We shall perform regression on the "Crest" data based on the model form,

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \epsilon$$

1. From R, we compute the least squares estimates as,

$$eta_0 = 30625.907$$
 $eta_1 = 3.893$
 $eta_2 = -29607.315$
 $eta_3 = 86.519$

- 2. SSE, s^2 and s
 - (a) We compute SSE, the sum of squared error, as

$$SSE = 894568667.411638$$

(b) In order to compute both s and s^2 , we must compute k. k is equal to the subscript of the highest β_i . As such our model has k = 3. Then, noting that n = 13, we compute,

$$s^2 = \frac{SSE}{n - (k + 1)} = \frac{894568667.411638}{13 - (3 + 1)} = \frac{894568667.411638}{9} \approx 99396519$$

(c) We may then compute,

$$s=\sqrt{s^2}\approx 9969.78$$

- 3. Variation
 - (a) We compute, $TotalVariation = \sum (y_i \bar{y})^2$. Then,

$$T.V. \approx 21050596826.9231$$

(b) Next, we compute ExplainedVariation = $\sum (\hat{y}_i - \bar{y})^2$. Then,

$$E.V. \approx 20156028159.5114$$

(c) Finally, we compute $UnexplainedVariation = \sum (y_i - \hat{y}_i)^2$. Then,

$$U.V. \approx 894568667.4116$$

- 4. Using R, we compute F(model) = 67.59 with a p-val = 1.709e-6. Thus, with $\alpha = .05$, Thus, we reject the null hypothesis that all of the $\beta_i = 0$. Then, our alternative that $\beta_i \neq 0$ becomes plausible.
- 5. Using the generated t-statistics and associated p-values, we shall test each of the computed β_i using the null hypotheses $H_0: \beta_i = 0$ versus the alternative $H_A: \beta_i \neq 0$ with $\alpha = .05$. The results follow,

Variable	Beta Value	t-value	p-value	Significance
Intercept	eta_0	1.546	0.15647	Not Significant at $\alpha = .05$
Budget	eta_1	1.871	.09420	Not Significant at $\alpha = .05$
Ratio	eta_2	-1.243	.24533	Not Significant at $\alpha = .05$
Income	eta_3	4.628	.00124	Significant at $\alpha = .05$

Similarly, we make the same comparison with $\alpha = .01$, and see that, again the income variable is the only one significant at this new level.

6. We shall now compute 90% confidence intervals for each β_i . The equation for our interval is,

$$\left[b_j \pm t_{\alpha/2}^{n-(k+1)} s_{b_j}\right]$$

Using our previously computed k and n, we find

$$t_{.05}^9 = 1.833$$

Then, we compute the intervals as follows:

β_j	b_{j}	$ s_{b_j} $	Interval
β_0	30625.907	19808.009	[-5682.173, 66933.988]
β_1	3.893	2.081	[.078527, 7.707473]
β_2	-29607.315	23822.087	[-73273.20047, 14058.57047]
β_3	86.519	18.693	[52.254731, 120.783269]

7. Next, we note that our equation of,

$$R^2 = \frac{Explained\ Variation}{Total\ Variation}$$

Thus,

$$R^2 \approx .9575039$$

To compute the adjusted R^2 value, we use the equation,

$$\bar{R}^2 = \left(R^2 - \frac{k}{n-1}\right) \left(\frac{n-1}{n-(k+1)}\right)$$

So,

$$\bar{R}^2 = \left(0.9575 - \frac{3}{13 - 1}\right) \left(\frac{13 - 1}{13 - (3 + 1)}\right) \approx 0.9433385$$

8. We now consider the computation of a new point prediction and interval based on the new data of Budget = 25,000, Ratio = 1.55 and Income = 1821.70. Using our computed linear model and R, we compute,

