

# Notes on Determinants and Matrix Inverse

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# 1 Definition of determinant

Determinant is a scalar that measures the “magnitude” or “size” of a square matrix.

Notice that conclusions presented below are focused on rows and row expansions. Similar results apply when the row-focus is changed to column-focus.

**Definition (verbal):** For each  $n \times n$  (square) matrix  $A$ ,  $\det A$  is the sum of  $n!$  terms, each is the product of  $n$  entries in which *one and only one* is taken from each row and column. The sum exhausts all permutations of integers  $\{1, 2, \dots, n\}$ . Terms with even permutations carry a “+” sign and those with odd permutations a “-” sign.

**Permutation:** any rearrangement of an ordered list of numbers, say  $S$ , is a permutation of  $S$ .

**Example 1.1:** For  $S = \{1, 2, 3\}$ , there are a total of  $3! = 6$  permutations:

$$\{123, 231, 312, 132, 213, 321\}.$$

**Definition:** For each permutation  $(j_1 j_2 \dots j_n)$  of the ordered list  $S = \{1, 2, \dots, n\}$ ,

$$\sigma(j_1 j_2 \dots j_n) = \text{min number of pairwise exchanges to recover order } (12 \dots n).$$

**Even/odd permutations:** a permutation  $(j_1 j_2 \dots j_n)$  is even/odd if the value of  $\sigma(j_1 j_2 \dots j_n)$  is even/odd.

**Example 1.2:** For  $S = \{1, 2, 3\}$ :

$$\begin{aligned} \sigma(123) &= 0, & \sigma(231) &= 2, & \sigma(312) &= 2 & \text{ are even;} \\ \sigma(132) &= 1, & \sigma(213) &= 1, & \sigma(321) &= 1 & \text{ are odd.} \end{aligned}$$

**Example 1.3.1:** Show that the determinant of diagonal and triangular matrices is equal to the product of diagonal entries.

**Answer:** In each one of these matrices, there is only one way of picking one and only one entry from each row and each column that is not a zero.

$$\begin{vmatrix} a_{11} & 0 & \cdots & 0 \\ 0 & a_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{nn} \end{vmatrix} = \begin{vmatrix} a_{11} & 0 & \cdots & 0 \\ * & a_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ * & * & \cdots & a_{nn} \end{vmatrix} = \begin{vmatrix} a_{11} & * & \cdots & * \\ 0 & a_{22} & \cdots & * \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{nn} \end{vmatrix} = a_{11}a_{22}\cdots a_{nn}.$$

**Example 1.3.2:** Show that if  $A$  has a zero column (or row),  $\det A = 0$ .

**Answer:** Since in every term of the determinant, there is one number from each row and each column. Thus, there is at least a 0 in each product which makes every term equal to 0!

$$\begin{vmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj} & \cdots & a_{nn} \end{vmatrix} = \begin{vmatrix} a_{11} & \cdots & 0 & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \cdots & 0 & \cdots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & 0 & \cdots & a_{nn} \end{vmatrix} = 0.$$

**Definition (formula):**  $\forall n \times n$  matrix  $A = [a_{ij}]$  ( $1 \leq i, j \leq n$ )

$$\det A = \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} = \sum_{(j_1 j_2 \cdots j_n)} (-1)^{\sigma(j_1 j_2 \cdots j_n)} a_{1j_1} a_{2j_2} \cdots a_{nj_n},$$

where the sum is taken over all possible permutations of integers  $(12 \cdots n)$  (a total of  $n!$ ) and that

$$(-1)^{\sigma(j_1 j_2 \cdots j_n)} = \begin{cases} +1, & \text{if } \sigma(j_1 j_2 \cdots j_n) \text{ is even,} \\ -1, & \text{if } \sigma(j_1 j_2 \cdots j_n) \text{ is odd.} \end{cases}$$

A column focused formula reads

$$\det A = \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} = \sum_{(i_1 i_2 \cdots i_n)} (-1)^{\sigma(i_1 i_2 \cdots i_n)} a_{i_1 1} a_{i_2 2} \cdots a_{i_n n},$$

**Example 1.4:** For a  $3 \times 3$  matrix  $A = [a_{ij}]$  ( $1 \leq i, j \leq 3$ ),

$$\det A = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \sum_{(j_1 j_2 j_3)} (-1)^{\sigma(j_1 j_2 j_3)} a_{1j_1} a_{2j_2} a_{3j_3}$$

$$= (-1)^{\sigma(123)=0} a_{11} a_{22} a_{33} + (-1)^{\sigma(231)=2} a_{12} a_{23} a_{31} + (-1)^{\sigma(312)=2} a_{13} a_{21} a_{32}$$

$$(-1)^{\sigma(132)=1} a_{11} a_{23} a_{32} + (-1)^{\sigma(213)=1} a_{12} a_{21} a_{33} + (-1)^{\sigma(321)=1} a_{13} a_{22} a_{31}$$

$$= a_{11} a_{22} a_{33} + a_{12} a_{23} a_{31} + a_{13} a_{21} a_{32} - a_{11} a_{23} a_{32} - a_{12} a_{21} a_{33} - a_{13} a_{22} a_{31}.$$

**Remark:** For matrices larger than  $3 \times 3$ , this formula is not practical to use by hand. So, it shall be mainly employed in demonstrating important properties of determinants.

