

A TAUBERIAN THEOREM FOR RANDOM WALK

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ABSTRACT

Let X_1, X_2, \dots be independent random variables, all with the same distribution symmetric about 0;

$$S_n = \sum_{i=1}^n X_i.$$

It is shown that if for some fixed interval I , constant $1 < \alpha \leq 2$ and slowly varying function M one has

$$\sum_{k=1}^n P\{S_k \in I\} \sim \frac{n^{1-1/\alpha}}{M(n)} \quad (n \rightarrow \infty),$$

then the X_i belong to the domain of attraction of a symmetric stable law.

1. Introduction. Let Y_1, Y_2, \dots be a Markov chain and $N_n(A)$ = number of visits to A by Y_k up till time n . A well known result of Darling and Kac ([2], especially §6) states that (under very mild conditions) $N_n(A)$ tends to infinity and has a nondegenerate limit distribution after proper normalization, only if⁽²⁾(³)

$$(1.1) \quad \lim_{n \rightarrow \infty} \frac{\sum_{k=1}^n P_x\{Y_k \in A\}}{n^{1-1/\alpha}\{M(n)\}^{-1}} = 1$$

for some fixed $1 \leq \alpha < \infty$ and slowly varying function M for which $n^{1-1/\alpha}[M(n)]^{-1} \rightarrow \infty$ ($n \rightarrow \infty$). If (1.1) holds uniformly for $x \in A$, then $n^{-1+1/\alpha}M(n)N_n(A)$ has a Mittag-Leffler distribution as limit distribution. If $Y_k, k \geq 1$, is a random walk, i.e. if $Y_k = S_k = \sum_{i=1}^k X_i$ for independent, identically distributed random variables X_i , and if A is a bounded interval, then (1.1) reduces to

Received March 10, 1968.

(1) Research supported by the National Science Foundation under grant GP 7128.

(2) $P_x[E]$ denotes the conditional probability of the event E given $Y_1 = x$.

(3) The condition in Theorem 5 of [2] is stated in terms of the generating function

$$\sum_{k=1}^{\infty} z^k P_x\{Y_k \in A\}.$$

However, by means of Karamata's Tauberian theorem this condition is easily translated into (1.1).

$$(1.2) \quad \lim_{n \rightarrow \infty} \frac{\sum_{k=1}^n P\{S_k \in A\}}{n^{1-1/\alpha} \{M(n)\}^{-1}} = 1.$$

The uniformity of the limit in (1.1) as x varies over a compact set is automatic for random walks by Corollary 1 in [8]; moreover by the estimates in §48 of [7] (see also [6]) α can take only values in the closed interval $[1, 2]$ in the case of random walks. Professor Spitzer raised the question whether (1.2) implies that X_i belongs to the domain of attraction of a stable distribution. The purpose of this note is to prove the theorem below, which answers the question affirmatively for $1 < \alpha \leq 2$ and symmetric X_i .

THEOREM. *Let X_1, X_2, \dots be independent random variables, all with the same distribution function $F(\cdot)$, symmetric about the origin and let $S_n = \sum_{i=1}^n X_i$. If for some fixed interval I (⁴), $1 < \alpha \leq 2$ and slowly varying function M*

$$(1.3) \quad \sum_{k=1}^n P\{S_k \in I\} \sim \frac{n^{1-1/\alpha}}{M(n)} \quad (n \rightarrow \infty),$$

then

$$(1.4) \quad \lim_{n \rightarrow \infty} P \left\{ \frac{CS_n}{n^{1/\alpha}M(n)} \leq x \right\} = F_\alpha(x)$$

where F_α is the symmetric stable distribution function with characteristic function $\exp -|t|^\alpha$ and C is a constant depending only on I and the support of F . If F is not a lattice distribution then(⁵)

$$C = \frac{\pi(\alpha - 1)}{\Gamma\left(\frac{1}{\alpha}\right) |I|}.$$

REMARK. The converse implication, i.e. from (1.4) to (1.3) is a special case of Stone's local limit theorem ([9], Theorem 1). Thus (1.3) and (1.4) are actually equivalent. The local limit theorem does not require symmetry assumptions and allows $\alpha = 1$ as well. It seems likely that the present theorem will also hold in this greater generality. However, our proof makes essential use of the symmetry and of $\alpha > 1$ and therefore offers little hope for generalization.

2. Proof of the theorem. We shall restrict ourselves to the case where F is not a lattice distribution. For a lattice distribution the proof is almost the same and actually simpler because Lemma 1(c) is not needed. We may also exclude the

(⁴) More generally, by Corollary 1 of [8], we could replace I by any bounded Borel set whose boundary has zero Lebesgue measure.

(⁵) $|A|$ denotes the Lebesgue measure of A .

case where $\sigma^2 = \int x^2 dF(x) < \infty$ for this case is covered by the central limit theorem. C_1, C_2, \dots will denote constants (which may depend on F, α, M and I though).

First we show that F may be assumed quite smooth.

