

# *Linear Algebra*



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## 2

# Vector Spaces (向量空间)

## 2.2

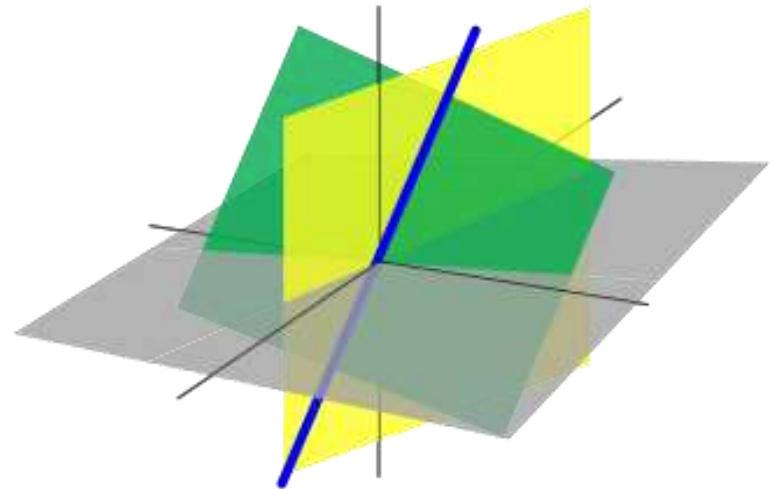
SOLVING  $Ax=0$  AND  $Ax=b$

(线性方程组的解)

Solving  $Ax=0$

Solving  $Ax=b$

Rank (秩)



# Introduction:

## A system of linear equations


$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2 \\ \cdots \quad \quad \quad \cdots \quad \quad \quad \cdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m \end{cases}$$

## 一般线性方程组的矩阵表示

$$Ax = b.$$

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix},$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, \quad b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}.$$

 **Solution**

## 齐次线性方程组的矩阵表示

*homogeneous*  $Ax = 0.$

## 解的结构:

在多解情况下, 讨论  
解与解之间的关系

## Introduction: Simple cases and more ...

- For a square invertible matrix  $A$ , there is only one solution to  $A\mathbf{x} = \mathbf{b}$ , and it is  $\mathbf{x} = A^{-1}\mathbf{b}$ .
- For a rectangular matrix  $A_{m \times n}$  ( $m \neq n$ ) or a square matrix without an inverse, there are new possibilities  
( $A \rightarrow$  echelon form  $U \rightarrow$  reduced echelon form  $R$ )

## Introduction: Simple cases and more ...

- For a square invertible matrix  $A$ :
  - The nullspace contains only  $\mathbf{x} = \mathbf{0}$ ; This zero solution is usually called the **trivial solution** (平凡解).  
(multiply  $A\mathbf{x} = \mathbf{0}$  by  $A^{-1}$ )
  - The column space is the whole space.  
( $A\mathbf{x} = \mathbf{b}$  has a solution for every  $\mathbf{b}$ )
- The new questions appear when the nullspace contains *more than the zero vector* (i.e., there exists a **nontrivial solution**) and/or the column space contains *less than all vectors*.

**For example,**

The matrix  $A = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}$  is not invertible.

$y + z = b_1$  and  $2y + 2z = b_2$  usually have no solution.

There is *no solution* unless  $b_2 = 2b_1$ . The column space of  $A$  contains only those  $\mathbf{b}$ 's, the multiples of  $(1, 2)^T$ .

When  $b_2 = 2b_1$  there are *infinitely many solutions*.

A particular solution to  $y + z = 2$  and  $2y + 2z = 4$  is  $\mathbf{x}_p = (1, 1)^T$ .

The nullspace of  $A$  contains  $(-1, 1)^T$  and all its multiples  $\mathbf{x}_n = (-c, c)^T$ ,  $c \in \mathbf{R}$ .

