Pruning a Lévy continuum random tree*

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Abstract

Given a general critical or sub-critical branching mechanism, we define a pruning procedure of the associated Lévy continuum random tree. This pruning procedure is defined by adding some marks on the tree, using Lévy snake techniques. We then prove that the resulting subtree after pruning is still a Lévy continuum random tree. This last result is proved using the exploration process that codes the CRT, a special Markov property and martingale problems for exploration processes. We finally give the joint law under the excursion measure of the lengths of the excursions of the initial exploration process and the pruned one.

Key words: continuum random tree, Lévy snake, special Markov property.

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Continuous state branching processes (CSBP) were first introduced by Jirina [25] and it is known since Lamperti [27] that these processes are the scaling limits of Galton-Watson processes. They hence model the evolution of a large population on a long time interval. The law of such a process is characterized by the so-called branching mechanism function \( \psi \). We will be interested mainly in critical or sub-critical CSBP. In those cases, the branching mechanism \( \psi \) is given by

\[
\psi(\lambda) = \alpha \lambda + \beta \lambda^2 + \int_{(0,+\infty)} \pi(d\ell) \left( e^{-\lambda \ell} - 1 + \lambda \ell \right), \quad \lambda \geq 0,
\]

with \( \alpha \geq 0, \beta \geq 0 \) and the Lévy measure \( \pi \) is a positive \( \sigma \)-finite measure on \((0, +\infty)\) such that \( \int_{(0, +\infty)} (\ell \wedge \ell^2) \pi(d\ell) < \infty \). We shall say that the branching mechanism \( \psi \) has parameter \((\alpha, \beta, \pi)\).

As for discrete Galton-Watson processes, we can associate with a CSBP a genealogical tree, see [30] or [22]. These trees can be considered as continuum random trees (CRT) in the sense that the branching points along a branch form a dense subset. We call the genealogical tree associated with a branching mechanism \( \psi \) the \( \psi \)-Lévy CRT (the term “Lévy” will be explained later). The prototype of such a tree is the Brownian CRT introduced by Aldous [9].

In a discrete setting, it is easy to consider and study a percolation on the tree (for instance, see [11] for percolation on the branches of a Galton-Watson tree, or [6] for percolation on the nodes of a Galton-Watson tree). The goal of this paper is to introduce a general pruning procedure of a genealogical tree associated with a branching mechanism \( \psi \) of the form (1), which is the continuous analogue of the previous percolation (although no link is actually made between both). We first add some marks on the skeleton of the tree according to a Poisson measure with intensity \( \alpha \lambda \) where \( \lambda \) is the length measure on the tree (see the definition of that measure further) and \( \alpha_1 \) is a non-negative parameter. We next add some marks on the nodes of infinite index of the tree: with such a node \( s \) is associated a “weight” say \( \Delta_s \) (see later for a formal definition), each infinite node is then marked with probability \( p(\Delta_s) \) where \( p \) is a non-negative measurable function satisfying the integrability condition

\[
\int_{(0,+\infty)} \ell \, p(\ell) \, \pi(d\ell) < +\infty.
\]

We then prune the tree according to these marks and consider the law of the pruned subtree containing the root. The main result of the paper is the following theorem:

**Theorem 0.1.** Let \( \psi \) be a (sub)-critical branching mechanism of the form (1). We define

\[
d\pi_0(x) := (1 - p(x))d\pi(x)
\]

\[
\alpha_0 := \alpha + \alpha_1 + \int_{(+\infty)} \ell p(\ell) \pi(d\ell)
\]

and set

\[
\psi_0(\lambda) = \alpha_0 \lambda + \beta \lambda + \int_{(0,+\infty)} \pi_0(d\ell) \left( e^{-\lambda \ell} - 1 + \lambda \ell \right)
\]

which is again a branching mechanism of a critical or subcritical CSBP.

Then, the pruned subtree is a Lévy-CRT with branching mechanism \( \psi_0 \).
In order to make the previous statement more rigorous, we must first describe more precisely the geometric structure of a continuum random tree and define the so-called exploration process that codes the CRT in the next subsection. In a second subsection, we describe the pruning procedure and state rigorously the main results of the paper. Eventually, we give some biological motivations for studying the pruning procedure and other applications of this work.

0.1 The Lévy CRT and its coding by the exploration process

We first give the definition of a real tree, see e.g. [24] or [28].

**Definition 0.2.** A metric space $(\mathcal{T}, d)$ is a real tree if the following two properties hold for every $v_1, v_2 \in \mathcal{T}$.

(i) There is a unique isometric map $f_{v_1, v_2}$ from $[0, d(v_1, v_2)]$ into $\mathcal{T}$ such that

$$f_{v_1, v_2}(0) = v_1 \quad \text{and} \quad f_{v_1, v_2}(d(v_1, v_2)) = v_2.$$ 

(ii) If $q$ is a continuous injective map from $[0, 1]$ into $\mathcal{T}$ such that $q(0) = v_1$ and $q(1) = v_2$, then we have

$$q([0, 1]) = f_{v_1, v_2}([0, d(v_1, v_2)]).$$ 

A rooted real tree is a real tree $(\mathcal{T}, d)$ with a distinguished vertex $v_0$ called the root.

Let $(\mathcal{T}, d)$ be a rooted real tree. The range of the mapping $f_{v_1, v_2}$ is denoted by $[[v_1, v_2]]$ (this is the line between $v_1$ and $v_2$ in the tree). In particular, for every vertex $v \in \mathcal{T}$, $[[v_0, v]]$ is the path going from the root to $v$ which we call the ancestral line of vertex $v$. More generally, we say that a vertex $v$ is an ancestor of a vertex $v'$ if $v \in [[v_0, v']]$. If $v, v' \in \mathcal{T}$, there is a unique $a \in \mathcal{T}$ such that $[[v_0, v]] \cap [[v_0, v']] = [[v_0, a]]$. We call $a$ the most recent common ancestor of $v$ and $v'$. By definition, the degree of a vertex $v \in \mathcal{T}$ is the number of connected components of $\mathcal{T} \setminus \{v\}$. A vertex $v$ is called a leaf if it has degree 1. Finally, we set $\lambda$ the one-dimensional Hausdorff measure on $\mathcal{T}$.

The coding of a compact real tree by a continuous function is now well known and is a key tool for defining random real trees. We consider a continuous function $g : [0, +\infty) \longrightarrow [0, +\infty)$ with compact support and such that $g(0) = 0$. We also assume that $g$ is not identically 0. For every $0 \leq s \leq t$, we set

$$m_g(s, t) = \inf_{u \in [s, t]} g(u),$$

and

$$d_g(s, t) = g(s) + g(t) - 2m_g(s, t).$$

We then introduce the equivalence relation $s \sim t$ if and only if $d_g(s, t) = 0$. Let $\mathcal{T}_g$ be the quotient space $[0, +\infty) / \sim$. It is easy to check that $d_g$ induces a distance on $\mathcal{T}_g$. Moreover, $(\mathcal{T}_g, d_g)$ is a compact real tree (see [21], Theorem 2.1). The function $g$ is the so-called height process of the tree $\mathcal{T}_g$. This construction can be extended to more general measurable functions.

In order to define a random tree, instead of taking a tree-valued random variable, it suffices to take a stochastic process for $g$. For instance, when $g$ is a normalized Brownian excursion, the associated real tree is Aldous’ CRT [10].
The construction of a height process that codes a tree associated with a general branching mechanism is due to Le Gall and Le Jan [30]. Let $\psi$ be a branching mechanism given by (1) and let $X$ be a Lévy process with Laplace exponent $\psi$: $\mathbb{E}[e^{-\lambda X_t}] = e^{\lambda \psi(\lambda)}$ for all $\lambda \geq 0$. Following [30], we also assume that $X$ is of infinite variation a.s. which implies that $\beta > 0$ or $\int_{(0,1]} \ell \pi(d\ell) = \infty$. Notice that these conditions are satisfied in the stable case: $\psi(\lambda) = \lambda^\epsilon$, $c \in (1,2]$ (the quadratic case $\psi(\lambda) = \lambda^2$ corresponds to the Brownian case).

We then set
\[
H_t = \liminf_{\varepsilon \to 0} -\frac{1}{\varepsilon} \int_0^t 1_{[X,s]} ds
\]
where for $0 \leq s \leq t$, $I_t^s = \inf_{s \leq r \leq t} X_r$. If the additional assumption
\[
\int_1^{+\infty} \frac{du}{\psi(u)} < \infty
\]
holds, then the process $H$ admits a continuous version. In this case, we can consider the real tree associated with an excursion of the process $H$ and we say that this real tree is the Lévy CRT associated with $\psi$. If we set $L_t^a(H)$ the local time of the process $H$ at level $a$ and time $t$ and $T_x = \inf\{ t \geq 0, L_t^0(H) = x \}$, then the process $(L_t^a(H), a \geq 0)$ is a CSBP starting from $x$ with branching mechanism $\psi$ and the tree with height process $H$ can be viewed as the genealogical tree of this CSBP. Let us remark that the latter property also holds for a discontinuous $H$ (i.e. if (7) doesn't hold) and we still say that $H$ describes the genealogy of the CSBP associated with $\psi$.

In general, the process $H$ is not a Markov process. So, we introduce the so-called exploration process $\rho = (\rho_t, t \geq 0)$ which is a measure-valued process defined by
\[
\rho_t(dr) = \beta 1_{[0,H_t]}(r) dr + \sum_{0 < s \leq t, X_s < I^s_{t-}} (I^s_{t-} - X_s) \delta_{H_t}(dr).
\]

The height process can easily be recovered from the exploration process as $H_t = H(\rho_t)$, where $H(\mu)$ denotes the supremum of the closed support of the measure $\mu$ (with the convention that $H(0) = 0$).

If we endow the set $\mathcal{M}_f(\mathbb{R}_+)$ of finite measures on $\mathbb{R}_+$ with the topology of weak convergence, then the exploration process $\rho$ is a càd-làg strong Markov process in $\mathcal{M}_f(\mathbb{R}_+)$ (see [22], Proposition 1.2.3).

To understand the meaning of the exploration process, let us use the queuing system representation of [30] when $\beta = 0$. We consider a preemptive LIFO (Last In, First Out) queue with one server. A jump of $X$ at time $s$ corresponds to the arrival of a new customer requiring a service equal to $\Delta_s := X_s - X_{s-}$. The server interrupts his current job and starts immediately the service of this new customer (preemptive LIFO procedure). When this new service is finished, the server will resume the previous job. When $\pi$ is infinite, all services will suffer interruptions. The customer (arrived at time $s$) will still be in the system at time $t > s$ if and only if $X_s < \inf_{s \leq r \leq t} X_r$ and, in this case, the quantity $\rho_t([H_t])$ represents the remaining service required by the customer $s$ at time $t$. Observe that $\rho_t([0,H_t])$ corresponds to the load of the server at time $t$ and is equal to $X_t - I_t$ where
\[
I_t = \inf\{X_u, 0 \leq u \leq t\}.
\]

In view of the Markov property of $\rho$ and the Poisson representation of Lemma 1.6, we can view $\rho_t$ as a measure placed on the ancestral line of the individual labeled by $t$ which gives the intensity
of the sub-trees that are grafted “on the right” of this ancestral line. The continuous part of the measure $\rho_t$ gives binary branching points (i.e. vertex in the tree of degree 3) which are dense along that ancestral line since the excursion measure $\mathcal{N}$ that appears in Lemma 1.6 is an infinite measure, whereas the atomic part of the measure $\rho_t$ gives nodes of infinite degree for the same reason.

Consequently, the nodes of the tree coded by $H$ are of two types: nodes of degree 3 and nodes of infinite degree. Moreover, we see that each node of infinite degree corresponds to a jump of the Lévy process $X$ and so we associate to such a node a “weight” given by the height of the corresponding jump of $X$ (this will be formally stated in Section 1.4). From now-on, we will only handle the exploration process although we will often use vocabulary taken from the real tree (coded by this exploration process). In particular, the theorems will be stated in terms of the exploration process and also hold when $H$ is not continuous.

0.2 The pruned exploration process

We now consider the Lévy CRT associated with a general critical or sub-critical branching mechanism $\psi$ (or rather the exploration process that codes that tree) and we add marks on the tree. There will be two kinds of marks: some marks will be set only on nodes of infinite degrees whereas the others will be 'uniformly distributed' on the skeleton on the tree.

0.2.1 Marks on the nodes

Let $p : [0, +\infty) \rightarrow [0, 1]$ be a measurable function satisfying condition (2). Recall that each node of infinite degree of the tree is associated with a jump $\Delta_s$ of the process $X$. Conditionally on $X$, we mark such a node with probability $p(\Delta_s)$, independently of the other nodes.

0.2.2 Marks on the skeleton

Let $\alpha_1$ be a non-negative constant. The marks associated with these parameters will be distributed on the skeleton of the tree according to a Poisson point measure with intensity $\alpha_1\lambda(dr)$ (recall that $\lambda$ denotes the one-dimensional Hausdorff measure on the tree).

0.2.3 The marked exploration process

As we don’t use the real trees framework but only the exploration processes that codes the Lévy CRTs, we must describe all these marks in term of exploration processes. Therefore, we define a measure-valued process

$$\mathcal{S} := (\rho_t, m_t^{nod}, m_t^{ske}, t \geq 0)$$

called the marked exploration process where the process $\rho$ is the usual exploration process whereas the processes $m^{nod}$ and $m^{ske}$ keep track of the marks, respectively on the nodes and on the skeleton of the tree.

The measure $m_t^{nod}$ is just the sum of the Dirac measure of the marked nodes (up to some weights for technical reasons) which are the ancestors of $t$.

To define the measure $m_t^{ske}$, we first consider a Lévy snake $(\rho_t, W_t)_{t \geq 0}$ with spatial motion $W$ a Poisson process of parameter $\alpha_1$ (see [22], Chapter 4 for the definition of a Lévy snake). We then
define the measure $m_t^{\text{ske}}$ as the derivative of the function $W_t$. Let us remark that in [22], the height process is supposed to be continuous for the construction of Lévy snakes. We explain in the appendix how to remove this technical assumption.

### 0.2.4 Main result

We denote by $A_t$ the Lebesgue measure of the set of the individuals prior to $t$ whose lineage does not contain any mark, i.e.,

$$A_t = \int_0^t \mathbf{1}_{\{m_{s \mod 2} = 0, m_s = 0\}} ds.$$ 

We consider its right-continuous inverse $C_t := \inf\{r \geq 0, A_r > t\}$ and we define the pruned exploration process $\tilde{\rho}$ by

$$\forall t \geq 0, \quad \tilde{\rho}_t = \rho_{C_t}.$$ 

In other words, we remove from the CRT all the individuals who have a marked ancestor, and the exploration process $\tilde{\rho}$ codes the remaining tree.

We can now restate Theorem 0.1 rigorously in terms of exploration processes.

**Theorem 0.3.** The pruned exploration process $\tilde{\rho}$ is distributed as the exploration process associated with a Lévy process with Laplace exponent $\psi_0$.

The proof relies on a martingale problem for $\tilde{\rho}$ and a special Markov property, Theorem 3.2. Roughly speaking, the special Markov property gives the conditional distribution of the individuals with marked ancestors with respect to the tree of individuals with no marked ancestors. This result is of independent interest. Notice the proof of this result in the general setting is surprisingly much more involved than the previous two particular cases: the quadratic case (see Proposition 6 in [7] or Proposition 7 in [16]) and the case without quadratic term (see Theorem 3.12 in [2]).

Finally, we give the joint law of the length of the excursion of the exploration process and the length of the excursion of the pruned exploration process, see Proposition 5.1.

### 0.3 Motivations and applications

A first approach for this construction is to consider the CSBP $Y^0$ associated with the pruned exploration process $\tilde{\rho}$ as an initial Eve-population which undergoes some neutral mutations (the marks on the genealogical tree) and the CSBP $Y$ denotes the total population (the Eve-one and the mutants) associated with the exploration process $\rho$. We see that, from our construction, we have

$$Y^0_0 = Y_0, \quad \text{and} \quad \forall t \geq 0, \quad Y^0_t \leq Y_t.$$ 

The condition

$$d\pi_0(x) = (1 - p(x)) d\pi(x)$$

means that, when the population $Y^0$ jumps, so does the population $Y$. By these remarks, we can see that our pruning procedure is quite general. Let us however remark that the coefficient diffusion $\beta$ is the same for $\psi$ and $\psi_0$ which might imply that more general prunings exist (in particular, we would like to remove some of the vertices of index 3).
As we consider general critical or sub-critical branching mechanism, this work extends previous work from Abraham and Serlet \cite{7} on Brownian CRT ($\pi = 0$) and Abraham and Delmas \cite{2} on CRT without Brownian part ($\beta = 0$). See also Bertoin \cite{14} for an approach using Galton-Watson trees and $p = 0$, or \cite{4} for an approach using CSBP with immigration. Let us remark that this paper goes along the same general ideas as \cite{2} (the theorems and the intermediate lemmas are the same) but the proofs of each of them are more involved and use quite different techniques based on martingale problem.

This work has also other applications. Our method separates in fact the genealogical tree associated with $Y$ into several components. For some values of the parameters of the pruning procedure, we can construct via our pruning procedure, a fragmentation process as defined by Bertoin \cite{13} but which is not self-similar, see for instance \cite{7,2,31}. On the other hand, we can view our method as a manner to increase the size of a tree, starting from the CRT associated with $\psi_0$ to get the CRT associated with $\psi$. We can even construct a tree-valued process which makes the tree grow, starting from a trivial tree containing only the root up to infinite super-critical trees, see \cite{5}.

0.4 Organization of the paper

We first recall in the next Section the construction of the exploration process, how it codes a CRT and its main properties we shall use. We also define the marked exploration process that is used for pruning the tree. In Section 2 we define rigorously the pruned exploration process $\tilde{\rho}$ and restate precisely Theorem 0.3. The rest of the paper is devoted to the proof of that theorem. In Section 3 we state and prove a special Markov property of the marked exploration process, that gives the law of the exploration process “above” the marks, conditionally on $\tilde{\rho}$. We use this special property in Section 4 to derive from the martingale problem satisfied by $\rho$, introduced in \cite{1} when $\beta = 0$, a martingale problem for $\tilde{\rho}$ which allows us to obtain the law of $\tilde{\rho}$. Finally, we compute in the last section, under the excursion measure, the joint law of the lengths of the excursions of $\rho$ and $\tilde{\rho}$. The Appendix is devoted to some extension of the Lévy snake when the height process is not continuous.

1 The exploration process: notations and properties

We recall here the construction of the height process and the exploration process that codes a Lévy continuum random tree. These objects have been introduced in \cite{30,29} and developed later in \cite{22}. The results of this section are mainly extracted from \cite{22}, but for Section 1.4.

We denote by $\mathbb{R}_+$ the set of non-negative real numbers. Let $\mathcal{M}(\mathbb{R}_+)$ (resp. $\mathcal{M}_f(\mathbb{R}_+)$) be the set of $\sigma$-finite (resp. finite) measures on $\mathbb{R}_+$, endowed with the topology of vague (resp. weak) convergence. If $E$ is a Polish space, let $\mathcal{B}(E)$ (resp. $\mathcal{B}_+(E)$) be the set of real-valued measurable (resp. and non-negative) functions defined on $E$ endowed with its Borel $\sigma$-field. For any measure $\mu \in \mathcal{M}(\mathbb{R}_+)$ and $f \in \mathcal{B}_+(\mathbb{R}_+)$, we write

$$\langle \mu, f \rangle = \int f(x) \mu(dx).$$
1.1 The underlying Lévy process

We consider a \( \mathbb{R} \)-valued Lévy process \( X = (X_t, t \geq 0) \) starting from 0. We assume that \( X \) is the canonical process on the Skorohod space \( \mathbb{D}(\mathbb{R}_+, \mathbb{R}) \) of càdlâg real-valued paths, endowed with the canonical filtration. The law of the process \( X \) starting from 0 will be denoted by \( \mathbb{P} \) and the corresponding expectation by \( \mathbb{E} \). Most of the following facts on Lévy processes can be found in [12].

In this paper, we assume that \( X \)

- has no negative jumps,
- has first moments,
- is of infinite variation,
- does not drift to \(+\infty\).

