The Relationship Between Transmission Type and Fuel Efficiency

Coursera - Regression Models

Executive Summary

From our analysis of the mtcars dataset, we have determined that in general manual transmissions are better in terms of miles per gallon than automatic transmissions. In a linear regression model with only transmission type as an explanatory variable, a change from automatic to manual transmission increased the mpg by 7.245 however, transmission type only explained 36% of the variation in mpg. A linear regression model of all significant variables (determined by ANOVA), explained 84% of the variation in mpg. It included only the variables weight and number of cylinders. Transmission type was determined to be an insignificant contributory variable to the model. Furthermore, when transmission type was included in the model, a Bootstrap Measures of Relative Importance showed that it only contributed only about 14% to the r^2 of 87%. It is recommended that the editors of *Motor Trend* consider the varibles weight, number of cylinders, and possibly horsepower as the most significant explanatory variables of miles per gallon.

Report

The Data

We are to investigate the relationship between miles per gallon (numerical class variable, mpg) and a set of explanatory variables. The explanatory variables and their classes are:

1. cyl: number of cylinders (factor, 4,6,8)
2. disp: displacement (cu.in.) (numerical)
3. hp: gross horsepower (numerical)
4. drat: rear axle ratio (numerical)
5. wt: weight (1000 pounds) (numerical)
6. qsec: 1/4 mile time (numerical)
7. vs: V/S, V-engine or Straight engine (factor, V,S)
8. am: transmission type (factor, automatic, manual)
9. gear: number of forwards gears (factor, 3,4,5)
10. carb: number of carburetors (factor, 1,2,3,4,5,6,7,8)

Data Processing

We will remove the variable qsec, 1/4 mile time, as this is not a reasonable explanatory variable for mpg and is better seen as another outcome variable. In addition, it is necessary to code the variables with their proper class (factor, numerical.)

```
mtcars$cyl=factor(mtcars$cyl)
mtcars$vs=factor(mtcars$vs)
mtcars$gear=factor(mtcars$gear)
mtcars$carb=factor(mtcars$carb)
mtcars$am=factor(mtcars$am,labels=c('Automatic','Manual'))
auto=subset(mtcars, am=="Automatic")
man=subset(mtcars, am=="Manual")
```

Exploratory Data Analysis

First we must determine whether or not there is actually a difference between automatic and manual transmissions in terms of mpg. The following boxplot appears to show that there is a difference.

Comparison of MPG of Automatic vs. Manual Transmission
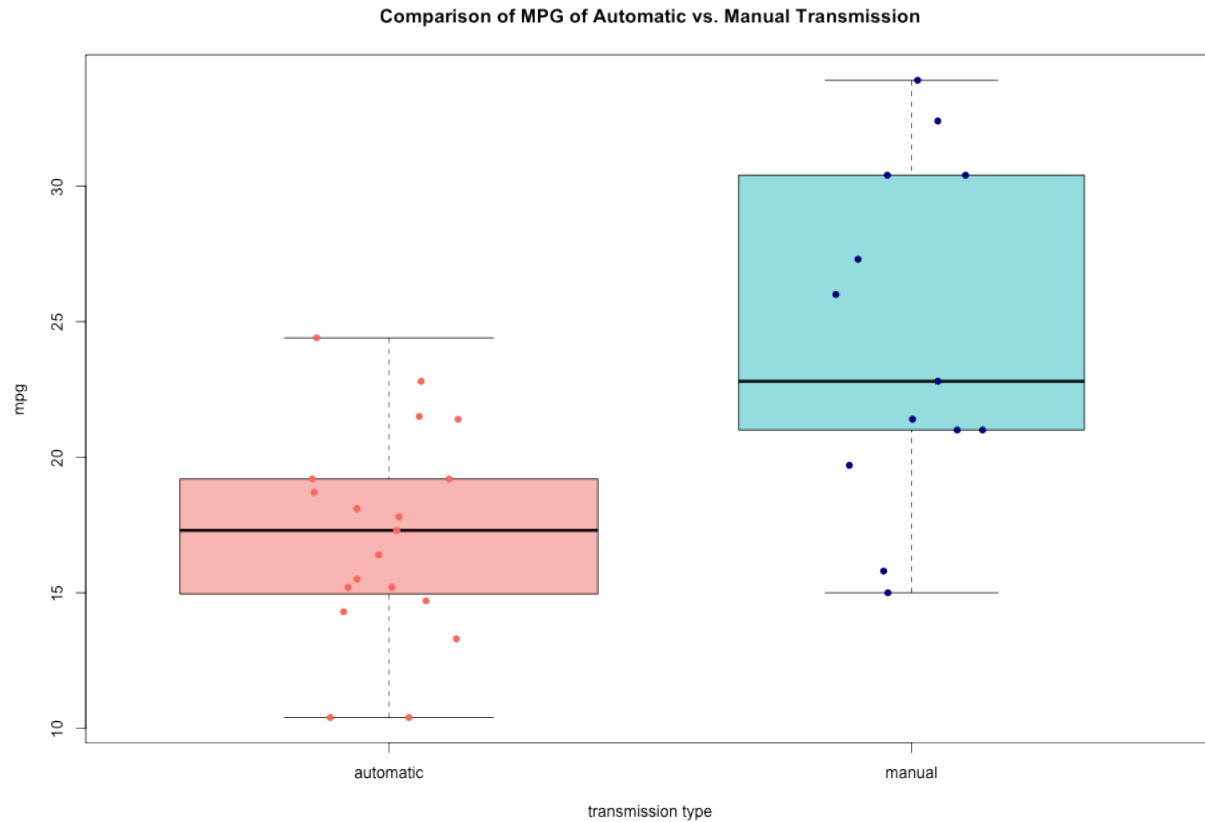
To be sure that the means are different, we can perform a t-test. The t-test results in a p-value of `r t`. This means that we reject the null hypothesis that the means are similar.

