### Plan

- Determinant
- Determinants of Elementary Matrices
- Laplace Expansion Theorem
- Cramer's Rule
- Subspaces Associated with Matrices
- Rank Nullity Theorem
- The Fundamental Theorem of Invertible Matrices

#### Definition

Let  $A = [a_{ij}]$  be an  $n \times n$  matrix.

• If A = [a], determinant of A is defined as det(A) = a.

#### Definition

Let  $A = [a_{ij}]$  be an  $n \times n$  matrix.

• If A = [a], determinant of A is defined as det(A) = a.

• If 
$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$
, then we define  $det(A) = ad - bc$ .

#### Definition

Let  $A = [a_{ij}]$  be an  $n \times n$  matrix.

- If A = [a], determinant of A is defined as det(A) = a.
- If  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ , then we define det(A) = ad bc.
- Let  $A_{ij}$  be the submatrix of A obtained by deleting the i-th row and the j-th column of A.

#### Definition

Let  $A = [a_{ii}]$  be an  $n \times n$  matrix.

- If A = [a], determinant of A is defined as det(A) = a.
- If  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ , then we define det(A) = ad bc.
- Let A<sub>ij</sub> be the submatrix of A obtained by deleting the *i*-th row and the *j*-th column of A.
- In general, det(A) is defined recursively as follows:

$$\det(A) = a_{11}\det(A_{11}) - a_{12}\det(A_{12}) + \ldots + (-1)^{1+n}a_{1n}\det(A_{1n}).$$

• Sometimes det(A) is also denoted by |A|.

- Sometimes det(A) is also denoted by |A|.
- We define  $det(A_{ij})$  to be the (i, j)-minor of A.

- Sometimes det(A) is also denoted by |A|.
- We define  $det(A_{ij})$  to be the (i, j)-minor of A.
- The number  $C_{ij} = (-1)^{i+j} \det(A_{ij})$  is called the (i,j)-cofactor of A.

- Sometimes det(A) is also denoted by |A|.
- We define  $det(A_{ij})$  to be the (i, j)-minor of A.
- The number  $C_{ij} = (-1)^{i+j} \det(A_{ij})$  is called the (i,j)-cofactor of A.
- Thus we can write

$$\det(A) = |A| = \sum_{j=1}^{n} (-1)^{1+j} a_{1j} \det(A_{1j}) = \sum_{j=1}^{n} a_{1j} C_{1j}.$$

 $\bullet \ \, \mathsf{Take} \; \mathsf{a} \; \mathsf{matrix} \; A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}.$ 

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j}.$$

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| = 0$$

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| =$$

● Take a matrix 
$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = (-1)^{i+j}$ .

Expand each of them.

• Take a matrix 
$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = (-1)^{i+j}$ .

• Expand each of them. Do you get 12 terms?

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| = 0$$

$$\sigma_{11} \sigma_{11} \sigma_{11} \sigma_{11} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{12} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{14}$$

• Expand each of them. Do you get 12 terms? Do you get a term  $(...)\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$ ?

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| =$$

$$\sigma_{11} \sigma_{11} \sigma_{11} \sigma_{11} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{12} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{14} \sigma_{15} \sigma_{15}$$

• Expand each of them. Do you get 12 terms? Do you get a term  $(...)\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$ ? Do you get this matrix twice?

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| =$$

$$\sigma_{11} \sigma_{11} \sigma_{11} \sigma_{11} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{12} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{14}$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\dots\right)\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| =$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\ldots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| =$$

$$\sigma_{11} \sigma_{11} \sigma_{11} \sigma_{11} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{14} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{14}$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\ldots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\frac{\sigma_{12}a_{12}}{\sigma_{12}a_{12}}$

• Take a matrix 
$$A = \begin{bmatrix} a_{11} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = 1$ 

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\ldots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\sigma_{12}a_{12}$

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| =$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\ldots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\frac{\sigma_{12}a_{12}}{\sigma_{12}a_{12}}$

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| =$$

$$\sigma_{11} \, \hat{\sigma}_{11} \, \hat{\sigma}_{11} \, \hat{\sigma}_{12} \, \quad \hat{\sigma}_{22} \, \quad \hat{\sigma}_{23} \, \quad \hat{\sigma}_{24} \, \\ \hat{\sigma}_{32} \, \quad \hat{\sigma}_{33} \, \quad \hat{\sigma}_{34} \, \quad \hat{\sigma}_{12} \, \hat{\sigma}_{12} \, \hat{\sigma}_{12} \, \quad \hat{\sigma}_{13} \, \quad \hat{\sigma}_{33} \, \quad \hat{\sigma}_{34} \, \\ \hat{\sigma}_{41} \, \quad \hat{\sigma}_{43} \, \quad \hat{\sigma}_{44} \, \quad \hat{\sigma}_{13} \, \hat{\sigma}_{13} \, \quad \hat{\sigma}_{31} \, \quad \hat{\sigma}_{22} \, \quad \hat{\sigma}_{24} \, \\ \hat{\sigma}_{31} \, \quad \hat{\sigma}_{32} \, \quad \hat{\sigma}_{34} \, \quad \hat{\sigma}_{14} \, \hat{\sigma}_{14} \, \hat{\sigma}_{14} \, \quad \hat{\sigma}_{42} \, \quad \hat{\sigma}_{43} \, \\ \hat{\sigma}_{41} \, \quad \hat{\sigma}_{42} \, \quad \hat{\sigma}_{44} \, \quad \hat{\sigma}_{14} \, \hat{\sigma}_{14} \, \quad \hat{\sigma}_{42} \, \quad \hat{\sigma}_{43} \, \\ \hat{\sigma}_{41} \, \quad \hat{\sigma}_{42} \, \quad \hat{\sigma}_{43} \, \quad \hat{\sigma}_{44} \, \quad \hat{\sigma}_{$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\ldots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\sigma_{12}a_{12} = \sigma_{12}a_{23}$

