

## STABILITY OF PERPETUITIES

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For a series of randomly discounted terms we give an integral criterion to distinguish between almost-sure absolute convergence and divergence in probability to  $\infty$ , these being the only possible forms of asymptotic behavior. This solves the existence problem for a one-dimensional perpetuity that remains from a 1979 study by Vervaat, and yields a complete characterization of the existence of distributional fixed points of a random affine map in dimension one.

**1. Introduction.** Let  $\Psi$  be a random affine map from  $\mathbb{R}$  to itself, given by

$$\Psi(t, \omega) := Q(\omega) + M(\omega)t, \quad \omega \in \Omega, t \in \mathbb{R},$$

where  $Q$  and  $M$  are r.v.s (random variables) on a probability space  $(\Omega, \mathcal{A}, P)$ . Of course we normally omit the argument  $\omega$  and write just  $\Psi(t) := Q + Mt$ . A (distributional) *fixed point* of  $\Psi$  is a probability law on  $\mathbb{R}$ , of a r.v.  $R$ , say, such that

$$(1.1) \quad R \stackrel{\text{L}}{=} Q + MR, \quad R \text{ independent of } (Q, M),$$

where  $\stackrel{\text{L}}{=}$  denotes equality of probability laws. In Section 3 of this paper we solve the problem of finding necessary and sufficient conditions for existence of a fixed point.

It is natural to attempt to approach a fixed point by iteration, and for that purpose we suppose the existence on the probability space of a sequence of random affine maps  $(\Psi_n)_{n \in \mathbb{N}}$ , mutually independent and identically distributed with  $\Psi$ . Thus for each  $n$ ,  $\Psi_n(t) := Q_n + M_n t$  for all  $t \in \mathbb{R}$ , and

$$\begin{pmatrix} Q \\ M \end{pmatrix}, \begin{pmatrix} Q_1 \\ M_1 \end{pmatrix}, \begin{pmatrix} Q_2 \\ M_2 \end{pmatrix}, \dots$$

are to be i.i.d. (independent, identically distributed) random elements of  $\mathbb{R}^2$ .

*Outer* iteration starts from a r.v.  $R_0$  independent of the sequence  $(\Psi_n)$  and often taken to be a nonrandom constant  $r$ , or just 0, and forms successively  $R_1, R_2, \dots$  by

$$(1.2) \quad R_{n+1} := \Psi_{n+1}(R_n) = Q_{n+1} + M_{n+1}R_n, \quad n = 0, 1, 2, \dots$$

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We write  $R_n$  as  $R_n(R_0)$  to bring out the dependence on  $R_0$ . Then

$$R_n(R_0) = \sum_{i=1}^n Q_i \prod_{j=i+1}^n M_j + R_0 \prod_{j=1}^n M_j.$$

We have that  $(R_n(R_0))_{n=0,1,\dots}$  is a Markov sequence, and we may ask about its recurrence, transience, etc. Our results lead to a complete characterization of its positive recurrence.

*Inner* iteration, by contrast, starts from an affine map  $Z_0(t) := t$  for all  $t$ , and forms successively the random affine maps  $Z_1(\cdot), Z_2(\cdot), \dots$  by

$$Z_{n+1}(t) := Z_n \circ \Psi_{n+1}(t) = Z_n(Q_{n+1} + M_{n+1}t), \quad n = 0, 1, \dots$$

We then have  $Z_n(t) = \sum_{k=1}^n \Pi_{k-1} Q_k + \Pi_n t$ , where throughout the paper we write

$$(1.3) \quad \Pi_n := \begin{cases} \prod_{k=1}^n M_k, & n = 1, 2, \dots, \\ 1, & n = 0. \end{cases}$$

Often we are interested just in

$$Z_n := Z_n(0) = \sum_{k=1}^n \Pi_{k-1} Q_k, \quad n = 0, 1, \dots$$

However it is useful also to allow for a randomized initial value, as with outer iteration discussed above, so we replace  $t$  in  $Z_n(t)$  by a r.v.  $Z_0$  independent of the sequence  $(\Psi_n)$ , and then have

$$(1.4) \quad Z_n(Z_0) = \sum_{k=1}^n \Pi_{k-1} Q_k + \Pi_n Z_0, \quad n = 0, 1, \dots$$

The sequence  $(Z_n)$  is not Markov but we have the possibility of its converging in some weak or strong sense, the candidate limit being a r.v.,

$$(1.5) \quad Z_\infty := \sum_{k=1}^\infty \Pi_{k-1} Q_k.$$

Now  $Z_\infty$ , if it exists, is the probabilistic formulation of the actuarial notion of a *perpetuity*, which represents the present value of a permanent commitment to make a payment at regular intervals, say annually, into the future forever. The  $Q_n$  represent annual payments, the  $M_n$  cumulative discount factors. Both are subject to random fluctuation. Our model allows the amount  $Q_n$  in a given year and the discount factor  $M_n$  for that year to be dependent, but requires independence between different years.

In our main result, Theorem 2.1, under a side condition that excludes degeneracies, we give a necessary and sufficient condition for  $Z_n(Z_0) \rightarrow Z_\infty$  a.s. and prove that the series for  $Z_\infty$  is a.s. absolutely convergent when the condition holds, whereas  $|Z_n(Z_0)| \xrightarrow{P} \infty$  when it does not.

Our results on the r.v.s  $Z_n(Z_0)$  and  $Z_\infty$  are in Section 2, and consequences for fixed points and positive recurrence are in Section 3. We specialize our

results to certain important particular cases in Section 4, and present our proofs in Section 5.

The current state of knowledge about convergence of iterations and existence and uniqueness of fixed points is largely as left by Vervaat (1979), at least in the one-dimensional case that we treat. On convergence, Vervaat dealt mainly with the case when  $E \log |M|$  exists. His results on fixed points are subsumed in our Theorem 3.1.

Two important later results are in Grincevičius (1980, 1981). In the 1981 paper a proof is sketched that, under the nondegeneracy assumption  $P(Q + Mc = c) < 1$  for all  $c$ , either  $|Z_n| \xrightarrow{P} \infty$  or the series (1.5) converges a.s. The proof is both technically brilliant and not wholly clear in a number of respects. In the course of giving an integral criterion for the a.s. absolute convergence of the series we re-prove Grincevičius's result.

In Grincevičius (1980) the random elements  $(Q_n, M_n)$  are assumed independent but not necessarily identically distributed, and a necessary and sufficient condition is found for the law of  $Z_\infty$  to have an atom. A corollary of this result is that when the  $(Q_n, M_n)$  are i.i.d., as assumed throughout the present paper,  $Z_\infty$  is in general a continuous r.v.

Both before and after the above papers, particular cases have been discussed in various parts of the literature, usually as a prior issue before a proof of some property of fixed points or iteration schemes. The picture up to 1994 is surveyed in Embrechts and Goldie (1994). Questions about recurrence and transience are treated in Kellerer (1992). Extensions of Vervaat's results to the case of ergodic  $(\Psi_n)$  sequences are discussed in Brandt (1986) and, multidimensionally, in Bougerol and Picard (1992). We mention the result in the latter reference that existence of a fixed point [see (1.1)] implies  $\Pi_n \rightarrow 0$  a.s. Stability results for random recursions that are not affine but only approximately so are treated in Letac (1986) Goldie (1991) and Glasserman and Yao (1995). Numerous applications of random affine maps in many diverse fields are given in the above references and in further references therein. We mention here a new application in the analysis of algorithms; see Grübel and Rösler (1996), Goldie and Grübel (1996) and Grübel (1998).

Notation to be used throughout the paper will be, in addition to the  $\Pi_n$  defined in (1.3),

$$(1.6) \quad X := -\log |M|, \quad X_n := -\log |M_n|, \quad Y := \log |Q|, \quad Y_n := \log |Q_n|, \\ n = 1, 2, \dots$$

We allow at this stage the possibility that  $M = 0$ , corresponding to  $X = +\infty$ , with positive probability, though this case turns out to be trivial (see Remark 2.4 below) and also the possibility that  $Q = 0$  with positive probability. The  $X_n$  form the steps of the random walk  $(S_n)$ , where

$$(1.7) \quad S_n := \sum_{k=1}^n X_k = -\log |\Pi_n|, \quad n = 0, 1, 2, \dots$$

We need the following truncated mean of  $X$ :

$$(1.8) \quad A_M(y) := E(X^+ \wedge y) = \int_0^y P(X > x) dx, \quad y > 0.$$

Any sum, maximum or minimum {respectively, product} over an empty range is, by convention,  $0\{1\}$ . Thus  $S_0$  is identically zero. We denote the indicator r.v. of an event  $A$  by  $1_A$  or  $1\{A\}$ . We write  $x^+ := x \vee 0$  and  $x^- := -(x \wedge 0)$ . Convergence in distribution is denoted by  $\rightarrow^L$ .

**2. Convergence and divergence.** We now state our results on the  $Z_n(Z_0)$  defined in (1.4). The main result is Theorem 2.1.

**THEOREM 2.1.** *Suppose  $P(Q = 0) < 1$  and  $P(M = 0) = 0$ . The following are equivalent:*

$$(2.1) \quad \Pi_n \xrightarrow{a.s.} 0 \ (n \rightarrow \infty) \quad \text{and} \quad \int_{(1, \infty)} \left( \frac{\log q}{A_M(\log q)} \right) dP(|Q| \leq q) < \infty,$$

$$(2.2) \quad P(|M| = 1) < 1 \quad \text{and} \quad \sup_{n \in \mathbb{N}} |\Pi_{n-1} Q_n| < \infty \quad a.s.,$$

$$(2.3) \quad \Pi_{n-1} Q_n \xrightarrow{a.s.} 0 \quad (n \rightarrow \infty),$$

$$(2.4) \quad \sum_{n=1}^{\infty} |\Pi_{n-1} Q_n| < \infty \quad a.s.,$$

$$(2.5) \quad \sum_{n \geq 1} P \left( \min_{1 \leq j \leq n-1} |\Pi_j| |Q_n| \geq e^{-x} \right) < \infty \quad \text{for all } x > 0.$$

Each of these implies

$$(2.6) \quad Z_n(Z_0) \xrightarrow{a.s.} Z_\infty, \quad n \rightarrow \infty,$$

where  $Z_\infty$  is given by (1.5), and the series in (1.5) is a.s. absolutely convergent. Conversely, assuming

$$(2.7) \quad P(Q + Mc = c) < 1 \quad \text{for all } c \in \mathbb{R},$$

if (2.1) does not hold [i.e., if  $\Pi_n$  does not converge to 0 a.s. as  $n \rightarrow \infty$  or if the integral in (2.1) diverges] then  $|Z_n(Z_0)| \rightarrow^P \infty$  as  $n \rightarrow \infty$ .

