

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

$$\begin{aligned} u \cdot v &= u_1 v_1 + u_2 v_2 + u_3 v_3 \\ &= 2(3) + (-5)(2) + (-1)(-3) \\ &= 6 - 10 + 3 \\ &= -1 \end{aligned}$$

$$\begin{aligned} v \cdot u &= v_1 u_1 + v_2 u_2 + v_3 u_3 \\ &= 3(2) + (2)(-5) + (-3)(-1) \\ &= 6 - 10 + 3 \\ &= -1 \end{aligned}$$

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

- i) $u \cdot v = v \cdot u$
- 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 $\sqrt{v \cdot v} = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2}$
(norm)

For any scalar c $\|cv\| = |c| \|v\|$
 $c = -1$ $\|cv\| = \|-v\| = |-1| \|v\| = \|v\|$

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.

$$\hat{v} = \frac{\vec{v}}{\|\vec{v}\|}$$

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

$$u \cdot v = \|u\| \|v\| \cos \theta \quad (or) \quad (1)$$

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

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

Orthogonal Vectors:

Two vectors u & v in R^n are orthogonal to each other if $u \cdot v = 0$ (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 says $\{v_1, v_2, \dots, v_p\}$ in R^n is said to be an orthogonal set if $v_i \cdot v_j = 0$ (or) 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 .

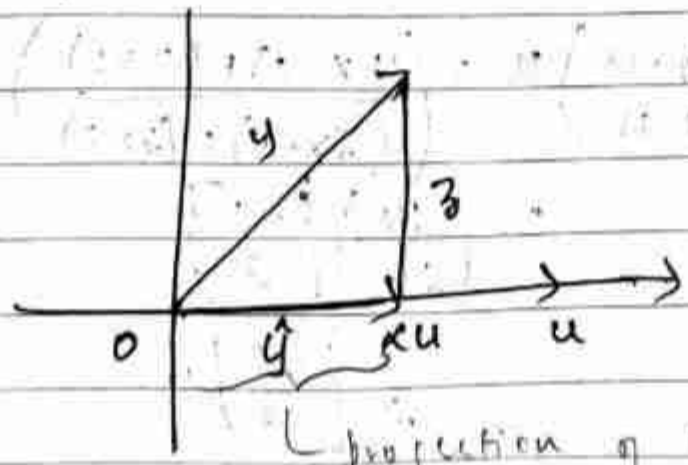
Date 03 / 07 / 2023

Orthogonal Basis for a subspace $W(R^n)$ is a basis for W that is also an orthogonal set.

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

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

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



projection of y on u

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

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

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

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

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

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

Soln:

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

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

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

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

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 linear independent.

Example:

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

$$v_1 = \begin{bmatrix} \frac{2}{\sqrt{11}} \\ \frac{1}{\sqrt{11}} \\ \frac{1}{\sqrt{11}} \end{bmatrix}, \quad v_2 = \begin{bmatrix} -\frac{1}{\sqrt{6}} \\ \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \end{bmatrix}, \quad v_3 = \begin{bmatrix} -\frac{1}{\sqrt{66}} \\ -\frac{4}{\sqrt{66}} \\ \frac{5}{\sqrt{66}} \end{bmatrix}$$

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

$$= 0$$

$$v_2 \cdot v_3 = \left(\frac{1}{\sqrt{66}} + \frac{-4}{\sqrt{66}} + \frac{5}{\sqrt{66}} \right)$$

$$= 0$$

$$v_1 \cdot v_3 = \left(-\frac{2}{\sqrt{11 \times 66}} - \frac{4}{\sqrt{11 \times 66}} + \frac{5}{\sqrt{11 \times 66}} \right)$$

$$= 0$$

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

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

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

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

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

$$\|v_1\| = 1$$

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

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

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

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

$$= \sqrt{\frac{66}{6}} = 1$$

\therefore The given is orthonormal basis of R^3

Gram-Schmidt's orthogonalization process

Given a basis $\{x_1, x_2, \dots, x_p\}$ for a subspace $W(R^n)$, the following algorithm namely gram-schmidt's algorithm forms a orthogonal basis for W namely $\{v_1, v_2, \dots, v_p\}$ where $v_1 = x_1, v_2 = x_2 - \left(\frac{x_2 \cdot v_1}{v_1 \cdot v_1}\right)v_1$

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

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

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

Example 2.3

Note: Normalizing each of the vectors v_1, v_2, \dots, v_p , we get an orthogonal basis $\{u_1, u_2, \dots, u_p\}$ where $u_i = \frac{v_i}{\|v_i\|}, 1 \leq i \leq p$.

Ques 1. Given a basis $\{x_1, x_2, x_3\}$ where

$$x_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \quad x_2 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad x_3 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} \text{ of a}$$

subspace W , construct an orthonormal basis.

