

# MtCars\_Prj

*Vijay Pulijala*

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## Executive Summary

Given the data provided in 'mtcars', a linear regression model was calculated using correlation, analysis of predictor variance, and significant linear regression coefficients.

The final resultant model found transmission type to be an insignificant predictor for MPG, therefore no confident conclusions can be made regarding transmission type's correlation with MPG, nor could quantified relationships between the two variables be calculated.

Instead, the only statistically significant parameters – given the provided data – were found to be 'number of cylinders' and 'weight', by which the formula holds true:

```
mpg = 39.6863 + (-1.5078)cyl + (-3.1910)wt
```

## Exploratory Analysis

We're interested in how data variables affect MPG. That said, let's take a quick look at the data & the correlations between the other variables and MPG:

```
##      cyl      disp      hp      drat      wt      qsec
## -0.8521620 -0.8475514 -0.7761684  0.6811719 -0.8676594  0.4186840
##      vs      am      gear      carb
##  0.6640389  0.5998324  0.4802848 -0.5509251
```

Number of cylinders, displacement, and weight have the strongest correlations with MPG. Let's quantify those relationships further.

## Analysis of the Variance of the Predictors

```
##      Df Sum Sq Mean Sq F value    Pr(>F)
## cyl      1   817.7    817.7 116.425 5.03e-10 ***
## disp      1    37.6     37.6   5.353 0.03091 *
## hp        1     9.4      9.4   1.334 0.26103
## drat      1    16.5     16.5   2.345 0.14064
## wt        1    77.5     77.5  11.031 0.00324 **
## qsec      1     3.9      3.9   0.562 0.46166
## vs        1     0.1      0.1   0.018 0.89317
## am        1    14.5     14.5   2.061 0.16586
## gear      1     1.0      1.0   0.138 0.71365
## carb      1     0.4      0.4   0.058 0.81218
## Residuals 21   147.5      7.0
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Analysis of the variance of the predictors as they relate to MPG further identifies engine configuration (vs) as the relatively least significant regressor. Let's remove it and reassess our pending model. We'll iteratively continue this until all predictors are significant. Then, we'll calculate coefficients.

## Linear Regression Model

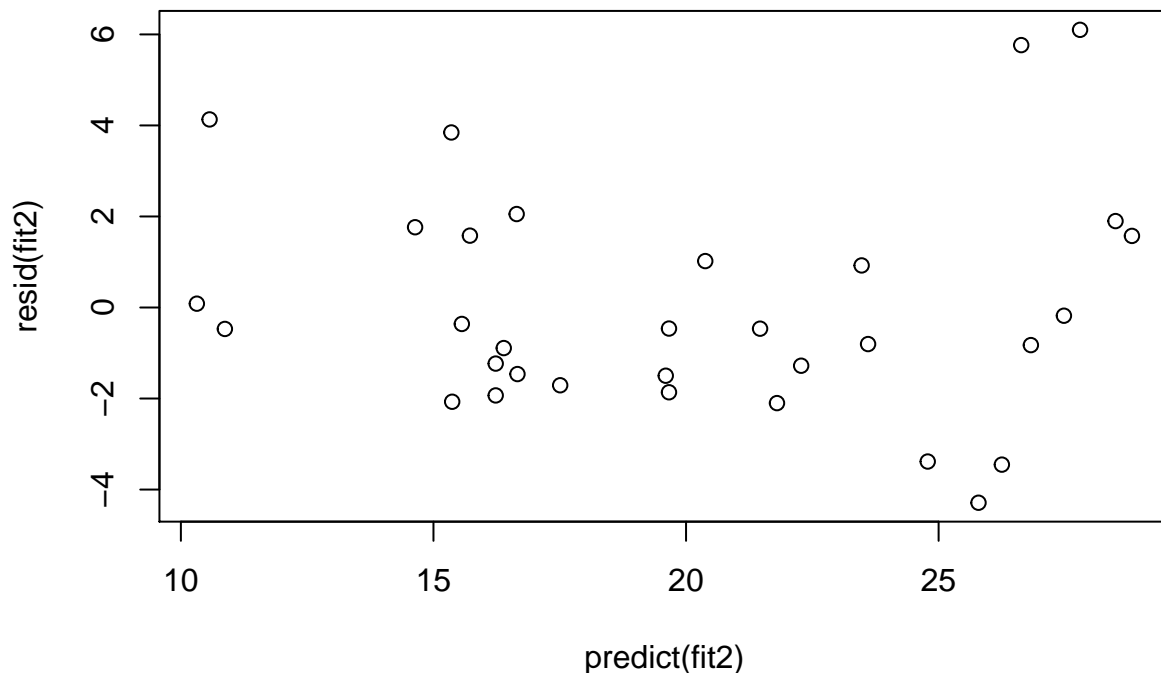
```
##           Estimate Std. Error   t value    Pr(>|t|)
## (Intercept) 41.107677641 2.84242604 14.4621802 1.620381e-14
## cyl        -1.784943519 0.60711048 -2.9400638 6.511676e-03
## disp        0.007472925 0.01184472  0.6309079 5.332173e-01
## wt         -3.635677016 1.04013753 -3.4953811 1.595519e-03
```

With this model, disp becomes insignificant once again, therefore we reduce the data again and reevaluate:

```
##           Estimate Std. Error   t value    Pr(>|t|)
## (Intercept) 39.686261  1.7149840 23.140893 3.043182e-20
## cyl        -1.507795  0.4146883 -3.635972 1.064282e-03
## wt         -3.190972  0.7569065 -4.215808 2.220200e-04
```

## Residuals

Plotting the residuals produces no obvious patterns:



Therefore, we have a linear model to help answer the pending questions:

1. Is an automatic or manual transmission better for MPG?

\_Given the transmission type does not have a significant impact on MPG.

1. Quantify the MPG difference between automatic and manual transmissions?

\_Unable to quantify this regressors impact because it's insignificant by the calculated model

## Appendix A

### Original context:

You 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?

## Appendix B

### Output of summary data:

```
summary(mtcars)
```

##	mpg	cyl	disp	hp
##	Min. :10.40	Min. :4.000	Min. : 71.1	Min. : 52.0
##	1st Qu.:15.43	1st Qu.:4.000	1st Qu.:120.8	1st Qu.: 96.5
##	Median :19.20	Median :6.000	Median :196.3	Median :123.0
##	Mean :20.09	Mean :6.188	Mean :230.7	Mean :146.7
##	3rd Qu.:22.80	3rd Qu.:8.000	3rd Qu.:326.0	3rd Qu.:180.0
##	Max. :33.90	Max. :8.000	Max. :472.0	Max. :335.0
##	drat	wt	qsec	vs
##	Min. :2.760	Min. :1.513	Min. :14.50	Min. :0.0000
##	1st Qu.:3.080	1st Qu.:2.581	1st Qu.:16.89	1st Qu.:0.0000
##	Median :3.695	Median :3.325	Median :17.71	Median :0.0000
##	Mean :3.597	Mean :3.217	Mean :17.85	Mean :0.4375
##	3rd Qu.:3.920	3rd Qu.:3.610	3rd Qu.:18.90	3rd Qu.:1.0000
##	Max. :4.930	Max. :5.424	Max. :22.90	Max. :1.0000
##	am	gear	carb	
##	Min. :0.0000	Min. :3.000	Min. :1.000	
##	1st Qu.:0.0000	1st Qu.:3.000	1st Qu.:2.000	
##	Median :0.0000	Median :4.000	Median :2.000	
##	Mean :0.4062	Mean :3.688	Mean :2.812	
##	3rd Qu.:1.0000	3rd Qu.:4.000	3rd Qu.:4.000	
##	Max. :1.0000	Max. :5.000	Max. :8.000	

## Appendix C

All code used in this report:

```
library(datasets)
data(mtcars)
cor(mtcars, mtcars$mpg)[-1,]
summary(aov(mpg ~ ., data=mtcars))

fit1 <- lm(mtcars$mpg ~ cyl + disp + wt, data=subset(mtcars, select=c(cyl, disp, wt)))
summary(fit1)$coefficients

fit2 <- lm(mtcars$mpg ~ cyl + wt, data=subset(mtcars, select=c(cyl, wt)))
summary(fit2)$coefficients
plot(resid(fit2) ~ predict(fit2))
```

## Appendix D

Confidence interval example:

```
coef <- summary(fit2)$coefficients
int95cyl <- (coef['cyl', 1] + c(-1, 1) * qt(0.975, df=fit2$df) * coef['cyl', 2]) * 2
int95wt <- coef['wt', 1] + c(-1, 1) * qt(0.975, df=fit2$df) * coef['wt', 2]
```

At 95% confidence, for each increase in cylinders (by 2), we find between 1.32 and 4.71 decrease in MPG.

At 95% confidence, for each 1000lb increase weight, we find between 1.64 and 4.74 decrease in MPG.