



# Elementary Row Operations and Linear Equations

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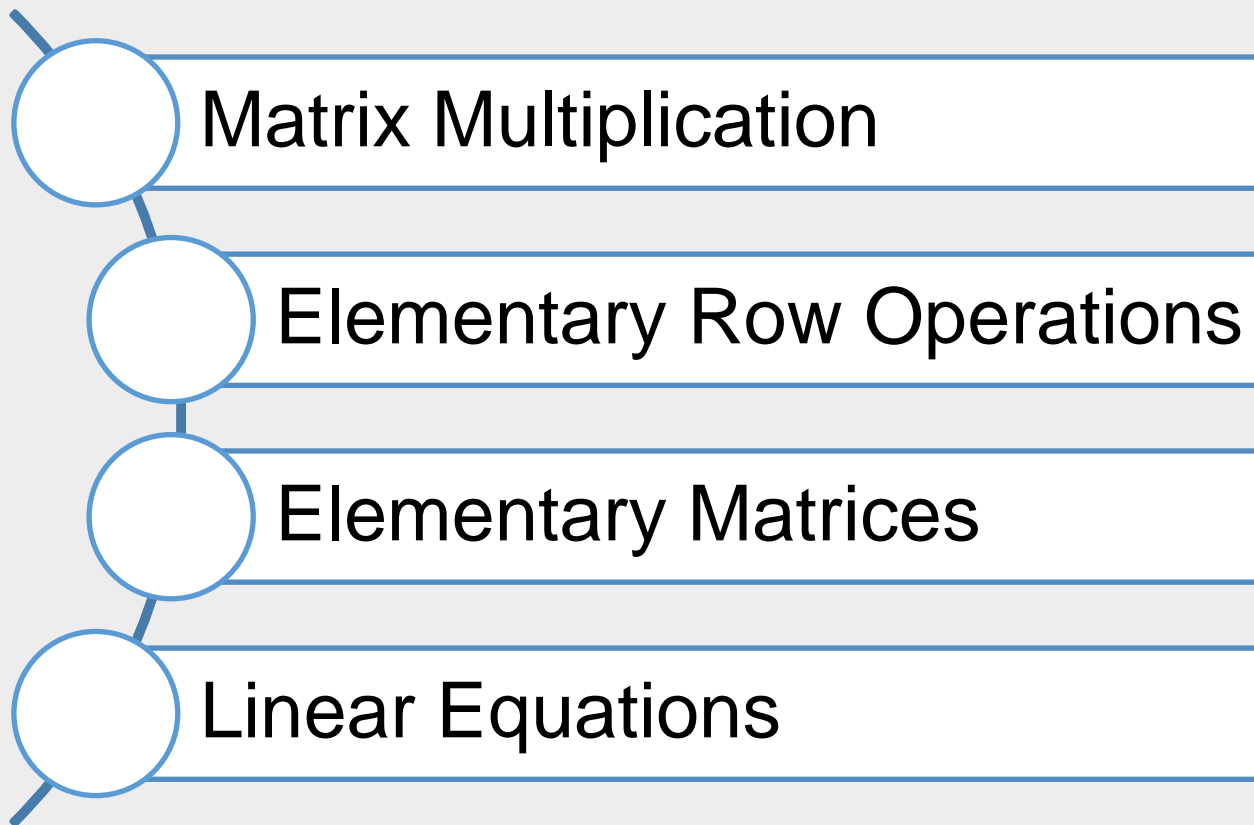
## Linear Algebra

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# Matrix Multiplication

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- By  $A \in \mathbb{R}^{m \times n}$  we denote a matrix with  $m$  rows and  $n$  columns, where the entries of  $A$  are real numbers.