Pairs Plot

The following is a chart plotting the more significant variables against each other (as determined by the models below). We can see that there are correlations between many variables therefore multicollinearity may be an issue. It is color-coded by transmission type.

```
library(ggplot2)
library(GGally)

pairsGG=mtcars[,c(1,2,3,4,6,9)]
ggpairs(pairsGG,lower = list(continuous="smooth",theme_set(theme_bw())),title="mtcars",colour = "am")
```
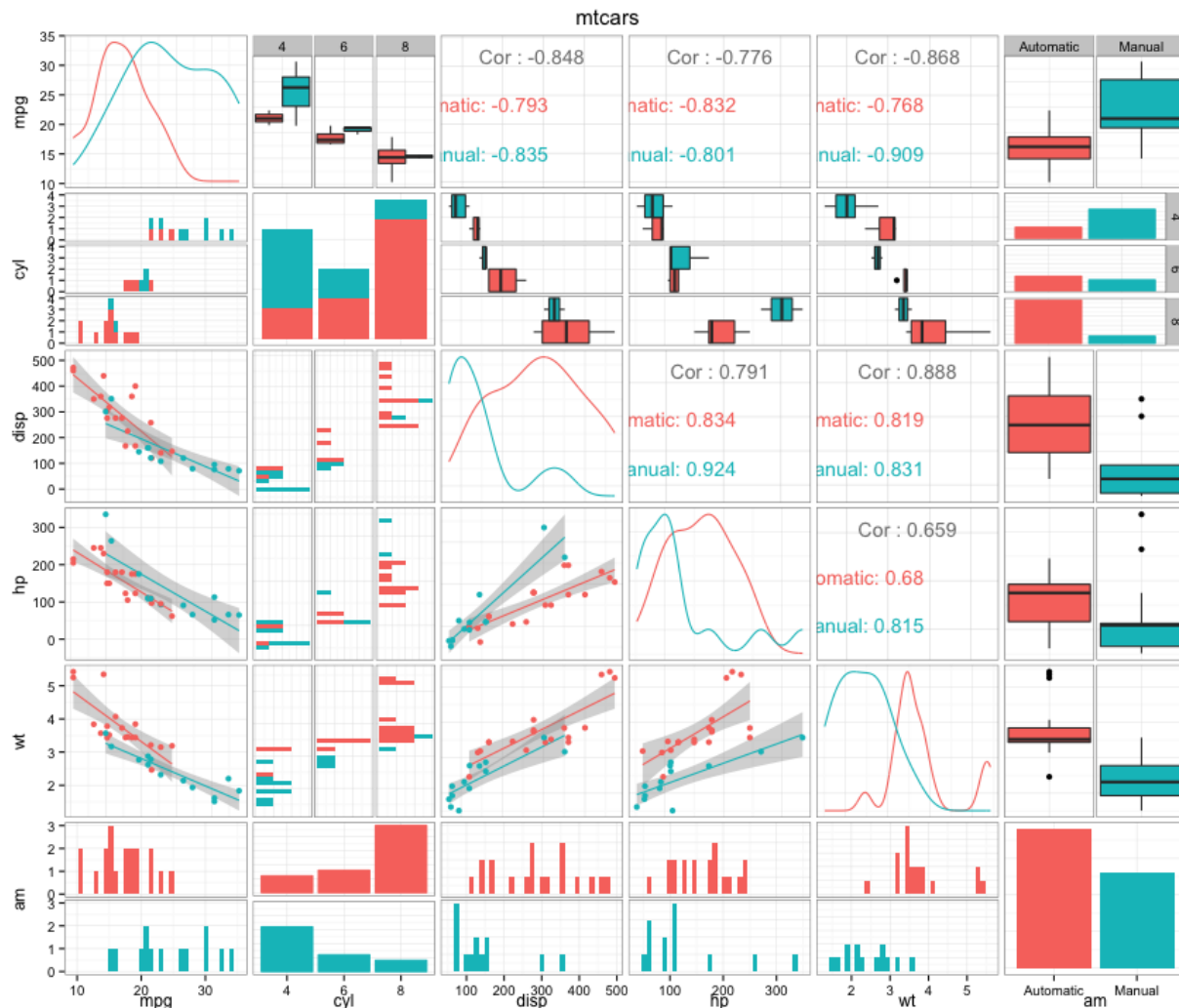
Building the Model

