STOCHASTIC REPRESENTATION OF A FRACTIONAL SUBDIFFUSION EQUATION. THE CASE OF INFINITELY DIVISIBLE WAITING TIMES, LÉVY NOISE AND SPACE-TIME-DEPENDENT COEFFICIENTS

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Abstract. In this paper we analyze a fractional Fokker-Planck equation describing subdiffusion in the general infinitely divisible (ID) setting. We show that in the case of space-time-dependent drift and diffusion and time-dependent jump coefficients, the corresponding stochastic process can be obtained by subordinating a two-dimensional system of Langevin equations driven by appropriate Brownian and Lévy noises. Our result solves the problem of stochastic representation of subdiffusive Fokker-Planck dynamics in full generality.

1. Preliminaries

Subdiffusion processes are characterized by the asymptotic power-law behavior of the variance $\text{Var}(X(t)) \sim ct^\alpha$ as $t \to \infty$, where $0 < \alpha < 1$. A physical equation describing subdiffusion in the presence of space-dependent force $F(x)$ is the fractional Fokker-Planck equation (FFPE) [28–31]

$$\frac{\partial w(x,t)}{\partial t} = 0D_t^{1-\alpha} \left[ -\frac{\partial}{\partial x} F(x) + \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2} \right] w(x,t),$$

with the scale parameter $\sigma > 0$ and the initial condition $w(x,0) = \delta(x)$. Here

$$0D_t^{1-\alpha} f(t) = \frac{1}{\Gamma(\alpha)} \frac{d}{dt} \int_0^t (t-s)^{\alpha-1} f(s) ds,$$

$0 < \alpha < 1, f \in C^1([0,\infty))$, is the fractional derivative of the Riemann-Liouville type [36]. Equation (1.1) specifies probability density function (PDF) $w(x,t)$ of the subdiffusion process. It was derived in [29] in the framework of continuous-time random walk with heavy-tailed waiting times [26,27].

Stochastic process corresponding to (1.1) has the subordination form

$$Y(t) = X(S_\alpha(t)), \quad t \geq 0,$$

where $X$ is given by the following stochastic differential equation:

$$dX(t) = F(X(t))dt + \sigma dB(t), \quad X(0) = 0,$$

driven by Brownian motion $B$. Moreover, $S_\alpha$ is the inverse $\alpha$-stable subordinator

$$S_\alpha(t) = \inf\{\tau > 0 : T_\alpha(\tau) > t\},$$

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which is assumed to be independent of \( B \). Here, \( T_\alpha \) is the \( \alpha \)-stable subordinator with Laplace transform \( \mathbb{E} \left[ e^{-uT_\alpha(\tau)} \right] = e^{-\tau u^\alpha}, \ 0 < \alpha < 1 \). As shown in \([19, 20]\), PDF of \( Y(t) = X(S_\alpha(t)) \) solves FFPE (1.1). The concept of subordination in terms of a coupled Langevin equation was originally introduced in \([7]\).

FFPE describing subdiffusion in the presence of time-dependent force \( F(t) \) was derived in \([37]\) and has the form

\[
\frac{\partial w(x,t)}{\partial t} = \left[ -F(t) \frac{\partial}{\partial x} + \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2} \right]_0 D_t^{1-\alpha} w(x,t), \tag{1.4}
\]

\( w(x,0) = \delta(x) \). Compared to (1.1), we can observe that the fractional operator \( _0D_t^{1-\alpha} \) in the above equation does not act on the force \( F(t) \), so the force is not influenced by the change of time.

Stochastic process corresponding to (1.4) has the form \([15]\)

\[
Y(t) = \int_0^t F(u) dS_\alpha(u) + \sigma B(S_\alpha(t)), \ t \geq 0, \tag{1.5}
\]

meaning that the PDF of \( Y(t) \) solves (1.4). \( Y(t) \) in (1.5) can be equivalently represented in the form of subordination

\[
Y(t) = X(S_\alpha(t)), \tag{1.6}
\]

where \( X \) is given by

\[
dX(t) = F(T_\alpha(t)) dt + \sigma dB(t), \ X(0) = 0,
\]

and \( S_\alpha \) is the inverse subordinator independent of \( B \). Note that in this case processes \( X \) and \( S_\alpha \) are not independent, since \( T_\alpha \) and its inverse \( S_\alpha \) are strongly dependent; see formula (1.3).

