POSITIVE SELF-SIMILAR MARKOV PROCESSES OBTAINED BY RESURRECTION

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Abstract. In this paper we study positive self-similar Markov processes obtained by (partially) resurrecting a strictly \( \alpha \)-stable process at its first exit time from \((0, \infty)\). We construct those processes by using the Lamperti transform. We explain their long term behavior and give conditions for absorption at 0 in finite time. In case the process is absorbed at 0 in finite time, we give a necessary and sufficient condition for the existence of a recurrent extension. The motivation to study resurrected processes comes from the fact that their jump kernels may explode at zero. We establish sharp two-sided jump kernel estimates for a large class of resurrected stable processes.

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1. Introduction

A \([0, \infty)\)-valued standard Markov process (see [2]) \(X = (X_t, \mathbb{P}_x), \ t \geq 0, \ x \geq 0\), is called a positive self-similar Markov process (pssMp) if there exists \( \alpha > 0 \) such that for any \( x > 0 \) and \( c > 0 \), the law of \((cX_{c^{-\alpha}t} : t \geq 0)\) under \( \mathbb{P}_x \) is equal to the law of \((X_t : t \geq 0)\) under \( \mathbb{P}_{cx} \). One refers to \( \alpha \) as the self-similarity index. We will say that \( X \) is a pssMp with the origin as a trap (or that \( X \) is absorbed at the origin) if once \( X \) hits the origin it stays there forever. Self-similar processes were introduced by Lamperti in [24] where he established a one-to-one correspondence between pssMps up to the first exit time from \((0, \infty)\) and possibly killed Lévy processes. A detailed description of this correspondence, usually called the Lamperti transform, is given in Section 2.

A canonical example of a pssMp with origin as a trap is an \( \alpha \)-stable process in \( \mathbb{R} \) absorbed at the origin upon exiting \((0, \infty)\). To be more precise, let \( \eta = (\eta_t)_{t \geq 0} \) be a strictly \( \alpha \)-stable process in \( \mathbb{R}, \alpha \in (0, 2) \). Its Lévy measure has a density

\[

\nu(x) = c_+ x^{-1-\alpha}1_{x>0} + c_- |x|^{-1-\alpha}1_{x<0}, \quad x \in \mathbb{R},

\]

where \( c_+, c_- \geq 0 \) and \( c_+ = c_- \) if \( \alpha = 1 \). Let \( \rho := \mathbb{P}(\eta_1 \geq 0) = \mathbb{P}(\eta_1 > 0) \) be the positivity parameter, and set \( \hat{\rho} := 1 - \rho \). The process \( \eta \) will be parameterized so that

\[

c_+ = \frac{\Gamma(\alpha + 1)}{\Gamma(\alpha \rho) \Gamma(1 - \alpha \rho)}, \quad c_- = \frac{\Gamma(\alpha + 1)}{\Gamma(\alpha \hat{\rho}) \Gamma(1 - \alpha \hat{\rho})}.

\]

Throughout the paper, we will exclude the cases of only one-sided jumps. More precisely, the set of permissible parameters \((\alpha, \rho)\) is given by

\[

\{(\alpha, \rho) : \alpha \in (0, 1), \rho \in (0, 1)\} \cup \{(\alpha, \rho) : \alpha \in (1, 2), \rho \in (1 - 1/\alpha, 1/\alpha)\} \cup \{(\alpha, \rho) = (1, 1/2)\},

\]

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We denote by $\mathbb{P}_x$, $x > 0$, the law of $\eta$ starting at $x$.

Let $\tau = \tau_{(0,\infty)} := \inf\{t > 0 : \eta_t \in (-\infty,0]\}$ be the first exit time of $\eta$ from $(0,\infty)$. At time $\tau$, we send the process to 0 where it stays forever, and thus arriving at the process $X_t^* := \eta_t 1_{(t<\tau)}$, $t \geq 0$. The process $X^* = (X^*_t, \mathbb{P}_x)$ is a pssMp of index $\alpha$, cf. [6] Section 3.1. If $T_0 := \inf\{t > 0 : X^*_t = 0\}$, then $T_0 = \tau_{(0,\infty)} < \infty$ and $X^*_{T_0} > 0$ a.s. We denote by $\xi^*$ the Lévy process associated with $X^*$ through the Lamperti transform.

In this paper, we introduce a large class of positive self-similar processes that can be obtained by (partially) resurrecting the strictly $\alpha$-stable process $\eta$ at the first exit time $\tau$. If $z = \eta_{\tau} < 0$ is the position where $\eta$ lands at the exit from $(0,\infty)$, we return the process into $[0,\infty)$ according to a probability distribution $p(z, \cdot)$. If the process is returned to 0, it stays there forever. More precisely, to ensure self-similarity, we consider probability kernels $p : (-\infty,0) \times \mathcal{B}([0,\infty)) \to [0,1]$ satisfying the scaling condition

$$p(\lambda z, \lambda A) = p(z, A) \quad \text{for all } z < 0, A \in \mathcal{B}([0,\infty)) \text{ and } \lambda > 0. \quad (1.3)$$

All such kernels arise in the following way: Let $\phi$ be a probability measure on $\mathcal{B}([0,\infty))$. Then

$$p(z, A) := \phi(|z|^{-1} A), \quad \text{for all } z < 0 \text{ and } A \in \mathcal{B}([0,\infty)) \quad (1.4)$$

satisfies (1.3). Conversely, if $p(\cdot, \cdot)$ satisfies (1.3) and if we set $\phi(A) = p(-1, A)$, then $p(\cdot, \cdot)$ is of the form (1.4). We call $p(\cdot, \cdot)$ the return kernel. Note that it follows from (1.3) that

$$p := 1 - p(z, \{0\})$$

is independent of $z < 0$.

Let $j(x, z) := \nu(z-x)$ be the jump kernel of $\eta$. Set

$$q_0(x, A) := \int_{(0,\infty)} j(x, z)p(z, A) \, dz, \quad x > 0, A \in \mathcal{B}([0,\infty)), \quad (1.5)$$

and note that

$$q_0(x, \{0\}) = \int_{(0,\infty)} j(x, z)p(z, \{0\}) \, dz = (1-p) \int_{-\infty}^0 c_-(x-z)^{-1-\alpha} \, dz = (1-p) \frac{c_+}{\alpha} x^{-\alpha}. \quad (1.6)$$

We define a resurrection kernel $q$ as the restriction of $q_0(x, \cdot)$ to $(0,\infty)$:

$$q(x, A) := \int_{(0,\infty)} j(x, z)p(z, A) \, dz, \quad x > 0, A \in \mathcal{B}((0,\infty)). \quad (1.6)$$

The idea behind the kernel $q_0(x, \cdot)$ is that if $x = \eta_{\tau-}$, then instead of sending $\eta$ to the origin at time $\tau$ (thus obtaining the pssMp $X^*$), with probability $1-p$ we send $\eta$ to the origin, and with probability $p = p(1, (0,\infty))$ we restart (or resurrect) it according to the normalized kernel $q(x, \cdot)$. If $p \in (0,1)$ we call the process partially resurrected, and when $p = 1$ we say that it is fully resurrected (or just resurrected). By Lemma 3.1, $q(x, \cdot)$ is absolutely continuous with respect to the Lebesgue measure, and its density will be denoted by $q(x, y)$, $x, y > 0$. By resurrecting according to the normalized density we arrive at a pssMp $\overline{X}$ absorbed at the origin with jump kernel $J(x, y) := j(x, y) + q(x, y)$. The precise construction will be carried out in Section 3 by means of the Lamperti transform.

Examples of pssMp that can be obtained by such resurrection include the path censored process from [23] (also called the trace process on $(0,\infty)$ of the stable process) in which case the return kernel is equal to the Poisson kernel of $\eta$ with respect to $(-\infty,0)$; the process from [11, 29] related to nonlocal problems with Neumann boundary condition in which case the return kernel is equal to $\int_0^\infty j(z, y)dy/\int_0^{\infty} j(z, y)du$; and the absolute value process $|\eta| = (|\eta_t|)_{t\geq 0}$ in which case $p(z, A) = \delta_{-z}(A)$, see Subsection 2.3. In all three examples above $p = 1$ so we have full resurrection of $\eta$. If $p(z, A) = (1-p)\delta_0(A) + pd_{-z}(A)$, we recover the ricocheted process from [5, 22].
One of the key questions that we will address in this paper is the behavior of $X$ at its absorption time at 0. In case $p < 1$ it is clear from the Lamperti trichotomy (see Section 2) that the partially resurrected process $X$ will be absorbed at 0 in finite time by a jump. In case $p = 1$, the absorption time may be infinite or finite, and in the latter case it turns out that $X$ is continuously absorbed at 0. We answer the question of finite or infinite lifetime by studying the behavior of the Lévy process $\xi = (\xi_t, P_x), t \geq 0, x \in \mathbb{R}$, associated to $X$ through the Lamperti transform. Let $\Psi$ denote the characteristic exponent of $\xi$. We will write $P_0$ as $P$ and denote expectation with respect to $P$ by $E$. When $p < 1$, it is obvious that $E|_{\xi_1} = -\infty$.

Our first main result is about the finiteness of $E|_{\xi_1}$ and provides an explicit expression of $E|_{\xi_1}$ when $p = 1$.

**Theorem 1.1.** Suppose $p = 1$. It holds that $E|_{\xi_1} < \infty$ if and only if

$$
\int_{(0,\infty)} |\log y| \phi(dy) < \infty.
$$

(1.7)

In this case

$$
E[\xi_1] = i\Psi'(0) = \Gamma(\alpha) \frac{\sin(\pi \alpha \hat{\rho})}{\pi} \left( \pi \cot(\pi \alpha \hat{\rho}) + \int_{(0,\infty)} (\log y) \phi(dy) \right).
$$

(1.8)

It follows from [19, Theorem 7.2] and the Lamperti trichotomy (see Section 2) that (i) If $E[\xi_1] \geq 0$, then $\limsup_{t \to \infty} \xi_t = +\infty$, hence the absorption time of $X$ is infinite, and (ii) If $E[\xi_1] < 0$, then $\lim_{t \to \infty} \xi_t = -\infty$, hence the absorption time of $X$ is finite $P_x$-a.s. and $X$ is continuously absorbed at 0. Therefore, since $\alpha \hat{\rho} \in (0,1)$, to deduce the long term behavior of $X$ it suffices to determine the sign of the expression in the parenthesis in (1.8), see Corollaries 3.8–3.9.

In case when $X$ is absorbed at zero in finite time, we give a definitive answer on the existence of its recurrent positive self-similar extensions.

**Theorem 1.2.** Suppose that $X$ is absorbed at zero in finite time. If

$$
\kappa_0 := \sup\{\kappa \in (0, \infty) : \int_{(0,\infty)} u^\kappa \phi(du) < \infty\} \in (0, \infty],
$$

(1.9)

then (1) $X$ has a positive self-similar recurrent extension which leaves 0 continuously; and (2) there exists $\kappa^* \in (0, \alpha)$ such that, for any $\beta \in (0, \kappa^*)$, $X$ has a positive self-similar recurrent extension which leaves 0 by a jump associated with an excursion measure of the form $c \beta x^{-1-\beta} dx, x > 0$.

Conversely, if (1.9) does not hold, then $X$ has no positive self-similar recurrent extension.

Positive self-similar Markov processes and their associated Lévy processes have been extensively studied in the last 15 years. We mention here the Lamperti stable processes [6, 7], hypergeometric processes [21, 18], double hypergeometric Lévy processes [22], and $\beta$-processes [17]. The interest in those families of processes was mostly motivated by the Wiener-Hopf factorization.

Our motivation for studying pssMps comes from our research program on the potential theory of Markov processes with jump kernels degenerate at the boundary. In [14, 15], we introduced a large class of symmetric Markov processes in $\mathbb{R}^d_+$ with jump kernels decaying at the boundary and systematically studied their potential theory. The trace process of a symmetric $\alpha$-stable process on $\mathbb{R}^d_+$ is degenerate in the sense that its jump kernel blows up at the boundary, see [4]. The same feature is true also for the process studied in [11, 29]. In [16] we studied the potential theory of a large class of symmetric Markov processes in $\mathbb{R}^d_+$ with jump kernels blowing up at the boundary. The main examples of such processes
are rotationally symmetric \( \alpha \)-stable processes resurrected upon exiting the upper-half-space. The current paper concentrates on the one dimensional case, but we do not assume that the processes are symmetric. The jump kernel \( J(x, y) = j(x, y) + q(x, y) \) of our resurrected process exhibits unusual and interesting behavior when \( y \to 0 \). Depending on the return kernel \( p(z, \cdot) \), \( J(x, y) \) may tend to \( \infty \) at various rates as \( y \to 0 \). Since \( j(x, y) \) is bounded away from the diagonal, this explosion is due to the resurrection kernel.

In order to state the precise result about behavior of the resurrection kernel, we first need a definition.

**Definition 1.3.** Let \( g : (0, \infty) \to (0, \infty) \) and \( \beta_1, \beta_2 \in \mathbb{R} \).

(i) We say that \( g \) satisfies the lower weak scaling condition at zero \( L_1(\beta_1) \) (resp. at infinity \( L^1(\beta_1) \)) if there exists \( c \in (0, 1] \) such that

\[
\frac{g(R)}{g(r)} \geq c \left( \frac{R}{r} \right)^{\beta_1} \quad \text{for all } \ r \leq R < 1 \ (\text{resp. } 1 \leq r \leq R).
\]

(ii) We say that \( g \) satisfies the upper weak scaling condition at zero \( U_1(\beta_2) \) (resp. at infinity \( U^1(\beta_2) \)) if there exists \( C \in [1, \infty) \) such that

\[
\frac{g(R)}{g(r)} \leq C \left( \frac{R}{r} \right)^{\beta_2} \quad \text{for all } \ r \leq R < 1 \ (\text{resp. } 1 \leq r \leq R).
\]

Here is our third main result in which we assume that the restriction of the measure \( \phi \) to \((0, \infty)\) is absolutely continuous with respect to the Lebesgue measure and, with slight abuse of notation, denote its density also by \( \phi \). The notation \( a \asymp b \) means that \( c \leq b/a \leq c^{-1} \) for some \( c \in (0, 1) \).

**Theorem 1.4.** Suppose that the density \( \phi \) is strictly positive. (1) If \( x \leq y \leq 5x \), then

\[
q(x, y) \asymp q(y, x) \asymp y^{1-\alpha} \asymp x^{1-\alpha}.
\]

(2) Suppose \( \phi \) satisfies the lower weak scaling condition \( L_1(\beta_1) \) at zero with \( \beta_1 > -1 - \alpha \). Then for \( 5x \leq y \),

\[
q(y, x) \asymp (y - x)^{-1-\alpha} \int_{\frac{x}{y}}^1 \phi(t) \frac{dt}{t} \asymp y^{-1-\alpha} \int_{\frac{x}{y}}^1 \phi(t) \frac{dt}{t}.
\]

Further, if \( \phi \) also satisfies the upper weak scaling condition \( U_1(\beta_2) \) at zero with \( \beta_2 < 0 \), for \( 5x \leq y \),

\[
q(y, x) \asymp (y - x)^{-1-\alpha} \phi \left( \frac{x}{y - x} \right) \asymp y^{-1-\alpha} \phi \left( \frac{x}{y} \right).
\]

(3) Suppose \( \phi \) satisfies the upper weak scaling condition \( U^1(\gamma_2) \) at infinity with \( \gamma_2 < 0 \). Then for \( 5x \leq y \),

\[
q(x, y) \asymp (y - x)^{-1-\alpha} \int_0^{\frac{y-x}{y}} t^\alpha \phi(t) dt \asymp y^{-1-\alpha} \int_1^{\frac{y}{1}} t^\alpha \phi(t) dt.
\]

Further, if \( \phi \) also satisfies the lower weak scaling condition \( L^1(\gamma_1) \) at infinity with \( \gamma_1 < -1 - \alpha \), for \( 5x \leq y \),

\[
q(x, y) \asymp (y - x)^{-1-\alpha} \left( \frac{x}{y - x} \right)^{-1-\alpha} \phi \left( \frac{y - x}{x} \right) \asymp x^{-1-\alpha} \phi \left( \frac{x}{y} \right).
\]

As a consequence of this theorem we can deduce, see Corollary 5.3, that the jump kernel \( J(x, y) \) of the path censored process goes to \( \infty \) at rate \( y^{-\alpha} \) as \( y \to 0 \), and that the jump kernel of the process with the resurrection kernel \( j(z, y)dy/ \int_0^\infty j(z, u)du \) goes to infinity at rate log \( (1/y) \) as \( y \to 0 \).
Stable process conditioned to stay positive and censored stable process can be regarded as resurrected stable processes, see [11] and [23, Remark 3.3]. However, they do not fall into the framework of resurrected stable processes of this paper. To cover these processes, we introduce a larger class of pssMps in Section 6 of this paper. This larger class incudes stable processes conditioned to stay positive, stable processes conditioned to hit 0 continuously, and censored stable processes as examples. The jump kernel \( J(x, y) = j(x, y) + q(x, y) \) of our resurrected stable process can be regarded as a modification of the original kernel \( j(x, y) \). The jump kernels of the class of pssMps in Section 6 are of the more general form \( j(x, y)B(x, y) \) where \( B : (0, \infty) \times (0, \infty) \to (0, \infty) \). In case of multidimensional isotropic stable process, the analogous procedure is quite standard when \( B \) is bounded from below and above by two positive constants, leading to the so-called stable-like processes. The situation when the function \( B(x, y) \) decays and vanishes at the boundary of the state space was recently studied in [14, 15] in the multidimensional case of the upper-half-space in \( \mathbb{R}^d \) and the symmetric jump kernel \( j(x, y) = |x - y|^{-d-\alpha} \). The paper [16] deals with the multidimensional case in the upper-half-space, the same jump kernel \( |x - y|^{-d-\alpha} \), with \( B(x, y) \) exploding at the boundary. The symmetric 1-dimensional case is also covered in that paper, but differently to the current paper which is mostly focused on self-similarity, the main concern of [16] is on potential-theoretic questions.

**Organization of the paper:** In the next section we recall some preliminary results related to pssMps and their connection to Lévy processes through the Lamperti transform, in particular the Lamperti trichotomy and the relationship between infinitesimal generators of those processes. We also introduce two families of return kernels and show how some of examples for pssMp from literature fit into these families.

Section 3 is central to the paper. Starting from the general return kernel described by the measure \( \phi \), we first look at the regular step process with the resurrection kernel \( q(x, y) \) as its jump kernel, and its counterpart through the Lamperti transform – a compound Poisson process \( \chi \). We compute the Lévy measure of \( \chi \) and its characteristic function. For particular examples given by (2.13) and (2.15) we obtain more precise expressions. Then we construct the resurrected process \( X \), and compute the characteristic exponent \( \Psi \) of the corresponding Lévy process \( \xi \). In Subsection 3.3 we study the behavior of \( X \) by analyzing the derivative of the characteristic exponent \( \Psi \) of \( \xi \) at zero and give a proof of Theorem 1.1. In Subsection 3.2 we prove Theorem 1.2. Finally, we provide several concrete examples illustrating the behavior of \( X \) at its absorption time.

