Diffusion equations for a Markovian jumping process

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We consider a Markovian jumping process which is defined in terms of the jump-size distribution and the waiting-time distribution with a position-dependent frequency, in the diffusion limit. We assume the power-law form for the frequency. For small steps, we derive the Fokker-Planck equation and show the presence of the normal diffusion, subdiffusion and superdiffusion. For the Lévy distribution of the step-size, we construct a fractional equation, which possesses a variable coefficient, and solve it in the diffusion limit. Then we calculate fractional moments and define fractional diffusion coefficient as a natural extension to the cases with the divergent variance. We also solve the master equation numerically and demonstrate that there are deviations from the Lévy stable distribution for large wave numbers.

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I. INTRODUCTION

The diffusion is called anomalous if the mean squared displacement of the Brownian particle does not rise with time linearly, as it is the case for the usual Brownian motion, but slower (subdiffusion) or faster (superdiffusion). There are many examples of physical systems which exhibit such anomalous transport; they involve complex systems, disordered systems, semiconductors, polymers, glasses, turbulent plasma, and many others (for a review see [1, 2]. Physical situations in which the deviation of the linear dependence is expected comprise: a porous and inhomogeneous environment with traps and barriers, a coupling to a fractal heat bath via a random matrix interaction [3], long-range and/or long-time correlations. Also the transport phenomena in dynamical hamiltonian systems often exhibit anomalous behaviour because the corresponding phase space can contain some regular structures which act as dynamical traps [4, 5]. A trajectory sticks to such self-similar hierarchy of islands in chaotic environment and abide on regular paths for a long time. The emergence of the anomalous diffusion means that the traditional approach, involving the standard Fokker-Planck equation (FPE) with the diffusion constant, is no longer valid. Usually, the anomalous diffusion is attributed to memory effects, which are especially pronounced in non-Markovian descriptions, e.g. by means of the generalized Langevin equation [6] and the continuous-time random walk models (CTRW) [7]. In the diffusive limit, those approaches resolve themselves to the fractional equations formalism [8] which contain the Riemann-Liouville operator as the fractional time-derivative; the step-size distribution is the Gaussian. As a direct consequence of the nonlocality in time and, therefore, of the lack of Markovian property, the uncoupled CTRW predicts subdiffusive solutions in this case [9, 10]. However, the anomalous behaviour is possible also for the Markovian processes; they are described by FPE with the position-dependent diffusion term and physically can correspond e.g. to the diffusion on fractal objects [11] or to the Langevin equation with the multiplicative noise [12]. In this paper we discuss another example of such Markovian process.

A more general approach considers the Lévy distribution which is defined, in its symmetric version, via the characteristic function \( \phi(k) = \exp(-\alpha k^\mu) \), where \( 0 < \mu \leq 2 \). If \( \mu \neq 2 \) the second moment is infinite and the generalized central limit theorem must be applied. Physically, Lévy processes can reflect self-similarity: they are the stable solutions of the renormalization group method and are invariant under the scaling of position and time. In contrast to the Gaussian distribution, they include large fluctuations. The Lévy distributions are present in many problems connected with various branches of science, including not only physics but also biology, economy, financial market research, etc. The diffusion equation for the Lévy process involves a fractional derivative over the process value \( \frac{\partial}{\partial \tau} \) and for \( \mu = 2 \) it resolves itself to the ordinary FPE. Since the variance diverges for \( \mu < 2 \) and the traditional description of the diffusion process is no longer valid, one has to introduce new concepts, e.g. to study fractional moments [13] or to restrict the integration in averaging to a finite box which scales with time in a prescribed way in [14, 15].