The law of \( X \) is characterized by its Laplace transform:

\[
\forall \lambda \geq 0, \quad \mathbb{E}\left[ e^{-\lambda X_t} \right] = e^{t\psi(\lambda)}
\]

where, as \( X \) does not drift to \(+\infty\), its Laplace exponent \( \psi \) can then be written as (1), where the Lévy measure \( \pi \) is a Radon measure on \( \mathbb{R}_+ \) (positive jumps) that satisfies the integrability condition

\[
\int_{(0, +\infty)} (\ell \wedge \ell^2) \pi(d\ell) < +\infty
\]

(\( X \) has first moments), the drift coefficient \( \alpha \) is non negative (\( X \) does not drift to \(+\infty\)) and \( \beta \geq 0 \). As we ask for \( X \) to be of infinite variation, we must additionally suppose that \( \beta > 0 \) or \( \int_{(0,1)} \ell \pi(d\ell) = +\infty \).

As \( X \) is of infinite variation, we have, see Corollary VII.5 in [12],

\[
\lim_{\lambda \to +\infty} \frac{\lambda}{\psi(\lambda)} = 0. \tag{9}
\]

Let \( I = (I_t, t \geq 0) \) be the infimum process of \( X \), \( I_t = \inf_{0 \leq s \leq t} X_s \), and let \( S = (S_t, t \geq 0) \) be the supremum process, \( S_t = \sup_{0 \leq s \leq t} X_s \). We will also consider for every \( 0 \leq s \leq t \) the infimum of \( X \) over \([s, t] \):

\[
I_s^t = \inf_{s \leq r \leq t} X_r.
\]

We denote by \( \mathcal{J} \) the set of jumping times of \( X \):

\[
\mathcal{J} = \{ t \geq 0, X_t > X_{t_-} \} \tag{10}
\]

and for \( t \geq 0 \) we set \( \Delta_t := X_t - X_{t_-} \) the height of the jump of \( X \) at time \( t \). Of course, \( \Delta_t > 0 \iff t \in \mathcal{J} \).

The point 0 is regular for the Markov process \( X - I \), and \( -I \) is the local time of \( X - I \) at 0 (see [12], chap. VII). Let \( \mathbb{N} \) be the associated excursion measure of the process \( X - I \) away from 0, and let
\[ \sigma = \inf\{ t > 0; X_t - I_t = 0 \} \]

be the length of the excursion of \( X - I \) under \( \mathbb{N} \). We will assume that under \( \mathbb{N} \), \( X_0 = I_0 = 0 \).

Since \( X \) is of infinite variation, 0 is also regular for the Markov process \( S - X \). The local time \( L = (L_t, t \geq 0) \) of \( S - X \) at 0 will be normalized so that

\[ \mathbb{E}[e^{-\int_t^{\sigma}}] = e^{-t\psi(\lambda)/\lambda}, \]

where \( L_t^{-1} = \inf\{ s \geq 0; L_s \geq t \} \) (see also [12] Theorem VII.4 (ii)).

### 1.2 The height process

We now define the height process \( H \) associated with the Lévy process \( X \). Following [22], we give an alternative definition of \( H \) instead of those in the introduction, formula (6).

For each \( t \geq 0 \), we consider the reversed process at time \( t \), \( \hat{X}(t) = (\hat{X}_s(t), 0 \leq s \leq t) \) by:

\[ \hat{X}_s(t) = X_t - X(t-s) - \] if \( 0 \leq s < t, \]

with the convention \( X_0 = X_0 \). The two processes \( (X_s(t), 0 \leq s \leq t) \) and \( (X_s, 0 \leq s \leq t) \) have the same law. Let \( \tilde{H}(t) \) be the supremum process of \( \hat{X}(t) \) and \( \tilde{L}(t) \) be the local time at 0 of \( \tilde{H}(t) - \tilde{X}(t) \) with the same normalization as \( L \).

**Definition 1.1.** ([22], Definition 1.2.1)

There exists a lower semi-continuous modification of the process \( (L(t), t \geq 0) \). We denote by \( (H(t), t \geq 0) \) this modification.

This definition gives also a modification of the process defined by (6) (see [22], Lemma 1.1.3). In general, \( H \) takes its values in \([0, +\infty]\), but we have, a.s. for every \( t \geq 0 \), \( H_s < +\infty \) for every \( s < t \) such that \( X_s \leq L^t_s \), and \( H_t < +\infty \) if \( \Delta_t > 0 \) (see [22], Lemma 1.2.1). The process \( H \) does not admit a continuous version (or even càd-làg) in general but it has continuous sample paths \( \mathbb{P} \)-a.s. iff (7) is satisfied, see [22], Theorem 1.4.3.

To end this section, let us remark that the height process is also well-defined under the excursion process \( \mathbb{N} \) and all the previous results remain valid under \( \mathbb{N} \).

### 1.3 The exploration process

The height process is not Markov in general. But it is a very simple function of a measure-valued Markov process, the so-called exploration process.

The exploration process \( \rho = (\rho_t, t \geq 0) \) is a \( \mathcal{M}_f(\mathbb{R}_+) \)-valued process defined as follows: for every \( f \in \mathcal{B}_+(\mathbb{R}_+) \), \( \langle \rho_t, f \rangle = \int_{[0,t]} d\lambda t_s f(H_s) \), or equivalently

\[ \rho_t(dr) = \beta 1_{[0,H_t]}(r) dr + \sum_{0 < s \leq t} (I^t_s - X_{t-s}) \delta_{H_t}(dr). \]

In particular, the total mass of \( \rho_t \) is \( \langle \rho_t, 1 \rangle = X_t - I_t \).

For \( \mu \in \mathcal{M}(\mathbb{R}_+) \), we set

\[ H(\mu) = \sup \text{Supp} \mu, \]

where \( \text{Supp} \mu \) is the closed support of \( \mu \), with the convention \( H(0) = 0 \). We have
Proposition 1.2. ([22], Lemma 1.2.2 and Formula (1.12))
Almost surely, for every \( t > 0 \),

- \( H(\rho_t) = H_t \),
- \( \rho_t = 0 \) if and only if \( H_t = 0 \),
- if \( \rho_t \neq 0 \), then \( \text{Supp} \rho_t = [0, H_t] \).
- \( \rho_t = \rho_t^- + \Delta_t \delta_{H_t} \), where \( \Delta_t = 0 \) if \( t \notin J \).

In the definition of the exploration process, as \( X \) starts from 0, we have \( \rho_0 = 0 \) a.s. To state the Markov property of \( \rho \), we must first define the process \( \rho \) started at any initial measure \( \mu \in M_f(\mathbb{R}_+) \).

For \( a \in [0, (\mu, 1)] \), we define the erased measure \( k_a \mu \) by

\[
    k_a \mu([0, r]) = \mu([0, r]) \land ((\mu, 1) - a), \quad \text{for } r \geq 0. \tag{13}
\]

If \( a > (\mu, 1) \), we set \( k_a \mu = 0 \). In other words, the measure \( k_a \mu \) is the measure \( \mu \) erased by a mass \( a \) backward from \( H(\mu) \).

For \( \nu, \mu \in M_f(\mathbb{R}_+) \), and \( \mu \) with compact support, we define the concatenation \([\mu, \nu] \in M_f(\mathbb{R}_+)\) of the two measures by:

\[
    \langle [\mu, \nu], f \rangle = \langle \mu, f \rangle + \langle \nu, f(H(\mu) + \cdot) \rangle, \quad f \in \mathcal{B}_+(\mathbb{R}_+). \tag{14}
\]

Finally, we set for every \( \mu \in M_f(\mathbb{R}_+) \) and every \( t > 0 \),

\[
    \rho_t^\mu = [k_{-I_t} \mu, \rho_t]. \tag{15}
\]

We say that \((\rho_t^\mu, t \geq 0)\) is the process \( \rho \) started at \( \rho_0^\mu = \mu \), and write \( \mathbb{P}_\mu \) for its law. Unless there is an ambiguity, we shall write \( \rho_t \) for \( \rho_t^\mu \).

Proposition 1.3. ([22], Proposition 1.2.3)
For any initial finite measure \( \mu \in M_f(\mathbb{R}_+) \), the process \((\rho_t^\mu, t \geq 0)\) is a càd-làg strong Markov process in \( M_f(\mathbb{R}_+) \).

Remark 1.4. From the construction of \( \rho \), we get that a.s. \( \rho_t = 0 \) if and only if \( -I_t \geq (\rho_0, 1) \) and \( X_t - I_t = 0 \). This implies that \( 0 \) is also a regular point for \( \rho \). Notice that \( \mathbb{N} \) is also the excursion measure of the process \( \rho \) away from 0, and that \( \sigma \), the length of the excursion, is \( \mathbb{N} \)-a.e. equal to \( \inf\{t > 0; \rho_t = 0\} \).

Exponential formula for the Poisson point process of jumps of the inverse subordinator of \(-I\) gives (see also the beginning of Section 3.2.2. [22]) that for \( \lambda > 0 \)

\[
    \mathbb{N} \left[ 1 - e^{-\lambda \sigma} \right] = \psi^{-1}(\lambda). \tag{15}
\]

1.4 The marked exploration process
As presented in the introduction, we add random marks on the Lévy CRT coded by \( \rho \). There will be two kinds of marks: marks on the nodes of infinite degree and marks on the skeleton.
1.4.1 Marks on the skeleton

Let $\alpha_1 \geq 0$. We want to construct a “Lévy Poisson snake” (i.e. a Lévy snake with spatial motion a Poisson process), whose jumps give the marks on the branches of the CR T. More precisely, we set $\mathcal{W}$ the space of killed càd-làg paths $w : [0, \zeta) \to \mathbb{R}$ where $\zeta \in (0, +\infty)$ is called the lifetime of the path $w$. We equip $\mathcal{W}$ with a distance $d$ (defined in [22] Chapter 4 and whose expression is not important for our purpose) such that $(\mathcal{W}, d)$ is a Polish space.

By Proposition 4.4.1 of [22] when $H$ is continuous, or the results of the appendix in the general case, there exists a probability measure $\tilde{\mathbb{P}}$ on $\tilde{\Omega} = \mathbb{D}(\mathbb{R}_+, \mathcal{M}_f(\mathbb{R}_+) \times \mathcal{W})$ under which the canonical process $(\rho_s, W_s)$ satisfies

1. The process $\rho$ is the exploration process starting at 0 associated with a branching mechanism $\psi$,

2. For every $s \geq 0$, the path $W_s$ is distributed as a Poisson process with intensity $\alpha_1$ stopped at time $H_s := H(\rho_s)$,

3. The process $(\rho, W)$ satisfies the so-called snake property: for every $s < s'$, conditionally given $\rho$, the paths $W_s(\cdot)$ and $W_{s'}(\cdot)$ coincide up to time $H_{s,s'} := \inf\{H_u, s \leq u \leq s'\}$ and then are independent.

So, for every $t \geq 0$, the path $W_t$ is a.s. càd-làg with jumps equal to one. Its derivative $m_{\text{ske}}$ is an atomic measure on $[0, H_t)$; it gives the marks (on the skeleton) on the ancestral line of the individual labeled $t$.

We shall denote by $\tilde{N}$ the corresponding excursion measure out of $(0, 0)$.

1.4.2 Marks on the nodes

Let $p$ be a measurable function defined on $\mathbb{R}_+$ taking values in $[0, 1]$ such that

$$\int_{(0, +\infty)} \ell p(\ell) \pi(d\ell) < \infty. \quad (16)$$

We define the measures $\pi_1$ and $\pi_0$ by their density:

$$d\pi_1(x) = p(x)d\pi(x) \quad \text{and} \quad d\pi_0(x) = (1 - p(x))d\pi(x).$$

Let $(\Omega', \mathcal{A}', \mathbb{P}')$ be a probability space with no atom. Recall that $\mathcal{F}$, defined by (10), denotes the jumping times of the Lévy process $X$ and that $\Delta_s$ represents the height of the jump of $X$ at time $s \in \mathcal{F}$. As $\mathcal{F}$ is countable, we can construct on the product space $\tilde{\Omega} \times \Omega'$ (with the product probability measure $\tilde{\mathbb{P}} \otimes \mathbb{P}'$) a family $(U_s, s \in \mathcal{F})$ of random variables which are, conditionally on $X$, independent, uniformly distributed over $[0, 1]$ and independent of $(\Delta_s, s \in \mathcal{F})$ and $(W_s, s \geq 0)$. We set, for every $s \in \mathcal{F}$:

$$V_s = 1_{\{U_s \leq p(\Delta_s)\}},$$

so that, conditionally on $X$, the family $(V_s, s \in \mathcal{F})$ are independent Bernoulli random variables with respective parameters $p(\Delta_s)$.  

1439
We set $\mathcal{J}^1 = \{ s \in \mathcal{J}, V_s = 1 \}$ the set of the marked jumps and $\mathcal{J}^0 = \mathcal{J} \setminus \mathcal{J}^1 = \{ s \in \mathcal{J}, V_s = 0 \}$ the set of the non-marked jumps. For $t \geq 0$, we consider the measure on $\mathbb{R}_+$, 
\[ m_t^{\text{nod}}(dr) = \sum_{0 < s \leq t; s \in \mathcal{J}^1} (I^s_t - X_{s-}) \delta_{H_1}(dr). \] (17)

The atoms of $m_t^{\text{nod}}$ give the marked nodes of the exploration process at time $t$.

The definition of the measure-valued process $m^{\text{nod}}$ also holds under $\bar{N} \otimes P'$. For convenience, we shall write $P$ for $\bar{P} \otimes P'$ and $N$ for $\bar{N} \otimes P'$.

1.4.3 Decomposition of $X$

At this stage, we can introduce a decomposition of the process $X$. Thanks to the integrability condition (16) on $p$, we can define the process $X^{(1)}$ by, for every $t \geq 0$,
\[ X_t^{(1)} = \alpha_1 t + \sum_{0 < s \leq t; s \in \mathcal{J}^1} \Delta_s. \]

The process $X^{(1)}$ is a subordinator with Laplace exponent $\phi_1$ given by:
\[ \phi_1(\lambda) = \alpha_1 \lambda + \int_{(0, +\infty)} \pi_1(d\ell) \left( 1 - e^{-\lambda \ell} \right), \] (18)

with $\pi_1(dx) = p(x)\pi(dx)$. We then set $X^{(0)} = X - X^{(1)}$ which is a Lévy process with Laplace exponent $\psi_0$, independent of the process $X^{(1)}$ by standard properties of Poisson point processes.

We assume that $\phi_1 \neq 0$ so that $\alpha_0$ defined by (14) is such that:
\[ \alpha_0 > 0. \] (19)

It is easy to check, using $\int_{(0, \infty)} \pi_1(d\ell) \ell < \infty$, that
\[ \lim_{\lambda \to \infty} \frac{\phi_1(\lambda)}{\lambda} = \alpha_1. \] (20)

1.4.4 The marked exploration process

We consider the process 
\[ \mathcal{S} = (\rho_t, m_t^{\text{nod}}, m_t^{\text{ske}}, t \geq 0) \]
on the product probability space $\bar{\Omega} \times \Omega'$ under the probability $P$ and call it the marked exploration process. Let us remark that, as the process is defined under the probability $P$, we have $\rho_0 = 0$, $m_0^{\text{nod}} = 0$ and $m_0^{\text{ske}} = 0$ a.s.

Let us first define the state-space of the marked exploration process. We consider the set $\mathcal{S}$ of triplet $(\mu, \Pi_1, \Pi_2)$ where
- $\mu$ is a finite measure on $\mathbb{R}_+$,
• $\Pi_1$ is a finite measure on $\mathbb{R}_+$ absolutely continuous with respect to $\mu$,

• $\Pi_2$ is a $\sigma$-finite measure on $\mathbb{R}_+$ such that
  
  - $\text{Supp}(\Pi_2) \subset \text{Supp}(\mu)$,
  - for every $x < H(\mu)$, $\Pi_2([0,x]) < +\infty$,
  - if $\mu([H(\mu)]) > 0$, $\Pi_2(\mathbb{R}_+) < +\infty$.

We endow $\mathbb{S}$ with the following distance: If $(\mu, \Pi_1, \Pi_2) \in \mathbb{S}$, we set

$$w(t) = \int_{]0,t]}(t)\Pi_2(d\ell)$$

and

$$\tilde{w}(t) = w(H(k_{(\mu,1)-t}(\mu))) \quad \text{for} \quad t \in [0,\langle \mu, 1 \rangle).$$

We then define

$$d'((\mu, \Pi_1, \Pi_2), (\mu', \Pi_1', \Pi_2')) = d((\mu, \tilde{w}), (\mu', \tilde{w}')) + D(\Pi_1, \Pi_1')$$

where $d$ is the distance defined by (52) and $D$ is a distance that defines the topology of weak convergence and such that the metric space $(\mathcal{M}_f(\mathbb{R}_+), D)$ is complete.

To get the Markov property of the marked exploration process, we must define the process $\mathcal{S}$ started at any initial value of $\mathbb{S}$. For $(\mu, \Pi^{\text{nod}}, \Pi^{\text{ske}}) \in \mathbb{S}$, we set $\Pi = (\Pi^{\text{nod}}, \Pi^{\text{ske}})$ and $H_{t}^\mu = H(k_{-1}, \mu)$. We define

$$(m_{\text{nod}})^{(\mu, \Pi)}_t = \left[ \Pi^{\text{nod}} 1_{]0,H_t^\mu]} + 1_{\mu(H_t^\mu)>0} \frac{k_{-1}(H_{t}^\mu)\Pi^{\text{nod}}(H_{t}^\mu)}{\mu(H_t^\mu)} \delta_{H_t^\mu}, m_{\text{nod}}^t \right]$$

and

$$(m_{\text{ske}})^{(\mu, \Pi)}_t = [\Pi^{\text{ske}} 1_{]0,H_t^\mu]}, m_{\text{ske}}^t].$$

Notice the definition of $(m_{\text{ske}})^{(\mu, \Pi)}_t$ is coherent with the construction of the Lévy snake, with $W_0$ being the cumulative function of $\Pi^{\text{ske}}$ over $[0,H_0]$.

We shall write $m_{\text{nod}}$ for $(m_{\text{nod}})^{(\mu, \Pi)}_t$ and similarly for $m_{\text{ske}}$. Finally, we write $m = (m_{\text{nod}}, m_{\text{ske}})$.

By construction and since $\rho$ is an homogeneous Markov process, the marked exploration process $\mathcal{S} = (\rho, m)$ is an homogeneous Markov process.

From now on, we suppose that the marked exploration process is defined on the canonical space $(\mathbb{S}, \mathcal{F})$ where $\mathcal{F}$ is the Borel $\sigma$-field associated with the metric $d'$. We denote by $\mathcal{S} = (\rho, m_{\text{nod}}, m_{\text{ske}})$ the canonical process and we denote by $\mathbb{P}_{\mu, \Pi}$ the probability measure under which the canonical process is distributed as the marked exploration process starting at time 0 from $(\mu, \Pi)$, and by $\mathbb{P}_{\mu, \Pi}$ the probability measure under which the canonical process is distributed as the marked exploration process killed when $\rho$ reaches 0. For convenience we shall write $\mathbb{P}_\mu$ if $\Pi = 0$ and $\mathbb{P}$ if $(\mu, \Pi) = 0$ and similarly for $\mathbb{P}_{\mu, \Pi}$. Finally, we still denote by $\mathbb{N}$ the distribution of $\mathcal{S}$ when $\rho$ is distributed under the excursion measure $\mathbb{N}$.

Let $\mathcal{F} = (\mathcal{F}_t, t \geq 0)$ be the canonical filtration. Using the strong Markov property of $(X, X^{(1)})$ and Proposition 6.2 or Theorem 4.1.2 in [22] if $H$ is continuous, we get the following result.

**Proposition 1.5.** The marked exploration process $\mathcal{S}$ is a càdlàg $\mathbb{S}$-valued strong Markov process.

Let us remark that the marked exploration process satisfies the following snake property:

$$\mathbb{P} - \text{a.s.} \quad \text{(or $\mathbb{N} - \text{a.e.}$,} \quad (\rho_t, m_t)(\cdot \cap [0,s]) = (\rho_{t'}, m_{t'})(\cdot \cap [0,s]) \quad \text{for every} \quad 0 \leq s < H_{t,t'}. \quad (21)$$
1.5 Poisson representation

We decompose the path of $\mathcal{S}$ under $\mathbb{P}^\mu_{h,\Pi}$ according to excursions of the total mass of $\rho$ above its past minimum, see Section 4.2.3 in [22]. More precisely, let $(a_i, b_i), i \in \mathcal{X}$ be the excursion intervals of $X - I$ above 0 under $\mathbb{P}^\mu_{h,\Pi}$. For every $i \in \mathcal{X}$, we define $h_i = H_{a_i}$ and $\mathcal{S}^i = (\tilde{\rho}^i, \tilde{m}^i)$ by the formulas: for $t \geq 0$ and $f \in \mathcal{B}_+(\mathbb{R}_+)$,

$$\langle \tilde{\rho}^i_t, f \rangle = \int_{(h_i, +\infty)} f(x - h_i)\rho_{(a_i+t)\wedge b_i}(dx)$$

$$\langle \tilde{m}^i_t, f \rangle = \int_{(h_i, +\infty)} f(x - h_i)m_{(a_i+t)\wedge b_i}(dx), \quad a \in \{\text{nond}, \text{ske}\},$$

with $\tilde{m}^i = ((\tilde{m}^\text{nond},i), (\tilde{m}^\text{ske},i))$. We set $\tilde{\sigma}^i = \inf\{s > 0; \langle \rho^i_s, 1 \rangle = 0\}$. It is easy to adapt Lemma 4.2.4. of [22] to get the following Lemma.

