

ON ONE-SIDED BOUNDEDNESS OF NORMED PARTIAL SUMS

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This paper gives a very general sufficient condition for the existence of constants $B(n)$, $C(n)$ for which either $\limsup_{n \rightarrow +\infty} S_n/B(n) = 1$ almost surely or $\liminf_{n \rightarrow +\infty} S_n/C(n) = -1$ almost surely, where $S_n = X_1 + X_2 + \dots + X_n$ and X_i are independent and identically distributed random variables. The theorem is closely connected with results of Klass and Teicher on the one-sided boundedness of S_n , with the relative stability of S_n , and with a generalised law of the iterated logarithm due to Kesten. For non negative X_i the sufficient condition is shown to be necessary, and the results are partially generalised to the case when X_i form a stationary m -dependent sequence. Some connections with a generalised type of regular variation and with domains of partial attraction are also noted.

1. Results

Let X_i, X be independent and identically distributed random variables with distribution F , and let $S_n = X_1 + X_2 + \dots + X_n$. Suppose $P(|X| \geq x) > 0$ for $x > 0$. The first result of this paper is the following sufficient condition for one-sided boundedness of S_n , normed in an appropriate manner.

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THEOREM 1. *If*

$$\liminf_{x \rightarrow +\infty} xP(|X| \geq x) / \left| \int_{-x}^x u dF(u) \right| = 0$$

there are nondecreasing sequences $B(n), C(n)$, such that either $\limsup_{n \rightarrow +\infty} S_n/B(n) = 1$ almost surely, or $\liminf_{n \rightarrow +\infty} S_n/C(n) = -1$ almost surely, or both.

This theorem is closely connected with the results of Klass [7], [8], and Klass and Teicher [9] on one-sided boundedness of S_n , and it is also connected with the relative stability of S_n . S_n is relatively stable if there is a sequence $b_n \rightarrow +\infty$ for which $S_n/b_n \xrightarrow{P} 1$ or $S_n/b_n \xrightarrow{P} -1$. It was shown by Rogozin [22] and Maller [13] that the relative stability of S_n is equivalent to

$$\lim_{x \rightarrow +\infty} xP(|X| \geq x) / \int_{-x}^x u dF(u) = 0,$$

which gives us the immediate corollary to Theorem 1 that, if S_n is relatively stable, there are nondecreasing sequences $B(n), C(n)$, such that either $\limsup_{n \rightarrow +\infty} S_n/B(n) = 1$ almost surely, or $\liminf_{n \rightarrow +\infty} S_n/C(n) = -1$ almost surely, or both. This also shows the connection of Theorem 1 with the results of Klass and Teicher, since in each of their theorems, S_n is relatively stable. However, their extra restrictions, which include some kind of moment assumption, allow them to conclude something about the form of $B(n)$; in fact, that $B(n)$ may be computed from quantities involving only the distribution F in a simple way. Due to the generality of Theorem 1, we cannot give any such representation for our norming sequence.

Theorem 1 is also closely connected with the result of Kesten [6], that if F is in the domain of partial attraction of the normal distribution, equivalently

$$\liminf_{x \rightarrow +\infty} x^2P(|X| > x) / \int_{-x}^x u^2 dF(u) = 0,$$

there are sequences $\alpha_n, \gamma(n)$, for which

$$-\infty < \liminf(S_n - \alpha_n)/\gamma(n) < \limsup(S_n - \alpha_n)/\gamma(n) < +\infty$$

almost surely. This is a very general form of the law of iterated logarithm, and its converse is known to be true by results of Heyde [5] and Rogozin [20]. It was extended by Maller [16], who showed that $\gamma(n)$ may be chosen so that the almost sure limit points of $(S_n - \alpha_n)/\gamma(n)$ are precisely the interval $[-1, 1]$. The methods of [6] and [16] are of great use in our proof of Theorem 1. We note the similarity of the conditions

$$\liminf_{x \rightarrow +\infty} xP(|X| > x) / \left| \int_{-x}^x u dF(u) \right| = 0$$

and

$$\liminf_{x \rightarrow +\infty} x^2 P(|X| \geq x) / \int_{-x}^x u^2 dF(u) = 0 ;$$

This is a reflection of the duality between the convergence to normality of S_n and its relative stability, as noticed by Raikov (see Gnedenko and Kolmogorov [3, p. 143] and Hall [4]).

We mention that the converse of Theorem 1 is not true, since of course any symmetric distribution in the domain of partial attraction of the normal distribution satisfies $\limsup S_n/B(n) = 1$ almost surely for some

$B(n)$ by Kesten's result; but $\int_{-x}^x u dF(u) \equiv 0$ for such a distribution.

