

SOME RESULTS ON THE TELEGRAPH PROCESS DRIVEN BY GAMMA COMPONENTS

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Abstract

We study the integrated telegraph process X_t under the assumption of general distribution for the random times between consecutive reversals of direction. Specifically, X_t represents the position, at time t , of a particle moving U time units upwards with velocity c and D time units downwards with velocity $-c$. The latter motions are repeated cyclically, according to independent alternating renewals. Explicit expressions for the probability law of X_t are given in the following cases: (i) (U, D) gamma-distributed; (ii) U exponentially distributed and D gamma-distributed. For certain values of the parameters involved, the probability law of X_t is provided in a closed form. Some expressions for the moment generating function of X_t and its Laplace transform are also obtained. The latter allows us to prove the existence of a Kac-type condition under which the probability density function of the integrated telegraph process, with identically distributed gamma intertimes, converges to that of the standard Brownian motion.

Finally, we consider the square of X_t and disclose its distribution function, specifying the expression for some choices of the distribution of (U, D) .

Keywords: Random motions with finite velocity; telegraph process; gamma distribution; moment generating function; Kac conditions

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1. Introduction

Random evolutions with finite velocity have drawn the attention of many scientists in the last decades. Indeed, they provide a good alternative to diffusion processes, which often appear unsuitable for describing natural phenomena in life sciences. For instance, the integrated telegraph random process, which is one of the basic models for the description of finite-velocity random motions, can be regarded as a finite-velocity counterpart of the one-dimensional Brownian motion, thanks to the relationship between the corresponding probability density functions. This explains the growing interest in this topic and also the numerous papers published in the literature in recent years. Indeed, starting from the seminal papers [23] and [28], many scientists have devoted their research to this theme by proposing generalizations of the

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integrated telegraph random process (see e.g. [34, 7, 10, 13, 16, 37, 21]) and also applications in many fields (see e.g. [52, 17, 48, 51]). An in-depth analysis of the one-dimensional telegraph process can be found in the books of Kolesnik and Ratanov [33] and Zacks [54]. The present paper is part of the aforementioned research and can be regarded as a further step in the path of investigation of random motions.

The standard integrated telegraph process X_t , $t > 0$, describes the position of a particle traveling at constant speed on the real line. The direction of the motion is reversed according to the arrival epochs of a homogeneous Poisson process. This results in exponentially distributed random times between consecutive reversals of direction of motion.

The process X_t is also called a piecewise linear Markov process or Markovian fluid [2, 46, 49]. Piecewise linear processes were first studied in [22] and represent a subclass of the family of piecewise deterministic processes [11]. A piecewise deterministic process Ξ is defined as $\Xi(t) = \phi_{\epsilon(\tau_n)}(t)$, $\tau_n \leq t < \tau_{n+1}$, where (1) $\epsilon = (\epsilon(t))_{t \geq 0}$ is an arbitrary measurable and adapted process with values in a finite space $\{1, \dots, N\}$, (2) ϕ_1, \dots, ϕ_N are N deterministic flows, and (3) $\{\tau_n\}_{n \geq 1}$ is the sequence of switching times of ϵ . Given a fixed starting point, a piecewise deterministic Markov process evolves according to the flow ϕ_i for an exponentially distributed random time; then a switch occurs, and the evolution is governed by another flow ϕ_j , $j \neq i$, for another exponential time; then it switches again. The standard integrated telegraph process X_t represents the simplest example of a piecewise linear process based on a two-state Markov process $\epsilon(t) \in \{0, 1\}$. Its sample paths are composed of straight lines whose slopes alternate between two values. The process X_t alone is not Markovian, whereas if we supply X_t with a second stochastic process keeping track of the driving flow, we obtain a two-dimensional Markov process.

Piecewise linear processes based on the distribution of inter-switching times different from the exponential ones are much less studied. Some examples can be found in [18] and [47]. Generally, for the integrated telegraph process, there have been many papers in the literature considering a more general setting, but most of them are Markovian. Unfortunately, the assumption of exponentially distributed intertimes is not suitable for many important applications in physics, biology, and engineering, since it gives higher probability to very short intervals. Hence, special attention should be reserved for non-Markovian cases. For instance, in [12], the author analyses the case when the random times separating consecutive velocity reversals of the particle have Erlang distribution with possibly unequal parameters. This hypothesis, if we recall that the sum of independent and identically distributed exponential random variables has Erlang distribution, can be interpreted as stating that the particle undergoes a fixed number of collisions arriving according to a Poisson process before reversing its motion direction. Some connections of the model with queueing, reliability theory, and mathematical finance are also presented in the same paper. The study of a one-dimensional random motion with Erlang distribution for the sojourn times is also performed in [40], where the authors apply the methodology of random evolutions to find the partial differential equations governing the particle motion and obtain a factorization of these equations. Moreover, motivated by applications in mathematical biology concerning randomly alternating motion of micro-organisms, in [14] the authors consider gamma-distributed random intertimes and obtain the probability law of the process, expressed in terms of series of incomplete gamma functions. Similarly, the case of exponentially distributed random intertimes with linearly increasing parameters has been treated in [15], thus modeling a damping behavior sometimes appearing in particle systems. We recall also the papers [41] and [42], where isotropic random motions in higher dimensions with, respectively, Erlang- and gamma-distributed steps are studied.

In the present paper, along the lines of [19] and [54], we provide a general expression for the probability law of X_t in terms of the distributions of the n -fold convolutions of the random times between motion reversals. The key idea is to relate the probability law of X_t to that of the occupation time, which gives the fraction of time that the motion moved with positive velocity in $[0, t]$. Starting from this result, Section 3 is devoted to the explicit derivations of the probability law of the integrated telegraph process in some special cases, when the upward U and downward D random intertimes are distributed as follows: (i) both are gamma-distributed; (ii) both are Erlang-distributed; (iii) U is exponentially distributed and D gamma-distributed; (iv) U is exponentially distributed and D Erlang-distributed. In Cases (i) and (ii), for some values of the parameters involved, we obtain closed-form results for the probability law of the process, which, to the best of our knowledge, is a new result in the field of finite-time random evolutions. This shows the great advantage of this novel expression for the probability law compared to previously existing results: it has good mathematical tractability. Hence, Section 3 offers a collection of explicit expressions for the probability law of the generalized telegraph process, which could be useful for all scholars interested in non-Markovian generalizations of the telegraph process.

In Section 4 we obtain the Laplace transform of the moment generating function $M_{X_t}^s(t)$ of the generalized telegraph process. This result allows us to do the following:

- (a) Prove the existence of a Kac-type condition for the integrated telegraph process with identically distributed gamma intertimes. Under such a condition, the probability density function of X_t converges to the probability density function of the standard Brownian motion, thus generalizing the well-known result for the standard integrated telegraph process.
- (b) Provide an explicit expression for $M_{X_t}^s(t)$ in the case of gamma- and Erlang-distributed random times between consecutive velocity reversals.

As further validation, we also derive the moment generating function for exponentially distributed intertimes, finding a well-known result. The first and second moment of X_t under the assumptions of gamma-distributed intertimes are also given.

Finally, Section 5 is devoted to the study of certain functionals of the generalized telegraph process. In particular, we provide an explicit expression for the distribution function of the square of X_t , under the assumption of exponential distribution for the upward intertimes U and gamma distribution for the downward intertimes D . We recall that the square of the standard integrated telegraph process has been studied in [35], where its relationship with the square of the Brownian motion is also stressed. See also [33, Chapter 7] and [32] for other relevant functionals of the telegraph processes.

2. The generalized telegraph process

Let $\{X_t; t \geq 0\}$ be a generalized integrated telegraph process. Such a process describes the position of a particle moving on the real axis with velocity c or $-c$ ($c > 0$), according to an independent alternating counting process $\{N_t; t \geq 0\}$. The latter is governed by sequences of positive independent random times $\{U_1, U_2, \dots\}$ and $\{D_1, D_2, \dots\}$, which in turn are assumed to be mutually independent. The random variable U_i (resp. D_i), $i = 1, 2, \dots$, describes the i th random period during which the motion proceeds with positive (resp. negative) velocity. A sample path of X_t with initial velocity $V_0 = c$ is shown in Figure 1.

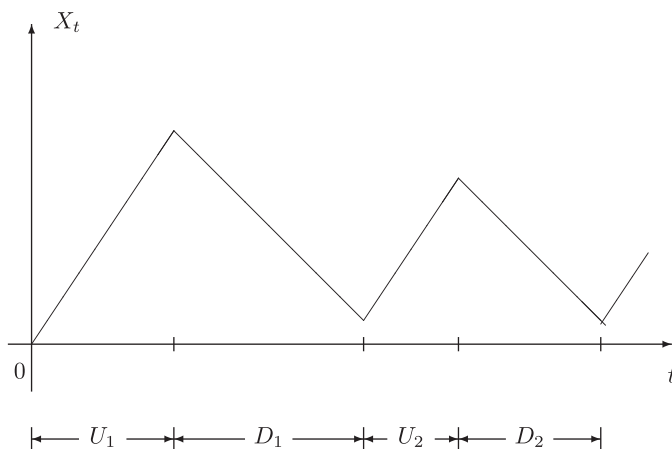


FIGURE 1. A sample path of X_t .

Let us denote by V_t the particle velocity at time $t \geq 0$. Assuming that $X_0 = 0$, and $V_0 \in \{-c, c\}$, with V_0 independent of N_t , we have

$$X_t = \int_0^t V_s \, ds, \quad V_t = \text{sgn}(V_0) c(-1)^{N_t},$$

where

$$N_t = \sum_{n=1}^{+\infty} \mathbf{1}_{\{T_n \leq t\}}, \quad N_0 = 0,$$

with

$$T_{2n} = U^{(n)} + D^{(n)}, \quad T_{2n+1} = T_{2n} + \begin{cases} U_{n+1} & \text{if } V_0 = c, \\ D_{n+1} & \text{if } V_0 = -c, \end{cases} \quad n = 0, 1, \dots,$$

and $U^{(0)} = D^{(0)} = 0$, $U^{(n)} := U_1 + \dots + U_n$, $D^{(n)} := D_1 + \dots + D_n$.

Hence, if the motion does not change velocity in $[0, t]$, we have $X_t = V_0 t$. Otherwise, if there is at least one velocity change in $[0, t]$, then $-ct < X_t < ct$ with probability 1.

Hereafter, according to the assumptions of the standard symmetric telegraph process [28], we assume that the initial velocity is random, i.e.

$$P(V_0 = c) = P(V_0 = -c) = \frac{1}{2}.$$

Let $F_{U_i}(\cdot)$ and $G_{D_i}(\cdot)$ be the absolutely continuous distribution functions of U_i and D_i ($i = 1, 2, \dots$) respectively, with densities $f_{U_i}(\cdot)$ and $g_{D_i}(\cdot)$, and denote by $\bar{F}_{U_i}(\cdot)$ and $\bar{G}_{D_i}(\cdot)$ their complementary distribution functions. In the sequel, we shall denote the distribution functions of $U^{(n)}$ and $D^{(n)}$ by $F_U^{(n)}(\cdot)$ and $G_D^{(n)}(\cdot)$, and their densities by $f_U^{(n)}(\cdot)$ and $g_D^{(n)}(\cdot)$, respectively. If the random variables U_i are independent and identically distributed for $i = 1, 2, \dots, n$, then $F_U^{(n)}(\cdot)$ is the n -fold convolution of $F_{U_1}(\cdot)$, and similarly for $G_D^{(n)}(\cdot)$. Moreover, we set $F_U^{(0)}(x) = G_D^{(0)}(x) = 1$ for $x \geq 0$.

In order to provide the probability law of X_t , we refer to the general method proposed by Zacks in [53]. The key idea is to relate the probability law of X_t to that of the occupation time

$$W_t = \int_0^t \mathbf{1}_{\{V_s=c\}}(s)ds, \quad t > 0,$$

which gives the fraction of time that the motion moved with positive velocity in $[0, t]$. Indeed, for all $t \geq 0$, X_t and W_t are linked by the following relationship:

$$X_t = 2c W_t - ct.$$

Hence, denoting by

$$\psi(x, t) := \frac{\partial}{\partial x} P(W_t \leq x), \quad 0 < x < t, \quad t > 0,$$

the absolutely continuous component of the probability law of W_t over $(0, t)$, we can express the probability law of X_t in terms of that of W_t , according to the following theorem.

Proposition 2.1. *For all $t > 0$ we have*

$$P(X_t = -ct) = P(W_t = 0), \quad P(X_t = ct) = P(W_t = t),$$

and

$$f(x, t) := \frac{\partial}{\partial x} \mathbb{P}(X_t \leq x) = \frac{1}{2c} \psi\left(\frac{x + ct}{2c}, t\right), \quad -ct < x < ct. \tag{1}$$

The distribution of W_t was derived by Perry et al. (see [38], [39]) in terms of the distribution of the first time at which a compound process crosses a linear boundary. Proceeding along the lines of Lemmas 5.1 and 5.2 of [54], in the following theorem we provide the probability law of W_t .

Theorem 2.1. *For all $t > 0$ it holds that*

$$P(W_t = 0) = \frac{1}{2} \bar{G}_{D_1}(t), \quad P(W_t = t) = \frac{1}{2} \bar{F}_{U_1}(t),$$

and, for $0 < x < t$,

$$\psi(x, t) = \frac{1}{2} [\psi_{-c}(x; t; c) + \psi_c(x; t; c) + \psi_{-c}(x; t; -c) + \psi_c(x; t; -c)],$$

where

$$\begin{aligned} \psi_c(x, t; c) &:= \frac{\partial}{\partial x} P(W_t \leq x, V_t = c \mid V_0 = c) = \sum_{n=1}^{+\infty} [F_U^{(n)}(x) - F_U^{(n+1)}(x)] g_D^{(n)}(t-x), \\ \psi_{-c}(x, t; c) &:= \frac{\partial}{\partial x} P(W_t \leq x, V_t = c \mid V_0 = -c) = \sum_{n=0}^{+\infty} [F_U^{(n)}(x) - F_U^{(n+1)}(x)] g_D^{(n+1)}(t-x), \\ \psi_c(x, t; -c) &:= \frac{\partial}{\partial x} P(W_t \leq x, V_t = -c \mid V_0 = c) = \sum_{n=0}^{+\infty} [G_D^{(n)}(t-x) - G_D^{(n+1)}(t-x)] f_U^{(n+1)}(x), \\ \psi_{-c}(x, t; -c) &:= \frac{\partial}{\partial x} P(W_t \leq x, V_t = -c \mid V_0 = -c) = \sum_{n=1}^{+\infty} [G_D^{(n)}(t-x) - G_D^{(n+1)}(t-x)] f_U^{(n)}(x). \end{aligned}$$

Using Theorem 2.1 and Proposition 2.1, we finally obtain the expression for the probability law of X_t .

Theorem 2.2. For all $t > 0$ it holds that

$$P(X_t = -ct) = \frac{1}{2} \bar{G}_{D_1}(t), \quad P(X_t = ct) = \frac{1}{2} \bar{F}_{U_1}(t), \tag{2}$$

and, for $-ct < x < ct$,

$$f(x, t) := \frac{\partial}{\partial x} \mathbb{P}(X_t \leq x) = \frac{1}{2} [f_c(x, t) + f_{-c}(x, t)], \tag{3}$$

where

$$\begin{aligned} f_c(x, t) := & \frac{\partial}{\partial x} \mathbb{P}(X_t \leq x \mid V_0 = c) = \frac{1}{2c} \left\{ \sum_{n=0}^{+\infty} \left[G_D^{(n)} \left(t - \frac{ct+x}{2c} \right) - G_D^{(n+1)} \left(t - \frac{ct+x}{2c} \right) \right] \right. \\ & \left. \times f_U^{(n+1)} \left(\frac{ct+x}{2c} \right) + \sum_{n=1}^{+\infty} \left[F_U^{(n)} \left(\frac{ct+x}{2c} \right) - F_U^{(n+1)} \left(\frac{ct+x}{2c} \right) \right] g_D^{(n)} \left(t - \frac{ct+x}{2c} \right) \right\}, \end{aligned} \tag{4}$$

and

$$\begin{aligned} f_{-c}(x, t) := & \frac{\partial}{\partial x} \mathbb{P}(X_t \leq x \mid V_0 = -c) = \frac{1}{2c} \left\{ \sum_{n=0}^{+\infty} \left[F_U^{(n)} \left(\frac{ct+x}{2c} \right) - F_U^{(n+1)} \left(\frac{ct+x}{2c} \right) \right] \right. \\ & \times g_D^{(n+1)} \left(t - \frac{ct+x}{2c} \right) + \sum_{n=1}^{+\infty} \left[G_D^{(n)} \left(t - \frac{ct+x}{2c} \right) \right. \\ & \left. \left. - G_D^{(n+1)} \left(t - \frac{ct+x}{2c} \right) \right] f_U^{(n)} \left(\frac{ct+x}{2c} \right) \right\}. \end{aligned} \tag{5}$$

An alternative approach to disclosing the probability law of the generalized integrated telegraph process is based on the resolution of the hyperbolic system of partial differential equations related to the probability density of (X_t, V_t) . For instance, in [15], for $t > 0$, $-ct < x < ct$, $j = 1, 2$, and $n = 1, 2, \dots$, the authors define the conditional densities of (X_t, V_t) , joint with $\{T_{2n-j} \leq t < T_{2n-j+1}\}$, and provide the relative system of partial differential equations. Unfortunately, the resolution of such a system is so hard a task that the authors are forced to follow a different approach.

3. Special cases of the probability law of X_t

In this section we make use of Theorem 2.2 to obtain an explicit expression for the probability law of the motion under suitable choices of $F_{U_i}(\cdot)$ and $G_{D_i}(\cdot)$, $i = 1, 2, \dots$. First of all, in Theorem 3.1 we assume that the random intertimes U_i and D_i are identically gamma-distributed. Figure 2 provides some simulated sample paths of the related integrated telegraph process for two different values of the coefficient of variation. We recall that the joint probability law of $\{(X_t, V_t), t \geq 0\}$, under the assumption of gamma-distributed intertimes, has been expressed in [14] in terms of a series of the incomplete gamma function. In the next theorem, we provide an expression for the probability law of X_t in terms of the generalized Wright function. The latter is a very simple and mathematically tractable expression, and, as shown in Propositions (3.2) and (3.3), it allows us to obtain closed-form results for the probability law

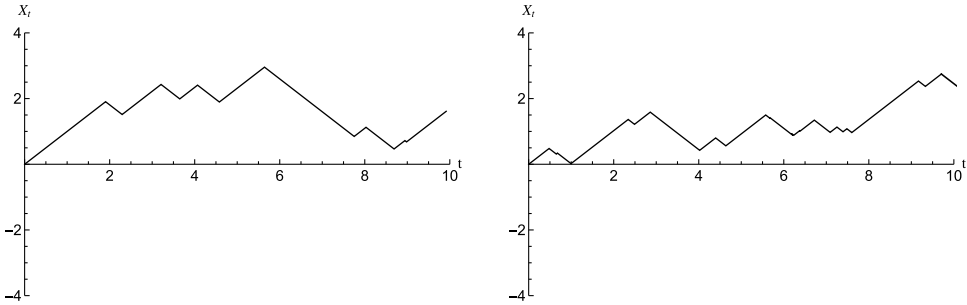


FIGURE 2. Simulated sample paths of X_t under the assumptions of gamma intertimes with coefficient of variation less than 1 (left-hand side) or greater than 1 (right-hand side).

of the generalized integrated telegraph process in the case of fixed values of the parameters involved. These results appear to be of great interest owing to the absence of similar outcomes in the literature for the one-dimensional case.

Theorem 3.1. For all $i = 1, 2, \dots$, let U_i and D_i be gamma-distributed with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$, and set

$$A_i^k(x, t) := {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \middle| (k + 1 + \alpha, \alpha) \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right], \tag{6}$$

where, for $z \in \mathbb{C}$, $a_i, b_j \in \mathbb{C}$, $\alpha_i, \beta_j \in \mathbb{R}$, $\alpha_i, \beta_j \neq 0$,

$${}_p\Psi_q \left[\begin{matrix} (a_1, \alpha_1)_{1,p} \\ (b_1, \beta_1)_{1,q} \end{matrix} \middle| z \right] = {}_p\Psi_q \left[\begin{matrix} (a_1, \alpha_1) \cdots (a_p, \alpha_p) \\ (b_1, \beta_1) \cdots (b_q, \beta_q) \end{matrix} \middle| z \right] := \sum_{k=0}^{+\infty} \frac{\prod_{l=1}^p \Gamma(a_l + \alpha_l k)}{q \prod_{j=1}^q \Gamma(b_j + \beta_j k)} \frac{z^k}{k!} \tag{7}$$

is the generalized Wright function. Then, for $t > 0$, we have

$$P(X_t = -ct \mid X_0 = 0) = P(X_t = ct \mid X_0 = 0) = \frac{1}{2} \left[1 - \frac{\gamma(\alpha, \beta t)}{\Gamma(\alpha)} \right],$$

with $\gamma(s, x)$ denoting the lower incomplete gamma function, and, for $-ct < x < ct$,

$$f(x, t) = \frac{\beta^\alpha e^{-\beta t}}{4c} \left\{ \sum_{k=0}^{+\infty} \beta^k A_0^k(x, t) \left[\left(\frac{ct+x}{2c} \right)^{\alpha-1} \left(\frac{ct-x}{2c} \right)^k + \left(\frac{ct-x}{2c} \right)^{\alpha-1} \left(\frac{ct+x}{2c} \right)^k \right] - \sum_{k=0}^{+\infty} \beta^{k+2\alpha} A_2^k(x, t) \left[\left(\frac{ct+x}{2c} \right)^{\alpha-1} \left(\frac{ct-x}{2c} \right)^{k+2\alpha} + \left(\frac{ct-x}{2c} \right)^{\alpha-1} \left(\frac{ct+x}{2c} \right)^{k+2\alpha} \right] \right\}. \tag{8}$$

Proof. The first result is due to Equation (2). Under the given assumptions, recalling that

$$f_U^{(n)}(x) = g_D^{(n)}(x) = \frac{\beta^{n\alpha} x^{n\alpha-1} e^{-\beta x}}{\Gamma(n\alpha)}, \quad F_U^{(n)}(x) = G_D^{(n)}(x) = \frac{\gamma(n\alpha, \beta x)}{\Gamma(n\alpha)}, \tag{9}$$

and by setting

$$\gamma^*(\alpha, x) = \frac{x^{-\alpha}}{\Gamma(\alpha)} \gamma(\alpha, x) \tag{10}$$

(cf. Equation 6.5.4 of [1]), from Equation (4) we get

$$\begin{aligned} f_c(x, t) &= \frac{1}{2c} \left\{ \sum_{n=0}^{+\infty} \left[\frac{\gamma(n\alpha, \beta(\frac{ct-x}{2c}))}{\Gamma(n\alpha)} - \frac{\gamma((n+1)\alpha, \beta(\frac{ct-x}{2c}))}{\Gamma((n+1)\alpha)} \right] \right. \\ &\quad \times \frac{\beta^{(n+1)\alpha} (\frac{ct+x}{2c})^{n\alpha+\alpha-1} e^{-\beta(\frac{ct+x}{2c})}}{\Gamma((n+1)\alpha)} \\ &\quad \left. + \sum_{n=1}^{+\infty} \left[\frac{\gamma(n\alpha, \beta(\frac{ct+x}{2c}))}{\Gamma(n\alpha)} - \frac{\gamma((n+1)\alpha, \beta(\frac{ct+x}{2c}))}{\Gamma((n+1)\alpha)} \right] \frac{\beta^{n\alpha} (\frac{ct-x}{2c})^{n\alpha-1} e^{-\beta(\frac{ct-x}{2c})}}{\Gamma(n\alpha)} \right\} \\ &= \frac{1}{2c} \left\{ \sum_{n=0}^{+\infty} \left[\beta\left(\frac{ct-x}{2c}\right) \right]^{n\alpha} \gamma^*\left(n\alpha, \beta\left(\frac{ct-x}{2c}\right)\right) \frac{\beta^{n\alpha+\alpha} (\frac{ct+x}{2c})^{n\alpha+\alpha-1} e^{-\beta(\frac{ct+x}{2c})}}{\Gamma((n+1)\alpha)} \right. \\ &\quad - \sum_{n=0}^{+\infty} \left[\beta\left(\frac{ct-x}{2c}\right) \right]^{n\alpha+\alpha} \gamma^*\left((n+1)\alpha, \beta\left(\frac{ct-x}{2c}\right)\right) \frac{\beta^{n\alpha+\alpha} (\frac{ct+x}{2c})^{n\alpha+\alpha-1} e^{-\beta(\frac{ct+x}{2c})}}{\Gamma((n+1)\alpha)} \\ &\quad + \sum_{n=1}^{+\infty} \left[\beta\left(\frac{ct+x}{2c}\right) \right]^{n\alpha} \gamma^*\left(n\alpha, \beta\left(\frac{ct+x}{2c}\right)\right) \frac{\beta^{n\alpha} (\frac{ct-x}{2c})^{n\alpha-1} e^{-\beta(\frac{ct-x}{2c})}}{\Gamma(n\alpha)} \\ &\quad \left. - \sum_{n=1}^{+\infty} \left[\beta\left(\frac{ct+x}{2c}\right) \right]^{n\alpha+\alpha} \gamma^*\left((n+1)\alpha, \beta\left(\frac{ct+x}{2c}\right)\right) \frac{\beta^{n\alpha} (\frac{ct-x}{2c})^{n\alpha-1} e^{-\beta(\frac{ct-x}{2c})}}{\Gamma(n\alpha)} \right\}. \tag{11} \end{aligned}$$

Let us focus on the first term on the right-hand side of (11). Recalling that

$$\gamma^*(a, z) = e^{-z} \sum_{n=0}^{+\infty} \frac{z^n}{\Gamma(a+n+1)}, \tag{12}$$

and using Equation (7), we obtain

$$\begin{aligned} &\sum_{n=0}^{+\infty} \left[\beta\left(\frac{ct-x}{2c}\right) \right]^{n\alpha} \gamma^*\left(n\alpha, \beta\left(\frac{ct-x}{2c}\right)\right) \frac{\beta^{n\alpha+\alpha} (\frac{ct+x}{2c})^{n\alpha+\alpha-1} e^{-\beta(\frac{ct+x}{2c})}}{\Gamma((n+1)\alpha)} \\ &= \beta^\alpha \left(\frac{ct+x}{2c}\right)^{\alpha-1} e^{-\beta(\frac{ct+x}{2c})} \sum_{n=0}^{+\infty} \frac{\left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2}\right) \right]^{n\alpha}}{\Gamma((n+1)\alpha)} e^{-\beta(\frac{ct-x}{2c})} \sum_{k=0}^{+\infty} \frac{\left[\beta\left(\frac{ct-x}{2c}\right) \right]^k}{\Gamma(n\alpha+k+1)} \\ &= \beta^\alpha \left(\frac{ct+x}{2c}\right)^{\alpha-1} e^{-\beta t} \sum_{k=0}^{+\infty} \left[\beta\left(\frac{ct-x}{2c}\right) \right]^k \sum_{n=0}^{+\infty} \frac{\left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2}\right) \right]^{n\alpha}}{\Gamma((n+1)\alpha) \Gamma(n\alpha+k+1)} \\ &= \beta^\alpha \left(\frac{ct+x}{2c}\right)^{\alpha-1} e^{-\beta t} \sum_{k=0}^{+\infty} \left[\beta\left(\frac{ct-x}{2c}\right) \right]^k \left[{}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha)(k+1, \alpha) \end{matrix} \middle| \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2}\right) \right]^\alpha \right] \right]. \end{aligned}$$

By applying similar reasoning for each term in the right-hand side of (11), we finally get

$$\begin{aligned}
 f_c(x, t) = & \frac{e^{-\beta t}}{2c} \left\{ \beta^\alpha \left(\frac{ct+x}{2c} \right)^{\alpha-1} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^k \right. \\
 & \times {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \quad (k+1, \alpha) \mid \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] \\
 & - \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \left(\frac{ct+x}{2c} \right)^{-1} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^k \\
 & \times {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \quad (\alpha+k+1, \alpha) \mid \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] \\
 & + \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \left(\frac{ct-x}{2c} \right)^{-1} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct+x}{2c} \right) \right]^k \\
 & \times {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \quad (\alpha+k+1, \alpha) \mid \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] \\
 & - \beta^\alpha \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \left(\frac{ct+x}{2c} \right)^\alpha \left(\frac{ct-x}{2c} \right)^{-1} \\
 & \times \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct+x}{2c} \right) \right]^k {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \quad (2\alpha+k+1, \alpha) \mid \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] \left. \right\}. \tag{13}
 \end{aligned}$$

Similarly, in the case of negative initial velocity, we have

$$\begin{aligned}
 f_{-c}(x, t) = & \frac{e^{-\beta t}}{2c} \left\{ \beta^\alpha \left(\frac{ct-x}{2c} \right)^{\alpha-1} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct+x}{2c} \right) \right]^k \right. \\
 & \times {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \quad (k+1, \alpha) \mid \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] \\
 & - \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \left(\frac{ct-x}{2c} \right)^{-1} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct+x}{2c} \right) \right]^k \\
 & \times {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \quad (\alpha+k+1, \alpha) \mid \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] \\
 & + \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \left(\frac{ct+x}{2c} \right)^{-1} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^k \\
 & \times {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \quad (\alpha+k+1, \alpha) \mid \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] \left. \right\}.
 \end{aligned}$$

$$\begin{aligned}
 &-\beta^\alpha \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \left(\frac{ct - x}{2c} \right)^\alpha \left(\frac{ct + x}{2c} \right)^{-1} \\
 &\times \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct - x}{2c} \right) \right]^k {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha) \end{matrix} \middle| \begin{matrix} (2\alpha + k + 1, \alpha) \end{matrix} \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] \Bigg\}.
 \end{aligned} \tag{14}$$

The proof finally follows from recalling the assumption of random initial velocity. □

Hereafter we analyze the behavior of the density $f(x, t)$ given in Equation (8) at the extreme points of the interval $(-ct, ct)$, for any fixed t .

Proposition 3.1. *For $t, \alpha, \beta > 0$, it holds that*

$$\lim_{x \rightarrow (ct)^-} f(x, t) = \lim_{x \rightarrow (-ct)^+} f(x, t) = \begin{cases} \frac{1}{4c} \frac{e^{-\beta t} \beta^\alpha t^{\alpha-1}}{\Gamma(\alpha)}, & \text{if } \alpha \geq 1, \\ +\infty, & \text{if } \alpha < 1. \end{cases}$$

Proof. Because of the symmetry properties of X_t , it is enough to analyze the behavior of $f(x, t)$ as $x \rightarrow (ct)^-$.

Let us fix $t, \alpha, \beta > 0$ and assume $-ct < x < ct$. By Corollary 1.1 of [30], the generalized Wright functions appearing in the density (8) are entire functions for all nonnegative integers k . Hence they are continuous, and, recalling their analytical expression (7), we obtain

$$\begin{aligned}
 \lim_{x \rightarrow (ct)^-} {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha)(k + 1, \alpha) \end{matrix} \middle| \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] &= \frac{1}{\Gamma(\alpha) \Gamma(k + 1)}, \\
 \lim_{x \rightarrow (ct)^-} {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (\alpha, \alpha)(2\alpha + k + 1, \alpha) \end{matrix} \middle| \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha \right] &= \frac{1}{\Gamma(\alpha) \Gamma(2\alpha + k + 1)},
 \end{aligned}$$

so that, by Equation (12), after simple calculations we get

$$\begin{aligned}
 \lim_{x \rightarrow (ct)^-} f(x, t) &= \frac{\beta^\alpha}{4c} e^{-\beta t} \left\{ \frac{t^{\alpha-1}}{\Gamma(\alpha)} - \frac{(\beta t)^{2\alpha}}{\Gamma(\alpha)(2c)^{\alpha-1}} \sum_{k=0}^{+\infty} \frac{(\beta t)^k}{\Gamma(2\alpha + k + 1)} \lim_{x \rightarrow (ct)^-} (ct - x)^{\alpha-1} \right. \\
 &\quad \left. + \frac{1}{\Gamma(\alpha)} \left(\frac{1}{2c} \right)^{\alpha-1} \sum_{k=0}^{+\infty} \frac{(\beta t)^k}{\Gamma(k + 1)} \lim_{x \rightarrow (ct)^-} (ct - x)^{\alpha-1} \right\} \\
 &= \frac{e^{-\beta t} (\beta t)^\alpha}{4ct \Gamma(\alpha)} + \frac{1}{2\Gamma(\alpha)} \left(\frac{\beta}{2c} \right)^\alpha \left(1 - \frac{\gamma(2\alpha, \beta t)}{\Gamma(2\alpha)} \right) \lim_{x \rightarrow (ct)^-} (ct - x)^{\alpha-1}. \quad \square
 \end{aligned}$$

In the following proposition we provide a closed-form expression for the probability density function (8) in the case $\alpha = 1/2$.

Proposition 3.2. *In the case $\alpha = 1/2$ the probability density function (8) has the following expression:*

$$f(x, t) = \frac{1}{4c} \sqrt{\frac{\beta}{\pi}} e^{\frac{\beta}{c}(\sqrt{c^2 t^2 - x^2} - ct)} \left\{ \left(\frac{ct + x}{2c} \right)^{-\frac{1}{2}} + \left(\frac{ct - x}{2c} \right)^{-\frac{1}{2}} \right\}, \quad -ct < x < ct.$$

Proof. For $\alpha = 1/2$ and recalling that, for $n \in \mathbb{N}$,

$$\Gamma\left(n + \frac{1}{2}\right) = \frac{(2n)!}{4^n n!} \sqrt{\pi}, \quad (15)$$

one can easily prove that $A_2^k(x, t) = A_0^{k+1}(x, t)$, where the function $A_l^k(x, t)$ has been defined in (6). Hence we obtain

$$f(x, t) = \frac{\sqrt{\beta} e^{-\beta t}}{4c} A_0^0(x, t) \left\{ \left(\frac{ct+x}{2c}\right)^{-\frac{1}{2}} + \left(\frac{ct-x}{2c}\right)^{-\frac{1}{2}} \right\}. \quad (16)$$

Moreover, by setting

$$z := \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2} \right) \right]^\alpha,$$

and using Equation (15), we get

$$\begin{aligned} A_0^0(z, t) &= \sum_{n=0}^{+\infty} \frac{z^{2n}}{\Gamma\left(n + \frac{1}{2}\right) \Gamma(n+1)} + \sum_{n=0}^{+\infty} \frac{z^{2n+1}}{\Gamma(n+1) \Gamma\left(1 + \frac{1}{2} + n\right)} \\ &= \frac{1}{\sqrt{\pi}} \sum_{n=0}^{+\infty} \frac{(2z)^{2n}}{(2n)!} + \frac{2}{\sqrt{\pi}} \sum_{n=0}^{+\infty} \frac{(2z)^{2n+1} (n+1)!}{n! (2n+2)!} \\ &= \frac{1}{\sqrt{\pi}} \cosh(2z) + \frac{1}{\sqrt{\pi}} \sinh(2z) = \frac{1}{\sqrt{\pi}} e^{2z}. \end{aligned}$$

Finally, the proof follows from substituting the previous expression in (16) and recalling the definition of z . \square

Starting from Theorem 3.1, in the next proposition we provide the expression for the probability law of X_t under the assumption of identically Erlang-distributed random intertimes U_i and D_i , $i = 1, 2, \dots$. Such results are in agreement with those obtained in [19].

