THE RANDOM WALK’S GUIDE TO ANOMALOUS DIFFUSION: A FRACTIONAL DYNAMICS APPROACH

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To the memory of my father, H.W. Metzler

Contents

1. Prologue: the scope, and why bother at all 1
   1.1. What can fractional equations do, what can they do better, and why should one care at all? 1
   1.2. What is the scope of this report? 5

2. Introduction 5
   2.1. Anomalous dynamics in complex systems 5
   2.2. Historical remarks 7
   2.3. Anomalous diffusion: experiments and models 9

3. From continuous time random walk to fractional diffusion equations 13
   3.1. Revisiting the realm of Brownian motion 14
   3.2. The continuous time random walk model 15
   3.3. Back to Brownian motion 17
   3.4. Long rests: a fractional diffusion equation describing subdiffusion 18
   3.5. Long jumps: Lévy flights 25
   3.6. The competition between long rests and long jumps 29
   3.7. What’s the course, helmsman? 30

4. Fractional diffusion–advection equations 31
   4.1. The Galilei invariant fractional diffusion–advection equation 32
   4.2. The Galilei variant fractional diffusion–advection equation 33
   4.3. Alternative approaches for Lévy flights 35

5. The fractional Fokker–Planck equation: anomalous diffusion in an external force field 36
   5.1. The Fokker–Planck equation 37
   5.2. The fractional Fokker–Planck equation 37
   5.3. Separation of variables and the fractional Ornstein–Uhlenbeck process 39
   5.4. The connection between the fractional solution and its Brownian counterpart 43
   5.5. The fractional analogue of Kramers escape theory from a potential well 45
   5.6. The derivation of the fractional Fokker–Planck equation 46
   5.7. A fractional Fokker–Planck equation for Lévy flights 51
   5.8. A generalised Kramers–Moyal expansion 52

6. From the Langevin equation to fractional diffusion: dispersive transport close to thermal equilibrium 53
   6.1. Langevin dynamics and the three stages to subdiffusion 54

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6.2. The fractional Klein–Kramers equation
and related transport equations  
7. Conclusions  
Note added in proof  
Acknowledgements  
Appendix A. Fractional differentiation and integration  
A.1. The Riemann–Liouville fractional operator  
A.2. The Riesz/Weyl fractional operator  
A.3. Differentiable functions and an equivalent definition  
A.4. Examples  
A.5. The singular nature of the fractional operator  
Appendix B. Special functions: Mittag–Leffler and Fox functions  
B.1. The Mittag–Leffler function  
B.2. The Fox function  
Appendix C. Some remarks on Lévy distributions and their exact representation in terms of Fox functions  
Appendix D. Abbreviations used  
References

Abstract

Fractional kinetic equations of the diffusion, diffusion–advection, and Fokker–Planck type are presented as a useful approach for the description of transport dynamics in complex systems which are governed by anomalous diffusion and non-exponential relaxation patterns. These fractional equations are derived asymptotically from basic random walk models, and from a generalised master equation. Several physical consequences are discussed which are relevant to dynamical processes in complex systems. Methods of solution are introduced and for some special cases exact solutions are calculated. This report demonstrates that fractional equations have come of age as a complementary tool in the description of anomalous transport processes. © 2000 Elsevier Science B.V. All rights reserved.

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Keywords: Anomalous diffusion; Fractional diffusion equation; Fractional Fokker–Planck equation; Anomalous relaxation; Mittag–Leffler relaxation; Dynamics in complex systems
For the beginning is assuredly
the end – since we know nothing, pure
and simple, beyond
our own complexities
William Carlos Williams, Patterson

1. Prologue: the scope, and why bother at all

1.1. What can fractional equations do, what can they do better, and why should one care at all?

Before we start off with the Introduction, we would like to address some points which we believe
to strike many colleagues who are not familiar with the topic.

The universality: The detailed structure of the propagator \( W(r, t) \), i.e., the probability density
function (pdf) for the initial condition \( \lim_{t \to 0^+} W(r, t) = \delta(r) \), depends, in general, on the special
shape of the underlying geometry. However, the interesting part of the propagator has the
asymptotic behaviour \( \log W(r, t) \sim -c \zeta^n \) where \( \zeta \equiv r/t^{1/2} \gg 1 \) which is expected to be universal.
Here, \( u = 1/(1 - \alpha/2) \) with the anomalous diffusion exponent \( \alpha \) defined below. The fractional
equations we consider in the following are universal in this respect as we do not consider any form
of quenched disorder. Our results for anomalous diffusion are equivalent to
findings from random
walk models on an isotropic and homogeneous support.

The non-universality: In contrast to Gaussian diffusion, fractional diffusion is non-universal in
that it involves a parameter \( \alpha \) which is the order of the fractional derivative. Obviously, nature
often violates the Gaussian universality mirrored in experimental results which do not follow the
Gaussian predictions. Fractional diffusion equations account for the typical
anomalous
features
which are observed in many systems.

The advantage to random walk models: Within the fractional approach it is possible to include
external fields in a straightforward manner. Also the consideration of transport in the phase space
spanned by both position and velocity coordinate is possible within the same approach. Moreover,
the calculation of boundary value problems is analogous to the procedure for the corresponding
standard equations.

The comparison to other approaches: The fractional approach is in some sense equivalent
to the generalised master equation approach. The advantage of the fractional model again lies in
the straightforward way of including external force terms and of calculating boundary value
problems. Conversely, generalised Langevin equations lead to a different description as they
correspond to Fokker–Planck equations which are local in time and which contain time-dependent
coefficients. In most cases of Brownian transport, the deterministic Fokker–Planck equation is
employed for the description of stochastic dynamics in external fields. In analogy, we promote
to use the fractional Fokker–Planck equation for situations where anomalous diffusion underlies
the system.

The mathematical advantage: A very convenient issue is that standard techniques for solving
partial differential equations or for calculating related transport moments also apply to fractional
equations which is demonstrated in the text.
The relation between the fractional solution and its Brownian counterpart: There exists a transformation which maps the Brownian solution onto the corresponding fractional solution, an interesting relation which is useful for both analytic and numerical analysis.

It is a simple approach: The appearance of fractional equations is very appealing due to their proximity to the analogous standard equations. It has been demonstrated recently that the fractional Fokker–Planck equation can be derived from a Langevin equation with Gaussian white noise for systems where trapping occurs. This offers some insight into the physical mechanisms leading to fractional kinetics.

Fractional kinetic equations are not just another way of presenting old stories. We believe that they are a powerful framework which is of use for many systems. By relating our and others’ work in that field and putting it in some more general context, the present report may be the basis for some active research on complex dynamics using a tool which is as old and new as classical calculus.

1.2. What is the scope of this report?

We do not present a report on anomalous transport theory: Anomalous diffusion is an involved field with intriguing subtleties. Accordingly, parameters and exponents can change in the course of time or when an external force is switched on or off, etc. In our considerations we do not touch on these issues but assume a given anomalous diffusion exponent. The present knowledge on anomalous diffusion is compiled in a number of review articles quoted in the list of references.

The focus is on subdiffusion: The major body of the text is concerned with transport processes which are, in the force-free limit, slower than Brownian diffusion. In a lose sense we also call such processes subdiffusive or dispersive which take place in an external force field and whose force-free limit corresponds to subdiffusion. Occasionally, such processes are called fractional.

