## CMSE 381: HW8

- 1 (20 pts) Exercise 6.8.4
- 2 (20 pts) Exercise 6.8.6
- 3 (10 pts) Exercise 6.8.9 (e)
- 4 We will use ridge and lasso regressions to estimate the salary of various baseball players based on numerous predictor measurements. Download 'MLB\_test.csv' and 'MLB\_training.csv' file from d2l, which contains Major League Baseball (MLB) data from the 1986 season. In this dataset, each observation represents one MLB player, and we want to build a model to predict the player's logsalary using the other features in this dataset. We will first define the grid of λs using

$$grid = 10^{\circ} seq(10, -2, length=100).$$

- a. (10 pts) For both ridge and Lasso regressions, perform 5-fold cross-validation on the training data to deter the best value of  $\lambda$  using the one standard error rule. Report the best  $\lambda$ s.
- b. (10 pts) Report the coefficients estimates at the appropriate values of  $\lambda$  for both ridge and Lasso regression using training data.
- c. (20 pts) Evaluate the predictive capacity of ridge and lasso regression by computing  $R^2$  for the testing data.
- d. (10 pts) Suppose that you were coaching a young baseball player who wanted to strike it rich in the major leagues. What handful of attributes would you tell this player to focus on?