

# 114-2

## 線性代數 (二)

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5.1



### Orthogonality in $\mathbb{R}^n$

<https://hmwu.idv.tw>

generalize the notion of orthogonality of vectors in  $\mathbb{R}^n$  from two vectors to sets of vectors.

## Orthogonal and Orthonormal Sets of Vectors

**Definition** A set of vectors  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  in  $\mathbb{R}^n$  is called an *orthogonal set* if all pairs of distinct vectors in the set are orthogonal—that is, if

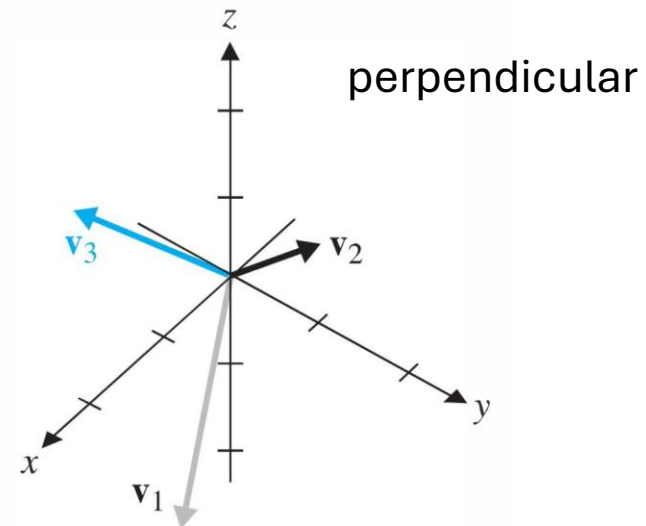
$$\mathbf{v}_i \cdot \mathbf{v}_j = 0 \quad \text{whenever} \quad i \neq j \quad \text{for} \quad i, j = 1, 2, \dots, k$$

The standard basis  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  of  $\mathbb{R}^n$  is an orthogonal set.

**Example 5.1**

Show that  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is an orthogonal set in  $\mathbb{R}^3$  if

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



**Figure 5.4**

An orthogonal set of vectors

**Theorem 5.1**

If  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  is an orthogonal set of nonzero vectors in  $\mathbb{R}^n$ , then these vectors are linearly independent.

**Proof** If  $c_1, \dots, c_k$  are scalars such that  $c_1\mathbf{v}_1 + \dots + c_k\mathbf{v}_k = \mathbf{0}$ , then

$$(c_1\mathbf{v}_1 + \dots + c_k\mathbf{v}_k) \cdot \mathbf{v}_i = \mathbf{0} \cdot \mathbf{v}_i = 0$$

$$c_1(\mathbf{v}_1 \cdot \mathbf{v}_i) + \dots + c_i(\mathbf{v}_i \cdot \mathbf{v}_i) + \dots + c_k(\mathbf{v}_k \cdot \mathbf{v}_i) = 0 \quad (1)$$

**Definition** An *orthogonal basis* for a subspace  $W$  of  $\mathbb{R}^n$  is a basis of  $W$  that is an orthogonal set.

**Example 5.2**

The vectors

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

from Example 5.1 are orthogonal and, hence, linearly independent.

Since any three linearly independent vectors in  $\mathbb{R}^3$  form a basis for  $\mathbb{R}^3$ ,

by the Fundamental Theorem of Invertible Matrices,

it follows that  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is an orthogonal basis for  $\mathbb{R}^3$ .

**Example 5.3**

Find an orthogonal basis for the subspace  $W$  of  $\mathbb{R}^3$  given by

$$W = \left\{ \begin{bmatrix} x \\ y \\ z \end{bmatrix} : x - y + 2z = 0 \right\}$$

**Solution** Section 5.3 gives a general procedure for problems of this sort.

**Theorem 5.2**

Let  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  be an orthogonal basis for a subspace  $W$  of  $\mathbb{R}^n$  and let  $\mathbf{w}$  be any vector in  $W$ . Then the unique scalars  $c_1, \dots, c_k$  such that

$$\mathbf{w} = c_1\mathbf{v}_1 + \cdots + c_k\mathbf{v}_k$$

are given by

$$c_i = \frac{\mathbf{w} \cdot \mathbf{v}_i}{\mathbf{v}_i \cdot \mathbf{v}_i} \quad \text{for } i = 1, \dots, k$$

**Proof** Since  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  is a basis for  $W$ , we know that there are unique scalars  $c_1, \dots, c_k$  such that  $\mathbf{w} = c_1\mathbf{v}_1 + \cdots + c_k\mathbf{v}_k$  (from Theorem 3.29).

