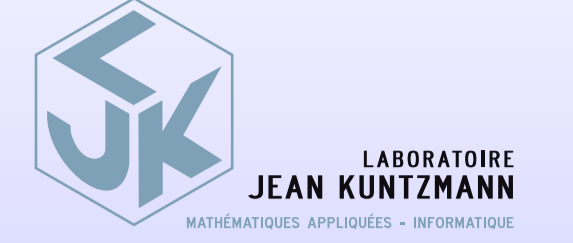


ESTIMATION OF A NEW PARAMETER DISCRIMINATING BETWEEN WEIBULL TAIL-DISTRIBUTIONS AND HEAVY-TAILED DISTRIBUTIONS



Jonathan El Methni ⁽¹⁾, Laurent Gardes ⁽¹⁾, Stéphane Girard ⁽¹⁾ & Armelle Guillou ⁽²⁾



⁽¹⁾ Team Mistis, INRIA Rhône-Alpes & LJK, <http://mistis.inrialpes.fr> – ⁽²⁾ Université de Strasbourg & CNRS.
Contact: jonathan.elmethni@inrialpes.fr <http://mistis.inrialpes.fr/people/elmethni>

I. Statistical Framework

- Let X_1, \dots, X_n be a sample of independent and identically distributed random variables driven from X with cumulative distribution function F , and let $X_{1,n} \leq \dots \leq X_{n,n}$ denote the order statistics associated to this sample.
- We want to estimate the extreme quantile x_{p_n} of order p_n associated to the random variable $X \in \mathbb{R}$ defined by:

$$x_{p_n} = \bar{F}^{\leftarrow}(p_n) = \inf\{x, \bar{F}(x) \leq p_n\},$$

with $p_n \rightarrow 0$ when $n \rightarrow \infty$. The function \bar{F}^{\leftarrow} is the generalized inverse of the non-increasing function $\bar{F} = 1 - F$.

- In [2], a family of distributions is introduced, it encompasses the whole Fréchet maximum domain of attraction as well as Weibull tail-distributions. These distributions depend on two parameters $\tau \in [0, 1]$ and $\theta > 0$.

II. Model: Gumbel/Fréchet

Let us consider the family of survival distribution functions defined as

$\bar{F}(x) = \exp(-K_\tau^{\leftarrow}(\log H(x)))$ for $x \geq x_* > 0$ where

- $K_\tau(y) = \int_1^y u^{\tau-1} du$ where $\tau \in [0, 1]$,
- H an increasing function such that $H^{\leftarrow} = x^{-\theta} \ell(x)$ where $\theta > 0$ and $\ell(x)$ is a slowly varying function.

Proposition

- $\tau = 0 \iff F$ is a Weibull-tail distribution function with Weibull tail-coefficient θ .
- $\tau \in [0, 1)$ and H is twice differentiable $\implies F$ belongs to the Gumbel maximum domain of attraction.
- $\tau = 1 \iff F$ is in the Fréchet maximum domain of attraction with tail-index θ .

III. Estimators depending on τ

Denoting by (k_n) an intermediate sequence of integers, the following estimator of θ is considered in [2]:

$$\hat{\theta}_{n,\tau}(k_n) = \frac{H_n(k_n)}{\mu_\tau(\log(n/k_n))}$$

where $H_n(k_n)$ is the Hill estimator [3],

$$H_n(k_n) = \frac{1}{k_n - 1} \sum_{i=1}^{k_n-1} \log(X_{n-i+1,n}) - \log(X_{n-k_n+1,n}),$$

and with, for all $t > 0$, $\mu_\tau(t) = \int_0^\infty (K_\tau(x+t) - K_\tau(t)) e^{-x} dx$.
An estimator of the extreme quantile x_{p_n} is proposed in [2]:

$$\hat{x}_{p_n, \hat{\theta}_{n,\tau}(k_n)} = X_{n-k_n+1,n} \exp\left(\hat{\theta}_{n,\tau}(k_n) (K_\tau(\log(1/p_n)) - K_\tau(\log(n/k_n)))\right).$$

