

Harmonic Renewal Measures and Bivariate Domains of Attraction in Fluctuation Theory

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Summary. Each probability measure C on a first orthant is associated with a harmonic renewal measure G . Specifically we consider (N, S_N) the ladder (time, place) of a random walk S_n . Using bivariate G we show that when S_1 is in a domain of attraction so is (N, S_N) . This unifies and generalizes results of Sinai, Rogosin.

Introduction

Let $\{S_n\}_1^\infty$ be a random walk generated by a random variable with step distribution F . Let $N = \inf\{n: S_n > 0\}$ be the first ladder index and S_N the first ladder height. Our object here is to show that if S_1 is in a domain of attraction, then the bivariate (N, S_N) is in an associated bivariate domain of attraction. The marginal results are theorems of Rogosin for N [11] and of Sinai [14] and Rogosin [12] for S_N .

We offer a unified treatment using harmonic renewal measures and double exponential representations. We rely heavily on the univariate results in [6].

§ 1a. Univariate Case: Harmonic Renewal Measure

Since we will heavily rely on the results for the univariate case we now formulate some theorems for that situation. We treat these results in such a way that they are optimally stated for future application in the bivariate case; the main methodology of the proof in the latter situation can be viewed as a generalisation from the univariate to the bivariate case.

So let X_1, X_2, \dots be i.i.d. with distribution C on $(0, \infty)$. We say that C is in a *domain of attraction* iff we can find constants $a_n \uparrow \infty$ such that $a_n^{-1}(X_1 + \dots + X_n)$ converges in distribution to a r.v. nondegenerate at 0.

Alternative formulations are contained in the well-known lemma. Lower case letters stand for LST (Laplace-Stieltjes transform) of corresponding capital letters.

Lemma 1.1. *The following statements are equivalent for $a_n \uparrow \infty$*

- (i) $a_n^{-1}(X_1 + \dots + X_n) \xrightarrow{\mathcal{D}} U$, non-degenerate at 0;
- (ii) $\forall x \geq 0: C^{(n)}(a_n x) \rightarrow H(x) = P\{U \leq x\}$, U non-degenerate at 0;
- (iii) $\forall s > 0, n\{1 - c(s/a_n)\} \rightarrow -\log h(s) \neq 0$.

The conditions of Lemma 1.1 can be transformed into a variety of equivalent statements involving the tail of C or the harmonic renewal measure G associated to C . Recall that $G(x) = \sum_1^\infty n^{-1} C^{(n)}(x)$ and that $1 - c(s) = \exp -g(s)$.

Lemma 1.2. *Let $0 < \beta \leq 1$ and L° s.v. Put $R(x) = x^{-\beta} L^\circ(x)$. The following statements are equivalent:*

- (i) $\int_0^x [1 - C(y)] dy \sim x R(x) / \Gamma(2 - \beta)$ as $x \rightarrow \infty$;
- (ii) $\forall t > 0: G(xt) + \log R(x) \rightarrow \beta \gamma + \beta \log t$ as $x \rightarrow \infty$;
- (iii) $\forall \theta > 0, g(\theta s) + \log R(1/s) \rightarrow -\beta \log \theta$ as $s \downarrow 0$.

Proof. (i) \Leftrightarrow (ii): It follows from Theorem 1 and 2 in [6] that (i) is equivalent to

$$G(x) + \log R(x) \rightarrow \beta \gamma. \tag{1}$$

Change x into xt and use $R(xt)/R(x) \rightarrow t^{-\beta}$.

(ii) \Leftrightarrow (iii): Again in [6] it was shown that (1) is equivalent to

$$g(s) + \log R(1/s) \rightarrow 0.$$

Put s equals θs to obtain (iii). \square

As a guideline for the bivariate case we briefly indicate one possible procedure to link the above two lemmas.

Assume (ii) of Lemma 1.2 holds. Then for any sequence $a_n \uparrow \infty$ and $0 < x < u < \infty$

$$\int_x^u G(a_n dp) = G(a_n u) - G(a_n x) \rightarrow \beta \log \frac{u}{x} \equiv \int_x^u L(dp). \tag{2}$$

But then for $\theta \in (0, \infty), x \in (0, \infty)$

$$\begin{aligned} g\left(\frac{\theta}{a_n}\right) - G(a_n x) &= \int_0^x (e^{-\theta u} - 1) G(a_n du) + \int_x^\infty e^{-\theta u} G(a_n du) \\ &\rightarrow \int_0^\infty (e^{-\theta p} - I_{(0,x)}(p)) L(dp). \end{aligned}$$

