B27

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```
Stats 110 - HW5
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```

1. 3.23

```
# Load data
diamonds = read.csv("../data/Diamonds.csv")
```

1-(a). Quadratic model using Depth

```
model_a = lm(TotalPrice ~ Depth + I(Depth^2), data=diamonds)
summary(model_a)
##
## Call:
## lm(formula = TotalPrice ~ Depth + I(Depth^2), data = diamonds)
##
## Residuals:
##
     Min
              1Q Median
                            3Q
                                  Max
   -9323 -4251 -2676
                          2134 45513
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept) -28406.783 112211.790 -0.253
                                                0.800
                 766.369
                            3353.222
                                       0.229
                                                0.819
## I(Depth^2)
                   -3.233
                              24.869 -0.130
                                                0.897
##
## Residual standard error: 7616 on 348 degrees of freedom
## Multiple R-squared: 0.04748,
                                   Adjusted R-squared:
## F-statistic: 8.673 on 2 and 348 DF, p-value: 0.0002111
```

1-(b). Two-predictor model using Carat and Depth

```
model_b = lm(TotalPrice ~ Carat + Depth, data=diamonds)
summary(model_b)
```

```
##
## Call:
## lm(formula = TotalPrice ~ Carat + Depth, data = diamonds)
##
## Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
## -9234.7 -1223.7 -274.3 1161.0 16368.6
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1059.24
                          1918.36
                                     0.552
                                              0.581
## Carat
              15087.01
                           320.96 47.006 < 2e-16 ***
## Depth
               -134.94
                             30.92 -4.364 1.68e-05 ***
## ---
```

```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2809 on 348 degrees of freedom
## Multiple R-squared: 0.8704, Adjusted R-squared: 0.8696
## F-statistic: 1168 on 2 and 348 DF, p-value: < 2.2e-16
1-(c). A three-predictor model that adds interaction for Carat and Depth
model_c = lm(TotalPrice ~ Carat + Depth + Carat*Depth, data=diamonds)
summary(model_c)
##
## Call:
## lm(formula = TotalPrice ~ Carat + Depth + Carat * Depth, data = diamonds)
## Residuals:
##
      Min
               1Q Median
                               3Q
## -8254.4 -1311.5 -157.2 1131.8 14513.9
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 31171.41 4219.58
                                   7.387 1.13e-12 ***
## Carat
              -11827.73
                           3436.47 -3.442 0.000648 ***
                             65.47 -9.137 < 2e-16 ***
## Depth
                -598.18
## Carat:Depth
                 408.45
                             51.96
                                   7.861 4.84e-14 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2592 on 347 degrees of freedom
## Multiple R-squared: 0.89, Adjusted R-squared: 0.889
## F-statistic: 935.7 on 3 and 347 DF, p-value: < 2.2e-16
1-(d). Complete second-order model using Carat and Depth.
model_d = lm(TotalPrice ~ Carat + Depth + I(Carat^2) + I(Depth^2) + Carat*Depth,
            data=diamonds)
summary(model_d)
##
## lm(formula = TotalPrice ~ Carat + Depth + I(Carat^2) + I(Depth^2) +
##
      Carat * Depth, data = diamonds)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
## -12196.1
             -652.7
                       -38.5
                                485.7 10582.2
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 24338.820 30297.912
                                   0.803
                                             0.4223
                         3040.787
                                    2.491
                                             0.0132 *
## Carat
               7573.620
## Depth
               -728.700
                           904.439 -0.806
                                            0.4210
## I(Carat^2)
               4761.592
                           330.246 14.418
                                            <2e-16 ***
```

0.4333

0.784

6.727

I(Depth^2)

5.276

```
## Carat:Depth -83.891 53.530 -1.567 0.1180
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2053 on 345 degrees of freedom
## Multiple R-squared: 0.9313, Adjusted R-squared: 0.9304
## F-statistic: 936.1 on 5 and 345 DF, p-value: < 2.2e-16</pre>
```

The quadratic model where CaratSq is the quadratic term provides the better model. The model captures most of the variance yet it is a more parsimonious model than the second order model and the three predictor model with the interaction term.

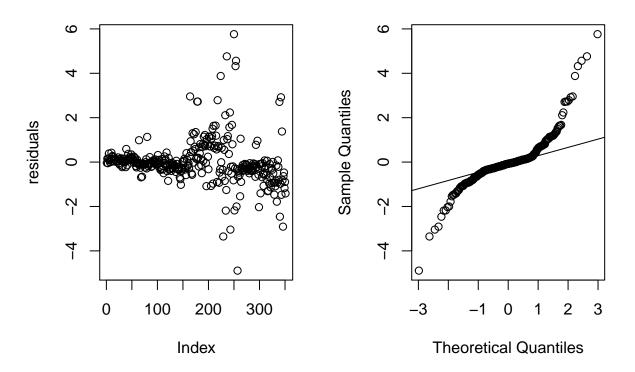
2. Exercise 3.24

```
my_model = lm(TotalPrice ~ Carat + I(Carat^2), data=diamonds)
summary(my_model)
##
## Call:
## lm(formula = TotalPrice ~ Carat + I(Carat^2), data = diamonds)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
## -10207.4
             -711.6
                      -167.9
                                355.0 12147.3
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                -522.7
                            466.3 -1.121 0.26307
## Carat
                2386.0
                            752.5
                                    3.171 0.00166 **
## I(Carat^2)
                4498.2
                            263.0 17.101 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2127 on 348 degrees of freedom
## Multiple R-squared: 0.9257, Adjusted R-squared: 0.9253
## F-statistic: 2168 on 2 and 348 DF, p-value: < 2.2e-16
```

2-(a). Check for homoscedacity and normality

```
par(mfrow=c(1,2))
residuals = rstandard(my_model)
plot(residuals)
qqnorm(residuals)
qqline(residuals)
```

Normal Q-Q Plot



There is heteroskedacity and a lack of normality in my model.