## 2 Row and column expansion formulas

As an engineer, you may not need to know how to prove each formula that we learn. In this section, I will focus on the “how-to” without saying “why”.

For all practical purposes, an engineer should know how to calculate the determinant of a large matrix using either row or column expansions. We have already learned how to calculate a  $3 \times 3$  matrix using expansion in the first row of a matrix.

As a matter of fact, the first row is not always the best to use in an expansion formula. So, we here extend that formulas to expansions in any row of a matrix.

$$\begin{vmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj} & \cdots & a_{nn} \end{vmatrix} = a_{i1}(-1)^{i+1} \begin{vmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj} & \cdots & a_{nn} \end{vmatrix} + \cdots \\
 + a_{ij}(-1)^{i+j} \begin{vmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj} & \cdots & a_{nn} \end{vmatrix} + \cdots + a_{in}(-1)^{i+n} \begin{vmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj} & \cdots & a_{nn} \end{vmatrix} \\
 = a_{i1}c_{i1} + \cdots + a_{ij}c_{ij} + \cdots + a_{in}c_{in} \quad \leftarrow \quad \text{row expansion formula,}$$

where

- $A_{ij}$  is the  $(n-1) \times (n-1)$  matrix obtained by deleting the  $i^{th}$  row and  $j^{th}$  column of  $A$ :

$$A_{ij} = \begin{bmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj} & \cdots & a_{nn} \end{bmatrix}, \quad \det A_{ij} = \begin{vmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj} & \cdots & a_{nn} \end{vmatrix}.$$

- $\det A_{ij}$  is called the  $(i, j)^{th}$  *minor* of  $A$ .
- $c_{ij} = (-1)^{i+j} \det A_{ij}$  is called the  $(i, j)^{th}$  *cofactor* of  $A$ .

Similarly, we can expand in the  $j^{th}$  column of a matrix.

$$\begin{vmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nj} & \cdots & a_{nn} \end{vmatrix} = a_{1j}(-1)^{1+j} \begin{vmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{in} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{vmatrix} + \cdots \\
 + a_{ij}(-1)^{i+j} \begin{vmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{in} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{vmatrix} + \cdots + a_{nj}(-1)^{n+j} \begin{vmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{i1} & \cdots & a_{in} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{vmatrix}$$

$$= a_{1j}c_{1j} + \cdots + a_{ij}c_{ij} + \cdots + a_{nj}c_{nj}. \quad \leftarrow \quad \text{column expansion formula.}$$

**Example 2.1:**

$$\det A = \begin{vmatrix} 2 & 0 & 2 & 4 \\ 0 & 0 & 3 & 2 \\ 2 & 2 & 4 & 4 \\ 3 & 0 & 6 & 2 \end{vmatrix} = ?$$

**Ans:** Using expansion in the first row, we obtain

$$\begin{aligned} \det A &= (-1)^{1+1}2 \begin{vmatrix} 0 & 3 & 2 \\ 2 & 2 & 4 \\ 3 & 0 & 6 \end{vmatrix} + (-1)^{1+3}2 \begin{vmatrix} 2 & 0 & 4 \\ 0 & 0 & 2 \\ 2 & 2 & 6 \end{vmatrix} + (-1)^{1+4}4 \begin{vmatrix} 2 & 0 & 2 \\ 0 & 0 & 3 \\ 2 & 2 & 4 \end{vmatrix} \\
 &= 2 \begin{vmatrix} 0 & 3 & 2 \\ 2 & 2 & 4 \\ 3 & 0 & 6 \end{vmatrix} + 2 \begin{vmatrix} 2 & 0 & 4 \\ 0 & 0 & 2 \\ 2 & 2 & 6 \end{vmatrix} - 4 \begin{vmatrix} 2 & 0 & 2 \\ 0 & 0 & 3 \\ 2 & 2 & 4 \end{vmatrix}. \end{aligned}$$

Using expansion in the second row, we obtain

$$\det A = (-1)^{2+3}3 \begin{vmatrix} 2 & 0 & 2 & 4 \\ 0 & 0 & 3 & 2 \\ 2 & 2 & 4 & 4 \\ 3 & 0 & 6 & 2 \end{vmatrix} + (-1)^{2+4}2 \begin{vmatrix} 2 & 0 & 2 & 4 \\ 0 & 0 & 3 & 2 \\ 2 & 2 & 4 & 4 \\ 3 & 0 & 6 & 2 \end{vmatrix} = -3 \begin{vmatrix} 2 & 0 & 4 \\ 2 & 4 & 4 \\ 3 & 0 & 2 \end{vmatrix} + 2 \begin{vmatrix} 2 & 0 & 2 \\ 2 & 2 & 4 \\ 3 & 0 & 6 \end{vmatrix}.$$

