# LINEAR ALGEBRA II Chapter one Characterístíc Equatíon

# 1.1. *Eígen values and Eígen vectors*

**Definition1.1.1**: Let A be an  $n \times n$  matrix. A Number  $\lambda$  is called an Eigen

value of A if there exists a non zero vector  $v \in F^n$  such that  $Av = \lambda v$ .

The vector v is then called an eigenvector of A corresponding to the Eigen value  $\lambda$ .

**Example:** Let 
$$A = \begin{bmatrix} 1 & 2 \\ 4 & 3 \end{bmatrix}$$
. Then taking  $\lambda = 5$  and  $v = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,

# Cont...

- we have  $Av = \begin{bmatrix} 1 & 2 \\ 4 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 5 \\ 10 \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \lambda v$  so 5 is an eigenvalue of A and  $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$  is an eigenvector of A corresponding to the eigenvalue 5
- **Definition1.1.2**: Let A be an  $n \times n$  matrix. The set of all eigenvalues of A is called the spectrum of A.
- **Theorem 1.1.1**: Let *A* be an  $n \times n$  matrix. A Number  $\lambda$  is an Eigen value of *A* if and only if  $det(A \lambda I) = 0$ , where *I* denotes the  $n \times n$  identity matrix.

# **Properties of Eigen Values**

- i. The sum of the eigen values of a matrix is the sum of the elements of the principal diagonal.
- ii. The product of the eigen values of a matrix A is equal to its determinant.
- iii. If  $\lambda$  is an eigen value of a matrix A, then  $1/\lambda$  is the eigen value of A<sup>-1</sup>
- *iv.* If  $\lambda$  is an eigen value of an orthogonal matrix, then  $1/\lambda$  is also its eigen value.
- v. The eigen values of a triangular matrix are precisely the entries of diagonal entries.

## Cont...

- vi. If  $\lambda_1, \lambda_2, ..., \lambda_n$  are the eigenvalues of A, then
- a.  $k\lambda_1, k\lambda_2, ..., k\lambda_n$  are the eigenvalues of the matrix kA, where k is a non zero scalar.

b.  $\frac{1}{\lambda_1}, \frac{1}{\lambda_2}, \dots, \frac{1}{\lambda_n}$  are the eigenvalues of the inverse matrix  $A^{-1}$ .

c.  $\lambda_1^{p}, \lambda_2^{p}...\lambda_n^{p}$  are the eigenvalues of  $A^{P}$ , where p is any positive integer.

# **1.2 Characteristic polynomial**

- **Definition 1.2.1:**Polynomial of degree n in x is an expression of the form  $a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$ , where  $a_0, a_{1,\dots} a_n \in F$  and  $a_n \neq 0$ , n is non negative integer.
- We can write  $p(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$ , if we replace x every where by a given number  $\lambda$  and we obtain  $p(\lambda) = a_n \lambda^n + a_{n-1} \lambda^{n-1} + \dots + a_1 \lambda + a_0$
- Definition 1.2.2: Let p(x) be a polynomial in x. A number  $\lambda$  is called a root of p(x) if  $p(\lambda) = 0$ .

#### Cont....

• **Definition 1.2.3**: Let  $A = (a_{ij})$  be an  $n \times n$  matrix. Then the polynomial

$$det(A - \lambda I) = det \begin{bmatrix} a_{11} - \lambda & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} - \lambda & \cdots & a_{21} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} - \lambda \end{bmatrix}$$
 is called the

characteristic polynomial of A.

• Denoted as  $C_A(x)$ , the leading coefficient is  $(-1)^n$  and the constant term is *detA*.

# Cont...

- The equation  $det(A \lambda I) = 0$  or equivalently  $det(\lambda I A) = 0$  is called characteristic equation of A.
- **Definition 1.2.4**: If an eigenvalue  $\lambda$  occur k times as a root of the characteristic polynomial  $C_A(x)$ , then k is called the multiplicity of the eigenvalue .

