

# Slide 16-Determinants II

MAT2040 Linear Algebra

### Property 16.1 (Determinant for Row or Column Interchange)

Suppose that  $A$  is a square matrix, and  $A \xrightarrow{R_i \rightarrow R_j} B$ , or  $A \xrightarrow{C_i \rightarrow C_j} B$ . Then  $\det(B) = -\det(A)$ .

**Proof.** This can also be proved by mathematical induction.

### Remark for elementary row operation I:

Recall elementary row operation I:

$I_n \xrightarrow{R_i \rightarrow R_j} E_{R_i R_j}$ . Thus  $\det(E_{R_i R_j}) = -\det(I_n) = -1$ . And  $\det(E_{R_i R_j} A) = -\det(A) = \det(E_{R_i R_j}) \det(A)$ .

## Examples 16.2

$$\det(A) = \begin{vmatrix} 0 & 0 & 0 & a_{14} \\ 0 & 0 & a_{23} & a_{24} \\ 0 & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{vmatrix}$$

$$\begin{vmatrix} 0 & 0 & 0 & a_{14} \\ 0 & 0 & a_{23} & a_{24} \\ 0 & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{vmatrix} (R_1 \leftrightarrow R_4) = - \begin{vmatrix} a_{41} & a_{42} & a_{43} & a_{44} \\ 0 & 0 & a_{23} & a_{24} \\ 0 & a_{32} & a_{33} & a_{34} \\ 0 & 0 & 0 & a_{14} \end{vmatrix} (R_2 \leftrightarrow R_3)$$
$$= \begin{vmatrix} a_{41} & a_{42} & a_{43} & a_{44} \\ 0 & a_{32} & a_{33} & a_{34} \\ 0 & 0 & a_{23} & a_{24} \\ 0 & 0 & 0 & a_{14} \end{vmatrix}$$
$$= a_{14} a_{32} a_{23} a_{41}$$

**Property 16.3** Suppose that  $A$  is a square matrix. Let  $A \xrightarrow{R_i \rightarrow \alpha R_i (\alpha \neq 0)} B$  or  $A \xrightarrow{C_i \rightarrow \alpha C_i (\alpha \neq 0)} B$ . Then  $\det(B) = \alpha \det(A)$ .

**Proof.** If the  $i$ th row of  $A$  is multiplied by  $\alpha (\alpha \neq 0)$  to obtain  $B$ , then expand  $B$  along  $i$ th row, one obtains that

$$\begin{aligned}\det(B) &= b_{i1}A_{i1} + b_{i2}A_{i2} + \cdots + b_{in}A_{in} \\ &= \alpha a_{i1}A_{i1} + \alpha a_{i2}A_{i2} + \cdots + \alpha a_{in}A_{in} \\ &= \alpha (a_{i1}A_{i1} + a_{i2}A_{i2} + \cdots + a_{in}A_{in}) \\ &= \alpha \det(A).\end{aligned}$$

The proof for column operation is similar.

## Remark for elementary row operation II:

Recall elementary row operation II:  $I_n \xrightarrow{R_i \rightarrow \alpha R_i (\alpha \neq 0)} E_{\alpha R_i}$ .

Thus,  $\det(E_{\alpha R_i}) = \alpha \det(I_n) = \alpha (\alpha \neq 0)$ , and

$$\det(E_{\alpha R_i} A) = \alpha \det(A) = \det(E_{\alpha R_i}) \det(A) (\alpha \neq 0).$$

**Lemma 16.4** Let  $A = (a_{ij})_{n \times n}$ ,  $A_{ij}$  is the cofactor of  $a_{ij}$ , then

$$a_{i1}A_{j1} + a_{i2}A_{j2} + \cdots + a_{in}A_{jn} = \begin{cases} \det(A), & \text{if } i = j, \\ 0, & \text{if } i \neq j \end{cases}$$

**Proof.** Define a new determinant

$$\det(A^*) = \begin{vmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \vdots & \vdots \\ a_{i1} & \cdots & a_{in} \\ \vdots & \vdots & \vdots \\ a_{i1} & \cdots & a_{in} \\ \vdots & \vdots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{vmatrix} \begin{matrix} \text{\textit{ith row}} \\ \\ \text{\textit{jth row}} \end{matrix} = \begin{cases} \det(A), & \text{if } i = j, \\ 0, & \text{if } i \neq j, \end{cases}$$

