

## Chapter 3 Vector Spaces

### Section 3.4 Basis and Dimension

## Example Re-visit

There are **infinitely many** sets of vectors that span the same vector space! e.g.

$$\mathbf{R}^3 = \text{Span}\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\} = \text{Span}\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3, (1, 2, 3)^T\} = \dots$$

What is the **minimal** spanning set for a vector space  $V$ ?

By "minimal", not too many vectors in a spanning set, also not too few.

Answer: **Basis!**

**Definition (Basis, plural form: bases)** The vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  form a *basis* of a vector space  $V$  if

1.  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  are linearly independent, and
2.  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a spanning set for  $V$ .

In this case,  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a *basis* of  $V$ .

**Example**  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$  is a basis of  $\mathbf{R}^3$ .

**Reason**

1. If  $c_1 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + c_2 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + c_3 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$ , then  $c_1 = c_2 = c_3 = 0$ .

2. For any vector  $\begin{pmatrix} a \\ b \\ c \end{pmatrix} \in \mathbf{R}^3$ , if  $\alpha_1 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + \alpha_2 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + \alpha_3 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$ , then  $\begin{cases} \alpha_1 = a \\ \alpha_2 = b \\ \alpha_3 = c \end{cases}$ .

**Example** Is  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3, (1, 1, 1)^T\}$  a basis of  $\mathbf{R}^3$ ?

**Solution** Solving  $c_1\mathbf{e}_1 + c_2\mathbf{e}_2 + c_3\mathbf{e}_3 + c_4(1, 1, 1)^T = \mathbf{0}$ , we get  
 $(c_1, c_2, c_3, c_4) = (-\alpha, -\alpha, -\alpha, \alpha)$ .

Hence,  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3, (1, 1, 1)^T$  are **linearly dependent**.

So,  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3, (1, 1, 1)^T\}$  is not a basis of  $\mathbf{R}^3$ .

**Example** Is  $\{\mathbf{e}_1, \mathbf{e}_2\}$  a basis of  $\mathbf{R}^3$ ?

**Solution**

Let  $(a, b, c)^T \in \mathbf{R}^3$ . Solving  $\alpha_1\mathbf{e}_1 + \alpha_2\mathbf{e}_2 = (a, b, c)^T$ , we have 
$$\begin{cases} \alpha_1 = a \\ \alpha_2 = b \\ 0 = c \end{cases} .$$

Since  $0 = c$  is inconsistent when  $c \neq 0$ , there is no solution for  $\alpha_1\mathbf{e}_1 + \alpha_2\mathbf{e}_2 = (a, b, c)^T$  when  $c \neq 0$ .

Since  $\{\mathbf{e}_1, \mathbf{e}_2\}$  is **not a spanning set** of  $\mathbf{R}^3$ ,  $\{\mathbf{e}_1, \mathbf{e}_2\}$  is not a basis of  $\mathbf{R}^3$ .

**Example** Is  $\{\mathbf{e}_1, \mathbf{e}_2, (1, 1, 1)^T\}$  a basis of  $\mathbf{R}^3$ ?

**Solution**

1. If  $c_1 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + c_2 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + c_3 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$ , then  $\begin{pmatrix} c_1 + c_3 \\ c_2 + c_3 \\ c_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$   
and so  $c_1 = c_2 = c_3 = 0$ . So,  $\mathbf{e}_1, \mathbf{e}_2, (1, 1, 1)^T$  are **linearly independent**.

2. Let  $(a, b, c)^T \in \mathbf{R}^3$ . Solving  $\alpha_1 \mathbf{e}_1 + \alpha_2 \mathbf{e}_2 + \alpha_3 (1, 1, 1)^T = (a, b, c)^T$ , we  
have  $\begin{cases} \alpha_1 = a - c \\ \alpha_2 = b - c \\ \alpha_3 = c \end{cases}$ .  $\{\mathbf{e}_1, \mathbf{e}_2, (1, 1, 1)^T\}$  is **a spanning set** for  $\mathbf{R}^3$ .

**Conclusion** Since  $\mathbf{e}_1, \mathbf{e}_2, (1, 1, 1)^T$  are linearly independent and form a spanning set for  $\mathbf{R}^3$ , they form a basis of  $\mathbf{R}^3$ .

Example (Standard basis)  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  is called the standard basis of  $\mathbf{R}^n$ .