Even if we restrict attention to nonsymmetric distributions outside the domain of partial attraction of the normal distribution, there is no possibility of a converse to Theorem 1, as is shown by the example on page 167 of Klass [7], together with his Theorem 1.2. (The fact that the distribution of Klass's random variable does not belong to the domain of partial attraction of the normal distribution follows from Theorem 1 of [15].)

Since the preparation of this paper a manuscript of a paper to appear by Pruitt [19] has been circulated. Theorem 5.2 of Pruitt's paper gives a result and a converse which includes our result of Theorem 1. The methods

and proofs of the two papers are quite different, and although Pruitt gives much more detail on the almost sure boundedness problem, he does not relate it to the concept of relative stability and its generalisations, nor does he consider any case other than independent and identically distributed random variables (*cf.* our Theorems 2 and 3). Thus our approach to Theorem 1 may be of independent interest.

For positive random variables, we can give a good summary.

THEOREM 2. *Suppose $P(X \geq 0) = 1$. Then the following are equivalent:*

- (i) $\liminf_{x \rightarrow +\infty} xP(X \geq x) / \int_0^x u dF(u) = 0$;
- (ii) $\liminf_{x \rightarrow +\infty} P(X \geq x\lambda) / P(X \geq x) \leq \lambda^{-1}$ for $\lambda \geq 1$;
- (iii) $\liminf_{x \rightarrow +\infty} \int_0^{x\lambda} [1-F(u)] du / \int_0^x [1-F(u)] du = 1$ for $\lambda \geq 1$;
- (iv) there are sequences $n_i \rightarrow +\infty$, $b_{n_i} \rightarrow +\infty$, for which

$$S_{n_i} / b_{n_i} \xrightarrow{P} 1 ;$$
- (v) there is a sequence $B(n) \rightarrow +\infty$ for which

$$\limsup_{n \rightarrow +\infty} S_n / B(n) = 1 \text{ almost surely.}$$

Conditions (i), (ii), (iii), are purely analytical equivalences and are proved by some of the methods of the theory of regular variation. A curious consequence of (ii) holding is that F belongs to some domain of partial attraction ([15], Theorem 2). It is interesting to compare Theorem 2 with Theorem 2 of Rogozin [21], which pertains to the fluctuation properties of a relatively stable S_n . Then it will be seen that there are gaps in Theorem 2 concerned with properties of the characteristic function (or Laplace transform) which it would be useful to fill (in this context see also [14]). The equivalence of (i) and (iv) in Theorem 2 is proved in Theorem 3 of [13]. Note that (ii) holds if $P(X \geq x)$ is regularly varying with index less than or equal to -1 .

As an application of Theorem 2 and Kesten's generalised law of the iterated logarithm it is easy to show that there are sequences $\alpha_n, \gamma(n)$, for which

$$-\infty < \liminf_{n \rightarrow +\infty} (S_n - \alpha_n) / \gamma(n) < \limsup_{n \rightarrow +\infty} (S_n - \alpha_n) / \gamma(n) < +\infty$$

almost surely if and only if there is a sequence $B(n)$ for which

$\limsup_{n \rightarrow +\infty} \sum_{i=1}^n X_i^2 / B(n) = 1$ almost surely, where the X_i are not necessarily positive. This is an example of the "duality" principle of Raikov mentioned earlier.

Let $\nu(x) = \int_0^x u dF(u)$; in the case $X \geq 0$ almost surely, it is easy to deduce from (iv) of Theorem 2 that $\liminf_{x \rightarrow +\infty} \nu(x\lambda) / \nu(x) = 1$ for $\lambda \geq 1$.

However, the converse to this is not true as may be shown by a minor modification of an example in [15], where an F is given for which

$$\liminf_{x \rightarrow +\infty} \int_{-x\lambda}^{x\lambda} u^2 dF(u) / \int_{-x}^x u^2 dF(u) = 1 \text{ for } \lambda \geq 1,$$

but $1 - F(x) + F(-x)$ is slowly varying. (Simply take the same tail for a positive distribution.) It can be shown that

$$\liminf_{x \rightarrow +\infty} \int_0^x u^2 dF(u) / x\nu(x) = 0$$

if and only if

$$\liminf_{x \rightarrow +\infty} \nu(x\lambda) / \nu(x) = 1 \text{ for } \lambda \geq 1,$$

so

$$\liminf_{x \rightarrow +\infty} \int_0^x u^2 dF(u) / x\nu(x) = 0$$

is necessary but not sufficient for (iv) of Theorem 2. (See Theorem 2 of [14].)

