

## Chapters 7-8: Linear Algebra

Sections 7.1, 7.2 & 7.4

### 1. Matrices and vectors

- An  $m \times n$  **matrix** is an array with  $m$  rows and  $n$  columns. It is typically written in the form

$$A = [a_{ij}] = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix},$$

where  $i$  is the **row index** and  $j$  is the **column index**.

- A **column vector** is an  $m \times 1$  matrix. Similarly, a **row vector** is a  $1 \times n$  matrix.
- The entries  $a_{ij}$  of a matrix  $A$  may be **real or complex**.

## Matrices and vectors (continued)

- **Examples:**

- $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$  is a  $2 \times 2$  **square** matrix with **real entries**.

- $u = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$  is a **column vector** of  $A$ .

- $B = \begin{bmatrix} 1 & 0 & 0 \\ 0 & i & 0 \\ 0 & 0 & 3 - 7i \end{bmatrix}$  is a  $3 \times 3$  **diagonal** matrix, with **complex entries**.

- An  $n \times n$  diagonal matrix whose entries are all ones is called the  $n \times n$  **identity matrix**.

- $C = \begin{bmatrix} 1 & 2 & 3 & 10 \\ 1 & 6 & -8 & 0 \end{bmatrix}$  is a  $2 \times 4$  matrix with **real entries**.

## Matrix addition and scalar multiplication

Let  $A = [a_{ij}]$  and  $B = [b_{ij}]$  be two  $m \times n$  matrices, and let  $c$  be a scalar.

- The matrices  $A$  and  $B$  are **equal** if and only if they have the same entries,

$$A = B \iff a_{ij} = b_{ij}, \text{ for all } i, j, 1 \leq i \leq m, 1 \leq j \leq n.$$

- The **sum** of  $A$  and  $B$  is the  $m \times n$  matrix obtained by adding the entries of  $A$  to those of  $B$ ,

$$A + B = [a_{ij} + b_{ij}].$$

- The **product** of  $A$  with the scalar  $c$  is the  $m \times n$  matrix obtained by multiplying the entries of  $A$  by  $c$ ,

$$cA = [c a_{ij}].$$

## 2. Matrix multiplication

- Let  $A = [a_{ij}]$  be an  $m \times n$  matrix and  $B = [b_{ij}]$  be an  $n \times p$  matrix. The **product**  $C = AB$  of  $A$  and  $B$  is an  **$m \times p$  matrix** whose entries are obtained by multiplying each row of  $A$  with each column of  $B$  as follows:

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

- Examples:** Let  $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$  and  $C = \begin{bmatrix} 1 & 2 & 3 & 10 \\ 1 & 6 & -8 & 0 \end{bmatrix}$ .
  - Is the product  $AC$  defined? If so, evaluate it.
  - Same question with the product  $CA$ .
  - What is the product of  $A$  with the third column vector of  $C$ ?

## Matrix multiplication (continued)

- More examples:**
  - Consider the system of equations

$$\begin{cases} 3x_1 + 2x_2 - x_3 = 4 \\ x_2 - 7x_3 = 0 \\ -x_1 + 4x_2 - 6x_3 = -10 \end{cases}.$$

Write this system in the form  $AX = Y$ , where  $A$  is a matrix and  $X$  and  $Y$  are two column vectors.

- Let

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}.$$

Calculate the products  $AB$  and  $BA$ .

### 3. Rules for matrix addition and multiplication

- The rules for matrix addition and multiplication by a scalar are **the same** as the rules for addition and multiplication of real or complex numbers.
- In particular, if  $A$  and  $B$  are matrices and  $c_1$  and  $c_2$  are scalars, then

$$\begin{aligned}
 A + B &= B + A \\
 (A + B) + C &= A + (B + C) \\
 c_1(A + B) &= c_1A + c_1B \\
 (c_1 + c_2)A &= c_1A + c_2A \\
 c_1(c_2A) &= (c_1c_2)A
 \end{aligned}$$

whenever the above quantities make sense.

### Rules for matrix addition and multiplication (continued)

- The product of two matrices is **associative** and **distributive**, i.e.

$$\begin{aligned}
 A(BC) &= (AB)C = ABC \\
 A(B + C) &= AB + AC \quad (A + B)C = AC + BC.
 \end{aligned}$$

- However, the **product** of two matrices is **not commutative**. If  $A$  and  $B$  are two square matrices, we typically have

$$AB \neq BA$$

- For two square matrices  $A$  and  $B$ , the **commutator** of  $A$  and  $B$  is defined as

$$[A, B] = AB - BA.$$

In general,  $[A, B] \neq 0$ . If  $[A, B] = 0$ , one says that the matrices  $A$  and  $B$  **commute**.

## 4. Transposition

- The **transpose** of an  $m \times n$  matrix  $A$  is the  $n \times m$  matrix  $A^T$  obtained from  $A$  by switching its rows and columns, i.e.

$$\text{if } A = [a_{ij}], \quad \text{then } A^T = [a_{ji}].$$

- Example:** Find the transpose of  $C = \begin{bmatrix} 1 & 2 & 3 & 10 \\ 1 & 6 & -8 & 0 \end{bmatrix}$ .
- Some properties of transposition.** If  $A$  and  $B$  are matrices, and  $c$  is a scalar, then

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

whenever the above quantities make sense.

