

Automobile transmission type and fuel mileage

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```
## Warning: package 'knitr' was built under R version 3.2.5
```

Executive summary

Introduction

We analyze the fuel consumption of various automobiles from early 1970s in order to determine whether the transmission type, manual or automatic, significantly affects fuel consumption. The dataset is `mtcars` provided by the R `dataset` package, and it consists of mechanical and performance data on 32 different cars from the model years 1973 and 1974.

The aim of the analysis is to answer two questions: is an automatic or manual transmission better for gas mileage (measured in mpg, miles traveled per one US gallon of fuel consumed), and how the gas mileage is quantitatively affected by the transmission type.

Data description and exploratory analysis

```
library(datasets)
data("mtcars")

library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 3.2.4
```

```
library(GGally)
```

```
## Warning: package 'GGally' was built under R version 3.2.5
```

The data consists of design and performance data of 32, as published in the *Motor Trends* magazine published in the United States in 1974. According to the dataset documentation, the variables in the following table are included. Text in *italics* gives explanatory notes not present in the original data.

Column name	Variable description and units
mpg	Miles/(US) gallon (<i>gas mileage</i> ; $235 \text{ l}/100 \text{ km} = 1/(1 \text{ mpg})$)
cyl	Number of cylinders
disp	Displacement (cu.in.) ($1 \text{ litre} = 61.0 \text{ cu in}$)
hp	Gross horsepower ($1 \text{ kW} = 1.34 \text{ hp}$)
drat	Rear axle ratio (<i>driveshaft rpm/axle rpm</i> ¹)
wt	Weight (1000 lbs = 454 kg)
qsec	1/4 mile time
vs	V/S [<i>V engine (0) or straight (inline) engine (1)</i>]
am	Transmission (0 = automatic, 1 = manual)
gear	Number of forward gears
carb	Number of carburetors

The research question stated in the project assignment exclusively asks for analysis of the gas mileage (MPG), so for consistency reasons the data is not converted into SI units, even though the conversion would make the data more accessible to most parts of the world. Additionally, analysing the consumption via the amount of fuel consumed per fixed distance (e.g., litres/100 km) would be both physically and statistically more reasonable choice² than gas mileage. However, as the assignment explicitly asks for effect of transmission type on gas mileage, the data will not be converted for analysis. As suggested by Henderson and Velleman, a new variable **pwr** (hp/1000 lbs) is created for the power-to-weight ratio.

```
# Convert transmission type, (cylinder count, carburetor count), and engine type
# to factor variables
mtcars$am <- factor(mtcars$am, levels = c(0, 1),
                    labels = c("auto", "manual"))
#mtcars$cyl <- factor(mtcars$cyl)
#mtcars$carb <- factor(mtcars$carb)
mtcars$vs <- factor(mtcars$vs, levels = c(0, 1),
                    labels = c("v", "straight"))

# Create a new variable for gross power/weight ratio
# (hp/1000 lb; 1 hp/1000 lb 1.64 W/kg)
mtcars$pwr <- mtcars$hp/mtcars$wt

