# **DISCUSSION WORKSHEET 06**

**PSTAT 120B:** Mathematical Statistics, I **Summer Session A, 2024** with Instructor: Ethan P. Marzban



## Conceptual Review

- (a) What is a likelihood? What about a log-likelihood?
- (b) How do we obtain a **maximum likelihood estimator** for a parameter? What do we do if the likelihood is nondifferentiable in the parameter of interest?
- (c) What is a **sufficient statistic**? How can the **factorization theorem** help us find a statistic that is sufficient for a given parameter?

## Problem 1

Let  $Y_1,\cdots,Y_n \overset{\text{i.i.d.}}{\sim} \mathcal{N}(\mu,\sigma^2)$ , where both  $\mu \in \mathbb{R}$  and  $\sigma^2 > 0$  are unknown parameters.

(a) Derive an expression for  $\mathcal{L}_{\vec{Y}}(\mu, \sigma^2)$ , the likelihood of the sample  $\vec{Y}$ . Recall that, since our sample is assumed to be i.i.d.,

$$\mathcal{L}_{\vec{Y}}(\mu,\sigma^2) = \prod_{i=1}^n f_{Y_i}(y_i;\mu,\sigma^2)$$

(b) Derive an expression for  $\ell_{\vec{Y}}(\mu,\sigma^2)$ , the log-likelihood of the sample  $\vec{Y}.$  Also compute

 $\frac{\partial}{\partial \mu} \ell_{\vec{\boldsymbol{Y}}}(\mu,\sigma^2) \qquad \text{and} \qquad \frac{\partial}{\partial \sigma^2} \ell_{\vec{\boldsymbol{Y}}}(\mu,\sigma^2)$ 

- (c) In the two derivatives you found in part (b), replace all instances of  $\mu$  with  $\widehat{\mu}_{\text{MLE}}$ , all instances of  $\sigma^2$  with  $\widehat{\sigma^2}_{\text{MLE}}$ . Set the resulting expressions equal to zero and solve for  $\widehat{\mu}_{\text{MLE}}$  and  $\widehat{\sigma^2}_{\text{MLE}}$ .
- (d) The **equivariance** property of maximum likelihood estimators is as follows: given the MLE  $\hat{\theta}_{\text{MLE}}$  for a parameter  $\theta$ , the MLE of  $\tau(\theta)$  [where  $\tau(\cdot)$  is an appropriatelybehaved function] is  $\tau(\hat{\theta}_{\text{MLE}})$ . For example, the MLE of  $\theta^3$  is  $(\hat{\theta}_{\text{MLE}})^3$ .

Use the equivariance property and your answer to part (c) to derive an expression for the maximum likelihood estimator of  $\sigma$ , the population *standard deviation*.

#### Problem 2

Something's gone awry with *GauchoPop*'s newest bottling machine! Specifically, the new soda dispenser doesn't fill each bottle entirely - rather, the proportion  $Y_i$  of a randomly-selected bottle that is full of soda follows the distribution with density

$$f(y;\theta) = \theta y^{\theta-1} \cdot \mathbb{1}_{\{0 \le y \le 1\}}$$

where  $\theta > 0$  is an unknown parameter. Let  $Y_1, \dots, Y_n$  denote the proportion of fill contained in n randomly-selected *GauchoPop* bottles.

- (a) Find  $\hat{\theta}_{MLE}$ , the maximum likelihood estimator for  $\theta$ .

**Hint:** Use the equivariance property

#### Problem 3

Let  $Y_1, \cdots, Y_n \overset{\text{i.i.d.}}{\sim} \text{Unif}[\theta, 1]$  where  $\theta < 1$  is an unknown parameter.

- (a) Find  $\hat{\theta}_{MM}$ , the method of moments estimator for  $\theta$ .
- (b) Show that the likelihood of the sample  $ec{Y}$  is given by

$$\mathcal{L}_{\vec{\mathbf{Y}}}(\theta) = \left(\frac{1}{1-\theta}\right)^n \cdot \mathbb{1}_{\{\theta \leq Y_{(1)}\}} \cdot \mathbb{1}_{\{Y_{(n)} \leq 1\}}$$

where  $Y_{(1)}$  denotes the first order statistic (sample minimum) and  $Y_{(n)}$  denotes the  $n^{\rm th}$  order statistic (sample maximum). Justify your logic.

- (c) Find  $\widehat{\theta}_{MLE}$ , the maximum likelihood estimator for  $\theta$ .
- (d) Find the exact sampling distribution of  $\hat{\theta}_{MLE}$ , and use this to determine whether or not  $\hat{\theta}_{MLE}$  is an unbiased estimator for  $\theta$ .
- (e) Show that  $U := Y_{(1)}$ , the first order statistic, is a sufficient sufficient statistic for  $\theta$ .