**Lemma 1.6.** Let $(\mu, \Pi) \in \mathbb{S}$. The point measure $\sum_{i \in \mathcal{X}} \delta_{(h_i, \mathcal{S}^i)}$ is under $\mathbb{P}^\mu_{h,\Pi}$ a Poisson point measure with intensity $\mu(dr)\mathbb{N}[d\mathcal{S}]$.

1.6 The dual process and representation formula

We shall need the $\mathcal{M}_f(\mathbb{R}_+)$-valued process $\eta = (\eta_t, t \geq 0)$ defined by

$$\eta_t(dr) = \beta 1_{[0,h_t]}(r)dr + \sum_{0 < s \leq t, X_s < I_t^i} (X_s - I_t^i)\delta_{H_t}(dr).$$

The process $\eta$ is the dual process of $\rho$ under $\mathbb{N}$ (see Corollary 3.1.6 in [22]).

The next Lemma on time reversibility can easily be deduced from Corollary 3.1.6 of [22] and the construction of $m$.

**Lemma 1.7.** Under $\mathbb{N}$, the processes $((\rho^i, \eta^i, 1_{\{h^i = 0\}}), s \in [0,\sigma))$ and $((\eta_s(\sigma-s)\to, \rho(\sigma-s)\to, 1_{\{h^s(\sigma-s)\to = 0\}}), s \in [0,\sigma))$ have the same distribution.

We present a Poisson representation of $(\rho, \eta, m)$ under $\mathbb{N}$. Let $\mathcal{N}_0(dx d\ell du)$, $\mathcal{N}_1(dx d\ell du)$ and $\mathcal{N}_2(dx)$ be independent Poisson point measures respectively on $[0, +\infty)^3$, $[0, +\infty)^3$ and $[0, +\infty)$ with respective intensity

$$dx \ell \pi_0(\ell)(1_{[0,1]}(u)du, \quad dx \ell \pi_1(\ell)(1_{[0,1]}(u)du \quad \text{and} \quad \alpha_1 dx.$$ 

For every $a > 0$, let us denote by $\mathcal{M}_a$ the law of the pair $(\mu, \nu, m^\text{nond}, m^\text{ske})$ of measures on $\mathbb{R}_+$ with finite mass defined by: for any $f \in \mathcal{B}_+(\mathbb{R}_+)$

$$\langle \mu, f \rangle = \int (\mathcal{N}_0(dx d\ell du) + \mathcal{N}_1(dx d\ell du)) 1_{[0,a]}(x)uf(x) + \beta \int_0^a f(r)dr,$$

$$\langle \nu, f \rangle = \int (\mathcal{N}_0(dx d\ell du) + \mathcal{N}_1(dx d\ell du)) 1_{[0,a]}(x)(1 - u)\ell f(x) + \beta \int_0^a f(r)dr,$$

$$\langle m^\text{nond}, f \rangle = \int \mathcal{N}_1(dx d\ell du) 1_{[0,a]}(x)uf(x) \quad \text{and} \quad \langle m^\text{ske}, f \rangle = \int \mathcal{N}_2(dx) 1_{[0,a]}(x)f(x).$$

1442
Remark 1.8. In particular \( \mu(dr) + \nu(dr) \) is defined as \( 1_{[0,a]}(r)d\xi_r \), where \( \xi \) is a subordinator with Laplace exponent \( \psi' - \alpha \).

We finally set \( M = \int_{0}^{+\infty} da \, e^{-\alpha a} M_{\alpha} \). Using the construction of the snake, it is easy to deduce from Proposition 3.1.3 in [22], the following Poisson representation.

**Proposition 1.9.** For every non-negative measurable function \( F \) on \( \mathcal{M}_f(\mathbb{R}_+)^4 \),

\[
\mathbb{N} \left[ \int_{0}^{\sigma} F(\rho_t, \eta_t, m_t) \, dt \right] = \int \mathbb{M}(dm \, dv \, dm) F(\mu, \nu, m),
\]

where \( m = (m^{\text{nond}}, m^{\text{ske}}) \) and \( \sigma = \inf\{s > 0; \rho_s = 0\} \) denotes the length of the excursion.

## 2 The pruned exploration process

We define the following continuous additive functional of the process \( ((\rho_t, m_t), t \geq 0) \):

\[
A_t = \int_{0}^{t} 1_{[m_s = 0]} \, ds \quad \text{for} \quad t \geq 0.
\]  

**Lemma 2.1.** We have the following properties.

(i) For \( \lambda > 0 \), \( \mathbb{N}[1 - e^{-\lambda A_{\sigma}}] = \psi_{0}^{-1}(\lambda) \).

(ii) \( \mathbb{N}\)-a.e. 0 and \( \sigma \) are points of increase for \( A \). More precisely, \( \mathbb{N}\)-a.e. for all \( \varepsilon > 0 \), we have \( A_{\sigma} > 0 \) and \( A_{\sigma} - A_{(\sigma - \varepsilon)\wedge 0} > 0 \).

(iii) If \( \lim_{\lambda \to +\infty} \phi_{1}(\lambda) = +\infty \), then \( \mathbb{N}\)-a.e. the set \( \{s; m_s \neq 0\} \) is dense in \([0, \sigma]\).

**Proof.** We first prove (i). Let \( \lambda > 0 \). Before computing \( \nu = \mathbb{N}[1 - \exp -\lambda A_{\sigma}] \), notice that \( A_{\sigma} \leq \sigma \) implies, thanks to (15), that \( \nu \leq \mathbb{N}[1 - \exp -\lambda \sigma] = \psi_{1}^{-1}(\lambda) < +\infty \). We have

\[
\nu = \lambda \mathbb{N} \left[ \int_{0}^{\sigma} dA_t \, e^{-\lambda \int_{0}^{\sigma} dA_u} \right] = \lambda \mathbb{N} \left[ \int_{0}^{\sigma} dA_t \, \mathbb{E}_{(\rho_t, \gamma_t)}^{+}[e^{-\lambda A_{\sigma}}] \right],
\]

where we replaced \( e^{-\lambda \int_{0}^{\sigma} dA_u} \) in the last equality by \( \mathbb{E}_{(\rho_t, \gamma_t)}^{+}[e^{-\lambda A_{\sigma}}] \), its optional projection, and used that \( dA_t\)-a.e. \( m_t = 0 \). In order to compute this last expression, we use the decomposition of \( \mathcal{S} \) under \( P_{\mu}^{+} \) according to excursions of the total mass of \( \rho \) above its minimum, see Lemma [16]. Using the same notations as in this lemma, notice that under \( P_{\mu}^{+} \), we have \( A_{\sigma} = A_{\infty} = \sum_{i \in \mathcal{X}} \tilde{A}_{i\infty} \), where for every \( T \geq 0 \)

\[
\tilde{A}_{i_{T}} = \int_{0}^{T} 1_{[m_{t} = 0]} \, dt.
\]  

By Lemma [16], we get

\[
\mathbb{E}_{\mu}^{+}[e^{-\lambda A_{\sigma}}] = e^{- (\mu, 1) \mathbb{N}[1 - \exp -\lambda A_{\sigma}]} = e^{-\nu(\mu, 1)}.
\]
We have
\[ v = \lambda N \left[ \int_0^\sigma dA_t \ e^{-\nu(\rho_t, 1)} \right] = \lambda N \left[ \int_0^\sigma dt \ 1_{\{m_t = 0\}} e^{-\nu(\rho_t, 1)} \right] \]
\[ = \lambda \int_0^{+\infty} da \ e^{-\alpha a} M_0[1_{\{m_0 = 0\}} e^{-\nu(\mu, 1)}] \]
\[ = \lambda \int_0^{+\infty} da \ e^{-\alpha a} \exp \left\{ -\alpha a - \int_0^a dx \int_0^1 du \int_{(0,\infty)} \ell_1 \pi_1(d\ell_1) \right\} \]
\[ \exp \left\{ -\beta v a - \int_0^a dx \int_0^1 du \int_{(0,\infty)} \ell_0 \pi_0(d\ell_0) \left( 1 - e^{-\nu t_0} \right) \right\} \]
\[ = \lambda \int_0^{+\infty} da \ exp \left\{ -a \int_0^1 du \psi'(u) \right\} \]
\[ = \lambda \frac{v}{\psi'(v)}. \] (31)

where we used Proposition 1.9 for the third and fourth equalities, and for the last equality that
\[ \alpha_0 = \alpha + \alpha_1 + \int_{(0,\infty)} \ell_1 \pi_1(d\ell_1) \] as well as
\[ \psi'(\lambda) = \alpha_0 + \int_{(0,\infty)} \pi_0(d\ell_0) \ell_0(1 - e^{-\lambda \ell_0}). \] (32)

Notice that if \( v = 0 \), then (30) implies \( v = \lambda / \psi'(0) \), which is absurd since \( \psi'(0) = \alpha_0 > 0 \) thanks to (19). Therefore we have \( v \in (0,\infty) \), and we can divide (31) by \( v \) to get \( \psi'(v) = \lambda \). This proves (i).

Now, we prove (ii). If we let \( \lambda \) goes to infinity in (i) and use that \( \lim_{r \to \infty} \psi_0(r) = +\infty \), then we get that \( N[A_\sigma > 0] = +\infty \). Notice that for \( (\mu, \Pi) \in \mathbb{S} \), we have under \( P_{\mu, \Pi}^* \), \( A_\infty = \sum_{i \in \mathcal{X}} \bar{A}^i \), with \( \bar{A}^i \) defined by (28). Thus Lemma 1.6 implies that if \( \mu \neq 0 \), then \( P_{\mu, \Pi}^* \)-a.e. \( \mathcal{X} \) is infinite and \( A_\infty > 0 \). Using the Markov property at time \( t \) of the snake under \( N \), we get that for any \( t > 0 \), \( N \)-a.e. on \( \{ \sigma > t \} \), we have \( A_\sigma - A_t > 0 \). This implies that \( \sigma \) is \( N \)-a.e. a point of increase of \( A \). By time reversibility, see Lemma 1.7 we also get that \( N \)-a.e. \( 0 \) is a point of increase of \( A \). This gives (ii).

If \( \alpha_1 > 0 \) then the snake \( ((\rho_s, W_s), s \geq 0) \) is non trivial. It is well known that, since the Lévy process \( X \) is of infinite variation, the set \( \{ s; \exists t \in [0, H_s), W_s(t) \neq 0 \} \) is \( N \)-a.e. dense in \( [0, \sigma] \). This implies that \( \{ s; m_s \neq 0 \} \) is \( N \)-a.e. dense in \( [0, \sigma] \).

If \( \alpha_1 = 0 \) and \( \pi_1((0,\infty)) = 0 \), then the set \( \mathcal{J}^1 \) of jumping time of \( X \) is \( N \)-a.e. dense in \( [0, \sigma] \). This also implies that \( \{ s; m_s \neq 0 \} \) is \( N \)-a.e. dense in \( [0, \sigma] \).

If \( \alpha_1 = 0 \) and \( \pi_1((0,\infty)) < 0 \), then the set \( \mathcal{J}^1 \) of jumping time of \( X \) is \( N \)-a.e. finite. This implies that \( \{ s; m_s \neq 0 \} \cap [0, \sigma] \) is \( N \)-a.e. a finite union of intervals, which, thanks to (ii), is not dense in \( [0, \sigma] \).

To get (iii), notice that \( \lim_{\lambda \to \infty} \phi_1(\lambda) = \infty \) if and only if \( \alpha_1 > 0 \) or \( \pi_1((0,\infty)) = 0 \). \( \square \)

We set \( C_t = \inf \{ r > 0; A_r > t \} \) the right continuous inverse of \( A \), with the convention that \( \inf \emptyset = +\infty \).

From excursion decomposition, see Lemma 1.6 (ii) of Lemma 2.1 implies the following Corollary.

**Corollary 2.2.** For any initial measures \( (\mu, \Pi) \in \mathbb{S} \), \( P_{\mu, \Pi} \)-a.s. the process \( (C_t, t \geq 0) \) is finite. If \( m_0 = 0 \), then \( P_{\mu, \Pi} \)-a.s. \( C_0 = 0 \).
We define the pruned exploration process \( \bar{\rho} = (\bar{\rho}_t = \rho_{C_t}, t \geq 0) \) and the pruned marked exploration process \( \bar{\mathcal{F}} = (\bar{\rho}, \bar{m}) \), where \( \bar{m} = (m_{C_t}, t \geq 0) = 0 \). Notice \( C_t \) is a \( \mathcal{F} \)-stopping time for any \( t \geq 0 \) and is finite a.s. from Corollary 2.2. Notice the process \( \bar{\rho} \), and thus the process \( \bar{\mathcal{F}} \), is càdlàg. We also set \( \bar{H}_t = H_{C_t} \) and \( \bar{s} = \inf\{t > 0; \bar{\rho}_t = 0\} \).

Let \( \mathcal{F} = (\mathcal{F}_t, t \geq 0) \) be the filtration generated by the pruned marked exploration process \( \bar{\mathcal{F}} \) completed the usual way. In particular \( \mathcal{F}_t \subset \mathcal{F}_{C_t} \), where if \( \tau \) is an \( \mathcal{F} \)-stopping time, then \( \mathcal{F}_\tau \) is the \( \sigma \)-field associated with \( \tau \).

We are now able to restate precisely Theorem 0.3. Let \( \rho^{(0)} \) be the exploration process of a Lévy process with Laplace exponent \( \psi_0 \).

**Theorem 2.3.** For every finite measure \( \mu \), the law of the pruned process \( \bar{\rho} \) under \( \mathbb{P}_{\mu,0} \) is the law of the exploration process \( \rho^{(0)} \) associated with a Lévy process with Laplace exponent \( \psi_0 \) under \( \mathbb{P}_{\mu} \).

The proof of this Theorem is given at the end of Section 4.

### 3 A special Markov property

In order to define the excursions of the marked exploration process away from \( \{s \geq 0; m_s = 0\} \), we define \( O \) as the interior of \( \{s \geq 0; m_s \neq 0\} \). We shall see that the complementary of \( O \) has positive Lebesgue measure.

**Lemma 3.1.** (i) If the set \( \{s \geq 0; m_s \neq 0\} \) is non empty then, \( \mathbb{N} \)-a.e. \( O \) is non empty.

(ii) If we have \( \lim_{\lambda \to \infty} \phi_1(\lambda) = \infty \), then \( \mathbb{N} \)-a.e. the open set \( O \) is dense in \([0, \sigma] \).

**Proof.** For any element \( s' \) in \( \{s \geq 0; m_s \neq 0\} \), there exists \( u \leq H_{s'} \) such that \( m_s([0,u]) \neq 0 \) and \( \rho_s([u, \infty)) > 0 \). Then we consider \( \tau_{s'} = \inf\{t > s', \rho_t([u, \infty)) = 0\} \). By the right continuity of \( \rho \), we have \( \tau_{s'} > s' \) and the snake property (21) implies that \( \mathbb{N} \) a.e. \( (s', \tau_{s'}) \subset O \).

Use (iii) of Lemma 2.1 to get the last part. \( \square \)

We write \( O = \bigcup_{i \in \mathcal{I}} (\alpha_i, \beta_i) \) and say that \( (\alpha_i, \beta_i)_{i \in \mathcal{I}} \) are the excursions intervals of the marked exploration process \( \mathcal{F} = (\rho, m) \) away from \( \{s \geq 0; m_s = 0\} \). For every \( i \in \mathcal{I} \), let us define the measure-valued process \( \mathcal{F}^i = (\rho^i, m^i) \). For every \( f \in \mathcal{B}_1(\mathbb{R}_+) \), \( t \geq 0 \), we set

\[
\langle \rho^i_t, f \rangle = \int_{[H_{\alpha_i},+\infty)} f(x - H_{\alpha_i}) \rho_{(\alpha_i + t)\wedge \beta_i}(dx)
\]

\[
\langle (m^a)^i_t, f \rangle = \int_{(H_{\alpha_i},+\infty)} f(x - H_{\alpha_i}) m^a_{(\alpha_i + t)\wedge \beta_i}(dx) \quad \text{with } a \in \{\text{nod, ske}\}
\]

and \( m^i_t = ((m^{\text{nod}})^i_t, (m^{\text{ske}})^i_t) \). Notice that the mass located at \( H_{\alpha_i} \) is kept, if there is any, in the definition of \( \rho^i \) whereas it is removed in the definition of \( m^i \). In particular, if \( \Delta_{\alpha_i} > 0 \), then \( \rho^i_0 = \Delta_{\alpha_i} \delta_0 \) and for every \( t < \beta_i - \alpha_i \), the measure \( \rho^i_t \) charges 0. On the contrary, as \( m^i_0 = 0 \), we have, for every \( t < \beta_i - \alpha_i \), \( m^i_t(\{0\}) = 0 \).
Let $\mathcal{F}_\infty$ be the $\sigma$-field generated by $\mathcal{F} = ((\rho_{C_t}, m_{C_t}), t \geq 0)$. Recall that $\mathbb{P}_\mu^\ast(d\mathcal{F})$ denotes the law of the marked exploration process $\mathcal{F}$ started at $(\mu, \Pi) \in \mathbb{S}$ and stopped when $\rho$ reaches 0. For $\ell \in (0, +\infty)$, we will write $\mathbb{P}_\ell^\ast$ for $\mathbb{P}_{\ell \delta_{\mu, 0}}^\ast$.

If $Q$ is a measure on $\mathbb{S}$ and $\varphi$ is a non-negative measurable function defined on the measurable space $\mathbb{R}^\times \times \mathbb{S} \times \mathbb{S}$, we denote by

$$Q[\varphi(u, \omega, \cdot)] = \int_{\mathbb{S}} \varphi(u, \omega, \mathcal{F})Q(d\mathcal{F}).$$

In other words, the integration concerns only the third component of the function $\varphi$.

We can now state the Special Markov Property.

**Theorem 3.2** (Special Markov property). Let $\varphi$ be a non-negative measurable function defined on $\mathbb{R}_+ \times \mathcal{M}_f(\mathbb{R}_+) \times \mathbb{S}$. Then, we have $\mathbb{P}$-a.s.

$$\mathbb{E}\left[ \exp\left( -\sum_{i \in \mathcal{F}} \varphi(A_{i}, \rho_{\mu_{i}}, \mathcal{F}') \right) \mid \mathcal{F}_\infty \right]$$

$$= \exp\left( -\int_0^\infty du \alpha_1 N\left[ 1 - e^{-\varphi(u, \mu_{i})} \right]_{\mu=\rho_{\mu}} \right)$$

$$\exp\left( -\int_0^\infty du \int_{(0, \infty)} \pi_1(d\ell) \left( 1 - \mathbb{E}_\ell^\ast [e^{-\varphi(u, \mu_{i})}]_{\mu=\rho_{\mu}} \right) \right). \ (34)$$

In other words, the law under $\mathbb{P}$ of the excursion process $\sum_{i \in \mathcal{F}} \delta_{(A_{i}, \rho_{\mu_{i}}, \mathcal{F}')}(du \mu d\mathcal{F})$, given $\mathcal{F}_\infty$, is the law of a Poisson point measure with intensity

$$1_{[u \geq 0]}du \delta_{\rho_{\mu}}(d\mu) \left( \alpha_1 N(d\mathcal{F}) + \int_{(0, \infty)} \pi_1(d\ell)\mathbb{P}_\ell^\ast(d\mathcal{F}) \right).$$

Informally speaking, this Theorem gives the distribution of the marked exploration process “above” the pruned CRT. The end of this section is now devoted to its proof.

Let us first remark that, if $\lim_{\lambda \to +\infty} \phi_1(\lambda) < +\infty$, we have $\alpha_1 = 0$ and $\pi_1$ is a finite measure. Hence, there is no marks on the skeleton and the number of marks on the nodes is finite on every bounded interval of time. The proof of Theorem 3.2 in that case is easy and left to the reader. For the rest of this Section, we assume that $\lim_{\lambda \to +\infty} \phi_1(\lambda) = +\infty$.

