

6.8 ORTHONORMAL BASES IN R^n

From our work with the natural bases for R^2 , R^3 , and, in general, for R^n , we know that when these bases are present, the computations are kept to a minimum. A subspace W of R^n need not contain any of the natural basis vectors, but in this section we want to show that it has a basis with the same properties. That is, we want to show that W contains a basis S such that every vector in S is of unit length and every two vectors in S are orthogonal. The method used to obtain such a basis is the Gram-Schmidt process, which is presented below.

DEFINITION A set $S = \{u_1, u_2, \dots, u_k\}$ in R^n is called **orthogonal** if any two vectors in S are orthogonal, that is, if $u_i \cdot u_j = 0$ for $i \neq j$. An orthogonal set of vectors is an orthogonal set of unit vectors. That is, $S = \{u_1, u_2, \dots, u_k\}$ is orthonormal if $u_i \cdot u_j = 0$ for $i \neq j$, and $u_i \cdot u_i = 1$ for $i = 1, 2, \dots, k$.

EXAMPLE 1

If $\mathbf{x}_1 = (1, 0, 2)$, $\mathbf{x}_2 = (-2, 0, 1)$, and $\mathbf{x}_3 = (0, 1, 0)$, then $\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$ is an orthogonal set in R^3 . The vectors

$$\mathbf{u}_1 = \left(\frac{1}{\sqrt{5}}, 0, \frac{2}{\sqrt{5}} \right) \quad \text{and} \quad \mathbf{u}_2 = \left(-\frac{2}{\sqrt{5}}, 0, \frac{1}{\sqrt{5}} \right)$$

are unit vectors in the directions of \mathbf{x}_1 and \mathbf{x}_2 , respectively. Since \mathbf{x}_3 is also of unit length, it follows that $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{x}_3\}$ is an orthonormal set. Also, $\text{span}\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$ is the same as $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{x}_3\}$. ■

EXAMPLE 2

The natural basis

$$\{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$$

is an orthonormal set in R^3 . More generally, the natural basis in R^n is an orthonormal set. ■

An important result about orthogonal sets is the following theorem.

THEOREM 6.16

Let $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$ be an orthogonal set of nonzero vectors in R^n . Then S is linearly independent.

Proof Consider the equation

$$c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_k\mathbf{u}_k = \mathbf{0}. \tag{1}$$

Taking the dot product of both sides of (1) with \mathbf{u}_i , $1 \leq i \leq k$, we have

$$(c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_k\mathbf{u}_k) \cdot \mathbf{u}_i = \mathbf{0} \cdot \mathbf{u}_i. \tag{2}$$

By properties (c) and (d) of Theorem 4.3, Section 4.2, the left side of (2) is

$$c_1(\mathbf{u}_1 \cdot \mathbf{u}_i) + c_2(\mathbf{u}_2 \cdot \mathbf{u}_i) + \dots + c_k(\mathbf{u}_k \cdot \mathbf{u}_i),$$

and the right side is 0. Since $\mathbf{u}_j \cdot \mathbf{u}_i = 0$ if $i \neq j$, (2) becomes

$$0 = c_i(\mathbf{u}_i \cdot \mathbf{u}_i) = c_i\|\mathbf{u}_i\|^2. \tag{3}$$

By (a) of Theorem 4.3, Section 4.2; $\|\mathbf{u}_i\| \neq 0$, since $\mathbf{u}_i \neq \mathbf{0}$. Hence (3) implies that $c_i = 0$, $1 \leq i \leq k$, and S is linearly independent. ■

COROLLARY 6.6

An orthonormal set of vectors in R^n is linearly independent.

Proof Exercise T.2. ■

DEFINITION

From Theorem 6.9 of Section 6.4 and Corollary 6.6, it follows that an orthonormal set of n vectors in R^n is a basis for R^n (Exercise T.3). An **orthogonal (orthonormal) basis** for a vector space is a basis that is an orthogonal (orthonormal) set.

