

Question 1

We need to compute the eigen values and eigen vectors of the following matrix

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$$

The characteristic polynomial is.

$$\begin{vmatrix} (1-\lambda) & 2 & 3 \\ 4 & (5-\lambda) & 6 \\ 7 & 8 & (9-\lambda) \end{vmatrix} = (1-\lambda)[(5-\lambda)(9-\lambda) - 48] - 2[4(9-\lambda) - 42] + 3[32 - 7(5-\lambda)]$$

$$(1-\lambda)[45 + \lambda^2 - 14\lambda - 48] - 2(36 - 4\lambda - 42) + 3[32 - 35 + 7\lambda]$$

$$= (1-\lambda)[\lambda^2 - 14\lambda - 3] - 2(-4\lambda - 6) + 3(7\lambda - 3)$$

$$= \lambda^2 - 14\lambda - 3 + \lambda^3 + 14\lambda^2 + 3\lambda + 8\lambda + 12 + 21\lambda - 9$$

$$= -\lambda^3 + 15\lambda^2 + 18\lambda$$

$$= -\lambda(\lambda^2 - 15\lambda - 18)$$

$$\lambda^2 - 15\lambda - 18 = 0.$$

$$D = b^2 - 4ac = 225 + 12 = 237 = 3 \times 3 \times 33$$

$$\sqrt{D} = 3\sqrt{33}$$

$$\lambda = \frac{15 \pm 3\sqrt{33}}{2}$$

Hence, the three λ values are.

$$\lambda_1 = 0.$$

$$\lambda_2 = \frac{15 + 3\sqrt{33}}{2}$$

$$\lambda_3 = \frac{15 - 3\sqrt{33}}{2}$$

\Rightarrow for $\lambda = 0$.

$$A - \lambda I = 0.$$

$$\begin{bmatrix} 1 & 2 & 3 & | & 0 \\ 4 & 5 & 6 & | & 0 \\ 7 & 8 & 9 & | & 0 \end{bmatrix} \quad R_2 - 4R_1 = \begin{bmatrix} 1 & 2 & 3 & | & 0 \\ 0 & -3 & -6 & | & 0 \\ 7 & 8 & 9 & | & 0 \end{bmatrix}$$

$$R_3 \Rightarrow R_3 - 7R_1 \rightarrow \begin{bmatrix} 1 & 2 & 3 & | & 0 \\ 0 & -3 & -6 & | & 0 \\ 0 & -6 & -12 & | & 0 \end{bmatrix} \xrightarrow{R_2 = R_2 / (-3)} \begin{bmatrix} 1 & 2 & 3 & | & 0 \\ 0 & 1 & 2 & | & 0 \\ 0 & -6 & -12 & | & 0 \end{bmatrix}$$

$$R_3 \rightarrow R_3 + 6R_2 \quad \begin{bmatrix} 1 & 2 & 3 & | & 0 \\ 0 & 1 & 2 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{bmatrix} \xrightarrow{R_1 - 2R_2 \rightarrow R_1} \begin{bmatrix} 1 & 0 & -1 & | & 0 \\ 0 & 1 & 2 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{bmatrix}$$

$$x_1 - x_3 = 0 \Rightarrow x_1 = x_3.$$

$$x_2 + 2x_3 = 0 \Rightarrow x_2 = -2x_3$$

hence

$$X_2 = \text{span} \left(\begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix} \right)$$

$$\Rightarrow \text{for } \lambda = \frac{15 - 3\sqrt{33}}{2}$$

$$A - \lambda I = \begin{bmatrix} \left(\frac{3\sqrt{33} - 13}{2} \right) & 2 & 3 \\ 4 & \left(\frac{3\sqrt{33} - 5}{2} \right) & 6 \\ 7 & 8 & \left(\frac{3\sqrt{33} + 3}{2} \right) \end{bmatrix} \rightarrow R_1 \rightarrow \left(\frac{3\sqrt{33} + 13}{64} \right) R_1$$