### 3.1 Preliminaries

Fix $t > 0$ and $\eta > 0$. For $\mathcal{F} = (\mathcal{F}_t = (\rho_{s_i}, m_{s_i}), s \geq 0)$, we set $B = \{\sigma(\mathcal{F}) = +\infty\} \cup \{T_{\eta}(\mathcal{F}) = +\infty\} \cup \{L_{\eta}(\mathcal{F}) = -\infty\}$ where $\sigma(\mathcal{F}) = \inf\{s > 0; \rho_{s_i} = 0\}$, $T_{\eta}(\mathcal{F}) = \inf\{s \geq \eta; \rho_{s_i, 1} \geq \eta\}$ and $L_{\eta}(\mathcal{F}) = \sup\{s \in [0, \sigma(\mathcal{F})]; \rho_{s_i, 1} \geq \eta\}$, with the convention $\inf \emptyset = +\infty$ and $\sup \emptyset = -\infty$.

We consider non-negative bounded functions $\varphi$ satisfying the assumptions of Theorem 3.2 and these four conditions:

$$(h_1) \quad \varphi(u, \mu, \mathcal{F}) = 0 \text{ for any } u \geq t.$$
Let \( \mathcal{M}(\mathbb{R}_+) \) be the Lebesgue measure. We denote by \( \mathcal{G}_\mathcal{F} \) the \( \mathcal{F} \)-field generated by \( \mathcal{G}_\mathcal{F} \).

(h_3) \( \varphi(u, \mu, \mathcal{F}) = 0 \) on \( B \); and if \( \mathcal{F} \in \mathcal{B}^c \) then \( \varphi(u, \mu, \mathcal{F}) \) depends on \( \mathcal{F} \) only through \( (\mathcal{F}_{u, u} \in [T, L]) \).

(h_4) The function \( \mu \mapsto \varphi(u, \mu, \mathcal{F}) \) is continuous with respect to the distance \( D(\mu, \mu') + |(\mu, 1) - (\mu', 1)| \) on \( \mathcal{M}_f(\mathbb{R}_+) \), where \( D \) is a distance on \( \mathcal{M}_f(\mathbb{R}_+) \) which defines the topology of weak convergence.

**Lemma 3.3.** Let \( \varphi \) satisfies (h_1 – h_3) and let \( w \) be defined on \( (0, \infty) \times [0, \infty) \times \mathcal{M}_f(\mathbb{R}_+) \) by

\[
w(\ell, u, \mu) = \mathbb{E}^*_t[e^{-\varphi(u, \mu, \cdot)}].
\]

Then, for \( N \) a.e. \( \mu \in \mathcal{M}_f(\mathbb{R}_+) \), the function \( (\ell, u) \mapsto w(\ell, u, \mu) \) is uniformly continuous on \( (0, \infty) \times [0, \infty) \).

**Proof.** Let \( u > 0 \) and \( \ell' > \ell \). If we set \( \tau_\ell = \inf\{ t \geq 0, \rho_t(\{0\}) = \ell \} \) we have, by the strong Markov property at time \( \tau_\ell \) and assumption (h_3), that

\[
\mathbb{E}^*_{\ell'}[e^{-\varphi(u, \mu, \cdot)}] = \mathbb{E}^*_{\ell'}[1_{\{T < \tau_\ell\}}\mathbb{E}^*_\ell[e^{-\varphi(u, \mu, \cdot)}]] + \mathbb{E}^*_{\ell'}[e^{-\varphi(u, \mu, \cdot)}1_{\{T \leq \tau_\ell\}}].
\]

Therefore,

\[
w(\ell', u, \mu) - w(\ell, u, \mu) \leq \mathbb{E}^*_\ell 1_{\{T \leq \tau_\ell\}} \mathbb{E}^*_\ell [e^{-\varphi(u, \mu, \cdot)}] + \mathbb{E}^*_\ell [e^{-\varphi(u, \mu, \cdot)}1_{\{T \leq \tau_\ell\}}]
\]

\[
\leq 2\mathbb{P}^*_\ell (T \leq \tau_\ell)
\]

\[
= 2\mathbb{P}^*_{\ell'-\ell}(T < +\infty).
\]

Using Lemma [1.6] for \( \ell' - \ell < \eta \), we get

\[
|w(\ell', u, \mu) - w(\ell, u, \mu)| \leq 2 \left( 1 - e^{-(\ell' - \ell)\mathbb{N}[T < \infty]} \right).
\]

Since \( \mathbb{N}[T < \infty] < \infty \), we then deduce there exists a finite constant \( c_\eta \) s.t. for all function \( \varphi \) satisfying (h_3),

\[
|w(\ell', u, \mu) - w(\ell, u, \mu)| \leq c_\eta |\ell' - \ell|.
\]

The absolute continuity with respect to \( u \) is a direct consequence of assumptions (h_1 – h_2).

### 3.2 Stopping times

Let \( R(dt, du) \) be a Poisson point measure on \( \mathbb{R}_+^2 \) (defined on \( (\mathbb{S}, \mathcal{G}) \)) independent of \( \mathcal{F}_\infty \) with intensity the Lebesgue measure. We denote by \( \mathcal{G}_r \) the \( \sigma \)-field generated by \( R(\cdot \cap [0, t] \times \mathbb{R}_+) \). For every \( \varepsilon > 0 \), the process \( R^\varepsilon_t := R([0, t] \times [0, 1/\varepsilon]) \) is a Poisson process with intensity \( 1/\varepsilon \). We denote by \( (e_k^\varepsilon, k \geq 1) \) the time intervals between the jumps of \( (R^\varepsilon_t, t \geq 0) \). The random variables \( (e_k^\varepsilon, k \geq 1) \) are i.i.d. exponential random variables with mean \( \varepsilon \), and are independent of \( \mathcal{F}_\infty \). They define a mesh of \( \mathbb{R}_+ \) which is finer and finer as \( \varepsilon \) decreases to 0.
For $\epsilon > 0$, we consider $T^\epsilon_0 = 0, M^\epsilon_0 = 0$ and for $k \geq 0$,

\[
M^\epsilon_{k+1} = \inf\{i > M^\epsilon_k; m^i_{T^\epsilon_{k+1} \land \sigma S^\epsilon_{k+1}} \neq 0\},
\]

\[
S^\epsilon_{k+1} = T^\epsilon_k + \sum_{j=M^\epsilon_k+1}^{M^\epsilon_{k+1}} e^j_{\epsilon}, \tag{36}
\]

\[
T^\epsilon_{k+1} = \inf\{s > S^\epsilon_{k+1}; m_s = 0\},
\]

with the convention $\inf\emptyset = +\infty$. For every $t \geq 0$, we set $\tau^\epsilon_t = \int_0^t ds 1_{\bigcup_{k \geq 1} \{T^\epsilon_k \land S^\epsilon_{k+1}\}}(s)$ and

\[
\mathcal{F}^\epsilon_t = \sigma(\mathcal{F}_t \cup \mathcal{G}^{\tau^\epsilon_t}). \tag{37}
\]

Notice that $T^\epsilon_k$ and $S^\epsilon_k$ are $\mathcal{F}^\epsilon$-stopping times.

Now we introduce a notation for the process defined above the marks on the intervals $[S^\epsilon_k, T^\epsilon_k]$. We set, for $a \geq 0$, $\tilde{H}_a$ the level of the first mark, $\rho^a_-$ the restriction of $\rho_a$ strictly below it and $\rho^a_+$ the restriction of $\rho_a$ above it:

\[
\tilde{H}_a = \sup\{t > 0; m_a([0, t]) = 0\}, \quad \rho^-_a = \rho_a(\cdot \cap [0, \tilde{H}_a)) \tag{38}
\]

and $\rho^+_a$ is defined by $\rho_a = [\rho^-_a, \rho^+_a]$, that is for any $f \in \mathcal{B}_+(\mathbb{R}_+)$,

\[
\langle \rho^+_a, f \rangle = \int_{[\tilde{H}_a, \infty)} f(r - \tilde{H}_a) \rho_a (dr). \tag{39}
\]

For $k \geq 1$ and $\epsilon > 0$ fixed, we define $\mathcal{G}^{k,\epsilon} = (\rho^{k,\epsilon}, m^{k,\epsilon})$ in the following way: for $s > 0$ and $f \in \mathcal{B}_+(\mathbb{R}_+)$

\[
\rho^{k,\epsilon}_s = \rho^+_s \chi_{(S^\epsilon_{k+1}, T^\epsilon_k]},
\]

\[
\langle (m^a)^{k,\epsilon}_s, f \rangle = \int_{(H^\epsilon_{S^\epsilon_k}, \infty)} f(r - H^\epsilon_{S^\epsilon_k}) m^a_s \chi_{(S^\epsilon_{k+1}, T^\epsilon_k]}(dr), \quad \text{with } a \in \{\text{nod, ske}\},
\]

and $m^{k,\epsilon}_s = ((m^{\text{nod}})^{k,\epsilon}_s, (m^{\text{ske}})^{k,\epsilon}_s)$. Notice that $\rho^{k,\epsilon}_s(\{0\}) = \rho_s(\{\tilde{H}_0^\epsilon\})$. For $k \geq 1$, we consider the $\sigma$-field $\mathcal{G}^{(e),k}$ generated by the family of processes $\left(\mathcal{G}_{(T^\epsilon_{l+1} \land S^\epsilon_{l+1} -)}, s > 0 \right)_{l \in \{0, \ldots, k-1\}}$.

Notice that for $k \in \mathbb{N}^*$

\[
\mathcal{G}^{(e),k} \subset \mathcal{G}^{\tau^\epsilon_k}_{S^\epsilon_k}. \tag{40}
\]

### 3.3 Approximation of the functional

Let $\mathcal{G}$ be a marked exploration process and $g$ be a function defined on $\mathcal{S}$. We decompose the path of $\rho$ according to the excursions of the total mass of $\rho$ above its minimum as in Section 1.5 with a slight difference if the initial measure $\mu$ charges $\{0\}$. More precisely, we perform the same decomposition as in Section 1.5 until the height process reaches 0. If $\mu(\{0\}) = 0$, then $H_t = 0 \iff \rho_t = 0$ and the decompositions are the same. If not, there remains a part of the process which
is not decomposed and is gathered in a single excursion (defined as \((a_i, b_i)\) in Figure 1). We set 
\[ Y_t = (\rho_t, 1) \] and \( J_t = \inf_{0 \leq u \leq t} Y_u \). Recall that \((Y_t, t \geq 0)\) is distributed under \( \mathbb{P}_{\mu}^{*} \) as a Lévy process with Laplace exponent \( \psi \) started at \( \langle \mu, 1 \rangle \) and killed when it reaches 0. Let \((a_i, b_i), i \in \mathcal{K}, \) be the intervals excursion of \( Y-J \) away from 0. For every \( i \in \mathcal{K} \), we define \( h_i = H_{a_i} = H_{b_i} \). We set \( \tilde{\mathcal{K}} = \{ i \in \mathcal{K}; h_i > 0 \} \) and for \( i \in \tilde{\mathcal{K}} \) let \( \mathcal{S}^i = (\tilde{\rho}^i, \tilde{m}^i) \) be defined by (22) and (23). If the initial measure \( \mu \) does not charge 0, we have \( \tilde{\mathcal{K}} = \mathcal{K} \) and we set \( \mathcal{K}^* = \mathcal{K} = \mathcal{K} \). If the initial measure \( \mu \) charges 0, we consider \( i_0 \notin \tilde{\mathcal{K}} \) and set \( \mathcal{K}^* = \tilde{\mathcal{K}} \cup \{ i_0 \} \), \( a_{i_0} = \inf \{ a_i; i \in \mathcal{K}, h_i = 0 \} \), \( b_{i_0} = \sup \{ b_i; i \in \mathcal{K}, h_i = 0 \} \) and \( \mathcal{S}^{i_0} = (\tilde{\rho}^{i_0}, \tilde{m}^{i_0}) \) with

\[
\langle \tilde{\rho}^{i_0}, f \rangle = \int_{[0, +\infty)} f(x) \rho_{(a_{i_0} + t) \wedge b_{i_0}}(dx)
\]

and \( \tilde{m}^{i_0} = ((\tilde{m}^{\text{nod}})^{i_0})_t, (\tilde{m}^{\text{ske}})^{i_0})_t \). See Figure 1 to get the picture of the different excursions.

![Figure 1: Definition of the different excursions](image)

We define

\[
g^*(\mathcal{S}) = \sum_{i \in \mathcal{K}^*} g(\mathcal{S}^i).
\]

**Lemma 3.4.** \( \mathbb{P}\)-a.s., we have, for \( \epsilon > 0 \) small enough,

\[
\sum_{i \in \mathcal{K}} \varphi(A_{a_i}, \rho_{a_i-}, \mathcal{S}^i) = \sum_{k=1}^{\infty} \varphi(A_{S^k}, \rho_{S^k}, \mathcal{S}^{k, \epsilon}) = \sum_{k=1}^{\infty} \varphi^*(A_{S^k}, \rho_{S^k}, \mathcal{S}^{k, \epsilon}),
\]

where the sums have a finite number of non zero terms.
Proof. First equality. By assumptions \( (h_1) \) and \( (h_2) \), as \( \mathbb{N} [T_\eta < +\infty] < +\infty \), the set

\[ \mathcal{S} = \{ i \in \mathcal{S}, \varphi(A_{\alpha}, \rho_{\alpha}, \mathcal{S}^j) \neq 0 \} \]

is finite. Therefore, for \( \varepsilon \) small enough, for every \( j \in \mathcal{S} \), the mesh defined by \( \mathcal{S}^j \) intersects the interval \( (a_j, \beta_j) \): in other words, there exists an integer \( k_j \) such that \( S_{k_j}^\varepsilon \in (a_j, \beta_j) \) (and that integer is unique).

Moreover, for every \( j \in \mathcal{S} \), we can choose \( \varepsilon \) small enough so that \( S_{k_j}^\varepsilon < T_\eta(\rho^j) \), which gives that, for \( \varepsilon \) small enough,

\[ \varphi(A_{\alpha}, \rho_{\alpha}, \mathcal{S}^j) = \varphi(A_{\alpha}, \rho_{\alpha}, \mathcal{S}^{k_j,\varepsilon}). \]

Finally, as the mark at \( \alpha_j \) is still present at time \( S_{k_j}^\varepsilon \), the additive functional \( A \) is constant on that time interval and \( \rho_{\alpha} = \rho_{S_{k_j}^\varepsilon} \). Therefore, we get

\[ \varphi(A_{\alpha}, \rho_{\alpha}, \mathcal{S}^j) = \varphi(A_{S_{k_j}^\varepsilon}, \rho_{S_{k_j}^\varepsilon}, \mathcal{S}^{k_j,\varepsilon}). \]

Second equality. Let \( j \in \mathcal{S} \). We consider the decomposition of \( \mathcal{S}^{k_j,\varepsilon} \) according to \( \rho^{k_j,\varepsilon} \) above its minimum described at the beginning of this Subsection. We must consider two cases:

First case: The mass at \( \alpha_j \) is on a node. Then, for \( \varepsilon > 0 \) small enough, we have \( T_\eta(\rho^j) > a_{i_0} \) and

\[ \varphi(A_{S_{k_j}^\varepsilon}, \rho_{S_{k_j}^\varepsilon}, \rho^{k_j,\varepsilon}) = \varphi(A_{S_{k_j}^\varepsilon}, \rho_{S_{k_j}^\varepsilon}, \rho^{j,i_0}) = \varphi^*(A_{S_{k_j}^\varepsilon}, \rho_{S_{k_j}^\varepsilon}, \rho^{k_j,\varepsilon}) \]

as all the terms in the sum that defines \( \varphi^* \) are zero but for \( i = i_0 \).

Second case: The mass at \( \alpha_j \) is on the skeleton. In that case, we again can choose \( \varepsilon \) small enough so that the interval \( [T_\eta, L_\eta] \) is included in one excursion interval above the minimum of the exploration total mass process of \( S^{k_j,\varepsilon} \). We then conclude as in the previous case. \( \square \)

3.4 Computation of the conditional expectation of the approximation

Lemma 3.5. For every \( \mathcal{F}_\infty \)-measurable non-negative random variable \( Z \), we have

\[ \mathbb{E} \left[ Z \exp \left( -\sum_{k=1}^{\infty} \varphi^* (A_{S_k^\varepsilon}, \rho_{S_k^\varepsilon}, \mathcal{S}^{k_j,\varepsilon}) \right) \right] = \mathbb{E} \left[ Z \prod_{k=1}^{\infty} K_k(A_{S_k^\varepsilon}, \rho_{S_k^\varepsilon}) \right], \]

where \( \gamma = \psi^{-1}(1/\varepsilon) \) and

\[ K_k(r, \mu) = \frac{\psi(\gamma)}{\phi_1(\gamma)} \left( \frac{\gamma - \nu(r, \mu)}{\psi(\gamma) - \psi(\nu(r, \mu))} \right) \left( \alpha_1 + \int_0^1 d\ell_1 \int_{(0, \infty)} \ell_1 \pi_1(d\ell_1) w(\ell_1, r, \mu) e^{-\gamma(1-\mu)\ell_1} \right) \]

where

\[ w(\ell, r, \mu) = \mathbb{E}_\ell^r \left[ e^{-\varphi(r, \mu, \cdot)} \right] \quad \text{and} \quad \nu(r, \mu) = \mathbb{N} \left[ 1 - e^{-\varphi(r, \mu, \cdot)} \right]. \]
Proof. This proof is rather long and technical. We decompose it in several steps.

Step 1. We introduce first a special form of the random variable $Z$.

Let $p \geq 1$. Recall that $H_{t, t'}$ denotes the minimum of $H$ between $t$ and $t'$ and that $\bar{H}_t$ defined by (38) represents the height of the lowest mark. We set

\[
\xi_{d}^{p-1} = \sup \left\{ t > T_{p-1}^e; \ H_t = H_{T_{p-1}^e, S_p^e} \right\},
\]

\[
\xi_{g}^{p} = \inf \left\{ t > T_{p-1}^e; \ H_t = \bar{H}_{S_p^e} \text{ and } H_{t, S_p^e} = H_t \right\}.
\]

$\xi_{d}^{p-1}$ is the time at which the height process reaches its minimum over $[T_{p-1}^e, S_p^e]$. By definition of $T_{p-1}^e$, $m_{T_{p-1}^e} = 0$ (there is no mark on the lineage of the corresponding individual). On the contrary, $m_{S_p^e} \neq 0$, $m_{S_p^e}(\{\bar{H}_{S_p^e}\}) \neq 0$ but $m_{S_p^e}(\{0, \bar{H}_{S_p^e}\}) = 0$. In other words, at time $S_p^e$, some mark exists and the lowest mark is situated at height $\bar{H}_{S_p^e}$. Roughly speaking, $\xi_{g}^{p}$ is the time at which this lowest mark appears, see figure 3.4 to help understanding. Let us recall that, by the snake property (21), $m_{\xi_{d}^{p-1}} = 0$ and consequently, $\xi_{d}^{p-1} < \xi_{g}^{p}$ a.s.

\[\text{Figure 2: Position of various random times}\]

We consider a bounded non-negative random variable $Z$ of the form $Z = Z_0 Z_1 Z_2 Z_3$, where $Z_0 \in \mathcal{G}^{(e), p-1}$, $Z_1 \in \sigma(\mathcal{A}_u, T_{p-1}^e \leq u < \xi_{d}^{p-1})$, $Z_2 \in \sigma(\mathcal{A}_u, \xi_{d}^{p-1} \leq u < \xi_{g}^{p})$ and $Z_3 \in \sigma(\mathcal{A}_{(T_{k+1}^e, k)} \mathcal{S}_{k+1}^e, s \geq 0, k \geq p)$ are bounded non-negative.