Further relevant studies of FFPE in the \( \alpha \)-stable case include: fractional Cauchy problems \([6, 23, 32]\), path properties of fractional diffusion \([16, 21, 33]\), Pearson diffusion \([13]\) and the case of general forces \([8, 14]\). In the recent paper \([18]\) the case of a general space-time-dependent force and diffusion coefficients is considered.

Extension of (1.5) (or equivalently (1.6)) to the ID case is the following: by \( T_\Psi(t), \ t \geq 0 \) we denote a subordinator (strictly increasing Lévy process) with Laplace transform \([2]\)

\[
\mathbb{E} \left[ e^{-uT_\Psi(t)} \right] = e^{-t\Psi(u)}. \tag{1.3}
\]

Here \( \Psi(u) \) is the Laplace exponent given by

\[
\Psi(u) = \lambda u + \int_0^\infty (1 - e^{-ux})\nu(dx).
\]

We assume for simplicity that \( \lambda = 0 \). The measure \( \nu(dx) \) is the Lévy measure supported in \((0, \infty)\) satisfying \( \int_0^\infty \min(1,x)\nu(dx) < \infty \). To exclude the case of compound Poisson processes, we will further assume that \( \nu((0, \infty)) = \infty \). The corresponding first-passage-time process

\[
S_\Psi(t) = \inf\{\tau > 0 : T_\Psi(\tau) > t\}, \ t \geq 0, \tag{1.7}
\]

is called an inverse subordinator. Inverse subordinators play an important role in probability theory \([2, 3]\), finance \([40]\) and physics \([11, 39]\). Since \( \nu((0, \infty)) = \infty \), the trajectories of \( S_\Psi(t) \) are continuous, which is relevant in physical applications.
Now, to extend (1.5) and (1.6) to the ID case, we replace the inverse stable subordinator $S_\alpha$ with $S_{\Psi}$ and obtain

\begin{equation}
Y(t) = \int_0^t F(u) dS_{\Psi}(u) + \sigma B(S_{\Psi}(t)), \quad t \geq 0,
\end{equation}

or equivalently

\begin{equation}
Y(t) = X(S_{\Psi}(t)).
\end{equation}

In this case $X$ is given by

\begin{equation}
dX(t) = F(T_{\Psi}(t)) dt + \sigma dB(t), \quad X(0) = 0,
\end{equation}

where the subordinator $T_{\Psi}$ is independent of Brownian motion $B$. The process $Y$ defined in (1.8) and (1.9) describes subdiffusion with ID waiting times in the presence of time-dependent force $F(t)$ \cite{17,37}. The corresponding fractional Fokker-Planck equation has the form \cite{17,37}

\begin{equation}
\frac{\partial w(x,t)}{\partial t} = \left[-F(t) \frac{\partial}{\partial x} + \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2}\right] \Phi_t w(x,t).
\end{equation}

The integro-differential operator $\Phi_t$ is defined as

\begin{equation}
\Phi_t f(t) = \frac{d}{dt} \int_0^t M(t-y)f(y)dy,
\end{equation}

where $f$ is a sufficiently smooth function. Moreover, the kernel $M(t)$ is defined via its Laplace transform

\begin{equation}
\mathcal{L}[M](u) = \int_0^\infty e^{-ut} M(t) dt = \frac{1}{\Psi(u)},
\end{equation}

where $\Psi(u)$ is the Laplace exponent of the underlying ID distribution of waiting times.

In addition to the previously mentioned $\alpha$-stable case, other important application examples include: the tempered-stable case \cite{17,22,25,35}, which corresponds to $\Psi(u) = (u+\lambda)^\alpha - \lambda^\alpha$ with $\lambda > 0$ and $0 < \alpha < 1$; the distributed order case \cite{5,12,24}, which corresponds to $\Psi(u) = \int_0^\infty (1 - e^{-ux}) \nu(dx)$ with $\nu(t,\infty) = \int_1^t e^{-x} \mu(dx)$ (here $\beta \in (0,1)$ and $\mu$ is some distribution supported in $[0,1]$).

