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# ON THE RUIN PROBABILITY OF THE GENERALISED ORNSTEIN-UHLENBECK PROCESS IN THE CRAMÉR CASE

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# ON THE RUIN PROBABILITY OF THE GENERALISED ORNSTEIN-UHLENBECK PROCESS IN THE CRAMÉR CASE

BY DAMIEN BANKOVSKY, CLAUDIA KLÜPPELBERG AND ROSS MALLER

#### **Abstract**

For a bivariate Lévy process  $(\xi_t, \eta_t)_{t\geq 0}$  and initial value  $V_0$ , define the generalised Ornstein–Uhlenbeck (GOU) process  $V_t := \mathrm{e}^{\xi_t}(V_0 + \int_0^t \mathrm{e}^{-\xi_{s-}} \,\mathrm{d}\eta_s),\ t\geq 0$ , and the associated stochastic integral process  $Z_t := \int_0^t \mathrm{e}^{-\xi_{s-}} \,\mathrm{d}\eta_s,\ t\geq 0$ . Let  $T_z := \inf\{t>0:\ V_t<0\mid V_0=z\}$  and  $\psi(z) := \mathrm{P}(T_z<\infty)$  for  $z\geq 0$  be the ruin time and infinite horizon ruin probability of the GOU process. Our results extend previous work of Nyrhinen (2001) and others to give asymptotic estimates for  $\psi(z)$  and the distribution of  $T_z$  as  $z\to\infty$ , under very general, easily checkable, assumptions, when  $\xi$  satisfies a Cramér condition.

Keywords: Exponential functionals of Lévy processes; generalised Ornstein–Uhlenbeck process; ruin probability; stochastic recurrence equation

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#### 1. Introduction

Let  $(\xi, \eta) = (\xi_t, \eta_t)_{t \ge 0}$  be a bivariate Lévy process on a filtered complete probability space  $(\Omega, \mathcal{F}, \mathbb{F}, P)$ , and define the generalised Ornstein–Uhlenbeck (GOU) process  $(V_t)_{t \ge 0}$  by

$$V_t := e^{\xi_t} \left( V_0 + \int_0^t e^{-\xi_{s-}} d\eta_s \right), \qquad t \ge 0,$$
 (1.1)

and the associated stochastic integral process  $Z = (Z_t)_{t \ge 0}$  by

$$Z_t := \int_0^t e^{-\xi_{s-}} d\eta_s.$$
 (1.2)

The random variable  $V_0$  is not necessarily independent of  $(V_t)_{t>0}$ . To avoid trivialities, assume that neither  $\xi$  nor  $\eta$  are identically 0.

Such processes have attracted attention over the last decade as continuous-time analogues of solutions to stochastic recurrence equations (SREs); cf. Carmona *et al.* [7], [8] and Erickson and Maller [12]. The link between SREs and the GOU process was made in [10]. GOU processes turn up naturally in stochastic volatility models (e.g. the continuous-time GARCH model of Klüppelberg *et al.* [21]), but most prominently as insurance risk models for perpetuities in life insurance or when the insurance company receives some stochastic return on investment; such investigations started with Dufresne [11] and Paulsen [27]. More references are given later.

This paper is intended to fill a gap left between Bankovsky [2] and Bankovsky and Sly [3], where more details on the insurance background can be found. Define

$$T_z := \inf\{t > 0 : V_t < 0 \mid V_0 = z\}, \quad z > 0$$

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(with the convention throughout that  $\inf \emptyset = \infty$ ), and let

$$\psi(z) := P\left(\inf_{t>0} V_t < 0 \mid V_0 = z\right) = P\left(\inf_{t>0} Z_t < -z\right) = P(T_z < \infty), \qquad z \ge 0,$$
 (1.3)

be the *infinite horizon ruin probability* for the GOU process. Note that  $\psi(z)$  is a nonincreasing function of z, and we can ask how fast it decreases as  $z \to \infty$ .

Our main result, Theorem 2.1, provides a very general asymptotic result for  $\psi(z)$  as  $z \to \infty$  for the case when  $\lim_{t\to\infty} Z_t$  exists as an almost-sure finite random variable and shows that, under a Cramér-like condition on  $\xi$ ,  $\psi(z)$  decreases approximately like a power law. This is an extension of a similar asymptotic result of Nyrhinen [26], who, like us, utilised a discrete-time result of Goldie [15] for the proof. We use more recent developments in the theory of discrete-time perpetuities and the continuous-time GOU process to update Nyrhinen's results. In Section 3 we provide some examples which cannot be dealt with by the prior results but satisfy the conditions of our theorem.

To conclude this introduction, we describe some previous literature relating to the GOU process and its ruin probability, beginning with those papers which examine the GOU process in its full generality. The process appears implicitly in the work of de Haan and Karandikar [10] as a continuous generalisation of an SRE. Basic properties are given by Carmona *et al.* [8]. A general survey of the GOU process and its applications is given by Maller *et al.* [24]. Exact conditions for no ruin ( $\psi(z) = 0$  for some  $z \ge 0$ ) are given by Bankovsky and Sly [3] whilst conditions for certain ruin ( $\psi(z) = 1$  for some  $z \ge 0$ ) are examined by Bankovsky [2].

The study of the GOU process is closely related to the study of integrals of the form Z, defined in (1.2). It was shown in Lindner and Maller [23] that stationarity of V is related to convergence of a stochastic integral constructed from  $(\xi, \eta)$  in a similar way to Z.

Among the few papers dealing with Z in its full generality, Erickson and Maller [12] gave necessary and sufficient conditions for the almost-sure convergence of  $Z_t$  to a random variable  $Z_{\infty}$  as  $t \to \infty$ , and Bertoin *et al.* [4] presented necessary and sufficient conditions for the continuity of the distribution of  $Z_{\infty}$ , when it exists. Fasen [13], using point process methods, gave an account of the extremal behaviour of a GOU process.

There are a larger number of papers dealing with V and Z when  $(\xi, \eta)$  is subject to restrictions. We discuss a selection of those papers which are relevant to ruin probability. Harrison [17] presented results on the ruin probability of V when  $\xi$  is a linear deterministic function and  $\eta$  is a Lévy process with finite variance. His approach is based on an exponential martingale argument, which corresponds to the Cramér case. The heavy-tailed case is investigated by Klüppelberg and Stadtmüller [20] and extended by Asmussen [1]. See also Klüppelberg  $et\ al.$  [22] and Maulik and Zwart [25].

Paulsen [27] generalised Harrison's results, and presented new ruin probability results for V, when  $\xi$  and  $\eta$  are independent with finite activities. This independent case is also treated in Kalashnikov and Norberg [18] and Paulsen [28], [29]. Chiu and Yin [34] generalised some of Paulsen's results to the case in which  $\eta$  is a jump-diffusion process. Cai [6] and Yuen *et al.* [36] presented results when  $\eta$  is a compound Poisson process.