On the other hand, expand along  $j$ th row for  $\det(A^*)$  gives

$$\det(A^*) = a_{i1}A_{j1} + a_{i2}A_{j2} + \cdots + a_{in}A_{jn}. \text{ This completes the proof.}$$

**Remark** The cofactor  $A_{ij}$  of matrix  $A$  and matrix  $A^*$  are the same only along  $j$ th row.

**Property 16.5** Suppose that  $A$  is a square matrix, and  $A \xrightarrow{R_j \rightarrow \beta R_i + R_j} B$  ( $i \neq j$ ). Then  $\det(B) = \det(A)$ , i.e.,

$$\begin{array}{l}
 \text{\textit{ith row}} \\
 \text{\textit{jth row}}
 \end{array}
 \left| \begin{array}{ccc}
 a_{11} & \cdots & a_{1n} \\
 \vdots & \cdots & \vdots \\
 a_{i1} & \cdots & a_{in} \\
 \vdots & \cdots & \vdots \\
 \beta a_{i1} + a_{j1} & \cdots & \beta a_{in} + a_{jn} \\
 \vdots & \cdots & \vdots \\
 a_{n1} & \cdots & a_{nn}
 \end{array} \right| = \left| \begin{array}{ccc}
 a_{11} & \cdots & a_{1n} \\
 \vdots & \cdots & \vdots \\
 a_{i1} & \cdots & a_{in} \\
 \vdots & \cdots & \vdots \\
 a_{j1} & \cdots & a_{jn} \\
 \vdots & \cdots & \vdots \\
 a_{n1} & \cdots & a_{nn}
 \end{array} \right|
 \begin{array}{l}
 \text{\textit{ith row}} \\
 \text{\textit{jth row}}
 \end{array}$$

**Proof.** Due to  $A \xrightarrow{R_j \rightarrow \beta R_i + R_j} B$  ( $i \neq j$ ). Expanding along  $j$ th row for  $B$ , one gets that

$$\begin{aligned}\det(B) &= b_{j1}A_{j1} + b_{j2}A_{j2} + \cdots + b_{jn}A_{jn} \\ &= (\beta a_{i1} + a_{j1})A_{j1} + (\beta a_{i2} + a_{j2})A_{j2} + \cdots + (\beta a_{in} + a_{jn})A_{jn} \\ &= \beta(a_{i1}A_{j1} + a_{i2}A_{j2} + \cdots + a_{in}A_{jn}) \\ &\quad + (a_{j1}A_{j1} + a_{j2}A_{j2} + \cdots + a_{jn}A_{jn}) \\ &= \det(A)\end{aligned}$$

where we have used the fact that  $a_{i1}A_{j1} + a_{i2}A_{j2} + \cdots + a_{in}A_{jn} = 0$  since  $i \neq j$ .

**Remark:** 1. The cofactor  $A_{ij}$  of matrix  $A$  and matrix  $B$  are the same only along  $j$ th row.

2. If  $A \xrightarrow{C_j \rightarrow \beta C_i + C_j} B$  ( $i \neq j$ ). Then  $\det(B) = \det(A)$ .



## Remark elementary row operation III:

Recall elementary row operation III:

$$I_n \xrightarrow{R_j \rightarrow \beta R_i + R_j} E_{\beta R_i + R_j}, \quad \det(E_{\beta R_i + R_j}) = \det(I_n) = 1. \text{ Thus,} \\ \det(E_{\beta R_i + R_j} A) = \det(A) = \det(E_{\beta R_i + R_j}) \det(A).$$

# Summary

Property 16.1, Property 16.3, Property 16.5 are the properties for determinants with respect to row operations (column operations).

- (1) When using row operation I ( $R_i \leftrightarrow R_j$ ) or column operation I ( $C_i \leftrightarrow C_j$ ), the determinant after the operation equals  $(-1)$  times the determinant before the operation.
- (2) When using row operation II ( $R_i \rightarrow \alpha R_i$ ) ( $\alpha \neq 0$ ) or column operation II ( $C_i \rightarrow \alpha C_i$ ) ( $\alpha \neq 0$ ), the determinant after the operation equals the  $\alpha$  times the determinant before the operation.
- (3) When using row operation III ( $R_j \rightarrow \beta R_i + R_j$ ) or column operation III ( $C_j \rightarrow \beta C_i + C_j$ ), the determinant after the operation equals the determinant before the operation.

