

# Lecture 07

**Chapter 3 Vector Spaces**

**3.2 Subspaces**

**3.3 Linear Independence**

## 3.2 Subspaces

Construct new vector spaces from a given vector space  $V$ .

**Example 1.** Let  $\mathcal{S} = \left\{ \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mid x_2 = 2x_1 \right\} \subset \mathbf{R}^2$ .

(1) If  $\begin{pmatrix} c \\ 2c \end{pmatrix} \in \mathcal{S}$  and  $\alpha$  is any scalar, then  $\alpha \begin{pmatrix} c \\ 2c \end{pmatrix} = \begin{pmatrix} \alpha c \\ 2\alpha c \end{pmatrix} \in \mathcal{S}$ .

(2) If  $\begin{pmatrix} a \\ 2a \end{pmatrix}, \begin{pmatrix} b \\ 2b \end{pmatrix} \in \mathcal{S}$ , their sum  $\begin{pmatrix} a+b \\ 2a+2b \end{pmatrix} = \begin{pmatrix} a+b \\ 2(a+b) \end{pmatrix} \in \mathcal{S}$ .

By the definition of vector space, the mathematical system consisting of the set  $\mathcal{S}$  together with the operations induced from  $\mathbf{R}^2$ , is **itself** a vector space.

# Definitions

**Definition 1.** If  $S$  is a nonempty subset of a vector space  $V$  and  $S$  satisfies the following conditions:

- (1)  $\alpha x \in S$  whenever  $x \in S$  and for any scalar  $\alpha$ ;
  - (2)  $x + y \in S$  whenever  $x \in S$  and  $y \in S$ ;
- then  $S$  is said to be a **subspace** [子空间] of  $V$ .

**Remark** (closure properties [封闭性])

- (1):  $S$  is closed under **scalar multiplication**;
- (2):  $S$  is closed under **addition**.

If we do arithmetic using operations from  $V$  and elements of  $S$ , we will always end up with elements of  $S$ . A subspace of  $V$ , is a subset  $S$  that is **closed** under the operations of  $V$ .

**Remark.** In a vector space  $V$ , it is easy to verify that  $\{\mathbf{0}\}$  and  $V$  are subspaces of  $V$ .

- Subspaces  $\{\mathbf{0}\}$  and  $V$  are called **trivial subspaces** [平凡子空间] of vector space  $V$ .
- All other subspaces are referred to as **proper subspaces** [真子空间].
- The subspace  $\{\mathbf{0}\}$  is also called **zero subspace** [零子空间].

**Example 2.** Let  $\mathcal{S} = \{(x_1, x_2, x_3)^T \mid x_1 = x_2\} \subset \mathbf{R}^3$ .

The subset  $\mathcal{S}$  is nonempty since  $\mathbf{x} = (1, 1, 0)^T \in \mathcal{S}$ .

(1) If  $\mathbf{x} = (a, a, b)^T \in \mathcal{S}$ , then  $\alpha\mathbf{x} = (\alpha a, \alpha a, \alpha b)^T \in \mathcal{S}$ .

(2) If  $(a, a, b)^T$  and  $(c, c, d)^T \in \mathcal{S}$ , then

$$(a, a, b)^T + (c, c, d)^T = (a + c, a + c, b + d)^T \in \mathcal{S}.$$

Since  $\mathcal{S}$  is nonempty and satisfies the two closure conditions, it follows that  $\mathcal{S}$  is a subspace of  $\mathbf{R}^3$ .

**Example 3.** Let  $W = \left\{ \begin{pmatrix} x \\ 1 \end{pmatrix} \mid x \text{ is a real number} \right\} \subset \mathbf{R}^2$ .

The subset  $W$  is **not** a subspace of  $\mathbf{R}^2$ . In this case, both conditions fail:

(1)  $W$  is **not** closed under scalar multiplication

since  $\alpha \begin{pmatrix} x \\ 1 \end{pmatrix} \notin S$  unless  $\alpha = 1$ .

