

## AMS 10/10A, Homework 10 Solutions

### Problem 1.

$$\begin{aligned}\hat{y} &= \frac{y^T u_1}{u_1^T u_1} u_1 + \frac{y^T u_2}{u_2^T u_2} u_2 \\ &= \frac{1}{3} u_1 + \frac{5}{7} u_2 \\ &= \begin{bmatrix} -8/21 \\ 52/21 \\ -23/21 \end{bmatrix}\end{aligned}$$

and

$$\begin{aligned}z &= y - \hat{y} \\ &= \begin{bmatrix} 2 \\ 2 \\ -3 \end{bmatrix} - \begin{bmatrix} -8/21 \\ 52/21 \\ -23/21 \end{bmatrix} \\ &= \begin{bmatrix} 50/21 \\ -10/21 \\ -40/21 \end{bmatrix}\end{aligned}$$

They satisfy that  $y = \hat{y} + z$ , where  $\hat{y}$  is a vector in  $H$  and  $z$  is a vector in  $H^\perp$ .

**Problem 2.** By Best Approximation Theorem the closest point in  $\text{span}\{v_1, v_2\}$  to  $y$  is given by the projection of  $y$  onto  $\text{span}\{v_1, v_2\}$ . Since  $v_1$  and  $v_2$  are orthogonal, this projection can be computed as

$$\frac{y^T v_1}{v_1^T v_1} v_1 + \frac{y^T v_2}{v_2^T v_2} v_2 = \frac{1}{5} v_1 + \frac{1}{13} v_2 = \begin{bmatrix} -7/65 \\ -21/65 \\ -1/5 \\ 41/65 \end{bmatrix}$$

### Problem 3-7.

- By applying elementary row operations on the augmented matrix  $[A | b]$ , we have

$$[A | b] \sim \left[ \begin{array}{cc|c} 1 & 5 & 1 \\ 0 & -14 & -3 \\ 0 & 0 & 4 \end{array} \right]$$

Since the last column is a pivot column, the equation  $Ax = b$  is inconsistent.

- Since the columns of  $A$  are an orthogonal set of non-zero vectors, they are a linearly independent set. Consequently, they form an orthogonal basis for  $Col(A)$ .
- The column of  $A$  are an orthogonal basis for  $Col(A)$ . Hence, projection of  $b$  onto  $Col(A)$  is given by

$$\hat{b} = \frac{b^T a_1}{a_1^T a_1} a_1 + \frac{b^T a_2}{a_2^T a_2} a_2 = \frac{1}{2} a_1 - \frac{1}{6} a_2 = \begin{bmatrix} -1/3 \\ 4/3 \\ 5/3 \end{bmatrix}$$

- The least square solution,  $\hat{x}$ , of  $Ax = b$  is given by

$$\begin{aligned} \hat{x} &= (A^T A)^{-1} A^T b \\ &= \begin{bmatrix} 14 & 0 \\ 0 & 42 \end{bmatrix}^{-1} \begin{bmatrix} 7 \\ -7 \end{bmatrix} = \begin{bmatrix} 1/2 \\ -1/6 \end{bmatrix} \end{aligned}$$

- $A\hat{x} = \begin{bmatrix} -1/3 \\ 4/3 \\ 5/3 \end{bmatrix} = \hat{b}.$

**Problem 8-9.** Let  $A$  be an  $m \times n$  matrix. Use the steps below to show that a vector  $x$  in  $\mathbb{R}^n$  satisfies  $Ax = 0$  if and only if  $A^T Ax = 0$ .

- Show that if  $Ax = 0$ , then  $A^T Ax = 0$ .

Proof: Let  $x$  be a vector such that  $Ax = 0$ . Multiplying  $A^T$  on both sides of the equation leads to  $A^T Ax = A^T 0 = 0$ .

- Suppose  $A^T Ax = 0$ . Show that  $x^T A^T Ax = 0$ , and use this to prove  $Ax = 0$ .

Proof: Let  $x$  be a vector such that  $A^T Ax = 0$ . Multiplying  $x^T$  on both sides of the equation leads to  $x^T A^T Ax = x^T 0 = 0$ . Therefore,  $x^T A^T Ax = (Ax)^T (Ax) = \|Ax\|^2 = 0$ . Since the norm of a vector equals to zero if and only if the vector itself is the zero vector, we have  $Ax = 0$ .

**Problem 10-11.** Let  $A$  be an  $m \times n$  matrix. Problem 8-9 implies that  $Nul(A) = Nul(A^T A)$ . Use this result to prove that

- $\text{rank}(A) = \text{rank}(A^T A)$ .

Proof: Matrix  $A$  is  $m \times n$  and matrix  $A^T A$  is  $n \times n$ . By the Rank Theorem, we have

$$\begin{aligned}\text{rank}(A) + \dim(\text{Nul } A) &= n \\ \text{rank}(A^T A) + \dim(\text{Nul } A^T A) &= n\end{aligned}$$

Hence,

$$\begin{aligned}\text{rank}(A) &= n - \dim(\text{Nul } A) \\ &= n - \dim(\text{Nul}(A^T A)) \quad (\text{since } \text{Nul}(A) = \text{Nul}(A^T A)) \\ &= \text{rank}(A^T A) \quad (\text{By The Rank Theorem})\end{aligned}$$

- If  $\text{rank}(A) = n$ , then  $A^T A$  is invertible.

Proof: If  $\text{rank}(A) = n$ , from the result in Problem 10,  $\text{rank}(A^T A) = \text{rank}(A) = n$ . Since matrix  $A^T A$  is a square matrix of  $n \times n$ , by Invertible Matrix Theorem,  $\text{rank}(A^T A) = n$  implies that  $A^T A$  is invertible.