LEMMA 1. *If F is not a lattice distribution and $\sigma^2 = \infty$ and (1.3) holds then*

(a) *For any fixed interval J*

$$(2.1) \quad P\{S_n \in J\} \sim \left(1 - \frac{1}{\alpha}\right) \frac{|J|}{|I|} \frac{1}{n^{1/\alpha} M(n)}.$$

(b) *For all sufficiently large n*

$$(2.2) \quad \sup_{\substack{h \geq |I| \\ -\infty < u < +\infty}} \frac{1}{h} P\{S_n \in [u, u+h]\} \leq \frac{12}{|I|} \left(1 - \frac{1}{\alpha}\right) \frac{1}{n^{1/\alpha} M(n)}.$$

(c) *For every $\varepsilon > 0$ and $B > 0$ there exists an $N = N(\varepsilon, B)$ such that for all $n \geq N$*

$$(2.3) \quad \sup_{|x| \leq B\sqrt{n}} \left| \frac{P\{S_n \in I\}}{P\{S_n \in x + I\}} - 1 \right| \leq \varepsilon.$$

Proof. Since F is a symmetric non-lattice distribution, the smallest closed subgroup containing the support of F is the whole group of reals. By Proposition 2 and Corollary 1 of [8]

$$(2.4) \quad \lim_{n \rightarrow \infty} \frac{P\{S_{n+k} \in J\}}{P\{S_n \in I\}} = \frac{|J|}{|I|}$$

for each fixed k . In particular

$$(2.5) \quad \begin{aligned} \int_0^h ds P\{S_{2\lfloor n/2 \rfloor} \in [-s, +s]\} &\sim \frac{P\{S_n \in I\}}{|I|} \int_0^h 2s ds \\ &= \frac{h^2}{|I|} P\{S_n \in I\}. \end{aligned}$$

On the other hand, if

$$\psi(t) = \int e^{itx} dF(x)$$

is the characteristic function of F , then (see [1], formula 10.3.3)

$$(2.6) \quad \int_0^h ds P\{S_m \in [u-s, u+s]\} = \frac{1}{\pi} \int_{-\infty}^{+\infty} \frac{1 - \cos ht}{t^2} e^{-iut} \psi^m(t) dt,$$

so that we conclude

$$(2.7) \quad P\{S_n \in I\} \sim \frac{|I|}{\pi h^2} \int_{-\infty}^{+\infty} \frac{1 - \cos ht}{t^2} \psi^{2[n/2]}(t) dt.$$

Since $0 \leq \psi^2(t) \leq 1$ (F is symmetric) the right hand side of (2.7) is decreasing in n . Thus $P\{S_n \in I\}$ is ‘‘approximately decreasing’’ and it is an easy consequence of this fact and (1.3) (see [3], proof of Hilfssatz 3 in Chapter 16.1 or [4], Theorem XIII.5.5) that

$$P\{S_n \in I\} \sim \left(1 - \frac{1}{\alpha}\right) \frac{1}{n^{1/\alpha} M(n)}.$$

(2.1) now follows from Corollary 1 in [8].

As for part (b) it suffices to show

$$\sup_u P \left\{ S_n \in \left[u - \frac{1}{2}|I|, u + \frac{1}{2}|I| \right] \right\} \leq 6 \left(1 - \frac{1}{\alpha}\right) \frac{1}{n^{1/\alpha} M(n)}.$$

since each interval of length $h \geq |I|$ can be written as the union of at most $2h/|I|$ intervals of length $|I|$. But by Theorem 1 of [8] there exists a $\delta > 0$ such that for all $n \geq n_0$ and all u

$$\begin{aligned} P\left\{S_n \in \left[u - \frac{1}{2}|I|, u + \frac{1}{2}|I|\right]\right\} &\leq \frac{3}{4} P \{S_{2[n/2]} \in (u - |I|, u + |I|)\} + e^{-\delta n} \\ &\leq \frac{3}{4|I|} \int_0^{2|I|} ds P\{S_{2[n/2]} \in [u - s, u + s]\} + e^{-\delta n} \\ (2.8) \quad &= \frac{3}{4\pi|I|} \int_{-\infty}^{+\infty} \frac{1 - \cos 2|I|t}{t^2} e^{-iut} \psi^{2[n/2]}(t) dt + e^{-\delta n} \\ &\leq \frac{3}{4\pi|I|} \int_{-\infty}^{+\infty} \frac{1 - \cos 2|I|t}{t^2} \psi^{2[n/2]}(t) dt + e^{-\delta n} \\ &\leq 6 \left(1 - \frac{1}{\alpha}\right) \frac{1}{n^{1/\alpha} M(n)} \qquad \text{(See (2.1) and (2.7)).} \end{aligned}$$

This proves (b).

To prove (c) we observe that one can decompose F as

$$(2.9) \quad F = a G_1 + (1 - a) G_2$$

for some symmetric distribution functions G_1, G_2 such that

$$(2.10) \quad \frac{1}{4} \leq a \leq \frac{3}{4}$$

and such that the support of $G_1(dx)$ is bounded. One can clearly find such functions by taking a $G_1(dx) = \alpha(x) F(dx)$ for some $0 \leq \alpha(x) \leq 1$, α symmetric and zero outside a compact interval and such that

$$\frac{1}{4} \leq \int \alpha(x) dF(x) \leq \frac{3}{4}.$$

$G_1(dx)$ is then obtained by normalizing $\alpha(x) F(dx)$ and $(1 - a)G_2(dx) = (1 - \alpha(x)) F(dx)$. It is clear from this construction and the fact that $\sigma^2 = \infty$ that we can make

$$\sigma_1^2 = \int_{-\infty}^{+\infty} x^2 dG_1(x)$$

as large as desired. From (2.9) we have for $n = 2m$ or $n = 2m + 1$ ⁽⁶⁾

$$(2.11) \quad F^{(n)} = \sum_{k=0}^m \binom{m}{k} a^k (1 - a)^{m-k} G_1^{(k)} * G_2^{(m-k)} * F^{(n-m)}.$$

We shall use the abbreviation H_k for the distribution function $G_2^{(m-k)} * F^{(n-m)}$ (suppressing the dependence on n, m). We shall use $G(A)$ for the measure assigned to A by a distribution function G . Then

$$(2.12) \quad F^{(n)}(x + I) = \sum_{k=0}^m \binom{m}{k} a^k (1 - a)^{m-k} \int H_k(dy) G_1^{(k)}(x - y + I).$$

Because of (2.10), there exists a $b > 0$ such that

$$(2.13) \quad \sum_{k \leq am/2} \binom{m}{k} a^k (1 - a)^{m-k} \leq e^{-bm}, \quad m \geq m_0.$$