$$\rightarrow \begin{bmatrix} 1 & \frac{3\sqrt{33} + 13}{32} & \frac{9\sqrt{33} + 39}{64} & | & 0 \\ 4 & (3\sqrt{33} - 5)/2 & 6 & | & 0 \\ 7 & 8 & (3\sqrt{33} + 3)/2 & | & 0 \end{bmatrix}$$

$$R_2 - 4R_1 \rightarrow \begin{bmatrix} 1 & \frac{(3\sqrt{33}+13)}{32} & \frac{(9\sqrt{33}+39)}{64} & 0 \\ 0 & \frac{(9\sqrt{33}-33)}{8} & \frac{(-9\sqrt{33}+51)}{16} & 0 \\ 0 & \frac{(3\sqrt{33}+3)}{2} & \frac{(33\sqrt{33}-177)}{64} & 6 \end{bmatrix}$$

$$R_3 - 7R_1 \rightarrow \begin{bmatrix} 1 & \frac{(3\sqrt{33}+13)}{32} & \frac{(9\sqrt{33}+39)}{64} & 0 \\ 0 & \frac{(9\sqrt{33}-33)}{8} & \frac{(-9\sqrt{33}+51)}{16} & 0 \\ 0 & \frac{(-21\sqrt{33}+165)}{2} & \frac{(33\sqrt{33}-177)}{64} & 6 \end{bmatrix}$$

$$R_2 \rightarrow R_2 \left(\frac{3\sqrt{33}+11}{64} \right) \rightarrow \begin{bmatrix} 1 & \frac{(3\sqrt{33}+13)}{32} & \frac{(9\sqrt{33}+39)}{64} & 0 \\ 0 & 1 & \frac{(3\sqrt{33}-11)}{44} & 0 \\ 0 & \frac{(-21\sqrt{33}+165)}{32} & \frac{(33\sqrt{33}-177)}{64} & 6 \end{bmatrix}$$

$$\left(\frac{21\sqrt{33}+165}{32} \right) R_3 + \left(\frac{21\sqrt{33}-165}{32} \right) R_2 \rightarrow \begin{bmatrix} 1 & \frac{(3\sqrt{33}+13)}{32} & \frac{(9\sqrt{33}+39)}{64} & 0 \\ 0 & 1 & \frac{(3\sqrt{33}-11)}{44} & 0 \\ 0 & 0 & 0 & 6 \end{bmatrix}$$

$$R_1 \rightarrow R_1 - \frac{(3\sqrt{33}+13)}{32} R_2 \rightarrow \begin{bmatrix} 1 & 0 & \frac{(3\sqrt{33}+11)}{22} & 0 \\ 0 & 1 & \frac{(3\sqrt{33}-11)}{44} & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$n_1 + \frac{3\sqrt{33}+11}{22} n_3 = 0.$$

$$n_2 + \frac{3\sqrt{33}-11}{44} n_3 = 0.$$

$$\Rightarrow X \in \text{span} \left(\begin{pmatrix} \frac{-3\sqrt{33}-11}{22} \\ \frac{-3\sqrt{33}+11}{44} \\ 1 \end{pmatrix} \right)$$

$$\Rightarrow \text{for } \lambda = \frac{(3\sqrt{33}+15)}{2}$$

$$A - \lambda I = 0.$$

$$\begin{bmatrix} \frac{(-3\sqrt{33}-13)}{2} & 2 & 3 & 0 \\ 1 & \frac{(-3\sqrt{33}-5)}{2} & 6 & 0 \\ 1 & \frac{(-3\sqrt{33}+3)}{2} & 0 & 0 \end{bmatrix} \rightarrow R_1 \rightarrow R_1 \times \left(\frac{-3\sqrt{33}+13}{64} \right)$$

$$\left[\begin{array}{cc|c} 1 & \frac{(-3\sqrt{33}+13)}{32} & \frac{-9\sqrt{33}+39}{64} & 0 \\ 4 & \frac{(-3\sqrt{33}-5)}{2} & 6 & 0 \\ 7 & 8 & \frac{(-3\sqrt{33}+3)}{2} & 0 \end{array} \right] \longrightarrow R_2 \rightarrow R_2 - 4R_1$$

