

1. a.

Using R. $\hat{\beta}_0 = 0.289523$ $\hat{\beta}_1 = 17.367688$
 $\text{Var}(\hat{\beta}_0 - 3\hat{\beta}_1 - 15) = \text{Var}(\hat{\beta}_0) + \text{Var}(3\hat{\beta}_1) - 6\text{cov}(\hat{\beta}_0, \hat{\beta}_1)$
 $= \frac{6^2}{n} + \frac{9^2 6^2}{\sum (x_i - \bar{x})^2} + \frac{96^2}{2 \sum (x_i - \bar{x})^2} + \cancel{\frac{6 \cdot 9 \cdot 6^2}{\sum (x_i - \bar{x})^2}}$

$$\frac{\hat{\beta}_0 - 3\hat{\beta}_1 - 15 - 0}{\sqrt{\text{Var}(\hat{\beta}_0 - 3\hat{\beta}_1 - 15)}} = \frac{\hat{\beta}_0 - 3\hat{\beta}_1 - 15}{\sqrt{\frac{(n-2)S_e^2}{S^2(n-2)}}} = \frac{\hat{\beta}_0 - 3\hat{\beta}_1 - 15}{\sqrt{S_e^2 \cdot \frac{\sum x_i^2 + 6x + 9}{\sum (x_i - \bar{x})^2}}}$$

$t = 0.002240767$ using R

1. b.

$$\frac{(SSE_R - SSE_F) / (df_R - df_F)}{SSE_F / df_F} \sim F_{df_R - df_F, df_F}$$

$SSE_F = \sum e_i^2$ $SSE_R = \sum e_i^{*2}$

$\beta_0 - 3\beta_1 - 15 = 0$ $\beta_0 = 3\beta_1 + 15$

$y_i = 3\beta_1 + 15 + \beta_1 x_i + e_i$

$y_i - 15 = \beta_1(x_i + 3) + e_i$

$\hat{\beta}_1^* = \frac{\sum x_i^* y_i^*}{\sum x_i^{*2}}$ $\sum x_i^* = (x_i + 3)$ $y_i^* = y_i - 15$

$df_R = n - 1$ $df_F = n - 2$ by previous hw.

$$\frac{(SSE_R - SSE_F)}{SSE_F / (n - 2)} = \frac{\sum e_i^{*2} - \sum e_i^2}{\sum e_i^2 / (n - 2)} \sim F_{1, n-2}$$

c.

The part $\frac{(\hat{\beta}_0 - 3\hat{\beta}_1 - 15)^2}{\text{Var}(\hat{\beta}_0 - 3\hat{\beta}_1 - 15)}$ is equal to $(SSE_R - SSE_F)$

2.

$$E(\hat{y}_{g_i}) = \hat{\beta}_0 + \hat{\beta}_1 x_{g_i} \quad \text{unbiased}$$

$$\begin{aligned} \text{Var}(\hat{y}_{g_i}) &= \text{Var}(\hat{\beta}_0 + \hat{\beta}_1 x_{g_i}) \\ &= \text{Var}(\bar{y} - \hat{\beta}_1 \bar{x} + \hat{\beta}_1 x_{g_i}) \\ &= \text{Var}(\bar{y}) + \text{Var}(\hat{\beta}_1)(x_{g_i} - \bar{x})^2 + \text{cov}(\bar{y}, \hat{\beta}_1)(x_{g_i} - \bar{x}) \\ &= \frac{\sigma^2}{n} + \frac{\sigma^2(x_{g_i} - \bar{x})^2}{\sum (x_i - \bar{x})^2} \end{aligned}$$

$$\frac{E(\hat{y}_{g_i}) - E(y_{g_i})}{\sqrt{\frac{\sigma^2}{n} + \frac{\sigma^2(x_{g_i} - \bar{x})^2}{\sum (x_i - \bar{x})^2}}} \sim t_{n-2}$$

\therefore let significance level be α .

