

Inner Product of Vectors (Dot Product | Scalar Product)

Inner product of 2 vectors $u = (u_1, u_2, \dots, u_n)$ and $v = (v_1, v_2, \dots, v_n)$ from the vector space \mathbb{R}^n is defined as $u \cdot v = u_1 v_1 + u_2 v_2 + \dots + u_n v_n$.

It's also defined as $u \cdot v = u^T v$.

Note: Inner product is a scalar. Compute $u \cdot v$ and

1. Compute $u \cdot v$ and $v \cdot u$ for $u = \begin{bmatrix} 2 \\ -5 \\ -1 \end{bmatrix}$ & $v = \begin{bmatrix} 3 \\ 2 \\ -3 \end{bmatrix}$

$$u \cdot v = u_1 v_1 + u_2 v_2 + u_3 v_3$$

$$= 2(3) + (-5)(2) + (-1)(-3)$$

$$= 6 - 10 + 3$$

$$v \cdot u = v_1 u_1 + v_2 u_2 + v_3 u_3$$

$$= 3(2) + (2)(-5) + (-3)(-1)$$

$$= 6 - 10 + 3$$

$$= -1$$

Theorem: Let u, v, w be vectors in \mathbb{R}^n and c be a scalar then

- i) $u \cdot u = \|u\|^2$
- ii) $(u + v) \cdot w = (u \cdot w) + (v \cdot w)$
- iii) $(cu) \cdot v = c(u \cdot v) = u \cdot (cv)$
- iv) $u \cdot u \geq 0$

$u \cdot u = 0$ if and only if $u = 0$

(norm)

Defn: Length of a vector $v = (v_1, v_2, \dots, v_n)$ is a non-negative scalar denoted by $\|v\|$ is defined as $\|v\| = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2}$

norm

For any scalar c $\|cv\| = |c| \|v\|$

$$c = -1 \Rightarrow \|cv\| = \|(-1)v\| = |-1| \|v\| = \|v\|$$

Unit Vector: A vector with unit length is called unit vector.

Given a non zero vector v we obtain the unit vector by dividing v by its length.

$$\text{i.e. } \hat{v} = \frac{\vec{v}}{\|v\|}$$

This process of creating a unit vector from a given non zero vector is called normalization.

Distance b/w 2 vectors:

If $u = (u_1, u_2, \dots, u_n)$ & $v = (v_1, v_2, \dots, v_n)$ are 2 vectors, then

$$\|u - v\| = \|v - u\| = \sqrt{(u_1 - v_1)^2 + (u_2 - v_2)^2 + \dots + (u_n - v_n)^2}$$

Angle b/w 2 vectors u and v is given by the formula

$$u \cdot v = \|u\| \|v\| \cos \theta \quad (01)$$

$$\theta = \cos^{-1} \left(\frac{u \cdot v}{\|u\| \|v\|} \right)$$

If $\theta = 90^\circ$; $\cos 90^\circ = 0$; $\therefore u \cdot v = 0$
and we say that u & v are orthogonal.

Orthogonal Vectors:

Two vectors \vec{u} & \vec{v} in R^n are orthogonal to each other if $\vec{u} \cdot \vec{v} = 0$ (zero).

Note: 0 vector is orthogonal to every vector in R^n .

Since $0 \cdot u = 0$ for all u .

Orthogonal Set:

A set of vectors $\{v_1, v_2, \dots, v_p\}$ in R^n is said to be an orthogonal set if $v_i \cdot v_j = 0$ (0's) each pair of distinct vectors from the set is orthogonal for $i \neq j$.

Theorem: If $S = \{v_1, v_2, \dots, v_p\}$ is an orthogonal set of non zero vectors in R^n then S is linearly independent and hence a basis for the subspace spanned by S .

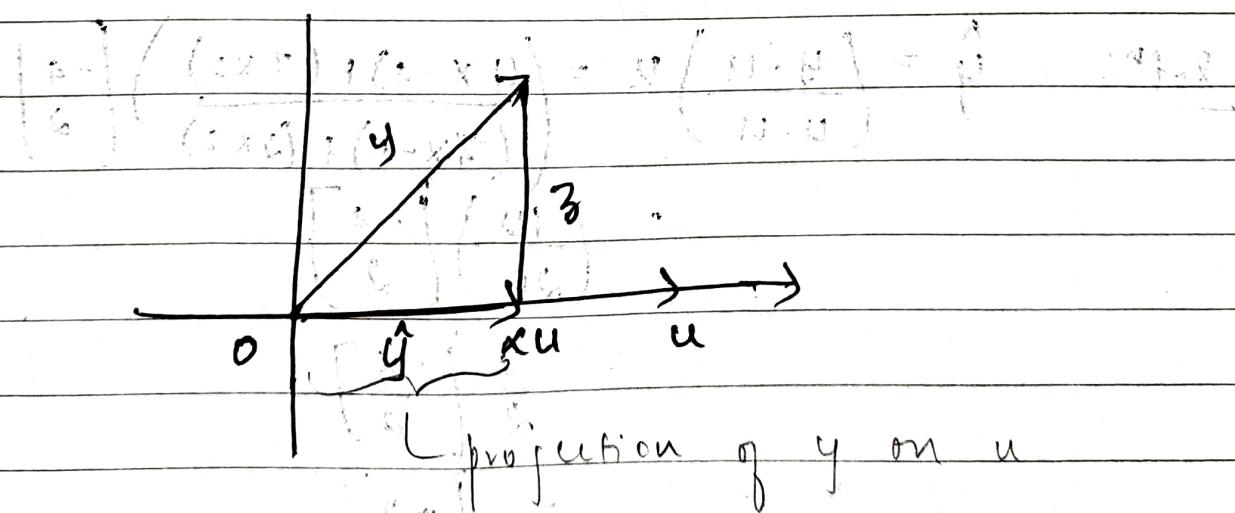
Date 08 / 07 / 2023

Orthogonal Basis for a subspace $\text{B}(1)(\mathbb{R}^n)$ is a basis for \mathbb{H} that is also a orthogonal set.

Orthogonal Projection: given a non zero vector u in \mathbb{R}^n consider the problem of decomposing it a vector y in \mathbb{R}^n into the sum of 2 vectors, one a multiple of u and the other orthogonal to u if we write $y = \hat{y} + z$.

Here \hat{y} is multiple of u and z is orthogonal to u .

The vector $\hat{y} = \frac{(y \cdot u)}{(u \cdot u)}u$ is called the orthogonal projection of y onto u .



Imp 1: Let $y = \begin{bmatrix} 7 \\ 4 \\ 6 \end{bmatrix}$ and $u = \begin{bmatrix} 4 \\ 2 \\ 2 \end{bmatrix}$ find the orthogonal projection of u onto u .

Soln: \hat{y} is orthogonal projection of y onto u

$$\hat{y} = \left(\frac{y \cdot u}{u \cdot u} \right) u = \left(\frac{(7 \times 4) + (6 \times 2)}{(4 \times 4) + (2 \times 2)} \right) \begin{bmatrix} 4 \\ 2 \\ 2 \end{bmatrix}$$

$$= \left(\frac{40}{20} \right) \begin{bmatrix} 4 \\ 2 \\ 2 \end{bmatrix}$$

$$= 2 \begin{bmatrix} 4 \\ 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 8 \\ 4 \\ 4 \end{bmatrix}$$

2. Compute the orthogonal projection of $\begin{bmatrix} 9 \\ 1 \\ 7 \end{bmatrix}$ onto the line through $\begin{bmatrix} -4 \\ 2 \\ 2 \end{bmatrix}$ and the origin.

