

- (1) **Elementary matrix:** An  $n \times n$  matrix is called elementary matrix if it can be obtained from the identity matrix  $I_n$  by performing a single elementary row operations.

Ex:

$$\mathbf{E}_1 = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \quad \text{and} \quad \mathbf{E}_2 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & -5 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad \text{are elementary.} \quad C = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \quad \text{is not elementary.}$$

- (2) **Properties:**

- (a) When a matrix  $\mathbf{A}$  is multiplied on the *left* by an elementary matrix  $\mathbf{E}$ , the result is the same as performing the same row operation on  $\mathbf{A}$ .

Thus, for every matrix  $A$ , there exist a sequence of elementary matrices  $E_1, E_2, \dots, E_k$  such that

$$E_k \cdots E_2 E_1 A = rref(A).$$

Ex:

$$A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 3 & -1 \\ 1 & -1 & 0 \end{bmatrix} \quad E_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \implies E_1 A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 3 & -1 \\ 0 & -3 & 0 \end{bmatrix}$$

$$E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \implies E_2 E_1 A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 3 & -1 \\ 0 & 0 & -1 \end{bmatrix}$$

$$E_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{3} & 0 \\ 0 & 0 & 1 \end{bmatrix} \implies E_3 E_2 E_1 A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & -\frac{1}{3} \\ 0 & 0 & -1 \end{bmatrix}$$

We can continue the process until we get  $rref(A)$ .

- (b) Elementary matrix is invertible and its inverse is also an elementary matrix.
- (3) **Theorem:** Let  $A$  be  $n \times n$  matrix. The following statements are equivalence:
- $A$  is invertible.
  - $A\mathbf{x} = \mathbf{b}$  has exactly one solution for every  $n \times 1$  matrix  $\mathbf{b}$ .
  - $A\mathbf{x} = \mathbf{0}$  has only trivial solution.
  - The  $rref(A)$  is  $I_n$
  - $A$  is expressible as a product of elementary matrices.
- (4) **Row equivalence:** If a matrix  $B$  can be obtained from a matrix  $A$  by performing a finite sequence of row operations, then  $A$  and  $B$  are row equivalence.
- (5) **Finding  $A^{-1}$ :** From part (d) in theorem above,  $A$  is invertible iff it is row equivalent to  $I_n$ . Therefore, if  $A$  is invertible, we have

$$E_k \cdots E_3 E_2 E_1 A = I_n \implies E_k \cdots E_3 E_2 E_1 = I_n A^{-1} \implies A^{-1} = E_k \cdots E_3 E_2 E_1 I_n.$$

This means we can perform a sequence of row operations on  $A$  to obtain  $I_n$  and do the same row of operation on  $I_n$  to get  $A^{-1}$ . Thus, to find  $A^{-1}$ , we do

$$\left[ \mathbf{A} : I_n \right] \cdots \text{row operations} \cdots \left[ I_n : A^{-1} \right]$$

If the *left* side has a row of zeros; i.e., does not look like  $I_n$ , then  $A$  does not have inverse.

Ex.:

$$\begin{aligned} \left[ \begin{array}{ccc|ccc} 1 & 2 & 0 & : & 1 & 0 & 0 \\ 0 & 3 & -1 & : & 0 & 1 & 0 \\ 1 & -1 & 0 & : & 0 & 0 & 1 \end{array} \right] &\longrightarrow \left[ \begin{array}{ccc|ccc} 1 & 2 & 0 & : & 1 & 0 & 0 \\ 0 & 3 & -1 & : & 0 & 1 & 0 \\ 0 & -3 & 0 & : & -1 & 0 & 1 \end{array} \right] &\longrightarrow \left[ \begin{array}{ccc|ccc} 1 & 2 & 0 & : & 1 & 0 & 0 \\ 0 & 0 & -1 & : & -1 & 1 & 1 \\ 0 & -3 & 0 & : & -1 & 0 & 1 \end{array} \right] \\ &\longrightarrow \left[ \begin{array}{ccc|ccc} 1 & 2 & 0 & : & 1 & 0 & 0 \\ 0 & -3 & 0 & : & -1 & 0 & 1 \\ 0 & 0 & -1 & : & -1 & 1 & 1 \end{array} \right] &\longrightarrow \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & : & 1/3 & 0 & 2/3 \\ 0 & 1 & 0 & : & 1/3 & 0 & -1/3 \\ 0 & 0 & 1 & : & 1 & -1 & -1 \end{array} \right] \end{aligned}$$

(6) **Linear system and invertability:**

(a) Suppose that  $\mathbf{A}$  is an invertible matrix. Then the system  $\mathbf{Ax} = \mathbf{b}$  has exactly one solution; namely,  $\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$ .

Note: this only applies to systems that have number of equations equals to number of variables.

Ex.:

$$\begin{aligned} x_1 + 2x_2 &= 1 \\ 3x_2 - x_3 &= -3 \\ x_1 - x_2 &= 4 \end{aligned} \implies \mathbf{A} = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 3 & -1 \\ 1 & -1 & 0 \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} 1 \\ -3 \\ 4 \end{bmatrix}$$

- Solve by Gauss-Jordan elimination:

$$\left[ \begin{array}{ccc|c} 1 & 2 & 0 & : & 1 \\ 0 & 3 & -1 & : & -3 \\ 1 & -1 & 0 & : & 4 \end{array} \right] \longrightarrow \dots \longrightarrow \left[ \begin{array}{ccc|c} 1 & 0 & 0 & : & 3 \\ 0 & 1 & 0 & : & -1 \\ 0 & 0 & 1 & : & 0 \end{array} \right] \implies x_1 = 3, x_2 = -1, x_3 = 0$$

- Solve by using inverse:

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b} = \begin{bmatrix} 1/3 & 0 & 2/3 \\ 1/3 & 0 & -1/3 \\ 1 & -1 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ -3 \\ 4 \end{bmatrix} = \begin{bmatrix} 3 \\ -1 \\ 0 \end{bmatrix}$$

(b) Systems with common coefficient matrix: To solve systems

$$\mathbf{Ax} = \mathbf{b}_1, \quad \mathbf{Ax} = \mathbf{b}_2, \quad \dots, \mathbf{Ax} = \mathbf{b}_k$$

we can use

- inverse:  $\mathbf{x}_1 = \mathbf{A}^{-1}\mathbf{b}_1, \quad \mathbf{x}_2 = \mathbf{A}^{-1}\mathbf{b}_2, \dots, \mathbf{x}_k = \mathbf{A}^{-1}\mathbf{b}_k$

- Gauss-Jordan elimination method

$$\left[ \mathbf{A} : \mathbf{b}_1 : \mathbf{b}_2 : \dots : \mathbf{b}_k \right]$$