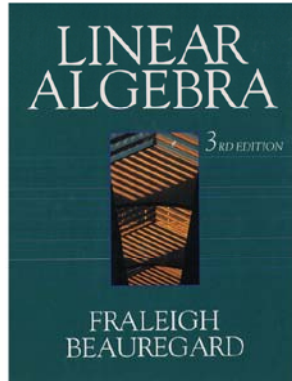


# Linear Algebra

## Chapter 6: Orthogonality

### Section 6.4. The Projection Matrix—Proofs of Theorems



#### Example 6.4.2

### Example 6.4.2

**Example 6.4.2.** Use the definition to find the projection matrix for the subspace in  $\mathbb{R}^3$  of the  $x_2x_3$ -plane (which is spanned by  $\hat{j} = [0, 1, 0]$  and  $\hat{k} = [0, 0, 1]$ ).

**Solution.** We have  $A = [\hat{j} \hat{k}] = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$  and  $A^T = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ . Then

$$A^T A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = (A^T A)^{-1}, \text{ so}$$

$$P = A(A^T A)^{-1} A^T = A A^T = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

Notice that  $P \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} 0 \\ b_2 \\ b_3 \end{bmatrix}$ , as expected.  $\square$

## Theorem 6.10

### Theorem 6.10. The Rank of $(A^T)A$ .

Let  $A$  be an  $m \times n$  matrix of rank  $r$ . Then the  $n \times n$  symmetric matrix  $(A^T)A$  also has rank  $r$ .

**Proof.** We work with nullspaces and the rank-nullity equation (Theorem 2.5). If  $\vec{v}$  is in the nullspace of  $A$  then  $A\vec{v} = \vec{0}$  and so  $A^T(A\vec{v}) = A^T\vec{0}$  or  $(A^T A)\vec{v} = \vec{0}$ . Hence  $\vec{v}$  is in the nullspace of  $A^T A$ . Conversely, suppose  $\vec{w}$  is in the nullspace of  $A^T A$  so that  $(A^T A)\vec{w} = \vec{0}$ . Then  $\vec{w}^T(A^T A)\vec{w} = \vec{w}^T\vec{0}$  or  $(\vec{w})^T(A\vec{w}) = \vec{w}^T\vec{0} = [0]$  (notice that  $\vec{w}$  is  $n \times 1$  and  $\vec{0}$  is  $n \times 1$  so that  $\vec{w}^T\vec{0}$  is  $1 \times 1$ ). Now  $(A\vec{w})^T(A\vec{w}) = [A\vec{w} \cdot A\vec{w}] = [\|A\vec{w}\|^2]$  and so  $\|A\vec{w}\| = 0$  or  $A\vec{w} = \vec{0}$ . That is,  $\vec{w}$  is in the nullspace of  $A$  and  $\text{nullity}(A) = \text{nullity}(A^T A)$ . Since both  $A$  and  $A^T A$  have  $n$  columns then by the rank-nullity equation (Theorem 2.5)  $\text{rank}(A) + \text{nullity}(A) = \text{number of columns}(A) = n = \text{number of columns}(A^T A) = \text{rank}(A^T A) + \text{nullity}(A^T A)$  and so  $\text{rank}(A) = \text{rank}(A^T A)$ , as claimed.  $\square$

#### Page 368 Number 6

### Page 368 Number 6

**Page 368 Number 6.** Find the projection for subspace  $3x + 2y + z = 0$  of  $\mathbb{R}^3$  and find the projection of  $\vec{b} = [4, 2, -1]$  onto the plane  $3x + 2y + z = 0$ .

**Solution.** We choose two (nonzero) linearly independent vectors in the plane as a spanning set of the plane. Say,  $\vec{a}_1 = [0, 1, -2]$  and

$\vec{a}_2 = [1, 0, -3]$ . We set  $A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \\ -2 & -3 \end{bmatrix}$  and so  $A^T = \begin{bmatrix} 0 & 1 & -2 \\ 1 & 0 & -3 \end{bmatrix}$ .

Hence  $A^T A = \begin{bmatrix} 0 & 1 & -2 \\ 1 & 0 & -3 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \\ -2 & -3 \end{bmatrix} = \begin{bmatrix} 5 & 6 \\ 6 & 10 \end{bmatrix}$ . So for

$(A^T A)^{-1}$ , consider

$$\left[ \begin{array}{cc|cc} 5 & 6 & 1 & 0 \\ 6 & 10 & 0 & 1 \end{array} \right] \xrightarrow{R_2 \rightarrow R_2 - R_1} \left[ \begin{array}{cc|cc} 5 & 6 & 1 & 0 \\ 1 & 4 & -1 & 1 \end{array} \right]$$

