

Chapter 13: Matrix

March 31, 2023

Contents

1	General approach	1
1.1	Definition	1
1.1.1	Definition of a matrix	1
1.1.2	Notation	1
1.1.3	Examples	1
1.2	Particular matrices	1
1.2.1	Null matrix	1
1.2.2	Column matrix	2
1.2.3	Row matrix	2
1.2.4	Square matrix	2
1.2.5	Diagonal matrix	2
1.2.6	Identity matrix	2
1.2.7	Triangular matrix	3
1.3	Transposed matrix	3
1.3.1	Definition	3
1.3.2	Notation	3
1.3.3	Example	3
1.4	Symmetric matrix	4
1.4.1	Symmetric	4
1.4.2	Anti-Symmetric	4
2	Operations on matrices	4
2.1	Addition and external product	4
2.1.1	Definition	4
2.1.2	Elementary matrix	5
2.1.3	Proposition	5
2.2	Internal product	5
2.2.1	Definition	5
2.2.2	Remarks	6
2.3	Properties of matrix calculus	6
2.3.1	Properties	6
2.3.2	Case of Square Matrices	7
2.4	Inverse of a matrix	7
2.4.1	Definition	7
2.4.2	Notation	7
2.4.3	How to find the inverse of a matrix	7
3	Matrices of Liner Maps	8
3.1	Definition and examples	8
3.1.1	Definition	8

3.1.2	Example	9
3.1.3	Definition	9
3.1.4	Example	9

1 General approach

1.1 Definition

1.1.1 Definition of a matrix

We call matrix of n rows and p columns any mapping in the following form:

$$\begin{array}{ccc} \llbracket 1, n \rrbracket \times \llbracket 1, p \rrbracket & \rightarrow & \mathbb{K} \\ i, j & & a_{ij} \end{array}$$

We denote such maps as tables of n rows and p columns, and we write:

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1p} \\ a_{21} & a_{22} & \cdots & a_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{np} \end{pmatrix}$$

$\forall (i, j) \in \llbracket 1, n \rrbracket \times \llbracket 1, p \rrbracket$, we call a_{ij} a coefficient of the matrix. In this case coefficient if i -th row and j -th column.

1.1.2 Notation

We denote $M_{np}(\mathbb{K})$ the set of matrix of n rows and p columns with coefficient from \mathbb{K} .

1.1.3 Examples

$$A = \begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix} \in M_{32}(\mathbb{R})$$

$$B = \begin{pmatrix} i \\ 1+i \\ 3 \end{pmatrix} \in M_{31}(\mathbb{C})$$

1.2 Particular matrices

Let $A \in M_{np}(\mathbb{K})$ then:

1.2.1 Null matrix

1. $[\forall (i, j) \in \llbracket 1, n \rrbracket \times \llbracket 1, p \rrbracket, a_{ij} = 0] \Rightarrow [A = 0_{np}]$ We say A is the null matrix 0_{np} .

1.2.1.1 Example

$$A' = \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{pmatrix} \in M_{32}(\mathbb{R})$$

1.2.2 Column matrix

2. $B \in M_{np}(\mathbb{K})$ and $p = 1 \Rightarrow B$ is a column matrix of n rows

1.2.2.1 Example

$$B' = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \in M_{31}(\mathbb{R})$$

1.2.3 Row matrix

3. $B \in M_{np}(\mathbb{K})$ and $n = 1 \Rightarrow C$ is a row matrix of p columns

1.2.3.1 Example

$$C' = (1 \ 2 \ 3) \in M_{13}(\mathbb{R})$$

1.2.4 Square matrix

We call square matrix any matrix with same number of rows and columns. We denote $M_n(\mathbb{K})$ the set of square matrix of n rows and columns with coefficient from \mathbb{K} .

4. $D \in M_{np}(\mathbb{K})$ and $n = p \Rightarrow D$ is a square matrix denote $M_n(\mathbb{K})$

1.2.4.1 Example

$$D' = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \in M_3(\mathbb{R})$$

1.2.5 Diagonal matrix

5. $\forall E \in M_n(\mathbb{R})$, if $\forall (i, j) \in \llbracket 1, n \rrbracket^2, i \neq j \Rightarrow a_{ij} = 0$ then we say E is a diagonal matrix

1.2.5.1 Example

$$E' = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \in M_2(\mathbb{R})$$

$$E'' = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix} \in M_3(\mathbb{R})$$

1.2.6 Identity matrix

6. $\forall I_n \in M_n(\mathbb{R})$, if $\forall (i, j) \in \llbracket 1, n \rrbracket^2, i \neq j \Rightarrow a_{ij} = 0$ and $i = j \Rightarrow a_{ij} = 1$ then we say I_n is a identity matrix