Soln. Using Gram-Schmidt's orthogonal process,

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

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

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

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

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

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

$$v_3 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

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

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

To obtain orthonormal basis we have to normalize each of v_1, v_2, v_3

$$\text{Now } u_1 = \frac{v_1}{\|v_1\|} = \frac{(1, 0, 1)}{\sqrt{1^2 + 0^2 + 1^2}} = \left(\frac{1}{\sqrt{2}}, \frac{0}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$

$$u_2 = \frac{v_2}{\|v_2\|} = \frac{(\frac{1}{2}, 0, -\frac{1}{2})}{\sqrt{(\frac{1}{2})^2 + 0^2 + (-\frac{1}{2})^2}} = \frac{(\frac{1}{2}, 0, -\frac{1}{2})}{\frac{\sqrt{2}}{2}} = \left(\frac{1}{2} \cdot \frac{2}{\sqrt{2}}, 0, -\frac{1}{2} \cdot \frac{2}{\sqrt{2}} \right) = \left(\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}} \right)$$

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

$$= \frac{1}{\sqrt{2}}$$

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

∴ Orthonormal basis for W is $\{u_1, u_2, u_3\}$

$$\text{is } \left\{ \begin{bmatrix} 1/3 \\ 0 \\ 1/3 \end{bmatrix}, \begin{bmatrix} 1/3 \\ 0 \\ -1/3 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$$

Ans 2.

Construct an orthonormal basis for a subspace W of \mathbb{R}^3 where $\{x_1, x_2, x_3\}$

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

$$\text{basis for } W.$$

Ans 1.

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

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

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

$$= \begin{bmatrix} -3/4 \\ 1/4 \\ 1/4 \end{bmatrix}$$

On scaling v_2 , we get $(v_2 x_2)$

$$v_2' = \begin{bmatrix} -3 \\ 1 \\ 1 \end{bmatrix}$$

(scaling is applied only to simplify the calculation multiply vector by suitable no. so

all components

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

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

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

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

∴ $\{v_1, v_2', v_3\}$ are orthogonal basis for W

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

$$= \left(\frac{1}{\sqrt{5}}, \frac{1}{\sqrt{5}}, \frac{1}{\sqrt{5}}, \frac{1}{\sqrt{5}}, \frac{1}{\sqrt{5}}\right)$$

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

$$= \left(\frac{-3}{\sqrt{7}}, \frac{1}{\sqrt{7}}, \frac{1}{\sqrt{7}}, \frac{1}{\sqrt{7}}, \frac{1}{\sqrt{7}}\right)$$

$$= \frac{v_3}{\|v_3\|}$$

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

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

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

Then $\{u_1, u_2, u_3\}$ form an orthonormal basis.

CO-V

Quadratic form: A Quadratic form on \mathbb{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 \mathbb{R}^n .

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

Examples:

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

In example 3 the term $2xy$ is called cross product term.

1. Let $x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$, compute $x^T A x$ for the

following matrices.

- i) $A = \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix}$
- ii) $A = \begin{bmatrix} 3 & -2 \\ -2 & 7 \end{bmatrix}$

$$\text{Sol}^n \Rightarrow A = \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix}$$

$$x^T A x$$

$$\Rightarrow Ax =$$

$$\begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$=$$

$$\begin{bmatrix} 4x_1 & 0 \\ 0 & 3x_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$=$$

$$\begin{bmatrix} 4x_1^2 & 0 \\ 0 & 3x_2^2 \end{bmatrix}$$

$$\Rightarrow$$

$$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 4 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 4x_1 & 0 \\ 0 & 3x_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{4x_1}{3x_2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 4x_1 \\ 3x_2 \end{bmatrix}$$

$$= \begin{bmatrix} 4x_1^2 \\ 3x_2^2 \end{bmatrix}$$

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

$\therefore A$ is called matrix of Quadratic form

$$\text{ii) } A =$$

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

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

$$= \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 3x_1 & -2x_2 \\ -2x_1 & 7x_2 \end{bmatrix}$$

$$= x_1(3x_1 - 2x_2) + x_2(-2x_1 + 7x_2)$$

$$= 3x_1^2 - 2x_1x_2 - 2x_1x_2 + 7x_2^2$$

$$= 3x_1^2 - 4x_1x_2 + 7x_2^2$$

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

$$14/04/23$$

$$2.$$

For $x \in \mathbb{R}^3$, let $Q(x) = 5x_1^2 + 3x_2^2 + 2x_3^2 - x_1x_2 + 8x_2x_3$, write the quadratic form as $x^T A x$

$$\text{Sol}^n:$$

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

$$x_1^2$$

$$= 8/5$$

$$= 8/5$$

$$\text{Matrix of } Q(x) \text{ is } A = \begin{bmatrix} 5 & -1/2 & 0 \\ -1/2 & 3 & 4 \\ 0 & 4 & 2 \end{bmatrix}$$

