

Features x_1 to x_6 ... Rate them by usefulness for decision making.

	x_1	x_2	x_3	x_4	x_5	x_6	
" 6 dimension data "	1	6	2	2	1	100	\rightarrow outlier
	2	5	3	4	1	1	
	3	4	4	6	1	2	
	4	3	5	8	1	1	
	5	2	6	10	1	2	
	6	1	7	12	1	1	
Negative correlation	P	Not useful	Correlated	Not useful	Not useful		
		P	Not useful	R	"Same for all examples"		
		(linear shift of x_1)			derived from x_1		

$x_1 \times x_2 =$ negatively correlated

$x_3, x_4 =$ Can be calculated from x_1

$x_5 =$ Same across all observation

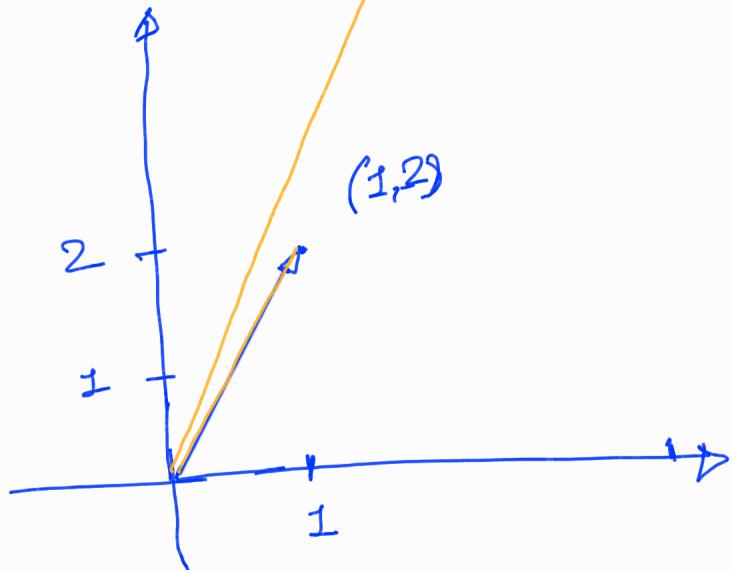
$x_6 = ?$ $\boxed{\quad}$ has an outlier

" if we know x_1, x_2, x_3, x_4 are not required; because they are not providing any more info, than x_1 ."

Why does matrix exist?

→ To torture vectors.

$$\begin{bmatrix} 2 & 3 \\ 3 & 4 \end{bmatrix} \times \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 8 \\ 11 \end{bmatrix}$$



If matrix is Thanos

for vectors; then there

is super hero vector. (Iron Man) / Captain America

- "Eigen Vector"
- When a matrix is multiplied with its Eigen vector, we get a scalar λ Eigen vector

$$Av = \lambda v \quad \begin{array}{l} \text{Eigen vector} \\ \text{Eigen value} \end{array}$$

$$A = \begin{bmatrix} -6 & 3 \\ 4 & 5 \end{bmatrix} \quad \text{find out Eigen values \& vectors.}$$

$$Av - \lambda v = 0$$

$$(A - \lambda I)v = 0$$

$$\therefore A - \lambda I = 0 \leftarrow \text{use this to find Eigen values.}$$

$$\begin{bmatrix} -6 & 3 \\ 4 & 5 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = 0$$

$$\therefore \begin{bmatrix} -6-\lambda & 3 \\ 4 & 5-\lambda \end{bmatrix} = 0$$

$$\therefore (-6-\lambda)(5-\lambda) - 12 = 0$$

$$\therefore \lambda^2 + \lambda - 42 = 0$$

$$\therefore \lambda_1 = -7, \quad \lambda_2 = 6$$

$$AV = \lambda_1 V$$

$$\begin{bmatrix} -6 & 3 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \lambda_1 \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$-6v_1 + 3v_2 = -7v_1 \therefore v_1 = -3v_2$$

