

Q1.

$\rightarrow \text{E.V.}$ $\rightarrow \text{trace}$ $\rightarrow \text{determinant}$
 $\rightarrow \text{E.V}$ $\rightarrow \text{Rank}$

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

$$|A - \lambda I| = 0 = \begin{vmatrix} 1-\lambda & 2 & 3 \\ 4 & 5-\lambda & 6 \\ 7 & 8 & 9-\lambda \end{vmatrix}$$

$$0 = (1-\lambda)[(5-\lambda)(9-\lambda) - 48] - 2[4(9-\lambda) - 42] + 3[32 - 7(5-\lambda)]$$

$$(1-\lambda) \left[\underset{\substack{\downarrow \\ \lambda^2 - 14\lambda - 3}}{45 - 14\lambda + \lambda^2 - 48} \right] - 2 \left[\underset{\substack{\downarrow \\ -4\lambda - 6}}{36 - 4\lambda - 42} \right] + 3 \left[\underset{\substack{\downarrow \\ 7\lambda - 3}}{32 - 35 + 7\lambda} \right] = 0$$

$$\lambda^2 - 14\lambda - 3 - \lambda^3 + 14\lambda^2 + 3\lambda + 8\lambda + 12 + 21\lambda - 9 = 0$$

$$-\lambda^3 + 15\lambda^2 + 18\lambda = 0$$

$$-\lambda (\lambda^2 - 15\lambda - 18) = 0$$

$$\begin{matrix} \parallel \\ 0 \end{matrix}$$

$$\begin{matrix} \parallel \\ 0 \end{matrix}$$

$$\lambda = \frac{15 \pm \sqrt{225 + 72}}{2}$$

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

$$\lambda_1 = 0$$

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

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

$$\Rightarrow \lambda = 0$$

$$Ax = 0$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$$\text{Solving } \left[\begin{array}{ccc|c} 1 & 2 & 3 & 0 \\ 4 & 5 & 6 & 0 \\ 7 & 8 & 9 & 0 \end{array} \right]$$

$$\left[\begin{array}{ccc|c} 1 & 2 & 3 & 0 \\ 0 & 3 & 6 & 0 \\ 0 & 6 & 12 & 0 \end{array} \right] \Rightarrow \left[\begin{array}{ccc|c} 1 & 2 & 3 & 0 \\ 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$x_3 = \text{free variable}$

$$x = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} x_3$$

$$x_2 = -2x_3$$

$$x_1 + 2x_2 + 3x_3 = 0$$

$$x_1 - 4x_3 + 3x_3 = 0$$

$$x_1 = x_3$$

$$V_1 = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$$

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

$$(A - \lambda I)x = 0$$

Using Cramer's rule

$$\left[\begin{array}{ccc|c} -3\sqrt{33} - 13 & 4 & 6 & 0 \\ 8 & -3\sqrt{33} - 5 & 6 & 12 \\ 14 & 16 & -3\sqrt{33} + 3 & 0 \end{array} \right]$$

$$\Rightarrow V_2 = \begin{bmatrix} \frac{3\sqrt{33} - 11}{22} \\ \frac{3\sqrt{33} + 11}{44} \\ 1 \end{bmatrix}$$

$$\left[\begin{array}{ccc|c} -3\sqrt{33} - 13 & 4 & 6 & 0 \\ 0 & -3\sqrt{33} - 5 & 6 & 12 \\ 0 & 48\sqrt{33} + 264 & -174 & 0 \end{array} \right]$$

$$\left[\begin{array}{ccc|c} -3\sqrt{33} - 13 & 4 & 6 & 0 \\ 0 & -3\sqrt{33} - 5 & 6 & 12 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

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

$$\Rightarrow V_3 = \begin{bmatrix} \frac{-3\sqrt{33} - 11}{22} \\ \frac{-3\sqrt{33} - 11}{44} \\ 1 \end{bmatrix}$$

Q.2.

