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MOMENTS OF ORDER STATISTICS

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Summary. Let $X_{1:n} \leq X_{2:n} \leq \dots \leq X_{n:n} = M_n$ denote the order statistics of a sample of size n . In this paper we investigate the asymptotic behaviour of $E(M_n)$ and $E(X_{n-k:n})$ as $n \rightarrow \infty$. We show that $\{E(M_n)\}_N$ and all its differentials $\{\Delta^i E(M_n)\}_N$ are regularly varying sequences if the underlying d.f. has a regularly varying tail.

Keywords: Order statistics, asymptotic results, rate of convergence, regularly varying functions.

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INTRODUCTION

Let $X_{1:n} \leq \dots \leq X_{n:n} = M_n$ denote the order statistics of a random sample X_1, X_2, \dots, X_n of size n from a distribution with a distribution function (d.f.) F which is concentrated on R_+ . In this paper we shall be concerned with the asymptotic behaviour of $E(M_n)$ and $E(X_{n-k:n})$ as $n \rightarrow \infty$. In Section 1 we show that $E(M_n)/n \rightarrow 0$ as $n \rightarrow \infty$ if $E(X_1) < \infty$. If not only $E(X_1)$ is finite but if X_1 belongs to the max-domain of attraction of a stable law, we show that $E(M_n)$ and $E(X_{n-k:n})$ have a very nice regularly varying behaviour. In Section 2 we discuss regularly varying and 0-regularly varying behaviour of $E(X_{n-k:n})$, $k = 0, 1, \dots$. Among others we show that $1 - F \in RV_{-\alpha}$ implies that

$$\lim_{n \rightarrow \infty} \frac{E(M_n)}{a_n} = \Gamma(1 - 1/\alpha)$$

for a (known) sequence $\{a_n\}$. In Section 3 some rate of convergence results are established.

1. MOMENT CONDITIONS

It is well-known that for i.i.d. non-negative random variables X_1, X_2, \dots we have $E(M_n) < \infty$ if and only if $E(X_1) < \infty$. In the next proposition we obtain some more precise information concerning $E(M_n)$.

Proposition 1.1. Assume that X_1, \dots, X_n are i.i.d. non-negative random variables and that $E(X_1) < \infty$. Define $\mu_n := E(M_n)/n$. Then

- (i) $\lim_{n \rightarrow \infty} \mu_n = 0$;
- (ii) μ_n is non-increasing and $\lim_{n \rightarrow \infty} (\mu_n/\mu_{n-1}) = 1$;
- (iii) $\lim_{n \rightarrow \infty} [E(M_n) - E(M_{n-1})] = \lim_{n \rightarrow \infty} n(\mu_n - \mu_{n-1}) = 0$.

Proof. (i) We have $0 \leq P(M_n > x) = 1 - F^n(x) \leq n(1 - F(x))$. Since $E(X_1) < \infty$ and $\lim_{n \rightarrow \infty} (1/n) \cdot P(M_n > x) = 0$, by Lebesgue's theorem on dominated convergence we obtain that

$$\lim_{n \rightarrow \infty} \frac{E(M_n)}{n} = \lim_{n \rightarrow \infty} \int_0^\infty \frac{P(M_n > x)}{n} dx = 0.$$

(ii) and (iii) A general result in the theory of order statistics [3], p. 37 states that

$$(1.1) \quad (n - r) E(X_{r:n}) + rE(X_{r+1:n}) = nE(X_{r:n-1}).$$

Applying (1.1) for $r = n - 1$ we obtain that

$$E(X_{n-1:n}) = n E(M_{n-1}) - (n - 1) E(M_n).$$

It follows that

$$\mu_n \leq \mu_{n-1} = \mu_n + \frac{E(X_{n-1:n})}{n(n-1)} \leq \mu_n + \mu_n \frac{1}{n-1}.$$

Hence μ_n is non-increasing and $\mu_n \leq \mu_{n-1} \leq \mu_n(n/n - 1)$. Now the results (ii) and (iii) easily follow. ■

The following corollary follows immediately.

Corollary 1.2. Assume that X_1, \dots, X_n are i.i.d. non-negative random variables and that $E(X_1^\beta) < \infty$ for some $\beta \geq 1$. Then

$$\lim_{n \rightarrow \infty} \frac{E(M_n)}{n^{1/\beta}} = 0.$$

Proof. Let $Y_1 = X_1^\beta$; from Proposition 1.1 (i) we obtain that $\lim_{n \rightarrow \infty} (E(M_n^\beta)/n) = 0$.

