

Eigen Values , Eigen Vectors & Diagonalization:

Motivation: Let A be a real symmetric matrix. Consider the following problem:-

Maximize (Minimize) $x^t A x$ s.t $x \in \mathbb{R}^n$ & $x^t x = 1$

To solve this, consider the Lagrangian.

$$\begin{aligned} L(x, \lambda) &= x^t A x - \lambda(x^t x - 1) \\ &= \sum_{i=1}^n \sum_{j=1}^n a_{ij} x_i x_j - \lambda \left(\sum_{i=1}^n x_i^2 - 1 \right) \end{aligned}$$

Partially differentiating $L(x, \lambda)$ w.r.t x_i , $1 \leq i \leq n$, we get

$$0 = \frac{\partial L}{\partial x_1} = 2a_{11}x_1 + 2a_{12}x_2 + \dots + 2a_{1n}x_n - 2\lambda x_1$$

$$0 = \frac{\partial L}{\partial x_2} = 2a_{21}x_1 + 2a_{22}x_2 + \dots + 2a_{2n}x_n - 2\lambda x_2$$

$$\vdots$$

$$0 = \frac{\partial L}{\partial x_n} = 2a_{n1}x_1 + 2a_{n2}x_2 + \dots + 2a_{nn}x_n - 2\lambda x_n$$

We get

$$\boxed{AX = \lambda X}$$

Q Consider a system of n -ordinary differential eqⁿ of the form

$$\frac{dy(t)}{dt} = Ay \quad t \geq 0, \quad \text{where } A \text{ is a real } n \times n \text{ matrix}$$

To obtain a solⁿ, let us assume that $y(t) = e^{et} c$ is a solⁿ of (1). Then, we get

$$e^{et} c' = Ae^{et} c \Leftrightarrow (A - etI)c = 0. \quad (2)$$

So, (2) is a solⁿ of (1) iff λ & c satisfy (3).

i.e given a matrix of order n , we are lead to find a pair (λ, c) s.t $c \neq 0$, & (3) has a solⁿ.

Suppose A be a matrix of order n . In general, we ask the question :-

For what values of $\lambda \in \mathbb{F}$, there exists a non-zero vector

$$x \in \mathbb{F}^n \text{ s.t } \boxed{Ax = \lambda x}. \quad (A)$$

Here \mathbb{F}^n stand for either vector space \mathbb{R}^n or \mathbb{C}^n

eqⁿ (A) equivalent to $\boxed{(A - \lambda I)x = 0}. \quad (B)$

We know that (B) has a non-zero solⁿ if $\text{rank}(A - \lambda I) < n$ or equivalently, $\det(A - \lambda I) = 0$.

We observe that $\det(A - \lambda I)$ is a polynomial of degree n .

We are therefore lead to the following definition:-

Definition: Let A be a matrix of order n . The polynomial $\det(A - \lambda I)$ is called the characteristic poly. of A and is denoted by $p(\lambda)$.

The eq
$$\boxed{p(\lambda) = 0}$$
 is called the characteristic eq of A .
 If $\lambda \in \mathbb{F}$ is a root of the characteristic eq $p(\lambda) = 0$, then λ is called characteristic root or ch. value of A or eigenvalue.

Result: Let $A = [a_{ij}]$; $a_{ij} \in \mathbb{F}$ for $1 \leq i, j \leq n$. Suppose $\lambda = \lambda_0 \in \mathbb{F}$ is a root of the ch. eq. Then \exists a non-zero $v \in \mathbb{F}^n$

$$s.t \quad Av = \lambda_0 v$$

Pf: λ_0 is a root of ch. eq. $\therefore \det(A - \lambda_0 I) = 0$

$\Rightarrow A - \lambda_0 I$ is singular.

\therefore The linear system has a non zero solⁿ.

#. We note that each scalar multiple kv of an eigenvector belonging to λ is also such an eigenvector.
 $A(kv) = kAv = \lambda(kv)$

Observation: The linear system $AX = \lambda X$ has a solⁿ $X=0$ for every $\lambda \in \mathbb{F}$. So, we consider only those $X \in \mathbb{F}^n$ that are non-zero and are solutions of the linear system $AX = \lambda X$.

