

## SOME ADMISSIBLE ESTIMATORS IN EXTREME VALUE DENSITIES

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Let  $X$  be a random variable having the extreme value density of the form

$$(1) \quad f(x; \theta) = \begin{cases} q(\theta)r(x), & 0 \leq x \leq \theta \\ 0, & \text{otherwise} \end{cases}$$

where  $r$  is assumed to be a positive Lebesgue measurable function of  $x$  and the function  $q$  is defined by

$$1/q(\theta) = \int_0^\theta r(x) dx < \infty$$

for all  $\theta$  in  $\Omega = (0, \infty)$ . It is further assumed that  $q(\theta)$  approaches zero as  $\theta \rightarrow \infty$ .

In this note we are concerned with estimating parametric functions  $g(\theta)$  of the form  $[1/q(\theta)]^\alpha$ ,  $\alpha$  any real number. The loss function is assumed to be squared error and the estimators are assumed to be functions of a single observation  $X$ . The case of estimators based on a sample of size  $n \geq 1$  is discussed in Remark 1.

In our search for a 'good' estimator for  $g(\theta) = [1/q(\theta)]^\alpha$  we calculate  $E[1/q(X)]^\alpha = \int_0^\theta [1/q(x)]^\alpha q(\theta)r(x) dx$ . Since  $r(x) = -q'(x)/q^2(x)$  almost everywhere we find that for every  $\alpha > -1$ ,  $E[1/q(X)]^\alpha$  exists and is given by  $E[1/q(X)]^\alpha = (1/\alpha + 1)[1/q(\theta)]^\alpha$ . This leads us to consider the class  $\Lambda_\alpha = \{\delta_K(X) = K[1/q(X)]^\alpha : K \text{ real}\}$  of estimators, which are constant multiples of  $[1/q(X)]^\alpha$ , for estimating the given parametric function  $[1/q(\theta)]^\alpha$ . Which of these estimators in  $\Lambda_\alpha$  has the smallest risk uniformly for all  $\theta$  in  $\Omega$ ? Since  $E[1/q(X)]^l = [1/(l+1)]^2 [1/q(\theta)]^l$  if  $l > -1$  and  $= \infty$  if  $l \leq -1$ , it follows easily that for any  $\delta_K$  in  $\Lambda_\alpha$ ,

$$(2) \quad R(\delta_K, \theta) = E[K(1/q(X))^\alpha - (1/q(\theta))^\alpha]^2 = \begin{cases} [1/q(\theta)]^{2\alpha}, & K = 0, \quad \text{all } \alpha \\ \left( \frac{K^2}{2\alpha+1} - \frac{2K}{\alpha+1} + 1 \right) [1/q(\theta)]^{2\alpha}, & \alpha > -\frac{1}{2}, \quad \text{all } K \\ \infty, & \alpha \leq -\frac{1}{2}, \quad K \neq 0 \end{cases}$$

where throughout this paper  $\infty$  stands for  $+\infty$ . If  $\alpha > -\frac{1}{2}$ , the quadratic expression  $[K^2/(2\alpha+1)] - [2K/(\alpha+1)] + 1$  in  $K$  achieves its minimum at  $K = (2\alpha+1)/(\alpha+1)$ . It follows from this that for estimating  $[1/q(\theta)]^\alpha$ ,  $\alpha > -\frac{1}{2}$ , the minimum risk estimator in  $\Lambda_\alpha$  is  $T_\alpha(X) = [(2\alpha+1)/(\alpha+1)][1/q(X)]^\alpha$  corresponding to  $K = (2\alpha+1)/(\alpha+1)$ .

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$(\alpha+1)$  with risk

$$(3) \quad R(\delta_K, \theta) = [\alpha/(\alpha+1)]^2 [1/q(\theta)]^{2\alpha}.$$

Is  $T_\alpha$  an admissible estimator of  $[1/q(\theta)]^\alpha$  for all  $\alpha$ ? We have the following

**THEOREM 1.** *Let the random variable  $X$  have density (1) and let the loss be quadratic. Then the estimator*

$$T_\alpha(X) = \frac{2\alpha+1}{\alpha+1} [1/q(X)]^\alpha$$

*is admissible for estimating  $[1/q(\theta)]^\alpha$  for every  $\alpha > -\frac{1}{2}$  and is inadmissible for all  $\alpha \leq -\frac{1}{2}$ .*

**Proof.** Assume  $\alpha > -\frac{1}{2}$ . Let  $T$  be any estimator satisfying the inadmissibility inequality for  $T_\alpha$ :

$$(4) \quad E[T - (1/q(\theta))^\alpha]^2 \leq E[T_\alpha - (1/q(\theta))^\alpha]^2$$