Now apply Hölder's inequality. ■

As to real random variables, we have

Corollary 1.3. Assume X_1, \dots, X_n are i.i.d. random variables and assume g is non-decreasing. If $E|g(X_1)| < \infty$, then $\lim_{n \rightarrow \infty} E(g(M_n))/n = 0$.

Proof. Since $|\mathbb{E}(g(M_n))| \leq \mathbb{E}(\max_{1 \leq i \leq n} |g(X_i)|)$ the result follows from Proposition 1.1 (i) with $Y_1 = |g(X_1)|$. ■

2. O-REGULARLY VARYING AND REGULARLY VARYING BEHAVIOUR

Recall that a measurable function $f: \mathbf{R}_+ \rightarrow \mathbf{R}_+$ is O-regularly varying (we write $f \in \text{ORV}$) if for all $x > 0$,

$$\limsup_{t \rightarrow \infty} \frac{f(tx)}{f(t)} < \infty.$$

Further, a measurable function $f: \mathbf{R}_+ \rightarrow \mathbf{R}_+$ is regularly varying with an index $\alpha \in \mathbf{R}$ (we write $f \in \text{RV}_\alpha$) if for all $x > 0$,

$$\lim_{t \rightarrow \infty} \frac{f(tx)}{f(t)} = x^\alpha.$$

It is well-known [1] that $f \in \text{ORV}$ implies the existence of constants α, A , and t_0 such that

$$\frac{f(tx)}{f(t)} \leq Ax^\alpha \quad \text{for all } x \geq 1, t \geq t_0.$$

We call α an upper index of f . For details we refer to [1], [2], [4], and [7].

Finally, we say that a sequence $\{a_n\}_N$ (N is the set of all positive integers) of non-negative real numbers is regularly varying if the function f defined by $f(x) = a_{[x]}$ is a regularly varying function where as usual $[x]$ is the integer part of x .

In order to estimate $\mathbb{E}(M_n)$ we start with some auxiliary results.

Lemma 2.1. *Let F be a distribution function concentrated on \mathbf{R}_+ and let a_n be defined as $a_n := \inf \{x: 1 - F(x) \leq 1/n\}$.*

(i) *If $1 - F \in \text{ORV}$ is such that for some $\beta > 1, A > 0$ and $t_0 > 0$,*

$$(2.1) \quad \frac{1 - F(tx)}{1 - F(t)} \leq Ax^{-\beta} \quad \text{for all } x \geq 1 \quad \text{and } t \geq t_0$$

holds, then there exists a number $n_0 \in N$ such that

$$(2.2) \quad \mathbb{P}\{M_n > a_n x\} \leq Ax^{-\beta} \quad \text{for } x \geq 1, n \geq n_0.$$

(ii) *If $1 - F \in \text{RV}_{-\alpha}$ for some $\alpha > 1$, then $\{a_n\}_N \in \text{RV}_{1/\alpha}$ and for all $x > 0$,*
 $\lim_{n \rightarrow \infty} n(1 - F(a_n x)) = x^{-\alpha}$ and

$$(2.3) \quad \lim_{n \rightarrow \infty} \mathbb{P}\{M_n \leq a_n x\} = e^{-x^{-\alpha}}$$

hold. Moreover, (2.2) holds with $\beta = \alpha - \varepsilon > 1$.

Proof. (i) Obviously

$$P\{M_n > a_n x\} = 1 - F^n(a_n x) \leq n(1 - F(a_n x)) \leq \frac{1 - F(a_n x)}{1 - F(a_n)}.$$

Since $a_n \rightarrow \infty (n \rightarrow \infty)$, the first part of (2.2) follows from (2.1). The second part of (2.2) is trivially true.

(ii) These results are well-known from the extreme value theory, see e.g. [5]. ■

Now we estimate $E(M_n)$ using the classes RV and ORV. As before we shall assume that X_1 is a nonnegative r.v. and that $\{a_n\}_N$ is defined as in Lemma 2.1.

Theorem 2.2. *Let F denote a d.f. on R_+ .*

(i) *If $1 - F \in \text{ORV}$ is such that (2.1) holds, then*

$$\limsup_{n \rightarrow \infty} \frac{E(M_n)}{a_n} < \infty.$$

(ii) *If $1 - F \in \text{RV}_{-\alpha}$, $\alpha > 1$, then $\{E(M_n)\}_N \in \text{RV}_{1/\alpha}$ and*

$$(2.4) \quad \lim_{n \rightarrow \infty} \frac{E(M_n)}{a_n} = \Gamma(1 - 1/\alpha)$$

where $\Gamma(\cdot)$ denotes the gamma function.