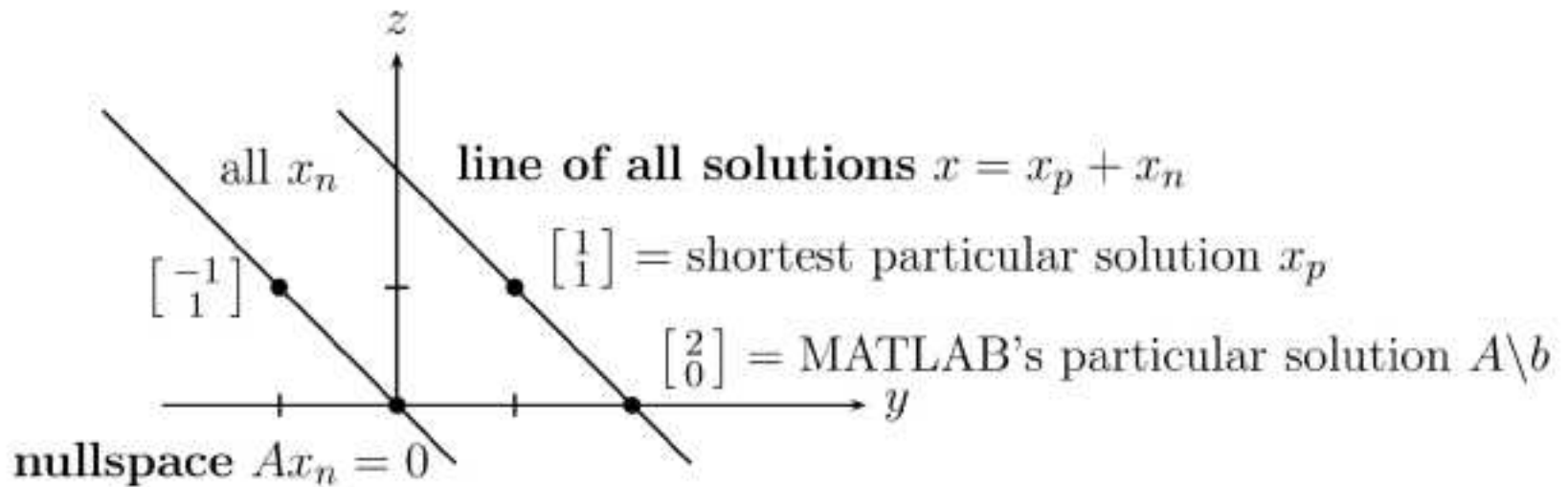
**Complete****solution to**

$$y + z = 2$$

$$2y + 2z = 4$$

is solved by  $\mathbf{x}_p + \mathbf{x}_n = \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 - c \\ 1 + c \end{bmatrix}$

**Complete solution:**  $A\mathbf{x}_p = \mathbf{b}$  and  $A\mathbf{x}_n = \mathbf{0}$  produce  $A(\mathbf{x}_p + \mathbf{x}_n) = \mathbf{b}$ .



## Complete

## solution to

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$$2y + 2z = 4$$

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**Complete solution:**  $A\mathbf{x}_p = \mathbf{b}$  and  $A\mathbf{x}_n = \mathbf{0}$  produce  $A(\mathbf{x}_p + \mathbf{x}_n) = \mathbf{b}$ .

# I. Solving $Ax = 0$

**Example 1** Find a spanning set for the nullspace of the matrix

$$A = \begin{bmatrix} -3 & 6 & -1 & 1 & -7 \\ 1 & -2 & 2 & 3 & -1 \\ 2 & -4 & 5 & 8 & -4 \end{bmatrix}.$$

**Solution:** The first step is to find the general solution of  $Ax=0$  in terms of free variables.

Row reduce the matrix  $A$  to *reduced echelon form* in order to write the basic variables in terms of the free variables:

$$\left[ \begin{array}{ccccc|c} 1 & -2 & 0 & -1 & 3 & 0 \\ 0 & 0 & 1 & 2 & -2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right], \quad \begin{array}{l} x_1 - 2x_2 - x_4 + 3x_5 = 0 \\ x_3 + 2x_4 - 2x_5 = 0 \\ 0 = 0 \end{array}$$

