

# MA582 Homework UC

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## 1 Problem 1

Suppose a parameter family has parameter  $\theta = E(X^3)$ , where  $\theta$  is real. Devise an estimator of  $\theta$  and show it is UC for  $\theta$ .

*Solution.* Let  $Y_n = X_n^3$  and consider

$$\hat{\theta}_n = \frac{Y_1 + \cdots + Y_n}{n} = \frac{X_1^3 + \cdots + X_n^3}{n}.$$

This estimator is unbiased for  $\theta$  because

$$E(\hat{\theta}_n) = \frac{1}{n} \sum_{i=1}^n E(Y_i) = \frac{1}{n} \sum_{i=1}^n \theta = \theta.$$

This estimator is consistent for  $\theta$  because the weak law of large numbers gives

$$\hat{\theta}_n = \frac{Y_1 + \cdots + Y_n}{n} \xrightarrow{p} \theta.$$

□

## 2 Problem 2

We say the rv  $X$  has the Weiner distribution with parameter  $\theta > 0$  if  $X$  has pdf  $f(x) = 2x/\theta^2$  for  $0 < x < \theta$  and  $f(x) = 0$  elsewhere.

Consider the parameterized Weiner family  $\{W(\theta) : \theta > 0\}$ .

- (a) Let  $Y_n$  be the maximum of the random sample of size  $n$  from  $W(\theta)$ . Show that  $Y_n$  is a consistent estimator of  $\theta$ .
- (b) Find the pdf of  $Y_n$ .
- (c) Show that  $Y_n$  is NOT an unbiased estimator of  $\theta$ .

- (d) Show how to “correct”  $Y_n$  here to make a new unbiased estimator  $T_n$ . Show  $T_n$  is also consistent, hence UC.
- (e) Obtain the asymptotic distribution of  $n(Y_n - \theta)$ .
- (f) Obtain the asymptotic distribution of  $n(T_n - \theta)$ .

*Solution.* (a) We must show  $Y_n \xrightarrow{P} \theta$ , meaning that for any  $\epsilon > 0$ , we have  $P(|Y_n - \theta| \geq \epsilon) \rightarrow 0$  as  $n \rightarrow \infty$ . As  $Y_n \leq \theta$ , this probability equals  $P(Y_n \leq \theta - \epsilon)$ . Since  $Y_n = \max(X_1, \dots, X_n)$ , the condition  $Y_n \leq \theta - \epsilon$  holds precisely when  $X_i \leq \theta - \epsilon$  for all  $i$ . As the  $X_i$ s are independent,

$$\begin{aligned} P(Y_n \leq \theta - \epsilon) &= \prod_{i=1}^n P(X_i \leq \theta - \epsilon) \\ &= \left( \int_0^{\theta - \epsilon} \frac{2x}{\theta^2} dx \right)^n \\ &= \left( \frac{(\theta - \epsilon)^2}{\theta^2} \right)^n \\ &= \left( 1 - \frac{\epsilon}{\theta} \right)^{2n}. \end{aligned}$$

As  $n \rightarrow \infty$ , this probability does indeed tend to 0.

- (b) By the same reasoning as before, the cdf of  $Y_n$  is

$$F_{Y_n}(y) = P(Y_n \leq y) = \left( \int_0^y \frac{2x}{\theta^2} dx \right)^n = \frac{y^{2n}}{\theta^{2n}}$$

for  $0 < y < \theta$ , with  $F_{Y_n}(y) = 0$  for  $y \leq 0$  and  $F_{Y_n}(y) = 1$  for  $y \geq \theta$ . The pdf of  $Y_n$  is then

$$f_{Y_n}(y) = F'_{Y_n}(y) = \frac{2n}{\theta^{2n}} \cdot y^{2n-1}$$

when  $0 < y < \theta$ , with  $f_{Y_n}(y) = 0$  for all other  $y$ .

- (c) The expected value of  $Y_n$  is

$$\begin{aligned} E(Y_n) &= \int_0^\theta y \cdot f_{Y_n}(y) dy = \int_0^\theta \frac{2n}{\theta^{2n}} y^{2n} dy \\ &= \frac{2n}{(2n+1)\theta^{2n}} \cdot \theta^{2n+1} = \frac{2n}{2n+1} \cdot \theta. \end{aligned}$$

In particular, this is not  $\theta$  itself, so  $Y_n$  is not an unbiased estimator for  $\theta$ .

- (d) If we define  $T_n = \frac{2n+1}{2n} Y_n$ , then  $E(T_n) = \frac{2n+1}{2n} E(Y_n) = \theta$ , so  $T_n$  is an unbiased estimator for  $\theta$ . Also, since  $\frac{2n+1}{2n} \rightarrow 1$  and  $Y_n \xrightarrow{P} \theta$  as  $n \rightarrow \infty$ , we have  $T_n \xrightarrow{P} 1 \cdot \theta = \theta$  as  $n \rightarrow \infty$ , so  $T_n$  is consistent for  $\theta$ . Hence  $T_n$  is UC for  $\theta$ .

- (e)

□