Problems related to generalized diffusion equations, which contain either anomalous behaviour of the variance or infinite fluctuations, are the subject of the present paper. We deal with a jumping process which is Markovian, defined in terms of a jump-size distribution \( Q(x) \) and the waiting time distribution \( P_\tau(\tau) \). A peculiarity of the distribution \( P_\tau(\tau) \) consists in a position-dependent jumping rate. The process is defined in Sec.II. In Sec.III we derive the FPE as a small step limit of the master equation. Sec.IV is devoted to the fractional diffusion equation which is an approximation of the master equation in the diffusion limit for \( Q \) given by the Lévy distribution; we solve this equation, discuss its diffusion properties and compare with numerical solution of the master equation. The results of our analysis are discussed in Sec.V.
II. DEFINITION OF THE PROCESS

The jumping process we are to deal with in this paper is a stepwise constant stochastic process $x(t)$ defined in terms of two probability distributions [19]. The waiting time density distribution determines the time intervals $\tau_i$ between consecutive jumps and it is assumed in the Poissonian form:

$$P_p(\tau) = \nu(x) e^{-\nu(x) \tau}, \quad (1)$$

where the jumping rate $\nu(x)$ depends on the process value (the position). The size of the jumps is determined by a given normalized distribution $Q(x = x' - x)$ and the master equation derived. It is of the form

$$\frac{\partial}{\partial t} p(x, t) = -\nu(x) p(x, t) + \int Q(x - x') \nu(x') p(x', t) dx'.$$

Because of the variable jumping rate $\nu(x)$, the above process can be regarded as a version of the kangaroo process [20, 21]. The difference consists in the jump-size dependence of $Q$ – in the kangaroo process $Q$ depends only on the current position. Due to that property, the Eq. (2) can describe transport phenomena [19]. On the other hand, the process constitutes a special case of the coupled CTRW. Taking into account the position dependence of jumping rate is an important generalization of traditional random walk approaches. Such dependence is expected in many phenomena in which inhomogeneity of the environment cannot be neglected [22].

The stationary solution of Eq. (2) can be easily obtained: $P(x) = 1/\nu(x)$. The normalization condition imposes restrictions on the function $\nu(x)$: it must rise sufficiently fast in the infinity and sufficiently slow near zero. If, in turn, $1/\nu(x)$ has poles at some $x$, the stationary solution also exists and it is in the form of a combination of the delta functions. In the other cases $P(x)$ does not exist. A special version of the stationary process, defined on the circle, exhibits long-time correlations and can serve as a model of the $1/f$ noise [23].

The general, time-dependent solution of the Eq. (2) can be obtained by using the Laplace transforms. The formal expression for the Laplace transform of $p(x, t)$ is the following [19]:

$$\tilde{p}(x, s) = \frac{p_0(x)}{s + \nu(x)} + \frac{1}{s + \nu(x)} \int \frac{\nu(x_0) p_0(x_0) Q(x - x_0)}{s + \nu(x_0)} dx_0 +$$

$$+ \frac{1}{s + \nu(x)} \sum_{k=2}^{\infty} \int \frac{\nu(x_0) p_0(x_0)}{s + \nu(x_0)} Q(x - x_{k-1}) \left[ \prod_{i=2}^{k} \frac{\nu(x_{i-1}) Q(x_{i-1} - x_{i-2})}{s + \nu(x_{i-1})} dx_{i-1} \right] dx_0, \quad (3)$$

where $p_0(x)$ stands for the initial condition.

By multiplying the Eq. (2) by $x^2$ and by integrating over $x$, one can obtain the equation which governs the time evolution of the variance. Assuming that $Q$ has the vanishing mean and finite second moment, we yield the following result:

$$\frac{\partial \langle x^2 \rangle_p}{\partial t} = \langle x^2 \rangle_Q \nu. \quad (4)$$

The simple case $\nu(x) = \text{const}$ can be solved completely and the Eq. (4) predicts the normal diffusion. However, if $\langle x \rangle_Q$ does not vanish, the diffusion becomes ballistic [19]. For $Q$ with divergent second moment, in turn, $\langle x^2 \rangle_p$ is infinite.