The general solution is  $x_1 = 2x_2 + x_4 - 3x_5$ ,  $x_3 = -2x_4 + 2x_5$ , with  $x_2$ ,  $x_4$ , and  $x_5$  free. ( $x_1, x_3$ : **basic variables**, also called **pivot variables**)



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Next, decompose the vector giving the general solution into a linear combination of *vectors where the weights are the free variables*. That is,

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} 2x_2 + x_4 - 3x_5 \\ x_2 \\ -2x_4 + 2x_5 \\ x_4 \\ x_5 \end{bmatrix} = x_2 \begin{bmatrix} 2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} 1 \\ 0 \\ -2 \\ 1 \\ 0 \end{bmatrix} + x_5 \begin{bmatrix} -3 \\ 0 \\ 2 \\ 0 \\ 1 \end{bmatrix} = x_2 \mathbf{u} + x_4 \mathbf{v} + x_5 \mathbf{w}$$

$\begin{bmatrix} 1 & -2 & 0 & -1 & 3 & 0 \\ 0 & 0 & 1 & 2 & -2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$

Every linear combination of  $\mathbf{u}$ ,  $\mathbf{v}$ , and  $\mathbf{w}$  is an element of  $N(A)$ . Thus  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is a spanning set for  $N(A)$ .

$\mathbf{u}, \mathbf{v}, \mathbf{w}$  : special solutions

The *nullspace* of  $\mathbf{A}$  can be spanned by a few *special solutions*, where a solution of  $\mathbf{Ax} = \mathbf{0}$  is called *special* if it is a solution for which each free variable takes value 1 or 0.

*The best way to find all solutions to  $\mathbf{Ax} = \mathbf{0}$  is from the special solutions:*

**Step 1.** After reaching  $\mathbf{Rx} = \mathbf{0}$ , identify the pivot variables (i.e., basic variables) and free variables. ( $\mathbf{R}$ : reduced echelon form of  $\mathbf{A}$ )

**Step 2.** Give one free variable the value 1, set the other free variables to 0, and solve  $\mathbf{Rx} = \mathbf{0}$  for the basic variables. This  $\mathbf{x}$  is a special solution.

**Step 3.** Every free variable produces its own “special solution” by step 2. The combinations of special solutions form the nullspace—all solutions to  $\mathbf{Ax} = \mathbf{0}$ .

## II. Solving $Ax = b$

- As observed above, all solutions of a homogeneous system of linear equations form a vector space (齐次线性方程组的解集构成一个向量空间). This enables us to write down the solutions in a nice way.
- However, solutions of a non-homogeneous system do not have such a nice property (非齐次线性方程组的解集不能构成向量空间) .
- A natural question is what we can say about solutions of a general system of linear equations.
- Consider the system of linear equations  $Ax = b$ ,  
and the homogeneous system  $Ax = 0$ .

**Lemma (引理)** *If  $u$  and  $w$  are two solutions of  $Ax = b$ , then  $u - w$  is a solution of  $Ax = 0$ .*

Let  $x_p$  be a **particular solution** of  $Ax = b$ . Then any solution  $x$  of  $Ax = b$  has the form

$$x_{\text{complete}} = x_{\text{nullspace}} + x_{\text{particular}} .$$

**Theorem** The solutions of a homogenous system  $A\mathbf{x} = \mathbf{0}$  form a subspace  $N(A)$ , and each solution  $\mathbf{x}$  of  $A\mathbf{x} = \mathbf{b}$  has the form

$$\mathbf{x} = \mathbf{x}_n + \mathbf{x}_p$$

where  $\mathbf{x}_p$  is a particular solution of  $A\mathbf{x} = \mathbf{b}$ , and  $\mathbf{x}_n \in N(A)$ .