Example (Standard basis)  $\{1, x, x^2, \dots, x^{n-1}\}$  is called the standard basis of  $P_n$ .

Example (Standard basis)  $\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$  is the standard basis of the vector space  $\mathbf{R}^{2 \times 2}$  of  $2 \times 2$  matrices.

**Example** Let  $A = \begin{pmatrix} 1 & -2 & 1 & 1 & 2 \\ -1 & 3 & 0 & 2 & -2 \\ 0 & 1 & 1 & 3 & 4 \\ 1 & 2 & 5 & 13 & 5 \end{pmatrix}$ . Find a basis of the null space  $N(A)$  of  $A$ .

**Solution** The reduced row echelon form of  $A$  is  $\begin{pmatrix} 1 & 0 & 3 & 7 & 0 \\ 0 & 1 & 1 & 3 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$ .

$$\begin{aligned} N(A) &= \{\mathbf{x} = (x_1, x_2, x_3, x_4, x_5)^T : A\mathbf{x} = \mathbf{0}\} \\ &= \left\{ (x_1, x_2, x_3, x_4, x_5)^T : \begin{pmatrix} 1 & 0 & 3 & 7 & 0 \\ 0 & 1 & 1 & 3 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \right\} \\ &= \left\{ \begin{pmatrix} -3\alpha - 7\beta \\ -\alpha - 3\beta \\ \alpha \\ \beta \\ 0 \end{pmatrix} \right\} = \left\{ \alpha \begin{pmatrix} -3 \\ -1 \\ 1 \\ 0 \\ 0 \end{pmatrix} + \beta \begin{pmatrix} -7 \\ -3 \\ 0 \\ 1 \\ 0 \end{pmatrix} \right\} \\ &\left\{ \begin{pmatrix} -3 \\ -1 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -7 \\ -3 \\ 0 \\ 1 \\ 0 \end{pmatrix} \right\} \text{ is a basis of } N(A). \end{aligned}$$

**Theorem 1** If  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a spanning set for  $V$ , then any collection of  $m$  vectors in  $V$ ,  $m > n$ , is linearly dependent.

**Corollary** If both  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  and  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$  are bases for a vector space  $V$ , then  $n = m$ .

**Proof of Corollary** Let both  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  and  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$  be bases for  $V$ . Since  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  span  $V$  and  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$  are linearly independent, it follows from the above theorem that  $m \leq n$ . By the same reasoning,  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$  span  $V$  and  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  are linearly independent, so  $n \leq m$ . The assertion of  $m = n$  is proved.



**Theorem 1** If  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a spanning set for  $V$ , then any collection of  $m$  vectors in  $V$ ,  $m > n$ , is linearly dependent.

**Proof** Let  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$  be  $m$  vectors in  $V$ , where  $m \geq n$ . Since  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  span  $V$ , we have  $\mathbf{u}_i = a_{i1}\mathbf{v}_1 + \dots + a_{in}\mathbf{v}_n$  and a linear combination

$$\begin{aligned} c_1\mathbf{u}_1 + \dots + c_m\mathbf{u}_m &= c_1 \sum_{j=1}^n a_{1j}\mathbf{v}_j + \dots + c_m \sum_{j=1}^n a_{mj}\mathbf{v}_j \\ &= \sum_{i=1}^m \left[ c_i \sum_{j=1}^n a_{ij}\mathbf{v}_j \right] \\ &= \sum_{j=1}^n \left[ \sum_{i=1}^m a_{ij}c_i \right] \mathbf{v}_j. \end{aligned}$$

**Proof (continuity)** Now consider the system of equations with unknown  $c_i$ 's as follows

$$\sum_{i=1}^m a_{ij}c_i = 0, j = 1, 2, \dots, n.$$

This is a homogeneous system with more unknowns than equations. Therefore, by a theorem in Sc 1.2 (An  $m \times n$  homogeneous system of linear equations has a nontrivial solution if  $m < n$ .) the system must have a nontrivial solution, denoted by  $c_1^*, \dots, c_m^*$ . But then

$$c_1^* \mathbf{u}_1 + \dots + c_m^* \mathbf{u}_m = \sum_{j=1}^n 0 \mathbf{v}_j.$$

Hence,  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$  are linearly dependent.

**Definition (Dimension)** Let  $V$  be a vector space. If  $V$  has a basis consisting of  $n$  vectors, we say that  $V$  has *dimension*  $n$  and write  $\dim V = n$ .