## Page 368 Number 6 (continued 1)

**Solution (continued).**

$$\underbrace{R_1 \leftrightarrow R_2}_{\left[ \begin{array}{cc|cc} 1 & 4 & -1 & 1 \\ 5 & 6 & 1 & 0 \end{array} \right]} \xrightarrow{R_2 \leftrightarrow R_2 - 5R_1} \left[ \begin{array}{cc|cc} 1 & 4 & -1 & 1 \\ 0 & -14 & 6 & -5 \end{array} \right] \xrightarrow{R_2 \leftrightarrow R_2 / (-14)}$$

$$\left[ \begin{array}{cc|cc} 1 & 4 & -1 & 1 \\ 0 & 1 & -6/14 & 5/14 \end{array} \right] \xrightarrow{R_1 \leftrightarrow R_1 - 4R_2} \left[ \begin{array}{cc|cc} 1 & 0 & 10/14 & -6/14 \\ 0 & 1 & -6/14 & 5/14 \end{array} \right]$$

and so  $(A^T A)^{-1} = \frac{1}{14} \begin{bmatrix} 10 & -6 \\ -6 & 5 \end{bmatrix}$ . So the projection matrix is

$$P = A(A^T A)^{-1} A^T = \begin{bmatrix} 0 & 1 \\ 1 & 0 \\ -2 & -3 \end{bmatrix} \left( \frac{1}{14} \begin{bmatrix} 10 & -6 \\ -6 & 5 \end{bmatrix} \right) \begin{bmatrix} 0 & 1 & -2 \\ 1 & 0 & -3 \end{bmatrix}$$

$$= \frac{1}{14} \begin{bmatrix} -6 & 5 \\ 10 & -6 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} 0 & 1 & -2 \\ 1 & 0 & -3 \end{bmatrix} = \boxed{\frac{1}{14} \begin{bmatrix} 5 & -6 & -3 \\ -6 & 10 & -2 \\ -3 & -2 & 13 \end{bmatrix}}.$$

()

## Page 368 Number 6 (continued 2)

**Page 368 Number 6.** Find the projection for subspace  $3x + 2y + z = 0$  of  $\mathbb{R}^3$  and find the projection of  $\vec{b} = [4, 2, -1]$  onto the plane  $3x + 2y + z = 0$ .

**Solution (continued).** So the projection of  $\vec{b} = [4, 2, -1]$  onto the plane is

$$\vec{b}_W = P\vec{b}^T = \frac{1}{14} \begin{bmatrix} 5 & -6 & -3 \\ -6 & 10 & -2 \\ -3 & -2 & 13 \end{bmatrix} \begin{bmatrix} 4 \\ 2 \\ -1 \end{bmatrix} = \boxed{\frac{1}{14} \begin{bmatrix} 11 \\ -2 \\ -29 \end{bmatrix}}.$$

□

()

## Theorem 6.11

**Theorem 6.11. Projection Matrix.**

Let  $W$  be a subspace of  $\mathbb{R}^n$ . There is a unique  $n \times n$  matrix  $P$  such that, for each column vector  $\vec{b} \in \mathbb{R}^n$ , the vector  $P\vec{b}$  is the projection of  $\vec{b}$  onto  $W$ . The projection matrix can be found by selecting any basis  $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_k\}$  for  $W$  and computing  $P = A(A^T A)^{-1} A^T$ , where  $A$  is the  $n \times k$  matrix having column vectors  $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_k$ .

**Proof.** We know from above that  $P = A(A^T A)^{-1} A^T$  satisfies the requirements of a projection matrix. For any vector  $\vec{x} \in \mathbb{R}^n$ , the transformation mapping  $\vec{x} \mapsto P(\vec{x})$  is a linear transformation by Theorem 1.3.A(7) and (10). Now with the  $i$ th standard basis vector of  $\mathbb{R}^n$  as  $\hat{e}_i$ , we have that  $P\hat{e}_i$  is the  $i$ th column of  $P$ ; that is,  $P$  is the standard matrix representation of the linear transformation. Since the standard matrix representation of a linear transformation is unique (assume  $P'$  is another standard matrix representation of the linear transformation and then  $P\hat{e}_i = P'\hat{e}_i$  is the  $i$ th column of both  $P$  and  $P'$ ). So the matrix  $P$  is unique, as claimed. □

()

## Theorem 6.12

**Theorem 6.12. Characterization Projection Matrices.**

The projection matrix  $P$  for a subspace  $W$  of  $\mathbb{R}^n$  is both idempotent (that is,  $P^2 = P$ ) and symmetric (that is,  $P = P^T$ ). Conversely, every  $n \times n$  matrix that is both idempotent and symmetric is a projection matrix (specifically, it is the projection matrix for its column space).

**Proof.** First, we show that a projection matrix  $P$  is idempotent and symmetric (this is Exercise 6.4.16). For  $P = A(A^T A)^{-1} A^T$  we have

$$P^2 = (A(A^T A)^{-1} A^T)(A(A^T A)^{-1} A^T) = A(A^T A)^{-1} (A^T A(A^T A)^{-1} A^T) = A(A^T A)^{-1} \mathcal{I} A^T = A(A^T A)^{-1} A^T = P$$

and so  $P$  is idempotent, as claimed. Next,

$$P^T = (A(A^T A)^{-1} A^T)^T = (A^T)^T ((A^T A)^{-1})^T (A^T)^T = A((A^T A)^T)^{-1} A^T \text{ since } (A^T)^T = A \text{ by Note 1.3.A} \\ \text{and } (A^{-1})^T = (A^T)^{-1} \text{ by Exercise 1.5.24} \\ = A(A^T A)^{-1} A^T = P \text{ and so } P \text{ is symmetric.}$$

