On the domain of attraction of an operator between supremum and sum

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Summary. Between the operations which produce partial maxima and partial sums of a sequence $Y_1, Y_2, ...$, lies the inductive operation: $X_n = X_{n-1} \lor (\alpha X_{n-1} + Y_n), n \ge 1$, for $0 < \alpha < 1$. If the Y_n are independent random variables with common distribution F, we show that the limiting behavior of normed sequences formed from $\{X_n, n \ge 1\}$, is, for $0 < \alpha < 1$, parallel to the extreme value case $\alpha = 0$. For $F \in D(\Phi_{\gamma})$ we give a full proof of the convergence, whereas for $F \in D(\Psi_{\gamma}) \cup D(\Lambda)$, we only succeeded in proving tightness of the involved sequence. The process X_n is interesting for some applied probability models.

1 Introduction

Let $Y_1, Y_2, ...$, be a sequence of independent random variables with distribution F. We define

(1.1) $X_0 := x_0 \in \mathbb{R}$ $X_n := \max \{ X_{n-1}, \alpha X_{n-1} + Y_n \}, \quad n \ge 1,$

where α is a parameter with range [0, 1). In this paper we study the existence of norming constants $a_n(\alpha)$ and $b_n(\alpha)$ such that

$$\{X_n-b_n(\alpha)\}/a_n(\alpha)$$

converges in distribution to a non-degenerate limit.

For $\alpha = 0$,

$$X_n = \max\{x_0, Y_1, \ldots, Y_n\}$$

In this case the limit laws are well known and the domain of attraction problem has been solved by Gnedenko [3] and by de Haan [6]. For $\alpha = 1$, a value *not* included in this study,

$$X_n = x_0 + \sum_{j=1}^n Y_j \mathbf{1}_{[0,\infty)}(Y_j)$$

in which case the limits are stable laws. A complete description of the domain of attraction problem for $\alpha = 1$ can be found in Gnedenko and Kolmogorov [4].

Instead of studying the sequence X_n in (1.1) directly it is more useful to study a stochastic process $X_n(\cdot)$ which we introduce below. We denote by D(G)the domain of attraction of the distribution G, where G is one of the extreme value distributions. For $F \in D(G)$ and $a_n > 0$, $b_n \in \mathbb{R}$ such that $F^n(a_n x + b_n) \to G(x)$, for all x, we define for $n \ge 1$,

(1.2)
$$Y_{n,j} := (Y_j - b_n)/a_n, \quad j = 1, 2, ...,$$

Then $X_n(\cdot) \in D[0, \infty)$ is defined by

(1.3)
$$X_n(t) := a_n^{-1} (x_0 - b_n/(1 - \alpha)), \quad 0 \le t < n^{-1},$$

 $X_n(t) := \max\left\{ X_n\left(\frac{j-1}{n}\right), \alpha X_n\left(\frac{j-1}{n}\right) + Y_{nj} \right\}, \quad \frac{j}{n} \le t < \frac{j+1}{n}, \quad j = 1, 2, ...,$

where $x_0 \in \mathbb{R}$. The process $X_n(\cdot)$ is called the α -sun process with input sequence $Y_{nj}, j=1, 2, ...,$ and initial value x_0 . The algebraic relation between X_n and $X_n(\cdot)$ is obviously,

(1.4)
$$X_n(j/n) = a_n^{-1} (X_j - b_n/(1-\alpha)).$$

In Sect. 2 we show that for $F \in D(\Phi_{\gamma})$, where $\Phi_{\gamma}(x) = \exp(-x^{-\gamma}) \mathbf{1}_{[0,\infty)}(x)$, $\gamma > 0$, and for all initial values x_0 the α -sun processes $X_n(\cdot)$ converge weakly on $D[0,\infty)$. In Sect. 3 we formulate the results we obtained for $F \in D(\Psi_{\gamma})$, where $\Psi_{\gamma}(x) = \exp(-(-x)^{\gamma}) \mathbf{1}_{(-\infty,0]}(x) + \mathbf{1}_{(0,\infty)}(x)$, $\gamma > 0$, and $F \in D(\Lambda)$, where $\Lambda(x) = \exp(e^{-x})$. Basically the only thing we are able to show in these cases is tightness. Because the proofs are long and technical, and the results unsatisfactory, they are omitted. Details of these proofs can be obtained on request through the second author.