1) x is $q \times 1$

y is $p \times 1$

$$y_{p \times 1} = A_{p \times q} x_{q \times 1}$$

dimension: $p \times q$

2)

$$y_1 = Ax_1$$

$$y_2 = Ax_2$$

distance of y_1 & y_2

= distance of x_1 & x_2

$$\Rightarrow (y_1 - y_2)^T (y_1 - y_2) = (x_1 - x_2)^T (x_1 - x_2)$$

$$\Rightarrow (Ax_1 - Ax_2)^T (Ax_1 - Ax_2)$$

$$(A(x_1 - x_2))^T \cdot A \cdot (x_1 - x_2)$$

$$(x_1 - x_2)^T A^T A (x_1 - x_2)$$

i) $A^T A = I$ then distance will be same

3) a) $p=2$ $q=2$

Any Orthogonal matrix will preserve distance

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

b) $q=2$

$p=1$

$A_{1 \times 2}$

~~$$\begin{bmatrix} a & b \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = 1$$~~

~~$$a^2 + b^2 = 1$$~~

~~$$\begin{bmatrix} a \\ b \end{bmatrix} \begin{bmatrix} a & b \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$~~

$$\begin{bmatrix} a \\ b \end{bmatrix}_{2 \times 1} \begin{bmatrix} a & b \end{bmatrix}_{1 \times 2} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

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

$$a^2 = 1$$

$$ab = 0$$

$$b^2 = 1$$

No such A is possible

c) $q=4$ $p=2$

$$A_{2 \times 4} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \end{bmatrix}$$

$$A^T A = \begin{bmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \\ a_{13} & a_{23} \\ a_{14} & a_{24} \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} a_{11}^2 + a_{21}^2 & a_{11}a_{12} + a_{21}a_{22} & a_{11}a_{13} + a_{21}a_{23} & a_{11}a_{14} + a_{21}a_{24} \\ a_{11}a_{12} + a_{21}a_{22} & a_{12}^2 + a_{22}^2 & a_{12}a_{13} + a_{22}a_{23} & a_{12}a_{14} + a_{22}a_{24} \\ a_{11}a_{13} + a_{21}a_{23} & a_{12}a_{13} + a_{22}a_{23} & a_{13}^2 + a_{23}^2 & a_{13}a_{14} + a_{23}a_{24} \\ a_{11}a_{14} + a_{21}a_{24} & a_{12}a_{14} + a_{22}a_{24} & a_{13}a_{14} + a_{23}a_{24} & a_{14}^2 + a_{24}^2 \end{bmatrix}$$

$$\begin{aligned} & a_{11}^2 + a_{21}^2 & a_{11}a_{12} + a_{21}a_{22} & a_{11}a_{13} + a_{21}a_{23} \\ & a_{12}^2 + a_{22}^2 & a_{12}a_{13} + a_{22}a_{23} & a_{12}a_{14} + a_{22}a_{24} \\ & a_{13}^2 + a_{23}^2 & a_{13}a_{14} + a_{23}a_{24} & a_{14}^2 + a_{24}^2 \end{aligned}$$

No such matrix exist
that will keep distance
same without putting any
condition on x .

Q-3.

1.

$$l: w_1 x_1 + w_2 x_2 + w_3 = 0$$

N points

$$D_i = \{x_i = [x_1^i, x_2^i]^T\}$$

$$\mu = [\mu_1, \mu_2]^T$$

$$A' = \begin{bmatrix} x_1^i - \mu_1 \\ x_2^i - \mu_2 \end{bmatrix} [x_1^i - \mu_1 \quad x_2^i - \mu_2]$$

= (column \times row)always rank 1. A' is rank 1.