But the simplest is to expand in the second column,

$$\det A = (-1)^{3+2}2 \begin{vmatrix} 2 & 0 & 2 & 4 \\ 0 & 0 & 3 & 2 \\ 2 & 2 & 4 & 4 \\ 3 & 0 & 6 & 2 \end{vmatrix} = (-1)^{5}2 \begin{vmatrix} 2 & 2 & 4 \\ 0 & 3 & 2 \\ 3 & 6 & 2 \end{vmatrix} = (-2)(12 + 12 - 36 - 24) = 72.$$

### Remarks:

- (i) Row (column) expansion is the sum of the products between all entries of that row (column) multiplied by the corresponding cofactors.
- (ii) The  $(i, j)^{th}$  cofactor  $c_{ij} = (-1)^{i+j} \det A_{ij}$ , where  $A_{ij}$  is the  $(n-1) \times (n-1)$  matrix obtained by deleting the  $i^{th}$  row and the  $j^{th}$  column of matrix  $A$ .

Derivation/proof of the expansion formula (Not required!):

$$\det A = \sum_{j=1}^n a_{i\bar{j}} (-1)^{i+\bar{j}} \det A_{i\bar{j}} = \sum_{j=1}^n a_{ij} c_{ij}.$$

**Proof:** Based on the definition:

$$\begin{aligned} \det A &= \sum_{(l_1 \dots l_i \dots l_n)} (-1)^{\sigma(l_1 \dots l_i \dots l_n)} [a_{1l_1} \dots a_{il_i} \dots a_{nl_n}] \\ &= \sum_{(l_1 \dots l_i \dots l_n)} (-1)^{\sigma(l_i l_1 \dots l_{(i-1)} l_{(i+1)} \dots l_n) - (i-1)} \left[ a_{il_i} a_{1l_1} \dots a_{(i-1)l_{(i-1)}} a_{(i+1)l_{(i+1)}} \dots a_{nl_n} \right] \\ &= \sum_{(l_1 \dots l_i \dots l_n)} (-1)^{\sigma(j l_1 \dots l_{(i-1)} l_{(i+1)} \dots l_n) - (i-1) - (j-1)} \left[ a_{ij} a_{1l_1} \dots a_{(i-1)l_{(i-1)}} a_{(i+1)l_{(i+1)}} \dots a_{nl_n} \right] \\ &= \sum_{j=1}^n a_{ij} (-1)^{i+j} \left( \sum_{(l_1 \dots l_{(i-1)} l_{(i+1)} \dots l_n)}^{l_k \neq j \text{ for all } k} (-1)^{\sigma(l_1 \dots l_{(i-1)} l_{(i+1)} \dots l_n)} \left[ a_{1l_1} \dots a_{(i-1)l_{(i-1)}} a_{(i+1)l_{(i+1)}} \dots a_{nl_n} \right] \right) \\ &= \sum_{j=1}^n a_{ij} (-1)^{i+j} \det A_{ij}. \end{aligned}$$

In the derivation above, we used the following identities:

$$(i) \quad \sigma(l_1 \dots l_i \dots l_n) = \sigma(l_i l_1 \dots l_{(i-1)} l_{(i+1)} \dots l_n) - (i-1),$$

because it takes  $(i-1)$  pairwise exchanges to turn  $(l_i l_1 \dots l_{(i-1)} l_{(i+1)} \dots l_n)$  into  $(l_1 \dots l_i \dots l_n)$ .

$$(ii) \quad \sigma(l_i l_1 \dots l_{(i-1)} l_{(i+1)} \dots l_n) = \sigma(j l_1 \dots l_{(i-1)} l_{(i+1)} \dots l_n) - (j-1),$$

because it takes  $(j-1)$  pairwise exchanges to turn  $(1 \dots j \dots n)$  into  $(j 1 \dots (j-1)(j+1) \dots l_n)$ .

$$(iii) \quad (-1)^{-(i+j)-2} = (-1)^{-(i+j)} = (-1)^{i+j}.$$

### 3 Important properties of determinants

#### 3.1 $\det \mathbf{A} = \det \mathbf{A}^T$

Because the “row-focused” and “column-focused” formulas are identical.

$$\begin{aligned}\det A &= \sum_{(j_1 j_2 \cdots j_n)} (-1)^{\sigma(j_1 j_2 \cdots j_n)} a_{1j_1} a_{2j_2} \cdots a_{nj_n} \\ &= \sum_{(i_1 i_2 \cdots i_n)} (-1)^{\sigma(i_1 i_2 \cdots i_n)} a_{i_1 1} a_{i_2 2} \cdots a_{i_n n} = \det A^T.\end{aligned}$$

**Example 3.1:**

$$\begin{vmatrix} 1 & 2 \\ 3 & 4 \end{vmatrix} = 4 - 6 = \begin{vmatrix} 1 & 3 \\ 2 & 4 \end{vmatrix}.$$

### 3.2 $\det \mathbf{A}_{k \leftrightarrow l} = -\det \mathbf{A}$

**Proof:**

Suppose that matrix  $A' = A_{k \leftrightarrow l}$  is obtained by swapping the  $k^{th}$  and  $j^{th}$  rows of  $A$ . Thus,  $a'_{ij} = a_{ij}$  except that  $a'_{kj} = a_{lj}$  and  $a'_{lj} = a_{kj}$  for all  $1 \leq j \leq n$ .