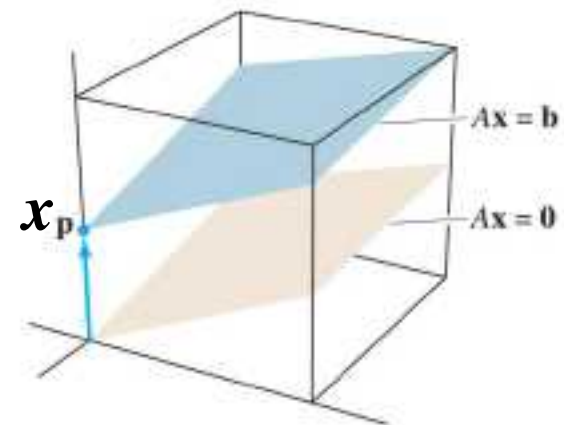
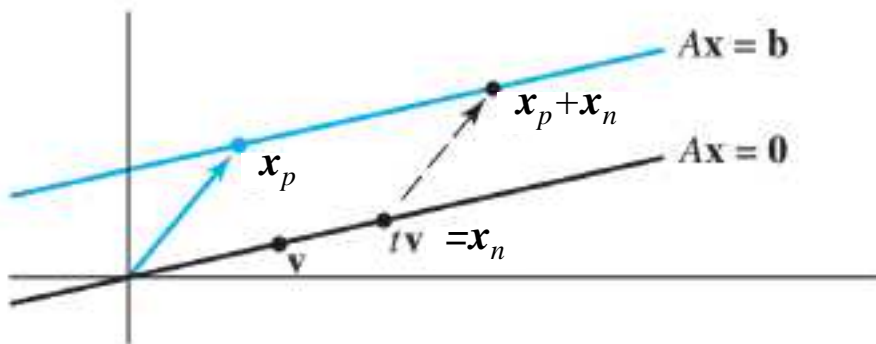
$A\mathbf{x}=\mathbf{0}$  的  
一般解  $\mathbf{x}_n$



$A\mathbf{x}=\mathbf{b}$  的  
特解  $\mathbf{x}_p$



$A\mathbf{x}=\mathbf{b}$  的  
一般解  $\mathbf{x}$



parallel solution sets of  $A\mathbf{x} = \mathbf{0}$  and  $A\mathbf{x} = \mathbf{b}$   
(Left: 1 free variable; Right: 2 free variables)

**Example 2** For the matrix  $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix}$ ,

the nullspace of  $A$  is the space spanned by the solutions of  $A\mathbf{x} = \mathbf{0}$ , which is

$$x = z, y = -2z.$$

So a solution vector is of the form

$$(x, y, z)^T = (z, -2z, z)^T = (1, -2, 1)^T z,$$

where  $z \in \mathbf{R}$ .

Consider the system of linear equations  $A\mathbf{x} = \mathbf{b}$ , where  $\mathbf{b} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ .

We observe that  $A\mathbf{x} = \mathbf{b}$  has a *particular solution*  $(0, 1, 0)^T$ .

Therefore, the solution set for  $A\mathbf{x} = \mathbf{b}$  is

$$\{(0, 1, 0)^T + (1, -2, 1)^T z \mid z \in \mathbf{R}\}.$$

$$\begin{bmatrix} 1 & 1 & 1 & | & 1 \\ 1 & 2 & 3 & | & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 1 & | & 1 \\ 0 & 1 & 2 & | & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -1 & | & 0 \\ 0 & 1 & 2 & | & 1 \end{bmatrix}$$

$$[A : \mathbf{b}] \rightarrow [U : \mathbf{c}] \rightarrow [R : \mathbf{d}]$$

$$\begin{aligned} x - z &= 0 \\ y + 2z &= 1 \end{aligned}$$

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} z + \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