**Example 5.4**

Find the coordinates of  $\mathbf{w} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$  with respect to the orthogonal basis  $\mathcal{B} = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  of Examples 5.1 and 5.2.

With the notation introduced in Section 3.5, we can also write the above equation as

$$[\mathbf{w}]_{\mathcal{B}} = \begin{bmatrix} \frac{1}{6} \\ \frac{5}{2} \\ \frac{2}{3} \end{bmatrix}$$

**Definition** A set of vectors in  $\mathbb{R}^n$  is an *orthonormal set* if it is an orthogonal set of unit vectors. An *orthonormal basis* for a subspace  $W$  of  $\mathbb{R}^n$  is a basis of  $W$  that is an orthonormal set.

**Remark** If  $S = \{\mathbf{q}_1, \dots, \mathbf{q}_k\}$  is an orthonormal set of vectors, then  $\mathbf{q}_i \cdot \mathbf{q}_j = 0$  for  $i \neq j$  and  $\|\mathbf{q}_i\| = 1$ . The fact that each  $\mathbf{q}_i$  is a unit vector is equivalent to  $\mathbf{q}_i \cdot \mathbf{q}_i = 1$ . It follows that we can summarize the statement that  $S$  is orthonormal as

$$\mathbf{q}_i \cdot \mathbf{q}_j = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

**Example 5.5**

Show that  $S = \{\mathbf{q}_1, \mathbf{q}_2\}$  is an orthonormal set in  $\mathbb{R}^3$  if

$$\mathbf{q}_1 = \begin{bmatrix} 1/\sqrt{3} \\ -1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix} \quad \text{and} \quad \mathbf{q}_2 = \begin{bmatrix} 1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{bmatrix}$$

**Example 5.6**

Construct an orthonormal basis for  $\mathbb{R}^3$  from the vectors in Example 5.1.

$\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is an orthogonal set in  $\mathbb{R}^3$  if

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

**Theorem 5.3**

Let  $\{\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_k\}$  be an orthonormal basis for a subspace  $W$  of  $\mathbb{R}^n$  and let  $\mathbf{w}$  be any vector in  $W$ . Then

$$\mathbf{w} = (\mathbf{w} \cdot \mathbf{q}_1)\mathbf{q}_1 + (\mathbf{w} \cdot \mathbf{q}_2)\mathbf{q}_2 + \cdots + (\mathbf{w} \cdot \mathbf{q}_k)\mathbf{q}_k$$

and this representation is unique.

**Proof** Apply Theorem 5.2 and use the fact that  $\mathbf{q}_i \cdot \mathbf{q}_i = 1$  for  $i = 1, \dots, k$ .

**Theorem 5.2**

Let  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  be an orthogonal basis for a subspace  $W$  of  $\mathbb{R}^n$  and let  $\mathbf{w}$  be any vector in  $W$ . Then the unique scalars  $c_1, \dots, c_k$  such that

$$\mathbf{w} = c_1\mathbf{v}_1 + \cdots + c_k\mathbf{v}_k$$

are given by

$$c_i = \frac{\mathbf{w} \cdot \mathbf{v}_i}{\mathbf{v}_i \cdot \mathbf{v}_i} \text{ for } i = 1, \dots, k$$

**Theorem 5.4**

The columns of an  $m \times n$  matrix  $Q$  form an orthonormal set if and only if  $Q^T Q = I_n$ .

**Definition**

An  $n \times n$  matrix  $Q$  whose columns form an orthonormal set is called an *orthogonal matrix*.

**Theorem 5.5**

A square matrix  $Q$  is orthogonal if and only if  $Q^{-1} = Q^T$ .

