

Modeling MPG by Transmission

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Monday, September 21, 2015

Summary

This paper aims to determine whether an automatic or manual transmission is better for miles per gallon, and give an estimated quantity for this value. Using the `mtcars` dataset, we will first explore the data and choose our regressors. We will look at two models, and then run some confirmation tests before presenting our conclusions. Please see the appendix for all figures and tables.

Exploratory Data Analysis

The `mtcars` dataset contains 11 variables on 32 different cars from 1973-74. The measured variables are 1) `mpg`, 2) `cyl` number of cylinders, 3) `disp` displacement, 4) `hp` horsepower, 5) `drat` rear axle ratio, 6) `wt` weight(lb/1000), 7) `qsec` 1/4 mile time, 8) `vs` engine orientation(straight or 'V'), 9) `am` transmission (0 = automatic, 1 = manual), 10) `gear` number of forward gears, and 11) `carb` number of carburetors. Table 1 in the appendix displays the original data..

While we ultimately want to know what impact, if any, transmission has on mpg, there are many variables that can have an impact on mpg. The linear model `lm(mpg ~ am, mtcars)` gives us a transmission coefficient of 7.245, which seems to indicate that a manual transmission greatly increases mpg. We also find a positive correlation of 0.599. However, this model is far too simplistic. Many factors outside of transmission can affect mpg - weight, cylinders, possibly others!

Our first task is determine which independent variables appear to have a relationship with the dependent variable, mpg. Then we will perform a simple check for multicollinearity among the independent variables, and based on what we find there we will be able to choose the regressors to build our model.

Table 2 in the appendix is a table with correlation values for each variable in the `mtcars` dataset. Using this table, we can tell that weight (-0.868), cylinders(-0.852), and displacement(-0.848) have the strongest correlations to mpg. Looking for redundancy, weight is strongly correlated with cylinders(0.782) and displacement(0.888). Because weight and displacement are so closely correlated, and weight is more closely correlated to mpg, we will drop displacement from our regression. Cylinders are slightly less correlated, so we will build two models - one including cylinders and one without. Figure 1 shows the relationships between our regressors with scatterplots and correlations.

Regression Models

The stage is set to build two multivariate linear regressions with outcome variable `mpg`. Model 1 contains predictors `am`, `wt`, and `cyl`. Model 2 leaves out cylinders.

```
model1 <- lm(mpg ~ am + wt + cyl, mtcars) ## First model, including cylinders.
model2 <- lm(mpg ~ am + wt, mtcars) ## Second model, without cylinders.

summary(model1)$coefficients
##              Estimate Std. Error    t value    Pr(>|t|)
## (Intercept) 39.4179334   2.6414573  14.9227979 7.424998e-15
## am           0.1764932   1.3044515   0.1353007 8.933421e-01
## wt          -3.1251422   0.9108827  -3.4308942 1.885894e-03
## cyl         -1.5102457   0.4222792  -3.5764148 1.291605e-03
```

```
summary(model2)$coefficients
##              Estimate Std. Error      t value      Pr(>|t|)
## (Intercept) 37.32155131  3.0546385  12.21799285 5.843477e-13
## am          -0.02361522  1.5456453  -0.01527855 9.879146e-01
## wt          -5.35281145  0.7882438  -6.79080719 1.867415e-07
```

We can interpret the coefficients for `am` (transmission) to mean that in **model 1**, we get a **-0.024 decrease in mpg for having a manual transmission**, while in **model 2** we see a **0.177 increase in mpg**. There is a slightly better adjusted r-squared value in model 1: 81% vs. 73%. This tells us that we have explained more of the total variation by including cylinders in our regression. Figure 2 in the appendix illustrates this phenomena, with model 1 having slightly less variation around the regression line than model 2.

95% confidence intervals for both models with $\hat{B}_1 \pm 1.96 * SE(\hat{B}_1)$ are:

```
sumCoef1 <- summary(model1)$coefficients
sumCoef1[2,1] + c(1, -1) * 1.96 * (sumCoef1[2,2])
## [1]  2.733218 -2.380232
sumCoef2 <- summary(model2)$coefficients
sumCoef2[2,1] + c(1, -1) * 1.96 * (sumCoef2[2,2])
## [1]  3.00585 -3.05308
```

In both models the intervals include 0. Thus, we cannot conclude with 95% confidence that there is a significant relationship between transmission and mpg. The p-values for transmission in these models are also extremely high:

```
sumCoef1[2,4] ## Model 1 transmission estimate p-value
## [1] 0.8933421
sumCoef2[2,4] ## Model 2 transmission estimate p-value
## [1] 0.9879146
```

Were there to be no relationship between mpg and transmission type, we would get these values 89% and 99% of the time. That's extremely poor for trying to establish a meaningful relationship between the two.

Running a `hatvalues` test and residual plot shows that, overall, the model fits the data well. Figure 3 illustrates that there is no discernible pattern in the residuals, and the hat values have no extreme impact on the model.

Conclusion

In conclusion, which kind of transmission one chooses has no significant effect on miles per gallon. At 32 cars, the sample size is somewhat small, but even so, the numbers aren't even close. Based on the two models in this analysis, weight and cylinders have a far greater impact on what kind of mpg you get.