• Example 1: Let 
$$A = \begin{bmatrix} 3 & -1 & -1 \\ -12 & 0 & 5 \\ 4 & -2 & -1 \end{bmatrix}$$
 then,

- i. Find characteristic polynomial and characteristic equation of A.
- ii. Eigenvalues and eigenvectors of A.
- iii. Eigen values of  $A^{-1}$
- iv. Eigen values of  $A^{10}$
- v. Eigen values of 10A.

#### **Solution:**

Characteristic polynomial is

$$det(A - \lambda I) = \begin{vmatrix} 3 & -1 & -1 \\ -12 & 0 & 5 \\ 4 & -2 & -1 \end{vmatrix} - \lambda \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{vmatrix}$$

$$= \begin{vmatrix} 3-\lambda & -1 & -1 \\ -12 & -\lambda & 5 \\ 4 & -2 & -1-\lambda \end{vmatrix} \\ = (3-\lambda)[(\lambda^2+\lambda)+10] + 1[(12\lambda+12)-20] - 1(24+4\lambda) \\ \text{characteristic polynomial is } -\lambda^3+2\lambda^2+\lambda-2 \text{ and characteristic} \\ \text{equation of A is } -\lambda^3+2\lambda^2+\lambda-2=0 \end{vmatrix}$$

ii. To find Eigenvalues of A we have to find root of

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

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

$$\lambda = -1 \text{ or } \lambda = 1 \text{ or } \lambda = 2$$

□So the eigenvalues are -1,1 and 2. □To find eigenvector corresponding to eigenvalues □write  $Av = \lambda v$ 

$$\Rightarrow$$
  $(A - \lambda I)v = 0$  that is

$$\Rightarrow \begin{bmatrix} 3-\lambda & -1 & -1 \\ -12 & -\lambda & 5 \\ 4 & -2 & -1-\lambda \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$
  
Then we can solve for  $x_1$ ,  $x_2$  and  $x_3$  by appropriate method.  
Let  $\lambda = -1$ , then we have

$$\begin{bmatrix} 4 & -1 & -1 \\ -12 & 1 & 5 \\ 4 & -2 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Solve by using Gaussian elimination .

Since the constant matrix is zero matrix, reduce the coefficient matrix to row echelon form

$$\begin{bmatrix} 4 & -1 & -1 \\ -12 & 1 & 5 \\ 4 & -2 & 0 \end{bmatrix} \xrightarrow{R_2 + 3R_1} \begin{bmatrix} 4 & -1 & -1 \\ 0 & -2 & 2 \\ 0 & -1 & 1 \end{bmatrix} \xrightarrow{R_3 - 1/2R_2} \begin{bmatrix} 4 & -1 & -1 \\ 0 & -2 & 2 \\ 0 & 0 & 0 \end{bmatrix}.$$

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

Thus the linear system becomes

 $4x_{1} - x_{2} - x_{3} = 0$ -2x<sub>2</sub> + 2x<sub>3</sub> = 0  $\Rightarrow x_{2} = x_{3}, x_{1} = \frac{1}{2}x_{3}$  $\Box \text{ Let } x_{3} = t \text{ then } x_{2} = t \text{ and } x_{1} = \frac{1}{2}t \text{ where t is arbitrary}$ 

Thus 
$$\mathbf{v} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \begin{bmatrix} 1/2 & t \\ t \\ t \end{bmatrix} = t \begin{bmatrix} 1/2 \\ 1 \\ 1 \end{bmatrix}$$
  
So  $\begin{bmatrix} 1/2 \\ 1 \\ 1 \end{bmatrix}$  is eigenvector corresponding to eigenvalue  $\lambda = -1$   
 $\Box$  Similarly, we obtain  $\begin{bmatrix} 3 \\ -1 \\ 7 \end{bmatrix}$  as an eigenvector corresponding to eigenvalue  $\lambda = 1$  as an eigenvector corresponding to eigenvalue 2.