Writing  $m(\theta)$  for  $E(T)$  and  $m^*(\theta)$  for  $E(T_\alpha)$  we have the following equivalent inequalities:

$$(5) \quad E[T - m(\theta)]^2 + [m(\theta) - (1/q(\theta))^\alpha]^2 \leq [\alpha/(\alpha+1)]^2 [q(\theta)]^{-2\alpha},$$

and

$$(6) \quad E(T - T_\alpha)^2 + 2E\{[T - T_\alpha][T_\alpha - (1/q(\theta))^\alpha]\} \leq 0.$$

Inequality (5) implies that

$$[m(\theta) - (1/q(\theta))^\alpha]^2 \leq [\alpha/(\alpha+1)]^2 [1/q(\theta)]^{2\alpha}$$

from which we get the bounds for the function  $m$  as

$$(7) \quad \begin{aligned} \frac{2\alpha+1}{\alpha+1} \frac{1}{[q(\theta)]^\alpha} \leq m(\theta) \leq \frac{1}{\alpha+1} \frac{1}{[q(\theta)]^\alpha} & \text{ if } -\frac{1}{2} < \alpha < 0 \\ \frac{1}{\alpha+1} \frac{1}{[q(\theta)]^\alpha} \leq m(\theta) \leq \frac{2\alpha+1}{\alpha+1} \frac{1}{[q(\theta)]^\alpha} & \text{ if } \alpha \geq 0. \end{aligned}$$

Since  $1/q(\theta)$  tends to zero as  $\theta$  tends to zero, it is clear from (7) that  $m(\theta)/[q(\theta)]^{\delta-\alpha} \rightarrow 0$  for every  $\delta > 0$ . Now the hypothesis  $\alpha > -\frac{1}{2}$  guarantees some  $\delta > 0$  such that  $\alpha = (\delta/2) - (\frac{1}{2})$  i.e.,  $2\alpha + 1 = \delta$ , i.e.,  $\alpha + 1 = \delta - \alpha$ . Thus it follows that

$$(8) \quad m(\theta)/[q(\theta)]^{\alpha+1} \rightarrow 0 \text{ as } \theta \rightarrow 0$$

The rest of the proof consists in showing that the only solution of the inadmissibility inequality (6) is  $m = m^*$ . For this it is enough to show that  $m = m^*$  is the only solution to the inequality

$$(9) \quad [m(\theta) - m^*(\theta)]^2 + 2E\{[T - T_\alpha][T_\alpha - (1/q(\theta))^\alpha]\} \leq 0$$

which is relaxation of (6) obtained after replacing its LHS by something smaller. But (9) still has  $T$  in it. To express it in terms of  $m$  we use the identity  $m(\theta) = q(\theta) \int_0^\theta T(x)r(x) dx$  to provide us the relation

$$(10) \quad T(x) = \frac{m'(x)}{q(x)r(x)} + m(x)$$

Substituting this value of  $T$  in (9) and performing the expectation of the expression therein, we obtain the inequality

$$(11) \quad [m(\theta) - m^*(\theta)]^2 + \frac{2\alpha}{\alpha+1} \frac{m(\theta)}{[q(\theta)]^\alpha} - \frac{2\alpha(2\alpha+1)}{\alpha+1} E\left\{ \frac{m(X)}{[q(X)]^\alpha} \right\} \leq 0$$

where in this derivation integration by parts and result (8) is used. This inequality still contains the integral  $E\{m(X)/[q(X)]^\alpha\}$ . If we write

$$u(\theta) = E\left\{ \frac{m(X)}{[q(X)]^\alpha} \right\} = q(\theta) \int_0^\theta \frac{m(x)}{[q(x)]^\alpha} r(x) dx$$

we have

$$(12) \quad m(\theta) = u(\theta)q^\alpha(\theta) - \left[ \frac{q^{\alpha+1}(\theta)}{q'(\theta)} \right] u'(\theta).$$

Introducing  $u(\theta)$  in (11) we have the inequality

$$(13) \quad [m(\theta) - m^*(\theta)]^2 - \left[ \frac{4\alpha^2}{\alpha+1} \right] u(\theta) - \frac{2\alpha}{\alpha+1} \left[ \frac{q(\theta)}{q'(\theta)} \right] u'(\theta) \leq 0$$

wherein  $m(\theta)$  is to be replaced by its value in terms of  $u(\theta)$  from (12). It is now shown that  $u^*(\theta) = [1/(1+\alpha)^2][1/q(\theta)]^{2\alpha}$ , corresponding to  $m=m^*$ , is the unique solution of (13). For convenience we write

$$[1/q(\theta)]^{2\alpha} v(\theta) = u(\theta) - \frac{1}{(1+\alpha)^2} [1/q(\theta)]^{2\alpha}$$

in (13) which becomes

$$(14) \quad \left[ (1+2\alpha)v(\theta) - \frac{q(\theta)}{q'(\theta)} v'(\theta) \right]^2 - \frac{2\alpha}{\alpha+1} \frac{q(\theta)}{q'(\theta)} v'(\theta) \leq 0.$$