Some care be taken with Lévy flights: Lévy flights do not possess a finite mean squared displacement. Their physical significance therefore has been questioned as particles with a finite mass should not execute long jumps instantaneously. We do consider them to some extent in the following as there are special cases whose description in terms of Lévy flights corresponds to physical principles.

Lévy walks: The proper way of considering systems which feature Lévy type jump length distributions is to introduce a finite speed for the test particle, a model referred to as Lévy walk. A first step towards a fractional dynamics formulation has been presented recently.

Since Newton integer order calculus has been used in physical modelling. Newton’s rival Leibniz’ prophesy already claimed in 1695: “Thus it follows that \( d^{1/2}x \) will be equal to \( x \sqrt{2dx}:x \), an apparent paradox, from which one day useful consequences will be drawn” – we believe that the following account represents an affirmative answer to this first statement towards a fractional calculus.

2. Introduction

2.1. Anomalous dynamics in complex systems

Complex systems and the investigation of their structural and dynamical properties have established on the physics agenda. These “structures with variations” [1] are characterised through
(i) a large diversity of elementary units, (ii) strong interactions between the units, or (iii) a non-predictable or anomalous evolution in the course of time [2]. Complex systems and their study play a dominant rôle in exact and life sciences, embracing a richness of systems such as glasses, liquid crystals, polymers, proteins, biopolymers, organisms or even ecosystems. In general, the temporal evolution of, and within, such systems deviates from the corresponding standard laws [3,4]. With the development of higher experimental resolutions, or the combination of different experimental techniques, these deviations have become more prominent, as the accessible larger data windows are more conclusive.

Thus, relaxation in complex systems deviates from the classical exponential Debye pattern [5–10]

\[ \Phi(t) = \Phi_0 \exp(-t/\tau) , \]  

(1)

and can often be described in terms of a Kohlrausch–Williams–Watts stretched exponential law [11–15]

\[ \Phi(t) = \Phi_0 \exp(- (t/\tau)^\alpha) \]  

for \( 0 < \alpha < 1 \), or by an asymptotic power law [9,13,14,16–19]

\[ \Phi(t) = \Phi_0 (1 + t/\tau)^{-n} \]  

(3)

with \( n > 0 \). It is also possible to observe transitions from the stretched exponential pattern (2) to the power-law behaviour (3) [20–22].

Similarly, diffusion processes in various complex systems usually no longer follow Gaussian statistics, and thus Fick’s second law fails to describe the related transport behaviour. Especially, one observes deviations from the linear time dependence of the mean squared displacement

\[ \langle x^2(t) \rangle \sim K_1 t \]  

(4)

which is characteristic of Brownian motion, and as such a direct consequence of the central limit theorem and the Markovian nature of the underlying stochastic process [23–39]. Instead, anomalous diffusion is found in a wide diversity of systems, its hallmark being the non-linear growth of the mean squared displacement in the course of time. In this report, we concentrate on the power-law pattern

\[ \langle x^2(t) \rangle \sim K_2 t^\alpha \]  

(5)

which is ubiquitous to a diverse number of systems [14,38–55]. There exists a variety of other patterns such as a logarithmic time dependence which we do not touch upon here. The anomalous diffusion behaviour manifest in Eq. (5) is intimately connected with the breakdown of the central limit theorem, caused by either broad distributions or long-range correlations. Instead, anomalous diffusion rests on the validity of the Lévy–Gnedenko generalised central limit theorem for such situations where not all moments of the underlying elementary transport events exist [38,42,47,56–58].

\[ \text{1 Actually, the notion of diverging moments goes back to the formulation of the “St. Petersburg paradox” by Nicolaas Bernoulli and its analysis by Daniel Bernoulli in 1724 [38,49].} \]
Fig. 1. Different domains of anomalous diffusion, defined through the mean squared displacement, Eq. (5), parametrised by the anomalous diffusion exponent $\alpha$: (a) subdiffusion for $0 < \alpha < 1$, (b) superdiffusion for $\alpha > 1$. On the threshold between sub- and superdiffusion is the normal Brownian diffusion located. Another special case is ballistic motion ($\alpha = 2$).

...propagators and a possibly non-Markovian time evolution of the system, the latter being a typical manifestation of non-local temporal phenomena encountered in a broad range of systems [1,3,4,59–63]. Note that the generalised diffusion coefficient $K_\alpha$ in Eq. (5) has the dimension $[K_\alpha] = \text{cm}^2 s^{-\alpha}$. According to the value of the anomalous diffusion exponent $\alpha$, defined in Eq. (5), one usually distinguishes several domains of anomalous transport, as sketched in Fig. 1. In what follows, the main emphasis will be laid on the description of subdiffusive phenomena, which correspond to $0 < \alpha < 1$.

2.2. Historical remarks

The stochastic formulation of transport phenomena in terms of a random walk process, as well as the description through the deterministic diffusion equation are the two fundamental concepts in the theory of both normal and anomalous diffusion. Indeed, the history of this dual description basing on erratic motion and on a differential equation for the probability density function is quite interesting and much worth a short digression.

Thus, small flickering of coal dust particles on the surface of alcohol was observed by the Dutch physician Jan Ingenhousz as early as in 1785. In 1827, the Scottish botanist Robert Brown [67] observed similar irregular movement of small pollen grain under a microscope. At about the same time, in 1822, Joseph Fourier came up with the heat conduction equation, on the basis of which...
Fig. 2. Recorded random walk trajectories by Jean Baptiste Perrin [72]. Left part: three designs obtained by tracing a small grain of putty (mastix, used for varnish) at intervals of 30 s. One of the patterns contains 50 single points. Right part: the starting point of each motion event is shifted to the origin. The figure illustrates the pdf of the travelled distance \( r \) to be in the interval \( (r, r + dr) \), according to \( (2\pi \xi^2)^{-1} \exp\left(-r^2/(2\xi^2)\right)2\pi r \, dr \), in two dimensions, with the length variance \( \xi^2 \). These figures constitute part of the measurement of Perrin, Dabrowski and Chaudesaigues leading to the determination of the Avogadro number. The result given by Perrin is \( 70.5 \times 10^{22} \). The remarkable œuvre of Perrin discusses all possibilities of obtaining the Avogadro number known at that time. Concerning the trajectories displayed in the left part of this figure, Perrin makes an interesting statement: “Si, en effet, on faisait des points de seconde en seconde, chacun de ces segments rectilignes se trouverait remplacé par un contour polygonal de 30 côtés relativement aussi compliqué que le dessin ici reproduit, et ainsi de suite”. [If, veritably, one took the position from second to second, each of these rectilinear segments would be replaced by a polygonal contour of 30 edges, each itself being as complicated as the reproduced design, and so forth.] This already anticipates Lévy’s cognisance of the self-similar nature, see footnote 9, as well as of the non-differentiability recognised by N. Wiener.