# Summary for determinants of elementary matrices

## Property 16.6 (Determinants of Three Elementary Matrices)

$$(1) \quad I_n \xrightarrow{R_i \rightarrow R_j} E_{R_i R_j}, \quad \det(E_{R_i R_j}) = -\det(I_n) = -1, \\ \det(E_{R_i R_j} A) = -\det(A) = \det(E_{R_i R_j}) \det(A).$$

$$(2) \quad I_n \xrightarrow{R_i \rightarrow \alpha R_i (\alpha \neq 0)} E_{\alpha R_i}, \quad \det(E_{\alpha R_i}) = \alpha \det(I_n) = \alpha (\alpha \neq 0), \\ \det(E_{\alpha R_i} A) = \alpha \det(A) = \det(E_{\alpha R_i}) \det(A) (\alpha \neq 0).$$

$$(3) \quad I_n \xrightarrow{R_j \rightarrow \beta R_i + R_j} E_{\beta R_i + R_j}, \quad \det(E_{\beta R_i + R_j}) = \det(I_n) = 1, \\ \det(E_{\beta R_i + R_j} A) = \det(A) = \det(E_{\beta R_i + R_j}) \det(A).$$

**Property 16.7 (Determinant for an Elementary Matrix Multiply a Matrix)** Suppose that  $A$  is a square matrix of size  $n$  and  $E$  is an elementary matrix of size  $n$ , then

$$\det(EA) = \det(E) \det(A)$$

$$\det(AE) = \det(A) \det(E)$$

### Proof

For elementary matrices type I, II, III,  $\det(EA) = \det(E) \det(A)$  has already been shown in previous discussion.

$E$  is an elementary matrix, thus  $E^T$  is also an elementary matrix. Therefore

$$\begin{aligned} \det(AE) &= \det((AE)^T) \\ &= \det(E^T A^T) \\ &= \det(E^T) \det(A^T) \\ &= \det(E) \det(A) \end{aligned}$$

**Example 16.8** (1). Given  $\det(A) = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = 1$ , find

$$\begin{vmatrix} 2a_{11} & 3a_{12} & 4a_{13} \\ 2a_{21} & 3a_{22} & 4a_{23} \\ 2a_{31} & 3a_{32} & 4a_{33} \end{vmatrix}$$

$$\begin{vmatrix} 2a_{11} & 3a_{12} & 4a_{13} \\ 2a_{21} & 3a_{22} & 4a_{23} \\ 2a_{31} & 3a_{32} & 4a_{33} \end{vmatrix} = 2 \begin{vmatrix} a_{11} & 3a_{12} & 4a_{13} \\ a_{21} & 3a_{22} & 4a_{23} \\ a_{31} & 3a_{32} & 4a_{33} \end{vmatrix}$$

$$= 2 * 3 \begin{vmatrix} a_{11} & a_{12} & 4a_{13} \\ a_{21} & a_{22} & 4a_{23} \\ a_{31} & a_{32} & 4a_{33} \end{vmatrix}$$

$$= 2 * 3 * 4 * \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = 24$$

(2). Compute

$$\begin{vmatrix} 1 & 1 & 1 & 1 & 1 \\ 2 & 2 & 1 & 2 & 2 \\ 1 & 2 & 1 & 2 & 3 \\ 1 & 1 & 1 & 3 & 2 \\ 1 & 1 & 1 & 1 & 4 \end{vmatrix} \begin{pmatrix} R_2 \rightarrow -R_1 + R_2 \\ R_3 \rightarrow -R_1 + R_3 \\ R_4 \rightarrow -R_1 + R_4 \\ R_5 \rightarrow -R_1 + R_5 \end{pmatrix}$$
$$= \begin{vmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 & 2 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 3 \end{vmatrix} \text{ (expand along 3rd column)}$$
$$= (-1)^{1+3} \begin{vmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 2 \\ 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 3 \end{vmatrix}$$
$$= (1)(1)(2)(3) = 6$$