- iii. Eigen values of  $A^{-1}$ . 1, -1 and  $\frac{1}{2}$  are eigenvalues of  $A^{-1}$  by the property of eigenvalue.
- iv. Eigen values of  $A^{10}$  are -1, 1 and  $2^{10}$
- v. Eigen values of 10A are -10,10 and 20 **Example 2**: Let  $A = \begin{bmatrix} 0 & 0 & -2 \\ 1 & 2 & 1 \\ 1 & 0 & 3 \end{bmatrix}$ Solution:
- The characteristic equation of matrix A is  $\lambda^3 5\lambda^2 + 8\lambda 4 = 0$ , or, in factored form  $(\lambda 1)(\lambda 2)^2 = 0$

• Thus the eigenvalues of *A* are  $\lambda = 1$  and  $\lambda_{2,3} = 2$ , so there are two eigenvalues of *A*.

 $v = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$ , is an eigenvector of *A* corresponding to  $\lambda$  if and only *v* is a nontrivial solution of  $(A - \lambda I)v = 0$ 

that is, of the form  $\begin{bmatrix}
-\lambda & 0 & -2 \\
1 & 2-\lambda & 1 \\
1 & 0 & 3-\lambda
\end{bmatrix}
\begin{bmatrix}
x_1 \\
x_2 \\
x_3
\end{bmatrix} = \begin{bmatrix}
0 \\
0 \\
0
\end{bmatrix}$ 

#### cont.

• If 
$$\lambda = 1$$
, then  $\begin{bmatrix} -1 & 0 & -2 \\ 1 & 1 & 1 \\ 1 & 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$ 

□Solving this system using Gaussian elimination yields  $x_1 = -2s, x_2 = s, x_3 = s$  (verify)

Thus the eigenvectors corresponding to  $\lambda = 1$  are the nonzero vectors of the form

$$\begin{bmatrix} -2s \\ s \\ s \end{bmatrix} = s \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$$
 so that 
$$\begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$$
 is the Eigen vector corresponding to eigenvalue 1

• If 
$$\lambda = 2$$
, then it becomes  $\begin{bmatrix} -2 & 0 & -2 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$ 

• Solving this system using Gaussian elimination yields

$$x_1 = -s, x_2 = t, x_3 = s$$
(verify)

- Thus, the eigenvectors of *A* corresponding to  $\lambda = 2$  are the nonzero vectors of the form
- Thus, the eigenvectors of *A* corresponding to  $\lambda = 2$  are the nonzero vectors of the form

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -s \\ t \\ s \end{bmatrix} = s \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} + t \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$
by shimelis Ayele

#### Cont...

 $\Box \operatorname{Since} \begin{bmatrix} -1\\0\\1\\0 \end{bmatrix} \operatorname{and} \begin{bmatrix} 0\\1\\0 \end{bmatrix} \text{ are linearly independent, these vectors form a basis for the Eigen space corresponding to } \lambda = 2.$  $\Box \operatorname{So the Eigen vectors of A corresponding Eigen values 1 and 2 are \begin{bmatrix} -2\\1\\1 \end{bmatrix}, \begin{bmatrix} -1\\0\\1 \end{bmatrix} \operatorname{and} \begin{bmatrix} 0\\1\\0 \end{bmatrix}$ Exercise 1:

1. Find eigenvalues and eigenvectors of  $\begin{bmatrix} 1 & 1 & 2 & 3 \end{bmatrix}$ 

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

2. Let 
$$A = \begin{bmatrix} 3 & -2 & 2 \\ 4 & -4 & 6 \\ 2 & -3 & 5 \end{bmatrix}$$
, then find

- a. Find characteristic polynomial and characteristic equation of A.
- b. Eigenvalues and eigenvectors of A.
- c. Eigen values of  $A^{-1}$
- d. Eigen values of  $A^{11}$

e.

Eigen values of 
$$\begin{bmatrix} 30 & -20 & 20 \\ 40 & -40 & 60 \\ 20 & -30 & 50 \end{bmatrix}$$
.

# **1.3.Similarity of matrix**

- **Definition1.3.1:** If A and B are n xn matrices, then A is similar to B if there is an invertible matrix P such that  $P^{-1}AP = B$ , or, equivalently  $A = PBP^{-1}$
- Changing A into B is called a similarity transformation.
- **Theorem 1.3.1:** If nxn matrices A and B are similar, then they have the same characteristic polynomial and hence the same eigenvalues (with the same multiplicities).

#### **Proof:**

□Suppose A and B are similar matrices of order n.

