

# Executive Summary

The purpose of this report is to analyze `mtcars` data set to explore the relationship between a set of variables and miles per gallon (MPG). The data comes from the 1974 *Motor Trend* US magazine, and contains fuel consumption and 10 aspects of automobile design and performance for 32 automobiles (1973-74 models). I will use regression models and exploratory data analysis to explore if *MPG* is impacted by automatic or manual transmission feature.

Performing t-test shows there is performance difference between cars with automatic and manual transmission of approximately 7 MPG more for cars with manual transmission than those with automatic transmission. Several linear regression models are fitted and I selected the one with highest Adjusted R-squared value. If weight and 1/4 mile time are held constant, manual transmitted cars are  $14.079 + (-4.141) \times \text{weight}$  more MPG (miles per gallon) on average better than automatic transmitted cars. The summary of this report is that lighter cars with a manual transmission and heavier cars with an automatic transmission will have higher MPG values.

## Exploratory Data Analysis

Load the data set `mtcars` and change some variables from `numeric` class to `factor` class.

```
library(ggplot2)
data(mtcars)
mtcars[1:3, ] # Test Data
```

```
##           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  1    4    4
## Mazda RX4 Wag  21.0   6  160  110  3.90  2.875  17.02  0  1    4    4
## Datsun 710     22.8   4  108   93  3.85  2.320  18.61  1  1    4    1
```

```
dim(mtcars)
```

```
## [1] 32 11
```

```
mtcars$cyl <- as.factor(mtcars$cyl)
mtcars$vs <- as.factor(mtcars$vs)
mtcars$am <- factor(mtcars$am)
mtcars$gear <- factor(mtcars$gear)
mtcars$carb <- factor(mtcars$carb)
attach(mtcars)
```

```
## The following objects are masked from mtcars (pos = 3):
##
##      am, carb, cyl, disp, drat, gear, hp, mpg, qsec, vs, wt
##
## The following objects are masked from mtcars (pos = 4):
##
##      am, carb, cyl, disp, drat, gear, hp, mpg, qsec, vs, wt
##
## The following object is masked from package:ggplot2:
##
##      mpg
```

Let's start some xploratory data analysis. Plots are at the end of this report **Plots** . The Box plot tells us thatt manual transmission yields higher values of MPG in general. The pair graph shows some highe correlations between variables like “wt”, “disp”, “cyl” and “hp”.

## Inference

Our null hypothesis is that MPG of the automatic and manual transmissions are from the same population (assuming the MPG has a normal distribution). We use the two sample T-test to test it.

```
result <- t.test(mpg ~ am)
result$p.value
```

```
## [1] 0.001373638
```

```
result$estimate
```

```
## mean in group 0 mean in group 1
##      17.14737      24.39231
```

Since the p-value is 0.00137, we reject our null hypothesis. Therefore the automatic and manual transmissions are from different populations. And the mean for MPG of manual transmitted cars is about 7 more than that of automatic transmitted cars.

## Regression Analysis

We fit the full model as the following.

```
fullModel <- lm(mpg ~ ., data=mtcars)
summary(fullModel) # results hidden
```

This model has the Residual standard error as 2.833 on 15 degrees of freedom. And the Adjusted R-squared value is 0.779, which means that the model can explain about 78% of the variance of the MPG variable. However, none of the coefficients are significant at 0.05 significant level.

Then, we use backward selection to select some statistically significant variables.

```
stepModel <- step(fullModel, k=log(nrow(mtcars)))
summary(stepModel) # results hidden
```

This model is “mpg ~ wt + qsec + am”. It has the Residual standard error as 2.459 on 28 degrees of freedom. And the Adjusted R-squared value is 0.8336, which means that the model can explain about 83% of the variance of the MPG variable. All of the coefficients are significant at 0.05 significant level.

Please refer to the **Plots** section for the plots again. According to the scatter plot, it indicates that there appear to be an interaction term between “wt” variable and “am” variable, since automatic cars tend to weigh heavier than manual cars. Thus, we have the following model including the interaction term:

```
amIntWtModel<-lm(mpg ~ wt + qsec + am + wt:am, data=mtcars)
summary(amIntWtModel) # results hidden
```

This model has the Residual standard error as 2.084 on 27 degrees of freedom. And the Adjusted R-squared value is 0.8804, which means that the model can explain about 88% of the variance of the MPG variable. All of the coefficients are significant at 0.05 significant level. This is a pretty good one.

Next, we fit the simple model with MPG as the outcome variable and Transmission as the predictor variable.

```
amModel<-lm(mpg ~ am, data=mtcars)
summary(amModel) # results hidden
```

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. This model has the Residual standard error as 4.902 on 30 degrees of freedom. And 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.

