

# On the path structure of a semimartingale arising from monotone probability theory

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Abstract. Let X be the unique normal martingale such that  $X_0 = 0$  and

 $d[X]_t = (1 - t - X_{t-}) dX_t + dt$ 

and let  $Y_t := X_t + t$  for all  $t \ge 0$ ; the semimartingale Y arises in quantum probability, where it is the monotone-independent analogue of the Poisson process. The trajectories of Y are examined and various probabilistic properties are derived; in particular, the level set  $\{t \ge 0: Y_t = 1\}$  is shown to be non-empty, compact, perfect and of zero Lebesgue measure. The local times of Y are found to be trivial except for that at level 1; consequently, the jumps of Y are not locally summable.

**Résumé.** Soit *X* l'unique martingale normale telle que  $X_0 = 0$  et

 $d[X]_t = (1 - t - X_{t-}) dX_t + dt$ 

et soit  $Y_t := X_t + t$  pour tout  $t \ge 0$ ; la semimartingale *Y* se manifeste dans la théorie des probabilités quantiques, où c'est analogue du processus de Poisson pour l'indépendance monotone. Les trajectoires de *Y* sont examinées et diverses propriétés probabilistes sont déduites; en particulier, l'ensemble de niveau  $\{t \ge 0: Y_t = 1\}$  est montré être non vide, compact, parfait et de mesure de Lebesgue nulle. Les temps locaux de *Y* sont trouvés être triviaux sauf celui au niveau 1; par conséquent les sauts de *Y* ne sont pas localements sommables.

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# **0. Introduction**

The first Azéma martingale, that is, the unique (in law) normal martingale M such that  $M_0 = 0$  and

 $d[M]_t = -M_{t-} dM_t + dt,$ 

has been the subject of much interest since its appearance in [3], Proposition 118 (see, for example, [4,13] and [17], Section IV.6); it was the first example to be found of a process without independent increments which possesses the chaotic-representation property. It shall henceforth be referred to as *Azéma's martingale*.

From a quantum-stochastic viewpoint, the process M may be obtained by applying Attal's D transform ([1], Section IV) to the Wiener process. Furthermore, thanks to the factorisation of D provided by vacuum-adapted calculus [5],

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M appears as a natural object in monotone-independent probability theory; the distribution of  $M_t$  (the arcsine law) is a central-limit law which plays a rôle analogous to that played by the Gaussian distribution in the classical framework ([16], Theorem 3.1).

The Poisson distribution also occurs as a limit (the *law of small numbers*): if, for all  $n \ge 1$ ,  $(x_{n,m})_{m=1}^{n}$  is a collection of independent, identically distributed random variables and there exists a constant  $\lambda > 0$  such that

$$\lim_{n \to \infty} n \mathbb{E} \big[ x_{n,1}^k \big] = \lambda \quad \forall k \ge 1,$$

then  $x_{n,1} + \cdots + x_{n,n}$  converges in distribution to the Poisson law with mean  $\lambda$ . (A simple proof of this result is provided in Appendix A.) In the case where  $x_{n,1}, \ldots, x_{n,n}$  are Bernoulli random variables taking the values 0 and 1 with mean  $\lambda/n$ , this is simply the Poisson approximation to the binomial distribution ([8], Example 25.2).

A corresponding theorem holds in the monotone set-up ([16], Theorem 4.1), but now the limit distribution is related to the D transform of the standard Poisson process (with intensity 1 and unit jumps) in the same way as the arcsine law and Azéma's martingale are related above [6]. (This result also holds for free probability: see [20], Theorem 4.) The classical process Y which results is such that  $Y_t = X_t + t$  for all  $t \ge 0$ , where X is the unique normal martingale such that  $X_0 = 0$  and

$$d[X]_t = (1 - t - X_{t-}) dX_t + dt$$

This article extends the sample-path analysis of Y (and so X) which was begun in [7]. Many similarities are found between Y and Azéma's martingale M; for example, they are both determined by a random perfect subset of  $\mathbb{R}_+$  and a collection of binary choices, one for each interval in that subset's complement. In Section 1 some results from the theory of martingales are recalled; Section 2 defines the processes X and Y and presents their Markov generators. A random time  $G_{\infty}$  after which Y is deterministic is discussed in Section 3: by Proposition 3.1 and Corollary 3.5,  $G_{\infty} < \infty$  almost surely and, in this case,

$$Y_{t+G_{\infty}} = -W_{-1}(-\exp(-1-t)) \quad \forall t \ge 0,$$

where  $W_{-1}$  is a certain branch of the inverse to the function  $z \mapsto ze^z$  (see Notation below). In Section 4 the process X is decomposed into an initial waiting time  $S_0$  which is exponentially distributed and an independent normal martingale Z which satisfies the same structure equation as X but has the initial condition  $Z_0 = 1$ ; Lemma 4.2 implies that, for all  $t \ge 0$ ,

$$X_t = \begin{cases} -t & \text{if } t \in [0, S_0[, \\ Z_{t-S_0} - S_0 & \text{if } t \in [S_0, \infty[. \end{cases}) \end{cases}$$

Explicit formulae are found for the distribution functions of  $G_{\infty}$  and J, a random variable analogous to  $G_{\infty}$  but for Z rather than X. In Section 5 it is shown that  $(H_t := 1 - (Z_t + t)^{-1})_{t \ge 0}$  is a martingale which is related to Azéma's martingale M by a time change; this gives a simple way to find various properties of the level set  $\mathcal{U} := \{t \ge 0: Y_t = 1\}$  in Section 6. Finally, Section 7 presents some results on the local times of Y. The appendices contain various supplementary results which are not appropriate for the main text.

#### 0.1. Conventions

The underlying probability space is denoted  $(\Omega, \mathcal{F}, \mathbb{P})$  and is assumed to contain a filtration  $(\mathcal{F}_t)_{t\geq 0}$  which generates the  $\sigma$ -algebra  $\mathcal{F}$ . This filtration is supposed to *satisfy the usual conditions*: it is right continuous and the initial  $\sigma$ algebra  $\mathcal{F}_0$  contains all the  $\mathbb{P}$ -null sets. Each semimartingale which is considered below has *càdlàg* paths (that is, they are right-continuous with left limits) and two processes  $(X_t)_{t\geq 0}$  and  $(Y_t)_{t\geq 0}$  are taken equal if they are *indistinguishable*:  $\mathbb{P}(X_t = Y_t \text{ for all } t \geq 0) = 1$ . Any quadratic variation or stochastic integral has value 0 at time 0.

#### 0.2. Notation

The expression  $\mathbb{1}_P$  is equal to 1 if the proposition *P* is true and equal to 0 otherwise; the indicator function of a set *A* is denoted by  $\mathbb{1}_A$ . The set of natural numbers is denoted by  $\mathbb{N} := \{1, 2, 3, \ldots\}$ , the set of non-negative rational

numbers is denoted by  $\mathbb{Q}_+$  and the set of non-negative real numbers is denoted by  $\mathbb{R}_+$ . The branches of the Lambert W function (that is, the multi-valued inverse to the map  $z \mapsto ze^z$ ) which take (some) real values are denoted by  $W_0$  and  $W_{-1}$ , following the conventions of Corless et al. [10]:

$$W_0(0) = 0,$$
  $W_0(x) \in [-1, 0[$  and  $W_{-1}(x) \in ]-\infty, -1]$   $\forall x \in [-e^{-1}, 0[.$ 

If  $\Xi$  is a topological space then  $\mathcal{B}(\Xi)$  denotes the Borel  $\sigma$ -algebra on  $\Xi$ . The integral of the process X by the semimartingale R will be denoted by  $\int X_t dR_t$  or  $X \cdot R$ , as convenient; the differential notation  $X_t dR_t$  will also be employed. The process X stopped at T is denoted by  $X^T$ , that is,  $X_t^T := X_{t \wedge T}$  for all  $t \ge 0$ , where  $x \wedge y$  denotes the minimum of x and y. For all x, the positive part  $x^+ := \max\{x, 0\}$ , the maximum of x and 0.

# 1. Normal sigma-martingales and time changes

**Remark 1.1.** Let  $A \in \mathcal{F}$  be such that  $\mathbb{P}(A) > 0$ . If  $\mathcal{G}$  is a sub- $\sigma$ -algebra of  $\mathcal{F}$  such that  $A \in \mathcal{G}$  then

$$\widetilde{\mathcal{G}} := \{ B \subseteq \Omega \colon B \cap A \in \mathcal{G} \}$$

is a  $\sigma$ -algebra containing  $\mathcal{G}$ ; the map  $\mathcal{G} \mapsto \widetilde{\mathcal{G}}$  preserves inclusions and arbitrary intersections. If

$$\widetilde{\mathbb{P}} := \mathbb{P}(\cdot|A) : \widetilde{\mathcal{F}} \to [0,1]; \qquad B \mapsto \frac{\mathbb{P}(B \cap A)}{\mathbb{P}(A)},$$

then  $(\Omega, \widetilde{\mathcal{F}}, \widetilde{\mathbb{P}})$  is a complete probability space; if  $(\mathcal{G})_{t\geq 0}$  is a filtration in  $(\Omega, \mathcal{F}, \mathbb{P})$  satisfying the usual conditions then  $(\widetilde{\mathcal{G}}_t)_{t\geq 0}$  is a filtration in  $(\Omega, \widetilde{\mathcal{G}}, \widetilde{\mathbb{P}})$  which satisfies them as well.

If T is a stopping time for the filtration  $(\mathcal{G}_t)_{t\geq 0}$  then it is also one for  $(\widetilde{\mathcal{G}}_t)_{t\geq 0}$  and, if  $B \subseteq \Omega$ ,

$$\begin{split} B \in \widetilde{\mathcal{G}_T} & \iff \quad B \cap A \in \mathcal{G}_T & \iff \quad B \cap A \cap \{T \leq t\} \in \mathcal{G}_t \quad \forall t \geq 0, \\ & \iff \quad B \cap \{T \leq t\} \in \widetilde{\mathcal{G}}_t \; \forall t \geq 0 \quad \iff \quad B \in (\widetilde{\mathcal{G}})_T, \end{split}$$

so the notation  $\widetilde{\mathcal{G}}_T$  is unambiguous.

**Lemma 1.2.** If T is a stopping time such that  $\mathbb{P}(T < \infty) > 0$  and M is a local martingale then  $N: t \mapsto \mathbb{1}_{T < \infty}(M_{t+T} - M_T)$  is a local martingale for the conditional probability measure  $\widetilde{\mathbb{P}} := \mathbb{P}(\cdot|T < \infty)$  and the filtration  $(\widetilde{\mathcal{F}}_{t+T})_{t \geq 0}$ , such that

$$[N]_t = \mathbb{1}_{T < \infty} ([M]_{t+T} - [M]_T) \quad \forall t \ge 0.$$

**Proof.** If  $T < \infty$  almost surely and M is uniformly integrable then the first part is immediate, by optional sampling ([18], Theorem II.77.5), and holds in general by localisation and conditioning. The second claim may be verified by realising [N] as a limit of sums in the usual manner (see [17], Theorem II.22, for example).

**Definition 1.3.** A martingale M is normal if  $t \mapsto (M_t - M_0)^2 - t$  is also a martingale. (If  $M_0$  is square integrable then this is equivalent to  $t \mapsto M_t^2 - t$  being a martingale, but in general it is a weaker condition.)