Hence 1 non zero eigen values

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

 \downarrow A is sum of rank 1 matrix

$$\text{Rank of } A \leq \underbrace{1 + 1 + \dots + 1}_N$$

$$\leq N$$

 $N=2$

$$\begin{bmatrix} x_1^1 - \mu_1 \\ x_2^1 - \mu_2 \end{bmatrix} [x_1^1 - \mu_1 \quad x_2^1 - \mu_2]$$

$$+ \begin{bmatrix} x_1^2 - \mu_1 \\ x_2^2 - \mu_2 \end{bmatrix} [x_1^2 - \mu_1 \quad x_2^2 - \mu_2]$$

$$\mu_1 = \frac{x_1' + x_1^2}{2}$$

$$\mu_2 = \frac{x_2' + x_2^2}{2}$$

$$x_1' - \mu_1 = \frac{x_1' - x_1^2}{2}$$

$$\Rightarrow \frac{1}{2} \left(\begin{bmatrix} x_1' - x_1^2 \\ x_2' - x_2^2 \end{bmatrix} \begin{bmatrix} x_1' - x_1^2 & x_2' - x_2^2 \end{bmatrix} + \begin{bmatrix} x_1^2 - x_1' \\ x_2^2 - x_2' \end{bmatrix} \begin{bmatrix} x_1^2 - x_1' & x_2^2 - x_2' \end{bmatrix} \right)$$

$$\Rightarrow \frac{1}{2} \left(\begin{bmatrix} a \\ b \end{bmatrix} \begin{bmatrix} a & b \end{bmatrix} + \begin{bmatrix} -a \\ -b \end{bmatrix} \begin{bmatrix} -a & -b \end{bmatrix} \right)$$

$$= \frac{1}{2} \left(2 \begin{bmatrix} a \\ b \end{bmatrix} \begin{bmatrix} a & b \end{bmatrix} \right) = \frac{1}{4} \left(\begin{bmatrix} a \\ b \end{bmatrix} \begin{bmatrix} a & b \end{bmatrix} \right)$$

column row

rank 1

for N=2, one non zero eigenvalue

$$\begin{aligned} N=3 \Rightarrow & \begin{bmatrix} x_1' - \mu_1 \\ x_2' - \mu_2 \end{bmatrix} \begin{bmatrix} x_1' - \mu_1 & x_2' - \mu_2 \end{bmatrix} + \begin{bmatrix} x_1^2 - \mu_1 \\ x_2^2 - \mu_2 \end{bmatrix} \begin{bmatrix} x_1^2 - \mu_1 & x_2^2 - \mu_2 \end{bmatrix} \\ & + \begin{bmatrix} x_1^3 - \mu_1 \\ x_2^3 - \mu_2 \end{bmatrix} \begin{bmatrix} x_1^3 - \mu_1 & x_2^3 - \mu_2 \end{bmatrix} \end{aligned}$$

$$x_1' - \mu_1 = \frac{3x_1' - x_1' - x_1^2 - x_1^3}{3} = \frac{2x_1' - x_1^2 - x_1^3}{3}$$

$$x_1^2 - \mu_1 = \frac{2x_1^2 - x_1^2 - x_1^3}{3}$$

here all terms are different
Substituting...

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

$$\begin{bmatrix} aa & ba \\ ab & bb \end{bmatrix} + \begin{bmatrix} cc & dc \\ cd & dd \end{bmatrix} + \begin{bmatrix} ee & fe \\ ef & ff \end{bmatrix}$$

$$\begin{bmatrix} aa+cc+ee & ba+dc+fe \\ ab+cd+ef & bb+dd+ff \end{bmatrix}$$



independent columns.