Or for $\theta \in (0, \infty), x \in (0, \infty)$, (2) implies

$$g\left(\frac{\theta}{a_n}\right) - G(a_n x) \rightarrow -\beta(\gamma + \log \theta x) \equiv \gamma(\theta, x) \tag{3}$$

where we used the explicit form of L and one of the possible representations of Euler's constant [5, p. 946].

$$\gamma = \lim_{n \rightarrow \infty} \left\{ \sum_{m=1}^n \frac{1}{m} - \log n \right\} = \int_0^1 (1 - e^{-w}) \frac{dw}{w} - \int_1^{\infty} e^{-w} \frac{dw}{w}. \tag{4}$$

However (3) implies that for $\theta \in (0, \infty)$

$$g\left(\frac{1}{a_n}\right) - g\left(\frac{\theta}{a_n}\right) \rightarrow \gamma(1, 1) - \gamma(\theta, 1) = \beta \log \theta.$$

Choose a_n in such a way that $g(1/a_n) = \log n$ then using $-g(s) = \log(1 - c(s))$ we get

$$\log n + \log(1 - c(\theta/a_n)) \rightarrow \log[-\log h(\theta)] = \beta \log \theta \tag{5}$$

which is (iii) of Lemma 1.1 with $h(\theta) = \exp(-\theta^\beta)$.

The point of the above analysis is that in the bivariate case we do not have such a simple expression for the limit quantities L and h . In the bivariate case a relation of the form (2) is given in Lemma 3.1. The consequence (3) is treated in Lemma 2.2. The analogue of (5) is the content of Theorem 3.2.

§ 1b. Univariate Case: Double Exponential Representations of Infinitely Divisible Distributions

The limit h can be put in a form that generalizes to more dimensions. Suppose that C is infinitely divisible and let $C^{(t)}$ denote the t -th power of C , i.e. the distribution corresponding to the LST $c^t(s)$; equivalently $C^{(t)}$ is the distribution of X_t where X is a process with stationary independent increments such that X_1 has distribution C .

We will use the Frullani integral form of $\log(a/b)$, i.e. for $a > 0$ and $b > 0$ [5, p. 334].

$$-\log \frac{a}{b} = \int_0^{\infty} \frac{e^{-aw} - e^{-bw}}{w} dw.$$

Write $a = 1$ and $b = -\log c(s)$; then successively

$$\begin{aligned} \log[-\log c(s)] &= \int_0^{\infty} w^{-1} \{e^{-w} - [c(s)]^w\} dw \\ &= \int_0^{\infty} w^{-1} \left\{ e^{-w} - \int_0^{\infty} e^{-sz} P\{X_w \in dz\} \right\} dw \\ &= \int_0^{\infty} \int_0^{\infty} w^{-1} \{e^{-w} - e^{-sz}\} P\{X_w \in dz\} dw \end{aligned}$$

or

$$\log[-\log c(s)] = \int_0^{\infty} \int_0^{\infty} w^{-1} \{e^{-w} - e^{-sz}\} C^{(w)}(dz) dw. \tag{6}$$

Hence when C is infinitely divisible we have as a companion to

$$\log[1 - c(s)] = - \int_0^\infty e^{-sz} G(dx)$$

the “continuous time” representation (6). We call (6) the *double exponential representation* of the distribution C or of its transform c .

It is useful to note at this point that the above derivation of (6) is equally straightforward if C is a distribution on $(-\infty, \infty)$ and c its characteristic function, or, in fact if C is a distribution on a higher dimensional space. The bivariate version of (6) will be used in § 2.

Return to (5). Then as

$$\log\{-\log h(s)\} = \int_0^\infty \int_0^\infty w^{-1} \{e^{-w} - e^{-sz}\} B^{(w)}(dz) dw$$

we can write for $s \in (0, \infty)$

$$\begin{aligned} \log n - \sum_{m=1}^n \frac{1}{m} + \int_0^\infty (1 - e^{-sx}) \sum_{m=1}^n \frac{1}{m} C^{(m)}(a_n dx) - \int_0^\infty e^{-sx} \sum_{m=n+1}^\infty \frac{1}{m} C^{(m)}(a_n dx) \\ \rightarrow -\gamma + \int_0^\infty (1 - e^{-sx}) \int_0^1 \frac{dw}{w} B^{(w)}(dx) - \int_0^\infty e^{-sx} \int_1^\infty \frac{dw}{w} B^{(w)}(dx) \end{aligned} \tag{7}$$

where we performed some simple algebra on the double exponential representation of $h(s)$.