**Definition 1.4.** A semimartingale M is a sigma-martingale if it can be written as  $K \cdot N$ , where N is a local martingale and K is a predictable, N-integrable process. Equivalently, there exists an increasing sequence  $(A_n)_{n\geq 1}$  of predictable sets such that  $\bigcup_{n\geq 1} A_n = \mathbb{R}_+ \times \Omega$  and  $1_{A_n} \cdot M \in H^1$  for all  $n \geq 1$ , where  $H^1$  denotes the Banach space of martingales M with  $\|M\|_{H^1} := \mathbb{E}[[M]_{\infty}^{1/2}] < \infty$ . Every local martingale is a sigma-martingale and if M is a sigma-martingale then so is  $H \cdot M$  for any predictable, M-integrable process H. (The class of sigma-martingales, so named by Delbaen and Schachermayer in [11], was introduced by Chou in [9], where it is denoted  $(\Sigma_m)$ ; the equivalence mentioned above is due to Émery ([12], Proposition 2).)

**Theorem 1.5** ([14]). If M is a semimartingale with  $M_0 = 0$  then the following are equivalent:

(i) M and  $t \mapsto M_t^2 - t$  are sigma-martingales;

- (ii) *M* and  $t \mapsto [M]_t t$  are sigma-martingales;
- (iii) M and  $t \mapsto M_t^2 t$  are martingales;
- (iv) *M* and  $t \mapsto [M]_t t$  are martingales.

**Proof.** Since  $M^2 - [M] = 2M_- \cdot M$ , the equivalence of (i) and (ii) is immediate; it also follows from this that (iv) implies (iii) ([17], Corollary 3 to Theorem II.27). To complete the proof it suffices to show that (ii) implies (iv).

Suppose (ii) holds and let  $(A_n)_{n\geq 1}$  be an increasing sequence of predictable sets such that  $\bigcup_{n\geq 1} A_n = \mathbb{R}_+ \times \Omega$ and both  $1_{A_n} \cdot M \in H^1$  and  $1_{A_n} \cdot N \in H^1$  for all  $n \geq 1$ , where  $N: t \mapsto [M]_t - t$ . (Note that if  $X \in H^1$  and B is a predictable set then  $1_B \cdot X \in H^1$ .) Let T be a bounded stopping time; since  $1_{A_n} \cdot N$  is a martingale,

$$\mathbb{E}\left[\left(1_{A_{n}}\cdot[M]\right)_{T}\right] = \mathbb{E}\left[\left(1_{A_{n}}\cdot N\right)_{T}\right] + \mathbb{E}\left[\int_{0}^{T}1_{A_{n}}\,\mathrm{d}s\right] = \mathbb{E}\left[\int_{0}^{T}1_{A_{n}}\,\mathrm{d}s\right]$$
(1)

and therefore  $\mathbb{E}[[M]_T] = \mathbb{E}[T] < \infty$ , by monotone convergence. It follows that  $\mathbb{E}[[N]_T] \le \mathbb{E}[[M]_T] + \mathbb{E}[T] < \infty$ and  $\mathbb{E}[N_T] = \mathbb{E}[[M]_T - T] = 0$ , so *N* is a martingale. (Apply [17], Theorem I.21 to *N* stopped at *t* for any  $t \ge 0$ .) Furthermore, since  $(1_{A_n \setminus A_m} \cdot [M])_t \le (1_{A_m^c} \cdot [M])_t$  for all  $m \le n$  and  $t \ge 0$ , where  $A_m^c := (\mathbb{R}_+ \times \Omega) \setminus A_m$ , the sequence  $(1_{A_n \cap ([0,t] \times \Omega)} \cdot M)_{n \ge 1}$  is Cauchy in  $H^2$ , so convergent there; it follows (by [17], Theorem IV.32, say) that *M* stopped at *t* is an  $H^2$ -martingale.

**Theorem 1.6.** If M is a normal martingale and T is a stopping time such that  $\mathbb{P}(T < \infty) > 0$  then  $N: t \mapsto \mathbb{1}_{T < \infty}(M_{t+T} - M_T)$  is a normal martingale (for the measure  $\widetilde{\mathbb{P}} := \mathbb{P}(\cdot | T < \infty)$  and the filtration  $(\widetilde{\mathcal{F}}_{t+T})_{t \ge 0}$ ).

**Proof.** As M and  $t \mapsto (M_t - M_0)^2 - t$  are local martingales, so are N and

$$Q:t \mapsto \mathbb{1}_{T < \infty} \left( (M_{t+T} - M_0)^2 - (t+T) - (M_T - M_0)^2 + T \right)$$
  
=  $\mathbb{1}_{T < \infty} \left( (M_{t+T} - M_T)^2 - t + 2(M_T - M_0)(M_{t+T} - M_T) \right),$ 

by Lemma 1.2. Hence  $t \mapsto (N_t - N_0)^2 - t = Q_t - 2\mathbb{1}_{T < \infty}(M_T - M_0)N_t$  is also a local martingale (as local martingales form a module over the algebra of random variables which are measurable with respect to the initial  $\sigma$ -algebra) and the conclusion follows from Theorem 1.5.

**Lemma 1.7.** If A is a right-continuous, increasing process such that  $A_0 \ge 0$  and each  $A_t$  is a stopping time then  $(\mathcal{F}_{A_t})_{t\ge 0}$  is a filtration which satisfies the usual conditions.

**Proof.** This is a straightforward exercise.

**Lemma 1.8.** Let K and L be independent martingales and let A be a continuous, increasing,  $(\mathcal{F}_t^K)_{t\geq 0}$ -adapted process with  $A_0 = 0$  and  $A_{\infty} = \infty$ , where  $(\mathcal{F}_t^K)_{t\geq 0}$  denotes the smallest filtration satisfying the usual hypotheses to which K is adapted.

If  $\mathcal{G}_t := \mathcal{F}_{\infty}^{K} \lor \mathcal{F}_t^L$  for all  $t \ge 0$  then each  $A_t$  is a  $(\mathcal{G}_t)_{t\ge 0}$ -stopping time,  $(\mathcal{G}_{A_t})_{t\ge 0}$  is a filtration satisfying the usual conditions,  $L_A$  is a  $(\mathcal{G}_{A_t})_{t\ge 0}$ -local martingale and  $[L_A] = [L]_A$ . If H is an  $(\mathcal{F}_t^L)_{t\ge 0}$ -predictable process which is L integrable then  $H_A$  is  $(\mathcal{G}_A)_{t\ge 0}$  predictable and  $L_A$  integrable, with  $(H \cdot L)_A = H_A \cdot L_A$ .

If  $\mathcal{H}_t := \mathcal{F}_t^K \vee \mathcal{F}_t^{L_A}$  for all  $t \ge 0$  then  $\mathcal{H}_t \subseteq \mathcal{G}_{A_t}$  for all  $t \ge 0$ . If there exist disjoint,  $(\mathcal{H}_t)_{t\ge 0}$ -predictable sets B and C such that  $1_B \cdot [K] = [K]$  and  $1_C \cdot [L]_A = [L]_A$  and if  $([K] + [L]_A)^{1/2}$  is  $(\mathcal{H}_t)_{t\ge 0}$ -locally integrable then  $K + L_A$  is a  $(\mathcal{H}_t)_{t\ge 0}$ -local martingale and  $[K + L_A] = [K] + [L]_A$ .

**Proof.** This is immediate from Lemmes 1–3 and Théorème 1 of [21].

# 2. The processes *X* and *Y*

**Definition 2.1.** Let X be the normal martingale which satisfies the (time-inhomogeneous) structure equation

 $d[X]_t = (1 - t - X_{t-}) dX_t + dt$ 

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with initial condition  $X_0 = 0$  and let  $Y_t := X_t + t$  for all  $t \ge 0$ . (The process X was introduced in [7], where it was constructed from the quantum stochastic analogue of the Poisson process for monotone independence. Existence also follows directly from [23], Théorème 4.0.2; uniqueness (in law) and the chaotic-representation property hold by [2], Corollary 26.) Then  $Y_0 = 0$  and

$$d[Y]_t = (1 - Y_{t-}) dY_t + Y_{t-} dt,$$
(2)

which implies that  $\Delta Y_t \in \{0, 1 - Y_{t-}\}$  for all t > 0. If

$$G_t := \sup \{ s \in [0, t] \colon Y_s = 1 \} \in \{-\infty\} \cup ]0, t]$$
(3)

then (by [7], Theorem 24)

$$Y_t = -W_{\bullet} \left( -\exp(-1 - t + G_t) \right) \tag{4}$$

for all  $t \ge 0$ , where  $W_{\bullet} = W_{-1}$  if  $Y_t \ge 1$  and  $W_{\bullet} = W_0$  if  $Y_t \le 1$ ; a little more will be said in Proposition 6.3. (It follows from this description of the trajectories that X and Y are uniformly bounded on [0, t] for all  $t \ge 0$ .)

#### Definition 2.2. Let

$$a: \mathbb{R}_+ \to ]0, 1]; \qquad t \mapsto -W_0(-e^{-1-t}),$$
  
$$b: \mathbb{R}_+ \to [1, \infty[; \qquad t \mapsto -W_{-1}(-e^{-1-t})]$$

and

$$c: ]0, \infty[ \to \mathbb{R}_+; \qquad t \mapsto b'(t) - a'(t) = \frac{b(t)}{b(t) - 1} + \frac{a(t)}{1 - a(t)}.$$

Note that a(0) = b(0) = 1, both a and b are homeomorphisms (which may be verified by inspecting their derivatives on  $]0, \infty[$ ) and  $c(t) \searrow 1$  as  $t \to \infty$ .

**Lemma 2.3.** For all  $t \ge 0$  the random variable  $Y_t$  is distributed with an atom at 0 (of mass  $e^{-t}$ ) and a continuous part with support [a(t), b(t)]:

$$\mathbb{P}(Y_t \in A) = \mathbb{1}_{0 \in A} \mathrm{e}^{-t} + \frac{1}{\pi} \int_{A \cap [a(t), b(t)]} \mathrm{Im} \frac{1}{W_{-1}(-y \mathrm{e}^{t-y})} \,\mathrm{d}y \quad \forall A \in \mathcal{B}(\mathbb{R}).$$

Proof. See [7], Corollary 17.