Hence rank = 2.

$$N = 1, 2 \Rightarrow \text{rank} = 1$$

1 non zero eigen vector

$$N > 2 \Rightarrow \text{rank} = 2$$

2 non zero eigen vector

2).

line perpendicular to $l \Rightarrow l^\perp$ with m_2 slope

$$l: w_1 x_1 + w_2 x_2 + w_3 = 0$$

$$x_2 = \frac{-w_1}{w_2} x_1 - \frac{w_3}{w_2}$$

$$m_1 = \frac{-w_1}{w_2}$$

$$m_1 m_2 = -1 \text{ for } l^\perp$$

$$m_2 = \frac{w_2}{w_1}$$

~~and intercept can be anything until a point is given.~~



$$x_2 = mx_1 + c$$

$$x_2 = \frac{w_2 x_1}{w_1} + c$$

Can be any thing.

$$w_1 x_2 - w_2 x_1 = w_1 c$$

$$\mu = \begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}$$

$$w_1 \mu_2 - w_2 \mu_1 = w_1 c$$

$$c = \frac{w_1 \mu_2 - w_2 \mu_1}{w_1}$$

$$\text{line } L^H \Rightarrow \boxed{-w_2 x_1 + w_1 x_2 + (w_2 \mu_1 - w_1 \mu_2) = 0}$$

And for $B^T \Rightarrow \text{column} \times \text{row}$

Rank 1 \Rightarrow 1 non zero e.v.

$B \Rightarrow$ same explain as in previous.

3) The eigen values of the covariance matrix is the variability of data in orthogonal basis.

By finding e.v. and e.v. of covariance matrix, we find that the eigenvectors with the largest eigenvalues correspond to the dimensions that have the strongest correlation in dataset.

Since a covariance matrix is symmetric and positive definite, eigen vectors are orthogonal.

eigen value = Variance of data
of covariance
matrix

euler1.py

3.py

C:\Users\Shaily Mishra\Desktop\MS\Sem1\SMAT\HWs\3\3.py

```
1 import numpy
2 import matplotlib.pyplot as plt
3 import random
4 from numpy import linalg as LA
5 m, b = 5, 3
6 lower, upper = -50, 50
7 xstart = 0
8 xstop = 1000
9 x = numpy.linspace(-10,xstop,10)
10
11 samplepoints = 1000
12 x1 = [numpy.random.randint(xstart, xstop) for i in range(samplepoints)]
13 y1 = [numpy.random.randint(m*x+b+lower, m*x+b+upper) for x in x1]
14 meanx = numpy.mean(x1)
15 meany = numpy.mean(y1)
16 print('mean:', [meanx, meany])
17 X = numpy.stack((x1, y1), axis=0)
18 Covar = numpy.cov(X)
19 print('Covariance Matrix',Covar)
20 eigenvalue, eigenvector = LA.eig(Covar)
21 print('Eigen Values ', eigenvalue)
22 print('Eigen Vectors ', eigenvector)
23 plt.plot(x,m*x+b, label= "Line y =mx+c",linestyle='solid',color = "k")
24 plt.scatter(x1, y1, label="random points near line" , c='c')
25 plt.quiver(meanx,meany,eigenvector[0,0],eigenvector[0,1], color=['r'],scale=5, label="eigen vector[0]")
26 plt.quiver(meanx,meany,eigenvector[1,0],eigenvector[1,1], color=['g'],scale=5, label="eigen vector[1]")
27 plt.legend()
28 plt.show()
29
```

euler1.py3.py

c:\Users\Shaily Mishra\Desktop\MS\Sem1\SMAT\HWs\3> 3.py ...
4 from numpy import linalg as LA

PROBLEMSOUTPUTDEBUG CONSOLETERMINAL

PS C:\Users\Shaily Mishra\Desktop\MS\Sem1\SMAT\HWs> cd 3
PS C:\Users\Shaily Mishra\Desktop\MS\Sem1\SMAT\HWs\3> python 3.py
mean: [511.006, 2556.9]
Covariance Matrix [[80453.56753153 402146.48508509]
[402146.48508509 2010945.16116116]]
Eigen Values [3.15185488e+01 2.09136721e+06]
Eigen Vectors [[-0.98058407 -0.19609914]
[0.19609914 -0.98058407]]

Figure 1

