

# Regression Models Assignment

## Introduction

We will first do exploratory data analysis to find some relationship between the variables.

Then we will fit several linear models and analyse the quality of prediction, choose the best one and find the confidence interval for the predictor *am*.

## Exploratory Data Analysis

The **pair plots** show that *mpg* has a strong correlation ( $> 0.8$ ) with *cyl*, *disp* and *wt*. The correlation between *mpg* and *am* is only 0.6 which is not very significant, and also *am* has stronger correlation with *wt* and *drat*, indicating that *am* could be confounded by *wt* and *drat*.

The **box plot** shows that there is a difference between the distribution of *mpg* for different *am* value.

Another **plot** shows *mpg* by *am* treating *cyl* as confounder. It seems *mpg* and *am* are positively correlated in each *cyl* level.

## Regression Analysis

First fit a model with all predictors except *am*.

```
lm_fit_no_am <- lm(mpg ~ . - am, data = mtcars)
```

```
summary(lm_fit_no_am)
```

```
...
Residuals:
    Min       1Q   Median       3Q      Max
-2.9886 -1.6738 -0.3834  0.9796  5.4395
...
Residual standard error: 2.68 on 22 degrees of freedom
Multiple R-squared:  0.8596,    Adjusted R-squared:  0.8022
F-statistic: 14.97 on 9 and 22 DF,  p-value: 1.855e-07
```

Then fit a linear model with all predictors.

```
lm_fit <- lm(mpg ~ ., data = mtcars)
```

```
summary(lm_fit)
```

```
...
Residuals:
    Min       1Q   Median       3Q      Max
-3.4506 -1.6044 -0.1196  1.2193  4.6271
```

```

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
...
am          2.52023    2.05665   1.225   0.2340
...
Residual standard error: 2.65 on 21 degrees of freedom
Multiple R-squared:  0.869, Adjusted R-squared:  0.8066
F-statistic: 13.93 on 10 and 21 DF, p-value: 3.793e-07

```

Note  $p\text{-value}(am) = 0.2340 \gg 0.05$ , indicating that *am* is unlikely to be significant in predicting *mpg*.

And [here](#) is the residual plot of the two models (ignore the third one for now), and a [QQ plot](#) shows that the residuals almost follow a normal distribution.

Performing anova on `lm_fit_no_am` and `lm_fit`,

```

anova(lm_fit_no_am, lm_fit)
## ...
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1      22 158.04
## 2      21 147.49  1   10.547 1.5016  0.234

```

The F-value for `lm_fit` compared to `lm_fit_no_am` is  $0.234 \gg 0.05$ , indicating that *am* does not have strong influence on *mpg*.

Next, find the VIF of the predictors of `lm_fit`,

```

library(car)
vif(lm_fit)

##      cyl      disp      hp      drat      wt      qsec      vs
## 15.373833 21.620241  9.832037  3.374620 15.164887  7.527958  4.965873
##      am      gear      carb
##  4.648487  5.357452  7.908747

```

Finally, fit another model by taking out the top two predictors with  $VIF > 10$ .

```
lm_fit_adj_vif <- lm(mpg ~ . - disp - cyl, data = mtcars)
```

```
summary(lm_fit_adj_vif)
```

```

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
...
am          2.42418    1.91227   1.268   0.2176
...
Residual standard error: 2.566 on 23 degrees of freedom
Multiple R-squared:  0.8655, Adjusted R-squared:  0.8187
F-statistic: 18.5 on 8 and 23 DF, p-value: 2.627e-08

```

We use the model with the highest adjusted r-squared value, i.e., `lm_fit_adj_vif`.