*Remark 2.4.* The (classical) Poisson process is simpler when uncompensated; similarly, it is easier to work with Y than with X. These processes are strongly Markov (by [2], Theorem 37, for example) and Émery's Itô formula ([13], Proposition 2) implies that, if  $f : \mathbb{R} \to \mathbb{R}$  is twice continuously differentiable,

$$f(X_t) = f(0) + \int_0^t g(X_{s-}, s) \, \mathrm{d}X_s + \int_0^t h(X_{s-}, s) \, \mathrm{d}s \tag{5}$$

and

$$f(Y_t) = f(0) + \int_0^t g(Y_{s-}, 0) \, \mathrm{d}X_s + \int_0^t \left( h(Y_{s-}, 0) + f'(Y_{s-}) \right) \, \mathrm{d}s \tag{6}$$

for all  $t \ge 0$ , where  $g, h : \mathbb{R}^2 \to \mathbb{R}$  are such that

$$g(x,t) = \mathbb{1}_{x \neq 1-t} \frac{f(1-t) - f(x)}{1 - x - t} + \mathbb{1}_{x=1-t} f'(1-t)$$

and

$$h(x,t) = \mathbb{1}_{x \neq 1-t} \frac{f(1-t) - f(x) - (1-x-t)f'(x)}{(1-x-t)^2} + \mathbb{1}_{x=1-t} \frac{1}{2}f''(1-t)$$

for all  $x, t \in \mathbb{R}$ . It follows that

$$\lim_{\varepsilon \to 0+} \frac{1}{\varepsilon} \mathbb{E} \Big[ f(X_{t+\varepsilon}) - f(X_t) | \mathcal{F}_t \Big] = \Big( \Gamma_t^X f \Big) (X_t)$$

and

$$\lim_{\varepsilon \to 0+} \frac{1}{\varepsilon} \mathbb{E} \Big[ f(Y_{t+\varepsilon}) - f(Y_t) | \mathcal{F}_t \Big] = \big( \Gamma^Y f \big) (Y_t),$$

*for almost all*  $t \ge 0$ *, where* 

$$(\Gamma_t^X f)(x) := \begin{cases} \frac{f(1-t)-f(x)-(1-x-t)f'(x)}{(1-x-t)^2} & \text{if } x \neq 1-t, \\ \frac{1}{2}f''(1-t) & \text{if } x = 1-t, \end{cases}$$

$$= \mathbb{1}_{x=1-t} \frac{1}{2}f''(x) + \int_{\mathbb{R}\setminus\{x\}} (f(y) - f(x) - (y-x)f'(x)) \frac{\delta_{1-t}(\mathrm{d}y)}{(y-x)^2}, \tag{7}$$

$$(\Gamma^Y f)(x) := \begin{cases} \frac{f(1)-f(x)-x(1-x)f'(x)}{(1-x)^2} & \text{if } x \neq 1, \\ \frac{1}{2}f''(1) + f'(1) & \text{if } x = 1, \end{cases}$$

$$= \mathbb{1}_{x=1} \frac{1}{2}f''(x) + f'(x) + \int_{\mathbb{R}\setminus\{x\}} (f(y) - f(x) - (y-x)f'(x)) \frac{\delta_1(\mathrm{d}y)}{(y-x)^2}, \tag{8}$$

and  $\delta_z$  denotes the Dirac measure on  $\mathbb{R}$  with support  $\{z\}$ .

# 3. The final jump time

**Proposition 3.1.** If  $G_{\infty} := \sup\{G_t: t \ge 0\}$ , where  $G_t$  is defined in (3), then the random variable  $G_{\infty}$  (the final jump time of Y) is almost surely finite and has density

$$g_{\infty} : \mathbb{R} \to \mathbb{R}_{+}; \qquad x \mapsto \mathbb{1}_{x \ge 0} \frac{1}{\pi} \operatorname{Im} \frac{1}{W_{-1}(-e^{-1+x})}.$$
(9)

**Proof.** Note first that  $G_t = 1 + t - Y_t + \log Y_t$  for all  $t \ge 0$ , by (4), so  $G_t$  is  $\mathcal{F}_t$  measurable. As  $t \mapsto G_t$  is increasing, it is elementary to verify that

$$G_{\infty} = \sup\{G_t: t \ge 0\} = \sup\{G_n: n \ge 1\} = \lim_{n \to \infty} G_n;$$

in particular,  $G_{\infty}$  is  $\mathcal{F}$  measurable. If t > 0 then  $\mathbb{1}_{G_n \in [0,t]} \to \mathbb{1}_{G_{\infty} \in [0,t]}$ , because  $G_n \nearrow G_{\infty}$ , and the dominatedconvergence theorem implies that

$$\mathbb{P}(G_{\infty} \in ]0, t]) = \mathbb{E}[\mathbb{1}_{G_{\infty} \in ]0, t]}] = \lim_{n \to \infty} \mathbb{E}[\mathbb{1}_{G_{n} \in ]0, t]}] = \lim_{n \to \infty} \mathbb{P}(G_{n} \in ]0, t]).$$

Since  $\mathbb{P}(G_{\infty} = -\infty) = \mathbb{P}(Y \equiv 0) \le \mathbb{P}(Y_t = 0) = e^{-t} \to 0$  as  $t \to \infty$ , it follows that  $\mathbb{P}(G_{\infty} = -\infty) = 0$  and

$$\mathbb{P}(G_{\infty} \le t) = \lim_{n \to \infty} \mathbb{P}(G_n \in ]0, t])$$

for all  $t \ge 0$ . If  $n \ge 1$  and  $t \in [0, n]$  then

$$0 < 1 + n - Y_n + \log Y_n \le t \quad \Longleftrightarrow \quad -e^{-1-n} > -Y_n \exp(-Y_n) \ge -e^{-1-n+t}$$
$$\longleftrightarrow \quad Y_n \in \left] a(n), a(n-t) \right] \cup \left[ b(n-t), b(n) \right[$$

and, by Lemma 2.3,

$$\gamma_n(t) := \mathbb{P}(G_n \in ]0, t]) = \frac{1}{\pi} \int_{]a(n), a(n-t)] \cup [b(n-t), b(n)[} \operatorname{Im} \frac{1}{W_{-1}(-y e^{n-y})} \, \mathrm{d}y.$$

Note that  $\gamma_n$  is continuously differentiable on [0, n[, with

$$\gamma'_n(s) = \frac{1}{\pi} \operatorname{Im} \frac{1}{W_{-1}(-e^{-1+s})} (b'(n-s) - a'(n-s)) = c(n-s)g_{\infty}(s)$$

for all  $s \in [0, n[$ . If n > t and  $s \in [0, t]$  then, by the remarks in Definition 2.2,  $\gamma'_n(s) \searrow g_{\infty}(s)$  as  $n \to \infty$  and the monotone-convergence theorem implies that

$$\lim_{n \to \infty} \int_0^t \gamma'_n(s) \, \mathrm{d}s = \int_0^t g_\infty(s) \, \mathrm{d}s \quad \forall t \ge 0$$

This gives the result, because  $\int_0^\infty g_\infty(s) \, ds = 1$  (by Proposition B.1).

**Remark 3.2.** It follows from Proposition 3.1 that  $\mathbb{E}[G_{\infty}] = \infty$ ; a proof is given in Proposition B.1.

**Remark 3.3.** Calling  $G_{\infty}$  the final jump time is perhaps a little misleading, since it is not a stopping time; it is, however, almost surely the limit of a sequence of jump times. (See Corollary 6.2 and Corollary 6.4.)

**Proposition 3.4.**  $\lim_{t\to\infty} \mathbb{P}(Y_t \le 1) = 0.$ 

Proof. By Lemma 2.3,

$$\mathbb{P}(Y_t \le 1) = e^{-t} + \frac{1}{\pi} \int_{a(t)}^{1} \operatorname{Im} \frac{1}{W_{-1}(-ye^{t-y})} \, \mathrm{d}y \quad \forall t \ge 0.$$
(10)

If  $y \in [0, 1]$  then there exists  $x \in [0, \infty)$  such that y = a(x), and if  $t \ge x$  then

$$\operatorname{Im} \frac{1}{W_{-1}(-ye^{t-y})} = \operatorname{Im} \frac{1}{W_{-1}(-e^{-1+t-x})} = \pi g_{\infty}(t-x) \to 0$$

as  $t \to \infty$ . (This last claim follows from Proposition B.1.) Furthermore, as  $g_{\infty}$  is bounded, the integrand in (10) is bounded uniformly in y and t, so the result follows from the dominated-convergence theorem.

**Corollary 3.5.** As  $t \to \infty$ , the process  $Y_t \to \infty$  almost surely.

**Proof.** If  $G_{\infty} < \infty$  then, as  $t \to \infty$ , either  $Y_t \to 0$  or  $Y_t \to \infty$ ; furthermore,

$$\{G_{\infty} < \infty\} \cap \left\{\lim_{t \to \infty} Y_t = \infty\right\} = \{G_{\infty} < \infty\} \cap \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} \{Y_m > 1\}.$$

Since  $\mathbb{P}(G_{\infty} < \infty) = 1$  and  $\mathbb{P}(Y_n \le 1) \to 0$  as  $n \to \infty$ , it follows that

$$\mathbb{P}\left(\lim_{t \to \infty} Y_t = \infty\right) \ge \limsup_{n \to \infty} \mathbb{P}(Y_n > 1) = 1 - \lim_{n \to \infty} \mathbb{P}(Y_n \le 1) = 1.$$

(The inequality in the previous line holds by [8], Theorem 4.1(i).)

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# 4. The active period

**Proposition 4.1.** The stopping time  $S_0 := \inf\{t > 0: Y_t = 1\}$  is exponentially distributed and has mean 1.

**Proof.** Note that  $Y_t = 0$  only if  $Y_s = 0$  for all  $s \in [0, t[$ , by (4); the claim now follows from Lemma 2.3.

**Lemma 4.2.** If  $Z_t := X_{t+S_0} + S_0$  for all  $t \ge 0$  then Z is a normal martingale for the filtration  $(\mathcal{F}_{t+S_0})_{t\ge 0}$  such that  $Z_0 = 1$ , which satisfies the structure equation

$$d[Z]_t = dt + (1 - t - Z_{t-}) dZ_t$$
(11)

and which is independent of  $\mathcal{F}_{S_0}$ .

**Proof.** As  $Z_t = X_{t+S_0} - X_{S_0} + 1$  for all  $t \ge 0$ , Theorem 1.6 implies that Z is a normal martingale. Furthermore,

$$[Z]_t = [X]_{t+S_0} - [X]_{S_0} = \int_{S_0}^{t+S_0} (1 - r - X_{r-}) \, \mathrm{d}X_r = \int_0^t (1 - s - Z_{s-}) \, \mathrm{d}Z_s$$

for all  $t \ge 0$ . (The first equality is a consequence of Lemma 1.2; the last may be shown by expressing the integrals as the limit of Riemann sums, as in [17], Theorem II.21, for example.) It now follows from [2], Theorem 25, that, for all  $t \ge 0$ , the law of  $Z_t$  conditional on  $\mathcal{F}_{S_0}$  depends only on the initial value  $Z_0 = 1$  and the coefficient functions  $\alpha : s \mapsto 1 - s$  and  $\beta \equiv -1$  restricted to [0, t], so  $Z_t$  is independent of  $\mathcal{F}_{S_0}$ .

*Remark 4.3.* If  $t \ge 0$  then

$$Z_t + t = Y_{t+S_0} = -W_{\bullet} \left( -\exp(-1 - (t+S_0) + G_{t+S_0}) \right) \in [a(t), b(t)],$$

since  $G_{t+S_0} \ge S_0$ . Consequently, Z is uniformly bounded on [0, t] for all  $t \ge 0$ .