Soln: $\hat{y} = \left(\frac{y \cdot u}{u \cdot u} \right) u = \left(\frac{(1 \times -4) + (7 \times 2)}{(-4 \times -4) + (2 \times 2)} \right) \begin{bmatrix} -4 \\ 2 \\ 2 \end{bmatrix}$

$$= \left(\frac{10}{20} \right) \begin{bmatrix} -4 \\ 2 \\ 2 \end{bmatrix}$$

$$= \frac{1}{2} \begin{bmatrix} -4 \\ 2 \\ 2 \end{bmatrix}$$

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

Orthonormal set: A set $\{u_1, u_2, \dots, u_p\}$ is an orthonormal set if it is an orthogonal set of unit vectors.

If W is the subspace spanned by such a set then this set is called an orthonormal basis for W . Since orthonormal the set is automatically linearly independent.

$$f(1) + f(2) + f(-1) = 1 + 0 + 0 = 1$$

Example: $\{\begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix}\}$

The standard basis $\{e_1, e_2, \dots, e_n\}$ of \mathbb{R}^n is an orthonormal basis for \mathbb{R}^n .

1. Show that the set $\{v_1, v_2, v_3\}$ is an orthonormal basis for \mathbb{R}^3 where

$$v_1 = \begin{bmatrix} 3/\sqrt{11} \\ 1/\sqrt{11} \\ 1/\sqrt{11} \end{bmatrix} \quad v_2 = \begin{bmatrix} -1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{bmatrix} \quad v_3 = \begin{bmatrix} -1/\sqrt{66} \\ -4/\sqrt{66} \\ 7/\sqrt{66} \end{bmatrix}$$

$$\text{Soln: } v_1 \cdot v_2 = \left(\frac{-3}{\sqrt{66}} + \frac{2}{\sqrt{66}} + \frac{1}{\sqrt{66}} \right) \\ = 0$$

$$v_2 \cdot v_3 = \left(\frac{1}{\sqrt{66} \times 6} + \frac{-8}{\sqrt{66} \times 6} + \frac{7}{\sqrt{66} \times 6} \right) \\ = 0$$

$$v_1 \cdot v_3 = \left(\frac{-3}{\sqrt{11} \times 66} - \frac{4}{\sqrt{11} \times 66} + \frac{7}{\sqrt{11} \times 66} \right) \\ = 0$$

Saathi

$\{v_1, v_2, v_3\}$ is orthogonal.

$$\|v_1\| = \sqrt{v_1^2 + v_2^2 + v_3^2}$$

$$= \sqrt{\left(\frac{3}{\sqrt{6}}\right)^2 + \left(\frac{1}{\sqrt{6}}\right)^2 + \left(\frac{1}{\sqrt{6}}\right)^2}$$

$$= \sqrt{\frac{9}{6} + \frac{1}{6} + \frac{1}{6}}$$

$$= \sqrt{\frac{11}{6}} = 1$$

$$\|v_2\| = \sqrt{\left(\frac{-1}{\sqrt{6}}\right)^2 + \left(\frac{2}{\sqrt{6}}\right)^2 + \left(\frac{1}{\sqrt{6}}\right)^2}$$

$$= \sqrt{\frac{1}{6} + \frac{4}{6} + \frac{1}{6}}$$

$$= \sqrt{\frac{6}{6}} = 1$$

$$\|v_3\| = \sqrt{\frac{+1}{6} + \frac{16}{6} + \frac{49}{6}}$$

$$= \sqrt{\frac{66}{66}} = 1$$

04/04/23

\therefore The given is orthonormal basis for R^3 .

Gram-Schmidt's Orthogonalization process

Given a basis $\{x_1, x_2, \dots, x_p\}$ for a subspace $S \subseteq R^n$, The following algorithm namely gram-schmidt finds algorithm forms a orthogonal basis for S , namely $\{v_1, v_2, \dots, v_p\}$ where $v_i = x_i - \sum_{j=1}^{i-1} \frac{x_j \cdot v_j}{x_i \cdot v_j} v_j$.

$$v_3 = x_3 - \left(\frac{x_3 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{x_3 \cdot v_2}{v_2 \cdot v_2} \right) v_2$$

$$v_4 = x_4 - \left(\frac{x_4 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{x_4 \cdot v_2}{v_2 \cdot v_2} \right) v_2 - \left(\frac{x_4 \cdot v_3}{v_3 \cdot v_3} \right) v_3$$

$$v_p = x_p - \left(\frac{x_p \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \dots - \left(\frac{x_p \cdot v_{p-1}}{v_{p-1} \cdot v_{p-1}} \right) v_{p-1}$$

Note: Normalizing each of the vectors v_1, v_2, \dots, v_p , we get an orthogonal basis

$\{u_1, u_2, \dots, u_p\}$ when $u_i = \frac{v_i}{\|v_i\|}, 1 \leq i \leq p$.

Saathi

*Saathit**Ans*

Given a basis $\{x_1, x_2, x_3\}$ where
 $x_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$, $x_2 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$, $x_3 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ of a
 subspace W , construct an orthonormal
 basis.

Using Gram-Schmidt's orthogonal process,

$$v_1 = x_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

$$v_2 = x_2 - \left(\frac{x_2 \cdot v_1}{v_1 \cdot v_1} \right) v_1$$

$$v_2 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} - \left(\frac{1}{2} \right) \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} \\ 0 \\ -\frac{1}{2} \end{bmatrix}$$

$$v_3 = x_3 - \left(\frac{x_3 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{x_3 \cdot v_2}{v_2 \cdot v_2} \right) v_2$$

$$v_3 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} - \left(\frac{1}{2} \right) \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} - \left(\frac{1}{2} \right) \begin{bmatrix} \frac{1}{2} \\ 0 \\ -\frac{1}{2} \end{bmatrix}$$

$$v_3 = \begin{bmatrix} \frac{1}{2} \\ 1 \\ -\frac{1}{2} \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} - \frac{1}{2} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{2} \\ 1 \\ -\frac{1}{2} \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} - \begin{bmatrix} \frac{1}{2} \\ 0 \\ \frac{1}{2} \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ 1 \\ -\frac{1}{2} \end{bmatrix}$$

$$\text{Thus } \{v_1, v_2, v_3\} \text{ is an orthogonal basis}$$

$$\text{Ex: } \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} \frac{1}{2} \\ 0 \\ -\frac{1}{2} \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -\frac{1}{2} \end{bmatrix} \right\} \text{ is an orthogonal basis.}$$