Appendix

Table 1:

mtcars												
##		mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
##	Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
##	Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
##	Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
##	Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
##	Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
##	Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
##	Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
##	Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
##	Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
##	Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
##	Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
##	Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
##	Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
##	Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
##	Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
##	Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
##	Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
##	Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
##	Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
##	Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
##	Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
##	Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0	0	3	2
##	AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0	0	3	2
##	Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0	0	3	4
##	Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0	0	3	2
##	Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
##	Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
##	Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
##	Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0	1	5	4
##	Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0	1	5	6
##	Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0	1	5	8
##	Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2

Table 2:

##		mpg	cyl	disp	hp	wt
##	mpg	1.0000000	-0.8521620	-0.8475514	-0.7761684	-0.8676594
##	cyl	-0.8521620	1.0000000	0.9020329	0.8324475	0.7824958
##	disp	-0.8475514	0.9020329	1.0000000	0.7909486	0.8879799
##	hp	-0.7761684	0.8324475	0.7909486	1.0000000	0.6587479
##	wt	-0.8676594	0.7824958	0.8879799	0.6587479	1.0000000

Figure 1:

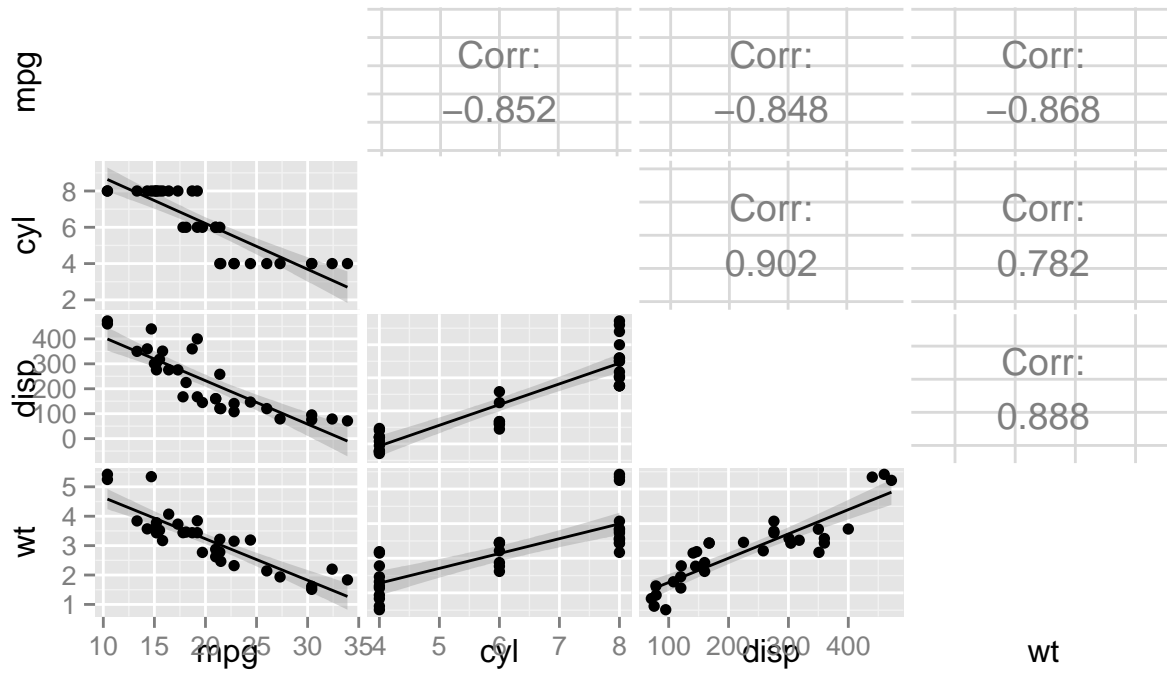
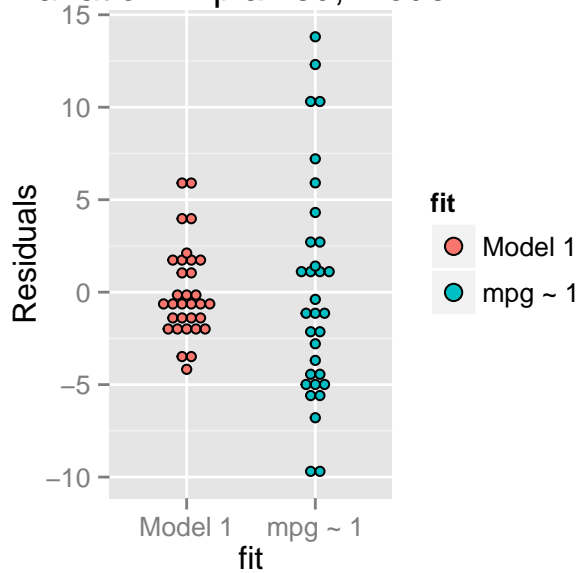


Figure 2:

Variation Explained, Model 1



Variation Explained, Model 2

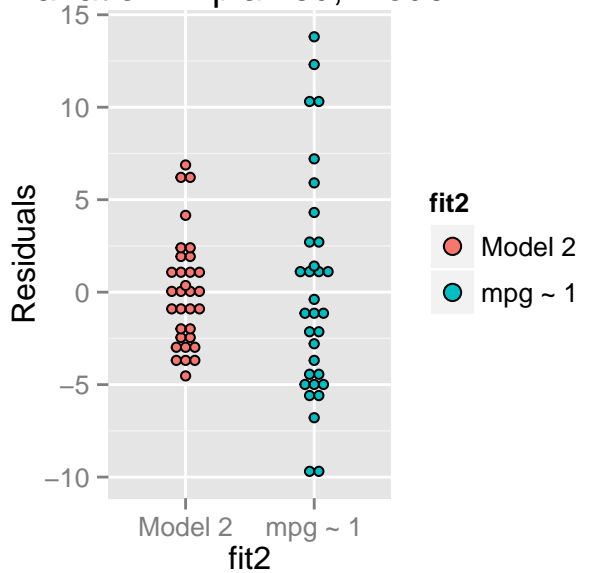


Figure 3:

