

1. (i) Use both the standard approach by setting up the vector equations, and some “shortcut” methods to determine whether the following sets are linearly independent.
- (ii) Which of these sets are bases for  $\mathbb{R}^3$ ?
- $S_1 = \{(1, 0, -1), (-1, 2, 3)\}$ .
  - $S_2 = \{(1, 0, -1), (-1, 2, 3), (0, 3, 0)\}$ .
  - $S_3 = \{(1, 0, -1), (-1, 2, 3), (0, 3, 3)\}$ .
  - $S_4 = \{(1, 0, -1), (-1, 2, 3), (0, 3, 0), (1, -1, 1)\}$ .

**Solution :**

**(i) Standard Approach**

(a) ① Set up the vector equation,

$$C_1(1, 0, -1) + C_2(-1, 2, 3) = (0, 0, 0)$$

② Convert them into linear system,

$$\begin{cases} 1 \cdot C_1 + (-1)C_2 = 0 \\ 0 \cdot C_1 + 2 \cdot C_2 = 0 \\ -1 \cdot C_1 + 3 \cdot C_2 = 0 \end{cases}$$

③ Write down the augmented matrix of this linear system,

$$\left( \begin{array}{cc|c} 1 & -1 & 0 \\ 0 & 2 & 0 \\ -1 & 3 & 0 \end{array} \right) \xrightarrow{\text{G.E.}} \left( \begin{array}{cc|c} 1 & -1 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{array} \right).$$

As the last column is non-pivot, and the rest columns are pivot, this homogeneous system has exactly one solution — trivial solution, i.e.  $C_1 = C_2 = 0$ .

Hence, the two vectors are linearly independent.

(b) ① Set up the vector equation,

$$C_1 \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} + C_2 \begin{pmatrix} -1 \\ 2 \\ 3 \end{pmatrix} + C_3 \begin{pmatrix} 0 \\ 3 \\ 0 \end{pmatrix} = 0$$

② write down the augmented matrix of this linear system directly,

$$\left( \begin{array}{ccc|c} 1 & -1 & 0 & 0 \\ 0 & 2 & 3 & 0 \\ -1 & 3 & 0 & 0 \end{array} \right) \xrightarrow{\text{G.E.}} \left( \begin{array}{ccc|c} 1 & -1 & 0 & 0 \\ 0 & 2 & 3 & 0 \\ 0 & 0 & -3 & 0 \end{array} \right).$$

As the last column is non-pivot, and the rest columns are pivot, this homogeneous system has exactly one solution — trivial solution, i.e.  $C_1 = C_2 = C_3 = 0$ .

Hence, the three vectors are linearly independent.

(c) ① Set up the vector equation,

$$C_1 \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} + C_2 \begin{pmatrix} -1 \\ 2 \\ 3 \end{pmatrix} + C_3 \begin{pmatrix} 0 \\ 3 \\ 3 \end{pmatrix} = 0$$

② write down the augmented matrix of this linear system directly,

$$\left( \begin{array}{ccc|c} 1 & -1 & 0 & 0 \\ 0 & 2 & 3 & 0 \\ -1 & 3 & 3 & 0 \end{array} \right) \xrightarrow{\text{G.E.}} \left( \begin{array}{ccc|c} 1 & -1 & 0 & 0 \\ 0 & 2 & 3 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right).$$

As the last column is non-pivot, and the third column is also non-pivot, this homogeneous system has non-trivial solutions, i.e.  $C_1, C_2, C_3$  are not all zero.

Hence, the three vectors are linearly dependent.

(d) ① Set up the vector equation,

$$C_1 \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} + C_2 \begin{pmatrix} -1 \\ 2 \\ 3 \end{pmatrix} + C_3 \begin{pmatrix} 0 \\ 3 \\ 0 \end{pmatrix} + C_4 \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} = 0$$

② write down the augmented matrix of this linear system directly,

$$\left( \begin{array}{cccc|c} 1 & -1 & 0 & 1 & 0 \\ 0 & 2 & 3 & -1 & 0 \\ -1 & 3 & 0 & 1 & 0 \end{array} \right) \xrightarrow{\text{G.E.}} \left( \begin{array}{cccc|c} 1 & -1 & 0 & 1 & 0 \\ 0 & 2 & 3 & -1 & 0 \\ 0 & 0 & -3 & 3 & 0 \end{array} \right).$$

As the last column is non-pivot, and the fourth column is also non-pivot, this homogeneous system has non-trivial solutions, i.e.  $C_1, C_2, C_3, C_4$  are not all zero.

Hence, the four vectors are linearly dependent.

### "Shortcut" methods

- (a)  $S_1 = \{(1, 0, -1), (-1, 2, 3)\}$ .
- (b)  $S_2 = \{(1, 0, -1), (-1, 2, 3), (0, 3, 0)\}$ .
- (c)  $S_3 = \{(\underline{1, 0, -1}), (\underline{-1, 2, 3}), (\underline{0, 3, 3})\}$ .
- (d)  $S_4 = \{(1, 0, -1), (-1, 2, 3), (0, 3, 0), (1, -1, 1)\}$ .