In Section 4 we give a necessary and sufficient condition on \( \phi \) making the resurrection kernel symmetric, i.e., \( q(x, y) = q(y, x) \) for all \( x, y > 0 \), cf. Theorem 1.3. If, in addition, the underlying \( \alpha \)-stable process \( \eta \) is symmetric, the resulting resurrected process \( X \) will be also symmetric.

In Section 5 we provide a proof of Theorem 1.4. The key estimates are given in Lemma 5.1 where it is assumed that \( |x - y| = 1 \). By using scaling of the resurrection kernel, this suffices to prove the theorem. A particular example of a slightly modified density from the family (2.13) gives additional insight of possible behaviors at zero and at infinity, see Corollary 5.3.

In Section 6 we put the resurrected process in a more general context of processes with modified jump kernel. Given a process with jump kernel \( j(x, y) \), we multiply it by a function \( B(x, y) \) which changes the behavior of the original kernel. We show that the jump kernel of the resurrected process can be thought of as being modified by the function which explodes at the boundary, namely at zero. We end the paper by looking at the symmetric \( \alpha \)-stable case modified by a function \( B(x, y) \), and establish the behavior of the modified process at its lifetime.
For the reader’s convenience, we summarize the known examples of resurrected stable processes that can be included in our framework. Path-censored stable processes, the processes studied in [11, 29], the absolute value of a stable process and the ricocheted stable processes are examples of our resurrected stable processes. Censored stable processes, stable processes conditioned to stay positive, and stable processes conditioned to hit 0 continuously are included in the more general framework of Section 6.

Notation: We use “:=” to indicate definitions. Define \( a \lor b := \min\{a, b\} \) and \( a \land b := \max\{a, b\} \). We write \( f \sim g \) if \( f, g \) are nonnegative functions, \( c^{-1}g \leq f \leq cg \) with some constant \( c \in (0, \infty) \). We call \( c \) the comparability constant. Lower case letters \( c_i, i = 1, 2, \ldots \) are used to denote the constants in the proofs and the labeling of these constants starts anew in each proof. \( \Gamma \) denotes the Gamma function defined as \( \Gamma(x) = \int_0^\infty y^{x-1}e^{-y}dy \), \( \psi \) denotes the digamma function defined as \( \psi(x) = \frac{d}{dx} \log \Gamma(x) \), and \( B \) denotes the beta function defined by \( B(x, y) = \Gamma(x)\Gamma(y)/\Gamma(x+y) \).

2. Preliminaries

2.1. Lamperti correspondence. We start this preliminary section by briefly describing the correspondence between positive self-similar Markov processes and 1-dimensional Lévy processes, usually called the Lamperti transform. Let \( \xi = (\xi_t, \mathbb{P}_x), t \geq 0, x \in \mathbb{R}, \) be a possibly killed Lévy process sent to \(-\infty\) at death. Define the integrated exponential process \( I = (I_t)_{t \geq 0} \) by

\[
I_t := \int_0^t e^{\alpha \xi_s}ds, \quad t \geq 0,
\]

and let \( \varphi \) be its inverse:

\[
\varphi(t) := \inf\{s > 0 : I_s > t\}, \quad t \geq 0.
\]

For each \( x > 0 \), define \( \mathbb{P}_x := \mathbb{P}_{\log x} \) and

\[
X_t := \exp\{\xi_{\varphi(t)}\}1_{(t < T_\infty)}, \quad t \geq 0.
\]

Then \( X = (X_t, \mathbb{P}_x), t \geq 0, x > 0, \) is a pssMp of index \( \alpha \) with absorption time \( \zeta = T_\infty \). (See [29] for an in-depth analysis of the absorption time.) Conversely, for a pssMp \( X = (X_t, \mathbb{P}_x), t \geq 0, x > 0, \) of index \( \alpha \), let

\[
S_t := \int_0^t X_u^{-\alpha}du
\]

and let \( T(\cdot) \) be its inverse

\[
T(t) := \inf\{u > 0 : S_u > t\}, \quad t \geq 0.
\]

For any \( x \in \mathbb{R} \), define \( \mathbb{P}_x := \mathbb{P}_{e^x} \) and \( \xi_s := \log X_{T_s} \). Then \( \xi = (\xi_t, \mathbb{P}_x), t \geq 0, x \in \mathbb{R}, \) is a possibly killed Lévy process. Moreover, we have the following three exhausting scenarios (see [19, Theorem 13.1]) that we refer to as the Lamperti trichotomy:

1. \( \mathbb{P}_x(\zeta = \infty) = 1 \) for all \( x > 0 \) in which case \( \limsup_{t \to \infty} \xi_t = \infty \);
2. \( \mathbb{P}_x(\zeta < \alpha, X_\zeta = 0) = 1 \) for all \( x > 0 \) in which case \( \lim_{t \to \infty} \xi_t = -\infty \);
3. \( \mathbb{P}_x(\zeta < \alpha, X_\zeta > 0) = 1 \) for all \( x > 0 \) in which case \( \xi \) is killed at an independent exponentially distributed random time.

In case (2), we will say that \( X \) is continuously absorbed at 0.

Now we recall a few facts about 1-dimensional Lévy processes. Let \( \xi = (\xi_t, \mathbb{P}_x), t \geq 0, x \in \mathbb{R}, \) be a 1-dimensional Lévy process with characteristic triple \((d, \sigma, \nu)\). We will write \( \mathbb{P}_0 \) as \( \mathbb{P} \) and denote expectation with respect to \( \mathbb{P} \) by \( \mathbb{E} \). Then

\[
\mathbb{E}[e^{i\theta \xi_t}] = e^{-\Psi(\theta)}, \quad \theta \in \mathbb{R},
\]

(2.1)
where the characteristic exponent $\Psi$ is given by
\[
\Psi(\theta) = d\theta + \frac{1}{2} \sigma^2 \theta^2 + \int_{\mathbb{R}} \left( 1 - e^{i\theta x} + i\theta x 1_{|x| \leq 1} \right) \nu(dx), \quad \theta \in \mathbb{R}.
\] (2.2)

Recall that $\xi_1$ has finite expectation if and only if $\int_{|y| \geq 1} |y| \nu(dy) < \infty$, cf. [28, Theorem 25.3, Example 25.12]. In this case, by differentiating (2.1) we get that $E[\xi_1] = i\Psi'(0)$.

If $\xi$ is killed at an independent exponential time of parameter $q \geq 0$ (when $\xi$ is sent either to a cemetery $\theta$ or to $-\infty$), the characteristic exponent becomes $\Psi(\theta) = \Psi(\theta) + q$. Thus for the killed Lévy process $\tilde{\xi}$, the killing rate is equal to $\Psi(0)$.

Let $\mathcal{A}$ be the infinitesimal generator of the semigroup of $\xi$ (possibly killed at rate $q \geq 0$) acting on $C_0(\mathbb{R})$ (continuous functions vanishing at infinity). Then, cf. [28, Theorem 31.5], $C_0^2(\mathbb{R}) \subset \mathcal{D}(\mathcal{A})$, and for $f \in C_0^2(\mathbb{R})$,
\[
\mathcal{A}f(x) = -qf(x) - df'(x) + \frac{1}{2} \sigma^2 f''(x) + \int_{\mathbb{R}} (f(x + y) - f(x) - f'(x)y 1_{|y| \leq 1}) \nu(dy).
\]

Let $X$ be the pssMp of index $\alpha$ corresponding to the Lévy process $\xi$. Its infinitesimal generator $\mathcal{L}$ can be described as follows (cf. [6, Theorem 1], where we take the usual cutoff function $\ell(y) = y 1_{[-1,1]}(y)$): If $f : [0, \infty] \to \mathbb{R}$ is such that $f, f'$ and $x^2 f''$ are continuous on $[0, \infty]$ then it belongs to the domain of $\mathcal{L}$ and
\[
\mathcal{L}f(x) = -qx^{-\alpha} f(x) + x^{1-\alpha} \left( -d + \frac{1}{2} \sigma^2 \right) f'(x) + \frac{1}{2} \sigma^2 x^{2-\alpha} f''(x)
\]
\[+ x^{-\alpha} \int_{(0,\infty)} \left( f(ux) - f(x) - xf'(x)(\log u) 1_{[-1,1]}(\log u) \right) \mu(du),
\]
where $\mu(du) = \nu(du) \circ \log u$. By the change of variables $y = \log u$, we get
\[
\mathcal{L}f(x) = -qx^{-\alpha} f(x) + x^{1-\alpha} \left( -d + \frac{1}{2} \sigma^2 \right) f'(x) + \frac{1}{2} \sigma^2 x^{2-\alpha} f''(x)
\]
\[+ x^{-\alpha} \int_{\mathbb{R}} \left( f(x e^y) - f(x) - xf'(x) 1_{[-1,1]}(y) \right) \nu(dy),
\]
(2.4) which corresponds to the formula in [25, p. 4]. In case $\nu$ has a density (which we also denote by $\nu$), the integral in (2.4) can be (after a change of variables) written in the form
\[
\int_0^\infty \left( f(z) - f(x) - xf'(x)(\log z/x) 1_{[-1,1]}(\log z/x) \right) \nu(\log z/x) \frac{dz}{z}
\]
showing that the intensity of jumps from $x$ to $z$ (i.e. the jump kernel of $X$) is given by $z^{-1} \nu(\log z/x)$.

2.2. Strictly stable process absorbed at 0 and censored process. Recall that $\eta = (\eta_t, \mathbb{P}_x)$ denotes a strictly $\alpha$-stable process in $\mathbb{R}$, $\alpha \in (0, 2)$. Thus $\eta$ is a Lévy process with characteristic exponent given by (2.2), where $\sigma = 0$ and the Lévy measure $\nu$ has density given by (1.1). Moreover, it holds that $d = a := (c_+ - c_-)/(\alpha - 1)$ when $\alpha \neq 1$, and we specify $d = a = 0$ when $\alpha = 1$, cf. [23, p. 398].

Recall also that $\tau = \tau_{(0,\infty)} := \inf\{ t > 0 : \eta_t \in (-\infty, 0] \}$, $X^*_t := \eta_t 1_{(t<\tau)}$, $t \geq 0$, and $\xi^*$ is the Lévy process associated to $X^*$ through the Lamperti transform. Then $\xi^*$ is a killed Lévy process. Its Lévy measure $\mu$ was computed in [6, Section 3.1], see also [23, (6)]. It holds that $\mu$ has a density
\[
\mu(x) = c_+ \frac{e^x}{(e^x - 1)^{1+\alpha}} 1_{(x>0)} + c_- \frac{e^x}{(1 - e^x)^{1+\alpha}} 1_{(x<0)}, \quad x \in \mathbb{R},
\]
(2.5)
and $\xi^*$ is killed at rate

$$\frac{c_+}{\alpha} = \frac{\Gamma(\alpha)}{\Gamma(\alpha + 1)}.$$

(2.6)

The characteristic exponent $\Psi^*$ of $\xi^*$ can be found in \cite{19} (13.46):

$$\Psi^*(\theta) = \frac{\Gamma(\alpha - i\theta)}{\Gamma(\alpha + 1 - i\theta)}, \quad \theta \in \mathbb{R}.$$

(2.7)

The infinitesimal generator of $\xi^*$ is given by

$$A^* f(x) = -\frac{c_+}{\alpha} f'(x) - bx^{1-\alpha} f'(x) + \int_{\mathbb{R}} \left( f(x + y) - f(x) - f'(x) y 1_{[-1,1]}(y) \right) \mu(dy),$$

(2.8)

where $b \in \mathbb{R}$ is a linear term which we will identify shortly. By \cite{4}, the infinitesimal generator of $X^*$ is equal to

$$L^* f(x) = -\frac{c_+}{\alpha} x^{-\alpha} f(x) - bx^{1-\alpha} f'(x) + \int_{\mathbb{R}} \left( f(xe^y) - f(x) - x f'(x) y 1_{[-1,1]}(y) \right) \mu(dy).$$

We have that

$$\mu \left( \log \frac{z}{x} \right) = c_+ \left( \frac{z}{x} - 1 \right)^{1+\alpha} 1_{z>x} + c_- \left( 1 - \frac{z}{x} \right)^{1+\alpha} 1_{z<x}$$

$$= x^{\alpha} \left( c_+ |z - x|^{-1-\alpha} 1_{z>x} + c_- |z - x|^{-1-\alpha} 1_{z<x} \right)$$

$$= x^\alpha \nu(z - x).$$

Therefore, with $j(x, z) := \nu(z - x),

$$L^* f(x) = -\frac{c_+}{\alpha} x^{-\alpha} f(x) - bx^{1-\alpha} f'(x)$$

$$+ \int_0^\infty \left( f(z) - f(x) - x f'(x) (\log z/x) 1_{[-1,1]}(\log(z/x)) \right) j(x, z) dz.$$

(2.9)

By comparing this expression with the form of the infinitesimal generator of $X^*$ given in \cite{5} Theorem 2], we see that

$$b = -a - \int_0^\infty \left( (\log u) 1_{[-1,1]}(\log u) - (u - 1) 1_{[-1,1]}(u - 1) \right) \nu(u - 1) du.$$  

(2.10)

Let $\Psi(\theta) := \Psi^*(\theta) - \Psi^*(0) = \Psi^*(\theta) - c_- / \alpha$. Then $\Psi$ is the characteristic exponent of an unskilled Lévy processes $\xi$. More precisely, $\xi^*$ is equal in distribution to $\xi$ killed at an independent exponential time with parameter $c_- / \alpha$. The infinitesimal generator of $\xi$ is

$$A f(x) = -b f'(x) + \int_{\mathbb{R}} \left( f(x + y) - f(x) - f'(x) y 1_{[-1,1]}(y) \right) \mu(dy).$$

(2.11)

It is immediate from \cite{2.3} that $\int_{|y| \geq 1} |y| \mu(dy) < \infty$, hence $\mathbb{E} |\xi_1| < \infty$. Let $X = (X_t, \mathbb{P}_x)$ be the pssMp of index $\alpha$ corresponding to $\xi$ through the Lamperti transform. The effect of removing the killing term from the generator of $\xi^*$ is to remove the killing term from the generator of $X^*$. Hence, the infinitesimal generator $L$ of $X$ is given by the right-hand side of \cite{2.9} with $-c_-/\alpha x^{-\alpha} f(x)$ removed:

$$L f(x) = -b x^{1-\alpha} f'(x) + \int_0^\infty \left( f(y) - f(x) - x f'(x) (\log y/x) 1_{[-1,1]}(\log(y/x)) \right) j(x, y) dy.$$  

(2.12)
Considered on \((0, \infty)\), \(X^*\) is a stable process killed upon exiting \((0, \infty)\). By removing the killing term in the infinitesimal generator, we end up with the process \(X\) – the (not necessarily symmetric) censored \(\alpha\)-stable process on \((0, \infty)\). The censored process \(X\) can be also regarded as a resurrected process with the resurrection kernel \(q(x, A) = \mathbb{P}_x(\eta^- \in A)\) – it is continued exactly at the position from which \(\eta\) has jumped out from \((0, \infty)\) (thus effectively suppressing this jump). Note that this type of resurrection does not fall into our setting. Censored processes were introduced in [33] in a more general multi-dimensional context for rotationally symmetric stable processes.

2.3. Examples of return kernels. An example of pssMp of index \(\alpha\) related to \(\eta\) is its absolute value process \(|\eta| = (|\eta_t|)_{t \geq 0}\). One can view \(|\eta|\) also as a resurrected process: at time \(\tau\), if \(z = \eta_\tau\), we resurrect at \(-z > 0\) according to the resurrection kernel \(q\) with \(p(z, A) = \delta_{-z}(A)\).

We discuss now two families of return kernels with \(p = 1\). First note that if the measure \(\phi\) is absolutely continuous with respect to the Lebesgue measure on \([0, \infty)\) with a density (which we denote by the same letter), then the return kernel \(p(z, \cdot)\) has a density \(p(z, y), y > 0\), and [1.3]–[1.4] imply that

\[
p(z, y) = \phi \left( \frac{y}{|z|} \right) \frac{1}{|z|}.
\]

In the first family of return kernels the probability measure \(\phi\) has a density which decays polynomially at infinity: For \(\beta > 0\) and \(\gamma > \beta\), let

\[
\phi(t) = \phi_{\beta, \gamma}(t) = \frac{\Gamma(\gamma)}{\Gamma(\beta) \Gamma(\gamma - \beta)} t^{\beta - 1} (1 + t)^{-\gamma}.
\]  

(2.13)

Motivation for this family comes from two particular examples. The first one is the path-censored process introduced in [23]. This process is obtained from \(\eta\) by removing parts of the path in \((-\infty, 0]\). More formally, define \(A_t := \int_0^t 1_{\{\eta_s > 0\}} ds\) and let \(\tau_t := \inf\{s > 0 : A_s > t\}\) be its right-continuous inverse. The process \(\theta = (\theta_t)_{t \geq 0}\), defined by \(\theta_t = \eta_{\tau_t}\), is a strong Markov process on \((0, \infty)\), called the path-censored process of \(\eta\) on \((0, \infty)\). The part of the process \(\theta\) until its first hitting time of 0 can be described in the following way: Let \(x = \eta_{\tau^-} \in (0, \infty)\) be the position from which \(\eta\) jumps out of \((0, \infty)\), and \(z = \eta_{\tau} < 0\) be the position where \(\eta\) lands at the exit from \((0, \infty)\). The distribution of the returning position of \(\eta\) to \((0, \infty)\) has the density \(P_{(-\infty, 0)}(z, y)\) called the Poisson kernel: If \(\sigma := \inf\{t > 0 : \eta_t \in [0, \infty)\}\), then \(\mathbb{P}_x(\eta_{\sigma} \in A) = \int_A P_{(-\infty, 0)}(z, y) dy, A \in \mathcal{B}([0, \infty))\). The exact formula for this Poisson kernel is given by (e.g. [20] Lemma 1.1) which contains a minor typo: the \(\alpha\) there should be \(\alpha \rho\),

\[
P_{(-\infty, 0)}(z, y) = \frac{1}{\Gamma(1 - \alpha \rho) \Gamma(\alpha \rho)} \left( \frac{y}{|z|} \right)^{-\alpha \rho} (|z| + y)^{-1} = \phi_{1 - \alpha \rho, 1} \left( \frac{y}{|z|} \right) \frac{1}{|z|}.
\]

In the second example the return kernel is equal to the normalized jump kernel, see [11] [29].

\[
p(z, y) = \frac{\int_0^\infty j(z, y) d\sigma}{\int_0^\infty j(z, u) d\sigma} = \alpha |z|^{\alpha} (|z| + y)^{-1 - \alpha} = \phi_{1, 1 + \alpha} \left( \frac{y}{|z|} \right) \frac{1}{|z|}.
\]  

(2.14)

Here we used that \(j(z, y) = c_+(y - z)^{-1-\alpha}\) for \(z < 0, y > 0\).