The characteristic polynomial of B is given by  

$$det(B - \lambda I)$$

$$= det(P^{-1}AP - \lambda I)$$

$$= det(p^{-1}(A - \lambda I)p)$$

$$= det(p^{-1})det(A - \lambda I)det(p)$$

$$= det(p^{-1}p) det(A - \lambda I)$$

$$= det(A - \lambda I)$$

**Definition 1.3.2**: The set of *distinct* eigenvalues, denoted by  $\sigma(A)$ , is called the *spectrum* of **A**.

**Definition 1.3.3:**For square matrices A, the number  $\rho(A) = \max |\lambda|, \lambda \in \sigma(A)$  is called the *spectral radius* of A.

# 1.4. Diagonalization

- **Definition1.4.1**: A square matrix A is called *diagonalizable* if there exist an invertible matrix P such that  $P^{-1}AP$  is a diagonal matrix; the matrix P is said to *diagonalize* A.
- **THEOREM 1.4.1** : *If*  $P_1$ ,  $P_2$ , ...,  $P_k$  are Eigen vectors of A corresponding to distinct eigenvalues
- , $\lambda_1$ ,  $\lambda_2$ , ..., $\lambda_k$  then {P<sub>1</sub>, P<sub>2</sub>, ..., P<sub>k</sub> } is a linearly independent set. • **Proof**
- $\Box$  Let ,  $P_1$  ,  $P_2$ , ...,  $P_k$  be eigenvectors of A corresponding to distinct eigenvalues  $\lambda_1$ ,  $\lambda_2$ , ...,  $\lambda_k$ .

- We shall assume that,  $P_1$ ,  $P_2$ , ...,  $P_k$  are linearly dependent and obtain a contradiction. We can then conclude that,  $P_1$ ,  $P_2$ , ...,  $P_k$  are linearly independent.
- Since an eigenvector is nonzero by definition,  $\{p_1\}$  is linearly independent. Let *r* be the largest integer such that  $P_1$ ,  $P_2$ , ...,  $P_r$ is linearly independent. Since we are assuming that  $P_1$ ,  $P_2$ , ...,  $P_k$  is linearly dependent, *r* satisfies  $1 \le r \le k$ .
- $\Box$  More over, by definition of r,  $P_1$ ,  $P_2$ , ...,  $P_{r+1}$  is linearly dependent.

 $\Box$ Thus there are scalars  $c_1$ ,  $c_2$ , ...,  $c_{r+1}$ , not all zero, such that  $P_1c_1 + P_2c_2 + \dots, P_{r+1}c_{r+1} = 0$ (4)Multiplying both sides of 4 by A and using  $AP_1 = \lambda_1 p_1, AP_2 = \lambda_2 P_2 \dots, AP_n = \lambda_{r+1} P_{r+1}$  $\Box \text{ we obtain } c_1 \lambda_1 p_1 + c_2 \lambda_2 p_2 + \dots + c_{r+1} \lambda_{r+1} p_{r+1} = 0 \dots$ (5) $\Box$ Multiplying both sides of 4 by  $\lambda_{r+1}$  and subtracting the resulting equation from 5 yields

 $\Box c_{1}(\lambda_{1} - \lambda_{r+1})p_{1} + c_{2}(\lambda_{2} - \lambda_{r+1})p_{2} + ... + c_{r}(\lambda_{r} - \lambda_{r+1})p_{r} = 0$ 

• Since {P<sub>1</sub>, P<sub>2</sub>, ..., P<sub>r</sub>} is a linearly independent set, this equation implies that  $c_1(\lambda_1 - \lambda_{r+1}) + c_2(\lambda_2 - \lambda_{r+1}) + ... + c_r(\lambda_r - \lambda_{r+1}) = 0$  and since ,  $\lambda_1$ ,  $\lambda_2$ , ...,  $\lambda_r$  are distinct by hypothesis, it follows that

• 
$$c_1 = c_2 = \dots = c_r = 0$$
 (6)

Substituting these values in 4 yields

•  $P_{r+1}c_{n+1} = 0$  (7)

 $\Box$ Since the eigenvector  $P_{r+1}$  is nonzero, it follows that  $c_{r+1} = 0$ 

 $\Box$ Equations 6 and 7 contradict the fact that  $c_1$ ,  $c_2$ , ...,  $c_{r+1}$  are not all zero; this completes the proof

**Theorem 1.4.2:** If A is an  $n \times n$  matrix, then the following are equivalent.

 $\checkmark$ a. A is diagonalizable.