(a). The two vectors  $(1, 0, -1)$ ,  $(-1, 2, 3)$  are not scalar multiples of each other,

So the 2 vectors are linearly independent.

(b). We can form the  $3 \times 3$  matrix  $A$ ,  $A = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 2 & 3 \\ -1 & 3 & 0 \end{pmatrix}$

$$\det(A) = 1 \times \begin{vmatrix} 2 & 3 \\ 3 & 0 \end{vmatrix} - (-1) \times \begin{vmatrix} 0 & 3 \\ -1 & 0 \end{vmatrix} + 0 \\ = 1 \times (-9) + 1 \times 3 = -6 \neq 0$$

So the three column vectors are linearly independent.

(C).

Method 1. By inspection,

$$\frac{3}{2}u_1 + \frac{3}{2}u_2 = \frac{3}{2}(u_1 + u_2) = \frac{3}{2} \cdot \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \\ 3 \end{pmatrix} = u_3.$$

So the three column vectors are linearly dependent.

Method 2. We can form the  $3 \times 3$  matrix  $B$ ,  $B = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 2 & 3 \\ -1 & 3 & 3 \end{pmatrix}$

$$\det(B) = 1 \times \begin{vmatrix} 2 & 3 \\ 3 & 3 \end{vmatrix} - (-1) \times \begin{vmatrix} 0 & 3 \\ -1 & 3 \end{vmatrix} + 0 \\ = 1 \times (-3) + 1 \times 3 + 0 = 0$$

So the three column vectors are linearly dependent.

(d). For the given four vectors, all vectors are from  $\mathbb{R}^3$ .

So these 4 vectors are linearly dependent.

(ii) Which of these sets are bases for  $\mathbb{R}^3$ ?

- (a)  $S_1 = \{(1, 0, -1), (-1, 2, 3)\}$ .
- (b)  $S_2 = \{(1, 0, -1), (-1, 2, 3), (0, 3, 0)\}$ .
- (c)  $S_3 = \{(1, 0, -1), (-1, 2, 3), (0, 3, 3)\}$ .
- (d)  $S_4 = \{(1, 0, -1), (-1, 2, 3), (0, 3, 0), (1, -1, 1)\}$ .

(iii) Since  $\dim \mathbb{R}^3 = 3$ ,

any basis for  $\mathbb{R}^3$  must contain 3 vectors.

And in fact, any three vectors in  $\mathbb{R}^3$  that are linearly independent will form a basis for  $\mathbb{R}^3$ .

As you see, (b) is a set of 3 linearly independent vectors.

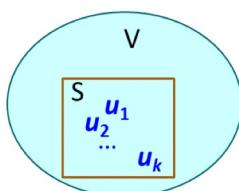
So (b) is indeed a basis for  $\mathbb{R}^3$ .

## 5. Basis and Dimension (slide 3)

Let  $S = \{u_1, u_2, \dots, u_k\}$  be a subset of a vector space  $V$ . (Recall that a vector space can either be  $\mathbb{R}^n$  or any subspace of the  $\mathbb{R}^n$ .) Then

$S$  is called a basis for  $V$  if

1.  $\text{span}\{u_1, u_2, \dots, u_k\} = V$ ; and
2.  $u_1, u_2, \dots, u_k$  are linearly independent.



In other words,  $S$  is a smallest possible set of vectors in  $V$  that generate every vector in  $V$ .

The number of vectors in a basis for  $V$  is called the dimension of  $V$  and is denoted by  $\dim V$ . In other words, if  $\{u_1, u_2, \dots, u_k\}$  is a basis for  $V$ , then  $\dim V = k$ . It is the smallest possible number of vectors that can generate  $V$ .

## ★ Remark

### 1. Linear Independence and Dependence (slide 4)

Let  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$  be a set of vectors in  $\mathbb{R}^n$ .

Set up the vector equation

$$c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \cdots + c_k\mathbf{u}_k = \mathbf{0}.$$

This is equivalent to a homogeneous linear system with variables  $c_1, c_2, \dots, c_k$ . (See examples below on vector equation form of linear system.)

Recall that a homogeneous system has either only the trivial solution (case I) or infinitely many solutions with non-trivial solutions (case II).

**Case I:** The system has only the trivial solution,  
i.e. the only possible scalars are:  $c_1 = 0, c_2 = 0, \dots, c_k = 0$

In this case, we say the set of vectors  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$  are linearly independent.

**Case II:** The system has non-trivial solutions,  
i.e. there are scalars  $c_1, c_2, \dots, c_k$  that are not all zero.

In this case, we say the set  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$  are linearly dependent.

2. In general, when there are  $n$  vectors  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  in  $\mathbb{R}^n$ , we can form an  $n \times n$  matrix  $\mathbf{A}$  using  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  (written in column form) as the  $n$  columns of the matrix.

- If  $\det(\mathbf{A}) = 0$ , then  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  are linearly dependent
- If  $\det(\mathbf{A}) \neq 0$ , then  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  are linearly independent.

