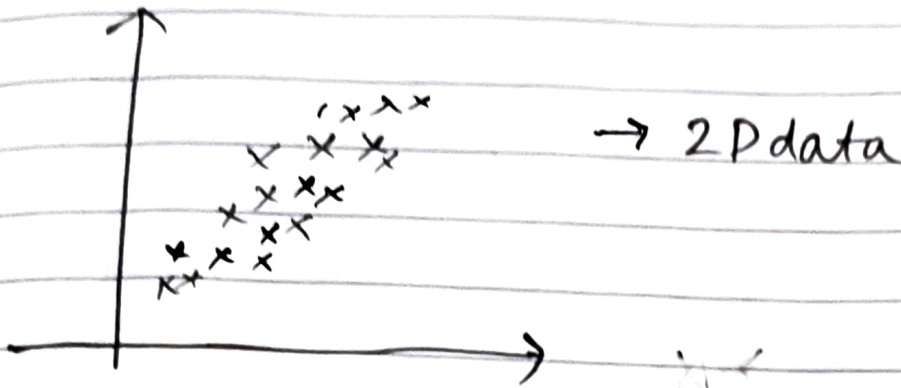
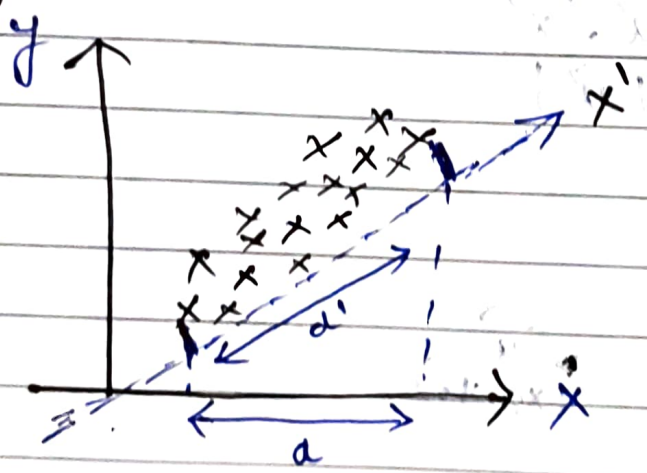


# Problem formulation



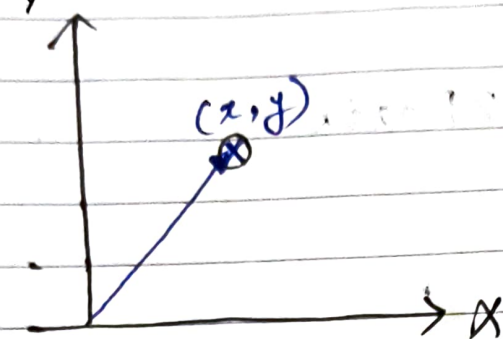
- You have to find that line in 1D, which will give as good results it would give in 2D
- Maybe, this axis can be that line  $x'$



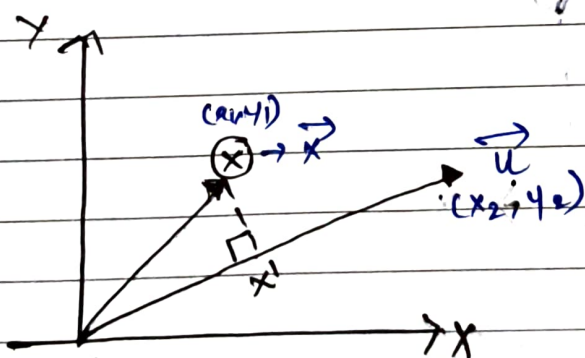
Because the variance on  $x$  axis is almost same as on  $x'$  axis

Exactly what does PCA solve understanding with single pt.

- Single pt



- $\otimes \rightarrow$  This pt is a vector  $\vec{x}$
- You have to find a unit vector  $\vec{u}$
- Project vector  $\vec{x}$  on  $\vec{u}$
- Projection =  $\frac{\vec{u} \cdot \vec{x}}{|\vec{u}|}$



- Since our  $\vec{u}$  is unit vector  
 $\therefore$  projection of  $\vec{x}$  on  $\vec{u}$   
 is  $\Rightarrow \frac{\vec{u} \cdot \vec{x}}{|\vec{u}|} = \vec{u} \cdot \vec{x}$
- $\vec{u} \cdot \vec{x} \Rightarrow \vec{u}^T \vec{x}$

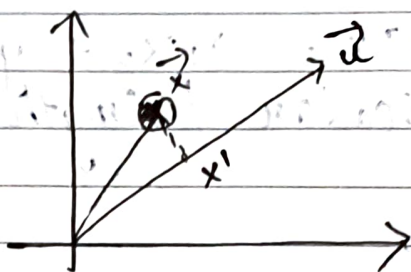
After projecting  $\vec{x}$  on  $\vec{u}$ , you get new vector. you can call it  $\vec{x}'$





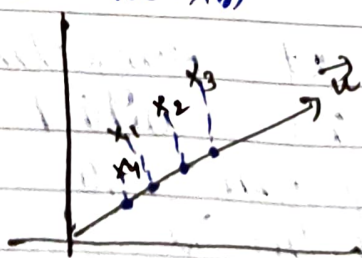
- There could be any unit vectors  
But we will choose <sup>that</sup> unit vector  
for which our variance is maximum.