Proof. (i) Since

$$\frac{E(M_n)}{a_n} = \int_0^\infty P\{M_n > a_n x\} dx,$$

the result is a consequence of (2.2).

(ii) Using (2.2), (2.3) and Lebesgue's theorem on dominated convergence we have

$$\lim_{n \rightarrow \infty} \frac{E(M_n)}{a_n} = \lim_{n \rightarrow \infty} \int_0^\infty P\{M_n > a_n x\} dx = \int_0^\infty (1 - e^{-x^{-\alpha}}) dx = \Gamma(1 - 1/\alpha).$$

Hence (2.4) follows; since $\{a_n\}_N \in \text{RV}_{1/\alpha}$, also $\{E(M_n)\}_N \in \text{RV}_{1/\alpha}$. ■

Corollary 2.3. *Let F denote a d.f. on R_+ , and let a_n be defined as before. If $1 - F \in \text{RV}_{-\alpha}$ with $\alpha > \beta$, then*

$$\lim_{n \rightarrow \infty} \frac{E(M_n^\beta)}{a_n^\beta} = \Gamma\left(1 - \frac{\beta}{\alpha}\right).$$

Proof. Use Theorem 2.2 (ii) with $Y_1 := X_1^\beta$. ■

In our next result we show that for each $k \in N \cup \{0\}$ the k -th differential $\Delta^k E(M_n)$

is regularly varying. To formulate the result we define

$$\begin{aligned}\Delta^0 E(M_n) &= E(M_n), \\ \Delta^{k+1} E(M_n) &= \Delta^k E(M_{n+1}) - \Delta^k E(M_n).\end{aligned}$$

Obviously, $\Delta^1 E(M_n) = \int_0^\infty F^n(x) (1 - F(x)) dx$ and by induction over k it follows that

$$\Delta^k E(M_n) = (-1)^{k+1} \int_0^\infty F^n(x) (1 - F(x))^k dx$$

so that

$$(2.5) \quad \frac{(-1)^{k+1} n^k \Delta^k E(M_n)}{a_n} = \int_0^\infty F^n(a_n x) [n(1 - F(a_n x))]^k dx.$$

Now we prove

Theorem 2.4. *If $1 - F \in \text{RV}_{-\alpha}$, $\alpha > 1$, then for each $k \in N$*

$$\begin{aligned}\text{(i)} \quad & \lim_{n \rightarrow \infty} \frac{(-1)^{k+1} n^k \Delta^k E(M_n)}{a_n} = \frac{1}{\alpha} \Gamma\left(k - \frac{1}{\alpha}\right); \\ \text{(ii)} \quad & \{(-1)^{k+1} \Delta^k E(M_n)\}_N \in \text{RV}_{1/\alpha - k}; \\ \text{(iii)} \quad & \lim_{n \rightarrow \infty} \frac{n[\Delta^k E(M_{n+1}) - \Delta^k E(M_n)]}{\Delta^k E(M_n)} = \frac{1}{\alpha} - k.\end{aligned}$$

Proof.

(i) Let $f(z, n, k) = z^n [n(1 - z)]^k$ ($0 \leq z \leq 1$). It is easily seen that $0 \leq f(z, n, k) \leq f(n/(n+k), n, k) \leq k^k$ ($0 \leq z \leq 1$); substituting $z = F(a_n x)$ we have

$$(2.6) \quad 0 \leq F^n(a_n x) (n(1 - F(a_n x)))^k \leq k^k \quad \text{for all } x \geq 0, \quad k \in N.$$

Also, from (2.1) with $\beta = \alpha - \varepsilon > 1$ we have

$$(2.7) \quad 0 \leq F^n(a_n x) [n(1 - F(a_n x))]^k \leq A^k x^{-k\beta} \quad \text{for all } x \geq 1, \quad n \geq n_0.$$

Now combine (2.5), (2.6) and (2.7) and Lebesgue's theorem on dominated convergence to obtain

$$\lim_{n \rightarrow \infty} \frac{(-1)^{k+1} n^k \Delta^k E(M_n)}{a_n} = \int_0^\infty e^{-x^{-\alpha}} x^{-ak} dx,$$

which proves (i).