### 3. Linear Independence of Two Vectors (slide 9)

Let  $\mathbf{u}, \mathbf{v}$  be a set with two vectors in  $\mathbb{R}^n$ .

This gives us a very simple way to decide whether two vectors are linearly independent by inspecting their components:

- If  $\mathbf{u}$  and  $\mathbf{v}$  are scalar multiples of each other, they are linearly dependent;
- If  $\mathbf{u}$  and  $\mathbf{v}$  are not scalar multiples of each other, they are linearly independent.

### 4. A Condition for Linear Dependence (slide 10)

Let  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$  be a set of  $k$  vectors in  $\mathbb{R}^n$ .

If  $k > n$ , then the set of vectors are linearly dependent.

2. Find a basis for and the dimension of the solution space of each of the following homogeneous systems.

$$(a) \begin{cases} x_1 + 3x_2 - x_3 + 2x_4 = 0 \\ -3x_2 + x_3 = 0. \end{cases}$$

$$(b) \begin{cases} x_1 + 3x_2 - x_3 + 2x_4 = 0 \\ -3x_2 + x_3 = 0 \\ x_1 - x_4 = 0. \end{cases}$$

**Solution:**

(a). the augmented matrix of the homogeneous system is

$$\left( \begin{array}{cccc|c} 1 & 3 & -1 & 2 & 0 \\ 0 & -3 & 1 & 0 & 0 \end{array} \right), \text{ which is already in REF.}$$

As the last column, and the 3rd, 4th column of REF are non-pivot columns, there are infinitely many solutions for the system.

We set  $x_3 = s$ ,  $x_4 = t$ . where  $s$  and  $t$  are arbitrary parameters.

Then we perform the back substitution to get the general solution,

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} -2t \\ \frac{s}{3} \\ s \\ t \end{pmatrix}$$

$$= \begin{pmatrix} 0 \\ \frac{s}{3} \\ s \\ 0 \end{pmatrix} + \begin{pmatrix} -2t \\ 0 \\ 0 \\ t \end{pmatrix} = s \cdot \begin{pmatrix} 0 \\ \frac{1}{3} \\ 1 \\ 0 \end{pmatrix} + t \cdot \begin{pmatrix} -2 \\ 0 \\ 0 \\ 1 \end{pmatrix} \quad \text{where } s \in \mathbb{R}, t \in \mathbb{R}.$$

So  $\{(0, \frac{1}{3}, 1, 0), (-2, 0, 0, 1)\}$  is a basis for the solution space, and the dimension of the solution space is 2.

(b). The augmented matrix of this homogeneous system is

$$\left( \begin{array}{cccc|c} 1 & 3 & -1 & 2 & 0 \\ 0 & -3 & 1 & 0 & 0 \\ 1 & 0 & 0 & -1 & 0 \end{array} \right) \xrightarrow{G.E.} \left( \begin{array}{cccc|c} 1 & 3 & -1 & 2 & 0 \\ 0 & -3 & 1 & 0 & 0 \\ 0 & 0 & 0 & -3 & 0 \end{array} \right)$$

As the last column , and the 3rd column of REF are non-pivot columns , there are infinitely many solutions for the system .

We set  $x_3 = t$  , where  $t$  is an arbitrary parameter .

We perform the back substitution to obtain the general solution ,

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 0 \\ \frac{t}{3} \\ t \\ 0 \end{pmatrix}$$

$$= t \cdot \begin{pmatrix} 0 \\ \frac{1}{3} \\ 1 \\ 0 \end{pmatrix} \quad \text{where } t \in R.$$

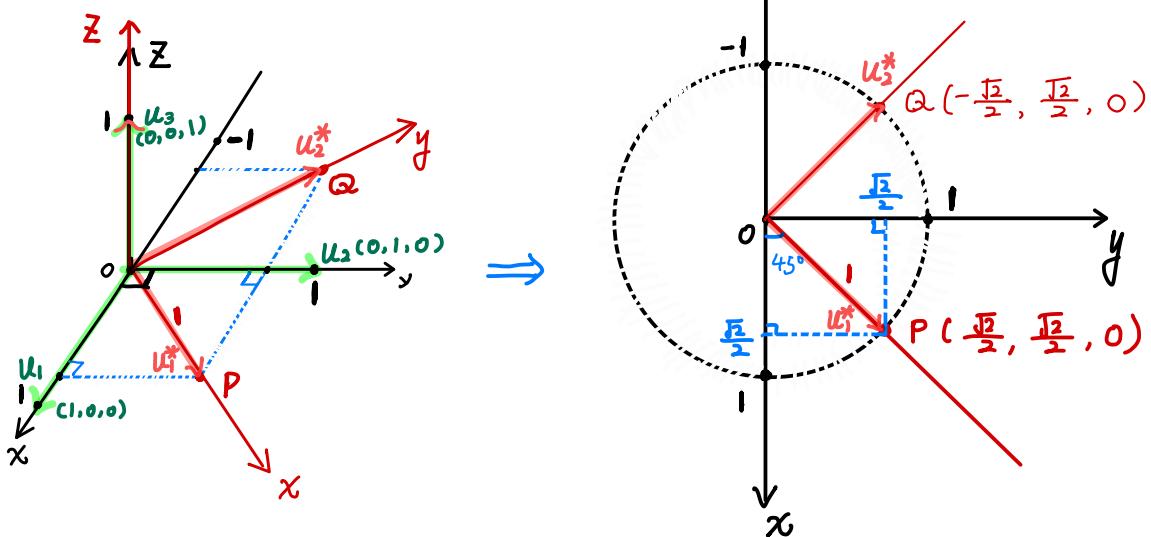
Hence ,  $\{(0, \frac{1}{3}, 1, 0)\}$  is a basis for the solution space , and the dimension of the solution space is 1 .

