

# Regression Modeling Course Project

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Summary: This analysis tries to find answers to following questions:

1 - Is an automatic or manual transmission better for MPG? 2 - Quantify the MPG difference between automatic and manual transmissions

The data was extracted from the 1974 Motor Trend US magazine, and comprises fuel consumption and 10 aspects of automobile design and performance for 32 automobiles (1973–74 models).

## Initializing

```
library(dplyr)

##
## Attaching package: 'dplyr'
##
## The following objects are masked from 'package:stats':
##
##   filter, lag
##
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

library(datasets)
data(mtcars)
```

Our dataset consists of 11 variables:

[, 1] mpg Miles/(US) gallon [, 2] cyl Number of cylinders [, 3] disp Displacement (cu.in.) [, 4] hp Gross horsepower [, 5] drat Rear axle ratio [, 6] wt Weight (1000 lbs) [, 7] qsec 1/4 mile time [, 8] vs Engine (0 = V-shaped, 1 = straight) [, 9] am Transmission (0 = automatic, 1 = manual) [,10] gear Number of forward gears [,11] carb Number of carburetors

```
mtcars$am <- as.factor(mtcars$am)
mtcars$vs <- as.factor(mtcars$vs)
mtcars$gear <- as.factor(mtcars$gear)
mtcars$cyl <- as.factor(mtcars$cyl)
head(mtcars)
```

```
##           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     4     4
## Mazda RX4 Wag  21.0    6  160  110  3.90  2.875  17.02  0   1     4     4
## Datsun 710     22.8    4  108   93  3.85  2.320  18.61  1   1     4     1
## Hornet 4 Drive  21.4    6  258  110  3.08  3.215  19.44  1   0     3     1
## Hornet Sportabout 18.7    8  360  175  3.15  3.440  17.02  0   0     3     2
## Valiant        18.1    6  225  105  2.76  3.460  20.22  1   0     3     1
```

## Exploratory Data Analysis

Checking out if the automatic and manual transmission groups have different mean miles per gallon

```
auto <- mtcars %>% filter(am == "0" )
manual <- mtcars %>% filter(am == "1")
summary(auto$mpg)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    10.40   14.95   17.30   17.15   19.20   24.40
```

```
summary(manual$mpg)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    15.00   21.00   22.80   24.39   30.40   33.90
```

T-Test

```
t.test(manual$mpg, auto$mpg, paired = F, var.equal = F)
```

```
##
##  Welch Two Sample t-test
##
## data:  manual$mpg and auto$mpg
## 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:
##   3.209684 11.280194
## sample estimates:
## mean of x mean of y
##  24.39231  17.14737
```

Our T-test gives that the true difference in means miles per gallon between manual and automatic transmission is not equal to 0.

## Linear Regression

Initial model with one variable:

```
fit1 = lm(mpg ~ am, data= mtcars)
summary(fit1)$coef
```

```
##              Estimate Std. Error  t value    Pr(>|t|)
## (Intercept) 17.147368   1.124603 15.247492 1.133983e-15
## am1         7.244939   1.764422  4.106127 2.850207e-04
```

The transmission effect seems significant at first glance. According to our initial model, it seems that switching from automatic to manual transmission increases the range of the car by 7.24 miles per gallon. But we need to analyze for possible confounders.

The most likely possible confounders that may affect the mpg are weight and hp. We also need to check for others in general.

```
fit1.1 = lm(mpg ~ am + hp, data = mtcars)
summary(fit1.1)$coef
```

```
##              Estimate Std. Error  t value    Pr(>|t|)
## (Intercept) 26.5849137 1.425094292 18.654845 1.073954e-17
```

```
## am1          5.2770853 1.079540576 4.888270 3.460318e-05
## hp          -0.0588878 0.007856745 -7.495191 2.920375e-08
```

```
fit1.2 = lm(mpg ~ am + hp + wt, data = mtcars)
fit1.3 = lm(mpg ~ am + hp + wt + cyl, data = mtcars)
anova(fit1, fit1.1, fit1.2, fit1.3)
```

```
## 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 81.8529 1.634e-09 ***
```

```
## 3      28 180.29  1     65.15 11.2157 0.002484 **
```

```
## 4      26 151.03  2     29.27  2.5191 0.099998 .
```