Definition: Eigenvalue & Eigen vector:

If the linear system $AX = \lambda X$ has a non-zero solⁿ $X \in \mathbb{F}^n$ for some $\lambda \in \mathbb{F}$, then

- 1) $\lambda \in \mathbb{F}$ is called an eigenvalue of A.
- 2) $0 \neq X \in \mathbb{F}^n$ is called eigen vector corresponding to the eigenvalue λ of A and
- 3) the tuple (λ, X) is called a eigenpair.

Remark: To understand the difference b/w a ch. value and a eigenvalue, we give the following example:-

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

$$\text{ch. pol}^n = \det(A - \lambda I) = \lambda^2 + 1. \quad \text{ch. polynomial is } \lambda^2 + 1 = 0$$

$$\text{ch. roots}, \quad \lambda = \pm i$$

1. If $\mathbb{F} = \mathbb{C}$, ie A is considered as a complex matrix. Then the roots of $p(\lambda) = 0$ in \mathbb{C} are $\pm i$.

$$\rightarrow \det(A - \lambda I) = (-1)^n \det(\lambda I - A)$$

Caley Hamilton Thm: Every matrix A is a root of its ch-eq.^w

Remark: [largest power of α (more than n)
3. Inverse]

Suppose $A = [a_{ij}]_{n \times n}$ is a triangular matrix.

Then $(\lambda I - A)$ is a triangular matrix with diagonal entries

$\lambda - a_{ii}$. Hence.

$$\text{ch. pol} = \det(\lambda I - A) = (\lambda - a_{11})(\lambda - a_{22}) \cdots (\lambda - a_{nn})$$

Observation: The roots of ch. poly are the diagonal elements of A .

or eigen values are diagonal elements of A .

Ex: $A = \begin{bmatrix} 1 & 3 \\ 4 & 5 \end{bmatrix}$.

$$\text{ch. poly} = |\lambda I - A| = \begin{vmatrix} \lambda - 1 & -3 \\ -4 & \lambda - 5 \end{vmatrix} = (\lambda - 1)(\lambda - 5) - 12 = \lambda^2 - 6\lambda + 7$$

By Caley Hamilton: $A^2 - 5A - 7I = 0$.

$$A^2 - 5A - 7I = \begin{bmatrix} 1 & 3 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 1 & 3 \\ 4 & 5 \end{bmatrix} - 5 \begin{bmatrix} 1 & 3 \\ 4 & 5 \end{bmatrix} - 7 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

? Find A^{-1} , using Caley Hamilton Thm,

We have $A^2 - 5A - 7I = 0$.

Premultiplying by A^{-1} , we obtain,

$$7A^{-1} = -5I + A$$

$$7A^{-1} = -\begin{bmatrix} 6 & 0 \\ 0 & 6 \end{bmatrix} + \begin{bmatrix} 1 & 3 \\ 4 & 5 \end{bmatrix} = \begin{bmatrix} -5 & 3 \\ 4 & -1 \end{bmatrix}$$

$$A^{-1} = \frac{-1}{7} \begin{bmatrix} 5 & -3 \\ -4 & 1 \end{bmatrix}.$$

③ $A^2 = 6A - 7I = \begin{bmatrix} 6 & 18 \\ 24 & 30 \end{bmatrix} - \begin{bmatrix} 7 & 0 \\ 0 & 7 \end{bmatrix} = \begin{bmatrix} -1 & 18 \\ 24 & 23 \end{bmatrix}.$

Ex: $\begin{bmatrix} -1 & -1 & 1 \\ 1 & -1 & 1 \\ 0 & 1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & -2 & -1 \\ -2 & 1 & -1 \\ 0 & -1 & 2 \end{bmatrix}$ (Find inverse using Caley Hamilton)

Similar matrix \Leftrightarrow Let A & B are two $n \times n$ matrix.

A is said to be similar to B if \exists non-singular A^{-1} matrix

$$B \text{ s.t } B = P^{-1}AP.$$

Diagonalizable matrix \Leftrightarrow Let A be an $n \times n$ matrix. The

matrix A is said to be diagonalizable if there exist a non-singular matrix P such that

$$B = P^{-1}AP \text{ is a diagonal matrix.}$$

Remark: 1) Suppose A and B are similar matrices.

