

Automatic vs Manual Transmission on MPG

Regression Models Course Project

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Executive Summary

High dimensional data analysis has become increasingly frequent and important in diverse fields of sciences, engineering, and humanities, ranging from genomics and health sciences to economics, finance and machine learning [1]. Generally the inclusion of multiple correlated variables in the model will make the individual regression coefficients to be estimated unstably, i.e. with big variance. If we only use a small subset of variables, then the models are likely to be biased. Often we would like to achieve a balance between the variance and the bias [2]. This project analyzes the Motor Trend Car Road Tests (mtcars) dataset, a collection of cars in exploring the relationship between a set of variables and miles per gallon (MPG) (outcome), We are trying to answer the following two questions:

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

Two models are selected to demonstrate the importance of model selection, one is automatic/manual as the only predictor, and the other is the minimal adequate model.

Exploratory Data Analysis

The data look clean and there are no missing values to impute (Appendix A). However we need to transform some of the variables to factor variables (cyl, vs, am, gear, carb). The table shows the first 5 observations of the mtcars dataset.

	mpg	cyl	displacement	hp	drat	wt	qsec	vs	am	gear	carb
1	21.0	6	160	110	3.90	2.620	16.46	0	1	4	4
2	21.0	6	160	110	3.90	2.875	17.02	0	1	4	4
3	22.8	4	108	93	3.85	2.320	18.61	1	1	4	1
4	21.4	6	258	110	3.08	3.215	19.44	1	0	3	1
5	18.7	8	360	175	3.15	3.440	17.02	0	0	3	2

Regression Analysis

If we naively took “am” (0 automatic, 1 manual) as the predictor and “mpg” as the response, we may mistakenly conclude that, on average, the manual transmission cars have 7.25 higher MPGs than automatic transmission cars as shown in the table below. However if you look at the “Residuals Fitted (1)” plot in Appendix B, there is a clear pattern. Also, the $R^2=0.36$ is small, in other words, the model considering single predictor “am” only explains 36% of the variability of the response data around its mean. There are confounders that we need to include into the regression model.

	Estimate	Std. Error	t value	Pr(> t)
<i>(Intercept)</i>	17.147	1.125	15.247	0
<i>am1</i>	7.245	1.764	4.106	0

Model Selection

We begin with throwing all the predictors into the regression model and performs stepwise model selection by AIC (Appendix C). The final model is using “cyl”, “hp”, “wt” and “am” as the predictors. The coefficients are shown in the following table. The “Residuals vs Fitted (2)” plot in Appendix B does not show any obvious patterns.

	Estimate	Std. Error	t value	Pr(> t)
<i>(Intercept)</i>	33.708	2.605	12.94	0
<i>cyl6</i>	−3.031	1.407	−2.154	0.041
<i>cyl8</i>	−2.164	2.284	−0.947	0.352
<i>hp</i>	−0.032	0.014	−2.345	0.027
<i>wt</i>	−2.497	0.886	−2.819	0.009
<i>am1</i>	1.809	1.396	1.296	0.206

The slope of “am” is 1.809, which means, on average, holding other variables constant the manual transmission cars have 1.809 higher MPGs than automatic transmission cars. However the “p-value” (0.206) is large, so we failed to reject the null-hypothesis that the manual transmission cars have the same MPGs as that of automatic transmission cars. The 95% confidence interval is [−1.061, 4.679] also includes 0.

Summary

After analyzing the “mtcars” dataset, we conclude:

1. There is no significant difference on MPG by automatic or manual transmission,
2. The MPG difference between automatic and manual transmissions is [−1.061, 4.679] at 95% C.L.

Reference

- [1] Jianqing Fan and Jinchi Lv, Stat Sin. 2010 Jan; 20(1): 101–148.
- [2] <http://www.ams.sunysb.edu/~xing/AMS578/Documents/Chp9.pdf>