III. SMALL JUMPS: THE FOKKER-PLANCK EQUATION

The expression (3) is difficult to handle and one has to resort to approximations. In the limit of small jumps, the process becomes continuous both in space and time and FPE may be a candidate for such approximation. In order to construct it, we apply the Kramers-Moyal expansion [24] of the master equation. In that method, one changes the integration variable in the Eq. (2) by introducing the step size $r = x - x'$ and, after the expansion of the function $p(x - r, t) \nu(x - r)$ around $r = x$, one obtains the master equation in a form of the following series

$$\frac{\partial p(x, t)}{\partial t} = \sum_{n=1}^{\infty} \frac{(-1)^n}{n!} \langle r^n \rangle_Q \left( \frac{\partial}{\partial x} \right)^n[p(x, t) \cdot \nu(x)] \quad (5)$$

where the jumping rate $\nu(x)$ depends on the process value (the position). The size of the jumps is determined by a given normalized distribution $Q(x = x' - x)$. Then the infinitesimal transition probability can be easily constructed and the master equation derived. It is of the form
which is still exact. The approximation consists in neglecting all terms of the order higher than two. Finally, we obtain the FPE:

$$\frac{\partial p(x, t)}{\partial t} = \frac{\sigma^2}{2} \frac{\partial^2 [\nu(x)p(x, t)]}{\partial x^2},$$

(6)

where $\sigma^2 = \langle r^2 \rangle_Q$. Therefore, the jumping rate may be interpreted as the position-dependent diffusion coefficient. The approximation is valid if the jumps are small and the function $\nu(x)$ is smooth [25]:

$$Q(r) \approx 0 \quad \text{for} \quad r > \delta$$

$$\nu(x + \Delta x) \approx \nu(x) \quad \text{for} \quad \Delta x < \delta,$$

(7)

where $\delta$ is a small parameter.

For $\nu(x) =$const, Eq.(6) takes the form of the ordinary diffusion equation which describes the Wiener process and is characterized by the normal diffusion. An interesting case is the power-law form of the jumping frequency:

$$\nu(x) = \gamma |x|^{-\theta},$$

(8)

where $\theta$ is a constant parameter and $\gamma$ ensures proper units; in the following we assume $\gamma = 1$. The diffusion coefficient in the form (8) has been used to describe the diffusion on fractal objects [11], the transport of fast electrons in a hot plasma [26], turbulent two-particle diffusion [27]. The FPE with that diffusion coefficient has been analyzed from the point of view of the first passage time in Ref. [28]. The FPE (6) with $\nu$ given by Eq.(8) can be solved exactly [29]:

$$p(x, t) = C_\theta \frac{|x|^{\theta} \exp(-\frac{2|x|^{2+\theta}}{\sigma^2(2+\theta)t})}{(\sigma^2t/2)^{\frac{\theta}{2+\theta}}},$$

(9)

where $C_\theta = \frac{1}{2\Gamma(1+\frac{\theta}{2+\theta})} |2 + \theta|^{\theta/2+\theta}$. Some values of $\theta$ must be excluded. Since for $\theta < -3$ the Eq.(2) has the stationary solution (in the form of the delta function), the approximation by the FPE is not valid; the diffusion in this case is local. Moreover, for $\theta \in (-2, -1)$ the distribution (9) is not normalized. Therefore, we have either $\theta \in (-3, -2)$ or $\theta > -1$.

Fig.1 presents the comparison of the FPE solutions $p(x, t)$ calculated according to the Eq.(9), with the solutions of the master equation (2), obtained from Monte Carlo simulations of the stochastic trajectories. The agreement is good already for $\sigma = 0.1$.

The diffusion properties of the system follow directly from the solution (9). The mean squared displacement for $\theta > -1$ is given by the power-law function of time by the following formula

$$\langle x^2(t) \rangle = \frac{\Gamma(\frac{3+\theta}{2+\theta})}{\Gamma(\frac{1+\theta}{2+\theta})} \left[ \frac{\sigma^2}{2} (2 + \theta)^2 t \right]^{\frac{\theta}{2+\theta}},$$

(10)
and the diffusion coefficient is:
\[ D = \frac{\langle x^2 \rangle}{2t} \sim t^{-\theta/(\mu+\theta)} \quad (t \to \infty). \] (11)

Then we can distinguish three cases. For \( \theta \in (-1, 0) \), \( D \) is infinite and the superdiffusion emerges. For \( \theta > 0 \), \( D = 0 \) and we get the subdiffusion. Finally, the normal diffusion takes place for \( \theta = 0 \). Therefore, the jumping process involves all kinds of diffusion.