• Take a matrix 
$$A = \begin{bmatrix} a_{11} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = (-1)^{i+j}$ .

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\dots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\sigma_{12}a_{12} = \sigma_{12}a_{23}$

$$\bullet \text{ Take a matrix } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix} \text{. Put } \sigma_{ij} = (-1)^{i+j} \text{. Then } |A| =$$

$$\sigma_{11} \sigma_{11} \sigma_{11} \sigma_{11} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{12} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{14}$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\ldots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\frac{\sigma_{12}a_{12}}{\sigma_{12}a_{23}}$  +

• Take a matrix 
$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = (-1)^{i+j}$ .

$$\sigma_{11} = \frac{1}{11} =$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\dots\right)\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\frac{\sigma_{12}a_{12}}{\sigma_{12}a_{23}}$  +

• Take a matrix 
$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = (-1)^{i+j}$ .

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\dots\right)\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\sigma_{12}a_{12} = \sigma_{12}a_{23} + \sigma_{13}a_{13} = \sigma_{12}a_{22}$

• Take a matrix 
$$A = \begin{bmatrix} 2 & 3 & & & \\ a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = (-1)^{i+j}$ .

$$\sigma_{11}\sigma_{11}\sigma_{11}\sigma_{11}\sigma_{12}\sigma_{13}\sigma_{22}\sigma_{33}\sigma_{34}\sigma_{34}\sigma_{34}\sigma_{12}\sigma_{12}\sigma_{12}\sigma_{12}\sigma_{33}\sigma_{34}\sigma_{41}\sigma_{12}\sigma_{12}\sigma_{33}\sigma_{41}\sigma_{43}\sigma_{43}\sigma_{44}\sigma_{13}\sigma_{13}\sigma_{41}\sigma_{13}\sigma_{22}\sigma_{34}\sigma_{1$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\dots\right)\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$ ? Do you get this matrix twice?
- So, the coefficient of  $\begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix}$  is  $\frac{\sigma_{12}a_{23}}{a_{23}} + \frac{\sigma_{13}a_{13}}{a_{12}a_{22}} = (-1)^{1+2+2+3} \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix}$ .

• Take a matrix 
$$A = \begin{bmatrix} 2 & 3 & & & \\ a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = (-1)^{i+j}$ .

$$\sigma_{11} \stackrel{1}{\text{ol}}_{11} \stackrel{1}{\text{ol}}_{12} = \stackrel{1}{\text{ol}}_{23} = \stackrel{2}{\text{ol}}_{23} = \stackrel{2}{\text{ol}}_{24} + \sigma_{12} \stackrel{1}{\text{ol}}_{21} = \stackrel{2}{\text{ol}}_{21} = \stackrel{2}{\text{ol}}_{23} = \stackrel{2}{\text{ol}}_{24} + \sigma_{13} \stackrel{1}{\text{ol}}_{21} = \stackrel{2}{\text{ol}}_{22} = \stackrel{2}{\text{ol}}_{24} + \sigma_{14} \stackrel{1}{\text{ol}}_{14} = \stackrel{2}{\text{ol}}_{14} = \stackrel{2}$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\dots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- $\bullet \text{ So, the coefficient of } \begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix} \text{ is } \qquad \qquad \begin{matrix} \sigma_{12} a_{12} & \sigma_{12} a_{23} & + & \sigma_{13} a_{13} & \sigma_{12} a_{22} = (-1)^{1+2+2+3} \\ a_{22} & a_{23} \end{vmatrix}.$
- The coefficient of  $|A_{1,2|1,3}| = \begin{vmatrix} a_{32} & a_{34} \\ a_{42} & a_{44} \end{vmatrix}$  is  $(-1)^{1+2+1+3} \begin{vmatrix} a_{11} & a_{13} \\ a_{21} & a_{23} \end{vmatrix}$ .