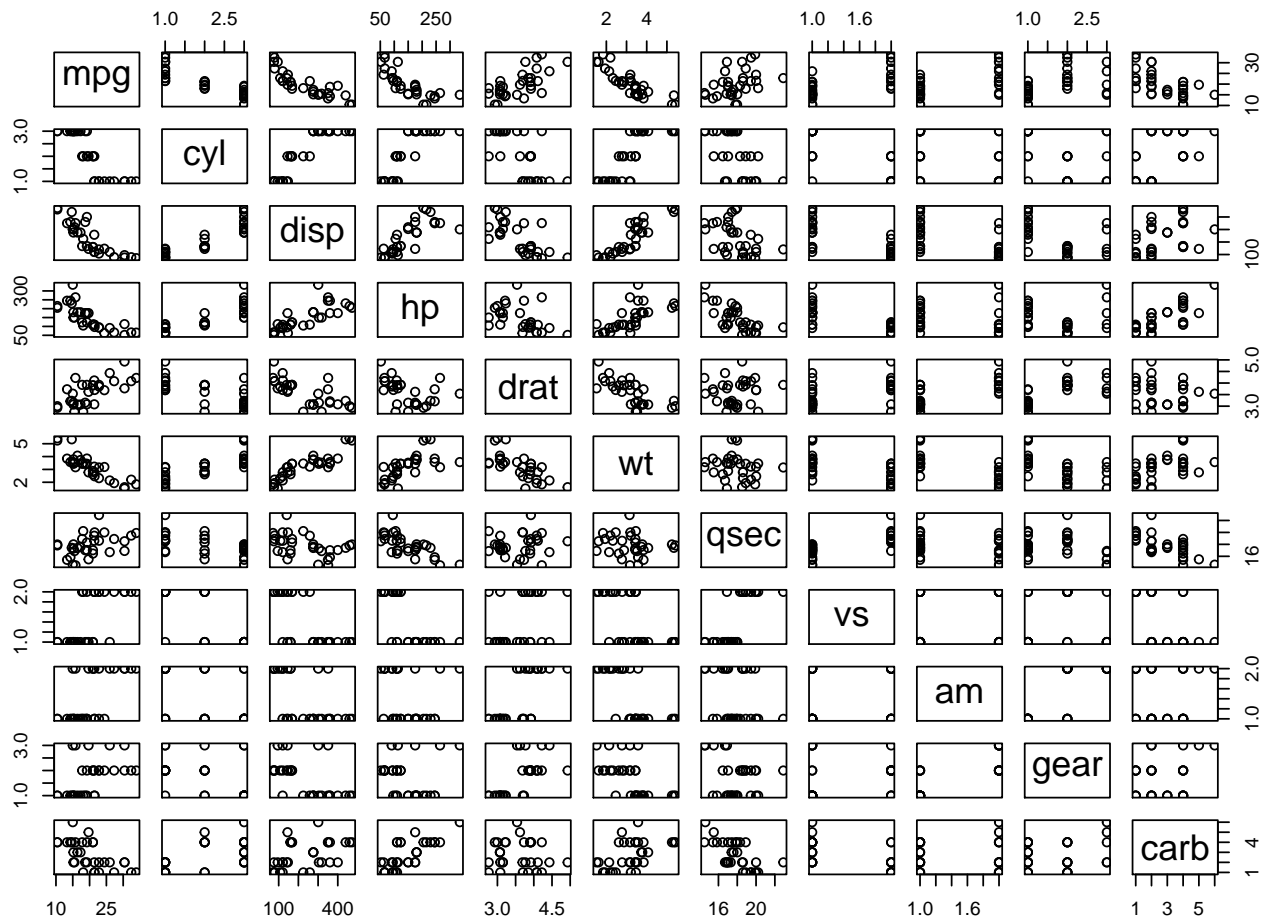
Apêndice

Apêndice A

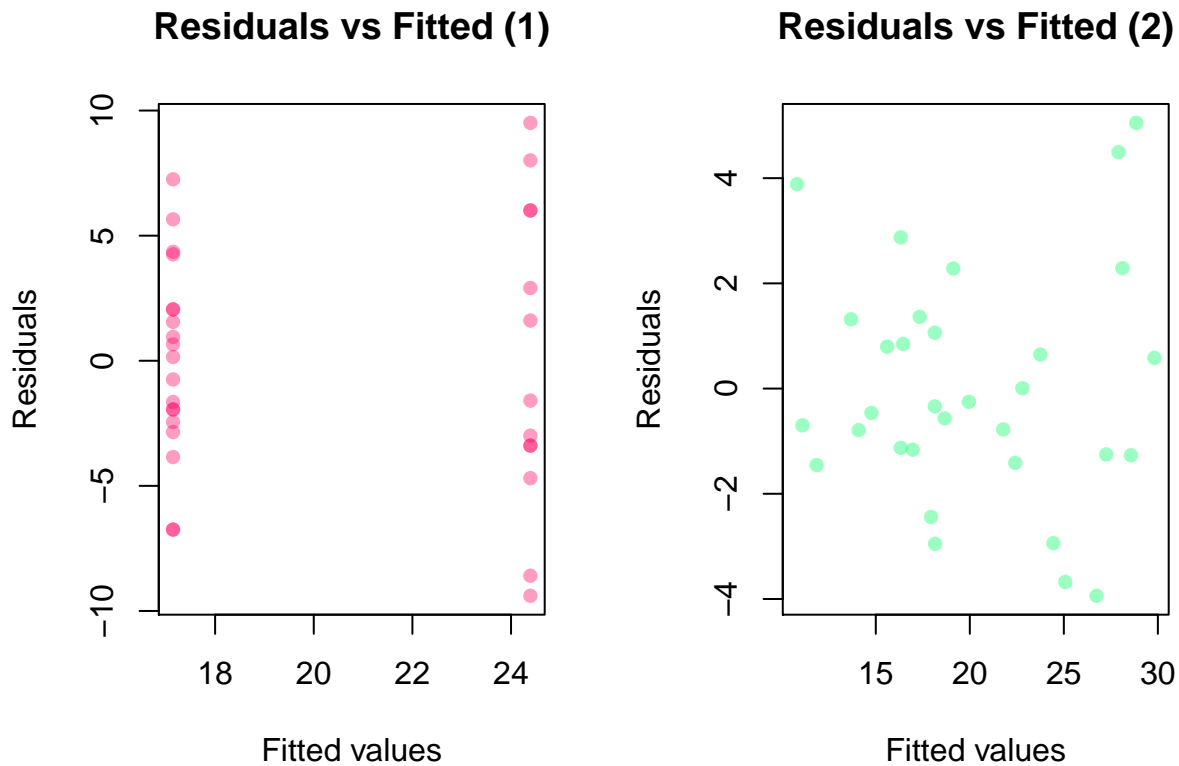
```
## 'data.frame': 32 obs. of 11 variables:
## $ mpg : num 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
## $ cyl : Factor w/ 3 levels "4","6","8": 2 2 1 2 3 2 3 1 1 2 ...
## $ disp: num 160 160 108 258 360 ...
## $ hp : num 110 110 93 110 175 105 245 62 95 123 ...
## $ drat: num 3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
## $ wt : num 2.62 2.88 2.32 3.21 3.44 ...
## $ qsec: num 16.5 17 18.6 19.4 17 ...
## $ vs : Factor w/ 2 levels "0","1": 1 1 2 2 1 2 1 2 2 2 ...
## $ am : Factor w/ 2 levels "0","1": 2 2 2 1 1 1 1 1 1 1 ...
## $ gear: Factor w/ 3 levels "3","4","5": 2 2 2 1 1 1 1 2 2 2 ...
## $ carb: Factor w/ 6 levels "1","2","3","4",...: 4 4 1 1 2 1 4 2 2 4 ...
```

##	mpg	cyl	disp	hp	drat
##	Min. :10.40	4:11	Min. : 71.1	Min. : 52.0	Min. :2.760
##	1st Qu.:15.43	6: 7	1st Qu.:120.8	1st Qu.: 96.5	1st Qu.:3.080
##	Median :19.20	8:14	Median :196.3	Median :123.0	Median :3.695
##	