(2)  $W$  is **not** closed under addition,

since  $\begin{pmatrix} x \\ 1 \end{pmatrix} + \begin{pmatrix} y \\ 1 \end{pmatrix} = \begin{pmatrix} x + y \\ 2 \end{pmatrix} \notin S$ .

**Example 4.** Let  $\mathcal{S} = \{A \in \mathbf{R}^{2 \times 2} \mid a_{12} = -a_{21}\} \subset \mathbf{R}^{2 \times 2}$ .

The set  $\mathcal{S}$  form a subspace of  $\mathbf{R}^{2 \times 2}$ , since

(1) If  $A \in \mathcal{S}$ , then  $A$  must be of the form

$$A = \begin{pmatrix} a & b \\ -b & c \end{pmatrix} \quad \text{and hence} \quad \alpha A = \begin{pmatrix} \alpha a & \alpha b \\ -\alpha b & \alpha c \end{pmatrix}.$$

Since the (2,1) entry of  $\alpha A$  is the negative of the (1,2) entry,  $\alpha A \in \mathcal{S}$ .

(2) If  $A, B \in \mathcal{S}$ , then they must be of the form

$$A = \begin{pmatrix} a & b \\ -b & c \end{pmatrix}, \quad B = \begin{pmatrix} d & e \\ -e & f \end{pmatrix}.$$

It follows that  $A + B = \begin{pmatrix} a + d & b + e \\ -(b + e) & c + f \end{pmatrix} \in \mathcal{S}$ .



**Example 5.** Let  $C^1[a, b]$  be the set of all functions  $f$  that have a continuous derivative on  $[a, b]$ . Then  $C^1[a, b]$  is a proper subspace of  $C[a, b]$ .

**Example 6.** Let  $S$  be the set of all polynomials of degree less than  $n$  with the property that  $p(0) = 0$ . Then  $S$  is a proper subspace of  $P_n$ .

# Nullspace of a Matrix

**Definition 2.** The **nullspace** [零空间] or **kernel** [核] of an  $m \times n$  matrix  $A$ , denoted by  $N(A)$ , is defined as the set of all solutions to the homogeneous system  $A\mathbf{x} = \mathbf{0}$ .

$$N(A) = \{\mathbf{x} \in \mathbf{R}^n \mid A\mathbf{x} = \mathbf{0}\}.$$

- Since  $\mathbf{0} \in N(A)$ ,  $N(A)$  is nonempty.
- If  $\mathbf{x} \in N(A)$  and  $\alpha$  is a scalar, then  $A(\alpha\mathbf{x}) = \alpha A\mathbf{x} = \alpha\mathbf{0} = \mathbf{0}$ ,  
and hence  $\alpha\mathbf{x} \in N(A)$ .
- If  $\mathbf{x}, \mathbf{y} \in N(A)$ , then  $A(\mathbf{x} + \mathbf{y}) = A\mathbf{x} + A\mathbf{y} = \mathbf{0} + \mathbf{0} = \mathbf{0}$ .  
Therefore  $\mathbf{x} + \mathbf{y} \in N(A)$ .

The nullspace  $N(A)$  is a subspace of  $\mathbf{R}^n$ .

**Example 7.** Determine  $N(A)$  if

$$A = \begin{pmatrix} 1 & 1 & 1 & 0 \\ 2 & 1 & 0 & 1 \end{pmatrix}.$$

**Solution.** Using Gauss-Jordan reduction to solve  $A\mathbf{x} = \mathbf{0}$ , we obtain

$$\begin{aligned} \left( \begin{array}{cccc|c} 1 & 1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 1 & 0 \end{array} \right) &\rightarrow \left( \begin{array}{cccc|c} 1 & 1 & 1 & 0 & 0 \\ 0 & -1 & -2 & 1 & 0 \end{array} \right) \\ &\rightarrow \left( \begin{array}{cccc|c} 1 & 0 & -1 & 1 & 0 \\ 0 & -1 & -2 & 1 & 0 \end{array} \right) \rightarrow \left( \begin{array}{cccc|c} 1 & 0 & -1 & 1 & 0 \\ 0 & 1 & 2 & -1 & 0 \end{array} \right). \end{aligned}$$

