# Regression Models Course Project

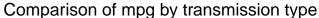
Teo Lo Piparo

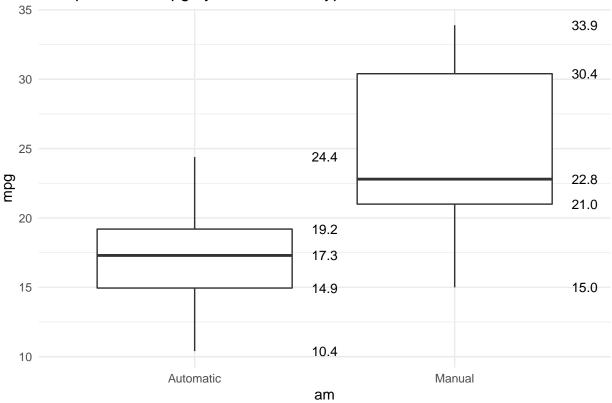
20/11/2020

```
library(tidyverse)
library(ggplot2)
library(broom)
```

### **Exploratory Analysis**

### **Boxplot**





• Manual transmissions, generally perform better in terms of miles/gallon, compared to Manual transmissions.

#### Quantified difference

```
mtcars %>%
    select(mpg, am) %>%
    group_by(am) %>%
    summarise("mu.mpg" = mean(mpg))

## # A tibble: 2 x 2
##    am mu.mpg
## <dbl> <dbl>
## 1    0    17.1
## 2    1    24.4
```

- For Automatic transmissions, the car has a mean of  ${\sim}17$  miles/gallon
- For Manual transmissions, the car has a mean of ~24 miles/gallon

### Regression analysis

#### Linear model for all variables

```
mtcars %>%
  mutate(am = factor(am, levels = c(0,1))) %>%
  group_by(am) %>%
  do(broom::tidy(lm(mpg ~ . -1, data = mtcars))) ## wt and qsec appear stat significant
```

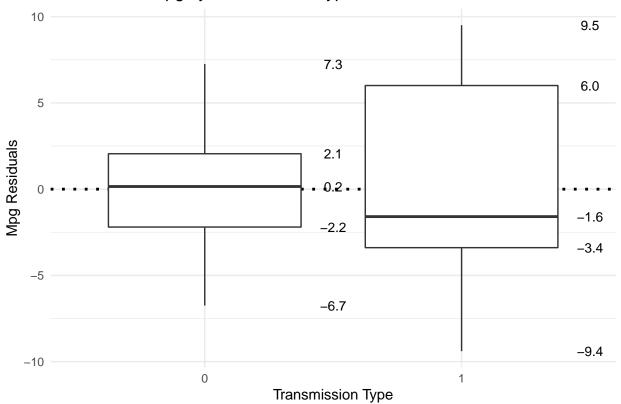
```
## # A tibble: 20 x 6
               am [2]
## # Groups:
##
            term
                   estimate std.error statistic p.value
##
      <fct> <chr>
                      <dbl>
                                 <dbl>
                                           <dbl>
                                                    <dbl>
##
    1 0
            cyl
                     0.351
                                0.763
                                          0.460
                                                   0.650
##
    2 0
                     0.0135
                                0.0176
                                          0.768
                                                   0.450
            disp
   3 0
                    -0.0205
                                0.0214
                                         -0.958
                                                   0.348
            hp
## 4 0
                     1.24
                                                   0.405
            drat
                                1.46
                                          0.849
## 5 0
            wt
                    -3.83
                                1.86
                                         -2.05
                                                   0.0520
##
  6 0
            qsec
                     1.19
                                0.459
                                          2.59
                                                   0.0166
  7 0
            vs
                     0.190
                                2.07
                                          0.0917 0.928
## 8 0
                     2.83
                                1.98
                                          1.43
                                                   0.166
## 9 0
                     1.05
                                1.35
                                          0.783
                                                   0.442
            gear
## 10 0
                                         -0.324
            carb
                    -0.263
                                0.812
                                                   0.749
## 11 1
                     0.351
                                0.763
                                          0.460
                                                   0.650
            cyl
## 12 1
            disp
                     0.0135
                                0.0176
                                          0.768
                                                   0.450
## 13 1
                    -0.0205
                                0.0214
                                         -0.958
                                                   0.348
            hp
## 14 1
                     1.24
                                1.46
                                          0.849
                                                   0.405
            drat
## 15 1
                    -3.83
                                1.86
                                         -2.05
                                                   0.0520
            wt
## 16 1
            qsec
                     1.19
                                0.459
                                          2.59
                                                   0.0166
## 17 1
            vs
                     0.190
                                2.07
                                          0.0917 0.928
## 18 1
                     2.83
                                1.98
                                          1.43
                                                   0.166
            am
## 19 1
                     1.05
                                1.35
                                          0.783
                                                   0.442
            gear
## 20 1
                    -0.263
                                0.812
                                         -0.324
                                                   0.749
            carb
```