**Remark 4.4.** Let  $m_n(t) := \mathbb{E}[(Z_t + t)^n]$  for all  $n \ge 1$  and  $t \ge 0$ , where Z is as in Lemma 4.2. It may be shown using *Émery's Itô formula* ([13], Proposition 2 and the subsequent remark) that

$$m_n(t) - m_{n-1}(t) = n \int_0^t m_{n-1}(s) \,\mathrm{d}s \tag{12}$$

for all  $n \ge 1$  and  $t \ge 0$  (where  $m_0 \equiv 1$ ). Hence (compare [6], Section 4)

$$\widehat{m}_n(p) = p^{-1} \prod_{j=1}^n (1+jp^{-1})$$

if  $n \ge 1$ , where  $\hat{f}$  denotes the Laplace transform of f, and so

$$m_n(t) = 1 + \sum_{k=1}^n \left( \sum_{1 \le j_1 < \dots < j_k \le n} j_1 \cdots j_k \right) \frac{t^k}{k!} = \sum_{k=0}^n \left[ \frac{n+1}{n+1-k} \right] \frac{t^k}{k!}$$
(13)

for all  $t \ge 0$ , where [:] denotes the unsigned Stirling numbers of the first kind [15]. (The final identity holds by [7], Proposition 3 and Remark 6, for example.)

**Theorem 4.5.** If t > 0 then  $Z_t + t = Y_{t+S_0}$  is continuously distributed, with density

$$f_{Z_t+t}: \mathbb{R} \to \mathbb{R}_+; \qquad z \mapsto \mathbb{1}_{z \in [a(t), b(t)]} \frac{1}{\pi} \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{t-z})}.$$
(14)

**Proof.** Let  $x \ge 0$ . Since  $Y_t = \mathbb{1}_{t \ge S_0}(Z_{(t-S_0)^+} + t - S_0)$  for all  $t \ge 0$ , it follows that

$$\mathbb{P}(0 < Y_t \le x) = \mathbb{P}(S_0 \le t \text{ and } Z_{(t-S_0)^+} + t - S_0 \le x)$$
  
=  $\int_0^t \int_{-\infty}^{x-t+s} dF_{Z_{t-s}}(z) e^{-s} ds$   
=  $e^{-t} \int_0^t \int_{-\infty}^{x-u} dF_{Z_u}(z) e^u du$ ,

where  $F_V$  denotes the distribution function of the random variable V. (For the second equality, note that

$$\mathbb{E}[\mathbb{1}_{S_0 \le t} \mathbb{1}_{Z_{(t-S_0)^+} + t - S_0 \le x}] = \mathbb{E}\big[\mathbb{1}_{S_0 \le t} \mathbb{E}[\mathbb{1}_{Z_{(t-S_0)^+} + t - S_0 \le x} | \mathcal{F}_{S_0}]\big]$$
$$= \mathbb{E}\big[\mathbb{1}_{S_0 \le t} \mathbb{E}[\mathbb{1}_{Z_{(t-s)^+} + t - s \le x}]|_{s=S_0}\big],$$

since Z is independent of  $\mathcal{F}_{S_0}$ .) Hence

$$\mathbb{P}(Z_t + t \le x) = e^{-t} \frac{\mathrm{d}}{\mathrm{d}t} \left( e^t \mathbb{P}(0 < Y_t \le x) \right) = \mathbb{P}(0 < Y_t \le x) + \frac{\mathrm{d}}{\mathrm{d}t} \mathbb{P}(0 < Y_t \le x)$$

Thus if t > 0 then either  $x \le a(t)$ , so that  $\mathbb{P}(Z_t + t \le x) = 0$ , or  $x \ge b(t)$ , whence  $\mathbb{P}(Z_t + t \le x) = 1 - e^{-t} + e^{-t} = 1$ , or  $x \in ]a(t), b(t)[$ , in which case

$$\pi \mathbb{P}(Z_t + t \le x) = \int_{a(t)}^x \operatorname{Im} \frac{1}{W_{-1}(-ze^{t-z})} dz - a'(t) \operatorname{Im} \frac{1}{W_{-1}(-a(t)e^{t-a(t)})} + \int_{a(t)}^x \frac{\partial}{\partial t} \operatorname{Im} \frac{1}{W_{-1}(-ze^{t-z})} dz$$
$$= \int_{a(t)}^x \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{t-z})} dz,$$

as claimed. (This formal working is a little awkward to justify: a rigorous proof is provided by Proposition C.2.)  $\Box$ 

**Proposition 4.6.** The random variables  $S_0$  and  $J := G_{\infty} - S_0$  are independent and J is continuous, with density

$$f_J : \mathbb{R} \to \mathbb{R}_+; \qquad x \mapsto \mathbb{1}_{x>0} \frac{1}{\pi} \operatorname{Im} \frac{1}{1 + W_{-1}(-e^{-1+x})}.$$
 (15)

**Proof.** To see that  $S_0$  and J are independent, note first that

$$J = \lim_{n \to \infty} G_{n+S_0} - S_0 = \lim_{n \to \infty} \left( 1 - Z_n + \log(Z_n + n) \right)$$

almost surely, where  $(Z_t)_{t\geq 0}$  is as defined in Lemma 4.2, which implies that  $G_{n+S_0} - S_0$  is independent of  $S_0$  for all  $n \geq 1$  and, therefore, so is J.

If  $F_J(z) := \mathbb{P}(J \le z)$  for all  $z \in \mathbb{R}$  then, by independence and Proposition 4.1,

$$\int_{-\infty}^{z} g_{\infty}(w) \, \mathrm{d}w = \mathbb{P}(J + S_0 \le z) = \iint_{\{(x, y) \in \mathbb{R}^2: x + y \le z\}} \mathrm{d}F_J(x) \, \mathbb{1}_{y \ge 0} \mathrm{e}^{-y} \, \mathrm{d}y$$
$$= \int_{-\infty}^{z} \mathrm{e}^{-v} \int_{-\infty}^{v} \mathrm{e}^u \, \mathrm{d}F_J(u) \, \mathrm{d}v$$

for all  $z \in \mathbb{R}$ , using the substitution (u, v) = (x, x + y). Thus, for almost all  $v \in \mathbb{R}$ ,

$$g_{\infty}(v) = \mathrm{e}^{-v} \int_{-\infty}^{v} \mathrm{e}^{u} \,\mathrm{d}F_{J}(v);$$

in fact, this holds for all  $v \in \mathbb{R}$ , as both functions are continuous, and, since  $g_{\infty}(0) = 0$ ,

$$g_{\infty}(t) = \mathrm{e}^{-t} \int_0^t \mathrm{e}^s \,\mathrm{d}F_J(s) \quad \forall t \ge 0.$$

Now  $g_{\infty}$  is continuously differentiable on  $\mathbb{R} \setminus \{0\}$  and  $f_J(x) = g_{\infty}(x) + \mathbb{1}_{x \neq 0} g'_{\infty}(x)$ , so if  $0 < \varepsilon < t$  then integration by parts yields the equality

$$\int_{\varepsilon}^{t} e^{s} f_{J}(s) ds = e^{t} g_{\infty}(t) - e^{\varepsilon} g_{\infty}(\varepsilon) \to \int_{0}^{t} e^{s} dF_{J}(s) \quad \text{as } \varepsilon \to 0 + .$$

Hence  $\int_0^t e^s f_J(s) ds$  exists for all  $t \ge 0$  (as does  $\int_0^t f_J(s) ds$ , by comparison) and

$$\mu: \mathcal{B}(\mathbb{R}_+) \to \mathbb{R}_+; \qquad A \mapsto \int_A e^s \, \mathrm{d}F_J(s) = \int_A e^s f_J(s) \, \mathrm{d}s$$

is a positive Borel measure on  $\mathbb{R}_+$ ; by [19], Theorem 1.29,

$$\int_0^t f_J(s) \, \mathrm{d}s = \int_0^t \mathrm{e}^{-s} \, \mathrm{d}\mu(s) = \int_0^t \, \mathrm{d}F_J(s) = F_J(t) - F_J(0)$$

for all  $t \ge 0$  and

$$1 = \lim_{t \to \infty} F_J(t) = F_J(0) + \int_0^\infty g_\infty(s) \, \mathrm{d}s + \lim_{t \to \infty} g_\infty(t) = F_J(0) + 1,$$

by Proposition B.1, so  $F_J(0) = 0$ . The result follows.

**Remark 4.7.** The distribution of J may also be found by imitating the proof of Proposition 3.1, with  $Z_t + t$  replacing  $Y_t$ , since J has the same relationship to Z as  $G_{\infty}$  does to X.

**Proposition 4.8.** *If*  $t \ge 0$  *then* 

$$\mathbb{P}(G_{\infty} \le t) = -\frac{1}{\pi} \operatorname{Im}\left(W_{-1}\left(-e^{-1+t}\right) + \frac{1}{W_{-1}\left(-e^{-1+t}\right)}\right)$$
(16)

and

$$\mathbb{P}(J \le t) = -\frac{1}{\pi} \operatorname{Im} W_{-1}(-e^{-1+t}) = \mathbb{P}(G_{\infty} \le t) + g_{\infty}(t).$$
(17)

Proof. These follow immediately from the identities

$$\int_0^t \frac{1}{W_{-1}(-e^{-1+x})} \, \mathrm{d}x = t - \frac{(1+W_{-1}(-e^{-1+t}))^2}{W_{-1}(-e^{-1+t})}$$

and

$$\int_0^t \frac{1}{1 + W_{-1}(-e^{-1+x})} \, \mathrm{d}x = t - W_{-1}(-e^{-1+t}) - 1,$$

which are valid for all  $t \ge 0$  and may be verified by differentiation. For brevity, let  $w = W_{-1}(-e^{-1+t})$  and  $w' = W'_{-1}(-e^{-1+t})$ ; note that  $dw/dt = -e^{-1+t}w'$  and  $-e^{-1+t}(1+w)w' = w$ , whence

$$\frac{\mathrm{d}}{\mathrm{d}t}\left(t - \frac{(1+w)^2}{w}\right) = 1 - \frac{-2\mathrm{e}^{-1+t}(1+w)w'w + \mathrm{e}^{-1+t}w'(1+w)^2}{w^2}$$
$$= 1 - \frac{-\mathrm{e}^{-1+t}(1+w)w'(2w - (1+w))}{w^2} = 1 - \frac{w - 1}{w} = \frac{1}{w}$$

and, if t > 0,

$$\frac{\mathrm{d}}{\mathrm{d}t}(t-w) = 1 + \mathrm{e}^{-1+t}w' = 1 - \frac{w}{1+w} = \frac{1}{1+w},$$

as required. (To see the existence of  $\int_0^t 1/(1 + W_{-1}(-e^{-1+x})) dx$ , note that if  $t \ge \varepsilon > 0$  then, letting  $W_{-1}(-e^{-1+x}) = -v \cot v + iv$ , where  $v \in ]-\pi, 0[$ ,

$$\int_{\varepsilon}^{t} \left| \frac{1}{1 + W_{-1}(-e^{-1+x})} \right| dx = \int_{v(t)}^{v(\varepsilon)} \sqrt{\frac{1 - 2v \cot v + v^2 \operatorname{cosec}^2 v}{v^2}} dx$$

and the function  $v \mapsto (1 - 2v \cot v + v^2 \csc^2 v)/v^2$  is continuous on  $]-\pi, 0[$  with limit 1 as  $v \to 0-$ .)

## 5. La martingale cachée

The martingale H discussed in this section was discovered by Émery [14].

**Theorem 5.1.** If  $H_t := 1 - (Z_t + t)^{-1}$  for all  $t \ge 0$  then H is a martingale such that  $H_0 = 0$ ,

$$d[H]_t = (1 - H_{t-})^2 dt - H_{t-} dH_t$$
(18)

and  $H_t \to H_\infty := 1$  almost surely as  $t \to \infty$ .