IV. Main Goal

The asymptotic distributions of $\hat{\theta}_{n,\tau}(k_n)$ and $\hat{x}_{p_n, \hat{\theta}_{n,\tau}(k_n)}$ have been established in [2] under a second-order condition on ℓ :

There exist $\rho < 0$, a function b satisfying $b(x) \rightarrow 0$ and $|b|$ asymptotically decreasing such that uniformly locally on $\lambda > 0$

$$\log\left(\frac{\ell(\lambda x)}{\ell(x)}\right) \sim b(x) K_\rho(\lambda), \text{ when } x \rightarrow \infty.$$

- The main goal of this work is to propose an estimator for τ independent from θ .

This parameter controls the behavior of the tail-distribution: the larger the value of τ , the heavier is the tail.

V. Estimator of τ

Let us consider for $t > t'$

$$\psi(x; t, t') : \mathbb{R} \rightarrow (-\infty, \exp(t - t')) \text{ such that } \psi(x; t, t') = \frac{\mu_x(t)}{\mu_x(t')}$$

Denoting by (k_n) and (k'_n) two intermediate sequences of integers such that $k'_n > k_n$, the following estimator of τ is considered:

$$\hat{\tau}_n = \begin{cases} \psi^{-1}\left(\frac{H_n(k_n)}{H_n(k'_n)}, \log(n/k_n), \log(n/k'_n)\right) & \text{if } \frac{H_n(k_n)}{H_n(k'_n)} < \frac{k'_n}{k_n} \\ u & \text{if } \frac{H_n(k_n)}{H_n(k'_n)} \geq \frac{k'_n}{k_n} \end{cases}$$

where $u \in [0, 1]$.

- $\hat{\tau}_n$ exists because $\psi(\cdot; \log(n/k_n), \log(n/k'_n)) : \mathbb{R} \rightarrow (-\infty, k'_n/k_n)$ is a bijection.

VI. Asymptotic distribution

Replacing τ by $\hat{\tau}_n$ we obtain:

- an estimator of θ : $\hat{\theta}_{n, \hat{\tau}_n}(k_n) = \frac{H_n(k_n)}{\mu_{\hat{\tau}_n}(\log(n/k_n))}$
- an estimator of x_{p_n} :

$$\hat{x}_{p_n, \hat{\theta}_{n, \hat{\tau}_n}(k_n)} = X_{n-k_n+1,n} \exp\left(\hat{\theta}_{n, \hat{\tau}_n}(k_n) (K_{\hat{\tau}_n}(\log(1/p_n)) - K_{\hat{\tau}_n}(\log(n/k_n)))\right).$$

