Regression Models Course Project

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Peer Graded Assignment: Regression Models Coursera Project

Answer the Following Questions:

1) "Is an automatic or manual transmission better for MPG"

The manual vehicle is better due to a p-value of .0006 from a two sample t-test in the means of the two samples.

2) "Quantify the MPG difference between automatic and manual transmissions"

We are 95% confident the true difference in the means of manual and automatic transmissions lies between 3.913 MPG and Positive Infinity.

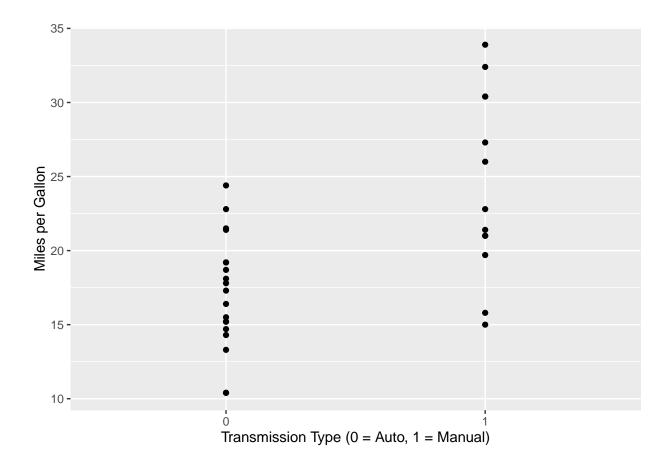
Step 1: Load the Data into R

```
library(datasets)
cars <- datasets::mtcars
head(cars)</pre>
```

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

Step 2: Exploratory Data Analysis

Now that we have the data loaded into R, it's time to graph the mpg based on automatic and manual transmissions.



As we can see, the engines of manual vehicles run more efficiently based on MPG than their automatic counterparts. To solidify this notion, we will run a two sample T test to see if the means of these two groups are in fact different.

Step 3: Split data by manual and automatic transmission type and find the mean of each groups MPG's.

```
autos <- subset(cars, am == "0")
mans <- subset(cars, am == "1")
mean(autos$mpg)

## [1] 17.14737

mean(mans$mpg)</pre>
```

Step 4: Check criteria to see if two-sample t-test is appropriate.

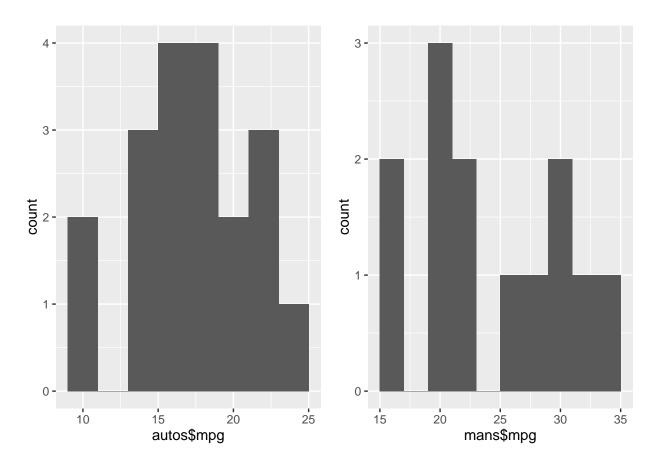
Criteria #1: Samples are Drawn from Normal distributions

[1] 24.39231

require(gridExtra)

Loading required package: gridExtra

```
plot1 <- qplot(autos$mpg, binwidth = 2)
plot2 <- qplot(mans$mpg, binwidth = 2)
grid.arrange(plot1, plot2, ncol=2)</pre>
```



Criteria #2: Samples must be independent.

Since the probability of being a manual car does not affect the probability of being an automatic vehicle, this condition is met.

