

Question 5 Solution:

Part a: Computing y, X, β and ϵ

$$y = \begin{bmatrix} 27.3 \\ 42.7 \\ 38.7 \\ 4.5 \\ 23.0 \\ 166.3 \\ 109.7 \\ 80.1 \\ 150.7 \\ 20.3 \\ 189.7 \\ 131.3 \\ 404.2 \\ 149 \end{bmatrix} \quad X = \begin{bmatrix} 1 & 13.1 \\ 1 & 15.3 \\ 1 & 25.8 \\ 1 & 1.8 \\ 1 & 4.9 \\ 1 & 55.4 \\ 1 & 39.3 \\ 1 & 26.7 \\ 1 & 47.5 \\ 1 & 6.6 \\ 1 & 94.7 \\ 1 & 61.1 \\ 1 & 135.6 \\ 1 & 47.6 \end{bmatrix} \quad \beta = \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix} \quad \epsilon = \begin{bmatrix} \epsilon_0 \\ \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \\ \epsilon_4 \\ \epsilon_5 \\ \epsilon_6 \\ \epsilon_7 \\ \epsilon_8 \\ \epsilon_9 \\ \epsilon_{10} \\ \epsilon_{11} \\ \epsilon_{12} \\ \epsilon_{13} \end{bmatrix}$$

$y = X\beta + \epsilon$ becomes,

$$\begin{bmatrix} 27.3 \\ 42.7 \\ 38.7 \\ 4.5 \\ 23.0 \\ 166.3 \\ 109.7 \\ 80.1 \\ 150.7 \\ 20.3 \\ 189.7 \\ 131.3 \\ 404.2 \\ 149 \end{bmatrix} = \begin{bmatrix} 1 & 13.1 \\ 1 & 15.3 \\ 1 & 25.8 \\ 1 & 1.8 \\ 1 & 4.9 \\ 1 & 55.4 \\ 1 & 39.3 \\ 1 & 26.7 \\ 1 & 47.5 \\ 1 & 6.6 \\ 1 & 94.7 \\ 1 & 61.1 \\ 1 & 135.6 \\ 1 & 47.6 \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \end{bmatrix} + \begin{bmatrix} \epsilon_0 \\ \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \\ \epsilon_4 \\ \epsilon_5 \\ \epsilon_6 \\ \epsilon_7 \\ \epsilon_8 \\ \epsilon_9 \\ \epsilon_{10} \\ \epsilon_{11} \\ \epsilon_{12} \\ \epsilon_{13} \end{bmatrix}$$

Part b: Solving the least squares estimator

```
```{r}
b = solve(t(X)%*%X, t(X)%*%y)
b
```
```

```
      [,1]
[1,] -1.233836
[2,]  2.701553
```

Part c:

```
```{r}
e = y - X%*%b
e #Residual errors
```
```

```

                                [,1]
[1,] -6.8565106
[2,]  2.3000724
[3,] -29.7662361
[4,]  0.8710405
[5,] 10.9962256
[6,] 17.8677893
[7,]  4.7627957
[8,]  9.2023660
[9,] 23.6100596
[10,]  3.7035852
[11,] -64.9032511
[12,] -32.5310639
[13,]  39.1032233
[14,] 21.6399042
```

```
```{r}
n = 14 #sample size
p = 2 #number of parameters
SSRes = sum(e^2)
ssquared = SSRes/(n-p)
ssquared
```

```
```
```

```
[1] 777.1528
```

Part d:

```
```{r}
```

```
c(1,28)*%b
```

```
```
```

```
[,1]
```

```
[1,] 74.40965
```

Part e:

```
```{r}
a = solve(t(X)%*%X)
a
```
```

| | [,1] | [,2] |
|------|--------------|---------------|
| [1,] | 0.163081936 | -2.230009e-03 |
| [2,] | -0.002230009 | 5.425812e-05 |

```
```{r}
H = X%*%a%*%t(X)
H
```
```

```
```{r}
z = e/sqrt(ssquared * (1 - diag(H)))
z[13]
```
```

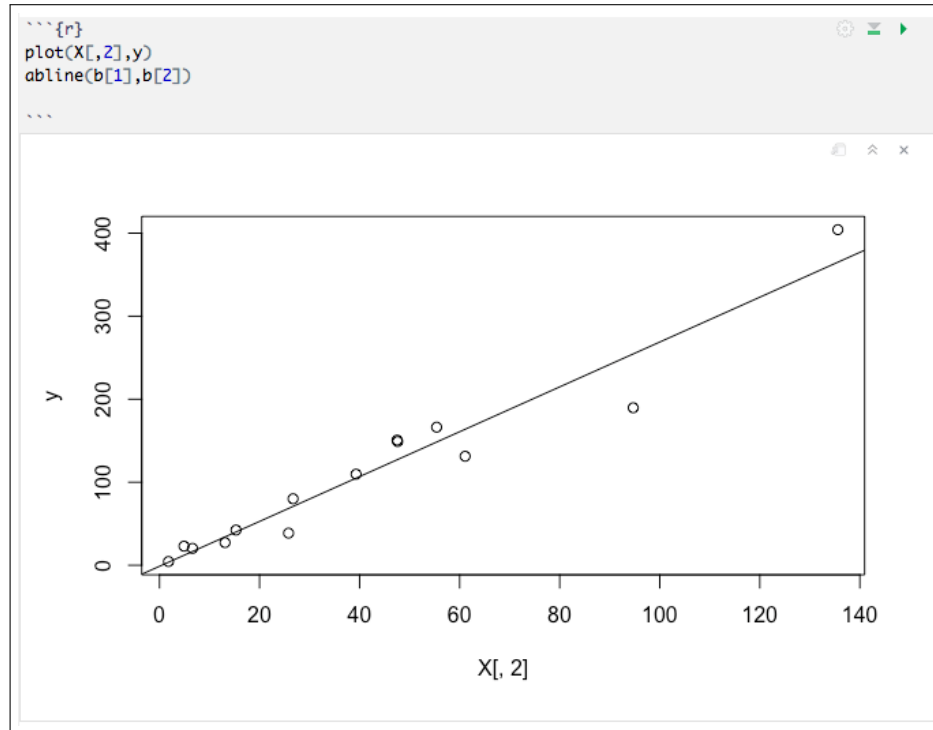
```
[1] 2.104999
```

Part f:

```
```{r}
k = 1
D = z^2 * (diag(H)/(1-diag(H))) * 1/(k+1)
D[13]
```
```

```
[1] 2.774008
```

Part g:



Full explanation: The Cook's distance certainly indicates it should be of some concern; however looking at the plot, it seems that the fit is actually okay. There is considerable evidence for heteroskedasticity — the variance increases with x (the design variable). Sea scallops has (by far) the largest x and so may be prone to a larger variance than the remaining points. The high Cook's distance therefore comes primarily from a very high leverage, rather than a bad fit to the model.

END OF ASSIGNMENT!!