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

$$= 8/5$$

$$= 8/5$$

$$= 8/5$$

$$= 8/5$$

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

$$= \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} 5x_1 - \frac{1}{2}x_2 \\ -\frac{1}{2}x_1 + 3x_2 + 4x_3 \\ 4x_2 + 2x_3 \end{bmatrix}$$

Classification (Nature of Quadratic form)

A Quadratic form $Q(x)$ is

- i) Positive definite if $Q(x) > 0$ for all $x \neq 0$
- ii) Negative definite if $Q(x) < 0$ for all $x \neq 0$
- iii) Indefinite if $Q(x) \neq 0$ assumes both +ve and -ve values
- iv) Positive semidefinite if $Q(x) \geq 0$ for all x
- v) Negative semidefinite if $Q(x) \leq 0$ for all x

Classification in terms of Eigen Values:

A Quadratic form $Q(x)$ is

- i) Positive definite if all its eigen values are +ve
- ii) Negative definite if all its eigen values are -ve
- iii) Indefinite if some eigen values are positive and some are negative
- iv) Positive semidefinite if its eigen values are ≥ 0 and at least one is 0
- v) Negative semidefinite if its eigen values are ≤ 0 and at least one is 0

1. Find the nature of following quadratic forms

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

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

$$\lambda^3 - (\text{tr}(A))\lambda^2 + (\text{sum of cofactors})\lambda - \det(A) = 0$$

(diagonal elements)

$$\lambda^3 - (1+5+1)\lambda^2 + (4-8+4)\lambda - (-36) = 0$$

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

$$\lambda_1 = -2, \lambda_2 = 6, \lambda_3 = 6$$

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

ii) $Q(x) = 3x^2 + 5y^2 + 3z^2 - 2yz + 2zx - 2xy$

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

$$\lambda^3 - (\text{tr}(A))\lambda^2 + (\text{sum of cofactors})\lambda - \det(A) = 0$$

(diagonal elements)

$$\lambda^3 - 11\lambda^2 + (14+8+14)\lambda - 36 = 0$$

$$\lambda^3 - 11\lambda^2 + 36\lambda - 36 = 0$$

$$\lambda = 6, 3, 2$$

\therefore Eigen values are 6, 3, 2
 $Q(x)$ is positive definite, since A has
 +ve eigen values.

iii) $Q(x) = 3x^2 + 2x^2 + x_3^2 + 4x_1x_2 + 4x_2x_3$ positive definite.

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

$$\lambda^3 - (\text{tr}(A))\lambda^2 + (\text{sum of cofactors})\lambda - \det(A) = 0$$

(Diagonal elements)

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

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

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

\therefore Eigen values are -1, 5, 2
 $Q(x)$ is indefinite since A has +ve & -ve
 eigen values.

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

Q1: If the quadratic form $Q(x)$ where x is a variable vector in R^n to write it in the canonical form (only the sum of square terms) we make change of variable in the equation

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

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 in quadratic form $Q(x)$ we get

$$Q(x) = x^T A x = (Py)^T A Py = y^T P^T A P y$$

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

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

If P orthogonally diagonalises A then $P^T = P^{-1}$

\therefore 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 g_A
Hence eigen values are the diagonal elements of D

Problems

1. Make the change of variable in the quadratic form $Q(x) = x_1^2 - 8x_1x_2 - 5x_2^2$ so that the new quadratic form is canonical form. (no cross product terms only square terms)

$$A = \begin{bmatrix} 1 & -4 \\ -4 & -5 \end{bmatrix}$$

Soln: The matrix of the given quadratic form is

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

$$\lambda^2 - (-4)\lambda + (-21) = 0$$

$$\lambda^2 + 4\lambda - 21 = 0$$

$$\lambda = 3, -7$$

\therefore eigen values are 3 & -7.

