

PCA - used for dimensionality reduction in ML

↳ high dim data to low dim.

Given the data in the table, reduce the dimension from 2 to 1 using PCA algorithm

Feature	Eg <sub>1</sub>	Eg <sub>2</sub>	Eg <sub>3</sub>	Eg <sub>4</sub>
X <sub>1</sub>	4	8	13	7
X <sub>2</sub>	11	4	5	14

Feature 1 → x<sub>1</sub>

Feature 2 → x<sub>2</sub>

Step 1:

Calculate Mean

$$\bar{x}_1 = \frac{4+8+13+7}{4} = 8\frac{1}{4}$$

$$\bar{x}_2 = \frac{11+4+5+14}{4} = 8\cdot5\frac{1}{4}$$

Step 2:

Calculate covariance matrix

$$S = \begin{bmatrix} \text{Cov}(x_1, x_1) & \text{Cov}(x_1, x_2) \\ \text{Cov}(x_2, x_1) & \text{Cov}(x_2, x_2) \end{bmatrix}$$

$$\text{Cov}(x_1, x_1) = \frac{1}{N-1} \sum_{k=1}^N (x_{1k} - \bar{x}_1)(x_{1k} - \bar{x}_1)$$

$\rightarrow$  no. of egss. =  $\frac{1}{3} ((4-8)^2 + (8-8)^2 + (13-8)^2 + (7-8)^2) = 14\frac{1}{3}$

$$\text{Cov}(x_1, x_2) = \frac{1}{N-1} \sum_{k=1}^N (x_{1k} - \bar{x}_1)(x_{2k} - \bar{x}_2)$$

$$\text{Cov}(x_1, x_2) = \frac{1}{3} ((4-8)^2 \cdot (11-8 \cdot 5) + (8-8)^2 \\ (4-8 \cdot 5)^2 + (18-8)^2 \cdot (5-8 \cdot 5)^2 + (7-8 \cdot 5)^2)$$

$$= \underline{-11},$$

$$\text{Cov}(x_2, x_1) = \text{Cov}(x_1, x_2) = \underline{-11}$$

$$\text{Cor}(x_2, x_2) = \frac{1}{3} ((11-8 \cdot 5)^2 + (4-8 \cdot 5)^2 + \\ (5-8 \cdot 5)^2 + (14-8 \cdot 5)^2) \\ = \underline{23}$$

$$S = \begin{bmatrix} 14 & -11 \\ -11 & 23 \end{bmatrix}$$

step 3: Eigenvalues of the covariance matrix.

$$S \rightarrow \text{covariance matrix } I \rightarrow \text{identity matrix}$$

$$0 = \det(S - \lambda I) \quad \lambda I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$S - \lambda I = \begin{vmatrix} 14-\lambda & -11-0 \\ 0-11 & 23-\lambda \end{vmatrix}$$

$$= \begin{vmatrix} 14-\lambda & -11 \\ -11 & 23-\lambda \end{vmatrix}$$

determinant

$$= (14-\lambda)(23-\lambda) - (-11 \times -11)$$

$$= 1^2 - 37\lambda + 201$$

Calculating roots

$$\lambda_1 = \underline{30.3849}$$

$$\lambda_2 = \underline{6.6151}$$

#### Step 4. Eigen vector calculation

$$U = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \quad \begin{bmatrix} 0 \\ 0 \end{bmatrix} = (S - \lambda I) U$$

$S \rightarrow$  covariance matrix  $I \rightarrow$  identity matrix

$$= \begin{bmatrix} 14-\lambda & -11 \\ -11 & 23-\lambda \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

$$= \begin{bmatrix} (14-\lambda)u_1 - 11u_2 \\ -11u_1 + (23-\lambda)u_2 \end{bmatrix}$$

Use any one equation to get value

Consider 1st eqn.

$$(14-\lambda)u_1 - 11u_2 = 0$$

$$(14-\lambda)u_1 = 11u_2$$

$$\frac{u_1}{11} = \frac{u_2}{14-\lambda} = t$$

$$\frac{u_1}{11} = \frac{u_2}{14-\lambda} = t$$

$$u_1 = 11t \quad u_2 = (14-\lambda)t$$

Assume  $t = 1$

$$u_1 = 11$$

$$u_2 = 14-\lambda$$

$$u = \begin{bmatrix} 11 \\ 14-\lambda \end{bmatrix}$$

$(\lambda_1, \lambda_2)$

Consider largest eigen value to calculate PCA.

here we want 1 principal component, we need one value  
here we use  $\lambda_1 = 130.3849$

$$u_1 = \begin{bmatrix} 11 \\ 14-\lambda_1 \end{bmatrix}$$

To find unit eigen vector, we compute length of  $U$ , which is given by

$$\begin{aligned}\|U\| &= \sqrt{\lambda^2 + (14-\lambda)^2} \\ &= \sqrt{\lambda^2 + (14 - 30.3849)^2} \\ &= \underline{19.7348}\end{aligned}$$

$$e_1 = \begin{bmatrix} \lambda / \|U\| \\ (14 - \lambda) / \|U\| \end{bmatrix}$$

$$= \begin{bmatrix} \lambda / 19.7348 \\ (14 - 30.3849) / 19.7348 \end{bmatrix}$$

$$= \begin{bmatrix} 0.5574 \\ -0.8303 \end{bmatrix}$$

Similarly by taking  $\lambda_2$  value as 6.6151

$$\text{We will get } e_2 = \begin{bmatrix} 0.8303 \\ 0.5574 \end{bmatrix}$$

Step 5: Computation of 1st principal component,  
for example 1.

$$e_1^T \times \begin{bmatrix} x_{1k} - \bar{x}_1 \\ x_{2k} - \bar{x}_2 \end{bmatrix}$$

$$e_1^T \begin{bmatrix} x_{1k} - \bar{x}_1 \\ x_{2k} - \bar{x}_2 \end{bmatrix} = [0.5574 - 0.8303] \times$$

$$\begin{bmatrix} x_{11} - \bar{x}_1 \\ x_{21} - \bar{x}_2 \end{bmatrix}$$

$$= 0.5574 (x_{11} - \bar{x}_1) - 0.8303 (x_{21} - \bar{x}_2)$$

$$= 0.5574 (4 - 8) - 0.8303 (11 - 8.5)$$

$$= -4.30535 //$$

for example 2.

$$e_1^T \begin{bmatrix} x_{12} - \bar{x}_1 \\ x_{22} - \bar{x}_2 \end{bmatrix}$$

$$= 3.7361$$

Calculate same for example 3 & 4

Feature	Eg1	Eg2	Eg3	Eg4
X1	4	8	13	7
X2	11	4	5	14
First principal component	-4.3052	3.7361	5.6928	-5.1238

we have 2 feature  $X_1, X_2$   
we reduced that to the first principal component.