**REMARK 2.2.** Relevant properties for the function  $A_M$  in (1.8) and (2.1) are that it is nondecreasing and concave, with  $A_M(0) = 0$  and  $A_M(\infty) = EX^+ \leq \infty$ . Also  $A_M(y) > 0$  for some (hence all)  $y > 0$  if and only if  $P(X > 0) = P(|M| < 1) > 0$ .

To bring out further properties of the integrand in (2.1) we write

$$\frac{A_M(y)}{y} = E\left(\frac{X^+}{y} \wedge 1\right) = \int_0^1 P(X > yz) dz,$$

which gives that  $A_M(y)/y$  is nonincreasing, with limits  $P(X > 0) = P(|M| < 1)$  at  $0+$ , and  $P(M = 0)$  at  $+\infty$ . Note that if  $\Pi_n \rightarrow 0$  a.s. then clearly we must have  $P(|M| < 1) > 0$ , so the integrand in (2.1) is bounded in the neighborhood of  $q = 1$  when the first part of (2.1) holds. If, on the other hand,  $|M| \geq 1$  a.s., then the first part of (2.1) fails so we do not need to evaluate the integral (in which the integrand is identically  $+\infty$ ). We will take  $A_M(y)/y$  to have the value  $P(|M| < 1)$  at  $0$ , and with that convention we could thus equivalently integrate over  $[1, \infty)$  rather than  $(1, \infty)$  in (2.1).

REMARK 2.3. Equation (2.7) is a *nondegeneracy* condition in the following sense. When  $Q + Mc = c$  a.s. for some real constant  $c$ , we call the random affine map  $\Psi$  a *tied-down line*: it is constrained to pass through the point  $(c, c)$  in  $\mathbb{R}^2$ , and only its slope is random. Observe that when  $Q_k = c - M_k c$ , (1.4) reduces to

$$(2.8) \quad Z_n(Z_0) = c + (Z_0 - c)\Pi_n, \quad n = 1, 2, \dots$$

So when  $Z_0$  is degenerate at  $c$  we have  $Z_n(c) = c$  a.s., whereas otherwise the asymptotic behavior of  $Z_n(Z_0)$  reduces to that of  $\Pi_n$ . Failure of (2.1) does not then imply  $|Z_n(Z_0)| \xrightarrow{P} \infty$ , so (2.7) is needed for the converse assertion in Theorem 2.1.

REMARK 2.4. We rule out the case when  $P(M = 0) > 0$  in Theorem 2.1, which is trivial in the following sense. When it occurs, the integrand in (2.1) is bounded by  $1/P(M = 0)$  (see Remark 2.2), so the integral is finite regardless of  $Q$ . Also, then,  $N := \min\{n: M_n = 0\} < \infty$  a.s. and  $\Pi_n = 0$  for all  $n \geq N$ , hence all of (2.1)–(2.6) hold, where in (2.6),  $Z_n(Z_0) = \sum_{k=1}^N \Pi_{k-1} Q_k$  for all  $n \geq N$ . There is in particular no content in the converse assertion of Theorem 2.1, for (2.1) cannot fail to hold in this case.

We also rule out the case when  $Q = 0$  a.s. in Theorem 2.1. When this occurs, (1.4) gives  $Z_n(Z_0) = \Pi_n Z_0$  a.s., and so  $Z_n(Z_0)$  converges a.s. if and only if  $\Pi_n$  converges a.s., as long as  $Z_0$  is not degenerate at  $0$ . Now  $\Pi_n$  converges a.s. if and only if  $\Pi_n$  converges to  $0$  a.s., since the random walk  $S_n$  can only drift to  $+\infty$ , to  $-\infty$ , or oscillate, and only the first of these corresponds to convergence of  $\Pi_n$  [see (1.72.1)]. Thus the behavior of  $Z_n(Z_0)$  is well understood in this case. The converse in Theorem 2.1, that  $|Z_n(Z_0)| \xrightarrow{P} \infty$  as  $n \rightarrow \infty$  if (2.1) does not hold, is not true when  $Q = 0$  a.s., but this is ruled out by (2.7) (with  $c = 0$ ) in any case.

REMARK 2.5. We noted above the equivalence that, when  $P(M = 0) = 0$ ,  $\Pi_n \rightarrow 0$  a.s. if and only if  $S_n \rightarrow \infty$  a.s. For the property  $S_n \rightarrow \infty$  a.s. we quote a necessary and sufficient condition which comes ultimately from Erickson

(1973) and which we give in the form derived in Kesten and Maller (1996), Lemma 1.1. In terms of

$$(2.9) \quad J_- := \int_{[0, \infty)} \left( \frac{y}{A_M(y)} \right) |dP(X \leq -y)|,$$

we have the following.

**PROPOSITION 2.6.** *When  $P(M = 0) = 0$ , for the assertion  $\Pi_n \rightarrow 0$  a.s. that forms the first half of (2.1), it is necessary and sufficient that*

$$(2.10) \quad J_- < E(X^+) = \infty \quad \text{or} \quad 0 < EX \leq E|X| < \infty.$$

We could even consider the case  $P(M = 0) > 0$  to be covered by  $J_- < E(X^+) = \infty$ , for it corresponds to  $X = -\log|M| = +\infty$  with positive probability, hence  $E(X^+) = \infty$ , and it makes the  $J_-$  integral finite because the integrand is bounded by  $1/P(M = 0)$  (see Remark 2.4). In this Proposition 2.6 remains true when  $P(M = 0) > 0$ , as do the conditions (2.1)–(2.6) of Theorem 2.1.

**REMARK 2.7.** In answer to a question put to us by Rainer Wittmann, we note that neither half of (2.1) is superfluous. Lemma 5.5 will show that finiteness of the integral is needed. On the other hand, Theorem 2.1 shows that  $\Pi_n \rightarrow 0$  a.s. is certainly needed for the other assertions of the theorem to hold, and  $\Pi_n \rightarrow 0$  a.s. is not implied by finiteness of the integral in (2.1), even if  $Q$  has unbounded support, by (2.10).

**REMARK 2.8.** Under the nondegeneracy condition (2.7), Theorem 2.1 establishes that  $Z_n(Z_0)$  has one of two extremely contrasted forms of asymptotic behavior: either  $Z_n(Z_0) \rightarrow Z_\infty$  a.s., or  $|Z_n(Z_0)| \rightarrow^P \infty$ . Consequently, any infringement of one form is enough to establish the other: for instance, if  $P(|Z_n(Z_0)| \leq A)$  is bounded away from 0 for some constant  $A$  along some infinite sequence of  $n$ , then  $Z_n(Z_0)$  converges a.s.

Theorem 2.1 is also remarkable in that the joint distribution of  $Q$  and  $M$  plays no role except in the nondegeneracy condition (2.7): thus (2.1) involves only the marginal laws of  $Q$  and  $M$ , and changing their joint law in a way consistent with these marginals has no effect on the convergence behavior of the  $Z_n(Z_0)$ .

It is even possible to weaken the independence assumed of the  $(Q_k, M_k)$ . So long as the  $M_k$  remain i.i.d. and the  $Q_k$  remain identically distributed, the  $Q_k$  can have arbitrary dependence on the  $M_k$  and each other for all except the converse assertion in Theorem 2.1. This will be clear from the proof.

**REMARK 2.9.** Convergence of the perpetuity (1.5) is unaffected by taking an arbitrary positive power of the  $Q_k$ , and a different arbitrary positive power of the  $\Pi_{k-1}$ . That is, we have the following.

COROLLARY 2.10. Assume (2.1). Then with probability 1,  $\sum_{k=1}^\infty |\Pi_{k-1}|^a \times |\mathbf{Q}_k|^b < \infty$  for each  $a > 0, b > 0$ .

REMARK 2.11. We can replace  $A_M(y)$  in (2.1) by

$$\widehat{A}_M(y) := \int_0^y P(|X| > x) dx = \int_0^y (P(X > x) + P(X < -x)) dx.$$

That is, we have the following.

PROPOSITION 2.12. Suppose  $P(Q = 0) < 1$  and  $P(M = 0) = 0$ . Then (2.1) is equivalent to

$$(2.11) \quad \Pi_n \xrightarrow{a.s.} 0 (n \rightarrow \infty) \quad \text{and} \quad \int_{(1, \infty)} \left( \frac{\log q}{\widehat{A}_M(\log q)} \right) dP(|\mathbf{Q}| \leq q) < \infty.$$

Again we can replace  $(1, \infty)$  by  $[1, \infty)$  in the integral if desired, by taking  $\widehat{A}_M(y)/y$  to have its limiting value  $P(|M| \neq 1)$  at 0, which value is positive if  $\Pi_n \rightarrow 0$  a.s.

