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## CONSTRUCTING MEDIAN-UNBIASED ESTIMATORS IN ONE-PARAMETER FAMILIES OF DISTRIBUTIONS VIA STOCHASTIC ORDERING

Abstract. If  $\theta \in \Theta$  is an unknown real parameter of a given distribution, we are interested in constructing an exactly median-unbiased estimator  $\hat{\theta}$ of  $\theta$ , i.e. an estimator  $\hat{\theta}$  such that a median  $\operatorname{Med}(\hat{\theta})$  of the estimator equals  $\theta$ , uniformly over  $\theta \in \Theta$ . We shall consider the problem in the case of a fixed sample size n (nonasymptotic approach).

**1. The model.** Let  $\mathcal{F}$  be a one-parameter family  $\{F_{\theta} : \theta \in \Theta\}$  of distributions, where  $\Theta$  is a (finite or infinite) interval on the real line. The family  $\mathcal{F}$  is assumed to be a family of distributions with continuous and strictly increasing distribution functions and stochastically ordered by  $\theta$  so that for every  $x \in \operatorname{supp} \mathcal{F} = \bigcup_{\theta \in \Theta} \operatorname{supp} F_{\theta}$  and for every  $q \in (0, 1)$ , the equation  $F_{\tau}(x) = q$  has exactly one solution in  $\tau \in \Theta$ . Given a sample  $X_1, \ldots, X_n$  from an  $F_{\theta}$ , we are interested in median-unbiased estimation of  $\theta$ ; here n is a fixed integer (nonasymptotic approach).

This model covers a wide range of one-parameter families of distributions.

EXAMPLE 1. The family of uniform distributions on  $(\theta, \theta + 1)$ , with  $-\infty < \theta < \infty$ .

EXAMPLE 2. The family of power distributions on (0, 1) with distribution functions  $F_{\theta}(x) = x^{\theta}, \theta > 0.$ 

EXAMPLE 3. The family of gamma distributions with probability distribution functions (pdf) of the form

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$$f_{\alpha}(x) = \frac{1}{\Gamma(\alpha)} x^{\alpha - 1} e^{-x}, \quad x > 0,$$

with  $\alpha > 0$ .

EXAMPLE 4. Consider the family of Cauchy distributions with pdf of the form

$$g_{\lambda}(y) = \frac{1}{\lambda} \frac{1}{1 + (y/\lambda)^2}, \quad -\infty < y < \infty,$$

and distribution function of the form

$$G_{\lambda}(y) = \frac{1}{2} + \frac{1}{\pi} \arctan \frac{x}{\lambda}$$

with  $\lambda > 0$ . The family  $\{F_{\lambda} : \lambda > 0\}$  of distribution functions of X = |Y| with

$$F_{\lambda}(x) = \frac{2}{\pi} \arctan \frac{x}{\lambda}$$

satisfies the model assumptions so that the problem of estimating  $\lambda$  from a sample  $Y_1, \ldots, Y_n$  amounts to estimating  $\lambda$  from the sample  $X_1, \ldots, X_n$ with  $X_i = |Y_i|, i = 1, \ldots, n$ .

EXAMPLE 5. Consider the one-parameter family of Weibull distributions with distribution functions of the form

$$G_{\alpha}(y) = 1 - e^{-y^{\alpha}}, \quad y > 0, \ \alpha > 0,$$

and let  $X = \max\{Y, 1/Y\}$ . The family  $\{F_{\alpha} : \alpha > 0\}$  of distributions of X with distribution functions

$$F_{\alpha}(x) = e^{-x^{-\alpha}} - e^{-x^{\alpha}}, \quad x > 1, \ \alpha > 0,$$

satisfies the model assumptions.

EXAMPLE 6 (Estimating the characteristic exponent of a symmetric  $\alpha$ -stable distribution). Consider the one-parameter family of  $\alpha$ -stable distributions with characteristic functions  $\exp\{-t^{\alpha}\}$ ,  $0 < \alpha \leq 2$ . The problem is to construct a median-unbiased estimator of  $\alpha$ . Some related results can be found in Fama and Roll (1971) and Zieliński (2000). We shall not consider this problem in this note because it needs (and deserves) a special treatment and will be discussed in detail elsewhere.