A paper of Miller [17] contains sufficient conditions for the

existence of $C(n)$ with $\liminf_{n \rightarrow +\infty} S_n/C(n) = 1$ almost surely. In fact, Miller's $C(n)$ may be taken as $nv(n)$ (he considers only the case $X \geq 0$ and $EX = +\infty$). Miller's distributions satisfy $x[1-F(x)]/\int_0^x u dF(u) \rightarrow 0$, so they are relatively stable, and in particular (i) of Theorem 2 holds. Thus they admit a sequence $B(n)$ for which $\limsup S_n/B(n) = 1$ almost surely, and this proves that $B(n)$ is not asymptotically equivalent to $nv(n)$; if it were, we would have $S_n/nv(n) \rightarrow 1$ almost surely, and, by the result of Chow and Robbins [1], $EX < +\infty$.

Consider the following proof, in essence due to Chow and Robbins [1], that (iv) implies (v) in Theorem 2: assume that the sequence in (iv) increases rapidly enough for $S_{n_i}/b_{n_i} \rightarrow 1$ almost surely, and let $B(j) = b_{n_i}$ whenever $n_{i-1} < j \leq n_i$. Then $S_{n_i}/B(n_i) \rightarrow 1$ almost surely, so $\limsup_{n \rightarrow +\infty} S_n/B(n) \geq 1$ almost surely, while if $j > n_1$ and i is such that

$$n_{i-1} < j \leq n_i, \quad S_j/B(j) \leq \max_{n_{i-1} < j \leq n_i} S_j/B(n_i) \leq S_{n_i}/b_{n_i} \rightarrow 1$$

almost surely, by the monotonicity of S_n when $X \geq 0$ almost surely, showing that $\limsup_{n \rightarrow +\infty} S_n/B(n) \leq 1$ almost surely. Hence $\limsup S_n/B(n) = 1$ almost surely.

This argument does not require the independence (or even the identity of distributions) of the X_i , and it suggests that versions of Theorems 1 and 2 may be true under some form of dependence among the X_i . We say that X_i forms a *stationary sequence* if the distribution of (X_1, X_2, \dots, X_n) is the same as that of $(X_{j+1}, X_{j+2}, \dots, X_{j+n})$ for $n \geq 1$ and $j \geq 1$, and that the X_j are *m-dependent* if there is an integer $m \geq 0$ such that X_i and X_j are independent whenever $|i-j| > m$. With this definition, an independent sequence is 0-dependent. We prove:

THEOREM 3. *Suppose X_i is a stationary m -dependent sequence with marginal distribution F for which*

$$xP(|X| \geq x) / \int_{-x}^x u dF(u) \rightarrow 0 \text{ as } x \rightarrow +\infty .$$

Then there are sequences $b_n \rightarrow +\infty$, $B(n) \rightarrow +\infty$, for which either

$S_n/b_n \xrightarrow{P} 1$ and $\limsup S_n/B(n) = 1$ almost surely, or $S_n/b_n \xrightarrow{P} -1$ and $\liminf S_n/B(n) = -1$ almost surely. The alternatives depend on the

ultimate sign of $\int_{-x}^x u dF(u)$, which is constant, and we have

$$b_n \sim n \left| \int_{-b_n}^{b_n} u dF(u) \right|$$

and b_n is regularly varying with index 1.

Theorem 3 partially generalises the results of [22], [7] and [8] to the m -dependent case.

2. Proofs of theorems

Proof of Theorem 1. We use the notations

$$H(x) = P(|X| \geq x) ,$$

$$v(x) = \int_{-x}^x u dF(u) ,$$

and

$$V(x) = \int_{-x}^x u^2 dF(u) - \left[\int_{-x}^x u dF(u) \right]^2 ,$$

where $x > 0$. When the condition of the theorem holds there is a sequence $x_k \uparrow +\infty$ for which $x_k H(x_k) / v(x_k) \rightarrow 0$, so $|v(x_k)| > 0$ for large k , X not being bounded above. By taking a subsequence, assume that $v(x_k) > 0$; (this will lead to $\limsup S_n/B(n) = 1$, whereas assuming that $v(x_k) < 0$

leads to $\liminf S_n/C(n) = -1$).

