

1. Use the same cars2010 dataset you have used in previous labs. To obtain this data, submit the following code:

```
library(AppliedPredictiveModeling)
data(FuelEconomy)
```

This dataset has variables pertaining to fuel economy of various cars. **Do not** create a training and test set. Just use the whole cars2010 dataset for the following analysis. The cars2011 and cars2012 datasets will be used at later time periods..

Perform the following analysis:

- a. Run a LASSO regression predicting the **FE** variable using all the remaining variables. Some of these predictor variables are coded as numeric, but should be treated as categorical. The only numeric variables in your dataset should be **EngDispl**. All remaining variables are categorical.
  - a. Plot the coefficients and how they change across different levels of lambda.
- b. Perform a CV LASSO to optimize the lambda value.
  - a. What is the value of lambda that minimizes the MSE?
  - b. What is the value of lambda one standard error above the minimum MSE value?
  - c. How many variables are left at the penalty value that is one standard error above the minimum MSE value (think of variables as a whole, not per category)? (HINT: Look at the coefficients from the model with `coef` function.)
- c. Obtain the variables from the LASSO regression at the penalty value that is one standard error above the minimum MSE value. The multiple linear regression with p-value selection (Lab 6) left the variables **EngDispl**, **NumCyl**, **Transmission**, **AirAspirationMethod**, **NumGears**, **TransLockup**, **DriveDesc**, **IntakeValvePerCyl**, **CarlineClassDesc**, and **VarValveLift**.
  - a. What variables were left in your LASSO at the 1SE above minimum MSE penalty value?
  - b. How does this compare against the variables from the multiple linear regression?