

**Problem Set 8**

Fall 2007

**Issued:** Thursday, October 25

**Due:** Thursday, November 1

**Reading:** Keener: Chapter 10, 11, §21.6, B & D: §3.5, §4.4

**Problem 8.1**

Find variance-stabilizing transformations for the following statistics and estimation problems:

- (a) The sample mean  $\bar{X}_n$  for i.i.d. samples from  $\text{Ber}(\theta)$ . (Here we want the transformed statistic to have asymptotic variance independent of  $\theta$ .)
- (b) The sample correlation coefficient  $r = \frac{\sum_{i=1}^n X_i Y_i}{\sqrt{\sum_{i=1}^n X_i^2} \sqrt{\sum_{i=1}^n Y_i^2}}$  for i.i.d. samples from the zero-mean bivariate normal with  $\sigma_X^2 = \sigma_Y^2 = 1$  and correlation coefficient  $\rho \in (-1, 1)$ . (Here we want the transformed statistic to have asymptotic variance independent of  $\rho$ .)

**Problem 8.2**

Let  $X_1, \dots, X_n$  be i.i.d. with  $\mathbb{E}[X_i] = \theta$ ,  $\text{var}(X_i) = 1$ , and  $\mathbb{E}[(X_i - \theta)^4] = \mu_4$ , and consider the unbiased estimators of  $\theta^2$  defined by

$$\delta_1(X) = \frac{1}{n} \sum_{i=1}^n X_i^2 - 1, \quad \text{and} \quad \delta_2(X) = (\bar{X}_n)^2 - \frac{1}{n}.$$

- (a) Determine the asymptotic relative efficiency (ARE) of  $\delta_1$  compared to  $\delta_2$ .
- (b) Show that  $\delta_2$  is superior in terms of ARE if the  $X_i$  are symmetric about  $\theta$ .
- (c) Find a distribution for which  $\delta_1$  is superior in terms of ARE.

**Problem 8.3**

Recall that a statistical function  $h$  is said to be continuous at  $F$  in the sup-norm topology if, for any sequence  $\{G_n\}$  such that  $\|G_n - F\|_\infty \rightarrow 0$ , there holds  $h(G_n) \rightarrow h(F)$ . Are the following statistical functionals continuous in this sense or not? Prove or disprove.

- (a) for fixed  $a \in \mathbb{R}$ , the evaluation  $h(F) = F(a)$ .
- (b) for fixed CDF  $F_0$  with density  $f_0$ , the Cramér-von Mises functional given by  $h(F) = \int [F(t) - F_0(t)]^2 f_0(t) dt$ .
- (c) the mean functional  $h(F) = \mathbb{E}_F[X]$ .
- (d) the quantile functional  $h_\alpha(F)$  that returns the level  $\alpha$  quantile of  $F$ , when the inverse CDF  $F^{-1}(t) = \inf\{x \in \mathbb{R} \mid F(x) \geq t\}$  is continuous at  $\alpha$ .

**Problem 8.4**

Let  $X_1, \dots, X_n$  be an i.i.d. sample of  $\text{Uni}[0, 2\theta]$  variables, and let  $X_{(1)}, \dots, X_{(n)}$  denote the associated order statistics. Define for any  $p \in (0, 1)$ , the  $p^{\text{th}}$ -sample quantile  $X_{(np)} \equiv X_{(\lceil np \rceil)}$ , where for any  $x \in \mathbb{R}$ , the notation  $\lceil x \rceil$  denotes the smallest integer greater than or equal to  $x$ .

- Find the asymptotic distribution of  $X_{(\frac{1}{2}n)}$ .
- Find the asymptotic distribution of  $\frac{2}{3}X_{(3n/4)}$ .
- Find the asymptotic distribution of the midquantile range  $\frac{1}{2}[X_{(n/4)} + X_{(3n/4)}]$ .
- Compare these three statistics as estimators of  $\theta$  in terms of their asymptotic relative efficiencies.

*Note:* See §10.4 of Keener for some useful results.

**Problem 8.5**

Suppose we form an empirical CDF  $\widehat{F}_{n-1}$  based on a set of  $(n-1)$  samples  $x_1, \dots, x_{n-1}$ . Now consider the new empirical CDF  $\widehat{F}_{n,x}$  formed when the point  $x$  is added to  $\{x_1, \dots, x_{n-1}\}$  as the  $n^{\text{th}}$  sample. The *sensitivity function*  $S_{h,F}(x) = [h(\widehat{F}_{n,x}) - h(\widehat{F}_{n-1})]$  measures the effect of adding the single new sample at  $x$ .

- Compute the sensitivity  $S_{h,F}(x)$  when  $h(F) = \mathbb{E}_F[X]$  is the mean functional.
- Suppose that  $n = 2k + 1$  is odd, and that the sample median of  $\{x_1, \dots, x_{n-1}\}$  is given by  $\frac{1}{2}(x_{(k)} + x_{(k+1)})$ . Prove that the sensitivity for the median functional is given by

$$S_{h,F}(x) = \begin{cases} x_{(k)} & \text{for } x \leq x_{(k)} \\ x & \text{for } x_{(k)} \leq x \leq x_{(k+1)} \\ x_{(k+1)} & \text{for } x \geq x_{(k+1)} \end{cases}$$

**Problem 8.6**

Given  $n$  i.i.d. samples  $X_1, \dots, X_n \sim F$ , the *Huber estimate*  $\theta(\widehat{F}_n)$  of location is defined implicitly as the solution of the equation  $\frac{1}{n} \sum_{i=1}^n \psi_K(X_i - \theta) = 0$ , where for each  $K \geq 0$ , the function  $\psi_K$  has the form

$$\psi_K(t) = \begin{cases} t & \text{if } |t| \leq K \\ \text{sign}(t) K & \text{if } |t| > K \end{cases}$$

It can be viewed as the empirical version of the statistical function  $h(F)$  that outputs the solution  $\theta$  to the equation  $\mathbb{E}_F[\psi_K(X - \theta)] = 0$ . Assume that the c.d.f.  $F$  has a density  $f_X$  that is strictly positive on all  $\mathbb{R}$ .

- Show for  $K \rightarrow +\infty$ , the Huber estimator tends to the sample mean  $\bar{X}_n$ , and that for  $K = 0$ , the Huber estimator is the median.
- Show that  $\theta(F)$  exists and is unique for  $K > 0$ .
- Show that for any  $K < +\infty$ , then  $\lim_{|x| \rightarrow +\infty} S_{\theta,F}(x)$  is a finite constant. (The sensitivity  $S$  was defined in the previous problem.)