Exploring the relationship between miles per gallon and transmission type

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Executive summary

The analysis has shown that miles per gallon does not depend on the transmission type.

The difference that can be observed without taking into account other predictors is explained by parameters which have quite obvious relation with the mpg outcome, such as the weight of the car (the more the weight of the car - the more gas is needed per mile), the horse power (the more powerful the engine - the more gas it uses per mile).

Exploratory analysis

All the figures related to exploratory analysis are provided in the Appendix (dataset structure, first rows of the data, and some plots).

Looking at parameters in the dataset, we may assume that number of cylinders (cyl), horse power (hp) and weight (wt) are highly correlated with each other and with the miles per gallon (mpg). There are other parameters which presumably have little effect on the outcome due to lack of correlation. To prove this point, we can look at correlations of parameters using ggcorr function (the plot is provided in Appendix).

Model selection

We'll fit several models, subsequently adding predictors one by one and test the models using the analysis of variance (anova).

Since our main questions are about the relationship of miles per gallon and transmission type, the first model will only include one predictor: factor(am). The model will show the mean mpg for cars with automatic transmission (Intercept) and the difference of the mean mpg for cars with manual transmission (factor(am)1 coefficient) from the atomatic ones.

```
## lm(formula = mpg ~ factor(am), data = mtcars)

## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 17.147368 1.124603 15.247492 1.133983e-15
## factor(am)1 7.244939 1.764422 4.106127 2.850207e-04
```

We can see that both coefficients are strongly significant (p-values are much less than the $\alpha = 0.05$). However we intentionally omitted all the other predictors and the model has high bias.

Next several models are nested, i.e. each subsequent model contains additional predictor. If new predictor in the next model did not improve the model (i.e. p-value for the F-statistic in anova test was bigger than 0.05) than we fit new model with another predictor instead. As a result we find a model which is cosidered the best.

```
## Analysis of Variance Table
##
## Model 1: mpg ~ factor(am)
## Model 2: mpg ~ factor(am) + hp
```

```
## Model 3: mpg ~ factor(am) + hp + wt
## Model 4: mpg ~ factor(am) + hp + wt + factor(cyl)
     Res.Df
               RSS Df Sum of Sq
## 1
         30 720.90
## 2
         29 245.44
                    1
                         475.46 81.8529 1.634e-09 ***
## 3
         28 180.29
                          65.15 11.2157 0.002484 **
                    1
         26 151.03
                          29.27 2.5191 0.099998 .
## ---
## Signif. codes:
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Predictors disp, drat, qsec, vs, gear and carb did not improve the model. Predictor factor(cyl) was useful at some point but trying to use it in different order in the nested test, it seems that it does not lead to significant improvement compared to the model with hp and wt only. It is seen from the parameters of the model 4 - the p-value is bigger than 0.05. So, we choose previous model 3 for further analysis.

Model diagnostics

We need to check the normality of the residuals in order to be sure that anova results can be trusted. We'll use shapiro.test on resuduals of the chosen model.

```
##
## Shapiro-Wilk normality test
##
## data: fit3$residuals
## W = 0.9453, p-value = 0.1059
```

We can see that the p-value is bigger than 0.05 which means that we fail to reject the null-hypothesis which states that the data is normally distributed. This means that the residuals are normally distributed and the model can be accepted.

If we look at model diagnostics plots (provided in Appendix), we can see that there are a few cars that have parameters with high leverage and influence but even removing them doesn't change the answers for the main questions of the analysis.

Questions and Answers

```
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 34.00287512 2.642659337 12.866916 2.824030e-13
## factor(am)1 2.08371013 1.376420152 1.513862 1.412682e-01
## hp -0.03747873 0.009605422 -3.901830 5.464023e-04
## wt -2.87857541 0.904970538 -3.180850 3.574031e-03
```

From the coefficients we can see that for manual transmission (factor(am)1), the estimated change of mpg is 2.0837101 compared to automatic transmission with hp and wt fixed. However, looking at p-value = 0.1412682 which is bigger than 0.05, or looking at 95% confidence interval for the estimate -0.7357587, 4.903179 which contains 0, the change in mpg is not statistically significant. Based on this result, the answers for the questions of the analysis are provided below:

- 1. Is an automatic or manual transmission better for MPG There is no difference.
- 2. Quantify the MPG difference between automatic and manual transmissions The MPG difference is not statistically significant.

Appendix

Exloratory summaries

Dataset structure:

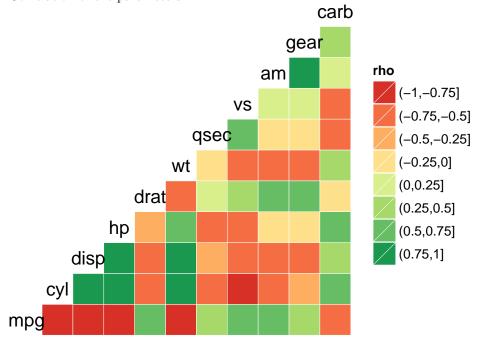
```
32 obs. of 11 variables:
  'data.frame':
   $ mpg : num 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
## $ cyl : num 6646868446 ...
               160 160 108 258 360 ...
  $ disp: num
## $ hp : num
               110 110 93 110 175 105 245 62 95 123 ...
                3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
##
   $ drat: num
##
   $ wt : num 2.62 2.88 2.32 3.21 3.44 ...
  $ qsec: num
               16.5 17 18.6 19.4 17 ...
  $ vs : num
               0 0 1 1 0 1 0 1 1 1 ...
               1 1 1 0 0 0 0 0 0 0 ...
   $ am : 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 ...
```

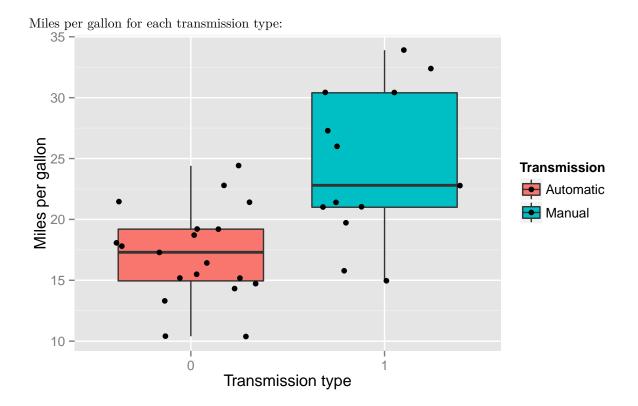
First few rows of the dataset:

```
##
                     mpg cyl disp hp drat
                                             wt qsec vs am gear carb
## Mazda RX4
                           6 160 110 3.90 2.620 16.46
                    21.0
                                                       0
                           6 160 110 3.90 2.875 17.02 0
## Mazda RX4 Wag
                    21.0
## Datsun 710
                    22.8 4 108 93 3.85 2.320 18.61
                          6 258 110 3.08 3.215 19.44
## Hornet 4 Drive
                    21.4
                                                                   1
                                                       1
## Hornet Sportabout 18.7
                          8 360 175 3.15 3.440 17.02
                                                                   2
                                                       0
## Valiant
                    18.1
                           6 225 105 2.76 3.460 20.22 1 0
                                                                   1
```

Exploratory plots

Correlation of the parameters:





Model diagnostics plots:

