

Chapter 2 Solution

2.8 (a) Since $E(Y|X = x) = \beta_0 + \beta_1 x$, then $\alpha = \beta_0 + \beta_1 \bar{x} = E(Y|X = \bar{x})$.

(b) Since $RSS(\alpha, \beta_1) = \sum (y_i - \alpha - \beta_1(x_i - \bar{x}))^2$, we have that

$$\begin{aligned}\frac{\partial}{\partial \alpha} RSS(\alpha, \beta_1) &= -2 \sum (y_i - \alpha - \beta_1(x_i - \bar{x})) = 0 \\ \frac{\partial}{\partial \beta_1} RSS(\alpha, \beta_1) &= -2 \sum (y_i - \alpha - \beta_1(x_i - \bar{x}))(x_i - \bar{x}) = 0\end{aligned}$$

where

$$\begin{aligned}\sum (y_i - \alpha - \beta_1(x_i - \bar{x})) &= \sum y_i - n\alpha - \beta_1 \sum (x_i - \bar{x}) = \sum y_i - n\alpha = 0 \\ \implies \hat{\alpha} &= \bar{y}\end{aligned}$$

and

$$\sum (y_i - \alpha - \beta_1(x_i - \bar{x}))(x_i - \bar{x}) = \sum (y_i(x_i - \bar{x}) - \alpha(x_i - \bar{x}) - \beta_1(x_i - \bar{x})^2) = 0$$

so that upon replacing α by $\hat{\alpha} = \bar{y}$, we obtain

$$\begin{aligned}\sum (y_i(x_i - \bar{x}) - \hat{\alpha}(x_i - \bar{x}) - \beta_1(x_i - \bar{x})^2) &= \sum (y_i(x_i - \bar{x}) - \bar{y}(x_i - \bar{x}) - \beta_1(x_i - \bar{x})^2) = 0 \\ \implies \sum (x_i - \bar{x})(y_i - \bar{y}) - \beta_1 \sum (x_i - \bar{x})^2 &= 0 \\ \implies \sum (x_i - \bar{x})(y_i - \bar{y}) &= \beta_1 \sum (x_i - \bar{x})^2 \\ \implies \hat{\beta}_1 &= \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} = \frac{SXY}{SXX}\end{aligned}$$

(c) Since $\text{Var}(y_i) = \sigma^2$, we have that

$$\text{Var}(\hat{\alpha}) = \text{Var}(\bar{y}) = \frac{\sigma^2}{n}$$

For β_1 , it is the same as OLS,

$$\text{Var}(\hat{\beta}_1) = \frac{\sigma^2}{SXX}$$

Now, the covariance between them is $\text{Cov}(\hat{\alpha}, \hat{\beta}_1) = \text{Cov}(\bar{y}, \hat{\beta}_1)$, which is found to be 0 (see Appendix), or we can derive (with $\hat{\beta}_1 = \sum c_i y_i$ where $c_i = (x_i - \bar{x})/SXX$ and $\sum c_i = 0$)

$$\begin{aligned}\text{Cov}(\bar{y}, \hat{\beta}_1) &= \text{Cov}\left(\frac{1}{n} \sum y_i, \sum c_i y_i\right) = \frac{1}{n} \sum c_i \text{Cov}(y_i, y_i) \\ &= \frac{1}{n} \sum c_i \text{Var}(y_i) = \frac{\sigma^2}{n} \sum c_i = 0\end{aligned}$$

2.12 We have

$$r_{xy} = \frac{s_{xy}}{SD_x SD_y} = \frac{SXY/(n-1)}{\sqrt{SXX/(n-1)}\sqrt{SYY/(n-1)}} = \frac{SXY}{\sqrt{SXX}\sqrt{SYY}}$$

and

$$\sqrt{1 - r_{xy}^2} = \sqrt{1 - \frac{SXY^2}{SXX \cdot SYY}} = \sqrt{\frac{SXX \cdot SYY - SXY^2}{SXX \cdot SYY}} = \frac{\sqrt{SXX \cdot SYY - SXY^2}}{\sqrt{SXX}\sqrt{SYY}}$$

so that (working backwards)

$$\begin{aligned} \sqrt{n-2} \frac{r_{xy}}{\sqrt{1 - r_{xy}^2}} &= \sqrt{n-2} \frac{SXY/(\sqrt{SXX}\sqrt{SYY})}{\sqrt{(SXX \cdot SYY - SXY^2)/(\sqrt{SXX}\sqrt{SYY})}} \\ &= \sqrt{n-2} \frac{SXY}{\sqrt{(SXX \cdot SYY - SXY^2)}} \\ &= \sqrt{n-2} \frac{SXY}{\sqrt{RSS \cdot SXX}} \quad \left(\text{since } RSS = SYY - \frac{SXY^2}{SXX} \right) \end{aligned}$$

whereas

$$\frac{\hat{\beta}_1}{\text{se}(\hat{\beta}_1)} = \frac{SXY/SXX}{\hat{\sigma}/\sqrt{SXX}} = \frac{SXY/\sqrt{SXX}}{\sqrt{RSS/(n-2)}} = \sqrt{n-2} \frac{SXY}{\sqrt{RSS \cdot SXX}}$$

from which we conclude that

$$t = \frac{\hat{\beta}_1}{\text{se}(\hat{\beta}_1)} = \sqrt{n-2} \frac{r_{xy}}{\sqrt{1 - r_{xy}^2}}$$

2.18 Here, we have that

$$\begin{aligned} RSS &= \sum \hat{e}_i^2 = \sum (y_i - \hat{y}_i)^2 = \sum (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2 = \sum (y_i - \bar{y} + \hat{\beta}_1 \bar{x} - \hat{\beta}_1 x_i)^2 \\ &= \sum ((y_i - \bar{y}) + \hat{\beta}_1 (x_i - \bar{x}))^2 = \sum [(y_i - \bar{y})^2 + \hat{\beta}_1^2 (x_i - \bar{x})^2 - 2(y_i - \bar{y})\hat{\beta}_1(x_i - \bar{x})] \\ &= \sum (y_i - \bar{y})^2 + \hat{\beta}_1^2 \sum (x_i - \bar{x})^2 - 2\hat{\beta}_1 \sum (y_i - \bar{y})(x_i - \bar{x}) \\ &= SYY + \hat{\beta}_1^2 SXX - 2\hat{\beta}_1 SXY = SYY + \left(\frac{SXY}{SXX} \right)^2 SXX - 2 \left(\frac{SXY}{SXX} \right) SXY \\ &= SYY + \frac{(SXY)^2}{SXX} - 2 \frac{(SXY)^2}{SXX} \\ &= SYY - \frac{(SXY)^2}{SXX} = SYY - \hat{\beta}_1^2 SXX \end{aligned}$$