Generally, every family of distributions  $F_{\theta}$  with continuous and strictly increasing  $F_{\theta}$  and a location parameter  $\theta$  satisfies the model assumptions. Similarly, every family of continuous and strictly increasing distributions on  $(0, \infty)$  with a scale parameter fits into the model.

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## 2. The method. The method consists in:

1) For a given  $q \in (0, 1)$ , estimating the *q*th quantile of the underlying distribution in a nonparametric setup; denote the estimator by  $\hat{x}_q$ . A restriction is that for a fixed *n* a median-unbiased estimator of the *q*th quantile exists iff  $\max\{q^n, (1-q)^n\} \leq \frac{1}{2}$ .

2) Solving the equation  $F_{\tau}(\hat{x}_q) = q$  with respect to  $\tau$ . The solution, to be denoted by  $\hat{\theta}_q$ , is considered as an estimator of  $\theta$ . The solution of the equation  $F_{\tau}(x) = q$  with respect to  $\tau$  will be denoted by  $\hat{\theta}_q(x)$  so that  $\hat{\theta}_q = \hat{\theta}_q(\hat{x}_q)$ .

In the model, if  $\hat{x}_q$  is a median-unbiased estimator of  $x_q$  then, due to monotonicity of  $\hat{\theta}_q(x)$  with respect to x,  $\hat{\theta}_q$  is a median-unbiased estimator of  $\theta$ . What is more, if  $\hat{x}_q$  is the median-unbiased estimator of  $x_q$  the most concentrated around  $x_q$  in the class of all median-unbiased estimators which are equivariant with respect to monotone transformations of data (briefly: the best estimator) then, due to monotonicity again,  $\hat{\theta}_q$  is the median-unbiased estimator of  $\theta$  most concentrated around  $\theta$  in the class of all median-unbiased estimators which are equivariant with respect to monotone transformations of data (briefly: the best estimator).

Given  $q \in (0, 1)$ , the best estimator  $\hat{x}_q$  of  $x_q$  is given by the formula

[E] 
$$\widehat{x}_q = X_{k:n} \mathbf{1}_{(0,\lambda]}(U) + X_{k+1:n} \mathbf{1}_{(\lambda,1)}(U)$$

where  $X_{k:n}$  is the kth order statistic,  $X_{1:n} \leq \ldots \leq X_{n:n}$ , from the sample  $X_1, \ldots, X_n$  and

$$k = k(q)$$

= the unique integer such that  $Q(k; n, q) \ge 1/2 \ge Q(k+1; n, q)$ ,

$$\lambda = \lambda(q) = \frac{1/2 - Q(k+1; n, q)}{Q(k; n, q) - Q(k+1; n, q)},$$
$$Q(k; n, q) = \sum_{j=k}^{n} {n \choose j} q^{j} (1-q)^{n-j};$$

here U is a random variable uniformly distributed on (0, 1) and independent of the sample  $X_1, \ldots, X_n$  (Zieliński 1988).

When estimating  $\theta$  in a parametric model  $\{F_{\theta} : \theta \in \Theta\}$ , the problem is to choose an "optimal" q. To define a criterion of optimality (or "an ordering in the class  $\hat{\theta}_q$ , 0 < q < 1, of estimators"), recall (e.g. Lehmann 1983, Sec. 3.1) that a median-unbiased estimator  $\hat{\theta}$  of a parameter  $\theta$  is the one for which

[K] 
$$E_{\theta}|\widehat{\theta} - \theta| \le E_{\theta}|\widehat{\theta} - \theta'|$$
 for all  $\theta, \theta' \in \Theta$ 

(the estimator is closer to the "true" value  $\theta \in \Theta$  than to any other value

 $\theta' \in \Theta$  of the parameter). According to this property, we shall choose  $q_{\text{opt}}$  as the one with minimal risk under the loss function  $|\hat{\theta} - \theta|$ , i.e. such that

$$E_{\theta}|\widehat{\theta}_{q_{\text{opt}}} - \theta| \le E_{\theta}|\widehat{\theta}_{q} - \theta|, \quad 0 < q < 1,$$

for all  $\theta \in \Theta$ , if possible.