A. Fick set up the diffusion equation in 1855 [68]. Subsequently, the detailed experiments by Gouy proved the kinetic theory explanation given by C. Weiner in 1863. After attempts of finding a stochastic footing like the collision model by von Nägeli and John William Strutt, Lord Rayleigh’s results, it was Albert Einstein who, in 1905, unified the two approaches in his treatises on the Brownian motion, a name coined by Einstein although he reportedly did not have access to Brown’s original work. Note that a similar description of diffusion was presented by the French...
Historically, often referred to as the Loschmidt number \[28\]. An important application of Einstein’s results was the independent measurement of the Avogadro number\(^7\) by Jean Baptiste Perrin, A. Westgren and Eugen Kappler [26,28,71–74], to a rather high accuracy. Some results of Perrin are shown in Fig. 2 and they are part of the work which won him the Nobel Prize in 1926. The random walk which can be experimentally observed, represents therefore a link between the microscopic dynamics of small atoms bombarding a larger particle in suspension, and macroscopic observables like the diffusion coefficient, or the Avogadro number. In Fig. 3, we reproduced data obtained by Kappler with his high-accuracy set-up using an optical detection method (a detailed explanation is given in the figure caption). Einstein’s ideas also set the scene for Langevin’s treatment [37,64] of Brownian motion with the assumption of an external erratic force, and the Fokker–Planck [78,79], Smoluchowski\(^8\) and Klein–Kramers [81,82] theories which culminated in the treatises of Ornstein and Uhlenbeck, Chandrasekhar and others, and later in the works of Elliott Montroll, and collaborators [24,83–86].

The mathematical treatment of Brownian motion is mainly due to Norbert Wiener who proved that the trajectory of a Brownian particle is (almost) everywhere continuous but nowhere differentiable \[87\]. This observation is related to the self-affine nature of the diffusion process whose resulting spatial trajectory is self–similar\(^9\) [27,38,88–93]. Further important mathematical contributions attribute to J. L. Doob, Mark Kac, W. Feller, and others.

2.3. Anomalous diffusion: experiments and models

Anomalous diffusion has been known since Richardson’s treatise on turbulent diffusion in 1926 \[94\]. Within transport theory it has been studied since the late 1960s. In particular, its theoretical investigation was instigated by Scher and Montroll in their description of dispersive transport in amorphous semiconductors, a system where the traditional methods proved to fail. The predictions of their continuous time random walk approach were very distinct from its Brownian counterpart and were shown to provide explanations for a variety of physical quantities and phenomena in numerous experimental realisations \[95–99\]. Important contributions are also due to Weiss [39] and Shlesinger [100]. Besides the random walk description, generalisations of the diffusion equation were developed which account for the anomalous transport statistics.

Today, the list of systems displaying anomalous dynamical behaviour is quite extensive \[14,40–55,101,102\] and hosts, among others, the following systems in the subdiffusive régime:

\(^7\) Historically, often referred to as the Loschmidt number \[28\].
\(^8\) The monovariate Fokker–Planck equation discussed here is often referred to as Smoluchowski equation \[80\].
\(^9\) “Le processus stochastique, que nous appellerons mouvement brownien linéaire, est une schématisation, qui représente bien les propriétés du mouvement brownien réel observables à une échelle assez petite, mais non infiniment petite, et qui suppose que les mêmes propriétés existent à n’importe quelle échelle”. (Paul Lévy [27]) [The stochastic process which we will call linear Brownian motion is a schematic representation which describes well the properties of real Brownian motion, observable on a small enough, but not infinitely small scale, and which assumes that the same properties exist on whatever scale.]
charge carrier transport in amorphous semiconductors [95–99, 101, 103, 104], nuclear magnetic resonance (NMR) diffusometry in percolative [105, 106], and porous systems [107, 108]. Rouse or reptation dynamics in polymeric systems [109–115], transport on fractal geometries [116, 117], the
diffusion of a scalar tracer in an array of convection rolls [118,119], or the dynamics of a bead in a polymeric network [120,121]. Superdiffusion or Lévy statistics are observed in special domains of rotating flows [122], in collective slip diffusion on solid surfaces [123], in layered velocity fields [124,125], in Richardson turbulent diffusion [94,126–129], in bulk-surface exchange controlled dynamics in porous glasses [130–132], in the transport in micelle systems and in heterogeneous rocks [133–135], in quantum optics [136,137], single molecule spectroscopy [138,139], in the transport in turbulent plasma [140], bacterial motion [141–145] and even for the flight of an albatross [146].

Anomalous diffusion in the presence or absence of an external velocity or force field has been modelled in numerous ways, including (i) fractional Brownian motion dating back to Benoît Mandelbrot [89–93,147], (ii) generalised diffusion equations [148], (iii) continuous time random walk models [52,95–99,101,102,149–155], (iv) Langevin equations [156–160], (v) generalised Langevin equations [62,161–163], (vi) generalised master equations [164–167], or (vii) generalised thermostatistics [168–172]. For anomalous diffusion, only the approaches (iii) and (v) incorporate the system’s memory and the special form which is to be expected for the pdf, in a consistent way. The disadvantage in the continuous time random walk and the generalised master equation approaches is that there is no straightforward way to incorporate force fields, boundary value problems, or to consider the dynamics in phase space.

The alternative approach to anomalous kinetics which we are going to present is given in terms of fractional equations which appear to be tailored for such kind of problems like the
consideration of external fields and boundary value problems. In the original works [173–175] it was realised that the replacement of the local time derivative in the diffusion equation by a fractional operator accounts for the memory effects which are connected with many complex systems.

Recently, a decade after their introduction, such fractional kinetic equations have attracted much interest. They are presently being extensively studied and recognised as important tools in the description of anomalous transport processes in both absence and presence of external velocity or force fields. Especially in the latter case, their mathematical structure allows for the application of known methods of solution. In the course of this development, a number of works has been published dealing with fractional relaxation equations and fractional rheological models [20,21,176–182], fractional diffusion equations (FDEs) [173–175,183–201], fractional diffusion–advection equations (FDAEs) [202–209], and fractional Fokker–Planck equations (FFPEs) [158–160,193,202,203,210–221]. Thereby, various generalisations to fractional order have been employed, i.e. different fractional operators have been introduced to replace either the time derivative or the occurring spatial derivatives, or both.

In first attempts of generalising the standard diffusion equation for the description of diffusion processes on fractal geometries of dimension \( d_f \), radius-dependent diffusion coefficients were assumed. O’Shaugnessy and Procaccia studied the generalised diffusion equation [148]

\[
\frac{\partial W}{\partial t} = R^{1-d_f} \frac{\partial}{\partial R} R^{d_f-1} K(R) \frac{\partial}{\partial R} W(R,t)
\]

(6)

with the radius-dependent diffusion coefficient \( K(R) = KR^{-\omega} \), and derived the corresponding propagator

\[
W(R,t) = A(\Theta,d_f) (K[2 + \Theta]^2t)^{-d_f/(2+\Theta)} \exp \left( -\frac{R^{2+\Theta}}{K(2+\Theta)^2t} \right).
\]

(7)

Here, \( \Theta = d_f + \sigma - 2 \) is connected to the power-law index \( \sigma \) of the radius-dependent, integrated electrical resistance, \( \Theta(R) \sim R^\sigma \) of the underlying fractal structure the mass density of which scales as \( \propto R^{d_f-3} \) in a 3-dimensional embedding. The mean squared displacement for this process, given by Eq. (5) with \( z = 2/(2+\Theta) \), can readily be inferred. As \( \Theta \) is positive, this result implies subdiffusion [148]. Further investigations indicated that the asymptotic form of the propagator on fractals such as the Sierpinski gasket, is given by the scaling form [222,223]

\[
W(R,t) \sim A(z,\beta) \xi^\beta \exp( - c \xi^u )
\]

(8)

with \( \xi = R/t^{u/2} \), \( u = 1/(1-z/2) \), and \( \beta \) is a system-dependent quantity [224]. Eq. (8) is at variance with the above result, Eq. (7). For subdiffusion, \( 0 < z < 1 \), which prevails on fractal structures, \( u \in (0,2) \) so that expression (8) is often referred to as a stretched Gaussian. Conversely, fractional

---

10 In Ref. [173], note that the author introduces the defining expression for a fractal operator, but does not make explicit use of fractional calculus, neither of the term fractional as such.