Also, by Esseen's form of the central limit theorem ([5], Theorem 42.2) or by Stone's local limit theorem ([9], Theorem 1)

$$(2.14) \quad \left| G_1^{(k)}(x - y + I) - \frac{1}{\sqrt{2\pi k} \sigma_1} \int_I e^{-(x-y+z)^2/2k\sigma_1^2} dz \right| = o\left(\frac{1}{\sqrt{k}}\right)$$

uniformly in x, y . Now take C_0 such that

$$(2.15) \quad \frac{2}{\sqrt{2\pi}} \int_{C_0/2}^{\infty} e^{-u^2/2} du \leq \frac{\varepsilon}{6 \cdot 24}.$$

By virtue of (2.14) we can then find $m_1 = m_1(B, \varepsilon)$ such that for $k \geq (a/2)m$, $m \geq m_1$, $|x| \leq B\sqrt{2m + 1}$, $|y| \leq C_0\sigma_1\sqrt{m}$

⁽⁶⁾ $F^{(r)}$ is the r -fold convolution of F .

$$\begin{aligned}
 & \left| G_1^{(k)}(x - y + I) - G_1^{(k)}(-y + I) \right| \leq \frac{\varepsilon |I|}{12\sqrt{2\pi k}\sigma_1} e^{-2C_0^2\sigma_1^2 m/2k\sigma_1^2} \\
 & \quad + \frac{1}{\sqrt{2\pi k}\sigma_1} \int_I \left| e^{-(x-y+z)^2/2k\sigma_1^2} - e^{-(y+z)^2/2k\sigma_1^2} \right| dz \\
 (2.16) \quad & \leq \frac{1}{\sqrt{2\pi k}\sigma_1} \int_I e^{-(-y+z)^2/2k\sigma_1^2} dz \left[\frac{\varepsilon}{12} + \left| \exp \left\{ \frac{4B^2m + 4B\sqrt{2m} \cdot C_0\sigma_1\sqrt{m}}{2k\sigma_1^2} \right\} - 1 \right| \right] \\
 & \leq 2G_1^{(k)}(-y + I) \left[\frac{\varepsilon}{12} + \left| \exp \left\{ \frac{4B^2m + 8BC_0\sigma_1 m}{2k\sigma_1^2} \right\} - 1 \right| \right].
 \end{aligned}$$

We already pointed out that σ_1 can be taken arbitrarily large; in particular we may assume that it is so large that the factor in square brackets in the last member of (2.16) does not exceed $\varepsilon/6$ for $k \geq am/2$. Note that the lower bound for σ_1 required here depends only on B, C_0 and a . Once σ_1 has been chosen in this way we have under the conditions for (2.16)

$$\begin{aligned}
 (2.17) \quad & \left| \int_{|y| \leq C_0\sigma_1\sqrt{m}} H_k(dy) G_1^{(k)}(x - y + I) - \int_{|y| \leq C_0\sigma_1\sqrt{m}} H_k(dy) G_1^{(k)}(-y + I) \right| \\
 & \leq \frac{\varepsilon}{3} \int H_k(dy) G_1^{(k)}(-y + I).
 \end{aligned}$$

To estimate the analogous integrals over $|y| > C_0\sigma_1\sqrt{m}$ we use the following inequality which is almost immediate from the definition of $H_k, n - m \geq m$ and (2.2) (see also [7], p. 90);

$$\begin{aligned}
 \sup_{x,z} H_k(x - z + I) & \leq \sup_u F^{(m)}([u, u + |I|]) \\
 & \leq 12 \left(1 - \frac{1}{\alpha} \right) \frac{1}{m^{1/\alpha} M(m)}, \quad m \geq m_2.
 \end{aligned}$$

This inequality implies for all $|u| \leq B\sqrt{2m+1}, m \geq m_2,$

$$\begin{aligned}
 & \int_{|y| \geq C_0\sigma_1\sqrt{m}} H_k(dy) G_1^{(k)}(u - y + I) \\
 (2.18) \quad & = \iint_{\substack{z+y \in u+I \\ |y| \geq C_0\sigma_1\sqrt{m}}} H_k(dy) G_1^{(k)}(dz) \leq \int_{|z| \geq (C_0\sigma_1 - 2B)\sqrt{m}} G_1^{(k)}(dz) H_k(u - z + I) \\
 & \leq 12 \left(1 - \frac{1}{\alpha} \right) \frac{1}{m^{1/\alpha} M(m)} \int_{|z| \geq (C_0\sigma_1 - 2B)\sqrt{m}} G_1^{(k)}(dx).
 \end{aligned}$$

Without loss of generality we may assume σ_1 so large that $2B \leq \frac{1}{2}C_0\sigma_1$ and then for $m \geq m_3, am/2 \leq k \leq m$

$$(21.9) \quad \int_{|z| \geq (C_0\sigma_1 - 2B)\sqrt{m}} G_1^{(k)}(dz) \leq \int_{|z| \geq C_0\sigma_1\sqrt{k}/2} G_1^{(k)}(dz) \leq \frac{2}{\sqrt{2\pi}} \int_{|u| \geq C_0/2} e^{-u^2/2} du \leq \frac{\varepsilon}{6.24}$$

(for the last two steps we used the central limit theorem and (2.15)). We now combine (2.12), (2.13), (2.17)–(2.19) to obtain for $n = 2m$ or $2m + 1, |x| \leq B\sqrt{2m + 1}, m \geq \max(m_0, m_1, m_2, m_3)$

$$(2.20) \quad \begin{aligned} & |F^{(n)}(x + I) - F^{(n)}(I)| \\ & \leq 2e^{-bm} + \sum_{am/2 \leq k \leq m} \binom{m}{k} a^k(1 - a)^{m-k} \left[\frac{\varepsilon}{3} \int H_k(dy) G_1^{(k)}(-y + I) \right. \\ & \quad \left. + \frac{\varepsilon}{6} \left(1 - \frac{1}{\alpha} \right) \frac{1}{m^{1/\alpha} M(m)} \right] \\ & \leq 2e^{-bm} + \frac{\varepsilon}{3} F^{(n)}(I) + \frac{\varepsilon}{3} \left(1 - \frac{1}{\alpha} \right) \frac{1}{n^{1/\alpha} M(n)}. \end{aligned}$$

In view of (2.1) this implies (2.3) for n large enough (recall $F^{(n)}(I) = P\{S_n \in I\}$) and the proof of Lemma 1 is therefore complete.