$$\left[\begin{array}{cc|c} 1 & \frac{(-3\sqrt{33}+13)}{32} & \frac{(-9\sqrt{33}+39)}{64} & 0 \\ 0 & \frac{(-9\sqrt{33}-33)}{8} & \frac{(9\sqrt{33}+51)}{16} & 0 \\ 7 & 8 & \frac{(-3\sqrt{33}+3)}{2} & 0 \end{array} \right] \longrightarrow R_3 - 7R_1 \rightarrow R_3$$

$$\left[\begin{array}{cc|c} 1 & \frac{(-3\sqrt{33}+13)}{32} & \frac{(-9\sqrt{33}+39)}{64} & 0 \\ 0 & \frac{(-9\sqrt{33}-33)}{8} & \frac{(9\sqrt{33}+51)}{16} & 0 \\ 0 & \frac{(21\sqrt{33}+165)}{32} & \frac{-33\sqrt{33}-177}{64} & 0 \end{array} \right] \longrightarrow R_2 \rightarrow R_2 \times \left(\frac{-3\sqrt{33}+11}{66} \right)$$

$$\left[\begin{array}{cc|c} 1 & \frac{(-3\sqrt{33}+13)}{32} & \frac{(-9\sqrt{33}+39)}{64} & 0 \\ 0 & 1 & \frac{(-3\sqrt{33}-11)}{44} & 0 \\ 0 & \frac{(21\sqrt{33}+165)}{32} & \frac{(-33\sqrt{33}-177)}{64} & 0 \end{array} \right] \longrightarrow R_3 - \frac{21\sqrt{33}+165}{32} R_2 \rightarrow R_3$$

$$\left[\begin{array}{cc|c} 1 & \frac{(-3\sqrt{33}+13)}{32} & \frac{(-9\sqrt{33}+39)}{64} & 0 \\ 0 & 1 & \frac{(-3\sqrt{33}-11)}{44} & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \longrightarrow R_1 + \frac{(-3\sqrt{33}-13)}{32} R_2$$

$$\left[\begin{array}{cc|c} 1 & 0 & \frac{(-3\sqrt{33}+11)}{22} & 0 \\ 0 & 1 & \frac{(-3\sqrt{33}-11)}{44} & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$$x_1 - \frac{(-3\sqrt{33}+11)}{22} x_3 = 0$$

$$x_2 - \frac{(-3\sqrt{33}-11)}{44} x_3 = 0$$

$$x_3 \text{ spans } \begin{pmatrix} \frac{(-3\sqrt{33}+11)}{22} \\ \frac{(-3\sqrt{33}-11)}{44} \\ 1 \end{pmatrix}$$

⇒ Determinant of A.

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} = 1(45 - 48) - 2(36 - 42) + 3(32 - 35).$$

Hence determinant = 0.

$$-3 + 12 - 9 = \underline{\underline{0}}$$

⇒ TRACE OF MATRIX

Sum of diagonals = sum of eigen values

$$15 = 0 - \frac{3\sqrt{33} + 15}{2} + \frac{3\sqrt{33} + 15}{2} = \underline{\underline{15}}$$

⇒ RANK OF MATRIX.

Rank is the number of non-zero eigen values we get. Hence the rank here is 2.

→ The determinant is the product of the eigen values, and since one is 0, the determinant here is 0.

→ The trace of the matrix is the sum of eigen values.

→ The rank of the matrix is the number of non-zero eigen values.



Question 2

i) Dimensions of matrix A.

$$y_i = Ax_i$$

$$y_i = p \times 1 \quad A = d \times b \quad x_i = q \times 1$$

$$d \times 1 = p \times 1$$

$$\Rightarrow d = p$$

For Ax_i to be possible.

$$b = q$$

Hence A is a $p \times q$ matrix.

ii) Can there exist a matrix A such that the Euclidean distance between y_1 & y_2 is same as x_1 & x_2