Most works containing restrictions on  $(\xi, \eta)$  focus on the case in which  $Z_t$  converges to  $Z_\infty$  as  $t \to \infty$ ; cf. Yor [35] and Carmona  $et\,al.$  [7]. Gjessing and Paulsen [14] studied the distribution of  $Z_\infty$  when  $\xi$  and  $\eta$  are independent with finite activity, and obtained exact distributions for some special cases. Hove and Paulsen [30] used Markov chain Monte Carlo methods to find the distribution of  $Z_\infty$  in some special cases. Klüppelberg and Kostadinova [19] and Brokate  $et\,al.$  [5] provided results on the tail distribution of  $Z_\infty$  when  $\eta$  is a compound Poisson process plus drift, independent of  $\xi$ .

## 2. Main results

Our main results apply under a Cramér-like condition on  $\xi$ : assume that

$$\mathrm{E}\,\mathrm{e}^{-w\xi_1} = 1$$
 for some  $w > 0$ . (2.1)

The following consequences of (2.1) are well known and easily verified. Condition (2.1) implies that  $\operatorname{E}\xi_1$  is well defined, with  $\operatorname{E}\xi_1^- < \infty$ ,  $\operatorname{E}\xi_1^+ \in (0,\infty]$ , and  $\operatorname{E}\xi_1 \in (0,\infty]$ , and so  $\lim_{t\to\infty}\xi_t = \infty$  almost surely (a.s.). Furthermore,  $\operatorname{E}\operatorname{e}^{-\alpha\xi_1}$  is finite and nonzero for all  $\alpha\in[0,w]$ , and  $c(\alpha):=\ln\operatorname{E}\operatorname{e}^{-\alpha\xi_1}$  is finite at least for all  $\alpha\in[0,w)$ . The derivatives  $c'(\alpha)$  and  $c''(\alpha)$  are finite at least for all  $\alpha\in[0,w)$ , and  $c''(\alpha)\in(0,\infty]$  for all  $\alpha\geq0$ . So  $c(\alpha)$  is strictly convex for  $\alpha\in[0,\infty)$  and  $\mu^*:=c'(w)=-\operatorname{E}[\xi_1\operatorname{e}^{-w\xi_1}]\in(0,\infty]$ .

We will need the Fenchel-Legendre transform of c, defined as

$$c^*(v) := \sup\{\alpha v - c(\alpha) \colon \alpha \in \mathbb{R}\}, \qquad v \in \mathbb{R}. \tag{2.2}$$

Next, let

$$\alpha_0 := \sup\{\alpha \in \mathbb{R} : c(\alpha) < \infty, \ \mathbb{E} |Z_1|^{\alpha} < \infty\} \in [0, \infty], \tag{2.3}$$

and define the constant

$$x_0 := \lim_{\alpha \to \alpha_0 -} \frac{1}{c'(\alpha)} \in [0, \infty].$$

A distribution is *spread out* if it has a convolution power with an absolutely continuous component.

**Theorem 2.1.** Suppose that the following conditions hold:

- (C1)  $\psi(z) > 0$  for all  $z \ge 0$ ,
- (C2) there exists w > 0 such that  $\operatorname{E} e^{-w\xi_1} = 1$  (i.e. (2.1) holds).
- (C3) there exist  $\varepsilon > 0$  and p, q > 1 with 1/p + 1/q = 1 such that

$$E[e^{-\max\{1, w+\varepsilon\}p\xi_1}] < \infty \quad and \quad E[|\eta_1|^{\max\{1, w+\varepsilon\}q}] < \infty.$$

Then  $0 < x_0 < 1/\mu^* < \infty$ , the function

$$R(x) := \begin{cases} xc^*(1/x) & \text{for } x \in (x_0, 1/\mu^*), \\ w & \text{for } x \ge 1/\mu^*, \end{cases}$$

is finite and continuous on  $(x_0, \infty)$  and strictly decreasing on  $(x_0, 1/\mu^*)$ , and we have

$$\lim_{z \to \infty} (\ln z)^{-1} \ln P(T_z \le x \ln z) = -R(x) \quad \text{for every } x > x_0.$$
 (2.4)

In addition.

$$\lim_{z \to \infty} (\ln z)^{-1} \ln \psi(z) = -w. \tag{2.5}$$

If, further, the distribution of  $\xi_1$  is spread out then there exist constants  $C_- > 0$  and  $\kappa > 0$  such that

$$z^{w}\psi(z) = C_{-} + o(z^{-\kappa}) \quad \text{as } z \to \infty.$$
 (2.6)

**Remark 2.1.** (i)  $\psi(z) > 0$  for all  $z \ge 0$  is of course a logical assumption to make in the context of Theorem 2.1, though not necessarily easy to verify. Necessary and sufficient conditions for it in terms of the Lévy measure of  $(\xi, \eta)$  are given in [3]. The moment conditions in Theorem 2.1 are also easily expressed in terms of the Lévy measure of  $(\xi, \eta)$ ; cf. Sato [32, p. 159]. They imply that  $\mathrm{E}[\sup_{0 \le t \le 1} |Z_t|^{\max\{1, w+\varepsilon\}}] < \infty$  (see Lemma 5.1 below). We also have  $\mathrm{E}[\ln(\max\{1, |\eta_1|\})] < \infty$  in Theorem 2.1, and  $\lim_{t \to \infty} \xi_t = \infty$  a.s., so  $Z_t$  converges a.s. to a finite random variable  $Z_\infty$  as  $t \to \infty$  by Proposition 2.4 of [23] or Theorem 2 of [12].

(ii) Let  $\bar{Z}_t := Z_t - \inf_{0 \le s \le t} Z_s$  be the process reflected in its minimum, and set

$$(M, Q, \bar{L}) := (e^{-\xi_1}, Z_1, -e^{\xi_1}\bar{Z}_1).$$

Then the value  $C_{-}$  in (2.6) is given by Equation (2.19) of Goldie [15], namely,

$$C_{-} = \frac{1}{w\mu^{*}} \operatorname{E}\left[\left(\left(Q + M \min\left\{\bar{L}, \inf_{t>0} Z_{t}\right\}\right)^{-}\right)^{w} - \left(\left(M \inf_{t>0} Z_{t}\right)^{-}\right)^{w}\right].$$

When  $\xi$  and  $\eta$  are independent, it was pointed out by Paulsen [29] that this constant can be written in a slightly different form, which, by Theorem 4 of [3], is also true in the dependent case. Namely, let  $G(z) := P(Z_{\infty} \le z)$ ,  $h(z) := E[G(-V_{T_z}) \mid T_z < \infty] \in [0, 1]$ , and  $h := \lim_{z \to \infty} h(z)$ . Then

$$C_{-} = \frac{1}{w\mu^*h} \operatorname{E}[((Q + MZ_{\infty})^{-})^w - ((MZ_{\infty})^{-})^w].$$

(iii) The requirement that  $\xi_1$  is spread out can be replaced with the less restrictive requirement that  $\xi_T$  be spread out, where T is uniformly distributed on [0, 1] and independent of  $\xi$ . We omit details of this, which can be carried out as in [29].