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} = \begin{bmatrix} | & | & \cdots & | \\ a_1 & a_2 & \cdots & a_n \\ | & | & \cdots & | \end{bmatrix} = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ \vdots & \vdots & \vdots \\ - & a_m^T & - \end{bmatrix}$$



- If we write  $A$  by rows, then we can express  $Ax$  as,

$$A \in \mathbb{R}^{m \times n} \quad y = Ax = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ & \vdots & \\ - & a_m^T & - \end{bmatrix} x = \begin{bmatrix} a_1^T x \\ a_2^T x \\ \vdots \\ a_m^T x \end{bmatrix}. \quad a_i^T x = \sum_{j=1}^n a_{ij} x_j$$

- If we write  $A$  by columns, then we have:

$$y = Ax = \begin{bmatrix} | & | & & | \\ a_1 & a_2 & \cdots & a_n \\ | & | & & | \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = [a_1]x_1 + [a_2]x_2 + \cdots + [a_n]x_n.$$

- $y$  is a linear combination of the columns  $A$ .

columns of  $A$  are linearly independent if  $Ax = 0$  implies  $x = 0$



It is also possible to multiply on the left by a row vector.

- If we write  $A$  by columns, then we can express  $x^T A$  as,

$$y^T = x^T A = x^T \begin{bmatrix} | & | & \cdots & | \\ a_1 & a_2 & \cdots & a_n \\ | & | & \cdots & | \end{bmatrix} = [x^T a_1 \quad x^T a_2 \quad \cdots \quad x^T a_n]$$

- expressing  $A$  in terms of rows we have:

$$y^T = x^T A = [x_1 \quad x_2 \quad \cdots \quad x_m] \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ & \vdots & \\ - & a_m^T & - \end{bmatrix}$$
$$= x_1 [- \quad a_1^T \quad -] + x_2 [- \quad a_2^T \quad -] + \cdots + x_m [- \quad a_m^T \quad -]$$

- $y^T$  is a linear combination of the rows of  $A$ .



## □ Properties

- $A(u + v) = Au + Av$
- $(A + B)u = Au + Bu$
- $(\alpha A)u = \alpha(Au) = A(\alpha u) = \alpha Au$
- $0u = 0$
- $A0 = 0$
- $Iu = u$



$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} = \begin{bmatrix} | & | & \cdots & | \\ a_1 & a_2 & \cdots & a_n \\ | & | & \cdots & | \end{bmatrix} = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ & \vdots & \\ - & a_m^T & - \end{bmatrix}$$

Example: Write in matrix-vector multiplication

- Column  $j$ :  $a_j =$
- Row  $i$ :  $a_i^T =$
- Vector sum of rows of  $A =$
- Vector sum of columns of  $A =$

$$A = \begin{bmatrix} -1 & 2 & 1 \\ 2 & 0 & -2 \end{bmatrix}$$





## Definition

Let  $A$  be an  $m \times n$  matrix over the field  $F$  and let  $B$  be an  $n \times p$  matrix over  $F$ . The product  $AB$  is the  $m \times p$  matrix  $C$  whose  $i, j$  entry is:

$$C_{ij} = \sum_{r=1}^n A_{ir} B_{rj}$$

# Matrix–Matrix Multiplication



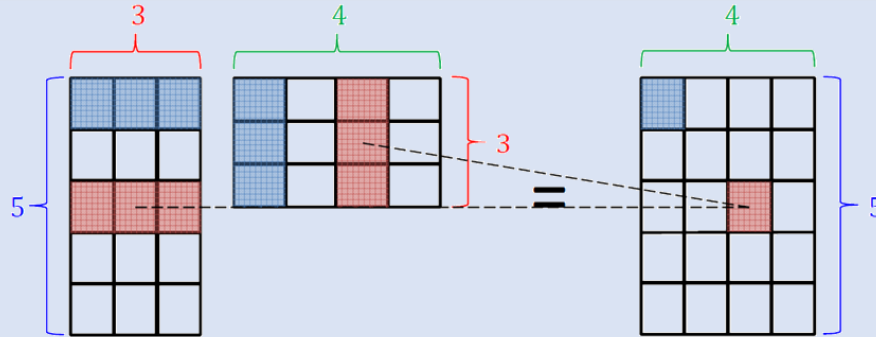
- $A \in \mathbb{R}^{m \times n}, B \in \mathbb{R}^{n \times p} \rightarrow \mathbb{R}^{m \times p}$ 
  - $a_i$  rows of A,  $b_j$  cols of B

