

$$11. \text{ a. } \alpha_n = \arccos \frac{\vec{u} \cdot \vec{v}}{\|\vec{u}\| \|\vec{v}\|} = \arccos \frac{1}{\sqrt{n}}$$

$$\alpha_2 = \arccos \frac{1}{\sqrt{2}} = \frac{\pi}{4} (= 45^\circ)$$

$$\alpha_3 = \arccos \frac{1}{\sqrt{3}} \approx 0.955 \text{ (radians)}$$

$$\alpha_4 = \arccos \frac{1}{2} = \frac{\pi}{3} (= 60^\circ)$$

b. Since  $y = \arccos(x)$  is a continuous function,  $\lim_{n \rightarrow \infty} \alpha_n = \arccos\left(\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}}\right) = \arccos(0) = \frac{\pi}{2} (= 90^\circ)$

16. You may be able to find the solutions by educated guessing. Here is the systematic approach: we first find all vectors  $\vec{x}$  that are orthogonal to  $\vec{v}_1$ ,  $\vec{v}_2$ , and  $\vec{v}_3$ , then we identify the unit vectors among them. Finding the vectors  $\vec{x}$  with  $\vec{x} \cdot \vec{v}_1 = \vec{x} \cdot \vec{v}_2 = \vec{x} \cdot \vec{v}_3 = 0$  amounts to solving the system

$$\begin{cases} x_1 + x_2 + x_3 + x_4 = 0 \\ x_1 + x_2 - x_3 - x_4 = 0 \\ x_1 - x_2 + x_3 - x_4 = 0 \end{cases}$$

(we can omit all the coefficients  $\frac{1}{2}$ ).

$$\text{The solutions are of the form } \vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} t \\ -t \\ -t \\ t \end{bmatrix}.$$

Since  $\|\vec{x}\| = 2|t|$ , we have a unit vector if  $t = \frac{1}{2}$  or  $t = -\frac{1}{2}$ . Thus there are two possible choices for  $\vec{u}_4$ :

$$\begin{bmatrix} \frac{1}{2} \\ -\frac{1}{2} \\ -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix} \text{ and } \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \\ -\frac{1}{2} \end{bmatrix}.$$

22. Let  $W = \{\vec{x} \text{ in } \mathbb{R}^n : \vec{x} \cdot \vec{v}_i = 0 \text{ for all } i = 1, \dots, m\}$ . We are asked to show that  $V^\perp = W$ , that is, any  $\vec{x}$  in  $V^\perp$  is in  $W$ , and vice versa.  
 If  $\vec{x}$  is in  $V^\perp$ , then  $\vec{x} \cdot \vec{v} = 0$  for all  $\vec{v}$  in  $V$ ; in particular,  $\vec{x} \cdot \vec{v}_i = 0$  for all  $i$  (since the  $\vec{v}_i$  are in  $V$ ), so that  $\vec{x}$  is in  $W$ .

Conversely, consider a vector  $\vec{x}$  in  $W$ . To show that  $\vec{x}$  is in  $V^\perp$ , we have to verify that  $\vec{x} \cdot \vec{v} = 0$  for all  $\vec{v}$  in  $V$ . Pick a particular  $\vec{v}$  in  $V$ . Since the  $\vec{v}_i$  span  $V$ , we can write  $\vec{v} = c_1 \vec{v}_1 + \dots + c_m \vec{v}_m$ , for some scalars  $c_i$ . Then  $\vec{x} \cdot \vec{v} = c_1(\vec{x} \cdot \vec{v}_1) + \dots + c_m(\vec{x} \cdot \vec{v}_m) = 0$ , as claimed.

24. Write  $T(\vec{x}) = \text{proj}_V(\vec{x})$  for simplicity.

To prove the linearity of  $T$  we will use the definition of a projection:  $T(\vec{x})$  is in  $V$ , and  $\vec{x} - T(\vec{x})$  is in  $V^\perp$ .

