

Unit 4

Orthogonalization, Eigenvalues and Eigenvectors

Orthogonal bases

A set of vectors q_1, q_2, \dots, q_n is called orthonormal if

$$q_i^T q_j = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$$

A matrix Φ that has ~~orthogonal~~ orthonormal columns will be written as Φ . For such a matrix, $\Phi^T \Phi = I$

i.e. Φ^T is a left inverse of Φ . In particular, if Φ is a square then Φ is called an orthogonal matrix.

In this case, $\Phi^T = \Phi^{-1}$

Example 1: $\Phi = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ dot product b/w col is 0
unit length of column.

Φ is orthogonal.

so that $\Phi^T \Phi = I$ $\Phi^T = \Phi^{-1}$

Example 2: $\Phi = \begin{bmatrix} c & -s \\ s & c \end{bmatrix}$ Φ is orthogonal
 $\Phi^T = \Phi^{-1}$, so that $\Phi^T \Phi = I$

Example 3: All permutation matrices are orthogonal

$P_{23} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$ is such that $P_{23}^T = P_{23}^{-1}$
so that $P_{23}^T P_{23} = I$

Example 4: $\Phi = \begin{bmatrix} 1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & -1/\sqrt{5} \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 1/\sqrt{5} & 2/\sqrt{5} & 0 \\ 2/\sqrt{5} & -1/\sqrt{5} & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

Φ^T is a left inverse of Φ .

* You can never have a g matrix of size $m \times n$ where $n > m$.

Advantages of g matrix:

(1) g preserves norm.

$$\begin{aligned}\text{Proof: } \|g\mathbf{x}\|^2 &= (g\mathbf{x})^T(g\mathbf{x}) \\ &= \mathbf{x}^T g^T g \mathbf{x} \\ &= \mathbf{x}^T I \mathbf{x} \\ &= \mathbf{x}^T \mathbf{x} \\ &= \|\mathbf{x}\|^2\end{aligned}$$

(2) g preserves angle b/w two vectors

$$\begin{aligned}\text{Proof: } (g\mathbf{x})^T(g\mathbf{y}) &= \mathbf{x}^T g^T g \mathbf{y} \\ &= \mathbf{x}^T \mathbf{y} \\ \therefore \cos \theta &= \frac{\mathbf{x}^T \mathbf{y}}{\|\mathbf{x}\| \|\mathbf{y}\|} = \frac{(g\mathbf{x})^T(g\mathbf{y})}{\|g\mathbf{x}\| \|g\mathbf{y}\|}\end{aligned}$$

(3) If v_1, v_2, \dots, v_n is a basis for a r/s V then any $b \in V$ is a linear combination of the v_i 's.

$$\therefore b = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$$

If q_1, q_2, \dots, q_n is an orthogonal basis for V . then

$$b = \alpha_1 q_1 + \alpha_2 q_2 + \dots + \alpha_n q_n.$$

$$q_1^T b = \alpha_1$$

$$q_2^T b = \alpha_2$$

$$q_n^T b = \alpha_n$$

\therefore The equation $b = \alpha_1 q_1 + \alpha_2 q_2 + \dots + \alpha_n q_n$

$$= \alpha_1 \begin{bmatrix} \uparrow \\ q_1 \\ \downarrow \end{bmatrix} + \alpha_2 \begin{bmatrix} \uparrow \\ q_2 \\ \downarrow \end{bmatrix} + \alpha_3 \begin{bmatrix} \uparrow \\ q_3 \\ \downarrow \end{bmatrix}$$

can be written as

$$\begin{bmatrix} \uparrow & \uparrow & \uparrow \\ q_1 & q_2 & q_3 \\ \downarrow & \downarrow & \downarrow \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} = g\mathbf{x} = b$$

i.e. $\mathbf{q} \mathbf{x} = \mathbf{b}$

The solution of the above system of eqn:

$$\mathbf{x} = \mathbf{q}^{-1} \mathbf{b} = \mathbf{q}^T \mathbf{b}$$

(4) $b = \alpha_1 q_1 + \alpha_2 q_2 + \dots + \alpha_n q_n$

where any α_i can be solved as $q_i^T b$.

$$b = (q_1^T b) q_1 + (q_2^T b) q_2 + \dots + (q_n^T b) q_n$$

$$= \frac{q_1^T b}{q_1^T q_1} q_1 + \frac{q_2^T b}{q_2^T q_2} q_2 + \dots + \frac{q_n^T b}{q_n^T q_n} q_n$$

i.e. b = projection of b onto line through q_1

+ projection of b onto line through q_2

+ projection of b onto line through q_n

$\therefore b$ = sum of projections of b onto ~~onto~~ q_i 's.

(5) If the columns of a square matrix \mathbf{q} are orthonormal, then its rows are automatically orthonormal.

$$\mathbf{q} = \begin{bmatrix} 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \\ 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \end{bmatrix}$$

(6) Consider a system of equation, $\mathbf{q} \mathbf{x} = \mathbf{b}$.

If \mathbf{q} is square, then $\mathbf{x} = \mathbf{q}^{-1} \mathbf{b}$. or $\mathbf{x} = \mathbf{q}^T \mathbf{b}$.

If \mathbf{q} is a matrix of size $m \times n$, with $m > n$. Then $\mathbf{q} \mathbf{x} = \mathbf{b}$ is solvable if and only if \mathbf{b} belongs to column space of \mathbf{q} $b \in C(\mathbf{q})$.

If not, we solve the system by the method of least squares.

