

# Transmission analysis on MPG

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## Executive Summary

This report answers the following two questions for Motor Trend:

1. Is an automatic or manual transmission better for miles per gallon (MPG)?
2. Quantify the MPG difference between automatic and manual transmissions

They key results from the analysis are:

- MPG of manual transmission cars is higher (by a factor 1.8) compared to automatic transmission cars.
- There is a significant difference between the mean MPG for automatic and manual transmission cars.

## Data processing

```
library(datasets)
data(mtcars)
str(mtcars)
```

Coerce the “cyl”, “vs”, “gear”, “carb” and “am” variables into factor variables.

```
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'))
```

## Exploratory Analysis

Boxplotting (Appendix - Figure 1) the distribution of MPG of am, it shows that manual transmission have a higher MPG than automatic transmissions. Exploring the relationships of the different variables in the dataset (Appendix - Figure 2), it shows that cyl, disp, hp, drat, wt, vs and am are correlated.

```
cor(datasets::mtcars)[ "mpg", ]
```

## Hypothesis Testing (Statistical Inference)

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

Manual transmissions mean is 7.24 MPGs higher than automatic transmissions. To determine significance a t-test is performed.

```
autoData <- mtcars[mtcars$am == "Automatic", ]
manualData <- mtcars[mtcars$am == "Manual", ]
t.test(autoData$mpg, manualData$mpg)
```

Null hypothesis is rejected as the p-value is 0.00137. There is significance difference in the mean MPG between automatic transmissions and manual transmissions.

# Regression Analysis

In this section a linear regression models based on the different variables are built.

## Model building and selection

Initially we include all variables of MPG. By performing a stepwise model selection we repeatedly select the most significant predictors and build the best model.

```
initialModel <- lm(mpg ~ ., data = mtcars)
bestModel <- step(initialModel, direction = "both")
summary(bestModel)
```

The adjusted R-squared value of 0.84 which means that 84% of the variability in the model is explained.

Next the base model will be compared with only am as the predictor variable with the best model.

```
baseModel <- lm(mpg ~ am, data = mtcars)
anova(baseModel, bestModel)
```

The p-value obtained is significant and we reject the null hypothesis.

## Residuals and Diagnostics

From the plots in Figure 3 (Appendix) there are following observations:

- Randomly scatterend points verify the indepedence condition.
- The Normal Q-Q plot indicates that the residuals are normally distributed.
- Scale-Locaton plot indicates constant variance.

Computation of some regression diagnostics will be performed to find out these leverage points.

```
leverage <- hatvalues(bestModel)
tail(sort(leverage), 3)
```

```
influential <- dfbetas(bestModel)
tail(sort(influential[, 6]), 3)
```

The above results show that the anayslsi was correct, as the same cars appear in the residual plots.

## Results

Based on the observations from our best fit model, we can conclude the following,

- Manual transmission cards get more MPG compared to automatic transmissions.
- MPG will decrease by 2.5 for every 1000 lb increase in weight (wt).
- MPG decreases with increase of horse power (hp).

## Appendix

Figure 1 - Boxplot | MPG by transmission type

```
plot(mpg ~ am, data = mtcars, main = "MPG by transmission type", xlab = "Transmission type", ylab = "MPG")
```

