Regression Models Course Project

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Based on the data set of a collection of cars (mtcars data set), we explore the relationship between a set of variables and miles per gallon (MPG) (outcome) and answer two questions:

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

Exploratory Analyses

Load the data and perform some basic exploratory data analyses

```
library (datasets)
data(mtcars)
dim(mtcars)
## [1] 32 11
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 ...
         : num
                 110 110 93 110 175 105 245 62 95 123 ...
                 3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
   $ drat: num
                2.62 2.88 2.32 3.21 3.44 ...
         : num
   $ qsec: num
                16.5 17 18.6 19.4 17 ...
                 0 0 1 1 0 1 0 1 1 1 ...
##
   $ vs
           num
                 1 1 1 0 0 0 0 0 0 0 ...
##
   $ am : num
                 4 4 4 3 3 3 3 4 4 4 ...
   $ gear: num
   $ carb: num
                4 4 1 1 2 1 4 2 2 4 ...
```

Data set consists consists of 11 variables and 32 observation for each variable.

Look at relationship between transmission type (am as factor variable (0 - automatic, 1 - manual) and miles per gallon (mpg) (Appendix A).

```
mtcars$am <- factor(mtcars$am, labels = c("automatic", "manual"))</pre>
```

Based on boxplot in **Appendix A** we can suppose that there is a significant difference in MPG for different transmission type.

Statistical Inference

Test our hypothesis: Null hypothesis is "the MPG means for different transmission type is equal" or "true difference in MPG means for different transmission type is equal to 0".

```
t.test(mpg ~ am, data = mtcars)

##

## Welch Two Sample t-test

##

## data: mpg by am

## t = -3.7671, df = 18.332, p-value = 0.001374

## alternative hypothesis: true difference in means is not equal to 0

## 95 percent confidence interval:

## -11.280194 -3.209684

## sample estimates:

## mean in group automatic mean in group manual

## 17.14737 24.39231
```

We can reject null hypothesis that the difference in MPG means for different transmission type is equal to 0 - value = 0.001374.

Regression Analysis

So MPG depends on transmission type, but define how other variables affect on MPG. Build multivariable regression model (results in **Appendix B**):

```
mtcars$cyl <- factor(mtcars$cyl)
mtcars$vs <- factor(mtcars$vs, labels = c("V", "S"))
mtcars$gear <- factor(mtcars$gear)
mtcars$carb <- factor(mtcars$carb)

fullModel <- lm(mpg ~ ., data = mtcars)
summary(fullModel)</pre>
```

So none of the coefficients have a p-value less than 0.05 (statistically significant). Find better model (based on removing variables from the model and evaluating the AIC):

```
AICModel <- step(fullModel, direction = "both")
summary(AICModel)
anova(AICModel, fullModel)</pre>
```

Comparing the AICModel with the fullModel we see that removing other predictors has not significantly affected the explanatory ability of the model.

The AICModel explains about 87% of the variance in MPG (R-squared is 0.8659). The coefficients conclude that increasing the number of cylinders from 4 to 6 with decrease the MPG by 3.031, but from 4 to 8 with decrease the MPG by 2.164. One additional horsepower is decreases MPG by 0.0321. Weight decreases the MPG by 2.497 for each 1000 lbs increase. A Manual transmission improves the MPG by 1.809.

Residual Analysis

Based on residuals plots (**Appendix C**) we can conclude:

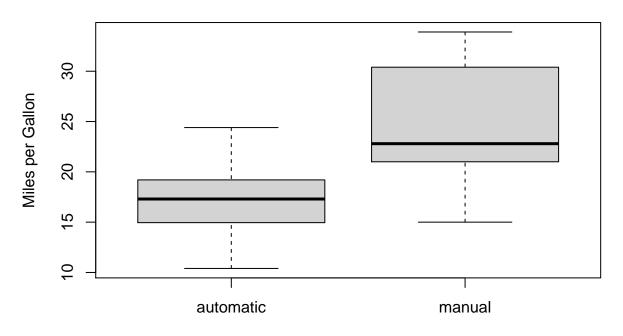
- the Residuals vs Fitted plot doesn't show pattern and confirms that residuals are independent;
- the Normal Q-Q plot confirms that the residuals are normally distributed (with some deviate from normality at the tails);
- the Scale-Location confirms the constant variance assumption;
- the Residuals vs Leverage confirms that there are no outliers (all values fall within the 0.5 bands).