We have already seen that the computational effort required to solve a given problem is often reduced when we are dealing with the natural basis for R^n . This reduction in computational effort is due to the fact that we are dealing with an orthonormal basis. For example, if $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ is a basis for R^n , then if \mathbf{v} is any vector in V , we can write \mathbf{v} as

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_n\mathbf{u}_n.$$

The coefficients c_1, c_2, \dots, c_n are obtained by solving a linear system of n equations in n unknowns. (See Section 6.3.)

However, if S is orthonormal, we can obtain the same result with much less work. This is the content of the following theorem.

THEOREM 6.17

Let $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ be an orthonormal basis for R^n and \mathbf{v} any vector in R^n . Then

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_n\mathbf{u}_n,$$

where

$$c_i = \mathbf{v} \cdot \mathbf{u}_i \quad 1 \leq i \leq n.$$

Proof Exercise T.4.

COROLLARY 6.7

Let $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ be an orthogonal basis for R^n and \mathbf{v} any vector in R^n . Then

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_n\mathbf{u}_n,$$

where

$$c_i = \frac{\mathbf{v} \cdot \mathbf{u}_i}{\mathbf{u}_i \cdot \mathbf{u}_i} \quad 1 \leq i \leq n.$$

Proof Exercise T.4(b).

EXAMPLE 3

Let $S = \{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ be an orthonormal basis for R^3 , where

$$\mathbf{u}_1 = \left(\frac{2}{3}, -\frac{2}{3}, \frac{1}{3}\right), \quad \mathbf{u}_2 = \left(\frac{2}{3}, \frac{1}{3}, -\frac{2}{3}\right), \quad \text{and} \quad \mathbf{u}_3 = \left(\frac{1}{3}, \frac{2}{3}, \frac{2}{3}\right).$$

Write the vector $\mathbf{v} = (3, 4, 5)$ as a linear combination of the vectors in S .

Solution We have

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + c_3\mathbf{u}_3.$$

Theorem 6.17 shows that c_1 , c_2 , and c_3 can be obtained without having to solve a linear system of three equations in three unknowns. Thus

$$c_1 = \mathbf{v} \cdot \mathbf{u}_1 = 1, \quad c_2 = \mathbf{v} \cdot \mathbf{u}_2 = 0, \quad c_3 = \mathbf{v} \cdot \mathbf{u}_3 = 7,$$

and $\mathbf{v} = \mathbf{u}_1 + 7\mathbf{u}_3$.

THEOREM 6.18

(Gram*-Schmidt** Process) Let W be a nonzero subspace of R^n with an orthogonal basis $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$. Then there exists an orthonormal basis $T = \{\mathbf{w}_1, \mathbf{w}_2, \dots, \mathbf{w}_m\}$ for W .

Proof The proof is constructive; that is, we develop the desired basis T . How we first find an orthogonal basis $T^* = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$ for W .

First, we pick any one of the vectors in S , say \mathbf{u}_1 , and call it \mathbf{v}_1 . $\mathbf{v}_1 = \mathbf{u}_1$. We now look for a vector \mathbf{v}_2 in the subspace W_1 of W spanned by $\{\mathbf{u}_1, \mathbf{u}_2\}$ that is orthogonal to \mathbf{v}_1 . Since $\mathbf{v}_1 = \mathbf{u}_1$, W_1 is also the subspace spanned by $\{\mathbf{v}_1, \mathbf{u}_2\}$. Thus

$$\mathbf{v}_2 = c_1\mathbf{v}_1 + c_2\mathbf{u}_2.$$

We try to determine c_1 and c_2 so that $\mathbf{v}_1 \cdot \mathbf{v}_2 = 0$. Now

$$0 = \mathbf{v}_2 \cdot \mathbf{v}_1 = (c_1\mathbf{v}_1 + c_2\mathbf{u}_2) \cdot \mathbf{v}_1 = c_1(\mathbf{v}_1 \cdot \mathbf{v}_1) + c_2(\mathbf{u}_2 \cdot \mathbf{v}_1).$$

*Jørgen Pederson Gram (1850–1916) was a Danish actuary.