Then,

$$\begin{aligned}
 \det A' &= \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots \textcolor{red}{j}_k \dots \textcolor{blue}{j}_l \dots j_n)} a'_{1j_1} a'_{2j_2} \dots a'_{kj_k} \dots a'_{lj_l} \dots a'_{nj_n} \\
 &= \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots \textcolor{red}{j}_k \dots \textcolor{blue}{j}_l \dots j_n)} a_{1j_1} a_{2j_2} \dots a_{lj_k} \dots a_{kj_l} \dots a_{nj_n} \\
 &= \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots \textcolor{red}{j}_k \dots \textcolor{blue}{j}_l \dots j_n)} a_{1j_1} a_{2j_2} \dots a_{kj_l} \dots a_{lj_k} \dots a_{nj_n} \\
 &= \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots \textcolor{blue}{j}_l \dots \textcolor{red}{j}_k \dots j_n) + 1} a_{1j_1} a_{2j_2} \dots a_{kj_l} \dots a_{lj_k} \dots a_{nj_n} = -\det A.
 \end{aligned}$$

Note that  $\sigma(j_1 j_2 \dots \textcolor{red}{j}_k \dots \textcolor{blue}{j}_l \dots j_n) = \sigma(j_1 j_2 \dots \textcolor{blue}{j}_l \dots \textcolor{red}{j}_k \dots j_n) + 1$  because the two differ by one pairwise swap.

**Example 3.2:**

$$\begin{vmatrix} 1 & 2 \\ 3 & 4 \end{vmatrix} = 4 - 6 = -2, \quad \text{but} \quad \begin{vmatrix} 3 & 4 \\ 1 & 2 \end{vmatrix} = \begin{vmatrix} 2 & 1 \\ 4 & 3 \end{vmatrix} = 6 - 4 = 2.$$

### 3.3 $\det \mathbf{A}_{k=1} = 0$

If two rows (columns) of a matrix  $A$  are identical (i.e.  $A = A_{k \leftrightarrow l}$ ), then

$$\det A = \det A_{k \leftrightarrow l} = -\det A = 0.$$

Alternatively, if two rows are identical or are constant multiples of each other, then the rows in  $A$  are not linearly independent. Thus,  $A$  is not invertible which implies  $\det A = 0$ .

**Example 3.3:**

$$\begin{vmatrix} 1 & 2 \\ 1 & 2 \end{vmatrix} = 2 - 2 = 0.$$

### 3.4 Determinant is linear in each row/column

Suppose a row vector is a linear combination of two row vectors, say the  $k^{th}$  row  $\mathbf{a}_k = s\mathbf{b} + t\mathbf{c}$  ( $k = 1, 2, \dots, n$ ), where  $s, t$  are scalars. Then,

$$\det \begin{bmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \vdots \\ s\mathbf{b} + t\mathbf{c} \\ \vdots \\ \mathbf{a}_n \end{bmatrix} = s \det \begin{bmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \vdots \\ \mathbf{b} \\ \vdots \\ \mathbf{a}_n \end{bmatrix} + t \det \begin{bmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \vdots \\ \mathbf{c} \\ \vdots \\ \mathbf{a}_n \end{bmatrix}.$$

Equivalently, we can also express this property in the following form

$$\begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ sb_1 + tc_1 & sb_2 + tc_2 & \cdots & sb_n + tc_n \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} = s \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ b_1 & b_2 & \cdots & b_n \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} + t \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ c_1 & c_2 & \cdots & c_n \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix}.$$

**Proof:**

$$\begin{aligned} \det A &= \sum_{(j_1 j_2 \cdots j_n)} (-1)^{\sigma(j_1 j_2 \cdots j_n)} a_{1j_1} a_{2j_2} \cdots a_{kj_k} \cdots a_{nj_n} \\ &= \sum_{(j_1 j_2 \cdots j_n)} (-1)^{\sigma(j_1 j_2 \cdots j_n)} a_{1j_1} a_{2j_2} \cdots (sb_{j_k} + tc_{j_k}) \cdots a_{nj_n} \\ &= s \sum_{(j_1 j_2 \cdots j_n)} (-1)^{\sigma(j_1 j_2 \cdots j_n)} a_{1j_1} a_{2j_2} \cdots b_{j_k} \cdots a_{nj_n} + t \sum_{(j_1 j_2 \cdots j_n)} (-1)^{\sigma(j_1 j_2 \cdots j_n)} a_{1j_1} a_{2j_2} \cdots c_{j_k} \cdots a_{nj_n}. \end{aligned}$$