## 3. Examples

In this section we provide examples of Lévy processes for which (C1)–(C3) of Theorem 2.1 are satisfied. Note that (C2) and (C3) involve only the marginal processes  $\xi$  and  $\eta$ , and they apply to all examples treated in the literature so far; cf. Klüppelberg and Kostadinova [19] for detailed references. The only condition which may involve dependence between  $\xi$  and  $\eta$  is (C1).

We denote the characteristic triplet of  $(\xi, \eta)$  by  $((\tilde{\gamma}_{\xi}, \tilde{\gamma}_{\eta}), \Sigma_{\xi, \eta}, \Pi_{\xi, \eta})$ . The characteristic triplet of the marginal process  $\xi$  is denoted by  $(\gamma_{\xi}, \sigma_{\xi}^2, \Pi_{\xi})$ , where

$$\gamma_{\xi} = \tilde{\gamma}_{\xi} + \int_{\{|x| < 1\} \cap \{x^2 + y^2 \ge 1\}} x \Pi_{\xi, \eta}(\mathbf{d}(x, y))$$
(3.1)

and  $\sigma_{\xi}^2$  is the upper-left entry in the matrix  $\Sigma_{\xi,\eta}$ . Similarly for  $\eta$ . The random jump measure and Brownian motion components of  $(\xi,\eta)$  will be denoted respectively by  $N_{\xi,\eta}$  and  $(B_{\xi},B_{\eta})$ ; see Section 1.1 of [3] for further details.

**Example 3.1.** (Bivariate compound Poisson process with drift.) Let  $(N_t)_{t\geq 0}$  be a Poisson process with intensity  $\lambda > 0$ , and, independent of it, let  $(X_i, Y_i)_{i\in \mathbb{N}}$  be an independent and identically distributed (i.i.d.) sequence of random 2-vectors. For  $\gamma_{\xi}, \gamma_{\eta} \in \mathbb{R}$ , set

$$(\xi_t, \eta_t) := (\gamma_{\xi}, \gamma_{\eta}) t + \sum_{i=1}^{N_t} (X_i, Y_i), \qquad t \ge 0,$$

with  $E|X| < \infty$ , and  $\lambda$ ,  $\gamma_{\xi}$ , and EX such that  $\gamma_{\xi} + \lambda EX > 0$ . For this process,

$$c(\alpha) = \ln \mathbf{E} \, \mathbf{e}^{-\alpha \xi_1} = -\alpha \gamma_{\xi} - \lambda (1 - \mathbf{E} \, \mathbf{e}^{-\alpha X}) < \infty$$

for  $\alpha \in \mathbb{R}$  such that  $\operatorname{E} e^{-\alpha X}$  is finite, with  $c'(0) = -\gamma_{\xi} - \lambda \operatorname{E} X < 0$ .

We consider the special case where (X, Y) is bivariate Gaussian with mean  $(m_X, m_Y)$  and positive definite covariance matrix

$$\Sigma_{X,Y} := \begin{pmatrix} \sigma_X^2 & \sigma_{X,Y} \\ \sigma_{X,Y} & \sigma_Y^2 \end{pmatrix}.$$

Then (C3) obviously holds. For (C2), note that

$$c(\alpha) = -\alpha \gamma_{\xi} - \lambda \left( 1 - \exp\left\{ -m_X \alpha + \frac{\sigma_X^2 \alpha^2}{2} \right\} \right) \to \infty \quad \text{as } \alpha \to \infty.$$
 (3.2)

Consequently, a Lundberg coefficient exists and (C2) is satisfied. To establish (C1), we note that  $(\xi, \eta)$  is a finite variation process and invoke Remark 2(2) of [3], also using the notation from that paper. In fact, by Remark 2(2),  $\psi(z) = 0$  for some z > 0 would imply that  $P_{X,Y}(A_3) = P(X \le 0, Y \le 0) = 0$ , which obviously is not the case. So condition (C1) holds.

**Example 3.2.** (Brownian motion with drift.) In this case

$$(\xi_t, \eta_t) = (\gamma_{\xi}, \gamma_{\eta})t + (B_{\xi,t}, B_{\eta,t}), \qquad t \ge 0$$

where  $\gamma_{\xi} > 0$ , and  $(B_{\xi}, B_{\eta})_t$  is bivariate Brownian motion with mean 0 and positive definite covariance matrix; it is easily seen that (C1)–(C3) are satisfied.

**Example 3.3.** (Jump diffusion  $\xi$  and Brownian motion  $\eta$ .) Let  $(B_t)_{t\geq 0}$  be Brownian motion with mean 0 and variance  $\sigma^2$ , let  $(N_t)_{t\geq 0}$  be a Poisson process with intensity  $\lambda > 0$ , and let  $(X_i)_{i\in\mathbb{N}}$  be i.i.d. random variables, all independent. Set

$$(\xi_t, \eta_t) = (\gamma_{\xi}, \gamma_{\eta})t + \left(B_t + \sum_{i=1}^{N_t} X_i, B_t\right), \qquad t \ge 0$$

where  $\gamma_{\xi} > 0$ , and assume that  $\gamma_{\xi} + \lambda E X > 0$ . Condition (C1) holds, since the Gaussian covariance matrix is of the form

$$\Sigma_{\xi,\eta} := \begin{pmatrix} \sigma^2 + \lambda E X^2 & 1 \\ 1 & \sigma^2 \end{pmatrix},$$

and, hence, is not of the form excluded by Theorem 1 of [3]. Moreover,  $c(\alpha)$  is the same as in (3.2) with the addition of a term  $\alpha^2 \sigma^2 / 2$ , so again  $c'(0) = -\gamma_{\xi} - \lambda E X < 0$ .

