

# Midterm Exam

## *Suggested Answers*

**Instructions:** This is a closed book exam, but you may refer to one sheet of notes. You have 80 minutes for the exam. Answer as many questions as possible. Partial answers get partial credit. Please write legibly. *Good luck!*

**Problem 1 (5 points).** Determine whether or not the statement below is correct and give a *brief* (e.g., a bluebook page or less) justification for your answer.

If  $(X, Y, Z)'$  is a (trivariate) random vector with  $E_X(|X|) < \infty$ , then

$$E_Y[|E_{X|Y}(X|Y)|] \leq E_{Y,Z}[|E_{X|Y,Z}(X|Y, Z)|].$$

*The statement is correct. By the law of iterated expectations,*

$$E_{X|Y}(X|Y) = E_{Z|Y}[E_{X|Y,Z}(X|Y, Z)|Y],$$

*and therefore*

$$|E_{X|Y}(X|Y)| = |E_{Z|Y}[E_{X|Y,Z}(X|Y, Z)|Y]| \leq E_{Z|Y}[|E_{X|Y,Z}(X|Y, Z)|Y|].$$

*Taking expectations, we have*

$$E_Y[|E_{X|Y}(X|Y)|] \leq E_Y(E_{Z|Y}[|E_{X|Y,Z}(X|Y, Z)|Y|]) = E_{Y,Z}[|E_{X|Y,Z}(X|Y, Z)|],$$

*where the last equality uses the law of iterated expectations.*

**Problem 2 (45 points, each part receives equal weight).** Let  $X_1, \dots, X_n$  be a random sample from a continuous distribution with pdf

$$f_X(x|\theta) = c(\theta)x^{-1/2}1(0 \leq x \leq \theta^2),$$

where  $\theta \in \Theta = (0, \infty)$  is an unknown parameter,  $1(\cdot)$  is the indicator function, and  $c(\cdot)$  is some function.

(a) Show that

$$c(\theta) = \frac{1}{2\theta}.$$

The function  $c(\cdot)$  is such that  $\int_{-\infty}^{\infty} f_X(x|\theta) dx = 1$ , so

$$c(\theta) = \frac{1}{\int_{-\infty}^{\infty} x^{-1/2}1(0 \leq x \leq \theta^2)dx} = \frac{1}{2\theta},$$

where the second equality uses

$$\int_{-\infty}^{\infty} x^{-1/2}1(0 \leq x \leq \theta^2)dx = \int_0^{\theta^2} x^{-1/2}dx = 2\sqrt{x}\Big|_{x=0}^{\theta^2} = 2\theta.$$

(b) Find  $F_X(\cdot|\theta)$ , the cdf of  $X$ .

For  $x \in [0, \theta^2]$ , we have:

$$\int_{-\infty}^x f_X(r|\theta) dr = \frac{1}{2\theta} \int_0^x r^{-1/2}dr = \frac{1}{2\theta} 2\sqrt{r}\Big|_{r=0}^x = \frac{1}{\theta}\sqrt{x}.$$

As a consequence,

$$F_X(x|\theta) = \begin{cases} 0 & \text{if } x < 0, \\ \frac{1}{\theta}\sqrt{x} & \text{if } 0 \leq x < \theta^2, \\ 1 & \text{if } x \geq \theta^2. \end{cases}$$

(c) Derive a method moments estimator  $\hat{\theta}_{MM}$  of  $\theta$ . Is  $\hat{\theta}_{MM}$  an unbiased estimator of  $\theta$ ?

Because

$$E_{\theta}(X_i) = \int_{-\infty}^{\infty} x f_X(x|\theta) dx = \frac{1}{2\theta} \int_0^{\theta^2} x^{1/2} dx = \frac{1}{2\theta} \left. \frac{2}{3} x^{3/2} \right|_{x=0}^{\theta^2} = \frac{1}{3} \theta^2,$$

a method of moments estimator satisfies

$$\hat{\theta}_{MM} = \sqrt{3\bar{X}}, \quad \bar{X} = \frac{1}{n} \sum_{i=1}^n X_i.$$

Because  $E_{\theta}(\hat{\theta}_{MM}^2) = \theta^2$  and  $\text{Var}_{\theta}(\hat{\theta}_{MM}) > 0$ ,

$$E_{\theta}(\hat{\theta}_{MM})^2 = E_{\theta}(\hat{\theta}_{MM}^2) - \text{Var}_{\theta}(\hat{\theta}_{MM}) = \theta^2 - \text{Var}_{\theta}(\hat{\theta}_{MM}) < \theta^2.$$

In particular,  $\hat{\theta}_{MM}$  is a biased estimator of  $\theta$ .

(d) Find the likelihood function. Does  $\theta$  admit a scalar sufficient statistic?

