

Analysis of MPG for Automatic vs. Manual Transmission

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Regression Models - Course Project

Executive Summary

We will analyze the mtcars dataset collected by the magazine “Motor Trend”. This report will leverage methods of exploratory and regression analysis. The analysis will focus on the relationship between miles per gallon (MPG) and transmission type.

“Is an automatic or manual transmission better for MPG”

“Quantify the MPG difference between automatic and manual transmissions”

Manual transmissions get more miles per gallon mpg.

The ratio of this is 1.8 adjusted by hp, cyl, and wt.

Multivariable Regression

Regression will all available variables.

```
full_fit <- lm(mpg ~ . ,data=mtcars)
best_fit <- step(full_fit, direction = "both", trace = FALSE)
```

Step method runs lm multiple times to build multiple regression models and select the best variables from them using both forward selection and backward elimination methods by the AIC algorithm.

```
summary(best_fit)
```

```
##
## Call:
## lm(formula = mpg ~ cyl + hp + wt + am, data = mtcars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.9387 -1.2560 -0.4013  1.1253  5.0513
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  33.70832    2.60489   12.940 7.73e-13 ***
## cyl6         -3.03134    1.40728   -2.154  0.04068 *
## cyl8         -2.16368    2.28425   -0.947  0.35225
## hp           -0.03211    0.01369   -2.345  0.02693 *
## wt           -2.49683    0.88559   -2.819  0.00908 **
## amManual      1.80921    1.39630    1.296  0.20646
```

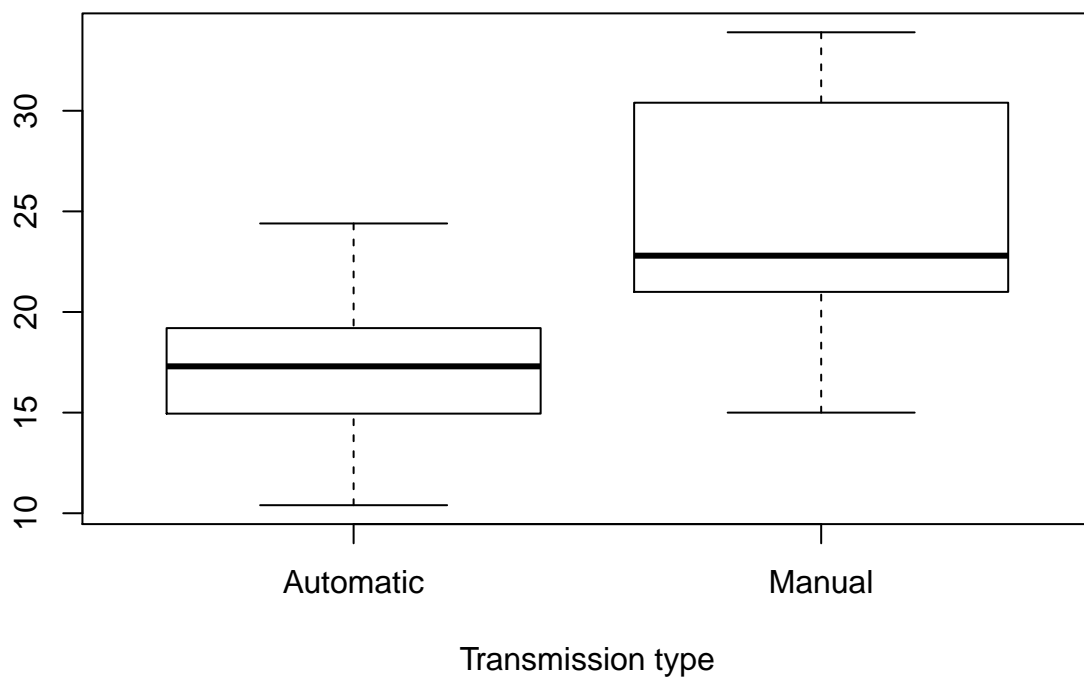
```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.41 on 26 degrees of freedom
## Multiple R-squared:  0.8659, Adjusted R-squared:  0.8401
## F-statistic: 33.57 on 5 and 26 DF,  p-value: 1.506e-10
```

We conclude that hp, wt, and cyl are confounding variables in the relationship between 'am' and 'mpg'. With 84% of the variance covered by this model it's a good fit (see analysis in appendix for further validation)

Appendix

Exploratory analysis

```
boxplot(mpg ~ am, data = mtcars, xlab = "Transmission type")
```



Clear separation of means between Automatic and Manual transmissions.

Inference

```
alpha <- 0.05
fit <- lm(mpg ~ am, data = mtcars)
summary(fit)
```

```
##
## 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
```

Simple Linear Regression: Adjusted R squared value is only 0.338 (or 33.8%) of the regression variance can be explained by our model. This indicates we should look into the other variables in a multivariate regression and potential confounding relationships.

```
analysis <- aov(mpg ~ ., data = mtcars)
summary(analysis)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## cyl           2  824.8   412.4   51.377 1.94e-07 ***
## disp          1   57.6    57.6    7.181  0.0171 *
## hp            1   18.5    18.5    2.305  0.1497
## drat          1   11.9    11.9    1.484  0.2419
## wt            1   55.8    55.8    6.950  0.0187 *
## qsec          1    1.5     1.5    0.190  0.6692
## vs            1    0.3     0.3    0.038  0.8488
## am            1   16.6    16.6    2.064  0.1714
## gear          2    5.0     2.5    0.313  0.7361
## carb          5   13.6     2.7    0.339  0.8814
## Residuals    15  120.4     8.0
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Using analysis of variance we see potential impact on mpg from am but higher on a number of other variables.

```
t.test(mpg ~ am, data = mtcars)
```

```
##
## Welch Two Sample t-test
##
## data:  mpg by am
## t = -3.7671, df = 18.332, p-value = 0.001374
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
```

```
## -11.280194 -3.209684
## sample estimates:
## mean in group Automatic    mean in group Manual
##                17.14737                24.39231
```

Transmission types are clearly different with respect to their relationship to mpg.

Residual Analysis and Diagnosis

The p-value analysis between simple regression and best allow us to reject the null hypothesis that confounding variable impact the relation between am and mpg. The residual analysis plots below validate the model as well.

```
anova(fit, best_fit)
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ cyl + hp + wt + am
##   Res.Df    RSS Df Sum of Sq    F    Pr(>F)
## 1      30 720.90
## 2      26 151.03  4    569.87 24.527 1.688e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
par(mfrow = c(2, 2))
plot(best_fit)
```