Mean :20.09		Mean :230.7	Mean :146.7	Mean :3.597
##	3rd Qu.:22.80		3rd Qu.:326.0	3rd Qu.:180.0	3rd Qu.:3.920
##	Max. :33.90		Max. :472.0	Max. :335.0	Max. :4.930

##	wt	qsec	vs	am	gear	carb
##	Min. :1.513	Min. :14.50	0:18	0:19	3:15	1: 7
##	1st Qu.:2.581	1st Qu.:16.89	1:14	1:13	4:12	2:10
##	Median :3.325	Median :17.71			5: 5	3: 3
##	Mean :3.217	Mean :17.85				4:10
##	3rd Qu.:3.610	3rd Qu.:18.90				6: 1
##	Max. :5.424	Max. :22.90				8: 1



Appendix B



Appendix C

```
## Stepwise Model Path
## Analysis of Deviance Table
##
## Initial Model:
## mpg ~ cyl + disp + hp + drat + wt + qsec + vs + am + gear + carb
##
## Final Model:
## mpg ~ cyl + hp + wt + am
##
##
```

##	Step	Df	Deviance	Resid. Df	Resid. Dev	AIC
##	1			15	120.4027	76.40339
##	2 - carb	5	13.5988573	20	134.0015	69.82769
##	3 - gear	2	5.0215145	22	139.0230	67.00492
##	4 - drat	1	0.9672159	23	139.9903	65.22678
##	5 - disp	1	1.2473996	24	141.2377	63.51066
##	6 - qsec	1	2.4420033	25	143.6797	62.05921
##	7 - vs	1	7.3459298	26	151.0256	61.65483