11 The capital \( R \) refers to the radius averaging over the fractal support which is necessary in order to obtain the smooth radius dependence.
diffusion equations for transport on fractal structures were shown to comply with the basic properties, like the propagator (8), the returning probability to the origin, the mean squared displacement, and the non-Markovian nature [183,184,225].

For homogeneous, isotropic systems which we are interested in, one knows from random walk models that the propagator behaves asymptotically like

\[ W(r, t) \sim b_0 t^{-\nu d/2} \xi^\beta \exp(-b_1 \xi^\nu) \]  

with \( \xi = d^{1/2} r / t^{1/2} \), \( \nu = 1/(1 - \alpha/2) \), \( \beta = -d(1 - \alpha)/(2 - \alpha) \), and \( b_0 \) and \( b_1 \) are constants depending on \( \alpha \) and the dimension \( d \) [238,239]. Such behaviour is reproduced by the fractional diffusion equations that were anticipated by Balakrishnan [173] and first formulated by Wyss [174], and Schneider and Wyss [175]. In Section 3 we show that the asymptotic behaviour (9) is consistent with the asymptotic expansion of the exact solution of the fractional diffusion equation. Fractional kinetic equations, their foundation and solution form the centre-piece of this report.

In what follows, we present the basic principles and physical properties connected with fractional kinetic equations. We show that fractional equations arise naturally in the diffusion limit of certain random walk schemes. By discussing methods of solution and deriving explicit solutions, we demonstrate the usefulness of the fractional approach. At first, we introduce FDEs basing on the continuous time random walk model where the transport events are subject to broad statistics. Subsequently, FDAEs and FFPEs are presented and they describe the transport in an external velocity or force field. In the final section, a physical derivation on the basis of a Langevin equation with white Gaussian noise is discussed which leads to a fractional Klein–Kramers equation. From the latter, the FFPE is consistently recovered. In the Appendices we have compiled some basic definitions and useful relations which are of relevance in the main text. Thus, we give a primer on fractional calculus and the special functions which emerge when dealing with fractional differential equations, as well as a short introduction to Lévy stable laws. Finally, we list the abbreviations and the notation used. Note that throughout the text we denote the Laplace and Fourier transforms of a function by stating the explicit dependence on the associated variable, e.g., \( W(k, u) = \mathcal{F}\{ \mathcal{L}\{W(x, t); t \rightarrow u\}; x \rightarrow k\} \).

The numerous illustrations spread throughout the text are to visualise the often striking differences in functional behaviour of the normal and anomalous cases, especially the perseverance of the initial condition in the subdiffusive domain.

The discussion in the remainder of this report is restricted to the one-dimensional case, with special emphasis on subdiffusion phenomena. The equations which describe subdiffusion presented in the text can be extended to higher dimensions through a replacement of the derivatives in respect of the position coordinate by corresponding orders of the \( \nabla \) operator. Some remarks on higher-dimensional systems are contained in Ref. [215]. For those equations which describe situations with Lévy distributed jump lengths and which therefore contain a generalised Laplacian, we refer to the definitions in Ref. [226] and the discussions in Refs. [156–158] for the multi-dimensional case.

3. From continuous time random walk to fractional diffusion equations

In our quest of establishing the fractional diffusion equation (FDE), we borrow from the ideas of connecting the random walk approach with the continuum description through the diffusion
equation, and we start off with the continuous time random walk scheme. The latter is flexible enough to account for the rich panel of transport régimes encountered in complex systems. After establishing the fundamental framework of random walks and recovering the standard diffusion equation, we move on to the discussion of the continuous time random walk framework, and the derivation of the FDE. This equation will be shown to enable an investigation of subdiffusive phenomena, and Lévy flights with the tools well-known from dealing with the standard diffusion equation.

3.1. Revisiting the realm of Brownian motion

A typical Brownian walk like Perrin’s original data, is schematically displayed on a two-dimensional lattice in Fig. 4. In discrete time steps of span $\Delta t$, the test particle is assumed to jump to one of its nearest neighbour sites, here displayed on a square lattice with lattice constant $\Delta x$, the direction being random. Such a process can be modelled by the master equation

$$W_j(t + \Delta t) = \frac{1}{2}W_{j-1}(t) + \frac{1}{2}W_{j+1}(t)$$

(10)

in the one-dimensional analogue, as the process is local in both space and time. In Eq. (10), the index denotes the position on the underlying one-dimensional lattice. Eq. (10) defines the pdf to be at position $j$ at time $t + \Delta t$ in dependence of the population of the two adjacent sites $j \pm 1$ at time $t$. The prefactor $1/2$ expresses the direction isotropy of the jumps. In the continuum limit $\Delta t \to 0$ and $\Delta x \to 0$, Taylor expansions in $\Delta t$ and $\Delta x$,

$$W_j(t + \Delta t) = W_j(t) + \Delta t \frac{\partial W_j}{\partial t} + O((\Delta t)^2)$$

(11)
and
\[
W_{j=1}(t) = W(x, t) \pm \Delta x \frac{\partial W}{\partial x} + \frac{(\Delta x)^2}{2} \frac{\partial^2 W}{\partial x^2} + O(\Delta x^3),
\]
lead to the diffusion equation
\[
\frac{\partial W}{\partial t} = K_1 \frac{\partial^2 W}{\partial x^2}(x, t),
\]
on taking along the lowest orders in \( \Delta t \) and \( \Delta x \). The continuum limit thereby has to be drawn such that the quotient
\[
K_1 \equiv \lim_{\Delta x, \Delta t \to 0} \frac{(\Delta x)^2}{2 \Delta t}
\]
is finite. \( K_1 \) is called the diffusion constant and is of dimension \([K_1] = \text{cm}^2 \text{s}^{-1}\).

The diffusion equation (13) is one of the most fundamental equations in physics, being a direct consequence of the central limit theorem \([27,42,58]\). Under the condition that the first two moments of the pdf, describing the appropriately normalised distance covered in a jump event and the variance, \( \bar{X} = \Sigma_i X_i \) and \( \bar{X}^2 \), as well as the mean time span \( \Delta t \) between any two individual jump events, exist, the central limit theorem assures that the random walk process is characterised by a mean velocity \( V = \bar{X}/\Delta t \) and a diffusion coefficient \( K = (2\Delta t)^{-1}[\bar{X}^2 - \bar{X}^2] \) \([27,42]\). Furthermore, for long times, i.e., a large enough number of steps, the pdf of being at a certain position \( x \) at time \( t \), is governed by the diffusion equation (13), and it is given by the Gaussian shape
\[
W(x, t) = \frac{1}{\sqrt{4\pi K_1 t}} \exp\left(-\frac{x^2}{4K_1 t}\right).
\]
\( W(x, t) \) from Eq. (15) is called the propagator, i.e., the solution of the diffusion equation (13) for the sharp initial condition \( W_0(x) = \lim_{t \to 0^+} W(x, t) = \delta(x) \). Individual modes of Eq. (13) decay exponentially in time,
\[
W(k, t) = \exp(-K_1 k^2 t),
\]
with the Fourier transformed diffusion equation,
\[
\frac{\partial W}{\partial t} = -K_1 k^2 W(k, t),
\]
being a relaxation equation, for a fixed wavenumber \( k \).