The proof now consists in showing that  $v(\theta) \equiv 0$  is the only solution of (14). This is done by using typical Hodges-Lehmann argument as follows:

(a)  $v'(\theta) \geq 0$  for  $-\frac{1}{2} < \alpha < 0$  and  $\leq 0$  for  $\alpha > 0$ . If  $v'(\theta) < 0$ , then, using the fact that  $q'(\theta) < 0$ , we find that the expression  $-[2\alpha/(\alpha+1)][q(\theta)/q'(\theta)]v'(\theta)$  is positive for  $-\frac{1}{2} < \alpha < 0$ . But then inequality (14) is violated.

Hence the assertion for  $-\frac{1}{2} < \alpha < 0$  follows. The conclusion for  $\alpha > 0$  follows likewise.

(b)  $v(\theta)$  is bounded. The inequality (7) for  $-\frac{1}{2} < \alpha < 0$  can be written as

$$\frac{2\alpha+1}{\alpha+1} \left[ \frac{1}{q(x)} \right]^\alpha \leq m(x) \leq \frac{1}{\alpha+1} \left[ \frac{1}{q(x)} \right]^\alpha$$

which after multiplying through by  $[q(\theta)r(x)]/[q^\alpha(x)]$  and integrating from 0 to  $\theta$  becomes

$$\frac{2\alpha+1}{\alpha+1} q(\theta) \int_0^\theta \frac{-q(x)}{[q(x)]^{2\alpha+2}} dx \leq u(\theta) \leq \frac{q(\theta)}{\alpha+1} \int_0^\theta \frac{-q(x)}{[q(x)]^{2\alpha+2}} dx$$

i.e.

$$\frac{1}{\alpha+1} [1/q(\theta)]^{2\alpha} \leq u(\theta) \leq \frac{1}{(\alpha+1)(2\alpha+1)} [1/q(\theta)]^{2\alpha}$$

Expressed in terms of  $v(\theta)$ , it becomes

$$\alpha[1+\alpha]^{-2} \leq v(\theta) \leq -\alpha[(1+2\alpha)(1+\alpha)]^{-2}$$

showing that  $v(\theta)$  is bounded. The boundedness of  $v(\theta)$  for  $\alpha \geq 0$  follows likewise.

(c)  $[q(\theta)/q'(\theta)]v'(\theta)$  is not bounded away from zero as  $\theta \rightarrow 0$ . For suppose there exists  $\varepsilon > 0$  and  $\theta_0 > 0$  such that  $[q(\theta)/q'(\theta)]v'(\theta) < -\varepsilon$  for  $\theta < \theta_0$ . That is,  $-v'(x) < \varepsilon[q'(x)/q(x)]$  for all  $x < \theta_0$ . Integrating this from  $\theta$  to  $\theta_0$  we get  $v(\theta) - v(\theta_0) < \varepsilon \ln[q(\theta_0)/q(\theta)]$  which shows that  $v(\theta) \rightarrow -\infty$  as  $\theta \rightarrow 0$ . This violates (b). Thus there exists a sequence  $\theta_i \rightarrow 0$  along which

$$[q(\theta_i)/q'(\theta_i)]v'(\theta_i) \rightarrow 0.$$

Similarly we can show

(d)  $[q(\theta)/q'(\theta)]v'(\theta)$  is not bounded away from zero as  $\theta \rightarrow \infty$ .

Now from (c) and (d) there are sequences  $\theta_i \rightarrow 0$  and  $\theta_i \rightarrow \infty$  along which  $[q(\theta)/q'(\theta)]v'(\theta) \rightarrow 0$ . From (14) it follows that  $v(\theta) \rightarrow 0$  along these sequences. Hence from (a) it follows that  $v(\theta) \equiv 0$ . This completes the proof of admissibility of  $T_\alpha$  for  $\alpha > -\frac{1}{2}$ . That  $T_\alpha$  is inadmissible for  $\alpha \leq -\frac{1}{2}$  follows from the fact that its risk (as shown in (2)) is finite for each such  $\alpha$ .