REMARK 2.13. The forward part of Theorem 2.1 extends to a multidimensional setting. Consider the case where  $M$  is a  $d \times d$  random matrix and  $\mathbf{Q}$  a random column  $d$ -vector, and  $(\mathbf{Q}, M), (\mathbf{Q}_1, M_1), (\mathbf{Q}_2, M_2), \dots$  are i.i.d. Set  $\Pi_0 := I, \Pi_k := M_1 M_2 \cdots M_k$  for  $k = 1, 2, \dots$ , and

$$\mathbf{Z}_n(\mathbf{Z}_0) := \sum_{k=1}^n \Pi_{k-1} \mathbf{Q}_k + \Pi_n \mathbf{Z}_0, \quad n = 1, 2, \dots,$$

where  $\mathbf{Z}_0$  is independent of the  $(\mathbf{Q}_k, M_k)$ . On  $\mathbb{R}^d$  use the Euclidean norm  $|\cdot|$  and on  $d \times d$  matrices the operator norm  $\|A\| := \sup_{|x|=1} |Ax|$ . Then  $\|AB\| \leq \|A\| \|B\|$  so, for  $m < n$ ,

$$\begin{aligned} |\mathbf{Z}_n(\mathbf{Z}_0) - \mathbf{Z}_m(\mathbf{Z}_0)| &\leq \sum_{k=m+1}^n |\Pi_{k-1} \mathbf{Q}_k| \\ &\leq \sum_{k=m+1}^n \|\Pi_{k-1}\| |\mathbf{Q}_k| \\ &\leq \sum_{k=m+1}^n \left( \prod_{j=1}^{k-1} \|M_j\| \right) |\mathbf{Q}_k|. \end{aligned}$$

Apply Theorem 2.1 to the i.i.d. sequence  $(|\mathbf{Q}_k|, \|M_k\|)$  to get the following, in which  $A_{\|M\|}(y) := \int_0^y P(-\log \|M\| > x) dx$ .

COROLLARY 2.14. Suppose  $P(|\mathbf{Q}| = 0) < 1$  and  $P(\|M\| = 0) = 0$ . If

$$\prod_{k=1}^n \|M_k\| \xrightarrow{a.s.} 0 (n \rightarrow \infty) \quad \text{and} \quad \int_{(1, \infty)} \left( \frac{\log q}{A_{\|M\|}(\log q)} \right) dP(|\mathbf{Q}| \leq q) < \infty,$$

then  $\mathbf{Z}_\infty := \sum_{k=1}^\infty \Pi_{k-1} \mathbf{Q}_k$  is a.s. absolutely convergent in norm, and  $\mathbf{Z}(\mathbf{Z}_0) \rightarrow \mathbf{Z}_\infty$  a.s. as  $n \rightarrow \infty$ .

Proof of a converse looks harder. From, for instance, tightness of  $\{\mathbf{Z}_n\}$  we can immediately deduce  $\{\Pi_{n-1} \mathbf{Q}_n\}$  tight. But to bring in a suitable nondegeneracy condition and deduce  $\Pi_n \rightarrow 0$  a.s. we would need detailed study of the consequences of our conditions on the random walk  $(\Pi_n)$  on the group  $GL(d)$ , rather than just norm calculations as above.

**3. Fixed points and positive recurrence.** In Vervaat (1979), Lemma 1.1, it is proved that if  $R_n(R_0) \rightarrow^L R$  then the distribution of  $R$  is a fixed point of (1.1), and Theorem 1.5 in the same paper characterizes fixed points when they exist. We assemble these results together with our convergence results of the previous section into the following complete characterization.

THEOREM 3.1.

- (a)  $P(M = 0) > 0$ . Then there is an integer-valued r.v.  $N$  such that  $Z_n(Z_0) = Z_N$  for all  $n \geq N$ , and the distribution  $\nu$  of  $Z_n$  is the unique fixed point of (1.1); also  $R_n(R_0) \rightarrow^L \nu$  whatever the distribution of  $R_0$ .
- (b)  $P(M = 0) = 0$  and  $P(Q + Mc = c) = 1$  for some  $c \in \mathbb{R}$ . Then

$$R_n(R_0) = c + (R_0 - c)\Pi_n \text{ a.s. for all } n = 0, 1, \dots,$$

and (2.8) holds, so that  $Z_n(Z_0)$  reduces to  $R_n(Z_0)$ .

- (i)  $M = 1$  a.s. In this case  $R_n(R_0) = R_0$  a.s., so all distributions for  $R_0$  are fixed points.
- (ii)  $M = -1$  a.s. Then

$$R_n(R_0) = c + (-1)^n(R_0 - c), \quad n = 0, 1, \dots$$

The fixed points consist of all distributions for  $R$  with  $R - c \stackrel{L}{=} -(R - c)$ , that is, for which  $R$  is symmetric about  $c$ , and  $R_n(R_0) \rightarrow^L R$  if and only if  $R_0 \stackrel{L}{=} R$ , in which case  $R_n(R_0) \stackrel{L}{=} R$  for all  $n$ .

- (iii)  $|M| = 1$  a.s. but  $0 < P(M = 1) < 1$ . The fixed points consist of all distributions for  $R$  with  $R - c \stackrel{L}{=} -(R - c)$ , and  $R_n(R_0) \rightarrow^L R$  whatever the distribution of  $R_0$ , the limit law being that of  $c + (R_0 - c)U$  where  $U = \pm 1$  with probabilities  $\frac{1}{2}, \frac{1}{2}$ , independent of  $R_0$ .
- (iv)  $\Pi_n \xrightarrow{P} 0$ . Then  $R = c$  a.s. is the only fixed point, and  $R_n(R_0) \xrightarrow{P} c$  whatever the distribution of  $R_0$ .
- (v)  $P(|M| = 1) < 1$  and  $\Pi_n$  does not tend to 0 in probability. Here  $R = c$  a.s. is the only fixed point, and  $R_n(R_0) \xrightarrow{P} c$  if and only if  $R_0 = c$  a.s.
- (c)  $P(M = 0) = 0$  and  $P(Q + Mc = c) < 1$  for all  $c \in \mathbb{R}$ .
- (i) (2.1) holds. Then the distribution  $\nu$  of  $Z_\infty$  is nondegenerate and is the only fixed point, and  $R_n(R_0) \rightarrow^L \nu$  whatever the distribution of  $R_0$ .
- (ii) (2.1) fails. Then there exists no fixed point, and  $|R_n(R_0)| \xrightarrow{P} \infty$  whatever the distribution of  $R_0$ .

A fixed-point distribution for (1.1) is a normalized invariant measure for the Markov sequence  $(R_n(R_0))$ , so existence of the fixed-point distribution says that this Markov sequence is “positive recurrent” by the usual definition. It is natural then to ask about ergodicity of  $(R_n(R_0))$ , and this is proved in the positive recurrent case in Kellerer (1992), Theorem 9.3. The papers [Kellerer (1992)] treat the special case of (1.1) and (1.2) restricted to  $\mathbb{R}_+$ , but the proof (via mixing) of Kellerer’s Theorem 9.3 holds for the general case. We quote it as follows.

**THEOREM 3.2** [Kellerer (1992)]. *Assume  $P(M = 0) = 0$ ,  $P(Q = 0) < 1$  and (2.1), and let  $\nu$  denote the fixed-point law, that is, the distribution of  $Z_\infty$ . Let  $f: \mathbb{R} \rightarrow \mathbb{R}$  be such that  $E|f(Z_\infty)| < \infty$  and either (a)  $f$  is bounded and  $\nu$ -a.e. continuous, or (b)  $f$  is uniformly continuous. Then, whatever the distribution of  $R_0$ ,*

$$\frac{1}{n} \sum_{k=1}^n f(R_k(R_0)) \rightarrow Ef(Z_\infty) \quad \text{a.s. } (n \rightarrow \infty).$$

Another way of saying this is that with probability 1 the empirical laws  $n^{-1} \sum_{k=1}^n \delta_{R_k}$  converge weakly to  $\nu$ . The reason for the continuity requirements on  $f$  are that without special assumptions on the law of  $(Q, M)$  it is quite unclear in what subset of  $\mathbb{R}$  the limit law  $\nu$  “really” lives. A simple example shows this: let  $M := \frac{1}{2}$  a.s. and  $P(Q = 1) := p$ ,  $P(Q = 0) := 1 - p$ , where  $0 < p \leq 1$ . Then the three cases  $p := \frac{1}{2}$ ,  $p \in (0, \frac{1}{2}) \cup (\frac{1}{2}, 1)$ ,  $p := 1$  make  $\nu$ , respectively, an absolutely continuous distribution, a continuous singular distribution and a degenerate distribution.

**4. Special cases.** We first relate our main result to moment conditions on  $M$  and  $Q$ . To have well-defined moments of  $-\log |M|$  we assume throughout this section that  $P(M = 0) = 0$ , and for a simple formulation we assume the nondegeneracy condition (2.7) [which also rules out  $P(Q = 0) = 1$ ]. Let us agree to call the conclusions that  $Z_n(Z_0) \rightarrow Z_\infty$  a.s. given by (1.5), and that (2.2)–(2.4) hold, “convergence.” As we know from Theorem 2.1, precisely when convergence fails we have “divergence” in the sense that  $|Z_n(Z_0)| \xrightarrow{P} \infty$ . Furthermore, by Theorem 3.1 there is a fixed point of (1.1), namely the distribution of  $Z_\infty$ , precisely when convergence occurs.

We write  $\log^+ x := \log(x \vee 1)$ ,  $\log^- x := -\log(x \wedge 1)$  for  $x > 0$ . Then  $J_-$  (see Remark 2.5) can be expressed as

$$J_- = E \left( \frac{\log^+ |M|}{A_M(\log^+ |M|)} \right).$$

**COROLLARY 4.1.** *Assume  $P(M = 0) = 0$  and the nondegeneracy condition (2.7):*

- (a)  $0 \leq E \log |M| \leq \infty$ . *Convergence does not occur.*
- (b)  $-\infty < E \log |M| < 0$ . *Convergence occurs if and only if  $E \log^+ |Q| < \infty$ .*

- (c)  $E \log |M| = -\infty$ . Convergence occurs if and only if the integral in (2.1) is finite, and in particular if  $E \log^+ |Q| < \infty$ .
- (d)  $E \log |M|$  does not exist, that is,  $E \log^+ |M| = \infty = E \log^- |M|$ .
- (i)  $J_- < \infty$ . Convergence occurs if and only if the integral in (2.1) is finite, and in particular if  $E \log^+ |Q| < \infty$ .
- (ii)  $J_- = \infty$ . Convergence does not occur.