The reduced row echelon form involves two free variables,  $x_3$  and  $x_4$ .

$$\begin{aligned} x_1 &= x_3 - x_4, \\ x_2 &= -2x_3 + x_4. \end{aligned}$$

**Example 7.** Determine  $N(A)$  if

$$A = \begin{pmatrix} 1 & 1 & 1 & 0 \\ 2 & 1 & 0 & 1 \end{pmatrix}.$$

**Solution. (continue)** Thus if we set  $x_3 = \alpha$  and  $x_4 = \beta$ , then

$$\mathbf{x} = \begin{pmatrix} \alpha - \beta \\ -2\alpha + \beta \\ \alpha \\ \beta \end{pmatrix} = \alpha \begin{pmatrix} 1 \\ -2 \\ 1 \\ 0 \end{pmatrix} + \beta \begin{pmatrix} -1 \\ 1 \\ 0 \\ 1 \end{pmatrix}.$$

is a solution to  $A\mathbf{x} = \mathbf{0}$ . The vector space  $N(A)$  consists of vectors of the form

$$\alpha(1, -2, 1, 0)^T + \beta(-1, 1, 0, 1)^T$$

where  $\alpha$  and  $\beta$  are arbitrary scalars. Finish.

# The Span of vectors

**Definition 3.** Let  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  be vectors in a vector space  $V$ . A sum of the form

$$\alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n,$$

where  $\alpha_1, \alpha_2, \dots, \alpha_n$  are scalars, is called a **linear combination** [线性组合] of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ .

The set of all linear combinations of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  is called the **span** [张成的集合] of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ . The span of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  will be denoted by  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ .

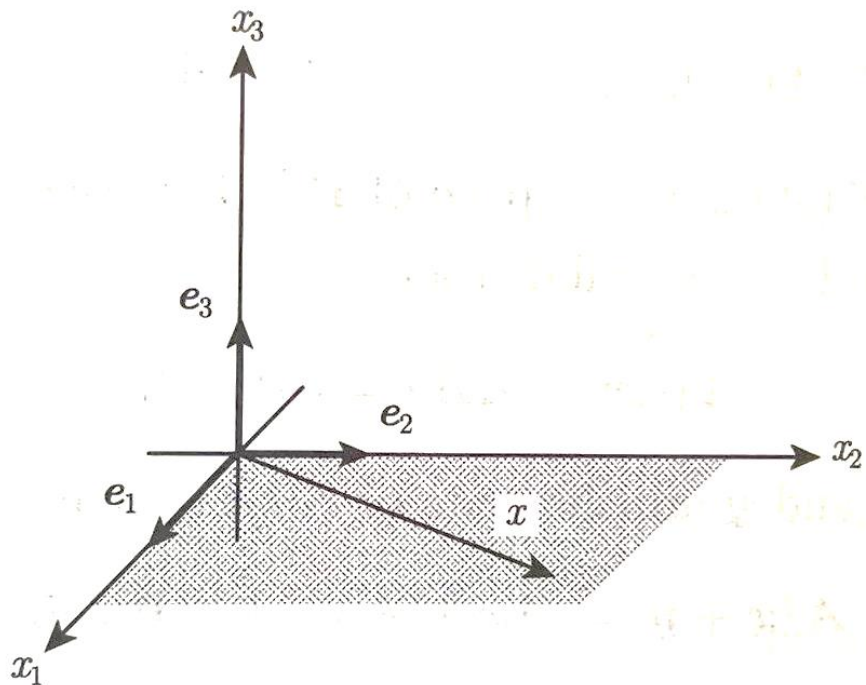
In **Example 7**, the nullspace of  $A$  is the span of the vectors

$$(1, -2, 1, 0)^T, (-1, 1, 0, 1)^T.$$

**Example 8.** In  $\mathbf{R}^3$ , the span of  $\mathbf{e}_1$  and  $\mathbf{e}_2$  is the set of all vectors

of the form  $\alpha \mathbf{e}_1 + \beta \mathbf{e}_2 = \begin{pmatrix} \alpha \\ \beta \\ 0 \end{pmatrix}$ .

