Miles per Gallon (MPG) – Automatic Transmission vs Manual Transmission

Executive Summary

The purpose of this project is to explore the relationship between a set of variables and miles per gallon (MPG). Specifically, we want to answer the following two questions:

- 1. Is an automatic or manual transmission better for MPG"?
- 2. How different is the MPG between automatic and manual transmissions?

This study shows that the answer depends on the car's weight (**wt**) and the 1/4 mile time (**qsec**), a performance benchmark for car acceleration. For light weighted cars, manual transmission is better for MPG. However, for heavier cars, automatic transmission is better for MPG. For the range of data considered in this study, the difference in MPG between automatic and manual transmission is between -7 and 7.

Data source

The mtcars data set in R's datasets package is used. mtcars has 11 variables and 32 observations, in which 19 cars have automatic transmission and 13 cars have manual transmission. It is a small data set. It is important that we don't overfit the model by using too many predictors. A new data frame mtcars2 is copied from the original data mtcars. mtcars2 is the same as mtcars except that the variables am and vs are factored.

The first model

Let \mathbf{Y} denote the car's MPG, \mathbf{T} the car's transmission (1 for automatic; 2 for manual), and \mathbf{X} the other predictor of the car's MPG. We can start with the following regression model:

```
\mathbf{Y} = \mathbf{C}_0 + \mathbf{C}_1 \mathbf{T}_1 + C_{2,1} \mathbf{X}, for cars with automatic transmission
```

```
\mathbf{Y} = \mathbf{C}_0 + \mathbf{C}_1 \mathbf{T} + \mathbf{C}_2 \mathbf{X}, for cars with manual transmission
```

Which variable should be used? It is desirable to pick the one that has the largest absolute correlation with MPG.

```
which.max(abs(cor(mtcars)[2:11, 1])) + 1
## wt
## 6
```

Consequently, the first model can be generated using the following codes:

```
mpg_formula <- mpg ~ am + am:wt
mpg_lm <- lm(mpg_formula, data = mtcars2)</pre>
summary(mpg_lm)
##
## Call:
## lm(formula = mpg formula, data = mtcars2)
##
## Residuals:
##
              1Q Median
                             3Q
      Min
                                   Max
   -3.600 -1.545 -0.533 0.901
                                6.091
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 31.416
                              3.020
                                      10.40 4.0e-11 ***
                              4.264
                                       3.49
                                              0.0016 **
## am1
                 14.878
## am0:wt
                 -3.786
                              0.786
                                      -4.82 4.6e-05 ***
                                      -7.49
                                             3.7e-08 ***
## am1:wt
                 -9.084
                              1.212
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
##
## Residual standard error: 2.59 on 28 degrees of freedom
## Multiple R-squared: 0.833, Adjusted R-squared: 0.815
## F-statistic: 46.6 on 3 and 28 DF, p-value: 5.21e-11
```

The plot is given in Fig 1 in Appendix.

Final Model and Variable Selection

Can the first model be improved? Let \mathbf{Y} denote the car's MPG, \mathbf{T} the car's transmission (1 for automatic; 2 for manual), \mathbf{X} _2 the car weight (\mathbf{wt}), and \mathbf{X} _3 the added variable. The new model may be written as:

```
\mathbf{Y} = \mathbf{C}_0 + \mathbf{C}_1 \mathbf{T}_1 + C_{2,1} \mathbf{X}_2 + \mathbf{C}_{3,1} \mathbf{X}_3, for cars with automatic transmission \mathbf{Y} = \mathbf{C}_0 + \mathbf{C}_1 \mathbf{T}_2 + C_{2,2} \mathbf{X}_2 + \mathbf{C}_{3,2} \mathbf{X}_3, for cars with manual transmission
```