Step 2. We apply the strong Markov property to get rid of terms which involve $S_p^e$ and $T_{p}^e$. We first apply the strong Markov property at time $T_{p}^e$ by conditioning with respect to $\mathcal{F}_{T_p^e}$. We obtain

\[
\mathbb{E} \left[ Z \exp \left( - \sum_{k=1}^{p} \phi^*(A_{S_k^e}, P_{S_k^e}, \mathcal{G}_k^e) \right) \right]
= \mathbb{E} \left[ Z_0 Z_1 Z_2 \exp \left( - \sum_{k=1}^{p} \phi^*(A_{S_k^e}, P_{S_k^e}, \mathcal{G}_k^e) \right) \right] \mathbb{E}_{\rho_{T_p^e}, 0}[Z_3].
\]
Recall notation (38) and (39). Notice that $\rho_{T^e_p} = \rho_{P^e_p}$, and consequently $\rho_{T^e_p}$ is measurable with respect to $\mathcal{F}_{S_p}^e$. So, when we use the strong Markov property at time $S_p^e$, we get thanks to (40)

$$
\mathbb{E} \left[ Z \exp \left( -\sum_{k=1}^{p} \varphi^*(A_{S_k}^e, \rho_{S_k}^e, \mathcal{G}^k) \right) \right]
= \mathbb{E} \left[ Z_0 Z_1 Z_2 \exp \left( -\sum_{k=1}^{p-1} \varphi^*(A_{S_k}^e, \rho_{S_k}^e, \mathcal{G}^k) \right) \mathbb{E}_{\rho_{S_p}^+}^e \left[ e^{-\varphi^*(A_{S_p}^e, \rho_{S_p}^e, \cdot)} \right] \mathbb{E}_{\rho_{S_p}^e}^e [Z_3] \right].
$$

Using the strong Markov property at time $T^e_{p-1}$, and the lack of memory for the exponential r.v., we get

$$
\mathbb{E} \left[ Z \exp \left( -\sum_{k=1}^{p} \varphi^*(A_{S_k}^e, \rho_{S_k}^e, \mathcal{G}^k) \right) \right]
= \mathbb{E} \left[ Z_0 \exp \left( -\sum_{k=1}^{p-1} \varphi^*(A_{S_k}^e, \rho_{S_k}^e, \mathcal{G}^k) \right) \phi \left( A_{S_p}^e, \rho_{S_p}^e, \rho_{T^e_p}^e \right) \right],
$$

with

$$
\phi(b, \mu, \nu) = \mathbb{E}_{\nu}^e \left[ Z_1 Z_2 \mathbb{E}_{\rho_{T^e_p}}^e \left[ e^{-\varphi^*(A_{T^e_p}, \mu, \cdot)} \right] \mathbb{E}_{\rho_{T^e_p}}^e [Z_3] \right],
$$

where $\tau'$ is distributed under $\mathbb{P}_{\nu}$ as $S_1^e$.

**Step 3.** We compute the function $\phi$ given by (45). To simplify the formulas, we set

$$
F(b', \mu') = \mathbb{E}_{\mu'}^e \left[ e^{-\varphi^*(b+b', \mu, \cdot)} \right], \quad G(\mu') = \mathbb{E}_{\mu'}^e [Z_3]
$$

(the dependence on $b$ and $\mu$ of $F$ is omitted) so that

$$
\phi(b, \mu, \nu) = \mathbb{E}_{\nu}^e \left[ Z_1 Z_2 F(A_{T^e_p}, \rho_{T^e_p}^e) G(\rho_{T^e_p}^e) \right].
$$

The proof of the following technical Lemma is postponed to the end of this Sub-section.

**Lemma 3.6.** We set $q(du, d\ell_1) = \alpha_1 \delta_{(0,0)}(du d\ell_1) + du \ell_1 \pi_1(d\ell_1)$ and by convention $\pi(\{0\}) = 0$. We have:

$$
\phi(b, \mu, \nu) = \mathbb{E}_{\nu}^e \left[ Z_1 Z_2 \frac{\Gamma_f(A_{T^e_p})}{\Gamma_1} G(\rho_{T^e_p}^e) \right],
$$

where for a non-negative function $f$ defined on $[0, \infty) \times \mathcal{M}_f(\mathbb{R}_+)$

$$
\Gamma_f(a) = \int_{[0,1] \times [0, \infty)} q(du, d\ell_1) \int \mathbb{M}(d\rho', d\eta', dm') \ e^{-\gamma(\rho', 1) - \gamma u \ell_1} f(a, \eta' + (1 - u)\ell_1 \delta_0)
$$

and for $f = 1$, $\Gamma_1$ does not depend on $a$.  

1452
We now use the particular form of $F$ to compute $\Gamma_F$. Using (41) and Lemma 1.6, we get

$$F(a, \mu') = \mathbb{E}_\mu^* \left[ e^{-\varphi'(b+a, \mu')} \right] = \mathbb{E}_\mu'([0,1]) \left[ e^{-\varphi'(0,\infty)} \right] \int [-e^{-\varphi'(b+a, \mu')}] .$$

Using $w$ and $v$ defined in (44), we get

$$F(a, \eta + (1-u)\ell_1 \delta_0) = w((1-u)\ell_1, b + a, \mu) e^{-\gamma u \ell_1} \mathbb{E}_\mu \left[ e^{-\gamma (\rho + 1) \eta} \left( \varphi(v(b+a, \mu)) \right) \right] = w((1-u)\ell_1, b + a, \mu) e^{-\gamma (1-u) \ell_1} .$$

We deduce that

$$\Gamma_F(a) = \frac{\gamma - \varphi(v(b+a, \mu))}{\psi(\gamma) - \psi(v(b+a, \mu))} \left( \alpha_1 + \int_0^1 \ell_1 \pi_1(d\ell_1) w(u\ell_1, b + a, \mu) e^{-\gamma (1-u) \ell_1} \right) ,$$

and with $F = 1$, $\Gamma_1 = \frac{\gamma}{\psi(\gamma)} = \phi_1(\gamma)/\psi(\gamma)$.

Finally, plugging this formula in (48) and using the function $K_\varepsilon$ introduced in (43), we have

$$\phi(b, \mu, v) = \mathbb{E}_v \left[ Z_{1} Z_{2} K_\varepsilon(b + A_v, \mu) G(\rho^-) \right] .$$

(49)

**Step 4. Induction.**

Plugging the expression (49) for $\phi$ in (45), and using the arguments backward from (45) we get

$$\mathbb{E} \left[ Z \exp \left( -\sum_{k=1}^p \varphi^+(A_{s_k}, \rho_{s_k}, \varphi^{k, \varepsilon}) \right) \right] = \mathbb{E} \left[ Z \exp \left( -\sum_{k=1}^{p-1} \varphi^+(A_{s_k}, \rho_{s_k}, \varphi^{k, \varepsilon}) \right) K_\varepsilon(A_{s_p}, \rho_{s_p}) \right] .$$

In particular, from monotone class theorem, this equality holds for any non-negative $Z$ measurable w.r.t. the $\sigma$-field $\mathcal{F}_\infty = \sigma(\mathcal{F}_t, t \in [A_{t+1}], k \geq 0)$. Notice that $K_\varepsilon(A_{s_p}, \rho_{s_p})$ is measurable w.r.t. $\mathcal{F}_\infty$. So, we may iterate the previous argument and let $p$ goes to infinity to finally get that for any non-negative random variable $Z \in \mathcal{F}_\infty$, we have

$$\mathbb{E} \left[ Z \exp \left( -\sum_{k=1}^\infty \varphi^+(A_{s_k}, \rho_{s_k}, \varphi^{k, \varepsilon}) \right) \right] = \mathbb{E} \left[ Z \prod_{k=1}^\infty K_\varepsilon(A_{s_k}, \rho_{s_k}) \right] .$$

Intuitively, $\mathcal{F}_\infty$ is the $\sigma$-field generated by $\mathcal{F}_\infty$ and the mesh $([A_{t_k}, A_{t_{k+1}}], k \geq 0)$. As $\mathcal{F}_\infty$ contains $\mathcal{F}_\infty$, the Lemma is proved.

**Proof of Lemma 3.6.** We consider the Poisson decomposition of $\mathcal{F}$ under $\mathbb{P}_v^+$ given in Lemma 1.6. Notice there exists a unique excursion $i_1 \in \mathcal{K}$ s.t. $a_{i_1} \prec \tau' < b_{i_1}$.

By hypothesis on $Z_1$, under $\mathbb{P}_v^+$, we can write $Z_1 = \mathcal{H}_1(v, \sum_{i \in \mathcal{K}, a_i < a_{i_1} \delta_i \prec \tau' \prec \varepsilon_i')$ for a measurable function $\mathcal{H}_1$. We can also write $Z_2 = \mathcal{H}_2(\rho_u, \xi_d^- \leq u < \xi_1^+)$ for a measurable function $\mathcal{H}_2$ as $m_u = 0$ for
We have, using the Poisson representation of Lemma 1.6 and (15), that

$$
\phi(b, \mu, \nu) = \mathbb{E} \left[ \int v(dv) \mathbf{1}_{\{\tau' > \sum_{s < v} \sigma(\mathcal{S}^s)\}} \mathcal{N} \left( \nu, \sum_{s < v} \delta_{(s, \mathcal{S}^s)} \right) h_P^v \left( r, \sum_{s < v} A_{\sigma(\mathcal{S}^s)}(\mathcal{S}^s) \right) \right],
$$

(50)

where $$\sum_{s} \delta_{(s, \mathcal{S}^s)}$$ is a Poisson point measure with intensity $$v(ds)\mathbb{N}[d\mathcal{S}]$$, $$\sigma(\mathcal{S}) = \inf\{r > 0, \mathcal{S}_r = 0\}$$, 

$$A_t(\mathcal{S}^s) = \int_0^t dv' \mathbf{1}_{\{m_{v'}(\mathcal{S}^s) = 0\}}$$

and

$$h_P^v(r, a) = \mathbb{N} \left[ F(a + A_{t'}, \rho^+_{t'}) G([k, r, \rho^-_{t'}]) \mathcal{N}_2([k, r, \rho_2], 0 \leq t < \xi_g^1) \mathbf{1}_{\{\tau' < \sigma\}} \right].$$

Let $$(R_k, k \geq 1)$$ be the increasing sequence of the jumping times of a Poisson process of intensity $$1/\varepsilon$$, independent of $$\mathcal{S}$$. Then, by time-reversal, we have

$$h_P^v(r, a) = \mathbb{N} \left[ \sum_{k=1}^{+\infty} \mathbf{1}_{\{m_{R_k} \neq 0\}} \mathbf{1}_{\{\forall k' > k, m_{R_{k'}} = 0\}} F(a + A_{R_k} - A_{R_{k'}}, \eta^+_R) \right]

G([k, r, \eta^-_R], \mathcal{N}_2([k, r, \eta_2], \tau_k < u \leq \sigma)],
$$

where $$\tau_k = \inf\{t > R_k; m_t = 0\}$$. We then apply the strong Markov property at time $$R_k$$ and the Poisson representation of the marked exploration process to get

$$h_P^v(r, a) = \mathbb{N} \left[ \sum_{k=1}^{+\infty} \mathbf{1}_{\{m_{R_k} \neq 0\}} G([k, r, \eta^-_R]) \right]

\mathbb{E}^+_R \left[ \mathbf{1}_{\{\forall k' > 0, m_{R_{k'}} = 0\}} F(a + A_{R_k}, \eta') \mathcal{N}_2([k, r, \eta_2], \tau_0 < u \leq \sigma) \right]_{\eta' = \eta^+_R},
$$

where $$\tau_0 = \inf\{t > 0; m_t = 0\}$$. Now, let us remark that, if $$m_0 \neq 0$$, then $$m_s \neq 0$$ for $$s \in [0, \tau_0]$$ and $$A_{\tau_0} = 0$$. Therefore, $$m_{R_1} = 0$$ implies $$R_1 > \tau_0$$. The strong Markov property at time $$\tau_0$$ gives, with $$\eta' = \eta^+_R$$,

$$1_{\{m_{R_k} \neq 0\}} \mathbb{E}^+_R \left[ \mathbf{1}_{\{\forall k' > 0, m_{R_{k'}} = 0\}} F(a + A_{R_k}, \eta') \mathcal{N}_2([k, r, \eta_2], \tau_0 < u \leq \sigma) \right]

= 1_{\{m_{R_k} \neq 0\}} \mathbb{E}^+_R \left[ (R_1 > \sigma) \mathbb{E}^+_R \left[ \mathbf{1}_{\{\forall k' > 0, m_{R_{k'}} = 0\}} F(a + A_{R_k}, \eta') \mathcal{N}_2([k, r, \eta_2], 0 < u \leq \sigma) \right] \right].
$$

We have, using the Poisson representation of Lemma 1.6 and (15), that

$$\mathbb{P}^+_R \left( R_1 > \sigma \right) = \mathbb{P}^+_R \left[ e^{-\gamma/\varepsilon} \right] = e^{-\gamma(\rho^+_R, 1)},$$

as $$\gamma = \psi^{-1}(1/\varepsilon)$$. We obtain

$$h_P^v(r, a) = \mathbb{N} \left[ \sum_{k=1}^{+\infty} 1_{\{m_{R_k} \neq 0\}} G(\rho^-_{R_k}, \eta^-_{R_k}, \rho^+_R, \eta^+_R) \right],
$$

1454
where
\[ \bar{G}(\rho, \eta, \rho', \eta') = G([k_r \nu, \eta]) e^{-\gamma'(\rho', 1)} \]
\[ \mathbb{E}^\rho \left[ 1_{\forall k' > 0, m_{\eta_k} = 0} F(a + A_\sigma, \eta') \mathcal{H}_2([k_r \nu, \eta], 0 < u \leq \sigma) \right]. \]

As \( \sum_{k \geq 1} \delta_{R_k} \) is a Poisson point process with intensity \( 1/\epsilon \), we deduce that
\[ h_x^\epsilon(r, a) = \frac{1}{\epsilon} \left[ \int_0^\sigma dt \ 1_{\{m, \mu \neq 0\}} \bar{G}(\rho^-, \eta^-, \rho^+, \eta^+) \right]. \]

Using Proposition 1.9 we get
\[ h_x^\epsilon(r, a) = \frac{1}{\epsilon} \int_0^\infty da \ e^{-\alpha a} \mathbb{M}_a \left[ 1_{\{m \neq 0\}} \bar{G}(\mu^-, \nu^-, \mu^+, \nu^+) \right]. \]

For \( r > 0 \) and \( \mu \) a measure on \( \mathbb{R}_+ \), let us define the measures \( \mu_{\geq r} \) and \( \mu_{<r} \) by
\[ \langle \mu_{\geq r}, f \rangle = \int f(x-r) 1_{\{x \geq r\}} \mu(dx) \quad \text{and} \quad \langle \mu_{<r}, f \rangle = \int f(x) 1_{\{x < r\}} \mu(dx). \]

Using Palm formula, we get
\[ \mathbb{M}_a \left[ 1_{\{m \neq 0\}} \bar{G}(\mu^-, \nu^-, \mu^+, \nu^+) \right] = \int_0^\sigma dr \int_{[0,1] \times [0,\infty)} q(du, d\ell_1) \mathbb{M}_a \left[ 1_{\{m(0, r) = 0\}} \bar{G}(\mu_{<r}, \nu_{<r}, \mu_{\geq r} + u\ell_1 \delta_0, \nu_{\geq r} + (1-u)\ell_1 \delta_0) \right]. \]

Using the independence of the Poisson point measures, we get
\[ \mathbb{M}_a \left[ 1_{\{m((0, r) = 0\)} \bar{G}(\mu_{<r}, \nu_{<r}, \mu_{\geq r} + u\ell_1 \delta_0, \nu_{\geq r} + (1-u)\ell_1 \delta_0) \right] = \int \mathbb{M}_r(d\mu, d\nu, dm) \int \mathbb{M}_{a-r}(d\rho', d\eta', dm') 1_{\{m = 0\}} \bar{G}(\mu, \nu, \rho' + u\ell_1 \delta_0, \eta' + (1-u)\ell_1 \delta_0). \]

We deduce that
\[ h_x^\epsilon(r, a) = \frac{1}{\epsilon} \int_{[0,1] \times [0,\infty)} q(du, d\ell_1) \int \mathbb{M}(d\rho, d\eta, dm) \int \mathbb{M}(d\rho', d\eta', dm') 1_{\{m = 0\}} \bar{G}(\rho, \eta, \rho' + u\ell_1 \delta_0, \eta' + (1-u)\ell_1 \delta_0). \]

Using this and (50) with similar arguments (in reverse order), we obtain (48). \( \Box \)
3.5 Computation of the limit

Recall notation of Section 3.2. Let $A^e_t$ be the Lebesgue measure of $[0,s] \cap \left( \bigcup_{k \geq 0} [T^e_k, S^e_{k+1}] \right)$. The process $t \mapsto \sup \{ i \in \mathbb{N} ; \sum_{j=1}^{i} e^e_j \leq A^e_t \}$ is a Poisson process with intensity $1/e$ and the process $s \mapsto N_{e,t}$, where

$$N_{e,t} = \sup \{ k \in \mathbb{N} ; A^e_{S^e_k} \leq t \} = \sup \{ k \in \mathbb{N} ; \sum_{j=1}^{M^e_k} e^e_j \leq A^e_t \},$$

is a marked Poisson process with intensity $\mathbb{P}(m_\gamma \neq 0)/e$, where $\tau$ is an exponential random variable with mean $e$ independent of $\mathcal{F}$.

We first study the process $t \mapsto N_{e,t}$.

**Lemma 3.7.** The process $t \mapsto N_{e,t}$ is a Poisson process with intensity $\frac{\phi_1(\gamma)}{e \psi_0(\gamma)}$, where $\gamma = \psi^{-1}(1/e)$.

**Proof.** We have, by the similar computations as in the proof of Lemma 3.5,

$$\mathbb{P}(m_\gamma = 0) = \frac{1}{e} \mathbb{E} \left[ \int_0^\infty dt \ e^{-t/e} 1_{\{m_\gamma = 0\}} \right]$$

$$= \frac{1}{e} \int_0^\infty ds \ e^{-\gamma s} N \left[ \int_0^\sigma dt \ e^{-t/e} 1_{\{m_\gamma = 0\}} \right]$$

$$= \frac{1}{e \gamma} N \left[ \int_0^\sigma dt \ e^{-t/e} 1_{\{m_\gamma = 0\}} \right].$$

By time reversibility and using optional projection and (15), we have

$$N \left[ \int_0^\sigma dt \ e^{-t/e} 1_{\{m_\gamma = 0\}} \right] = N \left[ \int_0^{\sigma} dt \ e^{-(\sigma-t)/e} 1_{\{m_\gamma = 0\}} \right]$$

$$= N \left[ \int_0^{\sigma} dt \ e^{-\gamma \rho_t^{1-1}} 1_{\{m_\gamma = 0\}} \right].$$

The proof of Lemma 2.1, see (29) and (31), gives that $\mathbb{P}(m_\gamma = 0) = \frac{1}{\psi_0(\gamma)} - \phi_1(\gamma)$. Since $e^{-1} = \psi(\gamma) = \psi_0(\gamma) - \phi_1(\gamma)$, we get $\frac{1}{e} \mathbb{P}(m_\gamma \neq 0) = \frac{\phi_1(\gamma)}{e \psi_0(\gamma)}$. \hfill $\square$

We then get the following Corollary.

**Corollary 3.8.** There exists a sub-sequence $(e_j, j \in \mathbb{N})$ decreasing to 0, s.t. $\mathbb{P}$-a.s. for any $t_0 \geq 0$ and any continuous function $h$ defined on $\mathbb{R}_+ \times \mathcal{M}_f(\mathbb{R}_+)$ such that $h(u, \mu) = 0$ for $u \geq t_0$, we have, with $\gamma_j = \psi^{-1}(1/e_j)$,

$$\lim_{j \to \infty} \phi_1(\gamma_j)^{-1} \sum_{k=1}^{\infty} h(A_{S^e_k}, \rho_{S^e_k}^{-1}) = \int_0^\infty h(u, \rho_0) \ du.$$
Proof. Notice that as a direct consequence of (9) and (20), we get
\[ \lim_{\varepsilon \to 0} \varepsilon \psi_0'(\gamma) = 1. \]
Recall that \((A^\varepsilon_{S_k}, k \geq 1)\) are the jumping time of the Poisson process \(t \mapsto N_{\varepsilon,t}\) with parameter \(\phi_1(\gamma)/\varepsilon \psi_0(\gamma)\). Standard results on Poisson process implies the vague convergence in distribution (see also Lemma XI.11.1 in \([18]\)) of \(\phi_1(\gamma)^{-1} \sum_{k=1}^{\infty} \delta_{A^\varepsilon_{S_k}}(dr)\) towards the Lebesgue measure on \(\mathbb{R}_+\) as \(\varepsilon\) goes down to 0. Since the limit is deterministic, the convergence holds in probability and a.s. along a decreasing sub-sequence \((\varepsilon_j, j \in \mathbb{N})\). In particular, if \(g\) is a continuous function on \(\mathbb{R}_+\) with compact support (hence bounded), we have that a.s.
\[ \lim_{j \to \infty} \phi_1(\gamma_j)^{-1} \sum_{k=1}^{\infty} g(A^\varepsilon_{S_k}) = \int_0^\infty g(u) \, du. \]
Notice that \(A^\varepsilon_{S_k} \geq A_k\) and that a.s. \(A^\varepsilon_{S_k} \to A_k\) as \(\varepsilon\) goes down to 0. This implies that a.s. \((A^\varepsilon_{S_k}, s \geq 0)\) converges uniformly on compacts to \((A_k, s \geq 0)\). Therefore, if \(g\) is continuous with compact support, we have a.s.
\[ \lim_{j \to \infty} \phi_1(\gamma_j)^{-1} \sum_{k=1}^{\infty} g(A^\varepsilon_{S_k}) = \int_0^\infty g(u) \, du, \]
and this convergence also holds for a càdlàg function \(g\) with compact support as the Lebesgue measure does not charge the point of discontinuity of \(g\).