IV. THE FRACTIONAL DIFFUSION EQUATION

Let us now assume the distribution \( Q \) in the form of the symmetric Lévy distribution, defined by its Fourier transform:
\[ \tilde{Q}(k) = \exp(-K^\mu|k|^\mu) \quad (K > 0), \] (12)
where \( 1 < \mu < 2 \). In contrast to the Gaussian distribution, \( Q(x) \), corresponding to the Eq.(12), has algebraic tails, \( Q(x) \sim |x|^{-1-\mu} (|x| \to \infty) \), which makes long jumps very probable. In the diffusion limit \( k \to 0 \), the Eq.(12) is given by \( \tilde{Q}(k) \approx 1 - K^\mu|k|^\mu \). We wish to derive an equation which could serve as an approximation to the master equation in the diffusion limit. We take the Fourier transform of the Eq.(2) and insert \( \tilde{Q}(k) \) in the above form, that yields
\[ \frac{\partial \tilde{p}(k, t)}{\partial t} = -K^\mu|k|^\mu \mathcal{F}[\nu(x)p(x, t)]. \] (13)
The Eq.(13) can be formally inverted by expressing the invert Fourier transform by a fractional Weyl-Riesz operator:
\[ \mathcal{F}^{-1}(-|k|\nu) = \frac{\partial^\mu}{\partial |x|^\mu}. \] (30)
The resulting equation is the following
\[ \frac{\partial p(x, t)}{\partial t} = K^\mu \frac{\partial^\mu[\nu(x)p(x, t)]}{\partial |x|^\mu}. \] (14)

Our aim is to solve the Eq.(14) for \( \nu \) given by Eq.(8), where \( \theta > -1 \), with the initial condition \( p(x, 0) = \delta(x) \). Obviously, the presence of the \( x \)-dependent term under the fractional derivative poses the main difficulty.

A. The case \( \theta = 0 \)

The Eq.(13) for this case can be easily solved:
\[ \tilde{p}(k, t) = \exp(-K^\mu t|k|^\mu). \] (15)
The above expression is the characteristic function of the Lévy process and its inversion produces the symmetric Lévy distribution. To handle those distributions, it is convenient to apply Fox functions. In fact, the Lévy distribution in its most general form can be expressed as a Fox function. That formalism is useful in describing stable processes because it reflects scaling properties of the underlying phenomena. The definition and some properties of the Fox functions are presented in Appendix.

The solution of the Eq.(14) can then be expressed in the following way:
\[ p(x, t) = \frac{1}{\mu|x|} H^{1,1}_{2,2} \begin{pmatrix} \frac{|x|}{(K^\mu t)^{1/\mu}} & (1, 1/\mu), (1, 1/2) \\ (1, 1), (1, 1/2) \end{pmatrix}. \] (16)
The asymptotics \( |x| \to \infty \) results from the expansion \( p(x, t) \sim t/|x|^{1+\mu} \) and implies that the variance, as well as all fractional moments of the order of \( \mu \) or higher, diverge.

B. The general case

The Eq.(14), as an approximation to the master equation in the diffusion limit, has been obtained from the Eq.(2) by dropping all terms of the order higher than \( k^\mu \) in the expansion of \( Q(k) \). Consequently, solving the Eq.(14), we
can neglect those terms without introducing any additional idealization. We look for a solution in the form of the Fox function

\[ p(x, t) = NaH_{2, 2}^{1, 1}[a|x| \begin{pmatrix} (a_1, A_1), (a_2, A_2) \\ (b_1, B_1), (b_2, B_2) \end{pmatrix}, \] (17)

similar to that for the case \( \theta = 0 \), where \( a = a(t) \) and \( N \) is the normalization constant. We want to determine the coefficients \( a_i, A_i, b_i \) and \( B_i \), as well as the function \( a(t) \), by expansion of both sides of the Eq. (13) in consecutive fractional powers of \( k \) and neglect the terms which are small for \( k \to 0 \).