The normal equation is: $\mathbf{q}^T \mathbf{q} \hat{\mathbf{x}} = \mathbf{q}^T \mathbf{b}$

$$\therefore \hat{\mathbf{x}} = \mathbf{q}^T \mathbf{b}$$

The point of projection is

$$p = g \hat{r}$$

$$p = gg^T b$$

$$gg^T \neq I$$

(Because g is a tall matrix)

it will never have a right inverse)

$$P = g(g^T g)^{-1} g^T$$

$$P = gg^T$$

Gram-Schmidt Process of orthogonalization

To construct an orthonormal set of vectors q_1, q_2, q_3 from a set of linearly independent vectors a, b, c

i) The first vector q_1 can go in any of the 3 directions (a, b, c) . Let us choose the direction of q_1 along a .

$$\therefore q_1 = \frac{\vec{a}}{\|a\|}$$

So that $\|q_1\| = 1$

ii) The second vector b is independent of a . If it is already orthogonal to a :
We can choose q_2 in this direction and write:

$$q_2 = \frac{b}{\|b\|}$$

so that $\|q_2\| = 1$

If b is not orthogonal to a :

b has components in the direction of q_1 and in \perp 's direction.

If we want q_2 to go in this \perp 's direction then, we need to subtract the component of b in the direction of q_1 from b .

Consider $B = b - \frac{(q_1^T b)}{(q_1^T q_1)} q_1$

Now, $q_2 = \frac{B}{\|B\|}$, so that $\|q_2\| = 1$

3) The third vector c is independent of both a and b . Hence, it is not in the plane of a and b . If it is already orthogonal, then we can choose q_3 in this direction and write $q_3 = \frac{c}{\|c\|}$, so that $\|q_3\| = 1$.

If c is not orthogonal to them:

If c has components in the direction of q_1, q_2 and in the L^\perp direction. If we want q_3 to go in this L^\perp direction, then we need to subtract the components of c in q_1 and q_2 directions from c .

Consider $C = c - \frac{(q_1^T c)}{(q_1^T q_1)} q_1 - \frac{(q_2^T c)}{(q_2^T q_2)} q_2$

Now, $q_3 = \frac{C}{\|C\|}$, so that $\|q_3\| = 1$

Proof:

$$\begin{aligned} q_1^T B &= q_1^T (b - \frac{(q_1^T b)}{(q_1^T q_1)} q_1) \\ &= q_1^T b - \frac{(q_1^T b)}{(q_1^T q_1)} q_1^T q_1 \\ &= q_1^T b - q_1^T b \\ &= 0 \end{aligned}$$

This implies that they are orthogonal. (B is orthogonal to q_1)

$$\begin{aligned}
 q_1^T c &= q_1^T (c - (q_1^T c)q_1 - (q_1^T c)q_2) \\
 &= q_1^T c - (q_1^T c)(q_1^T q_1) - 0 \\
 &= q_1^T c - q_1^T c \\
 &= 0
 \end{aligned}$$

This implies that q_1 is orthogonal to c .

Similarly, q_2 is orthogonal to c .

$A = QR$ factorisation

Given a set of linearly independent vectors a, b, c we construct a set of orthonormal vectors q_1, q_2, q_3 using the GS process. If A is a matrix whose columns are a, b, c and Q is the matrix whose col are q_1, q_2, q_3 we now find a relation b/w A & Q . To do this we express a, b, c as linear combination of q_1, q_2, q_3 .

$$\begin{aligned}
 a &= \text{projection of } a \text{ onto the line through } q_1 \\
 &= (q_1^T a) q_1
 \end{aligned}$$

$$\begin{aligned}
 b &= \text{sum of projection of } b \text{ onto lines through } q_1 \text{ and } q_2 \\
 &= (q_1^T b) q_1 + (q_2^T b) q_2
 \end{aligned}$$

$$\begin{aligned}
 c &= \text{sum of projection of } c \text{ onto lines through } q_1, q_2 \text{ and } q_3 \\
 &= (q_1^T c) q_1 + (q_2^T c) q_2 + (q_3^T c) q_3
 \end{aligned}$$

$$\therefore A = \begin{bmatrix} a & b & c \end{bmatrix} = \begin{bmatrix} q_1 & q_2 & q_3 \end{bmatrix} \begin{bmatrix} q_1^T a & q_1^T b & q_1^T c \\ q_2^T a & q_2^T b & q_2^T c \\ q_3^T a & q_3^T b & q_3^T c \end{bmatrix} \quad \boxed{A = QR}$$

A and Q will be equivalent (if A is \square Q is also \square , if A is \square Q is also \square)

R is always square matrix. (and upper triangular in nature)

When the system $Ax=b$ is inconsistent we solve it by the method of least squares. The normal equations are: $A^T A \hat{x} = A^T b$

$$(QR)^T QR \hat{x} = (QR)^T b$$

$$R^T Q^T Q R \hat{x} = R^T Q^T b$$

$$R^T R \hat{x} = R^T Q^T b$$

$$R \hat{x} = Q^T b \quad \text{this is of the form: } Ux = c$$

The above system can be solved by back substitution, since R is upper triangular

Q1: Find a 3rd column so that the matrix $Q = \begin{bmatrix} 1/\sqrt{3} & 1/\sqrt{4} & x \\ 1/\sqrt{3} & 2/\sqrt{4} & y \\ 1/\sqrt{3} & -3/\sqrt{4} & z \end{bmatrix}$ is orthogonal.

$$1/\sqrt{3}x + 1/\sqrt{3}y + 1/\sqrt{3}z = 0$$

$$x^2 + y^2 + z^2 = 1$$

$$x+y+z=0 \quad 1/\sqrt{4}x + 2/\sqrt{4}y + -3/\sqrt{4}z = 0$$

$$x+2y-3z=0 \quad x+y+1=0$$

$$x+2y-3=0$$

$$x+y=-1$$

$$x+2y=3$$

$$y=4 \quad x=-5$$

$$x, y, z = (-5, 4, 1)$$

$$\frac{-5}{\sqrt{42}}, \frac{4}{\sqrt{42}}, \frac{1}{\sqrt{42}}$$

$$(x, y, z) = \left(\frac{-5}{\sqrt{42}}, \frac{4}{\sqrt{42}}, \frac{1}{\sqrt{42}} \right) \text{ or } \left(\frac{5}{\sqrt{42}}, \frac{-4}{\sqrt{42}}, \frac{-1}{\sqrt{42}} \right)$$

