

## Linearly Independent Sets and Bases

Recall that an indexed set of vectors  $\{\mathbf{v}_1, \dots, \mathbf{v}_p\} \in V$  is *linearly independent* if the vector equation  $c_1\mathbf{v}_1 + \dots + c_p\mathbf{v}_p = \mathbf{0}$  has only the trivial solution  $c_1 = 0, \dots, c_p = 0$ . Also recall that the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_p\} \in V$  is *linearly dependent* if there exists weights  $c_1, \dots, c_p$ , not all zero, that satisfy  $c_1\mathbf{v}_1 + \dots + c_p\mathbf{v}_p = \mathbf{0}$  and, in such a case,  $c_1\mathbf{v}_1 + \dots + c_p\mathbf{v}_p = \mathbf{0}$  is called the *linear dependence relation* among  $\mathbf{v}_1, \dots, \mathbf{v}_p$ .

**Theorem** (Linearly Dependent Set of  $\geq 2$  Vectors).

An indexed set  $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$  of two or more vectors, where  $\mathbf{v}_1 \neq \mathbf{0}$ , is linearly dependent if and only if some  $\mathbf{v}_j$ , where  $j > 1$ , is a linear combination of the preceding vectors  $\mathbf{v}_1, \dots, \mathbf{v}_{j-1}$ . That is, the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ , where  $p \geq 2$  and  $\mathbf{v}_1 \neq \mathbf{0}$ , is linearly dependent if and only if  $\mathbf{v}_j = c\mathbf{v}_1 + \dots + c\mathbf{v}_{j-1}$ , where  $j > 1$ .

The general vector space  $V$  may contain vectors  $\mathbf{v}_1, \dots, \mathbf{v}_p$  that are not  $n$ -tuples (an ordered list on  $n$  numbers  $v_1, \dots, v_n$  in a column) and, in such a case, the homogeneous vector equation  $c_1\mathbf{v}_1 + \dots + c_p\mathbf{v}_p = \mathbf{0}$  cannot be expressed by the matrix equation  $A\mathbf{x} = \mathbf{0}$ , because the vectors  $\mathbf{v}_1, \dots, \mathbf{v}_p$  cannot be expressed as the columns of a matrix  $A$ . Therefore, the definition of linear dependence and the preceding theorem must be used to study a set  $\{\mathbf{v}_1, \dots, \mathbf{v}_p\} \in V$ , since the matrix equation  $A\mathbf{x} = \mathbf{0}$  cannot always be studied.

**Example.** Let  $\mathbf{p}_1(t) = 1$ ,  $\mathbf{p}_2(t) = t$ , and  $\mathbf{p}_3(t) = 4 - t$ . Then the set  $\{\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3\}$  is linearly dependent in  $\mathbb{P}$  because  $\mathbf{p}_3 = 4\mathbf{p}_1 - \mathbf{p}_2$ . ■

**Definition** (Basis).

Let  $U$  be a subspace of the vector space  $V$ . The set  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_p\} \in V$  is a **basis** of  $U$  if  $\mathcal{B}$  is linearly independent and  $U = \text{Span}\{\mathbf{b}_1, \dots, \mathbf{b}_p\}$ .

Every vector space  $V$  is a subspace of itself, so if  $U = V$ , then the *basis* of  $V$  is the linearly independent set  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_p\} \in V$ , where  $V = \text{Span}\{\mathbf{b}_1, \dots, \mathbf{b}_p\}$ . Let  $\mathbf{e}_1, \dots, \mathbf{e}_n$  be the columns of the  $n \times n$  identity matrix  $I_n$ . Then the set  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  is called the **standard basis** of  $\mathbb{R}^n$ .