## Both variables are equally changing regardless the transmission type.

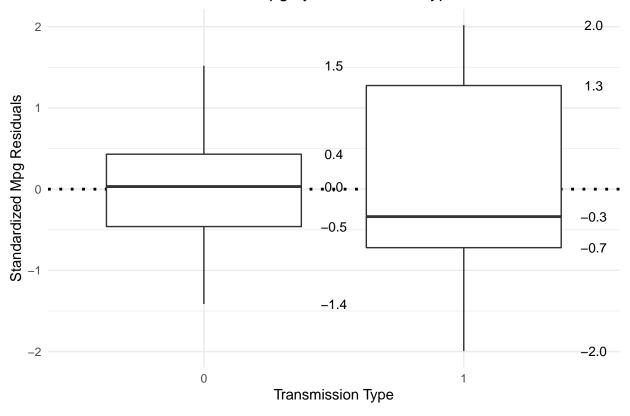
- For each increase of 1000 lbs the car decrease the miles/gallon by 3.83
- For each increase of sec necessary to travel 1/4 mile the miles/gallon increase by 1.19.

#### Residual analysis

## Residuals of Mpg by Transmission Type



### Standardized Residuals of Mpg by Transmission Type



• The standardized residual plot shows that the manual transmission presents a broader range of outliers and its mean value is not aligned with the mean value of the model, which indicate that the model accountability is way less reliable compared to automatic transmissions.

```
Nested lineam models with ANOVA

fit1 <- lm(mpg ~ factor(am) + factor(gear) -1, mtcars)
fit2 <- lm(mpg ~ factor(am) + factor(gear) + factor(carb) + factor(vs) -1, mtcars)
fit3 <- lm(mpg ~ factor(am) + factor(gear) + factor(carb) + factor(vs) + qsec + wt -1, mtcars)
fit4 <- lm(mpg ~ factor(am) + factor(gear) + factor(carb) + factor(vs) + qsec + wt + drat + hp -1, mtcar
fit5 <- lm(mpg ~ factor(am) + factor(gear) + factor(carb) + factor(vs) + qsec + wt + drat + hp + disp +
anova(fit1, fit2, fit3, fit4, fit5)

## Analysis of Variance Table
##
## Model 1: mpg ~ factor(am) + factor(gear) - 1
## Model 2: mpg ~ factor(am) + factor(gear) + factor(carb) + factor(vs) -
## 1
## Model 3: mpg ~ factor(am) + factor(gear) + factor(carb) + factor(vs) +
```