3.2. The continuous time random walk model

For the generalisations to anomalous transport, we choose the continuous time random walk (CTRW) scheme as the starting point. In parallel to the complementary, dual approach in the standard diffusion problem, we will then develop a generalised diffusion equation of fractional order on the basis of the CTRW.
Fig. 5. Continuous time random walk (CTRW) model. Left: CTRW process on a two-dimensional lattice, generalising the Brownian situation from Fig. 4. The waiting times are symbolised by the waiting circles the diameter of each is proportional to the waiting time which is to be spent on a given site before the next jump event occurs. The jump lengths are still equidistant. Right: $(x, t)$ diagram of a one-dimensional CTRW process where both jump lengths and waiting times are drawn from pdfs which allow for a broad variation of the corresponding random variables.

The CTRW model is based on the idea that the length of a given jump, as well as the waiting time elapsing between two successive jumps are drawn from a pdf $\psi(x, t)$ which will be referred to as the jump pdf. From $\psi(x, t)$, the jump length pdf

$$\lambda(x) = \int_0^\infty dt \, \psi(x, t)$$

(18)

and the waiting time pdf

$$w(t) = \int_{-\infty}^\infty dx \, \psi(x, t)$$

(19)

can be deduced. Thus, $\lambda(x) \, dx$ produces the probability for a jump length in the interval $(x, x + dx)$ and $w(t) \, dt$ the probability for a waiting time in the interval $(t, t + dt)$. If the jump length and waiting time are independent random variables, one finds the decoupled form $\psi(x, t) = w(t) \lambda(x)$ for the jump pdf $\psi(x, t)$. If both are coupled, i.e., $\psi(x, t) = p(x|t)w(t)$ or $\psi(x, t) = p(t|x)\lambda(x)$, a jump of a certain length involves a time cost or, vice versa; i.e., in a given time span the walker can only travel a maximum distance. In what follows, we employ the decoupled version. A schematic cartoon of the CTRW model is drawn in Fig. 5.

Different types of CTRW processes can be categorised by the characteristic waiting time

$$T = \int_0^\infty dt \, w(t)$$

(20)

and the jump length variance

$$\Sigma^2 = \int_{-\infty}^\infty dx \, \lambda(x)x^2$$

(21)
being finite or diverging, respectively. With these definitions, a CTRW process can be described through an appropriate generalised master equation [150,151,164,166,167], via a set of Langevin equations [159,160,216,217], or by the equation [151]

\[
\eta(x, t) = \int_{-\infty}^{\infty} dx' \int_{0}^{\infty} dt' \eta(x', t)\psi(x - x', t - t') + \delta(x)\delta(t)
\]  

(22)

which relates the pdf \(\eta(x, t)\) of just having arrived at position \(x\) at time \(t\), with the event of having just arrived at \(x'\) at time \(t'\), \(\eta(x', t')\). The second summand in Eq. (22) denotes the initial condition of the random walk, here chosen to be \(\delta(x)\). Consequently, the pdf \(W(x, t)\) of being in \(x\) at time \(t\) is given by

\[
W(x, t) = \int_{0}^{t} dt\eta(x, t')\Psi(t - t'),
\]

(23)
i.e., of arrival on that site at time \(t'\), and not having moved since. The latter is being defined by the cumulative probability

\[
\Psi(t) = 1 - \int_{0}^{t} dt' w(t')
\]

(24)

assigned to the probability of no jump event during the time interval \((0, t)\). In Fourier–Laplace space, the pdf \(W(x, t)\) obeys the algebraic relation [151]

\[
W(k, u) = \frac{1 - w(u)}{u} \frac{W_0(k)}{1 - \psi(k, u)},
\]

(25)

where \(W_0(k)\) denotes the Fourier transform of the initial condition \(W_0(x)\).

### 3.3. Back to Brownian motion

Consider now different cases of the CTRW model defined through the decoupled jump pdf \(\psi(x, t) = w(t)\lambda(x)\). If both characteristic waiting time and jump length variance, \(T\) and \(\Sigma^2\), are finite, the long-time limit corresponds to Brownian motion. Let us consider, for instance, a Poissonian waiting time pdf \(w(t) = \tau^{-1} \exp(-t/\tau)\) with \(T = \tau\), together with a Gaussian jump length pdf \(\lambda(x) = (4\pi\sigma^2)^{-1/2} \exp(-x^2/(4\sigma^2))\) leading to \(\Sigma^2 = 2\sigma^2\). Then, the corresponding Laplace and Fourier transforms are of the forms

\[
w(u) \sim 1 - u\tau + O(\tau^2),
\]

(26)

\[
\lambda(k) \sim 1 - \sigma^2 k^2 + O(k^4).
\]

(27)

In fact, any pair of pdfs leading to finite \(T\) and \(\Sigma^2\) leads to the same result, to lowest orders, and thus in the long-time limit [151]. Installing Eqs. (26) and (27) into Eq. (25), one readily recovers, for
the initial condition $W_0(x) = \delta(x)$, the Fourier–Laplace space transform of the propagator,

$$W(k,u) = \frac{1}{u + K_1 k^z},$$

with $K_1 \equiv \sigma^2 / \tau$. Back-transformed to $(x,t)$-coordinates, this is but the well-known Gaussian propagator, Eq. (15). After multiplication with the denominator in Eq. (28), and making use of the differentiation theorems of Fourier (i.e., $\mathcal{F}\{\partial^2 W(x,t)/\partial x^2\} = -k^2 W(k,t)$) and Laplace (i.e., $\mathcal{L}\{\partial W(x,t)/\partial t\} = uW(x,u) - W_0(x)$) transformations, Fick’s second law (13) is immediately obtained. Note that the notion long-time, equivalent to the diffusion limit, is only relative in respect to the time scale $\tau$. In Fourier–Laplace space, the diffusion limit is given through the assumption of $(k,u) \rightarrow (0,0)$ [150,223,227,228].

### 3.4. Long rests: a fractional diffusion equation describing subdiffusion

Consider the following situation, sometimes referred to as fractal time random walk [101], where the characteristic waiting time $T$ diverges, but the jump length variance $\Sigma^2$ is still kept finite. To this end, a long-tailed waiting time pdf with the asymptotic behaviour [83–85]

$$w(t) \sim A_x (t/\tau)^{1+z},$$

for $0 < x < 1$ is introduced, which has the corresponding Laplace space asymptotics [83–85,151,229–230]\(^{12}\)

$$w(u) \sim 1 - (ut)^x.$$  

Again, the specific form of $w(t)$ is of minor importance. Consequently, together with the Gaussian jump length pdf characterised through Eq. (27), the pdf in Fourier Laplace space becomes

$$W(k,u) = \frac{[W_0(k)/u]}{1 + K_x u^{-z}k^2}$$

in the $(k,u) \rightarrow (0,0)$ diffusion limit. Employing the integration rule for fractional integrals [226,232],

$$\mathcal{L}\{\partial^p W(x,t)/\partial t^p\} = u^{-p} W(x,u), \quad p \geq 0,$$

one infers the fractional integral equation

$$W(x,t) - W_0(x) = \partial_{t}^{-z} K_x \partial_{x}^2 W(x,t)$$

from relation (31). By application of the differential operator $\partial_{t}^{-z}$, one finally arrives at the FDE

$$\partial_{t}^{-z} K_x \partial_{x}^2 W(x,t).$$

\(^{12}\)Note that $\psi(u = 0)$, i.e. $\lim_{u \rightarrow 0} \int_0^u \ dt \ e^{-u} \psi(t)$ is but the normalisation of the waiting time pdf, i.e. $\psi(u = 0) = 1$.  


