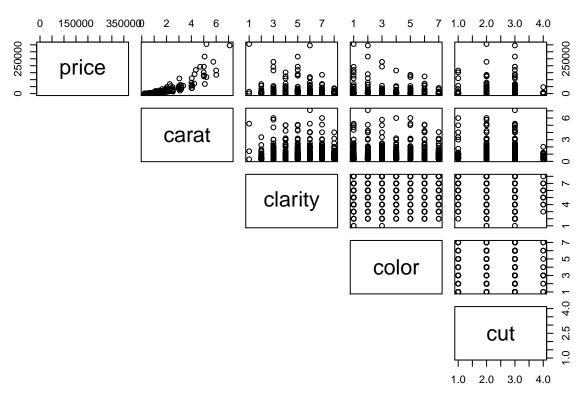
STAT 6021: Project 1

Greg Madden, Maxwell Levinson, Andrew Setaro, and Trey Hamilton

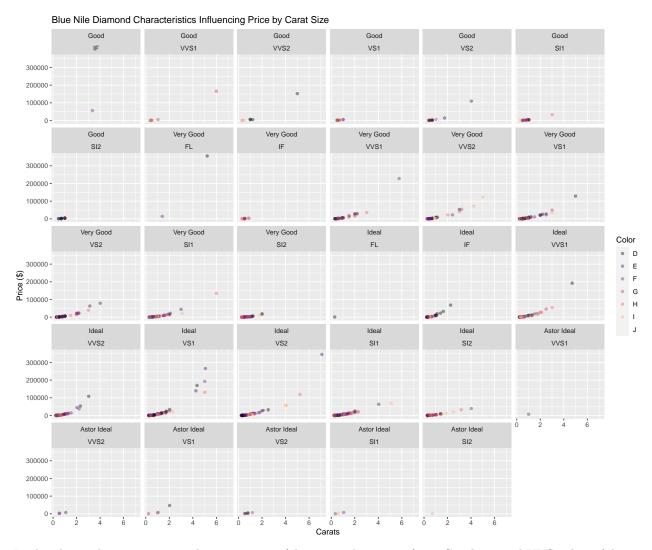
3/3/2022

We have been approached by Blue Nile to perform the following tasks:

1. Use data visualizations to explore how price is related to the other variables (carat, clarity, color, cut), as well as how the other variables may relate to each other.



Note that the x-axis above corresponds to increasing desireability of the factored categorical variables: clarity, color, and cut. In the above scatterplot matrix how price appears to have the clearest linear relationship with carats.



In the above plot, you can see that certain cut/clarity combinations (e.g., Good cut and VVS2 clarity) have different slopes in terms of their price \sim carat relationships. For example, ideal cut with FL (Flawless) clarity appears to have a higher price per additional carat size than the Ideal Cut with SI2 (Slightly Included). In addition, more desireable colors (i.e.D-F) appear to cluster at the lower price ranges.

Address the various claims on the diamond education page on Blue Nile.

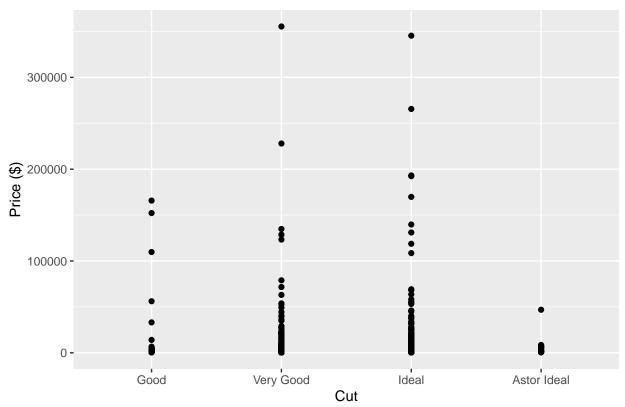
• Cut: https://www.bluenile.com/education/diamonds/cut +"A diamond's cut refers to how well-proportioned the dimensions of a diamond are, and how these surfaces, or facets, are positioned to create sparkle and brilliance. For example, what is the ratio of the diamond's diameter in comparison to its depth? These small, yet essential, factors determine the diamond's beauty and price."

