Regression models project

Mathew Erzoah

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Executive Summary

In this report, we will examine the mtcars data set and explore how miles per gallon (MPG) is affected by different variables. In particularly, we will answer the following two questions:

- 1. Is an automatic or manual transmission better for MPG, and
- 2. Quantify the MPG difference between automatic and manual transmissions.

From our analysis we can show that manual transmission has an MPG 1.8 greater than an automatic transmission.

Exploratory Analysis

```
library(ggplot2) #for plots
## Warning: package 'ggplot2' was built under R version 4.0.2
data(mtcars)
head(mtcars)
                      mpg cyl disp hp drat
##
                                                wt qsec vs am gear carb
## Mazda RX4
                     21.0
                               160 110 3.90 2.620 16.46
## Mazda RX4 Wag
                     21.0
                            6
                               160 110 3.90 2.875 17.02
## Datsun 710
                     22.8
                            4 108 93 3.85 2.320 18.61
                                                                        1
## Hornet 4 Drive
                     21.4
                            6 258 110 3.08 3.215 19.44
                               360 175 3.15 3.440 17.02
                                                                        2
## Hornet Sportabout 18.7
                            8
## Valiant
                     18.1
                               225 105 2.76 3.460 20.22 1
# Transform certain variables into factors
mtcars$cyl <- factor(mtcars$cyl)</pre>
           <- factor(mtcars$vs)
mtcars$vs
mtcars$gear <- factor(mtcars$gear)</pre>
mtcars$carb <- factor(mtcars$carb)</pre>
```

To help us understand the data, we build exploratory plots. Appendix - Plot 1, shows there is a definite impact on MPG by transmission with Automatic transmissions having a lower MPG.

<- factor(mtcars\$am,labels=c("Automatic","Manual"))</pre>

Regression Analysis

17.14737 24.39231

We've visually seen that automatic is better for MPG, but we will now quantify this difference.

```
aggregate(mpg~am, data = mtcars, mean)

## am mpg
## 1 Automatic 17.14737
## 2 Manual 24.39231
```

Thus we hypothesize that automatic cars have an MPG 7.25 lower than manual cars. To determine if this is a significant difference, we use a t-test.

```
D_automatic <- mtcars[mtcars$am == "Automatic",]
D_manual <- mtcars[mtcars$am == "Manual",]
t.test(D_automatic$mpg, D_manual$mpg)

##

## Welch Two Sample t-test
##

## data: D_automatic$mpg and D_manual$mpg
## t = -3.7671, df = 18.332, p-value = 0.001374

## alternative hypothesis: true difference in means is not equal to 0

## 95 percent confidence interval:
## -11.280194 -3.209684

## sample estimates:
## mean of x mean of y</pre>
```

The p-value is 0.001374, thus we can state this is a significant difference. Now to quantify this.

```
init <- lm(mpg ~ am, data = mtcars)
summary(init)</pre>
```

```
##
## lm(formula = mpg ~ am, data = mtcars)
##
## Residuals:
      Min
               1Q Median
                               3Q
                                      Max
## -9.3923 -3.0923 -0.2974 3.2439 9.5077
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                17.147
                            1.125 15.247 1.13e-15 ***
                 7.245
                            1.764
                                   4.106 0.000285 ***
## amManual
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 4.902 on 30 degrees of freedom
## Multiple R-squared: 0.3598, Adjusted R-squared: 0.3385
## F-statistic: 16.86 on 1 and 30 DF, p-value: 0.000285
```

This shows us that the average MPG for automatic is 17.1 MPG, while manual is 7.2 MPG higher. The R2 value is 0.36 thus telling us this model only explains us 36% of the variance. As a result, we need to build a multivariate linear regression.