Then A and B have same ch. poly.

$$\begin{aligned}
 \text{Character poly. of } B &= \det(\lambda I - B) \\
 &= \det(\lambda I - P^{-1}AP) \\
 &= \det(\lambda P^{-1}P - P^{-1}AP) \\
 &= \det(P^{-1}(\lambda I - A)P) \\
 &= \det P^{-1} \cdot \det(\lambda I - A) \cdot \det P \\
 &= \det(\lambda I - A) \\
 &= \text{ch. poly. of } A
 \end{aligned}$$

Hence, eigen values are also same.
 → Simplest form of characteristic poly. of degree 2 or 3.

Here $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$

$$\begin{aligned}
 \text{ch. poly.} &= \lambda^2 - (a+d)\lambda + ad - bc \\
 &= \lambda^2 - (\text{trace } A)\lambda + \det A
 \end{aligned}$$

$$\rightarrow A = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} \quad \text{ch. poly.} \rightarrow \left(\begin{array}{l} \lambda^3 - (\text{trace } A)\lambda^2 + (A_{11} + A_{22} + A_{33})\lambda \\ - \det A \end{array} \right)$$

where A_{ii} denote the cofactor of a_{ii} .

$$S_1 = \text{trace } A, \quad S_2 = A_{11} + A_{22} + A_{33}, \quad S_3 = \det A, \quad \left| \begin{array}{l} S_k \text{ is the sum of} \\ \text{the principal minors} \\ \text{of order } k. \end{array} \right|$$

* Characterization of Diagonal Matrix :-

Let A be $n \times n$ matrix, which is similar to a diagonal matrix iff A has n -linearly independent eigenvectors.

In this case, the diagonal elements of D are the corresponding eigenvalues and $D = P^T A P$, where P is the matrix whose columns are the eigenvectors.

Remark: (ii) "If $\lambda_1, \lambda_2, \dots, \lambda_k$ are distinct eigenvalues of a matrix A with corresponding eigenvectors x_1, x_2, \dots, x_k . Then the set $\{x_1, x_2, \dots, x_k\}$ is linearly independent".

Pf: This is done by induction.

As for $R=1$, obvious.

Assume for $R=m$, $1 \leq m \leq k$

We prove the result for $k=m+1$. Consider,

$$c_1 x_1 + c_2 x_2 + \dots + c_{m+1} x_{m+1} = 0 \quad \text{for } c_1, \dots, c_{m+1} \text{ are unknown.}$$

$$A(c_1 x_1 + c_2 x_2 + \dots + c_{m+1} x_{m+1}) = A(0) = 0.$$

$$\Rightarrow c_1 \lambda_1 x_1 + c_2 \lambda_2 x_2 + \dots + c_{m+1} \lambda_{m+1} x_{m+1} = 0 \quad \text{--- (2)}$$

$\Rightarrow c_1 \lambda_1 x_1 + c_2 (\lambda_2 - \lambda_1) x_2 + \dots + c_{m+1} (\lambda_{m+1} - \lambda_1) x_{m+1} = 0$

\therefore By induction hypothesis, $(\lambda_i - \lambda_1)(\lambda_{m+1} - \lambda_1) = 0$

These are n -vector,

Hence $c_1 \lambda_1 = 0$ But $\lambda_1 \neq \lambda_{m+1} \therefore c_1 = 0$

3) The eigenvectors corresponding to distinct eigenvalues of $n \times n$ matrix A are linearly independent.

Properties Suppose the ~~product~~ characteristic poly. $p(\lambda)$ of an- n -square matrix A is a product of n -distinct factor say $p(\lambda) = (\lambda - a_1)(\lambda - a_2) \dots (\lambda - a_n)$. Then A is similar to the diagonal matrix D with diagonal entries a_i .

Let A be a square matrix over the complex field \mathbb{C} . Then A has at least one eigenvalue.