(3). Compute

$$\begin{vmatrix} 1 & 1 & 1 & 1 & 1 \\ 2 & 2 & 1 & 2 & 2 \\ 1 & 2 & 1 & 2 & 3 \\ 1 & 1 & 1 & 3 & 2 \\ 1 & 1 & 1 & 1 & 4 \end{vmatrix} \begin{pmatrix} R_2 \rightarrow -2R_1 + R_2 \\ R_3 \rightarrow -R_1 + R_3 \\ R_4 \rightarrow -R_1 + R_4 \\ R_5 \rightarrow -R_1 + R_5 \end{pmatrix}$$
$$= \begin{vmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & -1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 2 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 3 \end{vmatrix} (R_2 \leftrightarrow R_3)$$
$$= - \begin{vmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 1 & 2 \\ 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 3 \end{vmatrix} = -(1)(1)(-1)(2)(3) = 6$$

(4)

$$\begin{aligned} & \begin{vmatrix} a & 1 & 1 & 1 \\ 1 & a & 1 & 1 \\ 1 & 1 & a & 1 \\ 1 & 1 & 1 & a \end{vmatrix} (R_1 \leftrightarrow R_2) = - \begin{vmatrix} 1 & a & 1 & 1 \\ a & 1 & 1 & 1 \\ 1 & 1 & a & 1 \\ 1 & 1 & 1 & a \end{vmatrix} \begin{pmatrix} R_2 \rightarrow -aR_1 + R_2 \\ R_3 \rightarrow -R_1 + R_3 \\ R_4 \rightarrow -R_1 + R_4 \end{pmatrix} \\ &= - \begin{vmatrix} 1 & a & 1 & 1 \\ 0 & 1-a^2 & 1-a & 1-a \\ 0 & 1-a & a-1 & 0 \\ 0 & 1-a & 0 & a-1 \end{vmatrix} \text{ (Take out } (1-a) \text{ from } R_2 \ R_3 \ R_4) \\ &= -(1-a)^3 \begin{vmatrix} 1 & a & 1 & 1 \\ 0 & 1+a & 1 & 1 \\ 0 & 1 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{vmatrix} (R_2 \leftrightarrow R_4) \end{aligned}$$



$$=(1-a)^3 \begin{vmatrix} 1 & a & 1 & 1 \\ 0 & 1 & 0 & -1 \\ 0 & 1 & -1 & 0 \\ 0 & 1+a & 1 & 1 \end{vmatrix} \begin{pmatrix} R_3 \rightarrow -R_2 + R_3 \\ R_4 \rightarrow -(1+a)R_2 + R_4 \end{pmatrix}$$

$$=(1-a)^3 \begin{vmatrix} 1 & a & 1 & 1 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 1 & a+2 \end{vmatrix} (R_4 \rightarrow R_3 + R_4)$$

$$=(1-a)^3 \begin{vmatrix} 1 & a & 1 & 1 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & a+3 \end{vmatrix}$$

$$=(1-a)^3 (1)(1)(-1)(a+3) = (a-1)^3 (a+3)$$

# Method to compute the determinant

Using row operations (column operations) to reduce the corresponding matrix into the upper triangular form or lower triangular form.

## Property 16.9

$$(I) \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{i-1,1} & a_{i-1,2} & \cdots & a_{i-1,n} \\ b_1 + c_1 & b_2 + c_2 & \cdots & b_n + c_n \\ a_{i+1,1} & a_{i+1,2} & \cdots & a_{i+1,n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} = \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{i-1,1} & a_{i-1,2} & \cdots & a_{i-1,n} \\ b_1 & b_2 & \cdots & b_n \\ a_{i+1,1} & a_{i+1,2} & \cdots & a_{i+1,n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} + \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{i-1,1} & a_{i-1,2} & \cdots & a_{i-1,n} \\ c_1 & c_2 & \cdots & c_n \\ a_{i+1,1} & a_{i+1,2} & \cdots & a_{i+1,n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix}$$