Q2: $Q = \begin{bmatrix} 1/\sqrt{2} & 0 \\ 1/\sqrt{2} & 0 \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}, \quad x = \begin{bmatrix} 1 \\ 3 \end{bmatrix}_{2 \times 1} \text{ and } y = \begin{bmatrix} -3\sqrt{2} \\ 6 \end{bmatrix}_{2 \times 1}$

Verify that: (i) $Q^T Q = I$

$$(ii) \|Qx\| = \|x\| \text{ and } \|Qy\| = \|y\|$$

$$(iii) (Qx)^T Qy = x^T y$$

$$\begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 1/\sqrt{2} & 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \\ 0 & 1/\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_2$$

$$\|\alpha\| = \sqrt{11}$$

$$\|y\| = \sqrt{54}$$

$$\|\beta\alpha\| = \begin{bmatrix} \sqrt{52} & 2/\sqrt{3} \\ \sqrt{52} & -2/\sqrt{3} \\ 0 & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} -3\sqrt{2} \\ 6 \end{bmatrix} = \begin{bmatrix} 1 \\ -7 \\ 2 \end{bmatrix} = \sqrt{54}$$

$$\|\beta y\| = \begin{bmatrix} \sqrt{52} & 2/\sqrt{3} \\ \sqrt{52} & -2/\sqrt{3} \\ 0 & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} \sqrt{2} \\ 3 \end{bmatrix} = \begin{bmatrix} 3 \\ -1 \\ 1 \end{bmatrix} = \sqrt{11}$$

$$(\beta\alpha)^T (\beta y) = \begin{bmatrix} 3 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -7 \\ 2 \end{bmatrix}_{3 \times 1} = 3 + 7 + 2 = 12$$

$$\alpha^T y = [\sqrt{2} \ 3] \begin{bmatrix} -3\sqrt{2} \\ 6 \end{bmatrix} = -6 + 18 = 12$$

Hence verified.

93. If w is a subspace spanned by the orthogonal vectors $(2, 5, -1)$ and $(-2, 1, 1)$. Find the point in w that is closest to $(1, 2, 3)$.

$$w = \begin{bmatrix} 2/\sqrt{30} & -2/\sqrt{6} \\ 5/\sqrt{30} & 1/\sqrt{6} \\ -1/\sqrt{30} & 1/\sqrt{6} \end{bmatrix}$$

$$p = A\hat{\alpha}$$

$$\hat{\alpha} = \frac{g^T b}{g^T g}$$

$$\hat{\alpha} = \begin{bmatrix} 2/\sqrt{30} & 5/\sqrt{30} & -1/\sqrt{30} \\ -2/\sqrt{6} & 1/\sqrt{6} & 1/\sqrt{6} \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

$$= \begin{bmatrix} 9/\sqrt{30} \\ 3/\sqrt{6} \end{bmatrix} \cdot \hat{\alpha}$$

$$p = \begin{bmatrix} 2/\sqrt{30} & -2/\sqrt{6} \\ 5/\sqrt{30} & 1/\sqrt{6} \\ -1/\sqrt{30} & 1/\sqrt{6} \end{bmatrix} \begin{bmatrix} 9/\sqrt{30} \\ 3/\sqrt{6} \end{bmatrix} = \begin{bmatrix} -12/\sqrt{30} \\ 2 \\ 6/\sqrt{30} \end{bmatrix} = \begin{bmatrix} -2/\sqrt{5} \\ 2 \\ 1/\sqrt{5} \end{bmatrix} = (-2/\sqrt{5}, 2, 1/\sqrt{5})$$

$$p = \frac{a_1^T b}{a_1^T a_1} a_1 + \frac{a_2^T b}{a_2^T a_2} a_2$$

$$p = [2 \ 5 \ -1] \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \frac{9}{30} a_1 + \frac{3}{10} (2, 5, -1) + \frac{1}{2} (2, 1, 1)$$

$$[-2, 1, 1] \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \frac{3}{6} a_2 \quad \left(\frac{3}{5} \cdot -1 \right), \left(\frac{3+1}{2} \right), \left(\frac{-3+1}{10} \right)$$

$$\left(\frac{2}{5}, 2, \frac{1}{5} \right)$$

Gram Schmidt Orthogonalization:

→ Given : A set of linearly independent vectors say a, b, c .

(note : necessary condition to apply Gram Schmidt orthogonalization is to have set of independent vectors)

→ Aim : To Find a set of orthonormal vectors say q_1, q_2, q_3

(note : number of orthonormal vectors obtained are equal to number of independent vectors given).

→ Procedure :-

(note : Always construct orthonormal vectors corresponding to the independent vectors in the given order.)

→ Objective :-

"Every vector in a given vector space say R^n , can always be expressed in terms of the linear combination of its one-Dimensional projection onto the orthonormal bases vectors".

e.g. vector $b \in R^n$

$$b = x_1 q_1 + x_2 q_2 + \dots + x_n q_n$$

where $x_i \in R$ q_i - orthonormal bases set

$$\text{s.t. } b = \frac{(q_1^T b) q_1}{q_1^T q_1} + \frac{(q_2^T b) q_2}{q_2^T q_2} + \dots + \frac{(q_n^T b) q_n}{q_n^T q_n}$$

$q_i^T q_j = 1$ whenever $i=j$

$$\therefore b = (q_1^T b) q_1 + (q_2^T b) q_2 + \dots + (q_n^T b) q_n$$

Gram Schmidt Process to solve Problems.

Step I: Find q_1 (consider only vector a)

$$q_1 = \frac{a}{\|a\|}$$



(both ' q_1 ' and ' a ' are in same direction).

