

## Team Evaluation

1. Brendan Rizzo
  - a. Worked with me, Eden Seo, to solve problem 2 part a and b, which is exercise 3.3
2. Richard Huang
  - a. Solved problem 1 part a by himself and explained to us how he got the answer.
3. Jiashang Cao
  - a. Solved problem 1 part b
4. **Eden Seo**
  - a. Worked with Brendan to solve problem 2 part a and b, and solved part c and d by myself.

1.

- a. *Derive the optimal  $w$  that minimizes  $E_2(w)$* 
  - i.  $E_2(w) = \|Xw - y\|^2 + \lambda \|w\|^2$
  - ii.  $E_2(w) = ((Xw)' - y')(Xw - y) + \lambda(w' * w) = (X'w' - y')(Xw - y) + \lambda(w' * w)$
  - iii.  $= X'w'Xw - X'w'y - y'Xw + y'y + \lambda(w' * w)$ , where  $y'Xw = X'w'y$
  - iv.  $= X'w'Xw - 2X'w'y + y'y + \lambda(w' * w)$
  - v. *To minimize  $E_2(w)$ , differentiate with respect to  $w$*
  - vi.  $2X'Xw - 2X'y + \lambda w$ , where  $\frac{d}{dx} X^T A X = 2AX$
  - vii.  $2X'Xw - 2X'y + \lambda w = 0 \Rightarrow 2X'Xw + \lambda w = 2X'y$
  - viii.  $w(2X'X + \lambda * I) = 2X'y$ , where  $I$  is an identity matrix
  - ix.  $w = \frac{2X'y}{2X'X + \lambda}$

Problem 1:

a.  $E_2(w) = \|Xw - y\|^2 + \lambda \|w\|^2$

$$E_2(w) = ((Xw)' - y')(Xw - y) + \lambda(w' * w)$$

$$= (X'w' - y')(Xw - y) + \lambda(w' * w)$$

$$= X'w'Xw - X'w'y - y'Xw + y'y + \lambda(w' * w), \quad y'Xw = X'w'y$$

To minimize  $E_2(w)$ , differentiate wrt  $w$

$$E_2(w) = w'X'Xw - 2w'X'y + y'y + \lambda(w' * w)$$

After differentiating wrt  $w$ :  $\frac{d}{dx} X^T A X = 2AX$

$$2X'Xw - 2X'y + \lambda w$$

$$2X'Xw - 2X'y + \lambda w = 0$$

$$2X'Xw + \lambda w = 2X'y$$

$$w(2X'X + \lambda \cdot I) = 2X'y$$

$$w = \frac{2X'y}{(2X'X + \lambda)}$$

optimal  $w = (2X'X + \lambda)^{-1} (2X'y)$

1.

(hand-written work is much easier to see, so here it is.)

- b. From 1a, we got optimal  $w = (2X^T X + \lambda)^{-1} (2X^T Y)$ .

In this case, the solution holds only when  $2X^T X + \lambda$  is

non singular, and since  $\lambda > 0$  and it is an user specific parameter, the new objective function overcame the singularity problem of  $X^T X$  using the parameter  $\lambda$ .

## 2. Exercise 3.3 in LFD

a. Show that  $H$  is symmetric.

$$\text{i. } H^T = (X(X^T X)^{-1} X^T)^T = (X^T (X^T X)^{-T} X) = (X^T (X X^T)^{-1} X) = \text{since } X^T X \text{ is invertible, } H^T = (X(X^T X)^{-1} X^T) = H$$

b. Show that  $H^K = H$  for any positive integer  $K$ .

$$\text{i. } H^K = (X(X^T X)^{-1} X^T)^K = \frac{X X^T}{X^T X} * \frac{X X^T}{X^T X} * \dots * \frac{X X^T}{X^T X} \text{ (for } K \text{ amount of time)}$$

$$\text{ii. since } X^T X \text{ is invertible, } \frac{X X^T}{X^T X} \text{ is equal to } \frac{X^T X}{X^T X} \text{ which is equal to } 1.$$

iii. So no matter how many time  $H$  is multiplied by itself, it will still be the same.

iv. If  $K = 2$

$$\begin{aligned} 1. & (X(X^T X)^{-1} X^T) * (X(X^T X)^{-1} X^T) = \\ & X(X^T X)^{-1} (X^T X) (X^T X)^{-1} X^T = X(X^T X)^{-1} X^T \end{aligned}$$

c. Show that Show that  $(I - H)^K = I - H$  for any positive integer  $K$ .

i. if  $K = 2$

$$1. (I - H)^2 = (I - H)(I - H) = II - IH - HI + H^2 = I - 2H + H^2, \text{ and since } H^2 = H, I - 2H + H = I - H$$

ii. Since  $I^K$ , where  $K$  is a positive integer, is equal to  $I$  and  $H^K$  is always equal to  $H$ ,

iii. it would always become  $I - H$

d. Show that  $\text{trace}(H) = d + 1$

i. Lets say that  $A = X(X^T X)^{-1}$  and  $B = X^T$

$$\begin{aligned} 1. & \text{trace}(AB) = \text{trace}(BA) \Rightarrow \text{trace}(X(X^T X)^{-1} X^T) = \\ & \text{trace}(X X^T (X^T X)^{-1}) \text{ which is equal to } \text{trace}(I_{(d+1) \times (d+1)}) = \\ & d + 1 \end{aligned}$$

$$2. \text{So } \text{trace}(H) = d + 1$$