● Take a matrix 
$$A = \begin{bmatrix} 1 \\ a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = 1$ 

$$\sigma_{11} \sigma_{11} \sigma_{11} \sigma_{11} \sigma_{12} \sigma_{13} \sigma_{13} \sigma_{14} \sigma_{14} \sigma_{12} \sigma_{12} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{13} \sigma_{14} \sigma_{14}$$

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\dots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- $\bullet \text{ So, the coefficient of } \begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix} \text{ is } \qquad \qquad \sigma_{12} a_{12} \qquad \sigma_{12} a_{23} \ + \ \sigma_{13} a_{13} \ \sigma_{12} a_{22} = (-1)^{1+2+2+3} \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix}.$
- The coefficient of  $|A_{1,2|1,3}| = \begin{vmatrix} a_{32} & a_{34} \\ a_{42} & a_{44} \end{vmatrix}$  is  $(-1)^{1+2+1+3} \begin{vmatrix} a_{11} & a_{13} \\ a_{21} & a_{23} \end{vmatrix}$ .
- In general, the coefficient of  $|A_{1,2|i,j}|$  in the double expansion of |A| is  $(-1)^{1+2+i+j}\begin{vmatrix} a_{1i} & a_{1j} \\ a_{2i} & a_{2j} \end{vmatrix}$ .

● Take a matrix 
$$A = \begin{bmatrix} 2 & 3 & \\ a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}$$
. Put  $\sigma_{ij} = (-1)^{i+j}$ . Then  $|A| = (-1)^{i+j}$ .

- Expand each of them. Do you get 12 terms? Do you get a term  $\left(\dots\right)\begin{vmatrix}a_{31}&a_{34}\\a_{41}&a_{44}\end{vmatrix}$ ? Do you get this matrix twice?
- $\bullet \text{ So, the coefficient of } \begin{vmatrix} a_{31} & a_{34} \\ a_{41} & a_{44} \end{vmatrix} \text{ is } \qquad \qquad \underbrace{\sigma_{12}a_{22}}_{\sigma_{12}a_{23}} \ + \ \underbrace{\sigma_{13}a_{13}}_{\sigma_{12}a_{22}} \ \underbrace{\sigma_{12}a_{23}}_{\sigma_{22}a_{23}} \end{vmatrix}.$
- The coefficient of  $|A_{1,2|1,3}| = \begin{vmatrix} a_{32} & a_{34} \\ a_{42} & a_{44} \end{vmatrix}$  is  $(-1)^{1+2+1+3} \begin{vmatrix} a_{11} & a_{13} \\ a_{21} & a_{23} \end{vmatrix}$ .
- ullet In general, the coefficient of  $|A_{1,2|i,j}|$  in the double expansion of |A| is  $(-1)^{1+2+i+j}\begin{vmatrix} a_{1i} & a_{1j} \\ a_{2i} & a_{2j} \end{vmatrix}$ .
- Hence  $|A| = \sum\limits_{i < j} (-1)^{1+2+i+j} \begin{vmatrix} a_{1i} & a_{1j} \\ a_{2i} & a_{2j} \end{vmatrix} |A_{1,2|i,j}|.$

• If B is obtained by interchanging the first two rows of A, then det(B) = -det(A).

- If B is obtained by interchanging the first two rows of A, then det(B) = -det(A).
- (By Induction) If B is obtained by interchanging any two consecutive rows of A, then det(B) = -det(A).

- If B is obtained by interchanging the first two rows of A, then det(B) = -det(A).
- (By Induction) If B is obtained by interchanging any two consecutive rows of A, then det(B) = -det(A).
- If B is obtained by interchanging any two rows of A, then det(B) = -det(A).

- If B is obtained by interchanging the first two rows of A, then det(B) = -det(A).
- (By Induction) If B is obtained by interchanging any two consecutive rows of A, then det(B) = -det(A).
- If B is obtained by interchanging any two rows of A, then det(B) = -det(A).
- If A has a zero row then det(A) = 0.

# Result (Properties of Determinants)

- If B is obtained by interchanging the first two rows of A, then det(B) = -det(A).
- (By Induction) If B is obtained by interchanging any two consecutive rows of A, then det(B) = -det(A).
- If B is obtained by interchanging any two rows of A, then det(B) = -det(A).
- If A has a zero row then det(A) = 0.
- If A has two identical rows then det(A) = 0.

# Result (**Properties of Determinants**)

- If B is obtained by interchanging the first two rows of A, then det(B) = -det(A).
- (By Induction) If B is obtained by interchanging any two consecutive rows of A, then det(B) = -det(A).
- If B is obtained by interchanging any two rows of A, then det(B) = -det(A).
- If A has a zero row then det(A) = 0.
- If A has two identical rows then det(A) = 0.
- If B is obtained by multiplying a row of A by k, then det(B) = kdet(A).

# Result (**Properties of Determinants**)

- If B is obtained by interchanging the first two rows of A, then det(B) = -det(A).
- (By Induction) If B is obtained by interchanging any two consecutive rows of A, then det(B) = -det(A).
- If B is obtained by interchanging any two rows of A, then det(B) = -det(A).
- If A has a zero row then det(A) = 0.
- If A has two identical rows then det(A) = 0.
- If B is obtained by multiplying a row of A by k, then det(B) = kdet(A).
- If B is obtained by adding a multiple of one row of A to another row, then det(B) = det(A).