g <- ggplot(mtcars, aes(x = wt, y = mpg))
g <- g + geom_point(aes(color = am))
g <- g + geom_hline(aes(yintercept = mpg, color = am),
                    data = aggregate(mpg ~ am, data = mtcars, mean),
```

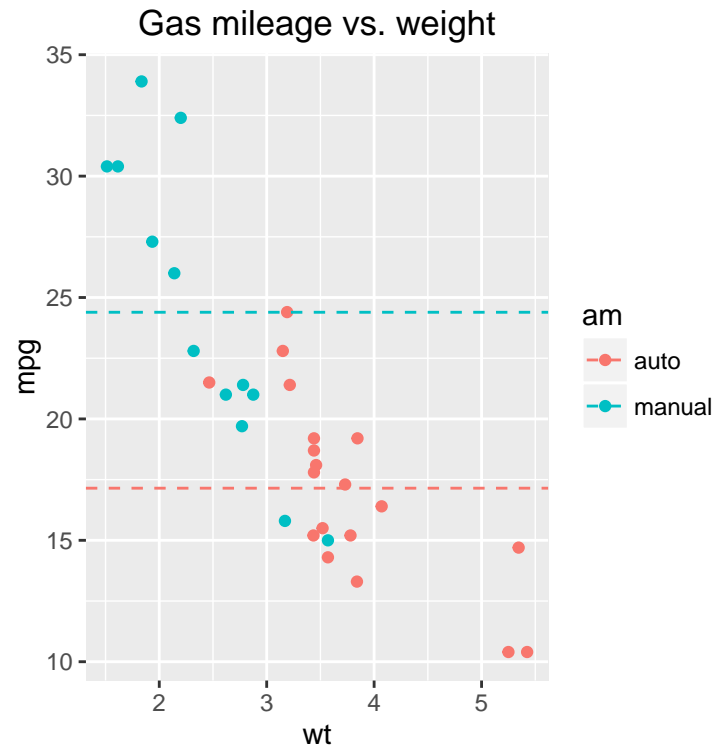
¹the drive-axle ratio is a comparison of the number of gear teeth on the ring gear of the rear axle and the pinion gear on the driveshaft. - - For example, a 4.11:1 ratio means there are 4.11 teeth on the axle's ring gear for each tooth on the driveshaft's pinion gear. Or, put another way, the driveshaft must turn 4.11 times to turn the rear wheels one full revolution. - - typical rule of thumb: The higher the numerical ratio, the slower the gear will be. This higher ratio gives a truck greater pulling power, but since the engine must work harder to spin the driveshaft more times for each turn of the rear wheels, top-end speed and fuel economy are sacrificed." From [worktruckonline.com][drat]

²As suggested also by Henderson and Velleman

```

linetype = "dashed")
g <- g + ggtitle("Gas mileage vs. weight")
print(g)

```



The plot shows that the mean gas mileage of cars with manual transmission is significantly higher (i.e., better) compared to cars with automatic transmission. However, the plot similarly illustrates that the gas mileage seems to have a clear negative correlation with the weight of the vehicle.

```

fit1 <- lm(mpg ~ am, mtcars)
fitall <- lm(mpg ~ . - hp, mtcars)
summary(fit1)

```

```

##
## Call:
## lm(formula = mpg ~ am, data = mtcars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.3923 -3.0923 -0.2974  3.2439  9.5077
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    17.147      1.125   15.247 1.13e-15 ***

```

```
## ammanual          7.245          1.764      4.106 0.000285 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.902 on 30 degrees of freedom
## Multiple R-squared:  0.3598, Adjusted R-squared:  0.3385
## F-statistic: 16.86 on 1 and 30 DF,  p-value: 0.000285
```

```
fit2.w <- lm(mpg ~ am + wt, mtcars)
fit2.d <- lm(mpg ~ am + disp, mtcars)
fit2.h <- lm(mpg ~ am + hp, mtcars) # greatest reduction in sum(~2)

anova(fit1, fit2.w)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + wt
##   Res.Df    RSS Df Sum of Sq      F       Pr(>F)
## 1      30 720.90
## 2      29 278.32  1    442.58 46.115 0.0000001867 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
anova(fit1, fit2.d)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + disp
##   Res.Df    RSS Df Sum of Sq      F       Pr(>F)
## 1      30 720.90
## 2      29 300.28  1    420.62 40.621 0.0000005748 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
anova(fit1, fit2.h)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + hp
##   Res.Df    RSS Df Sum of Sq      F       Pr(>F)
## 1      30 720.90
## 2      29 245.44  1    475.46 56.178 2.92e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
fit3.w <- lm(mpg ~ am + hp + wt, mtcars) # greatest reduction in sum(2)
fit3.d <- lm(mpg ~ am + hp + disp, mtcars)
fit3.c <- lm(mpg ~ am + hp + cyl, mtcars)
```

