

# Matrices

P. Danziger

## 1 Matrices

### 1.1 Definitions

#### Definition 1

1. A Matrix is an  $m \times n$  ( $m$  by  $n$ ) array of numbers.

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

2. The entries in a matrix are called the *components* of the matrix and can be written as  $a_{ij}$ , where  $i$  indicates the row number and runs from 1 to  $m$ , and  $j$  indicates the column number and runs from 1 to  $n$ .
3. A Vector is a  $1 \times n$  or  $n \times 1$  matrix. That is an ordered set of  $n$  numbers.  
We say that such a vector is of dimension  $n$ .
4. A scalar is a number (usually either real or complex).

#### Notation 2

- We generally use uppercase letters from the beginning of the alphabet ( $A, B, C \dots$ ) to denote matrices.
- A matrix is identified with its components, given a matrix  $A$  with components  $a_{ij}$ ,  $1 \leq i \leq m$ ,  $1 \leq j \leq n$ , we may write

$$A = [a_{ij}]$$

#### Example 3

Find the  $4 \times 4$  matrix  $A$  with components given by  $a_{ij} = i + j$ .

$$\begin{pmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \\ 4 & 5 & 6 & 7 \end{pmatrix}$$

- We generally use lowercase boldface letters from the end of the alphabet ( $\mathbf{u}, \mathbf{v}, \mathbf{w} \dots$ ) to denote vectors.
- We use the convention that  $\mathbf{u} = (u_1, u_2, \dots, u_n)$ ,  $\mathbf{x} = (x_1, x_2, \dots, x_n)$ , etc.
- If  $\mathbf{x} = (x_1, x_2, \dots, x_n)$  then the scalars  $x_1, x_2, \dots, x_n$  are called the components of  $\mathbf{x}$ .
- We denote the set of all vectors of dimension  $n$  whose components are real numbers by  $\mathbb{R}^n$ .
- We denote the set of all vectors of dimension  $n$  whose components are complex numbers by  $\mathbb{C}^n$ .

**Note** This definition of vector differs from the usual ‘High School’ definition involving magnitude and direction.

## 1.2 Special Matrices and Vectors

### 1. The Identity matrix

The identity matrix is a square matrix with 1’s down the diagonal, and zeros elsewhere. The  $n \times n$  identity matrix is denoted  $I_n$ .

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

### 2. The Zero Matrix

The zero matrix is an  $m \times n$  matrix, all of whose entries are 0.

$$\begin{pmatrix} 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 \end{pmatrix}$$

## 2 Operations on Matrices

### 1. Transpose

Given an  $m \times n$  matrix,  $A$ , the transpose of  $A$  is obtained by interchanging the rows and columns of  $A$ . We denote the transpose of  $A$  by  $A^t$ , or  $A^T$ .

**Notes:**

- If  $A$  is  $m \times n$  then  $A^T$  will be  $n \times m$ .
- If  $A = [a_{ij}]$ , then  $[a_{ij}]^T = [a_{ji}]$ .

**Example 4**

(a)

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}^t = \begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix}$$

(b)

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}^t = \begin{pmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{pmatrix}$$

(c)

$$(1, 2, 3)^t = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$

**2. Matrix Addition**Given two  $m \times n$  matrices

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}, \quad B = \begin{pmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{m1} & b_{m2} & \dots & b_{mn} \end{pmatrix}$$

We may define the sum of  $A$  and  $B$ ,  $A + B$ , to be the sum componentwise, i.e.

$$A + B = \begin{pmatrix} a_{11} + b_{11} & a_{12} + b_{12} & \dots & a_{1n} + b_{1n} \\ a_{21} + b_{21} & a_{22} + b_{22} & \dots & a_{2n} + b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} + b_{m1} & a_{m2} + b_{m2} & \dots & a_{mn} + b_{mn} \end{pmatrix}$$

Componentwise:  $[a_{ij}] + [b_{ij}] = [a_{ij} + b_{ij}]$ .

This works for vectors as well.

$$\begin{aligned} \mathbf{u} + \mathbf{v} &= (u_1, u_2, \dots, u_n) + (v_1, v_2, \dots, v_n) \\ &= (u_1 + v_1, u_2 + v_2, \dots, u_n + v_n) \end{aligned}$$

**Note** that matrix addition is only defined if  $A$  and  $B$  have the same size.

**Example 5**

(a)

$$\begin{aligned}
& \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} + \begin{pmatrix} 10 & 11 & 12 \\ 13 & 14 & 15 \\ 16 & 17 & 18 \end{pmatrix} \\
&= \begin{pmatrix} 1+10 & 2+11 & 3+12 \\ 4+13 & 5+14 & 6+15 \\ 7+16 & 8+17 & 9+18 \end{pmatrix} \\
&= \begin{pmatrix} 11 & 13 & 15 \\ 17 & 19 & 21 \\ 23 & 25 & 27 \end{pmatrix}
\end{aligned}$$