The limiting operation in (7) constitutes another alternate form of the conditions listed in Lemmas 1.1 and 1.2; it tells us that the convergence of the functional of a sample path of the random walk which can be read from either side of (7) is equivalent to the regular variation of C .

The analogous version of (7) for bivariate C is of additional interest and use because in the bivariate case we do not know how to write the characteristic exponent $-\log h$ for the limiting stable distribution in closed form.

§ 2. Bivariate Case

Let C be a distribution on $(0, \infty) \times (0, \infty)$ such that $C(0+, 0+) = 0$ and let c be its Laplace-Stieltjes transform; let G be the corresponding harmonic renewal measure with LST g . Hence

$$\begin{aligned} \forall \lambda \geq 0, \mu \geq 0 \quad c(\lambda, \mu) &= \int_0^\infty \int_0^\infty \exp -(\lambda x + \mu y) C(dx, dy) \\ \forall (x, y) \in \mathbb{R}_+^2 \quad G(x, y) &= \sum_{n=1}^\infty \frac{1}{n} C^{(n)}(x, y) \\ \forall \lambda \geq 0, \mu \geq 0 \quad 1 - c(\lambda, \mu) &= \exp -g(\lambda, \mu) \\ \forall \lambda \geq 0, \mu \geq 0 \quad g(\lambda, \mu) &= \int_0^\infty \int_0^\infty \exp -(\lambda x + \mu y) G(dx, dy). \end{aligned} \tag{8}$$

A bivariate distribution C is in the *domain of attraction* of a bivariate stable distribution H if there exist real sequences $a_n \uparrow \infty$, $b_n \uparrow \infty$ such that for $x \geq 0$, $y \geq 0$ as $n \rightarrow \infty$

$$C^{(n)}(a_n x, b_n y) \rightarrow H(x, y)$$

in the sense of convergence of distribution functions. The distribution H is, by definition, *bivariate stable* if it is such a limit.

Conditions equivalent to the bivariate domain of attraction condition in the more general setting where C is a distribution on \mathbb{R}^2 and centering constants may also appear, were obtained by Resnick and Greenwood [7]. The following additional result is also useful.

Lemma 2.1. *The following statements are equivalent*

- (i) $\forall (x, y) \in \mathbb{R}_+^2 : C^{(n)}(a_n x, b_n y) \rightarrow H(x, y)$,
- (ii) $\forall \lambda > 0, \mu > 0 : n \{1 - c(\lambda/a_n, \mu/b_n)\} \rightarrow -\log h(\lambda, \mu)$.

Going from (2) to (3) is generally possible in the bivariate case as shown in the following result. Let G be as defined in (8).

Lemma 2.2. *If for (x, y, u, v) satisfying $0 \leq x < u, 0 \leq y < v$ but $x > 0$ or $y > 0$*

$$\int_x^u \int_y^v G(a_n dp, b_n dq) \rightarrow \int_x^u \int_y^v L(dp, dq) \tag{9}$$

as $n \rightarrow \infty$, then for positive θ, η, x, y there exists a function $\gamma(\theta, \eta, x, y)$ such that

$$g\left(\frac{\theta}{a_n}, \frac{\eta}{b_n}\right) - G(a_n x, b_n y) \rightarrow \gamma(\theta, \eta, x, y). \tag{10}$$

If (9) holds with $v = \infty$, then (10) holds with $y = \infty$.

Proof. The hypothesis implies convergence of the measure G_n associated with $G(a_n x, b_n y)$ on Borelsets in \mathbb{R}_+^2 .

Let $A(x, y) = (0, x) \times (0, y)$ and $A(x, y)^c = A(\infty, \infty) - A(x, y)$. Then

$$g\left(\frac{\theta}{a_n}, \frac{\eta}{b_n}\right) - G(a_n x, b_n y) = \iint_{A(x, y)} [e^{-\theta u - \eta v} - 1] G(a_n du, b_n dv) + \iint_{A(x, y)^c} e^{-\theta u - \eta v} G(a_n du, b_n dv). \tag{11}$$

Denote the first integral on the right of (11) by $I_n(x, y)$, the second by $II_n(x, y)$.

