Manual or automatic: how does affect fuel performance?

Javier Santibáñez

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Summary

In this report we present the results of a comparison between manual and automatic vehicles, measured in miles per gallon of fuel (mpg). The results shows that automatic cars has a better performance than manual cars. The expected difference was estimated in XXX mpg.

Introduction

We used a data set that consists of the measurements of 11 variables from observations of cars from 32 cars. The variables are the following:

- mpg, miles(US)/gallon
- cyl, number of cylinders
- dip, displacement (cu.in.)
- hp, gross horsepower
- drat, rear axie ratio
- wt, weigth (lb/1000)
- qsec 1/4 mile time
- vs V/S
- am transmission (automatic, manual)
- gear, number of forward gears
- carb, number of carburetors

Additionally, we created a new variable, called am2, which is the factor version of am with levels automatic and manual.

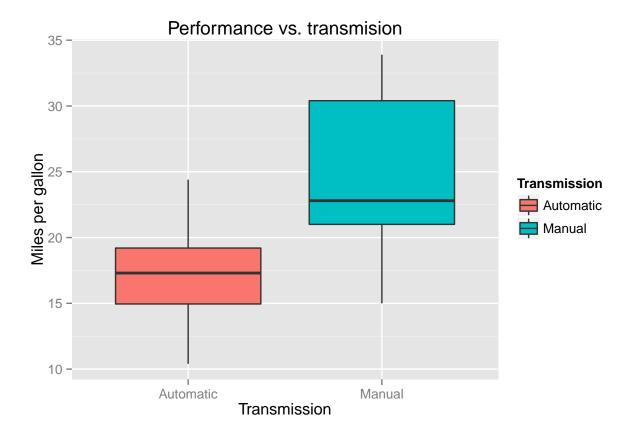
In this report we have to address the questions:

- 1. Is an automatic or manual transmission better for performance?
- 2. What is the difference in performance between automatic and manual transmissions?

To deal with this questions we used a statistical approach, in specific, we used regression models.

The first quotion can be easily answered. Figure *Performance vs. transmission* suggests that performance, measured in miles per gallon, depends of the car transmissions.

```
suppressWarnings(suppressMessages(library(ggplot2)))
suppressWarnings(suppressMessages(library(knitr)))
suppressWarnings(suppressMessages(library(corrplot)))
mtcars$am2<-factor(mtcars$am,labels=c("Automatic","Manual"))
g<-ggplot(data=mtcars,aes(factor(am2),mpg))
g+geom_boxplot(aes(fill=am2))+
    xlab("Transmission")+ylab("Miles per gallon")+labs(fill="Transmission")+
    ggtitle("Performance vs. transmision")</pre>
```



Despite the previous graphic, we must verify that manual trasmission is better than automatic trasmission. Also, it is important to consider other variables, because some of them can explain part of the variability.

Finally, we used the software statistical environment R and the packages ggplot2 and corrplot for graphics.

Methodology

The following is a brief description of the analysis performed:

- 1. Fit a regression model with mpg (performance) as output and am (transmission) as input.
- 2. Select other covariables from the data set, considering their correlation with mpg.
- 3. Fit a regression model with mpg as output and the set of variables from the preceding step as input.
- 4. Evaluate and compare models fitted in steps 1 and 4.

Results

Step 1. Basic model.

The basic model consists in only one input (mpg). We used the next code to fit the model.

```
fitMod1<-lm(mpg~am2,data=mtcars)
summary(fitMod1)</pre>
```