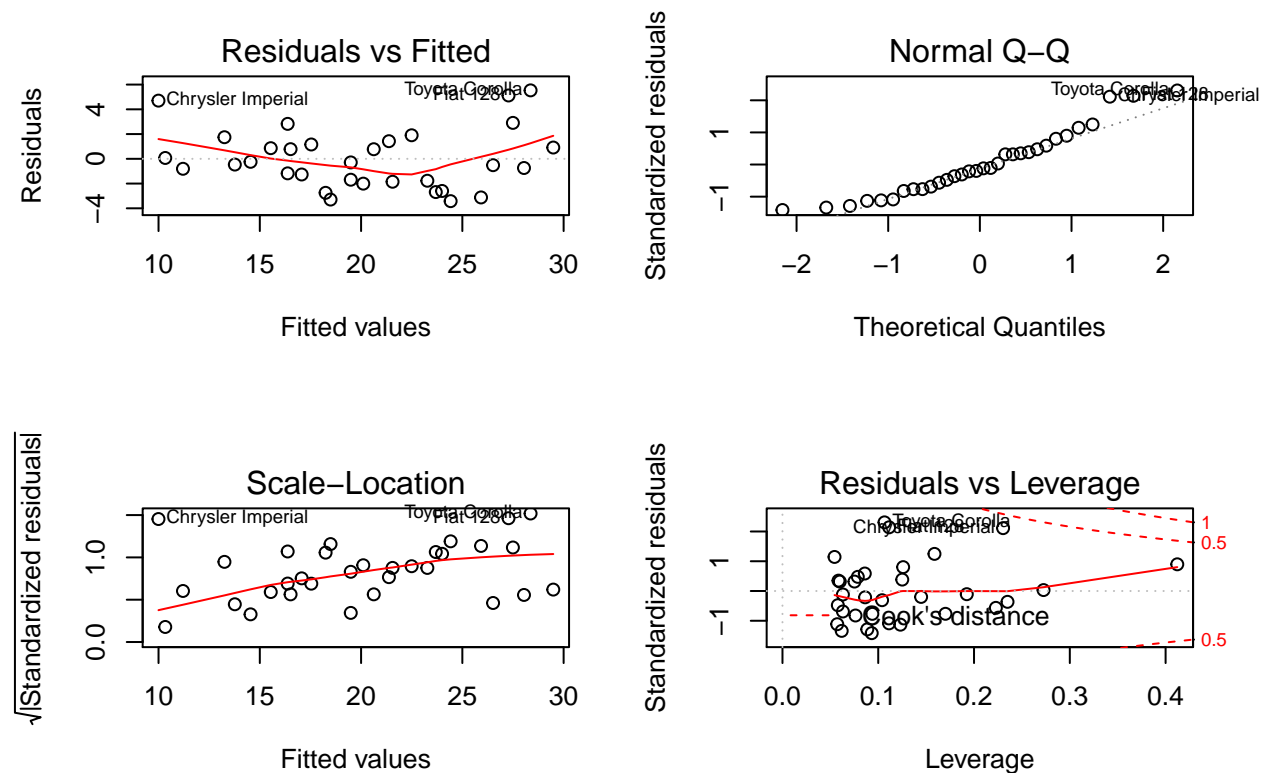
```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

comparing the nested models, it seems the cylinder variable is redundant. However, hp and wt variables influence the residual variance significantly.

Let's check the residuals for patterns for the hp and wt variables

```
par(mfrow = c(2,2))
plot(fit1.2)
```



It seems there is some non-linearity. We solve this problem by modeling the interaction between hp and wt variables.

```

fit1.4 = lm(mpg ~ am + hp*wt, data = mtcars)
summary(fit1.4)