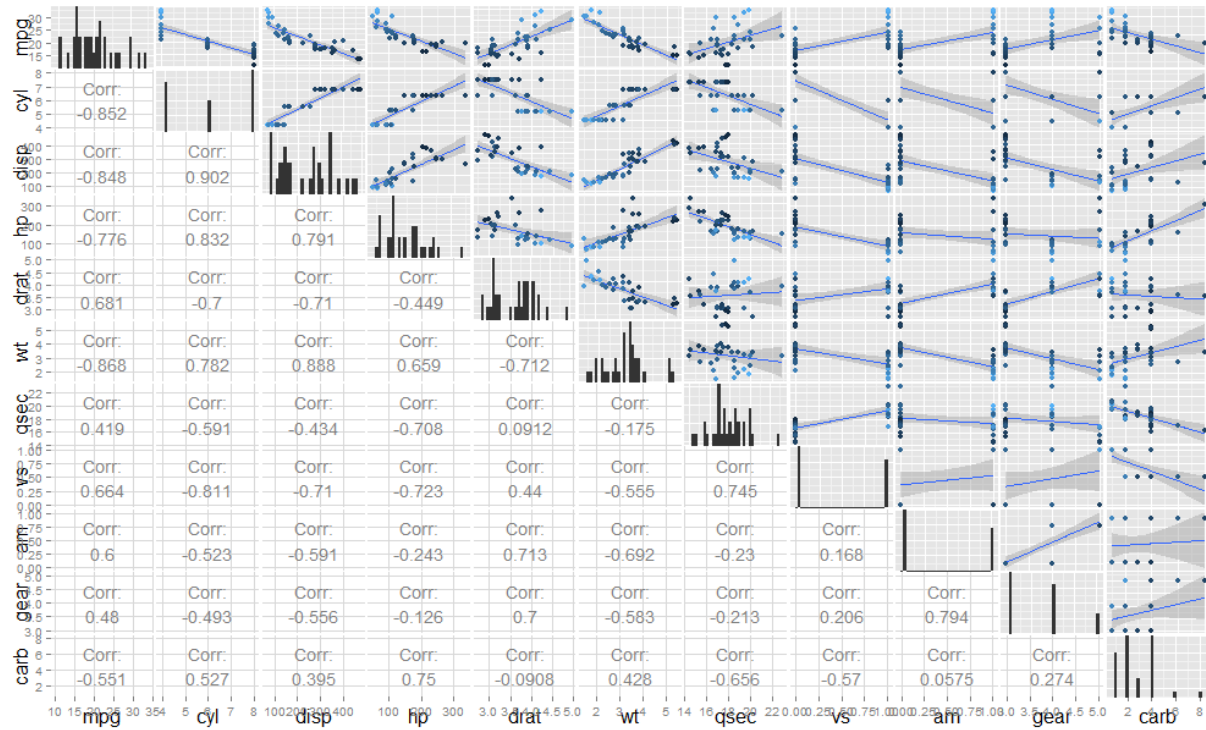
Assuming  $\text{coef}(am)$  follows a t-distribution, its  $\alpha$  confidence interval is  $[\hat{\theta} - z\sigma, \hat{\theta} + z\sigma]$ , where  $\hat{\theta}$  = estimate mean = 2.42418,  $\sigma$  = standard error = 1.91227,  $z = \Pr(t < \frac{\alpha}{2})$ , and the degree of freedom is 23. For  $\alpha = 0.05$ , the interval is [1.449181, 3.399179].

Overall, manual transmission seems to have a higher MPG on average, about 2.42mpg, with 95% confidence interval [1.449181, 3.399179].

