Motor Trend Cars Fuel Efficiency Analysis

Executive Summary

By studying the dataset mtcars, we found that the manual transmission is better for MPG than the automatic one, by a margin of 1.809. We reach this conclusion by finding the best model and exploring the relationship between a set of variables as predictors and the MPG as the output.

Exploratory Data Analysis

We begin the exploratory analysis by checking the structure and data class of the data frame mtcars.

```
str(mtcars)
```

```
'data.frame':
                    32 obs. of 11 variables:
   $ mpg : num
                 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
                 6 6 4 6 8 6 8 4 4 6 ...
   $ cyl : num
   $ disp: num
##
                 160 160 108 258 360 ...
                 110 110 93 110 175 105 245 62 95 123 ...
           num
                 3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
##
         : num
                 2.62 2.88 2.32 3.21 3.44 ...
##
                 16.5 17 18.6 19.4 17 ...
   $ qsec: num
                 0 0 1 1 0 1 0 1 1 1 ...
##
   $ vs
          : num
##
                 1 1 1 0 0 0 0 0 0 0 ...
         : num
   $ gear: num
                4 4 4 3 3 3 3 4 4 4 ...
   $ carb: num
                 4 4 1 1 2 1 4 2 2 4 ...
head(mtcars, 2)
##
                 mpg cyl disp hp drat
                                          wt qsec vs am gear carb
## Mazda RX4
                  21
                          160 110
                                   3.9 2.620 16.46
## Mazda RX4 Wag
                  21
                       6
                         160 110
                                   3.9 2.875 17.02
```

The data set has 32 observed samples and 11 variables, among which the "mpg" is the outcome and the other 10 are the predictors. It is noted that some variables, which should be considered as categorical factors, are listed as numeric. The conversion is conducted first.

```
mtcars$cyl<-as.factor(mtcars$cyl)
mtcars$vs<-as.factor(mtcars$vs)
mtcars$am<-as.factor(mtcars$am)
mtcars$gear<-as.factor(mtcars$gear)
mtcars$carb<-as.factor(mtcars$carb)</pre>
```

A boxplot is performed between "mpg" and "am" (transmission type, 0 for automatic, 1 for manual)[Fig-1], and it's obvious that the manual transmission has higher MPG. We also plot the "mpg" as the function of all other 10 predictors in Fig-2 and found that it has either positive or negative dependency on all predictors except the "gear".

Model Fitting and Selection

The visually observed dependency between "mpg" and almost all other predictors indicates the possible multicollinearity between the predictors. It is checked by calculating the variance inflation factor (vif) of each predictor in the complete fitting model.

```
library(car)
fitAll<-lm(mpg~., mtcars)
vif(fitAll)</pre>
```

```
GVIF Df GVIF<sup>(1/(2*Df))</sup>
##
## cyl 128.120962
                    2
                              3.364380
## disp 60.365687
                              7.769536
## hp
         28.219577
                    1
                              5.312210
## drat
          6.809663 1
                              2.609533
         23.830830 1
                              4.881683
## wt
## qsec 10.790189
                              3.284842
## vs
          8.088166
                   1
                              2.843970
## am
          9.930495
                              3.151269
## gear 50.852311
                              2.670408
                    2
## carb 503.211851
                              1.862838
```

The high VIF values of "disp", "hp" and "wt" suggest that a correlation exists between them and other predictors. The complete model with all 10 predictors is obviously not optimum and the number of predictors could be reduced. We use the Adjusted R-Squared, P values and backward elimination method to find the best model.

```
library(leaps)
subset<-regsubsets(mpg~., mtcars, nbest=2)</pre>
```

Fig-3 shows all the possible models and their corresponding Adjusted R-Squared values. We pick some top models with highest Adjusted R-Squared values for analysis, and try to find the best model among them.

```
fit1<-lm(mpg~cyl+hp+wt+am, mtcars)
fit2<-lm(mpg~cyl+hp+wt+am+vs, mtcars)
fit3<-lm(mpg~cyl+hp+wt+am+qsec, mtcars)</pre>
```