(II)

$$\begin{aligned} & \begin{vmatrix} a_{11} & \cdots & a_{1,i-1} & b_1 + c_1 & a_{1,i+1} & \cdots & a_{1n} \\ a_{21} & \cdots & a_{2,i-1} & b_2 + c_2 & a_{2,i+1} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & \cdots & a_{n,i-1} & b_n + c_n & a_{n,i+1} & \cdots & a_{nn} \end{vmatrix} \\ = & \begin{vmatrix} a_{11} & \cdots & a_{1,i-1} & b_1 & a_{1,i+1} & \cdots & a_{1n} \\ a_{21} & \cdots & a_{2,i-1} & b_2 & a_{2,i+1} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & \cdots & a_{n,i-1} & b_n & a_{n,i+1} & \cdots & a_{nn} \end{vmatrix} \\ + & \begin{vmatrix} a_{11} & \cdots & a_{1,i-1} & c_1 & a_{1,i+1} & \cdots & a_{1n} \\ a_{21} & \cdots & a_{2,i-1} & c_2 & a_{2,i+1} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & \cdots & a_{n,i-1} & c_n & a_{n,i+1} & \cdots & a_{nn} \end{vmatrix} \end{aligned}$$

**Proof.** (I) Expand along the  $i$ th row. (II) Expand along the  $i$ th column.

**Example** Given  $\det(A) = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = 1$ , then

$$\det(B) = \begin{vmatrix} a_{11} + 10a_{12} + 5a_{13} & a_{12} & a_{13} \\ a_{21} + 10a_{22} + 5a_{23} & a_{22} & a_{23} \\ a_{31} + 10a_{32} + 5a_{33} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} +$$

$$\begin{vmatrix} 10a_{12} & a_{12} & a_{13} \\ 10a_{22} & a_{22} & a_{23} \\ 10a_{32} & a_{32} & a_{33} \end{vmatrix} + \begin{vmatrix} 5a_{13} & a_{12} & a_{13} \\ 5a_{23} & a_{22} & a_{23} \\ 5a_{33} & a_{32} & a_{33} \end{vmatrix} = 1 + 0 + 0 = 1.$$

**Theorem 16.10 (Equivalent Condition for Singular Matrices)** Let  $A \in \mathbb{R}^{n \times n}$ ,  $A$  is singular if and only if

$$\det(A) = 0$$

**Proof.** The matrix  $A$  can be reduced into reduced row-echelon form  $B$  with a finite number of elementary row operations. Thus,

$$B = E_k E_{k-1} \cdots E_1 A,$$

where all  $E_i, i = 1, \dots, k$  are elementary matrices.

By property 16.7, one has

$$\det(B) = \det(E_k E_{k-1} \cdots E_1 A) = \det(E_k) \det(E_{k-1}) \cdots \det(E_1) \det(A)$$

Since the determinants of  $E_s(s = 1, \dots, k)$  are all nonzeros, it follows that  $\det(A) = 0$  if and only if  $\det(B) = 0$ . Thus, we only need to prove that  $A$  is singular if and only if  $\det(B) = 0$ .

If  $A$  is singular, then  $B$  has a row consisting entirely zeros, and hence  $\det(B) = 0$ .

If  $\det(B) = 0$ , suppose that  $A$  is nonsingular, then each row of  $B$  has a leading 1,  $B$  must be the identity matrix, and hence  $\det(B) = 1$ , which is a contradiction.

**Corollary (Equivalent condition for a nonsingular matrix)** Let

$A \in \mathbb{R}^{n \times n}$ ,  $A$  is nonsingular  $\Leftrightarrow \det(A) \neq 0$ .



**Theorem 16.11 (Determinants of Matrices Product)** If  $A$  and  $B$  are square matrices, then

$$\det(AB) = \det(A) \det(B)$$

**Proof.** (1) If  $A$  is singular, one has  $\det(A) = 0$  by theorem 16.10, and  $AB$  is also singular by using the Theorem 8.8. Thus,  $\det(AB) = 0$  by Theorem 16.10.

(2) If  $A$  is nonsingular, then  $A$  can be written as a product of elementary matrices by using the Theorem 8.7,  $A = E_1 E_2 \cdots E_k$ , where  $E_1, \dots, E_k$  are elementary matrices.

$$\begin{aligned} \det(AB) &= \det(E_1 E_2 \cdots E_k B) \\ &= \det(E_1) \det(E_2 \cdots E_k B) \\ &= \det(E_1) \det(E_2) \cdots \det(E_k) \det(B) \\ &= \det(E_1 E_2 \cdots E_k) \det(B) \\ &= \det(A) \det(B) \end{aligned}$$

where Theorem 16.7 is used.

