

# Lecture Notes: Matrix Definitions and Operators

Yufei Tao

Department of Computer Science and Engineering

Chinese University of Hong Kong

taoyf@cse.cuhk.edu.hk

## 1 Matrix Definitions

An  $m \times n$  *matrix* is defined as  $m$  rows of real values, where each row has length  $n$ . To represent a matrix, we typically write out all these numbers in a 2d array, enclosed by a pair of square brackets, e.g.:

$$\begin{bmatrix} 1 & 2 & 3 \\ 3 & 4 & 5 \\ 6 & 7 & 8 \\ 8 & 4 & 2 \end{bmatrix} \quad (1)$$

is a  $4 \times 3$  matrix. We will use capitalized bold symbols to denote arrays, e.g.,  $\mathbf{A}$ . The values  $m$  and  $n$  are called the *dimensions* of  $\mathbf{A}$ .

To refer to the number at the  $i$ -th row and  $j$ -th column of  $\mathbf{A}$ , we use  $ij$  as the subscript. For example, if  $a_{ij}$  is an element of the array in (1), then  $a_{12} = 2$  whereas  $a_{21} = 3$ . We use the notation  $\mathbf{A} = [a_{ij}]$  to indicate that  $\mathbf{A}$  is obtained by listing out all  $a_{ij}$ .

Let  $\mathbf{A}$  be an  $m \times n$  matrix. We sometimes will regard each row of  $\mathbf{A}$  as a  $1 \times m$  *row vector*, and likewise, each column of  $\mathbf{A}$  as an  $n \times 1$  *column vector*. For example, let  $\mathbf{A}$  be the matrix in (1). Then, the 3rd row of  $\mathbf{A}$  is a vector  $[6, 7, 8]$ , while the 2nd column is a vector:

$$\begin{bmatrix} 2 \\ 4 \\ 7 \\ 4 \end{bmatrix}$$

If  $m = n$ , then we say that  $\mathbf{A}$  is a *square matrix*, an example of which is:

$$\begin{bmatrix} 1 & 2 & 3 & 8 \\ 3 & 4 & 5 & 2 \\ 6 & 7 & 8 & 3 \\ 8 & 4 & 2 & 4 \end{bmatrix} \quad (2)$$

When  $\mathbf{A} = [a_{ij}]$  is an  $n \times n$  square matrix, we refer to the sequence  $\{a_{11}, a_{22}, \dots, a_{nn}\}$  as the *main diagonal* (or just *diagonal* for short). For example, if  $\mathbf{A}$  is the matrix in (2), then its main diagonal is  $\{1, 4, 8, 4\}$ .

Again, let  $\mathbf{A} = [a_{ij}]$  be a square matrix. Then, we say that

- $\mathbf{A}$  is *symmetric* if it always holds that  $a_{ij} = a_{ji}$ ;
- $\mathbf{A}$  is *skew-symmetric* if it always holds that  $a_{ij} = -a_{ji}$ .

It is easy to see that  $\mathbf{A}$  is skew-symmetric, then its main diagonal consists of only 0's. For example,

$$\begin{bmatrix} 1 & 2 & 3 & 8 \\ 2 & 4 & 5 & 2 \\ 3 & 5 & 8 & 3 \\ 8 & 2 & 3 & 4 \end{bmatrix}$$

is symmetric, while

$$\begin{bmatrix} 0 & 2 & -3 & -8 \\ -2 & 0 & 5 & 2 \\ 3 & -5 & 0 & 3 \\ 8 & -2 & -3 & 0 \end{bmatrix}$$

is skew-symmetric.

Still let  $\mathbf{A}$  be a square matrix. We say that  $\mathbf{A}$  is a *diagonal matrix* if it has non-zero values only at its main diagonal. If in addition all those non-zero values are 1, then we say that  $\mathbf{A}$  is an *identity matrix*, e.g.:

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

Finally, if all the values in an  $m \times n$  matrix  $\mathbf{A}$  are 0, then we say that  $\mathbf{A}$  as a *zero matrix*. We may denote the matrix as  $\mathbf{0}$  if its dimensions are clear from the context.

## 2 Matrix Operators

**Definition 1.** Let  $\mathbf{A} = [a_{ij}]$  and  $\mathbf{B} = [b_{ij}]$  be  $m \times n$  matrices. Then, we say that  $\mathbf{A}$  equals  $\mathbf{B}$  if  $a_{ij} = b_{ij}$  for all  $i \in [1, n]$  and  $j \in [1, m]$ .

If  $\mathbf{A}$  and  $\mathbf{B}$  are equal, then we write  $\mathbf{A} = \mathbf{B}$ ; otherwise, we write  $\mathbf{A} \neq \mathbf{B}$ .