- calculating variance using same example.



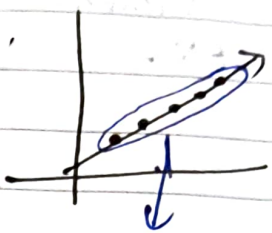
$$\rightarrow x' = u^T x$$

- ~~So~~ There will be other pts too
- for  $\vec{x}$ ; we got  $u^T x$
- Let's say this  $\vec{x}$  is pt  $x_1$ .
- Projection of  $x_1$  on  $\vec{u} \Rightarrow u^T x_1$
- Similarly there are other pts  $x_2, x_3, x_4$
- for  $x_2 = u^T x_2$   
 "  $x_3 = u^T x_3$   
 "  $x_4 = u^T x_4$   
 $x_n = u^T x_n$

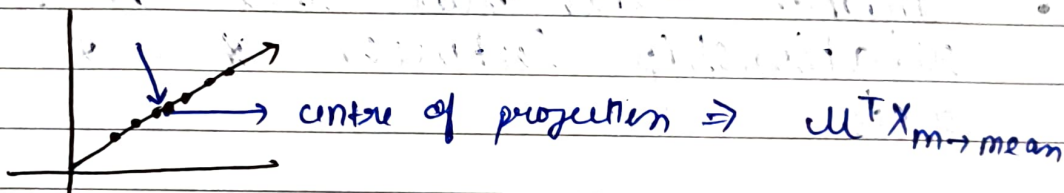


- Now we have to calculate variance  

$$\text{Variance} = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2$$

- $X_i \rightarrow$  each of individual projections  $\rightarrow (u^T X_i)$   
~~\* for  $X_i \rightarrow X$~~
  - for  $i=1 \Rightarrow u^T X_1$   
 $i=2 \Rightarrow u^T X_2$   
& all these are scalar
- 
- we are trying to calculate variance of this projection

- $\bar{X} \rightarrow$  mean / centre of projection



- $\Rightarrow$  Variance of projection of all points

$$= \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$$

$$= \boxed{\frac{1}{n} \sum_{i=1}^n (u^T X_i - u^T X_m)^2}$$

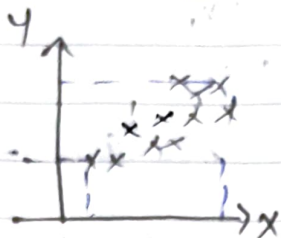
- Now, mathematically we have to find a unit vector for which our variance <sup>of projection</sup> is maximum.

- Optimization problem  $\rightarrow$  maximizing the variance.

## Problem with variance

- Variance is calculated for 1 axis.

Ex -



- Variance on x axis will different  
" " y axis "
- Variance will not tell the collective relationship between 2 or more axis.
- That's where covariance comes which tells the relation b/w  $x$  and  $y$ .

## Covariance

Ex ↑ We have 2 col  $x_1$  &  $x_2$

- It's covariance will be of  $[2 \times 2]$

$$\begin{matrix} x_1 & \begin{bmatrix} \text{cov}(x_1, x_1) & \text{cov}(x_1, x_2) \\ \text{cov}(x_2, x_1) & \text{cov}(x_2, x_2) \end{bmatrix} \\ x_2 & \end{matrix}$$

$x_1 \quad x_2$



# Covariance Matrix

$$\begin{matrix} X_1 & \\ X_2 & \end{matrix} \begin{bmatrix} \text{cov}(X_1, X_1) & \text{cov}(X_2, X_1) \\ \text{cov}(X_1, X_2) & \text{cov}(X_2, X_2) \end{bmatrix}$$

$X_1 \qquad X_2$

Note:-

① Covariance with itself is variance.

$\therefore$

$$= \begin{bmatrix} \text{var } X_1 & \text{cov}(X_2, X_1) \\ \text{cov}(X_1, X_2) & \text{var } X_2 \end{bmatrix}$$

②  $\text{cov}(X_1, X_2) = \text{cov}(X_2, X_1)$

$$= \begin{bmatrix} \text{var } X_1 & \text{cov}(X_1, X_2) \\ \text{cov}(X_1, X_2) & \text{var } X_2 \end{bmatrix}$$

- Now our matrix is square and symmetrical
- diagonal elements are variance
- other elements are covariance.

## Benefit of covariance matrix

- 1) tells about variance
- 2) " " covariance which tells the relationship b/w 2 axis is -ve or +ve

Ex 2

Find covariance matrix of 3 dimensions i.e.  $x, y, z$ .

Ans

$$\begin{matrix} x \\ y \\ z \end{matrix} \begin{bmatrix} \text{var}(x) & \text{cov}(x,y) & \text{cov}(x,z) \\ \text{cov}(y,x) & \text{var}(y) & \text{cov}(y,z) \\ \text{cov}(z,x) & \text{cov}(z,y) & \text{var}(z) \end{bmatrix}$$

$x$ 
 $y$ 
 $z$