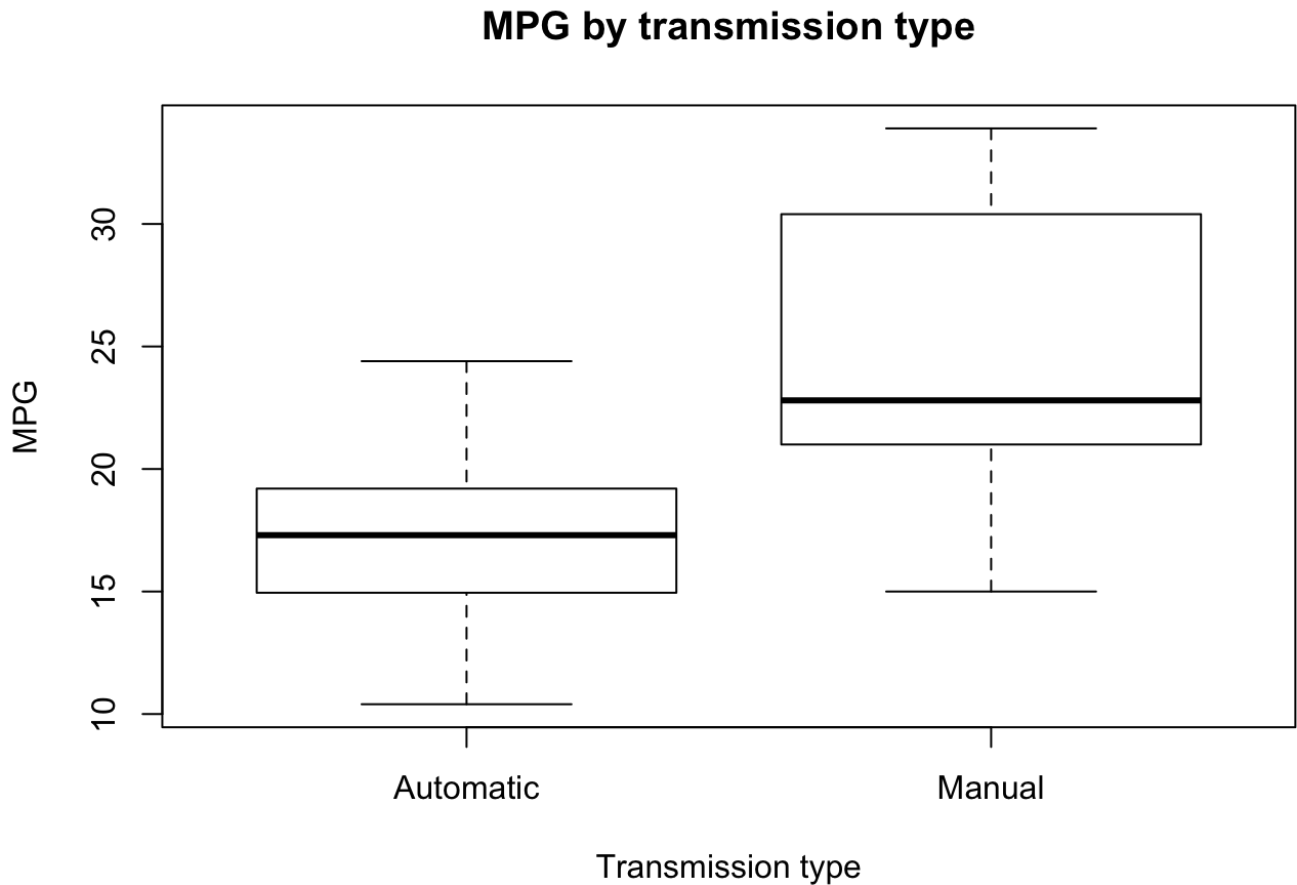


Figure 2 - Pairs | MPG by transmission type

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

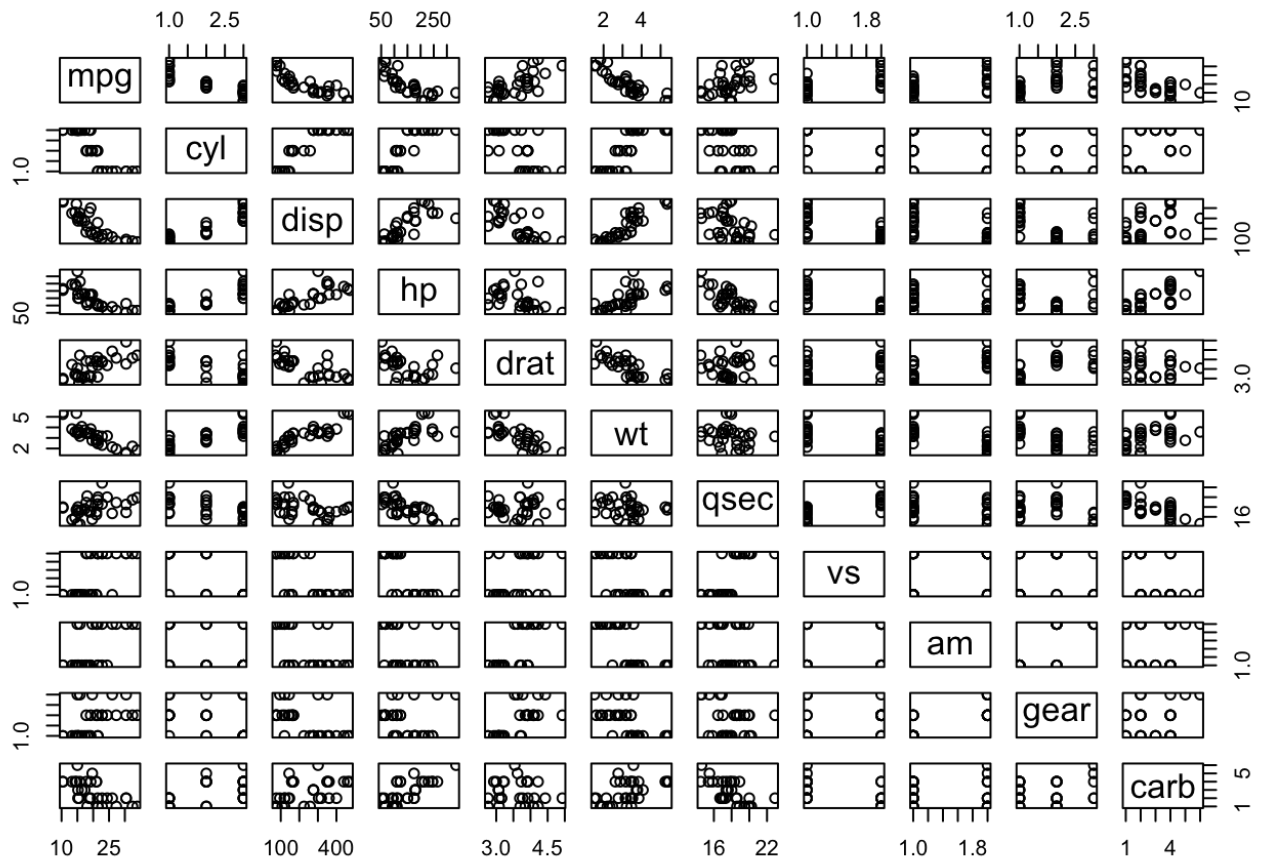


Figure 3 - Residuals & Diagnostics

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