**Definition 2.** Let  $\mathbf{A} = [a_{ij}]$  and  $\mathbf{B} = [b_{ij}]$  be  $m \times n$  matrices. We define:

- **(matrix addition)** the result of  $\mathbf{A} + \mathbf{B}$  to be the  $m \times n$  matrix  $\mathbf{C} = [c_{ij}]$  where  $c_{ij} = a_{ij} + b_{ij}$  for all  $i \in [1, n]$  and  $j \in [1, m]$ ;
- **(matrix subtraction)** the result of  $\mathbf{A} - \mathbf{B}$  to be the  $m \times n$  matrix  $\mathbf{C} = [c_{ij}]$  where  $c_{ij} = a_{ij} - b_{ij}$  for all  $i \in [1, n]$  and  $j \in [1, m]$ .

For example:

$$\begin{bmatrix} 1 & 2 & 3 \\ 3 & 4 & 5 \\ 6 & 7 & 8 \\ 8 & 4 & 2 \end{bmatrix} + \begin{bmatrix} 0 & 1 & 0 \\ 2 & -3 & 1 \\ 0 & -7 & 0 \\ 0 & -2 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 3 & 3 \\ 5 & 1 & 6 \\ 6 & 0 & 8 \\ 8 & 2 & 4 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 & 3 \\ 3 & 4 & 5 \\ 6 & 7 & 8 \\ 8 & 4 & 2 \end{bmatrix} - \begin{bmatrix} 0 & 1 & 0 \\ 2 & -3 & 1 \\ 0 & -7 & 0 \\ 0 & -2 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 3 \\ 1 & 7 & 4 \\ 6 & 14 & 8 \\ 8 & 6 & 0 \end{bmatrix}$$

**Definition 3. (Matrix Scalar Multiplication)** Let  $\mathbf{A} = [a_{ij}]$  be an  $m \times n$  matrices, and  $c$  be a real value. Then, we define  $c\mathbf{A}$  to be the  $m \times n$  matrix  $\mathbf{B} = [b_{ij}]$  where  $b_{ij} = c \cdot a_{ij}$  for all  $i \in [1, n]$  and  $j \in [1, m]$ .

For example:

$$2 \begin{bmatrix} 1 & 2 & 3 \\ 3 & 4 & 5 \\ 6 & 7 & 8 \\ 8 & 4 & 2 \end{bmatrix} = \begin{bmatrix} 2 & 4 & 6 \\ 6 & 8 & 10 \\ 12 & 14 & 16 \\ 16 & 8 & 4 \end{bmatrix}$$

**Definition 4. (Matrix Multiplication)** Let  $\mathbf{A} = [a_{ij}]$  be an  $m \times n$  matrix, and  $\mathbf{B} = [b_{ij}]$  be an  $n \times p$  matrix. We define  $\mathbf{AB}$  as the  $m \times p$  matrix  $\mathbf{C} = [c_{ij}]$  where

$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}.$$

for all  $i \in [1, m]$  and  $j \in [1, p]$ .

Note that matrix multiplication requires that the number of *columns* of the first matrix must equal the number of *rows* of the second matrix. For example:

$$\begin{bmatrix} 1 & 2 & 3 & 0 \\ 3 & 4 & 2 & 1 \\ 0 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 2 & 2 \\ -1 & 1 & 1 \\ -1 & 0 & 1 \\ 1 & -2 & 0 \end{bmatrix} = \begin{bmatrix} -5 & 4 & 7 \\ -5 & 8 & 12 \\ -1 & -1 & 2 \end{bmatrix}$$

More specifically, the value 4 in the first row of the result matrix equals the dot product of the first row of the first matrix and the second column of the second matrix on the left.

It is rudimentary to verify:

$$\begin{aligned} \mathbf{ABC} &= \mathbf{A(BC)} \\ (\mathbf{A+B})\mathbf{C} &= \mathbf{AC+BC} \\ \mathbf{C(A+B)} &= \mathbf{CA+CB} \end{aligned}$$

**Definition 5. (Matrix Transposition)** Let  $\mathbf{A} = [a_{ij}]$  be an  $m \times n$  matrix. Then, the **transpose** of  $\mathbf{A}$ , denoted as  $\mathbf{A}^T$ , is the  $n \times m$  matrix  $\mathbf{B} = [b_{ij}]$  where  $a_{ij} = b_{ji}$  for all  $i \in [1, n]$  and  $j \in [1, m]$ .

For example:

$$\begin{bmatrix} 1 & 2 & 3 \\ 3 & 4 & 5 \\ 6 & 7 & 8 \\ 8 & 4 & 2 \end{bmatrix}^T = \begin{bmatrix} 1 & 3 & 6 & 8 \\ 2 & 4 & 7 & 4 \\ 3 & 5 & 8 & 2 \end{bmatrix}$$

It is rudimentary to verify:

$$\begin{aligned} (\mathbf{A+B})^T &= \mathbf{A}^T + \mathbf{B}^T \\ (c\mathbf{A})^T &= c\mathbf{A}^T \\ (\mathbf{AB})^T &= \mathbf{B}^T \mathbf{A}^T \end{aligned}$$