

## Section 6.3:

**Exercise 6.3.14:** Find the best approximation to  $z$  by the vectors of the form  $c_1v_1 + c_2v_2$ .

$$z = \begin{bmatrix} 2 \\ 4 \\ 0 \\ -1 \end{bmatrix}, v_1 = \begin{bmatrix} 2 \\ 0 \\ -1 \\ -3 \end{bmatrix}, v_2 = \begin{bmatrix} 5 \\ -2 \\ 4 \\ 2 \end{bmatrix}$$

**Solution:** We are looking for the orthogonal projection of  $z$  onto the span of  $v_1, v_2$ . So using the orthogonal projection theorem we get,

$$\begin{aligned} \hat{z} &= \frac{z \cdot v_1}{v_1 \cdot v_1} \begin{bmatrix} 2 \\ 0 \\ -1 \\ -3 \end{bmatrix} + \frac{z \cdot v_2}{v_2 \cdot v_2} \begin{bmatrix} 5 \\ -2 \\ 4 \\ 2 \end{bmatrix} \\ &= \frac{7}{14} \begin{bmatrix} 2 \\ 0 \\ -1 \\ -3 \end{bmatrix} + \frac{0}{49} \begin{bmatrix} 5 \\ -2 \\ 4 \\ 2 \end{bmatrix} \\ &= \begin{bmatrix} 1 \\ 0 \\ \frac{-1}{2} \\ \frac{-3}{2} \end{bmatrix} + 0 \end{aligned}$$

So we can see that the best approximation of  $z$  in the form of  $c_1v_1 + c_2v_2$  is ,

$$\hat{z} = \begin{bmatrix} 1 \\ 0 \\ \frac{-1}{2} \\ \frac{-3}{2} \end{bmatrix}$$

**Exercise 6.3.22:** All vectors and subspaces are in  $R^n$ . Mark each statement True or False.

- (1) If  $W$  is a subspace of  $R^n$  and of  $v$  is in both  $W$  and  $W^\perp$  then  $v$  must be the zero vector.

**Solution:** True. The only vector that can be dotted with itself and still result in zero is the zero vector. Consider the definition of orthogonal and definition of orthogonal sets.

- (2) In the Orthogonal Decomposition Theorem, each term in formula (2) for  $\hat{y}$  is itself an orthogonal projection of  $y$  onto subspace  $W$

**Solution:** True. Consider this excerpt from page 351,

"When  $W$  is a one-dimensional subspace, the formula (2) for  $proj_W y$  contains just one term. Thus, when  $\dim W > 1$ , each term in (2) is itself an orthogonal projection of  $y$  onto a one-dimensional subspace spanned by one of the  $u$ 's in the basis for  $W$  "

- (3) If  $y = z_1 + z_2$ , where  $z_1$  is in the subspace of  $W$  and  $z_2$  is in  $W^\perp$ , then  $z_1$  must be the orthogonal projection of  $y$  onto  $W$

**Solution:** True. If  $y$  is composed of two orthogonal vectors,  $z_1$  and  $z_2$  then those vectors are the orthogonal projections of  $y$  onto their respective subspaces. For clarity reference figure 3 on page 351.

- (4) The best approximation to  $y$  by elements of a subspace  $W$  is given by the vector  $y - proj_W y$

**Solution:** False. The best approximation of  $y$  by the elements in the subspace  $W$  is given by just the  $proj_W y$ , also consider Theorem 9.

- (5) If an  $n \times p$  matrix  $U$  has orthonormal columns, then  $UU^T x = x$  for all  $x$  in  $R^n$

**Solution:** False. Statement is only true when  $x$  is an element of the column space of  $U$ . Consider Theorem 10 that gives the equality " $proj_W y = UU^T y$ , s.t  $U$  is a orthonormal basis for  $W$ ".