To show that  $T(\vec{x} + \vec{y}) = T(\vec{x}) + T(\vec{y})$ , note that  $T(\vec{x}) + T(\vec{y})$  is in  $V$  (since  $V$  is a subspace), and  $\vec{x} + \vec{y} - (T(\vec{x}) + T(\vec{y})) = (\vec{x} - T(\vec{x})) + (\vec{y} - T(\vec{y}))$  is in  $V^\perp$  (since  $V^\perp$  is a subspace, by Fact 5.1.5).

To show that  $T(k\vec{x}) = kT(\vec{x})$ , note that  $kT(\vec{x})$  is in  $V$  (since  $V$  is a subspace), and  $k\vec{x} - kT(\vec{x}) = k(\vec{x} - T(\vec{x}))$  is in  $V^\perp$  (since  $V^\perp$  is a subspace).

28. Since the three given vectors in the subspace are orthogonal, we have the orthonormal basis

$$\vec{v}_1 = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \vec{v}_2 = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ -1 \\ -1 \end{bmatrix}, \vec{v}_3 = \frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}$$

Now we can use Fact 5.1.6, with  $\vec{x} = \vec{e}_1 : \text{proj}_V \vec{x} = (\vec{v}_1 \cdot \vec{x})\vec{v}_1 + (\vec{v}_2 \cdot \vec{x})\vec{v}_2 + (\vec{v}_3 \cdot \vec{x})\vec{v}_3 = \frac{1}{4} \begin{bmatrix} 1 \\ 1 \\ -1 \\ 1 \end{bmatrix}$

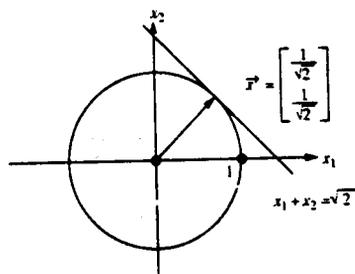
34. Let  $\vec{x}$  be a unit vector in  $\mathbb{R}^n$ , that is,  $\|\vec{x}\| = 1$ . Let  $\vec{y} = \begin{bmatrix} 1 \\ \dots \\ 1 \end{bmatrix}$  (all  $n$  components are 1). The Cauchy-

Schwarz inequality (Fact 5.1.10) tells us that  $|\vec{x} \cdot \vec{y}| \leq \|\vec{x}\| \|\vec{y}\|$ , or,  $|x_1 + \dots + x_n| \leq \|\vec{x}\| \sqrt{n} = \sqrt{n}$ . By Fact 5.1.10, the equation  $x_1 + \dots + x_n = \sqrt{n}$  holds if  $\vec{x} = k\vec{y}$  for positive  $k$ . Thus  $\vec{x}$  must be a unit vector

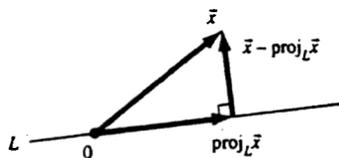
of the form  $\vec{x} = \begin{bmatrix} k \\ \dots \\ k \end{bmatrix}$  for some positive  $k$ . It is required that  $nk^2 = 1$ , or,  $k = \frac{1}{\sqrt{n}}$ . Thus  $\vec{x} = \begin{bmatrix} \frac{1}{\sqrt{n}} \\ \dots \\ \frac{1}{\sqrt{n}} \end{bmatrix}$

(all components are  $\frac{1}{\sqrt{n}}$ ).

The figure below illustrates the case  $n = 2$ .



39. No! By definition of a projection, the vector  $\vec{x} - \text{proj}_L \vec{x}$  is perpendicular to  $\text{proj}_L \vec{x}$ , so that  $(\vec{x} - \text{proj}_L \vec{x}) \cdot (\text{proj}_L \vec{x}) = \vec{x} \cdot \text{proj}_L \vec{x} - \|\text{proj}_L \vec{x}\|^2 = 0$  and  $\vec{x} \cdot \text{proj}_L \vec{x} = \|\text{proj}_L \vec{x}\|^2 \geq 0$ .



**Section 5.2**

4.  $\vec{w}_1 = \frac{1}{5} \begin{bmatrix} 4 \\ 0 \\ 3 \end{bmatrix}$  and  $\vec{w}_2 = \frac{1}{5} \begin{bmatrix} 3 \\ 0 \\ -4 \end{bmatrix}$  as in Exercise 3.