Pick $\varepsilon > 0$. By assumption for $x > \varepsilon, y > \varepsilon$

$$I_n(x, y) - I_n(\varepsilon, \varepsilon) \rightarrow \iint_{A(x, y) - A(\varepsilon, \varepsilon)} [e^{-\theta u - \eta v} - 1] L(du, dv).$$

But

$$0 \leq -I_n(\varepsilon, \varepsilon) \leq \iint_{A(\varepsilon, \varepsilon)} (\theta u + \eta v) G(a_n du, b_n dv).$$

For the first term of the integral we write $G_1(x) = \int_0^x u G(du, \infty)$; then

$$\int_0^\varepsilon u \int_0^\varepsilon G(a_n du, b_n dv) \leq \int_0^\varepsilon u \int_0^\infty G(a_n du, b_n dv) = a_n^{-1} G_1(a_n \varepsilon).$$

But G_1 satisfies the conditions of Lemma 3.1.1 in [6]. Hence uniformly in n , $G_1(a_n \varepsilon) \leq a_n \varepsilon$ and $|I_n(\varepsilon, \varepsilon)| \leq \varepsilon(\theta + \eta)$. Letting $n \rightarrow \infty$ and then $\varepsilon \downarrow 0$ we find

$$I_n(x, y) \rightarrow \iint_{A(x,y)} (e^{-\theta u - \eta v} - 1) L(du, dv).$$

We pass to II_n , and write $II_n(x, y) = II_n(x_o, y_o) + (II_n(x, y) - II_n(x_o, y_o))$ where (x_o, y_o) will be determined directly. By assumption the second part of $II_n(x, y)$ will converge. Now

$$\begin{aligned} II_n(x_o, y_o) &= \int_{x_o}^\infty \int_0^{y_o} e^{-(\theta u + \eta v)} G(a_n du, b_n dv) \\ &\quad + \int_0^\infty \int_{y_o}^\infty e^{-(\theta u + \eta v)} G(a_n du, b_n dv). \end{aligned} \tag{12}$$

The first integral on the right, say J_n , is estimated as follows:

$$\begin{aligned} J_n &\leq \int_{x_o}^\infty \int_0^\infty e^{-\theta u} \cdot 1 \cdot G(a_n du, b_n dv) = \int_{x_o}^\infty e^{-\theta u} G(a_n du, \infty) \\ &= \theta \int_{x_o}^\infty e^{-\theta u} \{G(a_n u, \infty) - G(a_n x_o, \infty)\} du. \end{aligned}$$

But as proved in Lemma 3.1.1 [6], $|G(u, \infty) - G(v, \infty)| \leq 1 + \log \left| \frac{u}{v} \right|$. Hence uniformly in n

$$J_n \leq \theta \int_{x_o}^\infty e^{-\theta u} \left\{ 1 + \log \frac{u}{x_o} \right\} du$$

and this can be made as small as we want by taking x_o large enough.

A similar argument estimates the second integral on the right of (12).

This proves the lemma. \square

The form of the limit in (10) can be conveniently written as

$$\gamma(\theta, \eta, x, y) = \int_0^\infty \int_0^\infty \{e^{-\theta p - \eta q} - I_{(0,x) \times (0,y)}(p, q)\} L(dp, dq). \tag{13}$$

For future reference we write down the double exponential representation for the case when C is infinitely divisible: for $\lambda > 0, \mu > 0$

$$\log \{-\log c(\lambda, \mu)\} = \int_0^\infty \int_0^\infty \int_0^\infty [e^{-w} - e^{-\lambda x - \mu y}] C^{(w)}(dx, dy) \frac{dw}{w}. \tag{14}$$

§ 3. The Bivariate Distribution of (N, S_N)

Now we consider a random walk $\{S_n\}_1^\infty$ generated by S_1 ; we let $N = \inf\{n: S_n > 0\}$ be the ladder index and S_N the ladder height. From now on

$$C(x, y) = P\{N \leq x, S_N \leq y\}.$$

The Spitzer-Baxter identity provides an identification of the harmonic renewal measure G associated with C in terms of the distributions of $\{S_m, m \in \mathbb{N}\}$. Indeed

$$\log\{1 - c(\lambda, \mu)\} = \int_0^\infty \int_0^\infty e^{-\lambda x - \mu y} d \left\{ \sum_{m=1}^\infty \frac{1}{m} P[S_m \in dy] \delta_m(dx) \right\}$$

where δ_m is a unit-step distribution at m . Hence

$$G(dx, dy) = \sum_{m=1}^\infty \frac{1}{m} P[S_m \in dy] \delta_m(dx)$$

or in integrated form

$$G(x, y) = \sum_{m=1}^{\lfloor x \rfloor} \frac{1}{m} P[0 < S_m \leq y]. \tag{15}$$

We first show that if S_1 is in a domain of attraction, G satisfies (9). From the resulting relation (10) we then show that (ii) of Lemma 2.1 is satisfied. On the basis of [6] and Lemma 1.2 one might expect that we should look for a bivariate version of deHaan’s result [1] linking the bivariate G to g and hence to c . We leave these questions aside and prove a result which looks like a deHaan-type theorem in two variates, but which is actually univariate along certain curves in 2-space.

