

Test number 01 (Solution)

Ex 01:

[a] We compute the characteristic polynomial of the matrix:

$$A_3 = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

From definition, we have:

$$P_{A_3}(\lambda) = \det(A - \lambda I_3)$$

$$= \begin{vmatrix} 1-\lambda & 1 & 1 \\ 1 & 1-\lambda & 1 \\ 1 & 1 & 1-\lambda \end{vmatrix} \begin{array}{l} C_1 \rightarrow C_1 - C_2 \\ C_2 \rightarrow C_2 - C_3 \end{array}$$

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

$$= \lambda^2 \begin{vmatrix} \overset{+}{-1} & \overset{-}{0} & \overset{+}{1} \\ 1 & -1 & 1 \\ 0 & 1 & 1-\lambda \end{vmatrix}$$

$$= \lambda^2 [-(-1 + \lambda - 1) + 1] = \lambda^2 (3 - \lambda).$$

Hence, $f_{A_3}(x) = x^2(3-x)$.

[b] We deduce the characteristic polynomial of the matrix :

$$A_n = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \in M_n(\mathbb{R})$$

First, we compute the characteristic polynomial of the matrix A_4 . In fact, we have :

$$\begin{aligned} f_{A_4}(x) &= \begin{vmatrix} 1-x & 1 & 1 & 1 \\ 1 & 1-x & 1 & 1 \\ 1 & 1 & 1-x & 1 \\ 1 & 1 & 1 & 1-x \end{vmatrix} \\ &= \begin{vmatrix} -x & 0 & 0 & 1 \\ x & -x & 0 & 1 \\ 0 & x & -x & 1 \\ 0 & 0 & x & 1-x \end{vmatrix} \end{aligned}$$

$$= x^3 \cdot \begin{vmatrix} \overset{+}{-1} & \overset{-}{0} & \overset{+}{0} & \overset{-}{1} \\ 1 & -1 & 0 & 1 \\ 0 & 1 & -1 & 1 \\ 0 & 0 & 1 & 1-x \end{vmatrix}$$

$$= x^3 \cdot (-1) \cdot \begin{vmatrix} \overset{+}{-1} & \overset{-}{0} & \overset{+}{1} \\ 1 & -1 & 1 \\ 0 & 1 & 1-x \end{vmatrix} + x^3 \cdot (-1) \cdot \begin{vmatrix} \overset{+}{1} & \overset{-}{-1} & \overset{+}{0} \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{vmatrix}$$

$$= -x^3 \left[-(-1+x-1)+1 \right] - x^3 \left[1 \right]$$

$$= -x^3(3-x) - x^3 = x^3(x-4)$$

Thus,

$$f_{A_4}(x) = x^3(x-4).$$

Conclusion: We deduce that:

$$f_{A_n}(x) = \begin{cases} x^{n-1}(x-n), & \text{if } n \text{ is even} \\ x^{n-1}(n-x), & \text{if } n \text{ is odd} \end{cases}$$

even = pair

odd = impair

Ex 02: We find the characteristic polynomial of the matrix:

$$A = \begin{bmatrix} 7 & -6 & -2 \\ 2 & 0 & -1 \\ 2 & -3 & 2 \end{bmatrix}$$

By definition, we have:

$$f_A(x) = \begin{vmatrix} 7-x & -6 & -2 \\ 2 & -x & -1 \\ 2 & -3 & 2-x \end{vmatrix} \begin{array}{l} L_1 \\ \downarrow \\ L_1 - 2L_2 \end{array}$$

$$= \begin{vmatrix} 3-x & -2(3-x) & 0 \\ 2 & -x & -1 \\ 2 & -3 & 2-x \end{vmatrix}$$

$$= (x-3) \cdot \begin{vmatrix} -1 & 2 & 0 \\ 2 & -x & -1 \\ 2 & -3 & 2-x \end{vmatrix}$$

$$L_2 \rightarrow (-1)L_2 + L_3$$

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$$= \begin{vmatrix} 3-x & -2(3-x) & 0 \\ 2 & -x & -1 \\ 2 & -3 & 2-x \end{vmatrix}$$

$$= (x-3) \cdot \begin{vmatrix} -1 & 2 & 0 \\ 2 & -x & -1 \\ 2 & -3 & 2-x \end{vmatrix}$$

$$L_2 \rightarrow (-1)L_2 + L_3$$

$$= (x-3) \cdot \begin{vmatrix} -1 & 2 & 0 \\ 0 & x-3 & -(x-3) \\ 2 & -3 & 2-x \end{vmatrix}$$

$$= (x-3) \cdot \begin{vmatrix} \overset{+}{-1} & \overset{-}{2} & \overset{+}{0} \\ 0 & 1 & -1 \\ 2 & -3 & 2-x \end{vmatrix}$$

$$= (x-3)^2 \cdot [-(2-x-3) - 2(2)]$$

$$= (x-3)^3. \text{ That is,}$$

$$P_A(x) = (x-3)^3.$$

Ex 03: Let $A \in M_n(\mathbb{R})$ be a square matrix. We will prove that

$$P_{rA}(x) = r^n \cdot P_A\left(\frac{x}{r}\right), \text{ for } r \neq 0.$$

where;

P_A $\equiv$ is the characteristic polynomial of A .

$r \equiv$ is a nonzero real number.

In fact, by definition, we have:

$$p_{r \cdot A}(\lambda) = \det(r \cdot A - \lambda \cdot I_n)$$

$$= \begin{vmatrix} r a_{11} - \lambda & r a_{12} & \dots & r a_{1n} \\ r a_{21} & r a_{22} - \lambda & \dots & r a_{2n} \\ \dots & \dots & \dots & \dots \\ r a_{n1} & r a_{n2} & \dots & r a_{nn} - \lambda \end{vmatrix}$$

$$= \begin{vmatrix} r(a_{11} - \frac{\lambda}{r}) & r a_{12} & \dots & r a_{1n} \\ r a_{21} & r(a_{22} - \frac{\lambda}{r}) & \dots & r a_{2n} \\ \dots & \dots & \dots & \dots \\ r a_{n1} & r a_{n2} & \dots & r(a_{nn} - \frac{\lambda}{r}) \end{vmatrix}$$

$$= r^n \cdot p_A\left(\frac{\lambda}{r}\right), \text{ where } r \neq 0.$$

This completes the proof. $\square$

Ex 04: (i) Let A and B be two matrices such that:

$$A^2 = B^2 = (AB)^2 = I_n \dots (*)$$

We show that $AB = BA$.

In fact, from (*), we get:

$$\begin{cases} A = \bar{A}^{-1} \\ B = \bar{B}^{-1} \\ AB = (AB)^{-1} \end{cases}$$

on the other hand, we know that:

$$(AB)^{-1} = \bar{B}^{-1} \bar{A}^{-1}, \text{ and so}$$

$$AB = \bar{B}^{-1} \bar{A}^{-1} = BA. \text{ The proof is finished } \square$$

(ii) Let A and B be two square matrices. We prove that if A is **invertible**, then

$$p_{AB}(x) = p_{BA}(x).$$

Proof: By definition, we have:

$$\begin{aligned} p_{AB}(x) &= \det(AB - xI) \\ &= \det(A B A \bar{A}^{-1} - x A \bar{A}^{-1}) \\ &= \det[A(BA - xI) \bar{A}^{-1}] \\ &= \det(A) \cdot \det(BA - xI) \cdot \det(\bar{A}^{-1}) \end{aligned}$$

therefore, $P_{AB}(x) = \det(BA - xI) = P_{BA}(x)$.
The proof is finished $\square$.

Ex 05: Consider the matrix:

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

We find the eigenvalues and corresponding eigenvectors of A .

- eigenvalues $\equiv$ Valeurs propres
- eigenvectors $\equiv$ Vecteurs propres

First, we compute the characteristic polynomial

$$\begin{aligned} P_A(x) &= \det(A - xI_2) \\ &= \begin{vmatrix} 1-x & 2 \\ 3 & 2-x \end{vmatrix} \quad \begin{matrix} C_1 \\ \downarrow \\ C_1 - C_2 \end{matrix} \\ &= \begin{vmatrix} -1-x & 2 \\ 1+x & 2-x \end{vmatrix} \\ &= (1+x) \cdot \begin{vmatrix} -1 & 2 \\ 1 & 2-x \end{vmatrix} \end{aligned}$$

$$= (1+x)(-2+x-2)$$

$$= (1+x)(x-4).$$

Thus, the eigenvalues of A are $\lambda_1 = -1$, $\lambda_2 = 4$.

• We find the eigenspaces E_{λ_1} and E_{λ_2} :

By definition, we have:

$$E_{\lambda_1} = \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} x + 2y = -x \\ 3x + 2y = -y \end{array} \right\}$$

$$= \left\{ (x, y) \in \mathbb{R}^2 \mid y = -x \right\}$$

$$= \left\{ (x, -x) ; x \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ (1, -1) \right\}. \text{ That is, } v_1 = (1, -1).$$

We use the same manner to find v_2 :

$$E_{\lambda_2} = \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} x + 2y = 4x \\ 3x + 2y = 4y \end{array} \right\}$$

$$= \left\{ (x, y) \in \mathbb{R}^2 \mid y = \frac{3}{2}x \right\}$$

$$= \left\{ \left(x, \frac{3}{2}x\right) ; x \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ \left(1, \frac{3}{2}\right) \right\} = \text{Vect} \left\{ (2, 3) \right\}.$$

Hence, $v_2 = (2, 3).$

Ex 06: We determine the eigenvalues and eigenvectors of the following matrices:

$$\text{II} \quad A = \begin{bmatrix} 1 & 2 \\ 0 & 5 \end{bmatrix}$$

We see that A is an upper triangular matrix. So its eigenvalues are:

$$\lambda_1 = 1, \lambda_2 = 5.$$

Recall that an upper triangular matrix U is given by:

$$U = \begin{bmatrix} \times & \times & \times & \cdots & \times & \times \\ & \times & \times & & \times & \times \\ & & \times & \times & \times & \times \\ & & & \times & \times & \times \\ & \bigcirc & & & \times & \times \\ & & & & & \times \end{bmatrix}$$

• We compute the eigenvectors of A .

$$E_{\lambda_1} = \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} x + 2y = x \\ 5y = y \end{array} \right\}$$

$$= \left\{ (x, y) \in \mathbb{R}^2 \mid y = 0 \right\}$$

$$= \left\{ (x, 0) ; x \in \mathbb{R} \right\} = \text{Vect} \{ (1, 0) \}.$$

Hence, $v_1 = (1, 0)$.

$$\begin{aligned} \bullet E_{\lambda_2} &= \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} x + 2y = 5x \\ 5y = 5y \end{array} \right\} \\ &= \left\{ (x, y) \in \mathbb{R}^2 \mid y = 2x \right\} \\ &= \left\{ (x, 2x) ; x \in \mathbb{R} \right\} \\ &= \left\{ x(1, 2) ; x \in \mathbb{R} \right\} \\ &= \text{Vect} \{ (1, 2) \}. \text{ Hence, } v_2 = (1, 2). \end{aligned}$$

2 For the matrix;

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

Since A is an upper triangular matrix, then the eigenvalues are: $\lambda = 2$ (double).

The corresponding eigenspace is given by:

$$\begin{aligned} E_{\lambda} &= \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} 2x + 6y = 2x \\ 2y = 2y \end{array} \right\} \\ &= \left\{ (x, y) \in \mathbb{R}^2 \mid y = 0 \right\} \\ &= \text{Vect} \{ (1, 0) \} \end{aligned}$$

That is, $v = (1, 0)$.

[3] For the matrix

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

Since A is an upper triangular matrix, then the eigenvalues are:

- $\lambda_1 = 1$
- $\lambda_2 = 2$
- $\lambda_3 = -5$

Now, we find the corresponding eigenvectors:

$$E_{\lambda_1} = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} x + 2y + 3z = x \\ 2y + 3z = y \\ -5z = z \end{array} \right. \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid z = 0, y = 0 \right\}$$

$$= \left\{ (x, 0, 0) ; x \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ (1, 0, 0) \right\}. \text{ Hence, } \mathbf{v}_1 = (1, 0, 0).$$

$$E_{\lambda_2} = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} x + 2y + 3z = 2x \\ 2y + 3z = 2y \\ -5z = 2z \end{array} \right. \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid z = 0, x = 2y \right\}$$

$$= \left\{ (2y, y, 0) ; y \in \mathbb{R} \right\} = \text{Vect} \left\{ (2, 1, 0) \right\}.$$

Hence, $v_2 = (2, 1, 0)$.

$$E_{\lambda_3} = \{(x, y, z) \in \mathbb{R}^3 \mid \left. \begin{aligned} x + 2y + 3z &= -5x \\ 2y + 3z &= -5y \\ -5z &= -5z \end{aligned} \right\}$$

$$= \{(x, y, z) \in \mathbb{R}^3 \mid \left. \begin{aligned} z &= -\frac{7}{3}y \\ x + 2y - 7y &= -5x \end{aligned} \right\}$$

$$= \{(x, y, z) \in \mathbb{R}^3 \mid \left. \begin{aligned} y &= \frac{6}{5}x \\ z &= -\frac{42}{15}x \end{aligned} \right\}$$

$$= \left\{ \left(x, \frac{6}{5}x, -\frac{42}{15}x \right) ; x \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ \left(1, \frac{6}{5}, -\frac{42}{15} \right) \right\}$$

$$= \text{Vect} \left\{ (5, 6, -14) \right\}. \text{ Hence,}$$

$$v_2 = (5, 6, -14).$$

[4] For the matrix

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

which is a lower triangular matrix, and so the eigenvalues are: $\lambda_1 = 1$ and $\lambda_2 = 2$.

