# CHAPTER 1

## Mathematical Review

## 1.1 Linear Algebra

## 1.1.1 Three-Dimensional Vector Algebra

#### Exercise 1.1

a) Show that  $O_{ij} = \hat{e}_i \cdot \mathcal{O}\hat{e}_j$ . b) If  $\mathcal{O}\vec{a} = \vec{b}$  show that  $b_i = \sum_i O_{ij} a_j$ .

## Solution 1.1

1. Using (1.7) and (1.13), we get that

$$\hat{e}_i \cdot \mathcal{O}\hat{e}_j = \hat{e}_i \cdot \sum_{k=1}^3 \hat{e}_k O_{kj} = \sum_{k=1}^3 \hat{e}_i \cdot \hat{e}_k O_{kj} = \sum_{k=1}^3 \delta_{ik} O_{kj} = O_{ij}. \tag{1.1-1}$$

2. Similarly,

$$\vec{b} = \sum_{i=1}^{3} b_i \hat{e}_i = \mathscr{O}\vec{a} = \mathscr{O}\sum_{j=1}^{3} a_j \hat{e}_j = \sum_{j=1}^{3} a_j \mathscr{O}\hat{e}_j = \sum_{j=1}^{3} a_j \sum_{i=1}^{3} \hat{e}_i O_{ij} = \sum_{i=1}^{3} \left(\sum_{j=1}^{3} O_{ij} a_j\right) \hat{e}_i.$$

From the uniqueness of linear expression by a basis, we arrive at

$$b_i = \sum_{j=1}^3 O_{ij} a_j. (1.1-2)$$

#### Exercise 1.2

Calculate  $[\mathbf{A}, \mathbf{B}]$  and  $\{\mathbf{A}, \mathbf{B}\}$  when

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

## Solution 1.2

$$[\mathbf{A}, \mathbf{B}] \equiv \mathbf{A}\mathbf{B} - \mathbf{B}\mathbf{A} = \begin{pmatrix} 0 & -1 & 1 \\ 1 & -1 & 3 \\ -3 & 0 & -1 \end{pmatrix} - \begin{pmatrix} 0 & 1 & -3 \\ -1 & -1 & 0 \\ 1 & 3 & -1 \end{pmatrix} = \begin{pmatrix} 0 & -2 & 4 \\ 2 & 0 & 3 \\ -4 & -3 & 0 \end{pmatrix},$$

$$\{\mathbf{A}, \mathbf{B}\} \equiv \mathbf{A}\mathbf{B} + \mathbf{B}\mathbf{A} = \begin{pmatrix} 0 & -1 & 1 \\ 1 & -1 & 3 \\ -3 & 0 & -1 \end{pmatrix} + \begin{pmatrix} 0 & 1 & -3 \\ -1 & -1 & 0 \\ 1 & 3 & -1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & -2 \\ 0 & -2 & 3 \\ -2 & 3 & -2 \end{pmatrix}.$$

### 1.1.2 Matrices

#### Exercise 1.3

If **A** is an  $N \times M$  matrix and **B** is a  $M \times K$  matrix show that  $(\mathbf{AB})^{\dagger} = \mathbf{B}^{\dagger} \mathbf{A}^{\dagger}$ .

#### Solution 1.3

It is obvious that

$$(\mathbf{B}^{\dagger}\mathbf{A}^{\dagger})_{ij} = \sum_{k=1}^{M} (\mathbf{B}^{\dagger})_{ik} (\mathbf{A}^{\dagger})_{kj} = \sum_{k=1}^{M} B_{ki}^{*} A_{jk}^{*} = \left(\sum_{k=1}^{M} A_{jk} B_{ki}\right)^{*} = [(\mathbf{A}\mathbf{B})^{*}]_{ji} = [(\mathbf{A}\mathbf{B})^{\dagger}]_{ij}, \quad (1.3-1)$$

which means that  $(\mathbf{A}\mathbf{B})^{\dagger} = \mathbf{B}^{\dagger}\mathbf{A}^{\dagger}$ .

#### Exercise 1.4

Show that

- a.  $tr(\mathbf{AB}) = tr(\mathbf{BA})$ .
- b.  $(\mathbf{AB})^{-1} = \mathbf{B}^{-1}\mathbf{A}^{-1}$ .
- c. If U is unitary and  $\mathbf{B} = \mathbf{U}^{\dagger} \mathbf{A} \mathbf{U}$ , then  $\mathbf{A} = \mathbf{U} \mathbf{B} \mathbf{U}^{\dagger}$ .
- d. If the product C = AB of two Hermitian matrices is also Hermitian, then A and B commute.
- e. If **A** is Hermitian then  $A^{-1}$ , if it exists, is also Hermitian.

f. If 
$$\mathbf{A} = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}$$
, then  $\mathbf{A}^{-1} = \frac{1}{(A_{11}A_{22} - A_{12}A_{21})} \begin{pmatrix} A_{22} & -A_{12} \\ -A_{21} & A_{11} \end{pmatrix}$ .