**Exercise 6.3.24:** Let  $W$  be a subspace of  $R^N$  with an orthogonal basis  $\{w_1, \dots, w_p\}$  and let  $\{v_1, \dots, v_q\}$  be an orthogonal basis for  $W^\perp$

- (1) Explain why  $\{w_1, \dots, w_p, v_1, \dots, v_q\}$  is an orthogonal set.

**Solution:** Note, each vector in the set  $\{w_1, \dots, w_p\}$  is pair-wise orthogonal, by the definition of an orthogonal basis, and likewise for the set  $\{v_1, \dots, v_q\}$ . By our definition of orthogonal complements we know that any pair of vectors taken from  $W^\perp$  and  $W$  will be orthogonal, it must be true that the union of both  $\{w_1, \dots, w_p\}$  and  $\{v_1, \dots, v_q\}$  creates an orthogonal set as well.

- (2) Explain why the set in part (a) spans  $R^n$

**Solution:** Consider the Orthogonal Decomposition Theorem that states any  $y$  in  $R^n$  can be written as  $y = \hat{y} + z$  such that  $\hat{y}$  is from  $W$  and  $z$  is from  $W^\perp$ . Since  $\hat{y}$  and  $z$  can be written as linear combinations of the vectors in  $W^\perp$  and  $W$  respectively, it must follow that  $y$  can be written as a linear combination of the vectors in  $W$  union  $W^\perp$ .

- (3) Show that  $\dim W^\perp + \dim W = n$

**Solution:** Since the set  $\{w_1, \dots, w_p, v_1, \dots, v_q\}$  is an orthogonal set, it has linearly independent vectors. We know that  $\{w_1, \dots, w_p, v_1, \dots, v_q\}$  spans  $R^n$ , therefore we know that  $\dim W^\perp + \dim W = q + p = n$ .

## Section 6.4:

**Exercise 6.4.4:** the given set is a basis for the subspace  $W$ . Use the Gram-Schmidt process to produce the orthogonal basis  $W$ ,

$$x_1 = \begin{bmatrix} 3 \\ -4 \\ 5 \end{bmatrix}, x_2 = \begin{bmatrix} -3 \\ 14 \\ -7 \end{bmatrix}$$

**Solution:** First we let,

$$v_1 = x_1 = \begin{bmatrix} 3 \\ -4 \\ 5 \end{bmatrix}$$

Then by the Gram-Schmidt Process we calculate  $v_2$ ,

$$\begin{aligned} v_2 &= x_2 - \frac{x_2 \cdot v_1}{v_1 \cdot v_1} v_1 \\ &= \begin{bmatrix} -3 \\ 14 \\ -7 \end{bmatrix} - \frac{-100}{50} \begin{bmatrix} 3 \\ -4 \\ 5 \end{bmatrix} \\ &= \begin{bmatrix} 3 \\ 6 \\ 3 \end{bmatrix} \end{aligned}$$

So then the orthogonal basis  $W$  is,

$$\left\{ \begin{bmatrix} 3 \\ -4 \\ 5 \end{bmatrix}, \begin{bmatrix} 3 \\ 6 \\ 3 \end{bmatrix} \right\}$$

**Exercise 6.4.12:** Find the orthogonal basis for the column space of each matrix,

$$\begin{bmatrix} 1 & 3 & 5 \\ -1 & -3 & 1 \\ 0 & 2 & 3 \\ 1 & 5 & 2 \\ 1 & 5 & 8 \end{bmatrix}$$

**Solution:** First let,

$$v_1 = x_1 = \begin{bmatrix} 1 \\ -1 \\ 0 \\ 1 \\ 1 \end{bmatrix}$$

Then calculating  $v_2$ ,

$$\begin{aligned} v_2 &= x_2 - \frac{x_2 \cdot v_1}{v_1 \cdot v_1} v_1 \\ &= \begin{bmatrix} 3 \\ -3 \\ 2 \\ 5 \\ 5 \end{bmatrix} - \frac{16}{4} \begin{bmatrix} 1 \\ -1 \\ 0 \\ 1 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} -1 \\ 1 \\ 2 \\ 1 \\ 1 \end{bmatrix} \end{aligned}$$