Step II: Find q_2 (consider vectors b and q_1)

project vector ' b ' onto ' q_1 '

s.t. error vector $\perp q_1$



$$e \perp q_1$$

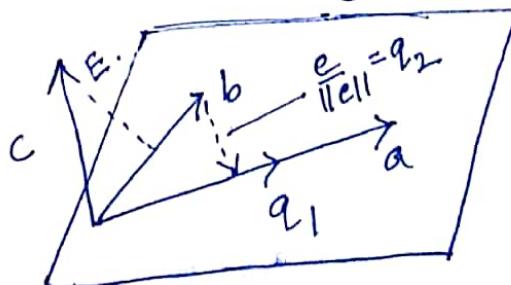
$$e = b - p = b - \hat{x}q_1 = b - (q_1^T b) q_1$$

$$e = b - (q_1^T b) q_1$$

normalise vector ' e'

$$q_2 = \frac{e}{\|e\|}$$

Step III: Find q_3 (consider vectors c and q_1 and q_2)



project 'c' onto the plane spanned
by q_1 and q_2

$$\therefore c = (q_1^T c) q_1 + (q_2^T c) q_2$$

$E \perp q_1$ and q_2

$$E = c - p$$

$$E = c - (\hat{x} q_1 + \hat{y} q_2)$$

$$\hat{x} = \frac{(q_1^T c)}{q_1^T q_1} \quad \hat{y} = \frac{(q_2^T c)}{q_2^T q_2}$$

$$E = c - (q_1^T c) q_1 - (q_2^T c) q_2$$

$$E = c - (q_1^T c) q_1 - (q_2^T c) q_2$$

Normalise vector 'E'

$$q_3 = \frac{E}{\|E\|}$$

Step IV : Vectors q_1, q_2, q_3 are
the required orthonormal vectors
w.r.t to the given independent
vectors a, b, c .

4.

Problems on Gram Schmidt Process.

Use GS process to find a set of orthonormal vectors q_1, q_2, q_3 from the given independent vectors a, b, c

$$a = (1, 1, 1), \quad b = (0, 1, 1), \quad c = (0, 0, 1)$$

Step I : $q_1 = \frac{a}{\|a\|}$

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

$$q_1 = \frac{1}{\sqrt{3}} (1, 1, 1) = \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right)$$

$$\boxed{q_1 = \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}}$$

Step II : $e = b - (q_1^T b) q_1$

$$e = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} - \left\{ \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right) \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\} \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}$$

$$e = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} - \frac{2}{\sqrt{3}} \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}$$

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

$$q_2 = \frac{e}{\|e\|} = \begin{pmatrix} -2/\sqrt{6} \\ 1/\sqrt{6} \\ 1/\sqrt{6} \end{pmatrix}$$

$$\therefore \|e\| = \sqrt{(-2/3)^2 + (1/3)^2 + (1/3)^2} = \frac{\sqrt{6}}{3}$$

$$\boxed{q_2 = \frac{-1}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}}}$$

Step III: $E = c - (q_1^T c) q_1 - (q_2^T c) q_2$

$$(q_1^T c) q_1 = \left\{ \begin{pmatrix} 1 \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right\} \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}$$

$$(q_1^T c) q_1 = \frac{1}{\sqrt{3}} \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}$$

$$(q_2^T c) q_2 = \left\{ \begin{pmatrix} -\frac{2}{\sqrt{6}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right\} \begin{pmatrix} -\frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \end{pmatrix}$$

$$(q_2^T c) q_2 = \frac{1}{\sqrt{6}} \begin{pmatrix} -\frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \end{pmatrix} = \begin{pmatrix} -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}$$

$$E = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} - \left\{ \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix} + \begin{pmatrix} -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix} \right\}$$

$$E = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} - \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$E = \begin{pmatrix} 0 \\ -\frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$$

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

$$q_3 = \frac{E}{\|E\|} = \begin{pmatrix} 0 \\ -\frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$$

$$\boxed{q_3 = \left(0, \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right)}$$

Given vectors :-

$$a = (1, 1, 1), b = (0, 1, 1), c = (0, 0, 1)$$

Apply G.S process

Obtained Orthonormal vectors :-

$$q_1 = \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}\right), q_2 = \left(\frac{2}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}}\right)$$

$$q_3 = \left(0, -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right).$$

Problems to solve :-

1. Apply Gram Schmidt process to

$$a = (1, 0, 1), b = (1, 0, 0), c = (2, 1, 0)$$

Solution :-

$$q_1 = \left(\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right), q_2 = \left(\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}}\right), q_3 = (0, 1, 0)$$

2. Find orthonormal vectors by G.S.

from $a = (1, -1, 0, 0)$

$$b = (0, 1, -1, 0)$$

$$c = (0, 0, 1, -1)$$

Need for Gram Schmidt Process.

Factorize $A = QR$.

$Q \rightarrow$ orthogonal Matrix
(square matrix with orthonormal vectors)

$R \rightarrow$ upper triangular matrix.

Given :- a set of independent vectors
say a, b, c , Find q_1, q_2, q_3 by G.S.

then $\therefore a = (q_1^T a) q_1$

$$b = (q_1^T b) q_1 + (q_2^T b) q_2$$

$$c = (q_1^T c) q_1 + (q_2^T c) q_2 + (q_3^T c) q_3$$

Matrix equivalent .

$$\begin{pmatrix} A \\ | & | & | \\ a & b & c \\ | & | & | \end{pmatrix} = \begin{pmatrix} Q \\ q_1 & q_2 & q_3 \\ | & | & | \end{pmatrix} \begin{pmatrix} R \\ q_1^T a & q_1^T b & q_1^T c \\ 0 & q_2^T b & q_2^T c \\ 0 & 0 & q_3^T c \end{pmatrix}$$

Problems :-

1. Apply G.S. to

$$a = (1, 1, 1), b = (0, 1, 1), c = (0, 0, 1)$$

and obtain QR factorization .

$$A = QR$$

$$R = Q^T A$$

$$R = \begin{pmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} & -\frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 & 2 \\ 1 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}$$

$$R = \begin{pmatrix} \sqrt{3} & \frac{2}{\sqrt{3}} & \frac{2}{\sqrt{3}} \\ 0 & \frac{2}{\sqrt{6}} & \frac{2}{\sqrt{6}} \\ 0 & 0 & \sqrt{2} \end{pmatrix}$$

$$R = \begin{pmatrix} q_1^T a & q_1^T b & q_1^T c \\ 0 & q_2^T b & q_2^T c \\ 0 & 0 & q_3^T c \end{pmatrix}$$

Live Session notes :-

16/3/20

Eigen Values and Eigen vectors :-

Page no. 1.