Corollary 3.1. *If U_i and D_i , $i = 1, 2, \dots$, both have Erlang distribution with parameters $m \in \mathbb{N}$ and $\beta > 0$, then for $t > 0$, we have*

$$P(X_t = -ct \mid X_0 = 0) = P(X_t = ct \mid X_0 = 0) = \frac{1}{2} \left[1 - \frac{\gamma(m, \beta t)}{(m-1)!} \right],$$

and for $-ct < x < ct$, we have

$$\begin{aligned} f(x, t) &= \frac{\beta}{4c} e^{-\beta t} \left\{ \sum_{j=0}^{2m-1} S_{(m-1, j)}^{(m, m)} \left[\beta \left(\frac{ct+x}{2c} \right), \beta \left(\frac{ct-x}{2c} \right) \right] \right. \\ &\quad \left. + \sum_{j=0}^{2m-1} S_{(j, m-1)}^{(m, m)} \left[\beta \left(\frac{ct+x}{2c} \right), \beta \left(\frac{ct-x}{2c} \right) \right] \right\}, \end{aligned}$$

where

$$S_{i, j}^{(k, r)}(x, y) = \sum_{l=0}^{+\infty} \frac{x^{kl+i} y^{rl+j}}{(kl+i)! (rl+j)!}, \quad k, r \geq 1, \quad i, j \geq 0, \quad (17)$$

is a two-index pseudo-Bessel function.

Proof. The result follows from Theorem 3.1 by letting $\alpha = m \in \mathbb{N}$. For instance, for the first term in (8), by (7), and recalling Equation (17), we have

$$\begin{aligned} & \frac{\beta^m}{4c} e^{-\beta t} \left(\frac{ct+x}{2c}\right)^{m-1} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct-x}{2c}\right)\right]^k {}_1\Psi_2 \left[\begin{matrix} (1, 1) \\ (m, m)(k+1, m) \end{matrix} \middle| \left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2}\right)\right]^m \right] \\ &= \frac{\beta^m}{4c} e^{-\beta t} \left(\frac{ct+x}{2c}\right)^{m-1} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct-x}{2c}\right)\right]^k \sum_{h=0}^{+\infty} \frac{\Gamma(1+h)}{\Gamma(m+mh)\Gamma(k+1+mh)} \frac{\left[\beta^2 \left(\frac{c^2 t^2 - x^2}{4c^2}\right)\right]^{mh}}{h!} \\ &= \frac{\beta}{4c} e^{-\beta t} \sum_{k=0}^{+\infty} S_{(m-1,k)}^{(m,m)} \left[\beta \left(\frac{ct+x}{2c}\right), \beta \left(\frac{ct-x}{2c}\right) \right]. \end{aligned}$$

Using similar reasoning, we obtain

$$\begin{aligned} f(x, t) &= \frac{\beta}{4c} e^{-\beta t} \left\{ \sum_{k=0}^{+\infty} S_{(m-1,k)}^{(m,m)} \left[\beta \left(\frac{ct+x}{2c}\right), \beta \left(\frac{ct-x}{2c}\right) \right] \right. \\ &\quad - \sum_{k=0}^{+\infty} S_{(m-1,2m+k)}^{(m,m)} \left[\beta \left(\frac{ct-x}{2c}\right), \beta \left(\frac{ct+x}{2c}\right) \right] \\ &\quad + \sum_{k=0}^{+\infty} S_{(m-1,k)}^{(m,m)} \left[\beta \left(\frac{ct-x}{2c}\right), \beta \left(\frac{ct+x}{2c}\right) \right] \\ &\quad \left. - \sum_{k=0}^{+\infty} S_{(m-1,2m+k)}^{(m,m)} \left[\beta \left(\frac{ct+x}{2c}\right), \beta \left(\frac{ct-x}{2c}\right) \right] \right\}. \end{aligned}$$

Hence, straightforward calculations and the identities (10) and (12) give

$$\begin{aligned} f(x, t) &= \frac{\beta}{4c} e^{-\beta t} \left\{ e^{\beta \left(\frac{ct-x}{2c}\right)} \sum_{h=0}^{+\infty} \frac{\left[\beta \left(\frac{ct+x}{2c}\right)\right]^{mh+m-1}}{(mh+m-1)!} \left[\frac{\gamma(mh, \beta \left(\frac{ct-x}{2c}\right))}{\Gamma(mh)} \right. \right. \\ &\quad \left. \left. - \frac{\gamma(m(h+2), \beta \left(\frac{ct-x}{2c}\right))}{\Gamma(m(h+2))} \right] \right. \\ &\quad \left. + e^{\beta \left(\frac{ct+x}{2c}\right)} \sum_{h=0}^{+\infty} \frac{\left[\beta \left(\frac{ct-x}{2c}\right)\right]^{mh+m-1}}{(mh+m-1)!} \left[\frac{\gamma(mh, \beta \left(\frac{ct+x}{2c}\right))}{\Gamma(mh)} - \frac{\gamma(m(h+2), \beta \left(\frac{ct+x}{2c}\right))}{\Gamma(m(h+2))} \right] \right\}. \end{aligned}$$

By setting

$$e_n(x) := \sum_{j=0}^n \frac{x^j}{j!} \tag{18}$$

(cf. Equation 6.5.11 of [1]) and recalling that

$$\frac{\gamma(n, x)}{\Gamma(n)} = 1 - e_{n-1}(x) e^{-x} \tag{19}$$

(cf. Equations 6.5.2 and 6.5.13 of [1]), we have

$$\begin{aligned}
 f(x, t) &= \frac{\beta}{4c} e^{-\beta t} \left\{ \sum_{h=0}^{+\infty} \frac{\left[\beta \left(\frac{ct-x}{2c}\right)\right]^{mh+m-1}}{(mh+m-1)!} \left[e_{mh+2m-1} \left(\beta \left(\frac{ct-x}{2c}\right)\right) - e_{mh-1} \left(\beta \left(\frac{ct-x}{2c}\right)\right) \right] \right. \\
 &\quad \left. + \sum_{h=0}^{+\infty} \frac{\left[\beta \left(\frac{ct+x}{2c}\right)\right]^{mh+m-1}}{(mh+m-1)!} \left[e_{mh+2m-1} \left(\beta \left(\frac{ct+x}{2c}\right)\right) - e_{mh-1} \left(\beta \left(\frac{ct+x}{2c}\right)\right) \right] \right\} = \frac{\beta}{4c} e^{-\beta t} \\
 &\quad \times \left\{ \sum_{h=0}^{+\infty} \sum_{j=0}^{2m-1} \frac{\left[\beta \left(\frac{ct+x}{2c}\right)\right]^{mh+m-1}}{(mh+m-1)!} \frac{\left[\beta \left(\frac{ct-x}{2c}\right)\right]^{j+mh}}{(j+mh)!} \right. \\
 &\quad \left. + \sum_{h=0}^{+\infty} \sum_{j=0}^{2m-1} \frac{\left[\beta \left(\frac{ct-x}{2c}\right)\right]^{mh+m-1}}{(mh+m-1)!} \frac{\left[\beta \left(\frac{ct+x}{2c}\right)\right]^{j+mh}}{(j+mh)!} \right\} \\
 &= \frac{\beta}{4c} e^{-\beta t} \left\{ \sum_{j=0}^{2m-1} S_{(m-1,j)}^{(m,m)} \left[\beta \left(\frac{ct+x}{2c}\right), \beta \left(\frac{ct-x}{2c}\right) \right] \right. \\
 &\quad \left. + \sum_{j=0}^{2m-1} S_{(m-1,j)}^{(m,m)} \left[\beta \left(\frac{ct-x}{2c}\right), \beta \left(\frac{ct+x}{2c}\right) \right] \right\} \\
 &= \frac{\beta}{4c} e^{-\beta t} \left\{ \sum_{j=0}^{2m-1} S_{(m-1,j)}^{(m,m)} \left[\beta \left(\frac{ct+x}{2c}\right), \beta \left(\frac{ct-x}{2c}\right) \right] \right. \\
 &\quad \left. + \sum_{j=0}^{2m-1} S_{(j,m-1)}^{(m,m)} \left[\beta \left(\frac{ct+x}{2c}\right), \beta \left(\frac{ct-x}{2c}\right) \right] \right\},
 \end{aligned}$$

where the last equality comes from the symmetry property $S_{(i,j)}^{(k,r)}(x, y) = S_{(j,i)}^{(r,k)}(y, x)$ of the pseudo-Bessel function. □

Some properties of the two-index pseudo-Bessel function (17) can be found in Remark 3.2 of [12].

In the following proposition we provide a closed-form expression for the probability density function (8) in the case $\alpha = 2$.

Proposition 3.3. *In the case $\alpha = 2$ the probability density function (8), for $-ct < x < ct$, has the following expression:*

$$\begin{aligned}
 f(x, t) &= \frac{\lambda}{8c} e^{-\lambda t} \sum_{j=0}^3 \left[\left(\frac{ct-x}{ct+x}\right)^{\frac{j-1}{2}} + \left(\frac{ct-x}{ct+x}\right)^{-\frac{j-1}{2}} \right] \left[I_{j-1} \left(\frac{\lambda}{c} \sqrt{c^2 t^2 - x^2}\right) \right. \\
 &\quad \left. - J_{j-1} \left(\frac{\lambda}{c} \sqrt{c^2 t^2 - x^2}\right) \right],
 \end{aligned}$$

where, for $\nu \in \mathbb{R}$,

$$J_\nu(z) = \sum_{l=0}^{+\infty} \frac{(-1)^l}{l! \Gamma(l + \nu + 1)} \left(\frac{z}{2}\right)^{2l + \nu}, \tag{20}$$

is the Bessel function of the first kind, and

$$I_\nu(z) = \sum_{k=0}^{+\infty} \frac{1}{k! \Gamma(\nu + k + 1)} \left(\frac{z}{2}\right)^{2k+\nu} \tag{21}$$

is the modified Bessel function of the first kind.

Proof. Note that, by (20) and (21),

$$\begin{aligned} I_1(2\lambda\sqrt{xy}) + J_1(2\lambda\sqrt{xy}) &= \sum_{m=0}^{+\infty} \frac{(\lambda\sqrt{xy})^{2m+1}}{m! \Gamma(m+2)} + \sum_{m=0}^{+\infty} \frac{(-1)^m (\lambda\sqrt{xy})^{2m+1}}{m! \Gamma(m+2)} \\ &= 2\lambda\sqrt{xy} \sum_{n=0}^{+\infty} \frac{(\lambda^2 xy)^{2n}}{(2n)! (2n+1)!}. \end{aligned}$$

Hence, setting $\tilde{x} := \frac{ct+x}{2c}$ and $y := \frac{ct-x}{2c}$, we can write the two-index pseudo-Bessel function (17) as

$$\begin{aligned} S_{(1,0)}^{(2,2)}[\lambda\tilde{x}, \lambda y] &= \sum_{l=0}^{+\infty} \frac{(\lambda\tilde{x})^{2l+1} (\lambda y)^{2l}}{(2l+1)! (2l)!} \\ &= (\lambda\tilde{x}) \sum_{l=0}^{+\infty} \frac{(\lambda^2 \tilde{x}y)^{2l}}{(2l+1)! (2l)!} = \frac{1}{2} \left(\frac{y}{\tilde{x}}\right)^{-\frac{1}{2}} \left[I_1(2\lambda\sqrt{\tilde{x}y}) + J_1(2\lambda\sqrt{\tilde{x}y}) \right] \\ &= \frac{1}{2} \left(\frac{y}{\tilde{x}}\right)^{-\frac{1}{2}} \left[I_{-1}(2\lambda\sqrt{\tilde{x}y}) - J_{-1}(2\lambda\sqrt{\tilde{x}y}) \right], \end{aligned}$$

where the last equality follows from recalling that $I_{-n}(x) = I_n(x)$ and $J_{-n}(x) = (-1)^n J_n(x)$.

Moreover, for $j \geq 1$,

$$\begin{aligned} I_{j-1}(2\lambda\sqrt{\tilde{x}y}) - J_{j-1}(2\lambda\sqrt{\tilde{x}y}) &= (\lambda\sqrt{\tilde{x}y})^{j-1} \left(\sum_{m=0}^{+\infty} \frac{(\lambda^2 \tilde{x}y)^m}{m! \Gamma(m+j)} - \sum_{m=0}^{+\infty} \frac{(-\lambda^2 \tilde{x}y)^m}{m! \Gamma(m+j)} \right) \\ &= 2(\lambda\sqrt{\tilde{x}y})^{j-1} \sum_{n=0}^{+\infty} \frac{(\lambda^2 \tilde{x}y)^{2n+1}}{(2n+1)! \Gamma(2n+1+j)} \\ &= 2(\lambda^2 \tilde{x}y)^{\frac{j+1}{2}} \sum_{n=0}^{+\infty} \frac{(\lambda^2 \tilde{x}y)^{2n}}{(2n+1)! (2n+j)!}. \end{aligned}$$

Hence,

$$\begin{aligned} S_{(1,j)}^{(2,2)}[\lambda\tilde{x}, \lambda y] &= \lambda^{j+1} \tilde{x}y^j \sum_{l=0}^{+\infty} \frac{(\lambda^2 \tilde{x}y)^{2l}}{(2l+1)! (2l+j)!} \\ &= \frac{\lambda^{j+1} \tilde{x}y^j (\lambda^2 \tilde{x}y)^{-\frac{j+1}{2}}}{2} \left[I_{j-1}(2\lambda\sqrt{\tilde{x}y}) - J_{j-1}(2\lambda\sqrt{\tilde{x}y}) \right] \\ &= \frac{1}{2} \left(\frac{y}{\tilde{x}}\right)^{\frac{j-1}{2}} \left[I_{j-1}(2\lambda\sqrt{\tilde{x}y}) - J_{j-1}(2\lambda\sqrt{\tilde{x}y}) \right]. \end{aligned}$$

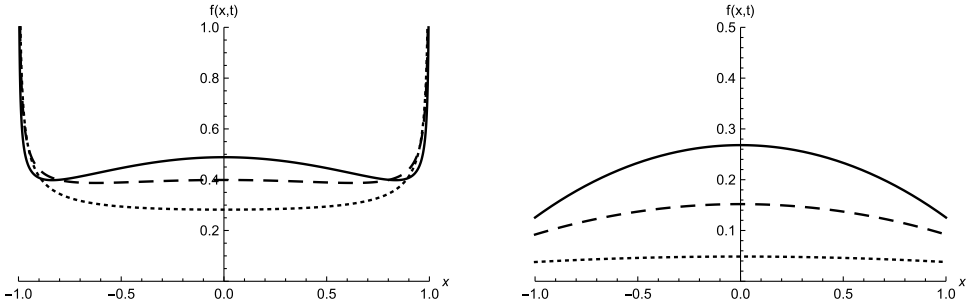


FIGURE 3. Probability density function of X_t under the assumption of gamma-distributed intertimes for $t = 1, c = 1, \beta = 0.5$ (dotted line), $\beta = 1$ (dashed line), $\beta = 1.5$ (solid line), with $\alpha = 1/2$ (left-hand side) and $\alpha = 2$ (right-hand side).

Similarly, we have

$$S_{(0,1)}^{(2,2)}[\lambda\tilde{x}, \lambda y] = \frac{1}{2} \left(\frac{y}{\tilde{x}}\right)^{\frac{1}{2}} \left[I_{-1}(2\lambda\sqrt{\tilde{x}y}) - J_{-1}(2\lambda\sqrt{\tilde{x}y}) \right],$$

and for $j \geq 1$,

$$S_{(j,1)}^{(2,2)}[\lambda\tilde{x}, \lambda y] = \frac{1}{2} \left(\frac{y}{\tilde{x}}\right)^{-\frac{j-1}{2}} \left[I_{j-1}(2\lambda\sqrt{\tilde{x}y}) - J_{j-1}(2\lambda\sqrt{\tilde{x}y}) \right].$$

The proof finally follows from Corollary (3.1) and the definition of \tilde{x} and y . □

Some plots of the probability density function (8) are shown in Figure 3 for different values of the parameters involved.

The following proposition deals with the case in which the random times U_i are exponentially distributed, whereas the random variables D_i are gamma-distributed, $i = 1, 2, \dots$

Corollary 3.2. *For $i = 1, 2, \dots$, let us assume that the random times U_i are exponentially distributed with parameter $\lambda > 0$ and that the random variables D_i have gamma distribution with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$. For $t > 0$, it holds that*

$$P(X_t = -ct | X_0 = 0) = \frac{1}{2} \left[1 - \frac{\gamma(\alpha, \beta t)}{\Gamma(\alpha)} \right], \quad P(X_t = ct | X_0 = 0) = \frac{1}{2} e^{-\lambda t},$$

and, for $-ct < x < ct$,

$$\begin{aligned} f(x, t) = & \frac{\lambda}{4c} e^{-\frac{1}{2}(\beta+\lambda)\frac{x}{2c}(\beta-\lambda)} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^k \left\{ W_{\alpha, k+1} \left[\lambda \beta^\alpha \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^\alpha \right] \right. \\ & \left. - \left[\beta \left(\frac{ct-x}{2c} \right) \right]^{2\alpha} W_{\alpha, 2\alpha+k+1} \left[\lambda \beta^\alpha \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^\alpha \right] \right\} \\ & + \frac{\beta}{4c} e^{-\frac{1}{2}(\beta+\lambda)\frac{x}{2c}(\beta-\lambda)} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^{-1} \left\{ W_{\alpha, 0} \left[\lambda \beta^\alpha \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^\alpha \right] \right. \\ & \left. + \left[\beta \left(\frac{ct-x}{2c} \right) \right]^{\alpha-1} W_{\alpha, \alpha} \left[\lambda \beta^\alpha \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^\alpha \right] \right\}, \end{aligned} \tag{22}$$

where

$$W_{\rho,\theta}(z) = \sum_{k=0}^{+\infty} \frac{z^k}{k! \Gamma(\rho k + \theta)}, \quad \rho > -1, \theta \in \mathbb{C}, \tag{23}$$

is the Wright function.