• We compute the eigenvalues of A :

$$E_{h_1} = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} x = x \\ x + 2y = y \\ x + 2z = z \end{array} \right. \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} x = -y \\ z = y \end{array} \right. \right\}$$

$$= \left\{ (-y, y, y) ; y \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ (-1, 1, 1) \right\}. \text{ Hence, } v_1 = (-1, 1, 1).$$

$$E_{h_2} = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} x = 2x \\ x + 2y = 2y \\ x + 2z = 2z \end{array} \right. \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} x = 0 \\ y, z \in \mathbb{R} \end{array} \right. \right\}$$

$$= \left\{ (0, y, z) ; y, z \in \mathbb{R} \right\}$$

$$= \left\{ y(0, 1, 0) + z(0, 0, 1) ; y, z \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ (0, 1, 0), (0, 0, 1) \right\}$$

Then h_2 has two eigenvectors:

$$v_2 = (0, 1, 0), \quad v_3 = (0, 0, 1).$$

5 For the matrix :

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

We see that A has one eigenvalue: $\lambda = 0$.

• The corresponding eigenspace:

$$E_{\lambda} = \left\{ (x, y, z) \in \mathbb{R}^3 \mid y + z = 0 \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid z = -y \right\}$$

$$= \left\{ (x, y, -y) ; x, y \in \mathbb{R} \right\}$$

$$= \left\{ x(1, 0, 0) + y(0, 1, -1) ; x, y \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ (1, 0, 0), (0, 1, -1) \right\}$$

Thus, $\lambda = 0$ has two eigenvectors:

$$v_1 = (1, 0, 0),$$

$$v_2 = (0, 1, -1).$$

6 For the matrix:

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

It is clear that A is a lower triangular matrix. That is, $\lambda = 2$ is the only eigenvalue.

Now, we compute the corresponding eigen-space. In fact, we have:

$$\begin{aligned} E_{\lambda} &= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} 2x = 2x \\ x + 2y = 2y \\ 3y + 2z = 2z \end{array} \right\} \\ &= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} x = 0 \\ y = 0 \end{array} \right\} \\ &= \{ (0, 0, z); z \in \mathbb{R} \} \\ &= \text{Vect} \{ (0, 0, 1) \}. \text{ Hence, } v = (0, 0, 1). \end{aligned}$$

7 For the matrix:

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

- The matrix A is neither upper triangular nor lower triangular. Thus, we must find $P_A(\lambda)$ using the definition: $P_A(\lambda) = \det(A - \lambda I)$.

In fact, we have:

$$\begin{aligned} p_A(x) &= \det(A - x I_3) \\ &= \begin{vmatrix} \overset{+}{1-x} & \overset{-}{1} & \overset{+}{0} \\ 1 & 1-x & 0 \\ 0 & 0 & 2-x \end{vmatrix} \end{aligned}$$

$$= (1-x) [(1-x)(2-x) - 0] - (2-x)$$

$$= (2-x) \cdot [(1-x)^2 - 1]$$

$$= (2-x)(1-2x+x^2-1)$$

$$= x(2-x)(x-2) = -x(x-2)^2.$$

Thus, the eigenvalues of A are:

$$\bullet \quad \lambda_1 = 0$$

$$\bullet \quad \lambda_2 = 2 \quad (\text{double}).$$

We compute the eigenvectors:

$$E_{\lambda_1} = \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} x+y=0 \\ 2z=0 \end{array} \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid x = -y, z = 0 \right\}$$

$$= \left\{ (-y, y, 0) ; y \in \mathbb{R} \right\}$$

$$= \text{Vect} \{ (-1, 1, 0) \}. \quad \text{Hence, } v_1 = (-1, 1, 0).$$

$$E_{h_2} = \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} x+y = 2x \\ x+y = 2y \\ 2z = 2z \end{array} \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid x = y \right\}$$

$$= \left\{ (x, x, z) ; x, z \in \mathbb{R} \right\}$$

$$= \left\{ x(1, 1, 0) + z(0, 0, 1) ; x, z \in \mathbb{R} \right\}$$

$$= \text{Vect} \{ (1, 1, 0), (0, 0, 1) \}$$

Hence, the eigenvalues $h = 2$ has 2 eigenvectors:

$$v_2 = (1, 1, 0) \text{ and } v_3 = (0, 0, 1).$$

8 For the matrix :

$$A = \begin{bmatrix} a & 2 & 3 \\ 0 & 2a & 8 \\ 0 & 0 & 3a \end{bmatrix}, \quad a \in \mathbb{R}.$$

We see that A is an upper triangular matrix, and hence $h_1 = a$, $h_2 = 2a$ and $h_3 = 3a$ are the eigenvalues of A .

Remark. For $a = 0$, the matrix A has one eigenvalue; $h = 0$.

There are two cases to consider:

Case 1: Assume that $a = 0$. Hence,

$$E_1 = \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} 2y + 3z = 0 \\ 8z = 0 \end{array} \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} z = 0 \\ y = 0 \end{array} \right\}$$

$$= \{ (x, 0, 0) ; x \in \mathbb{R} \}$$

$$= \text{Vect} \{ (1, 0, 0) \}. \text{ That is, } v_1 = (1, 0, 0).$$

Case 2: Assume that $a \neq 0$. In this case, the matrix A has three distinct eigenvalues:

$$\lambda_1 = a, \lambda_2 = 2a \text{ and } \lambda_3 = 3a.$$

• We find the eigenspace for each eigenvalue:

$$E_{\lambda_1} = \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} ax + 2y + 3z = ax \\ 2ay + 8z = ay \\ 3az = az \end{array} \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} z = 0 \text{ (because } a \neq 0) \\ y = 0 \end{array} \right\}$$

$$= \{ (x, 0, 0) ; x \in \mathbb{R} \} = \text{Vect} \{ (1, 0, 0) \}$$

Hence,

$$v_1 = (1, 0, 0).$$

$$E_{h_2} = \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} ax + 2y + 3z = 2ax \\ 2ay + 8z = 2ay \\ 3az = 2az \end{array} \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} z = 0, \quad (a \neq 0) \\ y = \frac{a}{2} x \end{array} \right\}$$

$$= \left\{ \left(\frac{2}{a} y, y, 0 \right); y \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ \left(\frac{2}{a}, 1, 0 \right) \right\}.$$

That is, $v_2 = \left(\frac{2}{a}, 1, 0 \right)$ or $(2, a, 0)$.

$$E_{h_3} = \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} ax + 2y + 3z = 3ax \\ 2ay + 8z = 3ay \\ 3az = 3az \end{array} \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} y = \frac{8}{a} z \\ 9x + \frac{16}{a} z + 3z = 3ax \end{array} \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} y = \frac{8}{a} z \\ x = \frac{1}{2a} \left(\frac{16+3a}{a} \right) z \end{array} \right\}$$

$$= \left\{ \left(\frac{1}{2a^2} (16+3a) z, \frac{8}{a} z, z \right); z \in \mathbb{R} \right\}$$

$$= \text{Vect} \left\{ \left(\frac{1}{2a^2} (16+3a), \frac{8}{a}, 1 \right) \right\}.$$

Test number 2

Ex 01: The definition of two similar matrices:

Definition: Let A, B be two n by n matrices. We say that A is **similar** to B if there exists an invertible matrix P such that:

$$A = P \cdot B \cdot P^{-1}$$

or, equivalently,

$$B = P^{-1} \cdot A \cdot P.$$

• We write $A \sim B$ when A is similar to B .

[b] We prove that ?

$$A - hI \sim B \Rightarrow A \sim hI + B.$$

In fact, if $A - hI \sim B$, then

$A - hI = P \cdot B \cdot P^{-1}$ for some invertible matrix P . It follows that:

$$\begin{aligned} A &= hI + P \cdot B \cdot P^{-1} \\ &= h \underbrace{P P^{-1}} + P B P^{-1} \\ &= P (hI + B) P^{-1} \end{aligned}$$

Thus, $A \sim hI + B$.

That is, A is similar to $hI + B$.

[c] Assume that $A = P B P^{-1}$, where P is invertible.

If (λ, x) is an eigenpair of $A \stackrel{?}{\Rightarrow} (\lambda, P^{-1}x)$ is an eigenpair of B .

(α, v) is an eigenpair of $A \Rightarrow \begin{cases} \alpha = \text{eigenvalue} \\ v = \text{eigenvector} \end{cases}$

Since $Ax = \lambda x$ and $A = P B P^{-1}$, then

$$\begin{aligned} B(P^{-1}x) &= P^{-1} A P (P^{-1}x) \\ &= P^{-1} A \underbrace{P P^{-1}}_I x = P^{-1} (Ax) \\ &= P^{-1} (\lambda x) \\ &= \lambda (P^{-1}x). \end{aligned}$$

Hence,

$$B(P^{-1}x) = \lambda (P^{-1}x).$$

That is, $(\lambda, P^{-1}x)$ is an eigenpair of B .

[d] Let

$$f_r(x) = a_0 + a_1 x + \dots + a_r x^r \text{ be}$$

a polynomial of degree r . We prove that

$$\text{if } B = P^{-1} A P, \text{ then } f_r(B) = P^{-1} f_r(A) P$$

Since $f_r(x) = a_0 + a_1 x + \dots + a_r x^r$, then

$$\begin{aligned} f_r(B) &= a_0 I + a_1 B + \dots + a_r B^r \\ &= a_0 I + a_1 (\bar{P}^{-1} A P) + \dots + a_r (\bar{P}^{-1} A P)^r \\ &= a_0 \cdot \underbrace{\bar{P}^{-1} P}_I + a_1 \bar{P}^{-1} A P + \dots + a_r \bar{P}^{-1} A^r P \\ &= \bar{P}^{-1} \left[a_0 \cdot I + a_1 \cdot A + \dots + a_r A^r \right] \cdot P \\ &= \bar{P}^{-1} f_r(A) \cdot P. \end{aligned}$$

Remark. We have used the fact that $(\bar{P}^{-1} A P)^i = \bar{P}^{-1} A^i P$, for $i \geq 1$.

Ex 02: We show that similar matrices have the same eigenvalues.

Proof. Let $A, B \in M_n(\mathbb{R})$ be two similar matrices. That is, $A = P B P^{-1}$ for some invertible matrix P . It suffices to prove that: $P_A(x) = P_B(x)$. In fact, we have:

$$\begin{aligned} P_A(x) &= \det(A - x I) \\ &= \det(P B P^{-1} - x \underbrace{I}_I) \\ &= \det(P B P^{-1} - x P P^{-1}) \end{aligned}$$

$$= \det [P (B - \lambda I) P^{-1}]$$

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

$$= \det(B - \lambda I), \text{ since } \det(P^{-1}) = \frac{1}{\det(P)}$$

$$= f_B(\lambda).$$

Second method:

First, let $Sp(A)$ the set given by:

$$Sp(A) = \left\{ h \in \mathbb{R} \mid h \text{ is an eigenvalue of } A \right\}$$

$$\text{Let } h \in Sp(A)$$

$$\Rightarrow 0 \in Sp(A - hI)$$

on the other hand, we have

$$A - hI = P (B - hI) P^{-1}$$

$$\Rightarrow 0 \in Sp(B - hI)$$

$$\Rightarrow h \in Sp(B).$$

$$\Rightarrow Sp(A) \subset Sp(B).$$

We use the same manner to show that

$$Sp(B) \subset Sp(A). \text{ Hence, } Sp(A) = Sp(B).$$

Ex 03: We prove the following result:

$$A \sim B \Rightarrow e^A \sim e^B$$

e^A = the exponential of A .

Recall that for any square matrix $A \in M_n(\mathbb{R})$ we have:

$$\begin{aligned} e^A &= I + \frac{A}{1!} + \frac{A^2}{2!} + \dots + \frac{A^k}{k!} + \dots \\ &= \sum_{k=0}^{+\infty} \frac{A^k}{k!} \end{aligned}$$

Since A is similar to B , then $A = P \cdot B \cdot P^{-1}$ for some invertible matrix P . Therefore,

$$e^A = \sum_{k=0}^{+\infty} \frac{(P B P^{-1})^k}{k!}$$

$$= \sum_{k=0}^{\infty} P \cdot \frac{B^k}{k!} \cdot P^{-1}$$

$$= P \cdot \left(\sum_{k=0}^{+\infty} \frac{B^k}{k!} \right) \cdot P^{-1}$$

$$= P \cdot e^B \cdot P^{-1}, \text{ and so } e^A \sim e^B.$$

Ex 04: Consider the two matrices

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

We prove that A is similar to B (i.e, $A \sim B$).

From Simple Computation, we obtain:

$$\begin{aligned} P_A(x) &= \begin{vmatrix} 2-x & 1 \\ 2 & 3-x \end{vmatrix} \\ &= \begin{vmatrix} (1-x) & 1 \\ -(1-x) & 3-x \end{vmatrix} = (1-x) \cdot \begin{vmatrix} 1 & 1 \\ -1 & 3-x \end{vmatrix} \\ &= (1-x)(3-x+1) = (1-x)(4-x). \end{aligned}$$

For the matrix B, we also obtain:

$$\begin{aligned} P_B(x) &= \begin{vmatrix} -x & 2 \\ -2 & 5-x \end{vmatrix} = \begin{vmatrix} 2-x & 2 \\ 3-x & 5-x \end{vmatrix} \\ &= \begin{vmatrix} (4-x) & 2 \\ 2(4-x) & 5-x \end{vmatrix} \\ &= (4-x) \cdot \begin{vmatrix} 1 & 2 \\ 2 & 5-x \end{vmatrix} \\ &= (4-x)(5-x-4) = (4-x)(1-x). \end{aligned}$$

Thus, $P_A(x) = P_B(x)$ and so $\text{sp}(A) = \text{sp}(B)$. Since the eigenvalues of A (resp. B) are simple, then A and B are **diagonalizable**. That is, there exist two invertible matrices P and Q such that:

$$\begin{cases} A = P \cdot D \cdot P^{-1} \\ B = Q \cdot D \cdot Q^{-1} \end{cases}$$

It follows that

$$\begin{aligned} A &= P \cdot D \cdot P^{-1} \\ &= P \cdot (Q^{-1} B Q) \cdot P^{-1} \\ &= P \cdot Q^{-1} \cdot B \cdot Q \cdot P^{-1} \\ &= P \cdot Q^{-1} \cdot B \cdot (P \cdot Q^{-1})^{-1} \end{aligned}$$

We put $R = P \cdot Q^{-1}$ which is invertible. Then

$$A = R \cdot B \cdot R^{-1} \Rightarrow A \sim B.$$

Remark: The product of invertible matrices is an invertible matrix.

- We find the matrix R .

We must compute P and Q .

• Simple Computation, we get:

For the matrix A:

$$\lambda_1 = 1 \rightarrow v_1 = (-1, 1)$$

$$\lambda_2 = 4 \rightarrow v_2 = (1, 2)$$

For the matrix B:

$$\lambda_1 = 1 \rightarrow v'_1 = (2, 1)$$

$$\lambda_2 = 4 \rightarrow v'_2 = (1, 2)$$

Therefore,

$$\begin{aligned} R = P \cdot \tilde{Q}^{-1} &= \begin{bmatrix} -1 & 1 \\ 1 & 2 \end{bmatrix} \cdot \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}^{-1} \\ &= \begin{bmatrix} -1 & 1 \\ 1 & 2 \end{bmatrix} \cdot \begin{bmatrix} \frac{2}{3} & -\frac{1}{3} \\ \frac{-1}{3} & \frac{2}{3} \end{bmatrix} \\ &= \begin{bmatrix} -1 & 1 \\ 0 & 1 \end{bmatrix} \end{aligned}$$

Ex 05: We prove the following result:

Let $A \in \mathcal{M}_n(\mathbb{R})$ be a diagonalizable matrix which has one eigenvalue λ . Then $A = \lambda \cdot I$, where I is the identity matrix.

$$\text{That is, } I = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix}$$

proof: Since A is diagonalizable, then $A = P D P^{-1}$ for some invertible matrix P , where D is diagonal. Moreover, D includes the eigenvalues of A . Then,

$$A = P \cdot D \cdot P^{-1} = P \cdot \begin{bmatrix} h & & \\ & h & \\ & & \ddots \\ & & & h \end{bmatrix} \cdot P^{-1}$$

$$= h \cdot P \cdot \begin{bmatrix} 1 & & \\ & 1 & \\ & & \ddots \\ & & & 1 \end{bmatrix} \cdot P^{-1} = h P P^{-1} = h \cdot I.$$

Hence, $A = hI$. The proof is finished.

Recall that any polynomial $p(x)$ of degree n can be factorized as follows:

$$p(x) = (x - h_1)^{\alpha_1} (x - h_2)^{\alpha_2} \dots (x - h_k)^{\alpha_k},$$

where $\alpha_1 + \alpha_2 + \dots + \alpha_k = n$, h_i are the roots of p . In this case: α_i is called:

$\alpha_i \equiv$ the Algebraic multiplicity of h_i .

Example: Let $p(x) = x^3 - 3x + 2$.

It is clear that

$$p(x) = (x - 1)^2 (x + 2)$$

• The algebraic multiplicity of 1 is 2.

• " " " " - 2 is 1

Notation: $A_m(h)$ denotes the Algebraic multiplicity of h .

- We also denote by $G_m(h)$ for the dimension of E_h . That is,

$$G_m(h) = \dim E_h$$

From the Course, we have

(A is diagonalizable) iff $A_m(h) = G_m(h)$ for any $h \in Sp(A)$.

Ex 06: We study the diagonalization of the matrix :

$$A = \begin{bmatrix} 1 & a & b & c \\ 0 & 1 & d & e \\ 0 & 0 & 2 & f \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

this is an upper triangular matrix.

The eigenvalues of A are :

- $h_1 = 1$ with $A_m(h_1) = 2$
- $h_2 = 2$ with $A_m(h_2) = 1$
- $h_3 = 3$ with $A_m(h_3) = 1$

So, the matrix A is diagonalizable whenever $\dim E_{h_1} = 2$.

We compute E_{h_1} :

$$E_{\lambda_1} = \left\{ (x, y, z, t) \in \mathbb{R}^4 \mid \begin{array}{l} x + ay + bz + ct = x \\ y + dz + et = y \\ 2z + ft = z \\ 3t = t \end{array} \right\}$$

$$= \left\{ (x, y, z, t) \in \mathbb{R}^4 \mid \begin{array}{l} t = 0, z = 0 \\ ay = 0 \end{array} \right\}$$

we distinguish two cases:

Case 1: If $a \neq 0$, then

$$E_{\lambda_1} = \left\{ (x, y, z, t) \in \mathbb{R}^4 \mid \begin{array}{l} t = 0, z = 0 \\ y = 0 \end{array} \right\}$$

$$= \{ (x, 0, 0, 0); x \in \mathbb{R} \}$$

$$= \text{Vect} \{ (1, 0, 0, 0) \} \Rightarrow \dim E_{\lambda_1} = 1.$$

In this case, A is not diagonalizable.

Case 2: If $a = 0$, then

$$E_{\lambda_1} = \left\{ (x, y, z, t) \in \mathbb{R}^4 \mid \begin{array}{l} z = 0 \\ t = 0 \end{array} \right\}$$

$$= \{ (x, y, 0, 0); x, y \in \mathbb{R} \}$$

$$= \{ x(1, 0, 0, 0) + y(0, 1, 0, 0) \mid x, y \in \mathbb{R} \}$$

$$= \text{Vect} \{ (1, 0, 0, 0), (0, 1, 0, 0) \}$$

$$\text{Hence, } \dim E_{\lambda_1} = 2 = A_m(\lambda_1)$$

In this case, the matrix A is diagonalizable.

Conclusion: The matrix A is diagonalizable if and only if $a=0, b, c, d, e, f \in \mathbb{R}$.

Ex 07: Let $A, B \in M_n(\mathbb{R})$ be a diagonalizable matrix with $Sp(A) = \{-1, 1\}$. That is, A has two eigenvalues $\lambda_1 = -1$ and $\lambda_2 = 1$.

We prove that $A^{-1} = A$.

Proof: Since $A = P \cdot D \cdot P^{-1}$ for some invertible matrix P , where

$$D = \begin{bmatrix} \pm 1 & & \\ & \pm 1 & \\ & & \ddots \\ & & & \pm 1 \end{bmatrix}, \text{ then}$$

$$A^{-1} = (P \cdot D \cdot P^{-1})^{-1} = P \cdot D^{-1} \cdot P^{-1}. \text{ Moreover,}$$

$$D^{-1} = \begin{bmatrix} \frac{1}{\pm 1} & & \\ & \frac{1}{\pm 1} & \\ & & \ddots \\ & & & \frac{1}{\pm 1} \end{bmatrix} = \begin{bmatrix} \pm 1 & & \\ & \pm 1 & \\ & & \ddots \\ & & & \pm 1 \end{bmatrix} = D$$

$$\text{Thus, } A^{-1} = P \cdot D \cdot P^{-1} = A.$$

The proof is finished $\square$.

Ex 08: Let A and B be two diagonalizable matrices such that

- $A = P D_1 P^{-1}$
- $B = P D_2 P^{-1}$

where P is invertible, D_1 and D_2 are diagonal.

• We prove that $AB = BA$.

proof: It follows from the hypothesis that:

$$\begin{aligned} AB &= P D_1 \bar{P}^{-1} P D_2 \bar{P}^{-1} \\ &= P D_1 D_2 \bar{P}^{-1} \\ &= P D_2 D_1 \bar{P}^{-1} \\ &= P D_2 \bar{P}^{-1} P D_1 \bar{P}^{-1} \\ &= (P D_2 \bar{P}^{-1}) \cdot (P D_1 \bar{P}^{-1}) \\ &= B \cdot A. \end{aligned}$$

The proof is finished $\square$.

Remark: we have used the fact that $D_1 D_2 = D_2 D_1$ since the matrices D_1 and D_2 are diagonal.

Ex 09: we study whether the matrix

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

is diagonalizable or not.

1. We find the characteristic polynomial:

$$P_A(x) = \begin{vmatrix} 3-x & 0 & 0 \\ 4 & 1-x & 2 \\ a & 0 & 3-x \end{vmatrix}$$
$$= (3-x) \left[(1-x)(3-x) - 0 \right]$$
$$= (3-x)^2 (1-x).$$

Hence, $P_A(x) = (3-x)^2 (1-x)$

2. The eigenvalues of A are:

- $\lambda_1 = 1$ with $A_m(\lambda_1) = 1$
- $\lambda_2 = 3$ with $A_m(\lambda_2) = 2$.

3. We compute the eigenspace E_{λ_2} :

$$E_{\lambda_2} = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} 3x = 3x \\ 4x + y + 2z = 3y \\ ax + 3z = 3z \end{array} \right. \right\}$$
$$= \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} y = 2x + z \\ ax = 0 \end{array} \right. \right\}$$

There are two cases to consider:

Case 1: $a = 0$.

$$E_{\lambda_2} = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ y = 2x + z \right. \right\}$$

$$= \{ (x, 2x+z, z) ; x, z \in \mathbb{R} \}$$

$$= \{ x(1, 2, 0) + z(0, 1, 1) ; x, z \in \mathbb{R} \}$$

$$= \text{Vect} \{ (1, 2, 0), (0, 1, 1) \}$$

Thus, $\dim E_{h_2} = 2 = A_m(h_2)$, and so A is diagonalizable.

Case 2: $a \neq 0$. In this case, we have:

$$E_{h_2} = \{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} x = 0 \\ y = z \end{array} \}$$

$$= \{ (0, y, y) ; y \in \mathbb{R} \}$$

$$= \text{Vect} \{ (0, 1, 1) \}$$

Thus, $\dim E_{h_2} = 1 \neq A_m(h_2)$.

Therefore, A is not diagonalizable.

b Let

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

We find the values $\alpha \in \mathbb{R}$ for which the matrix A_α is diagonalizable.

First, it is clear that A_α is an upper triangular matrix.

So, the eigenvalues of A_α are:

$$\lambda_1 = 2 \text{ and } \lambda_2 = \alpha.$$

There are two Cases:

Case 1: $\lambda_2 = 2$. In this Case, the matrix A_α has one eigenvalue, say $\lambda = 2$.

From exercise 05, if $A \neq \lambda I$, then A is not diagonalizable. But, we see that

$$A_\alpha = A_2 = \begin{bmatrix} 2 & 2 & 1 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \neq 2I$$

$\Rightarrow A_2$ is not diagonalizable.

Case 2: $\alpha \neq 2$: In this Case, the matrix A_α has two eigenvalues:

• $\lambda_1 = 2$ with $A_m(2) = 2$

• $\lambda_2 = \alpha$ with $A_m(\alpha) = 1$.

Let us compute the eigenspace E_{λ_1} :

$$\begin{aligned} E_{\lambda_1} &= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} 2x + \alpha y + z = 2x \\ 2y = 2y \\ \alpha z = 2z \end{array} \right\} \\ &= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} z = 0 \\ \alpha \cdot y = 0 \end{array} \right\} \end{aligned}$$

We distinguish two Cases:

Case 1: For $\alpha \neq 0$, we have:

$$E_{\lambda_1} = \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} z = 0 \\ y = 0 \end{array} \right\},$$

which gives $\dim E_{\lambda_1} = 1$, and so A_α is not diagonalizable.

Case 2: For $\alpha = 0$. In this Case, we have:

$$\begin{aligned} E_{\lambda_1} &= \left\{ (x, y, z) \in \mathbb{R}^3 \mid z = 0 \right\} \\ &= \left\{ (x, y, 0) ; x, y \in \mathbb{R} \right\} \\ &= \text{Vect} \left\{ (1, 0, 0), (0, 1, 0) \right\} \end{aligned}$$

That is, $\dim E_{\lambda_1} = 2 = A_m(\lambda_1)$.

$\Rightarrow A_\alpha$ is diagonalizable.

Ex 10: Consider the matrix:

$$A = \begin{bmatrix} 1 & c \\ 0 & 1 \end{bmatrix}, \quad c \in \mathbb{R}$$

We find the values c for which the matrix A is diagonalizable.

Since A is an upper triangular matrix, then the eigenvalues of A are $\lambda = 1$ (double)

- If $c \neq 0$ (See exercise 5), then A is not diagonalizable since $A \neq hI$.

Second method:

We see that $h = 1$ is the eigenvalue of A with multiplicity $A_m(h) = 2$.

Let us find the corresponding eigenspace:

$$\begin{aligned} E_h &= \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} x + cy = x \\ y = y \end{array} \right\} \\ &= \left\{ (x, y) \in \mathbb{R}^2 \mid c \cdot y = 0 \right\} \end{aligned}$$

For $c \neq 0$, we have:

$$\begin{aligned} E_h &= \left\{ (x, y) \in \mathbb{R}^2 \mid y = 0 \right\} \\ &= \text{Vect} \{ (1, 0) \} \end{aligned}$$

$$\Rightarrow \dim E_h = 1 \neq A_m(h)$$

$\Rightarrow A$ is not diagonalizable.