Topics :-

- 1) What are eigen values and vectors?
- 2) Geometrical meaning.
- 3) Why "eigen"
- 4) Properties
- 5) Largest eigen value.
- 6) Applications $\begin{cases} \text{Diagonalization. } (A = S \Delta S^{-1}) \\ \text{powers and products.} \end{cases}$

Later, $A = U \Sigma V^T \rightarrow \text{SVD. (Unit 5)}$

$$A_{n \times n} x_{n \times 1} = b_{n \times 1}$$
$$= \lambda x$$

$\Rightarrow Ax = \lambda x$, λ = eigenvalue
 x = eigen vector.

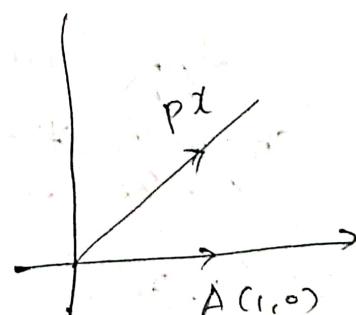
$$Ax - \lambda x = 0$$

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

$$\text{If } \lambda = 0 \Rightarrow b = 0$$

$$\lambda = 1 \Rightarrow b = x$$

λ = scale factor.



eigen-specific, characteristic.

If $\lambda_i > 0 \Rightarrow |A| \geq 0$.

If $\lambda_i < 0 \Rightarrow |A| < 0$.

$$Ax = \lambda x$$

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \lambda \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$$(A - \lambda I)x = 0 \Rightarrow |A - \lambda I| = 0.$$

$$\Rightarrow \begin{vmatrix} a_{11} - \lambda & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - \lambda & \dots & a_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} - \lambda \end{vmatrix} = 0.$$

$$\Rightarrow (-1)^n \lambda^n + c_1 \lambda^{n-1} + c_2 \lambda^{n-2} + \dots + c_n = 0 \quad \textcircled{1}$$

LHS is called characteristic polynomial.

Solving $\textcircled{1}$, we get $\lambda_1, \lambda_2, \dots, \lambda_n$ are called characteristic roots or latent roots or eigenvalues.

$$\lambda_1 + \lambda_2 + \dots + \lambda_n = \sum a_{ii} = \text{Trace of the matrix.}$$

$$\lambda_1 \cdot \lambda_2 \cdots \lambda_n = |A|.$$

Problem 1

$$\text{Let } A = \begin{bmatrix} 8 & -6 & 2 \\ -6 & 7 & -4 \\ 2 & -4 & 3 \end{bmatrix}$$

Find the eigen values and eigenvectors of A.

Solution:-

The characteristic eqn of A is

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

$$\Rightarrow \begin{vmatrix} 8-\lambda & -6 & 2 \\ -6 & 7-\lambda & -4 \\ 2 & -4 & 3-\lambda \end{vmatrix} = 0$$

$$\Rightarrow (8-\lambda)[(7-\lambda)(3-\lambda) - 16] + 6[-6(3-\lambda) + 8] + 2[24 - 2(7-\lambda)] = 0.$$

$$\Rightarrow \lambda^3 - 18\lambda^2 + 45\lambda = 0.$$

$$\Rightarrow \lambda(\lambda^2 - 18\lambda + 45) = 0.$$

$$\Rightarrow \lambda = 0, 3, 15.$$

Sum of the eigenvalues $= 0 + 3 + 15 = 18$

Sum of the diagonal elements $= 8 + 7 + 3 = 18$.

Case(i) $\lambda = 0$.

$$Ax = \lambda x \Rightarrow Ax = 0.$$

$$\Rightarrow \begin{bmatrix} 8 & -6 & 2 \\ -6 & 7 & -4 \\ 2 & -4 & 3 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Using the method of cross-multiplication,
Consider first two rows (independent)

$$\frac{x}{\begin{vmatrix} -6 & 2 \\ 7 & -4 \end{vmatrix}} = \frac{-y}{\begin{vmatrix} 8 & 2 \\ -6 & -4 \end{vmatrix}} = \frac{z}{\begin{vmatrix} 8 & -6 \\ -6 & 7 \end{vmatrix}}$$

$$\frac{x}{24 - 14} = \frac{-y}{-32 + 12} = \frac{z}{56 - 36}$$

$$\frac{x}{10} = \frac{-y}{-20} = \frac{z}{20}$$

$$\frac{x}{1} = \frac{y}{2} = \frac{z}{2}$$

Eigen vector corresponding to $\lambda = 0$ is
 $K(1, 2, 2)$, where $K \neq 0$.

III by For $\lambda = 3$, $Ax = 3x$

$$\Rightarrow (A - 3I)(x) = 0$$

$$\Rightarrow \begin{pmatrix} 5 & -6 & 2 \\ -6 & 4 & -4 \\ 2 & -4 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Solving, we get the eigen vector is
 $K(2, 1, -2)$.

Also For $\lambda = 15$, the eigen vector is
 $K(2, -2, 1)$.

Solving, we get "the eigenvalues are
 $\lambda = 1, 1, 5$.
Also, for $\lambda = 1$, the eigen vector is
 $x = (1, -2, 0)$.

Note:

Page no. 3

For a given eigen value, there are infinite eigen vectors.

Problem: 2 (Repeated eigen values).