We can now turn towards building the regression model for this dataset. We will first use multiple linear regression and the R "step" function, which chooses the optimal model by AIC (Aikake Information Criterion). With mpg as the outcome variable and all other variables, except 1/4 mile, as the explanatory variables we arrive at the model:

```
## (Intercept)        cyl6        cyl8          hp          wt      amManual
##      33.708      -3.031      -2.164      -0.032      -2.497        1.809
```

and the coefficients' corresponding p-values:

```
## (Intercept)        cyl6        cyl8          hp          wt      amManual
##     0.00000     0.04068     0.35225     0.02693     0.00908       0.20646
```

The model here is: $\text{mpg} \sim \text{cyl} + \text{hp} + \text{wt} + \text{am}$. The r^2 for this model is `r round(sumAIC$r.squared,3)` which means that this model explains `r round(sumAIC$r.squared,3)*100`% of the variation in mpg. The model, arrived at by AIC, includes the variable am, or transmission type. The coefficient for am is `r round(fitAIC$coefficients[6],3)`, which we can interepret as, when other variables are held constant, a change from automatic to manual transmission will increase the mpg by `r`

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`round(fitAIC$coefficients[6],3)`. However, we must note the p-value associated with the transmission type variable. It is `r round(sumAIC$coefficients[24],3)` which is well above our usual 0.05 significant level.

If we remove the transmission type from this model, and refit it we obtain a model mpg ~ cyl + hp + wt. The coefficients and corresponding p-values are:-

```
## (Intercept)         cyl6         cyl8           hp           wt
##      35.846       -3.359       -3.186       -0.023       -3.181


## (Intercept)         cyl6         cyl8           hp           wt
##     0.00000      0.02375      0.15370      0.06361      0.00014
```

An ANOVA test between this model and this previous results in a p-value of `r pANOVA1` which means that we should choose the simpler model. However, we can further simplify the model by removing the hp or horsepower variable since the p-value for its coefficient is above 0.05. This results in the model mpg ~ cyl + wt. An ANOVA test between this model and previous two models resulted in p-values of `r pANOVA2` and `r pANOVA3`, respectively. Both are above 0.05 so we can choose the simpler model.
The final model is:-

```
##
## Call:
## lm(formula = mpg ~ cyl + wt, data = mtcars)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -4.5890 -1.2357 -0.5159  1.3845  5.7915
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  33.9908     1.8878  18.006  < 2e-16 ***
## cyl6         -4.2556     1.3861  -3.070 0.004718 **
## cyl8         -6.0709     1.6523  -3.674 0.000999 ***
## wt           -3.2056     0.7539  -4.252 0.000213 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.557 on 28 degrees of freedom
## Multiple R-squared:  0.8374, Adjusted R-squared:   0.82
## F-statistic: 48.08 on 3 and 28 DF,  p-value: 3.594e-11
```

Here, all of the coefficients are highly significant. This model has an r^2 of `r round(sumNOhp$r.squared,3)` which means that it explains `r round(sumNOhp$r.squared,3)*100`% of the variation in mpg. This is very close to our original model with 2 more variables. Note that transmission type is not included in the final model.