The Riemann–Liouville operator \( _0D_t^{1-a} = (\partial / \partial t) _0D_t^{-a} \), for \( 0 < a < 1 \), is defined through the relation [226,232–235]

\[
_0D_t^{1-a}W(x, t) = \frac{1}{\Gamma(a)} \int_0^t \frac{d\tau}{(t - \tau)^{1-a}} W(x, \tau) .
\]  

(35)

Its fundamental property is the fractional order differentiation of a power,

\[
_0D_t^{1-a}t^p = \frac{\Gamma(1 + p)}{\Gamma(p + a)} t^{p + a - 1} .
\]  

(36)

In fact, it can be shown that the more general relation

\[
_0D_t^{q}t^p = \frac{\Gamma(1 + p)}{\Gamma(1 + p - q)} t^{p - q}
\]  

holds, for any real \( q \). Especially, the Riemann–Liouville fractional differintegration of a constant becomes

\[
_0D_t^{q}1 = \frac{1}{\Gamma(1 - q)} t^{-q} .
\]  

(38)

The special cases of integer order differentiation of a constant, \( d^n1/ dt^n = 0 \), are included through the poles of the Gamma function for \( q = 1, 2, \ldots \). A more detailed introduction to fractional differintegration is given in Appendix A.

Thus, the integrodifferential nature of the Riemann–Liouville fractional operator \( _0D_t^{1-a} \) according to Eq. (35), with the integral kernel \( M(t) \propto t^{x-1} \), ensures the non-Markovian nature of the subdiffusive process defined by the FDE (34). Indeed, calculating the mean squared displacement from relation (31) via the relation

\[
\langle x^2 \rangle = \lim_{k \to 0} \{ - (d^2 / dk^2)W(k, u) \}
\]  

and subsequent Laplace inversion, the result

\[
\langle x^2(t) \rangle = \frac{2K_x}{\Gamma(1 + a)} t^a
\]  

(39)

is obtained. Alternatively, it can be inferred from the FDE (34) through integration over \( \int_{-\infty}^{\infty} dx \, x^2 \), leading to \( (d/dt)\langle x^2(t) \rangle = _0D_t^{1-a}2K_x = 2K_xt^{a-1}/\Gamma(a) \).

Rewriting the FDE (34) in the equivalent form

\[
_0D_t^{x}W - \frac{t^{-x}}{\Gamma(1 - x)} W_0(x) = K_x \frac{\partial^2}{\partial x^2} W(x, t) ,
\]  

(40)

the initial value \( W_0(x) \) is seen to decay with the inverse power-law form \( (t^{-x}/\Gamma(1 - x))W_0(x) \), and not exponentially fast as for standard diffusion [215]. Note that in the limit \( x \to 1 \), the FDE (34) reduces to Fick’s second law, as it should. The generalised diffusion constant \( K_x \) which appears in the FDE (34), is defined by

\[
K_x \equiv \sigma^2 / \tau^a
\]  

(41)
in terms of the scales $\sigma$ and $\tau$, leading to the dimension $[K_a] = \text{cm}^2 \text{s}^{-3}$. The FDE (34) was first considered in the integral form (33) by Schneider and Wyss [175]. An equivalent form was considered by Balakrishnan [173], and a differential form by Wyss [174].

A closed-form solution for the FDE (34) can be found in terms of Fox functions, the result being

$$W(x, t) = \frac{1}{\sqrt{4\pi K_\tau t^2}} H_{1, 2}^{2, 0} \left[ \frac{x^2}{4K_\tau t^2} \right] \frac{(1 - \sigma/2, \sigma/2)}{(0, 1, (\sigma/2, 1))} ,$$

introducing the Fox function $H_{1, 2}^{2, 0}$. The Fox function is defined in Appendix B. Note that the result (42) can be rewritten in the alternative form

$$W(x, t) = \frac{1}{\sqrt{4K_\tau t^2}} H_{1, 1}^{1, 0} \left[ \frac{|x|}{\sqrt{K_\tau t^2}} \right] \frac{(1 - \sigma/2, \sigma/2)}{(0, 1)} ,$$

employing the definition of the Fox function and the duplication rule of the Gamma function [236].

Due to the occurrence of non-integer powers of the Laplace variable $u$ in the expression for $W(k, u)$, Eq. (31), a direct Laplace inversion is not tabled. There are three basic methods to compute the inversion: (i) First applied by Wyss [174], and Schneider and Wyss [175], the Mellin technique can overcome this problem by the roundabout way through Mellin space. Thereby, the path integral defining the Mellin inversion has a similar structure as the definition of the Fox functions, Eq. (B.8), so that the result can be directly inferred from its Mellin transform. (ii) One can identify the expression for $W(k, u)$, Eq. (31), with its corresponding Fox function, and then use the existing rules for the Fox functions to calculate the Laplace and Fourier inversions, see Refs. [180,237]. The result is again a Fox function, which can be simplified by standard rules, to obtain the above results. (iii) One can first Fourier invert $W(k, u)$, to obtain

$$W(x, u) = \frac{1}{2} u^{\sigma/2 - 1} \exp(- |x| u^{\sigma/2}) ,$$

expand the exponential function in its Taylor series, and invert term-by-term, using the rule (37). The final result is a power series, which can be shown to be identical with expression (43) [215]. Without the identification as a Fox function, the obtained series does not render any straightforward information on the stretched exponential asymptotics (45) derived below from standard properties of the Fox function.

Employing some standard theorems of the Fox function, one can derive the asymptotic stretched Gaussian behaviour

$$W(x, t) \sim \frac{1}{\sqrt{4\pi K_\tau t^2}} \left[ \frac{1}{2 - \sigma} \left( \frac{\sigma}{2} \right)^{(1 - \sigma)/(2 - \sigma)} \left( \frac{|x|}{\sqrt{K_\tau t^2}} \right)^{-1} \right] \exp \left( - \frac{2 - \sigma}{2} \left( \frac{|x|}{\sqrt{K_\tau t^2}} \right)^{1/(1 - \sigma/2)} \right) ,$$

valid for $|x| > \sqrt{K_\tau t^2}$. The functional form of the result (45) is equivalent to the CTRW findings reported by Zumofen and Klafter [238,239], see Eq. (9). Furthermore, $W(x, t)$ can be represented
Fig. 6. Propagator $W(x, t)$ for subdiffusion with anomalous diffusion exponent $\alpha = 1/2$, drawn for the consecutive times $t = 0.1, 1, 10$. The cusp shape of the pdf is distinct.