Proof. Under the assumptions concerning the distribution of U_i and D_i , $i = 1, 2, \dots$, for $x > 0$ it holds that

$$f_U^{(n)}(x) = \frac{\lambda^n x^{n-1} e^{-\lambda x}}{(n-1)!}, \quad F_U^{(n)}(x) - F_U^{(n+1)}(x) = p(n, \lambda x), \tag{24}$$

where $p(j, \eta)$ denotes the Poisson probability mass function with mean η evaluated at j , and

$$g_D^{(n)}(x) = \frac{\beta^{n\alpha} x^{n\alpha-1} e^{-\beta x}}{\Gamma(n\alpha)}, \quad G_D^{(n)}(x) = \frac{\gamma(n\alpha, \beta x)}{\Gamma(n\alpha)}. \tag{25}$$

Substituting Equations (24) and (25) in Equation (4), we get

$$\begin{aligned} f_c(x, t) &= \frac{1}{2c} \left\{ e^{-\lambda \left(\frac{ct+x}{2c}\right)} \sum_{n=0}^{+\infty} \left[G_D^{(n)}\left(\frac{ct-x}{2c}\right) - G_D^{(n+1)}\left(\frac{ct-x}{2c}\right) \right] \frac{\lambda^{n+1} \left(\frac{ct+x}{2c}\right)^n}{n!} \right. \\ &\quad \left. + \sum_{n=1}^{+\infty} p\left(n, \lambda \left(\frac{ct+x}{2c}\right)\right) g_D^{(n)}\left(\frac{ct-x}{2c}\right) \right\} = \frac{1}{2c} \left\{ \lambda \sum_{n=0}^{+\infty} p\left(n, \lambda \left(\frac{ct+x}{2c}\right)\right) \right. \\ &\quad \left. \times \left[G_D^{(n)}\left(\frac{ct-x}{2c}\right) - G_D^{(n+1)}\left(\frac{ct-x}{2c}\right) \right] + \sum_{n=1}^{+\infty} p\left(n, \lambda \left(\frac{ct+x}{2c}\right)\right) g_D^{(n)}\left(\frac{ct-x}{2c}\right) \right\} \\ &= \frac{\lambda}{2c} \sum_{n=0}^{+\infty} p\left(n, \lambda \left(\frac{ct+x}{2c}\right)\right) \left\{ \frac{\gamma(n\alpha, \beta \left(\frac{ct-x}{2c}\right))}{\Gamma(n\alpha)} - \frac{\gamma((n+1)\alpha, \beta \left(\frac{ct-x}{2c}\right))}{\Gamma((n+1)\alpha)} \right\} \\ &\quad + \frac{1}{2c} e^{-\beta \left(\frac{ct-x}{2c}\right)} \sum_{n=1}^{+\infty} p\left(n, \lambda \left(\frac{ct+x}{2c}\right)\right) \frac{\beta^{n\alpha} \left(\frac{ct-x}{2c}\right)^{n\alpha-1}}{\Gamma(n\alpha)}. \end{aligned}$$

Hence, recalling Equation (10), we obtain

$$\begin{aligned} f_c(x, t) &= \frac{\lambda}{2c} \sum_{n=0}^{+\infty} p\left(n, \lambda \left(\frac{ct+x}{2c}\right)\right) \left[\beta \left(\frac{ct-x}{2c}\right) \right]^{n\alpha} \left\{ \gamma^*\left(n\alpha, \beta \left(\frac{ct-x}{2c}\right)\right) \right. \\ &\quad \left. - \left[\beta \left(\frac{ct-x}{2c}\right) \right]^\alpha \gamma^*\left((n+1)\alpha, \beta \left(\frac{ct-x}{2c}\right)\right) \right\} + \frac{1}{2c} e^{-\beta \left(\frac{ct-x}{2c}\right)} \left[\frac{ct-x}{2c} \right]^{-1} \\ &\quad \times \sum_{n=1}^{+\infty} p\left(n, \lambda \left(\frac{ct+x}{2c}\right)\right) \frac{[\beta^\alpha \left(\frac{ct-x}{2c}\right)^\alpha]^n}{\Gamma(n\alpha)} = \frac{\lambda}{2c} \sum_{n=0}^{+\infty} p\left(n, \lambda \left(\frac{ct+x}{2c}\right)\right) \\ &\quad \times \left[\beta \left(\frac{ct-x}{2c}\right) \right]^{n\alpha} e^{-\beta \left(\frac{ct-x}{2c}\right)} \left\{ \sum_{k=0}^{+\infty} \frac{[\beta \left(\frac{ct-x}{2c}\right)]^k}{\Gamma(n\alpha + k + 1)} - \sum_{k=0}^{+\infty} \frac{[\beta \left(\frac{ct-x}{2c}\right)]^{k+\alpha}}{\Gamma(n\alpha + \alpha + k + 1)} \right\} \end{aligned}$$

$$\begin{aligned}
 & + \frac{1}{2c} e^{-\beta(\frac{ct-x}{2c})} e^{-\lambda(\frac{ct+x}{2c})} \left[\frac{ct-x}{2c} \right]^{-1} \sum_{n=1}^{+\infty} \frac{[\lambda(\frac{ct+x}{2c}) \beta^\alpha (\frac{ct-x}{2c})^\alpha]^n}{n! \Gamma(n\alpha)} \\
 = & \frac{\lambda}{2c} e^{-\lambda(\frac{ct+x}{2c})} e^{-\beta(\frac{ct-x}{2c})} \left\{ \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^k \sum_{n=0}^{+\infty} \frac{[\lambda(\frac{ct+x}{2c}) \beta^\alpha (\frac{ct-x}{2c})^\alpha]^n}{n! \Gamma(n\alpha + k + 1)} \right. \\
 & \left. - \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^{k+\alpha} \sum_{n=0}^{+\infty} \frac{[\lambda(\frac{ct+x}{2c}) \beta^\alpha (\frac{ct-x}{2c})^\alpha]^n}{n! \Gamma((n+1)\alpha + k + 1)} \right\} \\
 & + \frac{1}{2c} \left[\frac{ct-x}{2c} \right]^{-1} e^{-\beta(\frac{ct-x}{2c})} e^{-\lambda(\frac{ct+x}{2c})} \sum_{n=1}^{+\infty} \frac{[\lambda(\frac{ct+x}{2c}) \beta^\alpha (\frac{ct-x}{2c})^\alpha]^n}{n! \Gamma(n\alpha)}.
 \end{aligned}$$

From the definition (23), it finally results that

$$\begin{aligned}
 f_c(x, t) = & \frac{\lambda}{2c} e^{-\frac{t}{2}(\beta+\lambda) + \frac{x}{2c}(\beta-\lambda)} \sum_{k=0}^{+\infty} \left\{ \left[\beta \left(\frac{ct-x}{2c} \right) \right]^k W_{\alpha, k+1} \left[\lambda \beta^\alpha \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^\alpha \right] \right. \\
 & \left. - \left[\beta \left(\frac{ct-x}{2c} \right) \right]^{k+\alpha} W_{\alpha, k+\alpha+1} \left[\lambda \beta^\alpha \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^\alpha \right] \right\} \\
 & + \frac{1}{2c} \left[\frac{ct-x}{2c} \right]^{-1} e^{-\frac{t}{2}(\beta+\lambda) + \frac{x}{2c}(\beta-\lambda)} W_{\alpha, 0} \left[\lambda \beta^\alpha \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^\alpha \right].
 \end{aligned}$$

Substituting Equations (24) and (25) in Equation (5), and proceeding in a similar way, we easily obtain the expression for $f_{-c}(x, t)$, so that the thesis immediately follows from recalling the assumption of random initial velocity. □

As an immediate consequence of Proposition 3.2, we obtain the following result.

Proposition 3.4. *If the random times U_i are exponentially distributed with parameter $\lambda > 0$ and the D_i are Erlang-distributed with parameters $m \in \mathbb{N}$ and $\beta > 0$, it holds that*

$$P(X_t = -ct \mid X_0 = 0) = \frac{1}{2} \left[1 - \frac{\gamma(m, \beta t)}{(m-1)!} \right], \quad P(X_t = ct \mid X_0 = 0) = \frac{1}{2} e^{-\lambda t},$$

and

$$\begin{aligned}
 f(x, t) = & \frac{\lambda}{4c} e^{-\frac{t}{2}(\beta+\lambda) + \frac{x}{2c}(\beta-\lambda)} \sum_{k=0}^{2m-1} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^k W_{m, k+1} \left[\lambda \beta^m \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^m \right] \\
 & + \frac{\beta}{4c} e^{-\frac{t}{2}(\beta+\lambda) + \frac{x}{2c}(\beta-\lambda)} \left[\beta \left(\frac{ct-x}{2c} \right) \right]^{-1} \left\{ W_{m, 0} \left[\lambda \beta^m \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^m \right] \right. \\
 & \left. + \left[\beta \left(\frac{ct-x}{2c} \right) \right]^{m-1} W_{m, m} \left[\lambda \beta^m \left(\frac{ct+x}{2c} \right) \left(\frac{ct-x}{2c} \right)^m \right] \right\}.
 \end{aligned}$$

Figure 4 shows some plots of the probability density function given in Proposition 3.4 for different values of the parameters involved.

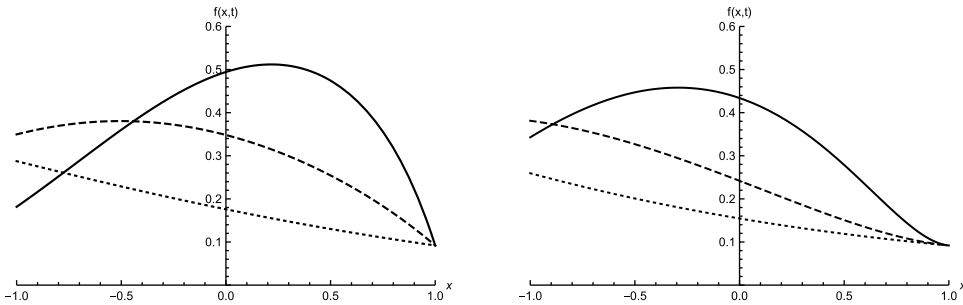


FIGURE 4. Density $f(x, t)$ given in Proposition 3.4 for $t = 1, c = 1, \lambda = 1, \beta = 0.5$ (dotted line), $\beta = 2$ (dashed line), $\beta = 4$ (solid line), with $m = 2$ (left-hand side) and $m = 3$ (right-hand side).

4. The moment generating function

Let us consider the moment generating function of the generalized telegraph process,

$$M_X^s(t) := \mathbb{E}\left[e^{sX_t} | X_0 = 0\right], \quad s \in \mathbb{R}, \quad t > 0. \tag{26}$$

Denoting by

$$\mathcal{L}_p[w] = \int_0^{+\infty} e^{-pt} w(t) dt, \quad p \geq 0, \tag{27}$$

the Laplace transform of an arbitrary function $w(t)$, in this section we provide some results on the Laplace transform $\mathcal{L}_p[M_X^s]$ of the moment generating function. Note that $\mathcal{L}_p[M_X^s] = \frac{1}{p} \mathbb{E}\left[e^{sXe_p}\right]$, where e_p is an exponential random variable with parameter p , independent of X_t .

The following theorem, providing an explicit expression for $\mathcal{L}_p[M_X^s]$, represents a useful tool for further analysis. For instance, it allows us to prove the existence of a Kac-type condition under which the probability density function of the generalized integrated telegraph process, with identically distributed gamma intertimes, converges to that of the standard Brownian motion. We point out the novelty of such a result in the field of finite-velocity random evolutions, since it generalizes the one holding in the case of the standard telegraph process.

Theorem 4.1. *If the random variables $U_i (D_i), i \geq 1$, are independent copies of an absolutely continuous random variable $U (D)$, for $t > 0$ and under the assumption $\mathcal{L}_{p-sc}[f_U] \cdot \mathcal{L}_{p+sc}[g_D] < 1$, the Laplace transform of the moment generating function of the generalized telegraph process is given by*

$$\begin{aligned} \mathcal{L}_p [M_X^s] &= \frac{1}{2} \mathcal{L}_{p-sc} [\overline{F}_U] + \frac{1}{2} \mathcal{L}_{p+sc} [\overline{G}_D] \\ &+ \frac{1}{2(p+sc)} \mathcal{L}_{p-sc}[f_U] + \frac{1}{2(p-sc)} \mathcal{L}_{p+sc}[g_D] + \frac{sc}{(p+sc)(p-sc)} \\ &\times \left\{ \frac{1}{1 - \mathcal{L}_{p-sc}[f_U] \cdot \mathcal{L}_{p+sc}[g_D]} - 1 \right\} \{ \mathcal{L}_{p+sc}[g_D] - \mathcal{L}_{p-sc}[f_U] \}. \end{aligned} \tag{28}$$

Proof. Let us denote by $M_c^s(t)$ and $M_{-c}^s(t)$ the conditional moment generating functions of the integrated telegraph process under fixed initial velocity $V_0 = \pm c$, i.e.

$$M_{v_0}^s(t) = \mathbb{E}\left[e^{sX_t} | V_0 = v_0\right].$$

In the case $v_0 = c$, by Equations (2) and (4), we get

$$\begin{aligned}
 M_c^s(t) &:= \mathbb{E} \left[e^{sX_t} | V(0) = c \right] = e^{sct} \bar{F}_U(t) + \int_{-ct}^{ct} e^{sx} f_c(x, t) dx = e^{sct} \bar{F}_U(t) \\
 &+ \frac{1}{2c} \left\{ \sum_{n=0}^{+\infty} \int_{-ct}^{ct} e^{sx} \times G_D^{(n)} \left(t - \frac{ct+x}{2c} \right) f_U^{(n+1)} \left(\frac{ct+x}{2c} \right) dx \right. \\
 &- \sum_{n=0}^{+\infty} \int_{-ct}^{ct} e^{sx} G_D^{(n+1)} \left(t - \frac{ct+x}{2c} \right) f_U^{(n+1)} \left(\frac{ct+x}{2c} \right) dx \\
 &+ \sum_{n=1}^{+\infty} \int_{-ct}^{ct} e^{sx} F_U^{(n)} \left(\frac{ct+x}{2c} \right) g_D^{(n)} \left(t - \frac{ct+x}{2c} \right) dx \\
 &\left. - \sum_{n=1}^{+\infty} \int_{-ct}^{ct} e^{sx} F_U^{(n+1)} \left(\frac{ct+x}{2c} \right) g_D^{(n)} \left(t - \frac{ct+x}{2c} \right) dx \right\} \\
 &= e^{sct} \bar{F}_U(t) + e^{-sct} \left\{ \sum_{n=1}^{+\infty} \int_0^t e^{2scy} f_U^{(n+1)}(y) \left[G_D^{(n)}(t-y) - G_D^{(n+1)}(t-y) \right] dy \right. \\
 &\left. + \int_0^t e^{2scy} f_U(y) [1 - G_D(t-y)] dy + \sum_{n=1}^{+\infty} \int_0^t e^{2scy} g_D^{(n)}(t-y) \left[F_U^{(n)}(y) - F_U^{(n+1)}(y) \right] dy \right\}.
 \end{aligned}$$

We now consider the Laplace transform of the moment generating function

$$\begin{aligned}
 \mathcal{L}_p [M_c^s] &= \mathcal{L}_p [e^{sct} \bar{F}_U(t)] + \sum_{n=1}^{+\infty} \mathcal{L}_p \left[e^{-sct} \int_0^t e^{2scy} f_U^{(n+1)}(y) \left[G_D^{(n)}(t-y) - G_D^{(n+1)}(t-y) \right] dy \right] \\
 &+ \mathcal{L}_p \left[e^{-sct} \int_0^t e^{2scy} f_U(y) [1 - G_D(t-y)] dy \right] \\
 &+ \sum_{n=1}^{+\infty} \mathcal{L}_p \left[e^{-sct} \int_0^t e^{2scy} g_D^{(n)}(t-y) \left[F_U^{(n)}(y) - F_U^{(n+1)}(y) \right] dy \right].
 \end{aligned}$$

Denoting by $w \star h(t)$ the convolution between two functions w and h , and exploiting the properties of the Laplace transform, we obtain

$$\begin{aligned}
 \mathcal{L}_p [M_c^s] &= \mathcal{L}_{p-sc} [\bar{F}_U] + \sum_{n=1}^{+\infty} \mathcal{L}_{p+sc} \left[\left(G_D^{(n)} - G_D^{(n+1)} \right) \star \left(e^{2sct} f_U^{(n+1)}(t) \right) \right] \\
 &+ \mathcal{L}_{p+sc} \left[\int_0^t e^{2scy} f_U(y) dy \right] - \mathcal{L}_{p+sc} \left[G_D \star e^{2sct} f_U(t) \right] \\
 &+ \sum_{n=1}^{+\infty} \mathcal{L}_{p+sc} \left[g_D^{(n)} \star e^{2sct} \left(F_U^{(n)}(t) - F_U^{(n+1)}(t) \right) \right]
 \end{aligned}$$

$$\begin{aligned}
 &= \mathcal{L}_{p-sc} [\bar{F}_U] + \sum_{n=1}^{+\infty} \mathcal{L}_{p-sc} [f_U^{(n+1)}] \cdot \left[\mathcal{L}_{p+sc} (G_D^{(n)}) - \mathcal{L}_{p+sc} (G_D^{(n+1)}) \right] \\
 &\quad + \frac{1}{p+sc} \mathcal{L}_{p+sc} [e^{2sct} f_U(t)] - \mathcal{L}_{p+sc} [e^{2sct} f_U(t)] \cdot \mathcal{L}_{p+sc} [g_D] \\
 &\quad + \sum_{n=1}^{+\infty} \mathcal{L}_{p+sc} [g_D^{(n)}] \cdot \left[\mathcal{L}_{p-sc} (F_U^{(n)}) - \mathcal{L}_{p-sc} (F_U^{(n+1)}) \right].
 \end{aligned}$$

Hence, rearranging the terms, we have

$$\begin{aligned}
 \mathcal{L}_p [M_c^s] &= \mathcal{L}_{p-sc} [\bar{F}_U] + \sum_{n=1}^{+\infty} \mathcal{L}_{p-sc} [f_U^{(n+1)}] \cdot \left[\frac{\mathcal{L}_{p+sc} (g_D^{(n)})}{p+sc} - \frac{\mathcal{L}_{p+sc} (g_D^{(n+1)})}{p+sc} \right] \\
 &\quad + \frac{\mathcal{L}_{p-sc}[f_U]}{p+sc} - \frac{\mathcal{L}_{p-sc}[f_U]\mathcal{L}_{p+sc}[g_D]}{p+sc} + \sum_{n=1}^{+\infty} \mathcal{L}_{p+sc} [g_D^{(n)}] \\
 &\quad \times \left[\frac{\mathcal{L}_{p-sc} (f_U^{(n)})}{p-sc} - \frac{\mathcal{L}_{p-sc} (f_U^{(n+1)})}{p-sc} \right] = \mathcal{L}_{p-sc} [\bar{F}_U] + \frac{1}{p+sc} \mathcal{L}_{p-sc}[f_U] \\
 &\quad - \frac{2sc}{(p+sc)(p-sc)} \mathcal{L}_{p-sc}[f_U] \sum_{n=1}^{+\infty} [\mathcal{L}_{p-sc}[f_U]]^n [\mathcal{L}_{p+sc}[g_D]]^n \\
 &\quad + \frac{2sc}{(p+sc)(p-sc)} \sum_{n=1}^{+\infty} [\mathcal{L}_{p-sc}[f_U]]^n [\mathcal{L}_{p+sc}[g_D]]^n.
 \end{aligned}$$

Under the assumption $\mathcal{L}_{p-sc}[f_U] \cdot \mathcal{L}_{p+sc}[g_D] < 1$, we can express the Laplace transform of the conditional moment generating function in terms of the Laplace transform of the densities of the random times U_i and D_i :

$$\begin{aligned}
 \mathcal{L}_p [M_c^s] &= \mathcal{L}_{p-sc} [\bar{F}_U] + \frac{1}{p+sc} \mathcal{L}_{p-sc}[f_U] \\
 &\quad + \frac{2sc}{(p+sc)(p-sc)} [1 - \mathcal{L}_{p-sc}[f_U]] \cdot \left[\frac{1}{1 - \mathcal{L}_{p-sc}[f_U] \cdot \mathcal{L}_{p+sc}[g_D]} - 1 \right]. \tag{29}
 \end{aligned}$$

In the case $v_0 = -c$, through similar reasoning we obtain

$$\begin{aligned}
 \mathcal{L}_p [M_{-c}^s] &= \mathcal{L}_{p+sc} [\bar{G}_D] + \frac{1}{p-sc} \mathcal{L}_{p+sc}[g_D] \\
 &\quad + \frac{2sc}{(p+sc)(p-sc)} [\mathcal{L}_{p+sc}[g_D] - 1] \cdot \left[\frac{1}{1 - \mathcal{L}_{p-sc}[f_U] \cdot \mathcal{L}_{p+sc}[g_D]} - 1 \right], \tag{30}
 \end{aligned}$$

so that the claimed result immediately follows under the assumption of random initial velocity. □

Starting from the previous theorem, the following proposition provides the Laplace transform of the moment generating function of the integrated telegraph process, under the assumption of identically distributed gamma intertimes U_i and D_i .