(ii) This assertion immediately follows from (i) and the regular variation of $\{a_n\}_N$.

(iii) Using (i) with k replaced by $k + 1$ we have

$$\lim_{n \rightarrow \infty} \frac{(-1)^{k+2} n^{k+1} (\Delta^k E(M_{n+1}) - \Delta^k E(M_n))}{a_n} = \frac{1}{\alpha} \Gamma\left(k + 1 - \frac{1}{\alpha}\right).$$

Using (i) once again, we obtain

$$\lim_{n \rightarrow \infty} \frac{(-1)^n n (\Delta^k E(M_{n+1}) - \Delta^k E(M_n))}{\Delta^k E(M_n)} = \frac{\Gamma\left(k + 1 - \frac{1}{\alpha}\right)}{\Gamma\left(k - \frac{1}{\alpha}\right)},$$

from which the result (iii) follows. ■

Remark. The previous result shows that the sequence $\{E(M_n)\}_N$ is regularly varying together with all its "derivatives" $\Delta^k E(M_n)$. This illustrates that the operations $(X_1, X_2, \dots, X_n) \rightarrow M_n \rightarrow E(M_n)$ have very smoothing character.

In our next result we estimate $E(X_{n-k:n})$ for fixed k , as $n \rightarrow \infty$. We first express $E(X_{n-k:n})$ in terms of $\Delta^i E(M_j)$.

Lemma 2.5. *Let $n \in N$, $k \in N \cup \{0\}$, $k \leq n$, and let X_1, \dots, X_n be i.i.d. random variables. Then*

$$(2.8) \quad E(X_{n-k:n}) = \sum_{i=0}^k (-1)^i \binom{n}{i} \Delta^i E(M_{n-i})$$

and

$$(2.9) \quad E(X_{n-k+1:n+1}) - E(X_{n-k:n}) = (-1)^k \binom{n}{k} \Delta^{k+1} E(M_{n-k}).$$

Proof. The relation (2.8) is obviously true for $k = 0$ and $n \in N$. Suppose it holds for all $k \leq K$ and all $n \geq K$. We prove that the relation holds for $k = K + 1$ and all $n > K$. By (1.1) with $r = n - K - 1$ we have

$$E(X_{n-K-1:n}) = \frac{n}{K+1} E(X_{n-1-K:n-1}) - \frac{n-K-1}{K+1} E(X_{n-K:n}),$$

by (2.7) we obtain

$$\begin{aligned} E(X_{n-K-1:n}) &= \frac{1}{K+1} \sum_{i=0}^K (-1)^i \binom{n}{i} \{(n-i) \Delta^i E(M_{n-1-i}) - \\ &\quad - (n-K-1) \Delta^i E(M_{n-i})\} = \\ &= \frac{1}{K+1} \sum_{i=0}^K (-1)^i \binom{n}{i} \{(K+1-i) \Delta^i E(M_{n-i}) - \\ &\quad - (n-i) \Delta^{i+1} E(M_{n-i-1})\} = \\ &= \frac{1}{K+1} \sum_{i=0}^K (-1)^i \binom{n}{i} (K+1-i) \Delta^i E(M_{n-i}) + \\ &\quad + \frac{1}{K+1} \sum_{j=1}^{K+1} (-1)^j \binom{n}{j-1} (n-j+1) \Delta^j E(M_{n-j}) = \end{aligned}$$

$$\begin{aligned}
&= \frac{1}{K+1} \sum_{i=1}^K (-1)^i \left[\binom{n}{i} (K+1-i) + \binom{n}{i-1} (n-i+1) \right] \\
&\cdot \Delta^i E(M_{n-i}) + \Delta^0 E(M_n) + (-1)^{K+1} \binom{n}{K+1} \Delta^{K+1} E(M_{n-K-1}) = \\
&= \sum_0^{K+1} (-1)^i \binom{n}{i} \Delta^i E(M_{n-i}).
\end{aligned}$$

This proves (2.8).