We consider two cases. First, assume $x_k^2 H(x_k)/V(x_k) \geq a > 0$, and put $\zeta_k = x_k H(x_k)/\nu(x_k) \rightarrow 0$. By taking a subsequence if necessary, assume $\zeta_k \leq k^{-6}$. We have $V(x_k)/x_k \nu(x_k) \leq a^{-1} \zeta_k$. Define a sequence of integers r_k as the integer part of $\log_2 \left\{ x_k \zeta_k^{-3/4} / \nu(x_k) \right\}$, where \log_2 stands for the logarithm to base 2 . Since $\nu(x_k)/x_k \rightarrow 0$, $r_k \rightarrow +\infty$. Let X_i^k denote X_i truncated at $\pm x_k$, and let $S_n^k = X_1^k + X_2^k + \dots + X_n^k$. By Chebychev's inequality, if $y > 0$,

$$P \left\{ S_{2^{r_k}}^k - 2^{r_k} \nu(x_k) > y \sqrt{2^{r_k} V(x_k)} \right\} \leq y^{-2} ,$$

(1)

$$P \left\{ S_{2^{r_k}}^k - 2^{r_k} \nu(x_k) > -y \sqrt{2^{r_k} V(x_k)} \right\} \geq 1 - y^{-2} .$$

We apply these inequalities with y replaced by

$$y_k = \delta 2^{r_k} \nu(x_k) / \sqrt{2^{r_k} V(x_k)} , \quad \delta > 0 .$$

Note that

$$\begin{aligned} \delta^{-1} y_k &= \left\{ \nu^2(x_k) 2^{r_k} / V(x_k) \right\}^{1/2} \sim \left\{ \zeta_k^{-3/4} x_k \nu(x_k) / V(x_k) \right\}^{1/2} \geq a^{1/2} \zeta_k^{-7/8} \\ &\geq a^{1/2} k^{21/4} \rightarrow +\infty , \end{aligned}$$

so an immediate consequence of (1) is that $S_{2^{r_k}}^k / 2^{r_k} \nu(x_k) \xrightarrow{P} 1$. We have

$$2^{r_k} H(x_k) \sim \zeta_k^{-3/4} x_k H(x_k) / \nu(x_k) = \zeta_k^{1/4} \leq k^{-3/2} ,$$

which implies that the truncation may be disregarded to deduce that

$S_{2^{r_k}} / 2^{r_k \nu}(x_k) \xrightarrow{p} 1$. Define a sequence $B(n) \uparrow +\infty$ by

$$B(n) = 2^{r_k \nu}(x_k) \text{ whenever } 2^{r_{k-1}} < n \leq 2^{r_k};$$

then we have $S_{2^{r_k}} / B(2^{r_k}) \xrightarrow{p} 1$, so $\limsup S_n / B(n) \geq 1$ almost surely.

By a modification of Lévy's inequality (Lemma 1 below), if $\delta > 0$,

$$\begin{aligned} P \left\{ \max_{1 \leq j \leq 2^{r_k}} \left\{ S_{j^{-j\nu}}^k(x_k) \right\} \geq \delta B(2^{r_k}) \right\} &\leq 2P \left\{ S_{2^{r_k}}^k - 2^{r_k \nu}(x_k) \geq \delta B(2^{r_k}) \sqrt{2.2^{r_k \nu}(x_k)} \right\} \\ &\leq 2P \left\{ S_{2^{r_k}}^k - B(2^{r_k}) \geq \frac{1}{2} \delta B(2^{r_k}) \right\} \end{aligned}$$

if k is large enough, since, as we showed above, $B^2(2^{r_k}) / 2^{r_k \nu}(x_k) \rightarrow +\infty$.

An easy consequence of (1) is that

$$\sum P \left\{ S_{2^{r_k}}^k > (1+\delta)B(2^{r_k}) \right\}$$

converges when $\delta > 0$, so we see that

$$\sum P \left\{ \max_{1 \leq j \leq 2^{r_k}} \left\{ S_{j^{-j\nu}}^k(x_k) \right\} \geq \delta B(2^{r_k}) \right\}$$

converges when $\delta > 0$.

By the Borel-Cantelli Lemma, this means

$$\limsup_{k \rightarrow +\infty} \max_{1 \leq j \leq 2^{r_k}} \left\{ S_{j^{-j\nu}}^k(x_k) \right\} / B(2^{r_k}) = 0$$

almost surely, and so

$$\begin{aligned} \max_{1 \leq j \leq 2^{r_k}} S_j^k / B(2^{r_k}) &\leq \max_{1 \leq j \leq 2^{r_k}} \left(S_j^k - j \nu(x_k) \right) / B(2^{r_k}) + \max_{1 \leq j \leq 2^{r_k}} j \nu(x_k) / B(2^{r_k}) \\ &= o(1) + 1 \text{ almost surely as } k \rightarrow +\infty. \end{aligned}$$