Proposition 4.1. *Let us assume that both the random variables U_i and D_i ($i \geq 1$) have gamma distribution with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$. Then, recalling Equations (26) and (27), for $p > -\beta + \sqrt{\beta^2 + c^2 s^2}$, we have*

$$\begin{aligned} \mathcal{L}_p [M_X^s] &= \frac{1}{2(p+sc)} \left(\frac{\beta}{\beta+p-sc} \right)^\alpha + \frac{1}{2(p-sc)} \left(\frac{\beta}{\beta+p+sc} \right)^\alpha \\ &+ \frac{sc}{(p+sc)(p-sc)} \left\{ \left(\frac{\beta}{\beta+p+sc} \right)^\alpha - \left(\frac{\beta}{\beta+p-sc} \right)^\alpha \right\} \left(\frac{\beta^2}{(\beta+p-sc)(\beta+p+sc)} \right)^\alpha \\ &\times \left[1 - \left(\frac{\beta^2}{(\beta+p-sc)(\beta+p+sc)} \right)^\alpha \right]^{-1}. \end{aligned}$$

Proof. The proof can be immediately obtained from Equation (28) and by recalling the following expression for the Laplace transform of the gamma density function with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$:

$$\mathcal{L}_p \{f_U\} = \mathcal{L}_p \{g_D\} = \left(\frac{\beta}{p+\beta} \right)^\alpha, \quad p > 0. \quad \square$$

It is well known (see, for instance, [33, Section 2.6]) that under Kac scaling conditions the symmetric telegraph process weakly converges to the Brownian motion. The following theorem allows us to obtain a similar result for the symmetric telegraph process driven by gamma components.

Theorem 4.2. *Let us consider the integrated telegraph process X_t under the assumption of gamma-distributed intertimes U_i and D_i ($i \geq 1$) with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$. Under a Kac-type scaling condition, the probability density function of X_t converges to that of the standard Brownian motion, i.e., recalling Equation (1),*

$$\lim_{\substack{c, \beta \rightarrow +\infty \\ c^2/\beta \rightarrow 1}} f(x, t) = \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t}, \quad t > 0.$$

Proof. By Proposition (4.1), and recalling that

$$\lim_{\substack{c, \beta \rightarrow +\infty \\ c^2/\beta \rightarrow 1}} \left(-\beta + \sqrt{\beta^2 + c^2 s^2} \right) = \frac{s^2}{2},$$

for $p > s^2/2$, we have

$$\begin{aligned} \lim_{\substack{c, \beta \rightarrow +\infty \\ c^2/\beta \rightarrow 1}} \mathcal{L}_p [M_X^s] &= \lim_{c \rightarrow +\infty} \frac{cs \left(-\left(\frac{c^2}{c^2+p-sc} \right)^\alpha + \left(\frac{c^2}{c^2+p+cs} \right)^\alpha \right)}{(p-sc)(p+cs)} \\ &\times \left(-1 + \frac{1}{1 - \left(\frac{c^4}{(c^2+p-sc)(c^2+p+cs)} \right)^\alpha} \right) \end{aligned}$$

$$\begin{aligned}
 &= \lim_{c \rightarrow +\infty} \frac{cs \left(- \left(1 + \frac{p}{c^2} + \frac{s}{c} \right)^\alpha + \left(1 + \frac{p}{c^2} - \frac{s}{c} \right)^\alpha \right)}{\left(1 + \frac{p}{c^2} + \frac{s}{c} \right)^\alpha \left(1 + \frac{p}{c^2} - \frac{s}{c} \right)^\alpha (p - cs) (p + cs) \left[\left(1 + \frac{p-sc}{c^2} \right)^\alpha \left(1 + \frac{p+sc}{c^2} \right)^\alpha - 1 \right]} \\
 &= \lim_{c \rightarrow +\infty} \frac{cs \left(- \left(1 + \frac{p}{c^2} + \frac{s}{c} \right)^\alpha + \left(1 + \frac{p}{c^2} - \frac{s}{c} \right)^\alpha \right)}{\alpha \left(1 + \frac{p}{c^2} - \frac{s}{c} \right)^\alpha \left(1 + \frac{p}{c^2} + \frac{s}{c} \right)^\alpha \left(\frac{p}{c} - s \right) \left(\frac{p}{c} + s \right) \left(2p + \frac{p^2}{c^2} - s^2 \right)}.
 \end{aligned}$$

Hence, by making use of the binomial series, we obtain

$$\lim_{\substack{c, \beta \rightarrow +\infty, \\ c^2/\beta \rightarrow 1}} \mathcal{L}_p [M_X^s] = \frac{2}{2p - s^2}, \quad p > \frac{s^2}{2}. \tag{31}$$

The proof finally follows from recalling that the inverse Laplace transform of $\frac{2}{2p - s^2}$, $p > s^2/2$, is given by $e^{s^2 t/2}$. □

The following proposition provides the explicit expression for the moment generating function of the integrated telegraph process under the same assumptions as in Proposition (4.1).

Proposition 4.2. *If the random variables U and D both have gamma distribution with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$, for $t > 0$ and $s \in \mathbb{R}$ we have*

$$\begin{aligned}
 M_X^s(t) &= \frac{1}{2} \left\{ e^{-sct} \left(1 - \frac{\gamma(\alpha, \beta t)}{\Gamma(\alpha)} \right) + e^{sct} \left(1 - \frac{\gamma(\alpha, \beta t)}{\Gamma(\alpha)} \right) \right. \\
 &\quad + \frac{(\beta t)^\alpha}{\Gamma(\alpha + 1)} \left[e^{(sc-\beta)t} {}_1F_1(1; \alpha + 1; (\beta - 2sc)t) + e^{-(sc+\beta)t} {}_1F_1(1; \alpha + 1; (\beta + 2sc)t) \right] \\
 &\quad + (\beta t)^\alpha e^{-(sc+\beta)t} \sum_{n=1}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 1)} \left[\Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) \right. \\
 &\quad - \Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1; 2sct, \beta t) + \Phi_2(n\alpha + \alpha, 1; 2n\alpha + \alpha + 1; 2sct, \beta t) \\
 &\quad \left. - \Phi_2(n\alpha + \alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) \right] \left. \right\}, \tag{32}
 \end{aligned}$$

where

$${}_1F_1(a, b, z) = \sum_{n=0}^{+\infty} \frac{(a)_n}{(b)_n} \frac{z^n}{n!} \tag{33}$$

is the confluent hypergeometric function and

$$\Phi_2(b_1, b_2, c, w, z) = \sum_{m,n=0}^{+\infty} \frac{(b_1)_m (b_2)_n}{(c)_{m+n}} \frac{w^m}{m!} \frac{z^n}{n!}, \quad w, z \in \mathbb{C}, \tag{34}$$

denotes the Humbert series (cf. [26]).

Proof. The proof follows from Proposition 4.1. Let us set

$$A_1 := \frac{1}{2(p+sc)} \mathcal{L}_{p-sc}[fU], \tag{35a}$$

$$A_2 := \frac{1}{2(p-sc)} \mathcal{L}_{p+sc}[gD], \tag{35b}$$

$$A_3 := \frac{sc}{(p+sc)(p-sc)}, \tag{35c}$$

$$A_4 := \frac{1}{1 - \mathcal{L}_{p-sc}[fU]\mathcal{L}_{p+sc}[gD]}, \tag{35d}$$

$$A_5 := \mathcal{L}_{p+sc}[gD] - \mathcal{L}_{p-sc}[fU], \tag{35e}$$

so that Equation (28) can be expressed as

$$\mathcal{L}_p [M_X^s] = \frac{1}{2} \mathcal{L}_{p-sc} [\overline{F}U] + \frac{1}{2} \mathcal{L}_{p+sc} [\overline{G}D] + A_1 + A_2 + A_3 \cdot (A_4 - 1) \cdot A_5. \tag{36}$$

From the properties of the Laplace transform, Equation (35a) reads

$$\begin{aligned} A_1 &= \frac{1}{2(p+sc)} \left(\frac{\beta}{p-sc+\beta} \right)^\alpha \\ &= \frac{1}{2} \mathcal{L}_p \{ e^{-sc t} \} \mathcal{L}_p \left\{ \frac{e^{(sc-\beta)t} t^{\alpha-1} \beta^\alpha}{\Gamma(\alpha)} \right\} \\ &= \frac{1}{2} \mathcal{L}_p \left[\frac{\beta^\alpha}{\Gamma(\alpha)} e^{(sc-\beta)t} \int_0^t e^{(\beta-2sc)\tau} (t-\tau)^{\alpha-1} d\tau \right] \\ &= \frac{1}{2} \mathcal{L}_p \left[\frac{(\beta t)^\alpha}{\Gamma(\alpha)} e^{(sc-\beta)t} B(\alpha, 1) {}_1F_1(1; \alpha + 1; (\beta - 2sc)t) \right], \end{aligned}$$

where the last equality follows from the integral representation of the Kummer hypergeometric function (cf. 13.2.1 of [1]). Analogously, Equation (35b) can be written as

$$A_2 = \frac{1}{2} \mathcal{L}_p \left[\frac{(\beta t)^\alpha}{\Gamma(\alpha)} e^{-(sc+\beta)t} B(\alpha, 1) {}_1F_1(1; \alpha + 1; (\beta + 2sc)t) \right].$$

Moreover, we have

$$A_3 = sc \mathcal{L}_p \left[e^{sc t} \int_0^t e^{-2sc\tau} d\tau \right] = \frac{1}{2} \mathcal{L}_p [e^{sc t} - e^{-sc t}], \tag{37}$$

whereas

$$\begin{aligned} A_4 &= \sum_{n=0}^{+\infty} (\mathcal{L}_{p-sc}[fU]\mathcal{L}_{p+sc}[gD])^n \\ &= \sum_{n=0}^{+\infty} \left(\mathcal{L}_p \left[\frac{e^{-(\beta-sc)t} \beta^{n\alpha} t^{n\alpha-1}}{\Gamma(n\alpha)} \right] \right) \left(\mathcal{L}_p \left[\frac{e^{-(\beta+sc)t} \beta^{n\alpha} t^{n\alpha-1}}{\Gamma(n\alpha)} \right] \right) \\ &= \sum_{n=0}^{+\infty} \frac{\beta^{2n\alpha}}{[\Gamma(n\alpha)]^2} \mathcal{L}_p \left\{ e^{-(\beta+sc)t} \int_0^t e^{2sc\tau} \tau^{n\alpha-1} (t-\tau)^{n\alpha-1} d\tau \right\}. \end{aligned}$$

By Equation 3.383.2 of [27], the previous formula becomes

$$A_4 = \mathcal{L}_p \left\{ \sqrt{\pi} e^{-\beta t} \sum_{n=0}^{+\infty} \frac{\beta^{2n\alpha}}{\Gamma(n\alpha)} \left(\frac{t}{2sc}\right)^{n\alpha - \frac{1}{2}} I_{n\alpha - \frac{1}{2}}(sct) \right\}, \tag{38}$$

where $I_\nu(z)$ has been defined in Equation (21). It is worth noting that the term A_4 can be also expressed in terms of the moment generating function $\chi(s) := E(e^{sY})$ of a random variable Y characterized by a beta distribution with equal parameters given by $n\alpha$. Indeed,

$$A_4 = \mathcal{L}_p \left\{ e^{-(\beta+sc)t} \frac{\chi(2sct)}{t} \cdot E_{2\alpha,0} \left[(\beta t)^{2\alpha} \right] \right\},$$

where

$$E_{a,b}(z) = \sum_{k=0}^{+\infty} \frac{z^k}{\Gamma(ak + b)} \tag{39}$$

is the two-parametric Mittag-Leffler function (see, for instance, Equation 4.1.1 of [24]).

From Equations (37), (38), and (35e), we have

$$A_3 \cdot (A_4 - 1) \cdot A_5 = \frac{1}{2} \mathcal{L}_p \{ e^{sct} - e^{-sct} \} \left[\mathcal{L}_p \left\{ \sqrt{\pi} e^{-\beta t} \sum_{n=0}^{+\infty} \frac{\beta^{2n\alpha}}{\Gamma(n\alpha)} \left(\frac{t}{2sc}\right)^{n\alpha - \frac{1}{2}} I_{n\alpha - \frac{1}{2}}(sct) \right\} - 1 \right] \\ \times [\mathcal{L}_{p+sc}[gD] - \mathcal{L}_{p-sc}[fU]]. \tag{40}$$

Aiming to evaluate the previous equation, we set

$$B_1 := \mathcal{L}_p \{ e^{sct} \} \mathcal{L}_p \left\{ \sqrt{\pi} e^{-\beta t} \sum_{n=0}^{+\infty} \frac{\beta^{2n\alpha}}{\Gamma(n\alpha)} \left(\frac{t}{2sc}\right)^{n\alpha - \frac{1}{2}} I_{n\alpha - \frac{1}{2}}(sct) \right\}, \tag{41}$$

$$B_{11} := B_1 \cdot \mathcal{L}_{p+sc}[gD], \tag{42}$$

$$B_{12} := B_1 \cdot \mathcal{L}_{p-sc}[fU], \tag{43}$$

$$C_1 := \mathcal{L}_p \{ e^{-sct} \} \mathcal{L}_p \left\{ \sqrt{\pi} e^{-\beta t} \sum_{n=0}^{+\infty} \frac{\beta^{2n\alpha}}{\Gamma(n\alpha)} \left(\frac{t}{2sc}\right)^{n\alpha - \frac{1}{2}} I_{n\alpha - \frac{1}{2}}(sct) \right\}, \tag{44}$$

$$C_{11} := C_1 \cdot \mathcal{L}_{p+sc}[gD], \tag{45}$$

$$C_{12} := C_1 \cdot \mathcal{L}_{p-sc}[fU], \tag{46}$$

so that Equation (40) becomes

$$\frac{1}{2} (B_{11} - B_{12} - C_{11} + C_{12} - \mathcal{L}_p \{ e^{sct} \} \mathcal{L}_{p+sc}[gD] + \mathcal{L}_p \{ e^{sct} \} \mathcal{L}_{p-sc}[fU] \\ + \mathcal{L}_p \{ e^{-sct} \} \mathcal{L}_{p+sc}[gD] - \mathcal{L}_p \{ e^{-sct} \} \mathcal{L}_{p-sc}[fU]).$$

Note that

$$B_1 = \mathcal{L}_p \left\{ \sqrt{\pi} e^{sct} \sum_{n=0}^{+\infty} \frac{\beta^{2n\alpha}}{\Gamma(n\alpha)} \left(\frac{1}{2sc}\right)^{n\alpha - \frac{1}{2}} \int_0^t e^{-(sc+\beta)\tau} \tau^{n\alpha - \frac{1}{2}} I_{n\alpha - \frac{1}{2}}(sct) d\tau \right\}.$$

The inner integral can be evaluated by making use of the formula 7.11.1.5 in [44], so that

$$\begin{aligned} & \int_0^t e^{-(sc+\beta)\tau} \tau^{n\alpha-\frac{1}{2}} I_{n\alpha-\frac{1}{2}}(sc\tau) d\tau \\ &= \frac{1}{\Gamma\left(n\alpha + \frac{1}{2}\right)} \int_0^t e^{-(2sc+\beta)\tau} \tau^{n\alpha-\frac{1}{2}} \left(\frac{sc\tau}{2}\right)^{n\alpha-\frac{1}{2}} {}_1F_1(n\alpha; 2n\alpha; 2sc\tau) d\tau \\ &= \frac{t^{2n\alpha}}{\Gamma\left(n\alpha + \frac{1}{2}\right)} \left(\frac{sc}{2}\right)^{n\alpha-\frac{1}{2}} \int_0^1 e^{-t(2sc+\beta)y} y^{2n\alpha-1} {}_1F_1(n\alpha; 2n\alpha; 2scty) dy \\ &= \frac{t^{2n\alpha}}{\Gamma\left(n\alpha + \frac{1}{2}\right)} \left(\frac{sc}{2}\right)^{n\alpha-\frac{1}{2}} \frac{1}{2n\alpha} e^{-t(2sc+\beta)} \Phi_2(n\alpha, 1; 2n\alpha + 1; 2sct, 2sct + \beta t), \end{aligned}$$

where we have exploited the integral representation for the Humbert function Φ_2 (cf. Equation (4.6) of [4]). From this relationship, and recalling Equation 6.1.18 of [1], we finally get

$$B_1 = \mathcal{L}_p \left\{ e^{-(sc+\beta)t} \sum_{n=0}^{+\infty} \frac{(\beta^2 t^2)^{n\alpha}}{\Gamma(2n\alpha + 1)} \Phi_2(n\alpha, 1; 2n\alpha + 1; 2sct, 2sct + \beta t) \right\}. \quad (47)$$