$$\left(E(\hat{y}_{g_i}) - t_{\frac{\alpha}{2}, n-2} \sqrt{\sigma^2 \left(\frac{1}{n} + \frac{(x_{g_i} - \bar{x})^2}{\sum (x_i - \bar{x})^2} \right)}, E(\hat{y}_{g_i}) + t_{\frac{\alpha}{2}, n-2} \sqrt{\sigma^2 \left(\frac{1}{n} + \frac{(x_{g_i} - \bar{x})^2}{\sum (x_i - \bar{x})^2} \right)} \right)$$

Exe 3.

$$\left. \begin{array}{l} y_i = \beta_0 + \varepsilon_i \\ x_i = \beta_0 + \eta_i \end{array} \right\} \Rightarrow \begin{array}{l} \hat{y}_i = \hat{\beta}_0 = \bar{y} \\ \hat{x}_i = \hat{\beta}_0 = \bar{x} \end{array}$$

$$e_i = y_i - \bar{y}$$

$$e_i^* = x_i - \bar{x}$$

$$e_i = \alpha_0 + \alpha_1 e_i^* + r_i$$

$$\hat{\alpha}_1 = \frac{\sum (e_i^* - \bar{e}^*)(e_i - \bar{e})}{\sum (e_i^* - \bar{e})^2} = \frac{\sum e_i^* e_i}{\sum e_i^{*2}} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} = \frac{\sum (x_i - \bar{x}) y_i - 0}{\sum (x_i - \bar{x})^2}$$

$$\hat{\alpha}_0 = \bar{e}_i - \hat{\alpha}_1 \bar{e}_i^*$$

$$\therefore \hat{\alpha}_1 = \hat{\beta}_1$$

$$4 a. \begin{pmatrix} \lambda_1 \\ \lambda_2 \end{pmatrix} = \frac{6^2}{n \sum (x_i - \bar{x})^2} \begin{pmatrix} \sum x_i^2 - \sum x_i & 1 \\ -\sum x_i & n \end{pmatrix} \begin{pmatrix} 1 \\ x_0 \end{pmatrix}$$

$$= \begin{pmatrix} \frac{6^2}{n} + \frac{6^2 \bar{x}^2}{\sum (x_i - \bar{x})^2} - \frac{6^2 \bar{x} x_0}{\sum (x_i - \bar{x})^2} \\ \frac{-\bar{x} 6^2}{\sum (x_i - \bar{x})^2} + \frac{x_0 6^2}{\sum (x_i - \bar{x})^2} \end{pmatrix}$$

$$\therefore \lambda_1 = 6^2 \frac{\sum x_i^2 - n \bar{x} x_0}{n \sum (x_i - \bar{x})^2} \quad \lambda_2 = 6^2 \frac{x_0 - \bar{x}}{\sum (x_i - \bar{x})^2}$$

$$b. \sum a_i y_i = \sum \frac{\lambda_1 + \lambda_2 x_i}{6^2} y_i$$

$$= \sum \frac{\sum x_i^2 - n \bar{x} x_0 + n x_i x_0 - n \bar{x} x_i}{n \sum (x_i - \bar{x})^2} y_i$$

$$= \frac{n \bar{y} \sum x_i^2 - n \bar{x} x_0 \sum y_i + n x_0 \sum x_i y_i - n \bar{x} \sum x_i y_i + n \bar{x}^2 \bar{y}}{n \sum (x_i - \bar{x})^2}$$

$$= \frac{\bar{y} (\sum x_i^2 - n \bar{x}^2) + \sum x_i y_i (x_0 - \bar{x}) - \bar{x} x_0 \sum y_i + n \bar{x}^2 \bar{y}}{\sum (x_i - \bar{x})^2}$$

$$= \bar{y} + \frac{(x_0 - \bar{x}) \sum x_i y_i + \bar{x} (x_0 - \bar{x}) \sum y_i}{\sum (x_i - \bar{x})^2}$$

$$= \bar{y} + (x_0 - \bar{x}) \frac{\sum (x_i - \bar{x}) y_i}{\sum (x_i - \bar{x})^2}$$

$$= \bar{y} + \hat{\beta}_1 (x_0 - \bar{x})$$