REMARK 4.2. (b) is due to Grincevičius (1974), while (a) and the sufficiency parts of (c) and (d) are due to Vervaat (1979). The rest is new. We derive the whole result from our main Theorem 2.1 to demonstrate that the latter includes it.

Our second special case is when  $M$  is degenerate at  $m \neq 0$ . If  $|m| \geq 1$ , both parts of (2.1) fail, the function  $A_M$  being identically zero. Otherwise, when  $0 < |m| < 1$ , the first part of (2.1) holds, and for the second we have  $A_M(y) = x \wedge y$  where  $x := -\log |m| > 0$ ; thus finiteness of the integral is equivalent to  $E \log^+ |Q| < \infty$ . We thus deduce the next result.

COROLLARY 4.3. Consider the partial sum  $Z_n := \sum_{k=1}^n m^{k-1} Q_k$  of a random power series built from a constant  $m \neq 0$  and i.i.d.r.v.s  $Q, Q_1, Q_2, \dots$ , where  $P(Q = 0) < 1$ . If

$$|m| < 1 \quad \text{and} \quad E \log^+ |Q| < \infty$$

then  $\sum_{k=1}^{\infty} |m|^{k-1} |Q_k| < \infty$  a.s., and otherwise  $|Z_n| \xrightarrow{P} \infty$  as  $n \rightarrow \infty$ . In the former case the distribution of  $Z_{\infty} := \sum_{k=1}^{\infty} m^{k-1} Q_k$  is the unique solution of

$$(4.1) \quad R \stackrel{\perp}{=} Q + mR, \quad R \text{ independent of } Q,$$

while otherwise no solution exists.

REMARK 4.4. The observation seems to be due to Zakusilo (1975) that, either for all  $|m| < 1$  in the case  $E \log^+ |Q| < \infty$ , or for no  $m \neq 0$  in the opposite case,  $Z_{\infty} := \sum_{k=1}^{\infty} m^{k-1} Q_k$  converges a.s. and solves (4.1).

REMARK 4.5. From Corollary 4.3 we can deduce a corresponding result for  $Z_n := \sum_{k=1}^n M_0^{k-1} Q_k$ , where  $Q, Q_1, Q_2, \dots$  are conditionally i.i.d. given a r.v.  $M_0$ , with  $P(M_0 = 0) = 0$  and  $P(Q = 0 | M_0) < 1$  a.s. In fact, the tightness of  $\{Z_n\}$  implies

$$(4.2) \quad P(|M_0| < 1) = 1 \quad \text{and} \quad E(\log^+ |Q| | M_0) < \infty \quad \text{a.s.},$$

and, conversely, (4.2) implies  $\sum_{k=1}^{\infty} |M_0|^{k-1} |Q_k| < \infty$  a.s. We will omit the proof.

Our final special case is when  $Q$  is degenerate at some nonzero value, which without loss of generality we take to be 1. This case is important for the analysis of algorithms. We can satisfy the nondegeneracy condition (2.7) simply by assuming  $M$  nondegenerate. The following is then immediate from Theorems 2.1 and 3.1.

COROLLARY 4.6. Assume  $M$  is nondegenerate. If  $\Pi_n := \prod_{j=1}^n M_j \rightarrow 0$  a.s. as  $n \rightarrow \infty$  then the series  $Z_\infty := \sum_{k=1}^\infty \Pi_k$  is a.s. absolutely convergent. Further, the distribution  $\nu$  of  $Z_\infty$  is nondegenerate and is the only solution of

$$(4.3) \quad R \stackrel{L}{=} 1 + MR, R \text{ independent of } M,$$

and the sequence  $R_{n+1}(R_0) := 1 + M_{n+1}R_n(R_0)$  converges in distribution to  $\nu$ , whatever the distribution of  $R_0$ .

Conversely, if it is not the case that  $\Pi_n \rightarrow 0$  a.s., then

$$Z_n(Z_0) := \sum_{k=1}^n \Pi_{k-1} + \Pi_n Z_0$$

satisfies  $|Z_n(Z_0)| \xrightarrow{P} \infty$ ; also there is no solution to (4.3), and  $|R_n(R_0)| \xrightarrow{P} \infty$  whatever the distribution of  $R_0$ .

**5. Proofs.** We carry out most of the proof of Theorem 2.1 in the sequence of Lemmas 5.1–5.8 below. Our use of the function  $A_M$  that appears in our main condition (2.1) is a consequence of an estimate due to Erickson [Erickson (1973), Lemma 1]. Similar working in another context appears in Chow and Zhang (1986) and in Klass and Wittmann (1993). We will need another generalization of Erickson’s result, Lemma 5.1, which is easily deduced from Erickson’s lemma and Kesten and Maller (1996).

LEMMA 5.1. Let  $V_1, V_2, \dots$  be i.i.d. r.v.s, not degenerate at 0, and suppose  $V_1 + \dots + V_n \rightarrow \infty$  a.s. as  $n \rightarrow \infty$ . Then there is a finite positive constant  $c_+$ , depending on the distribution of  $V_1$ , such that for all  $y \geq 0$ ,

$$(5.1) \quad \frac{y}{E(V_1^+ \wedge y)} \leq \sum_{n=0}^\infty P\left(\max_{1 \leq j \leq n} (V_1 + \dots + V_j) \leq y\right) \leq \frac{c_+ y}{E(V_1^+ \wedge y)},$$

where  $y/E(V_1^+ \wedge y)$  is interpreted to have value  $1/P(V_1 > 0) < \infty$  at  $y = 0$ .

PROOF. When  $V_1 \geq 0$  a.s.,  $\max_{1 \leq j \leq n} (V_1 + \dots + V_j) = V_1 + \dots + V_n$ , and (5.1) follows from Erickson (1973), Lemma 1, with  $c_+ = 2$ . Since for all  $y \geq 0$ ,

$$P(V_1^+ + \dots + V_n^+ \leq y) \leq P\left(\max_{1 \leq j \leq n} (V_1 + \dots + V_j) \leq y\right),$$

the left-hand inequality in (5.1) is true as stated for general  $V_j$ . By Kesten and Maller (1996) [see their (4.3), (4.2) and (4.5)],  $V_1 + \dots + V_n \rightarrow \infty$  a.s. implies that the right-hand inequality of (5.1) holds for all large enough  $y$ . The limit of the nonincreasing function  $E(V_1^+ \wedge y)/y$  as  $y \downarrow 0$  is  $P(V_1 > 0)$ , which must be positive since  $V_1 + \dots + V_n \rightarrow \infty$  a.s. So by choosing  $c_+$  larger if necessary we can assume (5.1) holds for all  $y \geq 0$ .  $\square$

LEMMA 5.2. Suppose  $P(Q = 0) < 1$  and  $P(M = 0) = 0$ . If (2.1) holds, then for some  $c > 0$ ,

$$(5.2) \quad \Pi_{n-1} Q_n = o(e^{-cn}), \quad n \rightarrow \infty, \text{ a.s.}$$

Consequently, (2.1) implies (2.4) and (2.6).

PROOF. The hypotheses imply that for the r.v.s.  $X_j = -\log |M_j|$  and  $Y_j = \log |Q_j|$ , defined in (1.6), we have

$$P(|X_j| < \infty) = 1 = P(Y_j^+ < \infty), \quad j = 1, 2, \dots,$$

and we can rewrite (2.1) as

$$(5.3) \quad \Pi_n \xrightarrow{\text{a.s.}} 0 \ (n \rightarrow \infty) \quad \text{and} \quad \int_{[0, \infty)} \left( \frac{y}{E(X^+ \wedge y)} \right) dP(Y_1^+ \leq y) < \infty.$$

[ $\Pi_n \rightarrow 0$  a.s. implies  $P(|M| < 1) = P(X > 0) > 0$ , and we take  $y/E(X^+ \wedge y)$  to have value  $1/P(X > 0) < \infty$  at  $y = 0$  for the integrand in (5.3).]

With  $S_n = \sum_{j=1}^n X_j$ , we will now show that

$$(5.4) \quad S_{n-1} - Y_n - cn \xrightarrow{\text{a.s.}} \infty$$

for some  $c > 0$ , which implies (5.2). To prove (5.4), note that  $\Pi_n \rightarrow 0$  a.s. implies  $S_n \rightarrow \infty$  under  $P(M = 0) = 0$  (see Remark 2.5). Now consider some cases.

Case 1. Suppose  $E(X^+) = \infty$ . Then  $\sum_{j=1}^{n-1} X_j^+ / n \rightarrow \infty$  a.s. We can obtain from the convergence of the integral in (5.3) that

$$(5.5) \quad \frac{Y_n^+}{\sum_{j=1}^{n-1} X_j^+} \xrightarrow{\text{a.s.}} 0.$$

To see this, take an  $\varepsilon$  in (0,1) and write, noting the independence of  $Y_n$  and of  $\{X_j\}_{1 \leq j \leq n-1}$ ,

$$\begin{aligned} & \sum_{n \geq 1} P(\varepsilon(X_1^+ + \dots + X_{n-1}^+) \leq Y_n^+) \\ &= \int_{[0, \infty)} \sum_{n \geq 1} P(X_1^+ + \dots + X_{n-1}^+ \leq y/\varepsilon) dP(Y^+ \leq y) \\ &\leq \int_{[0, \infty)} \left( \frac{c_+ y/\varepsilon}{E(X^+ \wedge (y/\varepsilon))} \right) dP(Y^+ \leq y) \\ &\leq \frac{c_+}{\varepsilon} \int_{[0, \infty)} \left( \frac{y}{E(X^+ \wedge y)} \right) dP(Y^+ \leq y) < \infty. \end{aligned}$$

We used the right-hand inequality in (5.1) in the last estimate. An application of the Borel-Cantelli lemma now establishes (5.5). Now  $S_n \rightarrow \infty$  a.s. and  $E(X^+) = \infty$  imply  $J_- < \infty$  [see (2.10)], and hence  $\sum_{j=1}^{n-1} X_j^- = o(\sum_{j=1}^{n-1} X_j^+)$  a.s. by Pruitt [(1981), Lemma 8.1]. This together with  $n = o(\sum_{j=1}^{n-1} X_j^+)$  a.s. and (5.5) implies

$$(5.6) \quad \sum_{j=1}^{n-1} X_j^+ - n - Y_n^+ - \sum_{j=1}^{n-1} X_j^- = \left( \sum_{j=1}^{n-1} X_j^+ \right) (1 - o(1)) \xrightarrow{\text{a.s.}} \infty,$$

which implies (5.4), with  $c = 1$ .