Now calculating  $v_3$ ,

$$\begin{aligned} v_3 &= x_3 - \frac{x_3 \cdot v_1}{v_1 \cdot v_1} v_1 - \frac{x_3 \cdot v_2}{v_2 \cdot v_2} v_2 \\ &= \begin{bmatrix} 5 \\ 1 \\ 3 \\ 2 \\ 8 \end{bmatrix} - \frac{14}{4} \begin{bmatrix} 1 \\ -1 \\ 0 \\ 1 \\ 1 \end{bmatrix} - \frac{12}{2} \begin{bmatrix} -1 \\ 1 \\ 2 \\ 1 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} 3 \\ 3 \\ 0 \\ -3 \\ 3 \end{bmatrix} \end{aligned}$$

Finally we can see the the orthogonal basis for the column space is,

$$\left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 2 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ 3 \\ 0 \\ -3 \\ 3 \end{bmatrix} \right\}$$

**Exercise 6.4.18:** All vectors and subspaces are in  $R^n$ . Mark Each statement True, or False

- (a) If  $W = \text{Span}\{x_1, x_2, x_3\}$  with  $\{x_1, x_2, x_3\}$  linearly independent and if  $\{v_1, v_2, v_3\}$  is an orthogonal set in  $W$ , then  $\{v_1, v_2, v_3\}$  is a basis for  $W$

**Solution:** True. By our definition being  $W = \text{Span}\{x_1, x_2, x_3\}$  we know that any set of three linearly independent vectors that span  $W$  must also be a basis for  $W$ . An Orthogonal set is also defined as a set of linearly independent vectors, where pair - wise vectors are orthogonal. Thus  $\{v_1, v_2, v_3\}$  spans  $W$  and therefore is also a basis for  $W$ .

- (b) If  $x$  is not in a subspace  $W$ , then  $x - \text{proj}_W x$  is not zero.

**Solution:** True. If  $x$  is not in the subspace  $W$  then it can never have the same magnitude as  $\text{proj}_W x$  and therefore  $x - \text{proj}_W x \neq 0$ .

- (c) In a QR factorization, say  $A = QR$  (when  $A$  has linearly independent columns), the columns of  $Q$  form an orthonormal basis for the column space of  $A$ .

**Solution:** True. Consider Theorem 12 Chapter 6.4.

## Section 6.5:

**Exercise 6.5.14:** Let,

$$A = \begin{bmatrix} 2 & 1 \\ -3 & -4 \\ 3 & 2 \end{bmatrix}, b = \begin{bmatrix} 5 \\ 4 \\ 4 \end{bmatrix}, u = \begin{bmatrix} 4 \\ -5 \end{bmatrix}, v = \begin{bmatrix} 6 \\ -5 \end{bmatrix}$$

Compute  $Au$  and  $Av$ , and compare them with  $b$ . Is it possible that at least one of  $u$  or  $v$  could be a least-squares solution of  $Ax = b$ ? (Do not compute least-squares solution)

**Solution:** Computing  $Au$  and  $Av$ ,

$$Au = \begin{bmatrix} 2 & 1 \\ -3 & -4 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} 4 \\ -5 \end{bmatrix} = \begin{bmatrix} 3 \\ 8 \\ 2 \end{bmatrix}$$

$$Av = \begin{bmatrix} 2 & 1 \\ -3 & -4 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} 6 \\ -5 \end{bmatrix} = \begin{bmatrix} 7 \\ 2 \\ 8 \end{bmatrix}$$

Now we calculate  $b - Au$  and  $b - Av$ ,

$$b - Au = \begin{bmatrix} 5 \\ 4 \\ 4 \end{bmatrix} - \begin{bmatrix} 3 \\ 8 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 \\ -4 \\ 2 \end{bmatrix}$$

$$b - Av = \begin{bmatrix} 5 \\ 4 \\ 4 \end{bmatrix} - \begin{bmatrix} 7 \\ 2 \\ 8 \end{bmatrix} = \begin{bmatrix} -2 \\ 2 \\ -4 \end{bmatrix}$$

