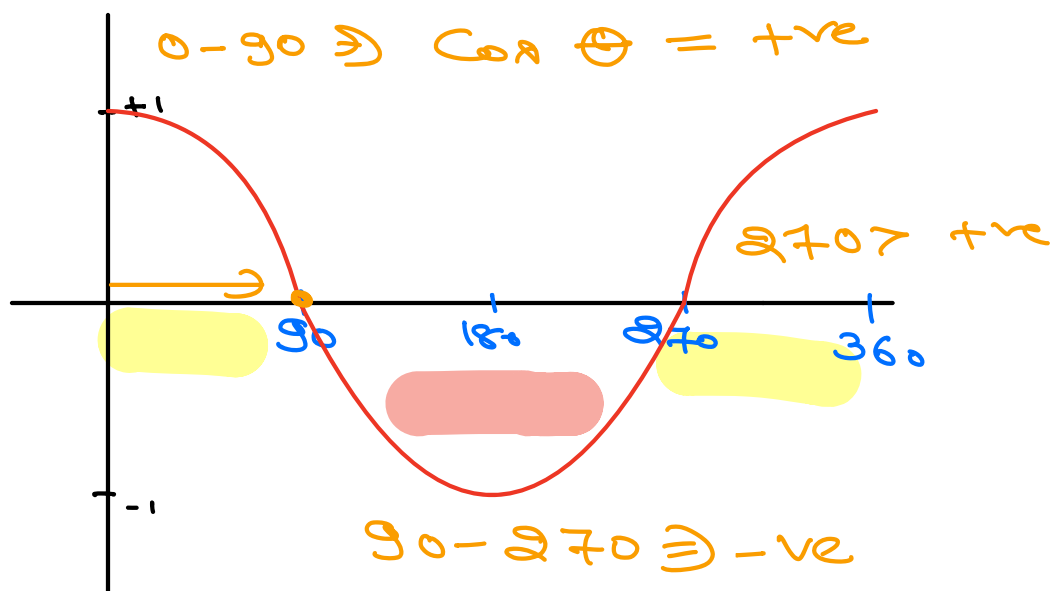
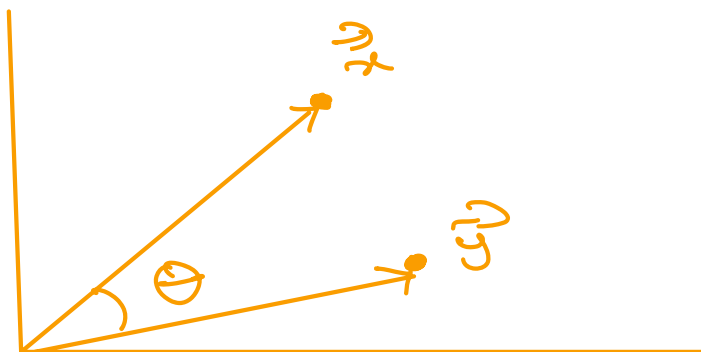
Angle b/w 2 Vectors

* Just by looking at dot product

$$\cos \theta = \frac{\vec{x} \cdot \vec{y}}{\|\vec{x}\| \|\vec{y}\|}$$

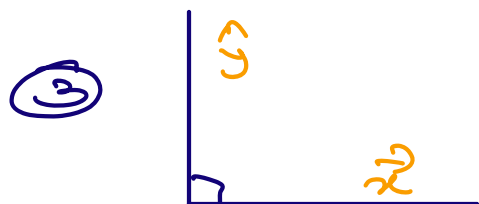
$\|\vec{x}\| \|\vec{y}\|$
L2 Norm

$\|\vec{x}\|$
L1 Norm
 $\|\vec{x}\|$
L2 Norm



$\theta < 90$ then $\vec{x} \cdot \vec{y} \Rightarrow +ve$?

$180 > \theta > 90$ then $\vec{x} \cdot \vec{y} -ve$?