**Erhard Schmidt (1876–1959) taught at several leading German Universities and was a student of both Hermann Amandus Schwarz and David Hilbert. He made important contributions to the study of integral equations and partial differential equations and, as part of this work, introduced the method for finding an orthonormal basis in 1907. In 1908 he wrote a paper on infinitely many linear equations in infinitely many unknowns, in which he founded the theory of Hilbert spaces and in which he again used his method.

Since $v_1 \neq 0$ (why?), $v_1 \cdot v_1 \neq 0$, and solving for c_1 and c_2 in (4), we have

$$c_1 = -c_2 \frac{u_2 \cdot v_1}{v_1 \cdot v_1}.$$

We may assign an arbitrary nonzero value to c_2 . Thus, letting $c_2 = 1$, we obtain

$$c_1 = -\frac{u_2 \cdot v_1}{v_1 \cdot v_1}.$$

Hence

$$v_2 = c_1 v_1 + c_2 u_2 = u_2 - \left(\frac{u_2 \cdot v_1}{v_1 \cdot v_1} \right) v_1.$$

Notice that at this point we have an orthogonal subset $\{v_1, v_2\}$ of W (see Figure 6.10).

Next, we look for a vector v_3 in the subspace W_2 of W spanned by $\{u_1, u_2, u_3\}$ that is orthogonal to both v_1 and v_2 . Of course, W_2 is also the subspace spanned by $\{v_1, v_2, u_3\}$ (why?). Thus

$$v_3 = d_1 v_1 + d_2 v_2 + d_3 u_3.$$

We now try to find d_1 and d_2 so that

$$v_3 \cdot v_1 = 0 \quad \text{and} \quad v_3 \cdot v_2 = 0.$$

Now

$$0 = v_3 \cdot v_1 = (d_1 v_1 + d_2 v_2 + d_3 u_3) \cdot v_1 = d_1 (v_1 \cdot v_1) + d_3 (u_3 \cdot v_1). \quad (5)$$

$$0 = v_3 \cdot v_2 = (d_1 v_1 + d_2 v_2 + d_3 u_3) \cdot v_2 = d_2 (v_2 \cdot v_2) + d_3 (u_3 \cdot v_2). \quad (6)$$

In obtaining the right sides of (5) and (6), we have used the fact that $v_1 \cdot v_2 = 0$. Observe that $v_2 \neq 0$ (why?). Solving (5) and (6) for d_1 and d_2 , respectively, we obtain

$$d_1 = -d_3 \frac{u_3 \cdot v_1}{v_1 \cdot v_1} \quad \text{and} \quad d_2 = -d_3 \frac{u_3 \cdot v_2}{v_2 \cdot v_2}.$$

We may assign an arbitrary nonzero value to d_3 . Thus, letting $d_3 = 1$, we have

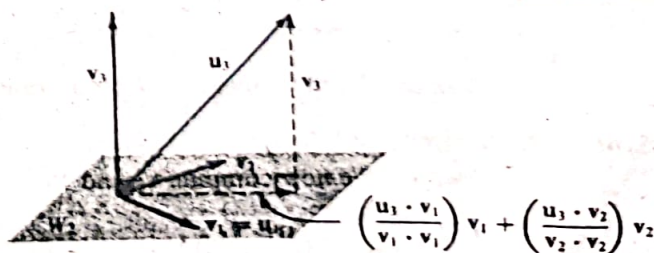
$$d_1 = -\frac{u_3 \cdot v_1}{v_1 \cdot v_1} \quad \text{and} \quad d_2 = -\frac{u_3 \cdot v_2}{v_2 \cdot v_2}.$$

Hence

$$v_3 = u_3 - \left(\frac{u_3 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{u_3 \cdot v_2}{v_2 \cdot v_2} \right) v_2.$$

Notice that at this point we have an orthogonal subset $\{v_1, v_2, v_3\}$ of W (see Figure 6.11).