Verify that  $\text{Span}\{\mathbf{e}_1, \mathbf{e}_2\}$  is a subspace of  $\mathbf{R}^3$ .



**Theorem 1.** If  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  are elements of a vector space  $V$ , then

$$\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$$

is a subspace of  $V$ .

**Proof.** We know that  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a subset of  $V$ .

- Let  $\mathbf{v} = \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n \in \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  and  $\beta$  be a scalar .

$$\begin{aligned}\beta \mathbf{v} &= (\beta \alpha_1) \mathbf{v}_1 + (\beta \alpha_2) \mathbf{v}_2 + \dots + (\beta \alpha_n) \mathbf{v}_n \\ &\in \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}.\end{aligned}$$

- Let  $\mathbf{v} = \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n$  and  $\mathbf{w} = \beta_1 \mathbf{v}_1 + \beta_2 \mathbf{v}_2 + \dots + \beta_n \mathbf{v}_n$ .

$$\begin{aligned}\mathbf{v} + \mathbf{w} &= (\alpha_1 + \beta_1) \mathbf{v}_1 + (\alpha_2 + \beta_2) \mathbf{v}_2 + \dots + (\alpha_n + \beta_n) \mathbf{v}_n \\ &\in \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}.\end{aligned}$$

Therefore,  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a subspace of  $V$ .

**Question:** Given a vector space  $V$ , can we find vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  such that they span  $V$ ?

**Definition 4.** The set  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a **spanning set** [生成集] for the vector space  $V$  **if and only if** every vector in  $V$  can be written as a linear combination of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ .

**Example 9.** For any vector  $\mathbf{x} \in \mathbf{R}^2$ ,

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix},$$

Therefore,  $\left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\}$  is a spanning set for the vector space  $\mathbf{R}^2$ .



**Example 10.** Which of the following are spanning sets for  $\mathbf{R}^3$ ?

- A.  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3, (1,2,3)^T\}$ ;
- B.  $\{(1,1,1)^T, (1,1,0)^T, (1,0,0)^T\}$ ;
- C.  $\{(1,0,1)^T, (0,1,0)^T\}$ ;
- D.  $\{(1,2,4)^T, (2,1,3)^T, (4, -1,1)^T\}$ .

To determine whether a set of vectors spans  $\mathbf{R}^3$ , we must determine whether an arbitrary vector  $(a, b, c)^T$  in  $\mathbf{R}^3$  can be written as a linear combination of the vectors in the set.

**Example 10.** Which of the following are spanning sets for  $\mathbf{R}^3$ ?

A.  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3, (1,2,3)^T\}$ ;

**Solution.** It is easily seen that  $(a, b, c)^T$  can be written as

$$(a, b, c)^T = a\mathbf{e}_1 + b\mathbf{e}_2 + c\mathbf{e}_3 + 0(1,2,3)^T$$

so that A is a spanning set for  $\mathbf{R}^3$ .

