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## # Numerical Topic : Dimensionality Reduction (PCA) → Principal Component Analysis

[1.] Problem = Given data =  $\{2, 3, 4, 5, 6, 7; 1, 5, 3, 6, 7, 8\}$   
Compute the principal component using  
PCA algorithm

(or)

Consider the two dimensional patterns  $(2, 1), (3, 5), (4, 3), (5, 6),$   
 $(6, 7), (7, 8)$ .

Compute the principal component using PCA algorithm.

Solution =

Given

class 1

$X = 2, 3, 4$

$Y = 1, 5, 3$

class 2

$X = 5, 6, 7$

$Y = 6, 7, 8$

⇒ Step = 1

The given features vectors are :-

→  $x_1 = (2, 1)$

→  $x_2 = (3, 5)$

→  $x_3 = (4, 3)$

→  $x_4 = (5, 6)$

→  $x_5 = (6, 7)$

→  $x_6 = (7, 8)$

(ie.)  $\begin{bmatrix} 2 \\ 1 \end{bmatrix} \begin{bmatrix} 3 \\ 5 \end{bmatrix} \begin{bmatrix} 4 \\ 3 \end{bmatrix} \begin{bmatrix} 5 \\ 6 \end{bmatrix} \begin{bmatrix} 6 \\ 7 \end{bmatrix} \begin{bmatrix} 7 \\ 8 \end{bmatrix}$

⇒ Step=2

calculate the mean vector ( $\mu$ )

$$\text{Mean vector } (\mu) = \left( \frac{(2+3+4+5+6+7)}{6}, \frac{(1+5+3+6+7+8)}{6} \right) \\ = (4.5, 5)$$

Thus,

$$\text{Mean vector } (\mu) = \begin{bmatrix} 4.5 \\ 5 \end{bmatrix}$$

⇒ Step=3

Subtract mean vector ( $\mu$ ) from the given feature vectors

$$\rightarrow x_1 - \mu = (2 - 4.5, 1 - 5) = (-2.5, -4)$$

$$\rightarrow x_2 - \mu = (3 - 4.5, 5 - 5) = (-1.5, 0)$$

$$\rightarrow x_3 - \mu = (4 - 4.5, 3 - 5) = (-0.5, -2)$$

$$\rightarrow x_4 - \mu = (5 - 4.5, 6 - 5) = (0.5, 1)$$

$$\rightarrow x_5 - \mu = (6 - 4.5, 7 - 5) = (1.5, 2)$$

$$\rightarrow x_6 - \mu = (7 - 4.5, 8 - 5) = (2.5, 3)$$

Feature vectors ( $x_i$ ) after subtracting mean vector ( $\mu$ ) are :-

$$\begin{bmatrix} -2.5 \\ -4 \end{bmatrix} \begin{bmatrix} -1.5 \\ 0 \end{bmatrix} \begin{bmatrix} -0.5 \\ -2 \end{bmatrix} \begin{bmatrix} 0.5 \\ 1 \end{bmatrix} \begin{bmatrix} 1.5 \\ 2 \end{bmatrix} \begin{bmatrix} 2.5 \\ 3 \end{bmatrix}$$

⇒ Step=4

→ calculate the covariance matrix

→ covariance matrix is given by,

$$\text{covariance matrix} = \frac{\sum (x_i - \mu)(x_i - \mu)^t}{n}$$

Now,

$$m_1 = (x_1 - \mu)(x_1 - \mu)^t = \begin{bmatrix} -2.5 \\ -4 \end{bmatrix} \begin{bmatrix} -2.5 & -4 \end{bmatrix} = \begin{bmatrix} 6.25 & 10 \\ 10 & 16 \end{bmatrix}$$

$$m_2 = (x_2 - \mu)(x_2 - \mu)^t = \begin{bmatrix} -1.5 \\ 0 \end{bmatrix} \begin{bmatrix} -1.5 & 0 \end{bmatrix} = \begin{bmatrix} 2.25 & 0 \\ 0 & 0 \end{bmatrix}$$