## Model 4: mpg ~ factor(am) + factor(gear) + factor(carb) + factor(vs) +

## Model 5: mpg ~ factor(am) + factor(gear) + factor(carb) + factor(vs) +

F

qsec + wt + drat + hp + disp + factor(cyl) - 1

##

##

##

Res.Df

qsec + wt - 1

qsec + wt + drat + hp - 1

RSS Df Sum of Sq

```
## 1
        28 570.00
## 2
        22 196.30 6
                        373.70 7.7594 0.0006273 ***
## 3
        20 155.44 2
                         40.87 2.5456 0.1117185
         18 144.18 2
## 4
                         11.26 0.7014 0.5114585
## 5
         15 120.40 3
                         23.77 0.9873 0.4252633
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## The linear models are not responding very well with inflated factorial variables. However, the conti
fit0 <- lm(mpg ~ factor(am) + wt + qsec -1, mtcars)
as.data.frame(summary(fit0)$coef) %>%
    select(Estimate, `Pr(>|t|)`) %>%
   filter(`Pr(>|t|)` <= 0.05)
##
                            Pr(>|t|)
               Estimate
## factor(am)1 12.553618 4.754335e-02
              -3.916504 6.952711e-06
```

• Transmission type is impacting the miles/gallon but not more than the weight variable and the 1/4 mile time variable.

#### **Predicted Probability**

1.225886 2.161737e-04

## qsec

```
fit1.2 <- glm(am ~ mpg + wt + qsec - 1, mtcars, family = "binomial")
am.predict = data.frame(mpg = mean(mtcars$mpg), wt=mean(mtcars$wt), qsec=mean(mtcars$qsec))
round(predict(fit1.2, am.predict, type="response"),2)
## 1
## 0.4</pre>
```

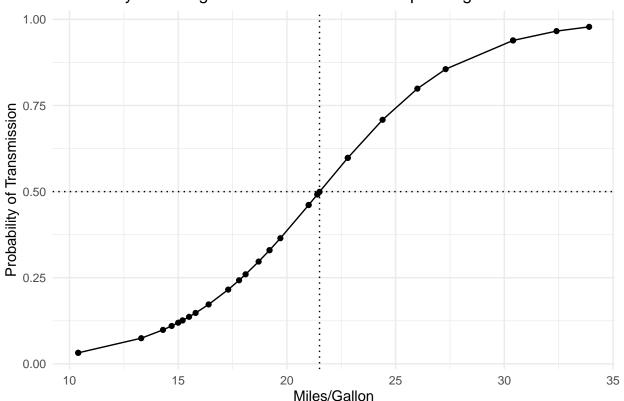
• Given a car with an average Miles/Gallon, an average Weight and an average 1/4 Miles Time the percentage that a car will have a manual transmission is estimated to be  $\sim 40\%$ 

#### Generalized linear model

```
fit1.1 <- glm(am ~ mpg, mtcars, family = "binomial")</pre>
summary(fit1.1)
##
## Call:
## glm(formula = am ~ mpg, family = "binomial", data = mtcars)
##
## Deviance Residuals:
      Min
##
                 10
                      Median
                                   3Q
                                           Max
## -1.5701 -0.7531 -0.4245
                               0.5866
                                        2.0617
##
## Coefficients:
              Estimate Std. Error z value Pr(>|z|)
##
## (Intercept) -6.6035
                            2.3514 -2.808 0.00498 **
## mpg
                 0.3070
                            0.1148
                                     2.673 0.00751 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
```

```
## (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 43.230 on 31 degrees of freedom
##
## Residual deviance: 29.675 on 30 degrees of freedom
## AIC: 33.675
##
## Number of Fisher Scoring iterations: 5
exp(fit1.1$coef)
## (Intercept)
                       mpg
## 0.001355579 1.359379288
exp(confint(fit1.1))
##
                      2.5 %
                                97.5 %
## (Intercept) 4.425443e-06 0.06255158
               1.129764e+00 1.79946863
ggplot(data = mtcars, aes(x=mpg, y=fit1.1$fitted.values)) +
   geom_line() +
   geom_point() +
   theme_minimal() +
   xlab(label = "Miles/Gallon") + ylab(label = "Probability of Transmission") +
    ggtitle(label = "Probability of having a Manual transmission depending on no. of MPG") +
   geom_hline(yintercept = 0.5, linetype = "dotted") +
    geom_vline(xintercept = 21.5, linetype = "dotted")
```

# Probability of having a Manual transmission depending on no. of MPG



 $\bullet$  Above 21.5 miles/gallon there is more probability that the car has a manual transmission and below there is more probability that the car is automatic