To obtain orthonormal basis we have

to normalize each of the

v_1, v_2, v_3

$$\text{Now } v_1 = v_1 = (1, 0, 1) = \left(\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right)$$

$$\|v_2\| = \sqrt{\left(\frac{1}{2}\right)^2 + 0^2 + \left(-\frac{1}{2}\right)^2} = \sqrt{\frac{1}{2}}$$

$$v_2 = \frac{v_2}{\|v_2\|} = \left(\frac{1}{2}, 0, -\frac{1}{2} \right) = \left(\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}} \right)$$

$$\|v_3\| = \sqrt{\left(\frac{1}{2}\right)^2 + 0^2 + \left(-\frac{1}{2}\right)^2} = \sqrt{\frac{1}{2}}$$

$$v_3 = \frac{v_3}{\|v_3\|} = \left(\frac{1}{2}, 0, -\frac{1}{2} \right) = \left(\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}} \right)$$

$$\begin{aligned} \mathbf{u}_3 &= \frac{\mathbf{v}_3}{\|\mathbf{v}_3\|} = \frac{(0, 1, 0)}{\sqrt{0^2 + 1^2 + 0^2}} = \frac{(0, 1, 0)}{1} \\ &= (0, 1, 0) \end{aligned}$$

∴ Orthonormal basis for \mathbb{W} is $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$

$$\text{Ex. } \left\{ \begin{bmatrix} \mathbf{u}_1 \\ 0 \end{bmatrix}, \begin{bmatrix} \mathbf{u}_2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}.$$

Answer

2. Construct an orthonormal basis for a

subspace \mathbb{W} of \mathbb{R}^4 where $\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$ with $\mathbf{x}_1 = \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix}, \mathbf{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \mathbf{x}_3 = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$ is a

$$\begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$

basis for \mathbb{W} .

Soln:

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix}$$

$$\mathbf{v}_2 = \mathbf{x}_2 - \left(\frac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \mathbf{v}_1$$

$$= \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \end{bmatrix} - \left(\frac{3}{4} \right) \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix}$$

$$\mathbf{v}_1' = \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix} \quad \begin{array}{l} \text{(Scaling is applied only to} \\ \text{simplify the calculation)} \end{array}$$

$$\begin{array}{l} \text{(multiply vector by} \\ \text{suitable no. so} \\ \text{all components} \\ \text{belonging to below)} \end{array}$$

$$\begin{aligned} \mathbf{v}_3 &= \mathbf{x}_3 - \left(\frac{\mathbf{x}_3 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \mathbf{v}_1 - \left(\frac{\mathbf{x}_3 \cdot \mathbf{v}_2'}{\mathbf{v}_2' \cdot \mathbf{v}_2'} \right) \mathbf{v}_2' \\ &= \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix} - \left(\frac{2}{2} \right) \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix} - \left(\frac{-3}{2} \right) \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix} \\ &= \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix} \end{aligned}$$

∴ $\{\mathbf{v}_1, \mathbf{v}_2', \mathbf{v}_3\}$ are orthogonal basis for \mathbb{W}

$$u_1 = \frac{v_1}{\|v_1\|} = \frac{(1, 1, 1, 1)}{\sqrt{1^2 + 1^2 + 1^2 + 1^2}}$$

$$= \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2} \right)$$

$$u_2 = \frac{v_2}{\|v_2\|} = \frac{(-3, 1, 1, 1)}{\sqrt{(-3)^2 + (1)^2 + 1^2 + 1^2}}$$

$$= \left(\frac{-3}{2\sqrt{3}}, \frac{1}{2\sqrt{3}}, \frac{1}{2\sqrt{3}}, \frac{1}{2\sqrt{3}} \right)$$

$$u_3 = \frac{v_3}{\|v_3\|} = \frac{(0, -2/3, 1/3, 1/3)}{\sqrt{0^2 + (-2/3)^2 + (1/3)^2 + (1/3)^2}}$$

$$= \left(0, -\frac{2}{3}, \frac{1}{3}, \frac{1}{3} \right)$$

$$= \left(0, -\frac{2}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}} \right)$$

Thus u_1, u_2, u_3 forms an orthonormal basis.

In example 3 the term xy is called cross product term.

1. $\det_{n \times n} x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$, compute $x^T A x$ for the following matrix.
 - i) $A = \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix}$
 - ii) $A = \begin{bmatrix} 3 & -2 \\ -2 & 7 \end{bmatrix}$

Quadratic forms: A quadratic form on R^n is a function $Q(x)$ defined as $Q(x) = x^T A x$ where A is an $n \times n$ symmetric matrix and x is a vector in R^n .

In other words a quadratic form is a homogeneous expression of 2nd degree in any number of variables.

1. $Q(x) = x^2$
2. $Q(x, y) = x^2 + y^2$
3. $Q(x, y) = x^2 + y^2 + 2xy$
4. $Q(x, y, z) = 2x^2 + 3y^2 + 5z^2 + 2xy + 6yz$ (cross product)

$$\text{Ans} \quad A = \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix}$$

$$A = \begin{bmatrix} 3 & -2 \\ -2 & 7 \end{bmatrix}$$

$$\Rightarrow x^T A x = [x_1 \ x_2] \begin{bmatrix} 3 & -2 \\ -2 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\Rightarrow Ax = \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [x_1 \ x_2] \begin{bmatrix} 4x_1 & 0 \\ 0 & 3x_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [x_1 \ x_2] \begin{bmatrix} 4x_1^2 & 0 \\ 0 & 3x_2^2 \end{bmatrix} = [x_1 \ x_2] \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [x_1 \ x_2] \begin{bmatrix} 4x_1 & 0 \\ 0 & 3x_2 \end{bmatrix} = 3x_1^2 + 7x_2^2$$

$$\Rightarrow [x_1 \ x_2] \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [x_1 \ x_2] \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \frac{4x_1}{3x_2} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \frac{4x_1^2}{3x_2^2}$$

$$= 4x_1^2 + 3x_2^2.$$

Ans: The cross product term $-4x_1x_2$ appears because of the off diagonal element of A .

i.e. A is called matrix quadratic form

2. For $x \in \mathbb{R}^3$, let $Q(x) = 5x_1^2 + 3x_2^2 + 2x_3^2 - x_1x_2 + 8x_1x_3$, write this quadratic form as $x^T A x$

$$\text{Soln: } x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad \text{Matrix of } Q(x) \text{ is } \begin{bmatrix} 5 & -1 & 8 \\ -1 & 3 & 0 \\ 8 & 0 & 2 \end{bmatrix} \quad \text{Diagonal elements need to write in 2 pos}$$

Saathī

$$g(\mathbf{x}) = \mathbf{x}^T A \mathbf{x} = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} 5 & -1 & 0 \\ -1 & 3 & 4 \\ 0 & 4 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$$= [x_1 \ x_2 \ x_3] \begin{bmatrix} 5x_1 \\ -x_1 + 3x_2 + 4x_3 \\ 4x_2 + 2x_3 \end{bmatrix}$$

Classifications (Nature of Quadratic Form):

A Quadratic form $Q(\mathbf{x})$ in

- i) Positive definite if $Q(\mathbf{x}) > 0$ for all $\mathbf{x} \neq 0$
- ii) Negative definite if $Q(\mathbf{x}) < 0$ for all $\mathbf{x} \neq 0$
- iii) Indefinite if some eigen values are +ve and some are -ve (no zero)
- iv) Positive semidefinite if its eigen values are ≥ 0 (if one is 0)
- v) Negative semidefinite if its eigen values are ≤ 0 (with at least one is 0)