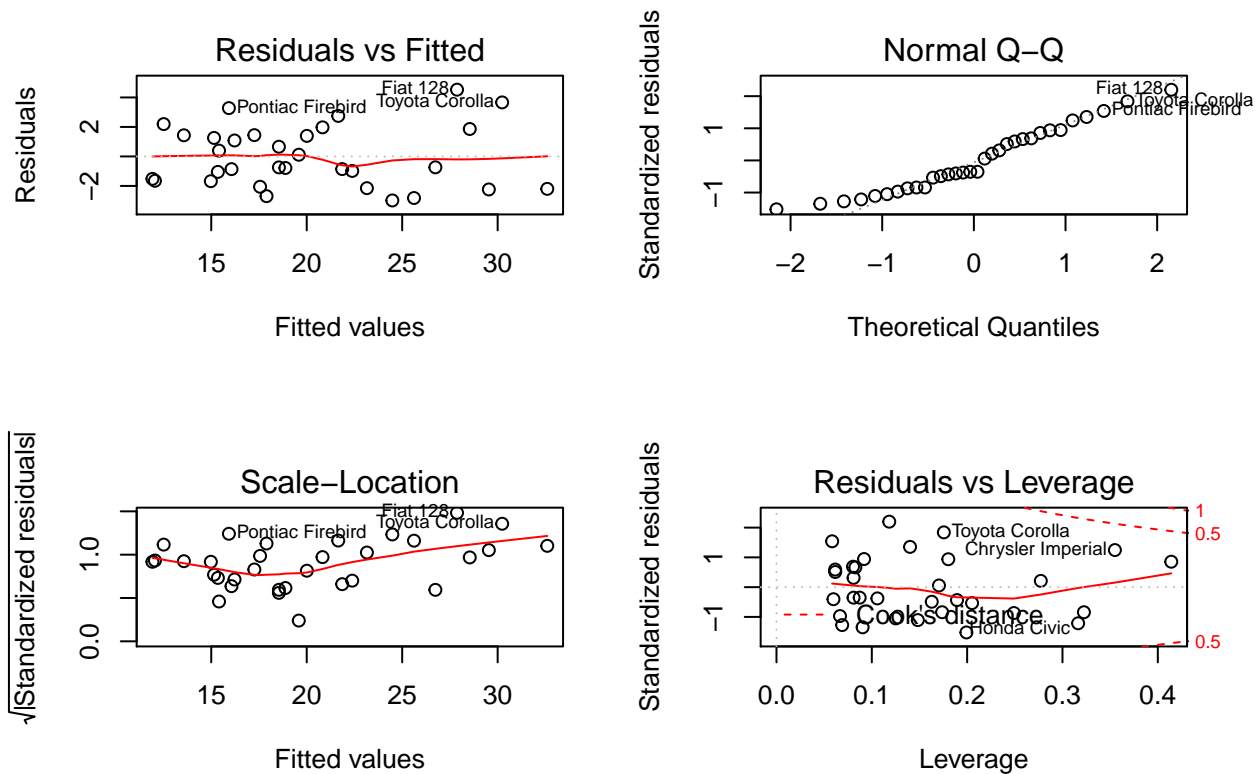
##
## Call:
## lm(formula = mpg ~ am + hp * wt, data = mtcars)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.9845 -1.6580 -0.7407  1.4362  4.5266
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 49.452241   5.280731   9.365 5.69e-10 ***
## am1          0.125107   1.333431   0.094 0.925942
## hp          -0.119303   0.026550  -4.494 0.000119 ***
## wt          -8.100558   1.789325  -4.527 0.000108 ***
## hp:wt         0.027488   0.008473   3.244 0.003130 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.192 on 27 degrees of freedom
## Multiple R-squared:  0.8848, Adjusted R-squared:  0.8677
## F-statistic: 51.84 on 4 and 27 DF,  p-value: 2.765e-12

confint(fit1.4)

##              2.5 %      97.5 %
## (Intercept) 38.61707633 60.28740526
## am1         -2.61086742  2.86108128
## hp          -0.17377926 -0.06482709
## wt         -11.77194963 -4.42916547
## hp:wt         0.01010407  0.04487245

par(mfrow = c(2,2))
plot(fit1.4)

```



Our final model consists of the coefficients of am, hp, wt and the interaction between hp:wt.

```
dfbetas(fit1.4)
```

##	(Intercept)	am1	hp	wt
## Mazda RX4	0.2229937740	-0.312719480	-0.149936156	-0.239169899
## Mazda RX4 Wag	0.1445912590	-0.177327541	-0.089536903	-0.153744133
## Datsun 710	0.0630466614	-0.243523618	-0.010845917	-0.091073466
## Hornet 4 Drive	-0.0053010247	-0.074748765	0.045559204	0.034426502
## Hornet Sportabout	0.0270406849	-0.116376753	0.057621386	-0.022827417
## Valiant	0.0473906064	0.002779453	-0.048100717	-0.067184891
## Duster 360	-0.1157407236	0.175403312	-0.005083592	0.136720055
## Merc 240D	-0.0864521529	-0.054597639	0.062062560	0.193554047
## Merc 230	0.0022441693	-0.107574183	0.043748776	0.047893378
## Merc 280	-0.0234790509	-0.017029520	0.034699317	0.035858725
## Merc 280C	0.0265555950	0.019260959	-0.039246093	-0.040557423
## Merc 450SE	-0.0353741886	-0.010652726	0.031631024	0.044522459
## Merc 450SL	-0.0003565963	-0.053153837	0.034565466	0.004768639
## Merc 450SLC	0.0037927751	0.036999472	-0.026540786	-0.007693670
## Cadillac Fleetwood	0.0489783149	-0.143818559	0.175266217	-0.072042317
## Lincoln Continental	-0.0274541972	-0.125972309	0.281747982	0.006638099
## Chrysler Imperial	0.2547611816	0.042056009	-0.589309239	-0.239573418
## Fiat 128	0.1823813720	0.305243229	-0.341848505	-0.094910944
## Honda Civic	-0.6864918866	0.232760800	0.654746754	0.638257214
## Toyota Corolla	0.6014043045	-0.094015963	-0.590863169	-0.541211197
## Toyota Corona	-0.5579251360	0.667280715	0.235609688	0.494313253
## Dodge Challenger	0.0375862716	0.086967597	-0.100132448	-0.060883017
## AMC Javelin	0.0187420157	0.150042822	-0.125467945	-0.047542506
## Camaro Z28	-0.1937882058	0.238923121	0.061783424	0.221367785
## Pontiac Firebird	-0.0621371938	-0.097193846	0.117278850	0.084797810

