

MATH 2101 LINEAR ALGEBRA I, FALL SEMESTER 2025  
CHAPTER 1: MATRIX ALGEBRA AND DETERMINANTS

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Goal of this section:

- (1) Basic terminology: matrices, entries, columns, rows, etc
- (2) Basic operations: additions, matrix multiplication, matrix addition, transpose
- (3) Properties of matrix multiplications
- (4) Determinants: definition, rules to compute
- (5) Inverse: computing inverse from a determinant, determining invertibility by determinants

## 1. MATRIX ALGEBRA

### 1.1. Basic terminology.

**Definition 1.1.** (1) We use  $\mathbb{R}$  to represent the set of real numbers. A **scalar** is usually referred to a number in  $\mathbb{R}$  (in this course).

- (2) An  $n \times m$  **matrix**  $A$  is a rectangular array of the form:

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \ddots & \ddots & & \ddots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{pmatrix}$$

where  $a_{ij}$  are in  $\mathbb{R}$ . We sometimes write  $(a_{ij})$  to represent an arbitrary matrix. The  $(i, j)$ -**entry** of  $A$  will be referred to the number  $a_{ij}$ .

- (3) For a matrix  $A$ , we also write  $A_{ij}$  to be the  $(i, j)$ -th entry of  $A$ .  
(4) The  $i'$ -th **row** of  $(a_{ij})$  is the  $1 \times m$  matrix

$$(a_{i'1} \ a_{i'2} \ \dots \ a_{i'm})$$

- (5) The  $j'$ -th **column** of  $(a_{ij})$  is the  $n \times 1$  matrix

$$\begin{pmatrix} a_{1j'} \\ a_{2j'} \\ \vdots \\ b_{nj'} \end{pmatrix}.$$

**Example 1.2.** Let  $A = \begin{pmatrix} 2 & 4 & 1 \\ 0 & 7 & 9 \end{pmatrix}$ . Then  $A_{13} = 1$ . The 2nd column of  $A$  is  $\begin{pmatrix} 4 \\ 7 \end{pmatrix}$  and the 1st row of  $A$  is  $(2 \ 4 \ 1)$ .

## 1.2. Operations of matrices.

(1) **Addition** of two  $n \times m$  matrices  $(a_{ij}) + (b_{ij})$  is defined as:

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{pmatrix} + \begin{pmatrix} b_{11} & b_{12} & \dots & b_{1m} \\ b_{21} & b_{22} & \dots & b_{2m} \\ \vdots & \vdots & & \vdots \\ b_{n1} & b_{n2} & \dots & b_{nm} \end{pmatrix} = \begin{pmatrix} a_{11} + b_{11} & a_{12} + b_{12} & \dots & a_{1m} + b_{1m} \\ a_{21} + b_{21} & a_{22} + b_{22} & \dots & a_{2m} + b_{2m} \\ \vdots & \vdots & & \vdots \\ a_{n1} + b_{n1} & a_{n2} + b_{n2} & \dots & a_{nm} + b_{nm} \end{pmatrix}$$

(2) **Scalar multiplication** on a matrix: for a scalar  $c \in \mathbb{R}$  and a  $n \times m$  matrix  $(a_{ij})$ ,  $c(a_{ij})$  is defined as:

$$c \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{pmatrix} = \begin{pmatrix} ca_{11} & ca_{12} & \dots & ca_{1m} \\ ca_{21} & ca_{22} & \dots & ca_{2m} \\ \vdots & \vdots & & \vdots \\ ca_{n1} & ca_{n2} & \dots & ca_{nm} \end{pmatrix}$$

(3) **Matrix multiplication:** Let  $A = (a_{ij})$  be a  $n \times m$  matrix and let  $B = (b_{ij})$  be a  $m \times p$  matrix. We define the multiplication  $AB$  such that the  $(i, j)$ -th entry of  $AB$  is:

$$(AB)_{ij} = \sum_{x=1}^m a_{ix} b_{xj}$$

**Remark 1.3.** Let  $A$  be a  $n \times m$  matrix and let  $B$  be a  $p \times q$  matrix. If  $m \neq p$ , then  $AB$  is *not* well-defined.

(Curious students may ask why to define matrix multiplication in such way. This is related to how to the concept of linear transformation, which we will come back later.)

