# practice midterm 1

Go Ito April 20, 2019

#### Question 1.

Suppose you are given a weighted least square. Somehow, its coefficients consist of three corners of a triangle, so we need to minimize our loss function subject to this "constraint"; that is, three coefficients must sum up to 180. Then, our loss function can be expressed as the following (neglect  $\frac{1}{N}$  part here):

$$L = (\mathbf{t} - \mathbf{X}\mathbf{w})^T \mathbf{A} (\mathbf{t} - \mathbf{X}\mathbf{w}) - 2\lambda (\mathbf{1}^T \mathbf{w} - 180)$$

The term  $2\lambda(\mathbf{1}^T\mathbf{w} - 180)$  is called a lagrange multiplier, which often appears in optimization problems (and Christou's 100C :P) What is the estimate of our coefficients? i.e. What is  $\hat{\mathbf{w}}$ ?

(For those who want to challenge furthermore, what is  $\hat{\mathbf{w}}$  without having  $\lambda$  term? Hint:  $\mathbf{1}^T \hat{\mathbf{w}} = 180$  still holds.)

#### Question 2.

Given a function:

$$f(w_1, w_2) = \sin(w_1) + \ln(w_2) + 4w_1w_2 + 90$$

Obtain a numeric gradient using  $\epsilon = 0.01$ .

## Question 3.

Using the function from Question 2., perform two iterations of gradient descent (get  $w_{(2)}$ ).

Start from  $w_{(0)} = [\pi, 1]$ , and use  $\gamma = 0.05$ 

# Question 4.

Given a dataset

$$\mathbf{X} = \begin{bmatrix} 5.1 & 3.5 \\ 4.9 & 3.0 \\ 4.7 & 3.2 \\ 4.6 & 3.1 \end{bmatrix}$$
$$\mathbf{t} = [1.4, 1.4, 1.3, 1.5]$$

Perform LOOCV, and obtain the average MSE(Loss) of all four cases.

For each "cases" (row 234, 134, 124, 123), the estimated coefficients are as follows (I purposefully did not add intercept)

$$\begin{split} \hat{\beta}_{234} &= [0.18, 0.17] \\ \hat{\beta}_{134} &= [3.14, -4.19] \\ \hat{\beta}_{124} &= [0.24, 0.07] \\ \hat{\beta}_{123} &= [0.37, -0.14] \end{split}$$

### Question 5.

Use the dataset in Question 4., obtain the estimated coefficients using OLS Loss and Ridge Loss ( $\lambda = 2$ ).

## Question 6.

- When  $\lambda$  is big, what would happen to Lasso estimated coefficients?
- Then, what is the advantage of using Lasso regresison?
- What's the point of gradient descent?
- Why is K-fold CV more preferred than LOOCV? (one reason)
- What is over-fitting? How is it related to bias-variance trade off?
- Mr. Avash was trying to do some gradient descent and obtained the plot below (shame on you Avash :P). How could he improve his algorithm?