Let *E* be an  $n \times n$  elementary matrix and *A* be any  $n \times n$  matrix. Then

 $0 \det(E) = -1, k \text{ or } 1.$ 

Let *E* be an  $n \times n$  elementary matrix and *A* be any  $n \times n$  matrix. Then

- $0 \det(E) = -1, k \text{ or } 1.$
- $oldsymbol{2}$  det(EA) = det(E)det(A).

Let *E* be an  $n \times n$  elementary matrix and *A* be any  $n \times n$  matrix. Then

- $0 \det(E) = -1, k \text{ or } 1.$
- $2 \det(EA) = \det(E)\det(A).$
- **3**  $E^t$  is also an elementary matrix and  $det(E) = det(E^t)$ .

Let *E* be an  $n \times n$  elementary matrix and *A* be any  $n \times n$  matrix. Then

- $0 \det(E) = -1, k \text{ or } 1.$
- $2 \det(EA) = \det(E)\det(A).$
- **3**  $E^t$  is also an elementary matrix and  $det(E) = det(E^t)$ .

• A square matrix A is invertible if and only if  $det(A) \neq 0$ .

- A square matrix A is invertible if and only if  $det(A) \neq 0$ .
- Let A be an  $n \times n$  matrix. Then  $det(kA) = k^n det(A)$ .

- A square matrix A is invertible if and only if  $det(A) \neq 0$ .
- Let A be an  $n \times n$  matrix. Then  $det(kA) = k^n det(A)$ .
- Let A and B be two n × n matrices. Then det(AB) = det(A)det(B).

- A square matrix A is invertible if and only if  $det(A) \neq 0$ .
- Let A be an  $n \times n$  matrix. Then  $det(kA) = k^n det(A)$ .
- Let A and B be two n × n matrices. Then det(AB) = det(A)det(B).
- If the matrix A is invertible then  $det(A^{-1}) = \frac{1}{det(A)}$ .



- A square matrix A is invertible if and only if  $det(A) \neq 0$ .
- Let A be an  $n \times n$  matrix. Then  $det(kA) = k^n det(A)$ .
- Let A and B be two n × n matrices. Then det(AB) = det(A)det(B).
- If the matrix A is invertible then  $det(A^{-1}) = \frac{1}{det(A)}$ .
- $\bigstar$  A matrix A is said to be singular or non-singular according as  $\det(A) = 0$  or  $\det(A) \neq 0$ .

• The determinant of a triangular matrix is the product of the diagonal entries. That is, if  $A = [a_{ij}]$  is an  $n \times n$  triangular matrix then  $det(A) = a_{11}a_{22} \dots a_{nn}$ .

- The determinant of a triangular matrix is the product of the diagonal entries. That is, if  $A = [a_{ij}]$  is an  $n \times n$  triangular matrix then  $det(A) = a_{11}a_{22} \dots a_{nn}$ .
- If R is in row echelon form having a zero row, then det(R) = 0 = det(R<sup>t</sup>).

- The determinant of a triangular matrix is the product of the diagonal entries. That is, if  $A = [a_{ij}]$  is an  $n \times n$  triangular matrix then  $det(A) = a_{11}a_{22} \dots a_{nn}$ .
- If R is in row echelon form having a zero row, then det(R) = 0 = det(R<sup>t</sup>).
- For any square matrix A,  $det(A^t) = det(A)$ .

- The determinant of a triangular matrix is the product of the diagonal entries. That is, if  $A = [a_{ij}]$  is an  $n \times n$  triangular matrix then  $det(A) = a_{11}a_{22} \dots a_{nn}$ .
- If R is in row echelon form having a zero row, then det(R) = 0 = det(R<sup>t</sup>).
- For any square matrix A,  $det(A^t) = det(A)$ .
- ★ Thus a determinant can be expanded column-wise.

- The determinant of a triangular matrix is the product of the diagonal entries. That is, if  $A = [a_{ij}]$  is an  $n \times n$  triangular matrix then  $det(A) = a_{11}a_{22} \dots a_{nn}$ .
- If R is in row echelon form having a zero row, then det(R) = 0 = det(R<sup>t</sup>).
- For any square matrix A,  $det(A^t) = det(A)$ .
- ★ Thus a determinant can be expanded column-wise.
- ★ All the previous results based on row-wise expansion of determinant are also valid for column-wise expansion.