Figure 6.11 ▶



We next seek a vector v_4 in the subspace W_3 of W spanned by the vectors $\{u_1, u_2, u_3, u_4\}$, and thus by $\{v_1, v_2, v_3, u_4\}$, that is orthogonal to v_1, v_2, v_3 . We can then write

$$v_4 = u_4 - \left(\frac{u_4 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{u_4 \cdot v_2}{v_2 \cdot v_2} \right) v_2 - \left(\frac{u_4 \cdot v_3}{v_3 \cdot v_3} \right) v_3.$$

We continue in this manner until we have an orthogonal set $T^* = \{v_1, v_2, \dots, v_m\}$ of m vectors. It then follows that T^* is a basis for W . If we normalize the v_i , that is, let

$$w_i = \frac{1}{\|v_i\|} v_i \quad (1 \leq i \leq m),$$

then $T = \{w_1, w_2, \dots, w_m\}$ is an orthonormal basis for W .

We now summarize the Gram-Schmidt process.

The Gram-Schmidt process for computing an orthonormal basis $T = \{w_1, w_2, \dots, w_m\}$ for a nonzero subspace W of R^n with basis $S = \{u_1, u_2, \dots, u_m\}$ is as follows.

Step 1. Let $v_1 = u_1$.

Step 2. Compute the vectors v_2, v_3, \dots, v_m , successively, one at a time, by the formula

$$v_i = u_i - \left(\frac{u_i \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{u_i \cdot v_2}{v_2 \cdot v_2} \right) v_2 - \dots - \left(\frac{u_i \cdot v_{i-1}}{v_{i-1} \cdot v_{i-1}} \right) v_{i-1}.$$

The set of vectors $T^* = \{v_1, v_2, \dots, v_m\}$ is an orthogonal set.

Step 3. Let

$$w_i = \frac{1}{\|v_i\|} v_i \quad (1 \leq i \leq m).$$

Then $T = \{w_1, w_2, \dots, w_m\}$ is an orthonormal basis for W .

Remark

It is not difficult to show that if u and v are vectors in R^n such that $u \cdot v = 0$, then $u \cdot (cv) = 0$ for any scalar c (Exercise T.7). This result can often be used to simplify hand computations in the Gram-Schmidt process. As a vector v_i is computed in Step 2, multiply it by a proper scalar to clear fractions that may be present. We shall use this approach in our computational work with the Gram-Schmidt process.

EXAMPLE 4

Consider the basis $S = \{u_1, u_2, u_3\}$ for R^3 , where

$$u_1 = (1, 1, 1), \quad u_2 = (-1, 0, -1), \quad \text{and} \quad u_3 = (-1, 2, 3).$$

Use the Gram-Schmidt process to transform S to an orthonormal basis for R^3 .

Solution

Step 1. Let $v_1 = u_1 = (1, 1, 1)$.

Step 2. We now compute v_2 and v_3 :

$$v_2 = u_2 - \left(\frac{u_2 \cdot v_1}{v_1 \cdot v_1} \right) v_1 = (-1, 0, -1) - \left(\frac{-2}{3} \right) (1, 1, 1) = \left(-\frac{1}{3}, -\frac{2}{3}, -\frac{4}{3} \right).$$

Multiplying v_2 by 3 to clear fractions, we obtain $(-1, 2, -1)$, which we now use as v_2 . Then

$$\begin{aligned} v_3 &= u_3 - \left(\frac{u_3 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{u_3 \cdot v_2}{v_2 \cdot v_2} \right) v_2 \\ &= (-1, 2, 3) - \frac{4}{3}(1, 1, 1) - \frac{2}{6}(-1, 2, -1) = (-2, 0, 2). \end{aligned}$$

Thus

$$T^* = \{v_1, v_2, v_3\} = \{(1, 1, 1), (-1, 2, -1), (-2, 0, 2)\}$$

is an orthogonal basis for R^3 .