Another extension of (1.9) is possible when we add a Lévy noise $dL(t)$ into equation (1.10). For the Lévy process $L(t)$ we have the Lévy-Khintchine formula

\begin{equation}
\mathbb{E}e^{iux(t)} = \exp \left[ t \left( ibu - \frac{a^2 u^2}{2} + \int_{\mathbb{R}\setminus\{0\}} (e^{iuy} - 1 - iuy\mathbf{1}_{\{|y|<1\}}) \mu(dy) \right) \right],
\end{equation}

where $b$ is a drift parameter, $a^2$ is connected with the Brownian part of $L(t)$ and $\mu$ is the Lévy measure with $\mu(dy)$ being the intensity of jumps of size $y$. Since in our stochastic differential equations the drift and Brownian part already exist, we may assume without loss of generality that $b = a = 0$.

The idea of stochastic representation of FFPE has become very popular in a physical society. In a very recent paper \cite{41} the authors found the stochastic representation in the case of tempered-stable subordination with space-time dependent coefficients. In the next section we significantly extend the results from \cite{17,18,41}; we consider the general case of space-time-dependent drift and diffusion coefficients, ID waiting times and Lévy noise. Moreover, we analyze the Fokker-Planck equation resulting from adding the Lévy noise $E(t)dL(t)$ to equation (1.10). The general idea...
of the proof of the main theorem is similar to [IS]; however, we focus our attention on necessary changes coming from the ID subordinator and the Lévy noise. The most important differences are presented in the proof of the main theorem. We also put our effort into delivering all necessary mathematical details that were lacking in the physical paper [IS].

2. MAIN RESULT

Now, let us look at the following fractional Fokker-Planck equation:

\[
\frac{\partial}{\partial t}w(x, t) = \left[-\frac{\partial}{\partial x}F(x, t) + \frac{1}{2} \frac{\partial^2}{\partial x^2} \sigma^2(x, t)\right] \Phi_t w(x, t) + \int_{\mathbb{R}\setminus \{0\}} \left(\Phi_t w(r, t) - \Phi_t w(x, t) + E(t)y \frac{\partial}{\partial x} \Phi_t w(x, t) 1_{\{|y|<1\}}\right) \nu(dy),
\]

with \(w(x, 0) = \delta(x)\). It describes the temporal evolution of \(w(x, t)\) – PDF of some anomalous diffusion process with ID waiting times, which is subjected to space-time-dependent force \(F(x, t)\) and diffusion \(\sigma(x, t)\) and with time-dependent jump coefficient \(E(t)\). In the theorem below, which is the main result of the paper, we derive the stochastic process corresponding to (2.1). The result encompasses the whole family of ID waiting times, arbitrary space-time-dependent drift and diffusion coefficients and Lévy jumps with time-dependent jump coefficient. From now on for the left limit process we will use the notation \(X^{-}(t) = \lim_{s \to t^{-}} X(s)\).

**Theorem 2.1.** Let \(T_\Psi(\tau)\) be the subordinator with the Laplace exponent \(\Psi(u)\) and \(S_\Psi(t)\) its inverse. Assume that the standard Brownian motion \(B(t)\), the Lévy process \(L(t)\) with Lévy measure \(\mu\) and \(T_\Psi(\tau)\) are independent. Let \(Y(t)\) be the solution of the stochastic equation

\[
\frac{dY(t)}{dt} = F\left(Y^-(t), T_\Psi(t)\right)dt + \sigma\left(Y^-(t), T_\Psi(t)\right)dB(t) + E(T_\Psi(t))dL(t), \ t \geq 0, \ Y(0) = 0, \ T_\Psi(0) = 0,
\]

where the functions \(F(x, t), \sigma(x, t), E \in C^2(\mathbb{R}^2)\) satisfy the Lipschitz condition. We assume that the PDF of the process \((Y(t), T_\Psi(t))\) - \(p_t(y, z)\) exists and so do \(\frac{\partial}{\partial t}p_t(y, z), \frac{\partial}{\partial y}p_t(y, z)\) and \(\frac{\partial^2}{\partial y^2}p_t(y, z)\). Additionally we require that:

\[
\int_{0}^{t_0} \int_{0}^{\infty} \left| \frac{\partial}{\partial t}p_s(x, t) \right| \ ds dt < \infty
\]

for each \(t_0 > 0\),

\[
\int_{x_1}^{x_2} \int_{0}^{\infty} \left| \frac{\partial}{\partial x}p_s(x, t) \right| \ ds dx < \infty, \ \int_{x_1}^{x_2} \int_{0}^{\infty} \left| \frac{\partial^2}{\partial x^2}p_s(x, t) \right| \ ds dx < \infty
\]

for each \(x_1, x_2 \in \mathbb{R}\) and finally

\[
\int_{0}^{\infty} \int_{\mathbb{R}\setminus \{0\}} \left| p_s(x - E(t)y, t) - p_s(x, t) + \frac{\partial}{\partial x} (E(t)y)p_s(x, t) 1_{\{|y|<1\}}(y) \right| \nu(dy) ds < \infty
\]

for each \(x \in \mathbb{R}, \ t > 0\). Then the PDF of the process \(X(t) = Y^-(S_\Psi(t))\) is the weak solution of FFPE (2.1), that is, (2.1) holds pointwise for \(t > 0\) with the required initial distribution for \(t = 0\).
Remark 2.2. If we impose additional conditions on the coefficients $F, \sigma, E$, the Lévy measures $\mu$ and $\nu$ (of $L(t)$ and $T_\Psi(t)$, respectively) namely: $F, \sigma \in C^3_b(\mathbb{R}^2)$, $E \in C^3_b(\mathbb{R})$ with bounded partial derivatives of the order $1 < 3$, $D(y, z) \neq 0$ for all $y, z \in \mathbb{R}$ and $\int_0^\infty x^p \nu(dx) < \infty$, $\int_\mathbb{R} |x|^p \mu(dx) < \infty$ for all $p \geq 2$, then the density $p_t(y, z)$ exists; see Theorem 2.14 in [1]. Other, less strict non-degeneracy conditions are also discussed there. The regularity assumptions of $p_t(y, z)$ are easy to check when $F(x, t)$ and $\sigma(x, t)$ depend only on $x$ and the processes $Y(t)$ and $T_\Psi(t)$ are independent. Notice that we do not assume anything about the PDF of $X(t)$.

Proof. In the proof we will essentially extend the techniques used in [8, 15, 18]. Equation (2.2) is equivalent to the following system of stochastic equations:

\[
\left( \begin{array}{c}
\frac{dY(t)}{dt} \\
\frac{dZ(t)}{dt}
\end{array} \right) = \left( \begin{array}{cc}
F(Y^-(t), Z^-(t)) & 0 \\
0 & 0
\end{array} \right) dt + \left( \begin{array}{cc}
\sigma(Y^-(t), Z^-(t)) & 0 \\
0 & 0
\end{array} \right) dB(t) \\
+ \left( \begin{array}{c}
E(Z^-(t)) dL(t) \\
0
dT_\Psi(t)
\end{array} \right).
\]

(2.6)

This system is subjected to Brownian and Lévy noise, therefore the infinitesimal generator for the process $(Y(t), Z(t))$ operates on functions $f \in C^3_b(\mathbb{R}^2)$ in the following way (see Theorem 6.7.4 in [1]):

\[
Af(y, z) = F(y, z) \frac{\partial}{\partial y} f(y, z) + \frac{1}{2} \sigma^2(y, z) \frac{\partial^2}{\partial y^2} f(y, z)
\]

\[
+ \int_{\mathbb{R} \setminus \{0\}} \left( f(y + E(z)x, z) - f(y, z) - E(z) x \frac{\partial}{\partial y} f(y, z) 1_{\{|x| < 1\}} \right) \mu(dx)
\]

\[
+ \int_0^\infty [f(y, z + u) - f(y, z)] \nu(du),
\]

