

Q1.

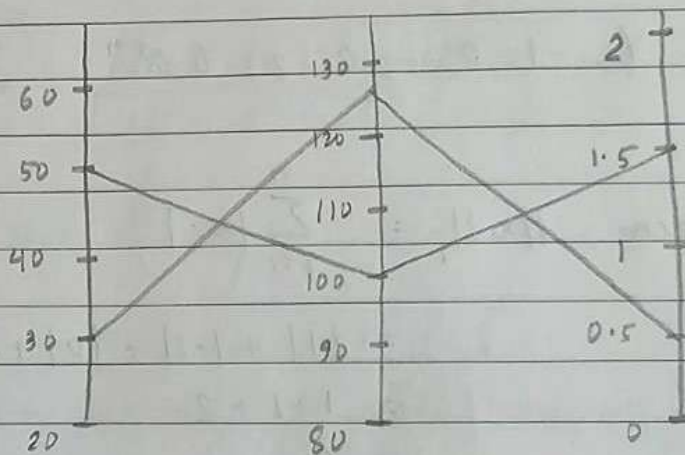
Parallel co-ordinate method is a method to visualize a set of points in  $n$ -dimension. In order to do so, equally spaced vertical lines are drawn in which points are represented as broken lines with vertices on the parallel axes.

For example:

if  $x = (30, 115, 0.5)$ ,

$y = (50, 100, 1.5)$  be two points  $x, y \in \mathbb{R}^3$ .

Then, using parallel co-ordinate axis, we plot them as follows:



Usage in data science

- Used for multivariate numerical data
- Ideal for comparing many variables together and see relationship between them.

Q2.

$L_1$ -norm on  $\mathbb{R}^n$ : If  $\|\cdot\|$  represents a mapping from  $\mathbb{R}^n \rightarrow \mathbb{R}^n$ , then  $L_1$ -norm, represented by  $\|x\|_1$ , is defined as

$$\|x\|_1 = \sum_{i=1}^n |x_i|$$

where,  $x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ .

$\ell_2$ -norm on  $\mathbb{R}^n$ : The  $\ell_2$ -norm, represented by  $\|x\|_2$ , is defined as

$$\|x\|_2 = \left( \sum_{i=1}^n x_i^2 \right)^{1/2}$$

where,  $x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ .

$\ell_\infty$ -norm on  $\mathbb{R}^n$ : The  $\ell_\infty$ -norm, represented by  $\|x\|_\infty$ , is defined as

$$\|x\|_\infty = \max_{i \in \{1, 2, \dots, n\}} |x_i|$$

where,  $x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ .

Given,  $x = (1, -1, 0, \dots, 0, 2) \in \mathbb{R}^n$ .

Then,

$$\ell_1\text{-norm, } \|x\|_1 = \sum_{i=1}^n |x_i|$$

$$\begin{aligned} &= |1| + |-1| + |0| + \dots + |0| + |2| \\ &= 1 + 1 + 2 \\ &= 4 \end{aligned}$$

$$\ell_2\text{-norm, } \|x\|_2 = \left( \sum_{i=1}^n x_i^2 \right)^{1/2}$$

$$\begin{aligned} &= \left( (1)^2 + (-1)^2 + (0)^2 + \dots + (0)^2 + (2)^2 \right)^{1/2} \\ &= (6)^{1/2} \\ &= 2.45 \end{aligned}$$

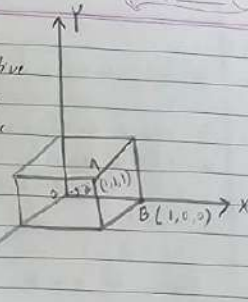
$$\ell_\infty\text{-norm, } \|x\|_\infty = \max_{i \in \{1, 2, \dots, n\}} |x_i|$$

$$\begin{aligned} &= \max \{ |1|, |-1|, |0|, \dots, |0|, |2| \} \\ &= 2 \end{aligned}$$

Q.3

Consider a unit cube in positive orthant as shown in diagram alongside. Let OA  $(1, 1, 1)$  be the diagonal of the cube and OB  $(1, 0, 0)$  be the standard basis vector along x-axis i.e.  $e_1$ .

Angle between OA & OB = ?