**Example**  $\mathbf{R}^n$  is of dimension  $n$ .

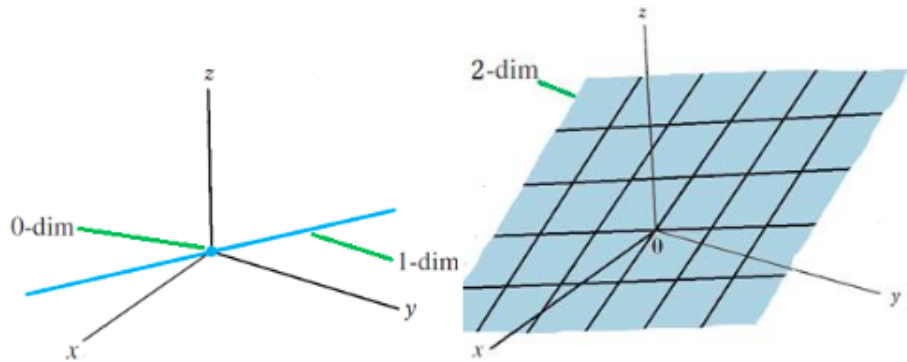
**Example**  $\mathbf{R}^{m \times n}$  is of dimension  $mn$ .

**Example**  $P_n$  is of dimension  $n$ .

**Example** The subspace  $\{\mathbf{0}\}$  of a vector space  $V$  is said to have dimension 0. (Some books would say that the basis of  $\{\mathbf{0}\}$  is  $\emptyset$ .)

**Example** The subspaces of  $\mathbf{R}^3$  can be classified by dimension.

- ▶ 0-dimensional subspaces: The zero subspace.
- ▶ 1-dimensional subspaces: Any subspace spanned by a single nonzero vector. Such subspaces are lines through the origin.
- ▶ 2-dimensional subspaces: Any subspace spanned by two linearly independent vectors. Such subspaces are planes through the origin.
- ▶ 3-dimensional subspaces:  $\mathbf{R}^3$



**Definition (Finite-dimensional / Infinite-dimensional)**  $V$  is said to be *finite-dimensional* if a basis of  $V$  has only finitely many elements. Otherwise,  $V$  is said to be *infinite-dimensional*.

**Example**  $\mathbf{R}^n$ ,  $\mathbf{R}^{m \times n}$ ,  $P_n$  are all finite-dimensional.

**Example**  $\{1, x, x^2, \dots\}$  is a basis of vector space  $P$  of all polynomials. So  $P$  is of infinite-dimensional.

**Example** The vector space of all continuous functions is infinite-dimensional.

**Example** The vector space of infinite sequences is infinite-dimensional.

**Theorem 2** If  $V$  is a vector space of **dimension  $n > 0$** , then

- (I) any set of  $n$  linearly independent vectors spans  $V$ ;
- (II) any  $n$  vectors that span  $V$  are linearly independent.

**Proof of (I)** Suppose that  $\mathbf{v}_1, \dots, \mathbf{v}_n$  are linearly independent and  $\mathbf{x}$  is any other vector in  $V$ . Since  $V$  has dimension  $n$ , it has a basis consisting of  $n$  vectors and these vectors span  $V$ . It follows that the set of  $\{\mathbf{v}_1, \dots, \mathbf{v}_n, \mathbf{x}\}$  must be linearly dependent. Thus, there exist scalars  $c_1, \dots, c_n, c_{n+1}$ , not all zero, such that

$$c_1\mathbf{v}_1 + \dots + c_n\mathbf{v}_n + c_{n+1}\mathbf{x} = \mathbf{0} \quad (\#)$$

The scalar  $c_{n+1} \neq 0$ , otherwise from  $(\#)$   $\mathbf{v}_1, \dots, \mathbf{v}_n$  are linearly dependent. Hence,  $(\#)$  can be solved for  $\mathbf{x}$ :

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

where  $\alpha_i = -c_i/c_{n+1}$ . Since  $\mathbf{x}$  is arbitrary, the assertion (I) follows.