8.  $\vec{w}_1 = \frac{1}{\|\vec{v}_1\|} \vec{v}_1 = \frac{1}{7} \begin{bmatrix} 5 \\ 4 \\ 2 \\ 2 \end{bmatrix}$

Since  $\vec{v}_3$  is orthogonal to  $\vec{w}_1$  and  $\vec{w}_2$ ,  $\vec{w}_3 = \frac{1}{\|\vec{v}_3\|} \vec{v}_3 = \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix}$

$\vec{w}_2 = \frac{\vec{v}_2 - (\vec{w}_1 \cdot \vec{v}_2)\vec{w}_1}{\text{length}} = \frac{1}{7} \begin{bmatrix} -2 \\ 2 \\ 5 \\ -4 \end{bmatrix}$

18.  $Q = \frac{1}{5} \begin{bmatrix} 4 & 3 & 0 \\ 0 & 0 & -5 \\ 3 & -4 & 0 \end{bmatrix}$ ,  $R = \begin{bmatrix} 5 & 5 & 0 \\ 0 & 35 & 0 \\ 0 & 0 & 2 \end{bmatrix}$

22.  $Q = \frac{1}{7} \begin{bmatrix} 5 & -2 \\ 4 & 2 \\ 2 & 5 \\ 2 & -4 \end{bmatrix}$ ,  $R = \begin{bmatrix} 7 & 7 \\ 0 & 7 \end{bmatrix}$

32. A basis of the plane is  $\vec{v}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$ ,  $\vec{v}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$

Now apply the Gram-Schmidt process.

$\vec{w}_1 = \frac{1}{\|\vec{v}_1\|} \vec{v}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$

$\vec{w}_2 = \frac{\vec{v}_2 - (\vec{w}_1 \cdot \vec{v}_2)\vec{w}_1}{\text{length}} = \frac{1}{\sqrt{6}} \begin{bmatrix} -1 \\ -1 \\ 2 \end{bmatrix}$

Your solution may be different if you start with a different basis  $\vec{v}_1, \vec{v}_2$  of the plane.

39.  $\vec{w}_1 = \frac{1}{\sqrt{14}} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ ,  $\vec{w}_2 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$ ,  $\vec{w}_3 = \vec{w}_1 \times \vec{w}_2 = \frac{1}{\sqrt{42}} \begin{bmatrix} -5 \\ 4 \\ -1 \end{bmatrix}$

**Section 5.3**

6. Yes! By Facts 5.3.9b and 5.3.7,  $(A^T)^{-1} = (A^{-1})^T = (A^T)^T$ .  
The equation  $(A^T)^{-1} = (A^T)^T$  shows that  $A^T$  is orthogonal, again by Fact 5.3.7.

Yes! If  $A$  is orthogonal, then so is  $A^T$ , by Exercise 6. Since the columns of  $A$  are orthonormal, so are the rows of  $A^T$ .

8. a. No! As a counterexample, consider  $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$  (see Exercise 4).

b. Yes! More generally, if  $A$  and  $B$  are  $n \times n$  matrices such that  $BA = I_n$ , then  $AB = I_n$ , by Fact 2.4.9c.

14. By Fact 5.3.9a,  $(A^T A)^T = A^T (A^T)^T = A^T A$ .  
The matrix  $AA^T$  is symmetric as well:  $(AA^T)^T = (A^T)^T A^T = AA^T$ .

18. a. The general form of a skew-symmetric  $3 \times 3$  matrix is  $A = \begin{bmatrix} 0 & a & b \\ -a & 0 & c \\ -b & -c & 0 \end{bmatrix}$  with

$A^2 = \begin{bmatrix} -a^2 - b^2 & -bc & ac \\ -bc & -a^2 - c^2 & -ab \\ ac & -ab & -b^2 - c^2 \end{bmatrix}$  a symmetric matrix.

b. By Fact 5.3.9.a,  $(A^2)^T = (A^T)^2 = (-A)^2 = A^2$ , so that  $A^2$  is symmetric.

