

§5.2 Algebra of Matrices

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Example

Why Matrices?

Algebra of Matrices would be the main tool to study and solve System of 1st-order Linear ODE. So, we provide a background of the Algebra of Matrices.

Definition

A matrix \mathbf{A} of size $m \times n$ is defined as an array, with m rows and n columns:

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \cdots & a_{3n} \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{m1} & a_{m2} & a_{m3} & \cdots & a_{mn} \end{pmatrix} \quad \text{Also written as } \mathbf{A} = (a_{ij}).$$

a_{ij} s are called **entries** of \mathbf{A} .

Continued

- ▶ A matrix of size $m \times 1$ is called column **vector**. A matrix of size $1 \times n$ is called row **vector**.
- ▶ Often, in lower level courses only matrices of real numbers are considered. But most of it works in a more generality.
- ▶ In this course, we will **also consider** matrices $\mathbf{A} = (a_{ij})$ where the entries a_{ij} are **complex numbers**.
- ▶ **Further**, we will consider matrices $\mathbf{A} = (a_{ij})$ where the entries $a_{ij} = a_{ij}(t)$ are **functions** of t .

Transpose and conjugate

- ▶ Given a matrix $\mathbf{A} = (a_{ij})$ the **transpose** matrix $\mathbf{A}^T = (a_{ji})$ is obtained by writing the columns of \mathbf{A} and rows as columns.
- ▶ If $\mathbf{A} = (a_{ij})$ is a matrix of **complex** numbers, then the **conjugate** matrix $\overline{\mathbf{A}} = (\overline{a_{ij}})$ is obtained by replacing each entry a_{ij} by its conjugate $\overline{a_{ij}}$.
- ▶ The **transpose conjugate** matrix $\mathbf{A}^* = \overline{\mathbf{A}}^T$ is obtained by taking conjugate and then transpose.

Equality and Zero

- ▶ Two matrices $A = (a_{ij})$, $B = (b_{ij})$ are equal, if they have same size ($m \times n$) and

$$a_{ij} = b_{ij} \quad \text{for } 1 \leq i \leq m, 1 \leq j \leq n.$$

- ▶ The symbol $\mathbf{0}$ denotes the matrix whose entries are 0.

Addition

If $A = (a_{ij})$, $B = (b_{ij})$ are two matrices of equal size ($m \times n$), then their **sum** is defined to be the $m \times n$ matrix given by

$$A + B = (a_{ij} + b_{ij}).$$

So, the sum is obtained by adding the respective entries.

If the sizes of two matrices are different, then the sum is NOT defined.

Scalar Multiplication

If $A = (a_{ij})$ is a $m \times n$ matrix and c is a **complex** number, then the **scalar multiplication** of A by c is the $m \times n$ matrix given by

$$cA = (ca_{ij}).$$

Example of scalar multiplication

Let

$$A = \begin{pmatrix} 1 & 1 & -3 \\ 10 & 7 & -3 \end{pmatrix} \implies 11A = \begin{pmatrix} 11 & 11 & -33 \\ 110 & 77 & -33 \end{pmatrix}$$

Also,

$$(1-i)A = \begin{pmatrix} 1-i & 1-i & -3+3i \\ 10-10i & 7-7i & -3+3i \end{pmatrix}$$

Matrix Multiplication

Suppose $A = (a_{ij})$ is a matrix of size $m \times n$ and $B = (b_{ij})$ is a matrix of size $n \times p$. The the **product** AB is an $m \times p$ matrix

$$AB = (c_{ij}) \quad \text{where} \quad c_{ij} = \sum_{k=1}^n a_{ik} b_{kj} = a_{i1} b_{1j} + a_{i2} b_{2j} + \cdots + a_{in} b_{nj}.$$