Let \(h\) be a continuous function defined on \(\mathbb{R}_+ \times \mathcal{M}_f(\mathbb{R}_+)\) such that \(h(u, \mu) = 0\) for \(u \geq t_0\). First let us remark that \(\rho^\varepsilon_{S_k} = \rho_{T^\varepsilon_k}\) and that \(m_{T^\varepsilon_k} = 0\). Using the strong Markov property at time \(T^\varepsilon_k\) and the second part of Corollary [2.2] we deduce that \(\mathbb{P}\)-a.s. for all \(k \in \mathbb{N}^*\),
\[ C_{A^\varepsilon_{T^\varepsilon_k}} = T^\varepsilon_k. \]
Therefore, as \(A^\varepsilon_{S_k} = A^\varepsilon_{T^\varepsilon_k}\), we have \(\mathbb{P}\)-a.s.
\[ \bar{\rho}_{A^\varepsilon_{S_k}} = \bar{\rho}_{A^\varepsilon_{T^\varepsilon_k}} = \rho_{T^\varepsilon_k} = \rho_{S_k}. \]
This gives
\[ \phi_1(\gamma_j)^{-1} \sum_{k=1}^{\infty} h(A^\varepsilon_{S_k}, \rho_{S_k}) = \phi_1(\gamma_j)^{-1} \sum_{k=1}^{\infty} h(A^\varepsilon_{S_k}, \bar{\rho}_{A^\varepsilon_{S_k}}) \]
and applying the convergence (51) to the càdlàg function
\[ g(u) = h(u, \bar{\rho}_u) \]
gives the result of the lemma. \(\square\)

We now study \(K_\varepsilon\) given by (43). We keep the same notation as in Lemma 3.5.
Lemma 3.9. There exists a deterministic function $\mathcal{R}$ s.t. $\lim_{\varepsilon \to 0} \mathcal{R}(\varepsilon) = 0$ and for all $\varepsilon > 0$ and $\mu \in \mathcal{M}_f(\mathbb{R}_+)$, we have:

$$\sup_{r \geq 0} \left| \phi_1(\gamma) \log(K_\varepsilon(r, \mu)) - \alpha_1 v(r, \mu) - \int_{(0, \infty)} \pi_1(\ell_1) \left( 1 - w(\ell_1, r, \mu) \right) \right| \leq \mathcal{R}(\varepsilon).$$

Proof. We have

$$K_\varepsilon(r, \mu) = \frac{\psi(\gamma)}{\psi(\gamma) - \psi(v(r, \mu))} \frac{\gamma - v(r, \mu)}{\gamma - v(r, \mu)} 1_{\phi_1(\gamma)}$$

$$\left( \alpha_1 \gamma + \gamma \int_0^1 \ell_1 \pi_1(\ell_1) w(u\ell_1, r, \mu) e^{-\gamma(1-u)\ell_1} \right)$$

$$= \frac{\psi(\gamma)}{\psi(\gamma) - \psi(v(r, \mu))} \frac{\gamma - v(r, \mu)}{\gamma - v(r, \mu)} 1_{\phi_1(\gamma)}$$

$$\left( \alpha_1 \gamma + \int_{(0, \infty)} \pi_1(\ell_1) \int_0^{\gamma \ell_1} e^{-s} ds w(\ell_1 - s \gamma, r, \mu) \right)$$

In particular, we have $\phi_1(\gamma) \log(K_\varepsilon(r, \mu)) = -A_1 + A_2 + A_3$, where

$$A_1(r) = \phi_1(\gamma) \log \left( 1 - \psi(v(r, \mu))/\psi(\gamma) \right),$$

$$A_2(r) = \phi_1(\gamma) \log(1 - v(r, \mu)/\gamma),$$

$$A_3(r) = \phi_1(\gamma) \log \left( 1 - \int_{(0, \infty)} \pi_1(\ell_1) \int_0^{\gamma \ell_1} e^{-s} ds \left( 1 - w(\ell_1 - s \gamma, r, \mu) \right)/\phi_1(\gamma) \right).$$

Thanks to (h$_3$), there exists a finite constant $a > 0$ s.t. $\mathbb{P}$-a.s. $v(r, \mu) < a$ for all $r \geq 0$. We deduce there exists $\varepsilon_0 > 0$ and a finite constant $c > 0$ s.t. $\mathbb{P}$-a.s for all $\varepsilon \in (0, \varepsilon_0]$,

$$\sup_{r \geq 0} |A_1(r)| \leq c \frac{\phi_1(\gamma)}{\psi(\gamma)} \quad \text{and} \quad \sup_{r \geq 0} |A_2(r) - \alpha_1 v(r, \mu)| \leq \frac{c}{\gamma} + c \left| \frac{\phi_1(\gamma)}{\gamma} - \alpha_1 \right|. \quad (53)$$

We have

$$\int_{(0, \infty)} \pi_1(\ell_1) \int_0^{\gamma \ell_1} e^{-s} ds \left( 1 - w(\ell_1 - s \gamma, r, \mu) \right) - \int_{(0, \infty)} \pi_1(\ell_1) \left( 1 - w(\ell_1, r, \mu) \right)$$

$$= \int_{(0, \infty)} \pi_1(\ell_1) e^{-\gamma \ell_1} (w(\ell_1, r, \mu) - 1)$$

$$+ \int_{(0, \infty)} \pi_1(\ell_1) \int_0^{\infty} e^{-s} ds \left( w(\ell_1, r, \mu) - w(\ell_1 - s \gamma, r, \mu) \right) 1_{\{s \leq \gamma \ell_1\}}.$$
It is then easy to get, using $(h_3)$ and $(35)$, that $\mathbb{P}$-a.s.
\[
\phi_2(\gamma) = \sup_{r \geq 0} \left| \int_{(0, \infty)} \pi_1(d\ell_1) \int_0^{\gamma \ell_1} e^{-s} ds \left( 1 - w(\ell_1 - \frac{s}{\gamma}, r, \mu) \right) - \int_{(0, \infty)} \pi_1(d\ell_1) \left( 1 - w(\ell_1, r, \mu) \right) \right|
\]
converges to 0 as $\gamma$ goes to infinity.

Recall that we assumed that $\lim_{\gamma \to \infty} \phi_1(\gamma) = +\infty$. Thus, there exist $\varepsilon_0 > 0$ and a finite constant $c > 0$ s.t. $\mathbb{P}$-a.s for all $\varepsilon \in (0, \varepsilon_0]$,
\[
\sup_{r \geq 0} \left| A_3(r) - \int_{(0, \infty)} \pi_1(d\ell_1) \left( 1 - w(\ell_1, r, \mu) \right) \right| \leq \frac{c}{\phi_1(\gamma)} + \phi_2(\gamma).
\]
(54)

Using $(53)$ and $(54)$, we get that there exists a deterministic function $\mathcal{R}$ s.t. $\mathbb{P}$-a.s
\[
\sup_{r \geq 0} \left| \phi_1(\gamma) \log(K_{\varepsilon}(\gamma, r, \mu)) - \alpha_1 v(r, \mu) - \int_{(0, \infty)} \pi_1(d\ell_1) \left( 1 - w(\ell_1, r, \mu) \right) \right| \leq \mathcal{R}(\varepsilon),
\]
where $\lim_{\varepsilon \to 0} \mathcal{R}(\varepsilon) = 0$, thanks to $(9)$ and $(20)$. \hfill \square

The previous results allow us to compute the following limit. We keep the same notation as in Lemma 3.5.

Lemma 3.10. Let $\varphi$ satisfying condition $(h_1)$–$(h_3)$. There exists a sub-sequence $(\varepsilon_j, j \in \mathbb{N})$ decreasing to 0, s.t. $\mathbb{P}$-a.s.
\[
\lim_{j \to \infty} \prod_{k=1}^{\infty} K_{\varepsilon_j}(A_{S_k^{\varepsilon_j}}, \rho_{S_k^{\varepsilon_j}}) = \exp - \int_0^{\infty} du \left( \alpha_1 v(u) + \int_{(0, \infty)} \pi_1(d\ell) \left( 1 - w(\ell, u, \mu) \right) \right).
\]

Proof. Notice that thanks to $(h_1)$, the functions $v$ and $(u, \mu) \mapsto w(\ell, u, \mu)$ are continuous and that for $r \geq \ell$, $v(r, \mu) = 0$ and $w(\ell, r, \mu) = 1$. The result is then a direct consequence of Corollary 3.8 and Lemma 3.9 \hfill \square

3.6 Proof of Theorem 3.2

Now we can prove the special Markov property in the case $\lim_{\gamma \to \infty} \phi_1(\gamma) = +\infty$.

Let $Z \in \mathcal{F}_\infty$ non-negative such that $\mathbb{E}[Z] < \infty$. Let $\varphi$ satisfying hypothesis of Theorem 3.2 $(h_1)$–$(h_3)$. We have, using notation of the previous sections
\[
\mathbb{E} \left[ Z \exp \left( -\sum_{i \in I} \varphi(A_{a_i}, \rho_{a_i}, \mathcal{S}^{-1}) \right) \right] = \lim_{j \to \infty} \mathbb{E} \left[ Z \exp \left( -\sum_{k=1}^{\infty} \varphi^* \left( A_{S_k^{\varepsilon_j}}, \rho_{S_k^{\varepsilon_j}}, \varphi^{k, \varepsilon_j} \right) \right) \right]
\]
\[
= \lim_{j \to \infty} \mathbb{E} \left[ Z \prod_{k=1}^{\infty} K_{\varepsilon_j}(A_{S_k^{\varepsilon_j}}, \rho_{S_k^{\varepsilon_j}}) \right]
\]
\[
= \mathbb{E} \left[ Z \exp {\int_0^{\infty} du \left( \alpha_1 v(u, \mu_u) + \int_{(0, \infty)} \pi_1(d\ell) \left( 1 - w(\ell, u, \mu_u) \right) \right) } \right],
\]
1459
where we used Lemma 3.4 and dominated convergence for the first equality, Lemma 3.5 for the second equality, Lemma 3.10 and dominated convergence for the last equality. By monotone class Theorem and monotonicity, we can remove hypothesis (h_1)–(h_3). To ends the proof of the first part, notice that \( \int_0^t du \left( \alpha_1 \nu(u) + \int_{(0,\infty)} \pi_1(d\ell) (1 - w(\ell,u)) \right) \) is \( \mathcal{F}_\infty \)-measurable and so this is \( \mathbb{P} \)-a.e. equal to the conditional expectation (i.e. the left hand side term of (34)).

\section{Law of the pruned exploration process}

Let \( \rho^{(0)} \) be the exploration process of a Lévy process with Laplace exponent \( \psi_0 \). The aim of this section is to prove Theorem 2.3.

\subsection{A martingale problem for \( \tilde{\rho} \)}

Let \( \tilde{\sigma} = \inf \{ t > 0, \tilde{\rho}_t = 0 \} \). In this section, we shall compute the law of the total mass process \( (\tilde{\rho}_{t \wedge \tilde{\sigma}}, 1), t \geq 0 \) under \( \mathbb{P}_\mu = \mathbb{P}_{\mu,0} \), using martingale problem characterization. We will first show how a martingale problem for \( \rho \) can be translated into a martingale problem for \( \tilde{\rho} \), see also [1]. Unfortunately, we were not able to use standard techniques of random time change, as developed in Chapter 6 of [23] and used for Poisson snake in [7], mainly because \( t^{-1} \left( \mathbb{E}_\mu [f(\rho_t)1_{\{m_i=0\}}] - f(\mu) \right) \) may not have a limit as \( t \) goes down to 0, even for exponential functionals.

Let \( F, K \in \mathcal{B}(\mathcal{M}_f(\mathbb{R}_+)) \) be bounded. We suppose that \( \mathbb{N} \left[ \int_0^\sigma |K(\rho_s)| \, ds \right] < \infty \), that for any \( \mu \in \mathcal{M}_f(\mathbb{R}_+) \), \( \mathbb{E}_\mu^* \left[ \int_0^\sigma |K(\rho_s)| \, ds \right] < \infty \) and that \( M_t = F(\rho_{t \wedge \tilde{\sigma}}) - \int_0^{t \wedge \tilde{\sigma}} K(\rho_s) \, ds \), for \( t \geq 0 \), defines an \( \mathcal{F} \)-martingale. In other words, if \( F \) belongs to the domain of the infinitesimal generator \( \mathcal{L} \) of \( \rho \), we have \( K = \mathcal{L} F \). We will see in the proof of Corollary 4.2 that these assumptions on \( F \) and \( K \) are in particular fulfilled for

\[ F(\nu) = e^{-c(\nu,1)} \quad K(\nu) = \psi(c)F(\nu). \]

Notice that we have

\[ |M_t| \leq \|F\|_\infty + \int_0^{t \wedge \tilde{\sigma}} |K(\rho_s)| \, ds \]

and thus \( \mathbb{E}_\mu^* \left[ \sup_{t \geq 0} |M_t| \right] < \infty \). Consequently, we can define for \( t \geq 0 \),

\[ N_t = \mathbb{E}_\mu^*[M_C,\mathcal{F}_t]. \]

\textbf{Proposition 4.1.} The process \( N = (N_t, t \geq 0) \) is an \( \mathcal{F} \)-martingale. And we have the representation formula for \( N_t \):

\[ N_t = F(\tilde{\rho}_{t \wedge \tilde{\sigma}}) - \int_0^{t \wedge \tilde{\sigma}} du \tilde{K}(\rho_u), \]

with

\[ \tilde{K}(\nu) = K(\nu) + \alpha_1 \mathbb{N} \left[ \int_0^\sigma K([\nu, \rho_s]) \, ds \right] + \int_{(0,\infty)} \pi_1(d\ell) \mathbb{E}_\mu^* \left[ \int_0^\sigma K([\nu, \rho_s]) \, ds \right]. \]
Proof. Notice that \( N = (N_t, t \geq 0) \) is an \( \mathcal{F} \)-martingale. Indeed, we have for \( t, s \geq 0 \),
\[
\mathbb{E}_\mu[N_{t+s}|\mathcal{F}_t] = \mathbb{E}_\mu[\mathbb{E}_\mu[M_{C_{t+s}}|\mathcal{F}_{t+s}]|\mathcal{F}_t] \\
= \mathbb{E}_\mu[M_{C_{t+s}}|\mathcal{F}_t] \\
= \mathbb{E}_\mu[\mathbb{E}_\mu[M_{C_{t+s}}|\mathcal{F}_s]|\mathcal{F}_t] \\
= \mathbb{E}_\mu[M_{C_t}|\mathcal{F}_t],
\]
where we used the optional stopping time Theorem for the last equality. To compute \( \mathbb{E}_\mu[M_{C_t}|\mathcal{F}_t] \),
we set \( N'_t = M_{C_t} + M'_{C_t} \), where for \( u \geq 0 \),
\[
M'_u = \int_0^{u \wedge \sigma} K(\rho_s)1_{[m_i \neq 0]} \, ds.
\]
Recall that \( C_0 = 0 \) \( \mathbb{P}_\mu \)-a.s. by Corollary 2.2. In particular, we get
\[
N'_t = M_{C_t} + M'_{C_t} = F(\rho_{C_t \wedge \sigma}) - \int_0^{C_t \wedge \sigma} K(\rho_s)1_{[m_i = 0]} \, ds \\
= F(\tilde{\rho}_{t \wedge \sigma}) - \int_0^{C_t \wedge \sigma} K(\rho_s) \, dA_s \\
= F(\tilde{\rho}_{t \wedge \sigma}) - \int_0^{t \wedge \sigma} K(\tilde{\rho}_u) \, du,
\]
where we used the time change \( u = A_t \) for the last equality. In particular, as \( \tilde{\sigma} \) is an \( \mathcal{F} \)-stopping

time, we get that the process \( (N'_t, t \geq 0) \) is \( \mathcal{F} \)-adapted. Since \( N_t = N'_t - \mathbb{E}_\mu[M'_{C_t}|\mathcal{F}_t] \),
we are left with the computation of \( \mathbb{E}_\mu[M'_{C_t}|\mathcal{F}_t] \).
We keep the notations of Section 3. We consider \( (\rho^i, m^i), i \in I \) the excursions of the process \( (\rho, m) \)
outside \( \{s, m_i = 0\} \) before \( \sigma \) and let \( (\alpha_i, \beta_i), i \in I \) be the corresponding interval excursions. In
particular we can write
\[
\int_0^{C_t \wedge \sigma} |K(\rho_s)|1_{[m_i \neq 0]} \, ds = \sum_{i \in I} \Phi(A_{\alpha_i}, \rho_{\alpha_i}, \rho^i),
\]
with
\[
\Phi(u, \mu, \rho) = 1_{\{u < t\}}\int_0^{\sigma(\rho)} |K([\mu, \rho_s])| \, ds,
\]
where \( \sigma(\rho) = \inf\{v > 0; \rho_v = 0\} \). We deduce from the second part of Theorem 3.2 that \( \mathbb{P}_\mu \)-a.s.
\[
\mathbb{E}_\mu \left[ \int_0^{C_t \wedge \sigma} |K(\rho_s)|1_{[m_i \neq 0]} \, ds \right] = \int_0^\tilde{\sigma} \int_{\{u < t\}} \hat{K}(\tilde{\rho}_u) \, du,
\]
(57)
with, \( \hat{K} \) defined for \( \nu \in \mathcal{M}(\mathbb{R}_+) \) by
\[
\hat{K}(\nu) = \alpha_1 \mathbb{N} \left[ \int_0^\sigma |K([\nu, \rho_s])| \, ds \right] + \int_{(0, \infty)} \pi_1(d\ell) \mathbb{E}_\nu^\ell \left[ \int_0^\sigma |K([\nu, \rho_s])| \, ds \right].
\]
Since \(\mathbb{E}_\mu \left[ \int_0^{C_1 + \sigma} |K(\rho_s)| \mathbf{1}_{[m, \infty)} \, ds \right] \leq \mathbb{E}_\mu \left[ \int_0^{\sigma} |K(\rho_s)| \, ds \right] < \infty\), we deduce from (57) that \(\mathbb{P}_\mu\)-a.s. \(d\mu\)-a.e. \(\mathbf{1}_{[\mu < \sigma]} \hat{K}(\hat{\rho}_u)\) is finite.

We define \(\bar{K} \in \mathcal{B}(\mathcal{M}_f(\mathbb{R}_+))\) for \(\nu \in \mathcal{M}_f(\mathbb{R}_+)\) by

\[
\bar{K}(\nu) = \alpha_1 \mathbb{E}_\mu \left[ \int_0^{\sigma} K([\nu, \rho_s]) \, ds \right] + \int_{(0, \infty)} \pi_1(d\ell) \mathbb{E}_\mu^\ast \left[ \int_0^{\sigma} K([\nu, \rho_s]) \, ds \right]
\]

(58)

if \(\bar{K}(\nu) < \infty\), or by \(\bar{K}(\nu) = 0\) if \(\bar{K}(\nu) = +\infty\). In particular, we have \(|\bar{K}(\nu)| \leq \bar{K}(\nu)\) and \(\mathbb{P}_\mu\)-a.s. \(\int_0^\sigma |\bar{K}(\hat{\rho}_u)| \, du\) is finite. Using the special Markov property once again (see (57)), we get that \(\mathbb{P}_\mu\)-a.s.,

\[
\mathbb{E}_\mu \left[ M'_{C_1} \big| \mathcal{F}_\infty \right] = \mathbb{E}_\mu \left[ \int_0^{C_1 + \sigma} K(\rho_s) \mathbf{1}_{[m, \infty)} \, ds \big| \mathcal{F}_\infty \right] = \int_0^{t + \sigma} \bar{K}(\hat{\rho}_u) \, du.
\]

Finally, as \(N_t = N'_t - \mathbb{E}_\mu \left[ M'_{C_1} \big| \mathcal{F}_\infty \right]\), this gives (55).

\[\square\]

**Corollary 4.2.** Let \(\mu \in \mathcal{M}_f(\mathbb{R}_+)\). The law of the total mass process \((\langle \hat{\rho}_t, 1 \rangle, \ t \geq 0)\) under \(\mathbb{P}_{\mu, 0}\) is the law of the total mass process \(\rho^{(0)}\) under \(\mathbb{P}_\mu^\ast\).

