# Regression Project \_ Motor Trend

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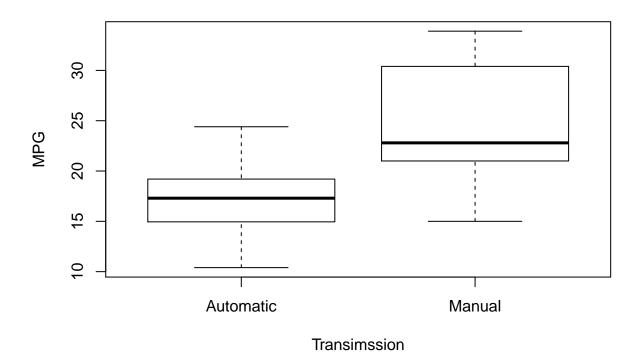
### 1. Synopsis

The work is based on data of Motor Trend, a magazine about the automobile industry. The data set of a collection of cars was reviewed. And the relationship between a set of variables and miles per gallon (MPG) (outcome) was analyzed. Two particular questions were answered: 1) whether an automatic or manual transmission is better for MPG; 2) whether the MPG difference between automatic and manual transmissions can be quantified.

#### 2.Data download and process

```
# Datasets
library(datasets)
data(mtcars)
# View of the data
head(mtcars)
                      mpg cyl disp hp drat
                                               wt qsec vs am gear carb
## Mazda RX4
                            6 160 110 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 1
                                                                       1
## Hornet 4 Drive
                     21.4 6 258 110 3.08 3.215 19.44
                                                          1
                                                                       1
## Hornet Sportabout 18.7 8 360 175 3.15 3.440 17.02 0
                                                                  3
                                                                       2
## Valiant
                     18.1
                            6 225 105 2.76 3.460 20.22
                                                                  3
                                                                       1
library(dplyr)
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
mtcars<-mutate(mtcars,Transmission=am)</pre>
mtcars$Transmission[mtcars$Transmission==1]<-'Manual'</pre>
mtcars$Transmission[mtcars$Transmission==0]<-'Automatic'</pre>
mtcars$Transmission<-factor(paste(mtcars$Transmission))</pre>
# Plot the mpg of manual vs automatic
boxplot(mtcars$mpg ~ mtcars$Transmission, data=mtcars, outpch = 19, ylab="MPG", xlab="Transimssion", ma
```

## **MPG vs Transmission**



From visuallization, the mpg of manual transmission is higher than that of automatic transmission. Now let's evaluate the significance.

```
auto<-mtcars[mtcars$Transmission=='Automatic',]
manual<-mtcars[mtcars$Transmission=='Manual',]
t.test(auto$mpg,manual$mpg)</pre>
```

```
##
## Welch Two Sample t-test
##
## data: auto$mpg and manual$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
## 17.14737 24.39231
```

The t value is negative and the confidence interval is absolutely below zero, which means the hypothesis that the automatic and manual transmission are the same (t=0) is rejected. The difference in the mpg of automatic and manual transmission is significant, and the mpg of automatic is lower than manual.

#### Regression Model

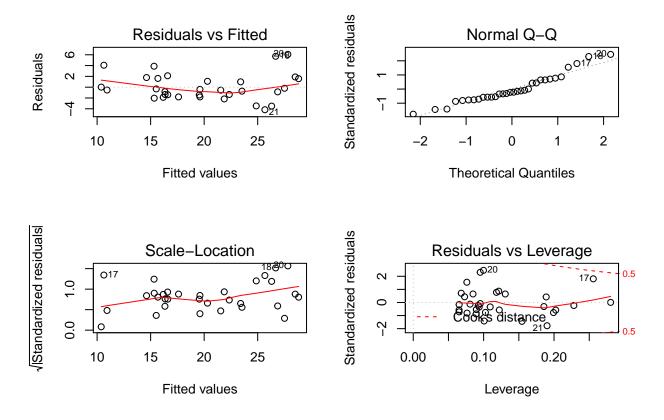
To quantify the difference between the automatic and manual transmission. The regression model is applied to evaluate the correlation of all other variables vs. transmission and mpg relationship.

```
fit1<-lm(mpg~am,mtcars)</pre>
summary(fit1)
##
## Call:
## 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|)
##
                 17.147
                             1.125 15.247 1.13e-15 ***
## (Intercept)
                  7.245
## am
                             1.764
                                    4.106 0.000285 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 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
The p-value is low but the R value is only 0.3385, which means other variables may influence the mpg. Now
let's consider other variables and perform the multivariable linear regression.
fit3<-lm(mpg~am+cyl+wt,mtcars)
summary(fit3)
##
## Call:
## lm(formula = mpg ~ am + cyl + wt, data = mtcars)
##
## Residuals:
##
                1Q Median
## -4.1735 -1.5340 -0.5386 1.5864 6.0812
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
                            2.6415 14.923 7.42e-15 ***
## (Intercept) 39.4179
## am
                 0.1765
                            1.3045
                                     0.135 0.89334
                            0.4223 -3.576 0.00129 **
                -1.5102
## cyl
                -3.1251
                            0.9109 -3.431 0.00189 **
## wt
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.612 on 28 degrees of freedom
## Multiple R-squared: 0.8303, Adjusted R-squared: 0.8122
## F-statistic: 45.68 on 3 and 28 DF, p-value: 6.51e-11
```

#### Residual and Diagnostics

Multivariable regression model residuals

```
par(mfrow=c(2,2))
plot(fit3)
```



#### par(mfrow=c(1,1))

From the above plots, we can make the following observations:

The residuals appear to be randomly scattered on the plot and verify the independence condition.

The points in Q-Q plot mostly fall on the line which indicates the normally distributed residuals.

The Scale-Location plot consists of points scattered in a constant band pattern, indicating constant variance.

The outliers or leverage points are limited and acceptible.