Write $S_j = S_j^k + Y_j^k$, and note that

$$\begin{aligned} \sum_k P \left\{ \max_{1 \leq j \leq 2^{r_k}} |Y_j^k| > 0 \right\} &= \sum_k P \left\{ |Y_j^k| > 0 \text{ for some } j \leq 2^{r_k} \right\} \\ &\leq \sum_k P \left\{ |X_i| > x_k \text{ for some } i \leq 2^{r_k} \right\} \\ &\leq \sum_k 2^{r_k} H(x_k) < +\infty \end{aligned}$$

so $\max_{1 \leq j \leq 2^{r_k}} |Y_j^k| \rightarrow 0$ almost surely as $k \rightarrow +\infty$. This means

$$\begin{aligned} \max_{1 \leq j \leq 2^{r_k}} S_j / B(2^{r_k}) &\leq \max_{1 \leq j \leq 2^{r_k}} S_j^k / B(2^{r_k}) + \max_{1 \leq j \leq 2^{r_k}} Y_j^k / B(2^{r_k}) \\ &\leq 1 + o(1) \text{ almost surely.} \end{aligned}$$

If $n \geq 1$ choose $k = k(n)$ so that $2^{r_{k-1}} < n \leq 2^{r_k}$; then

$B(n) = B(2^{r_k})$, and

$$\limsup_n S_n / B(n) \leq \limsup_n \sup_{2^{r_{k-1}} < j \leq 2^{r_k}} S_j / B(2^{r_k}) \leq 1 \text{ almost surely,}$$

and we conclude that $\limsup_n S_n / B(n) = 1$ almost surely.

We come now to the second case, when $x_k^2 H(x_k) / V(x_k) \rightarrow 0$. F is then in the domain of partial attraction of the normal distribution, and the methods we use are a modification of those in [16], which in turn is based on Kesten's work [6]. We put $\zeta_k = x_k^2 H(x_k) / V(x_k)$, and assume that

$\zeta_k \leq k^{-8}$. Let r_k be the integer part of $\log_2 \left\{ \zeta_k^{-3/4} x_k^2 / V(x_k) \right\}$, and again $r_k \rightarrow +\infty$. By the Berry-Esseen theorem (Feller [2, p. 542]), if, $r \geq 1$,

$$(2) \quad \sup_{-\infty < x < +\infty} \left| P \left\{ S_{2^r}^k - 2^r v(x_k) < x \sqrt{2^r V(x_k)} \right\} - \Phi(x) \right| \leq L_r^k$$

where

$$\Phi(x) = (2\pi)^{-1/2} \int_{-\infty}^x e^{-1/2 u^2} du \quad \text{and} \quad L_r^k = E \left| X_{2^r}^k - v(x_k) \right|^3 / 2^{3/2 r} V^{3/2}(x_k).$$

Exactly as in [16] we can show that

$$\sum_{k \geq 1} r_k^{-2} \log_2 \sum_{k < r \leq r_k} L_r^k \quad \text{and} \quad \sum_{k \geq 1} r_{k-1} \sum_{k < r \leq r_k} 2^{r H}(x_k)$$

are finite. Suppose firstly that

$$y_k = 2^{r_k v(x_k)} / \sqrt{2^{r_k} V(x_k)}$$

contains a subsequence $\rightarrow +\infty$. By taking a subsequence throughout the proof we can then assume that $y_k / \log k \rightarrow +\infty$. Since $\sum L_{r_k}^k$ and

$\sum 2^{r_k H}(x_k)$ are finite, we are in the situation of Case 1, with the inequality (2), with $r = r_k$, playing the same role as inequalities (1).

We thus find that $\limsup S_n / B(n) = 1$ for the same $B(n)$ as before.

Alternatively we have y_k bounded, in which case (2) implies

$$\sum_{k \geq 1} r_k^{-2} \log_2 \sum_{k < r \leq r_k} \sup_{-\infty < x < +\infty} \left| P \left\{ X_{2^r}^k \sqrt{2^{r_k} V(x_k)} < x + o(1) \right\} - \Phi(x) \right| < +\infty$$

because, when $r \leq r_k$,

$$2^{r v(x_k)} \sqrt{2^r V(x_k)} = 2^{1/2(r-r_k)} 2^{r_k v(x_k)} \sqrt{2^{r_k} V(x_k)} = o(1).$$