34. To write the general form of a skew-symmetric  $n \times n$  matrix  $A$ , we can place arbitrary constants above the diagonal, the opposite entries below the diagonal ( $a_{ij} = -a_{ji}$ ), and zeros on the diagonal (since  $a_{ii} = -a_{ii}$ ). See Exercise 33 for the case  $n = 3$ . Thus the dimension of the space equals the number of entries above the diagonal of an  $n \times n$  matrix. In Exercise 35 we will see that there are  $(n^2 - n)/2$  such entries. Thus  $\dim(V) = (n^2 - n)/2$ .

39. The kernel consists of all matrixes  $A$  such that  $L(A) = A - A^T = 0$ , that is,  $A^T = A$ ; those are the symmetric matrices.

The value  $L(A) = A - A^T$  is always skew-symmetric, since  $(A - A^T)^T = A^T - (A^T)^T = -(A - A^T)$ , and every skew-symmetric matrix  $B$  can be written as  $B = L(\frac{1}{2}B)$ . Thus the image consists of all skew-symmetric matrices.

$$\begin{array}{ccc} 16. \text{ If } A \text{ is an } m \times n \text{ matrix, then} & & \\ \dim(\operatorname{im}A) + \dim(\ker A) = m & = & m - \dim(\ker A) = m - \operatorname{rank}(A) \\ \uparrow & & \uparrow \\ \text{Fact 5.4.2a} & & \text{Fact 3.3.8} \end{array}$$

$$\text{and } \dim(\ker(A^T)) = m - \operatorname{rank}(A^T)$$

↑  
Fact 3.3.5

It follows that  $\operatorname{rank}(A) = \operatorname{rank}(A^T)$ , as claimed.

17. Yes! By Fact 5.4.3,  $\ker(A) = \ker(A^T A)$ . Taking dimensions of both sides and using Fact 3.3.5, we find that  $n - \operatorname{rank}(A) = n - \operatorname{rank}(A^T A)$ ; the claim follows.
18. Yes! By Exercise 17,  $\operatorname{rank}(A) = \operatorname{rank}(A^T A)$ . Substituting  $A^T$  for  $A$  in Exercise 17 and using Fact 5.3.9c, we find that  $\operatorname{rank}(A) = \operatorname{rank}(A^T) = \operatorname{rank}(AA^T)$ . The claim follows.
19.  $\vec{x}^* = (A^T A)^{-1} A^T \vec{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ , by Fact 5.4.7.

20. Using Fact 5.4.7, we find  $\vec{x}^* = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$  and  $\vec{b} - A\vec{x}^* = \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}$

Note that  $\vec{b} - A\vec{x}^*$  is perpendicular to the two columns of  $A$ .

21. Using Fact 5.4.7, we find  $\vec{x}^* = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$  and  $\vec{b} - A\vec{x}^* = \begin{bmatrix} -12 \\ 36 \\ -18 \end{bmatrix}$ , so that  $\|\vec{b} - A\vec{x}^*\| = 42$ .

22. Using Fact 5.4.7, we find  $\vec{x}^* = \begin{bmatrix} 3 \\ -2 \end{bmatrix}$  and  $\vec{b} - A\vec{x}^* = \vec{0}$ . This system is in fact consistent and  $\vec{x}^*$  is the exact solution; the error  $\|\vec{b} - A\vec{x}^*\|$  is 0.

23. Using Fact 5.4.7, we find  $\vec{x}^* = \vec{0}$ ; here  $\vec{b}$  is perpendicular to  $\operatorname{im}(A)$ .

24. Using Fact 5.4.7, we find  $\vec{x}^* = [2]$

25. In this case, the normal equation  $A^T A \vec{x} = A^T \vec{b}$  is  $\begin{bmatrix} 5 & 15 \\ 15 & 45 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 5 \\ 15 \end{bmatrix}$ , which simplifies to  $x_1 + 3x_2 = 1$ , or  $x_1 = 1 - 3x_2$ . The solutions are of the form  $\vec{x}^* = \begin{bmatrix} 1 - 3t \\ t \end{bmatrix}$ , where  $t$  is an arbitrary constant.