The interest in sequences of the type (1.1) originates from a storage problem for solar energy described by Haslett [7]. Additional studies are Daley and Haslett [1], Haslett [8], Hooghiemstra and Keane [9], Hooghiemstra and Scheffer [10] and Greenwood and Hooghiemstra [5]. In these studies the energy contents $X_n^{(\beta)}$ of a storage tank is analyzed, where $X_n^{(\beta)}$ satisfies the difference equation

(1.5)
$$X_n^{(\beta)} = \max \{\beta X_{n-1}^{(\beta)}, \alpha \beta X_{n-1}^{(\beta)} + Y_n\}, \quad \beta \in (0, 1).$$

In [10] the limit behavior of $X_{\infty}^{(\beta)}$, the stationary solution of (1.5), for $\beta \to 1$ and $F(x)=1-x^{-\gamma}(x>1)$ was obtained. It turns out that the limits obtained in this paper for $F \in D(\Phi_{\gamma})$ are of the same type as those in [10]. Intuitively this is because the order of the limit procedure has been interchanged, which apparently is allowed. The question whether this can be proved is not easy to answer and depends on the speed of convergence of $X_n^{(\beta)}$ to the stationary limit $X_{\infty}^{(\beta)}$. This matter will not be treated here.

The study of iterative sequences of the form (1.1) is also of some theoretical interest. The behavior of X_n is somewhere between sups and sums. In fact they turn out to be close to suprema of i.i.d. random variables. The limit processes are new examples of self-similar Markov processes. In a forthcoming paper we study the situation where α depends on n i.e., $\alpha_n = 1 - n^{-l}$, l > 0. For 0 < l < 1 sequences X_n satisfying (1.1) with α replaced by α_n behave similar to the sum of the n^l -largest order statistics.

2 Convergence for case I

We introduce an operator that enables us to write the α -sun process $X_n(\cdot)$, defined in (1.3) as a function of point processes.

Definition 1 For a countable collection $\xi = \{(x_j, y_j)\}_{j=1}^{\infty}$ of pairs of real numbers such that

(i) 0 < x₁ < x₂ < ...,
 (ii) card {x_j: x_j≤s} < ∞ for each positive s;

and for an arbitrary initial value $u \in \mathbb{R}$, we define the element $T_u(\xi)(\cdot) \in D[0, \infty)$ by

(2.1)
$$T_u(\xi)(t) = u, \quad x_0 = 0 \le t < x_1,$$

 $T_u(\xi)(t) = \max\{T_u(\xi)(x_{j-1}), \alpha T_u(\xi)(x_{j-1}) + y_j\}, \quad x_j \le t < x_{j+1}, \quad j \ge 1.$

The operator T_u will be called the α -sun operator. If u=0 we write T instead of T_0 ; occasionally we abbreviate T_{u_n} by T_n . The element $X_n(\cdot)$ defined in (1.3) can be expressed as

$$X_n(\cdot) = T_{u_n}(\xi_n)(\cdot),$$

where $u_n = a_n^{-1}(x_0 - b_n/(1 - \alpha))$ and $\xi_n = \{(j/n, Y_{nj})\}_{j=1}^{\infty}$ with Y_{nj} defined in (1.2). The following simple property of α -sun operators is used repeatedly.

Lemma 1 Let $\xi^{(1)} = \{(x_j, y_j^{(1)})\}_{j=1}^{\infty}$ and $\xi^{(2)} = \{x_j, y_j^{(2)}\}_{j=1}^{\infty}$ be two collections satisfying (i) and (ii) of Definition 1 and having identical x-values. Suppose that for some $j_0, y_j^{(1)} = y_j^{(2)}$ for all $j \ge j_0$. If $T_i, i = 1, 2$, denotes the α -sun operator applied to $\xi^{(i)}$ and if $T_1(x_{j_0}) - T_2(x_{j_0}) = c > 0$, then for all $t \ge x_{j_0}$,

$$0 \leq T_1(t) - T_2(t) \leq c,$$

i.e., the paths only become closer but do not cross.