$$\hat{y} = 239675.4$$

$$P.I. = [209507.3, 269843.5]$$

R-Code

```
> setwd("~/Desktop/PSM/Fall 2018/Regression/HW4")
> crest=read.table('crest.txt',header = TRUE)
> mod<-lm(Sales~Budget+Ratio+Income, data=crest)
> summary(mod)
lm(formula = Sales ~ Budget + Ratio + Income, data = crest)
Residuals:
  Min
           1Q Median
                         3Q
                               Max
                1195
-24447 -2561
                              9527
                       3885
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)
             30625.907 19808.009
                                    1.546 0.15647
Budget
                 3.893
                            2.081
                                    1.871 0.09420 .
Ratio
            -29607.315
                        23822.087
                                   -1.243 0.24533
                86.519
                           18.693
Income
                                    4.628 0.00124 **
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1
Residual standard error: 9970 on 9 degrees of freedom
Multiple R-squared: 0.9575, Adjusted R-squared: 0.9433
F-statistic: 67.59 on 3 and 9 DF, p-value: 1.709e-06
> test=data.frame(Budget=16300, Ratio=1.25, Income=547.9)
> predict(mod,test)
       1
104479.3
> 105000-104479.3
[1] 520.7
> mod$residuals
          1
                      2
                                  3
              1195.3334
  520.6815
                          9527.2175
                                     -2708.3077
                                                  -6270.0669
          6
                      7
                                  8
                                               9
                                                          10
  7144.6314
              3885.3353 -24446.9075
                                       868.6300
                                                   2758.4026
         11
                     12
                                 13
  1446.1279
              8639.8221
                        -2560.8997
> SSE=sum(mod$residuals^2)
> s2=SSE/(13-(3+1))
> s=sqrt(s2)
> s2
[1] 99396519
> s
[1] 9969.78
> print(s2)
[1] 99396519
> totvar=sum(crest$Sales-mean(crest$Sales))
> mean(crest$Sales)
[1] 148723.1
> diff=crest$Sales-mean(crest$Sales)
> diff
 [1] -43723.077 -43723.077 -27123.077 -34973.077 -34973.077
 [6] -19798.077 -6223.077 -22723.077 13276.923 42901.923
[11] 40276.923 61276.923 75526.923
> diff=diff^2
```

```
> sum(diff)
[1] 21050596827
> totvar=sum((crest$Sales-mean(crest$Sales))^2)
> totvar
[1] 21050596827
> X=data.frame(crest[3:5])
> test
  Budget Ratio Income
  16300 1.25 547.9
1
> X
   Budget Ratio Income
   16300 1.25 547.9
   15800 1.34 593.4
2
3
   16000 1.22 638.9
   14200 1.00 695.3
4
5
   15000 1.15 751.8
6
   14000 1.13 810.3
7
   15400 1.05 914.5
   18250 1.27 998.3
8
9
   17300 1.07 1096.1
10
   23000 1.17 1194.4
11 19300 1.07 1311.5
12 23056 1.54 1462.9
13 26000 1.59 1641.7
> yhat=predict(mod,X)
> yhat
                         3
104479.3 103804.7 112072.8 116458.3 120020.1 121780.4 138614.7
                        10
                                 11
                                          12
150446.9 161131.4 188866.6 187553.9 201360.2 226810.9
> expvar=sum((yhat-mean(crest$Sales))^2)
> unexvar=sum((crest$Sales-yhat)^2)
> totvar-expvar-unexvar
[1] -4.529953e-06
> expvar
[1] 20156028160
> unexvar
[1] 894568667
> anova(mod)
Analysis of Variance Table
Response: Sales
          Df
                 Sum Sq
                           Mean Sq F value
                                              Pr(>F)
Budget
           1 1.7008e+10 1.7008e+10 171.117 3.681e-07 ***
Ratio
           1 1.0184e+09 1.0184e+09 10.246
                                             0.01081 *
Income
           1 2.1292e+09 2.1292e+09
                                   21.421
                                             0.00124 **
Residuals 9 8.9457e+08 9.9397e+07
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1
> fmodel=(expvar/3)/(unexvar/(13-(3+1)))
> summary(mod)
Call:
lm(formula = Sales ~ Budget + Ratio + Income, data = crest)
Residuals:
  Min
           1Q Median
                         3Q
                               Max
-24447 -2561
                1195
                       3885
                              9527
```

```
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
(Intercept) 30625.907 19808.009
                                  1.546 0.15647
                                    1.871 0.09420 .
Budget
                 3.893
                            2.081
Ratio
            -29607.315
                        23822.087 -1.243 0.24533
                86.519
                           18.693
Income
                                    4.628 0.00124 **
Signif. codes:
0 *** 0.001 ** 0.01 * 0.05 . 0.1
Residual standard error: 9970 on 9 degrees of freedom
Multiple R-squared: 0.9575, Adjusted R-squared: 0.9433
F-statistic: 67.59 on 3 and 9 DF, p-value: 1.709e-06
> expvar/totvar
[1] 0.9575039
> (expvar/totvar-3/12)*(12/9)
[1] 0.9433385
> topred=data.frame(Budget=25000, Ratio=1.55, Income=1821.70)
> predict(mod, topred)
       1
239675.4
> predict.lm(mod, topred)
239675.4
> predict(mod, topred, interval = "prediction")
       fit
                lwr
                         upr
1 239675.4 209507.3 269843.5
```