Step 3. Let

$$\begin{aligned} w_1 &= \frac{1}{\|v_1\|} v_1 = \frac{1}{\sqrt{3}} (1, 1, 1) \\ w_2 &= \frac{1}{\|v_2\|} v_2 = \frac{1}{\sqrt{6}} (-1, 2, -1) \\ w_3 &= \frac{1}{\|v_3\|} v_3 = \frac{1}{\sqrt{8}} (-2, 0, 2) = \left(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right). \end{aligned}$$

Then

$$\begin{aligned} T &= \{w_1, w_2, w_3\} \\ &= \left\{ \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right), \left(-\frac{1}{\sqrt{6}}, \frac{2}{\sqrt{6}}, -\frac{1}{\sqrt{6}} \right), \left(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right) \right\} \end{aligned}$$

is an orthonormal basis for R^3 . ■

EXAMPLE 5

Let W be the subspace of R^4 with basis $S = \{u_1, u_2, u_3\}$, where

$$u_1 = (1, -2, 0, 1), \quad u_2 = (-1, 0, 0, -1), \quad \text{and} \quad u_3 = (1, 1, 0, 0).$$

Use the Gram-Schmidt process to transform S to an orthonormal basis for W .

Solution *Step 1.* Let $v_1 = u_1 = (1, -2, 0, 1)$.

Step 2. We now compute v_2 and v_3 :

$$\begin{aligned} v_2 &= u_2 - \left(\frac{u_2 \cdot v_1}{v_1 \cdot v_1} \right) v_1 = (-1, 0, 0, -1) - \left(\frac{-2}{6} \right) (1, -2, 0, 1) \\ &= \left(-\frac{2}{3}, -\frac{2}{3}, 0, -\frac{2}{3} \right). \end{aligned}$$

Multiplying v_2 by 3 to clear fractions, we obtain $(-2, -2, 0, -2)$, which we now use as v_2 . Then

$$\begin{aligned} v_3 &= u_3 - \left(\frac{u_3 \cdot v_1}{v_1 \cdot v_1} \right) v_1 - \left(\frac{u_3 \cdot v_2}{v_2 \cdot v_2} \right) v_2 \\ &= (1, 1, 0, 0) - \left(\frac{-1}{6} \right) (1, -2, 0, 1) - \left(\frac{-4}{12} \right) (-2, -2, 0, -2) \\ &= \left(\frac{1}{2}, 0, 0, -\frac{1}{2} \right). \end{aligned}$$

Multiplying v_3 by 2 to clear fractions, we obtain $(1, 0, 0, -1)$, which we now use as v_3 . Thus

$$T^* = \{(1, -2, 0, 1), (-2, -2, 0, -2), (1, 0, 0, -1)\}$$

is an orthogonal basis for W .

Step 3. Let

$$w_1 = \frac{1}{\|v_1\|} v_1 = \frac{1}{\sqrt{6}} (1, -2, 0, 1)$$

$$w_2 = \frac{1}{\|v_2\|} v_2 = \frac{1}{\sqrt{12}} (-2, -2, 0, -2) = \frac{1}{\sqrt{3}} (-1, -1, 0, -1)$$

$$w_3 = \frac{1}{\|v_3\|} v_3 = \frac{1}{\sqrt{2}} (1, 0, 0, -1).$$

Then

$$T = \{w_1, w_2, w_3\}$$

$$= \left\{ \left(\frac{1}{\sqrt{6}}, -\frac{2}{\sqrt{6}}, 0, \frac{1}{\sqrt{6}} \right), \left(-\frac{1}{\sqrt{3}}, -\frac{1}{\sqrt{3}}, 0, -\frac{1}{\sqrt{3}} \right), \left(\frac{1}{\sqrt{2}}, 0, 0, -\frac{1}{\sqrt{2}} \right) \right\}$$

is an orthonormal basis for W .

Remarks

1. In solving Example 5, as soon as a vector is computed we multiply by an appropriate scalar to eliminate any fractions that may be present. This optional step results in simpler computations when working by hand. If this approach is taken, the resulting basis, while orthonormal, differs from the orthonormal basis obtained by not clearing fractions. Computer implementations of the Gram–Schmidt process, including those developed with MATLAB, do not clear fractions.
2. We make one final observation with regard to the Gram–Schmidt process. In our proof of Theorem 6.18 we first obtained an orthogonal basis T^* and then normalized all the vectors in T^* to obtain the orthonormal basis T . Of course, an alternative course of action is to normalize each vector in T^* as soon as we produce it.

