# RegressionProject

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##Overview You work for Motor Trend, a magazine about the automobile industry. Looking at a data set of a collection of cars, they are interested in exploring the relationship between a set of variables and miles per gallon (MPG) (outcome). They 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" ##Dataset

```
library(datasets)
data(mtcars)
head(mtcars)
```

```
##
                      mpg cyl disp hp drat
                                               wt qsec vs am gear
## Mazda RX4
                     21.0
                            6 160 110 3.90 2.620 16.46
                                                         0
## Mazda RX4 Wag
                            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
                                                                      1
                                                                 3
                                                                      2
## Hornet Sportabout 18.7
                            8
                               360 175 3.15 3.440 17.02
                                                         0
## Valiant
                     18.1
                            6 225 105 2.76 3.460 20.22
                                                                      1
```

##Processing Data

```
mtcars$cyl <- as.factor(mtcars$cyl)
mtcars$vs <- as.factor(mtcars$vs)
mtcars$am <- as.factor(mtcars$am)
str(mtcars)</pre>
```

```
## 'data.frame':     32 obs. of 11 variables:
## $ mpg : num     21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
## $ cyl : Factor w/ 3 levels "4","6","8": 2 2 1 2 3 2 3 1 1 2 ...
## $ disp: num    160 160 108 258 360 ...
## $ hp : num    110 110 93 110 175 105 245 62 95 123 ...
## $ drat: num     3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
## $ wt : num     2.62 2.88 2.32 3.21 3.44 ...
## $ qsec: num    16.5 17 18.6 19.4 17 ...
## $ vs : Factor w/ 2 levels "0","1": 1 1 2 2 1 2 1 2 2 2 ...
## $ am : Factor w/ 2 levels "0","1": 2 2 2 1 1 1 1 1 1 1 ...
## $ gear: num    4 4 4 3 3 3 3 3 4 4 4 ...
## $ carb: num    4 4 1 1 2 1 4 2 2 4 ...
```

##Exploratory analysis with Appendix: Figures The mean of the variant cars (by transmission types) for their MPG:

```
mtcars$am<-as.factor(mtcars$am)
levels(mtcars$am)<-c("Auto", "Man.")
aggregate(mpg ~ am, data = mtcars, mean)</pre>
```

```
## am mpg
## 1 Auto 17.14737
## 2 Man. 24.39231
```

We can see from this simple view that the mean average of MPG's for Automatic cars outstrips that of Manual.

We should see if this variance is significant - via running a t-test:

```
Auto.Data <- mtcars[mtcars$am == "Auto",]
Manu.Data <- mtcars[mtcars$am == "Man.",]

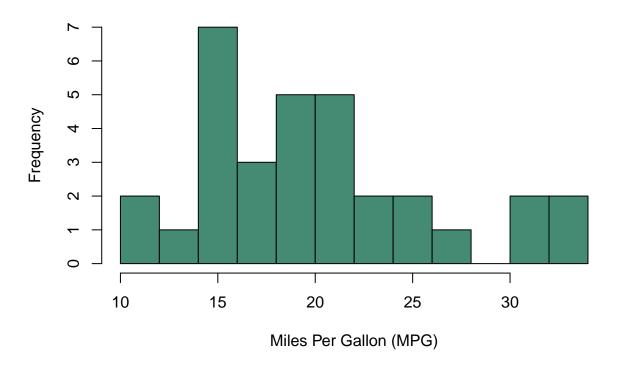
t.test(Auto.Data$mpg, Manu.Data$mpg)</pre>
```

```
##
## Welch Two Sample t-test
##
## data: Auto.Data$mpg and Manu.Data$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
```



## Histgram of MPG
hist(mtcars\$mpg, breaks=12, xlab="Miles Per Gallon (MPG)", main="MPG Distribution", col="aquamarine4")

### **MPG Distribution**



##Regression modelling

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

```
##
## 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
                                   15.247 1.13e-15 ***
## (Intercept)
                             1.125
## amMan.
                  7.245
                             1.764
                                     4.106 0.000285 ***
## ---
## 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
```

Since Adjusted  $R^2 = 0.3385$ , thus only 33.85% of the regression variance of this model is explained.

```
multiVariableRegressionModel <- lm(mpg ~ am + wt + hp, data = mtcars)
summary(multiVariableRegressionModel)</pre>
```

```
##
## Call:
## lm(formula = mpg ~ am + wt + hp, data = mtcars)
## Residuals:
      Min
              1Q Median
                             3Q
                                   Max
## -3.4221 -1.7924 -0.3788 1.2249 5.5317
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 34.002875 2.642659 12.867 2.82e-13 ***
## amMan.
             2.083710 1.376420 1.514 0.141268
## wt
             -2.878575 0.904971 -3.181 0.003574 **
             ## hp
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.538 on 28 degrees of freedom
## Multiple R-squared: 0.8399, Adjusted R-squared: 0.8227
## F-statistic: 48.96 on 3 and 28 DF, p-value: 2.908e-11
```

Compared to single variable regression model, the Adjusted  $R^2 = 0.823$ , thus approximately 82.3% of the regression variance of this model is explained.

#### anova(singleVariableRegressionModel, multiVariableRegressionModel)

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + wt + hp
## Res.Df RSS Df Sum of Sq F Pr(>F)
## 1 30 720.90
## 2 28 180.29 2 540.61 41.979 3.745e-09 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Since the p-value is 3.745e-09, we can now reject the H0 and state that our Multivariable Model is considerably different than the Single Linear Regression Model.

### t.test(mtcars\$mpg~mtcars\$am)

```
##
## Welch Two Sample t-test
##
## data: mtcars$mpg by mtcars$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 Auto mean in group Man.
## 17.14737 24.39231
```

With assumption that all other conditions remain unchanged. Since p-value = 0.001374 which is less than 0.05, we conclude that manual transmission is better than automatic transmission for MPG and reject the null hypothesis that there is no difference in MPG.

##Executive Summary The key finding of this analysis is that manual transmission on average has a better miles per gallon (mpg) than its counterpart i.e. automatic transmission.