The following method is used to search for X_3 . The model that produces the smallest p-value of the anova table may be desirable.

```
fit_cyl <- anova(mpg_lm, update(mpg_lm, mpg ~ am + am:(wt + cyl)))
fit_disp <- anova(mpg_lm, update(mpg_lm, mpg ~ am + am:(wt + disp)))
fit_hp <- anova(mpg_lm, update(mpg_lm, mpg ~ am + am:(wt + hp)))
fit_drat <- anova(mpg_lm, update(mpg_lm, mpg ~ am + am:(wt + drat)))
fit_qsec <- anova(mpg_lm, update(mpg_lm, mpg ~ am + am:(wt + qsec)))
fit_vs <- anova(mpg_lm, update(mpg_lm, mpg ~ am + am:(wt + vs)))
fit_gear <- anova(mpg_lm, update(mpg_lm, mpg ~ am + am:(wt + gear)))
fit_carb <- anova(mpg_lm, update(mpg_lm, mpg ~ am + am:(wt + carb)))</pre>
```

The model that including **qsec** produces the smallest p-value (0.002) among all. Furthermore, that p-value is significant small. To ensure that the 1/4 mile time (**qsec**) is a good choice, a plot of **MPG** vs **qsec** is given in Fig 2 in Appendix. Note that for a given **qsec**, manual transmission is always better for MPG.

The final model is as follows:

```
mpg_formula_fin <- mpg ~ am + am:(wt + qsec)</pre>
mpg_lm_fin <- lm(mpg_formula_fin, data = mtcars2)</pre>
summary(mpg_lm_fin)
##
## Call:
## lm(formula = mpg_formula_fin, data = mtcars2)
##
## Residuals:
##
      Min
              1Q Median
                            3Q
                                  Max
   -3.683 -1.322 -0.375 1.069
                               4.091
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                 11.249
                             6.992
                                      1.61 0.11974
                            12.666
                                      0.70 0.48723
## am1
                  8.926
## am0:wt
                 -2.996
                             0.691
                                     -4.34 0.00019 ***
                 -6.754
                                     -5.01 3.3e-05 ***
## am1:wt
                             1.349
## am0:qsec
                  0.945
                             0.307
                                      3.08 0.00481 **
## am1:qsec
                  1.181
                             0.464
                                      2.54 0.01729 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.12 on 26 degrees of freedom
## Multiple R-squared: 0.897, Adjusted R-squared: 0.877
## F-statistic: 45.1 on 5 and 26 DF, p-value: 5.39e-12
```

As it is shown in the summary statistics of the final model, the final model produces smaller residual standard error, higher R-square and adjusted R-square than the first model.

A comparison of residuals of the two models are plotted in Fig 3 and 4 (Appendix)

Is an automatic or manual transmission better for MPG?

Based on the plot of the first model (Fig 1, Appendix), the answer of question 1 depends on the car's weight. For cars with a weight less than 2.8081 tons, manual transmission is better for MPG. However, for cars with a weight greater than 2.8081 tons, automatic transmission is better for MPG.

To quantify the differences, the following routine is used:

```
wtarr \leftarrow seq(from = 1.5, to = 5.5, by = 0.5)
qsecarr \leftarrow seq(from = 15, to = 22, by = 0.5)
nwtarr <- length(wtarr)</pre>
nqsecarr <- length(qsecarr)</pre>
wtarr2 <- rep(rep(wtarr, each = nqsecarr), times = 2)</pre>
qsecarr2 <- rep(qsecarr, times = nwtarr * 2)</pre>
amarr2 <- factor(rep(c(0, 1), each = nwtarr * nqsecarr))</pre>
p_dataframe <- data.frame(mpg = rep(0, times = nwtarr * nqsecarr * 2), wt = wtarr2,
    qsec = qsecarr2, am = amarr2)
p_dataframe$predict <- predict(mpg_lm_fin, newdata = p_dataframe)</pre>
diffmat <- matrix(0, nrow = nwtarr, ncol = nqsecarr)</pre>
for (i in 1:nwtarr) {
    for (j in 1:nqsecarr) {
        diffmat[i, j] = p_dataframe$predict[(i - 1) * nqsecarr + j] - p_dataframe$predict[(i -
             1) * nqsecarr + j + nwtarr * nqsecarr]
    }
}
```

The contours of the difference (MPG of automatic transmission - MPG of manual transmission) matrix are plotted in Fig 5 (Appendix). The magnitude and the sign of the difference depend on the car's weight (wt) and 1/4 mile time (qsec). Within the scope of this data set, the difference ranges between (-7, 7).