1.2.6.1 Example

$$I'_n = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \in M_2(\mathbb{R})$$

$$I''_n = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \in M_3(\mathbb{R})$$

1.2.7 Triangular matrix

6. $\forall F \in M_n(\mathbb{R})$, if $\forall (i, j) \in \llbracket 1, n \rrbracket^2, i > j \Rightarrow a_{ij} = 0$ then we say F is a lower triangular matrix
7. $\forall G \in M_n(\mathbb{R})$, if $\forall (i, j) \in \llbracket 1, n \rrbracket^2, i < j \Rightarrow a_{ij} = 0$ then we say G is a upper triangular matrix

1.2.7.1 Example

$$F' = \begin{pmatrix} 1 & 0 \\ 2 & 3 \end{pmatrix} \in M_2(\mathbb{R})$$

$$G' = \begin{pmatrix} 1 & 2 \\ 0 & 3 \end{pmatrix} \in M_2(\mathbb{R})$$

1.3 Transposed matrix

1.3.1 Definition

Let $A \in M_{np}(\mathbb{K})$. We call transposed matrix of A (or A transpose) a matrix B from $M_{pn}(\mathbb{K})$ such as:

$$\forall (i, j) \in \llbracket 1, n \rrbracket \times \llbracket 1, p \rrbracket, a_{ij} = b_{ji}$$

1.3.2 Notation

We denote B as tA

1.3.3 Example

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \in M_{23}(\mathbb{R})$$

$${}^tA = \begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix} \in M_{32}(\mathbb{R})$$

1.4 Symmetric matrix

1.4.1 Symmetric

If ${}^tA = A$ then we say A is symmetric

1.4.1.1 Example

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 6 \end{pmatrix} = {}^tA = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 6 \end{pmatrix} \in M_3(\mathbb{R})$$

1.4.2 Anti-Symmetric

If ${}^tA = -A$ then we say A is Anti-symmetric

1.4.2.1 Example

$$A = \begin{pmatrix} 0 & -2 & 3 \\ 2 & 0 & -5 \\ -3 & 5 & 0 \end{pmatrix} = {}^tA = \begin{pmatrix} 0 & 2 & -3 \\ -2 & 0 & 5 \\ 3 & -5 & 0 \end{pmatrix} \in M_3(\mathbb{R})$$

2 Operations on matrices

2.1 Addition and external product

2.1.1 Definition

1. We call internal operation in $M_{np}(\mathbb{K})$ denoted \oplus "internal addition" the one defined as follows:

$$\forall A, B \in M_{np}^2(\mathbb{K}), A + B = (a_{ij} + b_{ij})_{\substack{1 \leq i \leq n \\ 1 \leq j \leq p}}$$

$$\text{Where } A = a_{ij} \substack{1 \leq i \leq n \\ 1 \leq j \leq p} \text{ and } B = b_{ij} \substack{1 \leq i \leq n \\ 1 \leq j \leq p}$$

2. We call "external multiplication" or "multiplication by a scalar" the one defined as follows:

$$\forall A \in M_{np}(\mathbb{K}), \forall \alpha \in \mathbb{K}, \alpha A = (\alpha a_{ij})_{\substack{1 \leq i \leq n \\ 1 \leq j \leq p}}$$

2.1.1.1 Example

$$(A, B) \in M_{2,3}(\mathbb{R})^2 \quad A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \text{ and } B = \begin{pmatrix} 7 & 8 & 9 \\ 10 & 11 & 12 \end{pmatrix}$$

$$A + B = \begin{pmatrix} 1+7 & 2+8 & 3+9 \\ 4+10 & 5+11 & 6+12 \end{pmatrix} = \begin{pmatrix} 8 & 10 & 12 \\ 14 & 16 & 18 \end{pmatrix}$$

$$\alpha = 3, \quad \alpha A = \begin{pmatrix} 3 \times 1 & 3 \times 2 & 3 \times 3 \\ 3 \times 4 & 3 \times 5 & 3 \times 6 \end{pmatrix} = \begin{pmatrix} 3 & 6 & 9 \\ 12 & 15 & 18 \end{pmatrix}$$

2.1.1.2 Proposition

(M_{np}, \oplus, \cdot) is a vector space over \mathbb{K}

2.1.2 Elementary matrix

For $(n, p) \in \mathbb{N}^2, (i, j) \in \llbracket 1, n \rrbracket \times \llbracket 1, p \rrbracket$; We denote E_{ij} the matrix from $M_{np}(\mathbb{K})$ such that the ij -th coefficient is 1 and all other coefficient are 0.