where $\nu$ is the Lévy measure of $Z(t)$ (or equivalently $T_\Psi(t)$). We have two separate integrals here, because $L(t)$ and $T_\Psi(t)$ are assumed to be independent which implies that $\varphi$ - Lévy measure of the two-dimensional process $(L(t), T_\Psi(t))$ is supported on the axes of $\mathbb{R}^2$, that is, $\varphi(A) = \mu(\{(y, z) \in \mathbb{R}^2 : (y, 0) \in A\}) + \nu(\{z \in \mathbb{R} : (0, z) \in A\})$ for $A \in Bor(\mathbb{R}^2)$. In the first step of the proof we will obtain an FFPE describing the temporal evolution of the density $p_t(y, z)$. To do this we take advantage of the fact that

\[
\frac{\partial}{\partial t} p_t(y, z) = A^+ p_t(y, z).
\]

(2.8)

Here $A^+$ is the $L^2$ Hermitian adjoint of $A$, meaning that it satisfies the following relation for all test functions $f \in C^3_c(\mathbb{R}^2)$:

\[
\int_{\mathbb{R}^2} Af(y, z) p_t(y, z) dydz = \int_{\mathbb{R}^2} A^+ p_t(y, z) f(y, z) dydz.
\]

(2.9)

If we substitute (2.7) into the above equation, then the left-hand side consists of four summands. The first two summands are dealt with in [18]. Now we turn our attention to the third one - connected with an integration with respect to the
measure $\mu$. Since $f$ has compact support we can apply Fubini’s theorem to get

$$\int_{\mathbb{R}^2} \int_{\mathbb{R}\setminus\{0\}} \left( f(y + E(z)x, z) - f(y, z) - E(z)x \frac{\partial}{\partial y} f(y, z) 1_{\{|x| < 1\}} \right) \mu(dx) p_t(y, z) d(y, z)$$

$$= \int_{\mathbb{R}\setminus\{0\}} \int_{\mathbb{R}} \left( f(y + E(z)x, z) - f(y, z) - E(z)x \frac{\partial}{\partial y} f(y, z) 1_{\{|x| < 1\}} \right) p_t(y, z) dydz\mu(dx).$$

For the inner integral (for each fixed $z$ and $x$) after a substitution and using integration by parts we get

$$\int_{\mathbb{R}} \left( f(y + E(z)x, z) - f(y, z) - E(z)x \frac{\partial}{\partial y} f(y, z) 1_{\{|x| < 1\}} \right) p_t(y, z) dy$$

$$= \int_{\mathbb{R}} f(y, z) \left( p_t(y - E(z)x, z) - p_t(y, z) + E(z)x \frac{\partial}{\partial y} p_t(y, z) 1_{\{|x| < 1\}} \right) dy.$$

Substituting (2.11) back into (2.10) and using Fubini’s theorem again we obtain

$$\int_{\mathbb{R}^2} \int_{\mathbb{R}\setminus\{0\}} \left( f(y + E(z)x, z) - f(y, z) - E(z)x \frac{\partial}{\partial y} f(y, z) 1_{\{|x| < 1\}} \right) \mu(dx) p_t(y, z) d(y, z)$$

$$= \int_{\mathbb{R}^2} f(y, z) \left( p_t(y - E(z)x, z) - p_t(y, z) + E(z)x \frac{\partial}{\partial y} p_t(y, z) 1_{\{|x| < 1\}} \right) \mu(dx) dy(z).$$

Now let us put

$$G(u) = \nu((u, \infty)).$$

Using the general formula for integration by parts (see (21.68) in [9]), we get for the last summand in equation (2.9)

$$\int_0^\infty \left[ f(y, z + u) - f(y, z) \right] \nu(du) = -\lim_{u \to \infty} \left( f(y, z + u) - f(y, z) \right) G(u)$$

$$+ \lim_{u \to 0} \left( f(y, z + u) - f(y, z) \right) G(u) + \int_0^\infty \frac{\partial}{\partial u} f(y, z + u) G(u) du.$$

It is easy to see that the first limit vanishes. For the second limit we also have

$$\lim_{u \to 0} \left( f(y, z + u) - f(y, z) \right) G(u) = \lim_{u \to 0} \frac{f(y, z + u) - f(y, z)}{u} uG(u) = 0,$$

since $f$ is differentiable and $\lim_{u \to 0} uG(u) = 0$. The last equality follows from the fact that if

$$\limsup_{u \to 0} uG(u) > 0,$$
then there exist a decreasing sequence $s_n \to 0$ and $d > 0$ such that for each $n \in \mathbb{N}$
$s_n G(s_n) > d$.