3. The regular coordinate axes for the  $xyz$ -space is rotated about the  $z$ -axis through  $45^\circ$  counterclockwise, viewing from the positive  $z$ -axis.

- Find a basis  $S$  consisting of unit vectors that determine the new coordinate axes.
- What are the coordinates of the vector  $\mathbf{v} = (1, 1, 1)$  relative to the new axes?
- Find a matrix  $M$  such that  $M\mathbf{v} = (\mathbf{v})_S$ .

**Solution:**

(a).



The unit vector  $(1, 0, 0)$  will be rotated to  $(\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}, 0)$   $u_1^*$

The unit vector  $(0, 1, 0)$  will be rotated to  $(-\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}, 0)$   $u_2^*$

The unit vector  $(0, 0, 1)$  remains unchanged, still  $(0, 0, 1)$ .  $u_3$

Hence, the basis  $S$  determining the new coordinate axes is

$$S = \left\{ \left( \frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}, 0 \right), \left( -\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}, 0 \right), (0, 0, 1) \right\}$$

(b). we need to find  $a, b, c$  such that

$$a \cdot \left( \frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}, 0 \right) + b \cdot \left( -\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}, 0 \right) + c \cdot (0, 0, 1) = (1, 1, 1) \quad (***)$$

That is

$$\begin{cases} \frac{\sqrt{2}}{2}a - \frac{\sqrt{2}}{2}b = 1 \\ \frac{\sqrt{2}}{2}a + \frac{\sqrt{2}}{2}b = 1 \\ 1 \cdot c = 1 \end{cases} \xrightarrow{\text{Substitution}} \begin{cases} a = \sqrt{2} \\ b = 0 \quad (*) \\ c = 1 \end{cases}$$

Hence, the coordinates of the vector  $(1, 1, 1)$  relative to the new axes are  $(\sqrt{2}, 0, 1)$ , i.e.  $(\mathbf{v})_S = (\sqrt{2}, 0, 1)$

(CC). From part (b),

the coordinate vector relative to the basis S is  $(\mathbf{v})_S = (\sqrt{2}, 0, 1)$ ,

Substitute  $(\mathbf{x})$  into  $(**)$ , and write the vectors in the column form,

$$\begin{pmatrix} \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \\ 0 \end{pmatrix} \cdot \sqrt{2} + \begin{pmatrix} -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \\ 0 \end{pmatrix} \cdot 0 + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \cdot 1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

Which can be further expressed in the matrix form,

$$\underbrace{\begin{pmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} & 0 \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \\ 0 & 0 & 1 \end{pmatrix}}_A \cdot \underbrace{\begin{pmatrix} \sqrt{2} \\ 0 \\ 1 \end{pmatrix}}_{(\mathbf{v})_S} = \underbrace{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}}_{\mathbf{v}} \Rightarrow A \cdot (\mathbf{v})_S = \mathbf{v}$$

$$\left( \begin{array}{ccc|ccc} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} & 0 & 1 & 0 & 0 \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \xrightarrow{R_1 \times (-1) + R_2} \left( \begin{array}{ccc|ccc} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} & 0 & 1 & 0 & 0 \\ 0 & \sqrt{2} & 0 & -1 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right)$$

$\overset{A}{\cancel{R}}$

$$\xrightarrow{R_2 \times \frac{1}{\sqrt{2}}} \left( \begin{array}{ccc|ccc} 1 & -1 & 0 & \sqrt{2} & 0 & 0 \\ 0 & 1 & 0 & -\frac{1}{2} & \frac{1}{2} & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \xrightarrow{R_2 + R_3} \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \\ 0 & 1 & 0 & -\frac{1}{2} & \frac{1}{2} & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \overset{A^{-1}}{\cancel{R}}$$

Pre-multiply  $A^{-1}$  to  $(\mathbf{v})_S$  to move it to the right-hand side, and

we get  $(\mathbf{v})_S = A^{-1}\mathbf{v}$ .

We require  $(\mathbf{v})_S = M\mathbf{v}$ .

So we have the matrix  $M$ ,

$$M = A^{-1} = \underbrace{\begin{pmatrix} \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \\ -\frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \\ 0 & 0 & 1 \end{pmatrix}}$$

4. Let  $A = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ -1 & 0 & 1 \\ 1 & 1 & 1 \end{pmatrix}$ ,  $x = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$  and  $b = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}$ .