- (a) Now assume that X is, as in the Merton model, normally distributed with mean  $m_X$  and variance  $\sigma_X$ . Then (C2) and (C3) are satisfied just as in Example 3.1.
- (b) The picture changes slightly when we consider Laplace distributed X with density  $f(x) = \rho e^{-\rho|x|}/2$  for  $x \in \mathbb{R}$  and  $\rho > 0$ . Then  $E e^{-\alpha X} = \rho((\rho + \alpha)^{-1} + (\rho \alpha)^{-1})/2$  for  $-\rho < \alpha < \rho$  with singularities at  $-\rho$  and  $\rho$ . Moreover,

$$c'(\alpha) = -\gamma_{\xi} + \alpha \sigma^2 + \lambda \frac{\rho}{2} \left( \frac{1}{(\rho - \alpha)^2} - \frac{1}{(\rho + \alpha)^2} \right),$$

implying that  $c'(0) = -\gamma_{\xi} < 0$ . So a Lundberg coefficient w > 0 exists. Since the normal random variable  $B_1$  has absolute moments of every order, for (C3) to hold, it suffices that  $w < \rho$ , which is guaranteed, since  $\rho$  is a singularity of c.

**Example 3.4.** (Subordinated Brownian motion  $\xi$  and spectrally positive  $\eta$ .) Let  $(B_t)_{t\geq 0}$  be a standard Brownian motion, and let  $(S_t)_{t\geq 0}$  be a driftless subordinator with  $\Pi_S\{\mathbb{R}\} = \infty$ . For constants  $\mu$ ,  $\gamma_{\xi}$ , and  $\gamma_{\eta}$ , define

$$(\xi_t, \eta_t) = (\gamma_{\xi}, \gamma_{\eta})t + (B(S_t) + \mu S_t, S_t), \qquad t \ge 0.$$

Subordinated Brownian motions play an important role in financial modeling; cf. Bankovsky and Sly [9, Chapter 4]. The bivariate process above has joint Laplace transform

$$\exp\{(\alpha_1 \gamma_{\xi} + \alpha_2 \gamma_{\eta})t\} \operatorname{E}[\exp\{\alpha_1 (B(S_t) + \mu S_t) + \alpha_2 S_t\}]$$

$$= \exp\{(\alpha_1 \gamma_{\xi} + \alpha_2 \gamma_{\eta})t\} \operatorname{E}[\exp\{\Psi_B(\alpha_1) S_t + (\alpha_1 \mu + \alpha_2) S_t\}]$$

$$= \exp\{t[\Psi_S(\Psi_B(\alpha_1) + \alpha_1 \mu + \alpha_2) + \alpha_1 \gamma_{\xi} + \alpha_2 \gamma_{\eta}]\},$$

where  $\Psi_B$  and  $\Psi_S$  are the Laplace exponents of B and S, respectively. Thus,  $\Psi_B(\alpha) = -\alpha^2/2$ . By setting  $\alpha_2 = 0$  and t = 1, we obtain

$$c(\alpha) = \ln \mathbf{E} \, \mathbf{e}^{-\alpha \xi_1} = \Psi_S(\Psi_B(-\alpha) - \alpha \mu) - \alpha \gamma_\xi = \Psi_S\left(-\frac{1}{2}\alpha^2 - \alpha \mu\right) - \alpha \gamma_\xi.$$

Consider the variance gamma model with parameters c,  $\lambda > 0$ , where S is a gamma subordinator with Lévy density  $\rho(x) = cx^{-1}e^{-\lambda x}$  for x > 0 and Laplace transform  $Ee^{-uS_t} = (1+u/\lambda)^{-ct}$ . Assume that  $\gamma_{\xi} + c\mu/\lambda > 0$  and  $\gamma_{\eta} \leq 0$ . Now,  $\gamma_{S}(u) = -c\ln(1-u/\lambda)$ , giving

$$c(\alpha) = -\alpha \gamma_{\xi} - c \ln \left( 1 + \frac{\alpha \mu}{\lambda} - \frac{\alpha^2}{2\lambda} \right).$$

Here  $c(\alpha)$  is well defined for  $\alpha \in (\mu - \sqrt{\mu^2 + 2\lambda}, \mu + \sqrt{\mu^2 + 2\lambda})$ , which includes 0, and  $c'(0) = -\gamma_{\xi} - c\mu/\lambda < 0$ . Then, since  $c(\mu + \sqrt{\mu^2 + 2\lambda}) = \infty$ , the Lundberg coefficient w exists.

In order to check (C1), we have, in the notation of Theorem 1 of [3],  $\Pi_{\xi,\eta}(A_2) = \Pi_{\xi,\eta}(A_3) = 0$ , since  $\eta$  has only positive jumps, and  $\theta_2 = 0$ . Now, with  $u \ge 0$ ,

$$A_4^u = \{x \le 0, y \ge 0 : y < u(e^{-x} - 1)\} = \{x \ge 0, y \ge 0 : y < u(e^x - 1)\}.$$

Since  $\Pi_{\eta}(\mathbb{R}) = \infty$ ,  $\eta$  has jumps arbitrarily close to 0, and we have  $\Pi_{\xi,\eta}(A_4^u) > 0$  for u > 0, while  $\Pi_{\xi,\eta}(A_4^0) = 0$ . Thus,  $\theta_4 := \inf\{u \geq 0 \colon \Pi_{\xi,\eta}(A_4^u) > 0\} = 0$ . There is no Gaussian component, so  $\sigma_{\xi}^2 = 0$ , which puts us in the situation of the second item of Theorem 1 of [3], and to verify that  $\psi(z) > 0$  for all  $z \geq 0$ , we only need (since  $\theta_2 = \theta_4 = 0$ )

$$g(0) = \widetilde{\gamma}_{\eta} - \int_{x^2 + y^2 < 1} y \Pi_{\xi, \eta}(dx, dy) < 0.$$

But, by (3.1),

$$\widetilde{\gamma}_{\eta} = \gamma_{\eta} - \int_{0 < y < 1, \ x^2 + y^2 > 1} y \Pi_{\xi, \eta}(\mathrm{d}x, \ \mathrm{d}y) \le \gamma_{\eta};$$

thus,  $g(0) < \gamma_{\eta} \le 0$ , since we chose  $\gamma_{\eta} \le 0$ . Hence, (C1) holds in this model.

## 4. Discrete-time background and preliminaries

Our continuous-time asymptotic results will be transferred across from discrete-time versions, and our first task in the present section is to show how  $(V_t)_{t\geq 0}$  can be expressed as a solution of one of two SREs, and give the associated discrete stochastic series for  $(Z_t)_{t\geq 0}$ . Earlier papers in this area also adopted this approach and we will tap into some of their results in proving Theorem 2.1.