**Example 16.12** Let

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 3 & 1 \\ -1 & 2 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} -1 & 1 & 1 \\ 2 & -3 & 1 \\ 1 & 2 & -1 \end{bmatrix}$$

$$AB = \begin{bmatrix} 2 & 0 & 1 \\ 5 & -5 & 4 \\ 6 & -5 & 0 \end{bmatrix}$$

$$\det(AB) = 45, \quad \det(A) = 5, \quad \det(B) = 9$$

$$\det(AB) = \det(A) \det(B)$$

## Question:

Is it true that  $\det(A + B) = \det(A) + \det(B)$ ? (No.)

**Definition 16.13 (Adjoint Matrix)** Let  $A = (a_{ij})_{n \times n}$ , then the adjoint matrix of  $A$  is defined as

$$\text{adj}(A) = \begin{bmatrix} A_{11} & A_{12} & \cdots & A_{1n} \\ A_{21} & A_{22} & \cdots & A_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1} & A_{n2} & \cdots & A_{nn} \end{bmatrix}^T = \begin{bmatrix} A_{11} & A_{21} & \cdots & A_{n1} \\ A_{12} & A_{22} & \cdots & A_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ A_{1n} & A_{2n} & \cdots & A_{nn} \end{bmatrix}$$

where  $A_{ij} = (-1)^{i+j} \det(M_{ij})$  is the cofactor of  $a_{ij}$ .

**Theorem 16.14 (Adjoint Matrix)** Let  $A = (a_{ij})_{n \times n}$ , then  $A \operatorname{adj}(A) = \det(A)I_n$ . If  $\det(A) \neq 0$ ,  $A^{-1} = \frac{1}{\det(A)} \operatorname{adj}(A)$ .

**Proof.** By Lemma 16.4, one has

$$a_{i1}A_{j1} + a_{i2}A_{j2} + \cdots + a_{in}A_{jn} = \begin{cases} \det(A), & \text{if } i = j, \\ 0, & \text{if } i \neq j, \end{cases}$$

And  $(i, j)$ -entry of  $A \operatorname{adj}(A)$  is  $a_{i1}A_{j1} + a_{i2}A_{j2} + \cdots + a_{in}A_{jn}$ . Thus

$$A \operatorname{adj}(A) = \det(A)I_n$$

In addition, if  $\det(A) \neq 0$ ,  $A^{-1} = \frac{1}{\det(A)} \operatorname{adj}(A)$ .

# Methods to find the inverse of $A$

$$(1) [A|I] \xrightarrow{\text{Row operations}} [I|A^{-1}]$$

$$(2) A^{-1} = \frac{1}{\det(A)} \text{adj}(A) \text{ (Straightforward but massive calculations)}$$

**Example 16.15** Let

$$A = \begin{bmatrix} 2 & 1 & 2 \\ 3 & 2 & 2 \\ 1 & 2 & 3 \end{bmatrix}$$

$$\text{adj}(A) = \begin{bmatrix} \begin{vmatrix} 2 & 2 \\ 2 & 3 \end{vmatrix} & -\begin{vmatrix} 3 & 2 \\ 1 & 3 \end{vmatrix} & \begin{vmatrix} 3 & 2 \\ 1 & 2 \end{vmatrix} \\ -\begin{vmatrix} 1 & 2 \\ 2 & 3 \end{vmatrix} & \begin{vmatrix} 2 & 2 \\ 1 & 3 \end{vmatrix} & -\begin{vmatrix} 2 & 1 \\ 1 & 2 \end{vmatrix} \\ \begin{vmatrix} 1 & 2 \\ 2 & 2 \end{vmatrix} & -\begin{vmatrix} 2 & 2 \\ 3 & 2 \end{vmatrix} & \begin{vmatrix} 2 & 1 \\ 3 & 2 \end{vmatrix} \end{bmatrix}^T = \begin{bmatrix} 2 & 1 & -2 \\ -7 & 4 & 2 \\ 4 & -3 & 1 \end{bmatrix}$$

$$\begin{vmatrix} 2 & 1 & 2 \\ 3 & 2 & 2 \\ 1 & 2 & 3 \end{vmatrix} = - \begin{vmatrix} 1 & 2 & 3 \\ 3 & 2 & 2 \\ 2 & 1 & 2 \end{vmatrix} = - \begin{vmatrix} 1 & 2 & 3 \\ 0 & -4 & -7 \\ 0 & -3 & -4 \end{vmatrix} \\
 = - \begin{vmatrix} -4 & -7 \\ -3 & -4 \end{vmatrix} = -((-4)(-4) - (-3)(-7)) = 5$$