This property can also be expressed in form of two related properties:

$$(1) \quad \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ sa_{k1} & sa_{k2} & \cdots & sa_{kn} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} = s \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{k1} & a_{k2} & \cdots & a_{kn} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix},$$

$$(2) \quad \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ b_1 + c_1 & b_2 + c_2 & \cdots & b_n + c_n \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} = \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ b_1 & b_2 & \cdots & b_n \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} + \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ c_1 & c_2 & \cdots & c_n \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix}.$$

**Example 3.4:**

$$\begin{vmatrix} a & b \\ 3 & 5 \end{vmatrix} = 5a - 3b, \quad \begin{vmatrix} 2a & 2b \\ 3 & 5 \end{vmatrix} = 10a - 6b = 2(5a - 3b) = 2 \begin{vmatrix} a & b \\ 3 & 5 \end{vmatrix}.$$

$$\begin{vmatrix} a + c & b + d \\ 3 & 5 \end{vmatrix} = \begin{vmatrix} a & b \\ 3 & 5 \end{vmatrix} + \begin{vmatrix} c & d \\ 3 & 5 \end{vmatrix}.$$

### 3.5 Adding a scalar multiple of one row to another

Let matrix  $A'$  be identical to  $A$  except that  $k^{th}$  row multiplied by scalar  $s$  ( $\neq 0$ ) is added to  $l^{th}$  row ( $1 \leq k < l \leq n$ ).

Thus,  $a'_{ij} = a_{ij}$  for all  $1 \leq i, j \leq n$  except that  $a'_{lj} = a_{lj} + sa_{kj}$ . Then,

$$\det A' = \det A.$$

**Proof:**

$$\begin{aligned} \det A' &= \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} a'_{1j_1} a'_{2j_2} \dots a'_{lj_k} \dots a'_{nj_n} \\ &= \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} a_{1j_1} \dots a_{kj_k} \dots (a_{lj_l} + sa_{kj_l}) \dots a_{nj_n} \\ &= \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} a_{1j_1} \dots a_{kj_k} \dots a_{lj_l} \dots a_{nj_n} + s \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} a_{1j_1} \dots a_{kj_k} \dots a_{kj_l} \dots a_{nj_n} \\ &= \det A + 0 = \det A. \end{aligned}$$

**Example 3.5:**

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc.$$

$$\begin{vmatrix} a & b \\ c+2a & d+2b \end{vmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix} + 2 \begin{vmatrix} a & b \\ a & b \end{vmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix}.$$

### 3.6 $\det \mathbf{AB} = \det \mathbf{A} \det \mathbf{B}$ .

**Example 3.6:**

$$\det A = \begin{vmatrix} 2 & 1 \\ 3 & 4 \end{vmatrix} = 8 - 3 = 5, \quad \det B = \begin{vmatrix} 1 & 1 \\ -4 & 3 \end{vmatrix} = 3 - (-4) = 7.$$

$$AB = \begin{bmatrix} -2 & 5 \\ -13 & 15 \end{bmatrix}, \quad \det AB = \begin{vmatrix} -2 & 5 \\ -13 & 15 \end{vmatrix} = -30 - (-65) = 35 = \det A \det B.$$

**Proof(1):** (Not required!)

**Case I:** If at least one of  $A$  and  $B$  is noninvertible, i.e. either  $\det A = 0$  or  $\det B = 0$  or both, then  $AB$  as the composite transformation of the two must also be noninvertible. Thus,

$$\det(AB) = 0 = \det A \det B.$$

**Case II:** If both  $A$  and  $B$  are invertible, then based on the results on elementary matrices:

$$A = A_1 A_2 \cdots A_k, \quad B = B_1 B_2 \cdots B_l,$$

where  $A_1, \dots, A_k$  and  $B_1, \dots, B_l$  are all elementary matrices. Therefore,

$$\det(AB) = \det(A_1 A_2 \cdots A_k B_1 B_2 \cdots B_l) = (\det A_1 \cdots \det A_k)(\det B_1 \cdots \det B_l) = \det A \det B.$$