*The best way to write all solutions to  $Ax = b$  :*

**Step 1.** Row reduce  $Ax = b$  to  $Ux = c$  or  $Rx = d$ .

**Step 2.** With free variables  $= 0$ , find a particular solution to  $Ax_p = b$  (or  $Ux_p = c$  or  $Rx_p = d$ ).

**Step 3.** Find the special solutions to  $Ax = 0$  (or  $Ux = 0$  or  $Rx = 0$ ). Each free variable, in turn, is 1. Then

$$x = x_p + (\text{any combination } x_n \text{ of special solutions}).$$

$A$ : coefficient matrix

$[A \ b]$ : augmented matrix

$U$ : the echelon form of  $A$

$R$ : the reduced echelon form of  $A$

**Example 3** Find the condition on  $b_1, b_2, b_3$  to have a solution;  
Solve  $A\mathbf{x} = \mathbf{b}$  when  $\mathbf{b} = (0,6,-6)^T$ .

$$1x_1 + 2x_2 + 3x_3 + 5x_4 = b_1$$

$$2x_1 + 4x_2 + 8x_3 + 12x_4 = b_2$$

$$3x_1 + 6x_2 + 7x_3 + 13x_4 = b_3$$

**Solution**

$$\left[ \begin{array}{cccc|c} 1 & 2 & 3 & 5 & b_1 \\ 2 & 4 & 8 & 12 & b_2 \\ 3 & 6 & 7 & 13 & b_3 \end{array} \right] \rightarrow \left[ \begin{array}{cccc|c} 1 & 2 & 3 & 5 & b_1 \\ 0 & 0 & 2 & 2 & b_2 - 2b_1 \\ 0 & 0 & -2 & -2 & b_3 - 3b_1 \end{array} \right]$$

$$\rightarrow \left[ \begin{array}{cccc|c} 1 & 2 & 3 & 5 & b_1 \\ 0 & 0 & 2 & 2 & b_2 - 2b_1 \\ 0 & 0 & 0 & 0 & b_3 + b_2 - 5b_1 \end{array} \right]$$

The last row shows the **solvability condition**  $b_3 + b_2 - 5b_1 = 0$ . (有解条件)

For  $\mathbf{b} = (0,6,-6)^T$ ,

$$\left[ U \quad c \right] = \left[ \begin{array}{cccc|c} 1 & 2 & 3 & 5 & 0 \\ 0 & 0 & 2 & 2 & 6 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

$$\begin{bmatrix} U & c \end{bmatrix} = \left[ \begin{array}{cccc|c} 1 & 2 & 3 & 5 & 0 \\ 0 & 0 & 2 & 2 & 6 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right] \longrightarrow \begin{bmatrix} R & d \end{bmatrix} = \left[ \begin{array}{cccc|c} \mathbf{1} & 2 & \mathbf{0} & 2 & -9 \\ 0 & 0 & \mathbf{1} & 1 & 3 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

pivot columns

***Everything is revealed by  $R\mathbf{x} = \mathbf{d}$***

The **special solutions to  $A\mathbf{x}=\mathbf{0}$**  have free variables  $x_2 = 1, x_4 = 0$  and  $x_2 = 0, x_4 = 1$ .

The **particular solution to  $A\mathbf{x}=\mathbf{b}$**  has free variables  $x_2 = 0, x_4 = 0$ .

The matrix with columns being special solutions is called the *nullspace matrix*:

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

Particular solution to  $A\mathbf{x}_p = \mathbf{b}$ :

$$\mathbf{x}_p = \begin{bmatrix} -9 \\ \mathbf{0} \\ 3 \\ \mathbf{0} \end{bmatrix}$$

The complete solution to  $A\mathbf{x} = \mathbf{b}$ :

$$\begin{aligned} \mathbf{x} &= \mathbf{x}_p + \mathbf{x}_n \\ &= \begin{bmatrix} -9 \\ 0 \\ 3 \\ 0 \end{bmatrix} + k_1 \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + k_2 \begin{bmatrix} -2 \\ 0 \\ -1 \\ 1 \end{bmatrix}, \\ &\text{where } k_1, k_2 \in \mathbf{R}. \end{aligned}$$



**Example 4** 设线性方程组 
$$\begin{cases} ax_1 + x_2 + x_3 = 4 \\ x_1 + bx_2 + x_3 = 3 \\ x_1 + 2bx_2 + x_3 = 4 \end{cases}$$

就参数  $a, b$  讨论方程组的解的情况, 有解时并求出解.