$$C = AB \quad \text{for } 1 \leq i \leq m, 1 \leq j \leq p$$

**inner product**  $(a_i \cdot b_j)$

$$C_{ij} = a_i^T b_j$$

## Example





1. As a set of vector–vector products

$$C = AB = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ & \vdots & \\ - & a_m^T & - \end{bmatrix} \begin{bmatrix} | & | & \dots & | \\ b_1 & b_2 & \dots & b_p \\ | & | & & | \end{bmatrix} = \begin{bmatrix} a_1^T b_1 & a_1^T b_2 & \dots & a_1^T b_p \\ a_2^T b_1 & a_2^T b_2 & \dots & a_2^T b_p \\ \vdots & \vdots & \ddots & \vdots \\ a_m^T b_1 & a_m^T b_2 & \dots & a_m^T b_p \end{bmatrix}$$

2. As a sum of outer products

$$C = AB = \begin{bmatrix} | & | & \dots & | \\ a_1 & a_2 & \dots & a_n \\ | & | & & | \end{bmatrix} \begin{bmatrix} - & b_1^T & - \\ - & b_2^T & - \\ & \vdots & \\ - & b_n^T & - \end{bmatrix} = \sum_{i=1}^n a_i b_i^T$$



### 3. As a set of matrix–vector products.

$$C = AB = A \begin{bmatrix} | & | & \cdots & | \\ b_1 & b_2 & \cdots & b_p \\ | & | & \cdots & | \end{bmatrix} = \begin{bmatrix} | & | & \cdots & | \\ Ab_1 & Ab_2 & \cdots & Ab_p \\ | & | & \cdots & | \end{bmatrix}$$

Here the  $i$ th column of  $C$  is given by the matrix–vector product with the vector on the right,  $c_i = Ab_i$ . These matrix–vector products can in turn be interpreted using both viewpoints given in the previous subsection.

### 4. As a set of vector–matrix products.

$$C = AB = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ & \vdots & \\ - & a_m^T & - \end{bmatrix} B = \begin{bmatrix} - & a_1^T B & - \\ - & a_2^T B & - \\ & \vdots & \\ - & a_m^T B & - \end{bmatrix}$$



## □ Properties:

- Associative

$$(AB)C = A(BC)$$

- Distributive

$$A(B + C) = AB + AC$$

- NOT commutative

$$AB \neq BA$$

– Dimensions may not even be conformable



## Theorem

If  $A, B, C$  are matrices over the field  $F$  such that the products  $BC$  and  $A(BC)$  is defined, then so are the products  $AB$ ,  $(AB)C$  and

$$A(BC) = (AB)C$$

Proof:

## Note

Linear combinations of linear combinations of the rows of  $C$  are again linear combinations of the rows of  $C$



- $A^k$ : repeated multiplication of a square matrix

$$A^1 = A, A^2 = AA, \dots, A^k = \underbrace{AA \cdots A}_{k \text{ matrices}}$$

- Properties:

- $A^j A^k = A^{j+k}$
- $(A^j)^k = A^{jk}$

where j and k are non-negative integers and  $A^0$  is assumed to be I

- For diagonal matrices:

$$D = \begin{bmatrix} d_1 & 0 & \cdots & 0 \\ 0 & d_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & d_n \end{bmatrix} \Rightarrow D^k = \begin{bmatrix} d_1^k & 0 & \cdots & 0 \\ 0 & d_2^k & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & d_n^k \end{bmatrix}$$

# Elementary Row Operations

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## □ Elementary Row Operations

1. **Scaling**: Multiply all entries in a row by a nonzero scalar.
2. **Replacement**: Replace one row by the sum of itself and a multiple of another row.
3. **Interchange**: Interchange two rows.