We now provide the bivariate analogue of (2).

Lemma 3.1. *Suppose $0 < c_n \rightarrow \infty$ is chosen in such a way that $S_n/c_n \xrightarrow{d} X_1$, stable with index α . Let G be defined by (15). Then for $0 \leq x < u < \infty, 0 \leq y < v \leq \infty$ but $x > 0$ or $y > 0$*

$$\int_x^u \int_y^v G(ndp, c_n dq) \rightarrow \int_x^u \int_y^v \frac{1}{t} P\{X_t \in dz\} dt \tag{16}$$

where $\{X_t, t \geq 0\}$ is the associated stable processes.

Proof. First assume $0 < x < u < \infty, 0 < y < v \leq \infty$; then the left hand side of (16) can be written as

$$\int_x^u \frac{n}{[nt]} P\{y < c_n^{-1} S_{[nt]} \leq v\} dt.$$

By assumption $P\{y < c_n^{-1} S_{[nt]} \leq v\} \rightarrow P\{y < X_t \leq v\}$; since $X_t \stackrel{d}{=} t^{1/\alpha} X_1$ and X_1 has a continuous distribution, the convergence is uniform [9, p.139]. Also uniformly $n/[nt] \rightarrow 1/t$. From the uniform convergence of the integrands to a continuous limit it follows that the integrals converge.

Let now $y=0, v=\varepsilon$, for a fixed $\varepsilon > 0$. Then an entirely similar argument yields the required convergence on $[x, u]$ as long as $x > 0$.

Finally, let $x=0, u=\varepsilon, v \leq \infty$ but $y>0$. The next argument has been inspired by a result of Heyde [8, p. 306] on large deviations. We look for an upper bound on $P[S_m > c_n y]$ where $1 \leq m \leq [n\varepsilon]$.

Let $Y_k \equiv S_k - S_{k-1}$ ($k \geq 1$); take Y_k^s as a symmetrized version of Y_k ; finally truncate Y_k^s at $z \equiv c_n y$:

$$Y'_k = \begin{cases} Y_k^s & \text{if } |Y_k^s| \leq z \\ 0 & \text{if not.} \end{cases}$$

Then

$$P \left\{ \left| \sum_{k=1}^m Y_k^s \right| > z \right\} \leq m P\{|Y_k^s| > z\} + P \left\{ \left| \sum_{k=1}^m Y'_k \right| > z \right\}.$$

Put for brevity $H(x) = 1 - F(x) + F(-x)$ and $U(x) = \int_0^x y^2 dF(y)$ where F is the d.f. of $Y_1 = S_1$. By a weak symmetrization inequality [10, p. 257] we obtain

$$P\{|Y_k^s| > z\} \leq 2P \left\{ |Y_k| > \frac{z}{2} \right\} \leq 2H(z/2).$$

By Chebyshev's inequality and part of Heyde's reasoning

$$\begin{aligned} P \left\{ \left| \sum_{k=1}^m Y'_k \right| > z \right\} &\leq z^{-2} E \left| \sum_{k=1}^m Y'_k \right|^2 \leq 2m z^{-2} \int_0^z x P\{|Y_1^s| > x\} dx \\ &\geq 16m z^{-2} \int_0^{z/2} y H(y) dy = 2m \{H(z/2) + 8z^{-2} U(z/2)\}. \end{aligned}$$

Combining the above inequalities we get

$$P \left\{ \left| \sum_{k=1}^m Y_k^s \right| > z \right\} \leq 4m \{H(z/2) + 2z^{-2} U(z/2)\}.$$