Practically the only reason for proving Lemma 1 is that it allows us to replace F by $F * \Phi$ where Φ is the standard normal distribution with density $1/\sqrt{2\pi} e^{-x^2/2}$. Indeed, let Y_1, Y_2, \dots be independent normal variables, each with distribution Φ and assume that the $\{Y_i\}_{i \geq 1}$ are independent of the $\{X_i\}_{i \geq 1}$. Then, by Lemma 1, there exists an $N_1(\varepsilon, B)$ such that for $n \geq N_1(\varepsilon, B)$

$$(2.21) \quad \begin{aligned} & \left| P\{S_n + \sum_{i=1}^n Y_i \in I\} - P\{S_n \in I\} \right| \\ & \leq \int \Phi^{(n)}(dy) |P\{S_n + y \in I\} - P\{S_n \in I\}| \\ & \leq \int_{|y| \leq B\sqrt{n}} \Phi^{(n)}(dy) \varepsilon P\{S_n \in I\} + \int_{|y| > B\sqrt{n}} \Phi^{(n)}(dy) 24 \left(1 - \frac{1}{\alpha} \right) \frac{1}{n^{1/\alpha} M(n)} \\ & \leq \left(1 - \frac{1}{\alpha} \right) \frac{1}{n^{1/\alpha} M(n)} \left(2\varepsilon + \frac{24}{\sqrt{2\pi}} \int_{|u| \geq B} e^{-u^2/2} du \right). \end{aligned}$$

Since B can be taken arbitrarily large and ε arbitrarily small we see from (2.1) and (2.21) that

$$P \left\{ S_n + \sum_{i=1}^n Y_i \in I \right\} \sim \left(1 - \frac{1}{\alpha} \right) \frac{1}{n^{1/\alpha} M(n)} \quad (n \rightarrow \infty)$$

as well. In other words the random variables $X_i + Y_i$ satisfy the hypothesis of the theorem. In addition

$$\lim_{n \rightarrow \infty} P \left\{ \frac{C \left(S_n + \sum_{i=1}^n Y_i \right)}{n^{1/\alpha} M(n)} \leq x \right\} = F_\alpha(x)$$

is equivalent to (1.4) because

$$\frac{\sum_{i=1}^n Y_i}{n^{1/\alpha} M(n)} \rightarrow 0 \quad \text{in probability}$$

(this is even true for $\alpha = 2$, for then $\sigma^2 = \infty$ implies that $M(n)$ must tend to ∞ if (1.3) is to hold).⁽⁷⁾ We therefore see that it suffices to prove the theorem for $X_i + Y_i$ instead of X_i . Rather than carry the Y_i along we change notation and write just X_i for $X_i + Y_i$ and F again for the distribution function of the new F . In the sequel we therefore have

$$(2.22) \quad |\psi(t)| = \left| \int e^{itx} dF(x) \right| \leq \left| \int e^{itx} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \right| = e^{-t^2/2}.$$

By the standard inversion formula, [5] Theorem 12.1, (2.1) now gives for any $\varepsilon > 0$

$$\begin{aligned} P \left\{ S_n \in \left[-\frac{1}{2}, +\frac{1}{2} \right] \right\} &= \frac{1}{2\pi} \int_{-\infty}^{+\infty} \frac{2 \sin \frac{t}{2}}{t} \psi^n(t) dt \\ &= \frac{1}{2\pi} \int_{-\varepsilon}^{+\varepsilon} \frac{2 \sin \frac{t}{2}}{t} \psi^n(t) dt + O \left(\int_{\varepsilon}^{\infty} e^{-nt^2/2} dt \right). \end{aligned}$$

Since this holds for each ε and $\psi(t) \geq 0$ for $|t|$ sufficiently small, we can translate the basic hypothesis of our theorem into

$$(2.23) \quad \frac{1}{2\pi} \int_{-\varepsilon}^{+\varepsilon} \psi^n(t) dt \sim \frac{1}{2\pi} \int_{-\infty}^{+\infty} \psi^n(t) dt \sim \left(1 - \frac{1}{\alpha} \right) \frac{1}{|I|} \frac{1}{n^{1/\alpha} M(n)} \quad (n \rightarrow \infty, \varepsilon > 0).$$

Next we show (and this is the crux of the proof) that

⁽⁷⁾ See for instance Theorem 4.1 in C. G. Esseen, On the concentration-function of a sum of independent random variables, *Z. Wahrscheinlichkeitstheorie verw. Gebiete* 9 (1968) 290-308.

$$A_n = \frac{1}{\left(1 - \frac{1}{\alpha}\right)} |I| n^{1/\alpha} M(n)$$

is of the right order to normalize S_n . For this purpose we define

$$d_n(q) = \inf \{L: P\{S_n \in [-L, +L]\} \geq q\}, \quad 0 \leq q < 1.$$

It follows directly from Theorem 3 in [6] and Lemma 1 above that for any $0 < q_1 \leq q_2 < 1$ there exists a $C_1(q_1, q_2)$ such that

$$(2.24) \quad \limsup_{n \rightarrow \infty} \frac{d_n(q_2)}{d_n(q_1)} \leq C_1(q_1, q_2) < \infty.$$

(We should point out that $d_n(q)$ is not exactly the same as the dispersion function $D(S_n; q)$ used in [6]. But clearly $D(S_n; q) \leq 2d_n(q)$ whereas for $q > \frac{1}{2}$, $d_n(q) \leq D(S_n; q)$ since any interval containing S_n with probability $q > \frac{1}{2}$ must contain the origin for symmetrically distributed S_n .) Now it is clear from (2.2) that for large n

$$(2.25) \quad d_n(q) \geq \frac{q}{24} A_n,$$

so that only an upper bound for d_n is needed.