(b)

$$\begin{aligned}
(1, 2, 3) + (4, 5, 6) &= (1+4, 2+5, 3+6) \\
&= (5, 7, 9)
\end{aligned}$$

**3. Matrix Multiplication****(a) Scalar Multiplication**

Given a matrix  $A$ , and a scalar  $k$ , we define the scalar product of  $k$  with  $A$ ,  $kA$  by multiplying each entry of  $A$  by  $k$ .

$$\begin{aligned}
kA &= k \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \\
&= \begin{pmatrix} ka_{11} & ka_{12} & \dots & ka_{1n} \\ ka_{21} & ka_{22} & \dots & ka_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ ka_{m1} & ka_{m2} & \dots & ka_{mn} \end{pmatrix}
\end{aligned}$$

Componentwise:  $k[a_{ij}] = [ka_{ij}]$ .

**Note** that this works for vectors as well.

$$k\mathbf{u} = k(u_1, u_2, \dots, u_n) = (ku_1, ku_2, \dots, ku_n)$$

**Example 6**

$$10 \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} = \begin{pmatrix} 10 & 20 & 30 \\ 40 & 50 & 60 \\ 70 & 80 & 90 \end{pmatrix}$$

(b) **Matrix Multiplication**

If  $A$  and  $B$  are two matrices where  $A$  has the same number of columns as  $B$  has rows (i.e.  $A$  is  $m \times n$  and  $B$  is  $n \times r$ ) we define the matrix product,  $AB$  to be the matrix in which the  $i, j^{\text{th}}$  entry is made up of the dot product of the  $i^{\text{th}}$  row of  $A$  with the  $j^{\text{th}}$  column of  $B$ .

$$\begin{aligned}
 A &= \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}, \\
 B &= \begin{pmatrix} b_{11} & b_{12} & \dots & b_{1r} \\ b_{21} & b_{22} & \dots & b_{2r} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \dots & b_{nr} \end{pmatrix} \\
 AB &= \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} & \dots & b_{1r} \\ b_{21} & b_{22} & \dots & b_{2r} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \dots & b_{nr} \end{pmatrix} \\
 &= \begin{pmatrix} a_{11}b_{11} + a_{12}b_{21} + \dots + a_{1n}b_{n1} & \dots & a_{11}b_{1r} + a_{12}b_{2r} + \dots + a_{1n}b_{nr} \\ a_{21}b_{11} + a_{22}b_{21} + \dots + a_{2n}b_{n1} & \dots & a_{21}b_{1r} + a_{22}b_{2r} + \dots + a_{2n}b_{nr} \\ \vdots & \ddots & \vdots \\ a_{m1}b_{11} + a_{m2}b_{21} + \dots + a_{mn}b_{n1} & \dots & a_{m1}b_{1r} + a_{m2}b_{2r} + \dots + a_{mn}b_{nr} \end{pmatrix}
 \end{aligned}$$

**Example 7**

$$\begin{aligned}
 &\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} 9 & 8 & 7 \\ 6 & 5 & 4 \\ 3 & 2 & 1 \end{pmatrix} = \\
 &\begin{pmatrix} 1 \times 9 + 2 \times 6 + 3 \times 3 & 1 \times 8 + 2 \times 5 + 3 \times 2 & 1 \times 7 + 2 \times 4 + 3 \times 1 \\ 4 \times 9 + 5 \times 6 + 6 \times 3 & 4 \times 8 + 5 \times 5 + 6 \times 2 & 4 \times 7 + 5 \times 4 + 6 \times 1 \\ 7 \times 9 + 8 \times 6 + 9 \times 3 & 7 \times 8 + 8 \times 5 + 9 \times 2 & 7 \times 7 + 8 \times 4 + 9 \times 1 \end{pmatrix} \\
 &= \begin{pmatrix} 9 + 12 + 9 & 8 + 10 + 6 & 7 + 8 + 3 \\ 36 + 30 + 18 & 32 + 25 + 12 & 28 + 20 + 6 \\ 63 + 48 + 27 & 56 + 40 + 18 & 49 + 32 + 9 \end{pmatrix} \\
 &= \begin{pmatrix} 30 & 24 & 18 \\ 84 & 69 & 54 \\ 138 & 114 & 90 \end{pmatrix}
 \end{aligned}$$

**Note** that Matrix multiplication is only defined if  $A$  has the same number of columns as  $B$  has rows.

**BIG Note**

Matrix multiplication is **NOT** commutative. i.e. It is **NOT** true that  $AB = BA$  (where defined).

**Example 8**

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

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