

A limit theorem for sums of i.i.d. random variables with slowly varying tail probability

By

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§1. Introduction.

Let X_1, X_2, \dots be nonnegative independent random variables with common distribution function $F(x)$ belonging to the domain of attraction of stable law with index α ($0 < \alpha < 1$); i.e.,

$$(1.1) \quad 1 - F(x) \sim 1/(x^\alpha L(x)) \quad \text{as } x \rightarrow \infty,$$

for some slowly varying $L(x)$ (≥ 0). Here $f(x) \sim g(x)$ means that $\lim f(x)/g(x) = 1$. Let $S_n = X_1 + X_2 + \dots + X_n$, $n = 1, 2, \dots$. It is well known as Skorohod's invariance principle that, for suitably chosen $v(t)$, $S_{[nt]}/v(n)$ converges in law to the one-sided stable process in the function space $D([0, \infty) \rightarrow \mathbf{R})$ endowed with J_1 -topology (see [7] for the definition and [9] for the result).

In this paper we will consider the extreme case of (1.1) as $\alpha \downarrow 0$; we treat the case where

$$(1.2) \quad 1 - F(x) \sim 1/L(x) \quad \text{as } x \rightarrow \infty.$$

As is well known, under condition (1.2), every linear normalization $a_n S_n + b_n$ leads to a degenerate limiting distribution. However, if we allow non-linear normalizations, we have

Darling's Theorem ([2]). *If (1.2) is satisfied, then*

$$\lim_{n \rightarrow \infty} P\left[\frac{1}{n} L(S_n) \leq x\right] = e^{-1/x}, \quad x > 0.$$

(In [2] some technical conditions are assumed but they may be removed.) The convergence of the semi-group of the process $t \wedge \rightarrow \frac{1}{n} L(S_{[nt]})$ was obtained by S. Watanabe [10], and the purpose of this paper is to show the weak convergence of this process in J_1 -topology. Our idea of the proof is quite different from that of Darling or Watanabe and so may be read independently.

§2. Main theorem.

Let $\xi=(\xi(t))_{t \geq 0}$ be the symmetric Cauchy process such that $E[e^{i\theta\xi(t)}]=e^{-\pi t|\theta|}$, $\theta \in \mathbf{R}$. In other words, ξ is the Lévy process with Lévy measure $x^{-2}dx$. Define

$$(2.1) \quad m(t) = \max_{0 < s \leq t} \Delta\xi(s) \quad \text{where} \quad \Delta\xi(s) = \xi(s) - \xi(s-).$$

Thus $m=(m(t))_{t \geq 0}$ ($m(0)=0$) is the maximum process of the Poisson point process with intensity measure $x^{-2}dx dt$, and therefore it is easy to compute the finite-dimensional marginal distributions; for $0 \leq t_1 < \dots < t_n$ and $0 \leq x_1 < \dots < x_n$,

$$(2.2) \quad P[m(t_1) \leq x_1, \dots, m(t_n) \leq x_n] = F(x_1)^{t_1} F(x_2)^{t_2-t_1} \dots F(x_n)^{t_n-t_{n-1}},$$

where $F(x)=e^{-1/x}$, $x > 0$. Processes with the property (2.2) are often referred to as *extremal processes* (see [3]). Our main result is

Theorem 2.1. *Let $L(x) \geq 0$, $x \geq 0$, be a nondecreasing function varying slowly at infinity. Then, under the assumption (1.2),*

$$\frac{1}{n} L(S_{[nt]}) \xrightarrow{\mathcal{D}} m(t) \quad \text{as } n \rightarrow \infty \quad \text{in } D([0, \infty) \rightarrow \mathbf{R})$$

where $\xrightarrow{\mathcal{D}}$ denotes the weak convergence in J_1 -topology.

Proof. Since the convergence of finite-dimensional marginal distributions are obtained by [10], it remains to prove the tightness of the processes. However, it does not seem easy to check the well-known conditions for tightness such as Chentsov's moment condition (see page 128 of [1]), so we will adopt a direct method: On the probability space where ξ (and hence m) is defined we will construct processes ζ_n , $n=1, 2, \dots$ which are distributed like $(1/n)L(S_{[nt]})$, $n=1, 2, \dots$ and which converge to m almost surely in J_1 -topology.