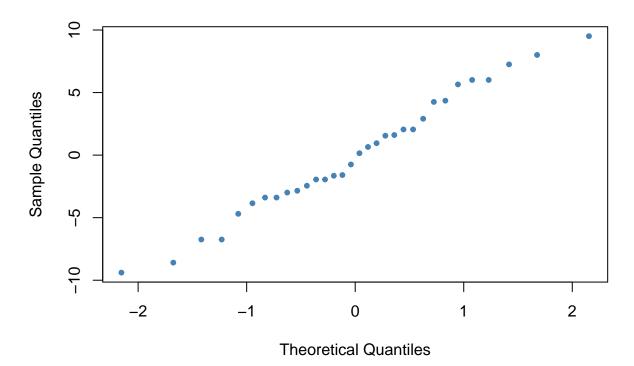
```
##
## Call:
  lm(formula = mpg ~ am2, data = mtcars)
##
##
  Residuals:
##
      Min
              1Q Median
                            3Q
                                   Max
   -9.392 -3.092 -0.297
                         3.244
##
##
  Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                                            1.1e-15 ***
##
   (Intercept)
                  17.15
                              1.12
                                      15.25
                   7.24
                                       4.11 0.00029 ***
   am2Manual
                              1.76
##
##
                   0 '***' 0.001 '**' 0.01 '*' 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
```

The previous results indicates that there is a highly significant difference in perfomance between automatic and manual transmission, estimated as 7.2449 miles per gallon, on average. With this information we can answer our two questions, but we improve them with the following steps.

We verify the accuracy of the normal assumption with a qq-plot, it shows that the residuals are approximately normal distributed.

```
qqnorm(fitMod1$residual,pch=20,col="steelblue")
```

Normal Q-Q Plot



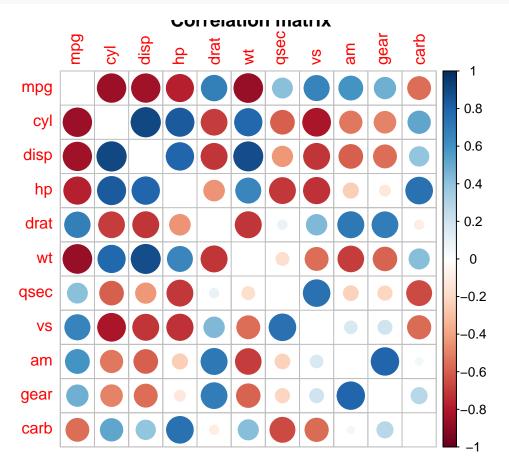
Step 2. Select new variables.

To select new variables to include in the model, we used the correlation matrix via the correlation. From Figure Correlation matrix we can identify a set of variables highly correlated:

- 6, 6, 4, 6, 8, 6, 8, 4, 4, 6, 6, 8, 8, 8, 8, 8, 8, 4, 4, 4, 4, 8, 8, 8, 8, 4, 4, 4, 8, 6, 8, 4
- 160, 160, 108, 258, 360, 225, 360, 146.7, 140.8, 167.6, 167.6, 275.8, 275.8, 275.8, 472, 460, 440, 78.7, 75.7, 71.1, 120.1, 318, 304, 350, 400, 79, 120.3, 95.1, 351, 145, 301, 121
- 110, 110, 93, 110, 175, 105, 245, 62, 95, 123, 123, 180, 180, 180, 205, 215, 230, 66, 52, 65, 97, 150, 150, 245, 175, 66, 91, 113, 264, 175, 335, 109
- 3.9, 3.9, 3.85, 3.08, 3.15, 2.76, 3.21, 3.69, 3.92, 3.92, 3.92, 3.07, 3.07, 3.07, 2.93, 3, 3.23, 4.08, 4.93, 4.22, 3.7, 2.76, 3.15, 3.73, 3.08, 4.08, 4.43, 3.77, 4.22, 3.62, 3.54, 4.11

Despite this result, we cannot include the whole set of variables because there are a high correlation between them. Then, we have the choose one variable from that set. We select the variable hp because we can use it as a confounder in a extended model, to reduce variability and to improve the estimation of difference in mpg due to transmission.

cor_mtcars<-cor(mtcars[,-12])
corrplot(cor_mtcars,diag=F,title="Correlation matrix")</pre>



Step 3. Extended model.