Find the eigen values and eigen vectors

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

Solution:

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

$$\Rightarrow \begin{vmatrix} 2-\lambda & 2 & 1 \\ 1 & 3-\lambda & 1 \\ 1 & 2 & 2-\lambda \end{vmatrix} = 0$$

$$\Rightarrow \lambda^3 - 7\lambda^2 + 11\lambda - 5 = 0.$$

Solving, we get $\lambda = 1, 1, 5$. (eigen values)
($\because 5+1+1 = 2+3+2$).

For $\lambda = 1$, $(A - \lambda I)x = 0$

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

$$\Rightarrow \begin{pmatrix} 1 & 2 & 1 \\ 1 & 2 & 1 \\ 1 & 2 & 1 \end{pmatrix} = 0.$$

$$\Rightarrow \begin{pmatrix} 1 & 2 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = 0.$$

The eigenvectors are,

$K(-2, 1, 0)$; $K(-1, 0, 1)$,
For $\lambda = 5$, $K(1, 1, 1)$, where $K \neq 0$.

Live Class : 17/3/2020. at 8.10 - 9.40.

①

Topic : Properties of Eigen values &
The Largest Eigen vectors & Values.

Properties:-

- 1] eigen values are unique.
- 2] For a λ , eigen vectors are not unique.
- 3] $A \in A^T$ have the same eigenvalues & eigen vectors.
- 4] Eigen values of a diagonal matrix are diagonal entries.
- 5] Products of the diagonal elements = determinant of the matrix.
- 6] Sum of the eigenvalues = Trace of the matrix = Sum of the elements of the principal diagonal.
- 7] If λ is an eigen value of A , Then $\frac{1}{\lambda}$ is the eigenvalue of A^{-1} .
- 8] If $\lambda_1, \lambda_2, \dots, \lambda_m$ are eigen values of A , then $\lambda_1^n, \lambda_2^n, \dots, \lambda_m^n$ are eigenvalues of A^n . where $n \in \mathbb{I}^+$.

9] If A is singular matrix, then zeros are the eigen values.

Problems:

Find the eigen values and eigenvectors of the matrix $A = \begin{bmatrix} 1 & -1 \\ 2 & 4 \end{bmatrix}$. Verify that

trace equals to the sum of the eigenvalues and the determinant equals to their product.

If we shift A to $A - \lambda I$, then what are the eigenvalues ~~of~~ of $A - \lambda I$, and how are they related to those of A.

Ans :-

The eigen values are given by

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

$$\begin{vmatrix} 1-\lambda & -1 \\ 2 & 4-\lambda \end{vmatrix} = 0$$

$$(1-\lambda)(4-\lambda) + 2 = 0$$

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

$$\lambda^2 - 5\lambda + 6 = 0$$

$$[\lambda^2 - (\text{sum of the diagonal elements})\lambda + |A| = 0]$$

$$(\lambda-2)(\lambda-3) = 0$$

$$\lambda = 2, \lambda = 3.$$

The eigenvalues are $\lambda = 2, \lambda = 3$.

The eigen vectors are given by ②

$$AX = \lambda X$$

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

$$\begin{pmatrix} 1-\lambda & -1 \\ 2 & 4-\lambda \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

When $\lambda = 2$,

$$\begin{pmatrix} -1 & -1 \\ 2 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = 0$$

$$R_2 \rightarrow R_2 + 2R_1$$

$$\begin{pmatrix} -1 & -1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$-x - y = 0 \Rightarrow x = -y$$

y is free variable. Let $y = 1$.

then $x = -1$.

∴ The eigen vector is $x_1 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$ or $K \begin{pmatrix} -1 \\ 1 \end{pmatrix}$.

When $\lambda = 3$,

$$\begin{pmatrix} -2 & -1 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = 0$$

$$R_2 \rightarrow R_2 + R_1$$

$$\begin{pmatrix} -2 & -1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = 0$$

$$-2x - y = 0$$

y is free variable. Let $y = 1$.

$$X_2 = \begin{bmatrix} -1/2 \\ 1 \end{bmatrix} \text{ or } K \begin{bmatrix} -1/2 \\ 1 \end{bmatrix},$$

$$\left. \begin{array}{l} \text{Trace} = 1+4 = 5 \\ \text{Sum of the eigenvalues} = 2+3 = 5 \\ \text{Product of the eigenvalues} = 2 \times 3 = 6 \end{array} \right\} \text{equal.}$$

Det. $|A| = \begin{vmatrix} 1 & -1 \\ 2 & 4 \end{vmatrix} = 4+2 = 6.$

Construct $A - \lambda I$.

$$A - \lambda I = \begin{pmatrix} 1 & -1 \\ 2 & 4 \end{pmatrix} - \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix} = \begin{pmatrix} 1-\lambda & -1 \\ 2 & 4-\lambda \end{pmatrix}$$

$$\underline{\lambda=7}$$

$$A - 7I = \begin{pmatrix} 1-7 & -1 \\ 2 & 4-7 \end{pmatrix}$$

$$A - 7I = \begin{pmatrix} -6 & -1 \\ 2 & -3 \end{pmatrix} = B \text{ (say)}.$$

The eigenvalues are given by

$$|B - \lambda I| = 0.$$

$$\begin{vmatrix} -6-\lambda & -1 \\ 2 & -3-\lambda \end{vmatrix} = 0$$

$$\Rightarrow \lambda^2 - (-9)\lambda + 20 = 0$$

$$\lambda^2 + 9\lambda + 20 = 0$$

$$(\lambda + 4)(\lambda + 5) = 0$$

$$\lambda = -4, -5$$

The eigenvectors are given by

$$(B - \lambda I) \mathbf{x} = 0.$$

$$\begin{pmatrix} -6 - \lambda & -1 \\ 2 & -3 - \lambda \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

When $\lambda = -4$, $\begin{pmatrix} -2 & -1 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

$$\Rightarrow R_2 + R_1 = R_2$$

$$\begin{pmatrix} -2 & -1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{aligned} -2x - y &= 0 & \text{Let } y = 1 \\ -2x &= 1 & \Rightarrow x = -\frac{1}{2} \end{aligned}$$

$$x_1 = \begin{bmatrix} -\frac{1}{2} \\ 1 \end{bmatrix} \text{ or } K \begin{bmatrix} -\frac{1}{2} \\ 1 \end{bmatrix}.$$

When $\lambda = -5$, $\begin{pmatrix} -1 & -1 \\ 2 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

$$-x - y = 0 \Rightarrow x = -y$$

$$\text{Let } y = 1, x = -1.$$

$$\therefore x_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \text{ or } K \begin{bmatrix} -1 \\ 1 \end{bmatrix}.$$

Eigenvalues of A are : 2, 3

Eigenvalues of $B = A - 7I$ are : -4, -5.

$$2 - (-4) = 2 -$$

Problems :-

Find the eigenvalues of the matrices A, A^2, A^{-1} and $A + 4I$, given $A = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$.