Suppose a matrix A can be diagonalizable. Then

$$\text{i.e } D = P^T A P$$

$$\text{Then } A = P D P^{-1}$$

$$A^2 = P D P^{-1} P D P^{-1} = P^T D^2 P$$

$$A^m = P^T D^m P^{-1} = P \text{ dia}\{d_1^m, d_2^m, \dots, d_n^m\} P^{-1}.$$

2) More generally, for any polynomial $f(t)$,

$$\begin{aligned} f(A) &= f(PDP^{-1}) \\ &= P f(D) P^{-1} \\ &= P \text{diag}(f(\lambda_1), f(\lambda_2), \dots, f(\lambda_n)) P^{-1}. \end{aligned}$$

3) Furthermore, if diagonal entries of D are non-negative,

$$B = P \text{diag}(\sqrt{\lambda_1}, \sqrt{\lambda_2}, \dots, \sqrt{\lambda_n}) P^{-1}$$

Then B is a non-negative square root of A .

i.e. $B^2 = A$ & the eigenvalues of B

are non-negative.

→ If " λ " is an eigenvalue of the matrix A , then the algebraic multiplicity of " λ " is defined to be the multiplicity of λ as a root of ch. poly. of A .

→ Geometric multiplicity of λ is defined to be the dimension of eigen space.

① Problem Set:

$\Rightarrow A = \text{diag}(d_1, d_2, \dots, d_n)$, $d_i \in \mathbb{R}$, $1 \leq i \leq n$.

Then $p(\lambda) = \prod_{i=1}^n (\lambda - d_i)$. is the ch. eqⁿ.

∴ eigen pair are

$(d_1, e_1), (d_2, e_2), \dots, (d_n, e_n)$

$e_i = (0, 0, \dots, \underset{i\text{th place}}{1}, \dots, 0)$

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

characteristic eqⁿ $(1-\lambda)^2 = 0$.

eigen value $- 1, 1$ (repeated eigen value).

eigen vector $- x_1(1, 0)$

$$\left| \begin{array}{l} (A - I_2)x = 0 \\ \left(\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right) \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \end{array} \right.$$

Two eigen values 1, but only one eigenvector.
(Not diagonalizable)

$$\left[\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right]$$

$$\boxed{x_2 = 0}$$

③

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \det(A - \lambda I_2) = (1-\lambda)^2$$

ch. roots $1, 1$.

Any two linearly independent vector of \mathbb{R}^2 are eigen vector.

$$4) \quad A = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}.$$

$$\det(A - \lambda I) = (\lambda - 3)(\lambda + 1).$$

ch. eqⁿ has roots 3, -1.

eigen ~~pair~~ — $(3, (1, 1)^T)$, $(-1, (1, -1)^T)$ — (Check).

In this case, we have Two distinct eigenvalues
and the corresponding eigen vectors are also
linearly independent (Diagonalizable).

$$5). \quad A = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}, \quad \det(A - \lambda I_2) = \lambda^2 - 2\lambda + 2.$$

The ch. roots are $1+i$, $1-i$

1) Hence A has no real value over \mathbb{R} .

2) It has two roots over \mathbb{C} .

corresponding eigen vectors are $(i, 1)$ & $(1, i)$

Eigen pair $(1+i, (i, 1))$ & $(1-i, (1, i))$.

So, A has $(1, (1, i)^T)$ & $(-i, (1, -i)^T)$ as eigen pair

$$Ax = i^{\circ}x$$

$$\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} i^{\circ}x_1 \\ ix_2 \end{bmatrix}$$

$$x_2 = i^{\circ}x_1$$

$$-x_1 = ix_2$$

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ ix_1 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ i \end{pmatrix}$$

$$Ax = -i^{\circ}x$$

$$x_2 = -i^{\circ}x_1$$

$$-x_1 = -ix_2$$

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ -ix_1 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ -i \end{pmatrix}$$

Thus $(i, (1))$, $(-i, (-i))$ are two eigen pair.

* Note: A & A^T have same ch. poly.