To prove (2.9) we use (2.8) twice to obtain

$$\begin{aligned}
&E(X_{n+1-k:n+1}) - E(X_{n-k:n}) = \\
&= \sum_0^k (-1)^i \binom{n+1}{i} \Delta^i E(M_{n+1-i}) - \sum_0^k (-1)^i \binom{n}{i} \Delta^i E(M_{n-i}).
\end{aligned}$$

Using $\Delta^i E(M_{n-i}) = \Delta^i E(M_{n+1-i}) - \Delta^{i+1} E(M_{n-i})$ we obtain

$$\begin{aligned}
&E(X_{n+1-k:n+1}) - E(X_{n-k:n}) = \\
&= \sum_0^k (-1)^i \binom{n}{i} \Delta^{i+1} E(M_{n-i}) + \sum_1^k (-1)^i \binom{n}{i-1} \Delta^i E(M_{n+1-i}) = \\
&= (-1)^k \binom{n}{k} \Delta^{k+1} E(M_{n-k}). \quad \blacksquare
\end{aligned}$$

Now we prove

Theorem 2.6. *If $1 - F \in RV_{-\alpha}$ with $\alpha > 1$, then for each $k \in N \cup \{0\}$*

$$(2.10) \quad \lim_{n \rightarrow \infty} \frac{E(X_{n-k:n})}{a_n} = \frac{\Gamma\left(k+1 - \frac{1}{\alpha}\right)}{k!}$$

and

$$(2.11) \quad \lim_{n \rightarrow \infty} \frac{E(X_{n+1-k:n+1}) - E(X_{n-k:n})}{a_n} n = \frac{\Gamma\left(k+1 - \frac{1}{\alpha}\right)}{\alpha k!}.$$

Proof. From Theorem 2.2, Theorem 2.4 and the regular variation of $\{a_n\}_N$ we obtain that

$$\lim_{n \rightarrow \infty} \frac{(-1)^i \binom{n}{i} \Delta^i E(M_{n-i})}{a_n} = \begin{cases} \Gamma\left(1 - \frac{1}{\alpha}\right) & \text{if } i = 0, \\ -\frac{1}{\alpha} \Gamma\left(i - \frac{1}{\alpha}\right) \frac{1}{i!} & \text{if } i \geq 1. \end{cases}$$

Using (2.8) we conclude that

$$\lim_{n \rightarrow \infty} \frac{\mathbb{E}(X_{n-k:n})}{a_n} = \Gamma\left(1 - \frac{1}{\alpha}\right) - \sum_{i=1}^k \frac{\Gamma\left(i - \frac{1}{\alpha}\right)}{i! \alpha} = \frac{1}{k!} \Gamma\left(k + 1 - \frac{1}{\alpha}\right).$$

To prove the second assertion, we use again Theorem 2.4 and (2.9). ■

3. RATES OF CONVERGENCE

In Theorem 2.2 we proved that a regular variation of $1 - F$ implies that

$$\frac{\mathbb{E}(M_n)}{a_n} \rightarrow \Gamma\left(1 - \frac{1}{\alpha}\right).$$

In this section, we consider the rate of convergence of

$$\frac{\mathbb{E}(M_n)}{a_n} \text{ to } \Gamma\left(1 - \frac{1}{\alpha}\right).$$

We shall start with the following lemma.

Lemma 3.1. *Let $\alpha > 1$ and let $\eta \in \text{RV}_s$, where $s \geq \alpha$. If X_1 has a distribution function F such that $\mathbb{P}(X_1 \geq 0) = 1$ and*

$$(3.1) \quad \varrho_{\eta, F, \alpha} = \sup_{x \geq 0} \eta(x) |F(x) - e^{-x^{-\alpha}}| < \infty$$

then

$$(3.2) \quad \limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \mathbb{E}\left(\frac{M_n}{n^{1/\alpha}}\right) - \Gamma\left(1 - \frac{1}{\alpha}\right) \right| < \infty.$$

Proof. Introduce a function $\psi: \mathbb{R}_+ \rightarrow \mathbb{R}$ determined by

$$\psi(x) = \eta(x^{1/\alpha}) \text{ for all } x \geq 0,$$

and random variables $Y_i = X_i^\alpha$ for $i \in N$. We find that

$$\psi \in \text{RV}_u, \text{ where } u = \frac{s}{\alpha} \geq 1$$

and

$$N_n = \max\{Y_1; \dots; Y_n\} = M_n^\alpha.$$

Further, since X_1, X_2, \dots are i.i.d. with a common distribution function F , the random variables Y_1, Y_2, \dots are i.i.d. as well and their common distribution function denoted

by G has the form

$$G(x) = \begin{cases} F(x^{1/\alpha}) & \text{for } x \geq 0, \\ 0 & \text{for } x < 0. \end{cases}$$

