

# Regression Analysis

*Michael Bristow*

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## R Markdown

Regression Models - Peer Assessment 1

## Issue/Problem

Take the mtcars data set and write up an analysis to answer their question using regression models and exploratory data analyses.

Your report must be:

Written as a PDF printout of a compiled (using knitr) R markdown document.

Brief. Roughly the equivalent of 2 pages or less for the main text.

Supporting figures in an appendix can be included up to 5 total pages including the 2 for the main report. The appendix can only include figures. Include a first paragraph executive summary.

## Executive Summary

This report looks at historic data and tries to find drivers for MPG. It will look at two main items

- Is an automatic or manual transmission better for MPG
- Quantify the MPG difference between automatic and manual transmissions

## Data Processing/Analysis

First load the data

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 3.4.3
```

```
data(mtcars)
xx <- mtcars
#convert "am" to a factor
mtcars$am <- as.factor(mtcars$am)
levels(mtcars$am) <- c("AT", "MT")

summary(mtcars)
```

```
##      mpg      cyl      disp      hp
##  Min.   :10.40   Min.    :4.000   Min.    : 71.1   Min.    : 52.0
##  1st Qu.:15.43   1st Qu.:4.000   1st Qu.:120.8   1st Qu.: 96.5
##  Median :19.20   Median :6.000   Median :196.3   Median :123.0
```

```
## Mean :20.09 Mean :6.188 Mean :230.7 Mean :146.7
## 3rd Qu.:22.80 3rd Qu.:8.000 3rd Qu.:326.0 3rd Qu.:180.0
## Max. :33.90 Max. :8.000 Max. :472.0 Max. :335.0
## drat wt qsec vs am
## Min. :2.760 Min. :1.513 Min. :14.50 Min. :0.0000 AT:19
## 1st Qu.:3.080 1st Qu.:2.581 1st Qu.:16.89 1st Qu.:0.0000 MT:13
## Median :3.695 Median :3.325 Median :17.71 Median :0.0000
## Mean :3.597 Mean :3.217 Mean :17.85 Mean :0.4375
## 3rd Qu.:3.920 3rd Qu.:3.610 3rd Qu.:18.90 3rd Qu.:1.0000
## Max. :4.930 Max. :5.424 Max. :22.90 Max. :1.0000
## gear carb
## Min. :3.000 Min. :1.000
## 1st Qu.:3.000 1st Qu.:2.000
## Median :4.000 Median :2.000
## Mean :3.688 Mean :2.812
## 3rd Qu.:4.000 3rd Qu.:4.000
## Max. :5.000 Max. :8.000
```

```
head(mtcars)
```

```
## mpg cyl disp hp drat wt qsec vs am gear carb
## Mazda RX4 21.0 6 160 110 3.90 2.620 16.46 0 MT 4 4
## Mazda RX4 Wag 21.0 6 160 110 3.90 2.875 17.02 0 MT 4 4
## Datsun 710 22.8 4 108 93 3.85 2.320 18.61 1 MT 4 1
## Hornet 4 Drive 21.4 6 258 110 3.08 3.215 19.44 1 AT 3 1
## Hornet Sportabout 18.7 8 360 175 3.15 3.440 17.02 0 AT 3 2
## Valiant 18.1 6 225 105 2.76 3.460 20.22 1 AT 3 1
```

Apply regsubsets to find best variable selection

```
library(leaps)
```

```
## Warning: package 'leaps' was built under R version 3.4.3
```

```
reg.best<- regsubsets(mpg~.,mtcars,nvmax=5)
summary(reg.best)
```

```
## Subset selection object
## Call: regsubsets.formula(mpg ~ ., mtcars, nvmax = 5)
## 10 Variables (and intercept)
## Forced in Forced out
## cyl FALSE FALSE
## disp FALSE FALSE
## hp FALSE FALSE
## drat FALSE FALSE
## wt FALSE FALSE
## qsec FALSE FALSE
## vs FALSE FALSE
## amMT FALSE FALSE
## gear FALSE FALSE
## carb FALSE FALSE
## 1 subsets of each size up to 5
## Selection Algorithm: exhaustive
## cyl disp hp drat wt qsec vs amMT gear carb
## 1 ( 1 ) " " " " " " " " "*" " " " " " " " "
## 2 ( 1 ) "*" " " " " " " "*" " " " " " " " "
## 3 ( 1 ) " " " " " " " " "*" "*" " " "*" " " " "
```

```
## 4 (1) " " " " " " " " " " " " " " " "
## 5 (1) " " " " " " " " " " " " " " " "
```

```
reg.summary<- summary(reg.best)
names(reg.summary)
```

```
## [1] "which" "rsq" "rss" "adjr2" "cp" "bic" "outmat" "obj"
```

Look at the best r squared