Conclusion

- There is a significant difference in MPG for different transmission type (MPG mean for manual type more automatic type at 7.24).
- Based on AICModel we can conclude that number of cylinders, weight and horsepower are more statistically significant then transmission type for determining MPG.

Appendix A

MPG vs. Transmission Type



Transmission Type

Appendix B

summary(fullModel)

```
##
## Call:
## lm(formula = mpg ~ ., data = mtcars)
##
## Residuals:
##
       Min
                1Q Median
                                ЗQ
                                       Max
## -3.5087 -1.3584 -0.0948 0.7745 4.6251
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
                                      1.190
## (Intercept) 23.87913
                          20.06582
                                             0.2525
## cyl6
               -2.64870
                           3.04089 -0.871
                                              0.3975
## cyl8
                           7.15954 -0.047
               -0.33616
                                             0.9632
```

```
## disp
              0.03555
                         0.03190
                                 1.114
                                          0.2827
## hp
             -0.07051 0.03943 -1.788 0.0939 .
## drat
             1.18283
                         2.48348
                                 0.476 0.6407
                         2.53875 -1.784 0.0946 .
## wt
             -4.52978
## qsec
              0.36784
                        0.93540
                                 0.393 0.6997
                       2.87126
                                 0.672 0.5115
## vsS
             1.93085
## ammanual
                         3.21355 0.377 0.7113
             1.21212
             1.11435
                         3.79952 0.293 0.7733
## gear4
## gear5
              2.52840
                         3.73636
                                 0.677
                                          0.5089
                         2.31797 -0.423 0.6787
## carb2
             -0.97935
## carb3
              2.99964
                         4.29355
                                 0.699 0.4955
                                 0.245 0.8096
## carb4
             1.09142
                         4.44962
## carb6
              4.47757
                         6.38406
                                 0.701 0.4938
## carb8
             7.25041
                         8.36057 0.867 0.3995
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.833 on 15 degrees of freedom
## Multiple R-squared: 0.8931, Adjusted R-squared: 0.779
## F-statistic: 7.83 on 16 and 15 DF, p-value: 0.000124
summary(AICModel)
##
## Call:
## lm(formula = mpg ~ cyl + hp + wt + am, data = mtcars)
##
## Residuals:
               1Q Median
      Min
                              3Q
                                    Max
## -3.9387 -1.2560 -0.4013 1.1253 5.0513
##
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 33.70832 2.60489 12.940 7.73e-13 ***
## cyl6
             -3.03134
                       1.40728 -2.154 0.04068 *
## cyl8
                         2.28425 -0.947 0.35225
             -2.16368
                         0.01369 -2.345 0.02693 *
## hp
             -0.03211
                         0.88559 -2.819 0.00908 **
## wt
             -2.49683
                         1.39630 1.296 0.20646
## ammanual
             1.80921
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.41 on 26 degrees of freedom
## Multiple R-squared: 0.8659, Adjusted R-squared: 0.8401
## F-statistic: 33.57 on 5 and 26 DF, p-value: 1.506e-10
anova(AICModel, fullModel)
## Analysis of Variance Table
## Model 1: mpg ~ cyl + hp + wt + am
## Model 2: mpg ~ cyl + disp + hp + drat + wt + qsec + vs + am + gear + carb
             RSS Df Sum of Sq
    Res.Df
                                 F Pr(>F)
## 1
        26 151.03
## 2
        15 120.40 11
                      30.623 0.3468 0.9588
```

Appendix C

Residuals plots for AICModel:

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