In the second family, the probability measure \(\phi\) has a density which decays exponentially at infinity: For \(a, \beta, \gamma > 0\), let

\[
\phi(t) = \phi_{a, \beta, \gamma}(t) = \frac{\gamma^a}{\Gamma \left( \frac{\gamma}{\beta} \right)} t^{\beta - 1} e^{-at^\gamma}, \quad t > 0.
\]  

(2.15)
Then
\[ p(z, y) = \frac{\gamma \alpha^x y^{\beta - 1}}{\Gamma \left( \frac{x}{\gamma} \right) |z|^{\beta}} e^{-\alpha(y/|z|)^{\gamma}}. \]

3. Resurrected process

Let us go back to the strictly \( \alpha \)-stable process \( \eta \). Instead of killing \( \eta \) upon exiting \((0, \infty)\) to get \( X^* \), or restarting at \( \eta_{\tau} \) to get the censored process \( X \), we look at the exit point \( \eta_{\tau} = z < 0 \) and (partially) resurrect according to some probability kernel \( p(z, A) \), \( A \in \mathcal{B}([0, \infty)) \). We assume that \( p(z, \cdot) \) satisfies (3.3). Recall from (1.4) that such kernels are in one-to-one correspondence with probability measures \( \phi \) on \( \mathcal{B}([0, \infty)) \) through the relation
\[ p(z, A) = \phi(|z|^{-1} A), \quad z < 0. \]
If the measure \( p(z, \cdot) \) has a density with respect to the Lebesgue measure, we will denote it by \( p(z, y) \), \( y > 0 \). It is immediate that
\[ p(\lambda z, \lambda y) = \lambda^{-1} p(z, y) \quad \text{for all } z < 0, \text{ a.e. } y > 0 \text{ and all } \lambda > 0. \] (3.1)
Recall that \( \nu(z) \) is the Lévy density given in (1.1). For \( x, z \in \mathbb{R} \), let \( j(x, z) := \nu(z - x) \). If \( x > 0 \) and \( z < 0 \), we have that
\[ j(x, z) = c_-(x - z)^{-1-\alpha}. \] (3.2)
Further, \( j \) enjoys the following scaling property:
\[ j(\lambda x, \lambda z) = \lambda^{-1-\alpha} j(x, z), \quad x > 0, z < 0, \lambda > 0. \] (3.3)
Recall that the resurrection kernel \( q(x, A) \) was defined in (1.6) as
\[ q(x, A) := \int_{(-\infty, 0)} j(x, z)p(z, A) \, dz = \int_{(-\infty, 0)} \frac{c_-}{(x - z)^{1+\alpha}} \phi \left( \frac{A}{|z|} \right) \, dz \quad x > 0, A \in \mathcal{B}((0, \infty)). \] (3.4)

3.1. Compound Poisson processes corresponding to resurrection kernels. In this subsection we study the pssMp defined through the resurrection kernel \( q \) and the corresponding Lévy process. From the scaling properties of \( p \) in (1.3) and \( j \) in (3.3) we get that the resurrection kernel \( q \) satisfies
\[ q(\lambda x, \lambda A) = \lambda^{-\alpha} q(x, A), \quad x > 0, A \in \mathcal{B}((0, \infty)), \lambda > 0. \]
In particular, \( q(1, A) = x^\alpha q(x, xA) \), implying that for any \( g : (0, \infty) \to \mathbb{R} \),
\[ \int g(y)q(1, dy) = x^\alpha \int g(y/x)q(x, dy). \] (3.5)

**Lemma 3.1.** For all \( x > 0 \), the measure \( q(x, \cdot) \) has a density \( q(x, y) \) given by
\[ q(x, y) := c_- \int_{(0, \infty)} \left( x + \frac{y}{t} \right)^{-1-\alpha} t^{-1} \phi(dt). \] (3.6)

**Proof.** By (3.4), for any \( A \in \mathcal{B}((0, \infty)) \) we have
\[ q(x, A) = c_- \int_0^\infty (x + z)^{-1-\alpha} \int_{(0, \infty)} 1_{A/z}(t) \phi(dt) \, dz \]
\[ = c_- \int_{(0, \infty)} \int_0^\infty (x + z)^{-1-\alpha} 1_{A}(zt) \, dz \, \phi(dt) \]
\[ = c_- \int_{(0, \infty)} \int_0^\infty \left( x + \frac{y}{t} \right)^{-1-\alpha} t^{-1} 1_A(y) \, dy \, \phi(dt) \]
= \int_A \left( c_- \int_{(0,\infty)} (x + \frac{y}{t})^{-1-\alpha} t^{-1} \phi(dt) \right) dy,

where in the second and last equalities we used Tonelli’s theorem and in the penultimate equality the change of variables \( y = tz \).

We record here a simple consequence of (3.6): For all \( y > 0 \),

\[
q(1, y) \leq c_- (y^{-1-\alpha} + 1).
\]

Indeed,

\[
q(1, y) = c_- \left( \int_{(0,1)} \left( 1 + \frac{y}{t} \right)^{-1-\alpha} \phi(dt) + \int_{[1,\infty)} \left( 1 + \frac{y}{t} \right)^{-1-\alpha} \phi(dt) \right)
\leq c_- \left( y^{-1-\alpha} \int_{(0,1)} t^\alpha \phi(dt) + \int_{[1,\infty)} t^{-1} \phi(dt) \right) \leq c_- (y^{-1-\alpha} + 1).
\]

Note that

\[
q(x) := q(x, (0, \infty)) = \int_{-\infty}^0 j(x, z)p(z, (0, \infty))dz = \int_{-\infty}^0 (x - z)^{-1-\alpha}dz = \frac{c_-}{\alpha} x^{-\alpha},
\]

so that

\[
Q(x, A) := \frac{q(x, A)}{q(x)}, \quad x > 0, A \in \mathcal{B}((0, \infty)),
\]

is a well-defined probability kernel satisfying

\[
Q(\lambda x, \lambda A) = Q(x, A), \quad x > 0, A \in \mathcal{B}((0, \infty)), \lambda > 0.
\]

It follows from Lemma 3.1 that both \( q(x, \cdot) \) and \( Q(x, \cdot) \) have densities \( q(x, y) \), resp. \( Q(x, y) \), satisfying

\[
q(\lambda x, \lambda y) = \lambda^{-1-\alpha} q(x, y), \quad Q(\lambda x, \lambda y) = \lambda^{-1} Q(x, y), \quad x, y > 0, \lambda > 0.
\]

We define now \( \Pi(x, \cdot) \) to be the image measure of \( Q(e^x, \cdot) \) under the mapping \( y \mapsto e^y \):
\[
\Pi(x, A) := Q(e^x, e^A). \quad \text{In particular, for every } g : \mathbb{R} \rightarrow \mathbb{R},
\]

\[
\int_{\mathbb{R}} g(y)\Pi(x, dy) = \int_{(0,\infty)} g(\log y)Q(e^x, dy). \quad (3.8)
\]

Note that \( \Pi(\cdot, \cdot) \) is translation invariant, that is, for all \( u \in \mathbb{R} \),

\[
\Pi(x + u, A + u) = Q(e^x e^u, e^ue^A) = Q(e^x, e^A) = \Pi(x, A).
\]

Let

\[
\Pi(A) := \Pi(0, A) = Q(1, e^A), \quad A \in \mathcal{B}(\mathbb{R}),
\]

and

\[
\pi(A) := q(1, e^A) = q(1)Q(1, e^A) = q(1)\Pi(A), \quad A \in \mathcal{B}(\mathbb{R}).
\]

Clearly, \( \Pi \) is a probability measure on \( \mathcal{B}(\mathbb{R}) \) and \( \pi \) a finite measure, hence a Lévy measure. Both \( \Pi \) and \( \pi \) have densities (which again by an abuse of notation we denote by the same letters) satisfying

\[
\Pi(y) = Q(1, e^y) e^y \quad \text{and} \quad \pi(y) = q(1, e^y) e^y = e^{-\alpha y} q(e^{-y}, 1). \quad (3.9)
\]

It follows from (3.7) that

\[
\pi(y) \leq c_- (e^{-(1+\alpha)y} + 1)e^y = c_- (e^{-\alpha y} + e^y).
\]

In particular, \( \pi(y) \) is bounded in every neighborhood of 0.
Let \( \chi \) be a compound Poisson process with characteristic exponent \( \Psi^\chi(\theta) = \int_\mathbb{R} (1 - e^{i\theta y}) \pi(dy) \). The jump distribution of \( \chi \) is given by the measure \( \Pi \) and the jump rate is \( q(1) \). The infinitesimal generator of \( \chi \) is given by

\[
Af(x) = \int_\mathbb{R} (f(x + y) - f(x)) \pi(dy).
\]

Let \( Y = (Y_t, \mathbb{P}_x) \) be the pssMp of index \( \alpha \) related to \( \chi \) through the Lamperti transform. The infinitesimal generator of \( Y \) is, according to (2.4), equal to

\[
\mathcal{L}^Y f(x) = x^{-\alpha} \int_\mathbb{R} (f(xe^y) - f(x)) \pi(du) = x^{-\alpha} \int_{(0,\infty)} (f(xu) - f(x)) q(1, du)
\]

\[
= \int_{(0,\infty)} (f(y) - f(x)) q(x, dy),
\]

where we used (3.8) in the second equality, and (3.5) in the third equality. This shows that \( Y = (Y_t, \mathbb{P}_x) \) is a regular step process defined by the Markov kernel \( Q(x, A), A \in \mathcal{B}((0, \infty)) \), and the holding function \( q(x) \), cf. [2, I.12].

**Theorem 3.2.** Suppose that \( p(z, \cdot) \) is given by (1.4). Then

\[ \tilde{\pi}(\theta) := \int_\mathbb{R} e^{i\theta y} \pi(dy) = c_- \frac{\Gamma(\alpha - i\theta)\Gamma(1 + i\theta)}{\Gamma(\alpha)} \int_{(0,\infty)} u^{i\theta} \phi(du). \]

**Proof.** We have for \( A \in \mathcal{B}(\mathbb{R}) \),

\[
\pi(A) = q(1, e^A) = c_- \int_0^\infty (1 + z)^{-1-\alpha} \phi \left( \frac{e^A}{z} \right) dz
\]

\[
= c_- \int_0^\infty (1 + z)^{-1-\alpha} \left( \int_{(0,\infty)} 1_{z/y}(y) \phi(dy) \right) dz
\]

\[
= c_- \int_0^\infty (1 + z)^{-1-\alpha} \left( \int_{(0,\infty)} 1_A(\log(yz)) \phi(dy) \right) dz
\]

\[
= c_- \int_{(0,\infty)} \left( \int_0^\infty 1_A(\log(yz))(1 + z)^{-1-\alpha} dz \right) \phi(dy).
\]

Therefore, by using [13, 8.380.1-3],

\[
\tilde{\pi}(\theta) = \int_\mathbb{R} e^{i\theta y} \pi(dy) = c_- \int_{(0,\infty)} \left( \int_0^\infty e^{i\theta \log(yz)} (1 + z)^{-1-\alpha} dz \right) \phi(dy)
\]

\[
= c_- \int_{(0,\infty)} \left( \int_0^\infty (yz)^{i\theta} (1 + z)^{-1-\alpha} dz \right) \phi(dy)
\]

\[
= c_- \left( \int_0^\infty z^{i\theta} (1 + z)^{-1-\alpha} dz \right) \left( \int_{(0,\infty)} y^{i\theta} \phi(dy) \right)
\]

\[
= c_- \frac{\Gamma(\alpha - i\theta)\Gamma(1 + i\theta)}{\Gamma(\alpha)} \int_{(0,\infty)} y^{i\theta} \phi(dy).
\]

This theorem allows us to rewrite the density of the jump distribution of \( \chi \) as a convolution of a subprobability and a probability distribution on \( \mathbb{R} \), cf. [23, p.411, 2nd paragraph]. Define

\[ \tau(A) := \phi(e^A), \ A \in \mathcal{B}(\mathbb{R}), \quad \text{and} \quad f(y) := \frac{\alpha e^y}{(1 + e^y)^{1+\alpha}}, \ y \in \mathbb{R}. \]
Then \( f \) is a probability density on \( \mathbb{R} \) and

\[
\hat{f}(\theta) := \int_{\mathbb{R}} e^{\theta y} f(y)dy = \frac{\Gamma(\alpha - i\theta)\Gamma(1 + i\theta)}{\Gamma(\alpha)} , \quad \hat{\tau}(\theta) := \int_{\mathbb{R}} e^{\theta y} \tau(dy) = \int_{(0,\infty)} y^\theta \phi(dy).
\]

**Corollary 3.3.** It holds that

\[
\Pi(y) = (f * \tau)(y) = \alpha e^y \int_{(0,\infty)} \frac{t^\alpha}{(t + e^y)^{1+\alpha}} \phi(dt) , \quad y \in \mathbb{R}.
\]

**Proof.** The first equality is an immediate consequence of the equality \( \hat{\Pi} = \hat{f} \hat{\tau} \). For the second equality, we rewrite

\[
\Pi(y) = \int_{\mathbb{R}} \frac{\alpha e^{y-u}}{(1 + e^{y-u})^{1+\alpha}} \tau(du) = \alpha e^y \int_{(0,\infty)} \frac{t^{-1}}{(1 + t^{-1}e^y)^{1+\alpha}} \phi(dt) = \alpha e^y \int_{(0,\infty)} \frac{t^\alpha}{(t + e^y)^{1+\alpha}} \phi(dt).
\]

**Corollary 3.4.** (a) Assume that \( \phi \) is given by (2.13). Then

\[
\hat{\pi}(\theta) = \frac{c_- \Gamma(\alpha - i\theta)\Gamma(1 + i\theta)\Gamma(\beta + i\theta)\Gamma(\gamma - \beta - i\theta)}{\Gamma(\beta)\Gamma(\gamma - \beta)}.
\] (3.13)

(b) Assume that \( \phi \) is given by (2.15). Then

\[
\hat{\pi}(\theta) = \frac{c_- \Gamma(\alpha - i\theta)\Gamma(1 + i\theta)\Gamma(\beta + i\theta)}{\Gamma(\beta)^2} \alpha^{-\frac{\theta}{\gamma}} \Gamma\left(\frac{\beta + i\theta}{\gamma}\right)
\] (3.14)

(c) Assume that \( \phi = \delta_a, \) \( a \in (0,\infty) \). Then

\[
\hat{\pi}(\theta) = \frac{c_- \Gamma(\alpha - i\theta)\Gamma(1 + i\theta)}{\Gamma(\alpha)} a^{\theta/g}.
\]

**Proof.** (a) It follows from Theorem 3.2 and [13] 8.380.1-3] that

\[
\hat{\pi}(\theta) = \frac{c_- \Gamma(\alpha - i\theta)\Gamma(1 + i\theta)}{\Gamma(\alpha)} \frac{\Gamma(\gamma)}{\Gamma(\beta)\Gamma(\gamma - \beta)} \int_0^\infty \frac{u^{\theta + \beta - 1}}{(1 + u)^\gamma} du
\]

(b) This follows from Theorem 3.2 by noting (after the change of variables \( u = au^\gamma \)) that

\[
\int_0^\infty u^{\theta + \beta - 1} e^{-au^\gamma} du = \gamma^{-1} a^{-\frac{\beta + i\theta}{\gamma}} \Gamma\left(\frac{\beta + i\theta}{\gamma}\right).
\]

(c) This is clear. \( \square \)

In case \( \phi \) is given by (2.13), we can also compute the density \( \Pi(y) \) of the jump distribution of \( \chi \). By using Corollary 3.3 and the change of variables \( u = t^{-1}e^y \) we have

\[
\Pi(y) = \frac{\alpha \Gamma(\gamma)}{\Gamma(\beta)\Gamma(\gamma - \beta)} e^y \int_0^\infty (t + e^y)^{-1-\alpha} e^{\beta + \alpha - 1} (1 + t)^{-\gamma} dt
\]

\[
= \frac{\alpha \Gamma(\gamma)}{\Gamma(\beta)\Gamma(\gamma - \beta)} e^{-(\gamma - \beta)y} \int_0^\infty u^{-\beta}(1 + ue^{-y})^{-\gamma}(1 + u)^{-1-\alpha} du
\]

\[
= \frac{\alpha \Gamma(\gamma)B(1 + \gamma - \beta, \beta - \alpha)}{\Gamma(\beta)\Gamma(\gamma - \beta)} e^{-(\gamma - \beta)y} 2\mathcal{F}_1(\gamma, 1 + \gamma - \beta; 1 + \alpha + \gamma; 1 - e^{-y}), \quad (3.15)
\]

where the last line follows from [13] 3.197.5. Here \( 2\mathcal{F}_1 \) is the hypergeometric function.
Example 3.5. (a) Recall that $\alpha \rho < 1$. Take $\beta = 1 - \alpha \rho$ and $\gamma = 1$ in (2.13), we get

$$p(z, y) = \frac{1}{\Gamma(1 - \alpha \rho)\Gamma(\alpha \rho)} |z|^{\alpha \rho}(|z| + y)^{-1}. \quad (3.16)$$

Since $\Pi(y) = (a/c_\pi)\pi(y)$, it follows from Corollary 3.4 that the characteristic function of the jump distribution $\Pi$ of $\chi$ is equal to

$$\frac{\Gamma(\alpha - i\theta)\Gamma(1 + i\theta)\Gamma(1 - \alpha \rho + i\theta)\Gamma(\alpha \rho - i\theta)}{\Gamma(1 - \alpha \rho)\Gamma(\alpha \rho)\Gamma(\alpha)\Gamma(1 - \alpha \rho + \alpha + 2; 1 - e^{-y}), \quad y \in \mathbb{R}},$$

Since $\Gamma(1 - \alpha \rho)\Gamma(\alpha \rho) = \frac{\pi}{\sin(\pi \alpha \rho)},$ cf. [13, 8.334.3], we see that the characteristic function of the jump distribution of $\chi$ coincides with the one in [23, Proposition 4.2, (9)]. This means that the return kernel given in (3.16) corresponds to the path censored (or trace) process studied in [23]. This can be also seen by recognizing $p(z, y)$ from (3.16) as the Poisson kernel $P_{(-\infty,0)}(z,y)$ of the stable process $\eta$ (see Subsection 2.3 above). By using (3.15) with $\beta = 1 - \alpha \rho$ and $\gamma = 1$, we find

$$\Pi(y) = \frac{\alpha \Gamma(\alpha + 1)\Gamma(\alpha \rho + 1)}{\Gamma(1 - \alpha \rho)\Gamma(\alpha \rho)\Gamma(\alpha + 2)} e^{-\alpha \rho y} S_{1}(1, \alpha \rho + 1; \alpha + 2; 1 - e^{-y}), \quad y \in \mathbb{R},$$

cf. [23, (13)].

