- 1. (a) First compute the statistics $\bar{x}=3$, $\bar{y}=(6+c)/5$. Since the regression line passes through (\bar{x},\bar{y}) , we have $\bar{y}=1.65$ indicating c=2.25. Then $\sum (y_i-\bar{y})^2=\sum y_i^2-n\bar{y}^2=3.075$.
 - (b) A 98% confidence interval for β_0 is given by $\hat{\beta}_0 \mp t_{0.01,23} \hat{\sigma} \sqrt{1/25 + 0}$, whose length is $2 \cdot t_{0.01,23} \hat{\sigma}/5 = 2 \times 2.499867 \times 10/5 \approx 9.9995$.
 - (c) We have Res.SS = TSS Reg.SS = 425 225 = 200, and then $\hat{\sigma}^2 = Res.SS/8 = 25$.

Since $Reg.SS = \hat{\beta}_1^2 S_{x,x}$, we have $|\hat{\beta}_1| = \sqrt{225/400} = 3/4$.

The absolute test statistic is given by

$$|t| = \frac{|\hat{\beta}_1|}{\hat{\sigma}\sqrt{c_{11}}} = \frac{3/4}{5/20} = 3.$$

Then $|t| > t_{0.025,8} = 2.306004$. Hence at the significant level of $\alpha = 0.05$, we reject H_0 .

2. (a) We compute

$$RSS = Y^{\top}Y - \hat{\beta}^{\top}X^{\top}Y$$

= 1324 - (1.237088, -1.007608, 2.089488) \cdot (138, 962, 1002)^{\tau} = 28.934.

The unbiased estimate of σ^2 is thus given by $\hat{\sigma}^2 = RSS/(n-3) = 1.702$.

(b) The test statistic is given by

$$t = \frac{\hat{\beta}_2 - 2}{\hat{\sigma}c_{22}} = \frac{2.089488 - 2}{\sqrt{1.702} \times \sqrt{0.014734}} \approx 0.5651$$

Compute the critical value $t_{0.025,17} = 2.109816 > |t|$. Hence at the significant level of $\alpha = 0.05$, we do not reject H_0 .

Assume that $\beta_0 = 2$.

(c) (2 marks) Write down X^TX , $(X^TX)^{-1}$ and X^TY in terms of values of summary statistics.

$$\mathcal{X}^{T}\mathcal{X} = \begin{pmatrix} \sum_{i=1}^{n} x_{i1}^{2} & \sum_{i=1}^{n} x_{i1}x_{i2} \\ \sum_{i=1}^{n} x_{i1}x_{i2} & \sum_{i=1}^{n} x_{i2}^{2} \end{pmatrix} = \begin{pmatrix} 1012 & 875 \\ 875 & 834 \end{pmatrix}, \text{ then } \\
(\mathcal{X}^{T}\mathcal{X})^{-1} = \begin{pmatrix} 0.010640 & -0.011163 \\ -0.011163 & 0.012911 \end{pmatrix}. \\
\mathcal{X}^{T}\mathcal{Y} = \begin{pmatrix} \sum_{i=1}^{n} x_{i1}y_{i} - 2\sum_{i=1}^{n} x_{i1} \\ \sum_{i=1}^{n} x_{i2}y_{i} - 2\sum_{i=1}^{n} x_{i2} \end{pmatrix} = \begin{pmatrix} 714 \\ 774 \end{pmatrix}.$$

(d)
$$\hat{\beta} = \begin{pmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix} = (X^T X)^{-1} X^T Y = \begin{pmatrix} -1.043262 \\ 2.022607 \end{pmatrix}$$
.

Thus, the fitted line is given by

$$\hat{y} = 2 - 1.043262x_1 + 2.022607x_2.$$

(e) According to the formula on page 9 of chapter 1, we have

$$RSS = \mathcal{Y}^{\mathsf{T}}\mathcal{Y} - \hat{\beta}^{\mathsf{T}}\mathcal{X}^{\mathsf{T}}\mathcal{Y} = 852 - 820.6087 = 31.3913$$
$$\hat{\sigma}^{2} = \frac{RSS}{n - p'} = \frac{31.3913}{20 - 2} = 1.743961.$$

3. (a)

$$\hat{\beta}_1 = \frac{S_{xy}}{S_{xx}} = \frac{\sum_{i=1}^{3} (i-2)y_i}{\sum_{i=1}^{3} (i-2)^2} = \frac{1}{2} (y_3 - y_1)$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = \frac{1}{3} \sum_{i=1}^{3} y_i - \sum_{i=1}^{3} (i-2)y_i = \frac{4}{3}y_1 + \frac{1}{3}y_2 - \frac{2}{3}y_3$$

(b)

$$Var(\hat{\beta}_1) = \frac{\sum_{i=1}^{3} (i-2)^2 i\sigma^2}{\left(\sum_{i=1}^{3} (i-2)^2\right)^2} = \frac{\sigma^2 + 3\sigma^2}{4} = \sigma^2$$

$$Var(\hat{\beta}_0) = Var(\frac{y_1 + y_2 + y_3}{3} - \sum_{i=1}^{3} (i-2)y_i)$$

$$= Var(\sum_{i=1}^{3} (\frac{7}{3} - i)e_i) = \sum_{i=1}^{3} (\frac{7}{3} - i)^2 i\sigma^2$$

$$= \frac{16}{9}\sigma^2 + \frac{1}{9}2\sigma^2 + \frac{4}{9}3\sigma^2 = \frac{10}{3}\sigma^2$$

(c) Write $\hat{\boldsymbol{e}} = (\boldsymbol{I} - \boldsymbol{H})(\boldsymbol{X}\boldsymbol{\beta} + \boldsymbol{e}) = (\boldsymbol{I} - \boldsymbol{H})\boldsymbol{e}$ where

$$\mathbf{X} = \begin{pmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{pmatrix} \quad \mathbf{X}^T \mathbf{X} = \begin{pmatrix} 3 & 6 \\ 6 & 14 \end{pmatrix} \quad (\mathbf{X}^T \mathbf{X})^{-1} = \begin{pmatrix} \frac{7}{3} & -1 \\ -1 & \frac{1}{2} \end{pmatrix}
\Rightarrow \quad \mathbf{H} = \mathbf{X} (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T = \frac{1}{6} \begin{pmatrix} 5 & 2 & -1 \\ 2 & 2 & 2 \\ -1 & 2 & 5 \end{pmatrix}
\Rightarrow \quad \mathbf{I} - \mathbf{H} = \frac{1}{6} \begin{pmatrix} 1 & -2 & 1 \\ -2 & 4 & -2 \\ 1 & -2 & 1 \end{pmatrix}$$

$$E(\hat{\boldsymbol{e}}) = (\boldsymbol{I} - \boldsymbol{H})E(\boldsymbol{e}) = 0$$

$$Var(\hat{\boldsymbol{e}}) = \sigma^{2}(\boldsymbol{I} - \boldsymbol{H}) \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix} (\boldsymbol{I} - \boldsymbol{H}) = \frac{\sigma^{2}}{3} \begin{pmatrix} 1 & -2 & 1 \\ -2 & 4 & -2 \\ 1 & -2 & 1 \end{pmatrix}$$