Thus

$$A^{-1} = \frac{1}{\det(A)} \text{adj}(A) = \frac{1}{5} \begin{bmatrix} 2 & 1 & -2 \\ -7 & 4 & 2 \\ 4 & -3 & 1 \end{bmatrix}$$



**Theorem 16.16 (Cramer's Rule)** Let  $A = [\mathbf{a}_1, \dots, \mathbf{a}_n]$  ( $\mathbf{a}_1, \dots, \mathbf{a}_n$  are column vectors of  $A$ ) be a nonsingular  $n \times n$  matrix and let  $\mathbf{b} \in \mathbb{R}^n$  and  $A_i = [\mathbf{a}_1, \dots, \mathbf{a}_{i-1}, \mathbf{b}, \mathbf{a}_{i+1}, \dots, \mathbf{a}_n]$ , then the unique solution of  $A\mathbf{x} = \mathbf{b}$  is given by

$$x_i = \frac{\det(A_i)}{\det(A)}, \quad \text{for } i = 1, 2, \dots, n$$

**Proof.**

$$\mathbf{x} = A^{-1}\mathbf{b} = \frac{1}{\det(A)} \text{adj}(A)\mathbf{b}$$

It follows that

$$x_i = \frac{b_1 A_{1i} + b_2 A_{2i} + \dots + b_n A_{ni}}{\det(A)} = \frac{\det(A_i)}{\det(A)}$$

where  $\det(A_i) = b_1 A_{1i} + b_2 A_{2i} + \dots + b_n A_{ni}$  (expand along  $i$ th column).

**Example 16.17** Solve the following linear system by using the Cramer's rule:

$$\begin{aligned}x_1 + 2x_2 + x_3 &= 5, \\2x_1 + 2x_2 + x_3 &= 6, \\x_1 + 2x_2 + 3x_3 &= 9.\end{aligned}$$

We calculate:

$$\det(A) = \begin{vmatrix} 1 & 2 & 1 \\ 2 & 2 & 1 \\ 1 & 2 & 3 \end{vmatrix} = -4, \quad \det(A_1) = \begin{vmatrix} 5 & 2 & 1 \\ 6 & 2 & 1 \\ 9 & 2 & 3 \end{vmatrix} = -4$$

$$\det(A_2) = \begin{vmatrix} 1 & 5 & 1 \\ 2 & 6 & 1 \\ 1 & 9 & 3 \end{vmatrix} = -4 \quad \det(A_3) = \begin{vmatrix} 1 & 2 & 5 \\ 2 & 2 & 6 \\ 1 & 2 & 9 \end{vmatrix} = -8$$

Thus

$$x_1 = \frac{\det(A_1)}{\det(A)} = 1, \quad x_2 = \frac{\det(A_2)}{\det(A)} = 1, \quad x_3 = \frac{\det(A_3)}{\det(A)} = 2$$

## Appendix: proof for Theorem 16.7

**Theorem 16.7 (Determinants of Matrices Product)** If  $A$  and  $B$  are square matrices, then

$$\det(AB) = \det(A) \det(B)$$

**Proof.** (1) If  $A$  is singular, one has  $\det(A) = 0$  by theorem 16.6, and  $AB$  is also singular by using the Theorem 8.11. Thus,  $\det(AB) = 0$  by Theorem 16.6.