Thus, putting $x = a\sqrt{2 \log k}$, we see that

$$\sum_{k \geq 1} \sum_{r_{k-2 \log_2 k} < r \leq r_k} P\left\{S_{2^r}^k > aB(2^r)\right\}$$

is finite when $a > 1$, where we now define $B(n) \uparrow +\infty$ by

$$B^2(n) = 2 \cdot 2^{r_k} V(x_k) \log k \quad \text{when} \quad 2^{r_{k-1}} < n \leq 2^{r_k}.$$

By Chebychev's inequality

$$\begin{aligned} \sum_{k \geq 1} \sum_{r_{k-1} < r \leq r_k} P\left\{S_{2^r}^k - 2^r v(x_k) > aB(2^r)\right\} \\ \leq a^{-2} \sum_{k \geq 1} \sum_{r_{k-1} < r \leq r_k} B^{-2}(2^r) 2^{2r} V(x_k) < +\infty \end{aligned}$$

as in [16], and hence

$$\sum_{k \geq 1} \sum_{r_{k-1} < r \leq r_k} P\left\{S_{2^r}^k > aB(2^r)\right\} < +\infty \quad \text{when} \quad a > 1,$$

because when $r_{k-1} < r \leq r_k$,

$$2^r v(x_k) / B(2^r) = 2^r v(x_k) / B(2^{r_k}) \leq 2^{r_k} v(x_k) / \sqrt{2 \cdot 2^{r_k} V(x_k) \log k} \rightarrow 0.$$

Thus we have

$$\sum_{k \geq 1} \sum_{r_{k-1} < r \leq r_k} P\left\{S_{2^r}^k > aB(2^r)\right\} < +\infty \quad \text{if} \quad a > 1.$$

The same is also true with X_i replaced by $-X_i$, as a symmetric argument shows, and thus by symmetrising and ignoring the truncation,

$$\sum_{r \geq 1} P\left\{\left|S_{2^r}^S\right| > 2aB(2^r)\right\} < +\infty. \quad \text{By Lemma 2 of [6], this means}$$

$\limsup S_n^S / B(n) \leq 4$ almost surely, so by Theorem 7 of [6], $S_n^S / B(n) \xrightarrow{p} 0$,

and hence, as in Feller [2, p. 232], $(S_n - n v(B_n)) / B(n) \xrightarrow{p} 0$.

If n is large choose $k = k(n)$ so that $2^{r_{k-1}} < n \leq 2^{r_k}$; then

$$\begin{aligned}
 nB_n^{-1} |v(B(n))| &\leq 2^{r_k} |v(B(2^{r_k}))| / B(2^{r_k}) \leq 2^{r_k} v(x_k) / B(2^{r_k}) \\
 &\quad + 2^{r_k} \left| \int_{x_k \leq |u| \leq B(2^{r_k})} u dF(u) \right| / B(2^{r_k}) \\
 &\leq o(1) + 2^{r_k} P(|X| > x_k) = o(1) \text{ as } n \rightarrow +\infty,
 \end{aligned}$$

so we deduce that $S_n/B(n) \xrightarrow{p} 0$. We can now follow exactly the proof of [16] to obtain $\limsup S_n/B(n) = 1$ almost surely, and in fact that the limit points of $S_n/B(n)$ are precisely the interval $[-1, 1]$. This concludes the proof of Theorem 1.

Proof of Theorem 2. The equivalence of (i) and (ii) is a simple consequence of Theorem 2 of [12], while the equivalence of (i) and (iv) was shown in Theorem 3 of [13]. It is easy to see that (iv) implies (iii), while if (i) does not hold,

$$x[1-F(x)] / \int_0^x [1-F(u)] du \geq \delta > 0$$

for x large, so for such x and $\lambda > 1$,

$$\begin{aligned}
 \int_0^{x\lambda} [1-F(u)] du / \int_0^x [1-F(u)] du &= \exp \int_x^{x\lambda} [1-F(u)] du / \int_0^u [1-F(y)] dy \\
 &\geq \exp \delta \left(\int_x^{x\lambda} u^{-1} du \right) = \lambda^\delta,
 \end{aligned}$$

so (iii) does not hold. The fact that (i) implies (v) follows from Theorem 1, so it remains to show only that (v) implies (i). Suppose (v) holds and (i) does not, so (ii) also fails to hold. From Theorem 1 of [15] we then see that F is not in the domain of partial attraction of the normal distribution, equivalently, by Lévy [10, p. 113], $x^2 P(X \geq x) / v(x) \geq a > 0$, where now

$$v(x) = \int_0^x u^2 dF(u) - \left[\int_0^x u dF(u) \right]^2.$$

Since $S_n \geq X_n$, we have $\limsup X_n/B(n) \leq 1$ almost surely, so by the

Borel-Cantelli Lemma, $\sum P(X > 2B(n))$ converges, and hence

$nP(X > 2B(n)) \rightarrow 0$. If $\limsup nB^{-1}(n)v(2B(n)) > 0$, where

$v(x) = \int_0^x u dF(u)$, then (i) holds, so assume $nB^{-1}(n)v(2B(n)) \rightarrow 0$. This

means $n^{-1}B(n) \rightarrow +\infty$. Let X_i^i denote X_i truncated at $2B(i)$, and let