Classification in terms of Eigen Value:
A Quadratic form $g(\mathbf{x})$ is

- i) Positive definite if all its eigen values are +ve (> 0)

Saathī

Find the nature of following quadratic forms

$$Q(x) = x^3 + 5y^2 + z^2 + 2xy + 2yz + 6xz$$

$$A = \begin{bmatrix} 1 & 1 & 3 \\ 1 & 5 & 1 \\ 3 & 1 & 1 \end{bmatrix}$$

$$\lambda_3 = (-1)^{n+1} \lambda_1 + \left(\sum_{\text{of cofactors}} \lambda - \det(A) = 0 \right)$$

$$\lambda^3 - 7\lambda^2 + 36 = 0$$

$$\lambda = -2, 6$$

Eigen values are $-2, 6, 3$.
 $Q(x)$ is indefinite since A has one +ve & two -ve eigen values.

$$\text{iii) } Q(x) = 3x^2 + 5y^2 + 3z^2 - 2yz + 2zx - 2xy$$

卷之三

$$A = \begin{pmatrix} 3 & -1 & 1 \end{pmatrix}$$

$$\begin{pmatrix} -1 & 5 & -1 \\ -1 & -1 & 3 \end{pmatrix}$$

Eigen values are 6, 3, 2
 $B(A)$ is positive definite, since A has +ve eigen values

iii) $T_2 = g(x) = 3x_1^2 + 2x_2^2 + x_3^2 + 4x_1x_2 + 4x_2x_3$ positive

definiti

$$A = \begin{bmatrix} 3 & 2 & 0 \\ 2 & 2 & 2 \\ 0 & 2 & 1 \end{bmatrix}$$

$$\lambda^3 - (\text{tr}(A))\lambda^2 + \left(\sum_{\text{Diagonal}} \text{cofactors} \right) \lambda = \text{det}(A)$$

$$\lambda^3 - 6\lambda^2 + (-2+3+2)\lambda - (-10) = 0$$

$$\lambda = -1, 5, 2$$

eige

values are - i.e.,
Indefinite sine

A. H. A. + ve & - ve

Reduction of Quadratic form to Canonical form by change of variable:

Q(x) If the quadratic form $Q(x)$ where x is a variable vector in \mathbb{R}^n to write it in the canonical form (only the sum of square terms) we make change of variable in the equation

$$\text{i. } x = Py \quad (\text{or}) \quad y = P^{-1}x \quad \dots \quad (1)$$

Here P is an invertible matrix whose columns are eigen vectors of the matrix of the quadratic form.

so the change of variable (1) made

$$\begin{aligned} \text{In quadratic form } Q(x) &= x^T A x \\ Q(x) &= x^T A x = (Py)^T A y = y^T P^T A y \end{aligned}$$

$$(P^T A P) y = y^T (P^T A P) y$$

and the new matrix of the quadratic form is $P^T A P$

If P orthogonally diagonalizes A then

$$\therefore \text{eigen values are } 3 \text{ & } -4.$$

Problems
 i. Matrix of the new quadratic form
 $= P^T A P = P^{-1} A P = D$
 Thus the Matrix of the new quadratic form is Diagonal of A
 Here eigen values are the diagonal elements of D

Soln: The matrix of the given quadratic form is
 $A = \begin{bmatrix} 1 & 4 & -4 \\ 0 & -4 & -5 \\ 0 & 0 & 3 \end{bmatrix}$
 $\lambda^2 - \text{tr}(A)\lambda + \det(A) = 0 \dots$
 $\lambda^2 - (-4)\lambda + (-21) = 0$
 $\lambda^2 + 4\lambda - 21 = 0$
 $\lambda = 3, -7$

Let $\mathbf{x} = \mathbf{P}\mathbf{y}$ or $\mathbf{y} = \mathbf{P}^{-1}\mathbf{x}$
 \mathbf{P} contains the eigen vectors as its columns
 \therefore new quadratic form is written using the matrix $D = \begin{bmatrix} 3 & 0 \\ 0 & -7 \end{bmatrix}$

the new quadratic form is:

$$Q(\mathbf{y}) = \underline{\underline{3y_1^2 - 7y_2^2}}$$

$$D = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & -7 \end{bmatrix}$$

eigen values are 6, 3, 2. \mathbf{P} should be in T order

2. Reduce the quadratic form $3x_2^2 + 5y^2 + 3y^2 + 2y_3 + 2x_2 - 2xy$ to the canonical form by changing variables.

Soln: Here $Q(\mathbf{x}) = 3x_2^2 + 5y^2 + 3y^2 - 2y_3 + 2x_2 - 2xy$

where $\mathbf{x} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$

\mathbf{P} contains columns

of the eigen vector

3. Find the maximum and minimum values of the quadratic form $Q(\mathbf{x}) = 9x_1^2 + 4x_2^2 + 3x_3^2$ subject to the constraint $\mathbf{x}^T \mathbf{x} = 1$

Given $\mathbf{x}^T \mathbf{x} = 1$

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad \text{Then } \mathbf{x}^T \mathbf{x} = [x_1 \ x_2 \ x_3]^T \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

and \mathbf{y} is new vector

Given

$$\text{i.e. } x_1^2 + x_2^2 + x_3^2 = 1$$

$$\text{Let } \mathbf{y} = \begin{bmatrix} p \\ q \\ r \end{bmatrix}$$

$$\begin{aligned} A^3 - (\text{tr } A) \lambda^2 + \left(\frac{\text{sum of diagonal elements}}{\text{no. of factors}} \right) \lambda - \det(A) &= 0 \\ \lambda^3 - 11\lambda^2 + (14 + 8 - 14) \lambda - 36 &= 0 \\ \lambda^3 - 11\lambda^2 + 36\lambda - 36 &= 0 \\ \lambda = 6, 3, 2 & \end{aligned}$$

look for max coefficient

Saathi

Date _____

Date: 18 / 07 / 2023

observe that $4x_1^2 \leq 9x_2^2$, $3x_2^2 \leq 9x_3^2$

$$\begin{aligned} g(x) &= 9x_1^2 + 4x_2^2 + 3x_3^2 \\ &\leq 9x_1^2 + 9x_2^2 + 9x_3^2 \\ &= 9(9x_1^2 + x_2^2 + x_3^2) \end{aligned}$$

$$= 9l_1$$

$$= 9$$

$\therefore g(x) \leq 9$
 $\therefore g$ is the maximum value of $g(x)$

observe that $4x_1^2 \geq 3x_2^2$, $9x_1^2 \geq 3x_2^2$

$$\begin{aligned} g(x) &= 9x_1^2 + 4x_2^2 + 3x_3^2 \\ &\geq 3x_1^2 + 3x_2^2 + 3x_3^2 \\ &= 3(x_1^2 + x_2^2 + x_3^2) \\ &= 3l_1 + 3l_2 \\ &= 3l_1 + 3 \times 1 \\ &= 3l_1 + 3 \end{aligned}$$

$\therefore 3$ is the minimum value of $g(x)$

If A is $m \times n$ matrix with linearly independent columns, then by applying Gram-Schmidt process with normalization A can be factorised as

$A = QR$ where Q is an $m \times n$

matrix whose columns form an orthonormal basis for column space of A and R is an $n \times n$ matrix with positive entries on its diagonal.