Step 5: T-test

```
t.test(mans$mpg, autos$mpg, alternative = "greater")

##
## Welch Two Sample t-test
##
## data: mans$mpg and autos$mpg
## t = 3.7671, df = 18.332, p-value = 0.0006868
```

```
## alternative hypothesis: true difference in means is greater than 0
## 95 percent confidence interval:
## 3.913256    Inf
## sample estimates:
## mean of x mean of y
## 24.39231 17.14737
```

With a p-value of 0.0006, we can conclude: there is sufficient evidence that manual transmissions consume less gasoline and are more fuel efficient. There is evidence against the hypothesis that the means of the two groups are the same. Finally, to quantify this difference between the two types of vehicles, we can use our t-test to create a confidence interval. Recalling that our interval was (3.913, + Inf), we are 95% confident the true difference lies within these bounds.

```
fit <- lm(mpg ~ am, data = cars)
summary(fit)
##
## Call:
## lm(formula = mpg ~ am, data = cars)
##
## Residuals:
##
               10 Median
                               ЗQ
                                      Max
  -9.3923 -3.0923 -0.2974 3.2439
                                  9.5077
##
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
                            1.125 15.247 1.13e-15 ***
## (Intercept)
                17.147
                 7.245
                            1.764
                                   4.106 0.000285 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 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
```

Now that we see the R² valus is around 36% our goal is to find a model where a higher percent of variance can be explained.

```
library(car)
fit_all <- lm(mpg ~ ., cars)</pre>
vif(fit_all)
##
                   disp
                                        drat
         cyl
                               hp
                                                              qsec
## 15.373833 21.620241
                         9.832037
                                   3.374620 15.164887 7.527958 4.965873
          am
                   gear
    4.648487 5.357452 7.908747
sqrt(vif(fit_all))
##
        cyl
                 disp
                            hp
                                    drat
                                                wt
                                                       qsec
                                                                   VS
                                                                             am
```

```
## 3.920948 4.649757 3.135608 1.837014 3.894212 2.743712 2.228424 2.156035 ## gear carb ## 2.314617 2.812249
```

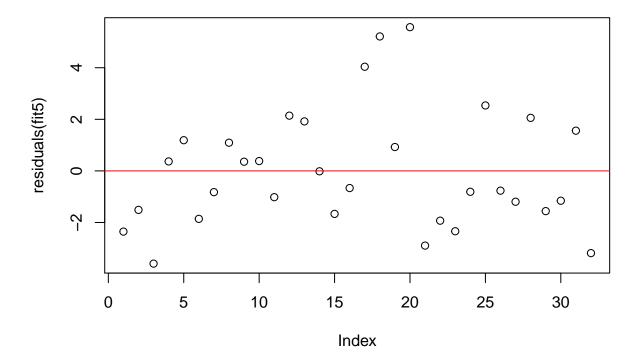
With the standard deviations of inflation factors stated. We will take the highest and update out model systematically to see which model will best predict MPG of vehicles and have the high R^2 value.

```
fit1 <- lm(mpg ~ am, data = cars)
fit2 <- update(fit1, mpg ~ am + cyl)
fit3 <- update(fit1, mpg ~ am + cyl + disp)
fit4 <- update(fit1, mpg ~ am + cyl + disp + hp)
fit5 <- update(fit1, mpg ~ am + cyl + disp + hp + wt)
anova(fit1, fit2, fit3, fit4, fit5)</pre>
```

```
## Analysis of Variance Table
##
## Model 1: mpg ~ am
## Model 2: mpg ~ am + cyl
## Model 3: mpg ~ am + cyl + disp
## Model 4: mpg ~ am + cyl + disp + hp
## Model 5: mpg ~ am + cyl + disp + hp + wt
    Res.Df
##
              RSS Df Sum of Sq
                                          Pr(>F)
## 1
        30 720.90
        29 271.36 1
## 2
                        449.53 71.6522 6.037e-09 ***
        28 252.08 1
                         19.28 3.0732 0.091376 .
## 3
## 4
        27 216.37 1
                         35.71 5.6925 0.024609 *
## 5
        26 163.12 1
                         53.25 8.4872 0.007257 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Based off the p-value from the F-statistics, Model 5 and Model 2 end up being the most statistically significant. With that being said, of the two values, a look at the RSS values yields Model 5 as the better model in this particular case. Finally, to make sure our model is not flawed, let's look at the residuals to see if there is a pattern.

```
plot(residuals(fit5))
abline(h = 0, col = "red")
```



Since there is no pattern of consistently over estimating or underestimating the residuals, this model appears ready for action.