# 605 Wk 12 Discussion - Predicting Diamond Price

Jen Abinette 2023-04-20

## **Assignment**

Build a multiple regression model for data that interests you. Include in this model at least one quadratic term, one dichotomous term, and one dichotomous vs. quantitative interaction term. Interpret all coefficients. Conduct residual analysis. Was the linear model appropriate? Why or why not?

#### Dichotomous variable

```
# New data frame that remove cases where cut is Fair
df <- subset (diamonds, cut != 'Fair')
# Create dichotomous variable with Good and Very Good as 0 and Premium and Ideal as 1
df$cut_dich <- ifelse(grepl("Good",df$cut),0,1)
# Quadratic variable
df$carat.sq <- df$carat**2</pre>
```

## Regression Modeling

Research Question - How can we predict diamond price?

```
df2.lm <- lm(price ~ carat + carat.sq + cut_dich + carat*cut_dich, data=df)
summary(df2.lm)</pre>
```

```
##
## Call:
## lm(formula = price ~ carat + carat.sq + cut_dich + carat * cut_dich,
       data = df
##
## Residuals:
       Min
##
                 1Q
                      Median
                                   3Q
                                           Max
## -20868.3
                     -39.6
            -677.6
                                397.6 13054.7
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 -1766.02
                               29.56 -59.739 < 2e-16 ***
## carat
                 6170.55
                               51.97 118.724 < 2e-16 ***
## carat.sq
                   772.89
                               21.79 35.471 < 2e-16 ***
## cut_dich
                   120.44
                               27.77 4.337 1.45e-05 ***
## carat:cut_dich 140.87
                               29.85 4.719 2.38e-06 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1483 on 52325 degrees of freedom
## Multiple R-squared: 0.8627, Adjusted R-squared: 0.8626
## F-statistic: 8.216e+04 on 4 and 52325 DF, p-value: < 2.2e-16
```

What about using squareroot of carat instead of carat squared?

```
df.lm <- lm(price ~ carat + sqrt(carat) + cut_dich + carat*cut_dich, data=df)
summary(df.lm)</pre>
```

```
##
## Call:
## lm(formula = price ~ carat + sqrt(carat) + cut_dich + carat *
       cut_dich, data = df)
##
##
## Residuals:
##
       Min
                 10 Median
                                   3Q
                                          Max
## -19591.9 -559.5
                       -38.7
                                279.7 13153.6
##
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                 1740.52
                              77.48 22.464 < 2e-16 ***
                               93.21 137.720 < 2e-16 ***
## carat
                 12836.40
## sqrt(carat)
                 -9513.59
                              169.59 -56.096 < 2e-16 ***
## cut dich
                              27.29 3.975 7.04e-05 ***
                   108.50
## carat:cut_dich 135.39
                               29.34 4.614 3.95e-06 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1457 on 52325 degrees of freedom
## Multiple R-squared: 0.8673, Adjusted R-squared: 0.8673
## F-statistic: 8.552e+04 on 4 and 52325 DF, p-value: < 2.2e-16
```

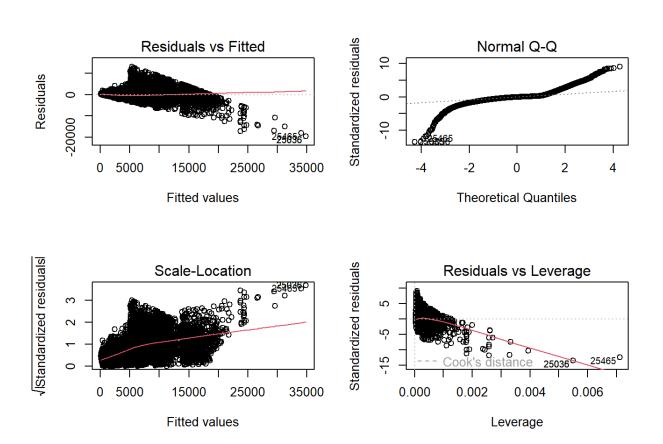
 $price = -1740.52 + 12836.40 imes carat - 9513.59 imes \sqrt{carat} + 108.5 imes cut. \ dich + 135.39 imes carat imes cut. \ dich$ 

Since the effect size is slightly larger and the residuals more normally distributed using the square root instead of quadratic version of carat, I chose to move forward with that model. Based on the summary of the model above, the coefficients for the model are all statistically significant with p-values less than .001 and minimal variability given the standard error for the intercept and slopes are much smaller than the coefficients. Additionally, Multiple R-squared indicates our model predicts 87% of the variability in diamond price.

### Residual Analysis

To assess whether the linear model is reliable, we need to check for (1) linearity, (2) nearly normal residuals, and (3) constant variability

par(mfrow=c(2,2))
plot(df.lm)



An analysis of the residuals plots indicates a problem with our model assumptions. The plots above show a lack of constant variability and normal distribution. In particular, the residuals are not uniformly scattered above and below zero on the residuals plot and the 2 ends diverge on the Q-Q plot. Although our model has a large effect size, it is still not the best model as it is not fully explaining the variability in price.