Find a QR factorization of $A = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix}$

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}_{3 \times 3}$$

observe that all the columns of A are linearly independent.
 \therefore These columns form a basis for a subspace W of \mathbb{R}^4 .

By applying Gram-Schmidt process to get an orthonormal basis

$$\det \mathbf{A} = \mathbf{x}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \quad \mathbf{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, \quad \mathbf{x}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\mathbf{v}_1 = \mathbf{x}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\mathbf{v}_3' = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$$

$$\mathbf{v}_2' = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\mathbf{v}_2 = \mathbf{x}_2 - \left(\frac{\mathbf{x}_2 \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|} \right) \mathbf{v}_1$$

$$\mathbf{u}_1 = \frac{1}{\sqrt{3}} \mathbf{v}_1 = \frac{1}{\sqrt{3}} (1, 1, 1, 1) = \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right)$$

$$\mathbf{v}_2' = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} - \left(\frac{3}{4} \right) \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\mathbf{u}_2 = \frac{\mathbf{v}_2'}{\|\mathbf{v}_2'\|} = \frac{(-3, 1, 1, 1)}{\sqrt{(-3)^2 + 1^2 + 1^2 + 1^2}} = \left(\frac{-3}{\sqrt{12}}, \frac{1}{\sqrt{12}}, \frac{1}{\sqrt{12}}, \frac{1}{\sqrt{12}} \right)$$

$$\mathbf{v}_2' = \begin{bmatrix} -3/4 \\ 1/4 \\ 1/4 \end{bmatrix} \text{ Multiply by } 4 = \begin{bmatrix} -3 \\ 1 \\ 1 \end{bmatrix} = \mathbf{v}_2'$$

$$\mathbf{u}_3 = \frac{\mathbf{v}_3'}{\|\mathbf{v}_3'\|} = \frac{(0, -2, 1, 1)}{\sqrt{(0)^2 + (-2)^2 + 1^2 + 1^2}} = \left(\frac{0}{\sqrt{6}}, \frac{-2}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}} \right)$$

$$\mathbf{v}_3 = \mathbf{x}_3 - \left(\frac{\mathbf{x}_3 \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|} \right) \mathbf{v}_1 - \left(\frac{\mathbf{x}_3 \cdot \mathbf{v}_2}{\|\mathbf{v}_2\|} \right) \mathbf{v}_2$$

$$= \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} - \left(\frac{25}{3} \right) \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \left(\frac{25}{3} \right) \begin{bmatrix} -3 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} -25/3 \\ 25/3 \\ 25/3 \end{bmatrix} = \begin{bmatrix} -25/3 \\ 25/3 \\ 25/3 \end{bmatrix}$$

on scaling \mathbf{v}_3 is multiply by 3.

Normalizing,

$$\therefore \mathbf{Q} = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ \frac{1}{2} & \frac{\sqrt{3}}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{\sqrt{3}}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{\sqrt{3}}{2} & \frac{1}{2} \end{bmatrix} = [\mathbf{u}_1 \ \mathbf{u}_2 \ \mathbf{u}_3]$$

$\det \mathbf{A} = \mathbf{Q}^T \mathbf{R}$ for some \mathbf{R} . Pre multiply by \mathbf{Q}^T

$$\mathbf{Q}^T \mathbf{A} = \mathbf{Q}^T \mathbf{Q} \mathbf{R} = (\mathbf{Q}^T \mathbf{Q}) \mathbf{R} = \mathbf{I} \mathbf{R} = \mathbf{R}$$

By applying Gram Schmidt process A

$$R = Q^T A = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ -\frac{\sqrt{3}}{2} & \frac{\sqrt{3}}{2} & \frac{1}{2}\sqrt{3} \\ 0 & -\frac{1}{2}\sqrt{6} & \frac{1}{2}\sqrt{6} \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

$$R = \begin{bmatrix} 2 & \frac{3}{2} & \frac{1}{2} \\ 0 & \sqrt{3}/2 & \frac{1}{2}\sqrt{3} \\ 0 & 0 & 2/\sqrt{6} \end{bmatrix} \text{ Thus } A = QR$$

$$\text{Verification: } QR = \begin{bmatrix} \frac{1}{2} & -\frac{\sqrt{3}}{2} & 0 \\ \frac{1}{2} & \frac{\sqrt{3}}{2} & \frac{1}{2}\sqrt{3} \\ 0 & 0 & \frac{1}{2}\sqrt{6} \end{bmatrix} \text{ where } Q \text{ & } R \text{ are}$$

$$x_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, x_2 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, x_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$v_1 = v_1$$

$$v_2 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \left(\frac{1}{3} \right) \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$v_3 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \left(\frac{1}{3} \right) \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$v_3 = x_3 - \left(\frac{x_3 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{x_3 \cdot v_2}{v_2 \cdot v_2} \right) v_2$$

$$2. \text{ Obtain QR factorization of } A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 2 & 1 & 2 \end{bmatrix}$$

$$\text{Matrix } A \text{ has columns } \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 2 & 1 & 2 \end{bmatrix} \text{ which give basis for } \text{span } W \text{ of } \mathbb{R}^4.$$

Observe that all the columns are

linearly independent. Columns form a basis for

$$W = \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \\ 1 \end{bmatrix} \right\}$$

$$= \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \\ 1 \end{bmatrix} \right\}$$

$$= \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \\ 1 \end{bmatrix} \right\}$$

Saathi

$$u_1 = \frac{v_1}{\|v_1\|} = \frac{(1, 2, 1, 1)}{\sqrt{1^2 + 2^2 + 1^2 + 1^2}} = \left(\frac{1}{\sqrt{7}}, \frac{2}{\sqrt{7}}, \frac{1}{\sqrt{7}}, \frac{1}{\sqrt{7}} \right)$$

$$u_2 = \frac{v_2}{\|v_2\|} = \frac{3, -1, 3, 4}{\sqrt{3^2 + (-1)^2 + 3^2 + 4^2}} = \left(\frac{3}{\sqrt{35}}, \frac{-1}{\sqrt{35}}, \frac{3}{\sqrt{35}}, \frac{4}{\sqrt{35}} \right)$$

$$u_3 = \frac{v_3}{\|v_3\|} = \frac{-9, -2, 11, 2}{\sqrt{(-9)^2 + (-2)^2 + 11^2 + 2^2}} = \left(\frac{-9}{\sqrt{210}}, \frac{-2}{\sqrt{210}}, \frac{11}{\sqrt{210}}, \frac{2}{\sqrt{210}} \right)$$

$$Q = \begin{bmatrix} \frac{1}{\sqrt{7}} & \frac{3}{\sqrt{35}} & \frac{-9}{\sqrt{210}} \\ \frac{1}{\sqrt{7}} & -\frac{1}{\sqrt{35}} & \frac{-2}{\sqrt{210}} \\ \frac{3}{\sqrt{7}} & \frac{3}{\sqrt{35}} & \frac{11}{\sqrt{210}} \\ \frac{1}{\sqrt{7}} & \frac{4}{\sqrt{35}} & \frac{2}{\sqrt{210}} \end{bmatrix}$$

Let $A = Q^T B$ for some $R \in \mathbb{R}^{4 \times n}$.
 Premultiply by Q^T

$$Q^T A = Q^T Q^T B = I R = R$$

$$R = Q^T B = \begin{bmatrix} \frac{1}{\sqrt{7}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{3}{\sqrt{35}} & -\frac{1}{\sqrt{35}} & \frac{3}{\sqrt{35}} & \frac{4}{\sqrt{35}} \\ \frac{9}{\sqrt{210}} & -\frac{2}{\sqrt{210}} & \frac{11}{\sqrt{210}} & \frac{2}{\sqrt{210}} \\ \frac{1}{\sqrt{210}} & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 2 \\ 2 & 1 & 4 \\ 1 & 1 & 4 \\ 1 & 0 & 1 \end{bmatrix}$$

The name least square arises from the fact that this $\|B - Ax\|$ is square root of sum of squares of the components.

If A is an ~~an~~ ^{an} ~~matrix~~ matrix and B is in \mathbb{R}^m , a least square solution of $AX = B$, is a solution denoted by \hat{x} in \mathbb{R}^n such that this $\|B - A\hat{x}\| \leq \|B - Ax\|$

The matrix equation $A^T A \hat{x} = A^T B$ is called normal equations for $AX = B$.

Least Square Solution:

Inconsistent systems $AX = B$ arise often in applications.

When a soln is demanded but now exists or the but one can do is to find a solution x such that the difference between AX & B is very very small.

The general least square problem is to find a solution x that makes $\|B - Ax\|$ as small as possible.

Solution of normal equations in the least square solution \hat{x} .

1. Find a least square solution of $Ax = B$

where $A = \begin{bmatrix} 4 & 0 \\ 0 & 2 \end{bmatrix}$ and $B = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$

$$\begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & 1 \end{bmatrix} \begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

 3×2 3×1

Soln: The normal equations of $Ax = B$ is

$$A^T A \hat{x} = A^T B.$$

$$\begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & 1 \end{bmatrix} \begin{bmatrix} 4 & 0 \\ 0 & 2 \\ 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix}$$

wrong

$$\begin{bmatrix} 1 & 7 & 17 \\ 1 & 5 & 19 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 19 \\ 11 \end{bmatrix}$$

$$6x_1 + 7x_2 = 19$$

$$x_1 + 5x_2 = 11$$

$$x_1 = -5x_2 + 11$$

\therefore least square soln $\hat{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

2. Find a least square soln in $Ax = B$ where $A = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$ and $B = \begin{bmatrix} -3 \\ -1 \\ 0 \\ 2 \end{bmatrix}$

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -3 \\ -1 \\ 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

 8×4

The normal equations of $Ax = B$ is

$$A^T A \hat{x} = A^T B$$

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -3 \\ -1 \\ 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

wrong

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} -3 \\ -1 \\ 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

 8×4

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} -3 \\ -1 \\ 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

 8×4

3. Find a least square solution of $Ax = B$.
 where $A = \begin{bmatrix} 1 & -6 \\ 1 & -2 \end{bmatrix}$ & $B = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

$$\begin{bmatrix} 1 & -6 \\ -6 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} 4 & 0 \\ 0 & 90 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 8 \\ 33 \end{bmatrix}$$

$$x^2 = n/3^k \quad x = \sqrt{n/3^k} \quad k = \lfloor \log_3 n \rfloor$$

$$\text{Least Squares Vector } \hat{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 33 \\ w_{3D} \end{bmatrix}$$

next require $\frac{1}{\sin x} = \left(\frac{1}{x_1} \right) \left(\frac{1}{1/3D} \right)$

(π) (w_{3D})

卷之三

卷之三

A HISTORY OF THE AMERICAN PEOPLE

2000 JOURNAL OF CLIMATE

卷之三

卷之三

Fill here
Unit III
Ques part - Q

Saathī

Date _____ / _____ / _____

Date _____ / _____ / _____

4. Let $A = \begin{bmatrix} 1 & -3 & -3 \\ 1 & 5 & 1 \\ 1 & 7 & 2 \end{bmatrix}$ & $B = \begin{bmatrix} 5 \\ -3 \\ -5 \end{bmatrix}$

Find least square sol'n of $Ax = B$

$$A^T A x = A^T B$$

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 5 & 7 \\ -3 & 5 & 7 \end{bmatrix} \begin{bmatrix} 1 & -3 & -3 \\ 1 & 5 & 1 \\ 1 & 7 & 2 \end{bmatrix} = \begin{bmatrix} 5 \\ -3 \\ -5 \end{bmatrix}$$

$$\begin{bmatrix} 3 & 5 & 7 \\ 3 & 5 & 7 \\ -3 & 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -3 \\ -32 \\ -32 \end{bmatrix}$$

$$R_2 \rightarrow R_2 - R_1$$

$$R_3 \rightarrow R_3 - 2R_1$$

$$\begin{bmatrix} 3 & 9 & 0 & -3 \\ 3 & 4 & 16 & -35 \\ 6 & 58 & 20 & -46 \end{bmatrix}$$

$$x_1 = 2 + \frac{3}{2}x_3$$

Thus the least square solution is

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 + \frac{3}{2}x_3 \\ -1 - \frac{1}{2}x_3 \\ x_3 \end{bmatrix}$$

$$x_3 = 0$$

$$\text{the L.S. sol'n is } x = \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}$$

x_1 & x_2 are basic variable.
 x_3 is free variable.

General solution is

$$3x_1 + 9x_2 = -32$$

$$3x_1 - 9 - 9x_3 = -3$$

$$x_2 = -1 - \frac{1}{2}x_3$$

$$3x_1 + 9\left(-1 - \frac{1}{2}x_3\right) = -3$$

$$3x_1 - 9 - \frac{9}{2}x_3 = -3$$

$$x_2 = 6 + \frac{9}{2}x_3$$

$$\begin{bmatrix} 3 & 9 & 0 & -3 \\ 3 & 4 & 16 & -35 \\ 6 & 58 & 20 & -46 \end{bmatrix}$$

Alternative method :
square soln of $Ax = B$

If A is $m \times n$ matrix with $n < m$,
column we can factorize A as $A = QR$
each B in \mathbb{R}^m the system
 $Ax = B$ has unique least square soln

$$\hat{x} = R^{-1}Q^T B$$

Apply Gram-Schmidt process,

$$A = \begin{bmatrix} 1 & 2 & 5 \\ 1 & 1 & 0 \\ 1 & 1 & 2 \\ 1 & 3 & 3 \end{bmatrix}, B = \begin{bmatrix} 3 \\ 5 \\ 4 \\ -3 \end{bmatrix}$$

$$v_1 = x_1 = \frac{1}{\sqrt{1+4+25}} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, v_2 = x_2 - \frac{(x_2 \cdot v_1)}{(v_1 \cdot v_1)} v_1 = \begin{bmatrix} 3 \\ 1 \\ -8 \\ 4 \end{bmatrix} - \frac{3}{4} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 3 \\ 1 \\ -2 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 1 \\ 1 \end{bmatrix}$$

$$v_3 = x_3 - \left(\frac{x_3 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{x_3 \cdot v_2}{v_2 \cdot v_2} \right) v_2$$

$$= \begin{bmatrix} 3 \\ 5 \\ 0 \\ 2 \end{bmatrix} - \left(\frac{10}{4} \right) \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} - \left(\frac{6}{4} \right) \begin{bmatrix} 1 \\ -1 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 5 \\ 0 \\ 2 \\ 3 \end{bmatrix} - \frac{5}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \frac{3}{2} \begin{bmatrix} 1 \\ -1 \\ 1 \\ 1 \end{bmatrix}$$

No, making $v_1, v_2, v_3 \rightarrow A$

$$u_1 = \frac{v_1}{\|v_1\|} = \frac{(1, 1, 1, 1)}{\sqrt{1^2 + 1^2 + 1^2 + 1^2}} = \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)$$

$$u_2 = \frac{v_2}{\|v_2\|} = \frac{(1, -1, -1, 1)}{\sqrt{1^2 + (-1)^2 + (-1)^2 + 1^2}} = \left(\frac{1}{2}, -\frac{1}{2}, -\frac{1}{2}, \frac{1}{2}\right)$$

$$u_3 = \frac{v_3}{\|v_3\|} = \frac{(1, -1, 1, -1)}{\sqrt{1^2 + (-1)^2 + 1^2 + (-1)^2}} = \left(\frac{1}{2}, -\frac{1}{2}, \frac{1}{2}, -\frac{1}{2}\right)$$

Least square soln of $AX = B$ is
i.e. $GRX = B$ is obtained by

$$R\hat{X} = Q^T B$$

$$Q = [u_1, u_2, u_3]$$

$$Q = \begin{bmatrix} v_1 & v_2 & v_3 \\ v_2 & -v_1 & -v_3 \\ v_3 & v_1 & -v_2 \end{bmatrix}$$

Now becomes

$$Q^T B = \begin{bmatrix} v_1 & v_2 & v_3 \\ v_2 & -v_1 & -v_3 \\ v_3 & v_1 & -v_2 \end{bmatrix} \begin{bmatrix} -3 \\ 5 \\ -3 \end{bmatrix} = \begin{bmatrix} 6 \\ -6 \\ 4 \end{bmatrix}$$

Since $A = QR$ we have $R = Q^{-1}A$

i.e. $R = Q^T A$ ($\because Q$ is orthogonal) $Q^{-1} = Q^T$

$$Q R = \begin{bmatrix} v_1 & v_2 & v_3 \\ v_2 & -v_1 & -v_3 \\ v_3 & v_1 & -v_2 \end{bmatrix} \begin{bmatrix} 1 & 3 & 5 \end{bmatrix}$$

$$2\hat{x}_3 = 4$$

$$\hat{x}_3 = 2$$

$$2\hat{x}_2 + 3\hat{x}_3 = -2$$

$$2\hat{x}_2 + 6 = -6$$

$$2\hat{x}_1 + 4\hat{x}_2 + 5\hat{x}_3 = 6$$

$$2\hat{x}_1 - 24 + 10 = 6$$

$$\hat{x}_1 = 10$$

Singular Value Decomposition (SVD)

Unfortunately Not all matrix can be diagonalizable or factorized as $A = PDP^{-1}$

However a factorization $A = QDP^{-1}$ is possible for any $m \times n$ matrix A . This special factorization is called singular value decomposition of A and this factorization is most useful in applied linear algebra. SVD is widely used in signal processing, noise deduction and image compressing are sum of the applications of SVD. In Data Science it helps to reduce the dimension of the data.

Singular Value of a Matrix :

Let A be an $m \times n$ matrix then $A^T A$ is symmetric and its eigen values are non negative, also it can be orthogonally diagonalized, the singular values of A (are the square roots of eigen values of $A^T A$).

$$\hat{x} = \begin{bmatrix} 10 \\ -6 \\ 2 \end{bmatrix}$$

i.e. Singular values of A are

$$\sigma_1 = \sqrt{\lambda_1} = \sqrt{360} = 6\sqrt{10}$$

$$\sigma_2 = \sqrt{\lambda_2} = \sqrt{90} = 3\sqrt{10}$$

$$\sigma_3 = \sqrt{\lambda_3} = \sqrt{0} = 0$$

These singular values are denoted by $\sigma_1, \sigma_2, \dots, \sigma_n$ and they are arranged in decreasing order.

$$\sigma_i = \sqrt{\lambda_i} \quad 1 \leq i \leq n$$

1. Obtain the singular value of a matrix

$$A = \begin{bmatrix} 4 & 11 & 14 \\ 8 & 7 & -2 \end{bmatrix}$$

$$\text{Solv: } A^T A = \begin{bmatrix} 4 & 8 \\ 11 & 7 \\ 14 & -2 \end{bmatrix} \begin{bmatrix} 4 & 11 & 14 \\ 8 & 7 & -2 \end{bmatrix}$$

Let A be an $m \times n$ matrix with a rank r , then there exists a $m \times n$ matrix $\Sigma = \begin{bmatrix} D & 0 \\ 0 & 0 \end{bmatrix}$ with the diagonal entries in D and which are singular values of matrix A .

$$= \begin{bmatrix} 80 & 100 & 40 \\ 100 & 170 & 140 \\ 40 & 140 & 200 \end{bmatrix}$$

$$\text{i.e. } \sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$$

there exists an $m \times m$ orthonormal

matrix V and an $n \times n$ orthonormal

matrix U , such that $A = V \Sigma V^T$

Here the matrices V and $-V$ are not uniquely determined by A .

$$\lambda^3 - 450\lambda^2 + (14400 + 14400 + 3600)\lambda - 0 = 0$$

$$\lambda = 360, 90, 0$$

$$\lambda_1, \lambda_2, \lambda_3$$

To find matrix V , we find the eigen vectors v_1, v_2 corresponding to eigen values $\lambda_1 = 45$ i.e.

$$\text{S.tn: } A^T A = \begin{bmatrix} 3 & 4 \\ 0 & 5 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 4 & 5 \end{bmatrix} = \begin{bmatrix} 25 & 20 \\ 20 & 25 \end{bmatrix}$$

$$(A^T A - \lambda_1 I) x = \lambda_1 x$$

characteristic eqn of $A^T A =$

$$\lambda^2 - \text{tr}(A^T A)\lambda + \det(A^T A) = 0$$

always

$$\lambda_1 = 45, \lambda_2 = 5$$

are eigenvalues
↓ order

\therefore singular values of A are $\sigma_1 = \sqrt{\lambda_1} = \sqrt{45}$

$$\sigma_2 = \sqrt{\lambda_2} = \sqrt{5}$$

~~equation form~~

$$\begin{bmatrix} -20 & 20 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-20x_1 + 20x_2 = 0$$

$$x_1 = x_2$$

$$\frac{x_1}{\|x_1\|} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \text{ unit eigen vector is } v_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$(A^T A - \lambda_2 I)x_2 = 0$$

$$\begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\therefore \begin{bmatrix} 0 & 0 \\ 0 & 5 \end{bmatrix} \text{ as singl of } A \text{ is } 2 \times 2$$

$$\begin{pmatrix} \begin{bmatrix} 25 & 20 \\ 20 & 25 \end{bmatrix} - \begin{bmatrix} 5 & 0 \\ 0 & 5 \end{bmatrix} \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\begin{bmatrix} 20 & 20 \\ 20 & 20 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