#### Solution 1.4

a. At this time, we assume that **A** is an  $N \times M$  matrix while **B** is a  $M \times N$  matrix. Then,

$$\operatorname{tr}(\mathbf{AB}) = \sum_{i=1}^{N} (\mathbf{AB})_{ii} = \sum_{i=1}^{N} \sum_{k=1}^{M} A_{ik} B_{ki} = \sum_{k=1}^{M} \sum_{i=1}^{N} B_{ki} A_{ik} = \sum_{k=1}^{M} (\mathbf{BA})_{kk} = \operatorname{tr}(\mathbf{BA}).$$
(1.4-1)

From this issue, we assume that both **A** and **B** are  $N \times N$  matrices.

b. We find that

$$AB(B^{-1}A^{-1}) = A(BB^{-1})A^{-1} = AA^{-1} = 1.$$

Since the inverse of a matrix is unique, we immediately get that

$$(\mathbf{A}\mathbf{B})^{-1} = \mathbf{B}^{-1}\mathbf{A}^{-1}. (1.4-2)$$

c. Due to  $\mathbf{B} = \mathbf{U}^{\dagger} \mathbf{A} \mathbf{U}$ , we can find

$$\mathbf{A} = \mathbf{1}\mathbf{A}\mathbf{1} = (\mathbf{U}\mathbf{U}^{\dagger})\mathbf{A}(\mathbf{U}\mathbf{U}^{\dagger}) = \mathbf{U}(\mathbf{U}^{\dagger}\mathbf{A}\mathbf{U})\mathbf{U}^{\dagger} = \mathbf{U}\mathbf{B}\mathbf{U}^{\dagger}. \tag{1.4-3}$$

d. Because C = AB of two Hermitian matrices is also Hermitian, we know that

$$\mathbf{C}^{\dagger} = (\mathbf{A}\mathbf{B})^{\dagger} = \mathbf{B}^{\dagger}\mathbf{A}^{\dagger} = \mathbf{C} = \mathbf{A}\mathbf{B}.$$

With  $\mathbf{A}^{\dagger} = \mathbf{A}$ ,  $\mathbf{B}^{\dagger} = \mathbf{B}$ , we find

$$\mathbf{B}^{\dagger} \mathbf{A}^{\dagger} = \mathbf{B} \mathbf{A} = \mathbf{A} \mathbf{B}. \tag{1.4-4}$$

In other words, **A** and **B** commute.

e. It is obvious that if  $A^{-1}$  exists,

$$(\mathbf{A}^{-1})^{\dagger} \mathbf{A}^{\dagger} = (\mathbf{A} \mathbf{A}^{-1})^{\dagger} = \mathbf{1}^{\dagger} = \mathbf{1}.$$

We know  $(\mathbf{A}^{\dagger})^{-1} = (\mathbf{A}^{-1})^{\dagger}$ . Then, with  $\mathbf{A} = \mathbf{A}^{\dagger}$ , we find that

$$(\mathbf{A}^{-1})^{\dagger} = (\mathbf{A}^{\dagger})^{-1} = \mathbf{A}^{-1}. \tag{1.4-5}$$

Namely,  $\mathbf{A}^{-1}$  is also Hermitian if it exists.

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f. If  $A_{11}A_{22} - A_{12}A_{21} \neq 0$ , we can find

$$\mathbf{A} \times \frac{1}{A_{11}A_{22} - A_{12}A_{21}} \begin{pmatrix} A_{22} & -A_{12} \\ -A_{21} & A_{11} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix},$$

which means that if  $A_{11}A_{22} - A_{12}A_{21} \neq 0$ ,

$$\mathbf{A}^{-1} = \frac{1}{A_{11}A_{22} - A_{12}A_{21}} \begin{pmatrix} A_{22} & -A_{12} \\ -A_{21} & A_{11} \end{pmatrix}. \tag{1.4-6}$$

#### 1.1.3 Determinants

#### Exercise 1.5

Verify the above properties for  $2 \times 2$  determinants.

### Solution 1.5

#### Exercise 1.6

Using properties (1)-(5) prove that in general

- 6. If any two rows (or columns) of a determinant are equal, the value of the determinant is zero.
- 7.  $|\mathbf{A}^{-1}| = (|\mathbf{A}|)^{-1}$ .
- 8. If  $AA^{\dagger} = 1$ , then  $|A|(|A|)^* = 1$ .
- 9. If  $\mathbf{U}^{\dagger}\mathbf{O}\mathbf{U} = \mathbf{\Omega}$  and  $\mathbf{U}^{\dagger}\mathbf{U} = \mathbf{U}\mathbf{U}^{\dagger} = \mathbf{1}$ , then  $|\mathbf{O}| = |\mathbf{\Omega}|$ .

## Solution 1.6

#### Exercise 1.7

Using Eq.(1.39), note that the inverse of a  $2 \times 2$  matrix **A** obtained in Exercise 1.4f can be written as

$$\mathbf{A}^{-1} = \frac{1}{|\mathbf{A}|} \begin{pmatrix} A_{22} & -A_{12} \\ -A_{21} & A_{11} \end{pmatrix}$$

and thus  $\mathbf{A}^{-1}$  does not exist when  $|\mathbf{A}| = 0$ . This result holds in general for  $N \times N$  matrices. Show that the equation

$$Ac = 0$$

where **A** is an  $N \times N$  matrix and **c** is a column matrix with elements  $c_i$ , i = 1, 2, ..., N can have a nontrivial solution ( $\mathbf{c} \neq 0$ ) only when  $|\mathbf{A}| = 0$ .

## **Solution 1.7**