$S_n^n = X_1^n + X_2^n + \dots + X_n^n$. Since $\sum P(X > 2B(n)) < +\infty$, $\{S_n - S_n^n\}/B(n) \rightarrow 0$

almost surely. We also have

$$\sum B^{-2}(n) \text{Var } X_n^n = \sum B^{-2}(n)V(2B(n)) \leq 4a^{-1} \sum P(X > 2B(n)) < +\infty$$

so by, for example, Loève [11, p. 252], $\{S_n^n - ES_n^n\}/B(n) \rightarrow 0$ almost surely.

But

$$ES_n^n = \sum_{i=1}^n EX_i^i = \sum_{i=1}^n \int_0^{2B(i)} u dF(u) \leq nv(2B(n)) = o(B(n))$$

so $S_n^n/B(n) \rightarrow 0$ almost surely and $S_n/B(n) \rightarrow 0$ almost surely. This

contradiction completes the proof of Theorem 2.

Proof of Theorem 3. Let $A(x) = \int_0^x [1-F(u)-F(-u)]du$. Since

$xP(|X| > x)/v(x) \rightarrow 0$, $xP(|X| > x)/A(x) \rightarrow 0$, and since A is continuous,

A is either positive or negative for large x , as in [13]. Suppose the

former; then, as in [13], A is slowly varying. Define a nondecreasing

sequence $b_n \rightarrow +\infty$ by $b_n = \sup\{x > 0 : x^{-1}A(x) \geq n^{-1}\}$.

As in [21], we can show that $nP(|X| > xb_n) \rightarrow 0$ and $nb_n^{-2}v(xb_n) \rightarrow 0$

for $x > 0$, that the sum of n independent copies of X is relatively

stable, and also from [21], that b_n is regularly varying with index 1.

Let n_m be the integer part of $n/(m+1)$, and let

$$S_{n_m}(t) = \sum_{i=1}^{n_m-1} X_{(m+1)i+t}, \text{ for } 0 \leq t \leq m.$$

Now $S_{n_m}(t)$ is the sum of $n_m - 1$ independent and identically distributed random variables and hence is relatively stable with norming sequence

b_{n_m-1} ; clearly then, $S_{n_m}(t)/b_{n_m} \xrightarrow{P} 1$. Since b_n is regularly varying

with index 1 , $S_{n_m}(t)/b_n \xrightarrow{P} 1/(m+1)$, which means

$$S_n/b_n = \sum_{t=0}^m S_{n_m}(t)/b_n + o(1) = 1 + o(1)$$

in probability. Clearly also $b_n \sim nv(b_n)$.

The remainder of Theorem 3 we prove as follows. Since $S_n/b_n \xrightarrow{P} 1$ and $nP(|X| > \epsilon b_n) \rightarrow 0$ for $\epsilon > 0$, we can choose a subsequence $n_i \rightarrow \infty$

for which $\sum P(|S_{n_i} - b_{n_i}| > \epsilon b_{n_i})$ and $\sum n_i P(|X| > \epsilon b_{n_i})$ converge for

every $\epsilon > 0$. Define $B(n)$ by $B(n) = b_{n_i}$ if $n_{i-1} < n \leq n_i$, so we

immediately have $\limsup S_n/B(n) \geq 1$ almost surely. Now by a version of

Lévy's inequality (Lemma 2 below), since $(S_n - b_n)/b_n \xrightarrow{P} 0$,

$$P\left\{ \max_{n_{i-1} < j \leq n_i} (S_j - b_j) > \epsilon B(n_j) \right\} \leq P\{S_{n_i} - b_{n_i} > \epsilon b_{n_i}\} + mn_i P\{|X| > \epsilon m^{-1} b_{n_i}\}.$$

The convergence of the series with terms on the right hand side implies the convergence of the series with terms on the left hand side, so by the Borel-Cantelli Lemma,

$$\limsup_i \max_{n_{i-1} < j \leq n_i} (S_j - b_j)/B(n_i) \leq 0 \text{ almost surely.}$$

Given $n > n_1$, choose $i = i(n)$ so that $n_{i-1} < n \leq n_i$. Then since

$$B(n) = B(n_i) = b_{n_i} \geq b_n,$$

$$\begin{aligned} \limsup_n S_n/B(n) &\leq \limsup_n (S_n - b_n)/B(n_i) + \limsup_n b_n/B(n_i) \\ &\leq \limsup_n \max_{n_{i-1} < j \leq n_i} (S_j - b_j)/B(n_i) + 1 \leq 1 \text{ almost surely,} \end{aligned}$$

showing that $\limsup S_n/B(n) = 1$ almost surely.