(b) Take $\beta = 1$ and $\gamma = 1 + \alpha$ in (2.13). Then

$$p(z, y) = \frac{\Gamma(1 + \alpha)}{\Gamma(\alpha)} |z|^{\alpha}(|z| + y)^{-1-\alpha} = \alpha |z|^\alpha(|z| + y)^{-1-\alpha}, \quad (3.17)$$

which is the normalized jump kernel in (2.14). It follows from Corollary 3.4 that the characteristic function of the jump distribution of $\chi$ is equal to

$$\hat{\Pi}(\theta) = \frac{\Gamma(\alpha - i\theta)^2\Gamma(1 + i\theta)^2}{\Gamma(\alpha)^2},$$

and the density is

$$\Pi(y) = \alpha^2 B(1 + \alpha, 1 - \alpha) e^{-\alpha y} S_{1}(1 + \alpha, 1 + \alpha; 2 + 2\alpha; 1 - e^{-y}), \quad y \in \mathbb{R},$$

(c) Suppose that $\phi = \delta_a, a > 0$. Then

$$\Pi(y) = \alpha e^y \frac{a^\alpha}{(a + e^y)^{1+\alpha}}, \quad y \in \mathbb{R}.$$

We end this subsection with a necessary and sufficient condition for $\chi_1$ to have finite expectation.

Proposition 3.6. $E|\chi_1| < \infty$ if and only if (1.7) holds true.

Proof. It follows from [28, Theorem 25.3, Example 25.12] that $E|\chi_1| < \infty$ if and only if $\int_{|y| \geq 1} |y| \pi(dy) < \infty$.

We first assume (1.7). We will show that $\int_{\mathbb{R}} |y| \pi(dy) < \infty$. By using (3.12) we have

$$\int_{\mathbb{R}} |y| \pi(dy) = \int_{(0,\infty)} \left( \int_{0}^{\infty} \log(yz)(1 + z)^{-1-\alpha} dz \right) \phi(dy)$$

$$\leq \int_{(0,\infty)} \left( \int_{0}^{\infty} (|\log|y| + |\log|z|)(1 + z)^{-1-\alpha} dz \right) \phi(dy)$$

$$= \int_{(0,\infty)} |\log|y|| \left( \int_{0}^{\infty} (1 + z)^{-1-\alpha} dz \right) \phi(dy).$$
\[ +c_\int_{(0,\infty)} \left( \int_0^\infty |\log z|(1 + z)^{-1-\alpha}dz \right) \phi(dy) < \infty. \]

Finiteness follows from the assumption and the fact that both integrals with respect to \( dz \) are finite.

We now assume that \( \int_{|y|\geq 1} |y|\pi(dy) < \infty \). Then by (3.12),
\[
\begin{align*}
\infty > \int_{y\leq -1} (-y)\pi(dy) & \geq c_- \int_{(0,1)} \left( \int_0^\infty (-\log(yz))(1 + z)^{-1-\alpha}1_{\log(yz)\leq -1}dz \right) \phi(dy) \\
& \geq c_- \int_{(0,1)} \left( \int_0^\infty (-\log y)(1 + z)^{-1-\alpha}1_{\log z\leq -1}dz \right) \phi(dy) \\
& = c_- \left( \int_0^{1/e} (1 + z)^{-1-\alpha}dz \right) \int_{(0,1)} (\log y)\phi(dy).
\end{align*}
\]

Thus, \( \int_{(0,1)} (\log y)\phi(dy) < \infty \). Similarly,
\[
\begin{align*}
\infty > \int_{y\geq 1} y\pi(dy) & \geq c_- \int_{(1,\infty)} \left( \int_0^\infty \log(yz)(1 + z)^{-1-\alpha}1_{\log(yz)\geq 1}dz \right) \phi(dy) \\
& \geq c_- \int_{(1,\infty)} \left( \int_0^\infty \log y(1 + z)^{-1-\alpha}1_{\log z\geq 1}dz \right) \phi(dy) \\
& = c_- \left( \int_\alpha^{\infty} (1 + z)^{-1-\alpha}dz \right) \int_{(1,\infty)} (\log y)\phi(dy).
\end{align*}
\]

Thus, \( \int_{(1,\infty)} (\log y)\phi(dy) < \infty \). We have shown that (1.7) holds. \( \square \)

3.2. Resurrected process. Let \( \Psi(\theta) = \Psi^*(\theta) - \Psi^*(0) = \Psi^*(\theta) - c_-/\alpha \) be the characteristic exponent of the Lévy process \( \xi \) corresponding to the censored \( \alpha \)-stable process \( X \) through the Lamperti transform. We define
\[
\Psi^\sharp(\theta) := \Psi(\theta) + (1 - p) \frac{c_-}{\alpha} = \Psi^*(\theta) - p\frac{c_-}{\alpha},
\]
and let \( \xi^\sharp \) be the (killed) Lévy process with the characteristic exponent \( \Psi^\sharp \). We will add to \( \xi^\sharp \) an independent compound Poisson process, denoted by \( \chi \), with characteristic exponent \( \Psi^\chi \) given by
\[
\Psi^\chi(\theta) = \int_{\mathbb{R}} (1 - e^{-i\theta y})\pi(dy) = p\frac{c_-}{\alpha} - \hat{\pi}(\theta), \quad \theta \in \mathbb{R}. \tag{3.18}
\]
The effect of this procedure is that, instead of completely removing the killing from \( \xi^* \), we remove part of the killing (i.e., \( pc_-/\alpha \)) and add jumps according to \( \Pi \) at the exponential rate \( \hat{\pi}(0) = pc_-/\alpha \). Let \( \bar{\xi} := \xi^\sharp + \chi \) be this new Lévy process. By (2.7), its characteristic exponent \( \overline{\Psi} \) is given by
\[
\overline{\Psi}(\theta) = \Psi^\sharp(\theta) + \Psi^\chi(\theta) = \frac{\Gamma(\alpha - i\theta)}{\Gamma(\alpha\hat{\rho} - i\theta)} \frac{\Gamma(i\theta + 1)}{\Gamma(i\theta + 1 - \alpha\hat{\rho})} - \hat{\pi}(\theta). \tag{3.19}
\]
Note that
\[
\overline{\Psi}(0) = \frac{\Gamma(\alpha)}{\Gamma(\alpha\hat{\rho})\Gamma(1 - \alpha\hat{\rho})} - \hat{\pi}(0) = \frac{c_-}{\alpha}(1 - p).
\]
If \( p = 1 \), then \( \overline{\Psi}(0) = 0 \), implying that in the Lamperti trichotomy the case (3) does not occur. If \( p < 1 \), \( \bar{\xi} \) is a killed Lévy process with rate \( \overline{\Psi}(0) \) and thus \( \mathbb{E}\bar{\xi}_1 = -\infty \).
Let $\overline{X} = (\overline{X}_t, \mathbb{P}_x)$ be the pssMp of index $\alpha$ with origin as a trap corresponding to $\overline{\xi}$ through the Lamperti transform. This process can be described as follows: Consider the strictly $\alpha$-stable process $\eta = (\eta_t, \mathbb{P}_x)$, and set as before $\tau = \inf\{t > 0 : \eta_t \leq 0\}$. Then $\tau < \infty$ a.s. At time $\tau$, with probability $p$ we resurrect according to the return kernel $p(\eta_t, y), y > 0$, and with probability $1 - p$ kill the process and send it to the origin. This amounts to adding the resurrection kernel $q(x, y)$ to the jump kernel $j(x, y)$ of $\eta$. The process $\overline{X}$ is a pssMp of index $\alpha$ with origin as a trap and jump kernel $J(x, y) := j(x, y) + q(x, y)$. Indeed, let $A^\xi$ be the infinitesimal generator of $\xi$ and $L^X$ the infinitesimal generator of $X$ given by (2.12). The infinitesimal generator of $\overline{\xi}$ is obtained by adding $\pi$ to the Lévy measure $\nu$ of $\xi$, and taking into account the killing term. Hence for $g \in C_c^0(\mathbb{R})$,$$
abla^\xi g = -\frac{c}{\alpha} (1 - p) g + A^\xi g + A^x g.$$By using the relation between generators of the Lévy process and the corresponding the Lamperti transformed pssMp of index $\alpha$, together with (3.11), we see that the infinitesimal generator of $\overline{X}$ is equal to$$L^X f(x) = -\frac{c}{\alpha} (1 - p) x^{-\alpha} + L^X f(x) + \int_0^\infty (f(y) - f(x))q(x, y)dy.$$Assume that $p(z, y)$ is given by (1.4). Then by (1.2), (3.19) and Theorem 3.2,$$\Psi(\theta) = \frac{\Gamma(\alpha - i\theta)\Gamma(i\theta + 1)}{\Gamma(\alpha \hat{\rho} - i\theta)\Gamma(i\theta + 1 - \alpha \hat{\rho})} - \frac{c}{\alpha} \frac{\Gamma(\alpha - i\theta)\Gamma(i\theta + 1)}{\Gamma(\alpha \hat{\rho} - i\theta)\Gamma(i\theta + 1 - \alpha \hat{\rho})} \int_{(0, \infty)} u^{i\theta} \phi(du)$$$$= \frac{\Gamma(\alpha - i\theta)\Gamma(i\theta + 1)}{\Gamma(\alpha \hat{\rho} - i\theta)\Gamma(i\theta + 1 - \alpha \hat{\rho})} \left( \frac{1}{\Gamma(\alpha \hat{\rho} - i\theta)\Gamma(i\theta + 1 - \alpha \hat{\rho})} \int_{(0, \infty)} u^{i\theta} \phi(du) \right)$$$$= \frac{\Gamma(\alpha - i\theta)\Gamma(i\theta + 1)}{\pi} \left( \frac{\sin(\pi(\alpha \hat{\rho} - i\theta)) - \sin(\pi \alpha \hat{\rho})}{\sin(\pi(\alpha \hat{\rho} - i\theta)) - \sin(\pi \alpha \hat{\rho})} \int_{(0, \infty)} u^{i\theta} \phi(du) \right).$$In the third line we used the identity $\Gamma(z)\Gamma(1 - z) = \pi / \sin(\pi z)$ twice.

**Remark 3.7.** For the ricocheted stable process from [22], the measure $\phi$ determining the return kernel is equal to $(1 - p)\delta_0 + p\delta_1$. In this case$$\Psi(\theta) = \frac{\Gamma(\alpha - i\theta)\Gamma(i\theta + 1)}{\pi} \left( \frac{\sin(\pi(\alpha \hat{\rho} - i\theta)) - \sin(\pi \alpha \hat{\rho})}{\sin(\pi(\alpha \hat{\rho} - i\theta)) - \sin(\pi \alpha \hat{\rho})} \int_{(0, \infty)} u^{i\theta} \phi(du) \right),$$which recovers [22] (4.2).

### 3.3. Behavior of $\overline{X}$ at absorption time.

If $p < 1$, it follows from the Lamperti trichotomy that case (3) occurs, hence $\overline{X}$ is absorbed at 0 by a jump. In the remaining part of this subsection we therefore assume that $p = 1$. Recall that $\mathbb{E}[\xi_1] < \infty$ and, under assumption (1.7), also $\mathbb{E}[x_1] < \infty$, cf. Proposition 3.6. Therefore, under assumption (1.7), $\mathbb{E}[\xi_1] < \infty$, $\Psi(0)$ exists, and $\mathbb{E}[\xi_1] = i\Psi(0)$. Thus, combining [19] Theorem 7.2 with the Lamperti trichotomy, we get that if $i\Psi(0) \geq 0$, then $\lim_{t \to \infty} \overline{\xi}_t = +\infty$, hence the absorption time of $\overline{X}$ is infinite; and if $i\Psi(0) < 0$, then $\lim_{t \to \infty} \overline{\xi}_t = -\infty$, hence the absorption time of $\overline{X}$ is finite $\mathbb{P}_x$-a.s. and $\overline{X}$ is continuously absorbed at 0.

**Proof of Theorem 1.1:** The equivalence of $\mathbb{E}[\xi_1] < \infty$ and (1.7) follows from Proposition 3.6. Put$$f_1(\theta) := B(\alpha - i\theta, 1 + i\theta), \quad f_2(\theta) := \int_{(0, \infty)} u^{i\theta} \phi(du)$$$$f_3(\theta) := \frac{\sin(\pi(\alpha \hat{\rho} - i\theta))}{\pi} - \frac{\sin(\pi \alpha \hat{\rho})}{\pi} f_2(\theta),$$
where \(B\) denotes the beta function. Then by \(3.20\), \(\Psi(\theta) = \Gamma(1+\alpha)f_1(\theta)f_3(\theta)\). Since \(f_2(0) = 1\) and \(f_3(0) = 0\), we have
\[
\Psi(0) = \Gamma(1 + \alpha)(f_1'(0)f_3(0) + f_1(0)f_3'(0)) = \Gamma(1 + \alpha)B(\alpha, 1)f_3'(0) = \Gamma(\alpha)f_3'(0).
\]
Using the assumption \(1.7\),
\[
f_3'(0) = i \int_{(0,\infty)} (\log u)u^\theta \phi(du)|_{\theta = 0} = i \int_{(0,\infty)} (\log u)\phi(du),
\]
and so
\[
f_3'(0) = -i \cos(\pi \hat{\alpha}) - \frac{\sin(\pi \alpha \hat{\rho})}{\pi}f_2'(0) = -i \cos(\pi \alpha \hat{\rho}) - \frac{\sin(\pi \alpha \hat{\rho})}{\pi} \int_{(0,\infty)} (\log u)\phi(du).
\]
Therefore by \(3.21\)
\[
i\Psi'(0) = \Gamma(\alpha) \left( \cos(\pi \alpha \hat{\rho}) + \frac{\sin(\pi \alpha \hat{\rho})}{\pi} \int_{(0,\infty)} (\log u)\phi(du) \right)
= \Gamma(\alpha) \frac{\sin(\pi \alpha \hat{\rho})}{\pi} \left( \pi \cot(\pi \alpha \hat{\rho}) + \int_{(0,\infty)} (\log u)\phi(du) \right).
\]

Since \((\alpha - 1)_+ < \alpha \hat{\rho} < 1\), we see from the display above that the sign of \(i\Psi'(0)\) depends on the sign of
\[
\pi \cot(\pi \alpha \hat{\rho}) + \int_{(0,\infty)} (\log u)\phi(du).
\]
Note that \(\int_{(0,\infty)} (\log u)\phi(du)\) may depend on \(\alpha\). For example, when \(\phi\) is given by \(2.13\) with \(\beta = 1\) and \(\gamma = 1 + \alpha\), it holds that \(\int_{(0,\infty)} (\log u)\phi(du) = \psi(1) - \psi(\alpha)\), where \(\psi\) is the digamma function.

Let \(L_\phi := -\int_{(0,\infty)} (\log u)\phi(du)\). Define \(\text{arccot} : \mathbb{R} \to (0, \pi)\) as a strictly decreasing and continuous function. Set
\[
a_\phi := \frac{1}{\pi} \arccot \left( \frac{L_\phi}{\pi} \right)
\]
and note that \(a_\phi \in (0, 1)\).

**Corollary 3.8.** Suppose \(p = 1\). (a) If \(\alpha \leq 1 + a_\phi\), then \(X\) is (continuously) absorbed at 0 at an a.s.-finite time if and only if \(\alpha \hat{\rho} > a_\phi\). (b) If \(\alpha > 1 + a_\phi\), then the absorption time of \(X\) is always finite \(P_x\)-a.s..

**Proof.** (a) If \(\alpha \leq 1\), since \(0 < \alpha \hat{\rho} < 1\), we see that \(\pi \cot(\pi \alpha \hat{\rho}) - L_\phi < 0\) if and only if
\[
\hat{\rho} > \frac{1}{\alpha \pi} \arccot \left( \frac{L_\phi}{\pi} \right) = \frac{a_\phi}{\alpha}.
\]
If \(1 < \alpha \leq 1 + a_\phi\), then
\[
\cot(\pi(\alpha - 1)) \geq \cot(\pi a_\phi) = \frac{L_\phi}{\pi}
\]
and so we also have that \(\pi \cot(\pi \alpha \hat{\rho}) - L_\phi < 0\) if and only if \(\hat{\rho} > \frac{a_\phi}{\alpha}\).

(b) If \(\alpha > 1 + a_\phi\) then \(\cot(\pi(\alpha - 1))) < \frac{L_\phi}{\pi}\) and we always have
\[
\pi \cot(\pi \alpha \hat{\rho}) - L_\phi < \pi \cot(\pi(\alpha - 1)) - L_\phi \leq 0.
\]
In case $L_\phi = -\int_{(0,\infty)} (\log u) \phi(du)$ is independent of $\alpha$, we can be slightly more precise. Let
\[
\rho(\alpha) := 1 - \frac{1}{\alpha \pi} \arccot \left( \frac{L_\phi}{\pi} \right),
\]
so that $\pi \cot \left( \pi \alpha (1 - \rho(\alpha)) \right) - L_\phi = 0$. Notice that $\rho(a_\phi) = 0$ (this need not be true in case $L_\phi$ depends on $\alpha$). Since $L_\phi$ does not depend on $\alpha$, the function $\alpha \mapsto \rho(\alpha)$ is strictly increasing, hence $a_\phi$ is the only zero of $\rho(\alpha)$.

**Corollary 3.9.** Suppose that $p = 1$ and $L_\phi$ does not depend on $\alpha$.

(i) If $\alpha \in (0, a_\phi)$, then the absorption time of $X$ is infinite;

(ii) If $\alpha \in [a_\phi, 1 + a_\phi]$, then the absorption time of $X$ is finite if and only if $\rho > \rho(\alpha)$;

(iii) If $\alpha \in (1 + a_\phi, 2)$, then the absorption time of $X$ is finite.

**Proof.** This is a direct consequence of Corollary 3.8 and the discussion above. \(\square\)

### 3.4. Recurrent extension

Recall that the origin is a trap for $X$. If $X$ is absorbed in 0 at finite time, one can ask if there exists a positive self-similar recurrent extension of $X$. The general result is given in [12, Theorem 1] and [27, Theorems 1 and 2]:

(i) There exists a unique positive self-similar recurrent extension of $X$ which leaves 0 continuously if and only if there exists $\kappa \in (0, \alpha)$ such that
\[
\mathbb{E} \left[ e^{\xi_1} \right] = 1, \tag{3.22}
\]
and (ii) For $\beta \in (0, \alpha)$ there exists a positive self-similar recurrent extension of $X$ which leaves 0 by a jump associated with an excursion measure of the form $c\beta x^{-1-\beta}dx$ if and only if
\[
\mathbb{E} \left[ e^{\beta \xi_1} \right] < 1. \tag{3.23}
\]

Note that $\mathbb{E}[e^{\kappa \xi_1}] = \mathbb{E}[e^{i(-i\kappa)\xi_1}] = e^{-\mathbb{E}(-i\kappa)}$ for all $\kappa \geq 0$ for which the expectation is finite. Let $\varphi : \mathbb{R} \to (\mathbb{-\infty}, +\infty]$ be defined by $\varphi(\kappa) := -\mathbb{E}(-i\kappa)$, so that
\[
\mathbb{E} \left[ e^{\kappa \xi_1} \right] = e^{\varphi(\kappa)}.
\]
Hence, (3.22) is equivalent to the existence of $\kappa \in (0, \alpha)$ such that $\varphi(\kappa) = 0$, and (3.23) is equivalent to $\varphi(\beta) < 0$. Note that, by Hölder’s inequality, $\varphi$ is convex.