* If λ is an eigen value of A then $\frac{1}{\lambda}$ is an eigen value of A^{-1} .

$$\begin{aligned} 0 &= \det(A - \lambda I) = \det(A - \lambda A^{-1}A) \\ &\subseteq \det(\lambda \lambda^T A - A A^{-1}A) \\ &= \det(\lambda (\cancel{\lambda^T A} A^{-1}) A) \\ &= \lambda^n \det(\lambda^T I - A^{-1}) \det A \end{aligned}$$

$$\Rightarrow \det(\lambda^T I - A^{-1}) = 0.$$

Ex :- Let $A = \begin{bmatrix} 3 & 1 \\ 2 & 2 \end{bmatrix}$.

$$\det v_1 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}, \quad v_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

Then

$$Av_1 = \begin{bmatrix} 3 & 1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \end{bmatrix} = v_1$$

$$Av_2 = 4v_2.$$

i.e. 1, 4 are eigen values corresponding to eigenvector v_1 & v_2 .

Since v_1 & v_2 are l.i. & hence form a basis of \mathbb{R}^2 .

Accordingly A is diagonalizable.

furthermore, let $P = [v_1 \ v_2]$ (ie whose columns are eigenvectors).

$$P = \begin{bmatrix} 1 & 1 \\ -2 & 1 \end{bmatrix}, \quad P^{-1} = \begin{bmatrix} \frac{1}{3} & -\frac{1}{3} \\ \frac{2}{3} & \frac{1}{3} \end{bmatrix}.$$

A is similar to Diagonal matrix.

$$D = P^{-1}AP = \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix}.$$

Accordingly,

$$A^4 = P D^4 P^{-1}$$
$$= \begin{bmatrix} y_3 & -y_3 \\ -y_2 & y_2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 256 \end{bmatrix} \begin{bmatrix} y_3 & -y_3 \\ y_3 & y_3 \end{bmatrix} = \begin{bmatrix} 171 & 85 \\ 170 & 86 \end{bmatrix}.$$

Moreover, suppose $f(t) = t^3 - 5t^2 + 3t + 6$.

$$f(1) = 5, \quad f(4) = 2.$$

$$f(A) = P f(D) P^{-1}$$
$$= \begin{bmatrix} 1 & 1 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 5 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} \frac{1}{3} & -y_3 \\ y_3 & y_3 \end{bmatrix} = \begin{bmatrix} 3 & -1 \\ -2 & 4 \end{bmatrix}.$$

→ Positive Square root of A

$$B = P \sqrt{D} P^{-1}$$
$$= \begin{bmatrix} 1 & 1 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} y_3 & -y_3 \\ y_3 & y_3 \end{bmatrix} = \begin{bmatrix} \sqrt{3} & y_3 \\ 2\sqrt{3} & 4y_3 \end{bmatrix}.$$

$$\rightarrow A = \begin{bmatrix} 2 & 1 & -2 \\ 2 & 3 & -4 \\ 1 & 1 & -1 \end{bmatrix} \quad |(9.27)$$

- λ :- 1) Find all eigenvalues of A .
 2) Eigen Vectors of A .
 3) Find a basis of each eigen space.
 4) Is A diagonalizable.

$$\text{Soln: Ch. poly of } A = \lambda^3 - (\text{trace } A)\lambda^2 + (A_{11} + A_{22} + A_{33})\lambda + \det A$$

$$\text{trace of } A = 4$$

$$A_{11} = \begin{vmatrix} 3 & -4 \\ 1 & -1 \end{vmatrix} = -3 + 4 = 1$$

$$A_{22} = \begin{vmatrix} 2 & -2 \\ 1 & -1 \end{vmatrix} = -2 + 2 = 0$$

$$A_{33} = \begin{vmatrix} 2 & 1 \\ 2 & 3 \end{vmatrix} = 6 - 2 = 4$$

$$|A| = 2(1) - 1(2) - 2(4) = 2.$$

$$\text{Ch. poly} = t^3 - 4t^2 + 5t - 2 = (t-1)^2(t-2)$$

$$\text{Eigen values, } \lambda = 1, 1, 2$$

$$\text{Eigen vector for } \lambda_1 = 1$$

$$(A - \lambda_1 I)x = 0 \Rightarrow \begin{pmatrix} 1 & 1 & -2 \\ 2 & 2 & -4 \\ 1 & 1 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$R_3 \rightarrow R_3 - R_1, \quad R_2 \rightarrow R_2 - 2R_1$$

$$\begin{pmatrix} 1 & 1 & -2 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