**Proof(2):** (Not required!)

$$\begin{aligned}
& \text{Note that } AB = [(AB)_{ij}], \text{ where } (AB)_{ij} = \sum_{l=1}^n a_{il}b_{lj}. \\
& \det AB = \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} (AB)_{1j_1} (AB)_{2j_2} \dots (AB)_{nj_n} \\
& = \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} \left( \sum_{l_1=1}^n a_{1l_1} b_{l_1 j_1} \right) \left( \sum_{l_2=1}^n a_{2l_2} b_{l_2 j_2} \right) \dots \left( \sum_{l_n=1}^n a_{nl_n} b_{l_n j_n} \right) \\
& \quad \xrightarrow{\text{All non-selfavoiding terms cancel (i.e. } l_1, l_2, \dots, l_n \text{ must be distinct)}} \\
& \text{because an identical term with opposite sign occurs when summed over all permutations of } (j_1 j_2 \dots j_n) \\
& = \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} \sum_{(l_1 l_2 \dots l_n)} (a_{1l_1} b_{l_1 j_1}) (a_{2l_2} b_{l_2 j_2}) \dots (a_{nl_n} b_{l_n j_n}) \\
& = \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} \sum_{(l_1 l_2 \dots l_n)} (a_{1l_1} a_{2l_2} \dots a_{nl_n}) (b_{l_1 j_1} b_{l_2 j_2} \dots b_{l_n j_n}) \\
& = \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} \sum_{(l_1 l_2 \dots l_n)} (a_{1l_1} a_{2l_2} \dots a_{nl_n}) (-1)^{\sigma(l_1 l_2 \dots l_n)} (b_{1j_1} b_{2j_2} \dots b_{nj_n}) \\
& = \sum_{(j_1 j_2 \dots j_n)} (-1)^{\sigma(j_1 j_2 \dots j_n)} (b_{1j_1} b_{2j_2} \dots b_{nj_n}) \sum_{(l_1 l_2 \dots l_n)} (-1)^{\sigma(l_1 l_2 \dots l_n)} (a_{1l_1} a_{2l_2} \dots a_{nl_n}) \\
& = \det B \det A.
\end{aligned}$$

The key step in the proof above is to eliminate all the non-selfavoiding terms. Here, a “non-selfavoiding term” refers to those terms in which  $l_i = l_j$  ( $i \neq j$ ) happens for at least one pair of  $i, j$  ( $1 \leq i, j \leq n$ ). To see this more clearly, we show the case when only two subscripts exists for  $2 \times 2$  matrices. In this case,

$$\begin{aligned}
& \det AB = \sum_{(j_1 j_2)}^{(12) \text{ or } (21)} (-1)^{\sigma(j_1 j_2)} (AB)_{1j_1} (AB)_{2j_2} \\
& = \sum_{(j_1 j_2)}^{(12) \text{ or } (21)} (-1)^{\sigma(j_1 j_2)} \left( \sum_{l_1=1}^2 a_{1l_1} b_{l_1 j_1} \right) \left( \sum_{l_2=1}^2 a_{2l_2} b_{l_2 j_2} \right) \\
& = \sum_{(j_1 j_2)}^{(12) \text{ or } (21)} (-1)^{\sigma(j_1 j_2)} (a_{11} b_{1j_1} + a_{12} b_{2j_1}) (a_{21} b_{1j_2} + a_{22} b_{2j_2}) \\
& = (a_{11} b_{11} + a_{12} b_{21}) (a_{21} b_{12} + a_{22} b_{22}) - (a_{11} b_{12} + a_{12} b_{22}) (a_{21} b_{11} + a_{22} b_{21})
\end{aligned}$$

Terms with identical color cancel each other.  
Self-avoiding colors!

$$\begin{aligned}
&= a_{11}b_{11}a_{22}b_{22} + a_{12}b_{21}a_{21}b_{12} - a_{11}b_{12}a_{22}b_{21} - a_{12}b_{22}a_{21}b_{11} \\
&= \sum_{l_1, l_2=1, (l_1 \neq l_2)}^2 (a_{1l_1}b_{l_11}a_{2l_2}b_{l_22} - a_{1l_1}b_{l_12}a_{2l_2}b_{l_21}) \\
&= \sum_{l_1, l_2=1, (l_1 \neq l_2)}^2 \sum_{(j_1 j_2)} (-1)^{\sigma(j_1 j_2)} a_{1l_1}b_{l_1 j_1}a_{2l_2}b_{l_2 j_2} = \sum_{(l_1 l_2)} \sum_{(j_1 j_2)} (-1)^{\sigma(j_1 j_2)} a_{1l_1}b_{l_1 j_1}a_{2l_2}b_{l_2 j_2} \\
&= \sum_{(l_1 l_2)} \sum_{(j_1 j_2)} (-1)^{\sigma(j_1 j_2)} (a_{1l_1}a_{2l_2})(b_{l_1 j_1}b_{l_2 j_2}) = \sum_{(l_1 l_2)} \sum_{(j_1 j_2)} (-1)^{\sigma(j_1 j_2)} (a_{1l_1}a_{2l_2})(-1)^{\sigma(l_1 l_2)} (b_{1j_1}b_{2j_2}) \\
&= \sum_{(l_1 l_2)} (-1)^{\sigma(l_1 l_2)} (a_{1l_1}a_{2l_2}) \sum_{(j_1 j_2)} (-1)^{\sigma(j_1 j_2)} (b_{1j_1}b_{2j_2}) \\
&= \det A \det B.
\end{aligned}$$

Here, non-selfavoiding terms are terms like  $a_{11}b_{11}a_{21}b_{12}$  and  $a_{12}b_{21}a_{22}b_{22}$  in which entries from the same column or row of a matrix occur twice in the same product. They all cancel out in the summation. Therefore, “self-avoiding” means in each term of the expression, there must be *one and only one* entries from each row and each column of each matrix. The same conclusion works in a similar way when the matrix is bigger. For any term that contains at least one pair of non-selfavoiding entries, we can use the same argument for this particular pair of entries in exactly the same way as we treat the  $2 \times 2$  case.