**Proof of (II)** Suppose that  $\mathbf{v}_1, \dots, \mathbf{v}_n$  span  $V$ . If  $\mathbf{v}_1, \dots, \mathbf{v}_n$  are linearly dependent, then one of the  $\mathbf{v}_i$ 's, say,  $\mathbf{v}_n$ , can be written as a linear combination of the others. It follows from Theorem 1 that  $\mathbf{v}_1, \dots, \mathbf{v}_{n-1}$  will still span  $V$ . If  $\mathbf{v}_1, \dots, \mathbf{v}_{n-1}$  are linearly dependent, we can eliminate another vector and still have a spanning set. We can continue eliminating vectors in this way until we arrive at a linearly independent spanning set with  $k < n$  elements. But this contradicts  $\dim V = n$ . Therefore,  $\mathbf{v}_1, \dots, \mathbf{v}_n$  must be linearly independent.



**Theorem 3** If  $V$  is a vector space of dimension  $n > 0$ , then

- (i) no set of fewer than  $n$  vectors can span  $V$ ;
- (ii) any subset of fewer than  $n$  linearly independent vectors can be extended to form a basis for  $V$ ;
- (iii) any spanning set containing more than  $n$  vectors can be pared down to form a basis for  $V$ .

**Proof of (i) and (ii)** Statement (i) follows by the same reasoning that was used to prove part (I) of Theorem 2.

To prove (ii), suppose that  $\mathbf{v}_1, \dots, \mathbf{v}_k$  are linearly independent with  $k < n$ . It follows from (i) that  $\text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  is a proper subspace of  $V$ , and hence there exists a vector  $\mathbf{v}_{k+1} \in V$  that is not in  $\text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ . It then follows that  $\mathbf{v}_1, \dots, \mathbf{v}_k, \mathbf{v}_{k+1}$  must be linearly independent. If  $k + 1 < n$ , then, in the same manner,  $\mathbf{v}_1, \dots, \mathbf{v}_k, \mathbf{v}_{k+1}$  can be extended to a set of  $k + 2$  linearly independent vectors. This extension process may be continued until a set  $\{\mathbf{v}_1, \dots, \mathbf{v}_k, \mathbf{v}_{k+1}, \dots, \mathbf{v}_n\}$  of  $n$  linearly independent vectors is obtained.

**Proof of (iii)** Suppose that  $\mathbf{v}_1, \dots, \mathbf{v}_m$  span  $V$  and  $m > n$ . Then, by Theorem 1,  $\mathbf{v}_1, \dots, \mathbf{v}_m$  must be linearly dependent. It follows that one of the vectors, say,  $\mathbf{v}_m$ , can be written as a linear combination of the others. Hence, if  $\mathbf{v}_m$  is eliminated from the set, the remaining  $m - 1$  vectors will still span  $V$ . If  $m - 1 > n$ , we can continue to eliminate vectors in this manner until we arrive at a spanning set containing  $n$  elements.

**Extra Exercise\*** Determine if the following set of vectors form a basis for  $\mathbf{R}^3$ .

1.  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$  in  $\mathbf{R}^3$ .
2.  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3, (1, 2, 3)^T\}$  in  $\mathbf{R}^3$ .
3.  $\{(1, 0, 0)^T, (0, 1, 1)^T\}$  in  $\mathbf{R}^3$ .

Answer:

(a). Yes, (b). No, (c). No

Extra Example\* Let

$$S = \{(2, -3, 5)^T, (8, -12, 20)^T, (1, 0, 2)^T, (0, 2, -1)^T, (7, 2, 0)^T\}.$$

$\text{Span}(S) = \mathbf{R}^3$ . Find a basis of  $\mathbf{R}^3$  providing  $S$ .

Answer:

1. Since  $(8, -12, 20)^T = 4(2, -3, 5)^T$ , then remove it

$$\text{Span}\{(2, -3, 5)^T, (1, 0, 2)^T, (0, 2, -1)^T, (7, 2, 0)^T\} = \mathbf{R}^3.$$

2. Since  $(7, 2, 0)^T = 2(2, -3, 5)^T + 3(1, 0, 2)^T + 4(0, 2, -1)^T$ , then remove it

$$\text{Span}\{(2, -3, 5)^T, (1, 0, 2)^T, (0, 2, -1)^T\} = \mathbf{R}^3.$$

3. Assume  $\alpha_1(2, -3, 5)^T + \alpha_2(1, 0, 2)^T + \alpha_3(0, 2, -1)^T = (0, 0, 0)^T$ , the only solution is  $\alpha_1 = \alpha_2 = \alpha_3 = 0$ , linear independent.

Therefore,  $\{(2, -3, 5)^T, (1, 0, 2)^T, (0, 2, -1)^T\}$  is a basis for  $\mathbf{R}^3$ .