For $c = 0$, we also have:

$$\begin{aligned} E_h &= \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} x = x \\ y = y \end{array} \right\} \\ &= \{ (x, y) ; x, y \in \mathbb{R} \} \\ &= \{ x(1, 0) + y(0, 1) ; x, y \in \mathbb{R} \} \\ &= \text{Vect} \{ (1, 0), (0, 1) \} \Rightarrow A \text{ is diago-} \end{aligned}$$

Remark. Any matrix of the form hI is diagonalizable. In fact, we have:

$$hI = I \cdot (hI) \cdot I^{-1} \quad (I^{-1} = I).$$

Ex 11: Consider the matrix

$$A(a) = \begin{bmatrix} 1 & 0 & 0 \\ a & -2 & 3 \\ 1 & -1 & 2 \end{bmatrix}; \quad a \in \mathbb{R}.$$

• The eigenvalues of $A(a)$:

First, we compute the characteristic polynomial of $A(a)$. In fact, we have:

$$\begin{aligned} \Delta_{A(a)}(\lambda) &= \begin{vmatrix} 1-\lambda & 0 & 0 \\ a & -2-\lambda & 3 \\ 1 & -1 & 2-\lambda \end{vmatrix} \\ &= \begin{vmatrix} 1-\lambda & 0 & 0 \\ a & 1-\lambda & 3 \\ 1 & 1-\lambda & 2-\lambda \end{vmatrix} \end{aligned}$$

$\begin{matrix} C_2 \\ \downarrow \\ C_2 + C_3 \end{matrix}$

$$= (1-x) \cdot \begin{vmatrix} \overset{+}{1-x} & \overset{-}{0} & \overset{+}{0} \\ a & 1 & 3 \\ 1 & 1 & 2-x \end{vmatrix}$$

$$= (1-x) \left[(1-x)(2-x-3) \right]$$

$$= (1-x)^2 (-1-x) = -(1+x)(1-x)^2.$$

• The eigenvalues of $A(a)$ are:

$$h_1 = -1 \text{ with } A_m(h_1) = 1,$$

$$h_2 = 1 \text{ with } A_m(h_2) = 2.$$

Now, we study the diagonalization of $A(a)$:

Let us compute E_{h_2} :

$$E_{h_2} = E_1 = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} x = x \\ ax - 2y + 3z = y \\ x - y + 2z = z \end{array} \right. \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} z = -x + y \\ ax + 3z = 3y \end{array} \right. \right\} \quad (*)$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \left/ (a-3)x = 0 \right. \right\}$$

• If $a \neq 3$, then by (*) we have

$$E_1 = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} x = 0 \\ z = y \end{array} \right. \right\}$$

$= \text{Vect} \{ (0, 1, 1) \} \Rightarrow A(a)$ is not diago-

- If $a = 3$, then by (*) we also have:

$$\begin{aligned}
 E_1 &= \left\{ (x, y, z) \in \mathbb{R}^3 \mid z = -x + y \right\} \\
 &= \left\{ (x, y, -x + y) ; x, y \in \mathbb{R} \right\} \\
 &= \left\{ x(1, 0, -1) + y(0, 1, 1) ; x, y \in \mathbb{R} \right\} \\
 &= \text{Vect} \{ (1, 0, -1), (0, 1, 1) \}.
 \end{aligned}$$

$$\Rightarrow \dim E_1 = 2 = A_m(1).$$

$$\Rightarrow A(3) \text{ is diagonalizable.}$$

Conclusion: The matrix $A(a)$ is diagonalizable if and only if $a = 3$.

- We diagonalize the matrix $A(3)$, i.e., we write the matrix $A(3)$ in the form:

$$A(3) = P \cdot D \cdot P^{-1}.$$

It suffices to find E_{λ_1} :

$$E_{\lambda_1} = E_{-1} = \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} x = -x \\ 3x - 2y + 3z = -y \\ x - y + 2z = -z \end{array} \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \mid \begin{array}{l} x = 0 \\ y = 3z \end{array} \right\}$$

$$= \left\{ (0, 3z, z) ; z \in \mathbb{R} \right\}$$

$$= \text{Vect} \{ (0, 3, 1) \} \rightarrow v_1 = (0, 3, 1).$$

Now, we put

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

and

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

Simple Computations, we obtain :

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

and so,

$$A(3) = \underbrace{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 3 \\ -1 & 1 & 1 \end{bmatrix}}_P \cdot \underbrace{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix}}_D \cdot \underbrace{\begin{bmatrix} 1 & 0 & 0 \\ \frac{3}{2} & -\frac{1}{2} & \frac{3}{2} \\ -\frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \end{bmatrix}}_{P^{-1}}$$

This is the diagonalization of $A(3)$, from which we can compute $\sqrt{A(3)}$, $e^{A(3)}$, ...

Ex 12: Let

$$f: P_3[x] \longrightarrow P_3[x]$$

$$p \mapsto f(p) = 3xp - (x^2 - 1)p'$$

p' is the derivative of p .

Let $B = \{1, x, x^2, x^3\}$ be the Canonical basis of $P_3[x]$.

① We Calculate $M_f(B)$

$M_f(B) \equiv$ the matrix of f with respect to the basis B .

First, we have:

- $f(1) = 3x = 0 + 3 \cdot x + 0 \cdot x^2 + 0 \cdot x^3$
- $f(x) = 1 + 2x^2 = 1 + 0 \cdot x + 2 \cdot x^2 + 0 \cdot x^3$
- $f(x^2) = 2x + x^3 = 0 + 2x + 0 \cdot x^2 + 1 \cdot x^3$
- $f(x^3) = 3x^2 = 0 + 0 \cdot x + 3x^2 + 0 \cdot x^3$

Hence,

$$M_f(B) = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 3 & 0 & 2 & 0 \\ 0 & 2 & 0 & 3 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

② We study whether $M_f(B)$ is diagonalizable or not.

* characteristic polynomial

$$p(x) = (x-1)(x+1)(x-3)(x+3)$$

* eigenvalues:

$$\lambda_1 = -1, \lambda_2 = 1, \lambda_3 = -3, \lambda_4 = 3$$

* eigenvectors:

$$v_1 = (1, -1, -1, 1), v_2 = (-1, -1, 1, 1)$$

$$v_3 = (-1, 3, -3, 1), v_4 = (1, 3, 3, 1)$$

The matrix $M_f(B)$ is diagonalizable since the eigenvalues λ are simple.

After few computation, we obtain:

$$M_f(B) = \begin{bmatrix} 1 & -1 & -1 & 1 \\ -1 & -1 & 3 & 3 \\ -1 & 1 & -3 & 3 \\ 1 & 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} -1 & & & \\ & 1 & & \\ & & -3 & \\ & & & 3 \end{bmatrix} \begin{bmatrix} \frac{3}{8} & -\frac{1}{8} & -\frac{1}{8} & \frac{3}{8} \\ -\frac{3}{8} & -\frac{1}{8} & \frac{1}{8} & \frac{3}{8} \\ -\frac{1}{8} & \frac{1}{8} & -\frac{1}{8} & \frac{1}{8} \\ \frac{1}{8} & \frac{1}{8} & \frac{1}{8} & \frac{1}{8} \end{bmatrix}$$

$P \quad \quad D \quad \quad P^{-1}$

This is the diagonalization formula of $M_f(B)$ or f .

Test number 3

Ex01

1) the definition of exponential of a square matrix A :

Definition: We know that:

$$\forall x \in \mathbb{R}: e^x = \sum_{k=0}^{+\infty} \frac{x^k}{k!}, \text{ then}$$

for any square matrix A , we have:

$$e^A = \sum_{k=0}^{+\infty} \frac{A^k}{k!}$$

$$= I + \frac{A}{1!} + \frac{A^2}{2!} + \dots + \frac{A^k}{k!} + \dots$$

2) Let $O = \text{zero matrix}$. Then

$$e^O = I + O + O + \dots$$

$= I$, where I is the unit matrix

$$I = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix}$$

مثال الأعداد
 $e^0 = 1$

3) Let $A \in M(\mathbb{R})$, and let $h \in \mathbb{R}$.
We prove that:

$$e^{h \cdot I} = e^h \cdot A$$

proof: By definition, we have

$$\begin{aligned} e^{hI} &= I + \frac{(hI)}{1!} + \frac{(hI)^2}{2!} + \dots + \frac{(hI)^k}{k!} + \dots \\ &= I + \frac{h}{1!} I + \frac{h^2}{2!} I + \dots + \frac{h^k}{k!} I + \dots \\ &= \left(1 + \frac{h}{1!} + \frac{h^2}{2!} + \dots + \frac{h^k}{k!} + \dots \right) \cdot I \\ &= e^h \cdot I. \end{aligned}$$

The proof is finished.

Example: Let

$$A = \begin{bmatrix} -2 & 0 \\ 0 & -2 \end{bmatrix} = -2 \cdot I.$$

Then

$$\begin{aligned} e^A &= e^{-2} \cdot I = e^{-2} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} e^{-2} & 0 \\ 0 & e^{-2} \end{bmatrix} \end{aligned}$$

Ex 02: Let A be the matrix

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

- We prove that A is diagonalizable.

Since A is an upper triangular matrix, then the eigenvalues are: $\lambda_1 = 1, \lambda_2 = 2$.

*) Now, we compute the eigenvectors:

$$\begin{aligned} E_1 &= \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} x + y = x \\ 2y = y \end{array} \right\} \\ &= \left\{ (x, y) \in \mathbb{R}^2 \mid y = 0 \right\} \\ &= \text{Vect} \{ (1, 0) \}. \text{ Hence, } v_1 = (1, 0). \end{aligned}$$

Using the same manner, we obtain:

$$\begin{aligned} E_2 &= \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} x + y = 2x \\ 2y = 2y \end{array} \right\} \\ &= \left\{ (x, y) \in \mathbb{R}^2 \mid y = x \right\} \\ &= \text{Vect} \{ (1, 1) \}. \text{ Hence, } v_2 = (1, 1). \end{aligned}$$

Since the eigenvalues of A are simple, i.e.,

$$A_m(1) = A_m(2) = 1, \text{ then the matrix}$$

A is **diagonalizable**.

Next, we put $D = \begin{bmatrix} \mathbf{1} & 0 \\ 0 & \mathbf{2} \end{bmatrix} = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$

and let $P = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} v_1 & v_2 \end{bmatrix}$
 it is clear that

$$AP = \begin{bmatrix} Av_1 & Av_2 \end{bmatrix}$$

$$= \begin{bmatrix} h_1 v_1 & h_2 v_2 \end{bmatrix}$$

$$= \begin{bmatrix} h_1 & h_2 \\ 0 & h_2 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} h_1 & 0 \\ 0 & h_2 \end{bmatrix}$$

$$= P \cdot D. \text{ Therefore, } AP = PD,$$

and so $A = P D P^{-1}$

• We compute e^A :

We have

$$\begin{aligned} e^A &= P \cdot e^D P^{-1} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} e & 0 \\ 0 & e^2 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} e & e^2 \\ 0 & e^2 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} e & e^2 - e \\ 0 & e^2 \end{bmatrix}. \end{aligned}$$

Exo 3: Let $A \in \mathcal{M}_n(\mathbb{R})$. We prove that the series $\sum_{k=0}^{+\infty} \frac{A^k}{k!}$

is absolutely Convergent.

For any $k \geq 0$, we have

$$\left\| \frac{A^k}{k!} \right\| \leq \frac{\|A\|^k}{k!}$$

We put

$$a_k = \frac{\|A\|^k}{k!}$$

Since $\lim_{k \rightarrow +\infty} \left| \frac{\frac{\|A\|^{k+1}}{(k+1)!}}{\frac{\|A\|^k}{k!}} \right| = \lim_{k \rightarrow +\infty} \frac{\|A\|}{k+1} = 0 < 1$

then $\sum_{k=0}^{\infty} \frac{\|A\|^k}{k!}$ is Convergent.

On the other hand, we have:

$$\left\| \sum_{k=0}^{\infty} \frac{A^k}{k!} \right\| \leq \sum_{k=0}^{\infty} \frac{\|A\|^k}{k!}$$

Thus, $\sum_{k=0}^{\infty} \frac{A^k}{k!}$ is absolutely Convergent.

Ex 04: Let A be a square matrix.
we prove that;

$$\lim_{\kappa \rightarrow 0} \frac{\kappa A (e^{\kappa A} - I)}{\kappa} = A.$$

Solution: By definition, we have;

$$e^{\kappa A} = I + \frac{\kappa A}{1!} + \frac{(\kappa A)^2}{2!} + \dots$$

Then

$$\begin{aligned} \|e^{\kappa A} - I - \kappa A\| &= \left\| \frac{(\kappa A)^2}{2!} + \frac{(\kappa A)^3}{3!} + \dots \right\| \\ &\leq \frac{\|\kappa A\|^2}{2!} + \frac{\|\kappa A\|^3}{3!} + \dots \\ &= e^{\|\kappa A\|} - 1 - \|\kappa A\|. \end{aligned}$$

Therefore,

$$\left\| \frac{e^{\kappa A} - I}{\kappa} - A \right\| \leq \left(\frac{e^{\|\kappa A\|} - 1 - \|\kappa A\|}{\|\kappa A\|} \right) \rightarrow 0$$

as κ tends to 0

$$\text{and so } \lim_{\kappa \rightarrow 0} \left\| \frac{e^{\kappa A} - I}{\kappa} - A \right\| \rightarrow 0$$

$$\Rightarrow \lim_{\kappa \rightarrow 0} \frac{e^{\kappa A} - I}{\kappa} \rightarrow A$$

Recall that

$$\lim_{\kappa \rightarrow 0} \frac{\kappa \cdot x (e^{\kappa x} - 1)}{\kappa} \rightarrow x$$

Ex 05: Consider the matrix

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

① We find the eigenvalues and eigenvectors:
After simple computation, we get:

- $P_A(x) = (x-2)(x+2)(x-4)$
- The eigenvectors are:

$$\begin{cases} \lambda_1 = -2 \\ \lambda_2 = 2 \\ \lambda_3 = 4 \end{cases}$$

- The eigenvectors are:

$$\begin{cases} \lambda_1 \rightarrow v_1 = (1, 0, 1) \\ \lambda_2 \rightarrow v_2 = (0, 1, 1) \\ \lambda_3 \rightarrow v_3 = (1, 1, 0) \end{cases}$$

② We deduce that A is diagonalizable.

