

## 1 Theorem 9 : The Best Approximation Theorem

Let  $W$  be a subspace of  $\mathbb{R}^n$ , let  $\mathbf{y}$  be any vector in  $\mathbb{R}^n$ , and let  $\hat{\mathbf{y}}$  be the orthogonal projection of  $\mathbf{y}$  onto  $W$ . Then  $\hat{\mathbf{y}}$  is the closest point in  $W$  to  $\mathbf{y}$ , in the sense that

$$\|\mathbf{y} - \hat{\mathbf{y}}\| < \|\mathbf{y} - \mathbf{v}\|$$

for all  $\mathbf{v}$  in  $W$  distinct from  $\hat{\mathbf{y}}$ .

What is the use of this theorem?

In fact, this theorem is significantly important to Statistics.

### 1.1 Least-Squares Problems

If  $A$  is  $m \times n$  and  $\mathbf{b}$  is in  $\mathbb{R}^m$ , a **least-squares solution** of  $A\mathbf{x} = \mathbf{b}$  is an  $\hat{\mathbf{x}}$  in  $\mathbb{R}^n$  such that

$$\|\mathbf{b} - A\hat{\mathbf{x}}\| \leq \|\mathbf{b} - A\mathbf{x}\|$$

for all  $\mathbf{x}$  in  $\mathbb{R}^n$ .

Theorem 13 says that the solution set of

$$A\mathbf{x} = \mathbf{b}$$

is the same as the solution set of

$$A^T A\mathbf{x} = A^T \mathbf{b}.$$

From this fact, we know that there exists a unique solution for such problem if and only if  $A^T A$  is invertible (or rank  $A = n$ ).

Statistical presumptions of Linear Models are based on this theorem.

$$X\beta = \mathbf{y}$$

1. Find an orthogonal basis for the column space of the matrix

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

2. Find the orthogonal projection of  $\mathbf{b}$  onto  $\text{Col } A$  and a least-squares solution of  $A\mathbf{x} = \mathbf{b}$ .

a.

$$A = \begin{pmatrix} 1 & 2 \\ -1 & 4 \\ 1 & 2 \end{pmatrix}, \quad B = \begin{pmatrix} 3 \\ -1 \\ 5 \end{pmatrix}$$

b.

$$A = \begin{pmatrix} 4 & 0 & 1 \\ 1 & -5 & 1 \\ 6 & 1 & 0 \\ 1 & -1 & -5 \end{pmatrix}, \quad B = \begin{pmatrix} 9 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

3. Mark each statement True or False. Justify your answer. ( $A$  is an  $m \times n$  matrix and  $\mathbf{b}$  is in  $\mathbb{R}^m$ .)

a. A least-squares solution of  $A\mathbf{x} = \mathbf{b}$  is a vector  $\hat{\mathbf{x}}$  that satisfies  $A\hat{\mathbf{x}} = \hat{\mathbf{b}}$ , where  $\hat{\mathbf{b}}$  is the orthogonal projection of  $\mathbf{b}$  onto  $\text{Col } A$ .

b. If the columns of  $A$  are linearly independent, then the equation  $A\mathbf{x} = \mathbf{b}$  has exactly one least-squares solution.

c. If  $\mathbf{b}$  is in the column space of  $A$ , then every solution of  $A\mathbf{x} = \mathbf{b}$  is a least-squares solution.

d. A least-squares solution of  $A\mathbf{x} = \mathbf{b}$  is a list of weights that, when applied to the columns of  $A$ , produces the orthogonal projection of  $\mathbf{b}$  onto  $\text{Col } A$ .

4. In Assignment 7, you are given a problem ; Show that  $\text{Nul } A = \text{Nul } A^T A$ . Using this fact without proof, show that

$$\text{rank } A^T A = \text{rank } A$$