Case 2. Suppose  $E(X^+) < \infty$ . Then  $E(Y^+) < \infty$  by (5.3). Since  $S_n \rightarrow \infty$  a.s., we must have  $E(X^-) < \infty$  and  $\mu := EX > 0$ . Now  $Y_n^+ = o(n)$  a.s., and  $(S_{n-1} - n\mu/2)/n \rightarrow \mu/2 > 0$  a.s., so

$$(5.7) \quad \frac{Y_n^+}{S_{n-1} - n\mu/2} \xrightarrow{\text{a.s.}} 0.$$

It follows that

$$S_{n-1} - n\mu/2 - Y_n^+ = (S_{n-1} - n\mu/2)(1 - o(1)) \xrightarrow{\text{a.s.}} \infty,$$

so (5.4) holds with  $c = \mu/2$ .

This establishes (5.2). It follows that the series (1.5),  $Z_\infty = \sum_{k=1}^\infty \Pi_{k-1} Q_k$ , converges absolutely a.s., which is (2.4).  $Z_\infty$  is the a.s. limit of its partial sums:  $Z_n \rightarrow Z_\infty$  a.s. Using  $\Pi_n \rightarrow 0$  a.s., which is the first part of (2.1), we have further that  $Z_n(Z_0) = Z_n + \Pi_n Z_0 \rightarrow Z_\infty$  a.s., which is (2.6).  $\square$

LEMMA 5.3. Assume that  $P(M = 0) = 0$  and suppose  $\Pi_n \rightarrow 0$  a.s. as  $n \rightarrow \infty$ . Define  $a_n := \sum_{j=1}^n X_j^- / \sum_{j=1}^n X_j^+$  and set  $\tilde{M}_j := e^{-X_j^+}$  and  $\tilde{\Pi}_n := \prod_{j=1}^n \tilde{M}_j$ . Then  $|\Pi_n| = \tilde{\Pi}_n^{1-a_n}$  for all  $n$ . Further, there exists a constant  $a < 1$  such that

$$(5.8) \quad \limsup_{n \rightarrow \infty} a_n < a < 1 \quad \text{a.s.},$$

so that as soon as  $a_n < a$  (and hence for all large  $n$ , a.s.),

$$(5.9) \quad \tilde{\Pi}_n \leq |\Pi_n| \leq \tilde{\Pi}_n^{1-a}.$$

PROOF. The assumption  $\Pi_n \rightarrow 0$  a.s. means that the random walk  $S_n = -\log |\Pi_n| \rightarrow \infty$  a.s., as  $n \rightarrow \infty$ , which is equivalent, as noted in Remark 2.5, to  $J_- < E(X^+) = \infty$  or  $0 < EX \leq E|X| < \infty$ . If  $J_- < \infty = E(X^+)$  then by Pruitt (1981), Lemma 8.1,  $a_n \rightarrow 0$  a.s. as  $n \rightarrow \infty$ . If on the other hand  $0 < EX \leq E|X| < \infty$  then  $a_n \rightarrow E(X^-)/E(X^+) < 1$  a.s. Thus in either case there is an  $a < 1$ , which by the Hewitt–Savage law we can take to be a constant, such that (5.8) holds.

We have

$$(5.10) \quad S_n = \sum_{j=1}^n (X_j^+ - X_j^-) = (1 - a_n) \sum_{j=1}^n X_j^+$$

and consequently,

$$|\Pi_n| = e^{-S_n} = \left( \prod_{j=1}^n e^{-X_j^+} \right)^{1-a_n} = \tilde{\Pi}_n^{1-a_n}.$$

Finally, the left-hand inequality of (5.9) holds as soon as  $a_n < 1$  and the right-hand as soon as  $a_n < a$ .  $\square$

LEMMA 5.4. Assume that  $P(M = 0) = 0$ ,  $P(0 \leq M \leq 1) = 1$ ,  $P(0 < M < 1) > 0$  and  $P(Q = 0) < 1 = P(Q \geq 0)$ . Suppose also that  $\limsup_{n \in \mathbb{N}} \Pi_{n-1} Q_n < \infty$  a.s. Then the integral in (2.1) converges.

PROOF. Define events  $E_n(u) := \{\Pi_{n-1}Q_n \geq u\}$  for  $n = 1, 2, \dots$ . By the Hewitt–Savage law the lim sup in the statement of the lemma is degenerate, so we have that for some constant  $u > 0$ ,

$$(5.11) \quad P(E_n(u) \text{ i.o.}) = 0.$$

Now  $E_i(u)$  implies, since  $Q \geq 0$  a.s. and  $0 \leq M \leq 1$  a.s., that  $Q_i \prod_{j=k+1}^{i-1} M_j \times M_j \geq \Pi_{i-1}Q_i \geq u$  when  $i > k$ . Thus, for  $i > k$ .

$$\begin{aligned} P(E_k(u) \cap E_i(u)) &\leq P(E_k(u))P\left(Q_i \prod_{j=k+1}^{i-1} M_j \geq u\right) \\ &= P(E_k(u))P(E_{i-k}(u)). \end{aligned}$$

This allows us to apply the generalized Borel–Cantelli lemma of Spitzer [(1976) page 317] to deduce from (5.11) that  $\sum_n P(E_n(u)) < \infty$ , for the given value of  $u$ . Since  $\sum_n P(E_n(u))$  is nonincreasing in  $u$ , we can further assume that  $u \geq 1$ .

Since we are assuming  $P(0 < M < 1) > 0$ , we have  $A_M(y) = \int_0^y P(-\log M > x) dx > 0$  for all  $y > 0$ , and since  $P(0 \leq M \leq 1) = 1$ ,  $\Pi_n = e^{-S_n}$ . We apply (5.1), noting that  $X = -\log M \geq 0$  a.s. here, and  $X$  is not degenerate at 0. Fix  $x_0 > 0$  and set  $C_0 := A_M(x_0) > 0$ . Then we calculate

$$\begin{aligned} \infty &> \sum_{n=1}^{\infty} P(E_n(u)) \geq \sum_{n=1}^{\infty} P(\Pi_{n-1}Q_n \geq u, Q_n > ue^{x_0}) \\ &= \int_{(ue^{x_0}, \infty)} \sum_{n=1}^{\infty} P\left(S_{n-1} \leq \log \frac{q}{u}\right) dP(Q \leq q) \\ &\geq \int_{(ue^{x_0}, \infty)} \left(\frac{\log(q/u)}{A_M(\log(q/u))}\right) dP(Q \leq q) \\ &\geq \int_{(ue^{x_0}, \infty)} \left(\frac{\log q}{A_M(\log q)}\right) dP(Q \leq q) - \frac{\log u}{C_0}. \end{aligned}$$

Here we used the facts that  $u \geq 1$  and  $A_M(y)$  is nondecreasing in  $y$ . Since  $(\log q)/A_M(\log q)$  is finite on  $(1, ue^{x_0})$  we see that the integral in (2.1) converges in this special case.  $\square$

LEMMA 5.5. *Suppose  $P(Q = 0) < 1$  and  $P(M = 0) = 0$ . If  $\Pi_n \rightarrow 0$  a.s. and  $|Z_n(Z_0)|$  does not tend in probability to  $\infty$ , then  $Z_n(Z_0)$  converges in distribution to a proper r.v.  $Z$  and the integral in (2.1) is finite.*

PROOF. Assume the conditions of the lemma. Then, since  $Z_n(Z_0) = Z_n + \Pi_n Z_0$  and  $\Pi_n Z_0 \rightarrow 0$  a.s., we may without loss of generality assume  $Z_0 = 0$  a.s. We first show that  $Z_n$  converges in distribution.

Since  $|Z_n|$  does not tend to  $\infty$  in probability, we can find a nonrandom sequence  $(n_s)$  such that, as  $s \rightarrow \infty$ ,  $n_s \uparrow \infty$  and  $Z_{n_s}$  converges in law to a

(possibly improper) d.f. (distribution function)  $F$  which has positive mass on  $(-\infty, \infty)$ . That is, with  $F(\infty) := \lim_{x \uparrow \infty} F(x)$  and  $F(-\infty) := \lim_{x \downarrow -\infty} F(x)$ , we have  $F(\infty) - F(-\infty) > 0$ .