# Result (Laplace Expansion Theorem)

The determinant of an  $n \times n$  matrix  $A = [a_{ij}]$ , where  $n \ge 2$ , can be computed as

$$det(A) = a_{i1}C_{i1} + a_{i2}C_{i2} + \ldots + a_{in}C_{in} = \sum_{j=1}^{n} a_{ij}C_{ij},$$

(this is the cofactor expansion along the i-th row),

# Result (Laplace Expansion Theorem)

The determinant of an  $n \times n$  matrix  $A = [a_{ij}]$ , where  $n \ge 2$ , can be computed as

$$det(A) = a_{i1}C_{i1} + a_{i2}C_{i2} + \ldots + a_{in}C_{in} = \sum_{j=1}^{n} a_{ij}C_{ij},$$

(this is the cofactor expansion along the i-th row),

# and also as

$$det(A) = a_{1j}C_{1j} + a_{2j}C_{2j} + \ldots + a_{nj}C_{nj} = \sum_{i=1}^{n} a_{ij}C_{ij},$$

(this is the cofactor expansion along the j-th column).

 If B is obtained by interchanging any two columns of A, then det(B) = -det(A).

- If B is obtained by interchanging any two columns of A, then det(B) = -det(A).
- If A has a zero column then det(A) = 0.

- If B is obtained by interchanging any two columns of A, then det(B) = -det(A).
- If A has a zero column then det(A) = 0.
- If A has two identical columns then det(A) = 0.

- If B is obtained by interchanging any two columns of A, then det(B) = -det(A).
- If A has a zero column then det(A) = 0.
- If A has two identical columns then det(A) = 0.
- If B is obtained by multiplying a column of A by k, then det(B) = kdet(A).

- If B is obtained by interchanging any two columns of A, then det(B) = -det(A).
- If A has a zero column then det(A) = 0.
- If A has two identical columns then det(A) = 0.
- If B is obtained by multiplying a column of A by k, then det(B) = kdet(A).
- If B is obtained by adding a multiple of one column of A to another column, then det(B) = det(A).

## Definition

Let A be an  $n \times n$  matrix and let  $\mathbf{b} \in \mathbb{R}^n$ . Then  $A_i(\mathbf{b})$  denotes the matrix obtained by replacing the i-th column of A by  $\mathbf{b}$ . That is, if  $A = [\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n]$ , then  $A_i(\mathbf{b}) = [\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_{i-1} \ \mathbf{b} \ \mathbf{a}_{i+1} \ \dots \ \mathbf{a}_n]$ .

### Definition

Let A be an  $n \times n$  matrix and let  $\mathbf{b} \in \mathbb{R}^n$ . Then  $A_i(\mathbf{b})$  denotes the matrix obtained by replacing the i-th column of A by  $\mathbf{b}$ . That is, if  $A = [\mathbf{a_1} \ \mathbf{a_2} \ \dots \ \mathbf{a_n}]$ , then  $A_i(\mathbf{b}) = [\mathbf{a_1} \ \mathbf{a_2} \ \dots \ \mathbf{a_{i-1}} \ \mathbf{b} \ \mathbf{a_{i+1}} \ \dots \ \mathbf{a_n}]$ .

# Result (Cramer's Rule)

Let A be an  $n \times n$  invertible matrix and let  $\mathbf{b} \in \mathbb{R}^n$ . Then the unique solution  $\mathbf{x} = [x_1, x_2, \dots, x_n]^t$  of the system  $A\mathbf{x} = \mathbf{b}$  is given by

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

The Adjoint of a Matrix: Let  $A = [a_{ij}]$  be an  $n \times n$  matrix and let  $C_{ij}$  be the (i, j)-cofactor of A. Then the adjoint of A, denoted adj(A), is defined as

$$adj(A) = \begin{bmatrix} C_{11} & C_{21} & \dots & C_{n1} \\ C_{12} & C_{22} & \dots & C_{n2} \\ \vdots & \vdots & \vdots & \vdots \\ C_{1n} & C_{2n} & \dots & C_{nn} \end{bmatrix} = [C_{ij}]^{t}.$$

The Adjoint of a Matrix: Let  $A = [a_{ij}]$  be an  $n \times n$  matrix and let  $C_{ij}$  be the (i, j)-cofactor of A. Then the adjoint of A, denoted adj(A), is defined as

$$adj(A) = \begin{bmatrix} C_{11} & C_{21} & \dots & C_{n1} \\ C_{12} & C_{22} & \dots & C_{n2} \\ \vdots & \vdots & \vdots & \vdots \\ C_{1n} & C_{2n} & \dots & C_{nn} \end{bmatrix} = [C_{ij}]^{t}.$$

## Result

Let A be an  $n \times n$  invertible matrix. Then  $A^{-1} = \frac{1}{\det(A)}$ .adj(A).

The Adjoint of a Matrix: Let  $A = [a_{ij}]$  be an  $n \times n$  matrix and let  $C_{ij}$  be the (i, j)-cofactor of A. Then the adjoint of A, denoted adj(A), is defined as

$$adj(A) = \begin{bmatrix} C_{11} & C_{21} & \dots & C_{n1} \\ C_{12} & C_{22} & \dots & C_{n2} \\ \vdots & \vdots & \vdots & \vdots \\ C_{1n} & C_{2n} & \dots & C_{nn} \end{bmatrix} = [C_{ij}]^{t}.$$

## Result

Let A be an  $n \times n$  invertible matrix. Then  $A^{-1} = \frac{1}{\det(A)} .adj(A)$ .