Ans :-

The eigenvalues of A are

$$\lambda = 1, \lambda = 3.$$

The eigenvalues of A^2 are : $1^2, 3^2 = 1, 9.$

$$" " " A^{-1} : \frac{1}{1}, \frac{1}{3}$$

$$" " " A + 4I : 1+4, 3+4 = 5, 7.$$

Problem :

Write the 3 different 2×2 matrices whose eigenvalues are 4 and 5.

Ans :- Given, eigenvalues are 4, 5.

$$\therefore |A| = 20.$$

1) $A = \begin{bmatrix} 4 & 0 \\ 0 & 5 \end{bmatrix}$ (diagonal matrix)

2) $A = \begin{bmatrix} 4 & 10 \\ 0 & 5 \end{bmatrix}$ (upper Δ matrix)

3) $A = \begin{bmatrix} 4 & 0 \\ -1 & 5 \end{bmatrix}$ (lower Δ matrix).

$\underbrace{\quad}_{\times} \underbrace{\quad}_{\times}$

(4)

Rayleigh's Power Method

- 1] What
- 2] Why
- 3] Procedure.
- 4] Adv. & Disadv.
- 5] Applications.

Finding largest eigen value and corresponding eigen vector.

→ It is an iterative process. Writing an algorithms is easy!

Procedure :-

1] Given :- A, initial approximation of eigen vector.

2] $Ax = \lambda x$

3] Applying initial eigen vector as x_0 .

$$Ax_0 = \lambda_1 x_1, Ax_1 = \lambda_2 x_2, \dots, Ax_{n-1} = \lambda_n x_n$$

4] Stop the process when two successive iterations are same.

Problem:-

Find the largest eigenvalue and eigenvector corresponding to the matrix $A = \begin{pmatrix} 4 & 1 & -1 \\ 2 & 3 & -1 \\ -2 & 1 & 5 \end{pmatrix}$.

Ans :-

Since initial eigen vector is not given, assume it one of the form $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$.

Let $x^{(0)} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$.

First Iteration :-

$$AX^{(0)} = \begin{bmatrix} 4 & 1 & -1 \\ 2 & 3 & -1 \\ -2 & 1 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \\ -2 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ 0.5 \\ -0.5 \end{bmatrix} = \lambda_1 x^{(1)}$$

Second Iteration :-

$$AX^{(1)} = \begin{bmatrix} 4 & 1 & -1 \\ 2 & 3 & -1 \\ -2 & 1 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 0.5 \\ -0.5 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \\ -4 \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 0.8 \\ -0.8 \end{bmatrix} = \lambda_2 x^{(2)}$$

Third Iteration :-

$$AX^{(2)} = \begin{bmatrix} 4 & 1 & -1 \\ 2 & 3 & -1 \\ -2 & 1 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 0.8 \\ -0.8 \end{bmatrix} = \begin{bmatrix} 5.6 \\ 5.2 \\ -5.2 \end{bmatrix} = 5.6 \begin{bmatrix} 1 \\ 0.9285 \\ -0.9285 \end{bmatrix} \\ = 5.6 \begin{bmatrix} 1 \\ 0.929 \\ -0.929 \end{bmatrix} = \lambda_3 x^{(3)}$$

Fourth Iteration :-

$$AX^{(3)} = \begin{bmatrix} 5.857 \\ 5.714 \\ -5.714 \end{bmatrix} = 5.857 \begin{bmatrix} 1 \\ 0.976 \\ -0.976 \end{bmatrix} = \lambda_4 x^{(4)}$$

Fifth Iteration :-

$$AX^{(4)} = \begin{bmatrix} 5.9512 \\ 5.9024 \\ -5.9024 \end{bmatrix} = 5.9512 \begin{bmatrix} 1 \\ 0.9908 \\ -0.9908 \end{bmatrix} = \lambda_5 x^{(5)}$$

Sixth Iteration :-

$$AX^{(5)} = \begin{bmatrix} 5.9816 \\ 5.9632 \\ -5.9632 \end{bmatrix} = 5.9816 \begin{bmatrix} 1 \\ 0.9969 \\ -0.9969 \end{bmatrix} = \lambda_6 x^{(6)}$$

Seventh Iteration:-

$$AX_6 = \begin{pmatrix} 5.9938 \\ 5.9876 \\ -5.9876 \end{pmatrix} = 5.9938 \begin{pmatrix} 1 \\ 0.999 \\ -0.999 \end{pmatrix} = \lambda_1 X^{(7)}$$

Stop the process

The Largest eigen value = 5.994 ≈ 6 .
" " " vector $\approx \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$.

* Google, twitter uses this method.

* Smallest eigen value of $A =$ Largest eigen value of A^{-1} .

Diagonalization.