## 5. Linear independence

- A **linear combination** of the  $n$  vectors  $a_1, a_2, \dots, a_n$  is an expression of the form

$$c_1 a_1 + c_2 a_2 + \dots + c_n a_n,$$

where the  $c_i$ 's are scalars.

- A set of vectors  $\{a_1, a_2, \dots, a_n\}$  is **linearly independent** if the only way of having a linear combination of these vectors equal to zero is by choosing all of the coefficients equal to zero. In other words,  $\{a_1, a_2, \dots, a_n\}$  is linearly independent if and only if

$$c_1 a_1 + c_2 a_2 + \dots + c_n a_n = 0 \implies c_1 = c_2 = \dots = c_n = 0.$$

## Linear independence (continued)

- **Examples:**

- Are the columns of the matrix  $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$  linearly independent?
- Same question with the columns of the matrix  $C = \begin{bmatrix} 1 & 2 & 3 & 10 \\ 1 & 6 & -8 & 0 \end{bmatrix}$ .
- Same question with the rows of the matrix  $C$  defined above.
- A set that is not linearly independent is called **linearly dependent**.
- Can you find a condition on a set of  $n$  vectors, which would guarantee that these vectors are linearly dependent?

## 6. Vector space

- A **real (or complex) vector space** is a non-empty set  $V$  whose elements are called vectors, and which is equipped with two operations called **vector addition** and **multiplication by a scalar**.
- The **vector addition** satisfies the following properties.
  - 1 The sum of two vectors  $a \in V$  and  $b \in V$  is denoted by  $a + b$  and is an element of  $V$ .
  - 2 It is **commutative**:  $a + b = b + a$ , for all  $a, b \in V$ .
  - 3 It is **associative**:  $(a + b) + c = a + (b + c)$  for all  $a, b, c \in V$ .
  - 4 There exists a unique **zero vector**, denoted by  $0$ , such that for every vector  $a \in V$ ,  $a + 0 = a$ .
  - 5 For each  $a \in V$ , there exists a unique vector  $(-a) \in V$  such that  $a + (-a) = 0$ .

## Vector space (continued)

- The **multiplication by a scalar** satisfies the following properties.
  - ① The multiplication of a vector  $a \in V$  by a scalar  $\alpha \in \mathbb{R}$  (or  $\alpha \in \mathbb{C}$ ) is denoted by  $\alpha a$  and is an element of  $V$ .

- ② Multiplication by a scalar is **distributive**:

$$\alpha(a + b) = \alpha a + \alpha b, \quad (\alpha + \beta)a = \alpha a + \beta a,$$

for all  $a, b \in V$  and  $\alpha, \beta \in \mathbb{R}$  (or  $\mathbb{C}$ ).

- ③ It is **associative**:  $\alpha(\beta a) = (\alpha\beta)a$  for all  $a \in V$  and  $\alpha, \beta \in \mathbb{R}$  (or  $\mathbb{C}$ ).

- ④ Multiplying a vector by 1 gives back that vector, i.e.

$$1 a = a,$$

for all  $a \in V$ .

## Bases and dimension

- The **span** of set of vectors  $\mathcal{U} = \{a_1, a_2, \dots, a_n\}$  is the set of all linear combinations of vectors in  $\mathcal{U}$ . It is denoted by

$$\text{Span}\{a_1, a_2, \dots, a_n\} \text{ or } \text{Span}(\mathcal{U})$$

and is a **subspace** of  $V$ .

- A **basis**  $\mathcal{B}$  of a subspace  $S$  of  $V$  is a set of vectors of  $S$  such that

- ①  $\text{Span}(\mathcal{B}) = S$ ;
- ②  $\mathcal{B}$  is a linearly independent set.

- **Theorem**: If a basis  $\mathcal{B}$  of a subspace  $S$  of  $V$  has  $n$  vectors, then all other bases of  $S$  have exactly  $n$  vectors.

- The **dimension** of a vector space  $V$  (or of a subspace  $S$  of  $V$ ) spanned by a finite number of vectors is the number of vectors in any of its bases.

## 7. Rank

- The **row space** of an  $m \times n$  matrix  $A$  is the span of the row vectors of  $A$ . If  $A$  has real entries, the row space of  $A$  is a subspace of  $\mathbb{R}^n$ .
- Similarly, the **column space** of  $A$  is the span of the column vectors of  $A$ , and is a subspace of  $\mathbb{R}^m$ .
- The **rank** of a matrix  $A$  is the dimension of its column space.
- **Theorem:** The dimensions of the row and column spaces of a matrix  $A$  are the same. They are equal to the rank of  $A$ .
- **Example:** Check that the row and column spaces of  $C = \begin{bmatrix} 1 & 2 & 3 & 10 \\ 1 & 6 & -8 & 0 \end{bmatrix}$  are vector subspaces, and find their dimension.

## The rank theorem

- The **null space** of an  $m \times n$  matrix  $A$ ,  $\mathcal{N}(A)$  is the set of vectors  $u$  such that  $Au = 0$ . If  $A$  has real entries, then  $\mathcal{N}(A)$  is a subspace of  $\mathbb{R}^n$ .
- The **rank theorem** states that if  $A$  is an  $m \times n$  matrix, then

$$\text{rank}(A) + \dim(\mathcal{N}(A)) = n.$$

- **Example:** Find the rank and the null space of the matrix  $C = \begin{bmatrix} 1 & 2 & 3 & 10 \\ 1 & 6 & -8 & 0 \end{bmatrix}$ . Check that the rank theorem applies.