We can now find a subsequence $s_{n_k}$ satisfying
$s_{n_k} \nu((s_{n_{k+1}}, s_{n_k}]) > \frac{d}{2}$
for each $k \in \mathbb{N}$. We reached a contradiction with $\int_{(0,\infty)} \min(1,x) \nu(dx) < \infty$.
Thus the limit equals 0. Taking into account the above calculations, we follow the reasoning from [18].
That gives us for the last summand in (2.9)
$$
\int_{\mathbb{R}^2} \int_0^\infty [f(y, z + u) - f(y, z)] p_t(y, z) \nu(du) dydz
$$
(2.14)
$$
= - \int_{-\mathbb{R}^2} f(y, w) \frac{\partial}{\partial w} \int_0^w G(w - z) p_t(y, z) dz dy dw.
$$

Let us introduce the following operator:
$$
(2.15) \Theta_w g(w) = \int_0^w G(w - z) g(z) dz.
$$

Laplace transform of its kernel equals
$$
(2.16) \mathcal{L}[G](u) = \int_0^\infty e^{-ut} G(t) dt = \int_0^\infty \int_{(t,\infty)} e^{-ut} \nu(ds) dt = \int_0^\infty \frac{1}{u} (1 - e^{-us}) \nu(ds) = \frac{\Psi(u)}{u}.
$$

Combining equations (2.12) and (2.14) we obtain the explicit formula for the adjoint operator $A^+$ with $p_t(y, z)$ in its domain, and from equation (2.8) we finally get
$$
\frac{\partial}{\partial t} p_t(y, z) = - \frac{\partial}{\partial y} (F(y, z) p_t(y, z)) + \frac{\partial^2}{\partial y^2} \left( \frac{1}{2} \sigma^2(y, z) p_t(y, z) \right)
+ \int_{\mathbb{R}\setminus\{0\}} \left( p_t(y - E(z)x, z) - p_t(y, z) + E(z) x \frac{\partial}{\partial y} p_t(y, z) 1_{\{|x| < 1\}} \right) \mu(dx)
- \frac{\partial}{\partial z} \Theta_z p_t(y, z).
$$

We derived the equation for temporal evolution of $p_t(y, z)$ which was our goal in the first step of the proof. Let $w(x, t)$ denote the probability density function of the process $X(t)$. In the next step we will find the relation between the densities $p_t(y, z)$ and $w(x, t)$. We start with denoting paths of the analyzed processes as $X(t, \omega)$ and $(Y(t, \omega), Z(t, \omega))$ for each $\omega \in \Omega$. For each fixed interval $I$, similarly as in [8,18], we define an auxiliary function
$$
(2.18) H_t(s, \omega, u) = \begin{cases} 
1_I(Y^-(s, \omega)) & \text{if } Z^-(s, \omega) \leq t \leq Z^-(s, \omega) + u, \\
0 & \text{otherwise},
\end{cases}
$$
and follow, with appropriate changes, the reasoning from the mentioned papers. This includes the observation
$$
(2.19) 1_I(X(t, \omega)) = \sum_{s>0} H_t(s, \omega, \Delta Z(s, \omega)),
$$
where $\Delta Z(s) = Z(s) - Z^-(s)$. In our case the above equation is also valid since we excluded the case of compound Poisson processes and jumping times of $Z(t, \omega)$
are dense in $[0, \infty)$ almost surely. The second key equation is the compensation formula (Ch. XII, Proposition (1.10) in [34])

\[ (2.20) \quad \mathbb{E} \left[ \sum_{s > 0} H_t(s, \omega, \Delta Z(s, \omega)) \right] = \mathbb{E} \left[ \int_0^\infty \int_0^\infty H_t(s, \omega, u) \nu(du) ds \right]. \]

