

CONVERGENCE IN DISTRIBUTION OF QUOTIENTS OF ORDER STATISTICS

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Let X_1, X_2, \dots be i.i.d. random variables with continuous distribution function $F < 1$. It is known that if $1 - F(x)$ varies regularly of order $-\rho$, the successive quotients of the order statistics in decreasing order of X_1, \dots, X_n are asymptotically independent, as $n \rightarrow \infty$, with distribution functions $x^{k\rho}$, $k = 1, 2, \dots$. A strong converse is proved, viz. convergence in distribution of this type of one of the quotients implies regular variation of $1 - F(x)$.

order statistics	partial maxima
limit theorem	regular variation
Wiener-Tauber theorem	

1. Introduction and results

Let X_1, X_2, \dots be independent random variables with common continuous distribution function F such that $F(x) < 1$ for all x . By $M_{n0}, M_{n1}, \dots, M_{n,n-1}$ we denote the order statistics in decreasing order of X_1, \dots, X_n , so that $X_{n0} = \max \{X_1, \dots, X_n\}$. The index n will be suppressed occasionally. The following result is contained in the work of Dwass [2, 3], Lamperti [6] and Polfeldt [8]:

Theorem 1. *If $1 - F(x)$ varies regularly of order $-\rho \leq 0$, as $x \rightarrow \infty$, then for $k = 1, 2, \dots$ and $0 < \xi_i < 1, i = 1, \dots, k$,*

$$\lim_{n \rightarrow \infty} P[M_{nj} M_{n,j-1}^{-1} < \xi_j, j = 1, \dots, k] = \xi_1^\rho \xi_2^{2\rho} \dots \xi_k^{k\rho}. \quad (1)$$

We give a proof in Section 2, but our main goal is proving the following strong converse:

Theorem 2. *If for some $j \geq 1$, $\xi \in (0, 1)$ and $\rho \geq 0$,*

$$\lim_{n \rightarrow \infty} \mathbf{P}[M_{nj} M_{n,j-1}^{-1} < \xi] = \xi^{\rho}, \quad (2)$$

then

$$\lim_{y \rightarrow \infty} (1 - F(\xi^{-1}y))/(1 - F(y)) = \xi^{\rho}. \quad (3)$$

Corollary. *If (2) holds for all $\xi \in (0, 1)$ (or even for ξ_1 and ξ_2 with $(\log \xi_1)/(\log \xi_2)$ irrational, see [5, Theorem 1.1.2]), then $1 - F(x)$ varies regularly of order $-\rho$ as $x \rightarrow \infty$.*

Let Y_1, Y_2, \dots be independent random variables with common continuous distribution function $G < 1$ and let $N_{n0}, \dots, N_{n,n-1}$ be the order statistics in decreasing order of Y_1, \dots, Y_n . By considering $X_i = \exp(Y_i)$ with distribution function $F(x) = G(\log x)$ and order statistics $M_{nj} = \exp(N_{nj})$, one immediately derives from Theorems 1 and 2:

Theorem 3. *If $(1 - G(y + \eta))/(1 - G(y)) \rightarrow \exp(-\rho\eta)$ as $y \rightarrow \infty$, for all $\eta > 0$, then for $k = 1, 2, \dots$ and $\eta_i > 0, i = 1, \dots, k$,*

$$\lim_{n \rightarrow \infty} \mathbf{P}[N_{n,j-1} - N_{n,j} > \eta_j, j = 1, \dots, k] = \exp(-\rho\eta_1 - 2\rho\eta_2 - \dots - k\rho\eta_k).$$

Theorem 4. *If for some $j \geq 1$, $\eta > 0$ and $\rho \geq 0$,*

$$\lim_{n \rightarrow \infty} \mathbf{P}[N_{n,j-1} - N_{n,j} > \eta] = \exp(-\rho j \eta),$$

then

$$(1 - G(y + \eta))/(1 - G(y)) \rightarrow \exp(-\rho\eta) \quad \text{as } y \rightarrow \infty.$$

Remarks. Theorems 1 and 2 with $\rho = 0$ mean that $M_{nj} M_{n,j-1}^{-1} \rightarrow 0$ in probability if and only if $1 - F(x)$ varies slowly at infinity. We may extend Theorems 1 and 2 to $\rho = \infty$, with $\xi^{\rho} = 0, 0 < \xi < 1$, by stating that $M_{nj} M_{n,j-1}^{-1} \rightarrow 1$ in probability if and only if $1 - F(x)$ varies rapidly at infinity, i.e. $(1 - F(ax))/(1 - F(x)) \rightarrow 0, a > 1$. No change in the proofs is required. An analogous remark applies to Theorems 3 and 4.