We begin by describing the discrete-time setup we use. For  $n \in \mathbb{N}$ , consider the SRE

$$Y_n = A_n Y_{n-1} + B_n, (4.1)$$

where  $(A_n, B_n)_{n \in \mathbb{N}}$  is an i.i.d. sequence of  $\mathbb{R}^2$ -valued random vectors independent of an initial random variable  $Y_0$ . The recursion in (4.1) can be solved in the form

$$Y_n = Y_0 \prod_{j=1}^n A_j + \sum_{i=1}^n \prod_{j=i+1}^n A_j B_i$$
 (4.2)

(with  $\prod_{j=n+1}^n = 1$ ). From (1.1) we can write, for  $n \in \mathbb{N}$ , denoting  $\int_{(n-1)+}^n := \int_{(n-1,n]}$ ,

$$V_n = e^{\xi_n - \xi_{n-1}} \left( e^{\xi_{n-1}} \left( V_0 + \int_0^{n-1} e^{-\xi_{s-1}} d\eta_s \right) \right) + e^{\xi_n} \int_{(n-1)+}^n e^{-\xi_{s-1}} d\eta_s.$$
 (4.3)

Thus, if we let  $Y_0 = V_0$  and define the  $\mathbb{R}^2$ -valued random vectors

$$(A_n, B_n) := \left(e^{\xi_n - \xi_{n-1}}, e^{\xi_n} \int_{(n-1)+}^n e^{-\xi_{s-1}} d\eta_s\right),$$

then  $V_n$  satisfies (4.1). An alternative formulation considers for  $n \in \mathbb{N}$  the SRE

$$Y_n = C_n Y_{n-1} + C_n D_n, (4.4)$$

where  $(C_n, D_n)_{n \in \mathbb{N}}$  is an i.i.d. sequence independent of  $Y_0$ . The solution is

$$Y_n = Y_0 \prod_{j=1}^n C_i + \sum_{i=1}^n \prod_{j=i}^n C_j D_i.$$

Using (4.3), it is clear that  $V_n$  is a solution of (4.4) if we let  $V_0 = Y_0$  and define

$$(C_n, D_n) := \left( e^{\xi_n - \xi_{n-1}}, e^{\xi_{n-1}} \int_{(n-1)+}^n e^{-\xi_{s-1}} d\eta_s \right).$$
 (4.5)

Then it is easily verified that

$$Z_n = \sum_{i=1}^n \prod_{j=1}^{i-1} C_j^{-1} D_i$$
 (4.6)

(with  $\prod_{j=1}^{0} = 1$ ). Note that even when  $\xi$  and  $\eta$  are independent, the random variables  $A_n$  and  $B_n$  may be dependent, and similarly for  $C_n$  and  $D_n$ . But we have the following result.

**Lemma 4.1.**  $(A_n, B_n)_{n \in \mathbb{N}}$  and  $(C_n, D_n)_{n \in \mathbb{N}}$  are i.i.d sequences.

*Proof.* We begin by proving that the sequence  $(C_n, D_n)_{n \in \mathbb{N}}$  is i.i.d. Fix  $n \in \mathbb{N}$ , and define the new Lévy process  $(\bar{\xi}_s, \bar{\eta}_s) := (\xi_{(n-1)+s} - \xi_{n-1}, \eta_{(n-1)+s} - \eta_{n-1})$  for  $s \ge 0$ . Thus,  $(\bar{\xi}_s, \bar{\eta}_s)_{s \ge 0} \stackrel{\text{\tiny D}}{=} (\xi_s, \eta_s)_{s \ge 0}$ . Note that we can bring the term  $e^{\xi_{n-1}}$  through the integral sign in (4.5) and write  $D_n = \int_{(n-1)+}^n e^{-(\xi_s - -\xi_{n-1})} d\eta_s$ . The Lévy process  $(\xi, \eta)$  has independent increments, so  $(C_n, D_n)$  is independent of  $(C_m, D_m)$  for every  $n \ne m$ . Now

$$(C_n, D_n) = \left( e^{\xi_n - \xi_{n-1}}, \int_{(n-1)+}^n e^{-(\xi_{s-} - \xi_{n-1})} d\eta_s \right)$$

$$= \left( e^{\bar{\xi}_1}, \int_{0+}^1 e^{-\bar{\xi}_{s-}} d\bar{\eta}_s \right)$$

$$\stackrel{\text{D}}{=} \left( e^{\bar{\xi}_1}, \int_{0+}^1 e^{-\xi_{s-}} d\eta_s \right)$$

$$= (C_1, D_1).$$

Thus, we have proved that  $(C_n, D_n)_{n \in \mathbb{N}}$  is an i.i.d. sequence. This implies that  $(C_n, C_n D_n)_{n \in \mathbb{N}}$  is also an i.i.d. sequence, and then  $(A_n, B_n)_{n \in \mathbb{N}}$  is also an i.i.d. sequence since

$$(C_n, C_n D_n) = \left( e^{\xi_n - \xi_{n-1}}, e^{\xi_n} \int_{(n-1)+}^n e^{-\xi_{s-1}} d\eta_s \right) = (A_n, B_n).$$

In order to directly access particular results from previous papers, when discretizing V, we will use the approach via recursion (4.1) and sequence (4.2), whereas when discretizing Z, we will use the approach via series (4.6). There has been significant attention paid to sequences of the form (4.2) and (4.6), and they are linked via the fixed point of the same SRE; see Goldie and Maller [16] and Vervaat [33].

Next we describe two important papers relating to the GOU process and its ruin time. In them,  $\xi$  and  $\eta$  are general Lévy processes, possibly dependent. The relevant papers are Nyrhinen [26] and Paulsen [29], which are very closely related to Theorem 2.1.

Nyrhinen [26] contains asymptotic ruin probability results for the GOU process, in which  $(\xi, \eta)$  is allowed to be an arbitrary bivariate Lévy process. He discretized the stochastic integral process Z and deduced asymptotic results in the continuous-time setting from similar discrete-time results. We describe Nyrhinen's results in some detail, and then make some comments.

Let  $(M_n, Q_n, L_n)_{n \in \mathbb{N}}$  be i.i.d. random vectors with P(M > 0) = 1 and  $(M, Q, L) \equiv (M_1, Q_1, L_1)$ . Define the sequence  $(X_n)_{n \in \mathbb{N}}$  by

$$X_n = \sum_{i=1}^n \prod_{j=1}^{i-1} M_j Q_i + \prod_{j=1}^n M_j L_n \quad \text{with } X_0 = 0.$$
 (4.7)

For u>0, define the passage time  $\tau_u^X:=\inf\{n\in\mathbb{N}\colon X_n>u\}$  and the function  $c_M(\alpha):=\ln E\,M^\alpha$ . Assume that there exists a  $w^+>0$  such that  $E\,M^{w^+}=1$ . Define

$$\alpha_0^+ := \sup\{\alpha \in \mathbb{R} : c_M(\alpha) < \infty, \ \operatorname{E}|Q|^{\alpha} < \infty, \ \operatorname{E}(ML^+)^{\alpha} < \infty\} \in [0, \infty]. \tag{4.8}$$

Also, let

$$\bar{y} := \sup \left\{ y \in \mathbb{R} \colon P\left(\sup_{n \in \mathbb{N}} X_n > y\right) > 0 \right\} \in (-\infty, \infty]. \tag{4.9}$$

Nyrhinen provided asymptotic results for  $X_n$  under the following hypothesis.