The difference between our case and [18] is that in the definition of $H_t$, instead of $Y(s, \omega)$, we put $Y^-(s, \omega)$. Due to the Lévy noise the previous process may no longer be continuous, but $Y^-$ is still left-continuous and therefore predictable, which is a condition for using the compensation formula. Another condition is that

\[ \sum_{s \in \mathbb{R}_+: \Delta Z(s) = 0} H_t(s, \omega, \Delta Z(s, \omega)) = 0 \text{ almost surely,} \]

which is also fulfilled. Indeed, assume to the contrary that for $\omega \in W \subset \Omega$ this sum does not vanish and $P(W) \neq 0$. This means that there exists $s > 0$ such that $Z^-(s, \omega) = t = Z(s, \omega)$. However the paths of the process $Z(s, \omega)$ hit the single point $t$ with probability 0 (see [38]), which contradicts the assumption. Consequently,

\[ (2.21) \quad w(x, t) = \int_0^\infty \Theta_t p_s(x, t) ds. \]

Detailed derivation of the above result (with the Riemann-Liouville fractional integral instead of $\Theta_t$) is in [18]. Notice that

\[ (2.22) \quad \frac{d}{dt} \Theta_t f(t) = \Theta_t \frac{d}{dt} f(t) + f(0) G(t) \]

for sufficiently smooth $f$. This can be proven calculating Laplace transform: for the left-hand side of (2.22) we get

\[ \mathcal{L} \left[ \frac{d}{dt} \Theta_t f(t) \right](s) = s \mathcal{L} \left[ \Theta_t f(t) \right](s) = \Psi(s) \mathcal{L} [f(t)](s), \]

whereas the right-hand side equals

\[ \mathcal{L} \left[ \Theta_t \frac{d}{dt} f(t) + f(0) G(t) \right](s) = \Psi(s) \mathcal{L} \left[ \frac{d}{dt} f(t) \right](s) + f(0) \frac{\Psi(s)}{s} = \Psi(s) \mathcal{L} [f(t)](s). \]

Thus

\[ \frac{\partial}{\partial t} \Theta_t p_s(x, t) = \Theta_t \frac{\partial}{\partial t} p_s(x, t) + p_s(x, 0) G(t) \]

and we have the following approximation:

\[ (2.23) \quad \int_0^{t_0} \int_0^\infty \left| \frac{\partial}{\partial t} \Theta_t p_s(x, t) \right| ds dt \leq \int_0^{t_0} \int_0^\infty \left| \Theta_t \frac{\partial}{\partial t} p_s(x, t) \right| ds dt + \int_0^{t_0} \int_0^\infty p_s(x, 0) G(t) ds dt. \]

We deal with both integrals separately. For the first one we observe that

\[ (2.24) \quad \int_0^{t_0} G(u) du = \int_0^{t_0} \int_{(u, \infty)} \mu(dw) du = \int_{(0, \infty)} \min(w, t_0) \mu(dw) = K < \infty, \]
where $K > 0$, and use Fubini’s theorem together with the assumption (2.25)

$$
\int_0^t \int_0^\infty \left| \Theta_t \frac{\partial}{\partial t} p_s(x, t) \right| ds dt \\
\leq \int_0^t \int_0^\infty \int_0^t G(t-u) \left| \frac{\partial}{\partial u} p_s(x, u) \right| duds dt \\
\leq K \int_0^t \int_0^\infty \left| \frac{\partial}{\partial u} p_s(x, u) \right| ds du < \infty.
$$

(2.25)

For the second integral in (2.23) we obtain

$$
\int_0^t \int_0^\infty p_s(x, 0) G(t) ds dt = K \int_0^\infty p_s(x, 0) ds < \infty.
$$

(2.26)

The last inequality is a consequence of Theorem 35.4 in [38]. Combining the approximations for both integrals we get

$$
\int_0^t \int_0^\infty \left| \frac{\partial}{\partial t} \Theta_t p_s(x, t) \right| ds dt < \infty.
$$

Therefore after differentiating equation (2.21) with respect to $t$ we can move the derivative inside the integral:

$$
\frac{\partial}{\partial t} w(x, t) = \int_0^\infty \frac{\partial}{\partial t} \Theta_t p_s(x, t) ds.
$$

(2.28)