**Proof of Theorem 1.2:** Let
\[
h(\kappa) := \sin(\pi (\alpha \hat{\rho} - \kappa)) - \sin(\pi \alpha \hat{\rho}) \int_{(0,\infty)} u^{\kappa} \phi(du).
\]
Clearly, $h(0) = (1 - p) \sin(\pi \alpha \hat{\rho}) \geq 0$ since $\alpha \hat{\rho} \in (0, 1)$, and note that from (3.20)
\[
\varphi(\kappa) = \varphi(-i\kappa) = \frac{\Gamma(\alpha - \kappa) \Gamma(\kappa + 1)}{\pi} h(\kappa). \tag{3.24}
\]
If $\kappa_0 < \alpha$, then $\kappa_0 + \epsilon < \alpha$ for all small $\epsilon > 0$. Since $\int_{(0,\infty)} u^{\kappa_0+\epsilon} \phi(du) = +\infty$ by definition of $\kappa_0$, we get that $h(\kappa_0 + \epsilon) = -\infty$.

Assume that (1.9) holds true, i.e., $\kappa_0 > 0$. If $\kappa_0 \geq \alpha$, then
\[
h(\alpha-) = \sin(\pi (\alpha \hat{\rho} - \alpha)) - \sin(\pi \alpha \hat{\rho}) \lim_{\kappa \uparrow \alpha} \int_{(0,\infty)} u^{\kappa} \phi(du)
\]
\[
= -\sin(\pi \alpha \rho) - \sin(\pi \alpha \hat{\rho}) \lim_{\kappa \uparrow \alpha} \int_{(0,\infty)} u^{\kappa} \phi(du) < 0,
\]
where in the last inequality we used the assumptions $\alpha \rho \in (0,1)$ and $\alpha \rho \in (0,1)$. Therefore, $h((\alpha \wedge (\kappa_0 + \epsilon))- < 0$ for all small $\epsilon > 0$. If $\kappa \in (0, \kappa_0)$,

$$h'(\kappa) = -\pi \cos(\pi (\alpha \rho - \kappa)) - \sin(\pi \alpha \rho) \left( \int_{(0, \infty)} (\log u) u^\kappa \phi(du) \right),$$

(3.25)

which is justified by (1.9).

Assume that $p = 1$. Since by the assumption that $\bar{X}$ is absorbed at 0 in finite time, this happens continuously, and therefore $E[\xi] \in [-\infty, 0)$. If (1.7) holds true, since $i\bar{W}(0) < 0$, by Theorem 1.1 and (3.25),

$$h'(0+) = -\pi \cos(\pi \alpha \rho) - \sin(\pi \alpha \rho) \int_{(0, \infty)} (\log u) \phi(du) > 0,$$

implying that $h$ is strictly positive in some neighborhood of zero. Note that by (1.9), we have $\int_{(1, \infty)} (\log u) u^\kappa \phi(du) < \infty$ for $\kappa \in (0, \kappa_0)$ and thus $\int_{(1, \infty)} (\log u) \phi(du) \leq \int_{(1, \infty)} (\log u) u^\kappa \phi(du) < \infty$. Consequently, if (1.7) does not holds, then we have $\int_{(0, 1)} (\log u) \phi(du) = -\infty$. By the monotone convergence theorem,

$$\lim_{\kappa \downarrow 0} \int_{(0, 1)} (\log u) u^\kappa \phi(du) = -\lim_{\kappa \downarrow 0} \int_{(0, 1)} (-\log u) u^\kappa \phi(du) = -\infty.$$

We now see from (3.25) that

$$\liminf_{\kappa \downarrow 0} h'(\kappa) = -\pi \cos(\pi \alpha \rho) - \sin(\pi \alpha \rho) \left( \lim_{\kappa \downarrow 0} \int_{(0, \infty)} (\log u) u^\kappa \phi(du) \right) = \infty,$$

implying again that $h$ is strictly positive in some neighborhood of zero.

If $p < 1$, then $h(0) > 0$, so again we see that $h$ is strictly positive in some neighborhood of zero.

Together with $h((\alpha \wedge (\kappa_0 + \epsilon))- < 0$ for all small $\epsilon > 0$, this implies the existence of $\kappa^* \in (0, \alpha)$ such that $h(\kappa^*) = 0$, hence also $\varphi(\kappa^*) = 0$. Thus $\bar{X}$ has a positive self-similar recurrent extension. Furthermore, by the convexity of $\varphi$, for every $\beta \in (0, \kappa^*)$ we have $\varphi(\beta) < 0$. This means that $E[e^{\beta \xi}] < 1$. By [24, Theorem 1], there exists a positive self-similar recurrent extension of $\bar{X}$ which leaves 0 by a jump associated with the excursion measure $e^{\beta \xi} x^{-1} - \beta dx, x > 0$.

Assume that (1.9) is false, that is $\kappa_0 = 0$. Then $h(\epsilon) = -\infty$ for all $\epsilon > 0$, and we see from (3.24) that $\varphi(\beta) = +\infty$ for all $\beta \in (0, \alpha)$ and consequently $E[e^{\beta \xi}] = +\infty$. Hence $\bar{X}$ does not have a recurrent extension.

\begin{remark}
It is easy to find examples of probability measures $\phi$ on $(0, \infty)$ satisfying (1.7) but not (1.9) giving rise to psSMP that are (continuously) absorbed at zero in finite time, but not having a positive self-similar recurrent extension. One such example is the measure with density $1_{(2, \infty)}(t)^{2(\log 2)^2 / (\log t)^\gamma}$. \hfill \ \square\
\end{remark}

3.5. Examples. In this subsection we analyze a list of examples. Recall that $\psi$ is the digamma function.

**Example 3.11.** We look at our main example in which $\phi$ is given by (2.13). Let $f(t) = (1 + t)^{-\gamma}$ with $\gamma > 0$. The Mellin transform of $f$ is, by [13, p.1131, 17.43.7], equal to

$$M_f(s) := \int_0^\infty f(t) t^{s-1} dt = \frac{\Gamma(s) \Gamma(\gamma - s)}{\Gamma(\gamma)}.$$
Since
\[ \int_0^\infty f(t)(\log t)t^{s-1}dt = M_p'(s) = \frac{1}{\Gamma(\gamma)} \left( \Gamma'(\gamma)\Gamma(\gamma-s) - \Gamma(\gamma)\Gamma'(\gamma-s) \right), \]
we get that
\[ \int_0^\infty (\log t)\phi(t)dt = \frac{\Gamma(\gamma)}{\Gamma(\beta)\Gamma(\gamma-\beta)} M_p'(\beta) = \psi(\beta) - \psi(\gamma - \beta). \]
Hence by (1.8),
\[ i\Psi(0) = \Gamma(\alpha)\frac{\sin(\pi\alpha\hat{\rho})}{\pi} \left( \pi \cot(\pi\alpha\hat{\rho}) + \psi(\beta) - \psi(\gamma - \beta) \right). \]
(3.26)
(a) Suppose that \( \gamma = 1 \) in (2.13). Then \( \beta \in (0, 1) \), and by the reflection formula for the digamma function, \( \psi(1 - \beta) - \psi(\beta) = \pi \cot(\pi\beta) \), see [13, 8.365.8]. Elementary calculation gives that
\[ i\Psi'(0) = -\Gamma(\alpha)\frac{\sin(\pi(\alpha\hat{\rho} - \beta))}{\sin(\pi\beta)}, \]
and the sign of \( i\Psi'(0) \) depends on the sign of \( \sin(\pi(\alpha\hat{\rho} - \beta)) \).
Case 1: \( \alpha = 1 \). Then \( \hat{\rho} = 1/2 \), and \( \sin(\pi(\alpha\hat{\rho} - \beta)) = \sin(\pi/2 - \pi\beta) = \cos(\pi\beta) \). Therefore, \( i\Psi(0) > 0 \) if \( \beta \in (1/2, 1) \), \( i\Psi(0) = 0 \) if \( \beta = 1/2 \), and \( i\Psi(0) < 0 \) if \( \beta \in (0, 1/2) \).
Case 2: \( \alpha \in (0, 1) \cup (1, 2) \). Since \( \alpha\hat{\rho} \in (0, 1) \), we have that \( \alpha\hat{\rho} - \beta \in (-1, 1) \). Thus, \( \sin(\pi(\alpha\hat{\rho} - \beta)) = 0 \) if and only if \( \rho = 1 - \frac{\beta}{\alpha} \). Since we must have that \( \rho > 0 \) and \( \rho < 1/\alpha \), we get two critical values: \( \alpha_* = \beta \) and \( 1 + \alpha_* = 1 + \beta \). If \( \alpha(0, \alpha_*) \), then \( i\Psi(0) > 0 \); if \( \alpha \in [\alpha_*, \alpha^*] \), then \( i\Psi(0) > 0 \) if \( \rho > 1 - \frac{\beta}{\alpha} \), \( i\Psi(0) = 0 \) if \( \rho = 1 - \frac{\beta}{\alpha} \), and \( i\Psi(0) < 0 \) if \( \rho < 1 - \frac{\beta}{\alpha} \). If \( \alpha \in (\alpha^*, 2) \), then \( i\Psi'(0) < 0 \).

**Figure 1.** Example 3.11 (a): Left: \( \beta = 1/3, \gamma = 1, \alpha_* = 1/3, \alpha^* = 4/3 \); Right: \( \beta = 2/3, \gamma = 1, \alpha_* = 2/3, \alpha^* = 5/3 \); \( i\Psi(0) > 0 \) in the shaded region, \( i\Psi(0) = 0 \) on the red line.

We further assume that \( \beta = 1 - \alpha\rho \). The pssMp \( \overline{X} \) is then the path process (until the first hitting of zero) of the trace process of the \( \alpha \)-stable process \( \eta \) in \( (0, \infty) \) (or the path-censored \( \alpha \)-stable process). We have that \( \alpha\hat{\rho} - \beta = \alpha\hat{\rho} + \alpha\rho - 1 = \alpha - 1 \). Thus, if \( \alpha < 1 \), then \( \alpha\hat{\rho} - \beta \in (-1, 0) \), and therefore \( i\Psi(0) > 0 \). If \( \alpha = 1 \), then \( \alpha\hat{\rho} - \beta = 0 \), and therefore \( i\Psi(0) = 0 \). If \( \alpha > 1 \), then \( \alpha\hat{\rho} - \beta \in (0, 1) \), and therefore \( i\Psi(0) < 0 \). This shows that \( \overline{X} \) has infinite absorption time in case \( \alpha \in (0, 1) \) and hits zero in finite time when \( \alpha \in (1, 2) \). Of course, since \( \overline{X} \) is the trace process, this fact is well known.

(b) Let us now consider the case \( \beta = 1 \) and \( \gamma = \alpha + 1 \). In this case
\[ i\Psi'(0) = \frac{\Gamma(\alpha)\sin(\pi\alpha\hat{\rho})}{\pi} \left( \pi \cot(\pi\alpha\hat{\rho}) + (\psi(1) - \psi(\alpha)) \right). \]

(3.27)
The equation \( \pi \cot(\pi(\alpha - 1)) = \psi(\alpha) - \psi(1) \) (obtained by formally taking \( \rho = 1/\alpha \)), has a unique solution \( \alpha^* \) in (1, 2) which can be numerically computed. It turns out that \( \alpha^* \approx 1.44386 \) with corresponding \( \rho^* = 1/\alpha^* \approx 0.692588 \). Further, solving \( \pi \cot(\pi\alpha(1-\rho)) + (\psi(1) - \psi(\alpha)) = 0 \) for \( \rho \), we get a unique solution

\[
\rho(\alpha) = 1 - \frac{1}{\alpha \pi} \arccot \left( \frac{\psi(1) - \psi(\alpha)}{\pi} \right).
\]

It is easy to see that \( \lim_{\alpha \downarrow 0} \rho(\alpha) = -\infty \), \( \lim_{\alpha \uparrow 2} \rho(\alpha) = 1 \), hence by continuity there exists \( \alpha_* \) such that \( \rho(\alpha_*) = 0 \), and consequently, \( \rho(\alpha) > 0 \) for \( \alpha \in (0, \alpha_*) \). Numerically we obtain \( \alpha_* \approx 0.596051 \).

Therefore, we conclude that (i) If \( \alpha \in (0, \alpha_*] \), then \( \Re \Psi(0) > 0 \) for all \( \rho \in (0, 1) \); (ii) if \( \alpha \in (\alpha_*, \alpha^*) \), then \( \Re \Psi(0) > 0 \) for \( \rho > \rho(\alpha) \), \( \Re \Psi(0) = 0 \) for \( \rho = \rho(\alpha) \), \( \Re \Psi(0) < 0 \) for \( \rho < \rho(\alpha) \); (iii) if \( \alpha \in [\alpha^*, 2) \), then \( \Re \Psi(0) < 0 \) for all admissible \( \rho \).

**Example 3.12.** In case \( \phi = \delta_a \),

\[
\Re \Psi'(0) = \Gamma(\alpha) \frac{\sin(\pi \alpha \hat{\rho})}{\pi} \left( \pi \cot(\pi \alpha \hat{\rho}) - \log a \right). \tag{3.28}
\]

If \( a = 1 \), the corresponding pssMp \( \mathbf{X} \) of index \( \alpha \) is the absolute value of the strictly \( \alpha \)-stable Lévy process. The sign of \( \Re \Psi'(0) \) depends on whether \( \alpha \hat{\rho} \) is less than, equal, or larger than 1/2. Similar analysis can be done for any \( a > 0 \).

**Figure 2.** Example 3.11 (b): \( \alpha_* \approx 0.596501 \), \( \alpha^* \approx 1.44386 \); \( \Re \Psi'(0) > 0 \) in the shaded region, \( \Re \Psi'(0) = 0 \) on the red line.

**Figure 3.** Example 3.12: \( a = 1 \); \( \Re \Psi'(0) > 0 \) in the shaded region, \( \Re \Psi'(0) = 0 \) on the red line.
Example 3.13. We now look at the example in which $\phi$ is given by (2.15). Then $\phi$ is a probability density on $(0, \infty)$. Its Mellin transform is given by

$$M_\phi(s) := \int_0^\infty \phi(t)t^{s-1}dt = \frac{\gamma a^{\frac{\beta}{\gamma}}}{\Gamma\left(\frac{\beta}{\gamma}\right)} \int_0^\infty t^{s+\beta-1}e^{-at}\,dt$$

$$= \frac{\gamma a^{\frac{\beta}{\gamma}}}{\Gamma\left(\frac{\beta}{\gamma}\right)} \gamma^{-1} a^{-\frac{\beta-1}{\gamma}} \Gamma\left(\frac{s+\beta-1}{\gamma}\right) = \frac{\Gamma\left(\frac{s+\beta-1}{\gamma}\right)}{\Gamma\left(\frac{\beta}{\gamma}\right)} a^{\frac{1-s}{\gamma}}.$$

Thus we have

$$\int_0^\infty \phi(t)(\log t)t^{s-1}dt = M'_\phi(1) = \frac{\Gamma\left(\frac{s+\beta-1}{\gamma}\right)}{\gamma \Gamma\left(\frac{\beta}{\gamma}\right)} \left(\psi\left(\frac{s+\beta-1}{\gamma}\right) - \log a\right) a^{\frac{1-s}{\gamma}},$$

and finally,

$$\int_0^\infty \phi(t)(\log t)dt = M'_\phi(1) = \frac{1}{\gamma} \left(\psi\left(\frac{\beta}{\gamma}\right) - \log a\right).$$

Therefore,

$$\im\Psi'(0) = \Gamma(\alpha)\frac{\sin(\pi\alpha\hat{\rho})}{\pi} \left(\pi \cot(\pi\alpha\hat{\rho}) - \frac{1}{\gamma} \left(\psi\left(\frac{\beta}{\gamma}\right) - \log a\right)\right).$$

![Figure 4. Example 3.13](image)

We end this subsection with the analysis of the behavior at lifetime (absorption time) of the censored process. To the best of our knowledge, this has not been completely done before, but see [6, p.976].

Example 3.14. In this example we consider the Lévy process with characteristic exponent

$$\Psi(\theta) = \frac{\Gamma(\alpha - i\theta)}{\Gamma(\alpha\hat{\rho} - i\theta)} - \frac{\Gamma(i\theta + 1)}{\Gamma(\alpha\hat{\rho} - i\theta)} - \frac{c_-}{\alpha}.$$

The corresponding pssMp $X$ is the (not necessarily symmetric) censored $\alpha$-stable process. It is straightforward to calculate that

$$i\Psi'(0) = \frac{\Gamma(\alpha)\sin(\pi\alpha\hat{\rho})}{\pi} \left(\pi \cot(\pi\alpha\hat{\rho}) - (\psi(1) - \psi(\alpha))\right).$$
Notice the similarity with the expression in Example 3.11 (b) and the difference being the sign in front of \((\psi(1) - \psi(\alpha))\). The equation \(\pi \cot(\pi(\alpha - 1)) = \psi(1) - \psi(\alpha)\) (obtained by formally taking \(\rho = 1/\alpha\)), has a unique solution \(\alpha^*\) in \((1, 2)\) which can be numerically computed. It turns out that \(\alpha^* \approx 1.56735\) with corresponding \(\rho^* = 1/\alpha^* \approx 0.63802\). For every \(\alpha \in (0, \alpha^*)\) equation \(\pi \cot(\pi(\alpha - 1)) = \psi(1) - \psi(\alpha)\) has a unique solution given by

\[
\rho(\alpha) = 1 - \frac{1}{\alpha \pi} \arccot \left( \frac{\psi(1) - \psi(\alpha)}{\pi} \right)
\]

which is strictly increasing in \(\alpha\). Moreover, it can be shown that \(\rho(\alpha) > 0\) for every \(\alpha \in (0, 2)\), and \(\lim_{\alpha \downarrow 0} \rho(\alpha) = 0\) (so formally we can take \(\alpha_* = 0\)). When \(\rho > \rho(\alpha)\) we have \(i\Psi'(0) > 0\), for \(\rho = \rho(\alpha)\) it holds that \(i\Psi'(0) = 0\), while for \(\rho < \rho(\alpha)\), \(i\Psi'(0) < 0\). When \(\alpha \in [\alpha^*, 2)\), for every admissible \(\rho\) we have that \(i\Psi'(0) < 0\). Note also that for every \(\rho \in [\rho^*, 1)\) it holds that \(i\Psi'(0) > 0\).

\[\text{Figure 5. Example 3.14: } \alpha^* \approx 1.56735; \ i\Psi'(0) > 0 \text{ in the shaded region, } i\Psi'(0) = 0 \text{ on the red line.}\]

4. Symmetric resurrection kernels

The goal of this section is to find a sufficient and necessary condition for the resurrection kernel to be symmetric, namely that it holds \(q(x, y) = q(y, x)\). When the resurrection kernel is symmetric and the strictly \(\alpha\)-stable process \(\eta\) is also symmetric (that is \(c_+ = c_-\), or, equivalently, \(\rho = 1/2\)), the jump kernel \(J(x, y) = j(x, y) + q(x, y)\) of the resurrected process \(\X\) is also symmetric. In particular, the process \(\X\) is symmetric with respect to the Lebesgue measure in \((0, \infty)\).