via the series expansion

$$W(x, t) = \frac{1}{\sqrt{4K_xt^2}} \sum_{n=0}^{\infty} \frac{(-1)^n}{n!} \left[ \frac{x^2}{n!} \right]^{n/2} \left( \frac{1}{n+1} \right)$$

in computable form. If $\alpha$ is a rational rather than a real number, the Fox function in Eqs. (42) and (43) can be simplified. Thus, for $\alpha = 1/2$, it can be rewritten in terms of the Meijer $G$-function

$$W(x, t) = \frac{1}{\sqrt{2\pi^2K_{1/2}}} H_{0,0}^{2,0} \left[ \frac{x^2}{8K_{1/2}} \right] \left[ 0, 1 \right] \left( \frac{1}{4}, \frac{1}{2} \right)$$

$$= \frac{1}{\sqrt{8\pi^3K_{1/2}}} H_{0,3}^{3,0} \left[ \frac{x^2}{16K_{1/2}} \right] \left[ 0, 1 \right] \left( \frac{3}{4}, \frac{1}{2} \right)$$

$$= \frac{1}{\sqrt{8\pi^3K_{1/2}}} G_{0,3}^{3,0} \left[ \frac{x^2}{16K_{1/2}} \right] \left[ 0, \frac{1}{4}, \frac{1}{2} \right]$$

by twice using the duplication formula of the Gamma function [236] in the Mellin–Barnes type integral, Eq. (B.8), defining the Fox function [237,240-243]. This representation is useful, as the Meijer $G$-function belongs to the implemented special functions of Mathematica [244], the notation being

$$\frac{1}{\sqrt{8\pi^3t^{1/2}}} G_{0,3}^{3,0} \left[ \left( \frac{x^2}{16t^{1/2}} \right) \right] \left[ 0, \frac{1}{4}, \frac{1}{2} \right]$$

$$= 1/(8\pi) \cdot 3 \cdot t^{-1/2}/(1/2)$$

$$\text{MeijerG}[[1, 1], [[0, 1/4, 1/2], [1]], x^{-4/(16 \cdot 2 \cdot t)}].$$

(48)
That way, the graphical representation of $W(x, t)$ for the subdiffusive case $\alpha = 1/2$ is obtained, which is displayed in Fig. 6. In comparison to the standard Gaussian result, shown in Fig. 7, the pronounced cusps of the subdiffusive propagator are distinct. Note that single modes of the FDE (34) decay in accordance to the Mittag-Leffler pattern

$$W(k, t) = E_\alpha(-K_\alpha k^2 t^\alpha)$$

with the asymptotic power-law behaviour $W(k, t) \sim (K_\alpha k^2 t^\alpha \Gamma(1 - \alpha))^{-1}$. This typical Mittag-Leffler behaviour of the mode relaxation replaces the exponential mode relaxation (16) occurring for normal diffusion, and it is discussed in more detail in Section 5. The Mittag-Leffler function is introduced in Appendix B.

Recently, it has been shown how boundary value problems for the FDE (34) can be solved [200]. As the jump length pdf in the subdiffusive case $0 < \alpha < 1$ is narrow, i.e., it possesses a finite variance $\Sigma^2$, one can apply the method of images due to Lord Kelvin which is summarised in the book of Feller [88]. Consider, for instance, the half space problem with a reflecting boundary at the origin. This situation is defined through the von Neumann condition, $(\partial Q(x, t)/\partial x)|_{x=0} = 0$, where $Q$, specified below, denotes the image solution of this boundary value problem. Suppose the initial condition to be a sharp distribution at $x_0$, $W_0(x) = \delta(x - x_0)$. Then the free solution can be ‘folded’ along a line through the origin, perpendicular to the $x$ axis, i.e., the unrestricted solution is taken, and the portion which spreads to the space region opposite to $x_0$, in respect to the origin, is reflected at this line; the final result fulfils the von Neumann condition. The solution is thus given by the function [88]

$$Q(x, t|x_0) = W(x - x_0, t) + W(-x - x_0, t),$$

for $x \leq 0$.
where \( W(x,t) \) denotes the solution of the FDE (34), for natural boundary conditions. A typical example is portrayed in Fig. 8, alongside with its Brownian counterpart in Fig. 9. Similarly, for an absorbing barrier, the problem is defined via the Dirichlet condition \( Q(x_0,t) = 0 \), and the half space solution for the sharp initial condition \( W_0(x) = \delta(x - x_0) \) takes on the form

\[
Q(x, t|x_0) = W(x - x_0, t) - W(-x - x_0, t).
\]

(51)

For subdiffusion in a box of extension \((-a, a)\), the propagator \( W(x,t) \) also suffices to determine the boundary value problem of two absorbing or two reflecting boundaries which are supposed to lie at \( x = \pm a \). There, the free solution with the initial value problem \( W_0(x) = \delta(x) \) is successively folded along the lines through \( x = \pm a \), perpendicular to the \( x \) axis, i.e., the exact solution is constructed with increasing accuracy according to the method of images, to result in the boundary value solution \([88,205]\)

\[
Q(x, t) = \sum_{m=-\infty}^{\infty} \left[ W(x + 4ma, t) \mp W(4ma - x + 2a, t) \right],
\]

(52)

where the minus sign stands for absorbing, the plus sign for reflecting boundaries at \( x = \pm a \). We note that the solution for the mixed condition of one absorbing and one reflecting boundary is obtained via a final folding at the origin of the solution for two absorbing boundaries. Employing the relation\(^{13}\)

\[
\sum_{m=-\infty}^{\infty} e^{-ikm} = e^{-ik/2} \sum_{m=-\infty}^{\infty} (-1)^m \delta(m + \frac{k}{2\pi}),
\]

(53)

we rewrite Eq. (52), after an additional Laplace transform, as

\[
Q(x, u) = (4a)^{-1} \sum_{m=-\infty}^{\infty} e^{im\pi/(2a)} \left[ W\left(k = \frac{m\pi}{2a}, u\right) \mp (-1)^m W\left(k = -\frac{m\pi}{2a}, u\right) \right].
\]

(54)

Making use of Eq. (31), the sums can be simplified and evaluated numerically. A typical result is shown in Fig. 10, in comparison with the Brownian counterpart.

An alternative solution procedure, the method of separation of variables is discussed in Section 5.

Being a direct generalisation of Fick’s second law, the fractional partial differential equation (34) can be solved by standard methods, like the Fourier–Laplace technique, or the method of images. The solution \( W(x,t) \), Eq. (42), of the FDE (34) has been expressed analytically, in closed form,

\(^{13}\) This relation can be easily proved by the Poisson summation formula \([247]\)

\[
\sum_{k=-\infty}^{\infty} f(k) = \sum_{m=-\infty}^{\infty} \int_{-\infty}^{\infty} \ dx f(x)e^{2\pi i k x}
\]

and the integral definition of the delta function \([248]\)

\[
\delta(x - x') = \frac{1}{2\pi} \int_{-\infty}^{\infty} dk e^{ik(x - x')},
\]

as well as some resummations.
Fig. 8. Subdiffusive image solution $Q(x,t)$ for a reflecting boundary at $x = 0$, portrayed for the times $t = 0.02, 0.2, 20$, for $z = 1/2$. The relatively long time passing by until the initial hump at $x_0 = 1$ disappears, in comparison to the Brownian case, is obvious on comparison to Fig. 9.

Fig. 9. Brownian image solution $Q(x,t)$ portrayed for the times $t = 0.01, 0.1$ and $1$. The initial condition of starting in $x_0 = 1$ decays much faster in comparison to the subdiffusive case graphed in Fig. 8.

by the Fox function $H^{1,0}_{1,2}$ which has been studied in detail; compare Appendix B. The propagator $W(x,t)$ possesses similar scaling properties like the Gaussian solution (15), i.e., it is given by the functional form $W(x,t) = \phi(t)g(\xi)$, where the scaling variable $\xi$ is defined through $\xi = x/t^{1/2}$ [249].
Fig. 10. Image solution $Q(x,t)$ for absorbing boundaries at $x = \pm 1$. Top: The subdiffusive case, $\alpha = 1/2$. Bottom: The Brownian case, $\gamma = 1$. Note, again, the cusp shape of the subdiffusive solution. The curves are shown for the times $t = 0.005, 0.1$ and $10$ on top and $t = 0.05, 0.1$ and $10$ on the bottom. The broad wings in the top graph are due to the stretched Gaussian shape of the free propagator in the subdiffusive regime.