$$4v_1 + 5v_2 = -7v_2 \quad v_1 = -3v_2$$

$$\boxed{V_1 = \begin{bmatrix} -3 \\ 1 \end{bmatrix} \quad \lambda_1 = -7}$$

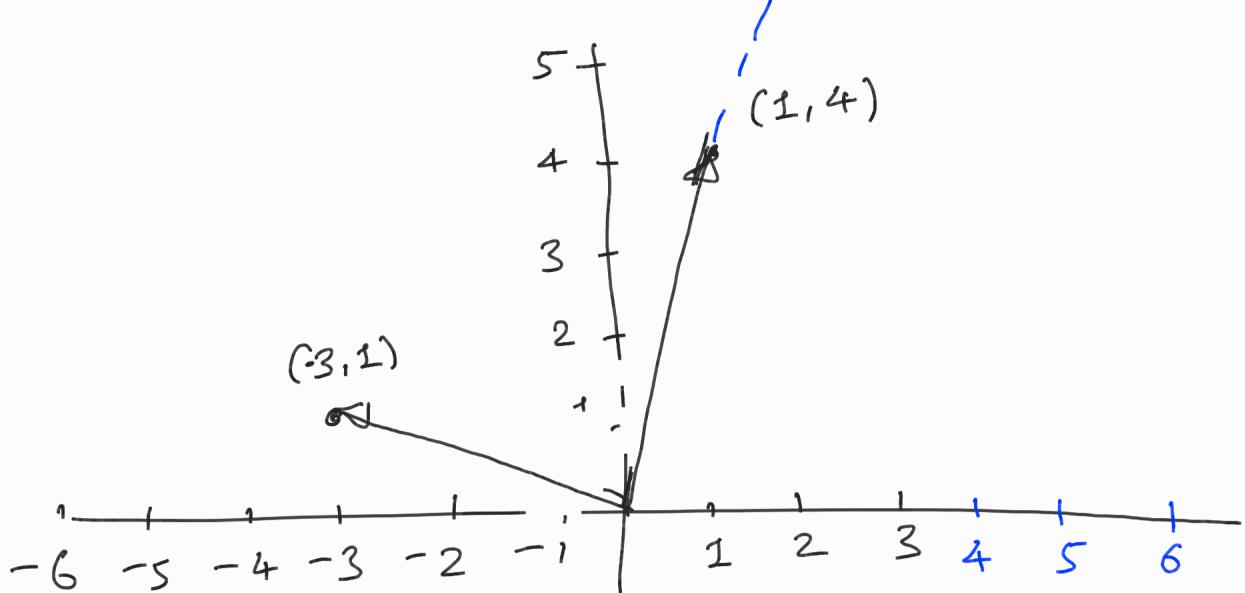
$$\begin{bmatrix} -6 & 3 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = 6 \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$-6v_1 + 3v_2 = 6v_1$$

$$3v_2 = 12v_1$$

$$\boxed{v_2 = 4v_1}$$

$$V_2 = \begin{bmatrix} 1 \\ 4 \end{bmatrix} \quad \lambda_2 = 6$$



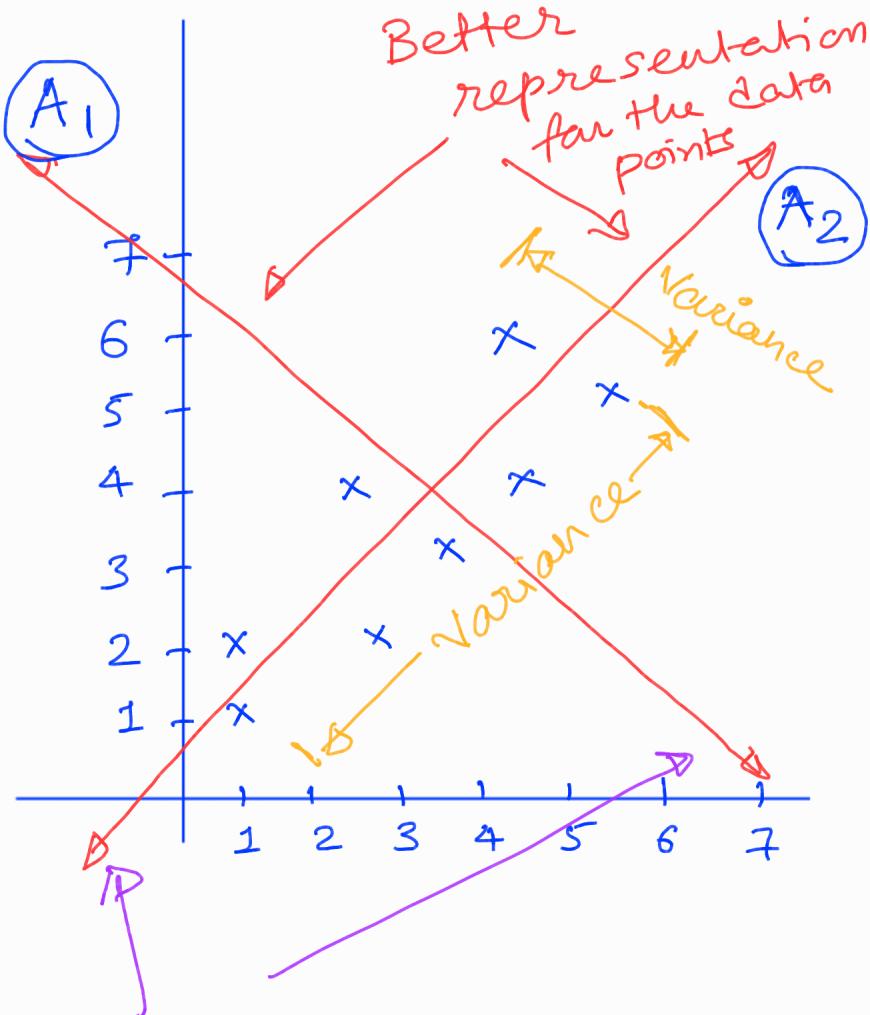
"Eigen vectors are orthogonal to each other."