LEMMA 2. For each $q < 1$ there exists a $C_2(q)$ such that

$$(2.26) \quad d_n(q) \leq C_2(q) A_n.$$

Proof. Since

$$\begin{aligned} P\{S_n \in dx\} \int_{-\epsilon}^{+\epsilon} \psi(t) \frac{1 - \cos xt}{1 - \psi(t)} dt &= \int_{-\epsilon}^{+\epsilon} \psi(t) \frac{1 - \psi^n(t)}{1 - \psi(t)} dt \\ &= \sum_{k=1}^n \int_{-\epsilon}^{+\epsilon} \psi^k(t) dt \sim \frac{2\pi}{|I|} n^{1-1/\alpha} \{M(n)\}^{-1} \end{aligned}$$

we have for sufficiently large n (from the positivity of the integrand)

$$P\left\{\int_{-\epsilon}^{+\epsilon} \psi(t) \frac{1 - \cos t S_n}{1 - \psi(t)} dt \geq \frac{10\pi}{|I|} n^{1-1/\alpha} (M(n))^{-1}\right\} \leq \frac{1}{4}.$$

In particular (because $P\{d_n(\frac{1}{4}) \leq S_n \leq d_n(\frac{3}{4})\} \geq \frac{1}{2}$ and $\psi(t) \geq \frac{1}{2}$ for $|t| \leq \pi\{d_n(\frac{1}{4})\}^{-1}$ eventually), for $n \geq n_1$ we can find an

$$(2.27) \quad x_n \in \left[d_n\left(\frac{1}{4}\right), d_n\left(\frac{3}{4}\right) \right]$$

for which

$$(2.28) \quad \frac{1}{\pi^2} \int_{|t| \leq \pi/x_n} \frac{t^2 x_n^2}{1 - \psi(t)} dt \leq \int_{-\varepsilon}^{+\varepsilon} \psi(t) \frac{1 - \cos t x_n}{1 - \psi(t)} dt.$$

$$\leq \frac{10\pi}{|I|} n^{1-1/\alpha} \{M(n)\}^{-1}.$$

To compare n with x_n we now define the function $K(\cdot)$ by

$$y^\alpha K(y) = \inf\{k: k^{1/\alpha} M(k) \geq y\} \quad y \geq 0.$$

From Karamata's well known representation for slowly varying functions, [4], Corollary to Theorem VIII.9.1, one easily sees that for suitable $n_2 = n_2(\varepsilon)$

$$(2.29) \quad \inf_{m \geq (1+\varepsilon)n \geq n_2} \frac{m^{1/\alpha} M(m)}{n^{1/\alpha} M(n)} \geq \left(1 + \frac{\varepsilon}{2}\right)^{1/\alpha}.$$

From this property it follows immediately that K is also a slowly varying function and that

$$(2.30) \quad K(n^{1/\alpha} M(n)) \sim M^{-\alpha}(n) \quad (n \rightarrow \infty)$$

as well as

$$(2.31) \quad M(y^\alpha K(y)) \sim K^{-1/\alpha}(y) \quad (y \rightarrow \infty).$$

We now observe that by (2.27) and (2.25)

$$x_n \geq d_n \left(\frac{1}{4}\right) \geq \frac{|I|}{96} \frac{1}{\left(1 - \frac{1}{\alpha}\right)} n^{1/\alpha} M(n).$$

By the definition of K and (2.29) this implies for $n \geq n_3$

$$(2.32) \quad n \leq \left\{ \frac{100}{|I|} x_n \left(1 - \frac{1}{\alpha}\right) \right\}^\alpha K \left(\frac{100}{|I|} x_n \left(1 - \frac{1}{\alpha}\right) \right).$$

In view of the slowly varying character of M and K and the analogue of (2.29) obtained by replacing $1/\alpha$ by $1 - 1/\alpha$ and M by M^{-1} , (2.32) implies

$$n^{1-1/\alpha} \{M(n)\}^{-1} \leq C_3 x_n^{\alpha-1} K(x_n)^{1-1/\alpha} \{M(x_n^\alpha K(x_n))\}^{-1}$$

$$\sim C_3 x_n^{\alpha-1} K(x_n) \quad (\text{See (2.31)}).$$

This estimate of the last member of (2.28) leads to

$$(2.33) \quad \int_{\pi/C_1^3 x_n \leq |t| \leq \pi/x_n} \frac{dt}{1 - \psi(t)} \leq C_1^6 C_4 x_n^{\alpha-1} K(x_n), \quad n \geq n_4.$$