**Example 1.4.** (1)  $\begin{pmatrix} 1 & 2 & 4 \\ -1 & 7 & 2 \end{pmatrix} + \begin{pmatrix} -1 & 3 & 2 \\ 1 & 3 & 8 \end{pmatrix} = \begin{pmatrix} 0 & 5 & 6 \\ 0 & 10 & 10 \end{pmatrix}$ .

(2)

$$(-2) \cdot \begin{pmatrix} 2 & 3 & 0 \\ 1 & 0.5 & 8 \end{pmatrix} = \begin{pmatrix} -4 & -6 & 0 \\ -2 & -1 & 0 \end{pmatrix}$$

(3)

$$(1 \ 2 \ 3) \begin{pmatrix} 4 & 3 \\ 0 & 1 \\ 9 & 2 \end{pmatrix} = (1(4) + 2(0) + 3(9) \ 1(3) + 2(1) + 3(2)) = (31 \ 11)$$

(4)

$$\begin{pmatrix} 1 & 2 & 3 \\ -1 & -5 & 0 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 0 & -16 \end{pmatrix}$$

is not well-defined.

### 1.3. Vectors.

**Definition 1.5.** (1) A **column vector** is a  $n \times 1$ -matrix i.e. of the form

$$\begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix}$$

(2) A **row vector** is a  $1 \times m$ -matrix i.e. of the form

$$(a_1, \dots, a_m).$$

(3) A **vector** simply refers to either a row vector or a column vector.

(4)  $\mathbb{R}^n$  refers to the set of all column vectors with  $n$  real entries.

(5) We sometimes simply write 0 for the row vector or column vector with all entries to be zero. We shall call such vector to be the **zero vector**.

**Exercise:** Visualize geometrically the additions and scalar multiplication for vectors in  $\mathbb{R}^2$ .

**1.4. Identity matrix.** The **identity matrix**  $I_n$  is defined as:

$$I_n = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & \vdots \\ \vdots & \dots & \dots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix},$$

that is

$$(I_n)_{ij} = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}$$

The main properties of the identity matrix are the following:

**Theorem 1.6.** Let  $A$  be an  $n \times m$  matrix. Then

- (1)  $I_n A = A$ ; and
- (2)  $A I_m = A$ .

*Proof.* Exercise. □

Indeed, one should do the following more general exercise:

**Exercise 1.7.** An  $n \times n$  matrix  $A$  is said to be **diagonal** if for any  $i, j$  with  $i \neq j$ ,  $A_{ij} = 0$ . For example,  $\begin{pmatrix} 1 & 0 \\ 0 & 7 \end{pmatrix}$  is diagonal but  $\begin{pmatrix} 1 & 3 \\ 0 & 2 \end{pmatrix}$  is not diagonal. The **diagonal entries** of a  $n \times n$  matrix  $(a_{ij})$  are referred to the entries  $a_{11}, a_{22}, \dots, a_{nn}$ .

Let  $A$  be a  $n \times m$  matrix.

- (1) For a diagonal  $n \times n$  matrix  $D$ , describe the matrix multiplication  $DA$ ;
- (2) For a diagonal  $m \times m$  matrix  $E$ , describe the matrix multiplication  $AE$ .

**1.5. Zero matrices.** The  $n \times m$  matrix is called the **zero matrix**  $0_{n \times m}$  (or sometimes simply 0) if all its entries are zero. The main properties of the zero matrix is the following:

**Theorem 1.8.** *Let  $A$  be an  $n \times m$  matrix. Then*

- (1)  $0_{p \times n}A = 0_{p \times m}$ ; and
- (2)  $A0_{m \times q} = 0_{n \times q}$ ; and
- (3)  $A + 0 = A$ .

*Proof.* Exercise. □

**Remark 1.9.** For simplicity, we usually write 0 for different things. One may need a bit thought to identify the meaning of 0 in a content. Can you explain the meaning of 0 in the following equations? Let  $A$  be a  $n \times p$ -matrix.

- (1) Then  $0_{m \times n}A = 0$ .
- (2) Let  $c = 0$  be the zero scalar. Then  $cA = 0$ .
- (3)  $A - A = 0$ .

## 1.6. Square matrices.

**Definition 1.10.** A  $n \times m$  matrix is called **square** if  $n = m$ .

**Example 1.11.** (1) Identity matrices are square. Diagonal matrices are also square.

- (2) The row vector  $(1 \ 3)$  is not square.
- (3) The matrix  $\begin{pmatrix} 1 & 2 \\ 2 & 0 \\ -1 & 1 \end{pmatrix}$  is not square.