Now let $\eta(t) = \sum_{s \leq t} (\Delta^+ \xi(s))^2$ where $\Delta^+ \xi(s) = \max\{\Delta\xi(s), 0\}$. In other words, $\eta(t) = \int_0^{t+} \int_{x>0} x^2 N(du dx)$, $N(du dx)$ being the Poisson random measure defined by the jumps of the Cauchy process ξ . Clearly $\eta=(\eta(t))_{t \geq 0}$ is a one-sided stable process with index 1/2 and the Lévy measure is given by $\mu[x, \infty) = x^{-1/2}$, $x > 0$. Observe that $m(t)$ may be expressed by using η instead of ξ ;

$$(2.3) \quad m(t) = \{\max_{0 \leq s \leq t} \Delta\eta(s)\}^{1/2}, \quad \Delta\eta(s) = \eta(s) - \eta(s-).$$

Indeed this is immediate from (2.1) and the definition of η . Let $F_0(x) = P[\eta(1) \leq x]$, $x \in \mathbf{R}$. It is well known that

$$(2.4) \quad 1 - F_0(x) \sim x^{-1/2} \quad \text{as } x \rightarrow \infty.$$

This may be confirmed as follows. Since $\eta(n)/n^2$ clearly converges in law to $\eta(1)$,

appealing to Theorem 4 in page 124 of [5] we have that for every $x > 0$,

$$n \{1 - F_0(n^2 x)\} \rightarrow x^{-1/2} \text{ as } n \rightarrow \infty,$$

which proves (2.4). We also define $\phi(x) = F^{-1}(F_0(x))$ ($x \in \mathbf{R}$). Throughout the inverse of a nondecreasing function is chosen so that it is right-continuous;

$$F^{-1}(x) = \inf \{u: F(u) > x\}, \quad 0 < x < 1.$$

Let, for $n = 1, 2, \dots$,

$$(2.5) \quad \eta_{ni} = \eta\left(\frac{i}{n}\right) - \eta\left(\frac{i-1}{n}\right), \quad i = 1, 2, \dots,$$

$$(2.6) \quad \zeta_n(t) = \frac{1}{n} L(\sum_{i \leq nt} \phi(n^2 \eta_{ni})), \quad t \geq 0,$$

$$(2.7) \quad m_n(t) = \begin{cases} (\max_{i \leq nt} \eta_{ni})^{1/2}, & t \geq 1/n \\ 0, & 0 \leq t < 1/n. \end{cases}$$

We are now ready to explain the story of the proof. We will prove the following three.

- (1) $\zeta_n = (\zeta_n(t))_{t \geq 0}$ is identical in law to $(\frac{1}{n} L(\sum_{i \leq nt} X_i))_{t \geq 0}$, $n = 1, 2, \dots$.
- (2) $m_n \rightarrow m$ (in J_1) a.s.
- (3) $\sup_{0 \leq t \leq T} |\zeta_n(t) - m_n(t)| \rightarrow 0$, a.s., for every $T > 0$.

Indeed, (2) and (3) imply that $\zeta_n \rightarrow m$ (in J_1) a.s., which together with (1) clearly proves our Theorem.

The proof of (1) and (2) are easy. To see (1) it suffices to show that $\phi(\eta(1))$ (and hence $\phi(n^2 \eta_{ni})$) is distributed like X_1 . However, this is obvious from the definition $\phi(x) = F^{-1}(F_0(x))$. In fact ϕ was defined so that (1) holds. To see (2) define $\eta_n(t) = \eta([nt]/n)$, $n = 1, 2, \dots$. Since $\eta_n(t) \rightarrow \eta(t)$ (in J_1), we have from the continuity theorem that

$$\max_{i \leq nt} \eta_{ni} = \max_{0 \leq s \leq t} \Delta \eta_n(t) \rightarrow \max_{0 \leq s \leq t} \Delta \eta(t) \text{ (in } J_1).$$

Therefore, considering the square root of each side, we obtain from (2.3) and (2.7) that $m_n(t) \rightarrow m(t)$ (in J_1) a.s., which is the desired result.

For the proof of (3) we need the next two key lemmas.

Lemma 2.2. For every $\lambda > 0$ and $T > 0$,

$$\lim_{n \rightarrow \infty} \sup_{0 \leq x \leq T} \left| \frac{1}{n} L(\lambda \phi(n^2 x^2)) - x \right| = 0.$$

Lemma 2.3. For every $\varepsilon > 0$ and $T > 0$ ($\varepsilon < T$) and for almost sure ω , there exist $K_1, K_2 > 0$ and n_0 depending on $\{\varepsilon, T, \omega\}$ such that, for every $t \in [\varepsilon, T]$ and $n \geq n_0$,

$$(2.8) \quad K_1 \phi(n^2 m_n(t)^2) \leq \sum_{i \leq n^t} \phi(n^2 \eta_{ni}) \leq K_2 \phi(n^2 m_n(t)^2).$$

