

# Generalized Jackknife Moment estimators of the tail index

M. Ivette Gomes\*

*D.E.I.O. and C.E.A.U.L., Faculty of Science of Lisbon*

*Bloco C2, Campo Grande*

*1749-016 Lisboa, Portugal*

*E-mail address: ivette.gomes@fc.ul.pt*

## 1. The Generalized Jackknife Moment estimators

In *Statistical Extreme Value Theory* we are mainly interested in the estimation of *parameters of rare events*, the basic parameter being the *tail index*  $\gamma = \gamma(F)$ , directly related to the tail weight of the model  $F(\cdot)$ . The *tail index*  $\gamma$  is the shape parameter in the *Generalized Extreme Value (GEV)* distribution function (d.f.), which appears as the non-degenerate limiting d.f. of the sequence of maximum values, linearly normalized.

Whenever there is a non-degenerate limit of the sequence of normalized maximum values we say that  $F$  is in the *domain of attraction* of  $G_\gamma$ , and write  $F \in D(G_\gamma)$ . Denoting by  $U(t) := F^\leftarrow(1-1/t)$ , the quantile function, and for heavy tails ( $\gamma > 0$ ),  $F \in D(G_\gamma)$  iff  $1-F \in RV_{-1/\gamma}$  iff  $U \in RV_\gamma$ , where  $RV_\alpha$  stands for the class of *regularly varying* functions at infinity with index of regular variation equal to  $\alpha$ . These conditions characterize completely the first order behaviour of  $F(\cdot)$  [Gnedenko (1943), de Haan (1970)], and the second order theory has been worked out in full generality by de Haan and Stadtmüller (1996) — for a large class of models there exists a function  $A(t)$  of constant sign for large  $t$ , such that  $\left[ \frac{U(tx)/U(t) - x^\gamma}{A(t)} \right] \xrightarrow{t \rightarrow \infty} x^\gamma(x^\rho - 1)/\rho$ , for every  $x > 0$ , where  $\rho(\leq 0)$  is a second order parameter.

In this paper we are going to work with the *Moment estimator* (Dekkers *et al*, 1989), a serious candidate to the *tail index* estimation,

$$(1) \quad \gamma_{n,M}(k) := M_n^{(1)}(k) + 1 - \frac{1}{2} \left\{ 1 - \frac{(M_n^{(1)}(k))^2}{M_n^{(2)}(k)} \right\}^{-1}, \quad M_n^{(j)}(k) := \frac{1}{k} \sum_{i=1}^k \{ \ln X_{n-i+1:n} - \ln X_{n-k:n} \}^j,$$

for which

$$(2) \quad \gamma_{n,M}(k) \stackrel{d}{=} \gamma + \sqrt{\frac{\gamma^2 + 1}{k}} Z_n + \frac{2 - \rho + (\gamma - 2)(1 - \rho)}{\gamma(1 - \rho)^2} A(n/k) + o_p(A(n/k)),$$

$Z_n$  asymptotically standard Normal (Dekkers and de Haan, 1993).

If we consider two intermediate levels  $k_1 = k$ ,  $k_2 = k/2$ ,  $A(n/(k/2)) = 2^\rho A(n/k)(1 + o(1))$ , and  $\frac{BIAS_\infty[\gamma_{n,M}(k)]}{BIAS_\infty[\gamma_{n,M}(k/2)]} = \frac{A(n/k)}{A(2n/k)} = 2^{-\rho}(1 + o(1))$ , the Generalized Jackknife random variable (Gray and Shucany, 1972) associated to the moment estimators at the two levels  $k/2$  and  $k$  is

$$(3) \quad \gamma_{n,M}^G(k) = \frac{\gamma_{n,M}(k) - 2^{-\rho} \gamma_{n,M}(k/2)}{1 - 2^{-\rho}}.$$

If we got to know  $\rho$  we would be able to remove completely the asymptotic bias, with the use of (3). In order to avoid the estimation of  $\rho$ , and since we are here going to work, with Fréchet and Cauchy parents, for which we have the second order parameter equal to  $-1$  and  $-2$ , respectively, we shall consider the following two alternative estimators to the Moment estimator,

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$$(4) \quad \gamma_{n,M}^{G_F}(k) := 2\gamma_{n,M}(k/2) - \gamma_{n,M}(k), \quad \text{and} \quad \gamma_{n,M}^{G_C}(k) := \frac{4}{3}\gamma_{n,M}(k/2) - \frac{1}{3}\gamma_{n,M}(k).$$

For a more general application of the *Generalized Jackknife* Theory to the estimation of parameters of *rare events*, see Gomes et al (1998).