(H) Suppose that  $0 < w^+ < \alpha_0^+ \le \infty$  and  $\bar{y} = \infty$ .

Under hypothesis (H), and assuming that P(M > 1) > 0, the following quantities are well defined:

$$\mu^+ := \frac{1}{c_M'(w)} \in (0, \infty) \quad \text{and} \quad x_0^+ := \lim_{t \to \alpha_0^+ -} \frac{1}{c_M'(t)} \in [0, \infty).$$

Let  $c_M^*(v)$  be the Fenchel–Legendre transform of  $c_M$ , as in (2.2). Define the function

$$R: (x_0^+, \infty) \to \mathbb{R} \cup \{\pm \infty\}$$

by

$$R(x) := \begin{cases} x c_M^*(1/x) & \text{for } x \in (x_0^+, 1/\mu^+), \\ w & \text{for } x \ge 1/\mu^+. \end{cases}$$

In our situation, R is finite and continuous on  $(x_0^+, \infty)$ , and strictly decreasing on  $(x_0^+, 1/\mu^+)$ .

**Proposition 4.1.** (Nyrhinen's main discrete results [26, Theorems 2 and 3].) Assume that hypothesis (H) holds. Then the following statements also hold.

(i) For every  $x > x_0$ ,

$$\lim_{u \to \infty} (\ln u)^{-1} \ln P(\tau_u^X \le x \ln u) = -R(x)$$
 (4.10)

and

$$\lim_{u \to \infty} (\ln u)^{-1} \ln P(\tau_u^X < \infty) = -w.$$
(4.11)

(ii) If the distribution of  $\ln M$  is spread out, there exist constants  $C_+ > 0$  and  $\kappa > 0$  such that

$$u^{w^{+}} P(\tau_{u}^{X} < \infty) = C_{+} + o(u^{-\kappa}) \quad as \ u \to \infty.$$
 (4.12)

The constant  $C_+$  can be obtained from the formula in Theorem 6.2 and Equation (2.18) of Goldie [15]. Nyrhinen continues in his Theorem 3 to give equivalences for the condition  $\bar{y} = \infty$ , but they are difficult to verify, as he admits. We discuss these more fully later.

Nyrhinen's continuous result is obtained by applying his discrete results to the case

$$(M_n, Q_n) = \left( e^{-(\xi_n - \xi_{n-1})}, e^{\xi_{n-1}} \int_{(n-1)+}^n e^{-\xi_{s-1}} d\eta_s \right) = (C_n^{-1}, D_n)$$
 (4.13)

(cf. (4.5)), and

$$L_n := e^{\xi_n} \left( \sup_{n-1 < t < n} \int_{(n-1)+}^t e^{-\xi_{s-}} d\eta_s - \int_{(n-1)+}^n e^{-\xi_{s-}} d\eta_s \right). \tag{4.14}$$

The sequence  $(M_n, Q_n, L_n)_{n \in \mathbb{N}}$  is i.i.d., as follows by an easy extension of our proof of Lemma 4.1. With these allocations,  $Z_n$  can be written via (4.6) in the form

$$Z_n = \sum_{i=1}^n \prod_{j=1}^{i-1} M_j Q_i = X_n - L_n \prod_{j=1}^n M_j.$$
 (4.15)

Nyrhinen proved the following result with equality in distribution.

**Proposition 4.2.** Let  $(M_n, Q_n, L_n)$ , and let  $Z_n$  be as defined in (4.14) and (4.15). Define  $X_n$  as in (4.7). Then

$$\sup_{n-1 < t \le n} Z_t = X_n \quad and \quad \sup_{0 \le t \le n} Z_t = \max_{m=1,\dots,n} X_m.$$

*Proof.* For  $n \in \mathbb{N}$ , we have

$$\sup_{n-1 < t \le n} Z_t = Z_{n-1} + \sup_{n-1 < t \le n} \int_{(n-1)+}^t e^{-\xi_{s-}} d\eta_s$$

$$= Z_{n-1} + \int_{(n-1)+}^n e^{-\xi_{s-}} d\eta_s + e^{-\xi_n} L_n$$

$$= X_n - \prod_{j=1}^n M_j L_n + e^{-\xi_n} L_n$$

$$= X_n.$$

This further implies that  $\sup_{0 \le t \le n} Z_t = \max_{m=1,...,n} X_m$ .

Define the first passage time of Z above u>0 by  $\tau_u^Z:=\inf\{t\geq 0\colon Z_t>u\}$ . Then Proposition 4.2 implies that, for all t>0,

$$P(\tau_u^Z \le t) = P(\tau_u^X \le t)$$
 and  $P(\tau_u^Z < \infty) = P(\tau_u^X < \infty)$ .

So (4.10) and (4.11) hold with  $\tau_u^X$  replaced by  $\tau_u^Z$ , when hypothesis (H) is satisfied for the associated values of  $(M_n, Q_n, L_n)$ . If, further, the distribution of  $\ln M$  is spread out then (4.12) holds with  $\tau_u^X$  replaced by  $\tau_u^Z$ . This is the content of Theorem 4 and Corollary 5 of [26].

## **Remark 4.1.** We make some comments on [26].

(i) We begin with the discrete results. Firstly, the sequence  $X_n$  defined in (4.7) converges as  $n \to \infty$  a.s. to a finite random variable under hypothesis (H). To see this, note that if we choose  $L_n = L$  then  $X_n$  is the inner iteration sequence  $I_n(L)$  for the random equation  $\phi(t) = Mt + Q$ . Goldie and Maller [16] proved that  $I_n(L)$  converges a.s. to a finite random variable if and only if  $\prod_{j=1}^n M_j \to 0$  a.s. as  $n \to \infty$  and  $I_{M,Q} < \infty$ , where  $I_{M,Q}$  is an integral involving the marginal distributions of M and Q. Since these conditions have no dependence on the distribution of L, it is clear that they are precisely those under which  $X_n$  converges a.s. for i.i.d.  $(M_n, Q_n, L_n)$ . We now show that these conditions are in fact satisfied under hypothesis (H), and, thus, the sequences  $X_n$  and  $\sum_{i=1}^n \prod_{j=1}^{i-1} M_j Q_i$  converge a.s., and to the same finite random variable.