 $\checkmark$ b. A has n linearly independent eigenvectors.

• Proof

 $\Box a \Rightarrow b$  Since A is assumed diagonalizable, there is an invertible matrix

$$\begin{bmatrix} p_{11} & p_{12} & \cdots & p_{1n} \\ p_{21} & p_{22} & \cdots & p_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ p_{n1} & p_{n2} & \cdots & p_{nn} \end{bmatrix}$$
  
such that  $P^{-1}AP$  is diagonal, say  $P^{-1}AP = D$ ,

where 
$$D = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix}$$
  
  
It follows from the formula  $P^{-1}AP = D$  that  $AP = PD$ ; that is,  

$$\Box AP = \begin{bmatrix} p_{11} & p_{12} & \cdots & p_{1n} \\ p_{21} & p_{22} & \cdots & p_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ p_{n1} & p_{n2} & \cdots & p_{nn} \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 p_{11} & \lambda_2 p_{12} & \cdots & \lambda_n p_{1n} \\ \lambda_1 p_{21} & \lambda_2 p_{22} & \cdots & \lambda_n p_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ \lambda_1 p_{n1} & \lambda_2 p_{n2} & \cdots & \lambda_n p_{nn} \end{bmatrix}$$

□ If we denote the column vectors of *P*, by  $P_1, P_2, ..., P_n$ , then from 1, the successive columns of PD are  $\lambda_1 P_1$ ,  $\lambda_2 P_2$ , ...,  $\lambda_n P_n$ .

The successive columns of AP are  $AP_1$ ,  $AP_2$ , ...,  $AP_n$ .

□Thus we must have

$$AP_1 = \lambda_1 p_1, AP_2 = \lambda_2 P_2 \dots, AP_n = \lambda_n P_n$$
<sup>2</sup>

#### Cont....

• Since *P* is invertible, its column vectors are all nonzero; thus, it follows from 2 that  $\lambda_1, \lambda_2, ..., \lambda_n$  are eigenvalues of *A*, and  $P_1$ ,  $P_2$ , ...,  $P_n$  are corresponding eigenvectors.

Since *P* is invertible,  $P_1, P_2, ..., P_n$  are linearly independent.  $\Box$  Thus *A* has *n* linearly independent eigenvectors.

 $\Box b \Rightarrow a$  Assume that A has n linearly independent eigenvectors  $P_1$ ,  $P_2$ , ...,  $P_n$  with corresponding eigenvalues,  $\lambda_1$ ,  $\lambda_2$ , ...,  $\lambda_n$ 

#### Cont....

• Let 
$$P = \begin{bmatrix} p_{11} & p_{12} & \cdots & p_{1n} \\ p_{21} & p_{22} & \cdots & p_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ p_{n1} & p_{n2} & \cdots & p_{nn} \end{bmatrix}$$
 be the matrix whose column vectors are

 $P_1$ ,  $P_2$ , ...,  $P_n$ 

• The column vectors of the product AP are  $AP_1$ ,  $AP_2$ , ...,  $AP_n$ 

$$\Box$$
But,  $AP_1 = \lambda_1 p_1$ ,  $AP_2 = \lambda_2 P_2 \dots$ ,  $AP_n = \lambda_n P_n$  Why?

$$AP = \begin{bmatrix} \lambda_1 p_{11} & \lambda_2 p_{12} & \cdots & \lambda_n p_{1n} \\ \lambda_1 p_{21} & \lambda_2 p_{22} & \cdots & \lambda_2 p_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ \lambda_1 p_{n1} & \lambda_2 p_{n2} & \cdots & \lambda_2 p_{nn} \end{bmatrix}$$

$$= \begin{bmatrix} p_{11} & p_{12} & \cdots & p_{1n} \\ p_{21} & p_{22} & \cdots & p_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ p_{n1} & p_{n2} & \cdots & p_{nn} \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} = PD$$
 3