Finally, we select the final model.

```
anova(amModel, stepModel, fullModel, amIntWtModel)
confint(amIntWtModel) # results hidden
```

We end up selecting the model with the highest Adjusted R-squared value, “mpg ~ wt + qsec + am + wt:am”.

```
summary(amIntWtModel)$coef
```

##	Estimate	Std. Error	t value	Pr(> t )
## (Intercept)	9.723053	5.8990407	1.648243	0.1108925394
## wt	-2.936531	0.6660253	-4.409038	0.0001488947
## qsec	1.016974	0.2520152	4.035366	0.0004030165
## am1	14.079428	3.4352512	4.098515	0.0003408693
## wt:am1	-4.141376	1.1968119	-3.460340	0.0018085763

Thus, the result shows that when “wt” (weight lb/1000) and “qsec” (1/4 mile time) remain constant, cars with manual transmission add 14.079 + (-4.141)\*wt more MPG (miles per gallon) on average than cars with automatic transmission. That is, a manual transmitted car that weighs 2000 lbs have 5.797 more MPG than an automatic transmitted car that has both the same weight and 1/4 mile time.

# Residual Analysis and Diagnostics

Please refer to the **Plots** section for the plots. According to the residual plots, we can verify the following underlying assumptions:

1. The Residuals vs. Fitted plot shows no consistent pattern, supporting the accuracy of the independence assumption.
2. The Normal Q-Q plot indicates that the residuals are normally distributed because the points lie closely to the line.
3. The Scale-Location plot confirms the constant variance assumption, as the points are randomly distributed.
4. The Residuals vs. Leverage argues that no outliers are present, as all values fall well within the 0.5 bands.

As for the Dfbetas, the measure of how much an observation has effected the estimate of a regression coefficient, we get the following result:

```
sum(abs(dfbetas(amIntWtModel)))>1)
```

```
## [1] 0
```

Therefore, the above analysis meet all basic assumptions of linear regression and we'll answer the questions.

## Plots

### 1. Boxplot of MPG vs. Transmission

```
boxplot(mpg ~ am, xlab="Transmission (0 = Automatic, 1 = Manual)", ylab="MPG",  
        main="Boxplot of MPG vs. Transmission")
```

## Data

mtcars 32 obs. of 11 variables

## values

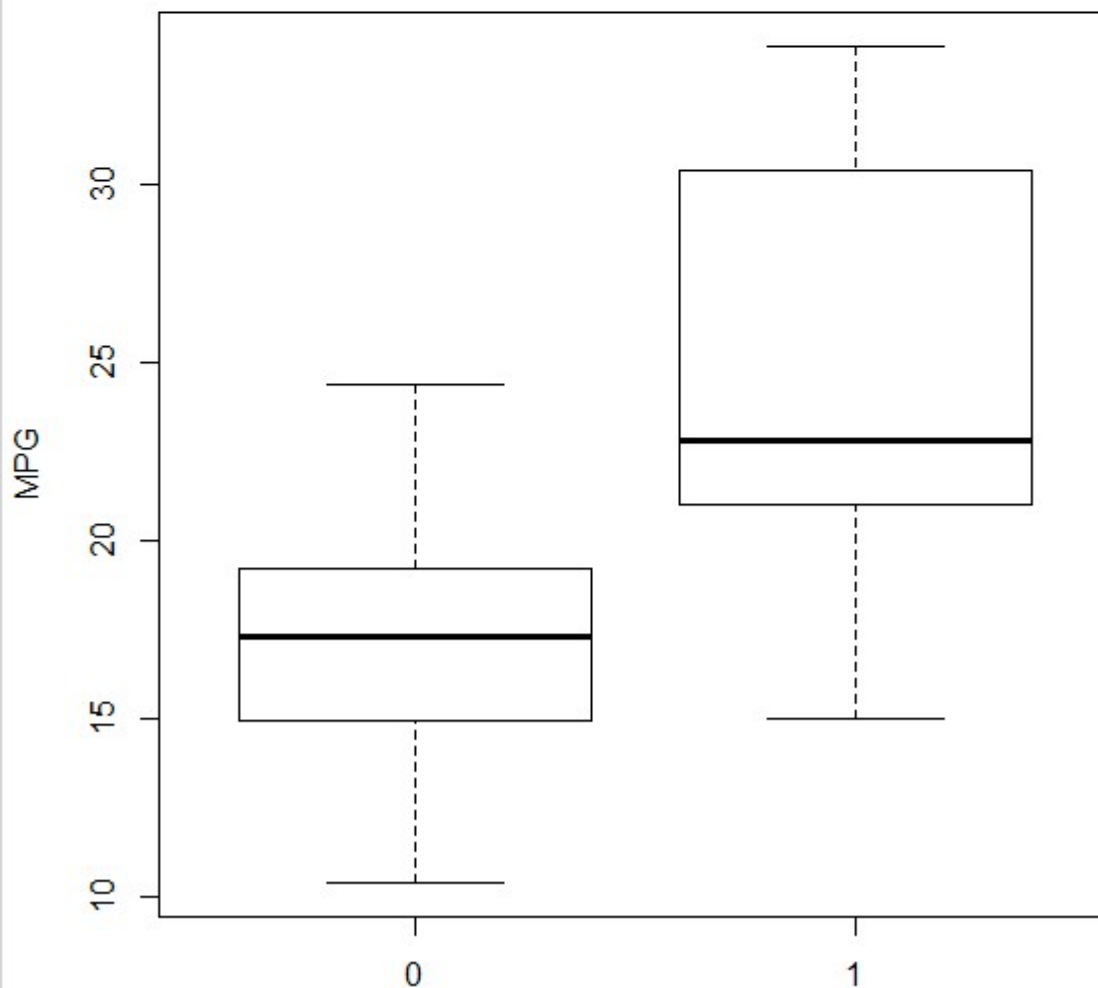
amIntwtModel	List of 13
amModel	List of 13
books	List of 3
fullModel	List of 13
getTermMatrix	function (book)
result	List of 9
stepModel	List of 14