## 1.7. Transpose of a matrix.

**Definition 1.12.** The **transpose**  $A^T$  of an  $n \times m$  matrix  $A$  is the  $m \times n$  matrix obtained from interchanging rows and columns i.e.

$$(A^T)_{\textcolor{red}{ij}} = A_{ji}$$

**Example 1.13.** (1) Let  $A = (3 \ 5 \ 0)$ . Then

$$A^T = \begin{pmatrix} 3 \\ 5 \\ 0 \end{pmatrix}.$$

$$(2) \text{ Let } B = \begin{pmatrix} 4 & 1 \\ 0 & 3 \end{pmatrix}. \text{ Then } B^T = \begin{pmatrix} 4 & 0 \\ 1 & 3 \end{pmatrix}.$$

There are some basic properties of the transpose:

**Theorem 1.14.** (1) (*Transpose commutes with addition*) For  $n \times m$ -matrices  $A$  and  $B$ ,

$$(A + B)^T = A^T + B^T.$$

(2) (*Transpose anti-commutes with matrix multiplication*) For an  $n \times m$ -matrix  $A$  and a  $m \times p$ -matrix  $B$ ,

$$(AB)^T = B^T A^T.$$

(3) (*Transpose commutes with scalar multiplication*) For an  $n \times m$ -matrix  $A$  and a scalar  $c$ ,

$$(cA)^T = cA^T.$$

**Example 1.15.** Let  $A = \begin{pmatrix} 1 & 2 \\ 3 & -1 \end{pmatrix}$  and let  $B = \begin{pmatrix} -3 \\ 2 \end{pmatrix}$ . Then

$$AB = \begin{pmatrix} 1 & 2 \\ 3 & -1 \end{pmatrix} \begin{pmatrix} -3 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \\ -11 \end{pmatrix},$$

and so

$$(AB)^T = (1 \quad -11),$$

$$B^T A^T = (-3 \quad 2) \begin{pmatrix} 1 & 3 \\ 2 & -1 \end{pmatrix} = (1 \quad -11).$$

**Theorem 1.16.** (*Transpose twice*) Let  $A$  be an  $n \times m$  matrix. Then  $(A^T)^T = A$ .

*Proof.*

$$((A^T)^T)_{ij} = (A^T)_{ji} = A_{ij}.$$

□

### 1.8. Symmetric matrices.

**Definition 1.17.** A matrix  $A$  is said to be *symmetric* if  $A^T = A$ .

**Example 1.18.** (1)  $\begin{pmatrix} 1 & 3 \\ 3 & -2 \end{pmatrix}$  is symmetric.

(2)  $\begin{pmatrix} 1 & 3 \\ -3 & -2 \end{pmatrix}$  is not symmetric.

(3) Any diagonal matrices are symmetric.

**Exercise 1.19.** Show that if a matrix is symmetric, then  $A$  is square.

**Exercise 1.20.** Write down the form of symmetric  $2 \times 2$  matrices and the form of symmetric  $3 \times 3$  matrices.

### 1.9. Upper triangular matrices.

**Definition 1.21.** (1) A square matrix  $(a_{ij})$  is said to be an **upper triangular matrix** if for any  $i > j$ ,  $a_{ij} = 0$ .

(2) A square matrix  $(a_{ij})$  is said to be a **lower triangular matrix** if for any  $i < j$ ,  $a_{ij} = 0$ .

**Example 1.22.** (1) The matrices  $\begin{pmatrix} 1 & 3 \\ 0 & 0 \end{pmatrix}$ ,  $\begin{pmatrix} 3 & -1 & 0 \\ 0 & 2 & -1 \\ 0 & 0 & -2 \end{pmatrix}$  are upper triangular.

(2) The matrices  $\begin{pmatrix} 1 & 0 & 0 \\ 2 & -1 & 0 \\ 1 & 0 & 2 \end{pmatrix}$  and  $\begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 2 & 0 & 0 \\ 1 & 2 & 0 & 0 \\ 1 & 2 & 0 & 0 \end{pmatrix}$  are lower triangular matrices.

(3) The matrix  $\begin{pmatrix} 1 & 2 \\ -1 & -2 \end{pmatrix}$  is neither upper triangular nor lower triangular.

**Exercise 1.23.** Show that a square matrix  $A$  is both upper triangular and lower triangular if and only if  $A$  is diagonal.