### Exercise

Use the adjoint method to find the inverse of the matrix

$$A = \left[ \begin{array}{rrr} 1 & 2 & -1 \\ 2 & 2 & 4 \\ 1 & 3 & -3 \end{array} \right].$$

#### Definition

Let A be an  $m \times n$  matrix.

The null space of A, denoted null(A), is the subspace of  $\mathbb{R}^n$  consisting of the solutions of the homogeneous linear system  $A\mathbf{x} = \mathbf{0}$ .

### Definition

Let A be an  $m \times n$  matrix.

1 The null space of A, denoted  $\operatorname{null}(A)$ , is the subspace of  $\mathbb{R}^n$ consisting of the solutions of the homogeneous linear system Ax = 0. In other words,

$$\mathsf{null}(A) = \{\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0}\}$$

#### Definition

Let A be an  $m \times n$  matrix.

- The null space of A, denoted null(A), is the subspace of  $\mathbb{R}^n$  consisting of the solutions of the homogeneous linear system  $A\mathbf{x} = \mathbf{0}$ . In other words, null(A) =  $\{\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0}\}$
- The column space of A, denoted col(A), is the subspace of  $\mathbb{R}^m$  spanned by the columns of A.

#### Definition

Let A be an  $m \times n$  matrix.

- The null space of A, denoted null(A), is the subspace of  $\mathbb{R}^n$  consisting of the solutions of the homogeneous linear system  $A\mathbf{x} = \mathbf{0}$ . In other words, null(A) = { $\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0}$ }
- The column space of A, denoted col(A), is the subspace of  $\mathbb{R}^m$  spanned by the columns of A. In other words,  $col(A) = \{A\mathbf{x} \mid \mathbf{x} \in \mathbb{R}^n\}$ .

#### Definition

Let A be an  $m \times n$  matrix.

- The null space of A, denoted null(A), is the subspace of  $\mathbb{R}^n$  consisting of the solutions of the homogeneous linear system  $A\mathbf{x} = \mathbf{0}$ . In other words, null(A) = { $\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0}$ }
- The column space of A, denoted col(A), is the subspace of  $\mathbb{R}^m$  spanned by the columns of A. In other words,  $col(A) = \{A\mathbf{x} \mid \mathbf{x} \in \mathbb{R}^n\}$ .
- The row space of A, denoted row(A), is the subspace of  $\mathbb{R}^n$  spanned by the rows of A.

#### Definition

Let A be an  $m \times n$  matrix.

- The null space of A, denoted null(A), is the subspace of  $\mathbb{R}^n$  consisting of the solutions of the homogeneous linear system  $A\mathbf{x} = \mathbf{0}$ . In other words, null(A) =  $\{\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0}\}$
- The column space of A, denoted col(A), is the subspace of  $\mathbb{R}^m$  spanned by the columns of A. In other words,  $col(A) = \{A\mathbf{x} \mid \mathbf{x} \in \mathbb{R}^n\}$ .
- The row space of A, denoted row(A), is the subspace of  $\mathbb{R}^n$  spanned by the rows of A. In other words,  $row(A) = \{\mathbf{x}^T A \mid \mathbf{x} \in \mathbb{R}^m\}$ 
  - [Here, elements of row(A) are row vectors. How can they be elements of  $\mathbb{R}^n$ ?



#### Definition

Let A be an  $m \times n$  matrix.

- The null space of A, denoted null(A), is the subspace of  $\mathbb{R}^n$  consisting of the solutions of the homogeneous linear system  $A\mathbf{x} = \mathbf{0}$ . In other words, null(A) =  $\{\mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0}\}$
- The column space of A, denoted col(A), is the subspace of  $\mathbb{R}^m$  spanned by the columns of A. In other words,  $col(A) = \{A\mathbf{x} \mid \mathbf{x} \in \mathbb{R}^n\}$ .
- The row space of A, denoted row(A), is the subspace of  $\mathbb{R}^n$  spanned by the rows of A. In other words,  $row(A) = \{\mathbf{x}^T A \mid \mathbf{x} \in \mathbb{R}^m\}$ 
  - [Here, elements of row(A) are row vectors. How can they be elements of  $\mathbb{R}^n$ ? In strict sense,  $row(A) := col(A^T)$ .]

• Let B be a matrix that is row equivalent to the matrix A. Then row(B) = row(A).

- Let B be a matrix that is row equivalent to the matrix A. Then row(B) = row(A).
- For any matrix A, row(A) = row(RREF(A)).

- Let B be a matrix that is row equivalent to the matrix A. Then row(B) = row(A).
- For any matrix A, row(A) = row(RREF(A)).
- For any A, the non-zero rows of RREF(A) forms a basis of row(A).

- Let B be a matrix that is row equivalent to the matrix A.
   Then row(B) = row(A).
- For any matrix A, row(A) = row(RREF(A)).
- For any A, the non-zero rows of RREF(A) forms a basis of row(A).

Suppose A and B are row-equivalent. Are col(A) and col(B) equal?

- Let B be a matrix that is row equivalent to the matrix A.
   Then row(B) = row(A).
- For any matrix A, row(A) = row(RREF(A)).
- For any A, the non-zero rows of RREF(A) forms a basis of row(A).

Suppose A and B are row-equivalent. Are col(A) and col(B) equal? No.

- Let B be a matrix that is row equivalent to the matrix A.
   Then row(B) = row(A).
- For any matrix A, row(A) = row(RREF(A)).
- For any A, the non-zero rows of RREF(A) forms a basis of row(A).

Suppose A and B are row-equivalent. Are col(A) and col(B) equal? No. Take  $A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ .

- Let B be a matrix that is row equivalent to the matrix A.
   Then row(B) = row(A).
- For any matrix A, row(A) = row(RREF(A)).
- For any A, the non-zero rows of RREF(A) forms a basis of row(A).

Suppose A and B are row-equivalent. Are col(A) and col(B) equal? No. Take  $A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ .

Suppose A and B are row-equivalent. Do col(A) and col(B) have same dimension?

- Let B be a matrix that is row equivalent to the matrix A.
   Then row(B) = row(A).
- For any matrix A, row(A) = row(RREF(A)).
- For any A, the non-zero rows of RREF(A) forms a basis of row(A).

Suppose A and B are row-equivalent. Are col(A) and col(B) equal? No. Take  $A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ .

Suppose A and B are row-equivalent. Do col(A) and col(B) have same dimension? Yes. We will see soon.

Let A be a given matrix and let R be the reduced row echelon form of A.

Use the non-zero rows of R to form a basis for row(A).

- Use the non-zero rows of R to form a basis for row(A).
- Solve the leading variables of Rx = 0 in terms of the free variables, set the free variables equal to parameters, substitute back into x, write the result as a linear combination of k vectors (where k is the number of free variables). These k vectors form a basis for null(A).

- Use the non-zero rows of R to form a basis for row(A).
- Solve the leading variables of Rx = 0 in terms of the free variables, set the free variables equal to parameters, substitute back into x, write the result as a linear combination of k vectors (where k is the number of free variables). These k vectors form a basis for null(A).
- 3 A basis for  $row(A^t)$  will also be a basis for col(A).

- Use the non-zero rows of R to form a basis for row(A).
- Solve the leading variables of Rx = 0 in terms of the free variables, set the free variables equal to parameters, substitute back into x, write the result as a linear combination of k vectors (where k is the number of free variables). These k vectors form a basis for null(A).
- 3 A basis for  $row(A^t)$  will also be a basis for col(A).
- Or, Use the columns of A that correspond to the columns of R containing the leading 1's to form a basis for col(A).

## Example

Find bases for the row space, column space and null space of the following matrix:

$$A = \left[ \begin{array}{rrr} 1 & 2 & -1 \\ 2 & 2 & 4 \\ 4 & 6 & 2 \end{array} \right],$$

## Example

Find bases for the row space, column space and null space of the following matrix:

$$A = \left[ \begin{array}{ccc} 1 & 2 & -1 \\ 2 & 2 & 4 \\ 4 & 6 & 2 \end{array} \right], \quad \textit{RREF}(A) = \left[ \begin{array}{ccc} 1 & 0 & 5 \\ 0 & 1 & -3 \\ 0 & 0 & 0 \end{array} \right].$$

## Example

Find bases for the row space, column space and null space of the following matrix:

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 2 & 4 \\ 4 & 6 & 2 \end{bmatrix}, \quad RREF(A) = \begin{bmatrix} 1 & 0 & 5 \\ 0 & 1 & -3 \\ 0 & 0 & 0 \end{bmatrix}.$$

#### Result

Let  $R = [\mathbf{b_1} \ \mathbf{b_2} \ \dots \ \mathbf{b_n}]$  be the reduced row echelen form of a matrix  $A = [\mathbf{a_1} \ \mathbf{a_2} \ \dots \ \mathbf{a_n}]$  of rank r. Let  $\mathbf{b_{j_1}}, \mathbf{b_{j_2}}, \dots, \mathbf{b_{j_r}}$  be the columns of R such that  $\mathbf{b_{j_k}} = \mathbf{e_k}$  for  $k = 1, \dots, r$ . Then  $\{\mathbf{a_{j_1}}, \mathbf{a_{j_2}}, \dots, \mathbf{a_{j_r}}\}$  is a basis for col(A).



The row space and the column space of a matrix A have the same dimension, and dim(row(A)) = dim(col(A)) = rank(A).

The row space and the column space of a matrix A have the same dimension, and dim(row(A)) = dim(col(A)) = rank(A).

#### Result

For any matrix A, we have  $rank(A^t) = rank(A)$ .

The row space and the column space of a matrix A have the same dimension, and dim(row(A)) = dim(col(A)) = rank(A).

#### Result

For any matrix A, we have  $rank(A^t) = rank(A)$ .

**Nullity:** The nullity of a matrix A is the dimension of its null space, and is denoted by nullity(A).

The row space and the column space of a matrix A have the same dimension, and dim(row(A)) = dim(col(A)) = rank(A).

#### Result

For any matrix A, we have  $rank(A^t) = rank(A)$ .

**Nullity:** The nullity of a matrix A is the dimension of its null space, and is denoted by nullity(A).

## Result (Rank Nullity Theorem)

Let A be an  $m \times n$  matrix. Then

$$rank(A) + nullity(A) = n.$$



## Result (Fundamental Theorem of Invertible Matrices: II)

# Result (Fundamental Theorem of Invertible Matrices: II) Let A be an $n \times n$ matrix. Then the following statements are

Let A be an  $n \times n$  matrix. Then the following statements are equivalent.

# Result (Fundamental Theorem of Invertible Matrices: II)

Let A be an  $n \times n$  matrix. Then the following statements are equivalent.

- A is invertible.
- $\triangle$  A<sup>t</sup> is invertible.
- **3**  $A\mathbf{x} = \mathbf{b}$  has a solution for every  $\mathbf{b}$  in  $\mathbb{R}^n$ .
- **4**  $\mathbf{A}\mathbf{x} = \mathbf{b}$  has a unique solution for every  $\mathbf{b}$  in  $\mathbb{R}^n$ .
- **3**  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.
- **1** The reduced row echelon form of A is  $I_n$ .
- The rows of A are linearly independent.
- The columns of A are linearly independent.

11. nullity(A) = 0.

- 11. nullity(A) = 0.
- 12. The column vectors of A span  $\mathbb{R}^n$ .

- 11. nullity(A) = 0.
- 12. The column vectors of A span  $\mathbb{R}^n$ .
- **13**. The column vectors of *A* form a basis for  $\mathbb{R}^n$ .

- 11. nullity(A) = 0.
- 12. The column vectors of A span  $\mathbb{R}^n$ .
- **13**. The column vectors of *A* form a basis for  $\mathbb{R}^n$ .
- **14.** The row vectors of *A* span  $\mathbb{R}^n$ .

- 11. nullity(A) = 0.
- 12. The column vectors of A span  $\mathbb{R}^n$ .
- **13**. The column vectors of *A* form a basis for  $\mathbb{R}^n$ .
- **14**. The row vectors of *A* span  $\mathbb{R}^n$ .
- **15**. The row vectors of *A* form a basis for  $\mathbb{R}^n$ .

Show that the vectors  $[1,2,3]^t$ ,  $[-1,0,1]^t$  and  $[4,9,7]^t$  form a basis for  $\mathbb{R}^3$ .

Show that the vectors  $[1,2,3]^t$ ,  $[-1,0,1]^t$  and  $[4,9,7]^t$  form a basis for  $\mathbb{R}^3$ .

#### Result

Let A be an  $m \times n$  matrix. Then

Show that the vectors  $[1,2,3]^t$ ,  $[-1,0,1]^t$  and  $[4,9,7]^t$  form a basis for  $\mathbb{R}^3$ .

#### Result

Let A be an  $m \times n$  matrix. Then

 $\bigcirc$  rank( $A^tA$ ) = rank(A).

Show that the vectors  $[1,2,3]^t$ ,  $[-1,0,1]^t$  and  $[4,9,7]^t$  form a basis for  $\mathbb{R}^3$ .

#### Result

Let A be an  $m \times n$  matrix. Then

- $\bullet$  rank( $A^tA$ ) = rank(A).
- 2 The  $n \times n$  matrix  $A^t A$  is invertible if and only if rank(A) = n.

Let A, B, T, S be matrices, where T are S are invertible.

If TA and AS are defined, then rank(TA) = rank(A) = rank(AS).

Let A, B, T, S be matrices, where T are S are invertible.

- If TA and AS are defined, then rank(TA) = rank(A) = rank(AS).
- ② If AB is defined then  $rank(AB) \le min\{rank(A), rank(B)\}$ .

Let A, B, T, S be matrices, where T are S are invertible.

- If TA and AS are defined, then rank(TA) = rank(A) = rank(AS).
- ② If AB is defined then  $rank(AB) \le min\{rank(A), rank(B)\}$ .
- ③ If A + B is defined then  $rank(A + B) \le rank(A) + rank(B)$ .

Let A, B, T, S be matrices, where T are S are invertible.

- If TA and AS are defined, then rank(TA) = rank(A) = rank(AS).
- ② If AB is defined then  $rank(AB) \le min\{rank(A), rank(B)\}$ .
- If A + B is defined then  $rank(A + B) \le rank(A) + rank(B)$ .

#### Result

Let A be an  $n \times n$  matrix of rank r, where  $1 \le r < n$ . Show that there exist elementary matrices  $E_1, \ldots, E_p$  and  $F_1, \ldots, F_q$  such that  $E_1 \ldots E_p A F_1 \ldots F_q = \begin{bmatrix} I_r & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{bmatrix}$ .