Now, applying equation (2.17), taking into account the facts that $\lim_{s \to \infty} p_s(x, t) = 0$ and $p_0(x, t) = \delta_{(0,0)}(x, t)$ - the two-dimensional Dirac delta, we obtain for $t > 0$

$$
\frac{\partial}{\partial t} w(x, t) = \int_0^\infty \left[ \frac{\partial^2}{\partial x^2} \left( \frac{1}{2} \sigma^2(x, t) p_s(x, t) \right) - \frac{\partial}{\partial x} \left( F(x, t) p_s(x, t) \right) \right] ds \\
+ \int_{\mathbb{R} \setminus \{0\}} \left( p_s(x - E(t)y, t) - p_s(x, t) + E(t)y \frac{\partial}{\partial x} p_s(x, t) 1_{\{|y| < 1\}} \right) \nu(dy) \\
- \frac{\partial}{\partial s} p_s(x, t) \right] ds \\
= \frac{\partial^2}{\partial x^2} \left( \frac{1}{2} \sigma^2(x, t) \right) \int_0^\infty p_s(x, t) ds \\
- \frac{\partial}{\partial x} \left( F(x, t) \int_0^\infty p_s(x, t) ds \right) \\
+ \int_{\mathbb{R} \setminus \{0\}} \left[ \int_0^\infty p_s(x - E(t)y, t) ds - \int_0^\infty p_s(x, t) ds \\
+ E(t)y \frac{\partial}{\partial x} \left( \int_0^\infty p_s(x, t) ds \right) 1_{\{|y| < 1\}} \right] \nu(dy).
$$

(2.29)

We can interchange the derivative and the integral here because of the assumptions (2.4). Changing the order of integration, $ds$ with $\nu(dy)$, is justified based on Fubini’s theorem and the assumption (2.5). Next we apply Fubini-Tonelli theorem again, this time to equation (2.21), obtaining

$$
w(x, t) = \int_0^\infty \int_0^t G(t-z) p_s(x, z) dz ds = \Theta_t \int_0^\infty p_s(x, t) ds,
$$

(2.30)
because the function $G(t-z)p_s(x,z)$ is non-negative. Therefore
\begin{equation}
(2.31) \quad \int_0^{\infty} p_s(x,t) ds = \Theta_t^{-1}w(x,t)
\end{equation}
and
\begin{equation}
(2.32) \quad \left. \int_0^{\infty} p_s(x-E(y)t,y) ds = (\Theta_t^{-1}w(r,t)) \right|_{r=x-E(y)t},
\end{equation}
where $\Theta_t^{-1}$ is the left-inverse of the operator $\Theta_t$ defined in (2.15). It turns out that $\Theta_t^{-1} = \Phi_t$ (see equation (1.12)). Indeed, one can easily notice that $\Theta_t$ is a convolution operator with the kernel $G(u)$ and, similarly, $\Phi_t$ is a composition of a derivative operator and a convolution operator with the kernel $M(u)$. Therefore, using equations (2.16) and (1.13) we get
\begin{equation*}
\mathcal{L}[\Phi_t \Theta_t f(t)](u) = u \mathcal{L}[f](u) \mathcal{L}[M](u) \mathcal{L}[G](u) = u \mathcal{L}[f](u) \frac{1}{\Psi(u)} \frac{\Psi(u)}{u} = \mathcal{L}[f](u)
\end{equation*}
for a sufficiently smooth function $f$, which implies that $\Phi_t \Theta_t f(t) = f(t)$. Consequently, from equations (2.29) and (2.31) we obtain the desired result. \hfill \square

The proof cannot be easily extended to the case where the jump coefficient $E$ is space dependent. In such situation the operator $A^+$ does not exist.

**Remark 2.3.** The above theorem can be used to approximate solutions of FFPE (2.1) using Monte Carlo methods based on realizations of the process $X(t)$. Indeed, to simulate trajectories of $X(t)$, one only needs to simulate the process $Y(t)$ (using the standard Euler scheme \[10\]) and the inverse subordinator $S_{\Psi}(t)$ \[15,17\].

It also opens the possibility of analyzing fractional Cauchy problems \[13,23–25\] in the general setting of ID subordinators and arbitrary space-time-dependent drift and diffusion coefficients.

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**References**