The plots below are the diagnostic plots for the model. There do not appear to be any problems with these plots; the residuals appear randomly, the standardized residuals appear normally distributed, and there are not any highly influential outliers.
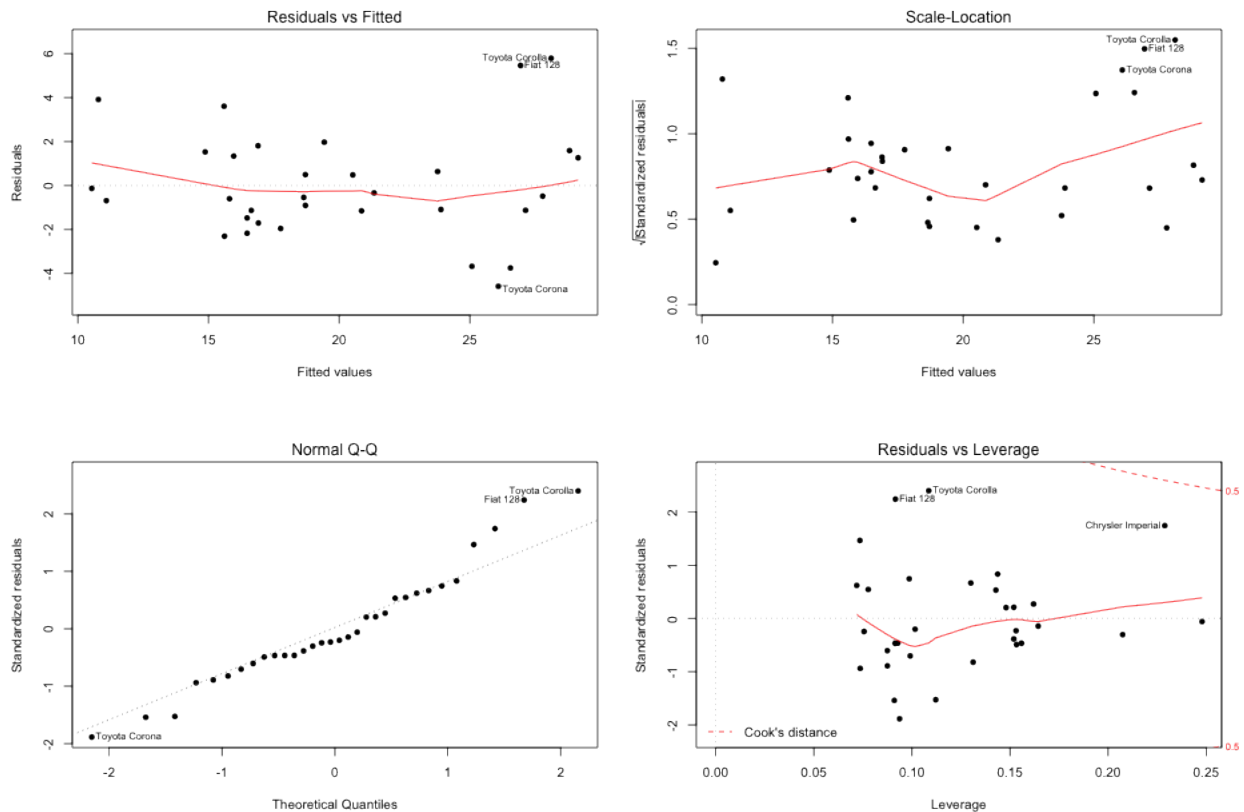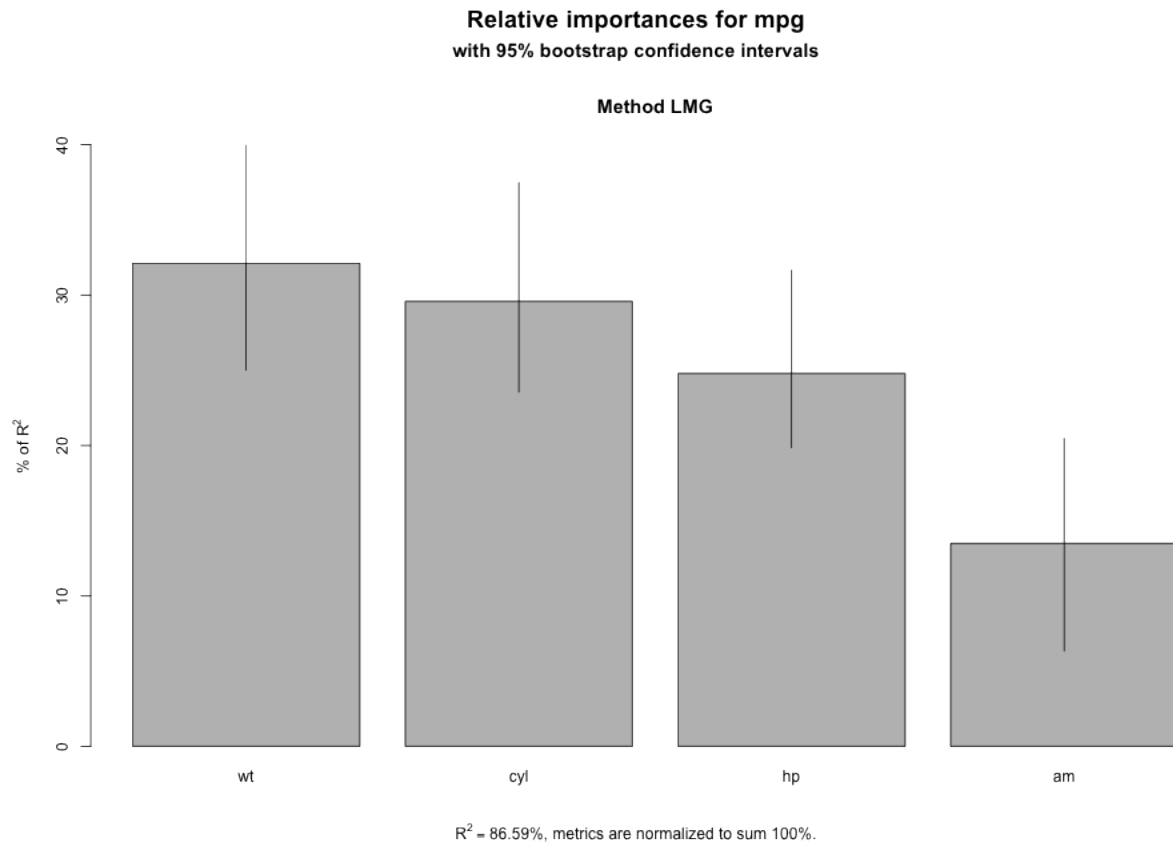
Model with only Transmission Type

