Regression Models Course Project

Executive Summary

Here we used the mtcars dataset which contains the set of variables with related to miles per gallon (MPG). They are particularly interested in the following two questions:

- "Is automatic or manual transmission better for MPG"
- "Quantify this difference"

Data Analysis

I analysis the dataset in dffrent statistical method.

Load and Test Data

```
library(datasets)
data(mtcars)
mpgData <- with(mtcars, data.frame(mpg, am))
mpgData$am <- factor(mpgData$am, labels = c("Automatic", "Manual"))
manual <- as.list(subset(mtcars, am == 1, select = mpg))[[1]]
automatic <- as.list(subset(mtcars, am == 0, select = mpg))[[1]]</pre>
```

Basic Analysis

```
summary(mpgData[mpgData$am == "Automatic",])
##
         mpg
##
   Min.
          :10.4
                   Automatic:19
##
   1st Qu.:14.9
                   Manual
   Median:17.3
##
   Mean
          :17.1
##
   3rd Qu.:19.2
           :24.4
  Max.
summary(mpgData[mpgData$am == "Manual",])
##
         mpg
##
          :15.0
                   Automatic: 0
   Min.
##
   1st Qu.:21.0
                   Manual
  Median:22.8
           :24.4
  Mean
##
   3rd Qu.:30.4
   Max.
           :33.9
```

The initial comparison is simply the summary statistics between automatic and manual. The "Manual" transmission cars on average get 7.3(24.4-17.1) MORE miles to the gallon than automatic cars do.

Basic Linear Model

```
fit <- lm(mtcars$mpg ~ as.integer(am), data=mpgData)</pre>
summary(fit)
##
## lm(formula = mtcars$mpg ~ as.integer(am), data = mpgData)
##
## Residuals:
     Min
              1Q Median
                            30
                                  Max
## -9.392 -3.092 -0.297 3.244
                                9.508
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                     9.90
                                 2.63
                                         3.77 0.00072 ***
                     7.24
                                 1.76
                                         4.11 0.00029 ***
## as.integer(am)
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 4.9 on 30 degrees of freedom
## Multiple R-squared: 0.36,
                                Adjusted R-squared: 0.338
## F-statistic: 16.9 on 1 and 30 DF, p-value: 0.000285
```

From summary Automatic transmission cars as the baseline, so the Intercept of 17.14. The coefficient for the Manual transmission is interepreted as 'how many more miles per gallon on average do you get by switching from an Automatic to a Manual' which is 7.24. The R-Squared is 0.338, which is quite low. That means this model does not fit the data terribly well.

Improved Linear Model

Now i try to improve the linear model by using ANOVA The Analysis Of Variance, popularly known as the ANOVA, can be used in cases where there are more than two groups.

```
fit2 <- update(fit, mtcars$mpg ~mtcars$am + mtcars$wt)
fit3 <- update(fit, mtcars$mpg ~mtcars$am + mtcars$wt + mtcars$qsec)
fit4 <- update(fit, mtcars$mpg ~ mtcars$am + mtcars$wt + mtcars$qsec + mtcars$cyl)
anova(fit, fit2, fit3, fit4)</pre>
```

```
## Analysis of Variance Table
##
## Model 1: mtcars$mpg ~ as.integer(am)
## Model 2: mtcars$mpg ~ mtcars$am + mtcars$wt
## Model 3: mtcars$mpg ~ mtcars$am + mtcars$wt + mtcars$qsec
## Model 4: mtcars$mpg ~ mtcars$am + mtcars$wt + mtcars$qsec + mtcars$cyl
    Res.Df RSS Df Sum of Sq
##
                                F Pr(>F)
## 1
        30 721
        29 278 1
                        443 71.22 4.7e-09 ***
## 2
## 3
        28 169 1
                        109 17.55 0.00027 ***
## 4
        27 168 1
                          2 0.24 0.62706
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

It is clear that the fit3 model, mpg being predicted by am, wt, and qsec gives us the strongest evidence to reject the null hypothesis.

```
summary(fit3)
```

```
##
## Call:
## lm(formula = mtcars$mpg ~ mtcars$am + mtcars$wt + mtcars$qsec,
       data = mpgData)
##
## Residuals:
##
     Min
             1Q Median
                            3Q
                                 Max
## -3.481 -1.556 -0.726 1.411 4.661
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                 9.618
                            6.960
                                     1.38 0.17792
## mtcars$am
                 2.936
                             1.411
                                     2.08 0.04672 *
## mtcars$wt
                -3.917
                            0.711
                                    -5.51
                                             7e-06 ***
## mtcars$qsec
                 1.226
                            0.289
                                     4.25 0.00022 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.46 on 28 degrees of freedom
                               Adjusted R-squared: 0.834
## Multiple R-squared: 0.85,
## F-statistic: 52.7 on 3 and 28 DF, p-value: 1.21e-11
```

This is much better. Now we have an adjusted R squared of 0.834, meaning this model explains 88.36% of the variance making for a much better fit.

```
bestfit <- update(fit, mtcars$mpg ~ mtcars$am + mtcars$wt + mtcars$qsec + mtcars$am*mtcars$wt)
anova(fit3, bestfit)</pre>
```

After trial and error (not detailed here for sake of space) I have discovered that the interaction am*wt should be included to create my bestfit model.

```
summary(bestfit)
```

```
##
## Call:
## lm(formula = mtcars$mpg ~ mtcars$am + mtcars$wt + mtcars$qsec +
## mtcars$am:mtcars$wt, data = mpgData)
```

```
##
## Residuals:
              1Q Median
##
      Min
                                   Max
  -3.508 -1.380 -0.559
##
                         1.063
                                 4.368
##
## Coefficients:
                        Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                           9.723
                                      5.899
                                               1.65
                                                     0.11089
## mtcars$am
                          14.079
                                      3.435
                                               4.10
                                                     0.00034 ***
## mtcars$wt
                          -2.937
                                      0.666
                                              -4.41
                                                      0.00015 ***
                                                      0.00040 ***
## mtcars$qsec
                           1.017
                                      0.252
                                               4.04
  mtcars$am:mtcars$wt
                                      1.197
                                              -3.46
                                                     0.00181 **
                          -4.141
##
                     '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
##
## Residual standard error: 2.08 on 27 degrees of freedom
## Multiple R-squared: 0.896, Adjusted R-squared: 0.88
## F-statistic: 58.1 on 4 and 27 DF, p-value: 7.17e-13
```

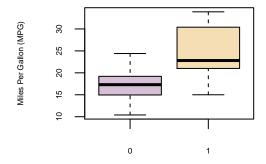
Now we can see we have an adjusted R squared of 0.8804 and small p-values for the coefficients, so this model explains 88.04% of the variance in our data providing a good fit with statistically significant coefficients.

Results

So from the Linear model and Basic analysis gave us manual transmission is better than automatic for MPG, which increased by 7.2449,and from Improved Linear Model difference between an automatic and manual car, a manual transmission car gets 14.079 more miles to the gallon than an Automatic - 4.141 * the weight. - Now answer to Q1 can say manual transmission is better than automatic - Now answer to Q2 manual transmission is better than automatic for MPG, which increased by 7.2449.

APPENDIX

Boxplot am vs mpg



Transmission (0 = Automatic, 1 = Manual)

Residuals

par(mfrow = c(2,2)) plot(bestfit)