2-(b). Transform response variable to natural log

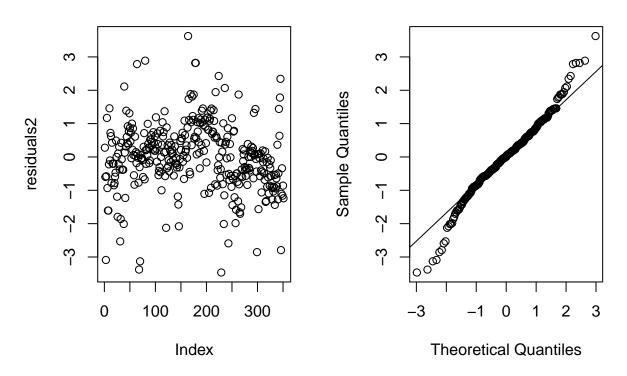
```
log_model = lm(log(TotalPrice) ~ Carat + I(Carat^2), data=diamonds)
summary(log_model)
##
  lm(formula = log(TotalPrice) ~ Carat + I(Carat^2), data = diamonds)
##
##
## Residuals:
##
      Min
                1Q
                   Median
                                3Q
                                       Max
##
  -0.8215 -0.1313 0.0003 0.1391
##
##
  Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
##
               6.13042
                           0.05218
                                    117.48
                                             <2e-16 ***
  (Intercept)
                                             <2e-16 ***
## Carat
                3.05963
                           0.08422
                                     36.33
## I(Carat^2)
               -0.52730
                           0.02944
                                    -17.91
                                             <2e-16 ***
## ---
                  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
## Residual standard error: 0.238 on 348 degrees of freedom
## Multiple R-squared: 0.925, Adjusted R-squared: 0.9246
## F-statistic: 2146 on 2 and 348 DF, p-value: < 2.2e-16
```

My "best" choice of predictors is still a reasonable choice for predicting logPrice.

2-(c). Check graphs for log model

```
par(mfrow=c(1,2))
residuals2 = rstandard(log_model)
plot(residuals2)
qqnorm(residuals2)
qqline(residuals2)
```

Normal Q-Q Plot



The log transformation helped with heteroskedacity and normality. The assumptions are now fairly reasonable.

4. Exercise 3.25

Use nested F-test for the complete second-order model to determine if all the terms in the model that involve Depth could be removed withought significantly impairing its effectiveness.

```
full_model = model_d
reduced_model = my_model
anova(reduced_model, full_model)

## Analysis of Variance Table
##
## Model 1: TotalPrice ~ Carat + I(Carat^2)
## Model 2: TotalPrice ~ Carat + Depth + I(Carat^2) + I(Depth^2) + Carat *
## Depth
```

```
## Res.Df RSS Df Sum of Sq F Pr(>F)
## 1 348 1574044410
## 2 345 1454702094 3 119342316 9.4345 5.24e-06 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.05 '.' 0.1 ' ' 1
```

There exist some term that involves information on Depth that is significant.

4. Exercise 3.26

Couple found a 0.5-carat diamond with depth of 62%.

4-(a). Determine average total price for quadratic model

```
new_diamond = data.frame(Carat=0.5, Depth=62.0)
predict(my_model, new_diamond)
## 1
```

1794.843

The quadratic model predicts \$1794.84 as the average total price.

4-(b). 95% confidence interval for the mean total price

```
predict(my_model, new_diamond, interval="c")
```

```
## fit lwr upr
## 1 1794.843 1424.296 2165.389
```

With 95% confidence, the mean total price for a 0.5-car at diamond with a depth of 62% somewhere between \$1424.30 and \$2165.40

4-(c). 95% prediction interval

```
predict(my_model, new_diamond, interval="predict")
```

```
## fit lwr upr
## 1 1794.843 -2404.462 5994.147
```

With 95% confidence, we predict that the total price for a 0.5-carat diamond with depth of 62% is somewhere between \$0 and \$5994.15

Note: Negative dollars or recieving money for purchasing a diamond doesn't make sense.

4-(d). Repeat two intervals for logPrice

```
exp(predict(log_model, new_diamond, interval="c"))
## fit lwr upr
## 1 1860.149 1784.588 1938.908
```

With 95% confidence, the mean total price for a 0.5-car at diamond with a depth of 62% somewhere between \$1784.59 and \$1938.91

```
exp(predict(log_model, new_diamond, interval="p"))
```

fit lwr upr ## 1 1860.149 1162.651 2976.09

With 95% confidence, we predict that the total price for a 0.5-car at diamond with depth of 62% is somewhere between \$1162.65 and \$2976.09