# Regression Models -

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## **Executive Summary**

Motor Trend, an automobile trend magazine is interested in exploring the relationship between a set of variables and miles per gallon (MPG) outcome. In this project, we will analyze the mtcars dataset from the 1974 Motor Trend US magazine to answer the following questions:

- Is an automatic or manual transmission better for MPG?

  Manual transmissions achieve a higher value of MPG compared to automatic transmission.
- Quantify the MPG difference between automatic and manual transmissions.

  This increase is approximately 1.8 MPG when switching from an automatic transmission to a manual one, with all other variables held constant.

```
library(caret)
```

### **Exploratory Data Analysis**

The data was extracted from the 1974 Motor Trend US magazine, and comprises 11 aspects of automobile design and performance for 32 automobiles (1973-1974 models). For a description of the data set please follow this link.

```
data(mtcars)
mtcarsRAW <- mtcars
mtcars$am <- factor(mtcars$am,labels=c("automatic","manual"))
mtcars$gear <- factor(mtcars$gear)
mtcars$cyl <- factor(mtcars$cyl)
mtcars$carb <- factor(mtcars$carb)
mtcars$vs <- factor(mtcars$vs)</pre>
```

To explore the relationship of the variable under study, mpg, with the other variables in the dataset, we plot mpg against each of the other variables (see appendix, figure 1). We try to find the highly correlated variables (>0.75), so we can excluded them from the model.

```
mtcarsCorrelation=cor(mtcarsRAW)
corIndices = findCorrelation(mtcarsCorrelation, cutoff = 0.75); print(corIndices)
```

```
## [1] 2 3 1 9
```

So the variables cyl, disp, mpg and am are highly correlated. As we are interested in the effect of am on mpg they will be part of the model, we will exclude cyl and disp. We will use the step function for the selection process.

Figure 2 in th appendix shows a first comparison of the effect of the type of transmission (am) on the fuel consumption (mpg),

#### Regression Model

First, we build a base regression model that includes all variables as predictors for mpg. Next we use the step function to refine our linear model, by selecting the most appropriate variables as predictors, and eliminating those that do not.

```
baseModel <- lm(mpg ~ ., data = mtcars)
bestModel <- step(baseModel, direction = "both")</pre>
```

As we can see in the summary below, the algorithm has selected cyl, hp, wt and am as predictors. To check residuals for normality and homoskedasticity, we plot the residuals (see appendix, figure 3). We see that the residuals are normally distributed and homoskedastic.

#### summary(bestModel)

```
##
## Call:
## lm(formula = mpg ~ cyl + hp + wt + am, data = mtcars)
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -3.9387 -1.2560 -0.4013 1.1253
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 33.70832
                          2.60489
                                   12.940 7.73e-13 ***
## cyl6
              -3.03134
                           1.40728
                                   -2.154
                                           0.04068 *
## cyl8
               -2.16368
                           2.28425
                                   -0.947
                                           0.35225
              -0.03211
                           0.01369
                                   -2.345 0.02693 *
## hp
## wt
               -2.49683
                           0.88559
                                   -2.819 0.00908 **
               1.80921
## ammanual
                           1.39630
                                     1.296
                                           0.20646
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.41 on 26 degrees of freedom
## Multiple R-squared: 0.8659, Adjusted R-squared: 0.8401
## F-statistic: 33.57 on 5 and 26 DF, p-value: 1.506e-10
```

#### Conclusions

best Model, has successfully explained more than 84% of the variability of the data. The model coefficients provide the following insights:

- Miles per gallon increases by a factor of 1.8 (1.8) with manual transmission.
- Miles per gallon decreases by a factor of 0.03 (-0.03) as horspower increases.
- Miles per gallon decreases by a factor of 2.5 (-2.5) for every increase of 1000 lb in weight.
- Miles per gallon decreases by a factor of 3.03 (-3.03) for 6 cylinders and by a factor of 2.16 (-2.16) for 8 cylinders.
- The intercept is at 33.7 mpg.

The overall p-value is very small  $(1.506 \times 10^{\circ} - 10)$  which means that there is a very small uncertainty for the sign of the coefficients.

# Appendix

Figure 1 - Relationships of the different variables

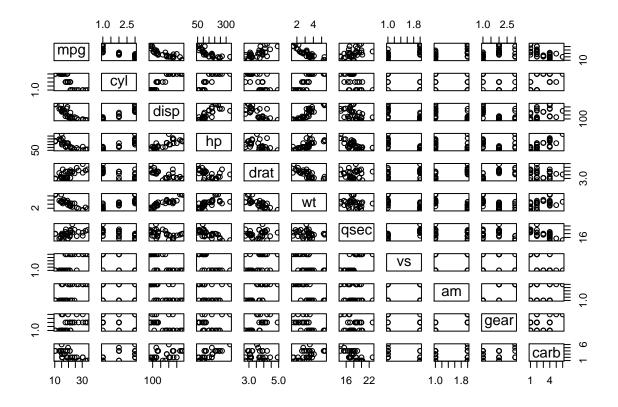


Figure 2 - MPG vs. Transmission Type

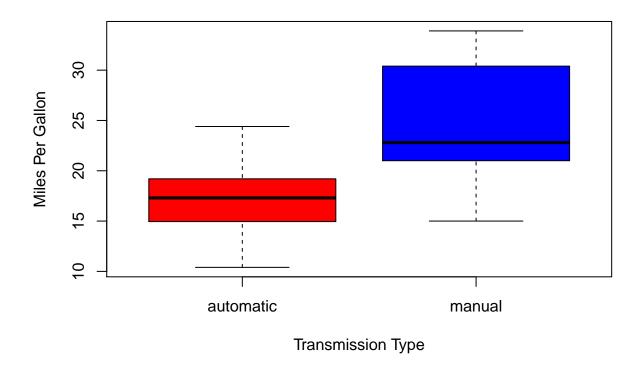


Figure 3 - Residuals