The project asks us about two specific points:

1. Is an automatic or manual transmission better for MPG?

2. Quantify the MPG difference between automatic and manual transmissions

We have seen in the explaoratory data analysis section that manual transmissions are generally better for mpg. To quantify the difference, a linear model was fit using only mpg as an explanatory variable. This model produced an `r round(sumfitT$r.squared,3)` which means that it explains `r round(sumfitT$r.squared,3)*100`% of the variation in mpg, much less than the final model arrived at above. The coefficient is significant, and at `r round(fitT$coefficients[2],3)`, which we can interepret as, automatic to manual transmission will increase the mpg by `r round(fitT$coefficients[2],3)`. Sounds like transmission type does have an impact on mpg, however when the previous models are considered, other variables are much more impartant than transmission type in explaining the variation in mpg. An ANOVA between this model and the final model above results in a p-value of less than 0.0000001 which means that we reject the null hypothesis that the models are similar. We must use the more complex model, as it explains a significantly higher proportion of variation in mpg than the simpler model.

Relative Importance of Variables

For our last analysis we will look at Bootstrap Measures of Relative Importance of explanatory variables. This is a nice way to interpret the importance of variables with regard to their contribution to the r^2 of a model. We used the model fit by AIC which includes the variables cylinder, horsepower, weight, and transmission type. The results are shown in the plot below. We can clearly see that am, transmission type, is the least important of these variables contributing only about 14% to the r^2.

Relative importances for mpg
with 95% bootstrap confidence intervals

Method LMG

R^2 = 86.59%, metrics are normalized to sum 100%.

Conclusion

We have determined that there is a difference in mpg in relation to transmission type and have quantified that difference. However, transmission type does not appear to be a very good explanatory variable for mpg; weight, horsepower, and number of cylinders are all more significant variables.

More information are available at: https://github.com/shilmi/Regression-Models/