- (a) Show that the linear system  $Ax = b$  has no solution.
- (b) Find the least squares solution to  $Ax = b$ .
- (c) Find the projection of  $b$  onto  $\text{span}\{(1, 0, -1, 1), (-1, 1, 0, 1), (0, -1, 1, 1)\}$ .
- (d) Use (c) to find a vector that is orthogonal to all the three vectors:  
 $\underbrace{(1, 0, -1, 1)}_{u_1}, \underbrace{(-1, 1, 0, 1)}_{u_2}, \underbrace{(0, -1, 1, 1)}_{u_3}$

**Solution :**

(a).  $(A | b) = \left( \begin{array}{ccc|c} 1 & -1 & 0 & 1 \\ 0 & 1 & -1 & 1 \\ -1 & 0 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{array} \right) \xrightarrow{\text{G.E.}} \left( \begin{array}{ccc|c} 1 & -1 & 0 & 1 \\ 0 & 1 & -1 & 1 \\ 0 & 0 & 3 & -2 \\ 0 & 0 & 0 & 3 \end{array} \right)$

Since the last column in REF is pivot, this linear system has no solution.

(b). Step 1. We start with the system  $Ax=b$ .

$$\underbrace{\begin{pmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ -1 & 0 & 1 \\ 1 & 1 & 1 \end{pmatrix}}_A \underbrace{\begin{pmatrix} x \\ y \\ z \end{pmatrix}}_X = \underbrace{\begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}}_b \Rightarrow Ax = b$$

Step 2. Form the new linear system  $A^T A x = A^T b$ ,

$$A^T = \begin{pmatrix} 1 & 0 & -1 & 1 \\ -1 & 1 & 0 & 1 \\ 0 & -1 & 1 & 1 \end{pmatrix}, \quad A^T A = \begin{pmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{pmatrix}, \quad A^T b = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

which is given by  $\begin{pmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{pmatrix} \cdot \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$

Step 3. Solve the above system to get its solution  $u$ ,

$$u = \begin{pmatrix} \frac{1}{3} \\ \frac{1}{3} \\ \frac{1}{3} \end{pmatrix}$$

Hence, this vector  $u$  is the least squares solution of  $Ax=b$ .

**Remark: Matlab**

```
Command Window
>> A=[1 -1 0; 0 1 -1; -1 0 1; 1 1 1];
b=[1;1;1;1];
rref([A'*A A'*b])

ans =
1.0000      0      0    0.3333
0    1.0000      0    0.3333
0      0    1.0000    0.3333
```

(C).  $\text{span}\{u_1, u_2, u_3\} = \text{span}\{u_1^T, u_2^T, u_3^T\}$  = the column space of  $A$ ,  
 So the required projection  $p$  is given by  $Au$ , where  $u$  is the least squares solution in part (b), which is

$$p = Au = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ -1 & 0 & 1 \\ 1 & 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} \frac{1}{3} \\ \frac{1}{3} \\ \frac{1}{3} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

(d) Since  $p$  is the projection of  $b$  onto the subspace  $= \text{span}\{u_1, u_2, u_3\}$ ,  
 then  $b-p$  is orthogonal to this subspace, which is,

$$b-p = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} - \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 0 \end{pmatrix}$$

As  $u_1, u_2, u_3 \in \text{span}\{u_1, u_2, u_3\}$ ,

this vector is orthogonal to the three vectors  $u_1, u_2, u_3$

## ★ Remark

### /. Finding Least Squares Solutions (slide 11)

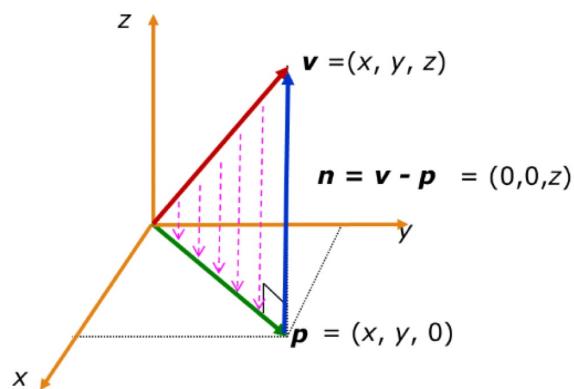
- Starting with the system  $\mathbf{Ax} = \mathbf{b}$  (1)
- Form the new linear system  $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$  (2)
- Solve the system (2)
- A solution of (2) gives a least squares solution of (1)

System (2)  $\mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$  is always consistent. It can have exactly one solution, or infinitely many solutions.

### 2. $\mathbf{Au} = \mathbf{p}$ where $\mathbf{p}$ is the projection of $\mathbf{b}$ onto the column space\* of $\mathbf{A}$ .

### 3. Projection of a Vector onto a Plane in $\mathbb{R}^3$ (slide 2)

For a 3-vector  $(x \ y \ z)$  in  $\mathbb{R}^3$ , its projection onto the xy-plane is given geometrically as follow.