We postpone the proof until next section and return to the proof of (3). Since $L(x)$ is nondecreasing, applying $L(\cdot)$ to each side of (2.8) and dividing by n , we obtain

$$\frac{1}{n} L(K_1 \phi(n^2 m_n(t)^2)) \leq \zeta_n(t) \leq \frac{1}{n} L(K_2 \phi(n^2 m_n(t)^2)).$$

Therefore, we have

$$(2.9) \quad \begin{aligned} & \sup_{\varepsilon \leq t \leq T} |\zeta_n(t) - m_n(t)| \\ & \leq \sup_{0 \leq t \leq T} \max_{i=1,2} \left| \frac{1}{n} L(K_i \phi(n^2 m_n(t)^2)) - m_n(t) \right| \\ & \leq \max_{i=1,2} \sup_{0 \leq x \leq m_n(T)} \left| \frac{1}{n} L(K_i \phi(n^2 x^2)) - x \right| \end{aligned}$$

which converges to 0 a.s. by Lemma 2.2. and (2). Since $\zeta_n(t)$ and $m_n(t)$ are monotone in t , we also see

$$\begin{aligned} & \lim_{\varepsilon \rightarrow 0} \limsup_{n \rightarrow \infty} \sup_{0 \leq t \leq \varepsilon} |\zeta_n(t) - m_n(t)| \\ & \leq \lim_{\varepsilon \rightarrow 0} \limsup_{n \rightarrow \infty} (\zeta_n(\varepsilon) + m_n(\varepsilon)) \\ & = \lim_{\varepsilon \rightarrow 0} 2m(\varepsilon) = 0 \text{ a.s.} \end{aligned}$$

This completes the proof of (3) and hence of Theorem 2.1.

Remark 2.4. We have a very simple proof for the special case where $L(x) = \log(x+1)$. Let $M_0 = 0$ and $M_n = \max(X_1, \dots, X_n)$, $n \geq 1$. Then we have

$$(2.10) \quad \frac{1}{n} L(M_{[nt]}) \leq \frac{1}{n} L(S_{[nt]}) \leq \frac{1}{n} L(nM_{[nt]}).$$

Since $\log(nx) = \log n + \log x$, it is easy to see that the first and the third process (and hence the second) should have the same limiting processes. Thus our problem may be reduced to the study of M_n , which was treated by [4] and [6]. However, for general slowly varying $L(x)$, we cannot apply this argument. For example, let $L(x) = \exp \sqrt{\log(x+1)}$. $L(x)$ is a slowly varying function and we can show that the first and the second processes in (2.10) converge to $m(t)$ but the third to $\sqrt{e} m(t)$.

§3. Proofs of Lemmas.

Proof of Lemma 2.2. Since $\phi(x) = F^{-1}(F_0(x))$, we have $1 - F(\phi(x)) = 1 - F_0(x)$ provided that $F(x)$ is continuous. In general it holds that

$$(3.1) \quad 1 - F(\phi(x)) \leq 1 - F_0(x) \leq 1 - F(\phi(x)/2).$$

By (1.2) both $1 - F(\phi(x))$ and $1 - F(\phi(x)/2)$ are asymptotically equal to $1/L(\phi(x))$ as $x \rightarrow \infty$, while $1 - F_0(x) \sim x^{-1/2}$ (see (2.4)). Therefore, (3.1) implies that $L(\phi(x)) \sim x^{1/2}$, or equivalently, $L(\phi(x^2)) \sim x$ as $x \rightarrow \infty$. This also proves that

$$(3.2) \quad \lim_{x \rightarrow \infty} L(\lambda \phi(x^2))/x = 1, \lambda > 0,$$

since $L(x)$ is a slowly varying function. Thus we have for every $\lambda > 0$ and $x \geq 0$,

$$(3.3) \quad \lim_{n \rightarrow \infty} \frac{1}{n} L(\lambda \phi(n^2 x^2)) = x.$$

Since $\frac{1}{n} L(\lambda \phi(n^2 x^2))$ is monotone in x and the limiting function is continuous, the convergence in (3.3) is automatically uniform for x on every finite interval, which completes the proof.