Solution

$$\|OA\|_2 = \sqrt{(1)^2 + (1)^2 + (1)^2} = \sqrt{3}$$

$$\|OB\|_2 = \sqrt{(1)^2 + 0^2 + 0^2} = 1$$

$$OA \cdot OB = \sum_{i=1}^3 x_i y_i$$

$$= 1 \times 1 + 1 \times 0 + 1 \times 0 = 1$$

From the property of dot product,

$$\cos \theta = \frac{OA \cdot OB}{\|OA\|_2 \|OB\|_2}$$

$$= \frac{1}{1 \times \sqrt{3}}$$

$$= \frac{1}{\sqrt{3}}$$

$$= \frac{1}{\sqrt{3}}$$

$$= \frac{1}{\sqrt{3}}$$

$$= \frac{1}{\sqrt{3}}$$

$$\Rightarrow \theta = \cos^{-1} \left( \frac{1}{\sqrt{3}} \right)$$

$$\therefore \theta = \cos^{-1} \left( \frac{1}{\sqrt{3}} \right)$$

$\therefore$  The angle between diagonal and  $e_1$  is  $\cos^{-1} \left( \frac{1}{\sqrt{3}} \right)$ .



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Given,  $u = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$ ,  $v = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$

To prove orthogonality, let us find the dot product.

$$u \cdot v = u_1 v_1 + u_2 v_2$$

$$= -1 \times -1 + 1 \times 1$$

$$= 1 + 1$$

$$= 2$$

Since,  $u \cdot v \neq 0$ ,  $u$  and  $v$  are orthogonal to each other.

To find orthonormal basis, we need to divide each vector by its corresponding  $L_2$ -norm.

Then,

$$\hat{u} = \frac{1}{\|u\|} u = \frac{1}{((1)^2 + (1)^2)^{1/2}} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$= \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$= \begin{pmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix}$$

$$\hat{v} = \frac{1}{\|v\|} v = \frac{1}{\sqrt{(-1)^2 + (-1)^2}} \begin{pmatrix} -1 \\ -1 \end{pmatrix}$$

$$= \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} -1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix}$$

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Given eigenvectors  $v_1$  and  $v_2$  correspond to distinct eigenvalues  $\lambda_1$  and  $\lambda_2$  of matrix  $A$ .

Let us assume,

$$c_1 v_1 + c_2 v_2 = 0 \quad \dots (1)$$

Multiplying eq<sup>n</sup> (1) by  $A$ , we get:

$$c_1 A v_1 + c_2 A v_2 = 0$$

$$\Rightarrow c_1 \lambda_1 v_1 + c_2 \lambda_2 v_2 = 0 \quad \dots (2) \quad \left[ \begin{array}{l} A v_1 = \lambda_1 v_1 \\ A v_2 = \lambda_2 v_2 \end{array} \right]$$

Multiplying eq<sup>n</sup> (1) by  $\lambda_1$ , we get:

$$c_1 \lambda_1 v_1 + c_2 \lambda_1 v_2 = 0 \quad \dots (3)$$

Subtracting eq<sup>n</sup> (3) from (2), we get:

$$c_2 (\lambda_2 - \lambda_1) v_2 = 0$$

Since,  $\lambda_1 \neq \lambda_2$ ,  $v_2 \neq 0$ , we must have  $c_2 = 0$ .

Similarly, we can show that  $c_1 = 0$ .

Thus, eigenvectors  $v_1$  and  $v_2$  are linearly independent.

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a

$$S$$

$$\text{Domain} = \mathbb{R}^3$$

$$\text{Co-domain} = \mathbb{R}^2$$

T

$$\text{Domain} = \mathbb{R}^2$$

$$\text{Co-domain} = \mathbb{R}^3$$

$$(S \circ T)(x) = S(T(x))$$

$$= S(Bx)$$

$$= ABx$$

Since, co-domain of  $T$  and domain of  $S$  are equal,  
the composite transformation  $SoT$  is defined

$$T: \mathbb{R}^2 \rightarrow \mathbb{R}^3 \quad S: \mathbb{R}^3 \rightarrow \mathbb{R}^2$$

Then,

$$SoT: \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

$$\text{Domain of } SoT = \mathbb{R}^2$$

$$\text{Co-domain of } SoT = \mathbb{R}^2$$

b. Let  $x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$ . Determine  $T(x)$

Solution

$$T(x) = Bx$$

$$= \begin{pmatrix} 3 & 0 \\ 5 & -2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= \begin{pmatrix} 3x_1 \\ 5x_1 - 2x_2 \\ x_2 \end{pmatrix}$$

c.

$$(SoT)(x) = S(T(x))$$

$$= S(Bx)$$

$$= ABx$$

$$= \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 3 & 0 \\ 5 & -2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 3x_1 \\ 5x_1 - 2x_2 \\ x_2 \end{pmatrix}$$

$$= \begin{pmatrix} 3x_1 + 10x_1 - 4x_2 \\ 5x_1 - 2x_2 + x_2 \end{pmatrix}$$

$$= \begin{pmatrix} 13x_1 - 4x_2 \\ 5x_1 - x_2 \end{pmatrix}$$

d.