# Appendix

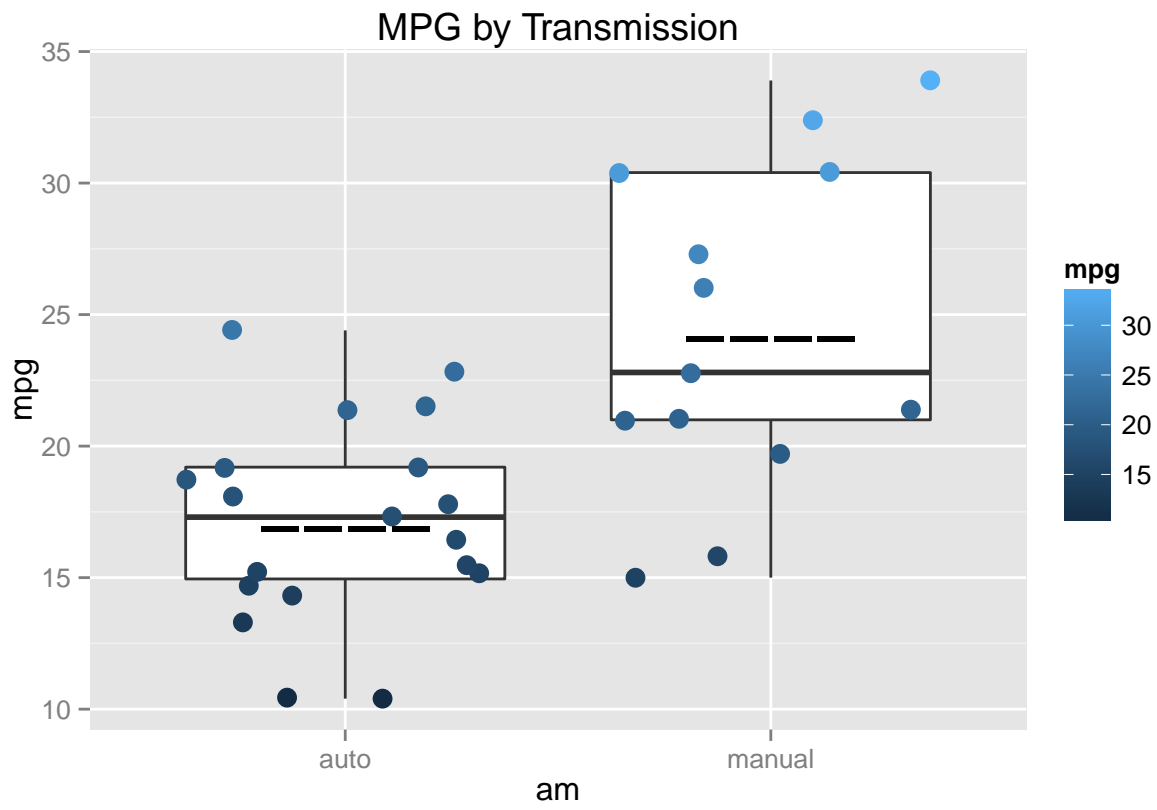
## Pair plot of features

- Lighter color means higher mpg

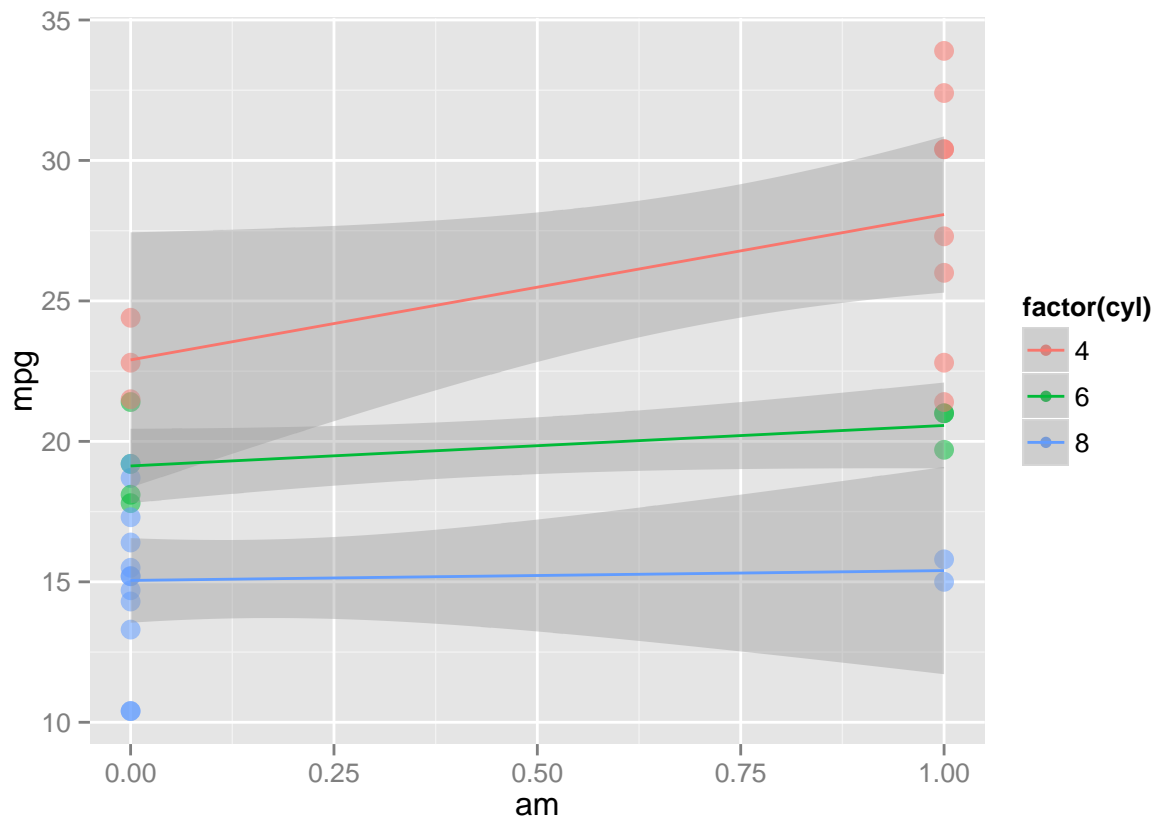


## Box plot of MPG and Transmission