Date

Let $x = Py$ or $y = P^{-1}x$.
 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.

$$Q(y) = 3y_1^2 - 7y_2^2$$

2. Reduce the quadratic form $3x^2 + 5y^2 + 3z^2 + 2yz + 2zx - 22y$ to the canonical form by changing variables.

Soln. Here $Q(x) = 3x^2 + 5y^2 + 3z^2 + 2yz + 2zx - 22y$
 where $x = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$ let $x = Py$ where

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

P contains columns of the eigen vector
 and y is new vector
 $P^{-1}x = y$

$$\text{Let } y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

Date

$$\lambda^3 - (\text{tr } A)\lambda^2 + (\text{sum of diagonal elements of } A)\lambda - \det(A) = 0$$

$$\lambda^3 - 11\lambda^2 + (14 + 8 + 14)\lambda - 36 = 0$$

$$\lambda^3 - 11\lambda^2 + 36\lambda - 36 = 0$$

$$\lambda = 6, 3, 2$$

eigen values are 6, 3, 2. D should be in T order

$$D = \begin{bmatrix} 6 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

New quadratic form $Q(y) = 6y_1^2 + 3y_2^2 + 2y_3^2$

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

Soln. Given $x^T x = 1$
 Then $x^T x = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 1$

Given $x_1^2 + x_2^2 + x_3^2 = 1$

look for max coefficient.

Saathie

Date

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

$$Q(x) = 9x_1^2 + 4x_2^2 + 3x_3^2$$

$$\leq 9x_1^2 + 9x_1^2 + 9x_1^2$$

$$= 9(9x_1^2 + x_2^2 + x_3^2)$$

$$= 9(1)$$

$$= 9$$

$$\therefore Q(x) \leq 9$$

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

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

$$Q(x) = 9x_1^2 + 4x_2^2 + 3x_3^2$$

$$\geq 3x_1^2 + 3x_2^2 + 3x_3^2$$

$$= 3(x_1^2 + x_2^2 + x_3^2)$$

$$= 3(x) + 3x_1$$

$$= 3$$

$$Q(x) \geq 3$$

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

Date 18/03/2023

Saathie

Q. QR factorization of matrix:

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

$$A = QR \quad \text{where } Q \text{ is } m \times n$$

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.

Ques 1.

Find a QR factorization of $A =$

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

observe that all the columns of A are linearly independent

\therefore These columns form a basis for a subspace of \mathbb{R}^3 .

By applying Gram-Schmidt's process we get an orthonormal basis

$$\text{Let } x_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad x_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \quad x_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

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

$$v_2 = x_2 - \left(\frac{x_2 \cdot v_1}{x_1 \cdot v_1} \right) v_1$$

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

$$v_2 = \begin{bmatrix} -3/4 \\ 1/4 \\ 1/4 \end{bmatrix} \quad \begin{array}{l} \text{Multiply by 4} \\ \text{(Or scaling)} \end{array} = \begin{bmatrix} -3 \\ 1 \\ 1 \end{bmatrix} = v_2'$$

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

$$= \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} - \left(\frac{2}{4} \right) \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \left(\frac{2}{12} \right) \begin{bmatrix} -3 \\ 1 \\ 1 \end{bmatrix}$$

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

on scaling v_2' is multiply by 3.

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

Normalizing,

$$u_1 = \frac{v_1}{\|v_1\|} = \frac{(1, 1, 1)}{\sqrt{1^2+1^2+1^2}} = \left(\frac{1}{3}, \frac{1}{3}, \frac{1}{3} \right)$$

$$u_2 = \frac{v_2'}{\|v_2'\|} = \frac{(-3, 1, 1)}{\sqrt{(-3)^2+1^2+1^2}} = \left(-\frac{3}{\sqrt{11}}, \frac{1}{\sqrt{11}}, \frac{1}{\sqrt{11}} \right)$$

$$u_3 = \frac{v_3'}{\|v_3'\|} = \frac{(0, -2, 1)}{\sqrt{(0)^2+(-2)^2+1^2}} = \left(0, -\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}} \right)$$

$$\therefore Q = \begin{bmatrix} 1/3 & -3/3 & 0 \\ 1/3 & 1/3 & -2/3 \\ 1/3 & 1/3 & 1/3 \end{bmatrix} = [u_1 \ u_2 \ u_3]$$

Let $A = QR$ for some R
Pre multiply by Q^T

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

$\therefore R$ is orthogonal matrix $\therefore Q^T = Q^{-1}$

Date: / /

$$R = Q^T A =$$

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

$$P = \begin{bmatrix} 2 & 3/2 & 1/2 \\ 0 & \sqrt{3}/2 & 1/2 \\ 0 & 0 & 3/2 \end{bmatrix} \text{ Then } A = QR$$

$$\text{where } Q^T R \text{ are as above}$$

Verification:

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

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

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

Observe that all the columns are linearly independent. So, the columns form a basis for the subspace of \mathbb{R}^3 .