Appendix

```
layout(matrix(1:2, ncol = 2))
plot(mpg ~ wt, data = mtcars2, col = as.integer(mtcars2$am), main = "Fig 1- MPG vs wt")
abline(lm(mpg ~ wt, data = mtcars2, subset = am == 0), col = 1)
abline(lm(mpg ~ wt, data = mtcars2, subset = am == 1), col = 2)
legend("topright", legend = c("auto", "manual"), col = c(1, 2), pch = 1, lty = 1)
plot(mpg ~ qsec, data = mtcars2, col = as.integer(mtcars2$am), main = "Fig 2- MPG vs qsec")
legend("topright", legend = c("auto", "manual"), col = c(1, 2), pch = 1)
layout(matrix(1:2, ncol = 2))
mpg_resid <- residuals(mpg_lm)</pre>
mpg_fitted <- fitted(mpg_lm)</pre>
plot(mpg_fitted, mpg_resid, xlab = "Fitted", ylab = "Residuals", main = "Fig 3- First model",
    col = as.integer(mtcars2$am), ylim = c(-3, 7))
mpg_resid <- residuals(mpg_lm_fin)</pre>
mpg_fitted <- fitted(mpg_lm_fin)</pre>
plot(mpg_fitted, mpg_resid, xlab = "Fitted", ylab = "Residuals", main = "Fig 4- Final model",
    col = as.integer(mtcars2$am), ylim = c(-3, 7))
layout(matrix(1))
contour(x = wtarr, y = qsecarr, z = diffmat, xlab = "wt", ylab = "qsec", labcex = 1.5,
    main = "Fig 5- difference in MPG (auto-manual)")
points(mtcars2$wt, mtcars2$qsec, col = as.integer(mtcars2$am))
legend("topright", legend = c("auto", "manual"), col = c(1, 2), pch = 1)
```

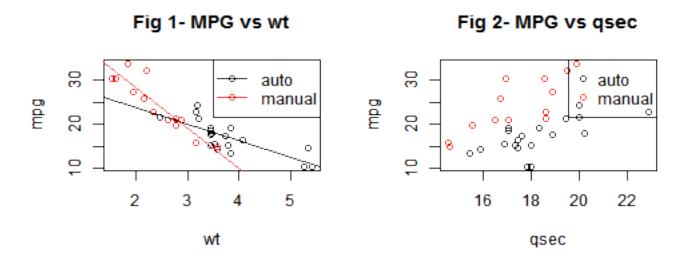


Figure 1: plot of chunk unnamed-chunk-7 $\,$

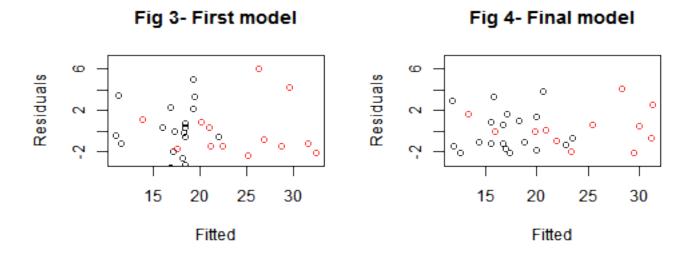


Figure 2: plot of chunk unnamed-chunk-8

Fig 5- difference in MPG (auto-manual)

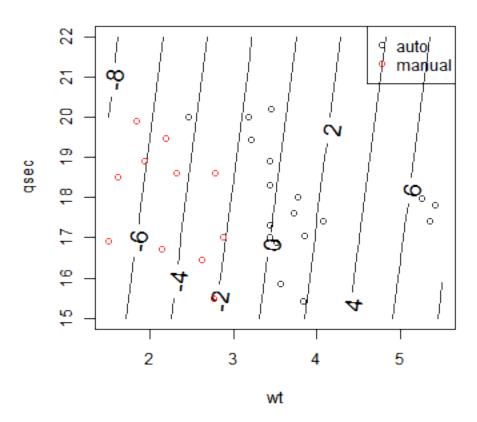


Figure 3: plot of chunk unnamed-chunk-9