6.8 Exercises

1. Which of the following are orthogonal sets of vectors?
 - (a) $\{(1, -1, 2), (0, 2, -1), (-1, 1, 1)\}$.
 - (b) $\{(1, 2, -1, 1), (0, -1, -2, 0), (1, 0, 0, -1)\}$.
 - (c) $\{(0, 1, 0, -1), (1, 0, 1, 1), (-1, 1, -1, 2)\}$.
2. Which of the following are orthonormal sets of vectors?
 - (a) $\left\{ \left(\frac{1}{3}, \frac{2}{3}, \frac{2}{3} \right), \left(\frac{2}{3}, \frac{1}{3}, -\frac{2}{3} \right), \left(\frac{2}{3}, -\frac{2}{3}, \frac{1}{3} \right) \right\}$.
 - (b) $\left\{ \left(\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}} \right), \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right), (0, 1, 0) \right\}$.
 - (c) $\{(0, 2, 2, 1), (1, 1, -2, 2), (0, -2, 1, 2)\}$.
3. Let $u = (1, 1, -2)$ and $v = (a, -1, 2)$. For what values of a are u and v orthogonal?
4. Let $u = \left(\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right)$, and $v = \left(a, \frac{1}{\sqrt{2}}, -b \right)$. For what values of a and b is $\{u, v\}$ an orthonormal set?
5. Use the Gram–Schmidt process to find an orthonormal basis for the subspace of R^3 with basis $\{(1, -1, 0), (2, 0, 1)\}$.
6. Use the Gram–Schmidt process to find an orthonormal basis for the subspace of R^3 with basis $\{(1, 0, 2), (-1, 1, 0)\}$.
7. Use the Gram–Schmidt process to find an orthonormal basis for the subspace of R^4 with basis $\{(1, -1, 0, 1), (2, 0, 0, -1), (0, 0, 1, 0)\}$.
8. Use the Gram–Schmidt process to find an orthonormal basis for the subspace of R^4 with basis $\{(1, 1, -1, 0), (0, 2, 0, 1), (-1, 0, 0, 1)\}$.
9. Use the Gram–Schmidt process to transform the basis $\{(1, 2), (-3, 4)\}$ for R^2 into (a) an orthogonal basis and (b) an orthonormal basis.

In Exercises 3 and 4, let $V = R^3$.

18. Find an orthonormal basis for the solution space of the homogeneous system

$$\begin{aligned} x_1 + x_2 - x_3 &= 0 \\ 2x_1 + x_2 + 2x_3 &= 0. \end{aligned}$$

19. Find an orthonormal basis for the solution space of the homogeneous system

$$\begin{bmatrix} 1 & 1 & -1 \\ 2 & 1 & 3 \\ 1 & 2 & -6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}.$$

20. Consider the orthonormal basis

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

for R^2 . Using Theorem 6.17 write the vector $(2, 3)$ as a linear combination of the vectors in S .

21. Consider the orthonormal basis

$$S = \left\{ \left(\frac{1}{\sqrt{5}}, 0, \frac{2}{\sqrt{5}} \right), \left(-\frac{2}{\sqrt{5}}, 0, \frac{1}{\sqrt{5}} \right), (0, 1, 0) \right\}$$

for R^3 . Using Theorem 6.17 write the vector $(2, -3, 1)$ as a linear combination of the vectors in S .

Exercises

Let the natural basis for R^n is an orthonormal

Corollary 6.6.

Let an orthonormal set of n vectors in R^n is a

Theorem 6.17.

Corollary 6.7.

Let v_1, v_2, \dots, v_n be vectors in R^n . Show that if u is orthogonal to v_1, v_2, \dots, v_n , then u is orthogonal to any vector in $\text{span}\{v_1, v_2, \dots, v_n\}$.

Let u be a fixed vector in R^n . Show that the set of all vectors in R^n that are orthogonal to u is a subspace.

Let u and v be vectors in R^n . Show that if $u \cdot v = 0$, then $(cu) \cdot v = 0$ for any scalar c .

Exercises

The Gram-Schmidt process takes a basis S for a subspace W and produces an orthonormal basis T for W . The

process produces the orthonormal basis T that is given by the following algorithm. The algorithm is implemented in MATLAB in routine `gschmidt`.

