Motor Trend - Effects of transmission on MPG

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Executive Summary

This report is part of a course project within the **Regression Models** course on the **Data Science Specialization** by **Johns Hopkins University** on **Coursera**. On this report we will analyze the **mtcars** data set and explore the relationship between the type of transmission (manual or automatic), among other variables, and miles per gallon (MPG), which will be our outcome.

We are particularly interested in the following two questions:

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

From our analysis we can conclude that cars with manual transmission get more miles per gallon than cars with automatic transmission by a rate of 1.8 adjusted by number of cylinders, gross horspower and weight.

Exploratory Data Analysis

In this section we explore some relationships between variables of the data set and our outcome. First we plot the relationship between all variables of the data set (see Figure 2 in the appendix). From this plot we can see a strong correlation between variables such as: **disp**, **hp**, **drat**, **wt**, and **am** with our outcome **mpg**.

Since we are actually interested in quantifying the MPG difference between automatic and manual transmissions we also plot a boxplot between **mpg** and **am** (see Figure 1 in the appendix). We can see that there is an increase in **mpg** when the transmission is **manual**.

Regression

In this section we build a linear regression using different variables as predictors and **mpg** as outcome. To find the best model we perform a stepwise selection using both forward selection and backward elimination. The code below takes care of this.

```
first_model <- lm(mpg ~., mtcars)
best_model <- step(first_model, direction = "both")</pre>
```

We can see that the best model uses the variables **cyl**, **hp** and **wt** as confounders and **am** as the independent variable.

```
best_model$call
```

```
## lm(formula = mpg ~ cyl + hp + wt + am, data = mtcars)
```

We can also use **anova** to compare a base model, that only uses **am** as a predictor, and the best model that was found performing stepwise selection.

```
base_model <- lm(mpg~am, mtcars)
anova(base_model, best_model)</pre>
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ cyl + hp + wt + am
##
    Res.Df
               RSS Df Sum of Sq
                                          Pr(>F)
## 1
         30 720.90
## 2
         26 151.03
                         569.87 24.527 1.688e-08 ***
## ---
                  0 '*** 0.001 '** 0.01 '* 0.05 '. ' 0.1 ' 1
## Signif. codes:
```

Since the **p-value** is highly significant we reject the null hypothesis that the variables **cyl**, **hp** and **wt** do not contribute to the model.

Diagnostics

Finally we study the residual plots of the regression model and do some diagnostics. From the residual plot (see Figure 3 in the appendix) we can see that the **Residuals x Fitted** plot seems randomly scattered, what verifies the independence condition. The **Normal Q-Q** plot indicates that the residuals are normally distributed and the **Scale-Location** plot indicates constant variance. We can also notice some outliers or leverage points from the plots, let's take a look in the hatvalues and dfbetas to see if they are the same as the ones showing in the plots.

```
hat_values <- hatvalues(best_model)</pre>
df betas <- dfbetas(best model)</pre>
tail(sort(hat_values),3)
##
         Toyota Corona Lincoln Continental
                                                     Maserati Bora
##
              0.2777872
                                    0.2936819
                                                         0.4713671
tail(sort(df_betas[,6]), 3)
## Chrysler Imperial
                                Fiat 128
                                              Toyota Corona
##
           0.3507458
                               0.4292043
                                                   0.7305402
```

We can see from the analysis above that the cars showing in the residual plots are the same as the ones found in our diagnosis.

Inference and Conclusion

By performing a t-test we can conclude that manual and automatic transmission are significatively different.

```
t.test(mpg~am, mtcars)

##
## Welch Two Sample t-test
##
```

```
## data: mpg by am
## 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 in group Automatic mean in group Manual
## 17.14737 24.39231
```

Let's take a look at the coefficients of our best model and draw some conclusions from it.

best_model\$coefficients

```
## (Intercept) cyl6 cyl8 hp wt amManual ## 33.70832390 -3.03134449 -2.16367532 -0.03210943 -2.49682942 1.80921138
```

From this coefficients we can conclude the following:

• Cars with manual transmission get more miles per gallon than cars with automatic transmission by a rate of 1.8, adjusted by number of cylinders, gross horspower and weight.

Appendix



Figure 1: Boxplot between mpg and am

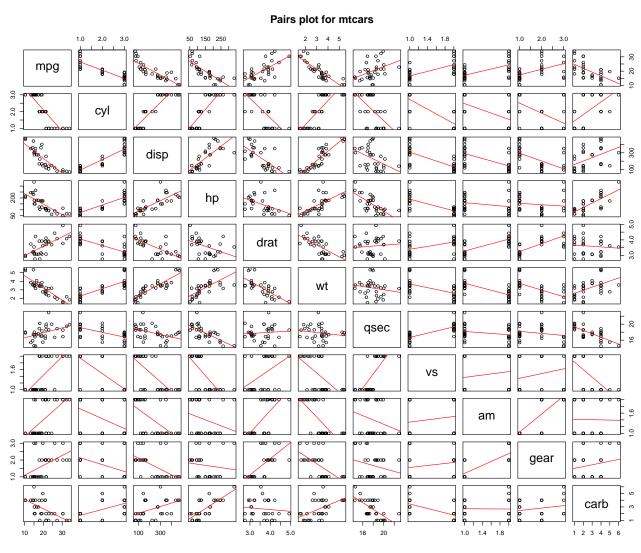


Figure 2: Pairs plot for mtcars

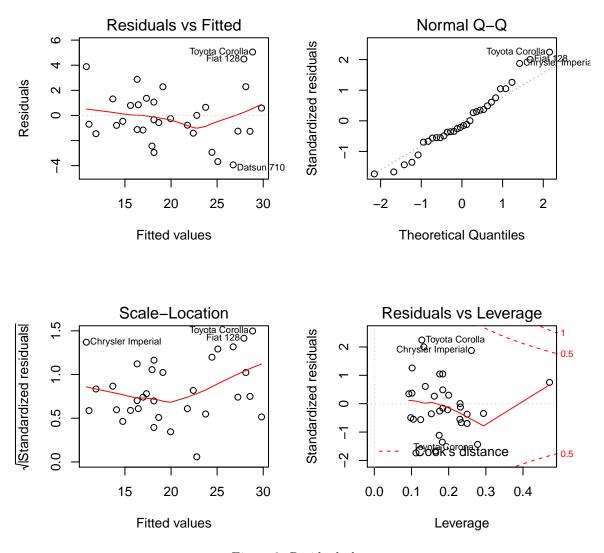


Figure 3: Residual plots