## 2. Distributional properties of estimators for Fréchet and Cauchy parents

In Table 1, we show the *Mean Values* ( $E_1, E_2, E_3$ ) and the *Mean Square Errors* ( $MSE_1, MSE_2, MSE_3$ ) of  $\gamma_n^{(1)}(k) \equiv \gamma_{n,M}(k)$ ,  $\gamma_n^{(2)}(k) \equiv \gamma_{n,M}^{G_F}(k)$  and  $\gamma_n^{(3)}(k) \equiv \gamma_{n,M}^{G_C}(k)$ , respectively, at the optimal levels (levels where minimum *MSE* is attained) for a Fréchet model, with  $\gamma = 1$ , and for a Cauchy model, which is also attracted towards a  $GEV(\gamma = 1)$ . The simulation was based on 20 replicas with 5000 runs each, and the smallest *BIAS* and the smallest *MSE*, were put in *italic* and underlined.

<i>F</i>	<i>n</i>	<i>E</i> <sub>1</sub>	<i>E</i> <sub>2</sub>	<i>E</i> <sub>3</sub>	<i>MSE</i> <sub>1</sub>	<i>MSE</i> <sub>2</sub>	<i>MSE</i> <sub>3</sub>
<i>R</i>	200	1.0574	0.9840	<u>1.0316</u>	0.0262	0.0296	<u>0.0215</u>
<i>É</i>	500	1.0483	<u>0.9948</u>	1.0251	0.0134	0.0103	<u>0.00932</u>
<i>C</i>	1000	1.0410	<u>0.9921</u>	1.0228	0.00815	<u>0.00487</u>	0.00511
<i>H</i>	5000	1.0264	<u>0.9982</u>	1.0221	0.00268	<u>0.000942</u>	0.00151
<i>E</i>	10000	1.0215	<u>0.9992</u>	1.0223	0.00167	<u>0.000472</u>	0.00102
<i>T</i>	20000	1.0173	<u>0.9995</u>	1.02271	0.00105	<u>0.000236</u>	0.000780
<i>C</i>	200	<u>1.0147</u>	0.8116	0.8992	<u>0.0500</u>	0.1404	0.0849
<i>A</i>	500	<u>1.0292</u>	0.8680	0.9518	<u>0.0208</u>	0.0516	0.0256
<i>U</i>	1000	<u>1.0282</u>	0.9118	0.9663	<u>0.0113</u>	0.0275	0.0117
<i>C</i>	5000	1.0192	0.9607	<u>0.9862</u>	0.00293	0.00677	<u>0.00209</u>
<i>H</i>	10000	1.0155	0.9715	<u>0.9932</u>	0.00166	0.00378	<u>0.00101</u>
<i>Y</i>	20000	1.0122	0.9790	<u>0.9966</u>	0.000946	0.00212	<u>0.000500</u>

Notice that for Fréchet parents  $\gamma_{n,M}^{G_F}(k)$  is the optimal estimator for sample sizes  $n \geq 1000$ , with a drastic reduction of *MSE*, relatively to the Moment estimator. For Cauchy parents  $\gamma_{n,M}^{G_C}(k)$  takes the role of  $\gamma_{n,M}^{G_F}(k)$ , but the improvement is not spectacular.

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## FRENCH RESUMÉ

*Nous considérons des statistiques généralisées de Jackknife associés à l'estimateurs des Moments de l'index de variation régulière, évaluées en deux niveaux, et nous étudions le gain en efficience, dans le cas de distributions parente Fréchet et Cauchy.*