**Proof.** If  $t \ge 0$  and  $\mathcal{E}(-Z)$  denotes the Doléans-Dade exponential of the normal martingale -Z then  $\mathcal{E}(-Z)$  is square integrable on [0, t] for all  $t \ge 0$  and (11) implies that

$$(Z_t + t)\mathcal{E}(-Z)_t$$
  
=  $Z_t + t - \int_0^t (dZ_s + ds) \int_0^t \mathcal{E}(-Z)_{s-} dZ_s$   
=  $Z_t + t - \int_0^t (Z_{s-} + s)\mathcal{E}(-Z)_{s-} dZ_s - \int_0^t (1 - \mathcal{E}(-Z)_{s-})(dZ_s + ds) - \int_0^t \mathcal{E}(-Z)_{s-} d[Z]_s$   
= 1.

Thus  $H = 1 - \mathcal{E}(-Z)$  is a martingale and  $dH_t = \mathcal{E}(-Z)_{t-} dZ_t = (1 - H_{t-}) dZ_t$ , whence

$$d[H]_t = (1 - H_{t-})^2 d[Z]_t$$
  
=  $(1 - H_{t-})^2 (dt + (1 - (1 - H_{t-})^{-1}) dZ_t)$   
=  $(1 - H_{t-})^2 dt - H_{t-} dH_t$ ,

as claimed. Since  $Y_t \to \infty$  almost surely as  $t \to \infty$ , by Corollary 3.5, so does  $Z_t + t = Y_{t+S_0}$ , and the final claim follows.

**Remark 5.2.** As  $H_t = 0$  if and only if  $Z_t + t = 1$ ,

$$\mathcal{U} := \{t \ge 0: Y_t = 1\} = \{s + S_0: Y_{s+S_0} = 1\} = \{s + S_0: H_s = 0\};\$$

the structure of  $\mathcal{U}$  is determined by the zero set of H.

Definition 5.3. Let

$$\tau: \mathbb{R}_+ \times \Omega \to \mathbb{R}_+; \qquad (t, \omega) \mapsto \tau_t(\omega) := \int_0^t \left(1 - H_{s-}(\omega)\right)^2 \mathrm{d}s$$

and note that  $\tau$  is adapted to the filtration  $(\mathcal{F}_t)_{t>0}$  and has paths which are continuous, strictly increasing and bi-*Lipschitzian on any compact subinterval of*  $\mathbb{R}_+$ *, since the derivative* 

$$\tau'_t = (1 - H_{t-})^2 = (Z_{t-} + t)^{-2} \in [b(t)^{-2}, a(t)^{-2}]$$

for all  $t \ge 0$ . Let

$$\tau_{\infty} := \int_0^{\infty} (1 - H_{s-})^2 \, \mathrm{d}s \in ]0, \infty]$$

and extend  $\tau^{-1}$  (defined pathwise) to all of  $\mathbb{R}_+$  by letting  $\tau_s^{-1} := \infty$  for all  $s \in [\tau_\infty, \infty[$ . If  $s \ge 0$  then  $\{\tau_s^{-1} \le t\} = \{s \le \tau_t\} \in \mathcal{F}_t$  for all  $t \ge 0$ , so  $\tau_s^{-1}$  is an  $(\mathcal{F}_t)_{t\ge 0}$  stopping time. Thus  $(\mathcal{G}_s := \mathcal{F}_{\tau_s^{-1}})_{s\ge 0}$  is a filtration which satisfies the usual conditions, by Lemma 1.7.

**Proposition 5.4.** The process  $K = (K_s := H_{\tau_s}^{-1})_{s \ge 0}$  is a martingale for the filtration  $(\mathcal{G}_s)_{s \ge 0}$  and satisfies the equation

$$[K]_s = s \wedge \tau_{\infty} - \int_0^s K_{r-} \,\mathrm{d}K_r \quad \forall s \ge 0.$$
<sup>(19)</sup>

**Proof.** Fix  $s \ge 0$ ; as  $\tau_s^{-1}$  is an  $(\mathcal{F}_t)_{t\ge 0}$  stopping time,  $H^{\tau_s^{-1}}$  is a martingale for this filtration ([18], Theorem II.77.4). Let  $(T_n)_{n\geq 1}$  be an increasing sequence of stopping times which reduces the local martingale  $H_- \cdot H$  and note that

$$\mathbb{E}\big[\tau_{\tau_s^{-1}\wedge T_n}-[H]_{\tau_s^{-1}\wedge T_n}\big]=\mathbb{E}\big[(H_-\cdot H)_{\tau_s^{-1}}^{T_n}\big]=0,$$

by the optional-sampling theorem. As  $\tau$  is increasing, the monotone-convergence theorem implies that

$$\mathbb{E}[s \wedge \tau_{\infty}] = \lim_{n \to \infty} \mathbb{E}[\tau_{\tau_s^{-1} \wedge T_n}] = \lim_{n \to \infty} \mathbb{E}[[H]_{\tau_s^{-1} \wedge T_n}] = \mathbb{E}[[H^{\tau_s^{-1}}]_{\infty}]$$

so  $H^{\tau_s^{-1}}$  is a square-integrable martingale ([17], Corollary 4 to Theorem II.27). Hence K is a martingale, by a further application of the optional-sampling theorem: if  $0 \le r \le s$  then

$$\mathbb{E}[K_s|\mathcal{G}_r] = \mathbb{E}\Big[H_{\infty}^{\tau_s^{-1}}|\mathcal{F}_{\tau_r^{-1}}\Big] = H_{\tau_r^{-1}} = K_r.$$

Moreover,

$$\int_0^s K_{r-} \, \mathrm{d}K_r = \int_0^{\tau_s^{-1}} K_{\tau_r^{-}} \, \mathrm{d}H_r = \int_0^{\tau_s^{-1}} H_{r-} \, \mathrm{d}H_r$$

(which follows from [17], Theorem II.21, for example), so

$$[K]_{s} = K_{s}^{2} - K_{0}^{2} - 2\int_{0}^{s} K_{r-} dK_{r} = H_{\tau_{s}^{-1}}^{2} - H_{0}^{2} - 2\int_{0}^{\tau_{s}^{-1}} H_{r-} dH_{r} = [H]_{\tau_{s}^{-1}}$$

and this equals

$$\tau_{\tau_s^{-1}} - \int_0^{\tau_s^{-1}} H_{r-} \,\mathrm{d}H_r = s \wedge \tau_\infty - \int_0^s K_{r-} \,\mathrm{d}K_r.$$

**Theorem 5.5.** Let M be Azéma's martingale, that is, the normal martingale such that  $M_0 = 0$  and

$$\mathrm{d}[M]_t = \mathrm{d}t - M_{t-} \,\mathrm{d}M_t.$$

If  $T := \inf\{t \ge 0: M_t = 1\}$  then  $M^T$  and K are identical in law.

**Proof.** Let L be a normal martingale which is independent of K such that  $L_0 = 1$  and

$$\mathrm{d}[L]_t = \mathrm{d}t - L_{t-} \,\mathrm{d}L_t,$$

that is, L is an Azéma's martingale started at 1; existence of such follows from [13], Proposition 5. For all  $t \ge 0$ , let

$$P_t := \mathbb{1}_{t \in [0, \tau_{\infty}[} K_t + \mathbb{1}_{t \in [\tau_{\infty}, \infty[} L_{t-\tau_{\infty}} = K_t + L_{(t-\tau_{\infty})^+} - 1.$$

In the notation of Lemma 1.8,  $\tau_{\infty} = \inf\{t \ge 0: K_t = 1\}$  is a  $(\mathcal{F}_t^K)_{t\ge 0}$ -stopping time, so  $]0, \tau_{\infty}]$  is  $(\mathcal{F}_t^K)_{t\ge 0}$  predictable and  $1_{]0,\tau_{\infty}]} \cdot [K] = [K]$  (since  $K = K^{\tau_{\infty}}$ ) whereas  $1_{]0,\tau_{\infty}]} \cdot [L_A] = 0$ , if  $A_t := (t - \tau_{\infty})^+$ , because  $(L_A)_t^{\tau_{\infty}} = L_{A_{t \land \tau_{\infty}}} =$ 0 for all  $t \ge 0$ . Since  $[K]_t = 2(t \land \tau_{\infty}) - K_t^2 \le 2t$  and  $[L]_{A_t} = 2A_t - L_{A_t}^2 \le 2t$ , Lemma 1.8 implies that P = $K + L_A - 1$  is a local martingale such that  $P_0 = 0$  and

$$[P]_t = [K]_t + [L]_{A_t} = t - (K_- \cdot K)_t - (L_- \cdot L)_{A_t}$$

However,

$$[P] = [K] + [L_A]$$
  
=  $K^2 - 2K_- \cdot K + L_A^2 - 1 - 2L_{A-} \cdot L_A$   
=  $(K + L_A - 1)^2 + 2K + 2L_A - 2 - 2KL_A - 2K_- \cdot K - 2(L_- \cdot L)_A$ 

and  $KL_A = P$ , so

$$P^{2} - 2P_{-} \cdot P = [P] = P^{2} - 2K_{-} \cdot K - 2(L_{-} \cdot L)_{A}.$$

Thus  $[P]_t = t - (P_- \cdot P)_t$ , so P is a normal martingale, by Theorem 1.5, and, by uniqueness ([13], Proposition 6), P is equal to M in law. Since  $\tau_{\infty} = \inf\{t \ge 0: P_t = 1\}$ , the processes  $K = P^{\tau_{\infty}}$  and  $M^T$  are identical in law, as claimed.

## 6. The level set $\mathcal{U}$

The level set

$$\mathcal{U} = \{t + S_0: H_t = 0\} = \tau^{-1} \left( \{s \in [0, \tau_\infty[: K_s = 0]\} + S_0, t_\infty[: K_s = 0] \right) + S_0,$$

where  $\tau$  is a homeomorphism between  $\mathbb{R}_+$  and  $[0, \tau_{\infty}[$  which is bi-Lipschitzian on compact subintervals. This observation, together with Theorem 5.5, leads immediately to the following theorem, thanks to well-known properties of the zero set of Azéma's martingale (or rather, by [17], Section IV.6, properties of the zero set of Brownian motion: see [8], Theorem 37.4 and [24]).

**Theorem 6.1.** The set  $\mathcal{U} := \{t \ge 0: Y_t = 1\}$  is almost surely non-empty, perfect (that is, closed and without isolated points), compact and of zero Lebesgue measure. If a > 0 then  $\mathcal{U} \cap [S_0, S_0 + a]$  has Hausdorff dimension 1/2.

**Corollary 6.2.** If T is a stopping time then  $\mathbb{P}(G_{\infty} = T) = 0$ . In particular, the final jump time  $G_{\infty}$  is not a stopping time.

**Proof.** If *T* is a stopping time then so is  $T' = \mathbb{1}_{Y_T=1}T + \mathbb{1}_{Y_T\neq 1}\infty$ ; let  $Z'_t := \mathbb{1}_{T'<\infty}(X_{t+T'} - X_{T'} + 1)$  for all  $t \ge 0$ . Conditional on  $T' < \infty$ , it holds that  $Z'_0 = 1$  and, working as in the proof of Lemma 4.2,

 $d[Z']_t = (1 - t - Z'_{t-}) dZ'_t + dt,$ 

so Z' is identical in law to Z. In particular, the set  $U \cap ]T, T + 1[$  is almost surely non-empty, given that  $Y_T = 1$ , but  $U \cap ]G_{\infty}, G_{\infty} + 1[ = \emptyset$  by definition.