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

$$x_1 = -x_2 + 2x_3$$

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -x_2 + 2x_3 \\ x_2 \\ x_3 \end{pmatrix} = x_2 \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} + x_3 \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

For $\lambda=2$,

$$\begin{pmatrix} 0 & 1 & -2 \\ 2 & 1 & -4 \\ 1 & 1 & -3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$R_2 \rightarrow R_2 - 2R_3$$

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

$$R_1 \rightarrow R_1 + R_2$$

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

$$-x_2 + 2x_3 = 0$$

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

$$x_2 = 2x_3$$

$$x_1 = 3x_3 - 2x_3 = x_3$$

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} +x_3 \\ 2x_3 \\ x_3 \end{pmatrix} = x_3 \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$$

Thus T is diagonalizable, because it has three independent eigenvectors.

$$\begin{aligned} S &= \{u_1, u_2, u_3\} \\ &= \left\{ \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} \right\} \end{aligned}$$

$$D = \text{dia}(1, 1, 2)$$

Ex ↗

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Diagonalizing Real Symmetric Matrices :-

There are many real matrices A that are not diagonalizable.

In fact, some real matrices may not have any (real) eigenvalues. However, if A is real symmetric matrix then these problem don't exist.

In fact, 1) If A is real symmetric matrix. Then each root λ of its characteristic polynomial is real.

2) A is real symmetric matrix. Suppose u & v are eigenvectors of A belonging to distinct eigenvalues λ_1, λ_2 . Then u & v are orthogonal.
i.e $\langle u, v \rangle = 0$.

1) & 2) gives the following fundamental result:-

Let A be real symmetric matrix. Then there exists an orthogonal matrix P such that $D = P^T A P$ is diagonal.

Remark: The orthogonal matrix P is obtained by normalizing the basis of orthogonal eigenvectors of A .

In such a case, we say A is "Orthogonally diagonalizable".

Ex :- Let $A = \begin{bmatrix} 2 & -2 \\ -2 & 5 \end{bmatrix}$, a real symmetric matrix.
 Find an orthogonal matrix P s.t P^TAP is diagonal.

Sol :- Characteristic polynomial of A is $\det(A - \lambda I)$

$$= \lambda^2 - 7\lambda + 6 = (\lambda - 6)(\lambda - 1)$$

Ch. eqn : $(\lambda - 6)(\lambda - 1) = 0$

Ch. roots $\Rightarrow \lambda = 6, \lambda = 1$

(a) Eigen vector (ch. vector) corresponding to $\lambda = 6$,

$$(A - 6I)x = 0$$

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

$$\Rightarrow \begin{cases} -4x_1 - 2x_2 = 0 \\ -2x_1 - x_2 = 0 \end{cases} \Rightarrow \begin{aligned} 2x_1 + x_2 &= 0 \\ x_2 &= -2x_1 \end{aligned}$$

$$\text{Eigen space} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ -2x_1 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$\mathcal{B}_1 = \left\{ \begin{bmatrix} 1 \\ -2 \end{bmatrix} \right\}$$

(b) Eigen vector corresponding to $\lambda = 1$

$$(A - I)x = 0 \Rightarrow \begin{bmatrix} 1 & -2 \\ -2 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$x_1 - 2x_2 = 0 \Rightarrow \boxed{x_1 = 2x_2}$$

$$\text{Eigenvector} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \pi_2 \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\text{Let } u_1 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}, \quad u_2 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}.$$

$$\text{As, we see } u_1 \cdot u_2 = 2 - 2 = 0$$

$\Rightarrow u_1 \text{ & } u_2 \text{ are orthogonal.}$

Normalizing u_1 & u_2 , we obtain orthonormal vector.
 $\|u_1\| = \sqrt{5}, \|u_2\| = \sqrt{5}$

$$\hat{u}_1 = \frac{u_1}{\|u_1\|} = \begin{bmatrix} 1/\sqrt{5} \\ -2/\sqrt{5} \end{bmatrix}, \quad \hat{u}_2 = \frac{u_2}{\|u_2\|} = \begin{bmatrix} 2/\sqrt{5} \\ 1/\sqrt{5} \end{bmatrix}$$

$$P = \begin{bmatrix} \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \\ -\frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{bmatrix}$$

$$P^T A P = \begin{bmatrix} 6 & 0 \\ 0 & 1 \end{bmatrix}$$

As expected, $P^T A P$ are the eigenvalues corresponding to the columns of P .