- Where *D* is the diagonal matrix having the eigenvalues,  $\lambda_1$ ,  $\lambda_2$ , ...,  $\lambda_n$  on the main diagonal.
- Since the column vectors of *P* are linearly independent, *P* is invertible. Thus 3 can be rewritten as  $P^{-1}AP = D$ ; that is, *A* is diagonalizable
- **THEOREM 1.4.3**: If an  $n \times n$  matrix A has n distinct eigenvalues, $\lambda_1$ ,  $\lambda_2$ , ..., $\lambda_n$  then A is diagonalizable
- **Proof:** If  $P_1$ ,  $P_2$ , ...,  $P_n$  are eigenvectors corresponding to the distinct eigenvalues , $\lambda_1$ ,  $\lambda_2$ , ..., $\lambda_n$

then by Theorem ,  $P_1$  ,  $P_2$ , ...,  $P_n$  are linearly independent.

• Thus A is diagonalizable by Theorem .

# Procedure to diagonalize a matrix

- $\checkmark$  Find the linearly independent eigenvectors
- $\checkmark$  Construct *P* from eigenvectors
- ✓ Construct  $D = P^{-1}AP$  from eigenvalues
- Example 1: Finding a Matrix *P* that diagonalize a matrix *A*

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

- Solution
- The characteristic equation of A is  $(\lambda 1)(\lambda 2)^2 = 0$  and we found the following bases for the Eigen spaces:

• 
$$\lambda = 2$$
:  $p_1 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ ,  $p_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$  and  $\lambda = 1$ :  $p_3 = \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$ 

• There are three basis vectors in total, so the matrix *A* is diagonalizable and

$$P = \begin{bmatrix} -1 & 0 & -2 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix}$$
 diagonalize A.

Verify that  $P^{-1}AP =$ 

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

• EXAMPLE 2:

• Find a matrix *P* that diagonalizable 
$$A = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 2 & 0 \\ -3 & 5 & 2 \end{bmatrix}$$

#### Solution

• The characteristic polynomial of A is

$$\Box \det(A - \lambda I) = \begin{pmatrix} 1 - \lambda & 0 & 0 \\ 1 & 2 - \lambda & 0 \\ -3 & 5 & 2 - \lambda \end{pmatrix}$$

 $= (\lambda - 1)(\lambda - 2)^2$ Thus the eigenvalues of A are  $\lambda_1 = 1$  and  $\lambda_{2,3} = 2$ .

• Verify that eigenvalues are corresponding to

$$\lambda_1 = 1 \text{ is } p_1 = \begin{bmatrix} -1 \\ 1 \\ -8 \end{bmatrix} \text{ and corresponding to } \lambda_{2,3} = 2 \text{ is } p_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

• Since *A* is a 3 × 3 matrix and there are only two basis vectors in total, *A* is not diagonalizable by theorem.

## **1.5.Cayley-Hamilton Theorem**

Theorem: If  $C_A(x)$  is characteristic polynomial of A, then  $C_A(x) = 0$ 

• Let 
$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$
 then characteristic polynomial of A

$$C_{A}(x) = \det(A - \lambda I) = \det \begin{bmatrix} a_{11} - \lambda & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - \lambda & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} - \lambda \end{bmatrix}$$

The characteristic equation is given by
$$C_A(x) = x^n + a_{n-1}x^{n-1} + a_{n-2}x^{n-2} + \dots + a_1x^1 + a_0 = 0$$
•  $C_A(x) = A^n + a_{n-1}A^{n-1} + a_{n-2}A^{n-2} + \dots + a_1A^1 + Ia_0 = 0$ 
• **Theorem**: Let A be a non singular n × n matrix, and let its characteristic polynomial be  $C_A(x) = A^n + a_{n-1}A^{n-1} + a_{n-2}A^{n-2} + \dots + a_1A^1 + Ia_0 = 0$ , then  $A^{-1} = -\frac{1}{a_0}(A^{n-1} + a_{n-1}A^{n-2} + \dots + Ia_1)$ 

