

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:

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

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

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

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

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

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)

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

(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.

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

Example 1.13. (1) Let $A = \begin{pmatrix} 3 & 5 & 0 \end{pmatrix}$. Then

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

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

$$(AB)^T = \begin{pmatrix} 1 & -11 \end{pmatrix},$$

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

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 2×2 matrices and the form of symmetric 3×3 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.