Hence, by Equation (47), from Equation (42) it follows that

$$\begin{aligned} B_{11} &= \mathcal{L}_p \left\{ e^{-(sc+\beta)t} \sum_{n=0}^{+\infty} \frac{(\beta^2 t^2)^{n\alpha}}{\Gamma(2n\alpha + 1)} \Phi_2(n\alpha, 1; 2n\alpha \right. \\ &\quad \left. + 1; 2sct, 2sct + \beta t) \right\} \mathcal{L}_p \left\{ \frac{e^{-(\beta+sc)t} t^{\alpha-1} \beta^\alpha}{\Gamma(\alpha)} \right\} \\ &= \mathcal{L}_p \left\{ \frac{e^{-(\beta+sc)t} \beta^\alpha}{\Gamma(\alpha)} \sum_{n=0}^{+\infty} \frac{\beta^{2n\alpha}}{\Gamma(2n\alpha + 1)} \int_0^t (t-\tau)^{\alpha-1} \tau^{2n\alpha} \Phi_2(n\alpha, 1; 2n\alpha \right. \\ &\quad \left. + 1; 2sct, 2sct + \beta\tau) d\tau \right\}. \end{aligned}$$

From the relation (3.19) of [6], i.e.

$$\int_0^w x^{c-1} (w-x)^{s-c-1} \Phi_2(b, b'; c; ux; vx) dx = B(c, s-c) w^{s-1} \Phi_2(b, b'; s; uw, vw), \quad (48)$$

we have

$$B_{11} = \mathcal{L}_p \left\{ (\beta t)^\alpha e^{-(\beta+sc)t} \sum_{n=0}^{+\infty} \frac{(\beta^2 t^2)^{n\alpha}}{\Gamma(2n\alpha + 1 + \alpha)} \Phi_2(n\alpha, 1; \alpha + 2n\alpha + 1; 2sct, 2sct + \beta t) \right\}.$$

A similar argument can be applied to Equation (43), yielding

$$\begin{aligned}
 B_{12} &= \mathcal{L}_p \left\{ e^{-(sc+\beta)t} \sum_{n=0}^{+\infty} \frac{(\beta^2 t^2)^{n\alpha}}{\Gamma(2n\alpha + 1)} \Phi_2(n\alpha, 1; 2n\alpha \right. \\
 &\quad \left. + 1; 2sct, 2sct + \beta t) \right\} \mathcal{L}_p \left\{ \frac{e^{(sc-\beta)t} t^{\alpha-1} \beta^\alpha}{\Gamma(\alpha)} \right\} \\
 &= \mathcal{L}_p \left\{ \frac{e^{(sc-\beta)t} \beta^\alpha}{\Gamma(\alpha)} \sum_{n=0}^{+\infty} \frac{\beta^{2n\alpha}}{\Gamma(2n\alpha + 1)} \int_0^t e^{-2sct} (t-\tau)^{\alpha-1} \tau^{2n\alpha} \Phi_2(n\alpha, 1; 2n\alpha \right. \\
 &\quad \left. + 1; 2sct, 2sct + \beta\tau) d\tau \right\} \\
 &= \mathcal{L}_p \left\{ \frac{e^{-(sc+\beta)t} \beta^\alpha}{\Gamma(\alpha)} \sum_{n=0}^{+\infty} \sum_{m=0}^{+\infty} \frac{\beta^{2n\alpha}}{\Gamma(2n\alpha + 1)} \frac{(2sc)^m}{m!} \right. \\
 &\quad \left. \times \int_0^t (t-\tau)^{\alpha+m-1} \tau^{2n\alpha} \Phi_2(n\alpha, 1; 2n\alpha + 1; 2sct, (2sc + \beta)\tau) d\tau \right\} \\
 &= \mathcal{L}_p \left\{ e^{-(sc+\beta)t} (\beta t)^\alpha \sum_{n=0}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + 1 + \alpha)} \right. \\
 &\quad \left. \times \sum_{m=0}^{+\infty} \frac{(\alpha)_m}{(2n\alpha + 1 + \alpha)_m} \frac{(2sct)^m}{m!} \Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1 + m; 2sct, 2sct + \beta t) \right\}.
 \end{aligned}$$

Finally, the application of the decomposition formula for the Humbert function Φ_2 (cf. (2.45) of [9]),

$$\Phi_2(\beta_1, \beta_2; \gamma; x, y) = \sum_{i=0}^{+\infty} \frac{(\beta_1 - \epsilon_1)_i}{(\gamma)_i} \frac{x^i}{i!} \Phi_2(\epsilon_1, \beta_2; \gamma + i; x, y),$$

yields

$$B_{12} = \mathcal{L}_p \left\{ e^{-(sc+\beta)t} (\beta t)^\alpha \sum_{n=0}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 1)} \Phi_2(n\alpha + \alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) \right\}.$$

Hence, the initial velocity being random, Equation (36) becomes

$$\begin{aligned}
 \mathcal{L}_p [M_X^s] &= \frac{1}{2} \mathcal{L}_p \left\{ e^{-sct} + e^{sct} + \frac{(\beta t)^\alpha}{\Gamma(\alpha)} e^{(sc-\beta)t} B(\alpha, 1) {}_1F_1(1; \alpha + 1; (\beta - 2sc)t) \right. \\
 &\quad \left. + \frac{(\beta t)^\alpha}{\Gamma(\alpha)} e^{-(sc+\beta)t} B(\alpha, 1) {}_1F_1(1; \alpha + 1; (\beta + 2sc)t) \right. \\
 &\quad \left. + (\beta t)^\alpha e^{-(\beta+sc)t} \sum_{n=0}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 1)} \Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) \right\}
 \end{aligned}$$

$$\begin{aligned}
& - (\beta t)^\alpha e^{-(\beta+sc)t} \sum_{n=0}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 1)} \Phi_2(n\alpha + \alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) \\
& - (\beta t)^\alpha e^{-(\beta+sc)t} \sum_{n=0}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 1)} \Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1; 2sct, \beta t) \\
& + (\beta t)^\alpha e^{-(\beta+sc)t} \sum_{n=0}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 1)} \Phi_2(n\alpha + \alpha, 1; 2n\alpha + \alpha + 1; 2sct, \beta t) \\
& - \left. \left(\frac{\beta}{\beta + 2sc} \right)^\alpha e^{sct} \frac{\gamma(\alpha, (\beta + 2sc)t)}{\Gamma(\alpha)} - \left(\frac{\beta}{\beta - 2sc} \right)^\alpha e^{-sct} \frac{\gamma(\alpha, (\beta - 2sc)t)}{\Gamma(\alpha)} \right\}.
\end{aligned}$$

The previous expression can be simplified by taking into account that, by (10) and (12),

$$\begin{aligned}
\Phi_2(0, 1; \alpha + 1; 2sct, \beta t) &= \sum_{k=0}^{+\infty} \sum_{r=0}^{+\infty} \frac{(0)_k (1)_r}{(\alpha + 1)_{k+r}} \frac{(2sct)^k (\beta t)^r}{k! r!} = \sum_{r=0}^{+\infty} \frac{(\beta t)^r}{(\alpha + 1)_r} = \alpha e^{\beta t} \frac{\gamma(\alpha, \beta t)}{(\beta t)^\alpha} \\
\Phi_2(0, 1; \alpha + 1; 2sct, 2sct + \beta t) &= \sum_{r=0}^{+\infty} \frac{(2sct + \beta t)^r}{(\alpha + 1)_r} = \alpha e^{(2sc + \beta)t} \frac{\gamma(\alpha, 2sct + \beta t)}{(2sct + \beta t)^\alpha},
\end{aligned}$$

and that, by making use of the integral representation for the Humbert function Φ_2 (cf. Equation (4.5) of [4]), we have

$$\begin{aligned}
\Phi_2(\alpha, 1; \alpha + 1; 2sct, \beta t) &= \frac{e^{\beta t}}{B(1, \alpha)} \int_0^1 x^{\alpha-1} e^{-\beta t x} {}_1F_1(\alpha; \alpha; 2sct x) dx \\
&= \alpha e^{\beta t} \frac{\gamma(\alpha, \beta t - 2sct)}{(\beta t - 2sct)^\alpha}
\end{aligned}$$

and

$$\Phi_2(\alpha, 1; \alpha + 1; 2sct, 2sct + \beta t) = \alpha e^{2sct + \beta t} \frac{\gamma(\alpha, \beta t)}{(\beta t)^\alpha}.$$

The final part of the proof is devoted to the investigation of the convergence of the series of Humbert functions appearing in the right-hand side of Equation (32). We start from the integral representation of the Humbert function Φ_2 (cf. Equation (4.5) of [4]),

$$\begin{aligned}
\Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) &= \frac{e^{2sct + \beta t}}{B(1, 2n\alpha + \alpha)} \\
&\quad \times \int_0^1 y^{2n\alpha + \alpha - 1} e^{-(2sct + \beta t)y} {}_1F_1(n\alpha; 2n\alpha + \alpha; 2sct y) dy,
\end{aligned}$$

and recall the following bound for the confluent hypergeometric function ${}_1F_1$ (cf. Equation 3.5 of [8]):

$${}_1F_1(n\alpha; 2n\alpha + \alpha; 2sct y) < 1 + \frac{n}{2n + 1} (e^{2sct y} - 1).$$

Hence, it follows that

$$\begin{aligned} \Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) &< \frac{n\alpha e^{2sct+\beta t}}{(\beta t)^{2n\alpha+\alpha}} \gamma(2n\alpha + \alpha, \beta t) \\ &+ \frac{(n+1)\alpha e^{2sct+\beta t}}{(\beta t + 2sct)^{2n\alpha+\alpha}} \gamma(2n\alpha + \alpha, \beta t + 2sct). \end{aligned}$$

Using Equation (4.1) of [36], we finally obtain

$$\Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) < \frac{e^{2sct+\beta t}}{2n\alpha + \alpha + 1} + \frac{n\alpha e^{2sct}}{2n\alpha + \alpha + 1} + \frac{(n+1)\alpha}{2n\alpha + \alpha + 1},$$

so that

$$\begin{aligned} &\sum_{n=1}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 1)} \Phi_2(n\alpha, 1; 2n\alpha + \alpha + 1; 2sct, 2sct + \beta t) \\ &< e^{2sct+\beta t} \sum_{n=1}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 2)} + e^{2sct} \sum_{n=1}^{+\infty} \frac{(n\alpha)(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 2)} + \sum_{n=1}^{+\infty} \frac{[(n+1)\alpha](\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 2)}. \\ &= e^{2sct+\beta t} {}_1\Psi_1 \left[(1, 1) (\alpha + 2, 2\alpha) \middle| (\beta t)^{2\alpha} \right] + e^{2sct} (\beta t)^{2\alpha} {}_1\Psi_1 \left[(2, 1) (3\alpha + 2, 2\alpha) \middle| (\beta t)^{2\alpha} \right] \\ &+ {}_1\Psi_1 \left[(2, 1) (\alpha + 2, 2\alpha) \middle| (\beta t)^{2\alpha} \right] - \frac{1}{\Gamma(\alpha + 2)} (e^{2sct+\beta t} + 1). \end{aligned}$$

The other series in (32) involving the Humbert functions can be treated in a similar way. □

As an immediate consequence of Theorem 4.2, we obtain the following result.

Proposition 4.3. *If the random variables U and D both have Erlang distribution with shape parameter m ∈ ℕ and rate parameter β > 0, for t > 0 and s ∈ ℝ we have*

$$\begin{aligned} M_X^s(t) = &\frac{1}{2} \left\{ (e^{-sct} + e^{sct}) e^{-\beta t} \sum_{j=0}^{m-1} \frac{(\beta t)^j}{j!} + \frac{(\beta t)^m}{m!} e^{sct-\beta t} {}_1F_1(1; m + 1; \beta t - 2sct) \right. \\ &+ \frac{(\beta t)^m}{m!} e^{-sct-\beta t} {}_1F_1(1; m + 1; 2sct + \beta t) + e^{sct-\beta t} \left[\beta t \sum_{n=1}^{+\infty} \sum_{j=0}^{nm+m-1} {}_1F_1(1; j + 2; \beta t - 2sct) \right. \\ &\times \left(\binom{nm + m - 1}{j} - \binom{nm - 1}{j} \right) \frac{(2sct)^j}{(j + 1)!} - \sum_{n=1}^{+\infty} (\beta t)^{2mn+m} (2sct + \beta t)^{1-m-2mn} \\ &\times \sum_{j=0}^{nm+m-1} \left(\binom{nm + m - 1}{j} - \binom{nm - 1}{j} \right) \frac{(2sct)^j}{(j + 1)!} {}_1F_1(1; j + 2; \beta t) + e^{-2sct} \sum_{n=1}^{+\infty} (\beta t)^{m+2mn} \\ &\times \sum_{j=0}^{2mn+m-2} \frac{[(\beta t)^{j+1-m-2mn} - (2sct + \beta t)^{j+1-m-2mn}]}{(j + 1)!} \\ &\left. \left. \times [{}_1F_1(nm; j + 2; 2sct) - {}_1F_1(nm + m; j + 2; 2sct)] \right] \right\}. \end{aligned}$$

Proof. Let $\alpha = m \in \mathbb{N}$ in Equation (32). The discrete component can be obtained from Equations (18) and (19). Aiming to evaluate the continuous component, let us recall that (cf. Equation (3.8) of [5])

$$\Phi_2(b, n+1; c; w, z) = (c-1) e^z \sum_{k=0}^n \frac{(-1)^k}{k!} z^k L_{n-k}^k(-z) \int_0^1 t^{c+k-2} e^{-tz} {}_1F_1(b; c-1; wt) dt,$$

where L_n^α is the generalized Laguerre polynomial of degree n , and that (cf. Equation (7.11.1.13) of [44])

$${}_1F_1(1; m; z) = (m-1)! z^{1-m} \left[e^z - \sum_{k=0}^{m-2} \frac{z^k}{k!} \right], \quad m = 1, 2, \dots$$

Hence

$$\begin{aligned} \Phi_2(1, nm; 2nm+m+1; 2sct+\beta t, 2sct) &= -e^{2sct} \Gamma(2nm+m+1) (2sct+\beta t)^{1-2nm-m} \\ &\times \left\{ \frac{1}{\beta t} \sum_{k=0}^{nm-1} \left(\frac{2sc}{\beta} \right)^k L_{nm-k-1}^k(-2sct) \left[1 - e^{\beta t} \sum_{l=0}^k \frac{(-\beta t)^l}{l!} \right] + \frac{1}{2sct} \sum_{k=0}^{nm-1} \sum_{j=0}^{2nm+m-2} \frac{(-1)^k}{k!} \right. \\ &\left. \times L_{nm-k-1}^k(-2sct) \frac{\left(\frac{2sc+\beta}{2sc} \right)^j}{j!} \gamma(k+j+1, 2sct) \right\}. \end{aligned}$$

Similar reasoning can be applied for the other functions involved in Equation (32). After some cumbersome calculations, and by Equations 5.3.10.20 of [3] and 2.19.3.6 of [43], the moment generating function (32) reads

$$\begin{aligned} M_X^s(t) &= \frac{1}{2} \left\{ (e^{-sct} + e^{sct}) e^{-\beta t} \sum_{j=0}^{m-1} \frac{(\beta t)^j}{j!} + \frac{(\beta t)^m}{m!} e^{sct-\beta t} {}_1F_1(1; m+1; \beta t - 2sct) \right. \\ &+ \frac{(\beta t)^m}{m!} e^{-sct-\beta t} {}_1F_1(1; m+1; 2sct + \beta t) + e^{sct} \sum_{n=1}^{+\infty} (\beta t)^{2nm+m} \sum_{k=0}^{nm+m-1} (-2sct)^k \\ &\times \left[L_{nm-k-1}^k(-2sct) - L_{nm+m-k-1}^k(-2sct) \right] \\ &\quad (2sct + \beta t)^{1-m-2mn} \frac{{}_1F_1(1; k+2; -\beta t)}{(k+1)!} - e^{-sct} \\ &\times \sum_{n=1}^{+\infty} \sum_{k=0}^{nm+m-1} (-2sct)^k \left[L_{nm-k-1}^k(-2sct) \right. \\ &\quad \left. - L_{nm+m-k-1}^k(-2sct) \right] (\beta t) \frac{{}_1F_1(1; k+2; 2sct - \beta t)}{(k+1)!} \\ &- e^{3sct-\beta t} \sum_{n=1}^{+\infty} (\beta t)^{2nm+m} \sum_{k=0}^{nm+m-1} (-2sct)^k \left[L_{nm-k-1}^k(-2sct) - L_{nm+m-k-1}^k(-2sct) \right] \\ &\times (2sct + \beta t)^{1-m-2mn} \sum_{j=0}^{2nm+m-2} (2sct + \beta t)^j \binom{k+j}{k} \frac{{}_1F_1(1; k+j+2; 2sct)}{(k+j+1)!} \end{aligned}$$

$$\begin{aligned}
 &+ e^{-sct-\beta t} (\beta t) \sum_{n=1}^{+\infty} \sum_{k=0}^{nm+m-1} (-2sct)^k \left[L_{nm-k-1}^k(-2sct) - L_{nm+m-k-1}^k(-2sct) \right] \\
 &\times \left. \sum_{j=0}^{2nm+m-2} (\beta t)^j \binom{k+j}{k} \frac{{}_1F_1(1; k+j+2; 2sct)}{(k+j+1)!} \right\}.
 \end{aligned}$$

Finally, the result follows from straightforward calculations. □

As further validation of Equation (4.3), in the following theorem we evaluate the moment generating function in the case of identically exponentially distributed random intertimes, finding a well-known result (see, for instance, Section 5 of [31]).