**Example.** Consider  $\mathbf{v}_1$ ,  $\mathbf{v}_2$ , and  $\mathbf{v}_3$ . Determine if  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is a basis of  $\mathbb{R}^3$ .

$$\mathbf{v}_1 = \begin{bmatrix} 3 \\ 0 \\ -6 \end{bmatrix} \quad \mathbf{v}_2 = \begin{bmatrix} -4 \\ 1 \\ 7 \end{bmatrix} \quad \mathbf{v}_3 = \begin{bmatrix} -2 \\ 1 \\ 5 \end{bmatrix}$$

There are three vectors in  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  which, themselves, are in  $\mathbb{R}^3$ , so the matrix  $A = [\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3]$  is  $3 \times 3$ . Performing two row replacement operations on  $A$  shows that  $A$  has three pivot positions, and hence is invertible. Thus, the columns of  $A$  form a basis of  $\mathbb{R}^3$  because by the Invertible Matrix Theorem, the columns of  $A$  are linearly independent and span  $\mathbb{R}^3$ , since  $A$  is invertible. ■

## The Spanning Set Theorem

Consider the sets  $S = \{\mathbf{v}_1, \dots, \mathbf{v}_p\} \in V$  and  $U = \text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ . Suppose that  $S$  is linearly dependent. Then by rearranging the vectors  $\mathbf{v}_1, \dots, \mathbf{v}_p \in S$ , the vector  $\mathbf{v}_p$  can be expressed by  $\mathbf{v}_p = a_1\mathbf{v}_1 + \dots + a_{p-1}\mathbf{v}_{p-1}$ . Also consider the arbitrary vector  $\mathbf{x} \in U$ , which can be expressed by  $\mathbf{x} = c_1\mathbf{v}_1 + \dots + c_{p-1}\mathbf{v}_{p-1} + c_p\mathbf{v}_p$  for suitable scalars  $c_1, \dots, c_p$ . Observe below that by substituting the expression for  $\mathbf{v}_p$  into the expression for  $\mathbf{x}$ , the arbitrary vector  $\mathbf{x} \in U$  can be expressed as a linear combination of the vectors  $\mathbf{v}_1, \dots, \mathbf{v}_{p-1}$ , and hence  $\{\mathbf{v}_1, \dots, \mathbf{v}_{p-1}\}$  spans  $U$ , because  $\mathbf{x}$  is an arbitrary element of  $U$ .

$$\begin{aligned}\mathbf{x} &= c_1\mathbf{v}_1 + \dots + c_{p-1}\mathbf{v}_{p-1} + c_p\mathbf{v}_p \\ &= (c_1 + \dots + c_p a_1)\mathbf{v}_1 + \dots + (c_{p-1} + \dots + c_p a_{p-1})\mathbf{v}_{p-1}\end{aligned}$$

If  $S$  is linearly dependent and  $U = \text{Span}\{S\} \neq \{\mathbf{0}\}$ , then  $S$  contains two or more vectors and, therefore, vectors in  $S$  can repeatedly be removed until  $S$  is linearly independent and hence a basis for  $U$ .

**Theorem** (The Spanning Set Theorem).

Let  $S = \{\mathbf{v}_1, \dots, \mathbf{v}_p\} \in V$  and let  $U = \text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ .

- If some vector  $\mathbf{v}_j \in S$  is a linear combination of the remaining vectors in  $S$ , then the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_{p-1}\}$  that is formed by removing  $\mathbf{v}_j$  from  $S$  still spans  $U$ . That is,  $U = \text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_p\} = \text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_{p-1}\}$ .
- If  $U \neq \{\mathbf{0}\}$ , then some subset of  $S$  is a basis for  $U = \text{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ .

**Example.** Consider the vectors  $\mathbf{v}_1$ ,  $\mathbf{v}_2$ , and  $\mathbf{v}_3$ , where  $U = \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  and  $\mathbf{v}_3 = 5\mathbf{v}_1 + 3\mathbf{v}_2$ . Show that  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} = \text{Span}\{\mathbf{v}_1, \mathbf{v}_2\}$ , and then find a basis for the subspace  $U$ .

$$\mathbf{v}_1 = \begin{bmatrix} 0 \\ 2 \\ -1 \end{bmatrix} \quad \mathbf{v}_2 = \begin{bmatrix} 2 \\ 2 \\ 0 \end{bmatrix} \quad \mathbf{v}_3 = \begin{bmatrix} 6 \\ 16 \\ -5 \end{bmatrix}$$