**Proof.** Let \(X = (X_t, \ t \geq 0)\) be under \(\mathbb{P}_\mu^\ast\) a Lévy process with Laplace transform \(\psi\) started at \(x > 0\) and stopped when it reached 0. Under \(\mathbb{P}_\mu\), the total mass process \((\langle \rho_t \wedge \sigma, 1 \rangle, \ t \geq 0)\) is distributed as \(X\) under \(\mathbb{P}_{\mu, 1}\). Let \(c > 0\). From Lévy processes theory, we know that the process \(e^{-cX_t} - \psi(c) \int_0^t e^{-cX_s} \, ds\), for \(t \geq 0\) is a martingale. We deduce from the stopping time Theorem that \(M = (M_t, \ t \geq 0)\) is an \(\mathcal{F}\)-martingale under \(\mathbb{P}_\mu\), where \(M_t = F(\rho_t \wedge \sigma) - \int_0^{t \wedge \sigma} K(\rho_s) \, ds\), with \(F, K \in \mathcal{B}(\mathcal{M}_f(\mathbb{R}_+))\) defined by \(F(\nu) = e^{-c(\nu, 1)}\) for \(\nu \in \mathcal{M}_f(\mathbb{R}_+)\) and \(K = \psi(c)F\). Notice \(K \geq 0\). We have by dominated convergence and monotone convergence.

\[
e^{-c(\mu, 1)} = \lim_{t \to \infty} \mathbb{E}_\mu[M_t] = \mathbb{E}_\mu[e^{-c(\rho_t, 1)}] - \psi(c) \mathbb{E}_\mu \left[ \int_0^{\sigma} e^{-c(\rho_s, 1)} \, ds \right].
\]

This implies that, for any \(\mu \in \mathcal{M}(\mathbb{R}_+)\), \(\mathbb{E}_\mu \left[ \int_0^{\sigma} |K(\rho_s)| \, ds \right]\) is finite. Using the Poisson representation, see Proposition 1.9, it is easy to get that

\[
\mathbb{N} \left[ \int_0^{\sigma} dt \ e^{-c(\rho_t, 1)} \right] = \frac{c}{\psi(c)}.
\]

(59)

In particular, it is also finite.

From Proposition 4.1, we get that \(N = (N_t, \ t \geq 0)\) is under \(\mathbb{P}_\mu\) an \(\mathcal{F}\)-martingale, where: for \(t \geq 0\),

\[
N_t = e^{-c(\rho_t \wedge \sigma, 1)} - \int_0^{t \wedge \sigma} \bar{K}(\hat{\rho}_u) \, du
\]
and \( \tilde{K} \) given by (56). We can compute \( \tilde{K} \):

\[
\tilde{K}(\nu) = \psi(c) e^{-c(\nu,1)} \left( 1 + \alpha_1 N \int_0^\sigma e^{-c(\nu,1)} \, ds \right) + \int_{(0,\infty)} \pi_1(d\ell) \mathbb{E}_\ell^\alpha \left( \int_0^\sigma e^{-c(\nu,1)} \, ds \right)
\]

\[
= \psi(c) e^{-c(\nu,1)} \left( 1 + \alpha_1 \frac{c}{\psi(c)} + \int_{(0,\infty)} \pi_1(d\ell) \int_0^\ell e^{-c\ell} \mathbb{E}_\ell^\alpha \left( \int_0^\sigma e^{-c(\nu,1)} \, ds \right) \right)
\]

\[
= e^{-c(\nu,1)} \left( \psi(c) + \alpha_1 c + \int_{(0,\infty)} \pi_1(d\ell)(1 - e^{-c\ell}) \right)
\]

\[
= \psi_0(c) e^{-c(\nu,1)},
\]

where we used (59) and the excursion decomposition for the second equality, and \( \psi_0 = \psi + \phi_1 \) for the last one.

Thus, the process \( (N_t, t \geq 0) \) with for \( t \geq 0 \)

\[
N_t = e^{-c(\hat{\rho}_{t,\sigma},1)} - \psi_0(c) \int_0^{t \wedge \hat{\sigma}} e^{-c(\hat{\rho}_{u,1})} \, du
\]

is under \( \mathbb{P}_\mu \) an \( \mathcal{F} \)-martingale.

Notice that \( \hat{\sigma} = \inf\{s \geq 0; \langle \hat{\rho}_{s}, 1 \rangle = 0\} \). Let \( X^{(0)} = (X^{(0)}_t, t \geq 0) \) be under \( \mathbb{P}^\alpha_x \) a Lévy process with Laplace transform \( \psi_0 \) started at \( x > 0 \) and stopped when it reached 0. The two non-negative càdlàg processes \( \langle \hat{\rho}_{t,\hat{\sigma}}, 1 \rangle, t \geq 0 \) and \( X^{(0)} \) solves the martingale problem: for any \( c \geq 0 \), the process defined for \( t \geq 0 \) by

\[
e^{-c Y_{t,\sigma'}} - \psi_0(c) \int_0^{t \wedge \sigma'} e^{-c Y_s} \, ds,
\]

where \( \sigma' = \inf\{s \geq 0; Y_s \leq 0\} \), is a martingale. From Corollary 4.4.4 in [23], we deduce that those two processes have the same distribution. To finish the proof, notice that the total mass process of \( \rho^{(0)} \) under \( \mathbb{P}_\mu^\alpha \) is distributed as \( X^{(0)} \) under \( \mathbb{P}^\alpha_{(\mu,1)} \).

\[ \square \]

4.2 Identification of the law of \( \hat{\rho} \)

To begin with, let us mention some useful properties of the process \( \hat{\rho} \).

Lemma 4.3. We have the following properties for the process \( \hat{\rho} \).

(i) \( \hat{\rho} \) is a càdlàg Markov process.

(ii) The sojourn time at 0 of \( \hat{\rho} \) is 0.

(iii) 0 is recurrent for \( \hat{\rho} \).

Proof. (i) This is a direct consequence of the strong Markov property of the process \( (\rho, m) \).

(ii) We have for \( r > 0 \), with the change of variable \( t = A_s \), a.s.

\[
\int_0^r 1_{\{\hat{\rho}_s = 0\}} \, dt = \int_0^r 1_{\{\rho_{C_r} = 0\}} \, dt = \int_0^{C_r} 1_{\{\rho_s = 0\}} \, dA_s = \int_0^{C_r} 1_{\{\rho_s = 0\}} \, ds = 0,
\]

1463
as the sojourn time of $\rho$ at 0 is 0 a.s.

(iii) Since $\bar{\sigma} = A_0$ and $\sigma < +\infty$ a.s., we deduce that 0 is recurrent for $\bar{\rho}$ a.s.

Since the processes $\bar{\rho}$ and $\rho^{(0)}$ are both Markov processes, to show that they have the same law, it is enough to show that they have the same one-dimensional marginals. We first prove that result under the excursion measure.

**Proposition 4.4.** For every $\lambda > 0$ and every non-negative bounded measurable function $f$,

$$
\mathbb{N} \left[ \int_0^{\bar{\sigma}} e^{-\lambda t - (\bar{\rho}, f)} \, dt \right] = \mathbb{N} \left[ \int_0^{\sigma^{(0)}} e^{-\lambda t - (\rho^{(0)}, f)} \, dt \right].
$$

**Proof.** On one hand, we compute, using the definition of the pruned process $\bar{\rho}$,

$$
\mathbb{N} \left[ \int_0^{\bar{\sigma}} e^{-\lambda t - (\bar{\rho}, f)} \, dt \right] = \mathbb{N} \left[ \int_0^{\sigma} e^{-\lambda A_0} e^{-(\rho,f) \, dA_0} \right] = \mathbb{N} \left[ \int_0^{\sigma} e^{-\lambda A_u} e^{-(\rho,f) \, 1_{\{m_u=0\}} \, du} \right].
$$

We now make the change of variable $t = A_u$ to get

$$
\mathbb{N} \left[ \int_0^{\bar{\sigma}} e^{-\lambda t - (\bar{\rho}, f)} \, dt \right] = \mathbb{N} \left[ \int_0^{\sigma} e^{-\lambda A_u} e^{-(\rho,f) \, dA_u} \right]
$$

By a time reversibility argument, see Lemma 1.7 we obtain

$$
\mathbb{N} \left[ \int_0^{\bar{\sigma}} e^{-\lambda t - (\bar{\rho}, f)} \, dt \right] = \mathbb{N} \left[ \int_0^{\sigma} e^{-\lambda (A_0 - A_u)} \, du \right]
$$

where we applied Lemma 2.1(i) for the last equality. Now, using Proposition 1.9 we have

$$
\mathbb{N} \left[ \int_0^{\bar{\sigma}} e^{-\lambda t - (\bar{\rho}, f)} \, dt \right] = \int_0^{\infty} da \, e^{-aa} \mathbb{M}_a \left[ 1_{\{m=0\}} e^{-(v,f) - \psi^{-1}(\lambda)(\mu,1)} \right].
$$

Using usual properties of point Poisson measures, we have, with $c = \alpha_1 + \int_{0,\infty} \ell \, \pi_1(d\ell)$,

$$
\mathbb{M}_a \left[ 1_{\{m=0\}} F(\mu, v) \right] = e^{-c} \mathbb{M}_a \left[ F(\mu^0, v^0) \right].
$$
where with the notations of Proposition 3.1.3 in [22] for any \( f \in \mathcal{B}_+(\mathbb{R}_+) \)

\[
\langle \mu^0, f \rangle = \int \mathcal{N}_0(dx \, d\ell \, du) \textbf{1}_{[0,a]}(x) u \ell f(x) + \beta \int_0^a f(r) \, dr,
\]

\[
\{v^0, f\} = \int \mathcal{N}_0(dx \, d\ell \, du) \textbf{1}_{[0,a]}(x)(1-u) \ell f(x) + \beta \int_0^a f(r) \, dr.
\]

As \( \alpha_0 = \alpha + c \), we have

\[
\mathbb{N} \left[ \int_0^\sigma e^{-\lambda t-(\hat{\rho}_t, f)} \, dt \right] = \int_0^\infty da \, e^{-a_0 a} \mathbb{M}_a \left[ e^{-(v^0, f)-\psi_0^{-1}(\lambda)(\mu^0, 1)} \right].
\]

Proposition 3.1.3 in [22] directly implies that the left-hand side of the previous equality is equal to

\[
\mathbb{N} \left[ \int_0^{\sigma^{(0)}} e^{-(\eta^{(0)}, f)-\psi_0^{-1}(\lambda)(\rho^{(0)}, t)} \, dt \right].
\]

On the other hand, similar computations as above yields that this quantity is equal to \( \mathbb{N} \left[ \int_0^{\sigma^{(0)}} e^{-\lambda t-(\hat{\rho}_t, f)} \, dt \right] \). This ends the proof. \( \square \)

Now, we prove the same result under \( \mathbb{P}^{\mu}_{\mu, 0} \), that is:

**Proposition 4.5.** For every \( \lambda > 0 \), \( f \in \mathcal{B}_+(\mathbb{R}_+) \) bounded and every finite measure \( \mu \),

\[
\mathbb{E}^{\mu}_{\mu, 0} \left[ \int_0^\sigma e^{-\lambda t-(\hat{\rho}_t, f)} \, dt \right] = \mathbb{E}^{\mu}_{\mu} \left[ \int_0^{\sigma^{(0)}} e^{-\lambda t-(\rho^{(0)}, f)} \, dt \right].
\]

**Proof.** From the Poisson representation, see Lemma 1.6 and using notations of this Lemma and of (28) we have

\[
\mathbb{E}^{\mu}_{\mu, 0} \left[ \int_0^\sigma e^{-\lambda t-(\hat{\rho}_t, f)} \, dt \right] = \mathbb{E}^{\mu}_{\mu, 0} \left[ \int_0^\sigma e^{-\lambda A_{\lambda t^+}-(\rho_{\lambda t}, f)} \, dA_u \right]
\]

\[
= \mathbb{E}^{\mu, 0}_{\mu, 0} \left[ \sum_{i \in I} e^{-\lambda A_{\lambda t^+}-(k_{i-}, f)} \int_{0}^{\sigma_i} e^{-\lambda A_{\lambda t^+}-(k_{i-}, f)} \, dA_s \right],
\]

where the function \( f_i \) is defined by \( f_i(x) = f(H_{\lambda t}^i + x) \) and \( H_{\lambda t}^i = H(k_{i, \lambda}) \) is the maximal element of the closed support of \( k_{i, \lambda} \) (see (12)). We recall that \(-I\) is the local time at 0 of the reflected process \( X - I \), and that \( \tau_r = \inf \{ s; -I_s > r \} \) is the right continuous inverse of \(-I\). From excursion formula, and using the time change \(-I_s = r \) (or equivalently \( \tau_r = s \)), we get

\[
\mathbb{E}^{\mu}_{\mu, 0} \left[ \int_0^\sigma e^{-\lambda t-(\hat{\rho}_t, f)} \, dt \right] = \mathbb{E}^{\mu}_{\mu, 0} \left[ \int_0^{\tau_r} d(-I_s) e^{-(k_{i-}, f)-\lambda A_{\lambda t}^{i} G(-I_s)} \right]
\]

\[
= \mathbb{E}^{\mu, 0}_{\mu, 0} \left[ \int_0^{\tau_r} dr \, e^{-(k_{i-}, f)-\lambda A_{\lambda r}^{i} G(r)} \right],
\]

(60)
where the function $G(r)$ is given by

$$G(r) = \mathbb{N} \left[ \int_0^r e^{-\lambda A_t} \, dA_t \right] = \mathbb{N} \left[ \int_0^r e^{-\lambda t - (\check{\rho}_t, f_t)} \, dt \right].$$

The same kind of computation gives

$$\mathbb{E}^\mu_\ast \left[ \int_0^{\sigma(0)} e^{-\lambda t - (\check{\rho}_t, f)} \, dt \right] = \mathbb{E} \left[ \int_0^{(\mu, 1)} d r \, e^{-\lambda \zeta_t} G(0)(r) \right]$$

where the function $G(0)$ is defined by

$$G(0)(r) = \mathbb{N} \left[ \int_0^{\sigma(0)} e^{-\lambda s - (\check{\rho}_s, f)} \, ds \right]$$

and $\sigma(0)$ is the right-continuous inverse of the infimum process $-I(0)$ of the Lévy process with Laplace exponent $\psi_0$.

Proposition 4.4 says that the functions $G$ and $G(0)$ are equal. Moreover, as the total mass processes have the same law (see Corollary 4.2), we know that the proposition is true for $f$ constant. And, for $f$ constant, the functions $G$ and $G(0)$ are also constant. Therefore, we have for $f$ constant equal to $c \geq 0$,

$$\mathbb{E}^\mu_\ast \left[ \int_0^{(\mu, 1)} d r \, e^{-c((\mu, 1) - r)} e^{-\lambda A_t} \right] = \mathbb{E} \left[ \int_0^{(\mu, 1)} d r \, e^{-c((\mu, 1) - r)} e^{-\lambda \tau} \right].$$

As this is true for any $c \geq 0$, uniqueness of the Laplace transform gives the equality

$$\mathbb{E}^\mu_\ast \left[ e^{-\lambda A_t} \right] = \mathbb{E} \left[ e^{-\lambda \tau} \right] \quad dr \text{ a.e.}$$

In fact this equality holds for every $r$ by right-continuity.

Finally as $G = G(0)$, we have thanks to (60) and (61), that, for every bounded non-negative measurable function $f$,

$$\int_0^{(\mu, 1)} d r \, e^{-(k, \mu, f) \mathbb{E}^\mu_\ast \left[ e^{-\lambda A_t} \right]} G(r) = \int_0^{(\mu, 1)} d r \, e^{-(k, \mu, f) \mathbb{E} \left[ e^{-\lambda \tau} \right]} G(0)(r)$$

which ends the proof.

\[ \square \]

**Corollary 4.6.** The process $\check{\rho}$ under $\mathbb{P}_\ast^\mu_0$ is distributed as $\rho(0)$ under $\mathbb{P}_\mu^\ast$.

**Proof.** Let $f \in \mathcal{B}_+(\mathbb{R})$ bounded. Proposition 4.5 can be re-written as

$$\int_0^{+\infty} e^{-\lambda t} \mathbb{E}^\mu_\ast \left[ e^{-(\check{\rho}_t, f)} 1_{t \leq \tilde{\sigma}} \right] \, dt = \int_0^{+\infty} e^{-\lambda t} \mathbb{E}^\mu \left[ e^{-(\rho_t, f)} 1_{t \leq \sigma(0)} \right] \, dt.$$

By uniqueness of the Laplace transform, we deduce that, for almost every $t > 0$,

$$\mathbb{E}^\mu_\ast \left[ e^{-(\check{\rho}_t, f)} 1_{t \leq \tilde{\sigma}} \right] = \mathbb{E}^\mu \left[ e^{-(\rho_t, f)} 1_{t \leq \sigma(0)} \right].$$
Proposition 5.1. For all non-negative \( \sigma \) unique non-negative solution of the equation

\[ \psi(\sigma) = \sigma + \psi_0(\sigma) \]

Thanks to (15), we get \( N_{bb} \) since \( \psi(\sigma) \) can compute the joint law of \( \sigma \). Notice that \( \psi(\sigma) \) is increasing and continuous, we get the result.

Proof of Theorem 2.3. \( 0 \) is recurrent for the Markov càdlàg processes \( \tilde{\rho} \) and \( \rho^{(0)} \). These two processes have no sojourn time at 0, and when killed on the first hitting time of 0, they have the same law, thanks to Lemma 4.6. From Theorem 4.2 of [17], Section 5, we deduce that \( \tilde{\rho} \) under \( P_{\mu,0} \) is distributed as \( \rho^{(0)} \) under \( P_{\mu} \).

5 Law of the excursion lengths

Recall \( \tilde{\sigma} = \int_0^\sigma 1_{(m_i = 0)} \, ds \) denotes the length of the excursion of the pruned exploration process. We can compute the joint law of \( (\tilde{\sigma}, \sigma) \). This will determine uniquely the law of \( \tilde{\sigma} \) conditionally on \( \sigma = r \).

Proposition 5.1. For all non-negative \( \gamma, \kappa \), the value \( v \) defined by

\[ v = N \left[ 1 - e^{-\gamma(\sigma - \kappa \tilde{\sigma})} \right] \]

is the unique non-negative solution of the equation

\[ \psi_0(v) = \kappa + \psi_0(\gamma). \]

Proof. Excursion theory implies that the special Markov property, Theorem 3.2, also holds under \( N \), with the integration of \( u \) over \([0, \tilde{\sigma} = A_\sigma]\) instead of \([0, \infty)\). Taking \( \phi(\tau) = \psi(\gamma)\sigma \), we have

\[ v = N \left[ 1 - e^{-\kappa \tilde{\sigma} - \psi(\gamma)\sigma} \right] = N \left[ 1 - e^{-(\kappa + \psi(\gamma))\tilde{\sigma} - \psi(\gamma)\sigma} \int_0^\sigma 1_{[m_i = 0]} \, ds \right] = N \left[ 1 - e^{-(\kappa + \psi(\gamma))\tilde{\sigma} - \psi(\gamma)\sigma} \left( a_1 N[1 - e^{-\psi(\gamma)\sigma}] + \int_{(0, \infty)} \pi_1(dt)(1 - e^{-\gamma(\sigma - \kappa \tilde{\sigma})}) \right) \right]. \]

Notice that \( \sigma \) under \( P^{\psi}_\ell \) is distributed as \( \tau_\ell \), the first time for which the infimum of \( X \), started at 0, reaches \(-\ell\). Since \( \tau_\ell \) is distributed as a subordinator with Laplace exponent \( \psi^{-1} \) at time \( \ell \), we have

\[ P^{\psi}_\ell[e^{-\psi(\gamma)\sigma}] = E[e^{-\psi(\gamma)\tau_\ell}] = e^{-\ell \gamma}. \]

Thanks to (15), we get \( N[1 - e^{-\psi(\gamma)\sigma}] = \gamma \). We deduce that

\[ v = N \left[ 1 - e^{-(\kappa + \psi(\gamma))\tilde{\sigma} - \psi_0(\sigma)} \right] = N \left[ 1 - e^{-(\kappa + \psi_0(\gamma))\tilde{\sigma} + \psi_0(\sigma)} \right] = \psi_0^{-1}(\kappa + \psi_0(\gamma)). \]

Since \( \psi_0 \) is increasing and continuous, we get the result.
6 Appendix

We shall present in a first subsection, how one can extend the construction of the Lévy snake from [22] to a weighted Lévy snake, when the height process may not be continuous and the lifetime process is given by the total mass of the exploration process (instead of the height of the exploration process in [22]). Then, using this construction, we can define in a second subsection a general Lévy snake when the height process is not continuous.

