

# POL 350C, Homework 7

May 19, 2016

Assigned: 5/19/2016

Due: 5/26/2016

In this problem set we're going to use machine learning to predict student's drinking habits. The data come from a public health study of Portuguese students. You can read more about the variables (and their interpretation) here:

<http://archive.ics.uci.edu/ml/datasets/STUDENT+ALCOHOL+CONSUMPTION#>

We're going to model the sum of weekday and weekend drinking activity.

The data are stored in `StudentDrinking.RData` on canvas. The dependent variable is, `alcohol`, which is a measure of alcohol consumption. The bigger `alcohol` is, the more students drink. The covariates are stored in `X`.

## 1 Comparing Coefficients from OLS, LASSO, Ridge, and Elastic Net

We first want to explore the behavior of OLS, LASSO, and Ridge applied to the data.

- i) Fit a linear regression of `alcohol` on the covariates in the included data
- ii) Using `cv.glmnet` fit a LASSO regression of `alcohol` on the covariates
- iii) Using `cv.glmnet` fit a Ridge regression of `alcohol` on the covariates
- iv) Using `cv.glmnet` fit an elastic-net regression of `alcohol` on the covariates, with  $\alpha = 0.5$ . Explain what  $\alpha = 0.5$  implies about the model you're fitting.
- v) Using your models from (i-iv) let's examine the behavior of the coefficient on `male` as  $\lambda$  increases
  - a) Suppose `glmnet.obj` contains the results from applying `cv.glmnet`. To obtain the coefficient values for the sequence of  $\lambda$  values tested in `cv.glmnet`, we use the coefficient function `coef(glmnet.obj, s = glmnet.obj$lambda)`.

Use this function to obtain a matrix of coefficients for the models used in (ii-iv).

- b) Using the matrix for each method, plot the coefficient on **male** against the value of  $\lambda$  from the models in ii-iv. Include the coefficient from OLS as a flat line. What do you notice as  $\lambda$  increases?

## 2 Cross-Validation, Super Learning and Ensembles

We're going to assess the performance of five models, an unweighted average, and a super-learning average of the methods.

- i) First, set the first 20 rows to the side for use as the validation set.
- ii) We'll first estimate the (unconstrained) super learner weights.
  - a) On the training data (all but the first 20 rows) perform ten fold cross validation, including (1) linear regression, (2) LASSO, (3) Ridge, (4) Elastic-Net, and (5) Random Forest. Obtain 5 predictions for each observation in the training set, one from each observation
  - b) Regress the dependent variable on the out of sample prediction, (without including an intercept).
  - c) Store those weights as **w**
- iii) Now, fit all 5 models from (ii)-(a) to the entire training data set and predict the drinking level from the validation set (the data put off to the side).
- iv) Obtain two ensemble predictions.
  - a) Take the unweighted average of the predictions from the methods
  - b) Take the weighted average, using the weights **w**.
- v) You should have 7 predictions. Store those in a matrix and report the correlation between the predictions
- vi) Using the average absolute difference as a loss function assess the performance of each method. Which method performs best? Which performs the worst?

The average absolute difference for method  $k$  is defined as

$$L(\mathbf{Y}, \hat{\mathbf{Y}}_k) = \sum_{i=1}^{N_{\text{validation}}} \frac{|Y_i - \hat{Y}_{ik}|}{N_{\text{validation}}}$$

where  $N_{\text{validation}}$  refers to the number of observations in the validation set  $\hat{\mathbf{Y}}_k$  refers to the predictions from the  $k^{\text{th}}$  method,