

PCA Steps

- $f_1 \quad f_2 \quad f_3 \quad \text{target}$

Step 1 : Mean centering

Step 2: find Covariance matrix

$$\begin{matrix} f_1 \\ f_2 \\ f_3 \end{matrix} \begin{bmatrix} \text{var}(f_1) & \text{cov}(f_1/f_2) & \text{cov}(f_1/f_3) \\ \text{cov}(f_2/f_1) & \text{var}(f_2) & \text{cov}(f_2/f_3) \\ \text{cov}(f_3/f_1) & \text{cov}(f_3/f_2) & \text{var}(f_3) \end{bmatrix}$$

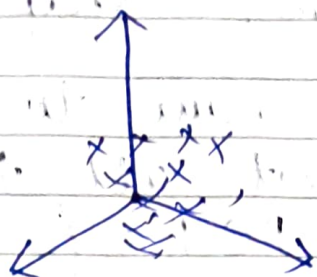
$f_1 \quad f_2 \quad f_3$

Step 3 find eigen vectors & values.

- you will get eigen vector & value for each dimension
- Eigen vector with highest eigen value $\rightarrow PC1$
- " " 2nd " " $\rightarrow PC2$
- " " 3rd " " $\rightarrow PC3$
- It's your choice, whether to choose only 1 PC ie bringing down 3d to 1D.
or choose 2 PC ie bring down 3d to 2D.

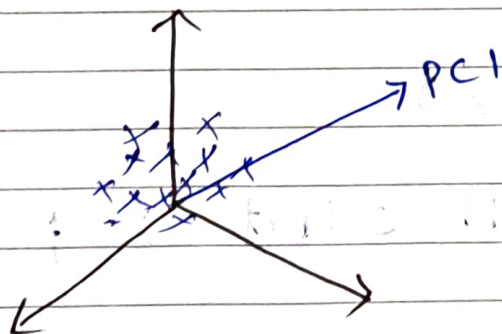
Q How to transform pts acc to no. of PC.

Ans ① Ex - you have f_1, f_2, f_3 (3D to 1D)



② You apply PCA, and you get 3 eigen vectors & value one for each dimension

③ You choose to bring data to 1D is selecting only one principle component and that PC is in below graph

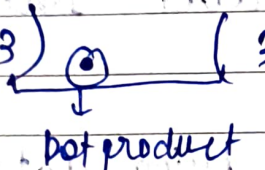


④ You want to project all the points onto PC1 that's how you will bring 3D to 1D

⑤ And you project on PC1 by

→ $u^T x$, where u is PC1 & x is data point

- Let's say our data shape is: $(1000, 3) \rightarrow 1000$ pts in 3D.
- And \vec{u} ie PC is also a vector in 3 dimension. \therefore shape $\rightarrow (1, 3)$
- Now you only have to dot product to project all those points on PC ($\cdot u^T x$)
- But you have to first transpose PC ($as \rightarrow u^T$) \rightarrow shape $\rightarrow (3, 1)$
- $(1000, 3) \otimes (3, 1)$



dot product
- You will get new data with shape $(1000, 1)$

Ex2 3D to 2D

- You will select 2 PC ie PC1 and PC2
- Data shape $\rightarrow (1000, 3)$
- Eigen vectors shape $(2, 3)$
- Transpose $\rightarrow (3, 2)$
- Dot product $\Rightarrow (1000, 3) \cdot (3, 2)$
- You will get $\rightarrow (1000, 2)$