**Example 10.** Which of the following are spanning sets for  $\mathbf{R}^3$ ?

B.  $\{(1,1,1)^T, (1,1,0)^T, (1,0,0)^T\};$

**Solution.** (continue) In part (B), we must determine whether it is possible to find constants  $\alpha_1, \alpha_2, \alpha_3$  such that

$$\begin{pmatrix} a \\ b \\ c \end{pmatrix} = \alpha_1 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} + \alpha_2 \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} + \alpha_3 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

This leads to the system of equations

$$\alpha_1 + \alpha_2 + \alpha_3 = a$$

$$\alpha_1 + \alpha_2 = b$$

$$\alpha_1 = c$$

**Example 10.** Which of the following are spanning sets for  $\mathbf{R}^3$ ?

B.  $\{(1,1,1)^T, (1,1,0)^T, (1,0,0)^T\};$

**Solution.** (continue)

$$\begin{aligned}\alpha_1 + \alpha_2 + \alpha_3 &= a \\ \alpha_1 + \alpha_2 &= b \\ \alpha_1 &= c\end{aligned}$$

Since the coefficient matrix of the system is nonsingular, the system has a unique solution. In fact, we find that  $\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} = \begin{pmatrix} c \\ b - c \\ a - b \end{pmatrix}$ . Thus

$$\begin{pmatrix} a \\ b \\ c \end{pmatrix} = c \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} + (b - c) \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} + (a - b) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

So the three vectors in B span  $\mathbf{R}^3$ .

**Example 10.** Which of the following are spanning sets for  $\mathbf{R}^3$ ?

$$C. \{(1,0,1)^T, (0,1,0)^T\};$$

**Solution.** (continue)

The span of the vectors in C is the set of vectors of the form

$$\begin{pmatrix} \alpha \\ \beta \\ \alpha \end{pmatrix}, \quad \alpha, \beta \in \mathbf{R},$$

which is clearly not the entire  $\mathbf{R}^3$ , since the vector  $(1,0,0)^T$  is not in the span.

Part D is left as **an exercise**.

**Remark.** The last example shows that

- the spanning set of a vector space  $V$  may **not** be unique, since the vector sets in A, B are both spanning set of  $\mathbf{R}^3$ ;
- the number of vectors of a spanning set of  $V$  may also be **not** unique, because in A, B the number of vectors is four and three respectively.

**Example 11.** Show that the vector set

$$S = \left\{ E_1 = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}, E_2 = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}, E_3 = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}, E_4 = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \right\}$$

is a spanning set of the vector space  $\mathbf{R}^{2 \times 2}$ .

**Proof.** See the textbook.

**Example 12.** Show that the vectors  $1 - x^2, x + 2, x^2$  span  $P_3$ .

**Proof.** Any vector  $p$  in  $P_3$  can be written as

$$p(x) = ax^2 + bx + c.$$

It is enough to show that  $p$  can be represented as a linear combination of the given vectors, i.e. to find constants  $\alpha_1, \alpha_2, \alpha_3$  such that

$$ax^2 + bx + c = \alpha_1(1 - x^2) + \alpha_2(x + 2) + \alpha_3x^2.$$

This leads to the following system

$$\begin{array}{lll} \alpha_3 - \alpha_1 = a, & & \alpha_1 = c - 2b \\ \alpha_2 = b, & \implies & \alpha_2 = b, \\ 2\alpha_2 - \alpha_1 = c. & & \alpha_3 = a + c - 2b \end{array}$$

Therefore, the three vectors span  $P_3$ .



# Linear System revisited

Let  $S$  be the solution set to the linear system  $A\mathbf{x} = \mathbf{b}$ , where  $A$  is  $m \times n$ ,  $\mathbf{b} \in \mathbf{R}^m$ .

- If  $\mathbf{b} = \mathbf{0}$ , then  $S = N(A)$  is a subspace of  $\mathbf{R}^n$ .
- If  $\mathbf{b} \neq \mathbf{0}$ , then  $S$  is **not** a subspace of  $\mathbf{R}^n$ .

**Theorem.** Assume that the linear system  $A\mathbf{x} = \mathbf{b}$  is consistent and  $\mathbf{x}_0$  is a **particular solution**. Then a vector  $\mathbf{y}$  is in the solution set  $S$  **if and only** if

$$\mathbf{y} = \mathbf{x}_0 + \mathbf{z},$$

where  $\mathbf{z} \in N(A)$ .

# Linear System revisited

Solution set of  $A\mathbf{x} = \mathbf{b}$