Date: / /

By applying Gram Schmidt process

$$x_1 = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \quad x_2 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad x_3 = \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix}$$

$$v_1 = x_1$$

$$v_2 = x_2 - \left(\frac{x_2 \cdot v_1}{v_1 \cdot v_1} \right) v_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} - \left(\frac{1}{2} \right) \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$$

$$v_2 = \begin{bmatrix} 1/2 \\ -1/2 \\ -1/2 \end{bmatrix} \text{ or scaling } v_2' = \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}$$

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

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

$$= \begin{bmatrix} -1/2 \\ 1/2 \\ 0 \end{bmatrix} \text{ or scaling } v_3' = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

$$u_1 = \frac{v_1}{\|v_1\|} = \frac{(1, 2, 1, 1)}{\sqrt{1^2 + 2^2 + 1^2 + 1^2}} = \left(\frac{1}{\sqrt{5}}, \frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}}, \frac{1}{\sqrt{5}}\right)$$

$$u_2 = \frac{v_2}{\|v_2\|} = \frac{(3, -1, 5, 4)}{\sqrt{3^2 + (-1)^2 + 5^2 + 4^2}} = \left(\frac{3}{\sqrt{55}}, -\frac{1}{\sqrt{55}}, \frac{5}{\sqrt{55}}, \frac{4}{\sqrt{55}}\right)$$

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

$$Q = \begin{bmatrix} \frac{1}{\sqrt{5}} & \frac{3}{\sqrt{55}} & -\frac{9}{\sqrt{210}} \\ \frac{2}{\sqrt{5}} & -\frac{1}{\sqrt{55}} & -\frac{2}{\sqrt{210}} \\ \frac{1}{\sqrt{5}} & \frac{5}{\sqrt{55}} & \frac{11}{\sqrt{210}} \\ \frac{1}{\sqrt{5}} & \frac{4}{\sqrt{55}} & \frac{2}{\sqrt{210}} \end{bmatrix}$$

Let $A = Q^T R$ per some R
 Multiply by Q^T

$$Q^T A = Q^T Q^T R = I R = \hat{R}$$

$$R = Q^T A = \begin{bmatrix} \frac{1}{\sqrt{5}} & \frac{1}{\sqrt{5}} & \frac{1}{\sqrt{5}} & \frac{1}{\sqrt{5}} \\ \frac{3}{\sqrt{55}} & -\frac{1}{\sqrt{55}} & \frac{5}{\sqrt{55}} & \frac{4}{\sqrt{55}} \\ -\frac{9}{\sqrt{210}} & -\frac{2}{\sqrt{210}} & \frac{11}{\sqrt{210}} & \frac{2}{\sqrt{210}} \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 2 & 1 & 2 \\ 1 & 1 & 4 \\ -1 & 0 & 1 \end{bmatrix}$$

Least Square Solution:

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

When a soln is demanded but none exists or the best 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 make $\|B-AX\|$ as small as possible.

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

If A is an $m \times n$ 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 $\|B-A\hat{x}\| \leq \sqrt{\|B-AX\|^2}$

The matrix equation $A^T A X = A^T B$ is called normal equation for $AX=B$

Solution of normal equations is the least square solution \hat{X} .

1. Find a least square solution of $AX=B$ where $A = \begin{bmatrix} 4 & 0 \\ 0 & 2 \\ 1 & 1 \end{bmatrix}$ and $B = \begin{bmatrix} 2 \\ 0 \\ 11 \end{bmatrix}$

$$3 \times 2$$

$$3 \times 1$$

Solⁿ: The normal equations of $AX=B$ is $A^TAX = A^TB$.

$$\begin{bmatrix} 4 & 0 \\ 0 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ 11 \end{bmatrix}$$

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

$$17x_1 + x_2 = 19$$

$$x_1 + 5x_2 = 11$$

$$x_1 = 5x_2 + 11$$

$$x_1 = 1, x_2 = 2$$

\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 of $AX=B$ where $A = \begin{bmatrix} 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \end{bmatrix}$ and $B = \begin{bmatrix} -3 \\ 0 \\ 2 \\ 5 \\ 1 \end{bmatrix}$

$$5 \times 4$$

$$5 \times 1$$

The normal equations of $AX=B$ is $A^TAX = A^TB$.

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -3 \\ -1 \\ 0 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -3 \\ -1 \\ 0 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -3 \\ -1 \\ 0 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -3 \\ -1 \\ 0 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -3 \\ -1 \\ 0 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} 6 & 2 & 2 & 2 & 2 \\ 2 & 0 & 0 & 0 & 0 \\ 2 & 0 & 2 & 0 & 0 \\ 2 & 0 & 0 & 0 & 2 \\ 2 & 0 & 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} 4 \\ -4 \\ 2 \\ 2 \\ 6 \end{bmatrix}$$

$$6x_1 + 2x_2 + 2x_3 + 2x_4 + 2x_5 = 4$$

$$2x_1 + 2x_2 = -4$$

$$2x_1 + 2x_3 = 2$$

$$2x_1 + 2x_4 = 6$$

$$\begin{bmatrix} 6 & 2 & 2 & 2 & 2 & 4 \\ 2 & 2 & 0 & 0 & 0 & -4 \\ 2 & 0 & 2 & 0 & 0 & 2 \\ 2 & 0 & 0 & 0 & 2 & 6 \end{bmatrix}$$