Under hypothesis (H) and our assumption P(M = 0) = 0,  $E \ln M$  is well defined and  $E \ln M \in [-\infty, 0)$ . Hence the random walk  $S_n := \sum_{j=1}^n (-\ln M_j) = -\ln \prod_{j=1}^n M_j$  drifts to  $\infty$  a.s., and it follows that  $\prod_{j=1}^n M_j \to 0$  a.s. as  $n \to \infty$ . Since  $\alpha_0^+ > 0$ , there exists s > 0 such that  $E |Q|^s < \infty$ ; thus,  $E \ln^+ |Q| < \infty$ . Hence, Corollary 4.1 of [16] implies that the integral condition  $I_{M,Q} < \infty$  is satisfied and the sequence  $\sum_{i=1}^n \prod_{j=1}^{i-1} M_j Q_i$  converges a.s.

(ii) Nyrhinen transfers his discrete results into continuous time, but the corresponding results are difficult to apply in general. The most problematic assumption is his condition  $\bar{y}=\infty$  (see (4.9)). In our notation, this is equivalent to the condition  $\psi(z)>0$  for all  $z\geq 0$ . Theorem 1 of [3] gives necessary and sufficient conditions on the Lévy measure of  $(\xi,\eta)$  for this, which are amenable to verification in special cases, as we showed in Section 3. Verifying Nyrhinen's condition  $0< w^+<\alpha_0^+\leq \infty$  requires finiteness of powers of  $\mathrm{E}\,|Z_1|$  and  $\mathrm{E}[\sup_{0< t\leq 1}|Z_t|]$ . These conditions would be more conveniently stated in terms of the characteristic triplet of  $(\xi,\eta)$  or (at least) the marginal distributions of  $\xi$  and  $\eta$ . In the special case that  $\xi$  and  $\eta$  are independent Lévy processes, Theorem 3.2 of [29] does exactly that. However, problems remain. In Paulsen [29], the condition  $\bar{y}=\infty$  is assumed to be true whenever  $\xi$  and  $\eta$  are independent and  $\eta$  is not a subordinator. However, this claim is false. To see this, let  $(\xi,\eta)_t:=(t+N_t,-t)$ , where N is a Poisson process with jump times  $0<\tau_0<\tau_1<\cdots$ . This example trivially satisfies

all the conditions in Paulsen's Theorem 3.2. However, using Itô's formula for semimartingales and some simple manipulation, we obtain  $Z_t = -1 + (e - 1) \sum_{i=1}^{N_t} e^{-\tau_i - i} + e^{-t - N_t}$ , and, hence,  $\inf_{t>0} Z_t \ge -1$  a.s. (The claim does hold if extra conditions are imposed, in line with Remark 2(3) of [3].) Finally, it would be desirable to remove the finite-mean assumption for  $\xi$  in [29] and replace the moment conditions in [29], which are sufficient for convergence of  $Z_t$ , with the precise necessary and sufficient conditions given in Goldie and Maller [16]. Our Theorem 2.1 addresses all of the above concerns in the most general setting.

#### 5. Proof of Theorem 2.1

The proof requires the following lemma, which was stated but not proved in [2].

**Lemma 5.1.** Suppose that there exist r > 0 and p, q > 1 with 1/p + 1/q = 1 such that  $\operatorname{E} \operatorname{e}^{-\max\{1,r\}p\xi_1} < \infty$  and  $\operatorname{E} |\eta_1|^{\max\{1,r\}q} < \infty$ . Then

$$\mathrm{E}\bigg[\sup_{0\leq t\leq 1}|Z_t|^{\max\{1,r\}}\bigg] = \mathrm{E}\bigg[\sup_{0\leq t\leq 1}\bigg|\int_0^t\mathrm{e}^{-\xi_{s-}}\,\mathrm{d}\eta_s\bigg|^{\max\{1,r\}}\bigg] < \infty. \tag{5.1}$$

*Proof.* For ease of notation, let  $k := \max\{1, r\}$ . Assume that there exist r > 0 and p, q > 1 with 1/p + 1/q = 1 such that  $\operatorname{E} \operatorname{e}^{-kp\xi_1} < \infty$  and  $\operatorname{E} |\eta_1|^{kq} < \infty$ . We prove the lemma first for the case in which  $\operatorname{E} \eta_1 = 0$ . Since  $\eta$  is a Lévy process, this implies that  $\eta$  is a càdlàg martingale. Since  $\xi$  is càdlàg,  $\operatorname{e}^{-\xi}$  is a locally bounded process and, hence, Z is a local martingale for  $\mathbb F$  by the construction of the stochastic integral (see, e.g. [31, Theorem 17, Section III.5]). Since, additionally,  $Z_0 = 0$ , the Burkholder–Davis–Gundy inequalities ensure that, for our choices of p, q, and k, there exists b > 0 such that

$$E\left[\sup_{0 \le t \le 1} \left| \int_{0}^{t} e^{-\xi_{s-}} d\eta_{s} \right|^{k} \right] \le b E\left[\left[\int_{0}^{z} e^{-\xi_{s-}} d\eta_{s}, \int_{0}^{z} e^{-\xi_{s-}} d\eta_{s}\right]_{z=1}^{z=k/2}\right] \\
= b E\left[\left(\int_{0}^{1} e^{-2\xi_{s-}} d[\eta, \eta]_{s}\right)^{k/2}\right] \\
\le b E\left[\left(\int_{0}^{1} \sup_{0 \le t \le 1} e^{-2\xi_{t}} d[\eta, \eta]_{s}\right)^{k/2}\right],$$

where in the second inequality recall that  $[\eta, \eta]_s$  is increasing. (The notation  $[\cdot, \cdot]$  denotes the quadratic variation process.) The last expression equals

$$b \, \mathrm{E} \Big[ \sup_{0 \le t \le 1} \mathrm{e}^{-k\xi_t} [\eta, \eta]_1^{k/2} \Big] \le b \Big( \mathrm{E} \Big[ \sup_{0 \le t \le 1} \mathrm{e}^{-pk\xi_t} \Big] \Big)^{1/p} (\mathrm{E} [[\eta, \eta]_1^{qk/2}])^{1/q},$$

where the inequality follows for our choices of p and q by Hölder's inequality. Since  $k \ge 1$  and q > 1, the Burkholder–Davis–Gundy inequalities give the existence of c > 0 such that (using Doob's inequality for the second inequality)

$$\mathrm{E}[[\eta,\eta]_1^{qk/2}] \leq \frac{1}{c} \mathrm{E}\Big[\sup_{0 < t < 1} |\eta_t|^{qk}\Big] \leq \frac{8}{c} \mathrm{E}[|\eta_1|^{qk}] < \infty.$$