- <u>Proof</u>:
- By Cayley-Hamilton theorem  $A^n + a_{n-1}A^{n-1} + a_{n-2}A^{n-2} + \dots + a_1A^1 + Ia_0 = 0$

- $a_0 = (-1)^n detA \neq 0$ , Since A is non singular.
- So the above equation can be written as

$$I = -\frac{1}{a_0} (A^n + a_{n-1}A^{n-1} + a_{n-2}A^{n-2} + \dots + a_1A^1)$$
$$= -\frac{1}{a_0} (A^{n-1} + a_{n-1}A^{n-2} + \dots + Ia_1)A$$

$$\Box A^{-1} = -\frac{1}{a_0} \left( A^{n-1} + a_{n-1} A^{n-2} + \dots + I a_1 \right)$$

#### Cont....

Example 1:Let A = 
$$\begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \\ 1 & -1 & 2 \end{bmatrix}$$

Verify Cayley -Hamilton theorem and compute  $A^{-1}$ 

- <u>Solution</u>
- The characteristic equation of A is  $|A \lambda I| = 0$  $\begin{vmatrix} 2 - \lambda & -1 & 1 \\ 1 & 2 & 2 \end{vmatrix}$

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

$$\Rightarrow -(\lambda^3 - 6\lambda^2 + 9\lambda - 4) = 0$$

To verify Cayley – Hamilton theorem, we have to show that  $A^3 - 6A^2 + 9A - 4I = 0$ 

$$\Box A^{2} = \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \\ 1 & -1 & 2 \end{bmatrix} \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \\ 1 & -1 & 2 \end{bmatrix} = \begin{bmatrix} 6 & -5 & 5 \\ -5 & 6 & -5 \\ 5 & -5 & 6 \end{bmatrix}$$

•  $A^3 = A^2 A = \begin{bmatrix} 6 & -5 & 5 \\ -5 & 6 & -5 \\ 5 & -5 & 6 \end{bmatrix} \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \\ 1 & -1 & 2 \end{bmatrix} = \begin{bmatrix} 22 & -21 & 21 \\ -21 & 22 & -21 \\ 22 & -21 & 22 \end{bmatrix}$ 

• Therefore  $A^3 - 6A^2 + 9A - 4I$ 

$$\checkmark = \begin{bmatrix} 22 & -21 & 21 \\ -21 & 22 & -21 \\ 22 & -21 & 22 \end{bmatrix} - 6 \begin{bmatrix} 6 & -5 & 5 \\ -5 & 6 & -5 \\ 5 & -5 & 6 \end{bmatrix} + 9 \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \\ 1 & -1 & 2 \end{bmatrix} - 6 \begin{bmatrix} 6 & -5 & 5 \\ -5 & 6 & -5 \\ 1 & -1 & 2 \end{bmatrix}$$

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

• Now multiplying both side of (1) by  $A^{-1}$  yields

• 
$$A^2 - 6A^1 + 9I - 4A^{-1} = 0$$

•  $4A^{-1} = (A^2 - 6A^1 + 9I)$ 

$$4A^{-1} = \begin{bmatrix} 6 & -5 & 5 \\ -5 & 6 & -5 \\ 5 & -5 & 6 \end{bmatrix} - 6\begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \\ 1 & -1 & 2 \end{bmatrix} + 9\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

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

• Example 2: given A =  $\begin{bmatrix} 1 & 2 & -1 \\ 0 & 1 & -1 \\ 3 & -1 & 1 \end{bmatrix}$  find AdjA by using Cayley-Hamilton theorem.

• Solution

- The characteristic equation of a matrix A is •  $|A - \lambda I| = 0$  i.e.,  $\begin{vmatrix} 1 - \lambda & 2 & -1 \\ 0 & 1 - \lambda & -1 \\ 3 & -1 & 1 - \lambda \end{vmatrix} = 0$ •  $\Rightarrow \lambda^3 - 3\lambda^2 + 5\lambda + 3 = 0$
- by Cayley-Hamilton theorem  $A^3 3A^2 + 5A + 3I = 0$
- Multiply both side by  $A^{-1}$ , we get  $A^2 3A + 5I + 3A^{-1} = 0$