## 4 Formula for the inverse of $n \times n$ matrices

### Definition of minor and cofactor:

$\forall n \times n$  matrices  $A = [a_{ij}]$ , the  $(i, j)^{th}$ -minor and  $(i, j)^{th}$ -cofactor are

$$m_{ij} = \det A_{ij}, \quad c_{ij} = (-1)^{i+j} m_{ij},$$

where  $A_{ij}$  is the  $(n-1) \times (n-1)$  matrix obtained by deleting the  $i^{th}$  row and  $j^{th}$  column of  $A$ .

### Definition of cofactor matrix:

$\forall n \times n$  matrices  $A = [a_{ij}]$ , its cofactor matrix is defined as

$$C = [c_{ij}] = \begin{bmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{bmatrix},$$

where  $c_{ij} = (-1)^{i+j} \det A_{ij}$  is the  $(i, j)^{th}$ -cofactor of matrix  $A$ .

### Formula for $A^{-1}$ :

$\forall$  invertible  $n \times n$  matrices  $A$  with a cofactor matrix  $C$ :

$$A^{-1} = \left( \frac{1}{\det A} \right) C^T.$$

**Example 4.1:** Consider the matrix  $A = \begin{bmatrix} 0 & 2 & 1 \\ -1 & 1 & 0 \\ 2 & -5 & 2 \end{bmatrix}$ .

- (a) Show that it is invertible using the row expansion formula;
- (b) Find  $A^{-1}$  using the cofactor formula.

**Ans:**

- (a) First calculate the cofactors for the expansion in row 1:

$$c_{11} = + \begin{vmatrix} 1 & 0 \\ -5 & 2 \end{vmatrix} = 2; \quad c_{12} = - \begin{vmatrix} -1 & 0 \\ 2 & 2 \end{vmatrix} = 2; \quad c_{13} = + \begin{vmatrix} -1 & 1 \\ 2 & -5 \end{vmatrix} = 3.$$

$$\implies \det A = (0)c_{11} + (2)c_{12} + (1)c_{13} = 4 + 3 = 7 \neq 0!$$

- (b) Then, calculate all the other cofactors:

$$c_{21} = - \begin{vmatrix} 2 & 1 \\ -5 & 2 \end{vmatrix} = -9; \quad c_{22} = + \begin{vmatrix} 0 & 1 \\ 2 & 2 \end{vmatrix} = -2; \quad c_{23} = - \begin{vmatrix} 0 & 2 \\ 2 & -5 \end{vmatrix} = 4.$$

$$c_{31} = + \begin{vmatrix} 2 & 1 \\ 1 & 0 \end{vmatrix} = -1; \quad c_{32} = - \begin{vmatrix} 0 & 1 \\ -1 & 0 \end{vmatrix} = -1; \quad c_{33} = + \begin{vmatrix} 0 & 2 \\ -1 & 1 \end{vmatrix} = 2.$$

Therefore,

$$C = \begin{bmatrix} 2 & 2 & 3 \\ -9 & -2 & 4 \\ -1 & -1 & 2 \end{bmatrix} \Rightarrow A^{-1} = \left( \frac{1}{\det A} \right) C^T = \frac{1}{7} \begin{bmatrix} 2 & -9 & -1 \\ 2 & -2 & -1 \\ 3 & 4 & 2 \end{bmatrix}.$$

Verify,

$$A^{-1}A = \frac{1}{7} \begin{bmatrix} 2 & -9 & -1 \\ 2 & -2 & -1 \\ 3 & 4 & 2 \end{bmatrix} \begin{bmatrix} 0 & 2 & 1 \\ -1 & 1 & 0 \\ 2 & -5 & 2 \end{bmatrix} = \frac{1}{7} \begin{bmatrix} 7 & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & 7 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = I_3.$$

Derivation/proof of the cofactor formula for inverse: (Not required!)

$$A^{-1} = \left( \frac{1}{\det A} \right) C^T.$$

**Proof:** Based on the definition of  $\det A$ :

$$\det A = \sum_{j=1}^n a_{ij} c_{ij} \quad \implies \quad \sum_{j=1}^n a_{ij} \frac{c_{ij}}{\det A} = 1.$$

