## ECE 532, Spring 2019

Homework #1 Due: Tue, Feb 5 @ 2:30 pm

**Problem 1.** We showed in class that when A is a symmetric matrix, we have  $\nabla_{\beta}(\beta^T A \beta) = 2A\beta$ . What is  $\nabla_{\beta}(\beta^T A \beta)$  when A is not necessarily a symmetric matrix? State and derive a general formula.

**Problem 2.** Consider the matrices  $X = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 1 & 2 \\ 2 & 3 & 5 \end{bmatrix}$  and  $y = \begin{bmatrix} 0 \\ 4 \\ 10 \end{bmatrix}$ .

- (a) Compute  $X^TX$ , and verify that  $\operatorname{rank}(X^TX) < 3$ , so  $X^TX$  is not invertible.
- (b) What is  $\min_{\beta} ||y X\beta||_2^2$ ? Show that multiple minimizers exist.

**Problem 3.** Suppose p=3, and we are performing linear regression to predict BMI with  $\beta_1$  corresponding to the intercept,  $\beta_2$  corresponding to height, and  $\beta_3$  corresponding to weight. We want to perform a variant of ridge regression, where only the coefficients  $\beta_2$  and  $\beta_3$  are encouraged to be close to 0. Furthermore, we think that  $\beta_2$  (the coefficient of weight) should be smaller than  $\beta_3$ , so we solve the problem

$$\hat{\beta} = \arg\min_{\beta} \left\{ \|y - X\beta\|_2^2 + \lambda \beta_2^2 + \frac{\lambda}{2} \beta_3^2 \right\}.$$

What is the generative model interpretation of this objective function? (What is the prior distribution  $p(\beta)$ ?)

## Problem 4.

- (a) Suppose  $x_i \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}\right)$ . Let  $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ , and define  $y_i = Ax_i$ . What is the mean of  $y_i$ ? What is the covariance matrix of  $y_i$ ?
- (b) Suppose instead that  $x_i \sim N\left(\begin{bmatrix} 1\\1 \end{bmatrix}, \begin{bmatrix} 1&0\\0&2 \end{bmatrix}\right)$ . Let  $y_i = Ax_i$ , where A is defined as in (i). What are the mean and covariance of  $y_i$ ?

## Problem 5.

(a) Generate n = 100 observations from the linear model  $y_i = x_i^T \beta + \epsilon_i$ , where  $\beta = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$ , the  $\epsilon_i$ 's are i.i.d. draws from a N(0, 12) distribution,

and the  $x_i$ 's are i.i.d. draws from a  $N\left(\begin{bmatrix}0\\0\\0\end{bmatrix},\begin{bmatrix}2&0&0\\0&4&0\\0&0&10\end{bmatrix}\right)$  distribution.

Compute  $\widehat{\beta}_{OLS}$ , and construct a 90% confidence interval for  $\widehat{\beta}_1$ . Does the interval cover  $\beta_1$ ?

- (b) Keeping X fixed, repeat the simulation in (a) 500 times by regenerating  $\epsilon$ . What percentage of the time do the confidence intervals cover  $\beta_1$ ?
- (c) Still keeping X fixed, repeat part (b), this time generating the  $\epsilon_i$ 's from a Uniform[-6,6] distribution. What percentage of the time do the confidence intervals cover  $\beta_1$ ?

**Problem 6.** The following question uses the dataset bikeshare.csv, collected from 2011–2012 from the Capital Bikeshare system in Washington, D.C.

- (a) Perform OLS regression of cnt (total count of rented bikes) against holiday (1 if holiday, 0 otherwise), temp (temperature in Celsius), hum (humidity), and windspeed (wind speed). Do not forget to include an intercept! What is  $\widehat{\beta}_{OLS}$ ? Are the coefficients what you would expect?
- (b) Now perform ridge regression for various values of  $\lambda$ : {0.1, 1, 10, 100, 1000} and report the values of  $\widehat{\beta}$ . What happens as  $\lambda$  increases?