Assertion above is that better cuts correlate with higher price. Let's check the scatterplot to see if that bears out in the data:

Increasing quality of diamond cut does not seem to have a linear relationship with price, contrary to the Blue Nile's claim.

'geom_smooth()' using formula 'y ~ x'

Blue Nile Diamonds: Cut and Price

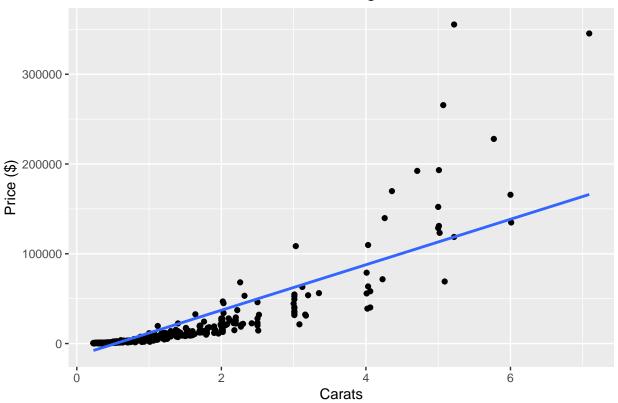


2. Fit an appropriate simple linear regression for price against carat.

First a scatterplot

'geom_smooth()' using formula 'y ~ x'

Blue Nile Diamonds: Factors Influencing Price

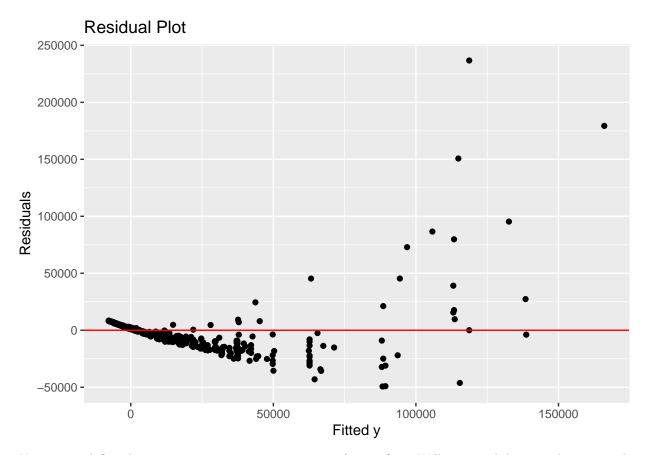


Model Summary

```
##
## Call:
## lm(formula = price ~ carat, data = Data)
##
## Residuals:
##
     Min
            1Q Median
                           ЗQ
                                 Max
## -49375 -5048
                         4965 236711
                  1867
##
## Coefficients:
##
              Estimate Std. Error t value
                                                     Pr(>|t|)
                            559.7 -24.21 <0.0000000000000000 ***
## (Intercept) -13550.9
## carat
               25333.9
                            494.4 51.24 <0.0000000000000000 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 13560 on 1212 degrees of freedom
## Multiple R-squared: 0.6842, Adjusted R-squared: 0.6839
## F-statistic: 2625 on 1 and 1212 DF, p-value: < 0.00000000000000022
```

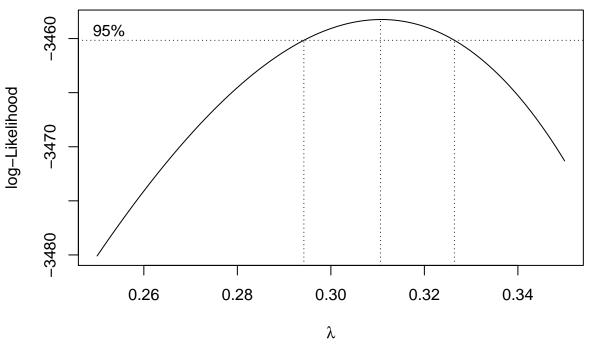
Plotting the residuals.