Proof. The result is obviously independent of the initial values of T_i . The proof goes by induction. Suppose that for some $j \ge j_0$,

$$0 \leq T_1(x_j) - T_2(x_j) \leq c.$$

If $y_{j+1}^{(1)} > (1-\alpha) T_1(x_j)$ then $y_{j+1}^{(2)} = y_{j+1}^{(1)} > (1-\alpha) T_2(x_j)$, both T_1 and T_2 jump at x_{j+1} and paths move closer by a factor α . If

$$(1-\alpha) T_1(x_j) \ge y_{j+1}^{(1)} = y_{j+1}^{(2)} > (1-\alpha) T_2(x_j),$$

then

$$\alpha T_2(x_j) + y_{j+1}^{(2)} < \alpha T_1(x_j) + y_{j+1}^{(1)} \le T_1(x_j),$$

so T_2 moves closer but does not cross. Finally if $y_{j+1}^{(1)} = y_{j+1}^{(2)} \leq (1-\alpha) T_2(x_j)$, then neither T_1 nor T_2 has a jump at x_{j+1} .

Theorem 1 For $F \in D(\Phi_{\gamma})$ the α -sun process $X_n(\cdot)$ with arbitrary initial value $x_0 \in \mathbb{R}$, converges weakly in $D[0, \infty)$ to a proper limit which will be denoted by $Z(\cdot)$.

Proof. Assume $x_0 = 0$, the case $x_0 \neq 0$ will be treated at the end. Put $b_n = 0$ and $a_n = \inf\{y: 1 - F(y) \le n^{-1}\}$, then $X_n(\cdot)$ is defined without ambiguity and $X_n(0) = 0$. By hypothesis

(2.2)
$$n(1-F(a_n x)) \to \mu(x, \infty), \quad x > 0,$$

where $\mu(x, \infty) := x^{-\gamma}$. Fix s > 0 and denote by ξ_n the collection $(j/n, Y_j/a_n), j = 1, 2, ..., [ns]$, and by $\xi_n(\delta), \delta > 0$, the collection of points $(j/n, Y_j/a_n) \in \xi_n$ with $Y_j \ge a_n \delta$. Further we denote by $\xi(\delta)$ the points (or better the support) of a Poisson random measure, with intensity $dt \times d\mu$, restricted to the set $[0, s] \times [\delta, \infty)$. Since $\mu(\delta, \infty) = \delta^{-\gamma} < \infty$ the number of points in $\xi(\delta)$ is a.s. finite. The point process (random measure) which gives unit measure to each of the points $(j/n, Y_j(\omega)/a_n) \in \xi_n(\delta)$ will also be notated by $\xi_n(\delta)$. According to Proposition 3.1 of Resnick [12] the convergence (2.2) implies that $\xi_n(\delta)$ converges vaguely to $\xi(\delta)$ on the space of point measures defined on $[0, s] \times [\delta, \infty)$. Since the map $(x, y) \to x \lor (\alpha x + y)$ from $\mathbb{R}^2 \to \mathbb{R}$ is continuous we conclude that for fixed $\delta > 0$,

(2.3)
$$T(\xi_n(\delta))(\cdot) \xrightarrow{d} T(\xi(\delta))(\cdot) \quad \text{on } D[0, s].$$

We will show that

- (a) $T(\xi(\delta))(\cdot)$ has an almost sure limit $Z(\cdot)$ as $\delta \to 0$, which is proper;
- (b) $\sup_{0 \le t \le s} |X_n(t) T(\xi_n(\delta))(t)| \le 2\delta/(1-\alpha).$

From (b), (a) and (2.3) it follows that $X_n(\cdot) \xrightarrow{d} Z(\cdot)$ on D[0, s], for each positive s.

Proof of (b). We have $X_n(t) = T(\xi_n)(t) \ge T(\xi_n(\delta))(t)$ for all $t \in [0, s]$, since ξ_n contains more points than $\xi_n(\delta)$. The points $(j/n, Y_j/a_n) \in \xi_n \setminus \xi_n(\delta)$ are discarded as soon as $X_n(t) > \delta/(1-\alpha)$. Hence we may assume that $X_n(\cdot)$ and $T(\xi_n(\delta))(\cdot)$ see the same input for $t > \tau(\omega) := \min \{t: X_n(t) > \delta/(1-\alpha)\}$.