It is useful to re-formulate the present theorem in terms of ψ , u , Y_i , G and N_n instead of η , s , X_i , F and M_n :

Let $\alpha > 1$ and let $\psi \in RV_u$ where $u \geq 1$. If Y_1 has a distribution function G such that $P(Y_1 \geq 0) = 1$ and

$$(3.1') \quad \begin{aligned} \varrho_{\eta, F, \alpha} &= \sup_{x \geq 0} \psi(x^\alpha) |G(x^\alpha) - e^{-x^{-\alpha}}| = \\ &= \sup_{x \geq 0} \psi(x) |G(x) - e^{-1/x}| < \infty \end{aligned}$$

then

$$(3.2') \quad \limsup_{n \rightarrow \infty} \frac{\psi(n)}{n} \left| E \left(\frac{N_n^{1/\alpha}}{n^{1/\alpha}} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| < \infty.$$

Notice that $\Gamma(1 - 1/\alpha)$ is equal to the mean value of a random variable having the distribution function

$$J_\alpha(x) = \begin{cases} e^{-x^{-\alpha}} & \text{for } x \geq 0, \\ 0 & \text{for } x < 0. \end{cases}$$

We have

$$\begin{aligned} \left| E \left(\frac{N_n^{1/\alpha}}{n^{1/\alpha}} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| &\leq \int_0^\infty \left| P \left(\frac{N_n}{n} \leq y^\alpha \right) - e^{-y^{-\alpha}} \right| dy \leq \\ &\leq \sup_{y \geq 0} \left| P \left(\frac{N_n}{n} \leq y^\alpha \right) - e^{-y^{-\alpha}} \right| + \int_1^\infty \left| G^n(ny^\alpha) - [e^{-1/ny^\alpha}]^n \right| dy \leq \\ &\leq \sup_{x \geq 0} \left| P \left(\frac{N_n}{n} \leq x \right) - e^{-1/x} \right| + \\ &+ \int_1^\infty \frac{n}{\psi(ny^\alpha)} \left[\sup_{z \geq 0} \psi(nz^\alpha) |G(nz^\alpha) - e^{-1/nz^\alpha}| \right] dy. \end{aligned}$$

Further, Rachev and Omev proved in [6] (Corollary 2.2) that

$$(3.3) \quad \limsup_{n \rightarrow \infty} \frac{\psi(n)}{n} \sup_{x \geq 0} \left| P \left(\frac{N_n}{n} \leq x \right) - e^{-1/x} \right| < \infty.$$

Since $\psi \in RV_u$ with $u \geq 1$ there exist positive constants n_0 and b such that

$$\frac{\psi(nx)}{\psi(n)} \geq bx^{1-(\alpha-1)/2\alpha} = bx^{(\alpha+1)/2\alpha} \quad \text{for all } x \geq 1 \text{ and } n \geq n_0$$

(see [4]). Finally, we obtain

$$\begin{aligned}
& \limsup_{n \rightarrow \infty} \frac{\psi(n)}{n} \left| \mathbb{E} \left(\frac{N_n^{1/\alpha}}{n^{1/\alpha}} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| \leq \\
& \leq \limsup_{n \rightarrow \infty} \frac{\psi(n)}{n} \sup_{x \geq 0} \left| \mathbb{P} \left(\frac{N_n}{n} \leq x \right) - e^{-1/x} \right| + \\
& + \sup_{x \geq 0} \psi(x) |G(x) - e^{-1/x}| \limsup_{n \rightarrow \infty} \int_1^\infty \frac{\psi(n)}{\psi(ny^\alpha)} dy \leq \\
& \leq \limsup_{n \rightarrow \infty} \frac{\psi(n)}{n} \sup_{x \geq 0} \left| \mathbb{P} \left(\frac{N_n}{n} \leq x \right) - e^{-1/x} \right| + \\
& + \varrho_{\eta, F, \alpha} \frac{1}{b} \int_1^\infty y^{-(\alpha+1)/2} dy < \infty.
\end{aligned}$$

Thus, the proof of (3.2') which is equivalent to (3.2) is completed.

We are now able to give the desired result concerning the rate of convergence of $\mathbb{E}(M_n)/a_n$ to $\Gamma(1 - 1/\alpha)$.