Now  $Z_{n_s+1} = Z_{n_s} + Q_{n_s+1}\Pi_{n_s}$ , and the term  $Q_{n_s+1}\Pi_{n_s}$ , tends to 0 in probability, so  $Z_{n_s+1}$  tends in law to  $F$  too. Note that, for  $1 \leq m < n$ ,

$$Z_n = Z_m + \Pi_m \sum_{j=1}^{n-m} \left( \prod_{k=1}^{j-1} M_{k+m} \right) Q_{j+m} = Z_m + \Pi_m (Z_{n-m} \circ \theta^m),$$

where  $\theta$  denotes the shift operator that adds 1 to the indices of the  $Q_j$  and the  $M_j$ . Take  $n := n_s + 1$ ,  $m := 1$ , to write

$$Z_{n_s+1} = Q_1 + M_1 \tilde{Z}_{n_s}, \quad \tilde{Z}_{n_s} \stackrel{L}{=} Z_{n_s} \text{ independent of } (Q_1, M_1),$$

If  $Z'$  is a (possibly improper) r.v. with d.f.  $F$ , this tells us that, on  $\{|Z'| < \infty\}$ ,

$$Z' \stackrel{L}{=} Q_1 + M_1 Z', \quad Z' \text{ independent of } (Q_1, M_1).$$

Now let  $Z$  have the distribution of  $Z'$  conditional on  $|Z'| < \infty$ ; that is,  $Z$  is a proper r.v. with d.f.  $G$  satisfying  $G(x) := (F(x) - F(-\infty))/(F(\infty) - F(-\infty))$  for all real  $x$ . Since  $\{|Z'| < \infty\} = \{|Q_1 + M_1 Z'| < \infty\}$ , we have

$$Z \stackrel{L}{=} Q + MZ, \quad Z \text{ independent of } (Q, M).$$

This says that  $G$  is a fixed-point distribution. Iterate this to get

$$Z \stackrel{L}{=} \sum_{k=1}^n Q_k \Pi_{k-1} + \Pi_n Z, \quad Z \text{ independent of } ((Q_k, M_k))_{k=1, \dots, n}.$$

The right-hand side is just  $Z_n + \Pi_n Z$ . Since  $\Pi_n \rightarrow 0$  a.s. and  $Z$  is proper,  $\Pi_n Z$  tends in probability to 0. It follows that  $Z_n$  tends in law to  $Z$ , that is, to the distribution  $G$ .

Now we will deduce that the integral in (2.1) is finite. We need to extend a method from Goldie [(1991), pages 136, 157] employing a maximal inequality of Grincevičius (1980). Define

$$\Pi_{j,n} := \prod_{k=j+1}^n M_k, \quad Z_{j,n} := \sum_{k=j+1}^n \Pi_{j,k-1} Q_k,$$

and let “med” denote a median. The inequality is

$$P\left(\max_{j=1, \dots, n} (Z_j + \Pi_j \text{med}(Z_{j,n} + \Pi_{j,n} y)) > x\right) \leq 2P(Z_n + \Pi_n y > x) \quad x, y \in \mathbb{R},$$

valid for  $n = 1, 2, \dots$ . We need only the  $y = 0$  case. We note also that  $Z_{j,n} \stackrel{L}{=} Z_{n-j}$ . Thus

$$P\left(\max_{j=1, \dots, n} (Z_j + \Pi_j \text{med } Z_{n-j}) > x\right) \leq 2P(Z_n > x), \quad x \in \mathbb{R}.$$

By applying this to the pairs  $(-Q_1, M_1), \dots, (-Q_n, M_n)$  we deduce a similar inequality with  $-Z_j$  in place of  $Z_j$ , so

$$P\left(\max_{j=1, \dots, n} |Z_j + \Pi_j \operatorname{med} Z_{n-j}| > x\right) \leq 2P(|Z_n| > x), \quad x \geq 0.$$

Write  $m_n := \operatorname{med} Z_n$ . The above implies, a fortiori, that for  $n \geq k \geq 1$ ,

$$(5.12) \quad P\left(\max_{j=1, \dots, k} |Z_j + \Pi_j m_{n-j}| > x\right) \leq 2P(|Z_n| > x), \quad x \geq 0.$$

Let  $n \rightarrow \infty$  with  $k$  fixed. The right-hand side converges, at least for  $x \geq 0$  in the set  $C_z$  of continuity points of  $Z$ , to  $2P(|Z| > x)$ . Because  $Z_n \xrightarrow{L} Z$  we can assume that  $m_n \rightarrow m_0$ , a median of  $Z$ . So (5.12) yields

$$P\left(\max_{j=1, \dots, k} |Z_j + \Pi_j m_0| > x\right) \leq 2P(|Z| > x), \quad x \in [0, \infty) \cap C_z.$$

Now let  $k \rightarrow \infty$ , to conclude that  $\sup_{j \in \mathbb{N}} |Z_j + \Pi_j m_0|$  is a.s. finite.

The a.s. finiteness of  $\sup_{j \in \mathbb{N}} |Z_j + \Pi_j m_0|$  together with  $\Pi_n \rightarrow 0$  a.s. yields that  $\limsup_{n \rightarrow \infty} |Z_n| < \infty$  a.s. We then have  $\limsup_{n \rightarrow \infty} |\Pi_{n-1}| |Q_n| < \infty$  a.s., because  $\Pi_{n-1} Q_n = Z_n - Z_{n-1}$ . With  $\tilde{\Pi}_n$  defined as in Lemma 5.3, we have the left-hand inequality of (5.9), and therefore  $\limsup_{n \rightarrow \infty} \tilde{\Pi}_{n-1} |Q_n| < \infty$  a.s. The a.s. convergence of  $\Pi_n$  to 0 implies that  $\tilde{M} := \min(|M|, 1)$  has  $P(\tilde{M} = 1) < 1$ . We now have all the assumptions of Lemma 5.4 satisfied by  $(|Q|, \tilde{M})$ , that is, by the i.i.d. sequence  $(|Q_n|, \tilde{M}_n)$ , so the lemma gives

$$\int_{(1, \infty)} \left( \frac{\log q}{\tilde{A}_M(\log q)} \right) dP(Q \leq q) < \infty,$$

where

$$\begin{aligned} \tilde{A}_M(y) &:= \int_0^y P(-\log \tilde{M} > x) dx = \int_0^y P(X^+ > x) dx \\ &= \int_0^y P(X > x) dx = A_M(y). \end{aligned}$$

We thus have the finiteness of the integral in (2.1).  $\square$

LEMMA 5.6. *Suppose  $P(Q = 0) < 1$  and  $P(M = 0) = 0$ . Then (2.1) holds if and only if (2.5) holds.*

PROOF. Define  $X_j, Y_j$  and  $S_n$  as in (1.6) and (1.7). Then (2.1) is equivalent to

$$(5.13) \quad \Pi_n \xrightarrow{\text{a.s.}} 0 \quad (n \rightarrow \infty)$$

and

$$\int_{(0, \infty)} \left( \frac{y}{\int_0^y P(X > z) dz} \right) dP(Y^+ \leq y) < \infty,$$

while (2.5) is equivalent to

$$(5.14) \quad S(x) := \sum_{n \geq 1} P\left(\max_{1 \leq j \leq n-1} S_j \leq x + Y_n\right) < \infty \quad \text{for all } x > 0.$$

Now let (5.13) hold. We have

$$(5.15) \quad \begin{aligned} S(x) &\leq \sum_{n \geq 1} P\left(\max_{1 \leq j \leq n-1} S_j \leq x\right) P(Y \leq 0) \\ &\quad + \int_{(0, \infty)} \sum_{n \geq 1} P\left(\max_{1 \leq j \leq n-1} S_j \leq x + y\right) dP(Y^+ \leq y). \end{aligned}$$

We assume  $S_n \rightarrow \infty$  a.s., and we must have  $P(|M| \geq 1) < 1$  here, since  $\Pi_n \rightarrow 0$  a.s. as  $n \rightarrow \infty$ . So  $P(X > 0) > 0$ , and we can apply Lemma 5.1 to get, for all  $x \geq 0$ ,

$$(5.16) \quad \frac{x}{E(X^+ \wedge x)} \leq \sum_{n \geq 1} P\left(\max_{1 \leq j \leq n-1} S_j \leq x\right) \leq \frac{c_+ x}{E(X^+ \wedge x)}.$$

But then, by (5.15),

$$S(x) \leq \frac{c_+ x}{E(X^+ \wedge x)} + \int_{(0, \infty)} \left(\frac{c_+(x+y)}{\int_0^{x+y} P(X > z) dz}\right) dP(Y^+ \leq y)$$

and the integral may be written as the sum of two integrals involving, respectively,  $x$  and  $y$  in the numerator, which are then seen to converge on applying (5.13). Hence (5.14) holds.

Conversely suppose (5.14) holds. Then for any  $x > 0$  and  $y_0 > 0$ ,

$$(5.17) \quad \begin{aligned} \infty &> \int_{[-y_0/y_0]} \sum_{n \geq 1} P\left(\max_{1 \leq j \leq n-1} S_j \leq x + y\right) dP(Y \leq y) \\ &\geq P(|Y| \leq y_0) \sum_{n \geq 1} P\left(\max_{1 \leq j \leq n-1} S_j \leq x - y_0\right). \end{aligned}$$

Since  $P(|Y| < \infty) > 0$  (not all of the mass of  $Y$  can be at  $-\infty$  because  $P(Q = 0) < 1$ ) we can choose  $y_0$  so large that  $P(|Y| \leq y_0) > 0$ . Then, with the choice  $x = y_0$ , (5.17) gives

$$\sum_{n \geq 1} P\left(\max_{1 \leq j \leq n-1} S_j \leq 0\right) < \infty.$$

Then  $S_n \rightarrow \infty$  a.s. by the case  $\alpha = 0$  of Kesten and Maller (1996), Theorem 2.1 [see also (1.4) of Kesten and Maller (1996)], and consequently  $\Pi_n \rightarrow 0$  a.s.

Thus we again have (5.16). Going back to (5.14) we then get

$$\begin{aligned} \infty &> \int_{(0,\infty)} \sum_{n \geq 1} P\left(\max_{1 \leq j \leq n-1} S_j \leq x + y\right) dP(Y^+ \leq y) \\ &\geq \int_{(0,\infty)} \left(\frac{x + y}{\int_0^{x+y} P(X > z) dz}\right) dP(Y^+ \leq y) \\ &\geq \int_{(0,\infty)} \left(\frac{y}{\int_0^y P(X > z) dz}\right) dP(Y^+ \leq y). \end{aligned}$$

The last step follows since  $x/\int_0^x P(X > z) dz$  increases in  $x$ . Hence (5.13) holds.  $\square$

For the next lemma we need the following inequality due to Esseen, to be found in Petrov (1995), Theorem 2.15, (2.54). It employs the *Lévy concentration function*

$$Q(X; \lambda) := \sup_{-\infty < x < \infty} P(x \leq X \leq x + \lambda),$$

of a r.v.  $X$ .