REMARKS 1. If  $X_1, \dots, X_n$  are independent random variables each having density (1) then the sufficient statistic  $T = \max X_i$  has density given by

$$[q(\theta)]^n n \left[ \int_0^t r(x) dx \right]^{n-1} r(t) \quad \text{for } 0 \leq t \leq \theta$$

which is a density of the form (1) with  $q(\theta)$  replaced by  $[q(\theta)]^n$  and  $r(x)$  replaced by  $n \left[ \int_0^x f(v) dv \right]^{n-1} r(x)$ . So from Theorem 1 we have the conclusion that

$$\frac{2\beta+1}{\beta+1} \left\{ n \left[ \int_0^x r(v) dv \right]^{n-1} r(X) \right\}^{-\beta}$$

is an admissible estimator of  $[q(\theta)]^{-n\beta}$  if and only if  $\beta > -\frac{1}{2}$ . That is, writing  $\alpha$  for  $n\beta$ , we conclude that

$$\frac{2\alpha+n}{\alpha+n} \left\{ n \left[ \int_0^x r(v) dv \right]^{n-1} r(x) \right\}^{-\alpha/n}$$

is an admissible estimator of  $[q(\theta)]^{-\alpha}$  if and only if  $\alpha > -n/2$ . So for a given  $\alpha$  we have admissibility for all sufficiently large sample sizes  $n$ .

2. Proof of Theorem 1 parallels the Blyth-Roberts [2] proof of the special case of the density (1) as

$$(15) \quad f(x; \theta) = \begin{cases} n\theta^{-n}x^{n-1}, & 0 \leq x \leq \theta \\ 0 & \text{otherwise} \end{cases}$$

In [2], the parametric function of interest is  $g(\theta) = \theta$ . If  $g(\theta) = \theta^s$  then according to Theorem 1 the estimator  $(n+2s)/(n+s)X^s$  is admissible (with respect to quadratic loss) for estimating  $\theta^s$  for every  $s > -n/2$  and is inadmissible for  $s \leq -n/2$ .

3. In [5] Karlin proved Theorem 1 (of this paper) for all  $\alpha > 0$  (see his Theorem 2, p. 418). His proof makes use of the fact that  $\alpha > 0$ . Theorem 1 of the present paper settles the question of the admissibility of  $T_\alpha$  for all values of  $\alpha$ .

4. An attempt was made in [6] to extend Karlin's Theorem 2 to all values of  $\alpha$  but this was successful only for some special extreme value densities such as (15). The approach there is the limiting Bayes method, used by Blyth [1] and Karlin [5].

5. The following theorem extends Theorem 3 of Karlin [5] to all other values of  $\alpha$ .

THEOREM 2. *Let  $X$  have density*

$$(16) \quad f(x; \theta) = \begin{cases} q(\theta)r(x), & x \geq \theta \\ 0, & \theta_0 < x < \theta, \end{cases}$$

where  $q^{-1}(\theta) = \int_{\theta_0}^{\theta} r(x) dx$  and  $q(\theta_0) = 0$ . Then (with quadratic loss) the estimator  $T_\alpha = [(2\alpha+1)/(\alpha+1)][1/q(x)]^\alpha$  is admissible for estimating  $[1/q(\theta)]^\alpha$  for all  $\alpha > -\frac{1}{2}$  and inadmissible for all  $\alpha \leq -\frac{1}{2}$ .

6. If the loss function is given by  $L_0(\delta, g) = [(\delta-g)/g]^2$ , the estimator  $T_\alpha$  is minimax and admissible for estimating  $[1/q(\theta)]^\alpha$  for all  $\alpha > -\frac{1}{2}$ .

7. The estimator  $(\alpha+1)[1/q(X)]^\alpha$  is the uniformly minimum variance unbiased estimator of  $[1/q(\theta)]^\alpha$  for all  $\alpha > -\frac{1}{2}$ . This estimator, however, is inadmissible for it is uniformly improved upon by the estimator  $T_\alpha$ .

8. In addition to the example of the density (15), Theorems 1 and 2 have the following applications:

(i) *Pareto distribution.* Let  $X$  have density of the form

$$(17) \quad f(x; \theta) = \begin{cases} c\theta^c \frac{1}{x^{c+1}}, & x \geq \theta \\ 0 & \text{otherwise,} \end{cases}$$

where  $c \geq 0$  is known and  $g(\theta) = \theta^s$ . If we take  $r(x) = c/x^{c+1}$  then (17) is a special case of (16).

(ii) Let  $X$  have density

$$(18) \quad f(x; \theta) = \begin{cases} e^{-(x-\theta)}, & x \geq \theta \\ 0 & \text{otherwise,} \end{cases}$$

where  $\theta \in (-\infty, \infty)$  and  $g(\theta) = \theta^3$ . If we set  $r(x) = e^{-x}$  the (18) is a special case of (16).

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