Theorem 3.2. *Let X_1 have a distribution function F such that $\mathbb{P}(X_1 \geq 0) = 1$ and $1 - F \in \text{RV}_{-\alpha}$ where $\alpha > 1$. Let $c > 0$ and let $\eta \in \text{RV}_s$ where $s \geq \alpha$. If*

$$(3.4) \quad \sup_{x \geq 0} \eta(x) |F(cx) - e^{-x^{-\alpha}}| < \infty$$

and

$$(3.5) \quad \limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \frac{a_n}{cn^{1/\alpha}} - 1 \right| < \infty$$

then

$$(3.6) \quad \limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \mathbb{E} \left(\frac{M_n}{a_n} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| < \infty.$$

Proof. Introduce auxiliary random variables $Z_i = (1/c)X_i$ for $i \in N$ and $K_n = \max \{Z_1; \dots; Z_n\} = M_n/c$. We find that Z_1, Z_2, \dots are i.i.d. random variables with a common distribution function $H(x) = F(cx)$. Our aim is to apply Lemma 3.1 with X_i, F and M_n substituted by Z_i, H and K_n . To this end we need to verify the validity of (3.1). We have

$$\begin{aligned}
\varrho_{\eta, H, \alpha} &= \sup_{x \geq 0} \eta(x) |H(x) - e^{-x^{-\alpha}}| = \\
&= \sup_{x \geq 0} \eta(x) |F(cx) - e^{-x^{-\alpha}}| < \infty
\end{aligned}$$

by (3.4). Thus, we know from Lemma 3.1 that

$$(3.7) \quad \limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \mathbb{E} \left(\frac{K_n}{n^{1/\alpha}} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| = \\ = \limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \mathbb{E} \left(\frac{M_n}{c \cdot n^{1/\alpha}} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| < \infty.$$

Finally, we obtain

$$\left| \frac{\mathbb{E}(M_n)}{a_n} - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| \leq \\ \leq \left| \frac{\mathbb{E}(M_n)}{a_n} - \frac{\mathbb{E}(M_n)}{c \cdot n^{1/\alpha}} \right| + \left| \mathbb{E} \left(\frac{M_n}{c \cdot n^{1/\alpha}} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right|$$

so that

$$\limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \frac{\mathbb{E}(M_n)}{a_n} - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| \leq \\ \leq \limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \frac{\mathbb{E}(M_n)}{a_n} \left| 1 - \frac{a_n}{c \cdot n^{1/\alpha}} \right| + \\ + \limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \mathbb{E} \left(\frac{M_n}{c \cdot n^{1/\alpha}} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| < \infty$$

by (2.4), (3.5) and (3.7).

We conclude the present section by demonstrating the contribution of Theorem 3.2 by the following example.

Example. Let

$$F(x) = \begin{cases} 0 & \text{for } x < 1, \\ 1 - \frac{x^2 + 1}{2x^4} & \text{for } x \geq 1. \end{cases}$$

We find that $1 - F \in \text{RV}_{-2}$, i.e. $\alpha = 2$ in this case, and by Theorem 2.2 (ii)

$$\lim_{n \rightarrow \infty} \frac{\mathbb{E}(M_n)}{a_n} = \Gamma \left(\frac{1}{2} \right) = \sqrt{\pi}.$$

The distribution function F is continuous on \mathbb{R} and the solution of the equation

$$1 - F(x) = \frac{1}{n}$$

results in

$$a_n = \frac{1}{2} \sqrt{(n + \sqrt{(n^2 + 8n)})}.$$

Thus,

$$\lim_{n \rightarrow \infty} \frac{a_n}{n^{1/2}} = \lim_{n \rightarrow \infty} \frac{1}{2} \sqrt{\left[1 + \sqrt{\left(1 + \frac{8}{n} \right)} \right]} = \frac{\sqrt{2}}{2},$$

i.e. we put $c = \frac{1}{2}\sqrt{2}$ in Theorem 3.2 – cf. (3.5). Further, $cx \geq 1$ if and only if $x \geq \sqrt{2}$ and

$$\begin{aligned} |F(cx) - e^{-x^{-\alpha}}| &= \left| 1 - x^{-2} - 2x^{-4} - \sum_{k=0}^{\infty} \frac{(-1)^k x^{-2k}}{k!} \right| = \\ &= \frac{5}{2}x^{-4} + x^{-4} \sum_{k=3}^{\infty} x^{-2k+4} \leq \frac{7}{2}x^{-4} \quad \text{for } x \geq \sqrt{2} \end{aligned}$$

and

$$|F(cx) - e^{-x^{-\alpha}}| = e^{-x^{-2}} \leq e^{-1/2} \quad \text{for } x \in [0; \sqrt{2}].$$