Type `help gschmidt` for directions.

- ML.1. Use `gschmidt` to produce an orthonormal basis for R^3 from the basis

$$S = \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}.$$

Your answer will be in decimal form; rewrite it in terms of $\sqrt{2}$.

ML.2. Use **gschmidt** to produce an orthonormal basis for R^4 from the basis $S = \{(1, 0, 1, 1), (1, 2, 1, 3), (0, 2, 1, 1), (0, 1, 0, 0)\}$.

ML.3. In R^3 , $S = \{(0, -1, 1), (0, 1, 1), (1, 1, 1)\}$ is a basis. Find an orthonormal basis T from S and then find $[v]_T$ for each of the following vectors.

(a) $v = (1, 2, 0)$.

(b) $v = (1, 1, 1)$.

(c) $v = (-1, 0, 1)$.

ML.4. Find an orthonormal basis for the subspace consisting of all vectors of the form

$$(a, 0, a + b, b + c),$$

where $a, b,$ and c are any real numbers.

6.9 ORTHOGONAL COMPLEMENTS

Let W_1 and W_2 be subspaces of a vector space V . Let $W_1 + W_2$ be the set of all vectors v in V such that $v = w_1 + w_2$, where w_1 is in W_1 and w_2 is in W_2 . In Exercise T.10 in Section 6.2, we asked you to show that $W_1 + W_2$ is a subspace of V . In Exercise T.11 in Section 6.2, we asked you to show that $V = W_1 + W_2$ and $W_1 \cap W_2 = \{0\}$, then V is the direct sum of W_1 and W_2 and we write $V = W_1 \oplus W_2$. Moreover, in this case every vector in V is uniquely written as $w_1 + w_2$, where w_1 is in W_1 and w_2 is in W_2 . In this section we show that if W is a subspace of R^n , then R^n can be written as the direct sum of W and another subspace of R^n . This subspace will be used to establish a basic relationship between four vector spaces associated with a matrix.

DEFINITION

Let W be a subspace of R^n . A vector u in R^n is said to be **orthogonal** to W if it is orthogonal to every vector in W . The set of all vectors in R^n orthogonal to all the vectors in W is called the **orthogonal complement** of W in R^n and is denoted by W^\perp (read “ W perp”).

EXAMPLE 1

Let W be the subspace of R^3 consisting of all multiples of the vector

$$w = (2, -3, 4).$$

Thus $W = \text{span}\{w\}$, so W is a one-dimensional subspace of R^3 . The vector u in R^3 belongs to W^\perp if and only if u is orthogonal to w , for any c . It can be shown that W^\perp is the plane with normal w .

Observe that if W is a subspace of R^n , then the zero vector always belongs to W^\perp (Exercise T.1). Moreover, the orthogonal complement of W^\perp is W (Exercise T.2).

THEOREM 6.19

Let W be a subspace of R^n . Then

- W^\perp is a subspace of R^n .
- $W \cap W^\perp = \{0\}$.

Proof

- Let u_1 and u_2 be in W^\perp . Then u_1 and u_2 are orthogonal to every vector in W . We now have

$$(u_1 + u_2) \cdot w = u_1 \cdot w + u_2 \cdot w = 0 + 0 = 0,$$

so $u_1 + u_2$ is in W^\perp . Also, let u be in W^\perp and let c be a real number. For any vector w in W , we have

$$(cu) \cdot w = c(u \cdot w) = c \cdot 0 = 0,$$

so cu is in W , which implies that W^\perp is closed under vector addition and scalar multiplication and hence is a subspace of R^n .

- (b) Let u be a vector in $W \cap W^\perp$. Then u is in both W and W^\perp , so $u \cdot u = 0$. From Theorem 4.3 in Section 4.2 it follows that $u = 0$. ■

In Exercise T.3 we ask you to show that if W is a subspace of R^n that is spanned by a set of vectors S , then a vector u in R^n belongs to W^\perp if and only if u is orthogonal to every vector in S . This result can be helpful in finding W^\perp , as shown in the next example.