E_{ij} are called elementary matrix

2.1.2.1 Example

$$\begin{aligned} E_{11} &= \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \\ E_{22} &= \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \\ E_{33} &= \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \end{aligned}$$

2.1.3 Proposition

1. $(E_{ij})_{\substack{1 \leq i \leq n \\ 1 \leq j \leq p}}$ is a basis of $M_{np}(\mathbb{K})$
2. $\dim((E_{ij})_{\substack{1 \leq i \leq n \\ 1 \leq j \leq p}}) = np$

Ex: $M_2(\mathbb{R})$ a (\mathbb{K}) -VS: $B = \left(\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right)$

B is a Standard basis of $M_2(\mathbb{R})$, $\dim(M_2(\mathbb{R})) = 2^2 = 4$

2.2 Internal product

2.2.1 Definition

Let $(n, p, q) \in \mathbb{N}^3$ and $A = a_{ij} \substack{1 \leq i \leq n \\ 1 \leq j \leq p} \in M_{np}(\mathbb{K}), B = b_{ij} \substack{1 \leq i \leq p \\ 1 \leq j \leq q} \in M_{pq}(\mathbb{K})$. We call product of A and B the matrix C form $M_{nq}(\mathbb{K})$ such that:

$$\forall (i, j) \in \llbracket 1, n \rrbracket \times \llbracket 1, q \rrbracket, c_{ij} = \sum_{k=1}^p a_{ik} b_{kj}$$

2.2.1.1 Example

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \in M_{2,3}(\mathbb{R}) \quad \text{and} \quad B = \begin{pmatrix} 1 & 0 & 1 & 3 \\ 0 & 0 & 2 & 4 \\ 0 & 0 & 1 & 0 \end{pmatrix} \in M_{3,4}(\mathbb{R})$$

$$C = A \cdot B = \begin{pmatrix} 1 & 0 & 8 & 11 \\ 4 & 0 & 20 & 32 \end{pmatrix} \in M_{2,4}(\mathbb{R})$$

$$C_{2,3} = 4 \times 1 + 5 \times 2 + 6 \times 1 = 20$$

2.2.2 Remarks

- (R1) If A, B two matrices: we only can multiply A by B if the number of column of A is equal to the number of row of B .
- (R2) AB can exists but BA not or the other way around.

2.2.2.1 Example

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \cdot B = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \Rightarrow M_{2,1}(\mathbb{R}), \text{ and exists but } BA \text{ does not exists}$$

- (R3) In the General case, where AB and BA exists:
- $AB \neq BA$ (multiplication of matrix is not commutative)
When $AB = BA$ we say A and B commute.

2.3 Properties of matrix calculus

2.3.1 Properties

- Let A, B two matrices such that AB exists.
We can have $AB = 0$ and ($A \neq 0$ or $B \neq 0$)

If $A = 0$ or $B = 0$ then $AB = 0$

$$AB = 0 \nRightarrow A = 0 \text{ or } B = 0$$

2.3.1.1 Example

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

2. $(n, p, q, r) \in \mathbb{N}^4$ and let $(A, B, C) \in M_{np}(\mathbb{K}) \times M_{pq}(\mathbb{K}) \times M_{qr}(\mathbb{K})$

$$A \cdot (B \cdot C) = (A \cdot B) \cdot C \otimes \text{ is commutative}$$

3. $(A, B, C) \in M_{np}(\mathbb{K}) \times M_{pq}^2(\mathbb{K})$

$$A \cdot (B + C) = A \cdot B + A \cdot C$$

\otimes (Matrices multiplication) is distributive over matrix addition (\oplus).

4. $A \in M_{np}(\mathbb{K})$ and $B \in M_{pq}(\mathbb{K})$ and $\lambda \in \mathbb{K}$

$$\lambda \cdot (A \cdot B) = A \cdot \lambda \cdot B = A \cdot B \cdot \lambda$$

2.3.2 Case of Square Matrices

1.

$$\forall A \in M_n(\mathbb{K}), A \cdot I_n = I_n \cdot A = A$$

2. Let $(A, B) \in M_n(\mathbb{K})^2$, such that $AB = BA$ Then:

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

By convention, $A^0 = B^0 = I_n$

3.

$$\forall (A, B) \in M_n(\mathbb{K})^2, {}^t(A \cdot B) = {}^tA \cdot {}^tB$$

2.4 Inverse of a matrix

2.4.1 Definition

Let $A \in M_n(\mathbb{K})$ we say that A is invertible if:

$$\exists B \in M_n(\mathbb{K}), AB = BA = I_n$$

Then we say that B is the inverse of A and denote $B = A^{-1}$ (B is unique) Hence we have (in case of A invertible): $A \cdot A^{-1} = A^{-1} \cdot A = I_n$

2.4.2 Notation

The set of invertible matrices of $M_n(\mathbb{K})$ is denoted $GL_n(\mathbb{K})$