$\theta = 90$ then $\vec{x} \cdot \vec{y} = 0$

* Just by looking at dot product we can tell which Quadrant the angle lies in

* Case 4: $\theta = 0^\circ$



Equation of Classifier

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

↓
Demo Classifier
 $x_1 + x_2 - 98 = 0$

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

$$\begin{bmatrix} w_1 \\ w_2 \\ w_0 \end{bmatrix}$$

weight Vector

$$\begin{bmatrix} x_1 \\ x_2 \\ 1 \end{bmatrix}$$

feature Vector

$$x_1 + x_2 - 98 = 0$$

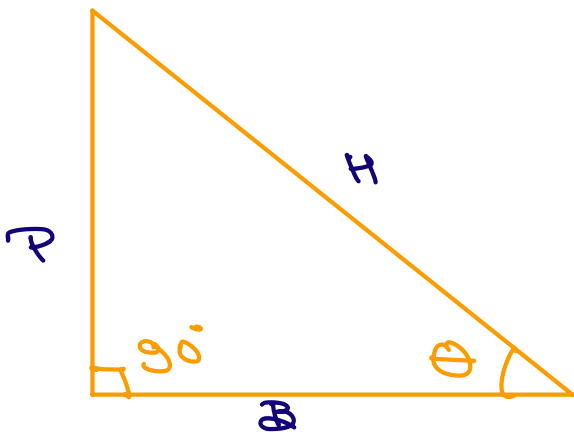
$$\begin{bmatrix} 1 \\ 1 \\ -98 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ 1 \end{bmatrix} = 0$$

$$w^T x + w_0 = 0$$

$$w^T \bar{x} = 0$$

Note: Weight Vector is always \perp to the classifier

Trigonometry of Angles

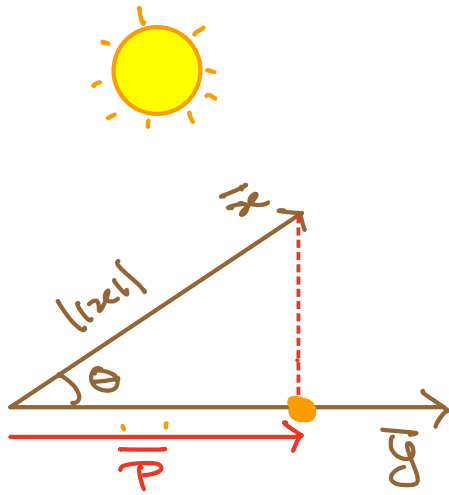


$$\tan \Theta \equiv \frac{P}{B}$$

$$\sin \Theta \equiv \frac{P}{H}$$

$$\cos \Theta \equiv \frac{B}{H}$$

Projection of Vector



Projection of Vector \vec{x} on \vec{y} is the shadow x casts on y

$$||P|| \equiv \vec{x}^T \hat{y}$$

Length of Projection is equal to Dot product b/w \vec{x} and \hat{y}

Projection of \vec{x}
on \vec{y}

Derivation of Projection

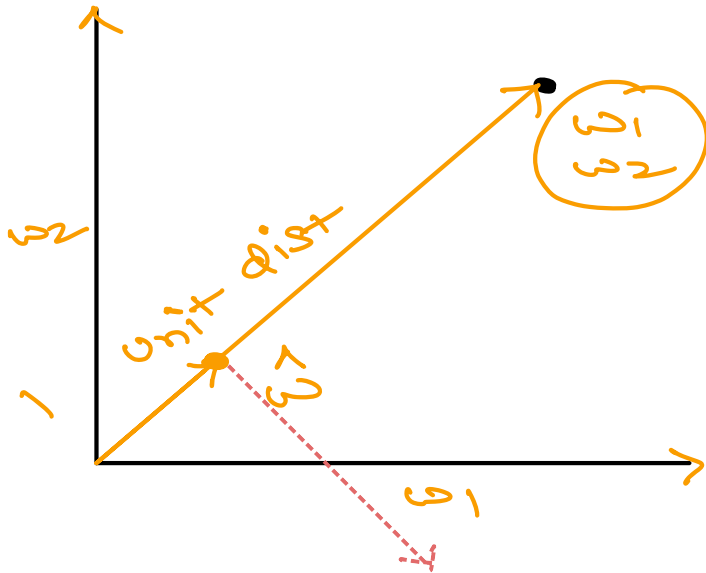
Hints

① (trigo) $\cos \theta =$

② (L.A) $\cos \theta =$

Unit Vector

Vector with Magnitude of 1



$$\text{mag} = \sqrt{w_1^2 + w_2^2}$$

$$\hat{w} \Rightarrow \left(\frac{w_1}{\sqrt{w_1^2 + w_2^2}}, \frac{w_2}{\sqrt{w_1^2 + w_2^2}} \right)$$

H.W 1

$$||\hat{w}|| \equiv ?$$

① A vector with magnitude and Same direction as Original Vector