Let  $A' = [a'_{ij}]$  be a matrix obtained by replacing the  $i^{th}$  row of  $A$  by its  $l^{th}$  row. Then,

$$\det A' = \sum_{j=1}^n a'_{ij} c_{ij} = \sum_{j=1}^n a_{lj} c_{ij}.$$

If  $l = i$  (no change), then  $A' = A$

$$\det A' = \sum_{j=1}^n a'_{ij} c_{ij} = \sum_{j=1}^n a_{lj} c_{ij} = \sum_{j=1}^n a_{ij} c_{ij} = \det A \quad \implies \quad \sum_{j=1}^n a_{lj} \frac{c_{ij}}{\det A} = 1 \quad (\text{for } l = i).$$

If  $l \neq i$ , then  $A'$  has two identical rows  $i$  and  $l$ , thus

$$\det A' = 0 = \sum_{j=1}^n a'_{ij} c_{ij} = \sum_{j=1}^n a_{lj} c_{ij} \quad \implies \quad \sum_{j=1}^n a_{lj} \frac{c_{ij}}{\det A} = 0 \quad (\text{for } l \neq i).$$

Therefore,

$$\sum_{j=1}^n a_{lj} \frac{c_{ij}}{\det A} = \begin{cases} 1, & \text{if } l = i, \\ 0, & \text{if } l \neq i. \end{cases}$$

Introduce a new matrix  $B = [b_{ij}]$  such that

$$b_{ij} = \frac{c_{ji}}{\det A}.$$

Now, let  $D = AB = [d_{ij}]$ . Based on definition of matrix multiplication

$$d_{li} = \sum_{j=1}^n a_{lj} b_{ji} = \sum_{j=1}^n a_{lj} \frac{c_{ji}}{\det A} = \begin{cases} 1, & \text{if } l = i, \\ 0, & \text{if } l \neq i. \end{cases}$$

Therefore,  $D = I = AB \Rightarrow$

$$A^{-1} = B = [b_{ij}] = \left[ \frac{c_{ji}}{\det A} \right] = \left( \frac{1}{\det A} \right) C^T.$$

## 5 Cramer's rule (not in syllabus)

**Cramer's Rule** on component-wise solution of matrix equation  $A\mathbf{x} = \mathbf{b}$ :

Let  $A = [\mathbf{c}_1 \ \mathbf{c}_2 \ \cdots \ \mathbf{c}_n]$  where  $\mathbf{c}_j$  ( $j = 1, 2, \dots, n$ ) are its column vectors.

If  $\det A \neq 0$ , then

$$\mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = A^{-1}\mathbf{b} = \frac{1}{\det A} C^T \mathbf{b}.$$

Thus, component-wise we have

$$x_i = \frac{1}{\det A} \sum_{j=1}^n c_{ij}^T b_j = \frac{1}{\det A} \sum_{j=1}^n b_j c_{ji}, \quad (i = 1, 2, \dots, n),$$

where

$$\sum_{j=1}^n b_j c_{ji} \quad - \quad \text{is } \det A \text{ in form of expansion in } i^{\text{th}} \text{ column with } a_{ji} \text{ replaced by } b_j!$$

Thus,

$$\sum_{j=1}^n b_j c_{ji} = \det A_{(i=\mathbf{b})} = \det[\mathbf{c}_1 \ \cdots \ \mathbf{c}_{i-1} \ \mathbf{b} \ \mathbf{c}_{i+1} \ \cdots \ \mathbf{c}_n] = \begin{vmatrix} a_{11} & \cdots & a_{1(i-1)} & b_1 & a_{1(i+1)} & \cdots & a_{1n} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & \cdots & a_{n(i-1)} & b_n & a_{n(i+1)} & \cdots & a_{nn} \end{vmatrix},$$

where  $A_{(i=\mathbf{b})}$  is obtained by replacing the  $i^{\text{th}}$  column of  $A$  by the vector  $\mathbf{b}$ !

Therefore,

$$x_i = \frac{\det A_{(i=\mathbf{b})}}{\det A}, \quad (i = 1, 2, \dots, n).$$

**Example 5.1:** Solve  $A\mathbf{x} = \mathbf{b}$  for  $\mathbf{x}$  component-wise using Cramer's Rule, where

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 10 \\ 20 \end{bmatrix}.$$

**Ans:** Using Cramer's Rule

$$x_1 = \frac{\det A_{(1=\mathbf{b})}}{\det A} = \frac{\begin{vmatrix} 10 & 1 \\ 20 & 1 \end{vmatrix}}{\begin{vmatrix} 2 & 1 \\ 1 & 1 \end{vmatrix}} = \frac{10 - 20}{2 - 1} = -10.$$
$$x_2 = \frac{\det A_{(2=\mathbf{b})}}{\det A} = \frac{\begin{vmatrix} 2 & 10 \\ 1 & 20 \end{vmatrix}}{\begin{vmatrix} 2 & 1 \\ 1 & 1 \end{vmatrix}} = \frac{40 - 10}{2 - 1} = 30.$$

Therefore,

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -10 \\ 30 \end{bmatrix}.$$