 $Constant\ variance\ and\ mean\ of\ error=0\ assumptions\ do\ not\ appear\ to\ be\ met.$



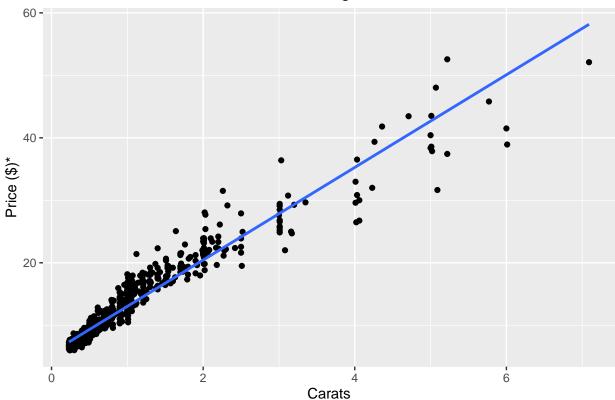
Variance is definitely not contant so attempting to transform y first. Will start with boxcox plot to see what the optimal lambda may be.

Looked like a lambda of 0.31 would be appropriate.



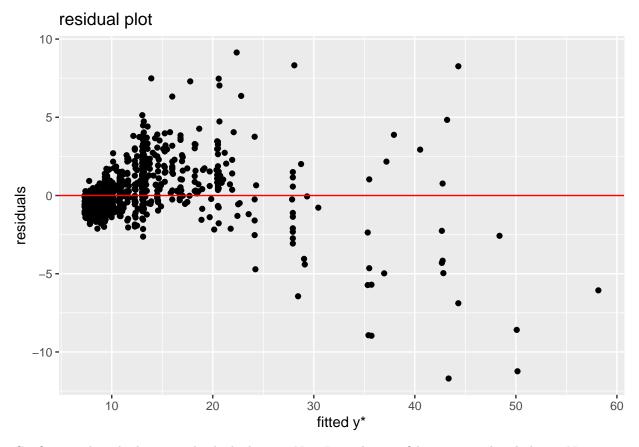
Replotting the scatter plot with the transformed y variable.

Blue Nile Diamonds: Factors Influencing Price

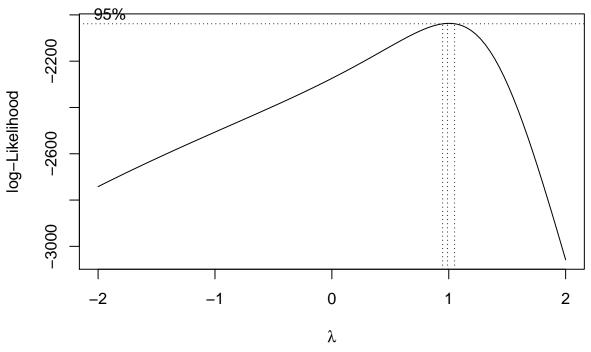


New residual plot:

variance looks better but mean of errors still not equal to zero over x so will attempt to transform the x variable. Given the curved appearance, will try a square root transformation.

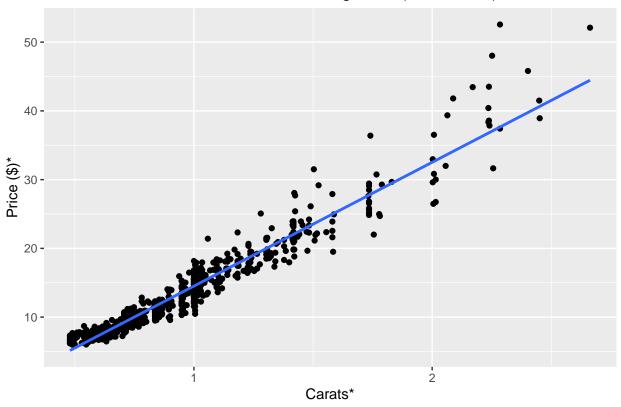


Confirming that the box \cos plot looks better. Now I see that confidence interval includes 1. Next step is to consider whether or not to transform x.



