

## Homework - 10

① To find the Method of moments estimators for the parameters  $\theta_1$  &  $\theta_2$  based on the given uniform distributions.

Given that  $Y_1, Y_2, \dots, Y_m$  are i.i.d uniform  $(0, \theta_1]$ , the first moment (mean) of  $Y$  is given by:

$$E[Y] = \frac{\theta_1}{2}$$

Similarly,  $Z_1, Z_2, \dots, Z_n$  are i.i.d uniform  $[\theta_1, \theta_2]$ , the first moment (mean) of  $Z$  is given by:

$$E[Z] = \frac{\theta_1 + \theta_2}{2}$$

Method of Moments for  $\theta_1$ :

Sample mean for  $Y$  is given by:  $\bar{Y} = \frac{1}{m} \sum_{i=1}^m Y_i$

By the method of moments, we equate the Sample mean to the population mean, we have  $\frac{\theta_1}{2} = \bar{Y}$

$$\Rightarrow \theta_1 = 2\bar{Y}$$

Method of Moments for  $\theta_2$ :

Sample mean for  $Z$  is given by  $\bar{Z} = \frac{1}{n} \sum_{i=1}^n Z_i$

from method of moments,  $\frac{\theta_1 + \theta_2}{2} = \bar{Z}$

we know  $\theta_1 = 2\bar{Y}$

$$2\bar{Y} + \theta_2 = 2\bar{Z}$$

$$\theta_2 = 2\bar{Z} - 2\bar{Y}$$

$$\theta_2 = 2(\bar{Z} - \bar{Y})$$

Therefore, the method of moments estimator for  $\theta_1$  is

$$\Rightarrow \boxed{\hat{\theta}_{1MM} = 2\bar{Y}}$$

The method of moments estimator for  $\theta_2$  is

$$\Rightarrow \boxed{\hat{\theta}_{2MM} = 2(\bar{Z} - \bar{Y})}$$

② Given PDF for a Rayleigh distributed random Variable  $X$  is :

$$f(x|\sigma^2) = \frac{x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}}$$

To find Maximum likelihood Estimator (MLE) of  $\sigma^2$

A function  $f(x)$  is a valid PDF if  $\int f(x)dx = 1$

Let  $x_1, x_2, \dots, x_n$  be a random sample. The likelihood function 'L' is the product of the individual PDFs of the observations:

$$L = \prod_{i=1}^n f(x_i)$$

$$L(\sigma^2) = \prod_{i=1}^n \left( \frac{x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}} \right)$$

Taking log on both sides

$$\log L(\sigma^2) = \sum_{i=1}^n \log \left( \frac{x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}} \right)$$

$$\log L(\sigma^2) = \sum_{i=1}^n \left( \log x_i - \log \sigma^2 - \frac{x_i^2}{2\sigma^2} \right)$$

$$\log L(\sigma^2) = \sum_{i=1}^n \log x_i - n \log \sigma^2 - \frac{1}{2\sigma^2} \sum_{i=1}^n x_i^2$$

Now, differentiating with respect to  $\sigma^2$  & setting it to zero to find the MLE

$$\frac{d}{d\sigma^2} \log L(\sigma^2) = 0$$

$$0 = \frac{n}{\sigma^2} + \frac{1}{2(\sigma^2)^2} \sum_{i=1}^n x_i^2 = 0$$

$$\sigma^2 = \frac{1}{2n} \sum_{i=1}^n x_i^2$$

Hence, the MLE of  $\sigma^2$  for a Rayleigh Distribution is

$$\Rightarrow \boxed{\hat{\sigma}^2 = \frac{1}{2n} \sum_{i=1}^n x_i^2}$$