**解** 用初等行变换将增广矩阵化为阶梯阵.

$$\begin{aligned} & \left[ \begin{array}{ccc|c} a & 1 & 1 & 4 \\ 1 & b & 1 & 3 \\ 1 & 2b & 1 & 4 \end{array} \right] \xrightarrow{\text{row exchange}} \left[ \begin{array}{ccc|c} 1 & b & 1 & 3 \\ 1 & 2b & 1 & 4 \\ a & 1 & 1 & 4 \end{array} \right] \xrightarrow[r_3 - ar_1]{r_2 - r_1} \left[ \begin{array}{ccc|c} 1 & b & 1 & 3 \\ 0 & b & 0 & 1 \\ 0 & 1 - ab & 1 - a & 4 - 3a \end{array} \right] \\ & \xrightarrow{r_3 + ar_2} \left[ \begin{array}{ccc|c} 1 & b & 1 & 3 \\ 0 & b & 0 & 1 \\ 0 & 1 & 1 - a & 4 - 2a \end{array} \right] \xrightarrow[r_2 \leftrightarrow r_3]{r_2 - br_3} \left[ \begin{array}{ccc|c} 1 & b & 1 & 3 \\ 0 & 1 & 1 - a & 4 - 2a \\ 0 & 0 & (a - 1)b & 1 - 4b + 2ab \end{array} \right] \end{aligned}$$

$$\rightarrow \left[ \begin{array}{ccc|c} 1 & b & 1 & 3 \\ 0 & 1 & 1-a & 4-2a \\ 0 & 0 & (a-1)b & 1-4b+2ab \end{array} \right]$$

(1) 当  $(a-1)b \neq 0$  时, 有唯一解

$$x_1 = \frac{2b-1}{(a-1)b}, \quad x_2 = \frac{1}{b}, \quad x_3 = \frac{1-4b+2ab}{(a-1)b}$$

(2) 当  $a=1$ , 且  $1-4b+2ab=1-2b=0$ , 即  $b=1/2$  时, 有无穷多解.

化为  $\left[ \begin{array}{ccc|c} 1 & 1/2 & 1 & 3 \\ 0 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 \end{array} \right] \rightarrow \left[ \begin{array}{ccc|c} 1 & 0 & 1 & 2 \\ 0 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 \end{array} \right]$  于是方程组的一般解为

$$\mathbf{x} = (2, 2, 0)^T + k(-1, 0, 1)^T$$

( $k$  为任意常数) .

(3) 当  $a=1, b \neq 1/2$  时,  $1-4b+2ab \neq 0$ , 方程组无解.

(4) 当  $b=0$  时,  $1-4b+2ab = 1 \neq 0$  时, 方程组无解.

(原方程组中后两个方程是矛盾方程)

### III. The Rank of a Matrix (矩阵的秩)

Recall the method for solving systems of linear equations  $A\mathbf{x} = \mathbf{b}$  we learnt before:

Covert the *augmented matrix*  $[A \mid \mathbf{b}]$

into *row echelon form*  $[U \mid \mathbf{c}]$

or further convert it into *reduced row echelon form*  $[R \mid \mathbf{d}]$

Since elementary row operations do not change the solutions of the system (初等行变换不改变方程组的解), the three systems of linear equations

$$A\mathbf{x} = \mathbf{b}, U\mathbf{x} = \mathbf{c}, R\mathbf{x} = \mathbf{d}$$

have same solutions (同解).