$$m_3 = (x_3 - \mu)(x_3 - \mu)^t = \begin{bmatrix} -0.5 \\ -2 \end{bmatrix} \begin{bmatrix} -0.5 & -2 \end{bmatrix} = \begin{bmatrix} 0.25 & 1 \\ 1 & 4 \end{bmatrix}$$

$$m_4 = (x_4 - \mu)(x_4 - \mu)^t = \begin{bmatrix} 0.5 \\ 1 \end{bmatrix} \begin{bmatrix} 0.5 & 1 \end{bmatrix} = \begin{bmatrix} 0.25 & 0.5 \\ 0.5 & 1 \end{bmatrix}$$

$$m_5 = (x_5 - \mu)(x_5 - \mu)^t = \begin{bmatrix} 1.5 \\ 2 \end{bmatrix} \begin{bmatrix} 1.5 & 2 \end{bmatrix} = \begin{bmatrix} 2.25 & 3 \\ 3 & 4 \end{bmatrix}$$

$$m_6 = (x_6 - \mu)(x_6 - \mu)^t = \begin{bmatrix} 2.5 \\ 3 \end{bmatrix} \begin{bmatrix} 2.5 & 3 \end{bmatrix} = \begin{bmatrix} 6.25 & 7.5 \\ 7.5 & 9 \end{bmatrix}$$

Now,

→ covariance matrix =  $\frac{(m_1 + m_2 + m_3 + m_4 + m_5 + m_6)}{6}$

→ On adding the above matrices and Dividing by 6, we get,

$$\begin{aligned} \text{covariance matrix} &= \frac{1}{6} \begin{bmatrix} 17.5 & 22 \\ 22 & 34 \end{bmatrix} \\ &= \begin{bmatrix} 2.92 & 3.67 \\ 3.67 & 5.67 \end{bmatrix} \end{aligned}$$

⇒

Step=5

→ Calculate the eigen values and eigen vectors of the covariance matrix

→  $\lambda$  is an eigen value for a matrix "M". If it is a solution of a characteristic equation  $|M - \lambda I| = 0$

So, we have

$$\begin{vmatrix} 2.92 & 3.67 \\ 3.67 & 5.67 \end{vmatrix} - \begin{vmatrix} \lambda & 0 \\ 0 & \lambda \end{vmatrix} = 0$$

$$\begin{vmatrix} 2.92 - \lambda & 3.67 \\ 3.67 & 5.67 - \lambda \end{vmatrix} = 0$$

$$(2.92 - \lambda)(5.67 - \lambda) - (3.67 \times 3.67) = 0$$

$$16.56 - 2.92\lambda - 5.67\lambda + \lambda^2 - (13.47) = 0$$

$$\lambda^2 - 8.59\lambda + 3.09 = 0$$

Solving we get,

$$\lambda = 8.22 > 0.38$$

$$\text{So, } \lambda_1 = 8.22 \text{ and } \lambda_2 = 0.38$$



then

→ we can use the following equation to find the eigen vector,

$$MX = \lambda X$$

where,

M = Covariance Matrix

X = Eigen vector

$\lambda$  = Eigen value

$$\begin{bmatrix} 2.92 & 3.67 \\ 3.67 & 5.67 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 8.22 \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

Solving this we get,

$$2.92 x_1 + 3.67 x_2 = 8.22 x_1$$

$$3.67 x_1 + 5.67 x_2 = 8.22 x_2$$

on simplification we get,  $5.3 x_1 = 3.67 x_2$  — (i)

$$3.67 x_1 = 2.55 x_2$$
 — (ii)

Solve (i) & (ii) we get,

$$\text{Eigen vector} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2.55 \\ 3.67 \end{bmatrix}$$

→ Thus principal component for the given set is,

$$\text{Principal Component} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2.55 \\ 3.67 \end{bmatrix}$$

→ Lastly we project the data points onto the new subspace as,