That is, we can write the matrix A as:

$$A = P \cdot D \cdot P^{-1}, \text{ where } P \text{ is invertible and } D \text{ is diagonal.}$$

now, we put

$$P = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{bmatrix}, D = \begin{bmatrix} -2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{bmatrix}$$

After Computation, we get

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

③ We Compute e^A :

Since $A = P \cdot D \cdot P^{-1}$, then

$$e^A = P \cdot e^D \cdot P^{-1}, \text{ where}$$

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

Ex 06: Consider the two matrices:

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

We verify that:

$$e^{A+B} \neq e^A \cdot e^B \neq e^B \cdot e^A$$

• We have:

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

$$\text{Then } e^{A+B} = \begin{bmatrix} e^2 & 0 \\ 0 & 1 \end{bmatrix}$$

• We Compute e^A :

The matrix A is diagonalizable, because it has two simple eigenvalues: $\lambda_1 = 1$, $\lambda_2 = 0$.

We have:

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

and so

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

After computation, we have:

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

• We compute e^B :

Also, the matrix B is diagonalizable. In fact:

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

and hence,

$$\begin{aligned} e^B &= \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} e^0 & 0 \\ 0 & e^1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} \\ &= \begin{bmatrix} e & 1-e \\ 0 & 1 \end{bmatrix} \end{aligned}$$

Finally, we see that:

$$e^A \cdot e^B = \begin{bmatrix} e^2 & 2e - e - 1 \\ 0 & 1 \end{bmatrix}$$

and

$$e^B \cdot e^A = \begin{bmatrix} e^2 & e - 2e + 1 \\ 0 & 1 \end{bmatrix}$$

Ex 07: Consider the matrix :

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

We calculate $\ln A$.

For computation, we obtain :

$$A = \underset{\text{P}}{\begin{bmatrix} -1 & 1 \\ 1 & 1 \end{bmatrix}} \cdot \underset{\text{D}}{\begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{3}{2} \end{bmatrix}} \cdot \underset{\text{P}^{-1}}{\begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}}$$

and so

$$\begin{aligned} \ln A &= P \cdot \ln D \cdot P^{-1} \\ &= \begin{bmatrix} -1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \ln \frac{1}{2} & 0 \\ 0 & \ln \frac{3}{2} \end{bmatrix} \begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{2} \ln 3 - \ln 2 & \frac{1}{2} \ln 3 \\ \frac{1}{2} \ln 3 & \frac{1}{2} \ln 3 - \ln 2 \end{bmatrix} \end{aligned}$$

Ex 08: Let A be a diagonalizable matrix. That is, $A = P \cdot D \cdot P^{-1}$, where P is invertible and D is diagonal. Then

$$A^k = P \cdot D^k \cdot P^{-1} \text{ for } k \geq 0$$

$$\Rightarrow \lim_{k \rightarrow +\infty} A^k = \lim_{k \rightarrow +\infty} (P \cdot D^k \cdot P^{-1})$$

$$= P \left(\lim_{k \rightarrow +\infty} D^k \right) P^{-1}.$$

• For example, if

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

At first, we compute the eigenvectors and the corresponding eigenvalues.

$$\bullet \lambda_1 = 1 \rightarrow v_1 = (1, 1)$$

$$\bullet \lambda_2 = \frac{1}{4} \rightarrow v_2 = (-2, 1)$$

We deduce that A is diagonalizable. That is, we can factorize A as:

$$A = P \cdot D \cdot P^{-1}, \text{ where } D = \begin{bmatrix} 1 & 0 \\ 0 & \frac{1}{4} \end{bmatrix}$$

and

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

Therefore,

$$A^k = P \cdot D^k \cdot P^{-1}, \text{ where } P^{-1} = \begin{bmatrix} \frac{1}{3} & \frac{2}{3} \\ -\frac{1}{3} & \frac{1}{3} \end{bmatrix}$$

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

Hence,

$$\lim_{k \rightarrow +\infty} A^k = P \left(\lim_{k \rightarrow +\infty} D^k \right) P^{-1}$$

$$= P \cdot \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} P^{-1}$$

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

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

Ex 09: Let

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

We calculate A^k for $k \geq 0$.

First, after computation we get

$$A = \underbrace{\begin{bmatrix} -1 & -1 \\ 2 & 1 \end{bmatrix}}_P \cdot \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}}_D \cdot \underbrace{\begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix}}_{P^{-1}}$$

and so

$$\begin{aligned} A^k &= P \cdot D^k \cdot P^{-1} \\ &= \begin{bmatrix} -1 & -1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2^k \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \end{aligned}$$

It follows that:

$$A^k = \begin{bmatrix} 2^{k+1} - 1 & 2^k - 1 \\ 2 - 2^{k+1} & 2 - 2^k \end{bmatrix}, \quad k \geq 0.$$

In particular, for $k=0$, we have:

$$A^0 = I.$$

Ex 10: Consider the matrix:

$$A = \begin{bmatrix} 9 & 0 & 0 \\ -5 & 4 & 0 \\ -8 & 0 & 1 \end{bmatrix}$$

We compute A^k , $k \geq 0$:

The eigenvalues of A are:

$$\lambda_1 = 9, \lambda_2 = 4, \lambda_3 = 1.$$

The corresponding eigenvectors are:

$$v_1 = (-1, 1, 1), v_2 = (0, 1, 0), v_3 = (0, 0, 1).$$

Thus,

$$A = \underset{\substack{\text{P} \\ \text{red}}}{\begin{bmatrix} 0 & 0 & -1 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix}} \underset{\substack{\text{D} \\ \text{red}}}{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 9 \end{bmatrix}} \underset{\substack{\text{P}^{-1} \\ \text{red}}}{\begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ -1 & 0 & 0 \end{bmatrix}}$$

$$\text{and so } A^k = P D^k P^{-1}$$

$$= \begin{bmatrix} 0 & 0 & -1 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 4^k & 0 \\ 0 & 0 & 9^k \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ -1 & 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 9^k & 0 & 0 \\ 4^k - 9^k & 4^k & 0 \\ 1 - 9^k & 0 & 1 \end{bmatrix}, \quad k \geq 0.$$

Test N° 04

Ex 01: We solve the system of linear recurrence sequences:

$$(S) \dots \begin{cases} x_{n+1} = 3x_n - y_n & , \quad x_0 = 1 \\ y_{n+1} = -x_n + 3y_n & y_0 = 2. \end{cases}$$

We can write the system (S) as follows:

$$\begin{bmatrix} x_{n+1} \\ y_{n+1} \end{bmatrix} = \begin{bmatrix} 3 & -1 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} x_n \\ y_n \end{bmatrix}$$
$$X_{n+1} = A \cdot X_n$$

Therefore,

$$\begin{aligned} X_n &= A \cdot X_{n-1} \\ &= A (A \cdot X_{n-2}) = A^2 \cdot X_{n-2} \\ &= \dots \\ &= A^n \cdot X_0 \end{aligned}$$

Thus,

$$X_n = A^n \cdot X_0 \dots (*)$$

After Computation, we show that

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

$A = P \cdot D \cdot P^{-1}$

it follows that:

$$A^n = P \cdot D^n \cdot P^{-1}$$

$$= \begin{bmatrix} \frac{2^n + 4^n}{2} & \frac{2^n - 4^n}{2} \\ \frac{2^n - 4^n}{2} & \frac{2^n + 4^n}{2} \end{bmatrix}$$

From (*), we obtain:

$$\begin{aligned} X_n = \begin{bmatrix} x_n \\ y_n \end{bmatrix} &= \begin{bmatrix} \frac{2^n + 4^n}{2} & \frac{2^n - 4^n}{2} \\ \frac{2^n - 4^n}{2} & \frac{2^n + 4^n}{2} \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} \\ &= \begin{bmatrix} \frac{3}{2} \cdot 2^n - \frac{1}{2} \cdot 4^n \\ \frac{3}{2} \cdot 2^n + \frac{1}{2} \cdot 4^n \end{bmatrix} \end{aligned}$$

Thus, $x_n = 2^{n-1}(3 - 2^n)$ and $y_n = 2^{n-1}(3 + 2^n)$, $n \geq 0$.

Ex 02: Consider the determinant

$$\Delta_n = \begin{vmatrix} 2 & -1 & & & \\ -1 & 2 & -1 & & \\ & 0 & -1 & 2 & \\ & & & \ddots & \ddots \\ & & & -1 & 2 & -1 \\ & & & & -1 & 2 \end{vmatrix}, \text{ n by n}$$

This is the determinant of a tridiagonal matrix.

① We show that $\Delta_n = 2\Delta_{n-1} - \Delta_{n-2}$.

In fact, we have:

$$\Delta_n = \begin{vmatrix} \overset{+}{2} & \overset{-}{-1} & \overset{+}{0} & \overset{-}{0} & \dots & 0 & 0 & 0 \\ -1 & 2 & -1 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 & \dots & 0 & 0 & 0 \\ 0 & 0 & -1 & 2 & -1 & \dots & 0 & 0 & 0 \\ & & & \ddots & \ddots & \ddots & & & \\ 0 & 0 & 0 & 0 & 0 & \dots & -1 & 2 & -1 \\ 0 & 0 & 0 & 0 & 0 & \dots & 0 & -1 & 2 \end{vmatrix}$$

Thus,

$$\Delta_n = 2 \Delta_{n-1} +$$

$$\begin{vmatrix} + & -1 & -1 & 0 & 0 & \dots & 0 & 0 \\ - & 0 & 2 & -1 & 0 & \dots & 0 & 0 \\ + & 0 & -1 & 2 & -1 & \dots & 0 & 0 \\ \vdots & 0 & 0 & -1 & 2 & \dots & 0 & 0 \\ \vdots & 0 & 0 & 0 & 0 & \dots & 2 & -1 \\ \vdots & 0 & 0 & 0 & 0 & \dots & -1 & 2 \end{vmatrix}$$

Δ

Using an expansion according to the first Column of Δ we get:

$$\Delta = - \Delta_{n-2}$$

Therefore,

$$\Delta_n = 2 \Delta_{n-1} - \Delta_{n-2}$$

Now, we Calculate Δ_n in terms of n .

• Simple Computation, we have:

$$\Delta_1 = \det [2] = 2$$

$$\Delta_2 = \det \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} = 3.$$

$$D_3 = \det \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix} = 4.$$

Since $D_n = 2D_{n-1} - D_{n-2} \dots \dots (S)$

in the matrix form, we get

$$\begin{bmatrix} D_n \\ D_{n-1} \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} D_{n-1} \\ D_{n-2} \end{bmatrix}$$

$X_n = A \cdot X_{n-1} \dots \dots (*)$

where, $X_2 = \begin{bmatrix} D_2 \\ D_1 \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$

It follows from (*) that:

$$X_n = A^{n-2} \cdot X_2 \quad \text{نتوقف عند } X_2$$

• We must compute A^{n-2} .

□ Characteristic polynomial:

$$P_A(\lambda) = \begin{vmatrix} 2-\lambda & -1 \\ 1 & -\lambda \end{vmatrix} \quad \begin{matrix} L_1 \\ \downarrow \\ L_1 - L_2 \end{matrix}$$

$$= \begin{vmatrix} 1-x & -(1-x) \\ 1 & -x \end{vmatrix}$$

$$= (1-x) \cdot \begin{vmatrix} 1 & -1 \\ 1 & -x \end{vmatrix}$$

$$= (1-x) \cdot [-x + 1]$$

$$= (1-x)^2$$

Hence, $P_A(x) = (1-x)^2$

- The Corresponding eigenvectors :
We see that the matrix A has one eigenvalue $\lambda = 1$.

$$\begin{aligned} E_\lambda &= \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{cases} 2x - y = x \\ x = x \end{cases} \right\} \\ &= \left\{ (x, y) \in \mathbb{R}^2 \mid y = x \right\} \\ &= \text{Vect} \{ (1, 1) \} \rightarrow v = (1, 1) \end{aligned}$$

- The matrix A is not diagonalizable.
That is, we can not write it as : $A = P \Lambda P^{-1}$.

Then, we try to write it as:

$$A = P \cdot T \cdot \bar{P}^{-1}, \text{ where}$$

P is invertible and T is upper triangular, i.e. $T = \begin{bmatrix} 1 & x \\ 0 & 1 \end{bmatrix}$

We choose $v' = (1, 0)$ and put

$$P = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \Rightarrow \bar{P}^{-1} \text{ exists.}$$

$\underbrace{\quad}_{v'}$ $\underbrace{\quad}_{v'}$

$$\text{After computation} \Rightarrow \bar{P}^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix}$$

Also, we get

$$\begin{aligned} \bar{P}^{-1} A P &= \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} = T \end{aligned}$$

Thus,

$$A = P \cdot T \cdot \bar{P}^{-1}, \text{ and so}$$

$$A^k = P \cdot T^k \cdot \bar{P}^{-1}$$

Next, we Compute T^k :

In fact

$$T = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}}_D + \underbrace{\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}}_N$$

Remark: we use the Binomial formula to compute T^k , where $T = D + N$ with $DN = ND$.