Since S_1 belongs to the domain of attraction of a stable law with index α , $u^2 H(u)/U(u) \rightarrow (2-\alpha) \cdot \alpha^{-1}$ as $u \rightarrow \infty$, [3, p. 313]; moreover $U(u) \sim u^{2-\alpha} L(u)$ for some s.v. L , [3, p. 312] while c_n is determined by $n c_n^{-2} U(c_n) \rightarrow 1$. Hence $z^{-2} U(z/2) = y^{-2} c_n^{-2} U(\frac{1}{2} y c_n) \sim y^{-2} 2^{\alpha-2} n^{-1}$. But then there exists a constant K_1 independent of m, n and ε such that

$$P \left\{ \left| \sum_{k=1}^m Y_k^s \right| > z \right\} \leq K_1 y^{-\alpha} \frac{m}{n}.$$

This is the desired inequality for the symmetrized sequence. Using another weak symmetrization inequality [10, p. 257] we find a constant K_2 .

$$P\{S_m - \mu_m > z\} \leq K_2 y^{-\alpha} \frac{m}{n} \tag{17}$$

where μ_m is a median for S_m .

In order to get rid of μ_m we first note that μ_m/c_m is a bounded sequence; for if not take a subsequence $m' \rightarrow \infty$ along which say $\mu_{m'}/c_{m'} \rightarrow \infty$. Then by as-

sumption

$$\frac{1}{2} \leq P\{S_{m'} \geq \mu_{m'}\} = P\{(S_{m'}/c_{m'} \geq (\mu_{m'}/c_{m'}))\} \rightarrow P\{X_1 \geq \infty\}$$

which is a contradiction. Hence for $1 \leq m \leq [n\varepsilon]$ and constants K_i ($i=3, 4, 5$) as before we have

$$|\mu_m|/c_n = K_3(c_m/c_n) = K_4(c_{[n\varepsilon]}/c_n) \leq K_5 \varepsilon^{1/\alpha} \equiv \varepsilon'.$$

Indeed c_n is a r.v. sequence with index α^{-1} and hence is asymptotically monotone. It follows that

$$P\{S_m > c_n y\} \leq P\{S_m - \mu_m > c_n y'\}$$

where $y' = y - \varepsilon'$. By the first part of the proof we can assume without loss of generality that $y' > 0$ by taking ε small enough. Then by (17) finally for $1 \leq m \leq [n\varepsilon]$

$$P\{S_m > c_n y\} \leq K_2(y')^{-\alpha} \frac{m}{n}.$$

Returning to (16) we see that

$$\begin{aligned} \int_0^\varepsilon \int_y^v G(ndp, c_n dq) &= \sum_{m=1}^{[n\varepsilon]} \frac{1}{m} P\{c_n y < S_m \leq c_n v\} \\ &\leq \sum_{m=1}^{[n\varepsilon]} \frac{1}{m} P[c_n y < S_m] \leq K_2(y')^{-\alpha} \varepsilon. \end{aligned}$$

On the right hand side of (16) we remark that by the stability of the limit $\{X_t\}$ there exists a constant K' such that

$$P\{y < X_t\} = P\{y t^{-1/\alpha} < X_1\} \leq K' y^{-\alpha} t.$$

Hence

$$\int_0^\varepsilon \frac{dt}{t} \int_y^v P\{X_t \in dz\} \leq K' y^{-\alpha} \varepsilon.$$

Letting $\varepsilon \downarrow 0$, (16) is fully proved. \square

As a consequence of the above lemma we obtain

$$\gamma(\theta, \eta, x, y) = \int_0^\infty \frac{dt}{t} \int_0^\infty \{e^{-\theta t - \eta z} - I_{(0,x) \times (0,y)}(t, z)\} P\{X_t \in dz\}.$$

We finally prove our main theorem. We will choose the $\{a_n\}$ sequence on the basis of the domain of attraction condition of N ; the sequence $\{b_n\}$ will then automatically be determined.

Theorem 3.2. *Suppose that S_1 is in a domain of attraction of a stable distribution with parameters α, p, q where $0 < \alpha \leq 2$, but $\alpha \neq 1$ unless $p = q = \frac{1}{2}$. Let $0 < c_n \rightarrow \infty$ be chosen in such a way that $S_n/c_n \xrightarrow{\mathcal{D}} X_1$. Then (N, S_N) is in a bivariate domain of attraction and the limiting Laplace-Stieltjes transform is $h(\lambda, \mu)$ where*

$$\log\{-\log h(\lambda, \mu)\} = \int_0^\infty \int_0^\infty \{e^{-w} - e^{-\lambda w - \mu z}\} P\{X_w \in dz\} \frac{dw}{w}.$$

Proof. Let $G_N(x) = G(x, \infty)$ be the harmonic renewal measure corresponding to N . Since S_n/c_n converges, N is in a domain of attraction of index $\beta = P(X_1 > 0)$. See [2, 10] or example 1 in [6]. Note that $\alpha \neq 1$ unless $p = 1/2$ is basically used here.