The "best" model is the one with fewest predictors yet still with one of the largest Adjusted R-Squared value among all models. The complete fitAll model has an Adjusted R-Squared value of ~0.807, and the fit1, fit2, and fit3 models have very similar Adjusted R-Squared value of ~0.84. Therefore we intend to go with the fit1 model. Now let's do the anova analysis bewteen fit1 and fit2, and bewteen fit1 and fit3, to test if they are substantially different.

```
anova(fit1, fit2)
```

```
## Analysis of Variance Table

##

## Model 1: mpg ~ cyl + hp + wt + am

## Model 2: mpg ~ cyl + hp + wt + am + vs

## Res.Df RSS Df Sum of Sq F Pr(>F)

## 1 26 151.03

## 2 25 143.68 1 7.3459 1.2782 0.269

anova(fit1, fit3)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ cyl + hp + wt + am
## Model 2: mpg ~ cyl + hp + wt + am + qsec
## Res.Df RSS Df Sum of Sq F Pr(>F)
## 1 26 151.03
## 2 25 143.98 1 7.0439 1.223 0.2793
```

When comparing the fit1 and fit2 models, the P value of 0.269 is larger than 0.05, so the null hypothesis that

the two models have no difference is tested false. Therefore there is a good reason to go with fit1 model which has fewer predictors. The same reasoning is behind the anova analysis of fit1 and fit3 models, which has a p value of 0.279. fit1 is the best model.

summary(fit1)\$coef

```
## (Intercept) 33.70832390 2.60488618 12.940421 7.733392e-13
## cyl6    -3.03134449 1.40728351 -2.154040 4.068272e-02
## cyl8    -2.16367532 2.28425172 -0.947214 3.522509e-01
## hp    -0.03210943 0.01369257 -2.345025 2.693461e-02
## wt    -2.49682942 0.88558779 -2.819404 9.081408e-03
## am1    1.80921138 1.39630450 1.295714 2.064597e-01

summary(fit1)$adj.r.squared
```

[1] 0.8400875

The slope coefficient 1.809 of the predictor am, the transmission type, shows that the mpg will increase by 1.809 when using a manual transmission instead of an automatic one. The p value of 2.065e-1 is not significant, so there is good reason to believe that the slope value is different from 0. The adjusted R squared value of 0.84 show that the model can explain 84% of the total variance.

Residual Analysis

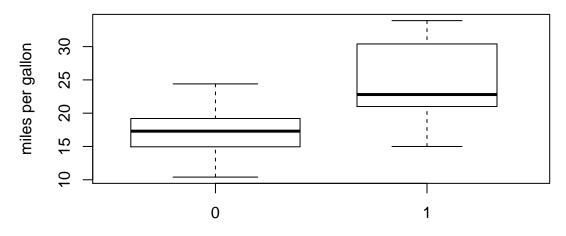
The plots of fit1 are attached as Fig-4. The "Residual vs Fitted" and "Standard Residual vs Fitted" curves show the residual is distributed randomly around 0 and no specified pattern can be seen. The "QQ" and "Cook's Distance" plots reveal, however, there may be some outliners toward the end of the sample data.

To Answer the Questions

- 1. The manual transmission is better for MPG
- 2. The expected MPG difference between the automatic and manual transmission is 1.809, with the manual one being higher.

Appendix

Fig-1 mpg vs transmission type



transmission type 0-auto 1-manual

Fig-2 mpg vs all predictors

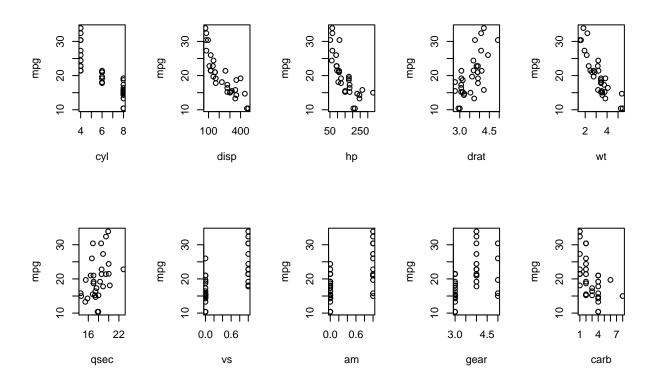
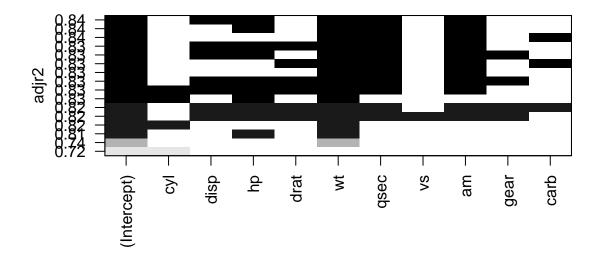


Fig-3 Adjusted R-Squared Plot for All Models



$Im(mpg \sim cyl + hp + wt + am)$