Theorem 4.3. *If the random variables U and D both have exponential distribution with parameter $\beta > 0$, for $t > 0$ and $s \in \mathbb{R}$ we have*

$$M_X^s(t) = e^{-\beta t} \left[\cosh \left(t\sqrt{s^2c^2 + \beta^2} \right) + \frac{\beta}{\sqrt{s^2c^2 + \beta^2}} \sinh \left(t\sqrt{s^2c^2 + \beta^2} \right) \right]. \tag{49}$$

Proof. The proof of Theorem 4.3 is provided in Appendix A. □

In the following proposition we give expressions for the first and second moment of X_t under the assumption of gamma-distributed intertimes.

Proposition 4.4. *If U_i and D_i ($i = 1, 2, \dots$) are gamma-distributed with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$, for $t > 0$ we have*

$$\begin{aligned}
 E(X_t) &= 0, \\
 E(X_t^2) &= c^2t^2 \left[1 - \frac{\gamma(\alpha, \beta t)}{\Gamma(\alpha)} \right] + c^2t^2 e^{-\beta t} \frac{(\beta t)^\alpha}{\Gamma(\alpha + 1)} \left\{ {}_1F_1(1, \alpha + 1, \beta t) - \frac{4}{\alpha + 1} {}_1F_1(2, \alpha + 2, \beta t) \right. \\
 &\quad \left. + \frac{8}{(\alpha + 1)(\alpha + 2)} {}_1F_1(3, \alpha + 3, \beta t) \right\} \\
 &\quad - \alpha(\beta t)^\alpha e^{-\beta t} (2ct)^2 \sum_{n=1}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 3)} {}_1F_1(2, 2n\alpha + \alpha + 3, \beta t). \tag{50}
 \end{aligned}$$

Proof. The proof follows immediately from Equation (32). □

Figure 5 shows some plots of the moment $E(X_t^2)$ given in Equation (50) for different values of (α, β) .

Proposition 4.5. *For $\alpha = 1$ and $t > 0$, Equation (50) reduces to the following well-known result (see, for instance, Equation (26) of [31]):*

$$E(X_t^2) = \frac{c^2}{2\beta^2} \left[2\beta t - (1 - e^{-2\beta t}) \right].$$

Proof. The proof of Proposition 4.5 is provided in Appendix B. □

5. Some results on the squared telegraph process

In this section we study the probability law of the stochastic process $Q_t := X_t^2, t > 0$, defined as the square of the generalized telegraph process. Hence, Q_t describes the square of the position of a particle performing a telegraph motion. As just stressed in [35], the sample paths of

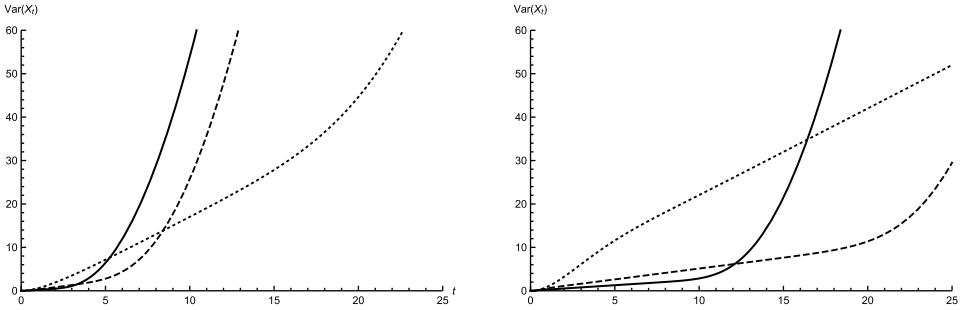


FIGURE 5. $E(X_t^2)$ for $c = 1$, $\beta = 0.5$ (dotted line), $\beta = 2$ (dashed line), $\beta = 4$ (solid line), with $\alpha = 1/2$ (left-hand side) and $\alpha = 2$ (right-hand side).

Q_t show motion reversals of the particle both when an event of the alternating counting process N_t occurs and when the underlying telegraph process X_t reaches the origin, which acts as a reflecting boundary. The interest in the functional Q_t arises in the context of establishing a link between finite-velocity random motions and diffusion processes. For instance, since it is well known that the standard Brownian motion is the limit, in some sense, of the telegraph process, the sum of squared telegraph processes can be treated as the analogue, in the setting of finite-velocity random motions, of the squared Bessel process.

Under the assumption of exponential distribution of the random variables U_i ($i = 1, 2, \dots$), the following theorem provides the distribution of Q_t , for all $t > 0$.

Proposition 5.1. *For all $i = 1, 2, \dots$, let us assume that the random variables U_i have exponential distribution with parameter $\lambda > 0$, and let $G_{D_i}(\cdot)$ be the distribution function of the random variables D_i . Denote by $G_D^{(n)}(\cdot)$, $n \geq 1$, the distribution function of $D^{(n)} := D_1 + \dots + D_n$. For $t > 0$ and $0 \leq z \leq c^2 t^2$, the distribution function of Q_t is given by*

$$F_Q(z, t) := P(Q_t \leq z) = F_X(\sqrt{z}, t) - F_X(-\sqrt{z}, t),$$

where

$$F_X(x, t) = 1 - H\left(\frac{ct - x}{2c}, \frac{ct + x}{2c}\right), \quad -ct < x \leq ct,$$

and

$$H(y, t) = \sum_{n=0}^{+\infty} e^{-\lambda t} \frac{(\lambda t)^n}{n!} G_D^{(n)}(y).$$

Proof. Let $N_t, t \geq 0$, be a Poisson process with parameter λ , and denote by

$$Y_t = \sum_{n=0}^{N_t} D_n$$

the compound Poisson process corresponding to N_t . Hence,

$$H_Y(y, t) := P(Y_t \leq y) = \sum_{n=0}^{+\infty} e^{-\lambda t} \frac{(\lambda t)^n}{n!} G_D^{(n)}(y).$$

Therefore, by Theorem 5.1 of [54], for $-ct \leq x \leq ct$, we have

$$F_X(x, t) := P(X_t \leq x | X_0 = 0) = 1 - H_Y\left(\frac{ct - x}{2c}, \frac{ct + x}{2c}\right),$$

so that the theorem immediately follows. □

Theorem 5.1. *If the random variables U_i ($i = 1, 2, \dots$) have exponential distribution with parameter $\lambda > 0$ and the random variables D_i ($i = 1, 2, \dots$) are gamma-distributed with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$, for $t > 0$ and $0 \leq z \leq c^2t^2$, we have*

$$F_Q(z, t) = e^{-\frac{(\lambda+\beta)t}{2}} \left\{ e^{\frac{(\lambda-\beta)\sqrt{z}}{2c}} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct + \sqrt{z}}{2c} \right) \right]^k W_{\alpha, k+1} \left(\lambda \beta^\alpha \left[\frac{ct - \sqrt{z}}{2c} \right] \left[\frac{ct + \sqrt{z}}{2c} \right]^\alpha \right) \right. \\ \left. - e^{-\frac{(\lambda-\beta)\sqrt{z}}{2c}} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct - \sqrt{z}}{2c} \right) \right]^k W_{\alpha, k+1} \left(\lambda \beta^\alpha \left[\frac{ct + \sqrt{z}}{2c} \right] \left[\frac{ct - \sqrt{z}}{2c} \right]^\alpha \right) \right\}, \tag{51}$$

where $W_{\rho, \theta}(z)$ is the Wright function (23).

Proof. By Theorem 5.1, assuming that the random variables D_i ($i = 1, 2, \dots$) have gamma distribution with shape parameter $\alpha > 0$ and rate parameter $\beta > 0$, it holds that

$$F_X(x, t) = 1 - \exp\left\{-\lambda \frac{ct + x}{2c}\right\} \left\{ 1 + \sum_{n=1}^{+\infty} \frac{\left[\lambda \frac{ct+x}{2c}\right]^n}{n!} \frac{\gamma(n\alpha, \beta \frac{ct-x}{2c})}{\Gamma(n\alpha)} \right\},$$

so that, for $0 \leq z \leq c^2t^2$, we have

$$F_Q(z, t) = F_X(\sqrt{z}, t) - F_X(-\sqrt{z}, t) = e^{-\frac{\lambda t}{2}} \left[2 \sinh\left(\frac{\lambda \sqrt{z}}{2c}\right) + e^{-\frac{\lambda t}{2}} \sum_{n=1}^{+\infty} \frac{\lambda^n}{n!} \frac{1}{\Gamma(n\alpha)} \right. \\ \left. \times \left\{ e^{\frac{\lambda \sqrt{z}}{2c}} \left(\frac{ct - \sqrt{z}}{2c}\right)^n \gamma\left(n\alpha, \frac{\beta}{2c}(ct + \sqrt{z})\right) \right. \right. \\ \left. \left. - e^{-\frac{\lambda \sqrt{z}}{2c}} \left(\frac{ct + \sqrt{z}}{2c}\right)^n \gamma\left(n\alpha, \frac{\beta}{2c}(ct - \sqrt{z})\right) \right\} \right]. \tag{52}$$

Hence, from (12), Equation (51) becomes

$$e^{-\frac{\lambda t}{2}} \left\{ 2 \sinh\left(\frac{\lambda \sqrt{z}}{2c}\right) + e^{\frac{\lambda \sqrt{z}}{2c} - \beta \left(\frac{ct + \sqrt{z}}{2c}\right)} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct + \sqrt{z}}{2c} \right) \right]^k \sum_{n=1}^{+\infty} \frac{\left[\lambda \beta^\alpha \left(\frac{ct - \sqrt{z}}{2c} \right) \left(\frac{ct + \sqrt{z}}{2c} \right)^\alpha \right]^n}{n! \Gamma(n\alpha + k + 1)} \right. \\ \left. - e^{-\frac{\lambda \sqrt{z}}{2c} - \beta \left(\frac{ct - \sqrt{z}}{2c}\right)} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct - \sqrt{z}}{2c} \right) \right]^k \sum_{n=1}^{+\infty} \frac{\left[\lambda \beta^\alpha \left(\frac{ct + \sqrt{z}}{2c} \right) \left(\frac{ct - \sqrt{z}}{2c} \right)^\alpha \right]^n}{n! \Gamma(n\alpha + k + 1)} \right\} \\ = e^{-\frac{\lambda t}{2}} \left\{ 2 \sinh\left(\frac{\lambda \sqrt{z}}{2c}\right) + e^{-\frac{\beta t}{2} + \frac{(\lambda - \beta)\sqrt{z}}{2c}} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct + \sqrt{z}}{2c} \right) \right]^k \right.$$

$$\begin{aligned} & \times \left[W_{\alpha,k+1} \left(\lambda \beta^\alpha \left[\frac{ct - \sqrt{z}}{2c} \right] \left[\frac{ct + \sqrt{z}}{2c} \right]^\alpha \right) - \frac{1}{\Gamma(k+1)} \right] \\ & - e^{-\frac{\beta t}{2} - \frac{(\lambda - \beta)\sqrt{z}}{2c}} \sum_{k=0}^{+\infty} \left[\beta \left(\frac{ct - \sqrt{z}}{2c} \right) \right]^k \left[W_{\alpha,k+1} \left(\lambda \beta^\alpha \left[\frac{ct + \sqrt{z}}{2c} \right] \left[\frac{ct - \sqrt{z}}{2c} \right]^\alpha \right) - \frac{1}{\Gamma(k+1)} \right] \Bigg\}, \end{aligned}$$

so that the proof immediately follows. □

Remark 5.1. Starting from Equation 6.5.29 of [1], the incomplete gamma function in (52) can be expressed in terms of the two-parametric Mittag-Leffler function (39). The latter, by Equation 4.4.6 of [24], can be written in terms of the Riemann–Liouville fractional integral of a suitable function (63).

Hence, recalling that the generalized Prabhakar fractional integral $\mathcal{E}_{\alpha, \beta; c^+}^{\omega, \rho, \kappa} f(x)$ (see e.g. [50] and also Appendix C for some details) can be expressed as a series of fractional integrals (cf. Theorem 2.1 of [20]), Equation (51) admits, for $t > 0$, the following alternative expression:

$$\begin{aligned} F_Q(z, t) = & e^{-\frac{\lambda t}{2} + \frac{\lambda\sqrt{z}}{2c} - \frac{\beta}{2c}(ct + \sqrt{z})} \left\{ W_{\alpha,1} \left(\frac{\lambda}{2c} (ct - \sqrt{z}) \left(\frac{\beta}{2c} (ct + \sqrt{z}) \right)^\alpha \right) \right. \\ & \left. + \mathcal{E}_{\alpha, 1; 0^+}^{\frac{\lambda}{2c}(ct - \sqrt{z}), \rho, 0} e^{\frac{\beta}{2c}(ct + \sqrt{z})} \right\} \\ - & e^{-\frac{\lambda t}{2} - \frac{\lambda\sqrt{z}}{2c} - \frac{\beta}{2c}(ct - \sqrt{z})} \left\{ W_{\alpha,1} \left(\frac{\lambda}{2c} (ct + \sqrt{z}) \left(\frac{\beta}{2c} (ct - \sqrt{z}) \right)^\alpha \right) \right. \\ & \left. + \mathcal{E}_{\alpha, 1; 0^+}^{\frac{\lambda}{2c}(ct + \sqrt{z}), \rho, 0} e^{\frac{\beta}{2c}(ct - \sqrt{z})} \right\}. \end{aligned}$$

Appendix A. Proof of Theorem 4.3

The present section is devoted to the proof of Theorem 4.3.

Let us set $m = 1$ in the statement of Theorem 4.3. The discrete component simplifies to

$$e^{-\beta t} (e^{-sct} + e^{sct}).$$

From Equations 7.11.1.13 of [44] and 13.4.4 of [1], we obtain

$$\begin{aligned} M_X^s(t) = & \frac{1}{2} \left\{ e^{-\beta t} (e^{-sct} + e^{sct}) \right. \\ & + \frac{\beta t e^{-sct - \beta t}}{4s^2 c^2 t^2 - \beta^2 t^2} \left[2sct (e^{\beta t} + 1) (e^{2sct} - 1) - \beta t (e^{\beta t} - 1) (e^{2sct} + 1) \right] \\ & + e^{sct - \beta t} \left[\beta t \sum_{n=1}^{+\infty} \sum_{r=0}^{n-1} \binom{n-1}{r} \frac{(2sct)^{r+1}}{(r+2)!} {}_1F_1(1; r+3; \beta t - 2sct) \right. \\ & - \beta t \sum_{n=1}^{+\infty} \left[\frac{(\beta t)^2}{(2sct + \beta t)^2} \right]^n \sum_{r=0}^{n-1} \binom{n-1}{r} \frac{(2sct)^{r+1}}{(r+2)!} {}_1F_1(1; r+3; \beta t) \\ & - e^{-2sct} 2sct \beta t \sum_{n=1}^{+\infty} \sum_{j=0}^{2n-1} \frac{(\beta t)^j}{(j+2)!} {}_1F_1(n+1; j+3; 2sct) \\ & \left. \left. + e^{-2sct} 2sct \beta t \sum_{n=1}^{+\infty} \left[\frac{(\beta t)^2}{(2sct + \beta t)^2} \right]^n \sum_{j=0}^{2n-1} \frac{(2sct + \beta t)^j}{(j+2)!} {}_1F_1(n+1; j+3; 2sct) \right] \right\}. \end{aligned}$$

Let us set

$$A := \sum_{n=1}^{+\infty} \sum_{j=0}^{2n-1} \frac{(\beta t)^j}{(j+2)!} {}_1F_1(n+1; j+3; 2sct), \tag{53}$$

$$B := \sum_{n=1}^{+\infty} \left[\frac{(\beta t)^2}{(2sct + \beta t)^2} \right]^n \sum_{j=0}^{2n-1} \frac{(2sct + \beta t)^j}{(j+2)!} {}_1F_1(n+1; j+3; 2sct), \tag{54}$$

$$C := \sum_{n=1}^{+\infty} \left[\frac{(\beta t)^2}{(2sct + \beta t)^2} \right]^n \sum_{r=0}^{n-1} \binom{n-1}{r} \frac{(2sct)^{r+1}}{(r+2)!} {}_1F_1(1; r+3; \beta t), \tag{55}$$

$$D := \beta t \sum_{n=1}^{+\infty} \sum_{r=0}^{n-1} \binom{n-1}{r} \frac{(2sct)^{r+1}}{(r+2)!} {}_1F_1(1; r+3; \beta t - 2sct). \tag{56}$$

Note that, by Equations 13.1.27 of [1], 13.6.9 of [1], and 7.11.1.5 of [44], we have

$$\begin{aligned} A &= e^{2sct} \sum_{n=2}^{+\infty} \sum_{j=0}^{n-2} \frac{(\beta t)^j}{(j+2)!} {}_1F_1(-n+2+j; j+3; -2sct) \\ &\quad + \sum_{n=1}^{+\infty} \sum_{j=n-1}^{2n-1} \frac{(\beta t)^j}{(j+2)!} {}_1F_1(n+1; j+3; 2sct) \\ &= e^{2sct} \sum_{n=2}^{+\infty} \sum_{j=0}^{n-2} \frac{(\beta t)^j (n-2-j)!}{n!} L_{n-2-j}^{j+2}(-2sct) \\ &\quad + \sum_{n=1}^{+\infty} \sum_{r=0}^{n-1} \frac{(\beta t)^{r+n-1}}{(r+n-1)!} {}_1F_1(n+1; r+n+2; 2sct) \\ &\quad + \sum_{n=1}^{+\infty} \frac{(\beta t)^{2n-1}}{(2n+1)!} \Gamma\left(n + \frac{3}{2}\right) 2^{2n+1} e^{sct} (2sct)^{-\frac{1}{2}-n} I_{n+\frac{1}{2}}(sct). \end{aligned}$$

