

1.3 Matrix Algebra

A **matrix** is a rectangular array of numbers arranged in rows and columns.

An $m \times n$ (this is often called the **size** or **dimension** of the matrix) matrix is a matrix with m rows and n columns and the element in the i^{th} row and j^{th} column is denoted by a_{ij} . A general $m \times n$ matrix is the following:

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & & & \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}_{m \times n} = (a_{ij})_{m \times n}.$$

The size or dimension of a matrix is subscripted as shown if required. If it's not required or clear from the problem the subscripted size is often dropped from the matrix.

Special matrices:

Square matrix has same number of rows and columns.

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & & & \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}_{n \times n}$$

Diagonal matrix is a square matrix with only zero entries off the main diagonal (that is $a_{ij} = 0$ if $i \neq j$):

$$\begin{bmatrix} a_{11} & 0 & \cdots & 0 \\ 0 & a_{22} & \cdots & 0 \\ \vdots & & & \\ 0 & 0 & \cdots & a_{nn} \end{bmatrix}$$

Identity matrix: $I = I_n = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & & & \\ 0 & 0 & \cdots & 1 \end{bmatrix}_{n \times n}$

Zero matrix: $0_{m \times n} = \begin{bmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & & & \\ 0 & 0 & \cdots & 0 \end{bmatrix}_{m \times n}$

Column matrix: $x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}_{n \times 1}$

Row matrix: $y = (y_1, y_2, \dots, y_n)_{1 \times n}$

we often refer to row matrix and column matrix as vectors.

Matrix addition, subtraction and Scalar multiplication:

$$A \pm B = (a_{ij})_{m \times n} \pm (b_{ij})_{m \times n} = (a_{ij} \pm b_{ij})_{m \times n}$$

$$cA = c(a_{ij})_{m \times n} = (ca_{ij})_{m \times n} \quad (c \text{ is a number})$$

Example 1: Given the following two matrices:

$$A = \begin{pmatrix} 2 & 5 \\ -1 & 6 \end{pmatrix}, \quad B = \begin{pmatrix} 0 & 2 \\ -1 & 0 \end{pmatrix}$$

Compute $2A - 3B$.

Matrix Multiplication:

Definition 1: Product of a $1 \times n$ row matrix and an $n \times 1$ column matrix is a 1×1 matrix given by

$$(y_1, y_2, \dots, y_n) \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = (x_1 y_1 + x_2 y_2 + \cdots + x_n y_n)$$

Example 2: Find the product $(2 \ 0 \ -1) \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$

$$(2 \ 0 \ -1) \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = 2 \times 1 + 0 \times 2 + (-1) \times 3 = -1$$

Definition 2: If A is an $m \times p$ matrix and B is a $p \times n$ matrix, then the **matrix product** of A and B , denoted by AB , is an $m \times n$ matrix whose element in the i th row and j th column is the number obtained from the product of the i th row of A and j th column of B . If the number of columns in A does not equal the number of rows in B , then the matrix product AB is not defined.

Example 3: Find the matrix product $\begin{pmatrix} 2 & 3 & -1 \\ -2 & 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & -3 \\ -2 & 0 \\ -1 & 2 \end{pmatrix}$

$$\begin{pmatrix} 2 & 3 & -1 \\ -2 & 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & -3 \\ -2 & 0 \\ -1 & 2 \end{pmatrix} = \begin{pmatrix} 2 \times 1 + 3 \times (-2) + (-1) \times (-1) & 2 \times (-3) + 3 \times 0 + (-1) \times 2 \\ -2 \times 1 + 1 \times (-2) + 2 \times (-1) & -2 \times (-3) + 1 \times 0 + 2 \times 2 \end{pmatrix} = \begin{pmatrix} -3 & -8 \\ -6 & 10 \end{pmatrix}$$

Example 4: Given $A = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$, $B = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$, compute AB and BA . Are they same?

$$AB = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}, BA = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

Matrix Inverse

Definition 3: For a given square matrix A if we can find another matrix of the same size B , such that $AB = BA = I$, then we call A is **invertible** or **nonsingular**, and B the inverse of A and denote it by $B = A^{-1}$, so $AA^{-1} = A^{-1}A = I$. A matrix that has no inverse is said to be **singular**.