```
## Fiat X1-9          -0.2679882518 -0.001644692  0.282719515  0.232420086
## Porsche 914-2     -0.0167904227 -0.035182315  0.018518867  0.010481086
## Lotus Europa      0.2553329897 -0.134207544 -0.088444398 -0.264284699
## Ford Pantera L    -0.0588654118  0.055208267  0.079139919  0.041264917
## Ferrari Dino      -0.0168494727  0.016736561  0.018008321  0.015371107
## Maserati Bora     -0.0094743721  0.110779351  0.126712083 -0.111118748
## Volvo 142E        0.1396981118 -0.178812613 -0.087625690 -0.149241813
##                  hp:wt
## Mazda RX4         0.20076249
## Mazda RX4 Wag     0.12237627
## Datsun 710        0.06123911
## Hornet 4 Drive    -0.05965579
## Hornet Sportabout -0.02847516
## Valiant           0.06772809
## Duster 360        -0.06688015
## Merc 240D         -0.17617027
## Merc 230          -0.07611081
## Merc 280          -0.04273871
## Merc 280C         0.04833891
## Merc 450SE        -0.03340471
## Merc 450SL        -0.02198193
## Merc 450SLC       0.01821290
## Cadillac Fleetwood -0.11854453
## Lincoln Continental -0.22746583
## Chrysler Imperial 0.55340692
## Fiat 128          0.21164818
## Honda Civic       -0.64812939
## Toyota Corolla    0.55891525
## Toyota Corona     -0.29761868
## Dodge Challenger  0.09769849
## AMC Javelin       0.11247753
## Camaro Z28        -0.15687446
## Pontiac Firebird  -0.10373377
## Fiat X1-9         -0.25348568
## Porsche 914-2     -0.01137519
## Lotus Europa      0.15155877
## Ford Pantera L    -0.05689438
## Ferrari Dino      -0.01707888
## Maserati Bora     0.04867646
## Volvo 142E        0.12000908
```

From a quick glance at the `dfbetas`, we can see that Honda Civic, Toyota Corolla, and Chrysler Imperial are top 3 influential observations. However, the data is clean and tidy. It is very unlikely that these observations are errors.

```
cooks.distance(fit1.4)
```

```
##          Mazda RX4      Mazda RX4 Wag      Datsun 710
##      0.0315169119      0.0088332478      0.0361539163
##      Hornet 4 Drive  Hornet Sportabout      Valiant
##      0.0079471836      0.0082362307      0.0033871388
##      Duster 360      Merc 240D      Merc 230
##      0.0148082243      0.0596024980      0.0180460527
##      Merc 280      Merc 280C      Merc 450SE
##      0.0017206544      0.0021989009      0.0045753632
```

##	Merc 450SL	Merc 450SLC	Cadillac Fleetwood
##	0.0034112694	0.0020847127	0.0501392462
##	Lincoln Continental	Chrysler Imperial	Fiat 128
##	0.0676055149	0.1707166771	0.1298085785
##	Honda Civic	Toyota Corolla	Toyota Corona
##	0.1363949262	0.1442959822	0.1152196007
##	Dodge Challenger	AMC Javelin	Camaro Z28
##	0.0134717409	0.0240311659	0.0298254696
##	Pontiac Firebird	Fiat X1-9	Porsche 914-2
##	0.0294347423	0.0426635950	0.0023794370
##	Lotus Europa	Ford Pantera L	Ferrari Dino
##	0.0386442240	0.0034148513	0.0001347569
##	Maserati Bora	Volvo 142E	
##	0.1032192228	0.0093545089	

With a final look at `cooks.distance`, we can see that there are no outliers in this dataset, creating bias in our model.

## Conclusion

We can conclude that ;

- We need more data to see if transmission type makes a difference in mpg variable. With the current data set, we can only conclude with %7.41 confidence that the transmission type changes miles per gallon between the values -2.61086742 and 2.86108128. We cannot conclude if transmission type affects mpg positively or negatively. With the current data set, the statistics test do not provide a significant result(%92.59 > %5). Maybe other variables can be included such as driven kms in city/in suburbs to find out the effect of traffic.
- 1 point increase in HP results in 0.12 decrease in miles per gallon.
- 1000 pounds increase in weight results in 8.10 decrease in miles per gallon.
- the effect of hp on mpg somewhat depends on the weight of the car.