3. Two lemmas

LEMMA 1. For each k suppose X_i^k are independent random variables with finite variance $\text{Var } X_i^k$. Then if $m(k)$ is any sequence of integers and $S_j^k = X_1^k + X_2^k + \dots + X_j^k$, for every real x ,

$$P\left\{ \max_{1 \leq j \leq m(k)} \left(S_j^k - \sum_{i=1}^k EX_i^k \right) \geq x \right\} \leq 2P\left\{ S_{m(k)}^k - \sum_{i=1}^{m(k)} EX_i^k \geq x - \left(2 \sum_{i=1}^{m(k)} \text{Var } X_i^k \right)^{1/2} \right\}.$$

Proof of Lemma 1. We omit this proof since it is similar to that of the ordinary Lévy inequality (cf. also Petrov [18]).

LEMMA 2. Let X_i be a stationary m -dependent sequence for which $(S_n - \alpha_n)/B(n) \xrightarrow{P} 0$ for constants $\alpha_n, B(n)$, $B(n) > 0$. Then for every $\epsilon > 0$, $\epsilon < 1/6$, there are constants $n_0(\epsilon), k_0(\epsilon)$, $n_0 > k_0$, for which $n \geq n_0$ implies for every real x ,

$$(i) \quad \max_{k_0 \leq k \leq n} |\alpha_n - \alpha_k - \alpha_{n-k}| \leq \epsilon B(n),$$

$$(ii) \quad (1-\epsilon)P\left\{ \max_{k_0 \leq k \leq n} (S_k - \alpha_k) \geq xB(n) \right\} \leq P\left\{ S_n - \alpha_n \geq (x-3\epsilon)B(n) \right\} + mnP\left\{ |X_i| > \epsilon m^{-1}B(n) \right\}.$$

Proof. (i) is given in Lemma 2 of [16], and it means that

$$\inf_{k_0 \leq k \leq n} P\{S_{n+m} - \alpha_n + \alpha_k - S_{k+m} > -\epsilon B(n)\} \geq \inf_{k_0 \leq k \leq n} P\{S_{n-k} - \alpha_{n-k} > -\frac{1}{2}\epsilon B(n)\} \geq 1 - \epsilon$$

if $n \geq n_0$ and n_0 is large enough, because $(S_n - \alpha_n)/B(n) \xrightarrow{P} 0$. By m -dependence and stationarity,

$$\begin{aligned}
 & (1-\varepsilon)P\left\{\max_{k_0 \leq k \leq n} (S_k - \alpha_k) > xB(n)\right\} \\
 &= (1-\varepsilon) \sum_{k=k_0}^n P\{S_k - \alpha_k > xB(n), \max_{k_0 \leq j < k} (S_j - \alpha_j) \leq xB(n)\} \\
 &\leq \sum_{k=k_0}^n P\{S_k - \alpha_k > xB(n), \max_{k_0 \leq j < k} (S_j - \alpha_j) \leq xB(n), S_{n+m} - \alpha_n + \alpha_k - S_{k+m} > -\varepsilon B(n)\} \\
 &\leq \sum_{k=k_0}^n P\{S_k - \alpha_k > xB(n), \max_{k_0 \leq j < k} (S_j - \alpha_j) \leq xB(n), \\
 &\hspace{20em} S_{n+m} - \alpha_n + S_k - S_{k+m} > (x-\varepsilon)B(n)\} \\
 &\leq P\{S_{n+m} - \alpha_n > (x-2\varepsilon)B(n)\} + nP\{S_m \leq -\varepsilon B(n)\} \\
 &\leq P\{S_n - \alpha_n > (x-3\varepsilon)B(n)\} + nP\{|S_m| \geq \varepsilon B(n)\} \\
 &\leq P\{S_n - \alpha_n > (x-3\varepsilon)B(n)\} + nmP\{|X_i| \geq \varepsilon m^{-1}B(n)\}.
 \end{aligned}$$

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