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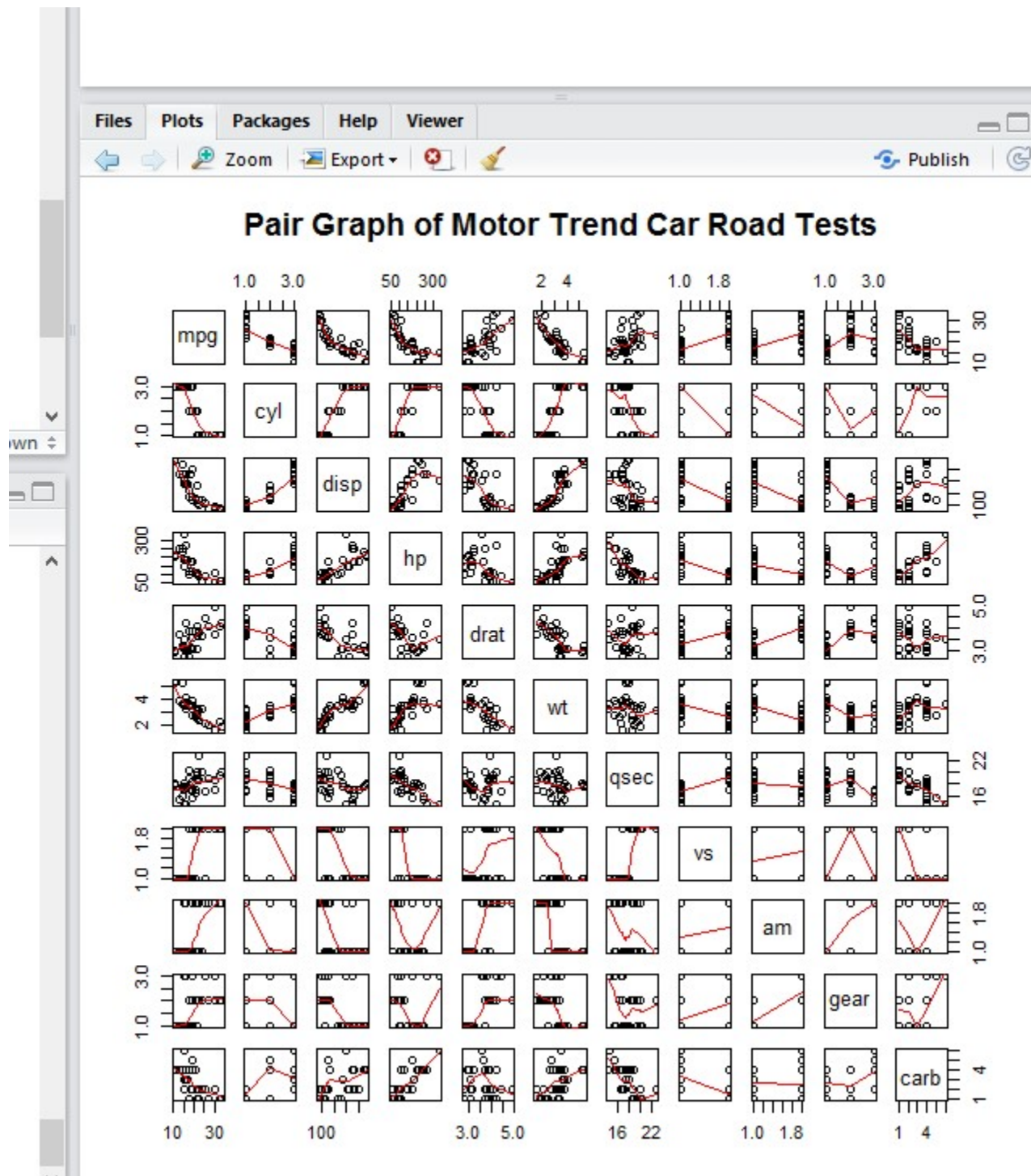
Publish

### Boxplot of MPG vs. Transmission



## 2. Pair Graph of Motor Trend Car Road Tests

```
pairs(mtcars, panel=panel.smooth, main="Pair Graph of Motor Trend Car Road Tests")
```



## 3. Scatter Plot of MPG vs. Weight by Transmission

```
ggplot(mtcars, aes(x=wt, y=mpg, group=am, color=am, height=3, width=3)) + geom_point(  
) +  
scale_colour_discrete(labels=c("Automatic", "Manual")) +  
xlab("weight") + ggtitle("Scatter Plot of MPG vs. Weight by Transmission")
```



#### 4. Residual Plots

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