Using the fact that  $\theta \in \Theta$  generates the stochastic ordering of the family  $\{F_{\theta} : \theta \in \Theta\}$ , we shall restrict our attention to finding  $q_{\text{opt}}$  which satisfies criterion [K] for a fixed  $\theta$ , for example  $\theta = 1$ ; then the problem reduces to minimizing

$$R(q) = E|\widehat{\theta}_q - 1|$$

with respect to  $q \in (0, 1)$ , where  $E = E_1$ .

By [E] we obtain

$$R(q) = \lambda(q)E|\widehat{\theta}_q(X_{k(q):n}) - 1| + (1 - \lambda(q))E|\widehat{\theta}_q(X_{k(q)+1:n}) - 1|$$

It is obvious that  $q_{\text{opt}}$  satisfies

$$\lambda(q_{\rm opt}) = 1$$

and

$$E|\widehat{\theta}_{q_{\text{opt}}}(X_{k(q_{\text{opt}}):n}) - 1| \le E|\widehat{\theta}_{q}(X_{k(q):n}) - 1|, \quad 0 < q < 1.$$

By the very definition of  $\lambda$ ,  $\lambda(q) = 1$  iff  $q \in \{q_1, \ldots, q_n\}$  where  $q_i$  satisfies  $Q(i; n, q_i) = 1/2$ , and the problem reduces to finding the smallest element of the finite set

$$\{E|\widehat{\theta}_{q_i}(X_{i:n}) - 1| : i = 1, \dots, n\}.$$

If  $X_{k:n}$  is the kth order statistic from the sample  $X_1, \ldots, X_n$  from a distribution function F, then  $U_{k:n} = F(X_{k:n})$  is the kth order statistic from the sample  $U_1, \ldots, U_n$  from the uniform distribution on (0, 1), which gives us

$$\begin{split} E|\widehat{\theta}_{q_i}(X_{i:n}) - 1| &= E|\widehat{\theta}_{q_i}(F^{-1}(U_{i:n})) - 1| \\ &= \frac{\Gamma(n)}{\Gamma(i)\Gamma(n-i+1)} \int_0^1 |\widehat{\theta}_{q_i}(F^{-1}(t)) - 1| t^{i-1}(1-t)^{n-i} dt. \end{split}$$

The latter can be easily calculated numerically.

## 3. Applications

EXAMPLE 1A. In the case of uniform distributions on  $(\theta, \theta + 1)$ , the solution  $\tau$  of the equation  $F_{\tau}(\hat{x}_q) = q$  takes the form  $\tau = \hat{x}_q - q$ . For example for n = 10 the best estimator is  $X_{1:10} - 0.067$  or  $X_{10:10} - 0.933$ .

EXAMPLE 2A. In the case of power distributions, the best estimator is the (unique) solution, with respect to  $\tau$ , of the equation  $\hat{x}_q^{\theta} = q_{\text{opt}}$ ; for n = 10 the estimator takes the form  $-1.81854/\text{Log}[X_{2:10}]$ .

EXAMPLE 3A. In the case of gamma distributions, the best estimator is the (unique) solution, with respect to  $\tau$ , of the equation  $F_{\tau}(\hat{x}_q) = q_{\text{opt}}$ ; for n = 10 this equation takes the form  $F_{\tau}(X_{3:10}) = 0.2586$ .

EXAMPLE 4A. In the case of Cauchy distributions the solution  $\tau$  of the equation  $F_{\tau}(\hat{x}_q) = q$  can be written in the form

$$\tau = \frac{\widehat{x}_q}{\tan\left(\frac{\pi}{2}q\right)}$$

and for n = 10 the best estimator is  $1.16456 \cdot X_{5:10}$ .

EXAMPLE 5A. In the case of Weibull distributions, the best estimator is the (unique) solution, with respect to  $\tau$ , of the equation  $F_{\tau}(\hat{x}_q) = q_{\text{opt}}$ ; for n = 10 this equation takes the form  $F_{\tau}(X_{8:10}) = 0.7414$ , which gives us the optimal estimator  $0.302/\text{Log}(X_{8:10})$ .

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