From this observation, we have:

$\mathbf{p}$  is a **projection** of  $\mathbf{v}$  onto  
the plane



$\mathbf{v} - \mathbf{p}$  is **orthogonal** to  
the plane

5. (MATLAB) To measure the takeoff performance of an airplane, the horizontal position of the plane was measured every second, from  $t = 0$  to  $t = 12$ . The positions (in meter) were: 0, 2.9, 9.8, 20.1, 34.3, 52.8, 74.0, 98.5, 126.7, 157.2, 190.3, 228.6, and 269.0.

- Find the least squares cubic curve:  $y = a + bt + ct^2 + dt^3$  for these data.
- Use the result of part (a) to estimate the velocity of the plane when  $t = 4.5$  second.

**Solution:**

(a). Substituting the 13 data pairs of  $(t, y)$  into the equation  $y = a + bt + ct^2 + dt^3$ ,

$$\left\{ \begin{array}{l} a = 0 \\ a + b + c + d = 2.9 \\ a + 2b + 4c + 8d = 9.8 \\ a + 3b + 9c + 27d = 20.1 \\ a + 4b + 16c + 64d = 34.3 \\ a + 5b + 25c + 125d = 52.8 \\ a + 6b + 36c + 216d = 74.0 \\ a + 7b + 49c + 343d = 98.5 \\ a + 8b + 64c + 512d = 126.7 \\ a + 9b + 81c + 729d = 157.2 \\ a + 10b + 100c + 1000d = 190.3 \\ a + 11b + 121c + 1331d = 228.6 \\ a + 12b + 144c + 1728d = 269.0 \end{array} \right.$$

$$\Rightarrow AX = b$$

where  $A$  denotes the coefficient matrix,  
 $b$  denotes the constant matrix,  
and  $X = \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix}$

Command Window

```
>> A=[1 0 0 0;1 1 1 1;1 2 4 8;1 3 9 27;1 4 16 64;1 5 25 125;1 6 36 216;1 7 49 343;1 8 64 512;
1 9 81 729;1 10 100 1000;1 11 121 1331;1 12 144 1728];
>> A
```

A =

1	0	0	0
1	1	1	1
1	2	4	8
1	3	9	27
1	4	16	64
1	5	25	125
1	6	36	216
1	7	49	343
1	8	64	512
1	9	81	729
1	10	100	1000
1	11	121	1331
1	12	144	1728

```
>> rref([A'*A A'*b])
```

ans =

$$\begin{array}{ccccc} 1.0000 & 0 & 0 & 0 & -0.1758 \\ 0 & 1.0000 & 0 & 0 & 1.1716 \\ 0 & 0 & 1.0000 & 0 & 1.9453 \\ 0 & 0 & 0 & 1.0000 & -0.0147 \end{array}$$

This will give us the least squares solution  $a = -0.1758$ ,  $b = 1.1716$   
 $c = 1.9453$ ,  $d = -0.0147$ .

Hence, the least squares cubic curve is

$$y = -0.1758 + 1.1716t + 1.9453t^2 - 0.0147t^3.$$

(b).  $\frac{dy}{dt} = 1.1716 + 1.9453 \times 2t - 0.0147 \times 3t^2$

when  $t = 4.5$ ,  $\frac{dy}{dt} = 17.9453 \approx 17.79 \text{ m/s}$ )

So the velocity at 4.5 second is given by 17.79 m/s.

6.

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \\ -1 & 3 & 6 \\ 2 & 1 & 0 \\ 3 & 1 & -1 \end{pmatrix}, \quad B = \begin{pmatrix} 2 & 1 & 4 & 1 & 2 \\ 4 & 2 & 2 & 3 & 2 \\ 2 & 1 & -2 & 2 & 0 \\ 6 & 3 & 6 & 4 & 4 \end{pmatrix}, \quad C = \begin{pmatrix} 1 & 4 & 5 & 8 \\ -1 & 4 & 3 & 0 \\ 2 & 0 & 2 & 1 \end{pmatrix}.$$

For each matrix, find

- (i) a basis for the row space and a basis for the column space;
- (ii) a basis for the nullspace;
- (iii) the rank and nullity of the matrix, and verify the Dimension Theorem for Matrices.

**Solution :**

In general,

the non-zero rows in the r.e.f.  $R$  of a matrix  $A \rightarrow$  a basis for the row space of  $A$ .

the columns of  $A$  corresponding to pivot columns of  $R \rightarrow$  a basis for the column space of  $A$ .

For  $A$ , perform Gaussian Elimination,

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \\ -1 & 3 & 6 \\ 2 & 1 & 0 \\ 3 & 1 & -1 \end{pmatrix} \xrightarrow{\text{G.E.}} \underbrace{\begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_{\text{REF of } A}$$

(i)  $\{(1, 2, 0), (0, 1, 1), (0, 0, 1)\}$  is a basis for the row space.

Since the pivot columns in REF of  $A$  are 1st, 2nd and 3rd column.