The function \( p_0 = x^{-\theta}p(x, t) \) on the right hand side of Eq. (13) can be expressed as the Fox function due to the multiplication rule (A5). Its Fourier transform, given by the Eq. (A6), expanded according to the formula (A3), takes the form

\[ \tilde{p}_0(k, t) = 2\pi(h^{(\theta)}_0 + h^{(\theta)}_1 k' + O(k^\mu)), \] (18)

where \( k' = Kk/a \). Deriving explicitly consecutive terms and utilizing simple properties of the gamma function, one can find conditions for the coefficients. First, to get the term of order \( k^0 \), we introduced the condition \( 1 - a_1 = A_1(1 - \theta) \). \( h^{(\theta)}_1 = 0 \) identically because the gamma function in the denominator is infinite. Similarly, for the function \( p(x, t) \) we have:

\[ \tilde{p}(k, t) = 2\pi(h_0 + h_1 k' - \theta + h_1 k' + h_2 k'^2 + h_3 k'^3 + O(k^2 + \theta)), \] (19)

where we imposed the condition \( 2 - a_1 = A_1(1 + \mu) \) to get the exponent \( \mu \) for the term \( k^\mu \). Then the above conditions determine the first two coefficients of the Fox function: \( a_1 \) and \( A_1 \). We need also \( h_{-\theta} = 0 \); this requirement can be satisfied if the argument of one of the gamma functions in the denominator assumes an integer and non-positive value. The condition for that is \( 1 - b_2 - B_2(1 - \theta) = 0 \) (alternatively, we could impose a similar condition for \( a_2 \) and \( A_2 \)). The same procedure allows us to satisfy the requirement \( b_2 = 0 \) and to determine \( B_2 = 1/(2 + \theta) \). Finally, since the coefficients \( a_2 \) and \( A_2 \) enter the above expressions in a similar way as \( b_2 \) and \( B_2 \), we want to preserve this symmetry, present for the case \( \theta = 0 \), and assume \( a_2 = b_2 \) and \( A_2 = B_2 \).

Then we insert the expansions Eqs. (18) and (19) into the Eq. (13) and separate the time-dependent term: \( (a/K)^{-\theta - 1}x/K = -\lambda \), where \( \lambda \) is a constant which scales the time and then it is not essential. We assume \( \lambda = 1/(\mu + \theta) \). By taking into account the initial condition for the Eq. (14), one can write down the solution for the function \( a(t) \) as:

\[ a = Kt^{-\mu/\theta} \] (20)

and then reduce the problem to a simple equation:

\[ -\frac{\mu}{\mu + \theta}h_{\mu} = h^{(\theta)}_0, \] (21)

where \( h^{(\theta)}_0 = a^{-\theta}h^{(\theta)}_0 \).

After implementing the coefficients we have evaluated, we obtain for the term \( h_{\mu} \) the following expression:

\[ h_{\mu} = \pi^{-2}(\mu + \theta)\Gamma(1)\Gamma(b_1 + B_1(1 + \mu))\cos(\mu\pi/2)\sin(\mu\pi/2). \] Unfortunately, the term \( h^{(\theta)}_0 \) cannot be obtained directly from the series expansion because the undetermined expression emerges. Instead, we proceed as follows. \( h^{(\theta)}_0 \) can be expressed as:

\[ h^{(\theta)}_0 = (2\pi)^{-1}a^{-\theta}\tilde{p}_0(k = 0) = (2\pi)^{-1}\int_0^\infty z^{-1}W(z)dz, \] where we obtained the function

\[ W(z) = H_{2, 2}^{1, 1}\left[ \begin{array}{c} \mu \pm \theta \\ \mu \pm \theta \end{array} \right] \begin{pmatrix} (1, 1, 1, 1) \\ (b_1 + B_1(1 - \theta), B_1, 1, 1) \end{pmatrix} \] (22)

by applying the relation (A5). The required term can now be easily evaluated as the Mellin transform \( M(W)(s) = \int_0^\infty W(z)z^{s-1}dz \):

\[ h^{(\theta)}_0 = \pi^{-1}M(W)(s = 0) = \pi^{-1}\chi(0) = \pi^{-1}\lim_{\epsilon \to 0} \chi(\epsilon) = \frac{1}{\pi} \frac{\mu + \theta}{2} \Gamma(b_1 + B_1(1 - \theta)). \] (23)
Inserting the expressions for $h_{\mu}$ and $h_{0}^{(\theta)\prime}$ to the Eq.(21), yields the relation between $b_{1}$ and $B_{1}$. Then, finally, the solution of the Eq. (24) reads

$$p(x,t) = NaH_{2,2}^{1,1}\left[ a|x| \left( 1 - \frac{1-\theta}{\mu+\theta}, 1 - \frac{1-\theta}{2+\theta} \right) \right]$$