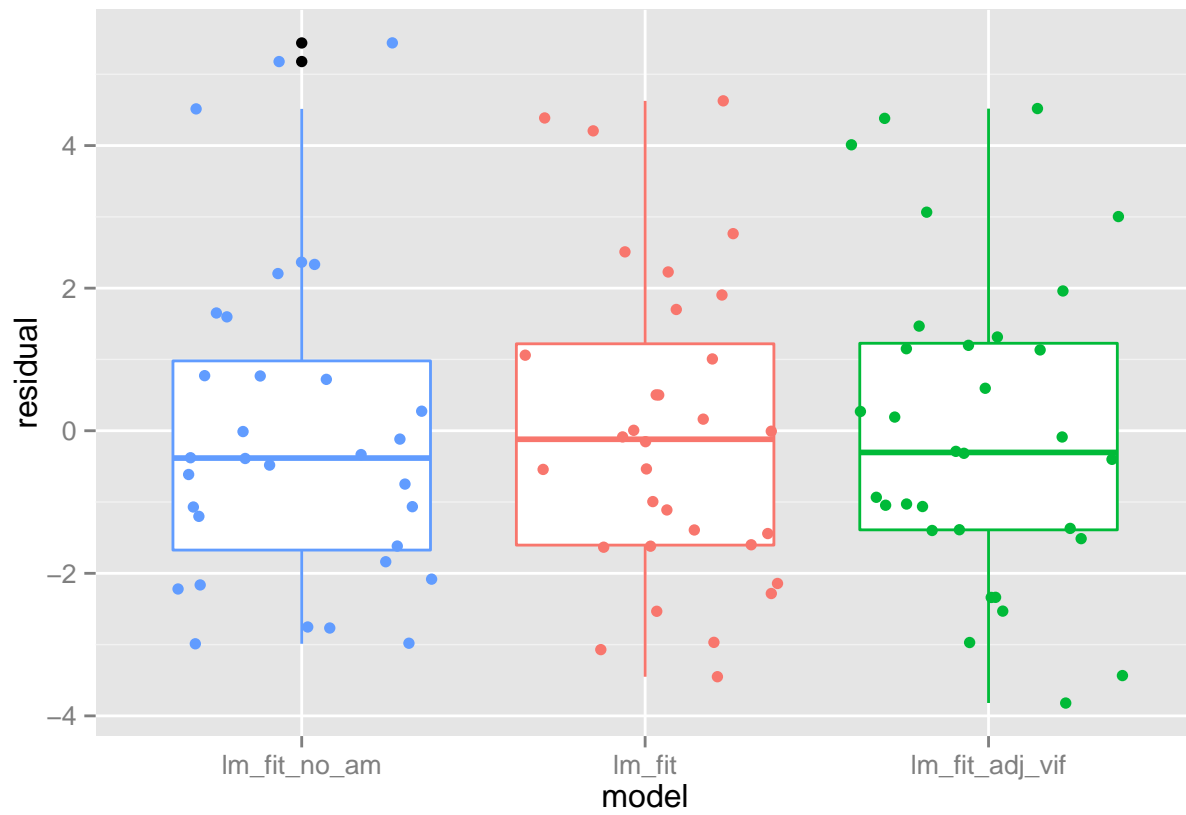
- ---- marks the mean



MPG by Transmission across Number of Cylinders



## Linear Regression Residual Plot



## QQ Plot for Residuals