## □ Elementary Row Operation is a special type of function $e$ on $m \times n$ matrix $A$ and gives an $m \times n$ matrix $e(A)$

1. **Scaling**:  $e(A)_{ij} = cA_{ij}$
2. **Replacement**:  $e(A)_{ij} = A_{ij} + cA_{kj}$
3. **Interchange**:  $e(A)_{ij} = A_{kj}$  ,  $e(A)_{kj} = A_{ij}$

In defining  $e(A)$ , it is not really important how many columns  $A$  has, but the number of rows of  $A$  is crucial.



## Theorem

The inverse operation (function) of an elementary row operation exists and is a elementary row operation of the same type.

**Proof:**



## Definition

If  $A$  and  $B$  are  $m \times n$  matrices over the field  $F$ , we say that  $B$  is **row-equivalent** to  $A$  if  $B$  can be obtained from  $A$  by a finite sequence of elementary row operations.

## Note (from pervious theorem and this definition)

- ☐ Each matrix is row-equivalent to itself
- ☐ If  $B$  is row-equivalent to  $A$ , then  $A$  is row-equivalent to  $B$ .
- ☐ If  $B$  is row-equivalent to  $A$ ,  $C$  is row-equivalent to  $B$ , then  $C$  is row-equivalent to  $A$

# Elementary Matrices

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## Definition

A  $m \times n$  matrix is an elementary matrix if it can be obtained from the  $m \times m$  identity matrix by means of a **single elementary row operation**.

## Example

Find all  $2 \times 2$  elementary matrices.



## Theorem

Let  $e$  be an elementary row operation and let  $E$  be the  $m \times m$  elementary matrix  $E = e(I)$ . Then, for every  $m \times n$  matrix  $A$ :

$$e(A) = EA$$

**Proof:**

**Multiplication of a matrix on the left by a square matrix performs row operations.**



## Example

(From [theorem](#))

$$M_4(M_3(M_2(M_1A))) = (M_4(M_3(M_2M_1)))A$$

Matrix	Elementary row operation	Elementary matrix
$\begin{bmatrix} 1 & 0 & 2 \\ -2 & 0 & -3 \\ 0 & 2 & 0 \end{bmatrix}$	$R_2 \leftarrow R_2 + 2R_1$	$M_1 = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$
$\begin{bmatrix} 1 & 0 & 2 \\ 0 & 0 & 1 \\ 0 & 2 & 0 \end{bmatrix}$	$R_2 \leftrightarrow R_3$	$M_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$
$\begin{bmatrix} 1 & 0 & 2 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$R_2 \leftarrow \frac{1}{2}R_2$	$M_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & 1 \end{bmatrix}$
$\begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$R_1 \leftarrow R_1 + (-2)R_3$	$M_4 = \begin{bmatrix} 1 & 0 & -2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$
$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$		



## Theorem

Let  $A$  and  $B$  be  $m \times n$  matrices over the field  $F$ . Then  $B$  is row-equivalent to  $A$  if and only if  $B = PA$ , where  $P$  is a product of  $m \times m$  elementary matrices.



# Linear Equations

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## Definition

A system of  $m$  linear equations with  $n$  unknowns:

□  $F$  is a field, we want to find  $n$  scalars (elements of  $F$ )  $x_1, \dots, x_n$  which satisfy the conditions: ( $A_{ij}, y_k$  are elements of  $F$ )

$$A_{11}x_1 + A_{12}x_2 + \dots + A_{1n}x_n = y_1$$

$$A_{21}x_1 + A_{22}x_2 + \dots + A_{2n}x_n = y_2$$

...

$$A_{m1}x_1 + A_{m2}x_2 + \dots + A_{mn}x_n = y_m$$

If  $y_1 = y_2 = \dots = y_m = 0$ , we say that the system is **homogeneous**.