(24)

with the condition

$$\frac{\mu}{\pi} \frac{\Gamma(-\mu)\Gamma(b_{1} + B_{1}(1 + \mu))}{\Gamma(b_{1} + B_{1})} \sin \left( \frac{\mu + \theta}{2 + \theta} \pi \right) \cos (\mu\pi/2) + \Gamma(b_{1} + B_{1}(1 - \theta)) = 0$$

(25)

which follows directly from the Eq.(21). The general theory of the Fox functions implies that $p(x,t)$ is the analytic function for all $x \neq 0$ if $B_{1} > 1/(\mu + \theta)$. Apart from that very weak condition, one parameter is arbitrary. We will return to this issue in Subsec. D. The normalization factor $N$ can be determined in a simple way by using the Mellin transform: $N = |2\chi(-1)|^{-1}$, that yields

$$N = -\frac{\pi}{2} \left[ \Gamma(b_{1} + B_{1})\Gamma \left( -\frac{\theta}{\mu + \theta} \right) \sin \left( \frac{\theta}{2 + \theta} \pi \right) \right]^{-1}. $$

(26)

In the $k$-space, the solution (24) is of the form

$$\tilde{p}(k,t) = 1 - \sigma^{\mu} k^{\mu} + \cdots \approx \exp(-\sigma^{\mu} k^{\mu})$$

(27)

where

$$\sigma^{\mu} = \frac{K^{-\mu} (\mu + \theta)^{2} \Gamma(b_{1} + B_{1}(1 - \theta))\Gamma(-\frac{\theta}{\mu + \theta})}{\mu \Gamma(b_{1} + B_{1})\Gamma(-\frac{\theta}{\mu + \theta})} |1 + \frac{\theta}{\mu + \theta}| \pi^{\mu/\pi}$$

(28)

and we neglected the terms of the order $k^{2\mu+\theta} (\theta < 0)$ or $k^{2\mu} (\theta > 0)$. The formula (24) means that within the scope of validity of our approximation the solution (24) is equivalent to the Lévy stable distribution.

The Fox functions can be evaluated from series expansions. Since they are poorly convergent, one needs the series expansions of validity of our approximation the solution (24) is equivalent to the Lévy stable distribution.

The characteristic feature of the Lévy distributions is the divergence of moments. In particular, the mean squared displacement is infinite for any time and then the transport phenomena require a modified formalism for the diffusion. One possibility is to substitute the variance by a moment of the order $\delta < \mu$.

In order to evaluate the moments of the distribution (24), we utilize properties of the Mellin transforms. A simple calculation yields:

$$\langle |x|^{\delta} \rangle = 2N \int_{0}^{\infty} x^{\delta} p(x,t)dx = 2Na^{-\delta} \chi(-\delta - 1) = -\frac{2NK^{-\mu}}{\pi} \Gamma(b_{1} + B_{1}(1 + \delta)) \Gamma \left( \frac{-\theta + \delta}{\mu + \theta} \right) \sin \left( \frac{\theta + \delta}{2 + \theta} \pi \right) \pi^{\delta/(\mu + \theta)}. $$

(31)

C. Diffusion

A characteristic feature of the Lévy distributions is the divergence of moments. In particular, the mean squared displacement is infinite for any time and then the transport phenomena require a modified formalism for the diffusion. One possibility is to substitute the variance by a moment of the order $\delta < \mu$.
Applying the above expression, one can compare individual cases in respect to the speed of transport. However, as long as the parameter $\delta$ is arbitrary, such formalism seems to be incomplete. Can it be fixed in some way? Clearly, the value $\delta = \mu$ is distinguished. Since that moment is divergent, we consider $\delta = \mu - \epsilon$, where $0 < \epsilon \ll \theta$ and then the case $\theta = 0$ is excluded. In the limit $\epsilon \to 0$, the gamma function can be expanded and Eq. (31) takes the form

$$\left< |x|^{\mu-\epsilon} \right> \approx \frac{2NK^{-\mu}}{\pi \epsilon} (\mu + \theta) \Gamma(b_1 + B_1(1 + \mu)) \sin \left( \frac{\mu + \theta}{2 + \theta} \pi \right) t^{\mu/(\mu+\theta)}.$$