**Proof** By Theorem 5.4,  $Q$  is orthogonal if and only if  $Q^T Q = I$ . This is true if and only if  $Q$  is invertible and  $Q^{-1} = Q^T$ , by Theorem 3.13.

**Example 5.7**

Show that the following matrices are orthogonal and find their inverses:

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

**Example 5.7**

Show that the following matrices are orthogonal and find their inverses:

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

Therefore,  $B$  is orthogonal, by Theorem 5.5, and

$$B^{-1} = B^T = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix}$$

The word *isometry* literally means “length preserving,” since it is derived from the Greek roots *isos* (“equal”) and *metron* (“measure”).

Matrix  $B$  is the matrix of a rotation through the angle  $\theta$  in  $\mathbb{R}^2$ .

Any rotation has the property that it is a *length-preserving* transformation (known as an *isometry* in geometry).

every orthogonal matrix transformation is an isometry.

**Theorem 5.6**

Let  $Q$  be an  $n \times n$  matrix. The following statements are equivalent:

- $Q$  is orthogonal.
- $\|Q\mathbf{x}\| = \|\mathbf{x}\|$  for every  $\mathbf{x}$  in  $\mathbb{R}^n$ .
- $Q\mathbf{x} \cdot Q\mathbf{y} = \mathbf{x} \cdot \mathbf{y}$  for every  $\mathbf{x}$  and  $\mathbf{y}$  in  $\mathbb{R}^n$ .

**Proof** We will prove that (a)  $\Rightarrow$  (c)  $\Rightarrow$  (b)  $\Rightarrow$  (a). To do so, we will need to make use of the fact that if  $\mathbf{x}$  and  $\mathbf{y}$  are (column) vectors in  $\mathbb{R}^n$ , then  $\mathbf{x} \cdot \mathbf{y} = \mathbf{x}^T \mathbf{y}$ .

(b)  $\Rightarrow$  (c)

Now if  $\mathbf{e}_i$  is the  $i$ th standard basis vector, then  $\mathbf{q}_i = Q\mathbf{e}_i$ . Consequently,

$$\mathbf{q}_i \cdot \mathbf{q}_j = Q\mathbf{e}_i \cdot Q\mathbf{e}_j = \mathbf{e}_i \cdot \mathbf{e}_j = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

Thus, the columns of  $Q$  form an orthonormal set, so  $Q$  is an orthogonal matrix.

$$\|\mathbf{x} \pm \mathbf{y}\|^2 = \|\mathbf{x}\|^2 \pm 2\mathbf{x} \cdot \mathbf{y} + \|\mathbf{y}\|^2$$

**Theorem 5.7**

If  $Q$  is an orthogonal matrix, then its rows form an orthonormal set.

**Proof** From Theorem 5.5, we know that  $Q^{-1} = Q^T$ . Therefore,

$$(Q^T)^{-1} = (Q^{-1})^{-1} = Q = (Q^T)^T$$

so  $Q^T$  is an orthogonal matrix. Thus, the columns of  $Q^T$ —which are just the rows of  $Q$ —form an orthonormal set.

**Theorem 5.8**

Let  $Q$  be an orthogonal matrix.

- $Q^{-1}$  is orthogonal.
- $\det Q = \pm 1$
- If  $\lambda$  is an eigenvalue of  $Q$ , then  $|\lambda| = 1$ .
- If  $Q_1$  and  $Q_2$  are orthogonal  $n \times n$  matrices, then so is  $Q_1 Q_2$ .

**Proof**

(c) Let  $\lambda$  be an eigenvalue of  $Q$  with corresponding eigenvector  $\mathbf{v}$ . Then  $Q\mathbf{v} = \lambda\mathbf{v}$ , and, using Theorem 5.6(b), we have

$$\|\mathbf{v}\| = \|Q\mathbf{v}\| = \|\lambda\mathbf{v}\| = |\lambda| \|\mathbf{v}\|$$

Since  $\|\mathbf{v}\| \neq 0$ , this implies that  $|\lambda| = 1$ .

**Exercises 5.1**

2, 8, 9, 17, 28