• We know that , 
$$A^{-1} = \frac{adjA}{|A|}$$
  
• Therefore  $adjA = |A|A^{-1} = -3 \begin{bmatrix} 0 & \frac{1}{3} & \frac{1}{3} \\ 1 & \frac{-4}{3} & -\frac{1}{3} \\ 1 & -\frac{7}{3} & -\frac{1}{3} \end{bmatrix}$   
 $= \begin{bmatrix} 0 & -1 & -1 \\ 1 & 4 & 1 \\ 1 & 7 & 1 \end{bmatrix}$ 

• Another application of Cayley-Hamilton theorem is to compute power of square matrix *A* 

• Example: Let 
$$A = \begin{bmatrix} 2 & 5 \\ 1 & -2 \end{bmatrix}$$
 and then find  $A^{735}$   
• Solution

- The characteristic polynomial of A is  $x^2 9$ . Eigen values are -3,3.
- Division algorithm applied to the polynomial  $x^{735}$ ,  $x^2 9$  will give equation of the form  $x^{735} = (x^2 - 9)q(x) + (a_0 + a_1x)$ , (1) Where  $(a_0 + a_1x)$  is remainder obtained by dividing  $x^{735}$  by  $x^2 - 9$ .

- Note that the degree of remainder is less than the degree of the divisor  $x^2 9$ .
- By Cayley-Hamilton theorem  $A^2 9I = 0$ .
- Inserting A for x in equation (1), we get
- $A^{735} = a_0 + a_1 A$  (2)
- Inserting eigen values 3 and -3 for x successively in equation (1), we get
- $3^{735} = a_0 + 3a_1$

## Cont....

• 
$$(-3)^{735} = a_0 - 3a_1$$

• This gives 
$$a_0 = 0, a_1 = 3^{734}$$
.

- Then  $A^{735} = a_0 + a_1 A$
- gives
- $A^{735} = 3^{735}A$ .

# 1.6.Minimal polynomial

- **Definition 1.6.1**: The minimal polynomial of a matrix A, denoted by  $m_A(\lambda)$  is the unique monic polynomial of least degree such that  $m_A(\lambda) = 0$ .
- Theorem: a scalar  $\lambda$  is Eigen values of a matrix A if and only if is root of minimal polynomial.
- proof(exercise)
- Example: find minimal polynomial  $m_A(\lambda)$  of  $A = \begin{bmatrix} 2 & 2 & -5 \\ 3 & 7 & -15 \\ 1 & 2 & -4 \end{bmatrix}$

#### • <u>Solution</u>

• First find the characteristic polynomial of A.

## Exercise 3

• Let 
$$A = \begin{bmatrix} 1 & 2 & -1 \\ 0 & 1 & -1 \\ 3 & -1 & 1 \end{bmatrix}$$
, then

- 1. Verify Cayley-Hamilton theorem
- 2. Find  $A^{-1}$  and AdjA by using Cayley-Hamilton theorem.

 $p(\lambda) = -\lambda^{3} + 5\lambda^{2} - 7\lambda + 3 = -(\lambda - 1)^{2}(\lambda - 3).$ The minimal polynomial must divide characteristic polynomial. Thus minimal polynomial is exactly one of the following  $p(\lambda) = (\lambda - 3)(\lambda - 1)^{2} \text{ or } g(\lambda) = (\lambda - 3)(\lambda - 1)$  $g(A) = (A - I)(A - 3I) = \begin{pmatrix} 1 & 2 & -5 \\ 3 & 6 & -15 \\ 1 & 2 & -5 \end{pmatrix} \begin{pmatrix} -1 & 2 & -5 \\ 3 & 4 & -15 \\ 1 & 2 & -7 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$ 

•

Thus  $g(t) = (\lambda - 1)(\lambda - 3) = \lambda^2 - 4\lambda + 3$  is a minimal polynomial of A

**Exercise 3** : Find minimal polynomial m ( $\lambda$ ) of

*i.* 
$$A = \begin{bmatrix} 3 & -2 & 2 \\ 4 & -4 & 6 \\ 2 & -3 & 5 \end{bmatrix}$$

$$i. \quad B = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$