- The matrix N is nilpotent عديمة القوى because: $N^2 = 0$. Then

$$T^k = (D + N)^k = \sum_{i=0}^k \binom{k}{i} D^{k-i} \cdot N^i$$

$$= \binom{k}{0} D^k N^0 + \binom{k}{1} D^{k-1} N + 0$$

$$= D^k + k D^{k-1} N \quad (D = I)$$

$$= I + k \cdot N$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 0 & k \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$$

$$\text{Hence, } T^k = \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}, \quad k \geq 0$$

Finally, we get

$$\begin{aligned} A^{n-2} &= P \cdot T^{n-2} \cdot P^{-1} \\ &= \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & n-2 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} \\ &= \begin{bmatrix} n-1 & 2-n \\ n-2 & 3-n \end{bmatrix} \end{aligned}$$

Since $X_n = A^{n-2} \cdot X_2$, then

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

That is,

$$\begin{bmatrix} D_n \\ D_{n-1} \end{bmatrix} = \begin{bmatrix} 3(n-1) + 2(2-n) \\ 3(n-2) + 2(3-n) \end{bmatrix}$$

and so

$$\begin{aligned} D_n &= 3(n-1) + 2(2-n) \\ &= n+1. \end{aligned}$$

Ex 03: Let (x_n) be the sequence given by :

$$x_{n+2} = \frac{2}{\frac{1}{x_n} + \frac{1}{x_{n+1}}} \quad \begin{matrix} x_0 > 0 \\ x_1 > 0 \end{matrix}$$

$y_0 = x_0$ and $\lim_{n \rightarrow +\infty} x_n$ $\rightarrow$ $\lim_{n \rightarrow +\infty} y_n$

We evaluate the limit as n tends to ∞ .

We put

$$y_n = \frac{1}{x_n}, \text{ then}$$

$$\frac{1}{y_{n+2}} = \frac{2}{y_n + y_{n+1}},$$

and so

$$2 y_{n+2} = y_n + y_{n+1}$$

that is,

$$y_{n+2} = \frac{1}{2} y_n + \frac{1}{2} y_{n+1}$$

In the matrix form, we get

$$\begin{bmatrix} y_{n+2} \\ y_{n+1} \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ 1 & 0 \end{bmatrix} \begin{bmatrix} y_{n+1} \\ y_n \end{bmatrix}$$

and so

$$\begin{bmatrix} y_n \\ y_{n-1} \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ 1 & 0 \end{bmatrix} \begin{bmatrix} y_{n-1} \\ y_{n-2} \end{bmatrix}$$

$$\begin{aligned} X_n &= A \cdot X_{n-1} \\ &= \dots \\ &= A^{n-1} \cdot X_1, \text{ where} \end{aligned}$$

$$X_1 = \begin{bmatrix} y_1 \\ y_0 \end{bmatrix} = \begin{bmatrix} \frac{1}{x_1} \\ \frac{1}{x_0} \end{bmatrix}$$

• We must compute A^k , $k \geq 0$:

By few Computation, we have

$$\begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ 1 & 0 \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & -1 \\ 1 & 2 \end{bmatrix}}_P \cdot \underbrace{\begin{bmatrix} 0 & 0 \\ 0 & -\frac{1}{2} \end{bmatrix}}_D \cdot \underbrace{\begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 1 \end{bmatrix}}_{P^{-1}}$$

and have $A^k = P \cdot D^k \cdot P^{-1}$

Thus,

$$A^k = \begin{bmatrix} \frac{1}{3} \left(-\frac{1}{2}\right)^k + \frac{2}{3} & \frac{1}{3} - \frac{1}{3} \left(-\frac{1}{2}\right)^k \\ \frac{2}{3} - \frac{2}{3} \left(-\frac{1}{2}\right)^k & \frac{2}{3} \left(-\frac{1}{2}\right)^k + \frac{1}{3} \end{bmatrix}$$

and so

$$y_n = \frac{1}{x_1} \left[\frac{1}{3} \left(-\frac{1}{2}\right)^{n-1} + \frac{2}{3} \right] + \frac{1}{x_0} \left[\frac{1}{3} - \frac{1}{3} \left(-\frac{1}{2}\right)^{n-1} \right]$$

$$\Rightarrow \lim_{n \rightarrow +\infty} y_n = \frac{2}{3x_1} + \frac{1}{3x_0}$$

and hence

$$\lim_{n \rightarrow +\infty} x_n = \frac{3}{\frac{2}{x_1} + \frac{1}{x_0}}$$

Ex 04: Consider the system of differential equations:

$$X' = A \cdot X, \text{ where}$$

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

Here,

$$X' = \begin{bmatrix} x' \\ y' \\ z' \end{bmatrix}$$

Thus,

$$\begin{cases} x' = x \\ y' = x + 2y \\ z' = x + 2z \end{cases}$$

الدوال x, y, z
و t بدلالة t

① We find the eigenvalues and eigenvectors.

$$\begin{cases} \lambda_1 = 1 \rightarrow v_1 = (-1, 1, 1) \\ \lambda_2 = 2 \rightarrow \begin{cases} v_2 = (0, 1, 0) \\ v_3 = (0, 0, 1) \end{cases} \end{cases}$$

In this case, the matrix A is diagonalizable

Thus,

$$X(t) = c_1 e^{\lambda_1 t} v_1 + c_2 e^{\lambda_2 t} v_2 + c_3 e^{\lambda_2 t} v_3$$

That is,

$$X(t) = c_1 e^{t} \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix} + c_2 e^{0 \cdot t} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + c_3 e^{0 \cdot t} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix},$$

$c_1, c_2, c_3 \in \mathbb{R}$

Example: Solve the differential system

$$\begin{cases} x' = x + 3y \\ y' = 3x + y \end{cases} \quad \dots (S')$$

the system (S') can be written as:

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$X' = A \cdot X$

• The eigenpairs of A are:

$$\lambda_1 = -2 \rightarrow v_1 = (-1, 1)$$

$$\lambda_2 = 4 \rightarrow v_2 = (1, 1)$$

• The matrix A is diagonalizable, and so

$$X(t) = c_1 e^{-2t} v_1 + c_2 e^{4t} v_2$$

$$= c_1 e^{-2t} \begin{bmatrix} -1 \\ 1 \end{bmatrix} + c_2 e^{4t} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

thus,

$$\begin{cases} x = -c_1 e^{-2t} + c_2 e^{4t} \\ y = c_1 e^{-2t} + c_2 e^{4t} \end{cases}, \quad c_1, c_2 = C^{te}$$

ثابتين

Ex 05: Using Cayley-Hamilton theorem
we calculate the inverse of the matrix:

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

First, we find the characteristic polynomial:

$$P_A(x) = \begin{vmatrix} \overset{+}{1-x} & \overset{-}{1} & \overset{+}{0} \\ -1 & \overset{-}{-x} & 0 \\ 2 & 0 & \overset{-}{-1-x} \end{vmatrix}$$

$$= x^3 + 1 \quad (\text{after computation})$$

By Cayley-Hamilton theorem, we get

$$P_A(A) = 0 \equiv \text{zero matrix.}$$

$$\Rightarrow A^3 + I = 0$$

$$\Rightarrow A^3 = -I \Rightarrow A^2 = -A^{-1}$$

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

$$\Rightarrow \boxed{A^{-1} = -A^2}$$

بعد الحساب

Ex 06: Let A be a square matrix. Assume that the characteristic polynomial of A is given by :

$$P_A(x) = c_0 + c_1 x + \dots + c_{n-1} x^{n-1} + c_n x^n,$$

where $c_0 \neq 0$.

For example, in Ex 05, we have :

$$P_A(x) = 1 + x^3, \text{ i.e. } c_0 = 1 \neq 0.$$

Applying Cayley-Hamilton theorem, we get

$$P_A(A) = 0. \text{ That is,}$$

$$c_0 I + c_1 A + \dots + c_{n-1} A^{n-1} + c_n A^n = 0.$$

It follows that

$$c_1 A + \dots + c_{n-1} A^{n-1} + c_n A^n = -c_0 I$$

and hence,

$$c_1 I + c_2 A + \dots + c_n A^{n-1} = -c_0 A^{-1}$$

Thus,

$$A^{-1} = \frac{-1}{c_0} (c_1 I + c_2 A + \dots + c_n A^{n-1})$$

The proof is finished. $\square$

Ex 07: Let $A \in M_n(\mathbb{R})$ be a square matrix with $\text{Sp}(A) = \{h\}$. That is, A has one eigenvalue h . We show that

$$e^{tA} = e^{ht} \sum_{k=0}^{n-1} \frac{(A-hI)^k t^k}{k!}.$$

Proof: First, we have: $p_A(x) = (x-h)^n$.

Then

$$e^{tA} = e^{tA + htI - htI}$$

$$= e^{htI + t(A-hI)}$$

$$= e^{htI} \cdot e^{t(A-hI)}$$

$$= e^{ht} \cdot e^{t(A-hI)}$$

$$\left(\alpha I \cdot B = \alpha \cdot B \right)$$

$$= e^{ht} \sum_{k=0}^{+\infty} \frac{(A-hI)^k t^k}{k!}$$

$$= e^{ht} \sum_{k=0}^{n-1} \frac{(A-hI)^k t^k}{k!}$$

because by Cayley-Hamilton theorem,

$$P_A(A) = (A - hI)^n = 0.$$

The proof is finished $\square$.

Now, we solve the system of differential equations:

$$\begin{cases} x' = 2x + y \\ y' = 2y \end{cases}$$

in the matrix form:

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$X' = A \cdot X$$

The solution is given by:

$$\begin{aligned} X(t) &= e^{tA} \cdot c, \text{ where } c = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} \\ &= e^{ht} \left[I + t(A - hI) \right] \\ &= e^{2t} \left[\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + t \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \right] \end{aligned}$$

$$= e^{2t} \begin{bmatrix} 1 & t \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} e^{2t} & t e^{2t} \\ 0 & e^{2t} \end{bmatrix}$$

Thus,

$$X(t) = \begin{bmatrix} c_1 e^{2t} + c_2 t e^{2t} \\ c_2 e^{2t} \end{bmatrix}$$

and so

$$\begin{cases} x = c_1 e^{2t} + c_2 t e^{2t} \\ y = c_2 e^{2t} \end{cases}, \quad c_1, c_2 \in \mathbb{R}.$$

- Solve the system of differential equations:

$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} -4 & 1 & 1 \\ 1 & -1 & -2 \\ -2 & 1 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

$$X' = A \cdot X$$

The characteristic polynomial of A is given by (after computation).

$$p_A(\lambda) = (\lambda + 2)^3.$$

Then A has one eigenvalue $\lambda = -2$,

and so,

$$e^{tA} = e^{ht} \left[I + t(A - hI) + \frac{t^2}{2}(A - hI)^2 \right]$$

The solution of $X' = AX$ is given by:

$$X(t) = e^{tA} \cdot c, \text{ where } c = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix},$$

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

$$\bullet A - hI = A + 2I = \begin{bmatrix} -2 & 1 & 1 \\ 1 & 1 & -2 \\ -2 & 1 & 1 \end{bmatrix}$$

$$\bullet (A - hI)^2 = (A + 2I)^2 = \begin{bmatrix} 3 & 0 & -3 \\ 3 & 0 & -3 \\ 3 & 0 & -3 \end{bmatrix}$$

! ~~side~~ S_i

The characteristic polynomial of A is given by (after computation).

$$p_A(\lambda) = (\lambda + 2)^3.$$

Then A has one eigenvalue $\lambda = -2$,

and so,

$$e^{tA} = e^{ht} \left[I + t(A - hI) + \frac{t^2}{2}(A - hI)^2 \right]$$

The solution of $X' = AX$ is given by:

$$X(t) = e^{tA} \cdot c, \text{ where } c = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix},$$

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

$$\bullet A - hI = A + 2I = \begin{bmatrix} -2 & 1 & 1 \\ 1 & 1 & -2 \\ -2 & 1 & 1 \end{bmatrix}$$

$$\bullet (A - hI)^2 = (A + 2I)^2 = \begin{bmatrix} 3 & 0 & -3 \\ 3 & 0 & -3 \\ 3 & 0 & -3 \end{bmatrix}$$

! ~~is the~~ S_i

- There is a few hand Case when $A \in M_3(\mathbb{R})$ with $Sp(A) = \{h, h, \mu\}$. For example:

$$A = \begin{bmatrix} 2 & -1 & 2 \\ 1 & -5 & 7 \\ 4 & -2 & 2 \end{bmatrix} \dots (*)$$

If the matrix A is not diagonalizable, then the solution of $X' = AX$ is:

$$X(t) = e^{tA} \cdot c, \text{ where } c = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix}$$

Using Cayley-Hamilton theorem (see the Course) we can prove that,

$$e^{tA} = e^{ht} \left[I + t(A - hI) \right] + \frac{e^{ht} - e^{\mu t}}{(\mu - h)^2} (A - hI)^2 - \frac{t e^{ht}}{\mu - h} (A - hI)^2 \dots (**)$$

In the matrix $(*)$, we have

- A is not diagonalizable
- $h = 0$ (with multiplicity 2)
- $\mu = -1$ (simple). So, we must compute e^{tA} using the long formula $(**)$. For the proof see the Course.

Test No 05

Ex 01: We calculate the minimal polynomial of the matrix:

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

Since A is an upper triangular matrix, then

$$p_A(x) = (1-x)^2. \text{ Thus,}$$

$$m_A(x) = \begin{cases} 1-x \\ \text{or} \\ (1-x)^2 \end{cases}$$

on the other hand, since $m_A(A) = 0$, then

$$m_A(x) = (1-x)^2 = p_A(x).$$

Recall that:

$$(A \text{ is diagonalizable}) \Leftrightarrow \left(\begin{array}{l} \text{the roots of} \\ m_A(x) \text{ are} \\ \text{simple.} \end{array} \right)$$

• Let $A = \begin{bmatrix} 1 & c \\ 0 & 1 \end{bmatrix}, c \neq 0$

Since $m_A(x) = (1-x)^2 \Rightarrow A$ is not diagonalizable.