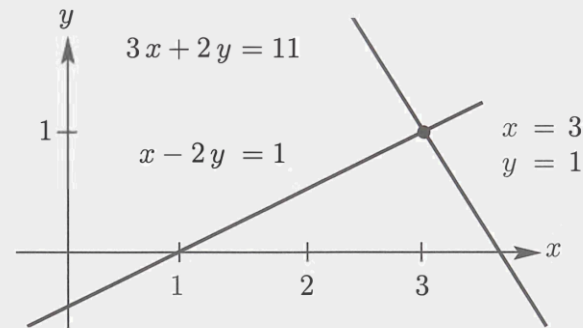
A **solution** of this **system of linear equations** is vector  $\begin{bmatrix} s_1 \\ \vdots \\ s_n \end{bmatrix}$  whose

components satisfy  $x_1 = s_1, \dots, x_n = s_n$



- Consider this simple system of equations,

$$\begin{aligned}x - y &= 1 \\ 3x + 2y &= 11\end{aligned}$$



- Can be expressed as a matrix-vector multiplication
- Matrix Equation:  $Ax=b$

$$\underbrace{\begin{bmatrix} 1 & -2 \\ 3 & 2 \end{bmatrix}}_A \underbrace{\begin{bmatrix} x \\ y \end{bmatrix}}_x = \underbrace{\begin{bmatrix} 1 \\ 11 \end{bmatrix}}_b$$

- $A$  is often called **coefficient matrix**:  $\begin{bmatrix} 1 & -2 \\ 3 & 2 \end{bmatrix}$
- $Ab$  is an **Augmented matrix**:  $\begin{bmatrix} 1 & -2 & 1 \\ 3 & 2 & 11 \end{bmatrix}$

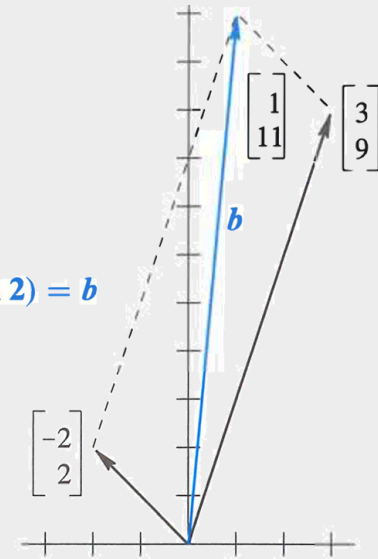
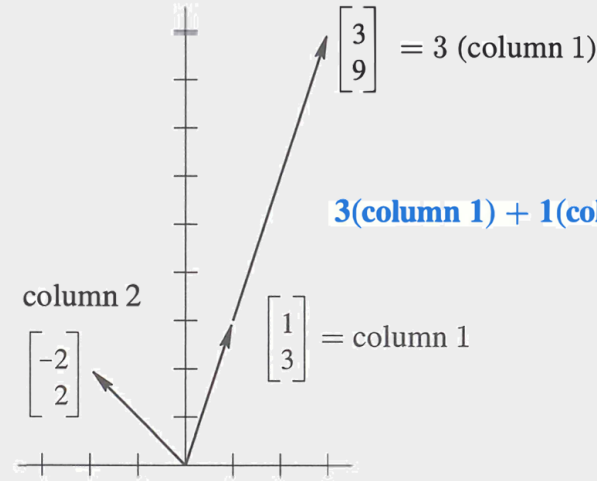


□ Also, Can be expressed as linear combination of cols:

$$\begin{aligned} x - 2y &= 1 \\ 3x + 2y &= 11 \end{aligned}$$

$$\underbrace{\begin{bmatrix} 1 & -2 \\ 3 & 2 \end{bmatrix}}_A \underbrace{\begin{bmatrix} x \\ y \end{bmatrix}}_x = \underbrace{\begin{bmatrix} 1 \\ 11 \end{bmatrix}}_b$$

$$x \begin{bmatrix} 1 \\ 3 \end{bmatrix} + y \begin{bmatrix} -2 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 11 \end{bmatrix} = b$$