For $t < \tau$, $0 \le X_n(t) - T(\xi_n(\delta))(t) \le \delta/(1-\alpha)$. For $t = \tau$, $X_n(\tau) = \alpha X_n(\tau) + Y_{nj}$, where $j/n = \tau(\omega)$. Either $Y_{nj} > \delta$ in which case $X_n(\tau) - T_n(\xi(\delta))(\tau) \le \alpha \delta/(1-\alpha)$, or $Y_{nj} \le \delta$ in which case $X_n(\tau) - T(\xi_n(\delta))(\tau) \le \delta/(1-\alpha) + \delta \le 2\delta/(1-\alpha)$. Now we apply Lemma 1 to obtain (b). Proof of (a). If $0 < \delta' < \delta$, then $T(\xi(\delta'))(\cdot)$ sees more input than $T(\xi(\delta))(\cdot)$, namely points of ξ with second coordinates in $[\delta', \delta)$; hence $T(\xi(\delta'))(t) \ge T(\xi(\delta))(t)$ for all $t \in [0, s]$. Because of monotonicity

$$Z = \lim_{\delta \downarrow 0} T(\xi(\delta)),$$

exists almost sure. We know that $T(\xi(\delta))(t)$ is proper, because $T(\xi(\delta))(t)$ operates on only finitely many points. For $\delta' < \delta$ we have by the argument in the proof of (b)

 $0 \leq T(\xi(\delta'))(t) - T(\xi(\delta))(t) \leq 2\delta/(1-\alpha),$

so that

$$\sup_{\delta' < \delta} T(\xi(\delta'))(t) \leq T(\xi(\delta))(t) + 2\,\delta/(1-\alpha).$$

Finally we turn to the case where $x_0 \neq 0$. Then $X_n(\cdot) = T_{u_n}(\xi_n)(\cdot)$, and $u_n = x_0/a_n \rightarrow 0$. $X_n(\cdot)$ and $T(\xi_n)(\cdot)$ have the same input so that according to Lemma 1,

$$0 \leq \sup_{t \leq s} |T_{u_n}(\xi_n)(t) - T(\xi_n)(t)| \leq u_n.$$

Hence $T_{u_n}(\xi_n)(\cdot)$ will also converge to $Z(\cdot)$ on D[0, s].

It is seen from the proof of Theorem 1 that the process Z(s), $s \ge t$, given Z(t) = z > 0, can be defined as an α -sun process with input the Poisson process ξ restricted to $[t, \infty) \times [(1-\alpha)z, \infty)$. Moreover, as Z(t) is the weak limit of $X_{[nt]}/a_{[nt]}$ and $\lim_{n \to \infty} a_{[nt]}/a_n = t^{1/\gamma}$, it follows that $Z(\cdot)$ is self-similar with index $H = \gamma^{-1}$. Self-similar Markov processes have been extensively studied by Lamperti [11]. We focus on the distribution of Z(1).

Theorem 2 For $F \in D(\Phi_{\gamma})$, $\gamma > 0$, the distribution of the limit random variable Z(1) has a density h_{α} concentrated on $(0, \infty)$. This density is the unique density solution of the equation

(2.4)
$$h_{\alpha}(x) = \gamma x^{-1} \int_{0}^{x} (x - \alpha u)^{-\gamma} h_{\alpha}(u) \, \mathrm{d} u, \quad x > 0.$$

In order to prove Theorem 2 we use the self-similarity together with the following relation between the law of $Z(\cdot)$ and the measure $\mu(x, \infty) = x^{-\gamma}, x > 0$.

Lemma 2 The law of $Z(\cdot)$ satisfies for t > 0 and x > 0,

(2.5)
$$P\{Z(t) > x\} = \int_{0}^{t} \int_{0}^{x} (x - \alpha u)^{-\gamma} P\{Z(s) \in du\} ds.$$