$$R_3 \leftrightarrow R_1$$

$$\begin{bmatrix} 2 & 0 & 0 & 2 & 2 & 6 \\ 2 & 2 & 0 & 0 & 0 & -4 \\ 2 & 0 & 2 & 0 & 0 & 2 \\ 2 & 2 & 2 & 2 & 2 & 4 \end{bmatrix}$$

$$R_1 \rightarrow R_1 - R_3$$

$$R_3 \rightarrow R_3 - R_1$$

$$R_2 \rightarrow R_2 - 3R_3$$

$$\begin{bmatrix} 2 & 0 & 0 & 2 & 2 & 6 \\ 0 & 2 & 0 & -2 & -10 & -4 \\ 0 & 0 & 2 & -2 & -4 & -4 \\ 0 & 2 & 2 & -4 & -14 & -4 \end{bmatrix}$$

$$R_4 \rightarrow R_4 - R_2$$

$$\begin{bmatrix} 2 & 0 & 0 & 2 & 2 & 6 \\ 0 & 2 & 0 & -2 & -10 & -4 \\ 0 & 0 & 2 & -2 & -4 & -4 \\ 0 & 0 & 2 & -2 & -4 & -4 \end{bmatrix}$$

$$R_4 \rightarrow R_4 - R_3$$

$$\begin{bmatrix} 2 & 0 & 0 & 2 & 2 & 6 \\ 0 & 2 & 0 & -2 & -10 & -4 \\ 0 & 0 & 2 & -2 & -4 & -4 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

x_1, x_2 & x_3 are basic variables

x_4 is free variable.

General soln in

$$2x_3 + 2x_4 = -4 \Rightarrow x_3 = -2 + x_4$$

$$+ 2x_2 - 2x_4 = -10 \Rightarrow x_2 = -5 + x_4$$

$$+ 2x_1 + 2x_4 = 6 \Rightarrow x_1 = 3 - x_4$$

Thus the least square solution is

$$\hat{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 3 - x_4 \\ -5 + x_4 \\ -2 + x_4 \\ x_4 \end{bmatrix}$$

where x_4 is free variable

If $x_4 = 0$, the l.s. soln is $\hat{x} = \begin{bmatrix} 3 \\ -5 \\ -2 \\ 0 \end{bmatrix}$

3. Find a least square solution of $AX=B$

where

$$A = \begin{bmatrix} 1 & -2 \\ 1 & -2 \\ 1 & 1 \\ 1 & 1 \end{bmatrix} \quad X = \begin{bmatrix} x \\ y \end{bmatrix} \quad B = \begin{bmatrix} -1 \\ 2 \\ 1 \\ 6 \end{bmatrix}$$

solⁿ. $A^T A X = A^T B$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ -2 & -2 & 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ -2 & -2 & 1 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \\ 1 \\ 6 \end{bmatrix}$$

$$\begin{bmatrix} A & B \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 2 \\ 38 \end{bmatrix}$$

$$AX = B$$

$$x = \frac{2}{11}$$

$$y = \frac{11}{20}$$

$$x = \frac{11}{20}$$

So least square solⁿ $X = \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \frac{2}{11} \\ \frac{11}{20} \end{bmatrix}$

+

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 solⁿ of $AX=B$

$A^T A X = A^T B$

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

$$\begin{bmatrix} 3 & 9 & 0 \\ 3 & 41 & 16 \\ 0 & 28 & 20 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -3 \\ -35 \\ -46 \end{bmatrix}$$

$$\begin{bmatrix} 3 & 9 & 0 & -3 \\ 3 & 41 & 16 & -35 \\ 0 & 28 & 20 & -46 \end{bmatrix} \begin{array}{l} R_2 \rightarrow R_2 - R_1 \\ R_3 \rightarrow R_3 - 2R_1 \end{array}$$

$$\begin{bmatrix} 3 & 9 & 0 & -3 \\ 0 & 22 & 16 & -32 \\ 0 & 40 & 20 & -40 \end{bmatrix} \begin{array}{l} \\ R_3 \rightarrow R_3 - 2R_2 \\ S_2 \end{array}$$

$$\begin{bmatrix} 3 & 9 & 0 & -3 \\ 0 & 22 & 16 & -32 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Date:

x_1 & x_2 are basic variables
 x_3 is free variable
General solution is

$$32x_1 + 16x_2 = -32$$

$$3x_1 + 4x_2 = -3$$

$$32x_1 = -32 - 16x_2$$

$$x_1 = \frac{-1 - \frac{1}{2}x_2}{2}$$

$$3x_1 + 4\left(\frac{-1 - \frac{1}{2}x_2}{2}\right) = -3$$

$$3x_1 - 2 - 2x_2 = -3$$

$$3x_1 = 2 - 2x_2$$

$$x_1 = \frac{2 - 2x_2}{3}$$

Then the least square solution is

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

$$X \in \mathbb{R}^3$$

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

Alternative method to find the least square soln of $AX=B$

If A is $m \times n$ matrix with k columns we can factorize A as $A=QR$ each R is in R^n the system $AX=B$ has unique least square soln