Yes, the Euclidean distance between x_1 & x_2 can be preserved even after linear transformation. For this, the matrix A should be an orthogonal matrix, i.e. $AA^T = I$

$$y_1 = Ax_1$$

$$y_2 = Ax_2$$

$$\|x_1\|^2 = x_1 \cdot x_1 = x_1^T x_1$$

$$\|Ax_1\|^2 = Ax_1 \cdot Ax_1 = (Ax_1)^T Ax_1$$

$$= x_1^T A^T A x_1$$

$$= x_1^T I x_1 = x_1^T x_1$$

$$\therefore \|Ax_1\|^2 = \|x_1\|^2 \quad \text{--- (1)}$$

$$\text{Similarly } \|Ax_2\|^2 = \|x_2\|^2 \quad \text{--- (2)}$$

From (1) & (2)

we can say that the lengths are preserved.



$$\cos \theta = \frac{x_1 \cdot x_2}{\|x_1\| \|x_2\|}$$

$$\cos \alpha = \frac{y_1 \cdot y_2}{\|y_1\| \|y_2\|}$$

$$= \frac{Ax_1 \cdot Ax_2}{\|Ax_1\| \|Ax_2\|} = \frac{(Ax_1)^T Ax_2}{\|x_1\| \|x_2\|}$$

$$\frac{n_1^T A^T A n_2}{\|n_1\| \|n_2\|}$$

$$\frac{n_1^T n_2}{\|n_1\| \|n_2\|} = \frac{n_1 \cdot n_2}{\|n_1\| \|n_2\|} = \cos \theta$$

Angle is preserved.

As angle between the vectors and lengths of vectors are preserved.
The distances between n_1 & n_2 will also be preserved.

In case A is not a matrix with $n \times n$ dimension, and with the dimension $m \times n$ dimensions.

We need $A^T A = I$ for the distance between n_1 & n_2 to be preserved.

i.e. A should be semi-orthogonal (left-invertible)
i.e. $m > n$ for $(A_{m \times n})$

iii) 2) $A_{p \times q}$, $p=2$, $q=2$

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$A^T A = I$$

$$\begin{bmatrix} a & c \\ b & d \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a^2+c^2 & ab+cd \\ ab+cd & b^2+d^2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$a^2+c^2=1, \quad b^2+d^2=1$$

$$ab+cd=0 \quad ab=-cd$$

$$A = \begin{bmatrix} 3/5 & 4/5 \\ 4/5 & -3/5 \end{bmatrix}$$

b) $A_{p \times q}$, $p=1$, $q=2$.

Here A is not a square matrix.

For the euclidean distance to be preserved, A should be a left-invertible matrix.

$$\text{ie } A^T A = I$$

for that $p > q$ is required.

but here $p < q$

Hence there are no examples. feasible $= A_{1 \times 2}$

a $p=2, q=4$.

same as in the case of (b) - Here $p < q$.

Hence there are no example feasible for $A_{2 \times 4}$.



Question 3

line: $w_1 x_1 + w_2 x_2 + w_3 = 0$.

a. Given N points on the line, $D = \{x_i^0 [x_1^0, x_2^0]^T\}$.

$$w_1 x_1^0 + w_2 x_2^0 + w_3 = 0.$$

$$\frac{1}{N} \sum_i (w_1 x_1^i + w_2 x_2^i + w_3) = 0.$$

$$\cancel{\frac{1}{N}} w_1 \mu_1 + \cancel{\frac{1}{N}} w_2 \mu_2 + w_3 = 0$$

(Hence, $[\mu_1 - \mu_2]^T$ lies on the line)

$$A' = \begin{bmatrix} x_1' - \mu_1 \\ x_2' - \mu_2 \end{bmatrix} \begin{bmatrix} x_1' - \mu_1 & x_2' - \mu_2 \end{bmatrix} = \begin{bmatrix} (x_1' - \mu_1)^2 & (x_1' - \mu_1)(x_2' - \mu_2) \\ (x_1' - \mu_1)(x_2' - \mu_2) & (x_2' - \mu_2)^2 \end{bmatrix}$$

$$R_1 \rightarrow R_1 / (x_1' - \mu_1)^2 = \begin{bmatrix} 1 & (x_2' - \mu_2) / (x_1' - \mu_1) \\ (x_1' - \mu_1) / (x_2' - \mu_2) & (x_2' - \mu_2)^2 \end{bmatrix}$$

$$R_2 \rightarrow R_2 / (x_1' - \mu_1) - (x_2' - \mu_2) R_1 = \begin{bmatrix} 1 & (x_2' - \mu_2) / (x_1' - \mu_1) \\ 0 & 0 \end{bmatrix}$$