Example 5: Given $A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$. Prove that $A^{-1} = A$.

Since $AA = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$, $A^{-1} = A$

Definition 4: The **transpose** of $m \times n$ matrix A is the $n \times m$ matrix B defined by $a_{ij} = b_{ji}$. The transpose of A is denoted by A^T

$$A = (a_{ij})_{m \times n} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}_{m \times n} \quad B = (b_{ij})_{n \times m} = \begin{bmatrix} a_{11} & a_{21} & \cdots & a_{m1} \\ a_{12} & a_{22} & \cdots & a_{m2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \cdots & a_{mn} \end{bmatrix}_{n \times m} = A^T$$

Example 6: $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$, then $A^T = \begin{pmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{pmatrix}$

Property of transpose:

- i. $(A^T)^T = A$
- ii. $(\alpha A)^T = \alpha A^T$
- iii. $(A+B)^T = A^T + B^T$
- iv. $(AB)^T = B^T A^T$

Definition: An $n \times n$ matrix A is said to be **symmetric** if $A = A^T$

Example 7: $A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 5 & 6 \\ 3 & 8 & 9 \end{pmatrix}$ is not symmetric, $A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 5 & 8 \\ 3 & 8 & 9 \end{pmatrix}$ is symmetric.

Definition: If $a_1 = \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix}$, $a_2 = \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix}$, \dots , $a_n = \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix}$ are vectors and c_1, c_2, \dots, c_n are

numbers, then the following

$$a_1 c_1 + a_2 c_2 + \cdots + a_n c_n$$

is called a **linear combination** of the vectors a_1, a_2, \dots, a_n

Example: Linear system

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m \end{cases}$$

can write as the following matrix form:

Let

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}, x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

We have $Ax = b$.

Also, we can write as the following matrix form:

$$\begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} x_1 + \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} x_2 + \cdots + \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix} x_n = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

or $a_1x_1 + a_2x_2 + \cdots + a_nx_n = b$

Thus, we have the following property:

A linear system $Ax = b$ is consistent (means it has at least one solution) if and only if b can be written as a linear combination of the column vectors of A .

Example: If A be a nonsingular matrix. Show that A^T is also nonsingular and

$$(A^T)^{-1} = (A^{-1})^T$$

Prove: $A^T(A^{-1})^T = (A^{-1}A)^T = I^T = I$

$$(A^{-1})^T A^T = (AA^{-1})^T = I^T = I$$

so, $(A^T)^{-1} = (A^{-1})^T$

Homework: 1. (d),(g),(h), 2. (e),(f), 5. (c), 7. (b), 10, 13, 15

You may use MATLAB to check your matrix computation.

Example 1:

```
>> A=[2 5;-1 6]
```

A =

```
 2  5  
-1  6
```

```
>> B=[0 2;-1 0]
```

B =

```
 0  2  
-1  0
```

```
>> 2*A-3*B
```

ans =

```
 4  4  
 1 12
```

Example 3:

```
>> A=[2 3 -1;-2 1 2]
```

A =

```
 2  3 -1  
-2  1  2
```

```
>> B=[1 -3;-2 0;-1 2]
```

B =

```
 1 -3  
-2  0  
-1  2
```

```
>> A*B
```

ans =

```
-3 -8
-6 10
```

Example 6:

```
>> A=[0 1 2;1 1 2;2 1 1]
```

A =

```
0 1 2
1 1 2
2 1 1
```

```
>> inv(A)
```

ans =

```
-1 1 0
3 -4 2
-1 2 -1
```

Example 7:

```
>> A=[1 -2;2 -4]
```

A =

```
1 -2
2 -4
```

```
>> inv(A)
```

Warning: Matrix is singular to working precision.

ans =

```
Inf Inf
Inf Inf
```