

# Probability and Statistics Homework 13

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1. Let  $X_1, \dots, X_n$  be a random sample from a distribution  $X$  with the density function

$$f(x; \theta) = \begin{cases} \frac{1}{2}(1 + \theta x), & x \in [-1, 1] \\ 0, & x \notin [-1, 1] \end{cases}$$

that depends on the parameter  $\theta$ . Find  $k$  such that  $k\bar{X}_n$  is an unbiased estimator for  $\theta$ , where

$$\bar{X}_n = \frac{X_1 + \dots + X_n}{n}$$

**Answer:**

Since  $k\bar{X}_n$  is an unbiased estimator for  $\theta$ , we have

$$E[k\bar{X}_n] - \theta = 0$$

$$\begin{aligned} E[k\bar{X}_n] - \theta &= kE\left[\frac{X_1 + \dots + X_n}{n}\right] - \theta \\ &= \frac{k}{n} \times nE[X_1] - \theta \quad (\text{i.i.d.}) \\ &= k \int_{-1}^1 \frac{1}{2}x(1 + \theta x) dx - \theta \\ &= \frac{\theta}{3}k - \theta = 0 \Rightarrow k = 3 \end{aligned}$$

Therefore,  $k = 3$  if  $k\bar{X}_n$  is an unbiased estimator for  $\theta$ .

2. Let  $X_1, \dots, X_n$  be a random sample from a distribution  $X$  with the density function (that depends on  $(\lambda, \theta)$ )

$$f(x; \theta) = \begin{cases} \lambda e^{-\lambda(x-\theta)}, & x > \theta \\ 0, & x \leq \theta \end{cases}$$

with  $\lambda > 0$  and  $\theta \in \mathcal{R}$ .

1. Find the methods of maximum likelihood estimator for  $(\lambda, \theta)$ .
2. Find the corresponding estimate of  $(\lambda, \theta)$  when  $n = 10$  and  $x_1 = 3, x_2 = 0.5, x_3 = 2.5, x_4 = 2, x_5 = 5, x_6 = 3.5, x_7 = 10, x_8 = 9, x_9 = 18, x_{10} = 1.5$ .

**Answer:**

1. Given the random sample  $X_1, \dots, X_n$ , the likelihood function is

$$L(\lambda, \theta) = \prod_{i=1}^n f(X_i; \lambda, \theta)$$

Since  $f(x; \lambda, \theta) = 0$  for  $x \leq \theta$ , we require  $\theta \leq \min(X_i)$ . For  $x_i > \theta$ , we have

$$\begin{aligned} L(\lambda, \theta) &= \prod_{i=1}^n f(X_i; \lambda, \theta) \\ &= \lambda^n e^{-\lambda \sum_{i=1}^n (X_i - \theta)} \end{aligned}$$

Taking the log:

$$\begin{aligned} l(\lambda, \theta) &= n \ln \lambda - \lambda \sum_{i=1}^n (X_i - \theta) \\ &= n \ln \lambda + \lambda n \theta - \lambda \sum_{i=1}^n X_i \end{aligned}$$

Since  $\lambda > 0$  and  $n > 0$ , increase  $\theta$  will increase the above formula. For that we have the constrain of the max value of  $\theta$ , thus  $\theta^* = \min(X_i)$ .

Then we maximize the likelihood with respect to  $\lambda$  given  $\theta^* = \min(X_i)$ . Consider the derivative of  $\lambda$  and set it to zero

$$\begin{aligned} \frac{\partial l}{\partial \lambda} &= \frac{n}{\lambda} + n\theta^* - \sum_{i=1}^n X_i = 0 \\ \Rightarrow \lambda^* &= \frac{n}{\sum_{i=1}^n (X_i - \theta^*)} \end{aligned}$$

Therefore, the maximum likelihood estimator for  $(\lambda, \theta)$  is  $\theta^* = \min(X_1, \dots, X_n), \lambda^* = \frac{n}{\sum_{i=1}^n (X_i - \theta^*)}$ .

**2.** First, for  $\theta$ , with the answer of part 1, we have

$$\theta^* = \min(X_i) = 0.5$$

Then we can find  $\lambda^*$  as

$$\begin{aligned}\lambda^* &= \frac{n}{\sum_{i=1}^n (X_i - \theta^*)} \\ &= \frac{10}{\sum_{i=1}^{10} X_i - 5} \\ &= \frac{10}{55 - 5} \\ &= 0.2\end{aligned}$$

Therefore, the MLE for  $(\lambda, \theta)$  given samples is  $\theta^* = 0.5, \lambda^* = 0.2$ .

3. Let  $x_1, x_2, \dots, x_n$  be independent and identically distributed samples from uniform distribution on the set  $[0, \theta]$ . (These values might look like  $x_1 = 2.325, x_2 = 1.1242, x_3 = 9.262$ , etc...) What is the MLE of  $\theta$ ?

**Answer:**

The uniform distribution on  $[0, \theta]$  has PDF

$$f(x; \theta) = \begin{cases} \frac{1}{\theta}, & 0 \leq x \leq \theta \\ 0, & \text{otherwise} \end{cases}$$

where  $\theta > 0$ . Given  $x_1, \dots, x_n$  i.i.d. samples, the likelihood function is

$$\begin{aligned} L(\theta) &= \prod_{i=1}^n f(x_i; \theta) \\ &= \prod_{i=1}^n \frac{1}{\theta} \times \mathbf{1}(0 \leq x_i \leq \theta) \\ &= \frac{1}{\theta^n} \mathbf{1}(\theta \geq \max(x_1, \dots, x_n)) \end{aligned}$$

To let  $L(\theta) > 0$ ,  $\theta$  must be large or equal than  $\max(x_1, \dots, x_n)$ . So that the formula turns to

$$L(\theta) = \frac{1}{\theta^n}$$

The formula increase with a decreasing  $\theta$ , since we have a constrain of  $\theta \geq \max(x_1, \dots, x_n)$ ,  $\theta^* = \max(x_1, \dots, x_n)$ .

Therefore, the MLE of  $\theta$  is  $\theta^* = \max(x_1, \dots, x_n)$