# Statistical Inference and Multivariate Analysis (MA324)

Lecture 14

Maximum Likelhood Estimator: Examples



Indian Institute of Technology Guwahati

Jan-May 2023

## Maximum Likelihood Estimator (MLE): More examples

Example 1: Let  $X_1, X_2, \ldots, X_n \overset{i.i.d.}{\sim} N(\mu, 1), \mu \leq 0$ . Thus, the parametric space is  $\Theta = (-\infty, 0]$ . Find the MLE of  $\mu$ .

Example 2: Let  $X_1$  be a sample of size one from  $Bernoulli(\frac{1}{1+e^{\theta}})$ , where  $\theta \geq 0$ . What is the MLE of  $\theta$ ?

Example 3: Let  $X_1, X_2, \ldots, X_n \stackrel{i.i.d.}{\sim} U(0, \theta), \theta > 0$ . What is the MLE of  $\theta$ ?

### Examples:

Example 4: Let  $X_1, X_2, \ldots, X_n \overset{i.i.d.}{\sim} U\left(\theta - \frac{1}{2}, \theta + \frac{1}{2}\right), \theta \in \mathbb{R}$ . The likelihood function is

$$\begin{split} L(\theta) &= 1 \text{ if } \theta - \frac{1}{2} \leq x_1, \, \dots, \, x_n \leq \theta + \frac{1}{2} \\ &= 1 \text{ if } x_{(n)} - \frac{1}{2} \leq \theta \leq x_{(1)} + \frac{1}{2}, \end{split}$$

where  $x_{(n)} = \max\{x_1, \ldots, x_n\}$  and  $x_{(1)} = \min\{x_1, \ldots, x_n\}$ . What is the MLE of  $\theta$ ?

#### Examples:

#### Theorem (Invariance Property of MLE)

If  $\hat{\theta}$  is MLE of  $\theta$ , then for any function  $\tau(\cdot)$  defined on  $\Theta$ , the MLE of  $\tau(\theta)$  is  $\tau(\hat{\theta})$ .

Example 5: Let  $X_1,\,X_2,\,\ldots,\,X_n\stackrel{i.i.d.}{\sim}P(\lambda),\,\lambda>0.$  To find the MLE of  $P(X_1=0),$  we can proceed as follows. Note that  $P(X_1=0)=e^{-\lambda}$  and we know that the MLE of  $\lambda$  is  $\bar{X}$ . Hence, the MLE of  $P(X_1=0)$  is  $e^{-\bar{X}}$ .

#### **Theorem**

Let T be a sufficient statistics for  $\theta$ . If a unique MLE exist for  $\theta$ , it is a function of T. If MLE of  $\theta$  exist but is not unique, then one can find a MLE that is a function of T only.

#### Examples:

Example 6: Let  $X_1, X_2, \ldots, X_n \overset{i.i.d.}{\sim} U(0, \theta), \theta > 0$ . We know that the MLE is unique and  $X_{(n)}$ , which is also sufficient. Thus, the unique MLE is a function of sufficient statistic in this case.

Example 7: Let  $X_1, X_2, \ldots, X_n \overset{i.i.d.}{\sim} U(\theta - \frac{1}{2}, \theta + \frac{1}{2}), \theta \in \mathbb{R}$ . Note that (can be shown) a sufficient statistic for  $\theta$  is  $\mathbf{T} = (X_{(1)}, X_{(n)})$ . Also, we have seen in Example 4 that MLE exists but is not unique. Any point in the interval  $\left[X_{(n)} - \frac{1}{2}, X_{(1)} + \frac{1}{2}\right]$  is a MLE of  $\theta$ . Hence,  $\frac{1}{2}\left(X_{(1)} + X_{(n)}\right)$  is a MLE and it is also a function of  $\mathbf{T}$ . On the other hand,  $Q = \left(\sin^2 X_1\right)\left(X_{(n)} - \frac{1}{2}\right) + \left(1 - \sin^2 X_1\right)\left(X_{(1)} - \frac{1}{2}\right)$  is also a MLE but not a function of  $\mathbf{T}$  only.