Let us now define the fractional diffusion coefficient $D^{(\mu)}(t)$:

$$D^{(\mu)}(\mu) \equiv \frac{1}{\Gamma(1 + \mu)} \lim_{\epsilon \to 0^+} \frac{1}{t} \epsilon \left< |x|^{\mu-\epsilon} \right> \quad (t \to \infty),$$

where $1 < \mu < 2$. According to (32), the limit is finite and $D^{(\mu)} \sim t^{-\theta/(\mu+\theta)}$.

The interpretation of the above result is straightforward. If $\theta < 0$ the coefficient $D^{(\mu)}$ rises with time to infinity and we have the "superdiffusion". Conversely, for $\theta > 0$ there is the "subdiffusion". Therefore we have obtained formally the same result as for FPE, Eq. (11), except the variance has been substituted by the fractional moment. The common conclusion, drawn both from the Gaussian and the fractional case, is that the sign of $\theta$ decides which kind of diffusion the system reveals.

D. Numerical examples

In this Subsection we evaluate the probability distributions for specific cases. They are compared with the solutions of the master equation (2), obtained by the Monte Carlo simulations of stochastic trajectories of the jumping process. The Lévy-distributed jump-size density has been generated by using the algorithm from Ref. [33].

It follows from the expansion (29) that $p(x, t) \sim |x|^{b_1/B_1}$ for small $|x|$. Therefore, the ratio of the coefficients $b_1$ and $B_1$ determines the shape of the distribution $p(x, t)$ there. Since our approximation is suited for large $|x|$, this ratio remains undetermined. The analysis of the master equation indicates that its solutions exhibit the power law dependence of the form $|x|^\theta$ for small $|x|$. We utilize this property and assume the relation $b_1 = \theta B_1$ in the present Subsection; $B_1$ follows from the numerical solving of the Eq. (25). The Fox functions have been evaluated from the series (29) and (30) with sufficient precision for all $|x| \in (0, \infty)$.

The solutions of the fractional equation (24), for a negative and a positive value of $\theta$, are presented in Fig. 2. They correspond to the superdiffusion and subdiffusion, respectively, and display very different shapes. All the distributions

![FIG. 2: The solutions of Eq. (2) for $\theta = -0.2$ (dots) and $\theta = 1$ (dashed line). The corresponding solutions (24) are marked by solid lines.](image-url)
with $\theta > 0$ rise at small $|x|$ and display a maximum, whereas those for $\theta < 0$ fall monotonically. The comparison with the solutions of the master equation is also presented in Fig. 2. The curves which represent both equations have similar shapes and their tails coincide.

Fig. 3 presents the comparison of the Fourier transforms for the solution of both equations; the same cases as in Fig. 2 are shown: $\theta = -0.2$ and $\theta = 1$. For that purpose, the master equation has been solved for all $x$ up to very large values, then the numerical integration with the cosine function has been performed. In the case of the master equation solutions, the figure reveals substantial deviations from the straight lines – which represent the shape of the Lévy distribution – for large $k$: for $\theta < 0$ that distribution stabilizes with $k$, whereas it falls rapidly to zero for $\theta > 0$. At small $k$ both solutions coincide with those of the fractional equation.

![Graph](image)

**FIG. 3**: Logarithm of the Fourier transform from the solution of Eq. (2) for $\theta = -0.2$ (dots) and $\theta = 1$ (dashed line). The same quantity for the solution (24) which corresponds to both cases, calculated according to Eq. (27), is marked by solid lines; $\xi = k/a$.

**V. DISCUSSION**

The jumping process we have discussed in this paper is Markovian because the waiting time probability distribution has the exponential form. However, since the jumping frequency depends on the process value, the system possesses some properties which are usually attributed to non-Markovian processes. In particular, we have demonstrated the presence of the anomalous diffusion.

If the step-size is small, the jumping process can be regarded as continuous and described in terms of the FPE. That approximation has been accomplished by applying the Kramers-Moyal expansion, on the assumption that all moments of the step-size distribution are finite. It has been demonstrated –by solving the FPE exactly for the power-law frequency $\nu(x)$– that both normal and anomalous diffusion emerges.