EXAMPLE 2

Let W be the subspace of R^4 with basis $\{w_1, w_2\}$, where

$$w_1 = (1, 1, 0, 1) \quad \text{and} \quad w_2 = (0, -1, 1, 1).$$

Find a basis for W^\perp .

Solution

Let $u = (a, b, c, d)$ be a vector in W^\perp . Then $u \cdot w_1 = 0$ and $u \cdot w_2 = 0$. Thus we have

$$u \cdot w_1 = a + b + d = 0 \quad \text{and} \quad u \cdot w_2 = -b + c + d = 0.$$

Solving the homogeneous system

$$\begin{aligned} a + b + d &= 0 \\ -b + c + d &= 0, \end{aligned}$$

we obtain (verify)

$$a = -r - 2s, \quad b = r + s, \quad c = r, \quad d = s.$$

Then

$$u = (-r - 2s, r + s, r, s) = r(-1, 1, 1, 0) + s(-2, 1, 0, 1).$$

Hence the vectors $(-1, 1, 1, 0)$ and $(-2, 1, 0, 1)$ span W^\perp . Since they are not multiples of each other, they are linearly independent and thus form a basis for W^\perp . ■

THEOREM 6.20

Let W be a subspace of R^n . Then

$$R^n = W \oplus W^\perp.$$

Proof Let $\dim W = m$. Then W has a basis consisting of m vectors. By the Gram-Schmidt process we can transform this basis to an orthonormal basis. Thus let $S = \{w_1, w_2, \dots, w_m\}$ be an orthonormal basis for W . If v is a vector in R^n , let

$$w = (v \cdot w_1)w_1 + (v \cdot w_2)w_2 + \dots + (v \cdot w_m)w_m$$

and

$$u = v - w.$$

Since w is a linear combination of vectors in S , w belongs to W . We show that u lies in W^\perp by showing that u is orthogonal to every vector in a basis for W . For each w_i in S , we have

$$\begin{aligned} u \cdot w_i &= (v - w) \cdot w_i = v \cdot w_i - w \cdot w_i \\ &= v \cdot w_i - [(v \cdot w_1)w_1 + (v \cdot w_2)w_2 + \cdots + (v \cdot w_m)w_m] \cdot w_i \\ &= v \cdot w_i - (v \cdot w_i)(w_i \cdot w_i) \\ &= 0 \end{aligned}$$

since $w_i \cdot w_j = 0$ for $i \neq j$ and $w_i \cdot w_i = 1$, $1 \leq i \leq m$. Thus u is orthogonal to every vector in W and so lies in W^\perp . Hence

$$v = w + u,$$

which means that $R^n = W + W^\perp$. From part (b) of Theorem 6.19, it follows that

$$R^n = W \oplus W^\perp.$$

Remark As pointed out at the beginning of this section, we also conclude that the vectors w and u defined by Equations (1) and (2) are unique.

THEOREM 6.21

If W is a subspace of R^n , then

$$(W^\perp)^\perp = W.$$

Proof First, if w is any vector in W , then w is orthogonal to every vector u in W^\perp , so w is in $(W^\perp)^\perp$. Hence W is a subspace of $(W^\perp)^\perp$. Conversely, let v be an arbitrary vector in $(W^\perp)^\perp$. Then, by Theorem 6.20, v can be written as

$$v = w + u,$$

where w is in W and u is in W^\perp . Since u is in W^\perp , it is orthogonal to w . Thus

$$0 = u \cdot v = u \cdot (w + u) = u \cdot w + u \cdot u = u \cdot u$$

or

$$u \cdot u = 0,$$

which implies that $u = 0$. Then $v = w$, so v belongs to W . Hence it follows that $(W^\perp)^\perp = W$.

Remark Since W is the orthogonal complement of W^\perp and W^\perp is also the orthogonal complement of W , we say that W and W^\perp are **orthogonal complements**.

Relations Among the Fundamental Vector Spaces Associated with a Matrix

If A is a given $m \times n$ matrix, we associate the following four fundamental vector spaces with A : the null space of A , the row space of A , the null space of A^T , and the column space of A . The following theorem shows that these four vector spaces are orthogonal complements.