2. Proof of Theorem 1

Since $F(x) < 1$, we have $M_{nj} \rightarrow \infty$ a.s. as $n \rightarrow \infty$. So it is no restriction to assume $F(0^+) = 0$. By symmetry we have, for $1 \leq r < n$,

$$P[M_{r-1} > a \mid M_{n-1}, \dots, M_r] = \left(\frac{1 - F(a)}{1 - F(M_r)} \right)^r, \quad a > M_r.$$

So, for $1 \leq r \leq k$,

$$P[M_k M_{k-1}^{-1} < \xi_k, \dots, M_r M_{r-1}^{-1} < \xi_r] = \int_A (1 - F(\xi_r^{-1} M_r))^r (1 - F(M_r))^{-r} dP,$$

where $A = \Omega$ if $r = k$ and $A = \{M_k M_{k-1}^{-1} < \xi_k, \dots, M_{r+1} M_r^{-1} < \xi_{r+1}\}$ if $r < k$. Since by the regular variation of $1 - F$,

$$\begin{aligned} & \left| \xi_r^{+\rho r} P[A] - \int_A (1 - F(\xi_r^{-1} M_r))^r (1 - F(M_r))^{-r} dP \right| \leq \\ & \leq \int_{\Omega} |\xi_r^{+\rho r} - (1 - F(\xi_r^{-1} M_r))^r (1 - F(M_r))^{-r}| dP \rightarrow 0. \end{aligned}$$

we have

$$\lim_{n \rightarrow \infty} |\xi_r^{+\rho r} P[A] - P[M_{nk}/M_{n,k-1} < \xi_k, \dots, M_{nr}/M_{n,r-1} < \xi_r]| = 0,$$

from which the theorem follows by induction with respect to r .

3. Proof of Theorem 2

Since $F(x) < 1$, we have $M_{nk} \rightarrow \infty$ a.s. as $n \rightarrow \infty$, so that it is no restriction to assume $F(0^+) = 0$. Put

$$\Theta(x) = 1/(1 - F(x)), \quad x \geq 0, \tag{4}$$

$$\beta(y) = \inf \{x: x \geq 0, \Theta(x) \geq y\}, \quad y \geq 1. \tag{5}$$

Then by the continuity of F , we have

$$\beta(\Theta(x)) = x \quad \text{a.e. } [F] \text{ on } [0, \infty), \tag{6}$$

$$\Theta(\beta(y)) = y, \quad y \geq 1. \quad (7)$$

For $n \geq j + 1 \geq 2$ we obtain

$$\begin{aligned} P[M_{j-1} > \xi^{-1} M_j] &= \\ &= n(n-1) \binom{n-2}{j-1} \int_0^\infty (F(u))^{n-j-1} dF(u) \int_{u\xi^{-1}}^\infty (1-F(v))^{j-1} dF(v) \\ &= \frac{n(n-1) \dots (n-j)}{j!} \int_0^\infty (F(u))^{n-j-1} (1-F(\xi^{-1}u))^j dF(u). \end{aligned}$$

Taking $\Theta(u)$ as new integration variable, we find, using (6),

$$P[M_{j-1} > \xi^{-1} M_j] = \frac{n(n-1) \dots (n-j)}{j!} \int_1^\infty (1-x^{-1})^{n-j-1} x^{-j-2} \gamma(x) dx, \quad (8)$$

with $\gamma(x)$ for $x \geq 1$ defined by

$$\gamma(x) = x^j \{1 - F(\xi^{-1}\beta(x))\}^j = \{\Theta(\beta(x))/\Theta(\xi^{-1}\beta(x))\}^j. \quad (9)$$

Defining the probability density q_j on $(0, \infty)$ by

$$q_j(x) = \frac{1}{j!} x^{-j-2} \exp(-x^{-1}), \quad x > 0, \quad (10)$$

we have

$$\lim_{n \rightarrow \infty} \int_1^\infty \left| n^{-1} q_j(n^{-1}x) - \frac{n(n-1) \dots (n-j)}{j!} (1-x^{-1})^{n-j-1} x^{-j-2} \right| dx = 0, \quad (11)$$

$$\lim_{\substack{\lambda \rightarrow \infty \\ |\lambda - \mu| \leq 1}} \int_0^\infty |\lambda^{-1} q_j(\lambda^{-1}x) - \mu^{-1} q_j(\mu^{-1}x)| dx = 0. \quad (12)$$

We thank the referee for the following short proof of (11) and (12). By putting $x = ny$ in (11) and $x = \lambda y$ in (12), the integrals are reduced to

$$\int |p_n(y) - p(y)| dy,$$

where $p_n(y) \rightarrow p(y)$ as $n \rightarrow \infty$, and p_n, p are probability densities. Now apply Scheffé's lemma ([9], [1, App. II]).