Defining  $X_{(n)} = \max_{1 \leq i \leq n} X_i$ , the likelihood function can be written as

$$L(\theta|X_1, \dots, X_n) = \prod_{i=1}^n f_X(X_i|\theta) = \prod_{i=1}^n \left\{ \frac{1}{2\theta} X_i^{-1/2} \underbrace{1(0 \leq X_i \leq \theta^2)}_{\equiv 1(X_i \leq \theta^2)} \right\} = 2^{-n} \theta^{-n} \left( \prod_{i=1}^n X_i \right)^{-1/2} \cdot 1(X_{(n)} \leq \theta^2).$$

It follows from the factorization criterion that the scalar  $X_{(n)}$  is a sufficient statistic for  $\theta$ .

(e) Show that

$$\hat{\theta}_{ML} = \sqrt{\max_{1 \leq i \leq n} X_i}$$

is the maximum likelihood estimator of  $\theta$ .

Because  $L(\theta|X_1, \dots, X_n) = 0$  for  $\theta < X_{(n)}$  and because  $\theta^{-n}$  is a decreasing function of  $\theta$ ,

$$\arg \max_{\theta \in \Theta} L(\theta|X_1, \dots, X_n) = \sqrt{X_{(n)}} = \hat{\theta}_{ML}.$$

(f) Find  $F_{ML}(\cdot|\theta)$ , the cdf of  $\hat{\theta}_{ML}$ .

Clearly,  $F_{ML}(x|\theta) = 0$  for  $x < 0$ . If  $x \geq 0$ , then

$$P_{\theta}(\hat{\theta}_{ML} \leq x) = P_{\theta}(\sqrt{X_{(n)}} \leq x) = P_{\theta}(X_{(n)} \leq x^2) = F_X(x^2|\theta)^n = \begin{cases} (x/\theta)^n & \text{if } 0 \leq x < \theta, \\ 1 & \text{if } x \geq \theta, \end{cases}$$

where the last equality uses part (b).

In other words,

$$F_{ML}(x|\theta) = \begin{cases} 0 & \text{if } x < 0, \\ (x/\theta)^n & \text{if } 0 \leq x < \theta, \\ 1 & \text{if } x \geq \theta. \end{cases}$$

It can be shown that  $\hat{\theta}_{ML}$  is complete.

(g) Find a uniform minimum variance unbiased estimator of  $\theta$ .

A pdf  $f_{ML}(\cdot|\theta)$  of  $\hat{\theta}_{ML}$  is given by

$$f_{ML}(x|\theta) = \frac{n}{\theta^n} x^{n-1} 1(0 \leq x \leq \theta).$$

As a consequence,

$$E_{\theta}(\hat{\theta}_{ML}) = \int_{-\infty}^{\infty} x f_{ML}(x|\theta) dx = \int_0^{\theta} \frac{n}{\theta^n} x^n dx = \frac{n}{\theta^n} \frac{1}{n+1} x^{n+1} \Big|_{x=0}^{\theta} = \frac{n}{n+1} \theta.$$

Therefore,

$$\hat{\theta}_{UMVU} = \frac{n+1}{n} \hat{\theta}_{ML}$$

is an unbiased estimator of  $\theta$ . In fact, because  $\hat{\theta}_{ML}$  is a complete sufficient statistic for  $\theta$ ,  $\hat{\theta}_{UMVU}$  is a uniform minimum variance unbiased estimator of  $\theta$ .

Let  $\theta_0 > 0$  be some constant and consider the one-sided testing problem

$$H_0 : \theta \leq \theta_0 \quad \text{vs.} \quad H_1 : \theta > \theta_0.$$

(h) Consider a test which rejects  $H_0$  if (and only if)  $\hat{\theta}_{ML} > c$ , where  $c$  is some positive constant (possibly depending on  $\theta_0$ ). Find  $c$  such that the test has 5% size.

The size of the test is  $\sup_{\theta \leq \theta_0} P_\theta(\hat{\theta}_{ML} > c)$ . Using part (f),

$$P_\theta(\hat{\theta}_{ML} > c) = 1 - P_\theta(\hat{\theta}_{ML} \leq c) = 1 - F_{ML}(c|\theta) = \begin{cases} 1 - (c/\theta)^n & \text{if } 0 \leq c < \theta, \\ 0 & \text{if } c \geq \theta, \end{cases} = \max[1 - (c/\theta)^n, 0]$$

for any  $c > 0$  and any  $\theta > 0$ . Since  $P_\theta(\hat{\theta}_{ML} > c)$  is non-decreasing in  $\theta$ ,

$$\sup_{\theta \leq \theta_0} P_\theta(\hat{\theta}_{ML} > c) = P_{\theta_0}(\hat{\theta}_{ML} > c) = \max[1 - (c/\theta_0)^n, 0],$$

so  $c$  must satisfy

$$(c/\theta_0)^n = 0.95 \quad \Leftrightarrow \quad c = \theta_0 \sqrt[n]{0.95}.$$

(i) Find the power function of the test derived in (h).

The power function is the function  $\beta : \Theta \rightarrow [0, 1]$  given by

$$\beta(\theta) = P_\theta[\hat{\theta}_{ML} > \theta_0 \sqrt[n]{0.95}] = \max[1 - (\theta_0 \sqrt[n]{0.95}/\theta)^n, 0] = \max[1 - 0.95(\theta_0/\theta)^n, 0],$$

where the second equality uses a result obtained in part (h).