2.4.3 How to find the inverse of a matrix

We will use the following system: Where $A \in M_n(\mathbb{K}), (U, V) \in M_{n,1}(K)^2$:

$$A \cdot U = V$$

By solving this system (Gauss elimination algorithm) when A is invertible, we will have:

$$U = A^{-1} \cdot V$$

2.4.3.1 Example

$$\begin{aligned}
 A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix} A \cdot X = U &\Rightarrow \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix} \cdot \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} x \\ y \\ z \end{pmatrix} \\
 &\Leftrightarrow \begin{pmatrix} \textcircled{1} & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 1 & -1 \end{pmatrix} = \begin{pmatrix} x \\ y \\ z - x \end{pmatrix} \\
 &\Leftrightarrow \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & -2 \end{pmatrix} = \begin{pmatrix} x \\ y \\ z - x - y \end{pmatrix} \\
 &\Leftrightarrow \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & -2 \end{pmatrix} = \begin{pmatrix} x - y + z \\ -x + y + z \\ z - x - y \end{pmatrix} \\
 &\Leftrightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} \frac{x-y+z}{2} \\ \frac{-x+y+z}{2} \\ \frac{x+y-z}{2} \end{pmatrix} \\
 &\Leftrightarrow \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} x \\ y \\ z \end{pmatrix} \\
 A^{-1} &= \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \end{pmatrix}
 \end{aligned}$$

3 Matrices of Linear Maps

3.1 Definition and examples

3.1.1 Definition

Let $f \in \mathcal{L}(E, F)$, E and F finite dimensional \mathbb{K} -vector spaces such that: $\dim(E) = p$ and $\dim(F) = n$ where $(p, n) \in \mathbb{N}^2$ and $B = (e_1, e_2, \dots, e_n)$ basis of E and $B' = (e'_1, e'_2, \dots, e'_n)$ basis of F

$$\forall U \in E, \exists! (\lambda_1, \lambda_2, \dots, \lambda_n) \in \mathbb{K}^p, \quad U = \sum_{i=1}^p \lambda_i e'_i$$

We say $\begin{pmatrix} U_1 \\ U_2 \\ \vdots \\ U_n \end{pmatrix}$ is the column matrix of coordinates $A \cdot U$ in B .

3.1.2 Example

Let $E = \mathbb{R}^2$, $U = \begin{pmatrix} 1 \\ 2 \end{pmatrix}_B$ with $B = (\begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix})$ and $B' = (\begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \end{pmatrix})$ Then:

$$U = \begin{pmatrix} 2 \\ 1 \end{pmatrix}_B = \begin{pmatrix} 1 \\ 1 \end{pmatrix}_B + \begin{pmatrix} 1 \\ 0 \end{pmatrix}_B \Rightarrow U = \begin{pmatrix} 1 \\ 1 \end{pmatrix}_{B'}$$

3.1.3 Definition

We call matrix of $f \in \mathcal{L}(E, F)$ with respect to basis B and B' denoted $Mat_{BB'}(f)$ the matrix whose j -th column is composed of the coordinates of $f(e_j)$ in B' , for all j from $\llbracket 1, p \rrbracket$. This is a matrix of p columns and n rows:

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1p} \\ a_{21} & a_{22} & \cdots & a_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{np} \end{pmatrix}, \forall j \in \llbracket 1, p \rrbracket, f(e_j) = \begin{pmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{nj} \end{pmatrix}_{B'} = \sum_{i=1}^n a_{ij} e_i$$

3.1.4 Example

$$f: \begin{matrix} \mathbb{R}^2 & \longrightarrow & \mathbb{R}^3 \\ \begin{pmatrix} x \\ y \end{pmatrix} & \longmapsto & \begin{pmatrix} x+y \\ 2x+4y \\ -3y \end{pmatrix} \end{matrix}$$

① basis for the domain (\mathbb{R}^2): $B = (\begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix})$

basis for the codomain (\mathbb{R}^3): $B' = (\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix})$

$$\textcircled{2} \quad \begin{pmatrix} x+y \\ 2x+4y \\ -3y \end{pmatrix} \Rightarrow \begin{cases} f(\begin{pmatrix} 1 \\ 0 \end{pmatrix}) = \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}_{B'} \\ f(\begin{pmatrix} 0 \\ 1 \end{pmatrix}) = \begin{pmatrix} 1 \\ 4 \\ -3 \end{pmatrix}_{B'} \end{cases}$$

③

$$\begin{aligned} \forall U \in \mathbb{R}^2, U = x \cdot \begin{pmatrix} 1 \\ 0 \end{pmatrix} + y \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix} &\Rightarrow f(U) = x \cdot f\left(\begin{pmatrix} 1 \\ 0 \end{pmatrix}\right) + y \cdot f\left(\begin{pmatrix} 0 \\ 1 \end{pmatrix}\right) \\ &\Rightarrow \begin{pmatrix} 1 & 1 \\ 2 & 4 \\ 0 & -3 \end{pmatrix} \end{aligned}$$