"For an $n \times n$ matrix, there are n Eigen vectors."

Image

$$\begin{bmatrix} -6 & 3 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 4 \end{bmatrix} = \begin{bmatrix} 6 \\ 24 \end{bmatrix} = 6 \begin{bmatrix} 1 \\ 4 \end{bmatrix}$$

X_1	X_2
1	1
1	2
3	2
3	4
4	3
5	4
5	6
6	5



- Eigen Vectors

- Eigen values is proportion
- al to variance

Eigen value corresponding
to $A_2 >$ Eigen value
corresponding to
 A_1

- Computations of any ML algo. increases with number of dimensions.
- By reducing dimension we can avoid unnecessary computations.
- We want to keep important attributes and remove redundant or unnecessary attributes.

Race Religion gender color appearance



"Don't use this features in decision making."

x_1	x_2	$\underline{x_1 - \mu_{x_1}}$	$\underline{x_2 - \mu_{x_2}}$	$(x_1 - \mu_{x_1})^2$	$(x_2 - \mu_{x_2})^2$	$(x_1 - \mu_{x_1})(x_2 - \mu_{x_2})$
S a m P e s	1	2	-1	-1	1	1
	2	3	0	0	0	0
	3	4	1	1	<u>1</u>	<u>1</u>
$\mu_{x_1} = 2$		$\mu_{x_2} = 3$		<u>$\frac{1}{2} = 1$</u>	<u>$\frac{2}{2} = 1$</u>	<u>$\frac{2}{2} = 1$</u>