Proof of Lemma 2.3. It is easy to find K_1 ; $K_1 = 1$ is the desired number. To find K_2 , define $\tilde{L}(x) = F_0^{-1}(F(x)) (= \phi^{-1}(x))$. By (2.4) we see that $F_0^{-1}(x) \sim 1/(1-x)^2$ as $x \rightarrow 1-0$. Therefore, $\tilde{L}(x) \sim 1/(1-F(x))^2 \sim L(x)^2$ as $x \rightarrow \infty$, from which it follows that $\tilde{L}(x)$ is also a slowly varying function, and so has the canonical representation $\tilde{L}(x) = c(x) \exp \int_1^x \frac{\varepsilon(t)}{t} dt$, where $c(x) \rightarrow c > 0$ and $\varepsilon(x) \rightarrow 0$ as $x \rightarrow \infty$ (see [8] page 2). Therefore $\phi(x) = \tilde{L}^{-1}(x)$ may be expressed as

$$(3.4) \quad \phi(x) = \exp \int_1^{q(x)x} \frac{\bar{\varepsilon}(t)}{t} dt$$

where $\bar{\varepsilon}(x) \rightarrow \infty$ and $q(x) \rightarrow 1/c$ as $x \rightarrow \infty$. Using (3.4) we easily see that, for every $a > 0$,

$$(3.5) \quad \lim_{n \rightarrow \infty} n^2 \phi(n^2 a) / \phi(2n^2 a) = 0.$$

Now let $a = a(\omega) = m(\varepsilon)^2/3$. Note that $a > 0$, a.s. Since $m_n(\varepsilon) \rightarrow m(\varepsilon)$ a.s., we may and do assume that there exists $n_1 > 0$ such that

$$(3.6) \quad m_n(t)^2 \geq m_n(\varepsilon)^2 \geq 2a, \quad t \geq \varepsilon, \quad n \geq n_1.$$

We divide $\sum_{i \leq nt} \phi(n^2 \eta_{ni})$ into three parts;

$$\begin{aligned} S_{1,n}(t) &= \sum_{i \leq nt} \phi(n^2 \eta_{ni}) I(\eta_{ni} < n^{-2}), \\ S_{2,n}(t) &= \sum_{i \leq nt} \phi(n^2 \eta_{ni}) I(n^{-2} \leq \eta_{ni} \leq a), \\ S_{3,n}(t) &= \sum_{i \leq nt} \phi(n^2 \eta_{ni}) I(\eta_{ni} > a). \end{aligned}$$

Here, $I(\cdot)$ denotes the indicator function. Since ϕ is monotone, we clearly have

$$(3.7) \quad S_{1,n}(t) \leq nT\phi(1), \quad t \leq T.$$

On the other hand, using (3.4), we see that $\lim_{n \rightarrow \infty} n/\phi(2n^2a) = 0$. Therefore, by (3.7) and (3.6) there exists $n_2 > 0$ such that

$$(3.8) \quad S_{1,n}(t) \leq \phi(2n^2a) \leq \phi(n^2 m_n(t)^2), \quad \varepsilon \leq t \leq T, n > n_2.$$

We next consider $S_{2,n}(t)$. If $n^{-2} \leq x \leq a$, then we have

$$(3.9) \quad 0 \leq \phi(n^2 x) \leq \phi(n^2 a) n^2 x.$$

Combining this with (3.5), we see that there exists $n_3 > 0$ such that

$$(3.10) \quad \phi(n^2 x) \leq \phi(2n^2 a) x, \quad n \geq n_3.$$

Therefore, we obtain that for $\varepsilon \leq t \leq T$ and $n \geq n_3$,

$$(3.11) \quad \begin{aligned} 0 \leq S_{2,n}(t) &\leq \phi(2n^2 a) \sum_{i \leq nt} \eta_{ni} \\ &= \phi(2n^2 a) \eta([nt]/n) \leq \phi(n^2 m_n(t)^2) \eta(T). \end{aligned}$$

In the last inequality we used (3.6). Finally let us consider $S_{3,n}(t)$. Notice that there are only finitely many $t \in [0, T]$ such that $\Delta\eta(t) > a$. Thus if we denote by $K_3 = K_3(\omega)$ the number of such t 's, then $\sum_{i \leq nt} I(\eta_{ni} > a) = K_3$ for all sufficiently large n (a.s.). Thus we have

$$(3.12) \quad \begin{aligned} 0 \leq S_{3,n}(t) &\leq K_3 \max_{i \leq nt} \phi(n^2 \eta_{ni}) \\ &= K_3 \phi(n^2 m_n(t)^2), \quad \varepsilon \leq t \leq T, n \geq n_4. \end{aligned}$$

See (2.7) for the last equality. Now combining (3.8), (3.11) and (3.12) we obtain

$$(3.13) \quad \sum_{i \leq nt} \phi(n^2 \eta_{ni}) \leq K_2 \phi(n^2 m_n(t)^2), \quad \varepsilon \leq t \leq T, n \geq n_0,$$

where $K_2 = 1 + \eta(T) + K_3$ and $n_0 = \max(n_2, n_3, n_4)$.

Thus Lemma 2.3 is proved.

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