# What type of transmission is better for the MPG

Alexander Alexandrov

#### Overview

Work for Motor Trend, a magazine about the automobile industry. Looking at a data set of a collection of cars, they are interested in exploring the relationship between a set of variables and miles per gallon (MPG) (outcome). They are particularly interested in the following two questions:

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

Its evident from my experience that *transimission type* is not the only variable that has impact on *mpg*. So several models should be analyzed and compared to each over. Following parameters should be considered: transmission type, weight, gross horsepower, number of cylinders, number of forward gears.

## **Exploratory Analysis**

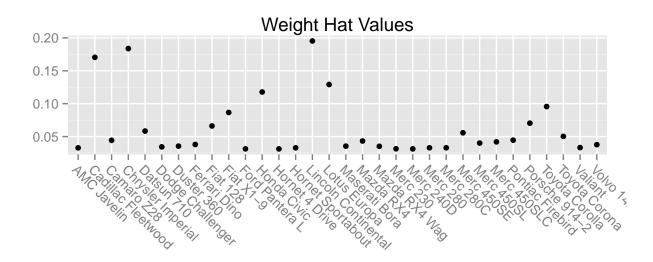
Load Motor Trend cars data. And adjust some variables.

```
##
  'data.frame':
                   32 obs. of 12 variables:
   $ mpg : num 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
   $ cyl : Factor w/ 3 levels "4","6","8": 2 2 1 2 3 2 3 1 1 2 ...
   $ disp: num 160 160 108 258 360 ...
                110 110 93 110 175 105 245 62 95 123 ...
   $ hp : num
                3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
   $ wt : num 2.62 2.88 2.32 3.21 3.44 ...
   $ qsec: num 16.5 17 18.6 19.4 17 ...
   $ vs : Factor w/ 2 levels "0","1": 1 1 2 2 1 2 1 2 2 2 ...
   $ am : Factor w/ 2 levels "0","1": 2 2 2 1 1 1 1 1 1 1 ...
   $ gear: Factor w/ 3 levels "3","4","5": 2 2 2 1 1 1 1 2 2 2 ...
   $ carb: num 4 4 1 1 2 1 4 2 2 4 ...
                "Mazda RX4" "Mazda RX4 Wag" "Datsun 710" "Hornet 4 Drive" ...
```

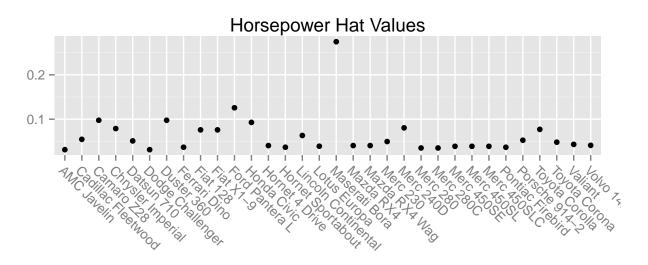
Variable	Type	Domain	Description
$\overline{\mathrm{mpg}}$	numeric	positive real number	Miles/(US) gallon
cyl	factor	4, 6, 8	Number of cylinders
$\operatorname{disp}$	numeric	positive real number	Displacement (cu.in.)
hp	numeric	positive real number	Gross horsepower
$\overline{\mathrm{drat}}$	numeric	positive real number	Rear axle ratio
$\mathbf{wt}$	numeric	positive real number	Weight (lb/1000)
qsec	numeric	positive real number	1/4 mile time
vs	factor	0, 1	V/S
am	factor	0, 1	Transmission $(0 = automatic, 1 = manual)$
gear	factor	3, 4, 5	Number of forward gears
carb	$\operatorname{numeric}$	positive integer number	Number of carburetors

# **Data Cleaning**

Outliers of the weight variable



Outliers of the horsepower variable



The only evident outliers are: "Cadillac Fleetwood", "Chrysler Imperial", "Lincoln Continental", "Maserati Bora". This is error or extraordinary cases. And they should be removed from the data before the regression analysis to avoid injurious effect.

mtcars <- mtcars[!(mtcars\$id %in% c("Cadillac Fleetwood", "Chrysler Imperial", "Lincoln Continental", "</pre>

#### Multivariate Regression Analysis

Following nested models will be treated against each over:

```
fit <- lm(mpg ~ am, data = mtcars)</pre>
fit.wt <- update(fit, mpg ~ am*wt)</pre>
fit.wt.hp <- update(fit, mpg ~ am*wt + am*hp)</pre>
anova(fit, fit.wt, fit.wt.hp)
## Analysis of Variance Table
## Model 1: mpg ~ am
## Model 2: mpg ~ am + wt + am:wt
## Model 3: mpg ~ am + wt + hp + am:wt + am:hp
     Res.Df
               RSS Df Sum of Sq
                                        F
## 1
         26 512.40
## 2
         24 165.49 2
                          346.92 36.0442 1.142e-07 ***
## 3
         22 105.87 2
                          59.61 6.1937 0.007349 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
According to p-values (they are less than 0.05), the best choi e is the last and more complicated model. Lets
test two additional models against the best one.
best.fit \leftarrow lm(mpg \sim am*wt + am*hp, data = mtcars)
fit.wt.hp.cyl <- update(best.fit, mpg ~ am*cyl*wt + am*cyl*hp)</pre>
anova(best.fit, fit.wt.hp.cyl)
## Analysis of Variance Table
##
## Model 1: mpg ~ am * wt + am * hp
## Model 2: mpg ~ am + cyl + wt + hp + am:cyl + am:wt + cyl:wt + am:hp +
##
       cyl:hp + am:cyl:wt + am:cyl:hp
     Res.Df
                RSS Df Sum of Sq
## 1
         22 105.872
         12 74.133 10
                            31.74 0.5138 0.8501
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am * wt + am * hp
## Model 2: mpg ~ am + gear + wt + hp + am:gear + am:wt + gear:wt + am:hp +
## gear:hp + am:gear:wt + am:gear:hp
## Res.Df RSS Df Sum of Sq F Pr(>F)
## 1 22 105.872
## 2 16 63.729 6 42.143 1.7634 0.1704
```