Finally we want to calculate the magnitude of  $b - Av$  and  $b - Au$  to see if either  $u$  or  $v$  could be the least squares solution.

$$\|b - Au\| = \sqrt{2^2 + (-4)^2 + 2^2} = \sqrt{24}$$

$$\|b - Av\| = \sqrt{(-2)^2 + 2^2 + (-4)^2} = \sqrt{24}$$

Since the least squares solution is unique neither  $u$  or  $v$  is the least squares solution.

**Exercise 6.5.18:**  $A$  is a an  $m \times n$  matrix and  $b$  is in  $R^m$ . Mark each statement True or False.

- (a) If  $b$  is in the column space of  $A$ , then every solution of  $Ax = b$  is a least-squares solution.

**Solution:** True. Consider this excerpt from page 362,

"So we seek an  $x$  that makes  $Ax$  the closest point in  $ColA$  to  $b$ . See Figure 1. (Of course, if  $b$  happens to be in  $ColA$ , then  $b$  is  $Ax$  for some  $x$ , and such an  $x$  is a "least-squares solution.")". the solution is also a vector not a point.

- (b) The least - squares solution of  $Ax = b$  is the point in the columns space of  $A$  that is closest to  $b$ .

**Solution:** True. By definition of Least Squares solution, normally the solution though is written as  $A\hat{x} = b$ , but it's the same thing the least squared solution is the vector in the span of  $A$  that is closest to  $b$ .

- (c) A least - squares solution of  $Ax = b$  is a list of weights that, when applied to the columns of  $A$  produces the orthogonal projections of  $b$  onto  $ColA$

**Solution:** True. Consider the following excerpt from page 363,

"Since  $\hat{b}$  is the closest point in  $ColA$  to  $b$ , a vector  $\hat{x}$  is a least-squares solution of  $Ax = b$  if and only if  $\hat{x}$  satisfies (1). Such an  $\hat{x}$  in  $R^n$  is a list of weights that will build  $\hat{b}$  out of the columns of  $A$ . See Figure 2."

- (d) If  $\hat{x}$  is a least-squares solution of  $Ax = b$ , then  $\hat{x} = (A^T A)^{-1} A^T b$  **Solution:** False. Columns of  $A$  must be linearly independent for this Theorem to work.

- (e) The normal equations always provide a reliable method for computing least - squares solutions.

**Solution:** False. Consider the excerpt from page 363

"In some cases, the normal equations for a least-squares problem can be ill- conditioned; that is, small errors in the calculations of the entries of  $A^T A$  can sometimes cause relatively large errors in the solution  $\hat{x}$ ."

- (f) If  $A$  has a  $QR$  factorization, say  $A = QR$ , then the best way to compute the least-square solution of  $Ax = b$  is to compute  $\hat{x} = R^{-1}Q^Tb$  **Solution:** False. The Numerical Note on page 367 says that the best way to compute the least-square solution is to use  $Rx = Q^Tb$ .

**Exercise 6.5.25:** Describe all least-squares solutions of the system,

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} b = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$

**Solution:** the least - squares solution can be found by solving,

$$A^T Ax = A^T b$$

So,

$$A^T b = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 4 \end{bmatrix} = \begin{bmatrix} 6 \\ 6 \end{bmatrix}$$

and,

$$A^T A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix}$$

Thus we solve the following,

$$\begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 6 \\ 6 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 2 & 6 \\ 2 & 2 & 6 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 3 \\ 0 & 0 & 0 \end{bmatrix}$$

So our solution is,

$$x = 3 - y$$

$$y = y$$

Where  $y$  is free. So the Least-Square Solution is,

$$\begin{bmatrix} 3 - y \\ y \end{bmatrix} \text{ such that } y \in \mathbb{R}$$