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

Alternative method:

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

Apply Gram-Schmidt process,

$$\begin{aligned} v_1 &= x_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} & n_1 &= x_2 - \frac{(x_2 \cdot v_1)}{(v_1 \cdot v_1)} v_1 = \begin{bmatrix} 3 \\ 1 \\ 1 \\ 1 \end{bmatrix} - \frac{2}{4} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} 3 \\ 1 \\ 1 \\ 1 \end{bmatrix} - 2 \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ -1 \\ -1 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} v_3 &= x_3 - \left(\frac{x_3 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{x_3 \cdot v_2}{v_2 \cdot v_2} \right) v_2 \\ &= \begin{bmatrix} 5 \\ 0 \\ 0 \\ 3 \end{bmatrix} - \left(\frac{10}{4} \right) \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} - \left(\frac{6}{4} \right) \begin{bmatrix} 1 \\ -1 \\ -1 \\ -1 \end{bmatrix} \\ &= \begin{bmatrix} 5 \\ 0 \\ 0 \\ 3 \end{bmatrix} - \frac{5}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} - \frac{3}{2} \begin{bmatrix} 1 \\ -1 \\ -1 \\ -1 \end{bmatrix} \\ &= \begin{bmatrix} 1 \\ -1 \\ -1 \\ -1 \end{bmatrix} \end{aligned}$$

Normalizing v_1, v_2, v_3 \downarrow

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

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

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

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

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

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

$$\text{As } R = Q^T A \quad (\because Q \text{ is orthogonal})$$

$$\text{matrix } Q^{-1} = Q^T$$

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

$$= \begin{bmatrix} 2 & 4 & 5 \\ 0 & 2 & 3 \\ 0 & 0 & 2 \end{bmatrix}$$

Thus $A = QR$ where Q and R are as above.

Least squares soln of $AX = B$ is \hat{x} . $QRX = B$ is obtained by

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

$$\text{i.e. } R\hat{x} = Q^T B \quad \text{--- (2)}$$

$$Q^T B = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 5 \\ -5 \end{bmatrix} = \begin{bmatrix} 6 \\ -6 \\ 4 \\ 4 \end{bmatrix}$$

③ now become

$$\begin{bmatrix} 2 & 4 & 5 \\ 0 & 2 & 3 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} \hat{x}_1 \\ \hat{x}_2 \\ \hat{x}_3 \end{bmatrix} = \begin{bmatrix} 6 \\ -6 \\ 4 \end{bmatrix}$$

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

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

$$2\hat{x}_2 + 3\hat{x}_3 = -6$$

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

$$2\hat{x}_2 = -12$$

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

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

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

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

∴ The least square solution

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

Singular Value Decomposition: (SVD)

Unfortunately Not all matrices 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 Values 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 $A^T A$.

The 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 values of a matrix

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

Soln:

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

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

Characteristic equation of $A^T A$ is $\lambda^3 - \text{tr}(A^T A)\lambda^2 + (\text{sum of cofactors})\lambda - \det(A^T A) = 0$ (of diagonal of $A^T A$)

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

$$= \lambda^3 - 450\lambda^2 + 32400\lambda = 0$$

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

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

\therefore Singular values of A are

$$\begin{aligned} \sigma_1 &= \sqrt{\lambda_1} = \sqrt{360} = 6\sqrt{10} \\ \sigma_2 &= \sqrt{\lambda_2} = \sqrt{90} = 3\sqrt{10} \\ \sigma_3 &= \sqrt{\lambda_3} = \sqrt{0} = 0 \end{aligned}$$

SVD Theorem:

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

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

there exists an $m \times m$ an orthogonal matrix U , and an $n \times n$ orthogonal matrix V , such that $A = UV^T$. Here the matrices U and V are not uniquely determined by A .