PROPOSITION 5.7. *Let  $X_1, \dots, X_n$  be independent r.v.s, with symmetrized versions  $X_1^s, \dots, X_n^s$ , and set  $S_n := \sum_{j=1}^n X_j$ . Let  $\lambda_1, \dots, \lambda_n$  be positive numbers and  $\lambda \geq \max\{\lambda_1, \dots, \lambda_n\}$ . Then*

$$Q(S_n; \lambda) \leq \frac{A\lambda}{\sqrt{\sum_{j=1}^n \lambda_j^2 P(|X_j^s| \geq \frac{1}{2}\lambda_j)}}$$

where  $A$  is an absolute constant.

We shall use this in the following way. Since  $Q(X^s; \lambda) \leq Q(X; \lambda)$  for all  $\lambda \geq 0$  [cf. Petrov (1995), Lemma 1.11] we may replace  $S_n$  by  $S_n^s$  in the left-hand side of the inequality. Then we pick  $z > 0$  and put  $\lambda_1 = \dots = \lambda_n = \lambda := 2z$ , and deduce that

$$(5.18) \quad P(|S_n^s| \leq z) \leq \frac{A}{\sqrt{\sum_{j=1}^n P(|X_j^s| \geq z)}}.$$

LEMMA 5.8. *Suppose that (2.7) holds,  $P(M = 0) = 0$  and  $\Pi_n$  does not converge to 0 a.s. Then  $|Z_n(Z_0)| \xrightarrow{P} \infty$ .*

PROOF. (2.7) implies  $P(Q = 0) < 1$ . Assume also that  $\Pi_n$  does not converge to 0 a.s. We need the following construction. Since  $Q$  and  $M$  are real random variables there exists a regular conditional distribution for  $Q$  given  $M$ , so we may define  $F_m(q) := P(Q \leq q | M = m)$ , such that for each  $q \in \mathbb{R}$ ,  $F_m(q)$  as a function of  $m$  is a version of the conditional probability  $P(Q \leq q | M = m)$ , while for each  $m \in \mathbb{R}$ ,  $F_m(q)$  as a function of  $q$  is a proper distribution function. For each  $m$  let  $F_m^{-1}(\cdot)$  be a left-continuous inverse of the function  $q \mapsto F_m(q)$ . By augmenting the probability space if necessary we may suppose it supports a sequence of independent r.v.s  $(U_1, U_2, \dots)$ , independent of the sequence  $((Q_1, M_1), (Q_2, M_2), \dots)$ , with each  $U_j$  uniformly distributed on

(0,1). For each  $j = 1, 2, \dots$  we let  $Q'_j := F_M^{-1}(U_j)$ . Thus we have constructed  $(Q'_j, M_j)$  which are i.i.d. with the same distribution as  $(Q_j, M_j)$ , and such that each  $Q'_j$  is conditionally independent of  $Q_j$ , given  $M_j$ . Now define conditionally symmetrized r.v.s by

$$Q_j^s := Q_j - Q'_j, \quad Z_n^s := \sum_{j=1}^n \Pi_{j-1} Q_j^s.$$

Note that the  $Q_j^s$  are degenerate at 0 only if  $Q$  is a Borel function of  $M$ ,  $Q = f(M)$ , say. Assume for now that that is not the case. Let  $\mathcal{S}_n := \sigma\{M_1, \dots, M_n\}$ . Applying (5.18) to the conditional distribution of  $Z_n^s$  given  $\mathcal{S}_{n-1}$  we obtain, for all  $z > 0$ ,

$$(5.19) \quad P(|Z_n^s| \leq z | \mathcal{S}_{n-1}) \leq \frac{A}{\sqrt{\sum_{j=1}^n P(|\Pi_{j-1} Q_j^s| \geq z | \mathcal{S}_{n-1})}} \quad \text{a.s.}$$

The denominator on the right-hand side is, everywhere on the probability space, nonzero for  $z > 0$  sufficiently small. To see this, note that under the conditioning on  $\mathcal{S}_{n-1}$  we may regard  $|\Pi_{n-1}|$  as fixed at  $m > 0$  say; then since  $Q_n^s$  is independent of  $\mathcal{S}_{n-1}$  and  $P(|Q_n^s| \geq z_0) > 0$  for some  $z_0 > 0$ , we have  $P(|\Pi_{n-1} Q_n^s| \geq z | \mathcal{S}_{n-1}) > 0$  for  $0 < z < z_0/m$ . From (5.19) we thus have, a.s.,

$$\begin{aligned} P^2(|Z_n^s| \leq z | \mathcal{S}_{n-1}) &\leq \frac{A^2}{E(\sum_{j=1}^n \mathbf{1}(|\Pi_{j-1} Q_j^s| > z) | \mathcal{S}_{n-1})} \\ &\leq A^2 E\left(\frac{1}{\sum_{j=1}^n \mathbf{1}(|\Pi_{j-1} Q_j^s| > z)} \middle| \mathcal{S}_{n-1}\right) \end{aligned}$$

by Jensen's inequality. Thus

$$(5.20) \quad \begin{aligned} (P(|Z_n^s| \leq z))^2 &= E^2(P(|Z_n^s| \leq z | \mathcal{S}_{n-1})) \leq E(P^2(|Z_n^s| \leq z | \mathcal{S}_{n-1})) \\ &\leq A^2 E\left(\frac{1}{\sum_{j=1}^n \mathbf{1}(|\Pi_{j-1} Q_j^s| > z)}\right). \end{aligned}$$

Suppose that for some  $z > 0$  the series in the denominator of (5.20) converges, with positive probability and hence a.s. Then  $P(|\Pi_{n-1} Q_n^s| > z \text{ i.o.}) = 0$ . This implies  $\Pi_n \rightarrow 0$  a.s. as follows. As we have assumed, the  $Q_j^s$  are not degenerate at 0, so we may choose  $\delta > 0$  so small that  $P(|Q_n^s| > \delta) > 0$ . Define events  $A_n := \{|\Pi_{n-1}| > z/\delta\}$  and  $B_n := \{|Q_n^s| > \delta\}$ .  $B_n$  is independent of  $A_n, A_{n-1}, \dots, A_1$ , so we have by the lemma for events [Loève (1977), Section 18] that

$$\begin{aligned} P\left(\bigcup_{n=m}^{\infty} A_n \cap B_n\right) &\geq P\left(\bigcup_{n=m}^{\infty} A_n\right) \inf_{n=m, m+1, \dots} P(B_n) \\ &= P(|Q_1^s| > \delta) P\left(\bigcup_{n=m}^{\infty} A_n\right). \end{aligned}$$

This shows that if  $P(A_n \text{ i.o.}) > 0$  then  $A_n \cap B_n$  occurs i.o. with positive probability, hence  $|\Pi_{n-1}| |Q_n^s| > z$  i.o. with positive probability, a contradiction. So we have  $P(|\Pi_{n-1}| > z/\delta \text{ i.o.}) = 0$ , and thus  $\Pi_n \rightarrow 0$  a.s. since the only alternative left for the random walk  $(S_n)$  is to diverge to  $+\infty$  a.s. But  $\Pi_n \rightarrow 0$  a.s. contradicts the initial hypothesis of the lemma, so it must be the case that the series in the denominator of (5.20) diverges a.s. for all  $z > 0$ . Then by monotone convergence we get  $|Z_n^s| \xrightarrow{P} \infty$ , as  $n \rightarrow \infty$ . But then

$$\begin{aligned}
 P(|Z_n^s| \leq z) &= E(P(|Z_n^s| \leq z | \mathcal{L}_n, Z_0)) \\
 &\geq E\left(P\left(|Z_n^s| \leq z, \left|\sum_{k=1}^n \Pi_{k-1} Q'_k + \Pi_n Z_0\right| \leq z/2 \mid \mathcal{L}_n, Z_0\right)\right) \\
 (5.21) \quad &\geq E\left(P\left(\left|\sum_{k=1}^n \Pi_{k-1} Q_k + \Pi_n Z_0\right| \leq z/2 \mid \mathcal{L}_n, Z_0\right)\right) \\
 &\quad \times P\left(\left|\sum_{k=1}^n \Pi_{k-1} Q'_k + \Pi_n Z_0\right| \leq z/2 \mid \mathcal{L}_n, Z_0\right) \\
 &= E\left(P^2(|Z_n(Z_0)| \leq z/2 | \mathcal{L}_n, Z_0)\right) \geq P^2(|Z_n(Z_0)| \leq z/2),
 \end{aligned}$$

which shows that  $|Z_n(Z_0)| \xrightarrow{P} \infty$  for this case.