Ex 02: We determine the minimal polynomial of the matrices:

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

- First, we see that $p_A(x) = x^3$. Then

$$m_A(x) = \begin{cases} x \\ \text{if} \\ x^2 \\ \text{if} \\ x^3 \end{cases}$$

Since

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

Hence,

$$\boxed{m_A(x) = x^2}$$

[6]

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

• Characteristic polynomial:

$$p_A(x) = (x-6)(x-3)^2. \text{ Thus,}$$

$$m_A(x) = \begin{cases} (x-6)(x-3) \\ \text{or} \\ (x-6)(x-3)^2 \end{cases}$$

$$\text{But, } (A-6I)(A-3I) =$$

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

Hence,

$$m_A(x) = (x-6)(x-3).$$

Recall that the minimal polynomial satisfies the following three conditions:

1. $m_A(x)$ divides $p_A(x)$ पूर्व.
2. $m_A(x)$ and $p_A(x)$ have the same roots.
3. $m_A(A) = p_A(A) = 0$.

Ex 1 Let

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

• Characteristic polynomial:

$$p_A(x) = (x-1)^2. \quad \text{Thus,}$$

$$m_A(x) = \begin{cases} x-1 \\ \text{or} \\ (x-1)^2 \end{cases}$$

Since $A - I = \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} \neq 0$, then

$$m_A(x) = (x-1)^2.$$

Ex 03: Assume that $m(x) = (x-a)(x-b)$, where $a, b \in \mathbb{R}$ ($a \neq b$).

We show that A^n can be only written in terms of A and I . In fact:

$$m_A(x) = x^2 - (a+b)x + ab.$$

$$\text{Thus, } A^2 - (a+b)A + ab \cdot I = 0.$$

$$\Rightarrow A^2 = (a+b)A - ab \cdot I$$

هذا يعني أن A^2 يمكن كتابته بدلالة A و I .
Thus, A^2 is written in terms of A and I .

Now, assume that

$$A^n = \alpha_n \cdot A + \beta_n \cdot I, \text{ where } \alpha_n, \beta_n \in \mathbb{R}$$

Therefore,

$$\begin{aligned} A^{n+1} &= A \cdot A^n = A (\alpha_n A + \beta_n \cdot I) \\ &= \alpha_n A^2 + \beta_n \cdot A \end{aligned}$$

$$= \alpha_n \left[(a+b) A - ab I \right] + \beta_n \cdot A$$

$$= \left[\underbrace{\alpha_n (a+b) + \beta_n}_{\in \mathbb{R}} \right] A - \underbrace{\alpha_n ab}_{\in \mathbb{R}} \cdot I$$

That is, A^{n+1} is written in terms of A and I .

The proof is finished $\square$.

Remark. In the proof of Ex 03, we have used the proof by induction. عاجل

Remark. If a matrix has the minimal polynomial:

$$m_A(x) = (x-a)(x-b),$$

then

$$f_A(x) = (x-a)^\alpha \cdot (x-b)^\beta, \text{ where}$$

$$\alpha + \beta = n.$$

Ex 04: We find the minimal polynomial of the matrix

$$A_n = \begin{bmatrix} h & & & \\ 1 & h & & \\ & 1 & h & \\ & & \ddots & \\ & & & 1 & h \end{bmatrix} \quad n \text{ by } n$$

• For $n = 2$:

$$A_2 = \begin{bmatrix} h & 0 \\ 1 & h \end{bmatrix}, \quad m_{A_2}(x) = (x - h)^2$$

• For $n = 3$,

$$A_3 = \begin{bmatrix} h & 0 & 0 \\ 1 & h & 0 \\ 0 & 1 & h \end{bmatrix}, \quad m_{A_3}(x) = (x - h)^3$$

• For $n = 4$,

$$A_4 = \begin{bmatrix} h & 0 & 0 & 0 \\ 1 & h & 0 & 0 \\ 0 & 1 & h & 0 \\ 0 & 0 & 1 & h \end{bmatrix},$$

$$m_{A_4}(x) = (x - h)^4$$

By induction, we deduce that

$$m_A^n(x) = (x - \lambda)^n.$$

Ex 05: Consider the matrix

$$A = \begin{bmatrix} a & b & b \\ b & a & b \\ b & b & a \end{bmatrix}, \quad a, b \in \mathbb{R}.$$

We show that A is diagonalizable.

In fact, we see that:

$$A = a \cdot \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} + b \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}$$

$$= a \cdot I + b \cdot C, \quad \text{where}$$

$$C = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}.$$

• The characteristic polynomial of the matrix C is:

$$p(x) = (1+x)^2(x-2)$$

This means that $m_C(x) = (1+x)(x-2)$ or
 $m_C(x) = (1+x)^2(x-2).$

But,

$$\begin{aligned} (I + C)(C - 2I) &= \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} -2 & 1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix} \\ &= \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}. \end{aligned}$$

Hence,

$m_C(x) = (1+x)(x-2)$ which has simple roots. Then C is diagonalizable. that is C can be written as:

$C = P \cdot D \cdot P^{-1}$, where P is invertible and D is diagonal. Therefore,

$$A = a \cdot I + b \cdot C$$

$$= a P \cdot P^{-1} + b P D P^{-1}$$

$$= P \left[\underbrace{a \cdot I + D}_{D'} \right] P^{-1} = P \cdot D' \cdot P^{-1}, \text{ where}$$

D' is diagonal.

Fact:

$(A \text{ is diagonalizable}) \iff \left(\begin{array}{l} \text{The roots of } m_A(x) \\ \text{are all simple.} \end{array} \right)$

Ex 06: Consider the matrix

$$A_n = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \dots & \dots & \dots & \dots \\ 1 & 1 & \dots & 1 \end{bmatrix} \quad n \text{ by } n$$

We show that A_n is diagonalizable.

We have proved before that:

$$f_{A_n}(x) = \begin{cases} x^{n-1}(x-n), & \text{if } n \text{ is even} \\ x^{n-1}(n-x), & \text{if } n \text{ is odd} \end{cases}$$

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Here, we see that:

$$\left\{ \begin{array}{l} A_n (A_n - nI) = 0 \\ \text{and} \\ A_n (nI - A_n) = 0. \end{array} \right. \quad \text{For example:}$$

$$\begin{aligned} A_n (A_n - nI) &= \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \dots & \dots & \dots & \dots \\ 1 & 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} 1-n & 1 & \dots & 1 \\ 1 & 1-n & & 1 \\ \dots & \dots & \dots & \dots \\ 1 & 1 & & 1-n \end{bmatrix} \\ &= \begin{bmatrix} 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 0 \end{bmatrix} \end{aligned}$$

Hence

$$A_n(u) = \begin{cases} u(u-n) \\ \text{or} \\ u(n-u) \end{cases}$$

and so A_n is diagonalizable.

Ex 07: Let

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

we diagonalize the matrix A .

- It seems that A is not diagonalizable. But, we try to write it as:

$$A = P \cdot T \cdot P^{-1}, \text{ where}$$

P is invertible

T is an upper triangular matrix
مشابهة عليا

That is, A is similar to the matrix T .

- From simple computation, we have

$$p_A(x) = (x - 3)^2.$$

This means that $\lambda = 3$ is the only eigenvalue of A with multiplicity 2. Hence A is not diagonalizable since $A \neq 3 \cdot I$.

Next, we find the corresponding eigenvector

In fact, we have:

$$E_\lambda = \left\{ (x, y) \in \mathbb{R}^2 \mid \begin{array}{l} 2x - y = 3x \\ x + 4y = 3y \end{array} \right\}$$

$$= \left\{ (x, y) \in \mathbb{R}^2 \mid y = -x \right\}$$

$$= \text{Vect} \{ (1, -1) \}. \text{ Hence, } v_1 = (1, -1).$$

Let v_2 be a nonzero vector for which $\{v_1, v_2\}$ is a basis of $\mathbb{R}^2$. For example, we put $v_2 = (1, 1)$, and let

$$P = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$

Therefore,

$$P^{-1} A P = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$

After Computation, we get

$$P^{-1} A P = \begin{bmatrix} 3 & -2 \\ 0 & 3 \end{bmatrix} = T.$$

That is, $A = P \cdot T \cdot P^{-1} \Rightarrow A$ is trigonalizable.

and so

$$A^n = P T^n P^{-1}$$

• It suffices to compute T^n .

We write the matrix T as follows:

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

$$T = D + N$$

where, $N^2 = 0$. Hence,

$$T^n = C_n^0 D^n + C_n^1 D^{n-1} \cdot N$$

$$= D^n + n D^{n-1} \cdot N$$

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

$$= \begin{bmatrix} 3^n & -2n \cdot 3^{n-1} \\ 0 & 3^n \end{bmatrix}, \quad n \geq 0.$$

Finally, we deduce that:

$$A^n = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 3^n & -2n \cdot 3^{n-1} \\ 0 & 3^n \end{bmatrix} \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$= \begin{bmatrix} 3^n - n \cdot 3^{n-1} & -n \cdot 3^{n-1} \\ n \cdot 3^{n-1} & n \cdot 3^{n-1} + 3^n \end{bmatrix}$$

Ex 08: We trigonalize the matrix

$$A = \begin{bmatrix} 2 & -1 & 2 \\ 10 & -5 & 7 \\ 4 & -2 & 2 \end{bmatrix}$$

That is, we write A in the form:

$$A = P \cdot T \cdot P^{-1}$$

مثلية عليا

• Characteristic polynomial:

$$P_A(x) = x^2(x+1)$$

• eigenvectors

$$\lambda_1 = 0 \rightarrow v_1 = (1, 2, 0)$$

$$\lambda_2 = -1 \rightarrow v_2 = (-1, 1, 2)$$

λ_2 has one eigenvector $\Rightarrow A$ is not diagonalizable.

Now, we choose $v_3 = (0, 0, 1)$, and put

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

thus,

$$\begin{aligned} P^{-1} A P &= \begin{bmatrix} \frac{1}{3} & \frac{1}{3} & 0 \\ -\frac{2}{3} & \frac{1}{3} & 0 \\ \frac{4}{3} & -\frac{2}{3} & 1 \end{bmatrix} \begin{bmatrix} 2 & -1 & 2 \\ 10 & -5 & 7 \\ 4 & -2 & 2 \end{bmatrix} \begin{bmatrix} 1 & -1 & 0 \\ 2 & 1 & 0 \\ 0 & 2 & 1 \end{bmatrix} \\ &= \begin{bmatrix} 0 & 0 & 3 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{bmatrix} = T. \end{aligned}$$

$\Rightarrow A$ is trigonalizable.

Ex 09 We trigonalize the matrix

$$A = \begin{bmatrix} 7 & -6 & -2 \\ 2 & 0 & -1 \\ 2 & -3 & 2 \end{bmatrix}$$

• Characteristic polynomial

$$P_A(x) = (x - 3)^3.$$

Thus, A has one eigenvalue $\lambda = 3$ with multiplicity 3.

• The Corresponding eigenvectors:

$$E_{\lambda} = \left\{ (x, y, z) \in \mathbb{R}^3 \left/ \begin{array}{l} 7x - 6y - 2z = 3x \\ 2x - z = 3y \\ 2x - 3y + 2z = 3z \end{array} \right. \right\}$$

$$= \left\{ (x, y, z) \in \mathbb{R}^3 \left/ z = 2x - 3y \right. \right\}$$

$$= \left\{ (x, y, 2x - 3y), x, y \in \mathbb{R} \right\}$$

$$= \left\{ x(1, 0, 2) + y(0, 1, -3); x, y \in \mathbb{R} \right\}$$

$$= \text{Vect} \{ (1, 0, 2), (0, 1, -3) \}$$

Hence $\lambda = 3$ has two eigenvectors:

$$\begin{cases} v_1 = (1, 0, 2) \\ v_2 = (0, 1, -3) \end{cases}$$

Thus, A is not diagonalizable.

• We choose a vector v_3 such that $\{v_1, v_2, v_3\}$ is a basis of $\mathbb{R}^3$. For example,

$$v_3 = (1, 0, 0).$$

We put

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

it follows that:

$$\begin{aligned} P^{-1} A P &= \begin{bmatrix} 0 & \frac{3}{2} & \frac{1}{2} \\ 0 & 1 & 0 \\ 1 & -\frac{3}{2} & -\frac{1}{2} \end{bmatrix} \begin{bmatrix} 7 & -6 & -2 \\ 2 & 0 & -1 \\ 2 & -3 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 2 & -3 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 3 & 0 & 4 \\ 0 & 3 & 2 \\ 0 & 0 & 3 \end{bmatrix} = T. \end{aligned}$$

$\Rightarrow A = P \cdot T \cdot P^{-1} \Rightarrow A$ is trigonalizable.

$$\Rightarrow A^n = P \cdot T^n \cdot P^{-1}$$

• It suffices to compute T^n :

T is an upper triangular matrix $\Rightarrow T^n$ is computed by using the Binomial formula:

$$T = \underbrace{\begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}}_D + \underbrace{\begin{bmatrix} 0 & 0 & 4 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix}}_N$$

where $DN = ND$.

$$\Rightarrow T^n = (D + N)^n = \sum_{i=0}^n C_n^i D^{n-i} N^i$$

• The matrix N is nilpotent ($N^2 = 0$)

Thus,

$$\begin{aligned} T^n &= C_n^0 D^n + C_n^1 D^{n-1} N \\ &= D^n + n \cdot D^{n-1} N \end{aligned}$$

$$\begin{aligned} &= \begin{bmatrix} 3^n & 0 & 0 \\ 0 & 3^n & 0 \\ 0 & 0 & 3^n \end{bmatrix} + n \begin{bmatrix} 3^{n-1} & 0 & 0 \\ 0 & 3^{n-1} & 0 \\ 0 & 0 & 3^{n-1} \end{bmatrix} \begin{bmatrix} 0 & 0 & 4 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 3^n & 0 & 4n \cdot 3^{n-1} \\ 0 & 3^n & 2n \cdot 3^{n-1} \\ 0 & 0 & 3^n \end{bmatrix}, \quad n \geq 0. \end{aligned}$$

Finally, we obtain:

$$A^n = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 2 & -3 & 0 \end{bmatrix} \begin{bmatrix} 3^n & 0 & 4n \cdot 3^{n-1} \\ 0 & 3^n & 2n \cdot 3^{n-1} \\ 0 & 0 & 3^n \end{bmatrix} \begin{bmatrix} 0 & \frac{3}{2} & \frac{1}{2} \\ 0 & 1 & 0 \\ 1 & -\frac{3}{2} & -\frac{1}{2} \end{bmatrix}$$

$$= \begin{bmatrix} 4n \cdot 3^{n-1} + 3^n & -2n \cdot 3^n & -2n \cdot 3^{n-1} \\ 2n \cdot 3^{n-1} & \frac{n}{3}(1-n) & -n \cdot 3^{n-1} \\ 2n \cdot 3^{n-1} & -n \cdot 3^n & 3^n - n \cdot 3^{n-1} \end{bmatrix}$$

Ex 10: Consider the matrix:

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

□ We verify that $A^2 = A + 2I$:

In fact:

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

Moreover, we have:

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

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

Thus, $A^2 = A + 2I$.