()

## Theorem 6.12 (continued 1)

**Proof (continued).** For the converse, let  $P$  be an  $n \times n$  symmetric and idempotent matrix. Let  $\vec{b} \in \mathbb{R}^n$ . If we show  $P\vec{b} \in W$  and  $\vec{b} - P\vec{b}$  is perpendicular to every vector in  $W$  where  $W$  is some subspace of  $\mathbb{R}^n$  (the subspace here will be the column space of matrix  $P$ ), that is the two conditions of the note after Theorem 6.10 are satisfied, then we know  $P$  is a projection matrix (based on the uniqueness of a projection matrix for a given subspace  $W$ , as given by Theorem 6.11). Now  $P\vec{b}$  is in the column space of  $P$  (see Note 1.3.A). Let  $p\vec{x}$  be any vector in the column space of  $P$ . Then

$$\begin{aligned} [(\vec{b} - P\vec{b}) \cdot P\vec{x}] &= (\vec{b} - P\vec{b})^T P\vec{x} = ((I - P)\vec{b})^T P\vec{x} \\ &= \vec{b}^T (I - P)^T P\vec{x} = \vec{b}^T (I^T - P^T) P\vec{x} \\ &= \vec{b}^T (I - P) P\vec{x} \text{ since } P \text{ is symmetric} \\ &= \vec{b}^T (P - P^2)\vec{x} = \vec{b}^T (P - P)\vec{x} \text{ since } P \text{ is idempotent} \\ &= \vec{b}^T (0)\vec{x} = [0]. \end{aligned}$$

()

## Page 369 Number 28

**Page 369 Number 28.** Find the projection matrix for the subspace  $W$  spanned by the orthonormal vectors  $\vec{a}_1 = [1/2, 1/2, 1/2, 1/2]$ ,  $\vec{a}_2 = [-1/2, 1/2, -1/2, 1/2]$ , and  $\vec{a}_3 = [1/2, 1/2, -1/2, -1/2]$ .

**Solution.** We have  $A = \frac{1}{2} \begin{bmatrix} 1 & -1 & 1 \\ 1 & 1 & 1 \\ 1 & -1 & -1 \\ 1 & 1 & -1 \end{bmatrix}$  and so

$$P = AA^T = \frac{1}{2} \begin{bmatrix} 1 & -1 & 1 \\ 1 & 1 & 1 \\ 1 & -1 & -1 \\ 1 & 1 & -1 \end{bmatrix} \left( \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ -1 & 1 & -1 & 1 \\ 1 & 1 & -1 & -1 \end{bmatrix} \right) \dots$$

()

## Theorem 6.12 (continued 2)

**Theorem 6.12. Characterization Projection Matrices.**

The projection matrix  $P$  for a subspace  $W$  of  $\mathbb{R}^n$  is both idempotent (that is,  $P^2 = P$ ) and symmetric (that is,  $P = P^T$ ). Conversely, every  $n \times n$  matrix that is both idempotent and symmetric is a projection matrix (specifically, it is the projection matrix for its column space).

**Proof (continued).** So  $(\vec{b} - P\vec{b}) \cdot P\vec{x} = 0$  and  $\vec{b} - P\vec{b}$  is orthogonal to every vector in the column space of  $P$ , as claimed.  $\square$

()

## Page 369 Number 28 (continued)

**Page 369 Number 28.** Find the projection matrix for the subspace  $W$  spanned by the orthonormal vectors  $\vec{a}_1 = [1/2, 1/2, 1/2, 1/2]$ ,  $\vec{a}_2 = [-1/2, 1/2, -1/2, 1/2]$ , and  $\vec{a}_3 = [1/2, 1/2, -1/2, -1/2]$ .

**Solution (continued).** ...

$$= \frac{1}{4} \begin{bmatrix} 3 & 1 & 1 & -1 \\ 1 & 3 & -1 & 1 \\ 1 & -1 & 3 & 1 \\ -1 & 1 & 1 & 3 \end{bmatrix}.$$

()

## Page 369 Number 32

**Page 369 Number 32.** Find the projection of  $\vec{b} = [4, -12, -4, 0]$  onto  $W$  the subspace of  $\mathbb{R}^4$  given in Exercise 6.4.28.

**Solution.** We use the projection matrix of Exercise 6.4.28:

$$\begin{aligned}\vec{b}_W = P\vec{b}^T &= \frac{1}{4} \begin{bmatrix} 3 & 1 & 1 & -1 \\ 1 & 3 & -1 & 1 \\ 1 & -1 & 3 & 1 \\ -1 & 1 & 1 & 3 \end{bmatrix} \begin{bmatrix} 4 \\ -12 \\ -4 \\ 0 \end{bmatrix} \\ &= \frac{1}{4} \begin{bmatrix} -4 \\ -28 \\ 4 \\ -20 \end{bmatrix} = \boxed{\begin{bmatrix} -1 \\ -7 \\ 1 \\ -5 \end{bmatrix}}.\end{aligned}$$

□