Now consider the case when  $Q = f(M)$  is a Borel function of  $M$ . Then

$$\begin{aligned}
 Z_{2n}(Z_0) &= \sum_{i=1}^{2n} \Pi_{i-1} Q_i + \Pi_{2n} Z_0 \\
 &= \sum_{i=1}^n \left(\prod_{j=1}^{i-1} \tilde{M}_j\right) \tilde{Q}_i + \left(\prod_{j=1}^n \tilde{M}_j\right) Z_0 \\
 &= \sum_{i=1}^n \tilde{\Pi}_{i-1} \tilde{Q}_i + \tilde{\Pi}_n Z_0 \text{ say,}
 \end{aligned}$$

where

$$(\tilde{Q}_i, \tilde{M}_i) := (Q_{2i-1} + M_{2i-1} Q_{2i}, M_{2i-1} M_{2i})$$

are i.i.d. and  $\tilde{\Pi}_n = \Pi_{2n}$ . Note that if  $\Pi_{2n} \rightarrow 0$  a.s. then  $\Pi_{2n+1} = M_1(\Pi_{2n} \circ \theta) \rightarrow 0$  a.s., so  $\Pi_n \rightarrow 0$  a.s. Since we are assuming that  $\Pi_n$  does not converge to 0 a.s., it follows that also  $\tilde{\Pi}_n$  does not converge to 0 a.s. If  $\tilde{Q}_i$  is a Borel function of  $\tilde{M}_i$  [which includes the possibility that  $\tilde{Q}_i = c(1 - \tilde{M}_i)$  for some  $c$ ],  $\tilde{Q}_i = g(\tilde{M}_i)$ , say, then  $Q_1 + M_1 Q_2 = g(M_1 M_2)$  and Proposition 1 of Grincevičius (1981) shows that either  $Q + cM = c$  a.s. for some  $c$  or  $(Q, M) = (c_1, 1)$  for some  $c_1$ . The first of these contradicts (2.7). If  $(Q, M) = (c_1, 1)$  we have  $c_1 \neq 0$  since  $P(Q = 0) < 1$ . Then by (1.4),  $|Z_n(Z_0)| = |nc_1 + Z_0| \xrightarrow{P} \infty$ . Alternatively,  $\tilde{Q}_i$  is not a Borel function of  $\tilde{M}_i$ , and the part of the present proof up to (5.21) then gives  $|Z_{2n}(Z_0)| = |\sum_{i=1}^n \tilde{\Pi}_{i-1} \tilde{Q}_i + \tilde{\Pi}_n Z_0| \xrightarrow{P} \infty$ . But the fact that

$Z_{2n+1}(Z_0) = Q_1 + M_1(Z_{2n} \circ \theta)(Z_0)$  together with  $P(M = 0) = 0$  then shows that  $|Z_{2n+1}(Z_0)| \xrightarrow{P} \infty$ , hence  $|Z_n(Z_0)| \xrightarrow{P} \infty$ .  $\square$

PROOF OF THEOREM 2.1. We have  $P(Q = 0) < 1$  and  $P(M = 0) = 0$ .

In Lemma 5.2 we proved that (2.1) implies (2.4) and (2.6). In turn, (2.4) implies (2.3), which implies (2.2).

Suppose (2.2) holds, so that  $\limsup_{n \rightarrow \infty} |\Pi_{n-1}|Q_n| < \infty$  a.s. By the Hewitt–Savage law this r.v. is degenerate, hence  $P(|\Pi_{n-1}|Q_n| > z \text{ i.o.}) = 0$  for some  $z > 0$ . Exactly as in Lemma 5.8, using the lemma for events, we deduce that  $\Pi_n \rightarrow 0$  a.s. Now we can use Lemma 5.3, and specifically the left-hand inequality in (5.9), to get  $\sup_{n \in \mathbb{N}} \tilde{\Pi}_{n-1}|Q_n| < \infty$  a.s. Again,  $\Pi_n \rightarrow 0$  a.s. implies that  $\tilde{M} := \min(|M|, 1)$  has  $P(\tilde{M} = 1) < 1$ , so we can use Lemma 5.4 on the pair  $(|Q|, \tilde{M})$ , that is, on the sequence  $(|Q_n|, \tilde{M}_n)$ . It gives, as at the end of the proof of Lemma 5.5, that the integral in (2.1) converges. So we have shown that (2.2) implies (2.1). Thus (2.1)–(2.4) are equivalent and imply (2.6).

The equivalence of (2.1) and (2.5) follows directly from Lemma 5.6.

For the converse, assume (2.7), and divide the failure of (2.1) into two cases: first when  $\Pi_n \rightarrow 0$  a.s. but the integral in (2.1) is infinite, second when it is not the case that  $\Pi_n \rightarrow 0$  a.s. In both cases we have proved that  $|Z_n(Z_0)| \xrightarrow{P} \infty$ , namely in Lemmas 5.5 and 5.8, respectively.  $\square$

PROOF OF COROLLARY 2.10. If  $Q = 0$  a.s. or  $P(M = 0) > 0$ , the series always converges, so assume  $P(Q = 0) < 1$  and  $P(M = 0) = 0$ . Then (2.1) gives  $\Pi_n \rightarrow 0$  a.s. so  $A_M(y) > 0$  when  $y > 0$  (see Remark 2.2), and by monotonicity and the definition of  $A_M(\cdot)$ ,

$$A_M(y) \leq cA_M(y/c) \leq cA_M(y), \quad y > 0$$

for all  $c \geq 1$ . Hence replacing  $|Q|$  by  $|Q|^b$  in the integral in (2.1) does no more than multiply the integral by a factor between 1 and  $b$ , if  $b > 1$ , and so has no effect on finiteness. Similarly if  $0 < b < 1$ .

The effect of replacing  $|M|$  by  $|M|^a$  is to replace  $X$  by  $aX$  in (1.8), and this simply multiplies  $A_M(y)$  by a factor between 1 and  $a$ , so again leaves the finiteness of the integral in (2.1) unaltered.  $\square$

PROOF OF PROPOSITION 2.12. Assume  $P(Q = 0) < 1$  and  $P(M = 0) = 0$ . Then  $A_M(y) > 0$  for all  $y > 0$  (Remark 2.2), and  $\hat{A}_M(y) \geq A_M(y)$ , so (2.1) implies (2.11). Conversely, let (2.11) hold. Instead of (5.10) we write

$$S_n = \sum_{j=1}^n (X_j^+ - X_j^-) = (1 - \hat{a}_n) \sum_{j=1}^n |X_j|,$$

where  $\hat{a}_n := 2 \sum_{j=1}^n X_j^- / \sum_{j=1}^n |X_j|$ . By the same reasoning as for (5.8) we have  $\limsup_{n \rightarrow \infty} \hat{a}_n < \hat{a}$  with probability 1, for some constant  $\hat{a} < 1$ . Thus

$$|\Pi_n| = \left( \prod_{j=1}^n \exp(-|X_j|) \right)^{1-\hat{a}_n} = \left( \prod_{j=1}^n \widehat{M}_j \right)^{1-\hat{a}_n},$$

where  $\widehat{M}_j := \exp(-|X_j|) \leq 1$  a.s. So if we set  $\widehat{\Pi}_n := \prod_{j=1}^n \widehat{M}_j$ , then we have  $\widehat{\Pi}_n \leq |\Pi_n| \leq \widehat{\Pi}_n^{1-\hat{a}}$  once  $\hat{a}_n < \hat{a}$ . Thus (5.3) holds with  $\Pi_n$  replaced by  $\widehat{\Pi}_n$  and  $X^+$  replaced by  $|X|$ . The working of Lemma 5.2 then gives, for Case 1, (5.5) with  $\sum_{j=1}^{n-1} X_j^+$  replaced by  $\sum_{j=1}^{n-1} |X_j|$ . But for Case 1,  $\sum_{j=1}^{n-1} X_j^- = o(\sum_{j=1}^{n-1} X_j^+)$  a.s., so (5.5) holds as stated and hence (5.6). In Case 2, (5.7) holds again. Thus (5.4) holds as stated and we again get (5.2).  $\square$

PROOF OF THEOREM 3.1. For (a) the existence of  $N$  is as noted in Remark 2.4, and the remaining conclusions follow by elementary deduction or as a special case of Vervaat [(1979), Theorem 1.5]. The whole of (b) is also given in the latter theorem.

For (c)(i), since  $R_n(R_0) \stackrel{L}{=} Z_n(R_0)$  marginally, and by our Theorem 2.1 the right-hand side converges to  $Z_\infty$  a.s., it follows that  $R_n(R_0) \rightarrow^L \nu$  and so  $\nu$  is a fixed point. By Vervaat (1979), Theorem 1.5, it is the only fixed point. The condition  $P(Q + Mc = c) < 1$  for all  $c$  prevents degeneracy.

Finally, in case (c)(ii), if there is a fixed point distribution  $\nu$  then on letting  $R_0$  have that distribution we find that  $R_n(R_0)$  has distribution  $\nu$  for all  $n$ . But Theorem 2.1 gives that  $|Z_n(R_0)| \rightarrow^P \infty$ . Since  $Z_n(R_0) \stackrel{L}{=} R_n(R_0)$  we arrive at a contradiction. Thus there is no fixed point. Theorem 2.1 and  $Z_n(R_0) \stackrel{L}{=} R_n(R_0)$  also give  $|R_n(R_0)| \rightarrow^P \infty$ .  $\square$

PROOF OF COROLLARY 4.1. Recall that  $P(Q = 0) < 1$  under (2.7) (see Remark 2.4).

(a) The condition says that  $-\infty \leq EX \leq 0$ , which prevents  $S_n \rightarrow \infty$ ; thus the first part of (2.1) does not hold and so Theorem 2.1 gives divergence.

(b)  $-\infty < E \log |M| < 0$  translates as  $0 < EX < \infty$ , so  $|\Pi_n| = e^{-S_n} \rightarrow 0$  a.s. and we have the first part of (2.1). By Theorem 2.1, convergence occurs if and only if the integral in (2.1) is finite. Since  $A_M(\infty) = EX^+ < \infty$ , this is equivalent to  $\int_{(C, \infty)} \log q \, dP(|Q| \leq q) < \infty$  for all  $C > 1$ , that is, to  $E \log^+ |Q| < \infty$ .

(c) Here,  $E \log |M| = -\infty$  means that  $EX^+ = \infty > EX^-$ , so we have  $|\Pi_n| = e^{-S_n} \rightarrow 0$  a.s. and again, by Theorem 2.1, convergence occurs if and only if the integral in (2.1) is finite. Since  $A_M(\infty) = EX^+ = \infty$ , the integrand is  $o(\log q)$  as  $q \rightarrow \infty$ , so it suffices for convergence that  $E \log^+ |Q| < \infty$  but this is no longer necessary.

(d) Here we are given that  $EX^+ = \infty = EX^-$ . By the theorem quoted in Remark 2.5,  $\Pi_n \rightarrow 0$  a.s. is in this case equivalent to  $J_- < \infty$ . Parts (i) and (ii) now are seen to be restatements of the possibilities for (2.1) to be satisfied, hence we complete (d) by applying Theorem 2.1.  $\square$

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