C_1 in (2.33) is taken as $C_1(\frac{1}{4}, \frac{3}{4})$, which we assume > 1 without loss of generality, whereas C_3 and C_4 are constants depending on α and $|I|$ only. Next observe that

$$(2.34) \quad d_{n+1} \left(\frac{1}{4} \right) \leq d_n \left(\frac{3}{4} \right) \quad \text{for } n \geq n_5,$$

because for each C_5

$$(2.35) \quad P \left\{ |S_{n+1}| \leq d_n \left(\frac{3}{4} \right) \right\} \geq P \left\{ |S_n| \leq d_n \left(\frac{3}{4} \right) - C_5 \right\} P \{ |X_{n+1}| \leq C_5 \};$$

in particular if C_5 is fixed so large that $P \{ |X_{n+1}| \leq C_5 \} \geq \frac{1}{2}$, then it follows from

$$\begin{aligned} \liminf_{n \rightarrow \infty} P \left\{ |S_n| \leq d_n \left(\frac{3}{4} \right) - C_5 \right\} \\ = \liminf_{n \rightarrow \infty} P \left\{ |S_n| \leq d_n \left(\frac{3}{4} \right) \right\} \quad (\text{see (2.2)}) \geq \frac{3}{4} \end{aligned}$$

and (2.35) that eventually

$$P \left\{ |S_{n+1}| \leq d_n \left(\frac{3}{4} \right) \right\} > \frac{1}{4},$$

whence (2.34), (2.27) together with (2.34) and (2.24) shows

$$x_{n+1} \leq \frac{d_n \left(\frac{3}{4} \right) d_{n+1} \left(\frac{3}{4} \right)}{d_n \left(\frac{1}{4} \right) d_{n+1} \left(\frac{1}{4} \right)} x_n \leq C_1^2 \left(\frac{1}{4}, \frac{3}{4} \right) x_n,$$

so that each interval of the form $[1/\pi C_1^k, 1/\pi C_1^{k-2}]$, $k \geq k_1$ contains at least one x_n^{-1} . This finally allows us to convert (2.33) into

$$\int_{C_1^{-k-1} \leq |t| \leq C_1^{-k}} \frac{dt}{1 - \psi(t)} \leq C_1^6 C_6 C_1^{k(\alpha-1)} K(C_1^k), \quad k \geq k_1,$$

an estimate which is free of x_n . It follows immediately that for $k \geq k_2$

$$\begin{aligned} \int_{C_1^{-k-1} \leq |t| \leq C_1^{-k}} |\psi^n(t)| dt &\leq \int_{C_1^{-k-1} \leq |t| \leq C_1^{-k}} \frac{e^{-n(1-\psi(t))} (1-\psi(t))}{1-\psi(t)} dt \\ &\leq \frac{1}{n} \max_{x \geq 0} x e^{-x} \int_{C_1^{-k-1} \leq |t| \leq C_1^{-k}} \frac{dt}{1-\psi(t)} \leq \frac{C_7}{n} C_1^{k(\alpha-1)} K(C_1^k). \end{aligned}$$

$k_2 \geq k_1$ only has to be chosen such that $\psi(t) \geq 0$ for $|t| \leq C_1^{-k_2}$. If we write C_8 for $C_1^{-k_2}$ we arrive at

$$\begin{aligned}
 (2.36) \quad \int_{C/A_n \leq |t| \leq C_8} |\psi^n(t)| dt &\leq \frac{C_7}{n} \sum_{C_1^k \leq A_n/C} C_1^{k(\alpha-1)} K(C_1^k) \\
 &\leq \frac{C_1}{n} \left(\frac{A_n}{C}\right)^{\alpha-1} K(A_n) \leq \frac{C_{10}}{A_n} \frac{1}{C^{\alpha-1}}, \quad n \geq n_6
 \end{aligned}$$

(see the definition of A_n and (2.30)); C_{10} is independent of C .

The proof of the lemma is now completed by an application of the inversion formula, [5] Theorem 12.1, which gives

$$\begin{aligned}
 (2.37) \quad P \left\{ |S_n| \leq \frac{\pi}{2} \frac{A_n}{C} \right\} &= \frac{1}{2\pi} \int_{-\infty}^{+\infty} \frac{2 \sin \frac{\pi}{2} A_n C^{-1} t}{t} \psi^n(t) dt \\
 &\geq \frac{2}{2\pi} \frac{A_n}{C} \int_{|t| \leq C/A_n} \psi^n(t) dt - \frac{1}{2} \frac{A_n}{C} \int_{|t| > C/A_n} |\psi^n(t)| dt \\
 &\geq \frac{A_n}{\pi C} \int_{-\infty}^{+\infty} \psi^n(t) dt - \frac{A_n}{C} \int_{C/A_n < |t| \leq C_8} |\psi^n(t)| dt \\
 &\quad - \frac{A_n}{C} \int_{|t| > C_8} e^{-n t^2/2} dt \geq \frac{2}{C} - \frac{2C_{10}}{C^\alpha} \quad (\text{See (2.23) and (2.36)}).
 \end{aligned}$$

Since $\alpha > 1$ we can choose $C = C_{11} > 0$ such that the last member of (2.37) exceeds C_{11}^{-1} so that $d(C_{11}^{-1}) \leq \pi A_n / 2C_{11}$. This proves (2.26) for $q = C_{11}^{-1}$ and for general q it then follows from (2.24) and the monotonicity of $d_n(\cdot)$.

To prove the theorem is easy enough now. By (2.25) and (2.26) every increasing sequence of integers contains a subsequence along which the distribution of DS_n/A_n converges weakly to a nondegenerate distribution (D a positive constant). Consider then any sequence $n_i \rightarrow \infty$ such that the weak limit

$$\lim_{i \rightarrow \infty} P \left\{ \frac{DS_{n_i}}{A_{n_i}} \leq x \right\} = G(x)$$

exists. Then also

$$\lim_{i \rightarrow \infty} \psi^{n_i} \left(\frac{Dt}{A_{n_i}} \right) = \gamma(t) = \int e^{itx} dG(x)$$

and it suffices to prove $\gamma(t) = \exp - |t|^\alpha$ when D is properly chosen⁽⁸⁾. To begin

⁽⁸⁾ If ε is taken such that $\psi(t) \geq 0$ for $|t| \leq \varepsilon$, then one easily deduces from (2.23) and Karamata's Tauberian theorem that

$|\{t: |t| \leq \varepsilon \text{ and } \psi(t) \geq y\}| \sim 2C(1-y)^{1/\alpha} M^{-1}(1/1-y)$ as $y \uparrow 1$ ($C = \frac{\pi(\alpha-1)}{\Gamma(1/\alpha)|I|}$) (compare also (2.43) and (2.44) below). One would like to conclude from this that

$$1 - \psi(t) \sim \frac{t^\alpha}{C^\alpha K(1/t)} \text{ as } t \downarrow 0,$$

which is equivalent to the main result (1.4). The author did not succeed in constructing a rigorous proof along these simple lines.

with we have the following simple estimate because X_i has a symmetric distribution ([4], Lemma V.5.2):

$$\begin{aligned}
 (2.38) \quad & \frac{1}{2} P \left\{ \max_{1 \leq i \leq n} |X_i| \geq 2C_2 \left(\frac{3}{4} \right) A_n \right\} \leq P \left\{ |S_n| \geq 2C_2 \left(\frac{3}{4} \right) A_n \right\} \\
 & \leq P \left\{ |S_n| \geq 2d_n \left(\frac{3}{4} \right) \right\} \leq \frac{1}{4}.
 \end{aligned}$$

In turn, (2.38) implies

$$n P \left\{ |X_1| \geq 2C_2 \left(\frac{3}{4} \right) A_n \right\} \leq C_{12}$$

or, using the definition of A_n and $K(y)$,

$$(2.39) \quad P\{|X_1| \geq y\} \leq \frac{C_{13}}{y^\alpha K(y)}, \quad y \geq y_1.$$

For $\alpha = 2$ this estimate will not be sharp enough, but for $\alpha < 2$ we obtain

$$\begin{aligned}
 (2.40) \quad 1 - \psi(t) &= \int_{-\infty}^{+\infty} (1 - e^{itx}) dF(x) = - \int_0^{\infty} (1 - \cos tx) dP\{|X_1| \geq x\} \\
 &\leq 2P\{|X_1| \geq |t|^{-1}\} + \int_0^{|t|^{-1}} t^2 x P\{|X_1| \geq x\} dx \leq C_{14} \frac{|t|^\alpha}{K\left(\frac{1}{|t|}\right)}.
 \end{aligned}$$

As a first estimate for γ we therefore have

$$\begin{aligned}
 (2.41) \quad 1 - \gamma(t) &\leq 1 - \liminf_{n \rightarrow \infty} \psi^n\left(\frac{Dt}{A_n}\right) \leq 1 - \liminf_{n \rightarrow \infty} \left[1 - \frac{C_{14} D^\alpha |t|^\alpha}{A_n^\alpha K\left(\frac{A_n}{|t|}\right)} \right]^n \\
 &\leq 1 - \liminf_{n \rightarrow \infty} \left(1 - C_{15} \frac{D^\alpha}{n} |t|^\alpha \right)^n = 1 - \exp\{-D^\alpha C_{15} |t|^\alpha\}.
 \end{aligned}$$

By means of standard estimates (e.g. [5], §13) for the tail of G in terms of the behavior of its characteristic function γ near the origin it is seen from (2.41) that

$$1 - G(x) + G(-x) = O(x^{-\alpha}) \quad (x \rightarrow \infty)$$

and since $\alpha > 1$ this implies that $\int |x| dG(x)$ is finite and that $\gamma(\cdot)$ is continuously differentiable. We also see from (2.41) that $\gamma(t) > 0$ for all t .

Much more precise information about γ is obtained by computing for any integer $k \geq 1$

$$\begin{aligned} \int_{-\infty}^{+\infty} \gamma^k(t) dt &= \lim_{T \rightarrow \infty} \int_{-T}^{+T} \lim_{i \rightarrow \infty} \psi^{kn_i} \left(\frac{Dt}{A_{ni}} \right) dt \\ &= \lim_{T \rightarrow \infty} \lim_{i \rightarrow \infty} \frac{A_{ni}}{D} \int_{|s| \leq DT A_{ni}^{-1}} \psi^{kn_i}(s) ds \\ &= \lim_{i \rightarrow \infty} \frac{A_{ni}}{D} \int_{-\infty}^{+\infty} \psi^{kn_i}(s) ds + \lim_{T \rightarrow \infty} \lim_{i \rightarrow \infty} O \left(\frac{A_{ni}}{A_{ki}} \frac{1}{T^{\alpha-1}} + A_{ni} \int_{C_8} e^{-kn_i s^2/2} ds \right) \end{aligned}$$

(see (2.36) and (2.22)) = $2\pi/Dk^{1/\alpha}$ (see (2.23)).

We now choose $D = a\pi/\Gamma(1/\alpha)$, which is indeed equivalent to taking

$$C = \frac{\pi(\alpha - 1)}{\Gamma\left(\frac{1}{\alpha}\right)|I|}$$

in (1.4). For this choice of D

$$(2.42) \quad \int_{-\infty}^{+\infty} \gamma^k(t) dt = \frac{2\Gamma\left(\frac{1}{\alpha}\right)}{\alpha k^{1/\alpha}} = \int_{-\infty}^{+\infty} e^{-k|t|^\alpha} dt.$$

Introduce

$$v(y) = |\{t: \gamma(t) \geq y\}|, \quad 0 \leq y \leq 1.$$

Because $0 \leq \gamma(t) \leq 1$ the left hand side of (2.42) can then be rewritten as

$$(2.43) \quad \int_{-\infty}^{+\infty} \gamma^k(t) dt = - \int_0^1 x^k dv(x), \quad k = 1, 2, \dots$$

Since the finite measure $-x dv(x)$ is uniquely determined by its moments (see [4], Chapter VII.3 or use Theorem 2.9.3 in [3] after an integration by parts), (2.42) and (2.43) imply

$$(2.44) \quad v(y) = |\{t: e^{-|t|^\alpha} \geq y\}| = 2 \left(\log \frac{1}{y} \right)^{1/\alpha}.$$

To complete the proof (for $\alpha < 2$) we show that $\gamma(t)$ is strictly decreasing on $t \geq 0$. Indeed, if there exist $0 \leq t_1 < t_2$ with $\gamma(t_1) \leq \gamma(t_2)$ then $\min_{0 \leq t \leq t_2} \gamma(t)$ is taken on at a point $t_3 \in (0, t_2)$ (because also $\gamma(0) = 1 > \min_{0 \leq t \leq t_2} \gamma(t)$ for any $t_2 > 0$ since G is non degenerate by (2.25); see [5], Theorem 14.2). At t_3 we must have $\gamma'(t_3) = 0$ and $0 < z = \gamma(t_3) < 1$ (recall that $\gamma(t) > 0$ for all t). But this is impossible for then

$$\limsup_{\varepsilon \downarrow 0} \frac{1}{\varepsilon} |\{t: z - \varepsilon \leq \gamma(t) \leq z + \varepsilon\}| = \infty$$

whereas this limit should have the finite value

$$\lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon} [v(z - \varepsilon) - v(z + \varepsilon)] = \frac{4}{\alpha z} \left(\log \frac{1}{z} \right)^{1/\alpha - 1}.$$

Thus $\gamma(t)$ is strictly decreasing on $t \geq 0$ and because γ is symmetric $v(y) = 2t(y)$ where $t(y)$ is the unique $t \geq 0$ with $\gamma(t) = y$. This means $t(y) = (\log 1/y)^{1/\alpha}$, $\gamma((\log 1/y)^{1/\alpha}) = y$ or $\gamma(t) = \exp - |t|^\alpha$ as desired.

For $\alpha < 2$ the proof is complete, but for $\alpha = 2$ an extra argument is needed because the last estimate in (2.40) could conceivably fail. We show that (2.40) is correct even for $\alpha = 2$. After (2.40) the proof did not rely on $\alpha < 2$ and can therefore be used also if $\alpha = 2$. To obtain (2.40) for $\alpha = 2$ we put

$$\sigma^2(T) = \int_{-T}^{+T} x^2 dF(x).$$

Clearly $\sigma^2(\cdot)$ is nondecreasing and

$$1 - \psi(t) = \int (1 - \cos xt) dF(x) \geq \frac{2}{\pi^2} \int_{|x| \leq \pi/t} t^2 x^2 dF(x) = \frac{2t^2}{\pi^2} \sigma^2\left(\frac{\pi}{t}\right).$$

Therefore

$$\begin{aligned} (2.45) \quad \int_{|t| \leq C/A_n} \psi^n(t) dt &\leq \int_{|t| \leq C/A_n} \exp\{-n(1 - \psi(t))\} dt \\ &\leq \int_{-\infty}^{\infty} \exp\left\{-\frac{2nt^2}{\pi^2} \sigma^2\left(\frac{\pi A_n}{C}\right)\right\} dt = \frac{\pi^{3/2}}{\sqrt{2n}} \frac{1}{\sigma\left(\frac{\pi A_n}{C}\right)} \end{aligned}$$

Together with (2.23), (2.36) and (2.22), (2.45) implies

$$\frac{\pi^{3/2}}{\sqrt{2n}} \frac{1}{\sigma\left(\frac{\pi A_n}{C}\right)} + \frac{C_{10}}{A_n} \frac{1}{C^{\alpha-1}} \geq \frac{\pi}{A_n}, \quad n \geq n_7.$$

Thus for $C = C_{16}$ say, C_{16} sufficiently large,

$$\sigma\left(\frac{\pi A_n}{C_{16}}\right) \leq C_{17} \frac{A_n}{\sqrt{n}} = C_{18} M(n) \quad (\alpha = 2),$$

or, in view of the definition of A_n and (2.31)

$$\sigma(y) \leq C_{19} K^{-1/2}(y).$$

Thus, by virtue of (2.39),

$$1 - \psi(t) = \int_{-\infty}^{+\infty} (1 - \cos tx) dF(x) \leq 2 P \left\{ |X_1| \geq \frac{1}{|t|} \right\} + \frac{1}{2} t^2 \sigma^2 \left(\frac{1}{|t|} \right)$$

$$\leq \frac{2 C_{13} t^2}{K \left(\frac{1}{|t|} \right)} + \frac{1}{2} t^2 \frac{C_{19}}{K \left(\frac{1}{|t|} \right)},$$

which is the desired replacement for (2.40).

REFERENCES

1. H. Cramér, *Mathematical methods of statistics*, Princeton, 1946.
2. D. A. Darling and M. Kac, *On occupation times for Markoff processes*, Trans. Amer. Math. Soc. **84** (1957) 444-458.
3. G. Doetsch, *Handbuch der Laplace Transformation*, Band I, Basel, 1950.
4. W. Feller, *An introduction to probability theory and its applications*, Vol. II, New York, 1966.
5. B. V. Gnedenko and A. N. Kolmogorov, *Limit distributions for sums of independent random variables*, Cambridge, Mass., 1954.
6. H. Kesten, *A sharper form of the Doeblin-Lévy-Kolmogorov-Rogozin inequality for concentration functions*, submitted to Math. Scand.
7. P. Lévy, *Théorie de l'addition des variables aléatoires*, 2^{me} éd., Paris, 1954.
8. C. Stone, *Ratio limit theorems for random walks on groups*, Trans. Amer. Math. Soc. **125** (1966) 86-100.
9. ———, *On local and ratio limit theorems*, Proc. Fifth Berkeley Symposium on Math. Stat. and Prob., Vol. II, Part II, 217-224, Berkeley, 1967.

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