① - Remove mean

$$\bar{x}_i/n$$

② - Calculate Variance

$$\frac{\sum (x_i - \mu_{x_i})^2}{(n-1)}$$

③ - Calculate Covariance

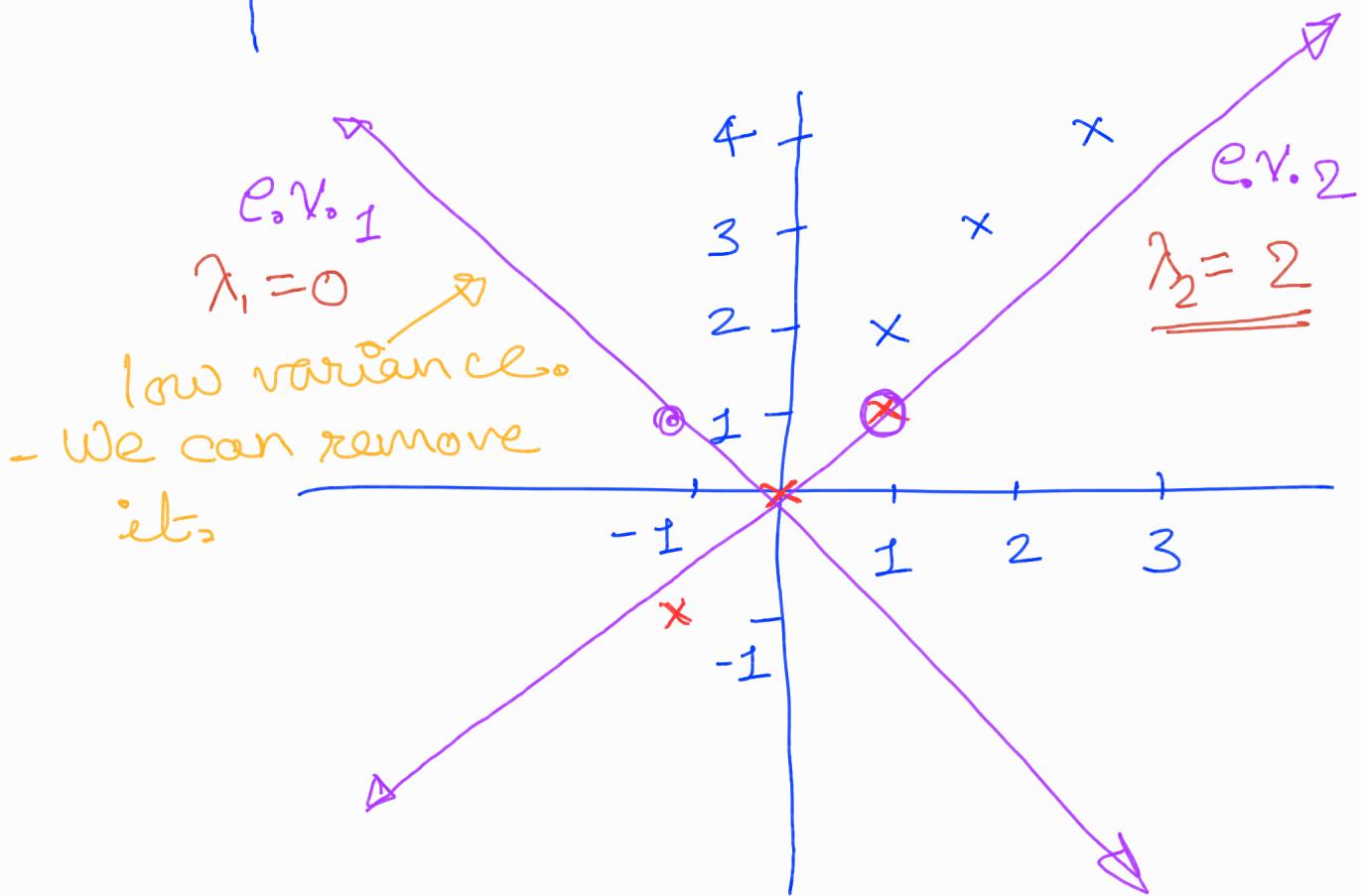
$$\frac{\sum_{i,j} (x_i - \mu_{x_i})(x_j - \mu_{x_j})}{(n-1)}$$

$$\text{Var}(x_1) = 1, \quad \text{Var}(x_2) = 1,$$

$$\text{Covar}(x_1, x_2) = 1$$

Covariance Matrix X_1, X_2

	X_1	X_2
X_1	1	1
X_2	1	1



④ Find Eigen values & vector of co-variance matrix

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = 0$$

$$\begin{bmatrix} 1-\lambda & 1 \\ 1 & 1-\lambda \end{bmatrix} = 0$$

$$\therefore (1-\lambda)^2 - 1 = 0$$

$$\therefore 1 - 2\lambda + \lambda^2 - 1 = 0$$

$$\therefore \lambda(\lambda - 2) = 0$$

Eigen values: $\lambda_1 = 0, \lambda_2 = 2$

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = 0 \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$v_1 + v_2 = 0$$

$$\boxed{v_1 = -v_2}$$

v_1

$$\begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

$$\left(-\frac{1}{\sqrt{2}}\right)^2 + \left(\frac{1}{\sqrt{2}}\right)^2 = 1$$

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = 2 \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$v_1 + v_2 = 2v_1$$

$$v_2 = v_1$$

v_2

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} \approx \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

$\lambda_1 = 0 \quad \begin{bmatrix} -1 \\ 1 \end{bmatrix}$

 $\lambda_2 = 2 \quad \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

Principal Component Analysis (PCA)

- ① Remove mean from the data
- ② Calculate Covariance matrix
- ③ Find Eigen values & vectors

from the Covariance matrix

- ④ Eigen vectors are the new principal components.
- ⑤ Arrange Eigen vectors in descending order of Eigen values.
- ⑥ Remove Eigen vector with low Eigen values to reduce dimension.

- Principal Components are the new features constructed from the original features.

- ① Eigen values & vectors are difficult to compute.
- ② Co-variance matrix is square matrix ($n \times n$)