We have also noticed that the number of free variables for  $A\mathbf{x} = \mathbf{b}$  depends on the number of non-zero rows of the row echelon form  $U$ .

This leads to an important parameter (参数) for a matrix.

**Definition 1** For a matrix  $A$ , let  $U$  be the row echelon form. Then the **rank of  $A$**  ( **$A$ 的秩**), denoted by  $\text{rank}(A)$ , is the number of non-zero rows of  $U$  (行阶梯形矩阵 $U$ 的非零行的行数).

*Obviously, the following properties hold.*

- The rank is not bigger than the number of rows.
- Suppose elimination reduces  $A\mathbf{x} = \mathbf{b}$  to  $U\mathbf{x} = \mathbf{c}$  and  $R\mathbf{x} = \mathbf{d}$ , with  $r$  pivot rows and  $r$  pivot columns. **The rank of those matrices is  $r$ .**

The last  $m-r$  rows of  $U$  and  $R$  are zero, so there is a solution only if the last  $m-r$  entries of  $\mathbf{c}$  and  $\mathbf{d}$  are also zero.

- The rank  $r$  is crucial. It counts the pivot rows in the “row space” and the pivot columns in the “column space”.

There are  $n - r$  special solutions in the nullspace.

There are  $m - r$  solvability conditions on  $\mathbf{b}$  or  $\mathbf{c}$  or  $\mathbf{d}$ .

Therefore, determining the rank of a given matrix is interesting for understanding the matrix.

To do this, we only need to *use elementary row operations to convert the matrix into row echelon form*.

**Example 5** Let

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

To find the rank of  $\mathbf{A}$ , we convert  $\mathbf{A}$  into row echelon form

$$\mathbf{U} = \begin{bmatrix} 1 & 1 & 2 \\ 0 & -1 & -1 \\ 0 & 0 & -4 \end{bmatrix}.$$

Thus  $\mathbf{A}$  has rank 3, called **full rank** (满秩).

It follows that for any  $\mathbf{b}$ , the system of linear equations  $\mathbf{Ax} = \mathbf{b}$  has a unique solution.

More properties hold for the rank of matrices:

$$\text{rank}(\mathbf{AB}) \leq \min\{\text{rank}(\mathbf{A}), \text{rank}(\mathbf{B})\}.$$

**Proof (hints)** Let  $\mathbf{A}$ ,  $\mathbf{B}$  be  $m \times n$  and  $n \times s$  matrices respectively. If we partition  $\mathbf{A}$  *by columns*:

$$\mathbf{AB} = [\alpha_1, \alpha_2, \dots, \alpha_n] \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1s} \\ b_{21} & b_{22} & \cdots & b_{2s} \\ \vdots & \vdots & & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{ns} \end{bmatrix} = \left[ \sum_{i=1}^n b_{i1} \alpha_i, \sum_{i=1}^n b_{i2} \alpha_i, \dots, \sum_{i=1}^n b_{is} \alpha_i \right]$$

Then every column of  $\mathbf{AB}$  is a combination of the column of  $\mathbf{A}$ .

Then the dimensions of the column spaces give:  $\text{rank}(\mathbf{AB}) \leq \text{rank}(\mathbf{A})$ .

Similarly, by partitioning  $\mathbf{B}$  *by rows*, we can prove that  $\text{rank}(\mathbf{AB}) \leq \text{rank}(\mathbf{B})$ .

对于线性方程组  $Ax=b$ , 下列命题等价:

- (1) 方程组有解(或相容);
- (2)  $b$ 可由 $A$ 的列向量组线性表示, 即  $b \in C(A)$ ;
- (3)  $\text{rank}([A \mid b]) = \text{rank}(A)$ , 即增广矩阵的秩等于系数矩阵的秩.

**Key words:** *free variables, basic variables (pivot variables), special solutions, particular solution, complete solution, rank*

## Homework

See Blackboard