We will look at the corresponding three columns of the original matrix  $A$ .

So  $\{(1, 0, -1, 2, 3)^T, (2, 1, 3, 1, 1)^T, (0, 1, 6, 0, -1)^T\}$  is a basis for the column space.

(Note: here  $(a, b, c, d, e)^T$  refer to the column form of the vector.)

(ii) Recall that the nullspace of  $A$  = the solution space of  $Ax = 0$ ,

Since every column is pivot in REF of  $A$ ,

then this system  $Ax=0$  has only trivial solution  $x=0$ .

As the vectors in a basis for a vector space must be non-zero,  
then there's no basis for this solution space.

So the basis for the nullspace is the empty set.

(iii).

Recall  $\text{nullity}(A) = \dim(\text{nullspace of } A)$

= the number of vectors in a basis for "nullspace of  $A$ "

So  $\text{nullity}(A) = 0$

Recall "the Dimension Theorem for Matrices"

If  $A$  is a matrix with  $n$  columns, then  $\text{rank}(A) + \text{nullity}(A) = n$ .

Since  $LHS = 3 + 0 = 3$ ,

$RHS = 3 = LHS$ .

Hence, verified.

For  $\mathbf{B} = \begin{pmatrix} 2 & 1 & 4 & 1 & 2 \\ 4 & 2 & 2 & 3 & 2 \\ 2 & 1 & -2 & 2 & 0 \\ 6 & 3 & 6 & 4 & 4 \end{pmatrix}$ , the row echelon form is  $\begin{pmatrix} 2 & 1 & 4 & 1 & 2 \\ 0 & 0 & -6 & 1 & -2 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$ .

- (i)  $\{(2, 1, 4, 1, 2), (0, 0, -6, 1, -2)\}$  is a basis for the row space.  
 $\{(2, 4, 2, 6)^T, (4, 2, -2, 6)^T\}$  is a basis for the column space.

(Using reduced row-echelon form:  $\begin{pmatrix} 1 & \frac{1}{2} & 0 & \frac{5}{6} & \frac{1}{3} \\ 0 & 0 & 1 & -\frac{1}{6} & \frac{1}{3} \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$ , basis for the row space is given by  $\{(1, \frac{1}{2}, 0, \frac{5}{6}, \frac{1}{3}), (0, 0, 1, -\frac{1}{6}, \frac{1}{3})\}$ .

- (ii) This is the same as finding basis for the solution space of the homogeneous system  $\mathbf{B}\mathbf{x} = \mathbf{0}$ .  
 $\{(-\frac{1}{2}, 1, 0, 0, 0)^T, (-\frac{5}{6}, 0, \frac{1}{6}, 1, 0)^T, (-\frac{1}{3}, 0, -\frac{1}{3}, 0, 1)^T\}$  is the basis for the nullspace.
- (iii)  $\text{rank}(\mathbf{B}) = 2$ ,  $\text{nullity}(\mathbf{B}) = 3$ .  
 $\text{rank}(\mathbf{B}) + \text{nullity}(\mathbf{B}) = 5 = \text{number of columns of } \mathbf{B}$ .

For  $\mathbf{C} = \begin{pmatrix} 1 & 4 & 5 & 8 \\ -1 & 4 & 3 & 0 \\ 2 & 0 & 2 & 1 \end{pmatrix}$ , the row echelon form is  $\begin{pmatrix} 1 & 4 & 5 & 8 \\ 0 & 8 & 8 & 8 \\ 0 & 0 & 0 & -7 \end{pmatrix}$ .

- (i)  $\{(1, 4, 5, 8), (0, 8, 8, 8), (0, 0, 0, -7)\}$  is a basis for the row space.  
 $\{(1, -1, 2)^T, (4, 4, 0)^T, (8, 0, 1)^T\}$  is a basis for the column space.

Using reduced row-echelon form:  $\begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$ .

the basis for row space is given by  $\{(1, 0, 1, 0), (0, 1, 1, 0), (0, 0, 0, 1)\}$ .

- (ii) This is the same as finding basis for the solution space of the homogeneous system  $\mathbf{C}\mathbf{x} = \mathbf{0}$ .  
 $\{(-1, -1, 1, 0)^T\}$  is a basis for the nullspace.
- (iii)  $\text{rank}(\mathbf{C}) = 3$ ,  $\text{nullity}(\mathbf{C}) = 1$ .  
 $\text{rank}(\mathbf{C}) + \text{nullity}(\mathbf{C}) = 4 = \text{number of columns of } \mathbf{C}$ .

7. Given two  $3 \times 4$  matrices  $\mathbf{A}$  and  $\mathbf{B}$  with respective row echelon forms

$$\mathbf{R}_A = \begin{pmatrix} 1 & 0 & 1 & 2 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad \text{and} \quad \mathbf{R}_B = \begin{pmatrix} 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

Do we have enough information to find the following? Justify your answers.

