Answers are provided for all even numbered problems and for some odd numbered problems. If you have a question about a problem that isn't included below, feel free to ask me. If you think you've spotted an error, please let me know.

## Section 5.3

- 2. Is orthogonal.
- 4. Is not orthogonal. The first column is not orthogonal to the third column, and the second column is not orthogonal to the third column.
- 28. There are a few ways to do this. The simplest is to use #27:  $L(\vec{v}) \cdot L(\vec{w}) = (A\vec{v}) \cdot (A\vec{w}) = \vec{v} \cdot (A^T A \vec{w}) = \vec{v} \cdot \vec{w}$  since  $A^T A = I_n$  for symmetric matrices.

Alternatively, you can rewrite the equation  $\|\vec{v} + \vec{w}\|^2 = \|\vec{v}\|^2 + 2(\vec{v} \cdot \vec{w}) + \|\vec{w}\|^2$  as  $\vec{v} \cdot \vec{w} = \frac{1}{2}(\|\vec{v} + \vec{w}\|^2 - \|\vec{v}\|^2 - \|\vec{w}\|^2)$ , so that  $L(\vec{v}) \cdot L(\vec{w}) = \frac{1}{2}(\|L(\vec{v}) + L(\vec{w})\|^2 - \|L(\vec{v})\|^2 - \|L(\vec{w})\|^2) = \frac{1}{2}(\|\vec{v} + \vec{w}\|^2 - \|\vec{v}\|^2 - \|\vec{w}\|^2) = \vec{v} \cdot \vec{w}$  since L is a linear transformation and preserves the length of vectors.

Or, a third method is to write everything in terms of an orthonormal basis and then compute both sides directly. That is, let  $\vec{u_1}, \ldots, \vec{u_n}$  form an orthonormal basis for  $\mathbb{R}^n$  (for example, take  $\vec{u_i} = \vec{e_i}$  for  $i = 1, \ldots, n$ ) and note that  $L(\vec{u_1}), \ldots, L(\vec{u_n})$  will also be an orthonormal basis of  $\mathbb{R}^n$  since L is orthogonal (see Theorem 5.3.2, or Theorem 5.3.3 for the special case  $\vec{u_i} = \vec{e_i}$ ). Write  $\vec{v} = c_1 \vec{u_1} + \cdots + c_n \vec{u_n}$  and  $\vec{w} = d_1 \vec{u_1} + \cdots + d_n \vec{u_n}$ . Then, compute  $\vec{v} \cdot \vec{w} = c_1 d_1 + \cdots + c_n d_n$  and  $L(\vec{v}) \cdot L(\vec{w}) = (c_1 L(\vec{u_1}) + \cdots + c_n L(\vec{u_n})) \cdot (d_1 L(\vec{u_1}) + \cdots + d_n L(\vec{u_n})) = c_1 d_1 + \cdots + c_n d_n$ , so that  $\vec{v} \cdot \vec{w} = L(\vec{v}) \cdot L(\vec{w})$ .

- 30. L preserves length, so the only solution to  $L(\vec{x}) = \vec{0}$  is  $\vec{x} = \vec{0}$ . Thus,  $\ker(L) = \{\vec{0}\}$ . By Rank-Nullity,  $\dim(\operatorname{im}(L)) = m$ . We saw back in chapter 1 that  $\operatorname{rank}(L) \leq n$ , so  $m \leq n$  as  $\operatorname{rank}(L) = m$ . Following the proofs of Theorems 5.3.2 and 5.3.3 (but dropping the assumption that m = n), we see that the columns of A will be orthonormal. Computing  $A^TA$  as in Theorem 5.3.7 (again, dropping the assumption m = n), we see that  $A^TA = I_m$ . From Theorem 5.3.10, we see that  $AA^T$  is the matrix of the orthogonal projection onto the space V spanned by the columns of A, and that the jth column of A is  $\operatorname{proj}_V(\vec{e_j})$ .
- 32. (a) Not necessarily. Counterexamples are easy to find. For example,  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  will work.
  - (b) This is necessarily true, as Summary 5.3.8 tells us that  $A^T = A^{-1}$  in this case.
- 36. Either find a basis of  $\mathbb{R}^3$  containing the first two columns of the matrix as basis vectors and use Gram-Schmidt, or note that  $A^TA = I_3$  and using the resulting equations to solve

for a,b, and c. Either way, you should come up with either  $\begin{bmatrix} 2/3 & 1/\sqrt{2} & 1/\sqrt{18} \\ 2/3 & -1/\sqrt{2} & 1/\sqrt{18} \\ 1/3 & 0 & -4/\sqrt{18} \end{bmatrix}$  or

$$\begin{bmatrix} 2/3 & 1/\sqrt{2} & -1/\sqrt{18} \\ 2/3 & -1/\sqrt{2} & -1/\sqrt{18} \\ 1/3 & 0 & 4/\sqrt{18} \end{bmatrix}.$$

42. Geometrically, an orthogonal projection does nothing on the space it projects onto, so that the second projection will do nothing new and  $A^2 = A$ . Algebraically, write  $A = QQ^T$  where Q is orthogonal, so that  $A^2 = QQ^TQQ^T = QQ^T = A$  since  $Q^TQ = I_n$ .

## Section 5.4

3. By Theorem 5.1.8(c), we know that p+q=n, so it is enough to either show that these vectors are linearly independent or that they span  $\mathbb{R}^n$ . I've provided arguments for both techniques. If you noted that p+q=n, you only need one of them. Otherwise, you need both halves.

## Linear Independence:

Write  $c_1\vec{v_1} + \cdots + c_p\vec{v_p} + d_1\vec{w_1} + \cdots + w_q\vec{w_q} = \vec{0}$ . We want to show that  $c_1 = \cdots = c_p = d_1 = \cdots = d_q = 0$ .

One strategy is to rewrite this relation as  $c_1\vec{v_1}+\dots+c_p\vec{v_p}=-d_1\vec{w_1}-\dots-w_q\vec{w_q}$ . The left-hand side is in V and the right-hand side is in  $V^\perp$ , so both sides are in  $V\cap V^\perp$ . But by Theorem 5.1.8(b), we know that  $V\cap V^\perp=\{\vec{0}\}$ , so that  $c_1\vec{v_1}+\dots+c_p\vec{v_p}=-d_1\vec{w_1}-\dots-w_q\vec{w_q}=0$ . Then, the linear independence of  $\vec{v_1},\dots,\vec{v_p}$  tells us that  $c_1=\dots=c_p=0$  and the linear independence of  $\vec{w_1},\dots,\vec{w_q}$  tells us that  $d_1=\dots=d_q=0$ . Thus,  $\vec{v_1},\dots,\vec{v_p},\vec{w_1},\dots,\vec{w_q}$  are linearly independent.

Another strategy is to take  $\operatorname{proj}_V$  of both sides of this relation to show that  $c_1\vec{v_1} + \cdots + c_p\vec{v_p} = \vec{0}$  and subtract this from the original relation to see that  $d_1\vec{w_1} + \cdots + w_q\vec{w_q} = \vec{0}$  as well, and then proceed as above. This is a special case of the solution to #5 on the 2nd midterm exam.

## Spanning:

Let  $\vec{x} \in \mathbb{R}^n$  be an arbitrary vector. We will show that  $\vec{x}$  is in the span of  $\vec{v_1}, \ldots, \vec{v_p}, \vec{w_1}, \ldots, \vec{w_q}$ . We know that we can uniquely write  $\vec{x} = \vec{x}^{\parallel} + \vec{x}^{\perp}$  where  $\vec{x}^{\parallel} \in V$  and  $\vec{x}^{\perp} \in V^{\perp}$ . As  $\vec{x}^{\parallel} \in V$ , we know that  $\vec{x}^{\parallel} = c_1 \vec{v_1} + \cdots + c_p \vec{v_p}$  for some choice of scalars  $c_1, \ldots, c_p$ , since  $\vec{v_1}, \ldots, \vec{v_p}$  form a basis of V. Similarly, the fact that  $\vec{w_1}, \ldots, \vec{w_q}$  form a basis of  $V^{\perp}$  tells us that we can write  $\vec{x}^{\perp} = d_1 \vec{w_1} + \cdots + w_q \vec{w_q}$  for some scalars  $d_1, \ldots, d_q$ . Then,  $\vec{x} = c_1 \vec{v_1} + \cdots + c_p \vec{v_p} + d_1 \vec{w_1} + \cdots + w_q \vec{w_q}$  so that  $\vec{x}$  is indeed in span $(\vec{v_1}, \ldots, \vec{v_p}, \vec{w_1}, \ldots, \vec{w_q})$ , and, as  $\vec{x}$  was an arbitrary vector in  $\mathbb{R}^n$ , we see that span $(\vec{v_1}, \ldots, \vec{v_p}, \vec{w_1}, \ldots, \vec{w_q}) = \mathbb{R}^n$ .

4. Start with  $(\operatorname{im}(A))^{\perp} = \ker(A^T)$ . Taking perps of both sides gives  $\operatorname{im}(A) = (\ker(A^T))^{\perp}$ . Now, this relation holds for any matrix A, so in particular we can replace the A on both sides by  $A^T$ . Then, since  $(A^T)^T = A$ , we have that  $(\ker(A))^{\perp} = \operatorname{im}(A^T)$ .