Matrix Multiplication

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \cdots & \cdots & \cdots & \cdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} & \cdots & b_{1p} \\ b_{21} & b_{22} & \cdots & b_{2p} \\ \cdots & \cdots & \cdots & \cdots \\ b_{n1} & b_{n2} & \cdots & b_{np} \end{pmatrix}$$
$$= \begin{pmatrix} c_{11} & c_{12} & \cdots & c_{1p} \\ c_{21} & c_{22} & \cdots & c_{2p} \\ \cdots & \cdots & \cdots & \cdots \\ c_{m1} & c_{m2} & \cdots & c_{mp} \end{pmatrix} \quad c_{12} = a_{11}b_{12} + a_{12}b_{22} + \cdots + a_{1n}b_{n2}$$

Example of matrix multiplication

Let

$$A = \begin{pmatrix} 1 & 1 & -3 \\ 10 & 7 & -3 \end{pmatrix}, \quad B = \begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 2 & 1 \end{pmatrix},$$

Since number of columns of A and number of rows of B are same, the product AB is defined.

We have

$$\begin{aligned} AB &= \begin{pmatrix} 1 * 1 + 1 * 1 + (-3) * 2 & 1 * 1 + 1 * 0 + (-3) * 1 \\ 10 * 1 + 7 * 1 + (-3) * 2 & 10 * 1 + 7 * 0 + (-3) * 1 \end{pmatrix} \\ &= \begin{pmatrix} -4 & -2 \\ 11 & 7 \end{pmatrix} \end{aligned}$$

Remark. BA is ALSO defined, which will be a 3×3 matrix.

Algebra of Matrices

Let A, B, C be $m \times n$ matrices and c, d be scalars. Then,

$$A + B = B + A$$

Commutativity of addition

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

Associativity of addition

$$(cd)A = c(dA)$$

Associativity of scalar multiplication

$$c(A + B) = cA + cB$$

a Distributive property

$$(c + d)A = cA + dA$$

a Distributive property

Continued

Let A, B, C be matrices and c is a constant. Assume all the matrix products below are defined. Then

$$\begin{array}{ll} A(BC) = (AB)C & \text{Associativity Matrix Product} \\ A(B + C) = AB + AC & \text{Distributive Property} \\ (A + B)C = AC + BC & \text{Distributive Property} \\ c(AB) = (cA)B = A(cB) & \end{array}$$

Proofs would be routine checking, which we would skip.

- ▶ Matrix multiplication is not necessarily commutative.
That means, often $AB \neq BA$

Vector Multiplication

- ▶ By a **vector**, we mean a column vector.
- ▶ Consider two vectors of (complex) numbers:

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_n \end{pmatrix} \quad \text{and} \quad \mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ \dots \\ y_n \end{pmatrix}$$

- ▶ Define a product $\mathbf{x}^T \mathbf{y} := \sum_{i=1}^n x_i y_i$. This is extension of **dot product**.

Continued

- ▶ The **inner product** or **scalar product** between \mathbf{x}, \mathbf{y} is defined as

$$(\mathbf{x}, \mathbf{y}) := \sum_{i=1}^n x_i \bar{y}_i$$

- ▶ If \mathbf{x}, \mathbf{y} are vectors of real numbers, then $\mathbf{x}^T \mathbf{y} = (\mathbf{x}, \mathbf{y})$.
- ▶ For another vector \mathbf{z} and $\alpha \in \mathbb{C}$ we have

$$(\mathbf{x}, \mathbf{y}) = \overline{(\mathbf{y}, \mathbf{x})}, \quad (\mathbf{x}, \mathbf{y} + \mathbf{z}) = (\mathbf{x}, \mathbf{y}) + (\mathbf{x}, \mathbf{z})$$

$$(\alpha \mathbf{x}, \mathbf{y}) = \alpha (\mathbf{x}, \mathbf{y}), \quad (\mathbf{x}, \alpha \mathbf{y}) = \bar{\alpha} (\mathbf{x}, \mathbf{y})$$

Continued

- ▶ Then, $(\mathbf{x}, \mathbf{x}) := \sum_{i=1}^n x_i \bar{x}_i = \sum_{i=1}^n |x_i|^2$ is a nonnegative real number.
- ▶ The **length** or **magnitude** of \mathbf{x} is defined as $\|\mathbf{x}\| := \sqrt{(\mathbf{x}, \mathbf{x})}$.
- ▶ It follows $\|\mathbf{x}\| = 0 \iff \mathbf{x} = \mathbf{0}$.
- ▶ We say \mathbf{x}, \mathbf{y} are **orthogonal**, if $(\mathbf{x}, \mathbf{y}) = 0$.