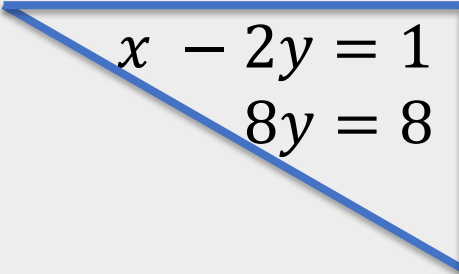
□ Same for  $n$  equation,  $n$  variable



- ❑ Subtract a multiple of equation (1) from (2) to eliminate a variable

$$\begin{aligned}x - 2y &= 1 \\ 3x + 2y &= 11\end{aligned}$$

multiply equation 1 by 3  
Subtract to eliminate  $3x$


$$\begin{aligned}x - 2y &= 1 \\ 8y &= 8\end{aligned}$$

$$\underbrace{\begin{bmatrix} 1 & -2 \\ 0 & 8 \end{bmatrix}}_U \underbrace{\begin{bmatrix} x \\ y \end{bmatrix}}_x = \underbrace{\begin{bmatrix} 1 \\ 8 \end{bmatrix}}_c$$

$A$  has become an upper triangle matrix  $U$



- The **pivots** are on the diagonal of the triangle after elimination (boldface 2 below is the first pivot)

$$\begin{array}{rcl} 2x + 4y - 2z & = & 2 \\ 4x + 9y - 3z & = & 8 \\ -2x - 3y + 7z & = & 10 \end{array} \quad \longrightarrow \quad \begin{array}{rcl} 2x + 4y - 2z & = & 2 \\ & 1y + 1z & = & 4 \\ & & 4z & = & 8 \end{array}$$

- Step 1: subtract (1) from (2) to eliminate x's in (2)
- Step 2: subtract (1) from (3) to totally eliminate x
- Step 3: subtract new (2) from new (3)

## Definition

The variables corresponding to pivot columns in the matrix are called **basic variables**.

The other variables are called a **free variable**.

$$\begin{bmatrix} x & x & x & x \\ 0 & x & x & x \\ 0 & 0 & x & x \\ 0 & 0 & 0 & x \end{bmatrix} \quad \begin{bmatrix} x & x & x & x \\ 0 & x & x & x \\ 0 & 0 & 0 & x \\ 0 & 0 & 0 & x \end{bmatrix}$$



## Theorem

If  $A$  and  $B$  are row-equivalent  $m \times n$  matrices, the homogenous systems of linear equations  $Ax = 0$  and  $Bx = 0$  have exactly the same solutions.

**Proof:**



## Example

Find the solution for this system.

Suppose  $F$  is the field of complex number and the coefficient matrix is:

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





## Definition

The two systems of linear equations are **equivalent** if each equation in each system is a linear combination of the equations in other system.

## Theorem

Equivalent systems of linear equations have exactly the same solutions.

**Proof:**

## Note

- It is important to note that row operations are reversible. If two rows are interchanged, they can be returned to their original positions by another interchange.
- If the augmented matrices of two linear systems are row equivalent, then the two systems have the same solution set.



□ A system of linear equations has:

- No solution
  - Exactly one solution
  - Infinitely many solutions
- inconsistent
- } → consistent

Next session:

1. Is the system consistent? That is, does at least one solution exist?
2. If a solution exists, is it the only one? That is, is the solution unique?



- ❑ Different view of matrix multiplication
  - ❑ Linear combination and matrix multiplication
  - ❑ Associativity of three matrices multiplication
  - ❑ Gaussian Elimination
  - ❑ Row-equivalent of two matrices
  - ❑ Elementary matrices
  - ❑ System of linear equations
  - ❑ Equivalent systems of linear equations have exactly the same solutions.
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