We first give a necessary and sufficient condition for symmetry in terms of the Lévy measure \(\pi\) of the compound Poisson process \(\chi\).

**Proposition 4.1.** Let \(\pi\) be the Lévy measure of the compound Poisson process \(\chi\). Then \(q(x, y) = q(y, x)\) for all \(x, y > 0\), if and only if,

\[
\pi(-y) = e^{\alpha - 1} y \pi(y). \tag{4.1}
\]

**Proof.** Recall that \(\pi(y) = e^{-\alpha y} q(e^{-y}, 1)\). Suppose that \(q\) is symmetric. Then by symmetry and scaling

\[
\pi(-y) = e^{\alpha y} q(y, 1) = e^{\alpha y} q(1, e^y) = e^{\alpha y} (e^y)^{-1-\alpha} q(e^{-y}, 1) = e^{-y} q(e^{-y}, 1) = e^{-y} e^{\alpha y} \pi(y) = e^{(\alpha - 1)y} \pi(y).
\]

The converse is similar. \(\square\)
Recall from Lemma 3.1 that
\[ q(x, y) = c_\alpha \int_{(0, \infty)} \left( x + \frac{y}{v} \right)^{-1-\alpha} \frac{\phi(\text{d}v)}{v}. \]

The proof of the next technical lemma is given in the appendix.

**Lemma 4.2.** Suppose that \( m \) is a signed Borel measure on \((0, \infty)\) such that
\[
\int_{(0, \infty)} (1 + xu)^{-1-\alpha} |m| (\text{d}u) < \infty, \quad \text{for all } x > 0
\]
and
\[
\int_{(0, \infty)} (1 + xu)^{-1-\alpha} m(\text{d}u) = 0, \quad \text{for all } x > 0. \tag{4.2}
\]
Then \( m \) is the zero measure on \((0, \infty)\).

**Theorem 4.3.** It holds that
\[ q(x, y) = q(y, x) \quad \text{for all } x, y > 0, \]
if and only if
\[ \phi_\alpha(\text{d}t) = t^{\alpha-1} \phi(\text{d}t), \quad \text{on } (0, \infty), \tag{4.3} \]
where \( \phi_\alpha \) is the pushforward of the restriction of the measure \( \phi \) to \((0, \infty)\) under the map \( x \mapsto 1/x \). In case when the restriction of the measure \( \phi(\text{d}t) \) to \((0, \infty)\) has a density \( \phi(t) \) with respect to the Lebesgue measure, the measure equality above reduces to \( \phi(t^{-1}) = t^{1+\alpha} \phi(t) \) for almost every \( t > 0 \).

**Proof.** We have that
\[
q(y, x) = c_\alpha \int_{(0, \infty)} \left( \frac{y}{u} + x \right)^{-1-\alpha} \frac{\phi(\text{d}u)}{u} = c_\alpha \int_{(0, \infty)} (yv + x)^{-1-\alpha} v^\alpha \phi(\text{d}v)
\]
\[
= c_\alpha \int_{(0, \infty)} \left( \frac{y}{u} + x \right)^{-1-\alpha} u^{-\alpha} \phi_\alpha(\text{d}u).
\]
If (4.3) holds, then the last integral in the display above is equal to
\[
c_\alpha \int_{(0, \infty)} \left( \frac{y}{u} + x \right)^{-1-\alpha} u^{-\alpha} \phi_\alpha(\text{d}u) = c_\alpha \int_{(0, \infty)} \left( \frac{y}{u} + x \right)^{-1-\alpha} \frac{\phi(\text{d}u)}{u} = q(x, y).
\]
Conversely, assume that \( q \) is symmetric. Then we must have that
\[
\int_{(0, \infty)} \left( \frac{y}{u} + x \right)^{-1-\alpha} u^{-\alpha} \phi_\alpha(\text{d}u) = \int_{(0, \infty)} \left( \frac{y}{u} + x \right)^{-1-\alpha} \frac{\phi(\text{d}u)}{u},
\]
for all \( x, y > 0 \). By taking \( y = 1 \) and rewriting, we get that
\[
\int_{(0, \infty)} (1 + xu)^{-1-\alpha} u \phi_\alpha(\text{d}u) = \int_{(0, \infty)} (1 + xu)^{-1-\alpha} \frac{\phi(\text{d}u)}{u}, \quad \text{for all } x > 0.
\]
The claim now follows from Lemma 4.2.

**Corollary 4.4.** Let \( \phi : (0, \infty) \to [0, \infty) \) be such that \( \int_0^\infty \phi(t) \text{d}t = 1 \). Then \( \phi \) satisfies (4.3) if and only if
\[
\phi(t) = f(t + t^{-1}) \left( 1 + t \right)^{1+\alpha}, \tag{4.4}
\]
for \( f : [2, \infty) \to [0, \infty) \).
Proof. Assume that $\phi$ is given by (4.4). Let $g(t) := f(t + t^{-1})$. Then $g(t^{-1}) = g(t)$ implying that $\phi$ satisfies (4.3).

Conversely, if $\phi$ satisfies (4.3), define $g(t) := \phi(t)(1 + t)^{1+\alpha}$. Then $g(t^{-1}) = g(t)$. For $s \geq 2$, solving $t + t^{-1} = s$, we get two solutions: $t = (s + \sqrt{s^2 - 4})/2 \geq 1$ and $t^{-1} = (s - \sqrt{s^2 - 4})/2 \leq 1$. Define $f(s) := g(t) = g(t^{-1})$. Since $s = t + t^{-1}$, we see that $g(t) = f(t + t^{-1})$. □

**Proposition 4.5.** Suppose that $p = 1$ and $q$ is symmetric. Then $E[\xi_1] < 0$ if $\alpha > 1$ and $\hat{\rho} \in [1/(2\alpha), 1/\alpha)$, $E[\xi_1] = 0$ if $\alpha = 1$ and $\hat{\rho} = 1/2$, and $E[\xi_1] > 0$ if $\hat{\rho} \in (0, 1/(2\alpha))$.

In particular, if $\alpha \geq 3/2$, then the absorption time of $X$ is finite $P_x$-a.s. and $X$ is continuously absorbed at 0. Also, if $\alpha \leq 1/2$, then the absorption time of $X$ is infinite.

Proof. By Theorem 4.3

$$\int_{(0,\infty)} (\log u)\phi(du) = \int_{(0,1)} (\log u)\phi(du) + \int_{(1,\infty)} (\log u)\phi(du)$$

$$= \int\limits_{(1,\infty)} (\log u)\phi^\ast(du) + \int_{(1,\infty)} (\log u)\phi(du)$$

$$= \int\limits_{(1,\infty)} (\log u)(-u^{\alpha-1}+1)\phi(du),$$

which is negative for $\alpha > 1$, zero for $\alpha = 1$, and positive for $\alpha < 1$. It follows from (1.8) that the sign of $i\overline{\Psi}'(0)$ depends on the sign of

$$\pi \cot(\pi\alpha\hat{\rho}) - \int\limits_{(1,\infty)} (\log u)(u^{\alpha-1}-1)\phi(du).$$

This expression is negative for $\alpha > 1$ and $\hat{\rho} \in [1/(2\alpha), 1/\alpha)$, zero for $\alpha = 1$ and $\hat{\rho} = 1/2$, and positive for $\alpha < 1$ and $\hat{\rho} \in (0, 1/(2\alpha)]$.

The absorption claim for $\alpha \geq 3/2$ (which is equivalent to $1 - 1/\alpha \geq 1/(2\alpha)$) follows from the assumption $\hat{\rho} > 1 - 1/\alpha$.

![Figure 6](image-url) Illustration of Proposition 4.5. In the shaded region $i\overline{\Psi}'(0) > 0$, in the white region $i\overline{\Psi}'(0) < 0$, in the yellow region the sign of $i\overline{\Psi}'(0)$ is undetermined.

**Example 4.6.** (a) Let $\phi$ be as in (2.13):

$$\phi(t) = \frac{\Gamma(\gamma)}{\Gamma(\beta)\Gamma(\gamma - \beta)} t^{\beta-1}(1 + t)^{-\gamma}.$$
Imposing condition (4.3) on \( \phi \) implies that \( \gamma = \alpha + 2\beta - 1 \). Since \( \gamma > \beta \), we get that \( \beta > 1 - \alpha \).

Thus
\[
\phi(t) = \frac{\Gamma(\alpha + 2\beta - 1)}{\Gamma(\beta)\Gamma(\alpha + \beta - 1)} t^{\beta - 1}(1 + t)^{1-\alpha-2\beta}.
\]

Note that if \( \beta = 1 \) then symmetry requires that \( \gamma = \alpha + 1 \), so we get Example 3.11 (b). In this case, the regions for the sign of \( \mu'(0) \) are completely determined, see Figure 2.

(b) If \( \phi \) is as in (2.15), then it cannot lead to a symmetric resurrection kernel.

(c) Let \( \phi = \delta_a, a > 0 \). Then the measure \( \phi \) satisfies (4.3) if and only if \( a = 1 \).

5. Sharp two-sided estimates of the resurrection kernel

In this section we establish sharp two-sided estimates of \( q(x, y) \) under minimal assumptions. First notice that it follows from (3.6) that \( q \) enjoys the following scaling
\[
q(\lambda x, \lambda y) = \lambda^{-1-\alpha} q(x, y), \quad x, y > 0, \quad \lambda > 0.
\]

This implies that for all \( 0 < x < y \) we have
\[
q(x, y) = (y - x)^{-1-\alpha} q \left( \frac{x}{y-x}, \frac{x}{y-x} + 1 \right)
\]
and
\[
q(y, x) = (y - x)^{-1-\alpha} q \left( \frac{x}{y-x} + 1, \frac{x}{y-x} \right).
\]

Thus it suffices to get the estimates of \( q(x, x+1) \) and \( q(x+1, x), x > 0 \).

We first look at the simple case when the measure \( \phi \) has compact support in \( (0, \infty) \). Then it is easy to see that
\[
q(x, x+1) = c_- \int_{(0, \infty)} \left( x + \frac{x}{t} \right)^{-1-\alpha} \phi(dt) \times 1
\]
and
\[
q(x+1, x) = c_- \int_{(0, \infty)} \left( x + \frac{x}{t} \right)^{-1-\alpha} \phi(dt) \times 1.
\]

Thus in this case by (5.1)-(5.2), we have
\[
q(x, y) \asymp |x - y|^{-1-\alpha}, \quad x, y \in (0, \infty).
\]

In the remainder of this section, we assume that \( \phi \) is absolutely continuous and has a strictly positive density.

Assume that \( |y - x| = 1 \) so that either \( y = x + 1 \) or \( y = x - 1 \). Then by Lemma 3.1
\[
q(x+1, x) = c_- \int_{0}^{\infty} \left( x + \frac{x}{t} \right)^{-1-\alpha} \phi(dt) \times 1
\]
and
\[
q(x, x+1) = c_- \int_{0}^{\infty} \left( x + \frac{x}{t} \right)^{-1-\alpha} \phi(dt) \times 1.
\]

**Lemma 5.1.** (1) If \( x \geq 1/4 \), then \( q(x+1, x) \asymp q(x, x+1) \asymp x^{-1-\alpha} \).

(2) Suppose \( \phi \) satisfies the lower weak scaling condition \( L_1(\beta_1) \) at zero with \( \beta_1 > -1 - \alpha \). Then for \( x \leq 1/4 \),
\[
q(x+1, x) \asymp \int_{x}^{1} \phi(t) \frac{dt}{t}.
\]

Further, if \( \phi \) also satisfies the upper weak scaling condition \( U_1(\beta_2) \) at zero with \( \beta_2 < 0 \), then
\[
q(x+1, x) \asymp \phi(x).
\]
(3) Suppose \( \phi \) satisfies the upper weak scaling condition \( U^1(\gamma_2) \) at infinity with \( \gamma_2 < 0 \). Then
\[
q(x, x + 1) \asymp \int_0^{1/x} t^{\alpha} \phi(t) dt \asymp \int_1^{1/x} t^{\alpha} \phi(t) dt.
\] (5.5)

Further, if \( \phi \) also satisfies the lower weak scaling condition \( L^1(\gamma_1) \) at infinity with \( \gamma_1 < -1 - \alpha \),
\[
q(x, x + 1) \asymp x^{-1-\alpha} \phi(1/x).
\] (5.6)

**Proof.** Without loss of generality, we neglect the constant \( c_- \) in the proof. First note that
\[
0 < \int_0^{\infty} \frac{t^{\alpha} \phi(t)}{(1 + t)^{\alpha + 1}} dt \leq \int_0^{\infty} \phi(t) dt = 1.
\] (5.7)

**(1) Case 1:** \( 1/4 \leq x \leq 4 \). Then
\[
x + 1 + \frac{x}{t} \geq 1 + \frac{1}{4t} \geq \frac{1}{4} \left( 1 + \frac{1}{t} \right) \quad \text{and} \quad x + 1 + \frac{x}{t} \leq 4 + 1 + \frac{1}{t} = 5 \left( 1 + \frac{1}{t} \right).
\]

Also,
\[
x + \frac{x + 1}{t} \geq \frac{1}{4} + \frac{1}{t} \geq \frac{1}{4} \left( 1 + \frac{1}{t} \right) \quad \text{and} \quad x + \frac{x + 1}{t} \leq 4 + \frac{4 + 1}{t} \leq 5 \left( 1 + \frac{1}{t} \right).
\]

Thus, by (5.7) we get
\[
q(x + 1, x) \asymp q(x, x + 1) \asymp \int_0^{\infty} \left( 1 + \frac{1}{t} \right)^{-1-\alpha} \phi(t) \frac{dt}{t} = \int_0^{\infty} \frac{t^{\alpha} \phi(t)}{(1 + t)^{\alpha + 1}} dt \asymp 1.
\]

**(Case 2):** \( x \geq 4 \). Then we have that
\[
x + \frac{x}{t} \leq x + 1 + \frac{x}{t} \leq 2x + 2 \frac{x}{t},
\]
and hence
\[
x + 1 + \frac{x}{t} \asymp x \left( 1 + \frac{1}{t} \right).
\]

Also,
\[
x + \frac{x + 1}{t} \leq (x + 1) \left( 1 + \frac{1}{t} \right) \quad \text{and} \quad x + \frac{x + 1}{t} \geq \frac{x + 1}{2} \left( 1 + \frac{1}{t} \right)
\]

imply
\[
x + \frac{x + 1}{t} \asymp (x + 1) \left( 1 + \frac{1}{t} \right) \asymp x \left( 1 + \frac{1}{t} \right).
\]

Thus, by (5.7)
\[
q(x + 1, x) \asymp q(x, x + 1) \asymp x^{-1-\alpha} \int_0^{\infty} \left( 1 + \frac{1}{t} \right)^{-1-\alpha} \phi(t) \frac{dt}{t} \asymp x^{-1-\alpha}.
\]

**(2) Assume \( x \leq 1/4 \) and let
\[
q(x + 1, x) = \int_0^{x} + \int_{x}^{1} + \int_1^{\infty} =: I + II + III.
\]

For \( 0 < t < x \), we have that
\[
x + 1 = \frac{t(x + 1) x}{x} \leq (x + 1) \frac{x}{t} \leq 2 \frac{x}{t},
\]
hence $x + 1 + \frac{x}{t} \asymp \frac{x}{t}$. Since $\phi$ satisfies the lower weak scaling condition at zero $L_1(\beta_1)$ with $\beta_1 > -1 - \alpha$, we have

$$\int_0^x t^\alpha \phi(t) dt = \phi(x) \int_0^x t^\alpha (\phi(t)/\phi(x)) dt \leq c^{-1} \phi(x) \int_0^x t^\alpha (t/x)^{\beta_1} dt = c_1 x^{\alpha + 1} \phi(x).$$

Thus,

$$I \asymp \int_0^x \left(\frac{x}{t}\right)^{-1-\alpha} \phi(t) \frac{dt}{t} = x^{-1-\alpha} \int_0^x t^\alpha \phi(t) dt \leq c_1 \phi(x).$$

For $x \leq t < \infty$, we have

$$1 \leq x + 1 + \frac{x}{t} \leq \frac{1}{4} + 1 + 1,$$

hence $x + 1 + x/t \asymp 1$. Therefore,

$$II \asymp \int_x^1 \phi(t) \frac{dt}{t}.$$

Finally,

$$III \asymp \int_1^\infty \phi(t) \frac{dt}{t} \leq \int_1^\infty \phi(t) dt \leq 1.$$

Since

$$\int_x^1 \phi(t) \frac{dt}{t} \geq \phi(x) \int_x^{2x} \frac{\phi(t) dt}{\phi(x) t} + 2 \int_{1/2}^1 \phi(t) dt \geq c\phi(x) \int_x^{2x} \frac{t^{\beta_1-1} dt}{x^{\beta_1}} + c_2 \asymp \phi(x) + 1,$$

we get (5.3). Moreover, if $\phi$ also satisfies $U_1(\beta_2)$ with $\beta_2 < 0$, then

$$c_3 \phi(x) = c_4 \phi(x) \int_x^1 (t/x)^{\beta_1} \frac{dt}{t} \leq \int_x^1 \phi(t) \frac{dt}{t} \leq \int_x^1 (t/x)^{\beta_2} \frac{dt}{t} = c_4 \phi(x),$$

so we get (5.4).