Proof. From Definition (1.3) with $x_0 = 0$ we have for x > 0,

$$P\left\{X_n(j/n) \leq x\right\} - P\left\{X_n\left(\frac{j-1}{n}\right) \leq x\right\}$$
$$= -\int_0^x \left\{1 - F(a_n(x-\alpha u))\right\} P\left\{X_n\left(\frac{j-1}{n}\right) \in \mathrm{d}u\right\},$$

and iteration of this formula yields

$$P\left\{X_n(j/n) \leq x\right\} - P\left\{X_n\left(\frac{j-k}{n}\right) \leq x\right\}$$
$$= -\sum_{m=1}^k \int_0^x \left\{1 - F(a_n(x-\alpha u))\right\} P\left\{X_n\left(\frac{j-m}{n}\right) \in \mathrm{d}u\right\}.$$

Fix t > 0 and put k = j = [nt], then for x > 0,

$$(2.6) \qquad P\{X_{n}(t) > x\} = P\{X_{n}\left(\frac{j}{n}\right) > x\} = P\{X_{n}\left(\frac{j}{n}\right) > x\} - P\{X_{n}(0) > x\} \\ = \sum_{m=1}^{[nt]} \int_{0}^{x} \{1 - F(a_{n}(x - \alpha u))\} P\{X_{n}\left(\frac{[nt] - m}{n}\right) \in du\} \\ = \int_{s=0}^{s = \frac{[nt] - 1}{n}} \int_{0}^{x} n\{1 - F(a_{n}(x - \alpha u))\} P\{X_{n}(s) \in du\}.$$

The final equality in (2.6) follows since $X_n(s)$ is a stepfunction.

Now fix $x \in \text{Cont}(Z(t))$, the set of continuity points of the distribution of Z(t). The left-hand side of (2.6) converges to $P\{Z(t) > x\}$, so we focus on the right-hand side. Note that for $u \in [0, x]$,

$$\lim_{n\to\infty}n\left\{1-F(a_n(x-\alpha u))\right\}=(x-\alpha u)^{-\gamma},$$

and by Karamata's theory this convergence is uniform for $u \in [0, x]$ (cf. [2], Theorem 1.3). Hence for each $\varepsilon > 0$ and *n* sufficiently large

(2.7)
$$\int_{0}^{\frac{\lfloor nt \rfloor - 1}{n}} \int_{0}^{x} |n\{1 - F(a_n(x - \alpha u))\} - (x - \alpha u)^{-\gamma}| P\{X_n(s) \in du\} ds$$
$$\leq \varepsilon \int_{0}^{\frac{\lfloor nt \rfloor - 1}{n}} \int_{0}^{x} P\{X_n(s) \in du\} ds \leq \varepsilon t.$$

For $s \in (0, t)$ we define

$$f_n(s) := \int_0^x (x - \alpha u)^{-\gamma} P\{X_n(s) \in \mathrm{d} u\},$$
$$f(s) := \int_0^x (x - \alpha u)^{-\gamma} P\{Z(s) \in \mathrm{d} u\}.$$

For $x \in \text{Cont}(Z(s))$ we have $f_n(s) \to f(s)$ (this is standard weak convergence theory and follows since $(x - \alpha u)^{-\gamma} \mathbb{1}_{[0, x]}(u)$ is bounded and is only discontinuous at u = x). Furthermore

$$|f_n(s)| = \int_0^x (x - \alpha u)^{-\gamma} P\{X_n(s) \in du\} \leq x^{-\gamma} (1 - \alpha)^{-\gamma},$$

and so

$$\int_{0}^{t} |f_{n}(s)| \, \mathrm{d} s \leq \int_{0}^{t} x^{-\gamma} (1-\alpha)^{-\gamma} \, \mathrm{d} s = t \, x^{-\gamma} (1-\alpha)^{-\gamma} < \infty.$$

By Lebesgues dominated convergence theorem $\int_{0}^{t} f_n(s) ds \rightarrow \int_{0}^{t} f(s) ds$ if we can prove that $f_n(s) \rightarrow f(s)$ Lebesgue – almost everywhere in (0, t). The latter statement is true because by self-similarity, the set,

 $\{s \in (0, t): \text{ the distribution function of } Z(s) \text{ is discontinuous at } x\}$

has the same cardinality as the set of all discontinuities of the distribution function of Z(1), which set evidently is countable. We conclude that

(2.8)
$$\lim_{n \to \infty} \int_{0}^{t} \int_{0}^{x} (x - \alpha u)^{-\gamma} P\{X_{n}(s) \in du\} ds = \int_{0}^{t} \int_{0}^{x} (x - \alpha u)^{-\gamma} P\{Z(s) \in du\} ds.$$