With respect to (3.4) and to the fact that $\alpha = 2$, we can take e.g. $\eta(x) = x^s$ where $s \in [2; 4]$. However, the greater is the exponent s the stronger will be the achieved result so that we choose $\eta(x) = x^4$. We find that

$$\sup_{x \geq 0} \eta(x) |F(cx) - e^{-x^{-\alpha}}| \leq \max \left\{ \frac{7}{2}; 4e^{-1/2} \right\} < \infty,$$

i.e. the assumption (3.4) is fulfilled. Finally, easy calculation yields

$$\begin{aligned} \limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \frac{a_n}{c \cdot n^{1/\alpha}} - 1 \right| &= \\ = \lim_{n \rightarrow \infty} n \left| -1 + \frac{1}{\sqrt{2}} \sqrt{1 + \sqrt{1 + \frac{8}{n}}} \right| &= 1 < \infty, \end{aligned}$$

i.e. the assumption (3.5) of Theorem 3.2 is fulfilled as well. Thus, its assertion reads

$$\limsup_{n \rightarrow \infty} n \left| \frac{E(M_n)}{a_n} - \sqrt{\pi} \right| < \infty,$$

i.e. the rate of convergence of $E(M_n)/a_n$ to its limit $\sqrt{\pi}$ is at least the same as the rate of convergence of $1/n$ to 0.

Remark. From Theorem 3.2 we find that the choice of the normalizing sequence $\{a_n\}_N$ is very important. We made the choice $a_n = \inf \{x; 1 - F(x) \leq 1/n\}$. On the other hand, if we make the choice $b_n = \inf \{x; -\log F(x) \leq 1/n\}$ we obviously obtain all results of Section 2 with a_n replaced by b_n . This follows from the obvious asymptotic equality $-\log F(x) \sim 1 - F(x)$ for $x \rightarrow \infty$ and from the assumption that $1 - F \in \text{RV}_{-\alpha}$ with the index $\alpha > 0$. Moreover, with the latter choice condition (3.4) alone implies

$$\limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| E \left(\frac{M_n}{b_n} \right) - \Gamma \left(1 - \frac{1}{\alpha} \right) \right| < \infty.$$

To see this, we show that (3.5) holds for the sequence $\{b_n\}_N$. From (3.4) we obtain

$$\limsup_{x \rightarrow \infty} \eta(x) |\log F(cx) + x^{-\alpha}| < \infty.$$

With our choice of b_n we have

$$\limsup_{n \rightarrow \infty} \eta \left(\frac{b_n}{c} \right) \left| \frac{1}{n} - \left(\frac{b_n}{c} \right)^{-\alpha} \right| < \infty .$$

Since $b_n \sim c \cdot n^{1/\alpha}$ for $n \rightarrow \infty$ and since $\eta(x)$ is regularly varying we find that

$$\limsup_{n \rightarrow \infty} \frac{\eta(n^{1/\alpha})}{n} \left| \frac{b_n^\alpha}{c^\alpha \cdot n} - 1 \right| < \infty$$

and hence (3.5) with a_n replaced by b_n follows.

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Souhrn

MOMENTY POŘÁDKOVÝCH STATISTIK

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Označme $X_{1:n} \leq \dots \leq X_{n:n} = M_n$ pořádkové statistiky z náhodného výběru X_1, \dots, X_n o rozsahu n a necht' rozložení náhodné veličiny X_1 s distribuční funkcí F je soustředěno na \mathbf{R}_+ . Článek pojednává o asymptotickém chování hodnot $E(M_n)$ a $E(X_{n-k:n})$ pro pevné k a $n \rightarrow \infty$. Speciální pozornost je věnována případu, že $1 - F$ je regulárně se měnící funkce.

Резюме

МОМЕНТЫ ПОРЯДКОВЫХ СТАТИСТИК

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Пусть $X_{1:n} \leq \dots \leq X_{n:n} = M_n$ — порядковые статистики из случайной выборки X_1, \dots, X_n размера n и пусть распределение случайной величины X_1 с функцией распределения F сосредоточено на \mathbf{R}_+ . Статья трактует асимптотическое поведение математических ожиданий $E(M_n)$ и $E(X_{n-k:n})$ для постоянного k и $n \rightarrow \infty$. Специальное внимание уделяется случаю, что $(1 - F)$ — регулярно меняющаяся функция.

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