We know,

$$(SoT)(x) = ABx$$

$$\Rightarrow Cx = ABx$$

$$\therefore C = AB$$

Then,

$$C = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 3 & 0 \\ 5 & -2 \\ 0 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \times 3 + 2 \times 5 + 0 \times 0 & 1 \times 0 + 2 \times -2 + 0 \times 1 \\ 0 \times 3 + 1 \times 5 + 1 \times 0 & 0 \times 0 + 1 \times -2 + 1 \times 1 \end{pmatrix}$$

$$= \begin{pmatrix} 13 & -4 \\ 5 & -1 \end{pmatrix}$$

e.

A transformation  $T$  is said to be linear if

$$T(x+y) = T(x) + T(y)$$

$$T(ax) = aT(x)$$

where, 'a' is a scalar and  $x, y$  are matrices.

Let  $a, b$  be two scalars and  $x, y$  be two matrices. Then,

$$(SoT)(ax+by) = S(T(ax+by))$$

$$\begin{aligned}
 &= S(B(ax+by)) \quad [\because T(x) = Bx] \\
 &= S(aBx + bBy) \\
 &= A(aBx + bBy) \quad [\because S(y) = Ay] \\
 &= aABx + bABy
 \end{aligned}$$

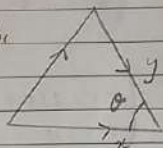
Put  $AB = C$  where  $C$  represents the transformation  $(SoT)$  i.e.  $(SoT)(x) = Cx$ , we get:

$$\begin{aligned}
 (SoT)(ax+by) &= aCx + bCy \\
 &= a(SoT)x + b(SoT)y
 \end{aligned}$$

This proves linearity of transformation  $SoT$ .

Q7

Let  $x, y \in \mathbb{R}^n$  and  $\theta$  be the angle between them. Then, by the law of cosines,



$$\|x-y\|^2 = \|x\|^2 + \|y\|^2 - 2\|x\|\|y\|\cos\theta \quad \dots (1)$$

Now,

$$\begin{aligned}
 \|x-y\|^2 &= (x-y) \cdot (x-y) \\
 &= x \cdot x - x \cdot y - y \cdot x + y \cdot y \\
 &= (x \cdot x) + (y \cdot y) - 2(x \cdot y) \\
 &= \|x\|^2 + \|y\|^2 - 2(x \cdot y) \quad \dots (2)
 \end{aligned}$$

Comparing eqn (1) & (2), we get:

$$x \cdot y = \|x\|\|y\|\cos\theta$$

Q8

Here,  $B = \left\{ \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \right\}$

Now,

$$1 + (-1) + 0 = 0$$

$$\& \quad 1 + 0 + (-1) = 0$$

Both points of  $B$  lie in  $V$  i.e.  $B \in V$

Also, the two points of  $B$  are linearly independent  
 ~~$\forall \lambda, \mu \in \mathbb{R} \quad \lambda \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} + \mu \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \Rightarrow \lambda = \mu = 0$~~

Thus,  $B$  is a basis for  $V$ .

To prove  $V$  is a subspace of  $\mathbb{R}^3$ . Let

$x = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$  be a point in  $\mathbb{R}^3$ .

Then,

$$x_1 + x_2 + x_3 = 0$$

$$\therefore x_1 = -x_2 - x_3$$

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -x_2 - x_3 \\ x_2 \\ x_3 \end{pmatrix}$$

$$= \begin{pmatrix} -x_2 \\ x_2 \\ 0 \end{pmatrix} + \begin{pmatrix} -x_3 \\ 0 \\ x_3 \end{pmatrix}$$

$$= x_2 \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} + x_3 \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \in \text{span} \left\{ \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \right\}$$



$C R^3$

$V$  is subspace of  $R^3$ .

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Here,

$$A = \begin{pmatrix} 5 & -2 \\ 7 & -4 \end{pmatrix}$$

To find eigenvalues:

$$\begin{aligned} |A - \lambda I| &= 0 \\ \Rightarrow \begin{vmatrix} 5-\lambda & -2 \\ 7 & -4-\lambda \end{vmatrix} &= 0 \\ \Rightarrow (5-\lambda)(-4-\lambda) + 14 &= 0 \\ \Rightarrow -20 - 5\lambda + 4\lambda + \lambda^2 + 14 &= 0 \\ \Rightarrow \lambda^2 - \lambda - 6 &= 0 \\ \Rightarrow \lambda^2 - 3\lambda + 2\lambda - 6 &= 0 \\ \Rightarrow \lambda(\lambda-3) + 2(\lambda-3) &= 0 \\ \Rightarrow (\lambda+2)(\lambda-3) &= 0 \\ \therefore \lambda &= -2, \lambda = 3 \end{aligned}$$