fit.wt.hp.gear <- update(best.fit, mpg ~ am\*gear\*wt + am\*gear\*hp)</pre>

anova(best.fit, fit.wt.hp.gear)

Adding of the extra predictors has no sensible impact according to p-values (gerater than 0.05). So the best model remains intact.

### Chosen Model Quality Analysis

```
summary(best.fit)
```

```
##
## Call:
## lm(formula = mpg ~ am * wt + am * hp, data = mtcars)
## Residuals:
      Min
               1Q Median
                               30
                                      Max
## -2.9793 -1.5338 -0.2435 1.1612 5.1652
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 29.28217
                          5.94151
                                    4.928 6.27e-05 ***
              15.92584
                          6.73612
                                    2.364 0.02731 *
## am1
## wt
              -1.26080
                          2.06659
                                   -0.610 0.54805
## hp
              -0.04515
                          0.01476
                                   -3.059 0.00575 **
## am1:wt
              -6.06647
                          2.72719
                                   -2.224 0.03669 *
               0.01707
                          0.02187
                                    0.781 0.44339
## am1:hp
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.194 on 22 degrees of freedom
## Multiple R-squared: 0.8757, Adjusted R-squared: 0.8474
## F-statistic: 30.98 on 5 and 22 DF, p-value: 2.955e-09
```

R-squared is very close to the 1, thus the most variance is explained by this model. Coefficients could be interpreted as follows:

1. Manual trans. intercept is greater than automatic trans. intercept by:

```
## [1] "15.926 MPG"
```

2. Change in MPG per one thousand lb automatic vs manual:

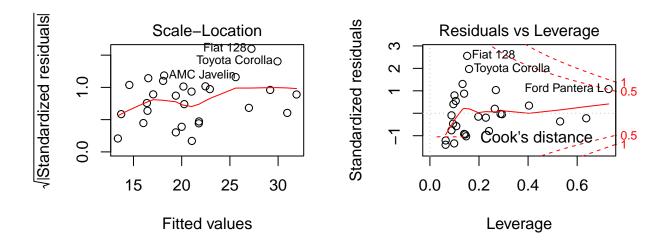
```
## [1] "-1.261 vs -7.327 MPG per 1000*lb"
```

3. Change in MPG per one horsepower automatic vs manual:

```
## [1] "-0.045 vs -0.028 MPG per horsepower"
```

4. Also according to the p-values of the wt and am1:hp coefficients. Null hypothesis about equality to zero could not be rejected.

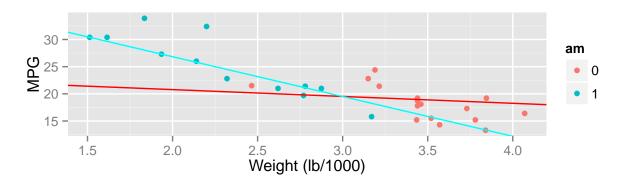
#### Residuals



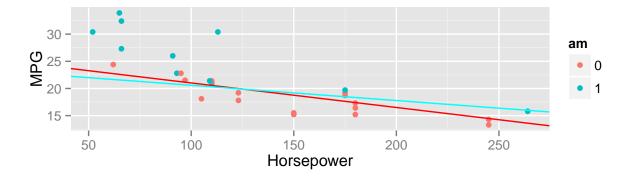
It seems that residuals do not form a pattern. So the conclusions could be made.

#### Conclusions

Greater MPG is better.



According to hypothetical relationship manual transmission is the best choice for lightwight cars, but automatic transmission is better for heavy cars. Thus nothing particular could be said. Also attention should be paid to the real points that form two clusters. Lightwieght cars with manual transmission and heavy cars with automatic transmission.



Hypothetical relationships between MPG and horsepower in case of automatic transmission is very close to the case of manual transmission.

So it's very hard to quantify the difference between automatic and manual transmission. At average they are very close to each over according to this model.

Thus t-test could be useful to quantify the difference at average.

```
test <- t.test(mpg ~ am, data = mtcars)
test</pre>
```

```
##
## Welch Two Sample t-test
##
## data: mpg by am
## t = -3.8331, df = 16.046, p-value = 0.00146
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -10.918977 -3.143523
## sample estimates:
## mean in group 0 mean in group 1
## 18.14375 25.17500
```

P-value is less than 0.05, so it's statistically significant difference between automatic and manual transmissions. Manual trasmission is better than autimatic. And change in MPG from manual to automatic could be estimated by the following confidence interval:

```
## [1] -10.918977 -3.143523
## attr(,"conf.level")
## [1] 0.95
centered at:
## [1] "-7.03125 Miles/(US) gallon"
```