By (ii) of Lemma 1.2 there exists a sequence of constants $0 < d_n \rightarrow \infty$ such that for $0 < y$ and $n \rightarrow \infty$

$$G_N(ny) - \log d_n \rightarrow \beta\gamma + \beta \log y.$$

By the definition of G_N then

$$G_N(n) - \log d_n = \sum_{m=1}^n \frac{1}{m} P(S_m > 0) - \log d_n \rightarrow \beta\gamma.$$

Define for all $n \geq 1$, $a_n = n$ and $b_n = c_n$; then by (16)

$$\lim_{n \rightarrow \infty} \int_{nx}^{nu} \int_{c_n y}^{c_n v} G(dp, dq) = \int_x^u \int_y^v \frac{1}{t} P\{X_t \in dz\} dt$$

and hence for $\lambda > 0, \mu > 0$ by (10)

$$g(\lambda/n, \mu/c_n) - G(n, \infty) \rightarrow \gamma(\lambda, \mu, 1, \infty).$$

We work towards (ii) of Lemma 2.1. Let us write

$$\begin{aligned} \log d_n - g(\lambda/n, \mu/c_n) &= \left\{ \log d_n - \sum_{m=1}^n \frac{1}{m} P[S_m > 0] \right\} \\ &+ \sum_{m=1}^n \int_0^\infty \int_0^\infty \frac{1}{m} \left\{ 1 - \exp - \left(\frac{\lambda x}{n} + \frac{\mu y}{c_n} \right) \right\} \delta_m(dx) P[S_m \in dy] \\ &- \sum_{m=n+1}^\infty \int_0^\infty \int_0^\infty \frac{1}{m} \exp - \left(\frac{\lambda x}{n} + \frac{\mu y}{c_n} \right) \delta_m(dx) P[S_m \in dy]. \end{aligned}$$

The first term has limit $-\beta\gamma$ as $n \rightarrow \infty$; in the two other terms one uses the effect of the point masses δ_m to write

$$\begin{aligned} &\lim_{n \rightarrow \infty} \{ \log d_n - g(\lambda/n, \mu/c_n) \} \\ &= -\beta\gamma + \lim_{n \rightarrow \infty} \left\{ \sum_{m=1}^\infty \int_0^\infty \int_0^\infty \frac{1}{m} \left\{ 1 - \exp - \left(\frac{\lambda x}{n} + \frac{\mu y}{c_n} \right) \right\} \delta_m(dx) P[S_m \in dy] \right. \\ &\quad \left. - \sum_{m=1}^\infty \int_n^\infty \int_0^\infty \frac{1}{m} \exp - \left(\frac{\lambda x}{n} + \frac{\mu y}{c_n} \right) \delta_m(dx) P[S_m \in dy] \right\} \\ &= -\beta\gamma + \lim_{n \rightarrow \infty} \{ G(n, \infty) - g(\lambda/n, \mu/c_n) \} \\ &= -\beta\gamma - \gamma(\lambda, \mu, 1, \infty). \end{aligned}$$

Replace $[d_n]$ by m then for some sequences $0 < a'_m \rightarrow \infty$ and $0 < b'_m \rightarrow \infty$

$$\log m + \log \{ 1 - c(\lambda/a'_m, \mu/b'_m) \} \rightarrow -\beta\gamma - \gamma(\lambda, \mu, 1, \infty). \tag{18}$$

It remains to identify the limit. Now by (13)

$$\begin{aligned} \log\{-\log h(\lambda, \mu)\} &= -\beta \gamma - \int_0^\infty \int_0^\infty [e^{-\lambda t - \mu z} - I_{(0,1)}(t)] P[X_t \in dz] \frac{dt}{t} \\ &= -\beta \gamma + \int_0^\infty \int_0^\infty \{e^{-t} - e^{-\lambda t - \mu z}\} P[X_t \in dz] \frac{dt}{t} \\ &\quad + \int_0^\infty \int_0^\infty \{I_{(0,1)}(t) - e^{-t}\} P\{X_t \in dz\} \frac{dt}{t}. \end{aligned}$$

The last integral is easily evaluated using (4); it equals $\gamma P[X_t > 0] = \gamma \beta$ since $\beta = P[X_t > 0] = P[X_1 > 0]$. \square

