

CS221 Fall 2017 Homework [Foundations]

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By turning in this assignment, I agree by the Stanford honor code and declare that all of this is my own work.

Problem 1

(a) Let $\theta^* = \arg \min_{\theta} f(\theta)$, then

$$\frac{\partial f(\theta^*)}{\partial \theta^*} = \sum_{i=1}^n w_i (\theta^* - x_i) = 0 \quad (1)$$

$$\theta^* \sum_{i=1}^n w_i = \sum_{i=1}^n w_i x_i \quad (2)$$

$$\theta^* = \frac{\sum_{i=1}^n w_i x_i}{\sum_{i=1}^n w_i} \quad (3)$$

Without loss of generality, we can assume $\sum_{i=1}^n w_i = 1$, then

$$\theta^* = \mathbb{E}_{\mathbf{w}}[\mathbf{x}] \quad (4)$$

In order to prove θ^* is a global minimum, we need to show

$$\frac{\partial^2 f(\theta^*)}{\partial \theta^{*2}} = \sum_{i=1}^n w_i \quad (5)$$

Note that Eqn(5) is strictly positive if all w_i are positive, so θ^* should be a global minimum, while when some w_i is negative, $\sum_{i=1}^n w_i$ may be negative, then the global minimum doesn't hold. \square

(b)

$$f(\mathbf{x}) = \sum_{i=1}^d \max_{s \in \{-1,1\}} s x_i = \sum_{i=1}^d |x_i| \quad (6)$$

$$g(\mathbf{x}) = \max_{s \in \{-1,1\}} \sum_{i=1}^d s x_i = \max_{s \in \{-1,1\}} s \sum_{i=1}^d x_i = \left| \sum_{i=1}^d x_i \right| \quad (7)$$

According to Triangular Inequality, we can conclude that $f(\mathbf{x}) \geq g(\mathbf{x})$ \square

(c) Let \mathbf{X}_i = (number of points you roll the i-th time), \mathbf{N} = (number of rolls), then

$$\begin{aligned}\mathbb{E}\left[\sum_{i=1}^{\mathbf{N}} \mathbf{X}_i\right] &= \sum_{n=0}^{\infty} f_{\mathbf{N}}(n+1) \mathbb{E}[\mathbf{X}|n+1] \\ &= \sum_{n=0}^{\infty} \frac{1}{6} \left(\frac{5}{6}\right)^n \frac{b-a}{5} n \\ &= \frac{b-a}{30} \sum_{n=0}^{\infty} n \left(\frac{5}{6}\right)^n \\ &= b-a\end{aligned}$$

□

(d) Let $M(p) = \log L(p)$, then

$$M(p) = 4 \log p + 3 \log (1-p) \quad (8)$$

At $M(p)$'s maximum, which is also $L(p)$'s maximum we have

$$\frac{\partial M(p)}{\partial p} = \frac{4}{p} - \frac{3}{1-p} = 0 \quad (9)$$

$$p = \frac{4}{7} \quad (10)$$

Intuitively, in order to get 4 Heads in 7 flips with max probability (on average $\frac{4}{7}$ Head in one flip), we should let the probability of Heads in one flip, which is the Expectation of Heads in one flip be exactly $\frac{4}{7}$ (because $L(p)$ is a continuous function and Expectation of Heads in one flip should be where the probability is the maximum. □

(e)

$$\frac{\partial f}{\partial w_k} = \sum_{i=1}^n \sum_{j=1}^n 2(a_{ik} - b_{jk})(\mathbf{a}_i^\top \mathbf{w} - \mathbf{b}_j^\top \mathbf{w}) + 2\lambda w_k \quad (11)$$

$$\frac{\partial f}{\partial \mathbf{w}} = 2\left(\sum_{i=1}^n \sum_{j=1}^n (\mathbf{a}_i^\top \mathbf{w} - \mathbf{b}_j^\top \mathbf{w})(\mathbf{a}_i - \mathbf{b}_j) + \lambda \mathbf{w}\right) \quad (12)$$

□

Problem 2

(a) Start from 1-dimensional case with only one line segment to be placed, then the number of cases is

$$\sum_{i=1}^n n = \frac{n(n+1)}{2} \quad (13)$$

With 6 independent parts and two independent axis, the total number should be

$$((\frac{n(n+1)}{2})^6)^2 = \frac{n(n+1)}{2}^{12} \quad (14)$$

And

$$\lim_{n \rightarrow \infty} \frac{n(n+1)}{2}^{12} = O(n^{24}) \quad (15)$$

□

(b) See prob2-2.py for the code. The complexity is $O(n^2)$. □

(c)

$$f(n) = \begin{cases} 1, n = 0, 1 \\ \sum_{i=0}^{n-1} f(i), n \geq 2 \end{cases} \quad (16)$$

Then, we can prove $f(n)$ has a more direct form:

$$f(n) = 2^{n-1}, n \in \mathbb{N}^+ \quad (17)$$

We can prove that by induction:

Base Case: Obviously, we have $f(1) = 2^0 = 1$

Inductive Step:

$$f(n) = \sum_{i=1}^{n-1} 2^{i-1} + f(0) = 2^{n-1} - 1 + 1 = 2^{n-1} \quad (18)$$

□

(d)

$$\sum_{i=1}^n \sum_{j=1}^n (\mathbf{a}_i^\top \mathbf{w} - \mathbf{b}_j^\top \mathbf{w})^2 + \lambda \|\mathbf{w}\|_2^2 = \sum_{i=1}^n \sum_{j=1}^n (\mathbf{w}^\top \mathbf{a}_i \mathbf{a}_i^\top \mathbf{w} - \mathbf{w}^\top \mathbf{a}_i \mathbf{b}_j^\top \mathbf{w} + \mathbf{w}^\top \mathbf{b}_j \mathbf{b}_j^\top \mathbf{w}) + \lambda \|\mathbf{w}\|_2^2 \quad (19)$$

$$= \mathbf{w}^\top \left(\sum_{i=1}^n \sum_{j=1}^n (\mathbf{a}_i \mathbf{a}_i^\top - \mathbf{a}_i \mathbf{b}_j^\top + \mathbf{b}_j \mathbf{b}_j^\top) \right) \mathbf{w} + \lambda \|\mathbf{w}\|_2^2 \quad (20)$$

See prob2-4.py for the code. □