It is the process of finding an invertible matrix S and diagonal matrix Λ , such that the given matrix A satisfies

$$A = S \Lambda S^{-1}$$

Example :- ① ^{A II} diagonal matrix ^{Eg:-} $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$ are diagonalizable

② ^{A II} symmetric Matrix ^{Eg:-} $A = \begin{bmatrix} 8 & -6 & 2 \\ -6 & 7 & -4 \\ 2 & -4 & 3 \end{bmatrix}$ are diagonalizable

Non Examples :- ① Shear matrix $\begin{bmatrix} 1 & a \\ 0 & 1 \end{bmatrix}$ ($a \neq 0$)
 $(k \begin{pmatrix} 1 \\ 0 \end{pmatrix})$ are eigenvectors)

② A Jordan block $\begin{bmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{bmatrix}$
 $(k \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix})$ are eigenvectors

③ A rotation matrix $\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$ $\theta \neq 0, \pi$
 $(\text{No } \begin{pmatrix} 1 \\ 0 \end{pmatrix} \text{ eigenvectors})$

$$A = S \Lambda S^{-1} \rightarrow \text{Eigen value matrix.}$$

Eigen
Vector

Diagonalizability of $A \Leftrightarrow$ Existence of a basis of Eigen vectors

$$A = S \Lambda S^{-1}$$

Column form a basis of Eigen ~~vector~~ spaces

Note: If A is $n \times n$ matrix has n distinct Eigen values, then A must be diagonalizable.

The condition is not necessarily $\forall \lambda = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix}$, $\lambda=1,1,2$

Problems:-

(1) Show that $A = \begin{bmatrix} 8 & -6 & 2 \\ -6 & 7 & -4 \\ 2 & -4 & 3 \end{bmatrix}$ is diagonalizable.

Soln:- For this matrix $\lambda_1=0, \lambda_2=3, \lambda_3=15$ (Eigen values)

$$v_1 = \begin{pmatrix} 1 \\ 2 \\ 2 \end{pmatrix}, v_2 = \begin{pmatrix} 2 \\ 1 \\ -2 \end{pmatrix}, v_3 = \begin{pmatrix} 2 \\ -2 \\ 1 \end{pmatrix} \text{ (Eigen vectors)}$$

$$\text{Set } S = \begin{pmatrix} 1 & 2 & 2 \\ 2 & 1 & -2 \\ 2 & -2 & 1 \end{pmatrix} \quad A = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 15 \end{pmatrix}$$

Scaling the column of S , we get the orthogonal matrix P ,

$$A = PAP^T$$

$\Rightarrow A$ is diagonalizable.

(2)

Show that $\begin{bmatrix} 1 & 3 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 3 \end{bmatrix}$ is not diagonalizable.

Since it is an upper triangular matrix $\lambda=1,1,3$ are the Eigen values

$$\lambda=1, (A-\lambda I)x=0$$

$$\Rightarrow \begin{bmatrix} 0 & 3 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Thus $(x_1, 0, 0)$ are the Eigen vectors

Thus there are not enough ~~values~~ Eigen vectors (we need two independent Eigen vectors)

The number of independent Eigen vectors is no more than the multiplicity of the respective Eigen value.
 \Rightarrow it is not diagonalizable.

(3) Show that $A = \begin{pmatrix} 1 & 0 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 3 \end{pmatrix}$ is diagonalizable.

Since it is upper diag matrix $\lambda = 1, 1, 3$ are the Eigen values.

$$\lambda=1 \quad (A - \lambda I)x = 0 \Rightarrow \begin{pmatrix} 0 & 0 & 3 \\ 0 & 0 & 2 \\ 0 & 0 & 2 \end{pmatrix}x = 0 \quad \text{Eigen vectors} \\ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$

$$\lambda=3 \quad (A - \lambda I)x = 0 \Rightarrow \begin{pmatrix} -2 & 0 & 3 \\ 0 & -2 & 2 \\ 0 & 0 & 0 \end{pmatrix}x = 0 \quad \text{Eigen vectors} \\ \begin{pmatrix} -3 \\ -2 \\ 1 \end{pmatrix}$$

$$S = \begin{pmatrix} 1 & 0 & -3 \\ 0 & 1 & -2 \\ 0 & 0 & 1 \end{pmatrix} \quad \Lambda = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

$\therefore A$ is diagonalizable.

We can compute power of a diagonalizable matrix A as follows,

$$A = S \Lambda S^{-1} \quad \text{then} \quad A^n = (S \Lambda S^{-1})^n \\ = S \Lambda S^{-1} \underbrace{S \Lambda S^{-1} \underbrace{S \Lambda S^{-1} \dots}_{(n \text{ times})} S \Lambda S^{-1}}_{S \Lambda S^{-1}}$$

$$= S \Lambda^n S^{-1}$$

Examples :-

$$\textcircled{1} \quad \det A = \begin{bmatrix} 3 & -1 & 1 \\ 0 & 2 & 0 \\ 1 & -1 & 3 \end{bmatrix}, \text{ find } A^4.$$

Solution :- Here $\lambda = 2, 2, 4$ are the Eigen values and $\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$ are the Eigen vectors.

$$S = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \quad A = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{pmatrix}$$

Matrix A satisfy $A = S \Lambda S^{-1}$

$$\begin{aligned} \therefore A^4 &= S \Lambda^4 S^{-1} = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 16 & 0 & 0 \\ 0 & 16 & 0 \\ 0 & 0 & 256 \end{pmatrix} \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \end{pmatrix} \\ &= \begin{pmatrix} 136 & -120 & 120 \\ 0 & 16 & 0 \\ 120 & -120 & 136 \end{pmatrix} \end{aligned}$$

If A and B are diagonalizable matrices can we compute the product the same way?

Not unless A and B have the same Eigen vectors.

Cayley Hamilton theorem :-

Every square matrix A, satisfies its own characteristic equation.

Eg:- If A has characteristic polynomial

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

then $A^2 - 3A + 2I = [0]$ → zero matrix

Thus $A - 3A + 2A^T = [0]$ (Multiply by A^T)

$$\text{Then } A^T = \frac{3I - A}{2}$$

If $A = SAS^{-1}$, $B = S\Gamma S^{-1}$ then $AB = BA = S\Lambda\Gamma S^{-1}$

Theorem :- $AB = BA$. \Leftrightarrow diagonalizable A and B iff
A & B have the same Eigen vectors.