First notice that  $c_1\mathbf{v}_1 + c_2\mathbf{v}_2 = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + 0\mathbf{v}_3$  for all scalars  $c_1, c_2 \in \mathbb{R}$  and, therefore, every vector in  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2\}$  is also an element of  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ . Let  $\mathbf{x}$  be an arbitrary vector in  $U$ , then  $\mathbf{x} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3$ . Observe that since  $\mathbf{v}_3 = 5\mathbf{v}_1 + 3\mathbf{v}_2$ , the vector  $\mathbf{v}_3$  can be substituted into  $\mathbf{x} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3$  to express  $\mathbf{x}$  as a linear combination of the vectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$ .

$$\begin{aligned}\mathbf{x} &= c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3(5\mathbf{v}_1 + 3\mathbf{v}_2) \\ &= (c_1 + 5c_3)\mathbf{v}_1 + (c_2 + 3c_3)\mathbf{v}_2\end{aligned}$$

Thus,  $\mathbf{x} \in \text{Span}\{\mathbf{v}_1, \mathbf{v}_2\}$  and hence every vector in  $U$  is also an element of  $\text{Span}\{\mathbf{v}_1, \mathbf{v}_2\}$ , indicating that  $U = \text{Span}\{\mathbf{v}_1, \mathbf{v}_2\}$ . Therefore, since the set  $\{\mathbf{v}_1, \mathbf{v}_2\}$  is linearly independent, it follows that  $\{\mathbf{v}_1, \mathbf{v}_2\}$  is a basis of  $U$ . ■

## Bases for $\text{Nul}(A)$ and $\text{Col}(A)$

Recall that for a given  $m \times n$  matrix  $A$ , solving the matrix equation  $A\mathbf{x} = \mathbf{0}$  amounts to explicitly defining  $\text{Nul}(A) = \{\mathbf{x} \in \mathbb{R}^n : A\mathbf{x} = \mathbf{0}\}$  in terms of its spanning set (the parametric vector equation, or linear combination, that describes the general solution to  $A\mathbf{x} = \mathbf{0}$ ), which is always linearly independent when it contains nonzero vectors. In such a case, it follows that the linearly independent spanning set of  $\text{Nul}(A)$  produces a *basis* for  $\text{Nul}(A)$ .

**Example.** Find a basis for  $\text{Col}(B)$ , where  $B$  is defined as follows.

$$B = [\mathbf{b}_1 \quad \mathbf{b}_2 \quad \cdots \quad \mathbf{b}_5] = \begin{bmatrix} 1 & 4 & 0 & 2 & 0 \\ 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

First recall that  $\text{Col}(B) = \text{Span}\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3, \mathbf{b}_4, \mathbf{b}_5\}$ . Observe that in the given matrix  $B$ , each non-pivot column is a linear combination of one or more pivot columns, where  $\mathbf{b}_2 = 4\mathbf{b}_1$  and  $\mathbf{b}_4 = 2\mathbf{b}_1 - \mathbf{b}_3$ . By the Spanning Set Theorem, the vectors  $\mathbf{b}_2$  and  $\mathbf{b}_4$  can be removed from  $\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3, \mathbf{b}_4, \mathbf{b}_5\}$  to produce the set  $\{\mathbf{b}_1, \mathbf{b}_3, \mathbf{b}_5\}$ , which still spans  $\text{Col}(B)$ . Let  $S$  be defined as follows.

$$S = \{\mathbf{b}_1, \mathbf{b}_3, \mathbf{b}_5\} = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} \right\}$$

Since  $\mathbf{b}_1 \neq \mathbf{0}$  and no vector  $\mathbf{b}_j \in S$  is a linear combination of the preceding vectors  $\mathbf{b}_1, \dots, \mathbf{b}_{j-1}$ , the set  $S$  is linearly independent and is hence a basis for  $\text{Col}(B)$ . ■

**Theorem** (The Basis of Column Space).

The pivot columns of a matrix  $A$  form a basis for  $\text{Col}(A)$ .

*Proof.* TO BE WRITTEN

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