Date / /

Saathi

$$20x_1 + 20x_2 = 0$$

$$x_1 = -x_2$$

$$\therefore \text{eigen vectors } x_2 = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -x_2 \\ x_2 \end{bmatrix} =$$

$$\text{say } x_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \text{ with } x_2 = 1$$

$$\therefore \text{unit eigen vector } v_2 = \begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

$$\therefore V = [v_1 \ v_2]$$

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

To find U :

$$u_1 = \frac{Av_1}{\|Av_1\|}$$

$$Av_1 = \begin{bmatrix} 3 & 0 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} \sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 3\sqrt{2} \\ \sqrt{2} \\ 9/\sqrt{2} \end{bmatrix}$$

$$= \frac{(3\sqrt{2}, 9/\sqrt{2})}{\sqrt{(3\sqrt{2})^2 + (9/\sqrt{2})^2}} = \frac{3\sqrt{2}, 9/\sqrt{2}}{3\sqrt{5}}$$

$$= \left(\frac{3}{\sqrt{2}}, \frac{9}{3\sqrt{5}} \right)$$

Date / /

Saathi

$$\therefore A = \left(\frac{1}{\sqrt{10}}, \frac{3}{\sqrt{10}} \right)$$

$$U_2 = AV_2 \Rightarrow AV_2 = \frac{3}{\sqrt{10}} \begin{bmatrix} 1 & 0 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} -3/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

$$= \frac{-3/\sqrt{2}, 1/\sqrt{2}}{\sqrt{(-3/\sqrt{2})^2 + (1/\sqrt{2})^2}} = \left(\frac{-3}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$

$$= \left(\frac{-3\sqrt{2} + 1}{\sqrt{10}}, \frac{1}{\sqrt{10}} \right)$$

$$\therefore U_2 = [u_1 \ u_2] = \begin{bmatrix} \sqrt{10} & -3/\sqrt{10} \\ 3/\sqrt{10} & 1/\sqrt{10} \end{bmatrix}$$

$$\text{SVD of } A = U \Sigma V^T$$

$$= \begin{bmatrix} \sqrt{10} & -3/\sqrt{10} \\ 3/\sqrt{10} & 1/\sqrt{10} \end{bmatrix} \begin{bmatrix} \sqrt{5} & 0 \\ 0 & \sqrt{5} \end{bmatrix} \begin{bmatrix} \sqrt{2} & \sqrt{2} \\ -1/\sqrt{2} & \sqrt{2} \end{bmatrix}$$

Note: To find U , alternate method is find the unit eigen vectors of AA^T

Principal Component Analysis (PCA):

PCA is a technique to analyse multi-variate data, it can be applied to any data that consists lists of measurements (attribute) made on a collection of objects or individuals.

Ex: consider a two dimensional data that represents weights and heights of n students

Let x_j denote the weight and height of j^{th} student.

This data can be represented by a $2 \times n$ matrix where the columns are x_1, x_2, \dots, x_n

$$\begin{bmatrix} w_1 & w_2 & \dots & w_n \\ h_1 & h_2 & \dots & h_n \end{bmatrix}$$

This data can be visualized as a 2D scatter plot.

In a similar way we can think of higher dimensional data which is difficult to visualize.

Mean & Covariance:

Let x_1, x_2, \dots, x_n represent p -dimensional data represented by a $p \times n$ matrix. The sample mean M of these observations x_1, x_2, \dots, x_n is given by

$$M = \frac{1}{n}(x_1 + x_2 + \dots + x_n)$$

for the above figure sample mean is the point in the centre of the scatter plot.

Let $\hat{x}_k = x_k - M$ for, $k = 1, 2, \dots, n$

The matrix $B = [\hat{x}_1, \hat{x}_2, \dots, \hat{x}_n]$ is known as mean deviation form of the data.

This has zero sample mean.

Date _____

Saathi

The sample covariance matrix is the $p \times p$ defined by $S = \frac{1}{N-1} \sum_{i=1}^N (x_i - M)(x_i - M)^T$

$$\text{or } S = \frac{1}{N-1} \sum_{i=1}^N (x_i^1)^2 + (x_i^2)^2 + \dots + (x_i^p)^2$$

1. Compute the sample mean and the covariance matrix of 3 measurement/attributes made on each of 4 individuals in a random sample given by $X_1 = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$, $X_2 = \begin{bmatrix} 4 \\ 2 \\ 3 \end{bmatrix}$,

$$X_3 = \begin{bmatrix} 7 \\ 8 \\ 5 \end{bmatrix}, X_4 = \begin{bmatrix} 8 \\ 4 \\ 5 \end{bmatrix}$$

$$\text{Sample Mean } \bar{x} = \frac{1}{4} (x_1 + x_2 + x_3 + x_4)$$

$$\bar{x} = \frac{1}{4} \begin{bmatrix} 1+4+7+8 \\ 2+2+8+4 \\ 1+3+1+5 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \\ 5 \end{bmatrix}$$

Subtract the sample mean from x_1, x_2, x_3, x_4

$$\hat{x}_1 = x_1 - \bar{x} = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} - \begin{bmatrix} 5 \\ 4 \\ 5 \end{bmatrix} = \begin{bmatrix} -4 \\ -2 \\ -4 \end{bmatrix}$$

Date _____

Saathi

$$\hat{x}_2 = \begin{bmatrix} -1 \\ -2 \\ 8 \end{bmatrix}, \hat{x}_3 = \begin{bmatrix} 2 \\ 4 \\ -4 \end{bmatrix}, \hat{x}_4 = \begin{bmatrix} -3 \\ 0 \\ 0 \end{bmatrix}$$

matrix of mean deviation $B = [\hat{x}_1, \hat{x}_2, \hat{x}_3, \hat{x}_4]$

$$B = \begin{bmatrix} -4 & -1 & 2 & 3 \\ -2 & -2 & 4 & 0 \\ -4 & 8 & -4 & 0 \end{bmatrix}$$

Covariance Matrix, $S = \frac{1}{N-1} B B^T$

$$S = \frac{1}{3} \begin{bmatrix} -4 & -1 & 2 & 3 \\ -2 & -2 & 4 & 0 \\ -4 & 8 & -4 & 0 \end{bmatrix} \begin{bmatrix} -4 & -2 & -4 \\ -1 & -2 & 8 \\ 2 & 4 & -4 \\ 3 & 0 & 0 \end{bmatrix}$$

$$= \frac{1}{3} \begin{bmatrix} 30 & 18 & 0 \\ 18 & 24 & -24 \\ 0 & -24 & 96 \end{bmatrix}$$

$$= \begin{bmatrix} 10 & 6 & 0 \\ 6 & 8 & -8 \\ 0 & -8 & 32 \end{bmatrix}$$

The diagonal entry s_{ii} in the covariance matrix is called the variance of the attribute x_i .

$$\text{Total Variance} = s_{11} + s_{22} + \dots + s_{pp} = \text{trace}(S)$$

The entry s_{ij} for $i \neq j$ in S is called covariance of x_i and x_j .

In the above problem covariance b/w x_1 & x_3 is 0 since the entry $s_{13} = 0$.
 x_1 & x_3 are uncorrelated.