```
reg.summary$rsq
```

```
## [1] 0.7528328 0.8302274 0.8496636 0.8578510 0.8637377
```

As can be seen, “weight” is the biggest “driver” of mpg followed by number of cylinders.

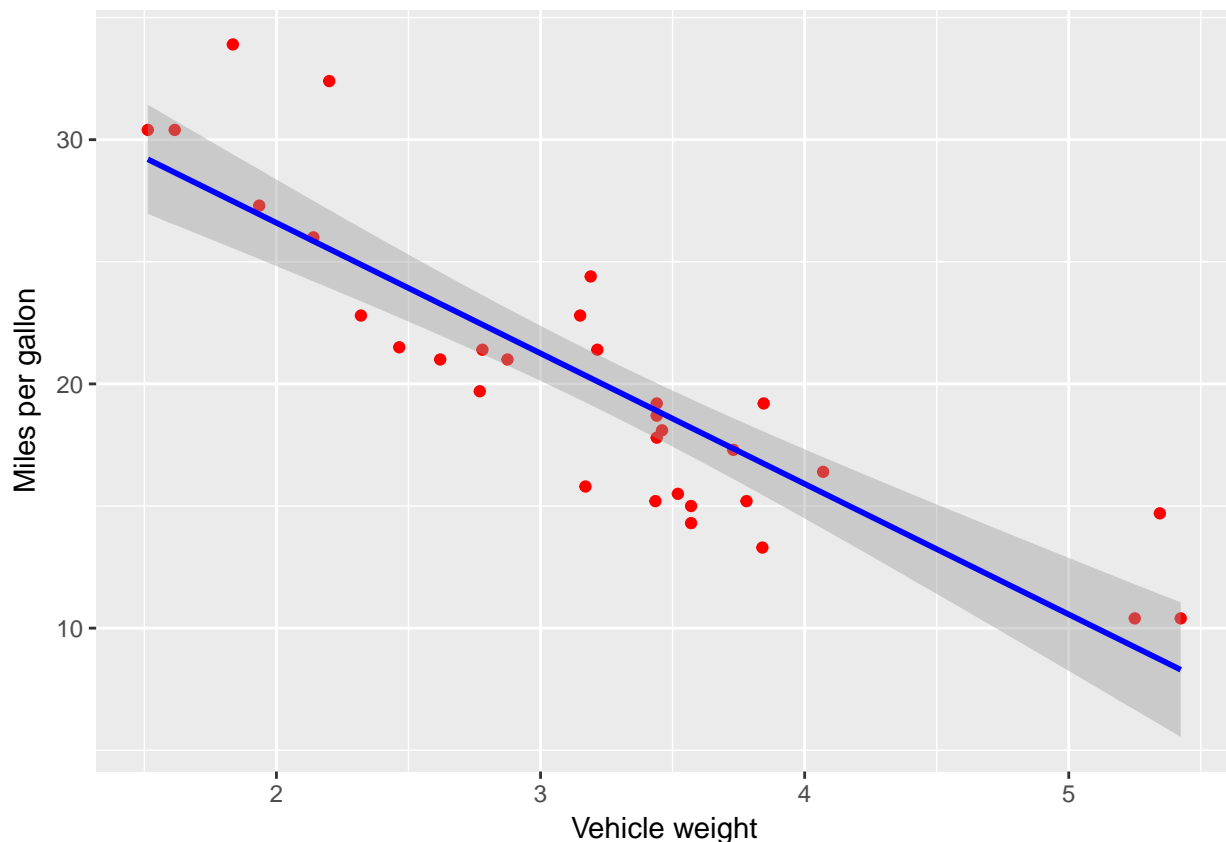
Our model should include

- weight
- cylinders

By adding more variables, the r-squared only increases by small amounts and we need to be wary of overfitting

Plot weight against mpg and show regression line

```
ggplot(data = xx, aes(x = wt, y = mpg)) +
  geom_point(color='red') +
  geom_smooth(method = "lm", color = "blue") +
  labs(x = "Vehicle weight") +
  labs(y = "Miles per gallon")
```



As can be seen there is a direct correlation between the vehicle weight and it mpg

```
wtdata <- lm(mpg~wt+cyl, data = mtcars)
summary(wtdata)
```

```
##
## Call:
## lm(formula = mpg ~ wt + cyl, data = mtcars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.2893 -1.5512 -0.4684  1.5743  6.1004
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  39.6863     1.7150  23.141 < 2e-16 ***
## wt          -3.1910     0.7569  -4.216 0.000222 ***
## cyl         -1.5078     0.4147  -3.636 0.001064 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.568 on 29 degrees of freedom
## Multiple R-squared:  0.8302, Adjusted R-squared:  0.8185
## F-statistic: 70.91 on 2 and 29 DF,  p-value: 6.809e-12
```

We can see that each extra ton in weight changes MPG by -3.1909721 when we use the full model using weight and cylinders

## Effect of Transmission type on MPG

Question is, which has better MPG - Manual or Automatic We do this by running a regression using ONLY transmission type

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

```
##
## 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 ***
## amMT         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
```

It shows that on average

- a car has 17.147 mpg with automatic transmission, and if it is manual transmission, 7.245 mpg is increased.
- has the Residual standard error as 4.902 on 30 degrees of freedom.
- the Adjusted R-squared value is 0.3385, which means that the model can explain about 34% of the variance of the MPG variable.
- The low Adjusted R-squared value also indicates that we need to add other variables to the model.

## Conclusion

Here are the conclusions from the analysis

- The primary driver of MPG is the cars weight
- The “optimal” model is to include weight and cylinders
- By adding more variables, the r-squared value only increases marginally and we are in danger of overfitting the model
- Transmission type also has an effect on MPG
- Having a manual transmission increases MPG by 7.245