By interchanging the order of summation in the previous formula, we get

$$\begin{aligned} A &= e^{2sct} \sum_{n=2}^{+\infty} \sum_{j=0}^{n-2} \frac{(\beta t)^j (n-2-j)!}{n!} L_{n-2-j}^{j+2}(-2sct) + \sum_{r=0}^{+\infty} (\beta t)^r \sum_{h=0}^{+\infty} \frac{(\beta t)^{h+r}}{(h+2r+2)!} \\ &\quad \times {}_1F_1(h+r+2; h+2r+3; 2sct) + \frac{2e^{sct}}{\beta t \sqrt{2sct}} \sum_{n=1}^{+\infty} \frac{\left(\frac{2\beta^2 t^2}{sct}\right)^n}{(2n+1)!} \Gamma\left(n + \frac{3}{2}\right) I_{n+\frac{1}{2}}(sct) \\ &= e^{2sct} \sum_{n=2}^{+\infty} \sum_{j=0}^{n-2} \frac{(\beta t)^j (n-2-j)!}{n!} L_{n-2-j}^{j+2}(-2sct) + e^{2sct} \sum_{r=0}^{+\infty} \frac{(\beta t)^{2r}}{\Gamma(2r+3)} \sum_{h=0}^{+\infty} \frac{(1)_h (\beta t)^h}{h! (2r+3)_h} \\ &\quad \times {}_1F_1(r+1; h+2r+3; -2sct) + \frac{e^{sct}}{\beta t \sqrt{2sct}} \sum_{n=1}^{+\infty} \frac{\left(\frac{2\beta^2 t^2}{sct}\right)^n}{(2n)!} \Gamma\left(n + \frac{1}{2}\right) I_{n+\frac{1}{2}}(sct). \tag{57} \end{aligned}$$

From Equation 6.6.1.1 of [44], and taking into account Equation 5.8.3.4 of [43] and the first equation in 10.2.13 of [1], we obtain

$$A = e^{2sct} \sum_{n=2}^{+\infty} \sum_{j=0}^{n-2} \frac{(\beta t)^j (n-2-j)!}{n!} L_{n-2-j}^{j+2}(-2sct) + \frac{e^{sct}}{2sct \cdot \beta t} \left(1 - \frac{\beta}{sc}\right) \left(1 + \frac{\beta^2}{s^2 c^2}\right)^{-\frac{1}{2}} \\ \times \sinh\left(sct \sqrt{1 + \frac{\beta^2}{s^2 c^2}}\right) - \frac{e^{sct}}{2sct \cdot \beta t} \cosh\left(sct \sqrt{1 + \frac{\beta^2}{s^2 c^2}}\right) + \frac{e^{2sct} (e^{\beta t} - 1) + 1}{2sct \cdot \beta t}.$$

Similarly, from Equation (54), it follows that

$$B = e^{2sct} \sum_{j=0}^{+\infty} \frac{(\beta t)^{2j+4}}{(j+2)! (2sct + \beta t)^{j+4}} \sum_{r=0}^{+\infty} \left[\frac{(\beta t)^2}{(2sct + \beta t)^2} \right]^r {}_1F_1(-r; j+3; -2sct) \\ + \sum_{h=0}^{+\infty} (2sct + \beta t)^{h-1} \left[\frac{(\beta t)^2}{2sct + \beta t} \right]^h \sum_{s=0}^{+\infty} \frac{\left[\frac{(\beta t)^2}{2sct + \beta t} \right]^s}{(s+2h+1)!} {}_1F_1(s+h+1; s+2h+2; 2sct) \\ - \frac{1}{2sct + \beta t} {}_1F_1(1; 2; 2sct). \quad (58)$$

By Equation 6.6.1.6 of [44], the first term in Equation (58) reads

$$\frac{e^{2sct}}{2sct \beta t (2sct + \beta t)} \left[-e^{\frac{(\beta t)^2}{2sct + 2\beta t}} (2sct + 2\beta t) + e^{\frac{(\beta t)^2}{2sct + \beta t}} (2sct + \beta t) + \beta t \right],$$

whereas, by Equations 6.6.1.1 and 7.11.1.7 of [44], the second term in Equation (58) equals

$$- \frac{\sqrt{\pi}}{4sct (2sct + \beta t)} \int_0^{-x} e^{-\tau} \left[\frac{(\beta t)^2}{2sct(2sct + \beta t)} + 1 \right] e^{\frac{2sct + \tau}{2}} (2sct + \tau)^{1/2} \\ \cdot \sum_{h=0}^{+\infty} \frac{\left[\frac{(\beta t)^2 (2sct + \tau)}{(2sct)^2} \right]^h}{h!} \left[I_{h-\frac{1}{2}} \left(\frac{2sct + \tau}{2} \right) - I_{h+\frac{1}{2}} \left(\frac{2sct + \tau}{2} \right) \right] d\tau.$$

Finally, Equations 5.8.3.4 and 5.8.3.6 of [43] and Equation 7.11.1.13 of [44] lead to

$$B = \frac{1}{2sct \cdot \beta t} \left\{ e^{sct} \left[\cosh\left(sct \sqrt{1 + \frac{\beta^2}{s^2 c^2}}\right) + \frac{sc + \beta}{\sqrt{s^2 c^2 + \beta^2}} \sinh\left(sct \sqrt{1 + \frac{\beta^2}{s^2 c^2}}\right) \right] \right. \\ \left. + \frac{1}{2sct + \beta t} \left[\beta t - (2sct + 2\beta t) e^{\frac{(2sct + \beta t)^2}{2sct + 2\beta t}} \right] \right\}.$$

Equation (55) can easily be simplified with the help of Equation 6.6.1.1 of [44], so that

$$C = \frac{1}{\beta t (2sct + \beta t)} \left[2sct + \beta t + \beta t e^{\beta t} - (2sct + 2\beta t) e^{\frac{\beta^2 t^2}{2sct + 2\beta t}} \right].$$

Since

$$D = 2sct \beta t e^{-2sct} \sum_{j=0}^{+\infty} \frac{(\beta t)^j}{\Gamma(j+3)} \sum_{n=1}^{+\infty} \sum_{r=0}^{n-1} \binom{n-1}{r} \frac{(2sct)^r}{(j+3)_r} {}_1F_1(r+2; j+r+3; 2sct),$$

by Equation 5.3.5.5 of [44], we have

$$D = sct \beta t e^{-2sct} \sum_{n=1}^{+\infty} \sum_{j=0}^{+\infty} \frac{(1)_j (\beta t)^j}{(3)_j j!} {}_1F_1(n+1; j+3; 2sct). \tag{59}$$

The inner series of Equation (59), by Equations 6.6.1.1 of [44] and 13.6.9 of [1], reduces to

$$2 (2sct)^{-2} e^{\beta t} \frac{(n-1)!}{\Gamma(n+1)} \int_0^{2sct} z e^{-z \left[\frac{\beta}{2sc} - 1 \right]} L_{n-1}^1(-z) dz.$$

Finally, recalling Equation 22.3.9 of [1], we get

$$\begin{aligned} D &= \frac{2sct \beta t e^{-2sct + \beta t}}{(\beta t - 2sct)^2} \sum_{n=1}^{+\infty} \sum_{k=0}^{n-1} \frac{\binom{n-1}{k}}{(k+1)!} \left(\frac{2sct}{\beta t - 2sct} \right)^k \gamma(k+2, \beta t - 2sct) \\ &= \frac{\beta}{2sc} e^{\beta t - 2sct} \sum_{n=2}^{+\infty} \sum_{k=0}^{n-1} \frac{\binom{n-1}{k}}{(k+1)!} \left[\frac{2sc}{\beta - 2sc} \right]^{k+2} \gamma(k+2, \beta t - 2sct) \\ &\quad + \frac{\beta}{2sc} e^{\beta t - 2sct} \left[\frac{2sc}{\beta - 2sc} \right]^2 \gamma(2, \beta t - 2sct). \end{aligned}$$

Hence, after some calculations, the moment generating function can be expressed as

$$\begin{aligned} M_X^s(t) &= \frac{1}{2} \left\{ e^{-\beta t} (e^{-sct} + e^{sct}) + \frac{\beta}{2sc} e^{-sct} \sum_{n=2}^{+\infty} \sum_{k=0}^{n-1} \frac{\binom{n-1}{k}}{(k+1)!} \left(\frac{2sct}{\beta t - 2sct} \right)^{k+2} \gamma(k+2, \beta t - 2sct) \right. \\ &\quad \left. - 2sct \beta t e^{sct - \beta t} \sum_{n=2}^{+\infty} \sum_{j=0}^{n-2} \frac{(\beta t)^j (n-2-j)!}{n!} L_{n-2-j}^{j+2}(-2sct) \right\} + \frac{e^{-\beta t}}{2} \frac{e^{-sct}}{(2sct - \beta t)^2} \\ &\quad \times \left[-(2sct - \beta t)^2 (e^{2sct + \beta t} + 1) + e^{2sct} 2sct \beta t (2sct - \beta t) - (e^{2sct} - e^{\beta t}) (\beta t)^2 \right] \\ &\quad + e^{-\beta t} \cosh \left[sct \sqrt{1 + \left(\frac{\beta}{sc} \right)^2} \right] + e^{-\beta t} 2\beta t \frac{\sinh \left[sct \sqrt{1 + \left(\frac{\beta}{sc} \right)^2} \right]}{\sqrt{(2sct)^2 + 4(\beta t)^2}}. \end{aligned}$$

Let us now note that by Equation 13.6.9 of [1] and the binomial theorem,

$$\begin{aligned} &2sct \beta t e^{sct - \beta t} \sum_{n=2}^{+\infty} \sum_{j=0}^{n-2} \frac{(\beta t)^j (n-2-j)!}{n!} L_{n-2-j}^{j+2}(-2sct) \\ &= 2sct \beta t e^{-sct} \sum_{l=0}^{+\infty} \frac{(2sct - \beta t)^l}{l!} \sum_{n=2}^{+\infty} \frac{(\beta t)^{n-2} (n+1)_l}{(l+n)!} \sum_{h=0}^{n-2} h! \left(\frac{1}{\beta t} \right)^h L_h^{l+n-h}(-\beta t). \tag{60} \end{aligned}$$

Hence, from the formulas 48.19.3 of [25] and 22.7.30 of [1], Equation (60) becomes

$$\begin{aligned}
 & 2sct \beta t e^{-sct} \sum_{l=0}^{+\infty} \frac{(2sct - \beta t)^l}{l!} \sum_{n=2}^{+\infty} \frac{(\beta t)^{n-2} (n+1)_l}{(l+n)!} \left\{ 1 + \frac{(n-1)! \left[L_{n-2}^{l+2}(-\beta t) + L_{n-1}^{l+1}(-\beta t) \right]}{(\beta t)^{n-2} (l+n+1)} \right. \\
 & \left. - \frac{l + \beta t + n + 1}{l + n + 1} \right\} = 2sct \beta t e^{-sct} \sum_{l=0}^{+\infty} \frac{(2sct - \beta t)^l}{l!} \sum_{n=2}^{+\infty} \frac{1}{n(l+n+1)} L_{n-1}^{l+2}(-\beta t) \\
 & - 2scte^{-sct} \sum_{l=0}^{+\infty} \frac{(2sct - \beta t)^l}{l!} \sum_{n=2}^{+\infty} \frac{(\beta t)^n}{n!(l+n+1)}.
 \end{aligned}$$

Finally, by taking into account Equations 6.5.29, 9.6.51, and 22.10.14 of [1], Equation 6.8.1.3 of [3], and the formulas 6.8.1.3 of [3], 3.38.1.1 of [45], and 5.2.3.1 of [43], after some algebraic manipulations we get

$$\begin{aligned}
 M_X^s(t) = & \frac{1}{2} \left\{ e^{-\beta t} (e^{-sct} + e^{sct}) + \frac{\beta e^{-sct}}{2sc} \sum_{n=2}^{+\infty} \sum_{k=0}^{n-1} \frac{\binom{n-1}{k}}{(k+1)!} \left(\frac{2sct}{\beta t - 2sct} \right)^{k+2} \gamma(k+2, \beta t - 2sct) \right. \\
 & \left. - 2sct \beta t e^{-sct - \beta t} \sum_{n=2}^{+\infty} \sum_{r=0}^{+\infty} \frac{(\beta t - 2sct)^r}{(r+2)!} {}_1F_1(n+r+2, r+3, 2sct) \right\} \\
 & + \cosh(sct) \times [\sinh(\beta t) - \cosh(\beta t)] \\
 & + e^{-\beta t} \left\{ \cosh \left[sct \sqrt{1 + \left(\frac{\beta}{sc} \right)^2} \right] + 2\beta t \frac{\sinh \left[sct \sqrt{1 + \left(\frac{\beta}{sc} \right)^2} \right]}{\sqrt{(2sct)^2 + 4(\beta t)^2}} \right\}.
 \end{aligned}$$

The proof follows from the formula 6.5.12 of [1], after straightforward calculations.

Appendix B. Proof of Proposition 4.5

In the present section we provide the proof of Proposition 4.5.

By Equation 7.11.2.14 of [44], and recalling Equation 6.5.2 of [1] and Equation (39), we have

$$\begin{aligned}
 & \sum_{n=1}^{+\infty} \frac{(\beta t)^{2n\alpha}}{\Gamma(2n\alpha + \alpha + 3)} {}_1F_1(2, 2n\alpha + \alpha + 3, \beta t) \\
 & = (\beta t)^{-\alpha-1} \int_0^{\beta t} e^{\beta t-x} x^\alpha \left[-\frac{1}{\Gamma(\alpha + 1)} + E_{2\alpha, \alpha+1}(x^{2\alpha}) \right] dx \\
 & - (\beta t)^{-\alpha-2} \int_0^{\beta t} e^{\beta t-x} x^{\alpha+1} \left[-\frac{1}{\Gamma(\alpha + 1)} + E_{2\alpha, \alpha+1}(x^{2\alpha}) \right] dx. \tag{61}
 \end{aligned}$$

Hence, from Equations 4.12.1.a and 3.9.2 of [24], Equation (61) can be written as

$$\frac{e^{\beta t}(\beta t)^{-\alpha-2}}{\Gamma(\alpha+1)} \{ \alpha(1+\alpha-\beta t)\Gamma(\alpha) + \beta t\Gamma(1+\alpha, \beta t) - \Gamma(\alpha+2, \beta t) \} + \frac{1}{2}(\beta t)^{-\alpha-1} \times \int_0^{\beta t} e^{\beta t-x} [E_\alpha(x^\alpha) - E_\alpha(-x^\alpha)] dx + \frac{1}{2}(\beta t)^{-\alpha-2} \int_0^{\beta t} e^{\beta t-x} x [E_\alpha(-x^\alpha) - E_\alpha(x^\alpha)] dx, \tag{62}$$

where $\Gamma(\delta, z)$ denotes the upper incomplete gamma function, and

$$E_\alpha(z) = \sum_{k=0}^{+\infty} \frac{z^k}{\Gamma(\alpha k + 1)}$$

is the Mittag-Leffler function (see, for instance, Equation 3.1.1 of [24]).

Recalling Equation 3.2.1 of [24], the formula (62) for $\alpha = 1$ becomes

$$\frac{e^{-\beta t}}{8\beta^3 t^3} \left\{ e^{2\beta t} [17 + 2\beta t(\beta t - 5)] - 1 - 8e^{\beta t}(2 + \beta t) \right\}.$$

Hence, the proof immediately follows from Equation (50) and recalling Equations 7.11.2.20, 7.11.2.38, and 7.11.2.56 of [44].

Appendix C. Some remarks on the fractional calculus

For any sufficiently well-behaved function f , the Riemann–Liouville fractional integral $I_{c^+}^\alpha f$ of order $\alpha > 0$ is defined as

$$I_{c^+}^\alpha f(x) := \frac{1}{\Gamma(\alpha)} \int_c^x \frac{f(t)}{(x-t)^{1-\alpha}} dt, \quad x > c, \quad \alpha > 0. \tag{63}$$

Note that the values of $I_{c^+}^\alpha f(x)$ for $\alpha > 0$ are finite if $x > c$, but it may happen that the limit (if it exists) of $I_{c^+}^\alpha f(x)$ is infinite as $x \rightarrow c^+$.

The fundamental property of the fractional integrals is the additive index law (semigroup property), according to which

$$I_{c^+}^\alpha \cdot I_{c^+}^\beta = I_{c^+}^{\alpha+\beta}, \quad \alpha, \beta > 0.$$

Among the various operators of fractional integration studied in the recent literature, Srivastava [50] introduced the generalized Prabhakar fractional integral defined, for $\rho, \omega \in \mathbb{C}$, $\Re(\alpha) > \max\{0, \Re(\kappa) - 1\}$, $\min\{\Re(\kappa), \Re(\beta)\} > 0$, as

$$\mathcal{E}_{\alpha, \beta; c^+}^{\omega, \rho, \kappa} f(x) := \int_c^x (x-t)^{\beta-1} E_{\alpha, \beta}^{\rho, \kappa} [\omega(x-t)^\alpha] f(t) dt, \quad x > c. \tag{64}$$

This operator contains in the kernel the generalized Mittag-Leffler function $E_{\alpha, \beta}^{\rho, \kappa}$, given by

$$E_{\alpha, \beta}^{\rho, \kappa}(z) := \sum_{n=0}^{+\infty} \frac{(\rho)_{\kappa n} z^n}{\Gamma(\alpha n + \beta)} = \sum_{n=0}^{+\infty} \frac{\Gamma(\rho + \kappa n) z^n}{\Gamma(\rho) \Gamma(\alpha n + \beta)}.$$

Note that, in the special case $\omega = 0$, the integral operator $\mathcal{E}_{\alpha, \beta; c^+}^{\omega, \rho, \kappa} f(x)$ reduces to the right-handed Riemann–Liouville fractional integral operator (63).

A series formula for the generalized Prabhakar integral defined by (64) is provided in Theorem 2.1 of [20]. Indeed for any interval $(c, d) \subset \mathbb{R}$ and any function $f \in L^1(c, d)$, we have

$$\mathcal{E}_{\alpha, \beta; c^+}^{\omega, \rho, \kappa} f(x) = \sum_{n=0}^{+\infty} \frac{\Gamma(\rho + \kappa n) \omega^n}{\Gamma(\rho) n!} I_{c^+}^{\alpha n + \beta} f(x).$$

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