1. Find the SVD of $A = \begin{bmatrix} 3 & 0 \\ 4 & 5 \end{bmatrix}$.

Soln: $A^T A = \begin{bmatrix} 3 & 4 \\ 0 & 5 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 4 & 5 \end{bmatrix} = \begin{bmatrix} 25 & 20 \\ 20 & 25 \end{bmatrix}$

Characteristic eqn of $A^T A =$

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

$$\lambda^2 - 50\lambda + 25 = 0$$

always write in order

$$\lambda_1 = 45, \lambda_2 = 5 \text{ are eigen values}$$

\therefore Singular values of A are $\frac{1}{\sqrt{\lambda_1}} = \sqrt{45}$
 $\frac{1}{\sqrt{\lambda_2}} = \sqrt{5}$

$$\Sigma = \begin{bmatrix} \sqrt{45} & 0 \\ 0 & \sqrt{5} \end{bmatrix}$$

$$D = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} \sqrt{45} & 0 \\ 0 & \sqrt{5} \end{bmatrix}$$

$$\therefore \Sigma = \begin{bmatrix} \sqrt{45} & 0 \\ 0 & \sqrt{5} \end{bmatrix} \text{ as size of } A \text{ is } 2 \times 2$$

Date

To find matrix V , we find the eigen vectors of $A^T A$ corresponding to eigen values.

$$\lambda_1 = 45 \text{ for}$$

$$(A^T A - \lambda_1 I)X = 0$$

$$\text{Soln of } (A^T A - \lambda_1 I)X = 0$$

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

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

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

$$x_1 = x_2$$

$$\therefore \text{eigen vector } X_1 = \begin{bmatrix} x_1 \\ x_1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\frac{x_1}{\|x_1\|} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\therefore \text{unit eigen vector is } v_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$(A^T A - \lambda_2 I)X_2 = 0$$

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

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

$$20x_1 + 20x_2 = 0$$

$$x_1 = -x_2$$

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

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

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

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

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

To find V :

$$u_1 = \frac{Av_1}{\|Av_1\|} \quad Av_1 = \begin{bmatrix} 3 & 0 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 3/\sqrt{2} \\ 7/\sqrt{2} \end{bmatrix}$$

$$= \frac{\left(\frac{3}{\sqrt{2}}, \frac{7}{\sqrt{2}} \right)}{\sqrt{\left(\frac{3}{\sqrt{2}} \right)^2 + \left(\frac{7}{\sqrt{2}} \right)^2}} = \frac{3/\sqrt{2}, 7/\sqrt{2}}{5/\sqrt{2}}$$

$$= \left(\frac{3}{5}, \frac{7}{5} \right)$$

Date: / /

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

$$u_2 = \frac{Av_2}{\|Av_2\|} \Rightarrow Av_2 = \begin{bmatrix} 3 & 0 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} -3/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

$$= \frac{-3/\sqrt{2}, 1/\sqrt{2}}{\sqrt{\left(\frac{-3}{\sqrt{2}} \right)^2 + \left(\frac{1}{\sqrt{2}} \right)^2}} = \left(\frac{-3}{\sqrt{5}}, \frac{1}{\sqrt{5}} \right)$$

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

$$V = [u_1 \ u_2]$$

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

SVD $A = V \Sigma V^T$

$$= \begin{bmatrix} 1/\sqrt{10} & -3/\sqrt{10} \\ 3/\sqrt{10} & 1/\sqrt{10} \end{bmatrix} \begin{bmatrix} \sqrt{45} & 0 \\ 0 & \sqrt{5} \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$$

Note: To find V , 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 list of measurements (attributes) made on a collection of objects or individuals.

Ex: consider a two dimensional data that represents weight and height of N^{th} student.

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 \bar{x} 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.



The sample covariance matrix is the $p \times p$ defined by $S = \frac{BB^T}{N-1}$

$$\text{or } S = \frac{B X_1^2 + X_2^2 + \dots + X_n^2}{N-1}$$

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

$$X_3 = \begin{bmatrix} 7 \\ 6 \\ 1 \end{bmatrix}, X_4 = \begin{bmatrix} 8 \\ 4 \\ 5 \end{bmatrix}$$

$$\text{Sample Mean } \bar{X} = \frac{X_1 + X_2 + X_3 + X_4}{4}$$

$$\bar{X} = \frac{1}{4} \begin{bmatrix} 1+4+7+8 \\ 2+2+6+4 \\ 1+1+1+5 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \\ 5 \end{bmatrix}$$

Subtract the sample mean from X_1, X_2, X_3, X_4

$$\hat{X}_1 = X_1 - \bar{X} = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} - \begin{bmatrix} 5 \\ 4 \\ 5 \end{bmatrix} = \begin{bmatrix} -4 \\ -2 \\ -4 \end{bmatrix}$$

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

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

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

$$\text{Covariance matrix, } S = \frac{BB^T}{N-1}$$

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

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

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

The diagonal entry S_{ii} in the covariance matrix S 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 $S_{13} = 0$.
i.e. x_1 & x_3 are uncorrelated.