Finally

(2.9)
$$\left|\int_{\frac{[nt]-1}{n}}^{t}\int_{0}^{x}(x-\alpha u)^{-\gamma}P\left\{X_{n}(s)\in du\right\}ds\right|\leq 2x^{-\gamma}(1-\alpha)^{-\gamma}/n\to 0.$$

Combining (2.7), (2.8) and (2.9) it follows that the right-hand side of (2.6) converges to

$$\int_{0}^{t} f(s) \, \mathrm{d}s = \int_{0}^{t} \int_{0}^{x} (x - \alpha u)^{-\gamma} P\{Z(s) \in \mathrm{d}u\} \, \mathrm{d}s. \quad \Box$$

Proof of Theorem 2. By self-similarity we have for each t > 0,

(2.10)
$$P\{Z(t) > x\} = P\{Z(1) > xt^{-1/\gamma}\}.$$

Using (2.10) we can write the integral equation (2.5) entirely in terms of the distribution of Z(1),

$$P\{Z(1) > xt^{-1/\gamma}\} = \int_{0}^{t} \int_{0}^{x} (x - \alpha u)^{-\gamma} P\{Z(1) \in s^{-1/\gamma} du\} ds.$$

Next differentiate with respect to t and set t = 1 afterwards to obtain (2.4). Finally, Hooghiemstra and Scheffer [10] showed that the integral equation

$$g_{\alpha}(x) = x^{-1} \int_{0}^{x} (x - \alpha u)^{-\gamma} g_{\alpha}(u) du,$$

has a unique density solution. From this we obtain the remaining conclusion of Theorem 2, since if X denotes an absolutely continuous random variable with density g_{α} , then $Y := \gamma^{1/\gamma} X$ is absolutely continuous with density h_{α} , satisfying (2.4). \Box

3 Further results

As mentioned in the introduction the results for $F \in D(\Psi_{\gamma})$ and $F \in D(\Lambda)$ are incomplete. In this section we formulate tightness results for the two indicated domains of attraction and convergence results for some special distributions F. We close the section with an elegant and short proof of thightness when F is the uniform distribution on [-1, 0]. We denote by r the right end point of F, i.e., $r = \sup \{x: F(x) < 1\}$.

Theorem 3 Let $F \in D(\Psi_{\gamma})$, i.e., $r < \infty$ and $1 - F(r - x^{-1}) = x^{-\gamma}L(x)$ for some positive γ and some slowly varying function L. Then there exist sequences $a_n > 0$ and $b_n \in \mathbb{R}$ such that for all initial values $x_0 < r/(1-\alpha)$ the sequence $a_n^{-1}(X_n - b_n/(1-\alpha))$ is tight on $(-\infty, 0)$. A possible choice of $\{a_n\}$ and $\{b_n\}$ is

$$a_n = r - \inf\{x: 1 - F(x) \leq n^{-1}\}, \quad b_n \equiv r. \quad \Box$$

Theorem 4 If $F \in D(A)$, then there exist sequences $\{a_n\}$ and $\{b_n\}$ such that for all initial values $x_0 < r/(1-\alpha)$ the sequence $a_n^{-1}(X_n - b_n/(1-\alpha))$ is tight on \mathbb{R} . A possible choice of $\{a_n\}$ and $\{b_n\}$ is

$$a_n = U^*(ne) - U^*(n), \quad b_n = U^*(n),$$

where for x > 1,

$$U^*(x) = \inf\{y: (1 - F(y))^{-1} \ge x\}. \quad \Box$$

Theorem 5 Let $Y_1, Y_2, ..., be$ independent with common distribution F(x)=1 $-|x|^{\gamma}, -1 \le x \le 0, F(x)=0, x < -1$ and F(x)=1, x > 0. Let $a_n = n^{-1/\gamma}, b_n = 0$. For each $x_0 \in [-1, 0)$ the α -sun processes $X_n(\cdot)$, with initial value $x_0 n^{1/\gamma}$ converge weakly in $D(0, \infty)$ to a proper limit $Z(\cdot)$. \Box