(3) Assume $x \leq 1/4$.

$$q(x, x + 1) = \int_0^{1/x} dt + \int_{1/x}^{\infty} dt = I + II.$$

For $0 < t < 2$,

$$x + \frac{x + 1}{t} \leq x + 1 + \frac{x + 1}{t} \leq 2 \frac{x + 1}{t} + (x + 1)t,$$

hence

$$x + \frac{x + 1}{t} \asymp \frac{x + 1}{t} \asymp \frac{1}{t}.$$

For $2 < t < 1/x$,

$$x + \frac{x + 1}{t} \leq \frac{1}{t} + \frac{2}{t} = \frac{3}{t},$$

hence $x + (x + 1)/t \asymp 1/t$. Therefore,

$$I \asymp \int_0^{1/x} \left(\frac{1}{t}\right)^{-1-\alpha} \phi(t) \frac{dt}{t} = \int_0^{1/x} t^\alpha \phi(t) dt.$$

When $t > 1/x$,

$$\frac{x + 1}{t} = \frac{x + 1}{x/t} \leq \frac{2}{x/t} \leq \frac{(2t)x}{t} = 2x.$$

Thus, $x + (x + 1)/t \leq 3x$, hence $x + (x + 1)/t \asymp x$. Moreover, using that fact that $\phi$ satisfies the upper weak scaling condition at infinity $U^1(\gamma_2)$ with $\gamma_2 < 0$, we have

$$\int_{1/x}^\infty \phi(t) \frac{dt}{t} = \phi(1/x) \int_{1/x}^\infty \frac{\phi(t)}{\phi(1/x) t} dt \leq C\phi(1/x) \int_{1/x}^\infty (tx)^{\gamma_2} \frac{dt}{t} = c_5 \phi(1/x).$$
Therefore,
\[ II \leq x^{-1-\alpha} \int_{(1/x)}^{\infty} \phi(t) \frac{dt}{t} \leq c_5 x^{-\alpha} \phi(1/x). \]

Now, using
\[
\int_{0}^{1/x} t^\alpha \phi(t) dt \geq \phi(1/x) \int_{1/(2x)}^{1/x} \frac{t^\alpha}{\phi(1/x)} t^\alpha dt + \int_{0}^{1} t^\alpha \phi(t) dt \geq C^{-1} \phi(1/x) \int_{1/(2x)}^{1/x} (xt)^{\gamma_2} t^\alpha dt + c_6 \propto x^{-\alpha} \phi(1/x) + 1,
\]
we get (5.5). Finally, if \( \phi \) also satisfies \( L^1(\gamma_1) \) with \( \gamma_1 < -1 - \alpha \), then
\[
c_7 x^{-\alpha} \phi(1/x) \leq c_8 \phi(1/x) \int_{1}^{1/x} t^\alpha(tx)^{\gamma_1} dt \leq \int_{1}^{1/x} t^\alpha \phi(t) dt \leq C \phi(1/x) \int_{1}^{1/x} t^\alpha(t)^{\gamma_2} dt \leq c_8 x^{-\alpha} \phi(1/x),
\]
and so we get (5.6).

Proof of Theorem 1.4: The result follows immediately from Lemma 5.1 and the scaling relations (5.1)–(5.2).

As a consequence of Theorem 1.4 we can get estimates of the Lévy density \( \pi(u) = q(1, e^u) e^u \).

We state the next corollary in its simple form.

**Corollary 5.2.** Suppose that \( \phi \) satisfies both the lower and upper scaling conditions at zero and infinity as in Theorem 1.4. Then
\[
\pi(u) \asymp \begin{cases} 
1, & |u| \leq \log 5, \\
e^u \phi(e^u), & |u| > \log 5.
\end{cases}
\]

We first apply Theorem 1.4 to a generalization of the function \( \phi \) given in 2.13. For \( \beta \geq 0, \gamma \geq \beta \) and \( \delta_+, \delta_- \in \mathbb{R} \), let
\[ \phi(t) \propto (\log(e + t))^{\delta_+} (\log(e + t^{-1}))^{\delta_-} t^{\beta-1} (1 + t)^{-\gamma}, \] (5.8)
and \( \int_{0}^{\infty} \phi(t) dt = 1 \), which implies that \( \delta_- < -1 \) if \( \beta = 0 \) and \( \delta_+ < -1 \) if \( \gamma = \beta \). Then
\[ q(x, y) \asymp \int_{0}^{\infty} \left( x + \frac{y}{t} \right)^{-1-\alpha} (\log(e + t))^{\delta_+} (\log(e + t^{-1}))^{\delta_-} \frac{dt}{t^{2-\beta} (1 + t)^{\gamma}}. \]

It is straightforward to check that the function \( \phi \) satisfies the scaling conditions assumed in Theorem 1.4. Hence we have the following result.

**Corollary 5.3.** Suppose \( \phi \) is a probability density on \( (0, \infty) \) satisfying (5.8). If \( x \leq y \leq 5x \), then
\[ q(x, y) \asymp q(y, x) \propto x^{-1-\alpha} \asymp y^{-1-\alpha}. \]

If \( 5x \leq y \), then
\[ q(y, x) \asymp y^{-1-\alpha} \begin{cases} 
(y/x)^{1-\beta} (\log(y/x))^{\delta_-}, & \beta < 1, \\
(\log(y/x))^{1+\delta_-}, & \delta_- > -1, \\
(\log(y/x)), & \delta_- = -1, \\
1, & \delta_- < -1, \\
1, & \beta > 1,
\end{cases} \] (5.9)
and

\[
q(x, y) \asymp y^{-1-\alpha} \begin{cases} 
(y/x)^{\alpha+\beta-\gamma} \log(y/x)^{\delta_+} & \gamma - \beta < \alpha, \\
\log(y/x)^{1+\delta_+} & \delta_+ > -1, \\
1 & \delta_+ = -1, \quad \gamma - \beta = \alpha, \\
1 & \delta_+ < -1, \quad \gamma - \beta > \alpha.
\end{cases}
\] (5.10)

Note that in the case \(\gamma - \alpha < \beta < 1\), or in the case \(\gamma - \alpha = \beta = 1\) and \(\delta_+ \wedge \delta_- \geq -1\), both \(q(x, y)\) and \(q(y, x)\) explode when \(x \to 0\). This leads to the following path interpretation: The intensity of jumps to and away from points near 0 is much higher than in case of the stable process. Thus, on average, large jumps to and away from points near 0 are more probable.

In the symmetric case, we have \(\gamma = \alpha + 2\beta - 1\) and \(\delta_+ = \delta_-\), see Example 4.6 (a). Hence \(-\alpha - \gamma = -1\), and the estimates for \(q(x, y)\) and \(q(y, x)\) coincide.

Now we apply Theorem 1.4 to the function \(\phi\) given in (2.15).  

**Corollary 5.4.** Suppose \(\phi\) is a probability density on \((0, \infty)\) satisfying

\[
\phi(t) \asymp t^{\beta-1}e^{-at^\gamma}, \quad t > 0
\] (5.11)

where \(a, \gamma > 0\) and \(\beta > 0\). If \(x \leq y \leq 5x\), then

\[
q(x, y) \asymp q(y, x) \asymp x^{-1-\alpha} \asymp y^{-1-\alpha}.
\]

If \(5x \leq y\), then

\[
q(y, x) \asymp y^{-1-\alpha} \begin{cases} 
(y/x)^{\beta+1}, & 0 < \beta < 1, \\
\log(y/x) & \beta = 1, \\
1 & \beta > 1,
\end{cases}
\] (5.12)

and

\[
q(x, y) \asymp y^{-1-\alpha} \int_1^y t^{\alpha+\beta-1}e^{-at^\gamma} dt \asymp y^{-1-\alpha}.
\] (5.13)

**Proof.** For the second comparison in (5.13), see (1.10) and

\[
0 < \int_1^4 t^{\alpha+\beta-1}e^{-at^\gamma} dt \leq \int_1^{y-1} t^{\alpha+\beta-1}e^{-at^\gamma} dt \leq \int_1^\infty t^{\alpha+\beta-1}e^{-at^\gamma} dt < \infty, \quad 5x \leq y.
\]

In case \(\beta \leq 1\), we see that \(q(y, x)\) explodes as \(x \to 0\), but \(q(x, y)\) stays bounded. This means that the process will have tendency for big jumps to points close to the origin.

As a consequence of Corollary 5.4 we can derive that, when a probability density \(\phi\) on \((0, \infty)\) satisfies (5.11), \(\pi(u) \asymp 1\) for \(|u| \leq \log 5\), \(\pi(u) \asymp e^{-ua}\) for \(u > \log 5\), and

\[
\pi(u) \asymp \begin{cases} 
eu^\beta, & 0 < \beta < 1, \\
|u|e^u, & \beta = 1, \\
e^u, & \beta > 1,
\end{cases}
\]

when \(u < -\log 5\).
6. Modified jump kernel

6.1. General case. Stable process conditioned to stay positive is a pssMp that can be regarded as a resurrected stable process. However, it does not fall into the framework of resurrected stable processes of this paper. In this section we introduce a larger class of pssMps by modifying the jump kernel of the pssMp $X = (X_t, \mathbb{P}_x)$ of index $\alpha$ defined in Subsection 2.2. Thus, $X$ is a not necessarily symmetric censored process. Let $j(x, y) = \nu(y - x)$, where $\nu$ is defined in (1.1). We define a new jump kernel by

$$J(x, y) := B(x, y)j(x, y), \quad x, y > 0,$$

where $B : (0, \infty) \times (0, \infty) \to (0, \infty)$ is a function satisfying the following properties:

(B1) Homogeneity: $B(\lambda x, \lambda y) = B(x, y)$ for all $x, y > 0$ and all $\lambda > 0$.

(B2) Integrability: (a) $y \mapsto e^{-\alpha y}B(1, e^y)$ is integrable at $\infty$ and $y \mapsto e^yB(1, e^y)$ is integrable at $-\infty$; (b) $y \mapsto B(1, e^y)|y|^{1-\alpha}$ is integrable at 0; (c) For all $x \in (0, \infty)$, $y \mapsto B(x, y)$ is locally integrable in $(0, \infty) \setminus \{x\}$.

(B3) Regularity: If $\alpha \in [1, 2)$, there exist $\theta > \alpha - 1$ and $C > 0$ such that

$$|B(x, x) - B(x, y)| \leq C \left( \frac{|x - y|}{x \wedge y} \right)^\theta$$

for $|x - y| \leq (x \wedge y)/4$. If $\alpha < 1$, there exists $C > 0$ such that $B(x, y) \leq C$ for $|x - y| \leq (x \wedge y)/4$.

Without loss of generality, from now on, we assume that $B(1, 1) = 1$.

For any $B(\cdot, \cdot)$ satisfying (B1)–(B3), we will construct a pssMp with the jump kernel $J$ above via the Lamperti transform of a certain Lévy process.

We first show that the jump kernel $J(x, y) = j(x, y) + q(x, y)$ of the resurrected process $\overline{X}$ is of the form introduced above. Indeed, $J(x, y)$ can be rewritten as

$$J(x, y) = j(x, y) + q(x, y) = j(x, y) \left( 1 + \frac{q(x, y)}{j(x, y)} \right) = B(x, y)j(x, y),$$

where we define $B(x, y) := 1 + q(x, y)/j(x, y)$ for $y \neq x$, and $B(x, x) = 1$. Clearly, $B(x, y)$ satisfies (B1). Next, by (2.5) and (3.9),

$$B(1, e^y) = 1 + \frac{q(1, e^y)}{j(1, e^y)} = 1 + \frac{\pi(y)}{e^{\alpha y} - 1} = 1 + \frac{\pi(y)}{\mu(y)},$$

so that $B(1, e^y)\mu(y) = \mu(y) + \pi(y)$ is a Lévy density, i.e., $\int_\mathbb{R}(1 \wedge y^2)B(1, e^y)\mu(y)dy < \infty$. Indeed, since $\mu(y) \asymp e^{-\alpha y}$ at $+\infty$, $\mu(y) \asymp e^{-y}$ at $-\infty$, and $\mu(y) \asymp |y|^{-1-\alpha}$ near zero, we know that (B2) holds. Finally, it follows from Theorem 1.4 that for $x < y < (5/4)x$ or $y < x < (5/4)y$ it holds that $q(x, y) \asymp q(y, x) \asymp x^{-1-\alpha} \asymp y^{-1-\alpha}$. Hence, if $|x - y| < (x \wedge y)/4$,

$$|B(x, y) - B(x, x)| = \frac{q(x, y)}{j(x, y)} \leq (c_+ \vee c_-)|x - y|^{1+\alpha}q(x, y) \leq C|x - y|^{1+\alpha}(x^{-1-\alpha} \vee y^{-1-\alpha}) = C \left( \frac{|y - x|}{x \wedge y} \right)^{1+\alpha}.$$

Thus (B3) holds with $\theta = 1 + \alpha$.

As examples of this general setting we also mention the $\alpha$-stable process conditioned to stay positive and the $\alpha$-stable process conditioned to hit 0 continuously, see [9, 10]. The jump kernel of the former is

$$J(x, y) = \frac{y^{\alpha \rho}}{x^{\alpha \rho}} j(x, y), \quad x, y > 0$$
and the latter

$$J(x, y) = \frac{y^{\alpha+1}}{x^{\alpha-1}} f(x, y), \quad x, y > 0.$$  

It is straightforward to show that for every $$\gamma \in (-1, 0)$$, the function $$B(x, y) = (y/x)^\gamma$$ satisfies conditions (B1)-(B3). In fact, (B1) and (B2)(b)-(c) clearly hold. For (B2)(a), $$B(1, e^y) = e^y$$, so $$e^{-\alpha y}B(1, e^y) = e^{-(\alpha-\gamma)y}$$ and is integrable at $$\infty$$ if and only if $$\gamma < \alpha$$. Also, $$e^{\delta}B(1, e^y) = e^{(1+\gamma)y}$$ and is integrable at $$-\infty$$ if and only if $$\gamma > -1$$. For (B3), without loss of generality assume that $$\gamma \neq 0$$ and consider $$x, y \in (0, \infty)$$ with $$|y - x| < (x \wedge y)/4$$. Then

$$|B(x, y) - B(x, x)| = \left| \left( \frac{y}{x} \right)^\gamma - 1 \right| = \left| \frac{\gamma u^{\gamma-1}|y - x|}{x^\gamma} \right|,$$

(6.1)

where $$u$$ is between $$x$$ and $$y$$. If $$x < y$$, then $$x = x \wedge y$$ and $$y \leq (5/4)x$$. If $$\gamma \geq 1$$, then the right-hand side above is less than $$(5/4)^{\gamma-1}(y - x)/x$$. If $$\gamma < 1$$, then we estimate the right-hand side with $$|\gamma|(y - x)/x$$. Thus both cases (B3) holds with $$\theta = 1$$. If $$y < x$$, then we replace $$x^\gamma$$ with $$y^\gamma$$ in (6.1) and argue as before.

See Proposition 6.7 for more on this example.

Given an arbitrary $$B(x, y)$$ satisfying (B1)-(B3), and the jump kernel $$j(x, y)$$ of the censored process $$X$$, we now construct a pssMp $$\overline{X} = (\overline{X}_t, \mathbb{P}_x)$$ of index $$\alpha$$ corresponding to the jump kernel $$J(x, y) = B(x, y)j(x, y)$$ via the Lamperti transform of a certain Lévy process. Define

$$\mu^B := B(1, e^y)\mu(y) = B(1, e^y) \left( c_+ \frac{e^y}{(e^y - 1)^{1+\alpha}} \mathbf{1}_{(y>0)} + c_- \frac{e^y}{(1-e^y)^{1+\alpha}} \mathbf{1}_{(y<0)} \right).$$

By the assumptions (B2)(a) and (c) we have that $$\int_{|y|>1} \mu^B(y) dy < \infty$$, while by (B2)(b) and (c) we get that $$\int_{|y|\leq 1} y^2 \mu^B(y) dy < \infty$$. Thus $$\mu^B$$ is a Lévy measure. Further, let $$\overline{\xi}$$ denote the Lévy process with infinitesimal generator

$$\overline{A}f(x) = -\overline{b}f'(x) + \int_\mathbb{R} (f(x + y) - f(x) - f'(x)y) \mathbf{1}_{[-1,1]}(y) \mu^B(y) dy,$$

(6.2)

where $$\overline{b} \in \mathbb{R}$$. Let $$\overline{X} = (\overline{X}_t, \mathbb{P}_x)$$ be the pssMp of index $$\alpha$$ obtained from $$\overline{\xi}$$ through the Lamperti transform. By using a calculation similar to the one we used to obtain $$\mathcal{L}^*$$ in Section 2 together with the homogeneity of $$B$$ and (B1), we see that the infinitesimal generator of $$\overline{X}$$ is

$$\mathcal{L} f(x) = -bx^{1-\alpha} f'(x) + x^{-\alpha} \int_\mathbb{R} (f(x e^y) - f(x) - x f'(x) y \mathbf{1}_{[-1,1]}(y)) \mu^B(y) dy$$

(6.3)

$$= -bx^{1-\alpha} f'(x) + \int_0^\infty (f(z) - f(x) - x f'(x) (\log z/x) \mathbf{1}_{[-1,1]}(\log(z/x))) B(x, z) j(x, z) dz.$$ 

This shows that the jump kernel of $$\overline{X}$$ is precisely $$J(x, y) = B(x, y) j(x, y)$$, see the last sentence of Subsection 2.2.