• Since $A^2 = A + 2I$, then $A^2 - A = 2I$
and so $A(A - I) = 2I$

$$\Rightarrow A - I = 2A^{-1}$$

$$\Rightarrow \boxed{A^{-1} = \frac{1}{2}(A - I)}$$

This is the expression of A^{-1} in terms of A and I .

[2] We use the proof by induction to show that;

$$A^n = a_n \cdot A + b_n \cdot I, \text{ where.}$$

$a_n, b_n \in \mathbb{N}$ and

$$(S) \dots \begin{cases} a_{n+1} = a_n + b_n \\ b_{n+1} = 2a_n \end{cases}, \quad n \geq 0.$$

- For $n = 1$, we have:

$$A^1 = 1 \times A + 0 \cdot I = a_1 A + b_1 I$$

- For $n = 2$, we also have:

$$A^2 = A + 2I = a_2 A + b_2 I$$

where:

$$\begin{cases} a_2 = a_1 + b_1 \\ b_2 = 2a_1 \end{cases}$$

Now, assume that $A^n = a_n A + b_n I$, where a_n, b_n satisfy the system (S). Then

$$A^{n+1} = A \cdot A^n$$

$$= A \left[a_n A + b_n I \right]$$

$$= a_n A^2 + b_n A$$

$$= a_n (A + 2I) + b_n A$$

$$= (a_n + b_n) A + 2a_n I$$

$$= a_{n+1} A + b_{n+1} I$$

The proof is finished $\square$.

[3] We write the system (S') in the matrix form :

$$(S') \Leftrightarrow \begin{bmatrix} a_{n+1} \\ b_{n+1} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 2 & 0 \end{bmatrix} \begin{bmatrix} a_n \\ b_n \end{bmatrix}$$

$$X_{n+1} = A \cdot X_n$$

Therefore,

$$X_n = A^{n-1} \cdot X_1$$

$$X_1 = \begin{bmatrix} a_1 \\ b_1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

[4] • We Compute A^k :

After Computation, we get

$$\begin{bmatrix} 1 & 1 \\ 2 & 0 \end{bmatrix} = \underbrace{\begin{bmatrix} -1 & 1 \\ 2 & 1 \end{bmatrix}}_P \underbrace{\begin{bmatrix} -1 & 0 \\ 0 & 2 \end{bmatrix}}_D \underbrace{\begin{bmatrix} -\frac{1}{3} & \frac{1}{3} \\ \frac{2}{3} & \frac{1}{3} \end{bmatrix}}_{P^{-1}}$$

and so

$$A^k = \begin{bmatrix} \frac{(-1)^k + 2^{k+1}}{3} & \frac{2^k - (-1)^k}{3} \\ \frac{2^{k+1} - 2(-1)^k}{3} & \frac{2(-1)^k + 2^k}{3} \end{bmatrix}$$

it follows that

$$\begin{cases} a_n = \frac{2^n - (-1)^n}{3}, & n \geq 0 \\ b_n = \frac{2(-1)^n + 2^n}{3}, & n \geq 0 \end{cases}$$

[5] We Calculate the minimal polynomial of A :

First, we have

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

• characteristic polynomial:

$$P_A(x) = (x-2)(x+1)^2. \text{ Then}$$

$$m_A(x) = \begin{cases} (x-2)(x+1) \\ \text{or} \\ (x-2)(x+1)^2 \end{cases}$$

Since

$$(A - 2I)(A + I) = \begin{bmatrix} -2 & 1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \\ = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Thus, $m_A(x) = (x-2)(x+1)$.

$\Rightarrow$ We deduce that A is diagonalizable
Since the roots of $m_A(x)$ are simple.

[6] We solve the system of differential equations:

$$\begin{cases} x' = y + z \\ y' = x + z \\ z' = x + y \end{cases} \quad \dots (S')$$

- We must Compute the eigenvectors:

$$\begin{cases} \lambda_1 = -1 \longrightarrow \begin{cases} v_1 = (-1, 1, 0) \\ v_2 = (-1, 0, 1) \end{cases} \\ \lambda_2 = 2 \longrightarrow v_3 = (1, 1, 1) \end{cases}$$

Then

$$\begin{aligned} X(t) &= c_1 e^{-t} v_1 + c_2 e^{-t} v_2 + c_3 e^{2t} v_3 \\ &= c_1 e^{-t} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} + c_2 e^{-t} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} + c_3 e^{2t} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \end{aligned}$$

where $c_1, c_2, c_3 \in \mathbb{R}$.

Hence,

$$\begin{cases} x = -c_1 e^{-t} - c_2 e^{-t} + c_3 e^{2t} \\ y = c_1 e^{-t} + c_3 e^{2t} \\ z = c_2 e^{-t} + c_3 e^{2t} \end{cases}$$



End

20 November 2021
By Bellaouar. Dj.

Some results with proof

1) If A is diagonalizable $\Rightarrow \det(A) = \prod \lambda_i$
 $0 \notin \text{Sp}(A) \Leftrightarrow A^{-1}$ exists

2) (λ, μ) eigenpair of $A \Rightarrow (e^\lambda, e^\mu)$ eigenpair of e^A .

3) $A = P \Lambda P^{-1} \Rightarrow e^A = P e^\Lambda P^{-1}, f(A) = P f(\Lambda) P^{-1}$

4) A nilpotent + diagonalizable $\Rightarrow A = 0$

5) A diagonalizable + $\text{Sp}(A) = \{\lambda\} \Rightarrow A = \lambda \cdot I$.

6) A, B diagonalizable

with $\left. \begin{aligned} A &= P \Lambda_1 P^{-1} \\ B &= P \Lambda_2 P^{-1} \end{aligned} \right\} \Rightarrow AB = BA.$

7) $A \cdot e^A = e^A \cdot A$, 8) A is idempotent $\Rightarrow e^A = I + (e-1) \cdot A$

9) $P_A(A) = 0$ (zero matrix): Cayley-Hamilton Theorem.

10) If A is strictly triangular $\Rightarrow A$ is nilpotent

11) $\sum \lambda_i = \text{tr}(A)$ and $\det(e^A) = e^{\text{tr}(A)}$

12) N is nilpotent $\Rightarrow \text{Sp}(N) = \{0\}$

13) N is nilpotent $\Rightarrow I - N$ is invertible i.e. $(I - N)^{-1}$ exists.

14) $A \cdot B = B \cdot A \Rightarrow e^{A+B} = e^A \cdot e^B = e^B \cdot e^A$

15) $A \sim 2A \Rightarrow A$ is nilpotent 16) $(e^A)^{-1} = e^{-A}$

17) If $AB = BA$, then

(Binomial formula) $(A+B)^n = \sum_{i=0}^n \binom{n}{i} A^i B^{n-i} = \sum_{i=0}^n \binom{n}{i} A^{n-i} B^i$

18) $A \cdot (\text{com}(A))^t = \det(A) \cdot I_n$

19) $\forall A \in M_n(\mathbb{C}) : A = P \cdot T \cdot P^{-1}$, where
 P is invertible, T is an upper triangular matrix.

20) (A is diagonalizable + $\text{Sp}(A) = \{-1, 1\}$) $\Rightarrow A^{-1} = A$

Ex 01: why? Finding powers of diagonalizable matrices is easy!

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

Find an expression for A^k for any positive integer k .

Let $B = \begin{bmatrix} -2 & 2 & 1 \\ 0 & -2 & 1 \\ 0 & 0 & -2 \end{bmatrix}$ Compute B^k for $k \geq 1$.

Deduce the minimal polynomial of the matrix:

$$C = \begin{bmatrix} a & x & x & \dots & x \\ & a & x & \dots & x \\ & & \ddots & x & x \\ 0 & & & a & x \\ & & & & a \end{bmatrix} \rightarrow \begin{matrix} a_{ij} > 0 & (i < j) \\ a \neq 0 \end{matrix}$$

Ex 02: Give the proof of Cayley-Hamilton Theorem for a matrix $A \in M(\mathbb{R})$.

(Use the Lemma: $A \cdot \text{Com}^t(A) = \det(A) \cdot I_n$)

• Let A, B be two square matrices such that $A = P \cdot D_1 \cdot P^{-1}$ and $B = P \cdot D_2 \cdot P^{-1}$, where D_1, D_2 are diagonal. Show that:

$$\begin{matrix} A+B & A & B \\ e & = & e \cdot e \end{matrix}$$

(Good. Luck..)

Ex 01: Consider the matrix

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 5 & 0 & 0 \end{bmatrix}$$

Verify that $A^3 = 5I$, and deduce the expression of A^{-1} .

Ex 02: Let

$$A = \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix} \text{ and } B = \begin{bmatrix} 5 & 2 \\ 4 & 1 \end{bmatrix}$$

Show that $A \not\sim B$.

Ex 03: Let $A, B \in M_n(\mathbb{R})$. prove the following results:

$$\textcircled{1} A \sim B \Rightarrow A^k \sim B^k \text{ for } k \geq 1.$$

$$\textcircled{2} A \sim B \Rightarrow e^A \sim e^B$$

$$\textcircled{3} \det(A) = \prod h_i \quad (h_i \text{ is an eigenvalue})$$

$$\textcircled{4} \det(e^A) = e^{\text{tr}(A)} \quad (\text{tr} \equiv \text{the trace})$$

Ex 04: Solve the system of recurrence sequences:

$$\begin{cases} x_{n+1} = a_{11}x_n + a_{12}y_n + c_1 \\ y_{n+1} = a_{21}x_n + a_{22}y_n + c_2 \end{cases}$$

$$x_0, y_0 \in \mathbb{R}.$$

Ex 05: Let

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

① Find the minimal polynomial of A .

② Let $n \in \mathbb{N}$. prove that $A^n = \alpha_n A + \beta_n I$, and find α_n, β_n in terms of n .

Good Luck

Homework № 03

Ex 01: Consider the matrix:

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

prove that A is diagonalizable and find A^k .

Ex 02: Let

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

prove that

$$A^k = \begin{bmatrix} (-1)^k & 0 & 0 \\ \frac{2^k - (-1)^k}{3} & 2^k & 0 \\ \frac{(-1)^k - 2^k}{3} & 3^k - 2^k & 3^k \end{bmatrix}$$

Ex 03: Let $(x_n), (y_n)$ be two sequences given by:

$$\begin{cases} x_{n+1} = 3x_n - y_n; & x_0 = 1 \\ y_{n+1} = 2x_n; & y_0 = -1 \end{cases}$$

Find x_n et y_n in terms of n .

End
Bellaouar. Dj.

Homework N° 04 : 21 | Jan 2021.

Ex 01: Consider the matrix :

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

prove that $p_A(x) = x^3 + 1$, and deduce A^{-1} .

Ex 02: Let

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

Compute the minimal polynomial of A .

Ex 03: Let $t \in \mathbb{R}^*$ and let

$$A_t = \begin{bmatrix} 0 & -t \\ t & 0 \end{bmatrix}.$$

Show that: $e^{A_t} = \begin{bmatrix} \cos t & -\sin t \\ \sin t & \cos t \end{bmatrix}.$

Ex 04: Solve the system of linear differential equations :

$$\begin{cases} x' = -x + 2y \\ y' = 2x - y \end{cases}$$

Good-luck;

Ex 01: Trigonalize the following matrix:

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

and compute A^n for $n \geq 0$.

Ex 02: Study the diagonalization of the matrix:

$$A = \begin{bmatrix} 1 & a & b & c \\ 0 & 1 & d & e \\ 0 & 0 & -1 & f \\ 0 & 0 & 0 & 2 \end{bmatrix}, \begin{matrix} a, b, c \\ d, e, f \in \mathbb{R}. \end{matrix}$$

Ex 03: Consider the n by n matrix given by:

$$A_n = \begin{bmatrix} 2 & -1 & & & \\ -1 & 2 & & & \\ & & \ddots & & \\ & & & \ddots & \\ & & & & -1 \\ & & & -1 & 2 \end{bmatrix}$$

Compute $\det(A_n)$ in terms of n .

Ex 04: Let $A \in M_n(\mathbb{R})$. Prove the following properties:

- 1) $\det(A) = 0 \Rightarrow 0$ is an eigenvalue of A .
- 2) $A^2 = A \Rightarrow A$ is diagonalizable.
- 3) $(A \text{ is diagonalizable} + \text{nilpotent}) \Rightarrow A = 0$.

(Good. Luck.)