6.1 Weighted Lévy snake

Let $D$ be a distance on $\mathcal{M}(\mathbb{R}_+)$ which defines the topology of weak convergence. Let us recall that $(\mathcal{M}(\mathbb{R}_+), D)$ is a Polish space, see [19], section 3.1.

Let $E$ be a Polish space, whose topology is defined by a metric $\delta$, and $\partial$ be a cemetery point added to $E$. Let $\mathcal{W}_x$ be the space of all $E$-valued weighted killed paths started at $x \in E$. An element $\tilde{w} = (\mu, w)$ of $\mathcal{W}_x$ is a mass measure $\mu \in \mathcal{M}(\mathbb{R}_+)$ and a càd-làg mapping $w : [0, (\mu, 1)) \to E$ s.t. $w(0) = x$. By convention the point $x$ is also considered as a weighted killed path with mass measure $\mu = 0$. We set $\mathcal{W} = \bigcup_{x \in E} \mathcal{W}_x$ and equip $\mathcal{W}$ with the distance

$$d((\mu, w), (\mu', w')) = \delta(w(0), w'(0)) + D(\mu, \mu') + \int_0^\wedge (\mu, 1) \land (\mu', 1) dt \left( d_t(w_{\leq t}, w'_{\leq t}) \wedge 1 \right) + |\langle \mu, 1 \rangle - \langle \mu', 1 \rangle|, \quad (62)$$

where $d_t$ is the Skorohod metric on the space $\mathcal{D}([0, t], E)$ and $w_{\leq t}$ denote the restriction of $w$ to the interval $[0, t]$. Notice $d$ is a distance on $\mathcal{W}$. Indeed, we have that:

- $d$ is symmetric.
- $d((\mu, w), (\mu', w')) = 0$ implies $D(\mu, \mu') = 0$, that is $\mu = \mu'$, and then $w = w'$ a.e. on $[0, (\mu, 1))$.
- $d$ satisfies the triangular inequality. We have for $(\mu, w), (\mu', w')$ and $(\mu'', w'') \in \mathcal{W}$:

$$d((\mu, w), (\mu', w')) \leq \delta(w(0), w''(0)) + \delta(w'(0), w''(0)) + D(\mu, \mu') + D(\mu'', \mu') + \int_0^\wedge (\mu, 1) \land (\mu', 1) dt \left( d_t(w_{\leq t}, w''_{\leq t}) \wedge 1 \right) + |\langle \mu, 1 \rangle - \langle \mu', 1 \rangle|$$

$$\leq d((\mu, w), (\mu'', w'')) + d((\mu'', w''), (\mu', w')) + (\langle \mu, 1 \rangle \land (\mu', 1) - (\mu, 1) \land (\mu'', 1) + (\mu, 1) \land (\mu'', 1) \land (\mu', 1) + |\langle \mu, 1 \rangle - \langle \mu', 1 \rangle|$$

$$\leq d((\mu, w), (\mu'', w'')) + d((\mu'', w''), (\mu', w')).$$
If we check that (\(\mathcal{M}_f(\mathbb{R}_+), D\)) is complete, we get that \(\mu_n\) converges to a limit say \(\mu\). If \(\mu = 0\), the result is clear. If not, for any \(\varepsilon > 0\) small enough, for \(n, n'\) large enough so that \(\langle \mu_n, 1 \rangle \wedge \langle \mu_n', 1 \rangle \geq \langle \mu, 1 \rangle - \varepsilon\) we deduce from (62) that, for \(\nu_{n \leq t}, n \geq n_\varepsilon\) large enough, \((w_{n \leq t}, n \geq n_\varepsilon)\) is a Cauchy sequence in \(\mathbb{D}([0, t\varepsilon), E)\) and hence converge to a limit \(w_{\leq t}\). Since this holds for any \(\varepsilon > 0\) small enough, we deduce that \(w\) is well defined on \([0, \langle \mu, 1 \rangle]\) and that \(w_n\) converges to \(w\) on any \(\mathbb{D}([0, t), E)\) for \(t < \langle \mu, 1 \rangle\). We deduce again from (62) that \((\langle \mu_n, w_n\rangle, n \in \mathbb{N})\) converges to \((\mu, w)\). We check that \((\mathcal{W}, d)\) is separable. Let \((\mu_n, n \in \mathbb{N})\) a dense subset of \((\mathcal{M}_f(\mathbb{R}_+), D)\), and for each \(n\), let \((w_{n,m}, m \in \mathbb{N})\) a dense subset of \(\mathbb{D}([0, \langle \mu_n, 1 \rangle], E), d_{\langle \mu_n, 1 \rangle}\). Then, it is easy to check that \(((\mu_n, w_{n,m}); n, m \in \mathbb{N})\) is a dense subset of \((\mathcal{W}, d)\).

Thus the space \((\mathcal{W}, d)\) is a Polish space.

We shall write \(\mu_\bar{w}\) instead of \(\mu\) when \(\bar{w} = (\mu, w)\). Recall (12). We consider a family of probability measures \(\tilde{\Pi}_{x,\mu}\), for \(x \in E\) and the mass measure \(\mu \in \mathcal{M}_f(\mathbb{R}_+)\) on \(\mathcal{W}_x\), s.t.

a) \(\mu_\bar{w} = \mu, \tilde{\Pi}_{x,\mu}(d\bar{w})\text{-a.s.}\);

b) \(w(0) = x, \tilde{\Pi}_{x,\mu}(d\bar{w})\text{-a.s.}\);

c) \(w\) has no fixed discontinuity: for all \(s \in [0, \langle \mu, 1 \rangle]\), \(\tilde{\Pi}_{x,\mu}(w(s) = w(s)) = 1\);

d) If \(H(\mu) < \infty\), then \(w(\langle \mu, 1 \rangle-)\) exists \(\tilde{\Pi}_{x,\mu}(d\bar{w})\text{-a.s.}\);

e) If \(H(\mu) < \infty\) and \(v \in \mathcal{M}_f(\mathbb{R}_+)\), then under \(\tilde{\Pi}_{x,\mu,v}\), \((w(r), r \in [0, \langle \mu, 1 \rangle])\) is distributed as \((w(r), r \in [0, \langle \mu, 1 \rangle])\) under \(\tilde{\Pi}_{x,\mu}\) and, conditionally on \((w(r), r \in [0, \langle \mu, 1 \rangle]), (w(r + \langle \mu, 1 \rangle), r \in [0, \langle v, 1 \rangle])\) is distributed as \((w(r), r \in [0, \langle v, 1 \rangle])\) under \(\tilde{\Pi}_{w(\langle \mu, 1 \rangle-),v}\).

The last property corresponds to the Markov property conditionally on the mass measure. We shall assume that the mapping \((x, \mu) \mapsto \tilde{\Pi}_{x,\mu}\) is measurable.

Let \(\rho\) be an exploration process starting at \(\mu\). We set \(Y_t = (\rho_t, 1)\). Recall that \((Y_t, t \geq 0)\) is distributed as a Lévy process with Laplace exponent \(\psi\) started at \((\mu, 1)\). For \(0 \leq s < t\), we set \(J_{s,t} = \inf_{0 \leq u \leq t} Y_u\) and \(\rho_{s,t} = k_{Y_t - J_{s,t}} \rho_s = k_{Y_t - J_{s,t}} \rho_t\), the last equality being a consequence of the construction of the exploration process. We also define \(\hat{\rho}_t^{(s)}\) as the unique measure \(v\) s.t. \([\rho_{s,t}, v] = \rho_t\).

Conditionally on \(\rho\), we define a probability transition semi-group \(R_{s,t}^\rho\) on \(\mathcal{W}_x\) as follows: for \(0 \leq s < t\) s.t. \(J_{s,t} < Y_s\) or \(w(\langle \rho_s, 1 \rangle-)\) exists and \(\mu_\bar{w} = \rho_s\), under \(R_{s,t}^\rho(\bar{w}, d\bar{w}')\) we have

i) \(\mu_\bar{w}' = \rho_t\),

ii) \((w'(r), r \in [0, \langle \rho_{s,t}, 1 \rangle))) = (w(r), r \in [0, \langle \rho_{s,t}, 1 \rangle)))\),

iii) \((w'(r), r \in [\langle \rho_{s,t}, 1 \rangle, \langle \rho_t, 1 \rangle))\) is distributed according to \(\tilde{\Pi}_{w(\langle \rho_{s,t}, 1 \rangle-),\hat{\rho}_t^{(s)}}\).
In (iii), by convention, if \( \rho_{s,t} = 0 \), then \( w(\langle \rho_{s,t}, 1 \rangle) = x \). Notice that for fixed \( s < t \), a.s. \( J_{s,t} < Y_s \) so that, with the previous convention \( w(\langle \rho_{s,t}, 1 \rangle) \) is a.s. well defined. Notice that if \( (\rho_s, w) \) is distributed as \( \Pi_{x,\rho_s} \) then \( (\rho_t, w') \) is distributed as \( \Pi_{x,\rho_t} \) thanks to condition \( e) \) on \( \Pi \). Thus we can use the Kolmogorov extension theorem to get that there exists a unique probability measure \( \mathbb{P}_{(\mu,w)} \) on \( (\mathcal{W})_{\mathbb{R}^+} \) s.t. for \( 0 = s_0 < s_1 < \cdots < s_n \),

\[
\mathbb{P}_{(\mu,w)}(W'_{s_0} \in A_0, \rho_{s_0} \in B_0, \ldots , W'_{s_n} \in A_n, \rho_{s_n} \in B_n)
= \mathbb{P}_{\mu} \left[ \mathbf{1}_{\{\rho_{s_0} \in B_0, \ldots , \rho_{s_n} \in B_n\}} \mathbf{1}_{\{w \in A_0\}} \int_{A_1 \times \cdots \times A_n} R^\rho_{s_0,s_1}(w, dw_{s_1}) \cdots R^\rho_{s_{n-1},s_n}(w_{s_n-1}, dw_{s_n}) \right].
\]

We set \( \tilde{W}_s = (\rho_s, W'_s) \). Notice that \( W'_s(r) = W'_t(r) \) for \( r \in [0, (\langle \rho_{s,t}, 1 \rangle) \) and thus that

\[
d(w_s, W_t) \leq D(\rho_s, \rho_t) + |Y_s - Y_t - J_{s,t}|.
\]

Since \( \rho \) and \( Y \) are \( \mathbb{P}_{\mu} \)-a.s. càdlàg, this implies that the mapping \( s \to \tilde{W}_s \) is \( \mathbb{P}_{(\mu,w)} \)-a.s. càdlàg on \( [0, \infty) \cap \mathbb{Q} \). Hence there is a unique càdlàg extension to the positive real line, we shall still denote by \( \mathbb{P}_{(\mu,w)} \) a time-homogeneous Markov process living in \( \mathcal{D}(\mathbb{R}_+, \mathcal{M}_f(\mathbb{R}_+) \times \mathcal{W}) \). We call this distribution the distribution of the weighted Lévy snake associated with \( \Pi \).

We set \( \mathcal{M}_f^0(\mathbb{R}_+) \) the set of \( \mu \in \mathcal{M}_f(\mathbb{R}_+) \) such that \( \text{supp} (\mu) = [0, H(\mu)] \) if \( H(\mu) < +\infty \) and \( \text{supp} (\mu) = (0, H(\mu)) \) if \( H(\mu) = +\infty \). We the define \( \Theta_x \) as the set of all pairs \( (\mu, w) \in \mathcal{W} \) such that \( \mu \in \mathcal{M}_f^0(\mathbb{R}_+) \), \( w(0) = x \) and at least one of the following three properties hold:

(i) \( \mu(H(\mu)) = 0 \);

(ii) \( w(\langle \mu, 1 \rangle) \) exists;

(iii) \( H(\mu) = +\infty \).

We denote by \( (\mathcal{F}_s, s \geq 0) \) the canonical filtration on \( \mathcal{D}(\mathbb{R}_+, \mathcal{M}_f(\mathbb{R}_+) \times \mathcal{W}) \). One can readily adapt the proofs of Propositions 4.1.1 and 4.1.2 of [22] to get the following result.

**Theorem 6.1.** The process \( (\tilde{W}_s, s \geq 0; \mathbb{P}_{(\mu,w)}), (\mu, w) \in \Theta_x \) is a càdlàg Markov process in \( \Theta_x \) and is strong Markov with respect to the filtration \( (\mathcal{F}_s, s \geq 0) \).

Let us remark that, when the family of probability measures \( \Pi_{x,\mu} \) is just the law of a homogeneous Markov process \( \xi \) starting at \( x \) and stopped at time \( \langle \mu, 1 \rangle \), the previous construction gives a snake with spatial motion \( \xi \) and lifetime process \( X - I \), which is the total mass of the exploration process. Notice that in [22] the lifetime process is given by the height of the exploration process.

### 6.2 The general Lévy snake

However, we need some dependency between the spatial motion and the exploration process \( \rho \) in order to recover the usual Lévy snake from the weighted Lévy snake. Informally, we keep the spatial motion from moving when time \( t \) is “on a mass” of \( \rho_t \). This idea can be compared to a subordination and has already been used in the snake framework by Bertoin, Le Gall and Le Jan in [16] in order to construct a kind of Lévy snake from the usual Brownian snake.
Let $\Pi_x$ be the distribution of $\xi$ a càdlàg Markov process taking values in $E$ with no fixed discontinuities and starting at $x$, such that the mapping $x \mapsto \Pi_x$ is measurable. Recall (13) and set $\tilde{\mu}_r = k_{\langle \mu, 1 \rangle - r} \mu$ for $r \in [0, \langle \mu, 1 \rangle)$. We define $\tilde{\Pi}_{x, \mu}$ as the distribution of $(\mu, w)$ with $w = (\xi_{H(\tilde{\mu}_r)}, r \in [0, \langle \mu, 1 \rangle))$ under $\Pi_x$. Notice that $\xi_r' = w((\mu, 1_{[0, r']}))$ for $r' \in [0, H(\mu))$. In particular, $w$ is on constant on intervals $\left(\mu([0, r]), \mu([0, r])\right)$ which corresponds to the atoms of $\mu$.

We have that $\tilde{\Pi}$ satisfies condition a)-e).

Let $((\rho_s, W'_s), s \geq 0)$ be the corresponding weighted Lévy snake. For $s \geq 0$, $r \geq 0$, we set $W_s(r) = W'_s((\rho_s, 1_{[0, r]}))$. When $H$ is continuous, the process $((\rho_s, W_s), s \geq 0)$ is the Lévy snake defined in Section 4 of [22] with underlying motion $\xi$. As a consequence of Theorem 6.1 we get that the (general) Lévy snake is strong Markov.

**Proposition 6.2.** The process $((\rho_s, W_s), s \geq 0; \mathbb{P}_{(\mu, w)}, (\mu, w) \in \Theta_x)$ is a càdlàg Markov process in $\Theta_x$ and is strong Markov with respect to the filtration $(\mathcal{F}_{s+}, s \geq 0)$.

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**References**

[1] R. ABRAHAM and J.-F. DELMAS. Feller property and infinitesimal generator of the exploration process. *J. Theor. Probab.*, 20(3):355–370, 2007. [MR2324536](https://doi.org/10.1007/s10959-007-0037-5)

[2] R. ABRAHAM and J.-F. DELMAS. Fragmentation associated to Lévy processes using snake. *Probab. Th. Rel. Fields*, 141:113–154, 2008. [MR2372967](https://doi.org/10.1007/s00440-007-0061-7)

[3] R. ABRAHAM and J.-F. DELMAS. Williams’ decomposition of the Lévy continuous random tree and simultaneous extinction probability for populations with neutral mutations. *Stoch. Proc. and Appl.*, 119:1124–1143, 2009. [MR2508567](https://doi.org/10.1016/j.spa.2009.05.001)

[4] R. ABRAHAM and J.-F. DELMAS. Changing the branching mechanism of a continuous state branching process using immigration. *Ann. Inst. Henri Poincaré*, 45:226–238, 2009. [MR2500236](https://doi.org/10.1214/09-AIHP264)

[5] R. ABRAHAM and J.-F. DELMAS. A continuum-tree-valued Markov process. *arxiv:0904.4175*, 2009.

[6] R. ABRAHAM and J.-F. DELMAS and H. HE. Pruning Galton-Watson Trees and Tree-valued Markov Processes. *In preparation*

[7] R. ABRAHAM and L. SERLET. Poisson snake and fragmentation. *Elect. J. of Probab.*, 7, 2002. [MR1943890](https://doi.org/10.1214/EJP.v7-120)

[8] D. ALDOUS. The continuum random tree II: an overview. In *Proc. Durham Symp. Stochastic Analysis*, pages 23–70. Cambridge univ. press edition, 1990. [MR1166406](https://doi.org/10.1017/CBO9780511626555.005)
[9] D. ALDOUS. The continuum random tree I. *Ann. Probab.*, 19(1):1–28, 1991. MR1085326

[10] D. ALDOUS. The continuum random tree III. *Ann. Probab.*, 21(1):248–289, 1993. MR1207226

[11] D. ALDOUS and J. PITMAN. Tree-valued Markov chains derived from Galton-Watson processes. *Ann. Inst. Henri Poincaré*, 34:637–686, 1998. MR1641670

[12] J. BERTOIN. *Lévy processes*. Cambridge University Press, Cambridge, 1996. MR1406564

[13] J. BERTOIN. *Random fragmentation and coagulation processes*. Cambridge University Press, Cambridge, 2006. MR2253162

[14] J. BERTOIN. The structure of the allelic partition of the total population for Galton-Watson processes with neutral mutations. *Ann. Probab.*, 37(4):1502–1523, 2009. MR2546753

[15] J. BERTOIN. A limit theorem for trees of alleles in branching processes with rare neutral mutations. *Stoch. Proc. and their Appl.*, 120(5):678–697, 2010. MR2603059

[16] J. BERTOIN, J.-F. LE GALL, and Yves LE JAN. Spatial branching processes and subordination. *Canad. J. of Math.*, 49(1):24–54, 1997. MR1437199

[17] R.M. BLUMENTHAL. *Excursions of Markov processes*. Birkhäuser, Boston, 1992. MR1138461

[18] D. J. DALEY and D. VERE-JONES. *An introduction to the theory of point processes. Vol. II*. Probability and its Applications (New York). Springer-Verlag, New York, second edition, 2008. General theory and structures. MR2371524

[19] D. A. DAWSON. Measure-valued Markov processes. In *École d’été de probabilité de Saint-Flour 1991*, volume 1541 of *Lect. Notes Math.*, pages 1–260. Springer Verlag, Berlin, 1993. MR1242575

[20] J.-F. DELMAS. Height process for super-critical continuous state branching process. *Markov Proc. and Rel. Fields.*, 14(2):309–326, 2008. MR2437534

[21] T. DUQUESNE and J.F. LE GALL. Probabilistic and fractal aspects of Lévy trees. *Probab. Th. rel. Fields*, 131:553–603, 2005. MR2147221

[22] T. DUQUESNE and J.-F. LE GALL. *Random trees, Lévy processes and spatial branching processes*, volume 281. Astérisque, 2002. MR1954248

[23] S. N. ETHIER and T. G. KURTZ. *Markov processes*. Wiley, 1986. MR0838085

[24] S. N. EVANS. *Probability and real trees*. Lectures from the 35th Summer School on Probability Theory held, Saint-Flour, 2005. Lecture Notes in Mathematics, 1920. Springer, Berlin, 2008. MR2351587

[25] M. JIRINA. Stochastic branching processes with continuous state space. *Czech. Math. J.*, 83(8):292–312, 1958. MR0101554

[26] A. LAMBERT. The allelic partition for coalescent point processes. *Markov Proc. Relat. Fields*, 15:359–386, 2009. MR2554367
[27] J. LAMPERTI. The limit of a sequence of branching processes. *Z. Wahrscheinlichkeitstheorie und Verw. Gebiete*, 7:271–288, 1967. MR0217893

[28] J. F. LE GALL. *Random real trees*. Ann. Fac. Sci. Toulouse Math., 15:35–62, 2006. MR2225746

[29] J.-F. LE GALL and Y. LE JAN. Branching processes in Lévy processes: Laplace functionals of snake and superprocesses. *Ann. Probab.*, 26:1407–1432, 1998. MR1675019

[30] J.-F. LE GALL and Y. LE JAN. Branching processes in Lévy processes: The exploration process. *Ann. Probab.*, 26:213–252, 1998. MR1617047

[31] G. VOISIN Dislocation measure of the fragmentation of a general Lévy tree. *arXiv:0903.4024*, accepted at *ESAIM : P&S*, 2009.