**Proposition 6.3.** If S and T are random variables such that  $0 \le S \le T \le \infty$  and Y is continuous on [S, T[ (both almost surely) then

$$Y_t = -W_{\bullet} \left( \exp(-1 - t + G_S) \right) \quad \forall t \in [S, T[$$

almost surely, where  $\bullet \equiv 0$  or  $\bullet \equiv -1$  on [S, T[.

**Proof.** Working pathwise, assume S < T and note that, almost surely for all  $n \ge 1$ , there exists  $T_n \in [S, S + 1/n]$  such that  $Y_{T_n} \ne 1$  (otherwise  $Y \equiv 1$  on [S, S + 1/n], contradicting the fact that  $\mathcal{U}$  almost surely has zero Lebesgue measure). Let

$$A := \{ R \in ]T_n, T ]: Y \neq 1 \text{ on } [T_n, R[ \};$$

since  $Y_{T_n} \neq 1$ , the right-continuity of *Y* at  $T_n$  implies that *A* is non-empty. Furthermore,  $R_{\infty} := \sup A \in A$ : there exists  $(R_n)_{n\geq 1} \subseteq A$  such that  $R_n \nearrow R_{\infty}$  and  $Y \neq 1$  on  $\bigcup_{n\geq 1} [T_n, R_n] = [T_n, R_{\infty}]$ .

If  $R \in A$  then, working as in [7], Proof of Theorem 24, it follows that Y is continuously differentiable on  $[T_n, R[$  (taking the right derivative at  $T_n$ ) with Y' = Y/(Y - 1) there. Hence, by [7], Lemma 25,

 $Y_t = -W_{\bullet} \left( -Y_{T_n} \exp(-t + T_n - Y_{T_n}) \right) = -W_{\bullet} \left( -\exp(-1 - t + G_{T_n}) \right)$ 

for all  $t \in [T_n, R[$ , where  $\bullet \equiv -1$  or  $\bullet \equiv 0$ . In particular,  $Y_{R-} \neq 1$ , so if  $R_{\infty} < T$  then *Y* is continuous at  $R_{\infty}$  and  $Y_{R_{\infty}} \neq 1$ , but then there exists  $\Delta > 0$  such that  $R_{\infty} + \Delta < T$  and  $Y \neq 1$  on  $[R_{\infty}, R_{\infty} + \Delta[$ , contradicting the definition of  $R_{\infty}$ . Thus *Y* has the desired form on  $[T_n, T[$ ; letting  $n \to \infty$ , so that  $T_n \searrow S$ , gives the result.

**Corollary 6.4.** If T is a random variable such that  $Y_T = 1$  almost surely then there exists a sequence  $(T_n)_{n\geq 1}$  of random variables such that  $T_n \nearrow T$  and  $\Delta Y_{T_n} \neq 0$  almost surely.

**Proof.** Let  $T_n := \sup\{t \in [0, T]: |\Delta Y_t| > 1/(n+1)\}$  for all  $n \ge 1$ ; the sequence  $(T_n)_{n\ge 1}$  is increasing, with each  $T_n$  almost surely finite and such that  $\Delta Y_{T_n} \ne 0$  (since *Y* has càdlàg paths, so only finitely many jumps of magnitude strictly greater than 1/(n+1) on any bounded interval). If  $S := \lim_{n\to\infty} T_n$  then *Y* is continuous on [S, T[ and Proposition 6.3 implies that S = T almost surely, as required.

# 7. Local time

This section is heavily influenced by [17], Section IV.6, hence the proofs are only sketched. Thanks to Theorem 5.5, the results may also be deduced simply from the corresponding properties of Azéma's martingale (except, perhaps, for (21)).

**Definition 7.1.** Let  $\mathcal{P}$  denote the predictable  $\sigma$ -algebra on  $\mathbb{R}_+ \times \Omega$ . Recall (see [22], Section I.6, for example) that there exists a  $\mathcal{B}(\mathbb{R}) \otimes \mathcal{P}$ -measurable function

 $L: \mathbb{R} \times \mathbb{R}_+ \times \Omega \to \mathbb{R}; \qquad (v, t, \omega) \mapsto L_t^v(\omega)$ 

such that, for all  $v \in \mathbb{R}$ ,  $L^v$  is a continuous, increasing process with  $L_0^v = 0$  and

$$|Y_t - v| = |v| + \int_0^t \operatorname{sgn}(Y_{s-} - v) \, dY_s + \sum_{0 < s \le t} \left( |Y_s - v| - |Y_{s-} - v| - \operatorname{sgn}(Y_{s-} - v) \Delta Y_s \right) + L_t^v$$
(20)

for all  $t \ge 0$  almost surely, where  $sgn(x) := \mathbb{1}_{x>0} - \mathbb{1}_{x \le 0}$  for all  $x \in \mathbb{R}$ .

**Remark 7.2.** Since X is purely discontinuous ([7], Lemma 23),  $[Y]^c = [X]^c = 0$ ; by the occupation-density formula ([17], Corollary 2 to Theorem IV.51), there exists a null set  $N \subseteq \Omega$  such that

$$0 = \int_0^\infty [Y]_t^c(\omega) \, \mathrm{d}t = \int_{-\infty}^\infty \int_0^\infty L_t^v(\omega) \, \mathrm{d}t \, \mathrm{d}v \quad \forall \omega \in \Omega \setminus N,$$

and so, almost surely,  $L^{v} \equiv 0$  on  $\mathbb{R}_{+}$  for almost all  $v \in \mathbb{R}$ . The following theorem gives a more exact result.

**Theorem 7.3.** If  $v \neq 1$  then the local time  $L^{v} = 0$ , whereas

$$\mathbb{E}\left[L_t^1\right] = 2\int_0^t g_\infty(x) \,\mathrm{d}x > 0 \tag{21}$$

and the random variable  $L_t^1$  is not almost surely zero for all t > 0.

**Proof.** If v = 0 then (20) implies that

$$|Y_{t+S_0}| = -\int_0^{S_0} dY_s + \int_{S_0}^{t+S_0} dY_s + 2\sum_{0 < s \le t+S_0} \mathbb{1}_{Y_{s-}=0} \Delta Y_s + L^0_{t+S_0}$$
$$= -1 + Y_{t+S_0} - 1 + 2 + L^0_{t+S_0}$$

for all  $t \ge 0$ , so  $L^0 = 0$ . (The first equality uses the local character of the stochastic integral ([17], Corollary to Theorem II.18).) If  $v \notin \{0, 1\}$  then the set  $\{s > 0: Y_{s-} = Y_s = v\}$  is countable and the claim follows as it does in [17], Proof of Theorem IV.63. For the remaining case, observe that the Meyer–Tanaka–Itô formula (or just [17], Theorem IV.49) yields, for all  $t \ge 0$ , the identity

$$(Y_t - 1)^+ = \int_0^t \mathbb{1}_{Y_{s-} > 1} \, \mathrm{d}Y_s + \frac{1}{2} L_t^1.$$

Since

$$\mathbb{E}\left[\int_0^t \mathbb{1}_{Y_{s-}>1} \,\mathrm{d}s\right] = \mathbb{E}\left[\int_0^t \mathbb{1}_{Y_s>1} \,\mathrm{d}s\right] = \int_0^t \mathbb{P}(Y_s>1) \,\mathrm{d}s,$$

as  $\{s > 0: Y_{s-} \neq Y_s\}$  is countable and thus has zero Lebesgue measure, it follows that

$$\mathbb{E}\left[L_t^1\right] = 2\mathbb{E}\left[\left(Y_t - 1\right)^+\right] - 2\int_0^t \mathbb{P}(Y_s > 1) \,\mathrm{d}s.$$

For all  $t \ge 0$  and  $x \ge 0$ , let  $F_{Y_t}(x) := \mathbb{P}(Y_t \le x)$ ; Lemma 2.3 implies that

$$\mathbb{E}[(Y_t-1)^+] = \int_1^\infty (x-1) \, \mathrm{d}F_{Y_t}(x) = \frac{1}{\pi} \int_1^{b(t)} \mathrm{Im} \, \frac{x-1}{W_{-1}(-x\mathrm{e}^{t-x})} \, \mathrm{d}x = \int_0^t b(t-y)g_\infty(y) \, \mathrm{d}y,$$

using the substitution x = b(t - y), and similarly

$$\int_0^t \mathbb{P}(Y_s > 1) \, \mathrm{d}s = \frac{1}{\pi} \int_0^t \int_1^{b(s)} \operatorname{Im} \frac{1}{W_{-1}(-xe^{s-x})} \, \mathrm{d}x \, \mathrm{d}s$$
$$= \int_0^t \int_0^s b'(s-y)g_\infty(y) \, \mathrm{d}y \, \mathrm{d}s = \int_0^t (b(t-y)-1)g_\infty(y) \, \mathrm{d}y.$$

Combining these calculations yields (21).

**Definition 7.4.** A semimartingale R has locally summable jumps (or satisfies Hypothesis A, in the terminology of [17]) if

$$\sum_{0 < s \le t} |\Delta R_s| < \infty \quad almost \ surely \ \forall t > 0.$$

Corollary 7.5. The martingale X does not have locally summable jumps.

**Proof.** Suppose for contradiction that *X* (and so *Y*) has locally summable jumps. By [17], Theorem IV.56, there exists a  $\mathcal{B}(\mathbb{R}) \otimes \mathcal{P}$ -measurable function

 $\widetilde{L} : \mathbb{R} \times \mathbb{R}_+ \times \mathcal{Q} \to \mathbb{R}_+; \qquad (v,t,\omega) \mapsto \widetilde{L}_t^v(\omega)$ 

such that  $(v, t) \mapsto \widetilde{L}_t^v(\omega)$  is jointly right continuous in v and continuous in t for all  $\omega \in \Omega$  and, for all  $v \in \mathbb{R}$ ,  $\widetilde{L}^v = L^v$ . This is, however, readily seen to contradict Theorem 7.3.

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## Appendix A. A Poisson limit theorem

The following theorem must be well known, but a reference for it (or a version with weaker hypotheses) has proved elusive.

**Theorem A.1.** For all  $n \ge 1$  let  $(x_{n,m})_{m=1}^n$  be a collection of independent, identically distributed random variables. *If there exists*  $\lambda > 0$  *such that* 

$$\lim_{n \to \infty} n \mathbb{E} \big[ x_{n,1}^k \big] = \lambda \quad \forall k \in \mathbb{N}.$$

then  $x_{n,1} + \cdots + x_{n,n}$  converges in distribution to a Poisson law with mean  $\lambda$ .