'geom_smooth()' using formula 'y ~ x'

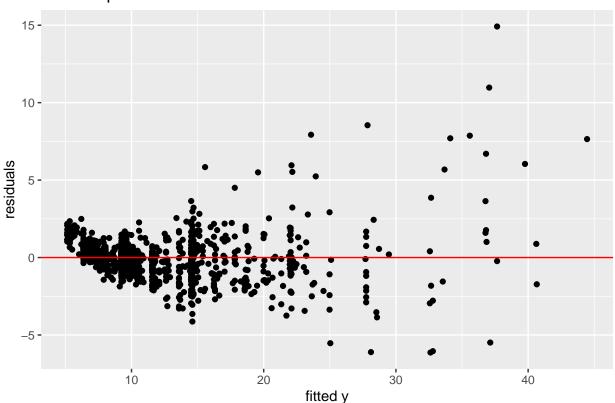
Blue Nile Diamonds: Factors Influencing Price (SLR Model)



Fitting new model and creating yet another residual plot.

Not a perfect fit but overall improved and adequate for prediction.





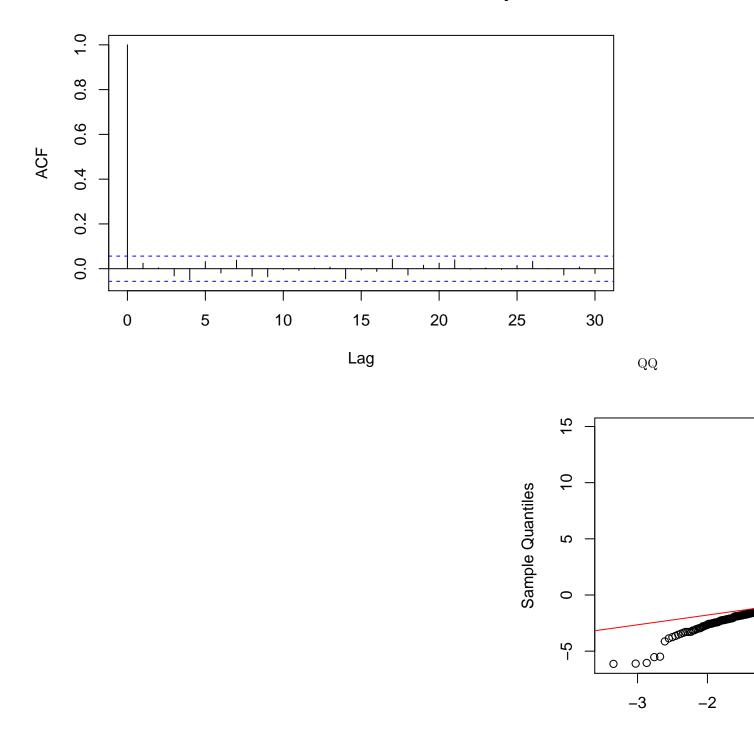
Summarizing the model below.

```
##
## Call:
## lm(formula = ystar ~ xstar, data = Data)
##
## Residuals:
##
       Min
                1Q Median
                               ЗQ
                                      Max
## -6.1320 -0.6377 0.0373 0.5315 14.9172
##
## Coefficients:
              Estimate Std. Error t value
                                                     Pr(>|t|)
##
## (Intercept) -3.4936
                            0.1137 -30.73 <0.0000000000000000 ***
                18.0085
                            0.1261 142.87 < 0.0000000000000000 ***
## xstar
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 1.436 on 1212 degrees of freedom
## Multiple R-squared: 0.9439, Adjusted R-squared: 0.9439
## F-statistic: 2.041e+04 on 1 and 1212 DF, p-value: < 0.000000000000000022
```

ACF Plot

Errors do not appear correlated to each other.

ACF Plot of Residuals with xstar and ystar



Plot Normality assumption is not met but this may be the least important.