Let  $x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$  be eigenvector. Then,

$$\begin{aligned} (A - \lambda I)x &= 0 \\ \Rightarrow \begin{pmatrix} 5-\lambda & -2 \\ 7 & -4-\lambda \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= 0 \quad \dots (1) \end{aligned}$$

Put  $\lambda = -2$  in (1), we get:

$$\begin{aligned} \begin{pmatrix} 7 & -2 \\ 7 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= 0 \\ \Rightarrow \begin{pmatrix} 7x_1 - 2x_2 \\ 7x_1 - 2x_2 \end{pmatrix} &= \begin{pmatrix} 0 \\ 0 \end{pmatrix} \end{aligned}$$

From above,

$$\begin{aligned} 7x_1 - 2x_2 &= 0 \\ \Rightarrow x_1 &= \frac{2}{7}x_2 \end{aligned}$$

$$\begin{aligned} \text{Put } x_2 &= t \in \mathbb{R}, \text{ then,} \\ x_1 &= \frac{2}{7}t \end{aligned}$$

$$\therefore V_{\lambda=-2} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2/7 t \\ t \end{pmatrix}$$

Put  $\lambda = 3$  in (1),

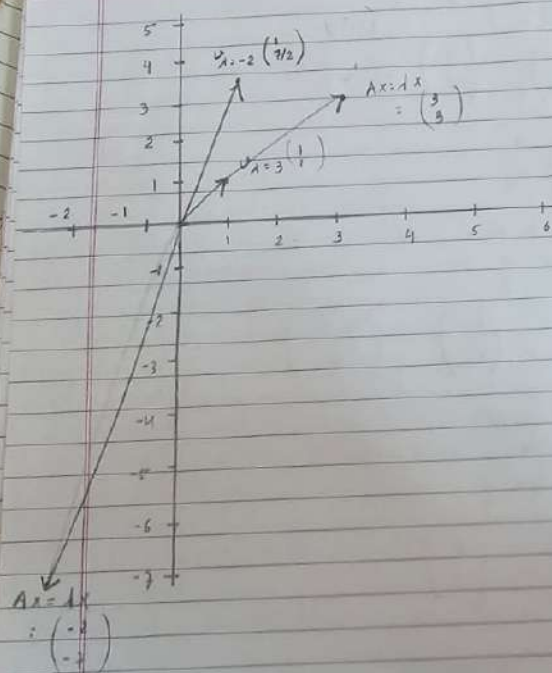
$$\begin{pmatrix} 2 & -2 \\ 7 & -7 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0$$

$$\Rightarrow \begin{pmatrix} 2x_1 - 2x_2 \\ 7x_1 - 7x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

From above,

$$\begin{aligned} x_1 &= x_2 \\ \text{Put } x_2 &= s \in \mathbb{R}, \text{ then,} \\ x_1 &= s \end{aligned}$$

$$\therefore V_{\lambda=3} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} s \\ s \end{pmatrix}$$



Q10.

a. Given,  $Au = \lambda u$  ... (i)

Multiplying eq<sup>n</sup> (i) by  $A$ , we get

$$A(Au) = A(\lambda u)$$

$$\Rightarrow A^2 u = \lambda(Au)$$

$$\Rightarrow A^2 u = \lambda(\lambda u) \quad (\text{From (i), } Au = \lambda u)$$

$$\Rightarrow A^2 u = \lambda^2 u \quad \dots (ii)$$

Performing (ii) twice multiplication of matrix  $A$  on both side of eq<sup>n</sup> (ii), we get

$$A^m u = \lambda^m u \quad \dots (iii)$$

Equation (iii) prove that  $\lambda^m$  is the eigenvalue of matrix  $A^m$  and corresponding eigenvector is  $u$ .

b.

Here, matrix  $A$  is invertible i.e.  $A^{-1}$  exists.

From (i), we have

$$Au = \lambda u$$

Multiplying both side by  $A^{-1}$ , we get

$$A^{-1}Au = A^{-1}\lambda u$$

$$\Rightarrow Iu = A^{-1}\lambda u$$

$$\Rightarrow u = A^{-1}\lambda u$$

$$\Rightarrow A^{-1}u = \frac{1}{\lambda} u$$

$$\therefore A^{-1}u = \lambda^{-1} u \quad \dots (iv)$$

Equation (iv) is in the form of (i), so which the eigen value of matrix  $A^{-1}$  is  $1/\lambda$  (or  $\lambda^{-1}$ ) and eigenvector is  $u$ .