```
anova(fit1, fit2.h, fit3.w)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + hp
## Model 3: mpg ~ am + hp + wt
##   Res.Df    RSS Df Sum of Sq      F    Pr(>F)
## 1      30 720.90
## 2      29 245.44  1    475.46 73.841 2.445e-09 ***
## 3      28 180.29  1     65.15 10.118 0.003574 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
anova(fit1, fit2.h, fit3.d)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + hp
## Model 3: mpg ~ am + hp + disp
##   Res.Df    RSS Df Sum of Sq      F    Pr(>F)
## 1      30 720.90
## 2      29 245.44  1    475.46 58.8793 2.332e-08 ***
## 3      28 226.10  1     19.34  2.3945    0.133
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
anova(fit1, fit2.h, fit3.c)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + hp
## Model 3: mpg ~ am + hp + cyl
##   Res.Df    RSS Df Sum of Sq      F    Pr(>F)
## 1      30 720.90
## 2      29 245.44  1    475.46 60.3610 1.834e-08 ***
## 3      28 220.55  1     24.89  3.1594 0.08636 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
fit4.d <- lm(mpg ~ am + hp + wt + disp, mtcars)
fit4.c <- lm(mpg ~ am + hp + wt + cyl, mtcars) # smallest sum(2), but fit4.c not re
anova(fit1, fit2.h, fit3.w, fit4.d)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + hp
## Model 3: mpg ~ am + hp + wt
## Model 4: mpg ~ am + hp + wt + disp
##   Res.Df    RSS Df Sum of Sq      F    Pr(>F)
## 1      30 720.90
## 2      29 245.44  1    475.46 71.3552 4.646e-09 ***
## 3      28 180.29  1     65.15  9.7773  0.0042 **
## 4      27 179.91  1      0.38  0.0576  0.8122
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
anova(fit1, fit2.h, fit3.w, fit4.c)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + hp
## Model 3: mpg ~ am + hp + wt
## Model 4: mpg ~ am + hp + wt + cyl
##   Res.Df    RSS Df Sum of Sq      F    Pr(>F)
## 1      30 720.90
## 2      29 245.44  1    475.46 75.5148 2.638e-09 ***
## 3      28 180.29  1     65.15 10.3472 0.003356 **
## 4      27 170.00  1     10.29  1.6348 0.211917
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
fit5 <- lm(mpg ~ am + hp + wt + disp + cyl, mtcars)
anova(fit1, fit2.h, fit3.w, fit4.d, fit5)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + hp
## Model 3: mpg ~ am + hp + wt
## Model 4: mpg ~ am + hp + wt + disp
## Model 5: mpg ~ am + hp + wt + disp + cyl
```

```
##   Res.Df    RSS Df Sum of Sq      F    Pr(>F)
## 1      30 720.90
## 2      29 245.44  1    475.46 75.7841 3.499e-09 ***
## 3      28 180.29  1     65.15 10.3841 0.003408 **
## 4      27 179.91  1      0.38  0.0611 0.806673
## 5      26 163.12  1     16.79  2.6758 0.113932
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
anova(fit1, fit2.h, fit3.w, fit4.c, fit5)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + hp
## Model 3: mpg ~ am + hp + wt
## Model 4: mpg ~ am + hp + wt + cyl
## Model 5: mpg ~ am + hp + wt + disp + cyl
##   Res.Df    RSS Df Sum of Sq      F    Pr(>F)
## 1      30 720.90
## 2      29 245.44  1    475.46 75.7841 3.499e-09 ***
## 3      28 180.29  1     65.15 10.3841 0.003408 **
## 4      27 170.00  1     10.29  1.6407 0.211542
## 5      26 163.12  1      6.88  1.0963 0.304719
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Sources

- Hocking (1971) <http://www.jstor.org/stable/2529336>
- Henderson and Velleman (1981) <http://www.jstor.org/stable/2530428>

Appendix

```
paircols <- c("mpg", "disp", "wt", "qsec", "pwr")
ggpairs(mtcars, aes(color = am),
        columns = match(paircols, names(mtcars)))
```