According to (14) we expect $\log\{-\log h(\lambda, \mu)\}$ to be of the form given in the double exponential representation for some $H^{(w)}$. Denote by $(Y_1(t), Y_2(t))$ the limiting bivariate stable processes associated with a random walk having steps distributed like (N, S_N) . Then we can write

$$\log\{-\log h(\lambda, \mu)\} = \int_0^\infty \int_0^\infty [e^{-t} - e^{-\lambda x - \mu y}] P\{(Y_1(t), Y_2(t)) \in (dx, dy)\} \frac{dt}{t}.$$

Theorem 3.2 gives alternatively

$$\log\{-\log h(\lambda, \mu)\} = \int_0^\infty \int_0^\infty [e^{-t} - e^{-\lambda x - \mu y}] \delta_t(dx) P\{X_t \in dy\} \frac{dt}{t}.$$

Hence we see that the measures $P\{(Y_1(t), Y_2(t)) \in (dx, dy)\}$ and $\delta_t(dx) P\{X_t \in dy\}$ play the same role in the representation of $\log\{-\log h(\lambda, \mu)\}$. This does not mean, of course, that the measures are the same on \mathbb{R}_+^2 ; it actually is the continuous time analogue of the fact that on $(0, \infty)$

$$\sum_{m=1}^\infty \frac{1}{m} C^{(m)}(dx) = \sum_{m=1}^\infty \frac{1}{m} P[S_m \in dx]$$

where $C(dx) = P[S_N \in dx]$.

The form of $\log\{-\log h(\lambda, \mu)\}$ in the theorem is the form of the characteristic exponent of the time-space maximal process of any Lévy process X_t found by Fristedt [4]. In the present setting, X_t is the limiting stable process associated with the random walk S_n .

Corollary 3.3. *Suppose S_1 is as in Theorem 3.2. Then (N, S_N) is in the domain of attraction of a stable law on \mathbb{R}_+^2 with parameters $(\beta, \alpha\beta)$ where $\beta = \frac{1}{2} + \frac{\delta}{2\alpha}$ and with δ defined by*

$$p - q = \left(\tan \frac{\pi \delta}{2} \right) \left/ \left(\tan \frac{\alpha \pi}{2} \right) \right.$$

Proof. The value for $\beta = P[X_1 > 0]$ is well known. See for example [2, 6]. By (ii) of Lemma 1.2

$$G_N(x) + \log R(x) \rightarrow \beta \gamma$$

where $R(x) = \{xL(x)\}^{-\beta}$ for some s.v. L . In the theorem we identified the sequence d_n by

$$G_N(n) - \log d_n \rightarrow \beta \gamma.$$

Hence by (ii) of Lemma 1.2 $R(n) \sim 1/d_n$ or $nL(n) \sim d_n^\beta$ or $n \sim d_n^{1/\beta} L^*(d_n^{1/\beta})$ where L^* is the conjugate of L [13]. However we defined m by $m = [d_n]$; hence we obtain similarly that $n \sim m^{1/\beta} L^*(m^{1/\beta})$.

On the other hand S_1 also is in a domain of attraction of a stable law with parameter α ; hence for some s.v. L_0

$$c_n = n^{1/\alpha} L_0(n)$$

as $n \rightarrow \infty$. Hence in (18)

$$a'_m = n \quad \text{or} \quad a'_m \sim m^{1/\beta} L^*(m^{1/\beta})$$

and

$$b'_m = c_n \quad \text{or} \quad b'_m \sim m^{1/(\alpha\beta)} L_1(m)$$

for some s.v. L_1 [13, p. 18].

Reversing the argument, used above for N , we find that S_N is in the domain of attraction of a stable law with index $\alpha\beta$. \square

The above corollary unifies some results of Rogosin [12]. Since $|\delta| \leq 1 - |\alpha\beta|$, $0 \leq \beta \leq 1$ if $0 < \alpha < 1$ and $1 - \frac{1}{\alpha} \leq \beta \leq \frac{1}{\alpha}$ if $1 < \alpha \leq 2$. Hence $\alpha\beta \leq 1$. If $\alpha = 2$, S_1 is in the domain of attraction of the normal law and the sample paths are continuous; then $\beta = \frac{1}{2}$ and $\alpha\beta = 1$. If $\alpha < 2$ and $\alpha\beta = 1$ then S_1 is in the domain of attraction of a spectrally negative stable process. If $\alpha(1-\beta) = 1$ then the index of S_N is $\alpha\beta = \alpha - 1$, a separate result of Rogosin [12].

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