

Project

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Regression Models - Course Project

Question

Take the mtcars data set and write up an analysis to answer the below two questions using regression models and exploratory data analyses.

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

Loading mtcars data

```
data(mtcars)
```

Convert required mtcars variables into factors

```
mtcars$cyl <- factor(mtcars$cyl)

mtcars$vs <- factor(mtcars$vs)

mtcars$am <- factor(mtcars$am, labels = c("Automatic", "Manual"))

mtcars$gear <- factor(mtcars$gear)

mtcars$carb <- factor(mtcars$carb)
```

Just analyzing data based on question - To check the impact on mpg with transmission mode, we can plot a basic box-plot as in plot 1 given in the Appendix, we can observe that there is an impact on MPG with automatic transmission which is having a lower MPG.

Model building using Regression

Mathematically check the impact on MPG with transmission type

```
aggregate(mpg ~ am, data = mtcars, mean)
```

```
##           am      mpg
## 1 Automatic 17.14737
## 2   Manual  24.39231
```

Checking the difference in MPG for automatic and manual transmission type

```
aggregate(mpg ~ am, data = mtcars, mean)[2,2] - aggregate(mpg ~ am, data = mtcars,
mean)[1,2]
```

```
## [1] 7.244939
```

So we can hypothesize that automatic cars have MPG 7.2449 lower than manual cars.

Test using t-test to check whether this difference is significant

```
T_automatic <- mtcars[mtcars$am=="Automatic",]

T_manual <- mtcars[mtcars$am=="Manual",]

t.test(T_automatic$mpg,T_manual$mpg)
```

```
##
## Welch Two Sample t-test
##
## data: T_automatic$mpg and T_manual$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 p-value obtained is 0.001374. So this is a significant difference.

Build a model using linear regression

```
first_model <- lm(mpg ~ am, data = mtcars)

summary(first_model)
```

```
##
## 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|)
## (Intercept)   17.147      1.125   15.247 1.13e-15 ***
## amManual       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
```

Summary shows that the average MPG for automatic is 17.15 MPG and manual is 7.2 MPG higher. The R^2 value is 0.36 means this model explains only 36 % of the variance. So we have to build a multivariate linear regression.

We can include other variables to make it more accurate. Selection of the new variables to be included is based on the pairs plot which shows the variable correlation with mpg. The pairs plot is given as plot - 3 in the Appendix.

From the pairs plot we can observe that cyl, disp, hp, wt have the strongest correlation with mpg. So we can build a new model based on multivariate linear regression and compare it to the initial model.

Build a second model and use anova function to compare it to first model

```
second_model <- lm(mpg ~ am + cyl + disp + hp + wt, data = mtcars)

anova(first_model, second_model)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + cyl + disp + hp + wt
##   Res.Df    RSS Df Sum of Sq    F    Pr(>F)
## 1      30 720.90
## 2      25 150.41  5    570.49 18.965 8.637e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

The new p-value is 8.637e-08. Which is far better than the initial model

We check again the residuals for non-normality and can observe that all normally distributed and homoskedastic. Please refer plot -3 given in the Appendix.

Check the parameters of second model

```
summary(second_model)
```

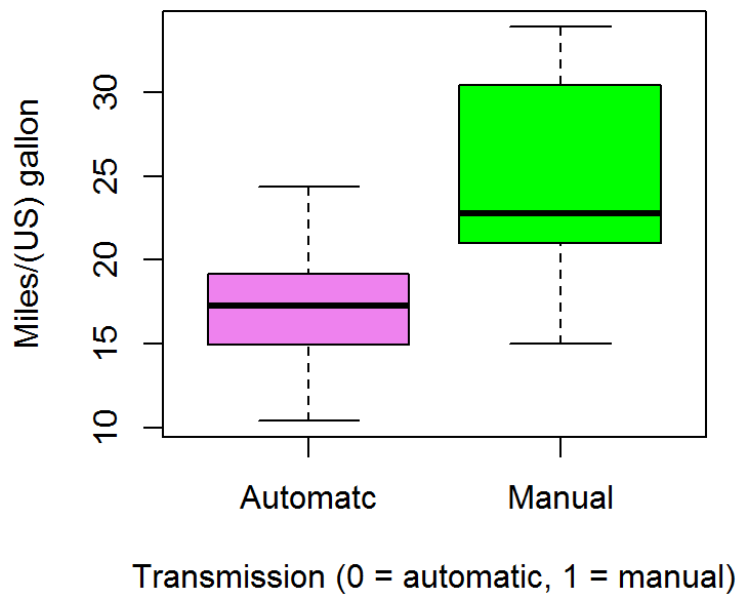
```
##
## Call:
## lm(formula = mpg ~ am + cyl + disp + hp + wt, data = mtcars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.9374 -1.3347 -0.3903  1.1910  5.0757
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 33.864276   2.695416  12.564 2.67e-12 ***
## amManual     1.806099   1.421079   1.271  0.2155
## cyl6        -3.136067   1.469090  -2.135  0.0428 *
## cyl8        -2.717781   2.898149  -0.938  0.3573
## disp         0.004088   0.012767   0.320  0.7515
## hp          -0.032480   0.013983  -2.323  0.0286 *
## wt          -2.738695   1.175978  -2.329  0.0282 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.453 on 25 degrees of freedom
## Multiple R-squared:  0.8664, Adjusted R-squared:  0.8344
## F-statistic: 27.03 on 6 and 25 DF,  p-value: 8.861e-10
```

So it is clear from the summary that model explains 86.64 % of the variance and as a result variables cyl, hp, disp, wt affected the correlation between mpg and am.

So we can say that the difference between automatic and manual transmissions is 1.81 MPG. So manual transmission is better in terms of MPG.

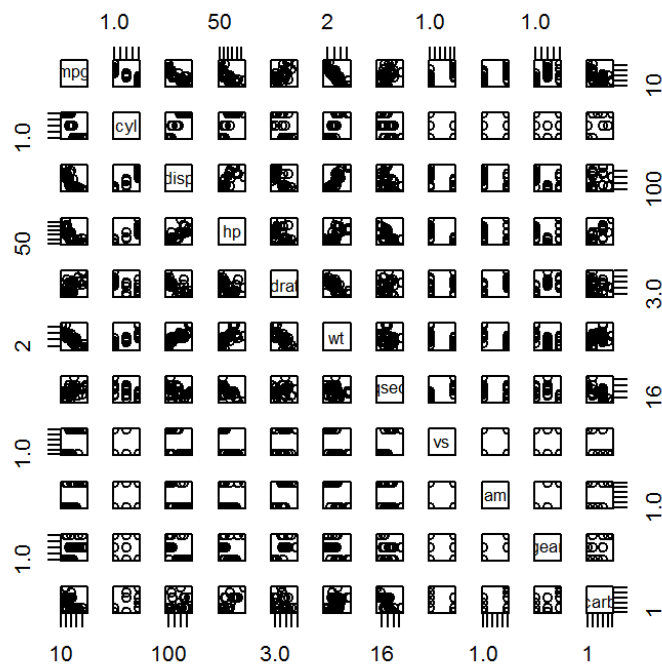
Appendix : Plot 1 Box-plot

```
boxplot(mpg ~ am, data = mtcars, col = (c("violet", "green")), ylab = "Miles/(US) gallon", xlab = "Transmission (0 = automatic, 1 = manual)")
```



Plot 2 - Pairs plot

```
pairs(mpg ~ ., data = mtcars)
```



Plot 3 - Residuals

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

```
plot(second_model)
```