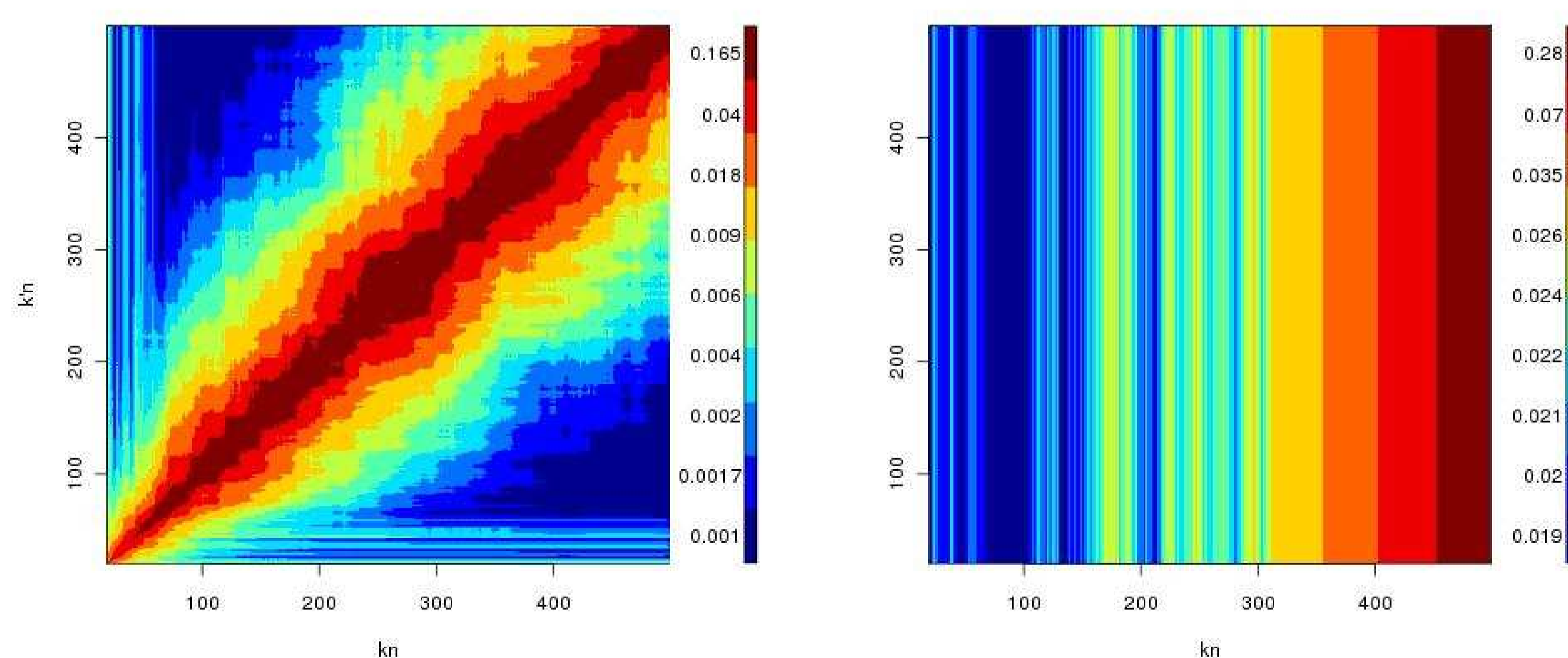
Under some assumptions on (k_n) and (k'_n) we establish the asymptotic normality of $\hat{\tau}_n$, $\hat{\theta}_{n, \hat{\tau}_n}(k_n)$ and $\hat{x}_{p_n, \hat{\theta}_{n, \hat{\tau}_n}(k_n)}$. In particular:

$$\frac{\sqrt{k_n} (\log_2(n/k_n) - \log_2(n/k'_n))}{\int_{\log(n/k_n)}^{\log(1/p_n)} \log(u) u^{\tau-1} du} \left(\frac{\hat{x}_{p_n, \hat{\theta}_{n, \hat{\tau}_n}(k_n)}}{x_{p_n}} - 1 \right) \xrightarrow{d} \mathcal{N}(0, \theta^2).$$

