

### Lab 14 Solution

#### 1. Multiple Linear Regression

Two objects are located on a line at points  $0 < \beta_1 < \beta_2$ . These points aren't precisely known. A surveyor makes the following measurements:

- (i) He stands at the origin and measures the 2 distances from there to  $\beta_1, \beta_2$ . Let these measurements be denoted  $y_1 = 10, y_2 = 11.5$ .
- (ii) He goes to  $\beta_1$  and measures the distance from there to  $\beta_2$ . Call this distance  $y_3 = 1$ .

- (a) Set up a linear model  $y = X\beta + e$

$$\begin{bmatrix} 10 \\ 11.5 \\ 1 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix}$$

- (b) Find the least square estimate of  $\beta_1, \beta_2$   
 $\hat{\beta} = (X^T X)^{-1} X^T y$  where

$$(X^T X)^{-1} = \left( \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 1 \end{bmatrix} \right)^{-1} = \left( \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \right)^{-1} = \frac{1}{3} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

and

$$X^T y = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 10 \\ 11.5 \\ 1 \end{bmatrix} = \begin{bmatrix} 9 \\ 12.5 \end{bmatrix}$$

Therefore

$$\hat{\beta} = (X^T X)^{-1} X^T y = \frac{1}{3} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 9 \\ 12.5 \end{bmatrix} = \begin{bmatrix} 10.17 \\ 11.33 \end{bmatrix} = \begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix}$$

- (c) Find the estimate of  $\sigma^2$

Our estimate for  $\sigma^2$  is  $\frac{RSS}{n-k-1} = \frac{RSS}{3-1-1} = RSS$  where  $RSS = (y - \hat{y})^T \cdot (y - \hat{y})$  and

$$y = \begin{bmatrix} 10 \\ 11.5 \\ 1 \end{bmatrix} \text{ and } \hat{y} = X\hat{\beta} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 10.17 \\ 11.33 \end{bmatrix} = \begin{bmatrix} 10.17 \\ 11.33 \\ 1.17 \end{bmatrix}$$

$$y - \hat{y} = \begin{bmatrix} -0.17 \\ 0.17 \\ 0.17 \end{bmatrix}$$

Therefore our estimate is  $\hat{\sigma}^2 = RSS/1 = 3 \cdot 0.17^2 = 0.0867$ .

(d) Find  $s_{\hat{\beta}_1}, s_{\hat{\beta}_2}$

$$Var(\hat{\beta}) = \sigma^2(X^T X)^{-1} \approx 0.0867 \cdot \frac{1}{3} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 0.0578 & 0.0289 \\ 0.0289 & 0.0578 \end{bmatrix}, \therefore \begin{bmatrix} se_{\hat{\beta}_1} \\ se_{\hat{\beta}_2} \end{bmatrix} = \begin{bmatrix} 0.2404 \\ 0.2404 \end{bmatrix}$$

(e) Estimate  $\beta_1 - \beta_2$  and its standard error

An unbiased estimator would be  $\hat{\beta}_1 - \hat{\beta}_2 = 10.1667 - 11.3333 = -1.167$ .

$$\begin{aligned} Var(\hat{\beta}_1 - \hat{\beta}_2) &= Var(\hat{\beta}_1) + Var(\hat{\beta}_2) - 2Cov(\hat{\beta}_1, \hat{\beta}_2) \\ &= 0.0578 + 0.0578 - 2 * 0.0289 = 0.0578. \end{aligned}$$

$$\therefore se_{\hat{\beta}_1 - \hat{\beta}_2} = 0.2404.$$

(f) Test the null  $H_0 : \beta_1 = \beta_2$

$$\begin{aligned} t_{n-2} = t_1 &= \frac{(\hat{\beta}_1 - \hat{\beta}_2) - (\beta_1 - \beta_2)}{se_{\hat{\beta}_1 - \hat{\beta}_2}} = \frac{(\hat{\beta}_1 - \hat{\beta}_2) - 0}{se_{\hat{\beta}_1 - \hat{\beta}_2}} \\ &= \frac{-1.167}{0.2404} = -4.854. \end{aligned}$$

$$2^*pt(-4.854, df = 1) = 0.1293 > 0.05$$

Therefore we do not reject the null.

## 2. Bayesian Statistic

Consider a biased coin with probability of landing heads equal to  $\theta$ . Also, let  $X$  be a random variable that is equal to 1 when the coin lands heads, and is otherwise equal to 0. In other words, we assume that  $P(X = 1|\Theta = \theta) = \theta$ , while  $P(X = 0|\Theta = \theta) = 1 - \theta$ . We consider the following prior for  $\Theta$  :

$$f_{\Theta}(\theta) = \begin{cases} \theta e^{\theta} & \text{for } 0 \leq \theta \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- (a) Find the probability that a coin toss results in heads: i.e., compute  $P(X = 1)$ .

**Solution:**

$$P(X = 1) = \int_0^1 P(X = 1|\Theta = \theta) f_{\Theta}(\theta) d\theta = \int_0^1 \theta * \theta e^{\theta} d\theta = \int_0^1 \theta^2 e^{\theta} d\theta$$

$$\text{By parts: } = [\theta^2 e^{\theta}]_0^1 - 2 \int_0^1 \theta e^{\theta} = [e^1 - 0] - 2 = e - 2 \approx 0.7183$$

- (b) Given that a coin toss results in heads, find the posterior density for  $\Theta$ .

**Solution:**

$$f_{\Theta|X}(\theta|X = 1) = \frac{P(X = 1|\Theta = \theta) f_{\Theta}(\theta)}{P(X = 1)} = \frac{\theta^2 e^{\theta}}{e - 2}$$

for  $\theta \in [0, 1]$ , zero otherwise.

- (c) Given that the first toss resulted in heads, find the conditional probability of heads on the second toss. (*Hint:* this is tricky. Use the posterior from part (b) as the new prior!)

**Solution:**

The "new" prior is the function from part (b).

$$P(X_2 = 1|X_1 = 1) = \int_0^1 P(X = 1|\Theta = \theta) f_{\Theta|X_1}(\theta|X_1 = 1) d\theta = \int_0^1 \frac{\theta^3 e^{\theta}}{e - 2} d\theta$$

$$\text{by parts: } = \frac{1}{e - 2} \{ [\theta^3 e^{\theta}]_0^1 - 3 \int_0^1 \theta^2 e^{\theta} d\theta \}$$

$$\text{from part (a): } = \frac{1}{e - 2} \{ e - 3(e - 2) \} = \frac{6 - 2e}{e - 2} \approx 0.7844$$