- (i) Matrices  $\mathbf{A}$  and  $\mathbf{B}$ .
- (ii) The row spaces of  $\mathbf{A}$  and  $\mathbf{B}$ .
- (iii) The column spaces of  $\mathbf{A}$  and  $\mathbf{B}$ .

Write down a basis for each of the row spaces and column spaces of  $\mathbf{A}$  and  $\mathbf{B}$  if possible.

**Solution:**

(i) No. As we do not know what elementary row operations performed on  $\mathbf{A}$  and  $\mathbf{B}$  to get the row echelon forms  $\mathbf{R}_A$  and  $\mathbf{R}_B$ , then we can not find the original  $\mathbf{A}$  and  $\mathbf{B}$  from  $\mathbf{R}_A$  and  $\mathbf{R}_B$  respectively.

(ii). **Recall:**

In general,

the non-zero rows in the r.e.f.  $\mathbf{R}$  of a matrix  $\mathbf{A} \rightarrow$  a basis for the row space of  $\mathbf{A}$ .

Hence,

a basis for the row space of  $\mathbf{A} = \{(1 \ 0 \ 1 \ 2), (0 \ 0 \ 1 \ 1)\}$

row space of  $\mathbf{A} = \text{span}\{(1 \ 0 \ 1 \ 2), (0 \ 0 \ 1 \ 1)\}$

a basis for the row space of  $\mathbf{B} = \{(0 \ 1 \ 0 \ 1), (0 \ 0 \ 1 \ -1), (0 \ 0 \ 0 \ 1)\}$

row space of  $\mathbf{B} = \text{span}\{(0 \ 1 \ 0 \ 1), (0 \ 0 \ 1 \ -1), (0 \ 0 \ 0 \ 1)\}$

(iii). **Recall:**

the columns of  $\mathbf{A}$  corresponding to pivot columns of  $\mathbf{R} \rightarrow$  a basis for the column space of  $\mathbf{A}$ .

In  $R_A$ , the 1st and 3rd columns are pivot, so  
 a basis of the column space of  $A = \{1\text{st column of } A, 3\text{rd column of } A\}$ .  
 Since we do not know  $A$ , we can not find a basis of the column space of  $A$ , and hence we can not get the column space of  $A$  either.

In  $R_B$ , the 2nd, 3rd and 4th columns are pivot, so

a basis of the column space of  $B$   
 $= \{2\text{nd column of } B, 3\text{rd column of } B, 4\text{th column of } B\}$ .  
 $= \{3 \text{ linearly independent vectors in } \mathbb{R}^3\}$

We know, the span of any 3 linearly independent vectors in  $\mathbb{R}^3$   
 $= \mathbb{R}^3$ .

Hence, the column space of  $B = \mathbb{R}^3$ .

Therefore, we can use the standard basis of  $\mathbb{R}^3$  as a basis of  
 the column space of  $B$ , which is

a basis of the column space of  $B = \left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right\}$ .

7. An engineer has found two solutions to a homogeneous system  $Ax = \mathbf{0}$  of 40 equations and 42 variables. These two solutions are not scalar multiple of each other, and every solution of the system can be constructed by a linear combination of these two solutions.

Suppose now the engineer needs to replace the homogeneous system with an associated non-homogeneous system  $Ax = b$  for some non-zero vector  $b$ . Can the engineer be certain that he will be able to find a solution?

**Solution:**

We assume that  $u_1, u_2$  are two solutions that the engineer found for homogeneous system  $Ax = 0$ .

Since  $u_1 \neq c \cdot u_2$  for all  $c \in \mathbb{R}$ ,

then  $u_1$  and  $u_2$  are linearly independent.

Since every solution of  $Ax = 0$  can be expressed as a linear combination of  $u_1, u_2$ ,

then the solution space =  $\text{span}\{u_1, u_2\}$

Hence,  $\{u_1, u_2\}$  is a basis for the solution space, which is also a basis for the nullspace of  $A$ .

Therefore,  $\text{nullity}(A) = 2$ .

Since  $Ax = 0$  has 40 equations and 42 variables,

then  $A$  is a  $40 \times 42$  matrix.

By the Dimension theorem,

$$\begin{aligned} \text{rank}(A) &= \text{total number of columns of } A - \text{nullity}(A) \\ &= 42 - 2 = 40. \end{aligned}$$

This implies, in the RREF of  $A$ , all the 40 rows are non-zero.  
(That is to say, in the RREF of  $A$ , the leading entry in the last row

is 1. Which implies, the system  $Ax = b$  is always consistent, no matter

what the value of the vector  $b$  is).

e.g. One possible RREF of the corresponding augmented matrix of  $Ax = b$  can be:

$$\left( \begin{array}{cccc|cc} 1 & 0 & \dots & 0 & a_{1,1} & a_{1,2} & b_1 \\ 0 & 1 & \dots & 0 & a_{2,1} & a_{2,2} & b_2 \\ \vdots & \vdots & & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & 0 & a_{40,1} & a_{40,2} & b_{40} \end{array} \right) \quad 40 \times 43$$

So for the associated non-homogeneous system  $Ax = b$ , it will have a solution regardless of the vector  $b$ .