VII. Numerical experiments on simulated data

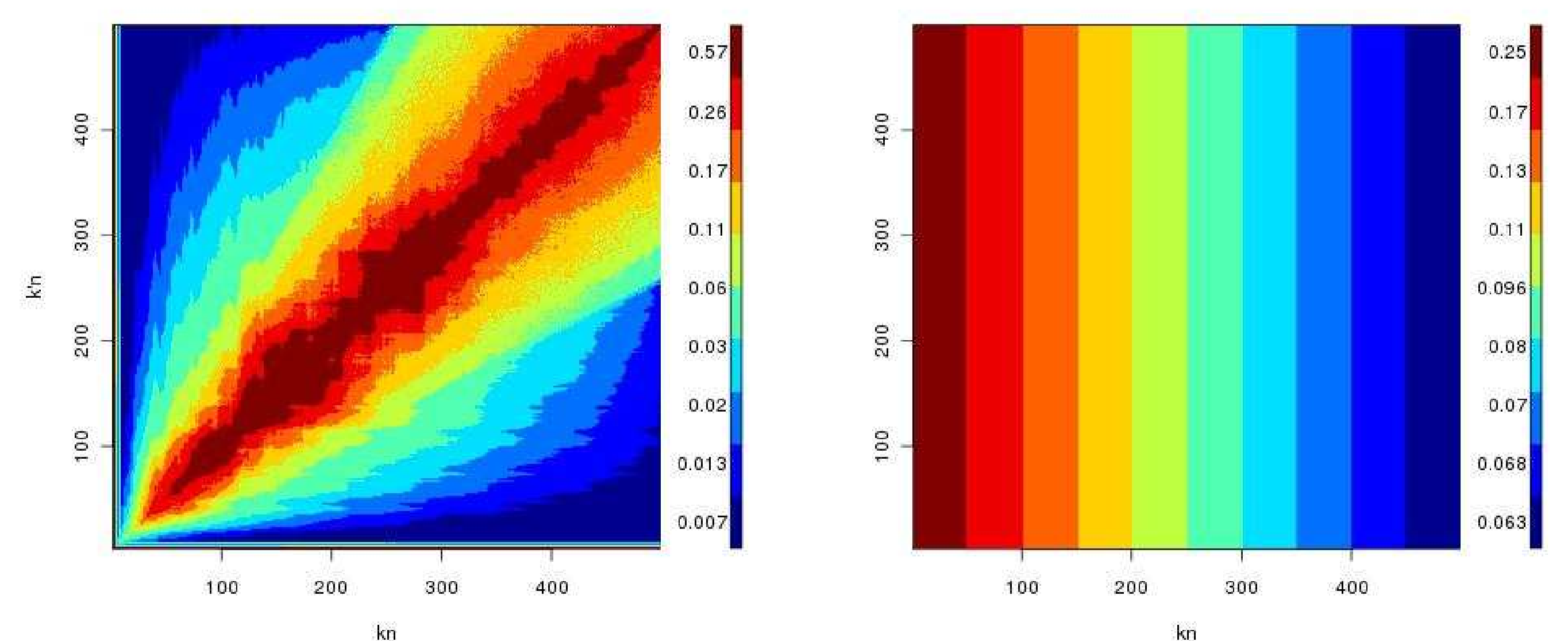
Gamma distribution $\mathcal{D}(\text{Gumbel})$

Pareto distribution $\mathcal{D}(\text{Fréchet})$



MSE $\hat{x}_{p_n, \hat{\theta}_{n, \hat{\tau}_n}(k_n)}$

MSE Dekkers et al.



MSE $\hat{x}_{p_n, \hat{\theta}_{n, \hat{\tau}_n}(k_n)}$

MSE Dekkers et al.

Bibliography

- The estimator $\hat{x}_{p_n, \hat{\theta}_{n, \hat{\tau}_n}(k_n)}$ is computed on $N = 100$ samples of size $n = 500$ for $k_n = 2, \dots, 499$ and $k'_n = k_n, \dots, 500$ where $p_n = 10^{-3}$.
- The associated deciles of the empirical Mean-Squared Error MSE are plotted
- Comparison with an estimator of A. L. M. Dekkers, J.H.J. Einmahl & L. de Haan [1].

- [1] Dekkers, A. L. M., Einmahl, J. H. J., De Haan, L., (1989). A Moment Estimator for the Index of an Extreme-Value Distribution, *The Annals of Statistics*, 17, 1833–1855.
- [2] Gardes, L., Girard, S., Guillou, A., (2011). Weibull tail-distributions revisited: a new look at some tail estimators, *Journal of Statistical Planning and Inference*, 141, 429–444.
- [3] Hill, B.M., (1975). A simple general approach to inference about the tail of a distribution, *The Annals of Statistics*, 3, 1163–1174.