Now we extend our basic model adding the variable hp. First we fit a model with an interection term with the following code.

```
fitMod2<-lm(mpg~am2+hp+am2*hp,data=mtcars)
summary(fitMod2)</pre>
```

```
##
## Call:
## lm(formula = mpg \sim am2 + hp + am2 * hp, data = mtcars)
## Residuals:
     Min
##
             1Q Median
                           3Q
                                 Max
## -4.382 -2.270 0.134 1.706 5.875
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 26.624848
                          2.182943
                                      12.20
                                               1e-12 ***
## am2Manual
                5.217653
                          2.665093
                                       1.96
                                                0.06 .
               -0.059137
                           0.012945
                                      -4.57
                                               9e-05 ***
## am2Manual:hp 0.000403
                          0.016460
                                       0.02
                                                0.98
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.96 on 28 degrees of freedom
## Multiple R-squared: 0.782, Adjusted R-squared: 0.759
## F-statistic: 33.5 on 3 and 28 DF, p-value: 2.11e-09
```

The previous results suggest that there is evidence to support the hipotesis that the interaction term is not necessary.

We fit a new model without the interaction term between am and hp, these are the results:

```
fitMod3<-lm(mpg~am2+hp,data=mtcars)
summary(fitMod3)</pre>
```

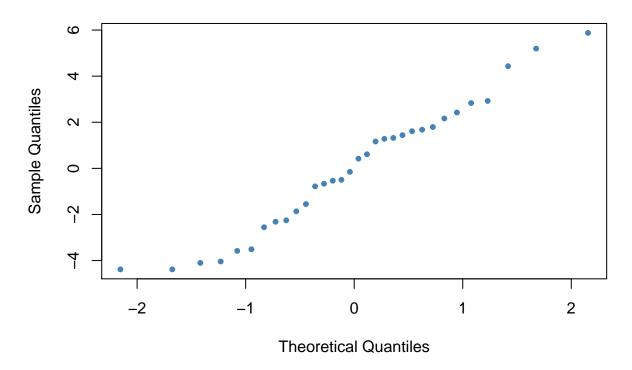
```
##
## Call:
## lm(formula = mpg ~ am2 + hp, data = mtcars)
## Residuals:
     Min
             1Q Median
                           30
## -4.384 -2.264 0.137 1.697 5.866
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                         1.42509
                                   18.65 < 2e-16 ***
## (Intercept) 26.58491
## am2Manual
               5.27709
                          1.07954
                                    4.89 3.5e-05 ***
## hp
              -0.05889
                          0.00786
                                    -7.50 2.9e-08 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.91 on 29 degrees of freedom
## Multiple R-squared: 0.782, Adjusted R-squared: 0.767
## F-statistic: 52 on 2 and 29 DF, p-value: 2.55e-10
```

These results suggest support again that there is a difference in performace between automatic and manual transmission, estimated in 5.2771 miles per gallon, but in the case, we take into account a confounder that explains some of the residual variability.

We evaluated the assuption of normality with a qq-plot and it seems that there is no problems with normality.

qqnorm(fitMod2\$residual,pch=20,col="steelblue")

Normal Q-Q Plot



Step 4. Comparing models.

We estimated two models. An analysis of variance (ANOVA) shows that the extended model is better than the basic model, with a high level of significance.

kable(anova(fitMod1,fitMod3))

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
30	720.9	NA	NA	NA	NA
29	245.4	1	475.5	56.18	0

Also, the ratio of the sum of squared residuals of the extended model to the basic model is 0.34, which means that the residual variation of the extended model is 34% smallet than the residual variation of the basic model.

Conclusions

According to the preceding results we can conclude that:

- $1.\ A$ manual transmission is better than an automatic one, in terms of performance, measured in miles per gallon.
- 2. There is a difference of 5.2771 miles per gallon between manual and automatic transmissions.