The number of eigen values (non-zero) gives the rank, i.e. $\text{rank}(A') = 1$.

$$A = \frac{1}{N} \sum_{i=1}^N \begin{bmatrix} x_1^i - \mu_1 \\ x_2^i - \mu_2 \end{bmatrix} \begin{bmatrix} x_1^i - \mu_1 & x_2^i - \mu_2 \end{bmatrix}.$$

$$= \frac{1}{N} \sum_{i=1}^N \begin{bmatrix} (x_1^i - \mu_1)^2 & (x_1^i - \mu_1)(x_2^i - \mu_2) \\ (x_1^i - \mu_1)(x_2^i - \mu_2) & (x_2^i - \mu_2)^2 \end{bmatrix}$$

$$R_1 \rightarrow R_1 / (x_1^i - \mu_1)^2 = \frac{1}{N} \sum_{i=1}^N \begin{bmatrix} 1 & (x_2^i - \mu_2) / (x_1^i - \mu_1) \\ (x_1^i - \mu_1) / (x_2^i - \mu_2) & (x_2^i - \mu_2)^2 \end{bmatrix}$$

$$R_2 \rightarrow R_2 / (x_1^i - \mu_1) - (x_2^i - \mu_2) R_1 \rightarrow \frac{1}{N} \sum_{i=1}^N \begin{bmatrix} 1 & (x_2^i - \mu_2) / (x_1^i - \mu_1) \\ 0 & 0 \end{bmatrix}$$

The eigen vectors of covariance matrix sorted in descending order of their corresponding eigen values represent the direction/component along which the data is in the order of principal components. Therefore variance is maximized along these vectors. Here, since all data is on the same line, $\text{rank} = 1$, and the max variance is along the line.

b). $B' = \begin{bmatrix} x_1' - \mu_1 \\ x_2' - \mu_2 \end{bmatrix} \begin{bmatrix} x_1' - \mu_1 & x_2' - \mu_2 \end{bmatrix}$

Follow the same steps as you've done for A' . we get B' as.

$$\begin{bmatrix} 1 & (x_2' - \mu_2)/(x_1' - \mu_1) \\ 0 & 0 \end{bmatrix} \Rightarrow \text{rank}(B') = 1 \text{ (no. of non-zero eigen values)}$$

$$B = \frac{1}{N} \sum_{i=1}^N \begin{bmatrix} x_1^i - \mu_1 \\ x_2^i - \mu_2 \end{bmatrix} \begin{bmatrix} x_1^i - \mu_1 & x_2^i - \mu_2 \end{bmatrix}$$

Follow the same steps you've followed for A . we get B as.

$$B = \frac{1}{N} \sum_{i=1}^N \begin{bmatrix} 1 & (x_2^i - \mu_2)/(x_1^i - \mu_1) \\ 0 & 0 \end{bmatrix} \Rightarrow \text{rank}(B) = 1 \text{ (no. of non-zero eigen values)}$$

c) Consider a vector U to be principal component and D be the data points. and μ is the mean of the points. To determine the component along which maximum information from the data is preserved.
 $\text{max}(\text{variance}(U^T D))$ $U^T U = 1$

$$E[(U^T D - U^T \mu)(U^T D - U^T \mu)^T] = E[U^T (D - \mu)(D - \mu)^T U]$$

$$= E[U^T S U]$$

$\text{max}(U^T S U)$, where $U^T U = 1$ } putting both condition together, we get the following.

$$\text{max}(U^T S U - \lambda(U^T U - 1))$$

Differentiating w.r.t U

$$\Rightarrow S U = \lambda U$$

where

λ is eigen value. U is eigen vector.

Eigen vector corresponding to larger eigen value captures more information than one with smaller eigen value.

Given that we have points around a line with mean

$\mu \Rightarrow$ we have 2 eigen vectors, perpendicular to each other.

Hence, $\Sigma \Rightarrow$ covariance matrix will have rank = 2 as there are 2 components and hence 2 non-zero eigen values.