The new model will use the other variables to make it more accurate. We explore the other variable via a pairs plot (Appendix - Plot 2) to see how all the variables correlate with mpg. From this we see that cyl, disp, hp, wt have the strongest correlation with mpg. We build a new model using these variables and compare them to the initial model with the anova function.

```
betterFit <- lm(mpg~am + cyl + disp + hp + wt, data = mtcars)
anova(init, betterFit)</pre>
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + cyl + disp + hp + wt
## Res.Df RSS Df Sum of Sq F Pr(>F)
## 1 30 720.90
## 2 25 150.41 5 570.49 18.965 8.637e-08 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

This results in a p-value of 8.637e-08, and we can claim the betterFit model is significantly better than our init simple model. We double-check the residuals for non-normality (Appendix - Plot 3) and can see they are all normally distributed and homoskedastic.

summary(betterFit)

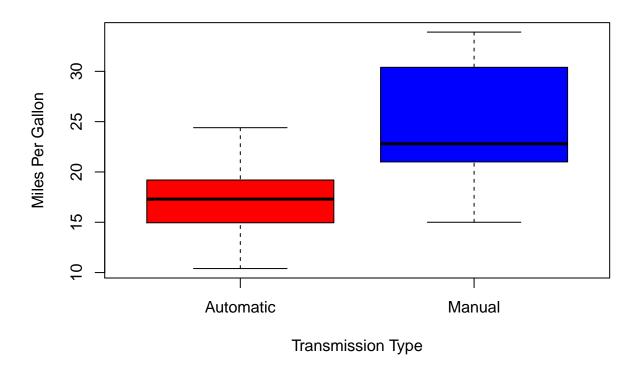
```
##
## Call:
## lm(formula = mpg ~ am + cyl + disp + hp + wt, data = mtcars)
## Residuals:
##
                1Q Median
       Min
                                3Q
                                        Max
  -3.9374 -1.3347 -0.3903
                           1.1910
                                    5.0757
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 33.864276
                           2.695416
                                     12.564 2.67e-12
## amManual
                1.806099
                           1.421079
                                      1.271
                                               0.2155
                                     -2.135
## cyl6
               -3.136067
                           1.469090
                                               0.0428 *
                                     -0.938
## cy18
               -2.717781
                           2.898149
                                               0.3573
                0.004088
## disp
                           0.012767
                                       0.320
                                               0.7515
               -0.032480
                           0.013983
                                      -2.323
                                               0.0286 *
## hp
## wt
               -2.738695
                           1.175978
                                     -2.329
                                               0.0282 *
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 2.453 on 25 degrees of freedom
## Multiple R-squared: 0.8664, Adjusted R-squared: 0.8344
## F-statistic: 27.03 on 6 and 25 DF, p-value: 8.861e-10
```

The model explains 86.64% of the variance and as a result, cyl, disp, hp, wt did affect the correlation between mpg and am. Thus, we can say the difference between automatic and manual transmissions is 1.81 MPG.

Appendix

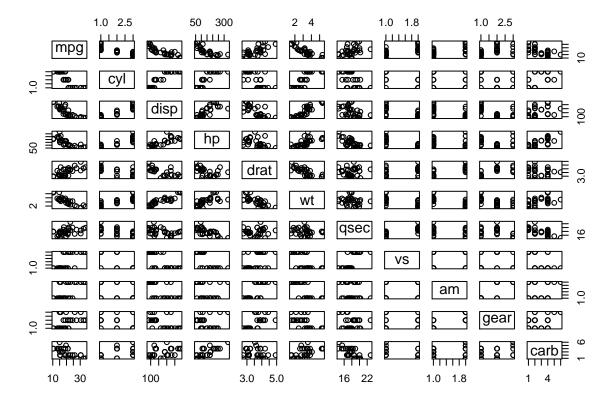
Plot 1: Boxplot of MPG by transmission type

```
boxplot(mpg ~ am, data = mtcars, col = (c("red","blue")),
   ylab = "Miles Per Gallon", xlab = "Transmission Type")
```



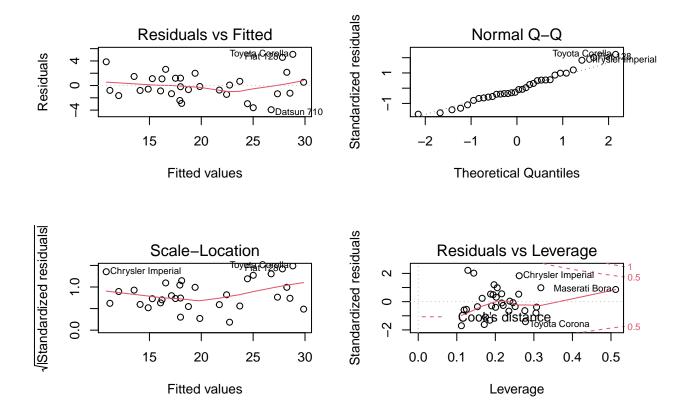
Plot 2: Pairs plot for the dataset

```
pairs(mpg ~ ., data = mtcars)
```



Plot 3: Check residuals

```
par(mfrow = c(2,2))
plot(betterFit)
```



Source of document

This report was created using the knitr package, source is available on the link My knitr code