The Identity Matrix

- ▶ For a positive integer, I_n would denote the square matrix of order n whose main diagonal (left to right) entries are 1 and rest of the entries are zero.
- ▶ So,

$$I_1 = (1), \quad I_2 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \quad I_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

- ▶ I_n is called the **identity** matrix of order n . Often, we write $I = I_n$.

Continued

- ▶ For matrices \mathbf{A} , \mathbf{B} , we have $\mathbf{A}I = \mathbf{A}$ and $I\mathbf{B} = \mathbf{B}$.

Inverse of a Matrix

- ▶ A square matrix \mathbf{A} is said to **invertible**, if there is a matrix \mathbf{B} such that $\mathbf{AB} = \mathbf{BA} = \mathbf{I}$.
- ▶ Such a matrix \mathbf{B} is unique, when there is one. In that case, denote $\mathbf{A}^{-1} := \mathbf{B}$. So,

$$\mathbf{A}^{-1}\mathbf{A} = \mathbf{AA}^{-1} = \mathbf{I}.$$

- ▶ An invertible matrix \mathbf{A} is also called **nonsingular**. If a matrix is not invertible, it is said to be a **singular** matrix.

Good News

- ▶ Both TI-84 and Matlab can compute inverses of matrices.

Analytic method

- ▶ The linear algebra course is not a prerequisite for this course.
- ▶ However, to give an analytic method to compute inverses of a matrix, we need to define determinant of a square matrix. This will be done in a separate note.

Minors, Cofactors and Inverses

- ▶ Let $A = (a_{ij})$ be an $n \times n$ -matrix.
- ▶ Let C_{ij} denote the cofactor of the the (i, j) -entry of A .
- ▶ Let $C = (C_{ij})$ be the cofactor matrix of A .
- ▶ Let $\text{Adj}(A) = C^t$ denote the transpose of C .
- ▶ If $\det(A) \neq 0$ then $A^{-1} = \frac{\text{Adj}(A)}{\det(A)}$

Matrix of Functions

We also consider matrices whose entries are functions of t and perform usual operation on them.

- ▶ We write

$$\mathbf{A}(t) = \begin{pmatrix} a_{11}(t) & a_{12}(t) & \cdots & a_{1n}(t) \\ a_{21}(t) & a_{22}(t) & \cdots & a_{2n}(t) \\ \cdots & \cdots & \cdots & \cdots \\ a_{m1}(t) & a_{m2}(t) & \cdots & a_{mn}(t) \end{pmatrix}$$

- ▶ Similarly, we write column/row matrices $\mathbf{X}(t)$.

Operations on matrices of functions

- ▶ Given a matrix of functions $\mathbf{A}(t) = (a_{ij}(t))$, define

$$\frac{d\mathbf{A}(t)}{dt} = \mathbf{A}'(t) = \left(\frac{da_{ij}(t)}{dt} \right)$$

- ▶ Similarly, define

$$\int \mathbf{A}(t)dt = \left(\int a_{ij}(t)dt \right), \quad \int_a^b \mathbf{A}(t)dt = \left(\int_a^b a_{ij}(t)dt \right)$$

Continued

By routine checking:

- ▶ $\frac{d(c\mathbf{A})}{dt} = \frac{cd(\mathbf{A})}{dt}$ for any matrix c of constants.
- ▶ $\frac{d(\mathbf{A}+\mathbf{B})}{dt} = \frac{d\mathbf{A}}{dt} + \frac{d\mathbf{B}}{dt}$
- ▶ $\frac{d(\mathbf{A}\mathbf{B})}{dt} = \frac{d\mathbf{A}}{dt}\mathbf{B} + \mathbf{A}\frac{d\mathbf{B}}{dt}$