On the other hand, we have considered the diffusion limit of the master equation for the step-size distribution which is stable and has divergent moments. The resulting equation is fractional and contains a variable coefficient, in contrast to usually studied equations which govern Lévy processes. We have demonstrated that in the diffusion limit the Eq. (14) is satisfied by the Fox function if $\nu$ depends algebraically on the position. The coefficients of the Fox function have been derived by inserting it to the equation and by comparing the terms. However, one parameter must remain undetermined if only the diffusion limit is considered because it is responsible for the behaviour of the solution at small $|x|$.

The solution, since it is expressed as the Fox function, depends on time in the scaling form. Due to that property, simple conclusions about the transport can be drawn. The fractional moments are given by complicated expressions but the time-dependence factorizes and it obeys a simple power law. We have defined the fractional diffusion coefficient
\(D^{(\nu)}\) which allows us to establish a correspondence with the standard description in terms of the FPE. Both approaches predict the diffusion coefficient, either \(D\) or \(D^{(\nu)}\), in the form \(\sim t^{-\theta/(\mu+\theta)}\), i.e. the subdiffusion (\(\theta > 0\)) and the superdiffusion (\(\theta < 0\)). We would like to emphasize that the above conclusions are independent of a specific choice of the free parameter in the solution.

An independent numerical analysis of the master equation (2) provides additional information about the jumping process. We can learn that it is not the Lévy stable process: substantial deviations from the Lévy distribution at large \(k\) are clearly visible. They become meaningless in the diffusion limit.

The comparison of the solutions of both diffusion equations we have discussed in this paper with the results of the numerical analysis of the master equation shows a reasonable agreement not only in the asymptotic limit of large \(|x|\). Since the expressions (9) and (24) are relatively simple, compared to the Eq.(3), both diffusion equations could serve as convenient approximations to the master equation (2).

**APPENDIX**

The Fox function \(\mathcal{F}_{pq}^{mn}\) (for a review of the most important properties see [14, 31]) is defined as an inverse Mellin transform in the following way:

\[
\mathcal{F}_{pq}^{mn}(z) = \int_L \chi(s) z^s ds,
\]

where

\[
\chi(s) = \frac{\prod_{i=1}^m (1 - b_i + A_j s) \prod_{j=1}^n \Gamma(1 - a_j + A_j s)}{\prod_{j=m+1}^p \Gamma(1 - b_j + B_j s) \prod_{j=n+1}^p \Gamma(a_j - A_j s)}. \tag{A2}
\]

Coefficients \(A_i\) and \(B_i\) are positive. The contour \(L\) is a straight line parallel to the imaginary axis which separates the poles of both gamma functions in Eq.(A2). If those poles are simple, the Fox function can be expressed in the form of the following series

\[
\mathcal{F}_{pq}^{mn}(z) = \sum_{h=1}^\infty \sum_{\nu=0}^\infty \prod_{i=1}^m \Gamma(1 - b_i + B_i s) \prod_{j=1}^n \Gamma(1 - a_j + A_j s) \frac{(1 - a_j + A_j \frac{b_i + \nu}{B_i})}{\nu! B_i} \frac{(-1)^\nu z^{(b_i + \nu)/B_i}}{\nu! B_i}.
\]

A similar expansion can be obtained for \(z \to \infty\) by using the property:

\[
\mathcal{F}_{pq}^{mn}(z) = \int_0^\infty \chi(s) z^s ds,
\]

Another useful property is the multiplication rule:

\[
\mathcal{F}_{pq}^{mn}(z) = \mathcal{F}_{pq}^{mn}(1) \mathcal{F}_{pq}^{mn}(1)
\]

The Fourier cosine transform of the Fox function yields:

\[
\int_0^\infty \mathcal{F}_{pq}^{mn}(x) \cos(kx) dx = \frac{1}{k} H_{q+1,p+2}^{n+1,m} \left[ (1 - b_q, B_q), (1, 1/2) \right] \left[ (1, 1), (1 - a_p, A_p), (1, 1/2) \right].
\]

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