Thus, it suffices to prove that  $\mathrm{E}[\sup_{0 \le t \le 1} \mathrm{e}^{-pk\xi_t}] < \infty$ . Now  $Y_t := \mathrm{e}^{-pk\xi_t}/c^t$ , where  $c := \mathrm{E}\,\mathrm{e}^{-pk\xi_1} \in (0,\infty)$  is a nonnegative martingale, and it follows by Doob's maximal inequality

that

$$\mathbb{E}\left[\sup_{0\leq t\leq 1} e^{-pk\xi_t}\right] \leq \max\{1,c\} \,\mathbb{E}\left[\sup_{0\leq t\leq 1} \frac{e^{-pk\xi_t}}{c^t}\right] \leq \max\left\{\frac{1}{c},1\right\} \left(\frac{pk}{pk-1}\right)^{pk} \,\mathbb{E}\,e^{-pk\xi_1} < \infty.$$

Hence, the lemma is proved for the case in which E  $\eta_1 = 0$ . In general, write

$$\begin{split} \mathbf{E} & \left[ \sup_{0 \le t \le 1} \left| \int_0^t \mathrm{e}^{-\xi_{s-}} \, \mathrm{d}\eta_s \right|^k \right] \\ &= \mathbf{E} \left[ \sup_{0 \le t \le 1} \left| \int_0^t \mathrm{e}^{-\xi_{s-}} \, \mathrm{d}(\eta_s - \mathbf{E}\,\eta_1 + \mathbf{E}\,\eta_1) \right|^k \right] \\ &\le \mathbf{E} \left[ \left( \sup_{0 \le t \le 1} \left| \int_0^t \mathrm{e}^{-\xi_{s-}} \, \mathrm{d}(\eta_s - \mathbf{E}\,\eta_1) \right| + |\mathbf{E}\eta_1| \sup_{0 \le t \le 1} \left| \int_0^t \mathrm{e}^{-\xi_{s-}} \, \mathrm{d}s \right| \right)^k \right], \end{split}$$

in which the first term on the right-hand side is finite by the first part of the proof. An application of Minkowski's inequality to the second term on the right-hand side completes the proof.

**Remark 5.1.** If  $\xi$  and  $\eta$  are independent, then Hölder's inequality is not required in the proof of Lemma 5.1, and a simpler independence argument shows that (5.1) holds if

$$\operatorname{E} e^{-\max\{1,r\}\xi_1} < \infty$$
 and  $\operatorname{E} |\eta_1|^{\max\{1,r\}} < \infty$  for some  $r > 0$ .

We can put further restrictions on  $\xi$  and  $\eta$ , such as in the example in Section 3 of Nyrhinen [26], which assumes that  $\xi$  is continuous and  $\eta$  is compound Poisson plus drift, which render the use of the Burkholder–Davis–Gundy inequalities unnecessary and further simplify the conditions. For general Lévy  $(\xi, \eta)$ , the above inequality is the sharpest we have found.

*Proof of Theorem 2.1.* We aim to use Proposition 4.1 for passage below rather than above. We can do this by replacing  $\eta$  by  $-\eta$ . Note that, for z > 0,

$$T_z = \inf\{t > 0 \colon Z_t < -z\} = \inf\{t > 0 \colon -Z_t > z\} = \inf\{t > 0 \colon \hat{Z}_t > z\},$$

where we denote  $Z_t$ , when  $\eta$  is replaced by  $-\eta$ , by  $\hat{Z}_t$  and similarly for the other quantities. Thus,  $\hat{Z}_t = -Z_t$ , and it is easily checked that, with  $(M_n, Q_n)$  as in (4.13),  $(\hat{M}_n, \hat{Q}_n) = (M_n, -Q_n)$ , and, with  $L_n$  as in (4.14),  $\hat{L}_n = -\bar{L}_n$ , where

$$\bar{L}_n := -e^{\xi_n} \left( \int_{(n-1)+}^n e^{-\xi_{s-}} d\eta_s - \inf_{n-1 < t \le n} \int_{(n-1)+}^t e^{-\xi_{s-}} d\eta_s \right).$$
 (5.2)

From (4.7) we obtain  $\hat{X}_n(\hat{L}_n) = -X_n(\bar{L}_n)$ . Then Proposition 4.1 ensures that (2.4) and (2.5) hold, if we can prove that the relevant conditions are satisfied for  $(\hat{M}, \hat{Q}, \hat{L})$ , i.e. we must show that hypothesis (H) holds for the hat variables.

The corresponding  $\hat{y}$  (see (4.9)) is

$$\sup \Big\{ y \in \mathbb{R} \colon \operatorname{P} \Big( \sup_{n \in \mathbb{N}} \hat{X}_n(\hat{L}_n) > y \Big) > 0 \Big\} = \inf \Big\{ z \in \mathbb{R} \colon \operatorname{P} \Big( \inf_{t > 0} Z_t < -z \Big) > 0 \Big\},$$

so  $\hat{y} = \infty$  if and only if  $\psi(z) > 0$  for all  $z \ge 0$ , which we have assumed.

We need a  $w^+ > 0$  such that  $E \hat{M}^{w^+} = 1$ , and this is the case with  $w^+ = w$  under (2.1) since  $\hat{M} = M = e^{-\xi_1}$ . Also,  $\hat{c}_M(\alpha) = \ln E \hat{M}^{\alpha} = c(\alpha)$ , so that  $\alpha_0^+$  in (4.8) here equals  $\alpha_0$  as defined

in (2.3). Note that the extra term  $E(\hat{M}\hat{L}^+)^{\alpha} = E(M\bar{L}^-)^{\alpha}$  required in (2.3) is superfluous here, since  $E(M\bar{L}^-)^{\alpha} = E\bar{Z}_1^{\alpha}$ , and this is finite for  $\alpha \ge 0$  if and only if  $E|Z_1|^{\alpha} < \infty$ .

Under the moment conditions of Theorem 2.1, the conditions of Lemma 5.1 hold with  $r=w^++\varepsilon$ , so  $\operatorname{E}|Z_1|^{\alpha}<\infty$  for  $\alpha=\max\{1,w+\varepsilon\}$ , and, hence,  $\alpha_0^+\geq w^++\varepsilon>w^+$ . Thus, indeed, hypothesis (H) is fulfilled in the present situation and Proposition 4.1 applies to give (2.4) and (2.5). Also,  $\alpha_0^+\geq w^++\varepsilon>w^+$  implies that  $c'(\alpha_0-)>c'(w)=\mu^*=-\operatorname{E}\xi_1\mathrm{e}^{-w\xi_1}$ , and this is finite since  $\operatorname{E}\mathrm{e}^{-(w+\varepsilon)\xi_1}$  is. So  $0\leq\alpha_0<1/\mu^*<\infty$ .

Suppose further that  $\xi_1$  is spread out. Then the dual version of (2.6) follows from Nyrhinen's comments in [26], which we expressed as Proposition 4.1.

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