So from (2), (8) and (11), since γ is bounded,

$$\lim_{n \rightarrow \infty} \int_1^{\infty} n^{-1} q_j(n^{-1}x) \gamma(x) dx = \xi^{j\rho} . \tag{13}$$

Defining $\gamma(x) = 1$, say, for $0 \leq x < 1$, we find from (13) and (12),

$$\lim_{\lambda \rightarrow \infty} \int_0^{\infty} \lambda^{-1} q_j(\lambda^{-1}x) \gamma(x) dx = \xi^{j\rho} . \tag{14}$$

We now use the Wiener–Tauber Theorem 3.2 given in [11, Section (8.3)]. Since γ is bounded and $q_j \in U$, i.e.

$$\int_0^{\infty} \mu^{-1} q_j(\mu^{-1}x) a(x) dx = 0, \quad \mu > 0 ,$$

implies $a(x) \equiv 0$ for bounded continuous a , we have

$$\lim_{\lambda \rightarrow \infty} \int_0^{\infty} \lambda^{-1} f(\lambda^{-1}x) \gamma(x) dx = \xi^{j\rho} , \tag{15}$$

for any probability density f . We now prove

$$\lim_{\lambda \rightarrow \infty} \lambda \{1 - F(\xi^{-1} \beta(\lambda y))\} = y^{-1} \xi^\rho, \quad y > 0 . \tag{16}$$

Since $\lambda \{1 - F(\xi^{-1} \beta(\lambda y))\}$ decreases with y and is bounded by y^{-1} , it is sufficient to show that $\lambda_k \rightarrow \infty$ and

$$\lim_{k \rightarrow \infty} \lambda_k \{1 - F(\xi^{-1} \beta(\lambda_k y))\} = \psi(y), \quad y \notin D(\psi) , \tag{17}$$

where $D(\psi)$ is the set of discontinuity points of ψ on $(0, \infty)$, imply $\psi(y) = y^{-1} \xi^\rho$. Now (17) implies, by (9) and (15), since γ is bounded and $D(\psi)$ is countable, that

$$\int_0^{\infty} f(x) x^j \psi^j(x) dx = \xi^{j\rho}$$

for every probability density f , so that

$$x^j \psi^j(x) = \xi^{j\rho} ,$$

and (16) follows. From (16) we easily derive (3), since $\beta(y) \rightarrow \infty$ as $y \rightarrow \infty$. See the proofs of [4, Lemma 2, Section VIII.8] and [5, Theorem 1.1.3].

Remark. Shorrock [10, Theorem 3] proved a result analogous to our Theorem 1 for quotients of upper record values. Attempts to prove a converse failed, even when using the strengthened version of Wiener's Tauberian theorem given by Moh [7].

References

- [1] P. Billingsley, *Convergence of Probability Measures* (Wiley, New York, 1968).
- [2] M. Dwass, *Extremal processes*, *Ann. Math. Statist.* 35 (1964) 1718–1725.
- [3] M. Dwass, *Extremal processes II*, *Illinois J. Math.* 10 (1966) 381–391.
- [4] W. Feller, *An Introduction to Probability Theory and its Applications*, Vol. II, 1st ed. (Wiley, New York, 1966).
- [5] L. de Haan, *On regular variation and its applications to the weak convergence of sample extremes*, *Math. Centre Tracts* 32 (Math. Centrum, Amsterdam, 1970).
- [6] J. Lamperti, *On extreme order statistics*, *Ann. Math. Statist.* 35 (1964) 1726–1736.
- [7] T.T. Moh, *On a general Tauberian theorem*, *Proc. Am. Math. Soc.* 36 (1972) 167–172.
- [8] Th. Polfeldt, *Skand. Aktuarietidskrift* (1970) Suppl. 1/2, 44.
- [9] H. Scheffé, *A useful convergence theorem for probability distributions*, *Ann. Math. Statist.* 18 (1947) 434–438.
- [10] R.W. Shorrock, *On record values and record times*, *J. Appl. Probab.* 9 (1972) 316–326.
- [11] D.V. Widder, *An Introduction to Transform Theory* (Academic Press, New York, 1971).