6.2. Symmetric case. In this subsection we assume that $$\eta$$ is a symmetric $$\alpha$$-stable process. Then $$\rho = 1/2$$, $$c_+ = c_- = c$$, and $$a = 0$$. For simplicity, we will assume that $$c = 1$$. We first consider the case that $$B(x, y)$$ is identically 1. Recall that $$X^*$$ is the process $$\eta$$ killed upon exiting $$(0, \infty)$$, and the constant $$b$$ in the linear term of its infinitesimal generator in (2.10) equal to

$$b = - \int_0^\infty \left( (\log u) \mathbf{1}_{[-1,1]}(\log u) - (u - 1) \mathbf{1}_{[-1,1]}(u - 1) \right) u^{-1-\alpha} du.$$
Lemma 6.1. It holds that

\[-b = \lim_{\epsilon \to 0} \int_{\mathbb{R},|y|>\epsilon} y \mathbf{1}_{[-1,1]}(y) \mu(y) \, dy = \text{p.v.} \int_{-1}^{1} y \mu(y) \, dy.\]

Proof. We first note that by using symmetry, for \( \epsilon \in (0,1) \) we have

\[
\int_{(0,\infty),|u-1|>\epsilon} (u-1) \mathbf{1}_{[-1,1]}(u-1)|u-1|^{-1-\alpha} \, du = \int_{\mathbb{R}} v \mathbf{1}_{|v|\leq 1}|v|^{-1-\alpha} \, dv = 0.
\]

Therefore

\[
I(\epsilon) := \int_{\mathbb{R},|y|>\epsilon} y \mathbf{1}_{[-1,1]}(y) \mu(y) \, dy
= \int_{(0,\infty),|u-1|>\epsilon} (\log u) \mathbf{1}_{[-1,1]}(\log u)|u-1|^{-1-\alpha} \, du
= \int_{(0,\infty),|u-1|>\epsilon} ((\log u) \mathbf{1}_{[-1,1]}(\log u) - (u-1) \mathbf{1}_{[-1,1]}(u-1))|u-1|^{-1-\alpha} \, du.
\]

By letting \( \epsilon \to 0 \) we obtain that \( \lim_{\epsilon \to 0} I(\epsilon) = -b \) which is the first equality in the statement. For the second,

\[
I(\epsilon) = \int_{-1}^{\log(1-\epsilon)} y \mu(y) \, dy + \int_{\log(1+\epsilon)}^{1} y \mu(y) \, dy
= \left( \int_{-1}^{\log(1-\epsilon)} y \mu(y) \, dy + \int_{-\log(1-\epsilon)}^{1-\log(1-\epsilon)} y \mu(y) \, dy \right) + \int_{\log(1+\epsilon)}^{0} y \mu(y) \, dy
=: I_1(\epsilon) + I_2(\epsilon).
\]

Suppose \( \alpha \in [1,2) \) (for \( \alpha \in (0,1) \) the integral \( I_2(\epsilon) \) is convergent). For \( y \in (0,1/2) \) it holds that \( y \mu(y) \leq c_1 y^{-\alpha} \) for some \( c_1 > 0 \), hence

\[
I_2(\epsilon) \leq c_1 \int_{\log(1+\epsilon)}^{-\log(1-\epsilon)} y^{-\alpha} \, dy
\leq c_2 \begin{cases} 
(\log (1+\epsilon)^{1-\alpha} - (\log (1-\epsilon))^{1-\alpha}) \leq c_3 \epsilon^{2-\alpha} & \text{for } \alpha \in (1,2) \\
\log \left( \frac{-\log(1-\epsilon)}{\log(1+\epsilon)} \right) & \text{for } \alpha = 1
\end{cases}
\to 0
\]
as \( \epsilon \to 0 \). Since we have already proved that \( \lim_{\epsilon \to 0} I(\epsilon) \) exists we can conclude that

\[
\lim_{\epsilon \to 0} I_1(\epsilon) = \text{p.v.} \int_{-1}^{1} y \mu(y) \, dy.
\]

\[
\square
\]

Remark 6.2. The existence of the principal value integral \( \text{p.v.} \int_{-1}^{1} y \mu(y) \, dy \) can be alternatively proved in the following way. First note that

\[
\int_{\mathbb{R},|y|\leq 1} y \mu(y) \, dy = \int_{\epsilon}^{1} y (\mu(y) - \mu(-y)) \, dy.
\]

Secondly, \( \mu(y) - \mu(-y) = y^{-\alpha}((1-\alpha) + O(y^2)) \) as \( y \downarrow 0 \), showing that the right-hand side above is convergent.
Let $\xi$ be a Lévy process with Lévy density $\mu^B(y) = \mu(y)B(1,e^y)$ and linear term
\[ b = b - \int_{-1}^{1} y(B(1,e^y) - 1)\mu(y)dy, \tag{6.4} \]
cf. (6.2). Note that the integral is convergent because of (B3). Let $\overline{X}$ be the corresponding pssMp of index $\alpha$. The jump kernel of $\overline{X}$ is $J(x,y) = B(x,y)|x-y|^{-1-\alpha}$ and the infinitesimal generator of $\overline{X}$ is given in (6.3). The following is an analog of Lemma [6.1]

**Lemma 6.3.** It holds that
\[ -\overline{b} = \lim_{\epsilon \to 0} \int_{R,|y| > \epsilon} y1_{|y| > \epsilon}(y)\mu^B(y)dy = p.v. \int_{-1}^{1} y\mu^B(y)dy. \]

**Proof.** We have
\[ \int_{R,|y| > \epsilon} y1_{|y| > \epsilon}(y)\mu^B(y)dy = \int_{R,|y| > \epsilon} y1_{|y| > \epsilon}(y)B(1,e^y)\frac{e^y}{|e^y - 1|^{1+\alpha}}dy \]
\[ = \int_{R,|y| > \epsilon} y1_{|y| > \epsilon}(y)B(1,1)\frac{e^y}{|e^y - 1|^{1+\alpha}}dy \]
\[ + \int_{R,|y| > \epsilon} y1_{|y| > \epsilon}(y)(B(1,e^y) - B(1,1))\frac{e^y}{|e^y - 1|^{1+\alpha}}dy \]
\[ =: J_1(\epsilon) + J_2(\epsilon). \]

By Lemma [6.1] and since $B(1,1) = 1$, $\lim_{\epsilon \to 0} J_1(\epsilon) = -b$. On the other hand, by using (B3) if $\alpha \geq 1$, we conclude that
\[ \lim_{\epsilon \to 0} J_2(\epsilon) = \int_{R} y1_{|y| < \epsilon}(y)(B(1,e^y) - 1)\frac{e^y}{|e^y - 1|^{1+\alpha}}dy = \int_{-1}^{1} y(B(1,e^y) - 1)\mu(y)dy. \]

This proves the first equality in the statement. For the second statement, note that for $u \to B(1,u)$ is by (B3) bounded in a neighborhood of 1. Hence, $y\mu^B(y) = y\mu(y)B(1,e^y) \leq c_1 y^{\alpha}$, and we obtain the conclusion in the same way as in Lemma [6.1]. In the case $\alpha \in (0,1)$, since the integral is absolutely convergent, we use the dominated convergence theorem. 

In the context of pssMps it is natural to write the generator in the form [6.3] which involves a cutoff function. On the other hand, in the multidimensional setting of regional non-local operators, such as the infinitesimal generator of a censored $\alpha$-stable process, generators are usually written as principal value integrals. In the context of jump kernels decaying at the boundary, such operators were studied in [14, Section 3.2] when $B$ is symmetric (see (B4) below). In the next result we reconcile these two approaches in the current setting. Let
\[ \hat{\mathcal{L}}f(x) := p.v. \int_{0}^{\infty} (f(z) - f(x))J(x,z)dz = \lim_{\epsilon \to 0} \int_{(0,\infty),|z-x| > \epsilon} (f(z) - f(x))B(x,z)|z-x|^{-\alpha}dz. \]

**Lemma 6.4.** If $f \in C^2_c((0,\infty))$, then $\hat{\mathcal{L}}f(x)$ is well defined and $\hat{\mathcal{L}}f = \mathcal{Z}f$.

**Proof.** By (B2)(c), for any compact set $K \subset (0,\infty)$ and $\epsilon > 0$,
\[ \int_{z \in K,|z-x| > \epsilon} B(x,z)dz \leq c(x,K,\epsilon) < \infty. \]
Using this and (B3), one can follow the proofs of [14, Lemma 3.3 and Proposition 3.4] and show that $\hat{\mathcal{L}}f$ is well defined for $f \in C^2_c((0,\infty))$. 

By the change of variables $z = xe^y$ we have:
\[
\int_{(0,\infty], |z-x| > \epsilon} (f(z) - f(x)) B(x, z) |x - z|^{-\alpha} dz
\]
\[
= \int_{\mathbb{R}, |xe^y - x| > \epsilon} (f(xe^y) - f(x)) B(x, xe^y) |x - xe^y|^{-\alpha} xe^y dy
\]
\[
= x^{-\alpha} \int_{\mathbb{R}, |e^y - 1| > \epsilon} (f(xe^y) - f(x)) B(1, e^y) |1 - e^y|^{-\alpha} e^y dy
\]
\[
= x^{-\alpha} \int_{\mathbb{R}, |e^y - 1| > \epsilon} (f(xe^y) - f(x)) \mu^B(y) dy
\]
\[
= x^{-\alpha} \left( \int_{\mathbb{R}, |e^y - 1| > \epsilon} (f(xe^y) - f(x) - xf'(x)y\mathbf{1}_{[-1,1]}(y)) \mu^B(y) dy \right)
\]
\[
+ x f'(x) \int_{\mathbb{R}, |e^y - 1| > \epsilon} y\mathbf{1}_{[-1,1]}(y) \mu^B(y) dy
\]
\[
= x^{-\alpha} (J_1(\epsilon) + x f'(x) J_2(\epsilon)).
\]

By the dominated convergence theorem,
\[
\lim_{\epsilon \to 0} J_1(\epsilon) = \int_{\mathbb{R}} \left( f(xe^y) - f(x) - xf'(x)y\mathbf{1}_{[-1,1]}(y) \right) \mu^B(y) dy.
\]

Since $\tilde{L} f(x)$ is well defined, we see that there also exists
\[
\lim_{\epsilon \to 0} J_2(\epsilon) = \lim_{\epsilon \to 0} \int_{\mathbb{R}, |e^y - 1| > \epsilon} y\mathbf{1}_{[-1,1]}(y) \mu^B(y) dy =: -\tilde{b}.
\]

Thus
\[
\tilde{L} f(x) = -\tilde{b} x^{-\alpha} f'(x) + x^{-\alpha} \int_{\mathbb{R}} \left( f(xe^y) - f(x) - xf'(x)y\mathbf{1}_{[-1,1]}(y) \right) \mu^B(y) dy.
\]

By Lemma 6.3 we see that $\tilde{b} = \bar{b}$ and thus $\tilde{L} = \mathcal{L}$. \hfill \Box

Now we turn to the question of the behavior of the pssMp $\overline{X}$ at its absorption time. We assume that
\[
\int_{-\infty}^{-1} |y|e^y B(1, e^y) dy + \int_{1}^{\infty} ye^{-\alpha y} B(1, e^y) dy < \infty. \tag{6.5}
\]

Then $\int_{\mathbb{R}, |y| \geq 1} |y| \mu^B(y) dy < \infty$, hence $\bar{x}_1$ has finite expectation given by
\[
\mathbb{E} \bar{x}_1 = -\tilde{b} + \int_{\mathbb{R}, |y| \geq 1} y \mu^B(y) dy, \tag{6.6}
\]
cf. [28] Theorem 25.3, Example 25.12].

For $\gamma \in \mathbb{R}$ let
\[
\sigma_\gamma(x) := \frac{e^{(1+\gamma)x}}{(e^x - 1)^{1+\alpha}} - \frac{e^{-(1+\gamma)x}}{(1 - e^{-x})^{1+\alpha}} = \frac{e^{-x}(e^{(\gamma-a+1)x} - 1)}{(1 - e^{-x})^{1+\alpha}}, \quad x > 0.
\]

The next lemma follows immediately from the second expression of $\sigma_\gamma$ above.

**Lemma 6.5.** For every $x > 0$ it holds that $\sigma_\gamma(x) > 0$ for $\alpha < 1 + \gamma$, $\sigma_\gamma(x) = 0$ for $\alpha = 1 + \gamma$, and $\sigma_\gamma(x) > 0$ for $\alpha > 1 + \gamma$.

In the next result, we will also assume that, in addition to (B1)-(B3), $B$ satisfies (B4) Symmetry: $B(x, y) = B(y, x)$ for all $x, y > 0$.
Proposition 6.6. Let $X$ be a pssMp with the infinitesimal generator $\mathcal{L}$ given in (6.3) where the jump kernel is $B(x, y)|x - y|^{-1-\alpha}$ and the linear term given in (6.4). Assume that $B$ satisfies (B1)-(B4) and (6.5). Let $\xi$ be the corresponding Lévy process through the Lamperti transform. Then $E_\xi > 0$ if $\alpha \in (0, 1)$, $E_\xi = 0$ if $\alpha = 1$, and $E_\xi < 0$ if $\alpha \in (1, 2)$.

**Proof.** Note that by Lemma 6.3 and the fact that $B(1, e^y) = B(e^y, 1) = B(1, e^{-y})$, it holds that

$$-\tilde{b} = \text{p.v.} \int_{-1}^{1} y\mu^B(y)dy = \lim_{\epsilon \to 0} \int_{\epsilon}^{1} y(\mu^B(y) - \mu^B(-y))dy = \lim_{\epsilon \to 0} \int_{\epsilon}^{1} y\sigma_0(y)B(1, e^y)dy.$$ 

Similarly,

$$\int_{\mathbb{R}, |y| \geq 1} y\mu^B(y)dy = \int_{1}^{\infty} y(\mu^B(y) - \mu^B(-y))dy = \int_{1}^{\infty} y\sigma_0(y)B(1, e^y)dy.$$ 

The claim now follows from (6.6) and Lemma 6.5.

We can also cover some cases with non-symmetric $B(x, y)$.

Proposition 6.7. Let $X$ be a pssMp with the infinitesimal generator $\mathcal{L}$ given in (6.3) where the jump kernel $J(x, y) = (y/x)\gamma|x - y|^{-1-\alpha}$ with $\gamma \in (-1, \alpha)$ and the linear term is given in (6.4). Let $\xi$ be the corresponding Lévy process through the Lamperti transform. Then $E_\xi > 0$ if $\alpha \in (0, 1 + \gamma)$, $E_\xi = 0$ if $\alpha = 1 + \gamma$, and $E_\xi < 0$ if $\alpha \in (1 + \gamma, 2)$.

In particular, if $\gamma \in [\alpha/2, \alpha)$ then $E_\xi > 0$.

**Proof.** Since $B(x, y) = (y/x)^\gamma$ with $\gamma \in (-1, \alpha)$, we have that (6.5) holds and that $\mu^B(y) - \mu^B(-y) = e^{y\gamma}\mu(y) - e^{-y\gamma}\mu(-y) = \sigma_\gamma(y)$ for $y > 0$. By Lemma 6.3, it holds that

$$-\tilde{b} = \text{p.v.} \int_{-1}^{1} y\mu^B(y)dy = \lim_{\epsilon \to 0} \int_{\epsilon}^{1} y(\mu^B(y) - \mu^B(-y))dy = \lim_{\epsilon \to 0} \int_{\epsilon}^{1} y\sigma_\gamma(y)dy.$$ 

Similarly,

$$\int_{\mathbb{R}, |y| \geq 1} y\mu^B(y)dy = \int_{1}^{\infty} y(\mu^B(y) - \mu^B(-y))dy = \int_{1}^{\infty} y\sigma_\gamma(y)dy.$$ 

The claim now follows from Lemma 6.5.

We end this subsection with a class of examples of modifying functions, satisfying (B1)-(B4), which appeared in our papers $[14, 15]$ on the potential theory of Dirichlet forms with jump kernels decaying at the boundary. For $\beta \geq 0$ and $\gamma \geq 0$ with $\gamma = 0$ if $\beta = 0$, we define

$$\tilde{B}(x, y) = \left(\frac{x \wedge y}{x \vee y}\right)^\beta \left(\log \left(1 + \frac{x \vee y}{x \wedge y}\right)\right)^\gamma. \quad (6.7)$$

It is easy to check that $\tilde{B}(x, y)$ satisfies (B1)-(B4). Since $\frac{x \wedge y}{x - y} \wedge 1 \cong \frac{x \wedge y}{x - y} \wedge 1 \cong 1$, $\tilde{B}(x, y)$ is comparable to the $B(x, y)$ in $[14, (1.8)]$ with $\beta = \beta_1$, $\gamma = \beta_3$ and $\beta_2 = \beta_4 = 0$.

When $B(x, y)$ is equal to $c\tilde{B}(x, y)$, we have that the Lévy measure $\mu^B(y)$ of $\xi$ is equal to $c$ times

\begin{align*}
    \frac{e^y}{|e^y - 1|^{1+\alpha}} &\left(1 \wedge \frac{e^y}{1 \vee e^y}\right)^\beta \left(\log \left(1 + \frac{1 \vee e^y}{1 \wedge e^y}\right)\right)^\gamma \\
    &= \frac{e^y}{|e^y - 1|^{1+\alpha}} \left(1_{(y < 0)}e^{y\beta}(\log(1 + e^{-y}))^\gamma + 1_{(y > 0)}e^{-y\beta}(\log(1 + e^y))^\gamma\right) \\
    &= \frac{e^y}{|e^y - 1|^{1+\alpha}}e^{-|y|\beta}(\log(1 + e^{|y|}))^\gamma.
\end{align*}
If $\gamma = 0$ (so there is no logarithmic term), we see that $\mu^B_*$ is the Lévy measure of a Lamperti stable process in the sense of [7].

Clearly, if $B(x, y)$ is comparable to $\tilde{B}(x, y)$, then the corresponding Lévy measure $\mu^B_*$ satisfies

$$
\mu^B_*(y) \asymp \mu^B(y), \quad y \in \mathbb{R} \setminus \{0\}.
$$

7. APPENDIX

Proof of Lemma 4.2: We claim that for any $0 \leq j \leq k$,

$$
\int_{(0, \infty)} u^j (1 + xu)^{-1-\alpha-k} m(du) = 0, \quad x > 0.
$$

(7.1) is valid for $k = 0$ by assumption. Note that, for $k = 1, 2, \ldots$,

$$
|\frac{\partial^k}{\partial x^k} (1 + xu)^{-1-\alpha}| \leq (1 + xu)^{-1-\alpha}, \quad x > 0, u > 0.
$$

Combining this with the integrability assumption of the lemma, we can exchange the order of the differentiation and integration when we take the derivative of the left hand side of (4.2). Taking derivative with respect to $x$ in (4.2) we get

$$
\int_{(0, \infty)} u (1 + xu)^{-1-\alpha-1} m(du) = 0, \quad x > 0,
$$

(7.2) and so (7.1) is valid for $k = j = 1$. Since

$$
\int_{(0, \infty)} (1 + xu)^{-1-\alpha-1} m(du)
= \int_{(0, \infty)} (1 + xu)^{-1-\alpha} m(du) - \int_{(0, \infty)} xu(1 + xu)^{-1-\alpha-1} m(du) = 0, \quad x > 0,
$$

(where the last equality follows from the assumption and (7.2)), we get that (7.1) is valid for $k = 1$ and $j = 0$. Now suppose that (7.1) is valid for $0 \leq j \leq k$. Taking derivative with respect to $x$ in (7.1), we get

$$
\int_{(0, \infty)} u^{j+1} (1 + xu)^{-1-\alpha-k-1} m(du) = 0, \quad x > 0.
$$

Thus (7.1) is valid for $1 \leq j \leq k + 1$. Noting that

$$
\int_{(0, \infty)} (1 + xu)^{-1-\alpha-k-1} m(du)
= \int_{(0, \infty)} (1 + xu)^{-1-\alpha-k} m(du) - \int_{(0, \infty)} xu(1 + xu)^{-1-\alpha-k-1} m(du) = 0, \quad x > 0,
$$

we get that (7.1) is valid for $0 \leq j \leq k + 1$.

Taking $x = 1$, we get that for any $0 \leq j \leq k$,

$$
\int_{(0, \infty)} \frac{u^j}{(1 + u)^k} (1 + u)^{-1-\alpha} m(du) = 0.
$$

Since the linear span of the set $\{\frac{u^j}{(1 + u)^k} : 0 < j < k\}$ is an algebra of real-valued continuous functions on $(0, \infty)$ which separates points of $(0, \infty)$ and vanishes at infinity, by the Stone-Weierstrass Theorem, the linear span of the set $\{\frac{u^j}{(1 + u)^k} : 0 < j < k\}$ is dense in $C_\infty(0, \infty)$ with respect to the uniform topology. Thus for all $g \in C_\infty(0, \infty)$,

$$
\int_{(0, \infty)} g(u)(1 + u)^{-1-\alpha} m(du) = 0,
$$
which implies \((1+u)^{-1-\alpha}m(du)\) is the zero measure on \((0, \infty)\). Therefore \(m\) is the zero measure on \((0, \infty)\).

\[\square\]

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