**Proof.** If  $n \ge 1$  and  $\theta \in \mathbb{R}$  then

$$\left|\mathbb{E}\left[\exp\left(\mathrm{i}\theta(x_{n,1}+\cdots+x_{n,n})\right)\right] - \left(1+\frac{\lambda}{n}\left(\mathrm{e}^{\mathrm{i}\theta}-1\right)\right)^n\right| \le n \left|\mathbb{E}\left[\mathrm{e}^{\mathrm{i}\theta x_{n,1}}\right] - 1 - \frac{\lambda(\mathrm{e}^{\mathrm{i}\theta}-1)}{n} \left|\left(1+\frac{2\lambda}{n}\right)^{n-1}\right|\right|$$

using the fact that  $|z^n - w^n| \le n|z - w| \max_{1 \le k \le n} \{|z|^{k-1}|w|^{n-k}\}$  for all  $z, w \in \mathbb{C}$ . Furthermore, because  $|e^{i\theta} - \sum_{k=0}^{2p-1} (i\theta)^k / k!| \le \theta^{2p} / (2p)!$  for all  $\theta \in \mathbb{R}$  and  $p \ge 1$ ,

$$n \left| \mathbb{E} \left[ e^{i\theta x_{n,1}} \right] - 1 - \frac{\lambda(e^{i\theta} - 1)}{n} \right|$$
  
$$\leq n \left| \mathbb{E} \left[ e^{i\theta x_{n,1}} - \sum_{k=0}^{2p-1} \frac{(i\theta x_{n,1})^k}{k!} \right] \right| + \sum_{k=1}^{2p-1} \frac{|\theta|^k}{k!} \left| n \mathbb{E} \left[ x_{n,1}^k \right] - \lambda \right| + \lambda \left| e^{i\theta} - \sum_{k=0}^{2p-1} \frac{(i\theta)^k}{k!} \right|$$

$$\leq \frac{|\theta|^{2p}(n\mathbb{E}[x_{n,1}^{2p}]+\lambda)}{(2p)!} + \sum_{k=1}^{2p-1} \frac{|\theta|^k}{k!} |n\mathbb{E}[x_{n,1}^k]-\lambda|.$$

Since  $(1 + 2\lambda/n)^{n-1} \to e^{2\lambda}$  as  $n \to \infty$ , this sequence is bounded by some constant *C*. Fix  $\varepsilon > 0$ , choose  $p \ge 1$  such that  $2|\theta|^{2p}\lambda/(2p)! < \varepsilon/(2C)$  and choose  $n_0$  such that

$$\frac{|\theta|^k}{k!} |n\mathbb{E}[x_{n,1}^k] - \lambda| < \frac{\varepsilon}{4pC} \quad \forall n \ge n_0, k = 1, \dots, 2p;$$

the previous working shows that

$$\left|\mathbb{E}\left[\exp\left(\mathrm{i}\theta(x_{n,1}+\cdots+x_{n,n})\right)\right] - \left(1 + \frac{\lambda}{n}\left(\mathrm{e}^{\mathrm{i}\theta}-1\right)\right)^n\right| < \frac{2|\theta|^{2p}\lambda C}{(2p)!} + \frac{\varepsilon}{4p} + (2p-1)\frac{\varepsilon}{4p} < \varepsilon \quad \forall n \ge n_0.$$

 $\square$ 

Hence

$$\lim_{n\to\infty} \mathbb{E}\Big[\exp\Big(\mathrm{i}\theta(x_{n,1}+\cdots+x_{n,n})\Big)\Big] = \lim_{n\to\infty} \left(1+\frac{\lambda}{n}(\mathrm{e}^{\mathrm{i}\theta}-1)\right)^n = \exp\big(\lambda(\mathrm{e}^{\mathrm{i}\theta}-1)\big),$$

and the result follows from the continuity theorem for characteristic functions ([8], Theorem 26.3).

*Remark A.2.* It follows from the working above that, if  $m \ge 1$  and  $\theta \in \mathbb{R}$ ,

$$\mathbb{E}\left[e^{\mathrm{i}\theta x_{n,m}}\right] = 1 + \left(\frac{\lambda}{n}\right)\left(e^{\mathrm{i}\theta} - 1\right) + \mathrm{o}\left(\frac{1}{n}\right) = \mathbb{E}\left[e^{\mathrm{i}\theta b_{n}}\right] + \mathrm{o}\left(\frac{1}{n}\right) \to 1$$

as  $n \to \infty$ , where  $\mathbb{P}(b_n = 0) = 1 - \lambda/n$  and  $\mathbb{P}(b_n = 1) = \lambda/n$ . Thus  $x_{n,m}$  converges to 0 in distribution, and so in probability, as  $n \to \infty$ , which explains why this result is a "law of small numbers".

## Appendix B. The probability density function $g_{\infty}$

Proposition B.1. The function

$$g_{\infty} : \mathbb{R} \to \mathbb{R}_+; \qquad x \mapsto \mathbb{1}_{x \ge 0} \frac{1}{\pi} \operatorname{Im} \frac{1}{W_{-1}(-e^{-1+x})}$$

has a global maximum  $g_{\infty}(x_0) \approx 0.2306509575$  at  $x_0 \approx 0.7376612533$ , is strictly increasing on  $[0, x_0]$ , is strictly decreasing on  $[x_0, \infty[$  with  $\lim_{x\to\infty} g_{\infty}(x) = 0$ ,

$$\int_0^\infty g_\infty(x) \, \mathrm{d}x = 1 \quad and \quad \int_0^\infty x g_\infty(x) \, \mathrm{d}x = \infty.$$

**Proof.** Let  $W_{-1}(-e^{-1+x}) = u(x) + iv(x)$  for all  $x \ge 0$ , where  $u(x) \in \mathbb{R}$  and  $v(x) \in [-\pi, 0]$ . Since

 $(u+\mathrm{i}v)\exp(u+\mathrm{i}v) = -\exp(-1+x) \quad \Longleftrightarrow \quad \begin{cases} \mathrm{e}^u(u\cos v - v\sin v) = -\mathrm{e}^{-1+x}, \\ u\sin v + v\cos v = 0, \end{cases}$ 

if v = 0 then  $ue^u = -e^{-1+x}$ , which has no solution for x > 0, so v = 0 if and only if x = 0. Suppose henceforth that x > 0; note that  $u = -v \cot v$ ,

$$e^{-v \cot v} (-v \cos v \cot v - v \sin v) = -e^{-1+x} \quad \Longleftrightarrow \quad x = 1 - v \cot v + \log(v \operatorname{cosec} v)$$

and  $\pi g_{\infty}(x) = -v/(u^2 + v^2) > 0$ . Observe that

$$\frac{du}{dv} = -\cot v + v\csc^2 v = \frac{1}{\sin^2 v}(v - \sin v\cos v) < \frac{1}{\sin^2 v}(0 - \sin 0\cos 0) = 0.$$

because  $(d/dv)(v - \sin v \cos v) = 1 - \cos 2v > 0$ , and

$$\frac{\mathrm{d}x}{\mathrm{d}v} = \frac{1}{v} - 2\cot v + v\csc^2 v = \frac{1}{v}(1 - v\csc v)^2 - \frac{2}{\sin v}(\cos v - 1) < 0,$$

so *u* is a strictly increasing function of *x*. As u(0) = -1, *u* takes its values in  $[-1, \infty[; \text{ as } v(0) = 0$ , letting  $x = 1 - v \cot v + \log(v \csc v) \rightarrow \infty$  shows that  $v \rightarrow -\pi$  (since this function of *v* is bounded on any proper subinterval of  $]-\pi, 0[$ ) and therefore  $u \rightarrow \infty$  as  $x \rightarrow \infty$ . (In particular,  $|g_{\infty}(x)| \le 1/u^2 \rightarrow 0$  as  $x \rightarrow \infty$ .) Since *u* is continuous, strictly increasing and maps  $[0, \infty[$  to  $[-1, \infty[$ , there exists  $x_0$  such that  $u(x_0) = -1/2$ . Moreover,

$$g'_{\infty}(x) = \operatorname{Im} \frac{\mathrm{d}}{\mathrm{d}x} \frac{1}{W_{-1}(-\mathrm{e}^{-1+x})} = \operatorname{Im} \frac{-1}{(u+\mathrm{i}v)(1+u+\mathrm{i}v)} = \frac{v(2u+1)}{(u^2+v^2)((1+u)^2+v^2)}$$

so  $g'_{\infty} > 0$  on  $]0, x_0[$  and  $g'_{\infty} < 0$  on  $]x_0, \infty[$ . (The approximate values for  $x_0$  and  $g_{\infty}(x_0)$  were determined with the use of Maple.)

For the integrals, the substitution x = v gives that

$$\pi \int_0^\infty g_\infty(x) \, \mathrm{d}x = \int_{-\pi}^0 \frac{\sin^2 v}{v} \left( \frac{1}{v} - 2\cot v + v \operatorname{cosec}^2 v \right) \mathrm{d}v$$
$$= \pi + \int_{-\pi}^0 \left( \frac{\sin^2 v}{v^2} - \frac{\sin 2v}{v} \right) \mathrm{d}v = \pi + \left[ -\frac{\sin^2 v}{v} \right]_{-\pi}^0 = \pi,$$

as required. Finally, if  $\varepsilon \in (0, \pi/2)$ 

$$\pi \int_0^\infty x g_\infty(x) \, \mathrm{d}x = \int_{-\pi}^0 \left( 1 - v \cot v + \log(v \operatorname{cosec} v) \right) \left( \frac{\sin^2 v}{v^2} - \frac{\sin 2v}{v} + 1 \right) \mathrm{d}v$$
$$\geq \int_{-\pi+\varepsilon}^{-\pi/2} -v \cot v \, \mathrm{d}v \geq \frac{\pi}{2} \int_{\varepsilon}^{\pi/2} \cot w \, \mathrm{d}w = -\log \sin \varepsilon \to \infty$$

as  $\varepsilon \to 0+$ .

**Remark B.2.** It follows from Propositions B.1 and 3.1 that the distribution of  $G_{\infty}$  is unimodal with mode  $x_0$ , that is,  $t \mapsto \mathbb{P}(G_{\infty} \leq t)$  is convex on  $]-\infty, x_0[$  and concave on  $]x_0, \infty[$ .

#### Appendix C. An auxiliary calculation

**Lemma C.1.** If  $f_J$  is as defined in Proposition 4.6 then

$$\pi f_J(t) = \operatorname{Im} \frac{1}{1 + W_{-1}(-e^{-1+t})} \sim \frac{1}{\sqrt{2t}} \quad as \ t \to 0 + t$$

and  $f_J$  is strictly decreasing on  $]0, \infty[$ .

**Proof.** For all  $t \ge 0$ , let  $p := -\sqrt{2(1 - e^t)} = -i\sqrt{2t} + O(t^{3/2})$  as  $t \to 0+$ ; recall that

$$-W_{-1}(-e^{-1+t}) = 1 - p + O(p^2) = 1 + i\sqrt{2t} + O(t)$$

as  $t \to 0+$ , by [10], (4.22), and this gives the first result. For the next claim, if t > 0 and  $W_{-1}(-e^{-1+t}) = -v \cot v + iv$ , where  $v \in ]-\pi, 0[$ , then

$$\pi f'_J(t) = \operatorname{Im} \frac{-W_{-1}(-e^{-1+t})}{(1+W_{-1}(-e^{-1+t}))^3} = \frac{((3-2v\cot v)v^2\csc^2 v - 1)v}{((1-v\cot v)^2 + v^2)^3}.$$

The result follows if

$$(3 - 2v \cot v)v^2 \csc^2 v - 1 > 0 \quad \iff \quad (v^2 - \sin^2 v) \sin v + 2v^2 (\sin v - v \cos v) < 0$$

for all  $v \in [-\pi, 0[$ , but since  $\sin^2 v < v^2$  and  $\sin v < v \cos v$  for such v, this is clear.