Theorem 6 Let $Y_1, Y_2, ..., be independent with common distribution F, given by <math>F(x)=1-e^{-x}$ on $[0,\infty)$, and let $a_n=1$, $b_n=\log n$. For each $x_0 \ge 0$ the α -sun processes $X_n(\cdot)$ with initial value $x_0 - \log n/(1-\alpha)$ converge weakly in $D(0,\infty)$ to a proper limit $Z(\cdot)$. \Box

Theorem 7 Suppose $Z(\cdot)$ is one of the limit processes of Theorem 5 or 6. (i) For $F(x)=1-|x|^{\gamma}$, on [-1,0], the process $Z(\cdot)$ is a self-similar Markov process with index $H=-\gamma^{-1}$. The law of Z(t) satisfies

(3.1)
$$P\{Z(t) > x\} = \int_{0}^{t} \int_{x/\alpha}^{x} |x - \alpha u|^{\gamma} P\{Z(s) \in du\} ds,$$

for t > 0 and x < 0. In particular we obtain from (3.1) that the law of Z is independent of $x_0 \in [-1, 0)$ and that the marginal distribution of Z(1) admits a density on $(-\infty, 0)$, denoted by h_{α} , where h_{α} is the unique density solution of

(3.2)
$$h_{\alpha}(x) = \gamma |x|^{-1} \int_{x/\alpha}^{x} |x - \alpha u|^{\gamma} h_{\alpha}(u) \, \mathrm{d} u, \quad x < 0.$$

(ii) For $F(x) = 1 - e^{-x}$, x > 0, the process $\exp\{Z(t)\}$, t > 0, is a self-similar Markov process with index $H = (1 - \alpha)^{-1}$. The law of Z(t) satisfies

(3.3)
$$P\{Z(t) > x\} = \int_{0}^{t} \int_{-\infty}^{x} e^{-(x-\alpha u)} P\{Z(s) \in du\} ds,$$

for t>0 and $x \in \mathbb{R}$. In particular we obtain from (3.3) that the law of Z is independent of $x_0 \in [0, \infty)$ and that the marginal distribution of Z(1) has density,

(3.4)
$$h_{\alpha}(x) = (1-\alpha)(\Gamma((1-\alpha)^{-1}))^{-1} \exp\{-x - e^{-x(1-\alpha)}\}, x \in \mathbb{R},$$

where $\Gamma(t) := \int_{0}^{\infty} x^{t-1} e^{-x} dx, t > 0.$

The following lemma is the basis for the proof of Theorem 3 and 4. It shows tightness for the special case where Y_1 is uniformly distributed on [-1, 0]. The proof of the lemma originates from Tom Liggett (private communication).

Lemma 3 If $X_1 = x_1 \in [-1, 0)$, $X_{n+1} = \max\{X_n, \alpha X_n + Y_{n+1}\}$, $n \ge 1$, where Y_1, Y_2, \ldots is an i.i.d. sequence with Y_1 uniformly distributed on [-1, 0], then

$$\inf_{n\geq 1} En X_n \geq -2(1-\alpha)^{-2}.$$

Proof. $E[X_n|X_{n-1}] = X_{n-1} + \frac{1}{2}(1-\alpha)^2 X_{n-1}^2$, hence taking double expectations

(3.5)
$$EX_n = EX_{n-1} + \frac{1}{2}(1-\alpha)^2 EX_{n-1}^2 \ge EX_{n-1} + \frac{1}{2}(1-\alpha)^2 (EX_{n-1})^2$$

Put $g(u) := u + \frac{1}{2}(1-\alpha)^2 u^2$, $u \in [-1, 0]$. Then $g'(u) = 1 + (1-\alpha)^2 u > 0$ on [-1, 0]and hence g is increasing on [-1, 0]. Set $u_n = EX_n$ and $v_n = \frac{-2(1-\alpha)^{-2}}{n}$.

We shall prove by induction that $u_n \ge v_n$ for all $n \ge 1$. The statement is trivially true for n=1, so assume $u_{n-1} \ge v_{n-1}$. Then from (3.5) and the monotonicity of g,

$$u_n \geq g(u_{n-1}) \geq g(v_{n-1}).$$

A simple calculation yields $g(v_{n-1}) \ge v_n$, and this completes the proof. \Box

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