$$\text{Ex: } \det A = \begin{bmatrix} 11 & -8 & 4 \\ -8 & -1 & -2 \\ 4 & -2 & -4 \end{bmatrix}$$

$$\underline{\text{Sol}}^{\text{a})} \text{ Ch. poly} - \lambda^3 - 6\lambda^2 - 135\lambda - 400.$$

$$\text{trace } A = 6, \quad |A| = 400, \quad A_{11} = 0, \quad A_{22} = -60, \quad A_{33} = -75$$

$$\sum A_{ii} = -135$$

$$\text{ch. roots, } (\lambda + 5)^2 (\lambda - 16) = 0$$

$$\lambda = -5 \text{ (Repeated twice)}, 16.$$

b) Orthogonal Basis

$$\lambda = -5$$

$$A + 5I = \begin{bmatrix} 16 & -8 & 4 \\ -8 & 4 & -2 \\ 4 & -2 & 1 \end{bmatrix}$$

$$(A + 5I)x = 0 \Rightarrow 4x - 2y + z = 0.$$

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} -4x+2y \\ x \\ y \\ -4x+2y \end{bmatrix} = y \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} + x \begin{bmatrix} 1 \\ 0 \\ -4 \end{bmatrix}$$

↓
 v_1

$$v_2 = \begin{pmatrix} a \\ b \\ c \end{pmatrix}, t \quad 0 \cdot a + b + 2c = 0 \quad 4a - 2b + c = 0$$

$$b + 2c = 0$$

$$\text{let } c = 4, \quad b = -8, \quad a = -5$$

$$v_2 = \begin{bmatrix} -5 \\ -8 \\ 4 \end{bmatrix}$$

$$\lambda = 16$$

$$A - 16I = \begin{bmatrix} -5 & -8 & 4 \\ -8 & -17 & -2 \\ 4 & -2 & -20 \end{bmatrix}$$

$$R_2 \rightarrow 5R_2 - 8R_1, \quad R_3 \rightarrow 5R_3 + 4R_1$$

$$\begin{bmatrix} -5 & -8 & 4 \\ 0 & -21 & -42 \\ 0 & -42 & 84 \end{bmatrix}$$

$$\sim R_3 \rightarrow \frac{1}{2}R_3 - R_2 \begin{bmatrix} -5 & -8 & 4 \\ 0 & -1 & -2 \\ 0 & 0 & 0 \end{bmatrix}$$

$$R_2 \rightarrow \frac{1}{2}R_2$$

$$v_3 = \begin{bmatrix} 4 \\ -2 \\ 1 \end{bmatrix}$$

$$\begin{aligned} -x_2 - 2x_3 &= 0 \Rightarrow x_2 = -2x_3 \\ -5x_1 - 8x_2 + 4x_3 &= 0 \\ -5x_1 + 16x_3 + 4x_3 &= 0 \\ x_1 &= 4x_3 \end{aligned}$$

Then $\{v_1, v_2, v_3\}$ form a maximal set of non-zero orthogonal eigenvectors of B .

(c) Normalizing v_1, v_2, v_3 to obtain the orthonormal basis:

$$\hat{v}_1 = \frac{v_1}{\sqrt{5}}, \quad \hat{v}_2 = \frac{v_2}{\sqrt{105}}, \quad \hat{v}_3 = \frac{v_3}{\sqrt{21}}$$

$$P = \begin{bmatrix} 0 & -5/\sqrt{105} & 4/\sqrt{21} \\ \sqrt{5} & -8/\sqrt{105} & -2/\sqrt{21} \\ 2/\sqrt{5} & 4/\sqrt{105} & \sqrt{21} \end{bmatrix}$$

$$D = P^{-1} A P = \begin{bmatrix} -5 & 0 & 0 \\ 0 & -5 & 0 \\ 0 & 0 & 16 \end{bmatrix}$$