**Proposition C.2.** *If*  $D := \{(t, x) \in \mathbb{R}^2_+ : a(t) \le x \le b(t)\},\$ 

$$f: D \to \mathbb{R}_+; \qquad (t, x) \mapsto \operatorname{Im} \frac{1}{W_{-1}(-xe^{t-x})}$$
$$F: D \to \mathbb{R}_+; \qquad (t, x) \mapsto \int_{a(t)}^x f(t, y) \, \mathrm{d}y$$

and  $(s, y) \in D^{\circ} := \{(t, x) \in \mathbb{R}^2_+: t > 0, a(t) < x < b(t)\}$  then

$$F(s, y) + \frac{\partial F}{\partial t}(s, y) = \int_{a(s)}^{y} \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{s-z})} \, \mathrm{d}z.$$
(22)

**Proof.** Note first that, since f is continuous, F is well defined. If h > 0 then

$$\frac{F(s+h, y) - F(s, y)}{h} = \frac{1}{h} \int_{a(s+h)}^{a(s)} f(s+h, z) \, \mathrm{d}z + \int_{a(s)}^{y} \frac{f(s+h, z) - f(s, z)}{h} \, \mathrm{d}z$$

and the intermediate-value theorem gives  $\zeta_h \in [a(s+h), a(s)]$  such that

$$\frac{1}{h} \int_{a(s+h)}^{a(s)} f(s+h,z) \, \mathrm{d}z = \frac{a(s) - a(s+h)}{h} f(s+h,\zeta_h) \to -a'(s) f\left(s,a(s)\right) = 0$$

as  $h \to 0+$ . For all  $z \in [a(s), b(s)]$  there exists  $\theta_{h,z} \in [0, 1[$  such that

$$\frac{f(s+h,z) - f(s,z)}{h} = \frac{\partial f}{\partial t}(s + \theta_{h,z}h, z)$$

by the mean-value theorem, since  $t \mapsto f(t, z)$  is continuous on [s, s + h] and differentiable on ]s, s + h[. Let

$$g: D^{\circ} \to \mathbb{R}_{+}; \qquad (t, x) \mapsto \frac{\partial f}{\partial t}(t, x) + f(t, x) = \begin{cases} \pi f_{J}\left(t - a^{-1}(x)\right) & \text{if } x \in \left]a(t), 1\right] \\ \pi f_{J}\left(t - b^{-1}(x)\right) & \text{if } x \in \left[1, b(t)\right] \end{cases}$$

where  $f_J$  is defined in Proposition 4.6. The continuity of f on  $[s, s + 1] \times [a(s), y]$  and the dominated-convergence theorem imply that

$$F(s, y) = \int_{a(s)}^{y} f(s, z) \, dz = \lim_{h \to 0+} \int_{a(s)}^{y} f(s + \theta_{h,z}h, z) \, dz,$$

so the right-hand limit in (22) has the correct value if  $\int_{a(s)}^{y} g(s, z) dz$  exists and

$$\lim_{h \to 0+} \int_{a(s)}^{y} g(s + \theta_{h,z}h, z) \, \mathrm{d}z = \int_{a(s)}^{y} g(s, z) \, \mathrm{d}z.$$

Fix  $r \in [0, s[$  such that y > a(r) and note that g is continuous on  $[s, s + 1] \times [a(r), y]$ , so the dominated-convergence theorem implies that

$$\lim_{h \to 0+} \int_{a(r)}^{y} g(s + \theta_{h,z}h, z) \, \mathrm{d}z = \int_{a(r)}^{y} g(s, z) \, \mathrm{d}z.$$

Next, note that if  $z \in [a(s), a(r)]$  and  $h \to 0+$  then

$$g(s + \theta_{h,z}h, z) = \pi f_J \left( s + \theta_{h,z}h - a^{-1}(z) \right) \nearrow \pi f_J \left( s - a^{-1}(z) \right) = g(s, z)$$

because  $f_J$  is strictly decreasing, by Lemma C.1. The first half of the result now follows from the monotoneconvergence theorem, once it is known that  $\int_{a(s)}^{a(r)} g(s, z) dz$  exists. However,

$$\int_{a(s)}^{a(r)} g(s,z) \, \mathrm{d}z = \pi \int_{s}^{r} f_{J}(s-u)a'(u) \, \mathrm{d}u = -\pi \int_{0}^{s-r} f_{J}(t)a'(s-t) \, \mathrm{d}t < \infty,$$

since, by Lemma C.1,  $\pi f_J(t) \sim 1/\sqrt{2t}$  as  $t \to 0+$ ,  $f_J$  is continuous on [0, s-r] and a' is continuous on [r, s]. Now suppose that h < 0 is such that s + h > 0 and b(s + h) > y > a(s + h). Then

$$\frac{F(s+h,y) - F(s,y)}{h} = \int_{a(s+h)}^{y} \frac{f(s+h,z) - f(s,z)}{h} \, \mathrm{d}z - \frac{1}{h} \int_{a(s)}^{a(s+h)} f(s,z) \, \mathrm{d}z$$

and the second term tends to 0 as  $h \to 0-$ . If  $z \in [a(s+h), b(s+h)]$  then  $t \mapsto f(t, z)$  is continuous on [s+h, s] and differentiable on ]s + h, s[, so there exists  $\theta_{h,z} \in ]0, 1[$  such that

$$\frac{f(s+h,z)-f(s,z)}{h} = \frac{\partial f}{\partial t}(s+\theta_{h,z}h,z).$$

Furthermore, as f is continuous, so bounded, on the compact set  $D \cap ([0, s] \times \mathbb{R}_+)$ , the dominated-convergence theorem implies that

$$F(s, y) = \lim_{h \to 0^-} \int_{a(s+h)}^{y} f(s + \theta_{h,z}h, z) \, \mathrm{d}z$$

and the result follows if

$$\lim_{h \to 0^-} \int_{a(s+h)}^{y} g(s + \theta_{h,z}h, z) \, \mathrm{d}z = \int_{a(s)}^{y} g(s, z) \, \mathrm{d}z$$

Fix  $0 < r_1 < r_2 < s$  such that  $a(r_1) < y$  and note that g is continuous on  $[r_2, s] \times [a(r_1), y]$ , so bounded there, and the dominated-convergence theorem implies that

$$\int_{a(r_1)}^{y} g(s + \theta_{h,z}h, z) \, \mathrm{d}z \to \int_{a(r_1)}^{y} g(s, z) \, \mathrm{d}z$$

as  $h \to 0-$ . A final application of the monotone-convergence theorem completes the result, since if h < 0 is such that  $r_2 < s + h$  then, letting  $h \to 0-$ ,

$$\begin{split} \mathbb{1}_{z \in [a(s+h), a(r_1)]} g(s + \theta_{h, z}h, z) &= \mathbb{1}_{z \in [a(s+h), a(r_1)]} \pi f_J \big( s + \theta_{h, z}h - a^{-1}(z) \big) \\ &\nearrow \mathbb{1}_{z \in ]a(s), a(r_1)]} \pi f_J \big( s - a^{-1}(z) \big) \\ &= \mathbb{1}_{z \in ]a(s), a(r_1)]} g(s, z). \end{split}$$

## Appendix D. A pair of Laplace transforms

**Theorem D.1.** If  $g_{\infty}$  is as defined in Proposition 3.1 and  $f_J$  is as defined in Proposition 4.6 then their Laplace transforms are as follows:

$$\widehat{g_{\infty}}(p) = \frac{e^{-p}p^p}{\Gamma(p+2)} \quad and \quad \widehat{f_J}(p) = (p+1)\widehat{g_{\infty}}(p) = \frac{e^{-p}p^p}{\Gamma(p+1)},$$
(23)

where  $\Gamma: p \mapsto \int_0^\infty z^{p-1} e^{-z} dz$  is the gamma function.

Proof. Let

$$f_1(t) := \frac{1}{\pi} \int_{a(t)}^{b(t)} \operatorname{Im} \frac{1}{W_{-1}(-y e^{t-y})} \, \mathrm{d}y \quad \forall t \ge 0.$$

Splitting the interval [a(t), b(t)] at 1 and using the substitutions y = a(t - x) and y = b(t - x), as appropriate,

$$f_1(t) = \frac{1}{\pi} \int_0^t \operatorname{Im}\left(\frac{1}{W_{-1}(-e^{-1+x})}\right) c(t-x) \, \mathrm{d}x = (g_\infty \star c)(t),$$

where  $\star$  denotes convolution of functions on  $\mathbb{R}_+$  and *c* is as in Definition 2.2. Furthermore,

$$\widehat{c}(p) := \int_0^\infty c(x) e^{-px} \, dx = \int_0^\infty b'(x) e^{-px} \, dx - \int_0^\infty a'(x) e^{-px} \, dx$$
$$= \int_1^\infty e^{-p(-1+y-\log y)} \, dy + \int_0^1 e^{-p(-1+y-\log y)} \, dy$$
$$= e^p \int_0^\infty \left(\frac{z}{p}\right)^p e^{-z} p^{-1} \, dz.$$

The second line follows from the substitutions y = b(x) and y = a(x). Thus, since  $f_1(t) = \mathbb{P}(Y_t > 0) = 1 - e^{-t}$ ,

$$\widehat{g_{\infty}}(p) = \frac{\widehat{f_1}(p)}{\widehat{c}(p)} = \frac{1}{p(p+1)} \frac{\mathrm{e}^{-p} p^{p+1}}{\Gamma(p+1)} = \frac{\mathrm{e}^{-p} p^p}{\Gamma(p+2)},$$

as claimed. If

$$f_2(t) := \frac{1}{\pi} \int_{a(t)}^{b(t)} \operatorname{Im} \frac{1}{1 + W_{-1}(-z e^{t-z})} \, \mathrm{d}y \quad \forall t > 0,$$

then, working as above,  $f_2 = f_J \star c$ . Moreover, since  $f_2 = f_1 + f'_1$  (by the working in the proof of Theorem 4.5), it follows that  $\hat{f}_2(p) = (p+1)\hat{f}_1(p)$  and

$$\widehat{f_J}(p) = \frac{\widehat{f_2}(p)}{\widehat{c}(p)} = \frac{(p+1)\widehat{f_1}(p)}{\widehat{c}(p)} = \frac{e^{-p}p^p}{\Gamma(p+1)}.$$

**Remark D.2.** The substitution  $x = 1 - v \cot v + \log(v \csc v)$  yields the identity

$$e^{p}\widehat{f_{J}}(p) = \frac{1}{\pi} \int_{0}^{\pi} \left(\frac{\sin v}{v}\right)^{p} \exp(pv \cot v) dv;$$
(24)

it should be possible to verify directly that the right-hand side of (24) equals  $p^p/\Gamma(p+1)$ . (This would give independent proof that

$$A \mapsto \mathbb{1}_{0 \in A} + \frac{1}{\pi} \int_{A \cap [a(t), b(t)]} \operatorname{Im} \frac{1}{W_{-1}(-y e^{t-y})} \, \mathrm{d}y$$

and

$$A \mapsto \frac{1}{\pi} \int_{A \cap [a(t), b(t)]} \operatorname{Im} \frac{1}{1 + W_{-1}(-ze^{t-z})} dz$$

are probability measures on  $\mathcal{B}(\mathbb{R})$ .)

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