(2) If  $A$  is nonsingular, then  $A$  can be written as a product of elementary matrices by using the Theorem 8.7,  $A = E_1 E_2 \cdots E_k$ , where  $E_1, \dots, E_k$  are elementary matrices. By property 16.15, one has

$$\begin{aligned} \det(AB) &= \det(E_1 E_2 \cdots E_k B) \\ &= \det(E_1) \det(E_2 \cdots E_k B) \\ &= \det(E_1) \det(E_2) \cdots \det(E_k) \det(B) \\ &= \det(E_1 E_2 \cdots E_k) \det(B) \\ &= \det(A) \det(B) \end{aligned}$$

## Appendix: Alternative Definition for Determinants

**Definition 16.18 (Regard Permutation Matrix as reordering columns of identity matrix)** A permutation matrix is a matrix formed from the identity matrix by reordering its columns or rows. Suppose that  $P$  is the permutation matrix formed by reordering the columns of  $I$  in the order  $(k_1, k_2, \dots, k_n)$ , then  $P = [\mathbf{e}_{k_1}, \mathbf{e}_{k_2}, \dots, \mathbf{e}_{k_n}]$ . Obviously,  $\det(P) = 1$  or  $\det(P) = -1$ . Denote:

$$\tau_C(k_1, \dots, k_n) = \det([\mathbf{e}_{k_1}, \mathbf{e}_{k_2}, \dots, \mathbf{e}_{k_n}]) = \det(P)$$

Not the permutation  $(k_1, k_2, \dots, k_n)$  is odd if  $\tau_C(k_1, \dots, k_n) = -1$ , and the permutation  $(k_1, k_2, \dots, k_n)$  is even if  $\tau_C(k_1, \dots, k_n) = 1$ .

## Appendix: Alternative Definition for Determinants

**Definition 16.19 (Regard Permutation Matrix as reordering rows of identity matrix)** Suppose that  $Q$  is the permutation matrix formed by

reordering the rows of  $I$  in the order  $(k_1, k_2, \dots, k_n)^T$ , then  $Q = \begin{bmatrix} \vec{e}_{k_1} \\ \vec{e}_{k_2} \\ \vdots \\ \vec{e}_{k_n} \end{bmatrix}$ .

Obviously,  $\det(Q) = 1$  or  $\det(Q) = -1$ . Denote:

$$\tau_R(k_1, \dots, k_n) = \det \left( \begin{bmatrix} \vec{e}_{k_1} \\ \vec{e}_{k_2} \\ \vdots \\ \vec{e}_{k_n} \end{bmatrix} \right) = \det(Q)$$

## Appendix: a closed formula for Determinants

**Definition 16.20 (Alternative Definition for Determinants)** Let

$A = (a_{ij})_{n \times n}$  be an  $n \times n$  matrix, then the determinant is defined as

$$\det(A) = \sum_{\text{permutations } (k_1, k_2, \dots, k_n) \text{ of } (1, 2, \dots, n)} \tau_C(k_1, \dots, k_n) a_{1k_1} a_{2k_2} \cdots a_{nk_n}$$

or

$$\det(A) = \sum_{\text{permutations } (k_1, k_2, \dots, k_n) \text{ of } (1, 2, \dots, n)} \tau_R(k_1, \dots, k_n) a_{k_1 1} a_{k_2 2} \cdots a_{k_n n}$$

which are called the **Leibnitz formulas**. The above two equalities are equivalent. These two formulas are big formulas since the permutation of  $(1, 2, \dots, n)$  has  $n!$  terms, which will result a lot of calculations. But in the case when the matrix has lots of zeros, the calculation may not that tedious.

**Remark.** Every term in the summation of the Leibnitz formula is a multiplication of  $n$  entries of  $A$  (with size  $n \times n$ ), where all the entries in the same term must be taken from different rows and different columns.



### Example 16.21 Calculate

$$\det(A) = \begin{vmatrix} 0 & 0 & 0 & a_{14} \\ 0 & 0 & a_{23} & a_{24} \\ 0 & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{vmatrix}$$

Observation: only one nonzero term in first row, which is  $a_{14}$  in the fourth column; besides the four column, only one nonzero term  $a_{23}$  in second row (in the third column); besides the third and four columns, only one nonzero term  $a_{32}$  in third row (in the second column); besides the second, third and four columns, only one nonzero term  $a_{41}$  in fourth row (in the first column). Thus, only one nonzero term in the summation when using the Leibnitz formula for this example.

$$\det(A) = \begin{vmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{vmatrix} a_{14}a_{23}a_{32}a_{41} = a_{14}a_{23}a_{32}a_{41}$$