Verifications of the FDE (34) have so far been discussed for NMR in disordered systems where the usually found Gaussian shape of the Fourier transformed propagator gets replaced by an inverse power-law pattern [225,250]. These predictions were corroborated by findings from NMR experiments in biological tissue [251] which were interpreted along the lines of a fractional diffusion equation.\textsuperscript{14} A further example is given by FRAP (Fluorescent Recovery After Photobleaching) experiments [253,254].

3.5. Long jumps: Lévy flights

The opposite case of finite characteristic waiting time $T$ and diverging jump length variance $\Sigma^2$ can be modelled by a Poissonian waiting time and a Lévy distribution for the jump length, i.e.,

$$\lambda(k) = \exp(-\sigma^\alpha |k|^\alpha) \sim 1 - \sigma^\alpha |k|^\alpha$$

\textsuperscript{14} It was shown that the results follow an asymptotic power-law in Fourier space which is neither consistent with the Gaussian approach nor a Lévy flight or walk approach which is of stretched Gaussian shape in Fourier space.
for $1 < \mu < 2$, corresponding to the asymptotic behaviour

$$\lambda(x) \sim A_\mu \sigma^{-\mu} |x|^{-1-\mu}$$

(56)

for $|x| \gg \sigma$. Due to the finiteness of $T$, this process is of Markovian nature. Substituting the asymptotic expansion from Eq. (55) into the relation (25), one obtains

$$W(k,u) = \frac{1}{u + K^\mu |k|^\mu}$$

(57)

from which, upon Fourier and Laplace inversion, the FDE [192,202]

$$\frac{\partial W}{\partial t} = K^\mu - \infty D_\alpha^\mu W(x,t)$$

(58)

is inferred. The Weyl operator $- \infty D_\alpha^\mu$ which in one dimension is equivalent to the Riesz operator $\nabla^\mu$, is defined in Section A.2. Here, the generalised diffusion constant is

$$K^\mu \equiv \sigma^\mu / \tau$$

(59)

and carries the dimension $[K^\mu] = \text{cm}^\mu \text{s}^{-1}$. The Fourier transform of the propagator can be readily computed, obtaining

$$W(k,t) = \exp(-K^\mu t |k|^\mu),$$

(60)

which is but the characteristic function of a centred and symmetric Lévy distribution, and as such used to generate Lévy flights [42]. Note that the Fourier space version of Eq. (58) was discussed in Ref. [134].

In Fig. 11 a computer simulation of a Lévy flight is shown on the right, in comparison to the trajectory of a walk with finite jump length variance $\Sigma^2$, for the same number of steps. Due to the asymptotic property (56) of the jump length pdf, very long jumps may occur with a significantly higher probability than for an exponentially decaying pdf like the formerly employed Gaussian jump length pdf. The scaling nature of the jump length pdf, as expressed by Eq. (56) leads to the clustering nature of the Lévy flights, i.e., local motion is occasionally interrupted by long sojourns, on all length scales. That is, one finds clusters of local motion within clusters. In fact, the Lévy flight trajectory can be assigned a fractal dimension $d_f = \mu [38,42,89]$ and is commonly supposed to be an efficient search strategy of living organisms [141,142,146]. Contrarily, the trajectory drawn on the left of Fig. 11, with $\Sigma^2 < \infty$, fills the two-dimensional space completely, and features no distinguishable clusters, as all jumps are of about the same length.

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15 The case $0 < \mu \leq 1$, though very similar, is not discussed here.

16 Note that we suppress the imaginary unit in Fourier space by adopting the slightly modified definition $\mathcal{F}\{-\infty D_\alpha^\mu f(x)\} = -|k|^\mu \hat{f}(k)$ instead of $\mathcal{F}\{-\infty D_\alpha^\mu f(x)\} = i|k|^\mu \hat{f}(k)$, following a convention initiated by Compte [192].

17 Note that we refer to the trajectory of the Lévy flight. The spatiotemporal behaviour of these organisms has a finite velocity of motion, see below.
Fig. 11. Comparison of the trajectories of a Brownian or subdiffusive random walk (left) and a Lévy walk with index $\mu = 1.5$ (right). Whereas both trajectories are statistically self-similar, the Lévy walk trajectory possesses a fractal dimension, characterising the island structure of clusters of smaller steps, connected by a long step. Both walks are drawn for the same number of steps (approx. 7000).

The solution of the FDE (58) in $(x, t)$ space can again be obtained analytically by making use of the Fox functions, the result being [195,217]

$$W(x, t) = \frac{1}{\mu|x|} H^{1.1}_{2,2} \left[ \frac{|x|}{(K^\mu t)^{1/\mu}} \right]^{(1,1/\mu),(1,1/2),(1,1)}.$$

This is a closed-form representation of a Lévy stable law, see Appendix C for details. For $\lim_{\mu \to 2}$, the classical Gaussian solution is recovered, by standard theorems of the Fox functions. As expected, one can infer from Eq. (61) the power-law asymptotics [47,217]

$$W(x, t) \sim \frac{K^\mu t}{|x|^{1+\mu}}, \quad \mu < 2,$$

typical for Lévy distributions. Due to this property, the mean squared displacement diverges:

$$\langle x^2(t) \rangle \to \infty$$
which has caused some controversy in that it is different from quantities obtained through cut-offs, scaling relations obtained from similarity methods, or rescaling of a fractional moment
\[
\langle |x|^\delta \rangle \propto t^{\delta/\mu}
\]
for \(0 < \delta < \mu \leq 2\), and loosely called “mean squared displacement” \([42,49,52,159,160,227,228]\).

Clearly, Eq. (63) cannot be valid for a particle with a non-diverging mass. For such massive particles, a finite velocity of propagation exists, making long instantaneous jumps impossible, see below.

The exact calculation of fractional moments profits from the definition of the Fox functions. Thus, writing
\[
\langle |x|^\delta \rangle = 2 \int_0^\infty dx x^\delta W(x,t)
\]
with \(0 < \delta\), it follows that this integral, due to Eq. (62), only converges for \(\delta < \mu\). In this case, note that through Eq. (61) the integral (65) defines the Mellin transformation
\[
\mathcal{M}\{f(t);s\} = \int_0^\infty dt t^{s-1}f(t)
\]
of the Fox function:
\[
\langle |x|^\delta \rangle = 2 \int_0^\infty dx x^{\delta-1} H^{1,2}_{1,1} \left[ \frac{|x|}{(K^\mu t)^{1/\mu}} \right] (1,1/(1+\mu),1,1/2) = 2 \frac{\mu}{\mu} \mathcal{M}\left\{ H^{1,2}_{1,2} \left( \frac{|x|}{(K^\mu t)^{1/\mu}} \right) \right\}.
\]

Employing the property \([237,242]\)
\[
\int_0^\infty dx x^{s-1} \mathcal{H}^{m,n}_{p,q} \left[ \frac{(a_p,A_p)}{(b_q,B_q)} \right] = \alpha^{-s} \chi(\alpha-s)
\]
of the Fox function, where \(\chi(s)\) is defined in Eq. (B.9), one readily infers
\[
\langle |x|^\delta \rangle = \frac{2}{\mu} (K^\mu t)^{\delta/\mu} \chi(-\delta) = \frac{2}{\mu} (K^\mu t)^{\delta/\mu} \frac{\Gamma(-\delta/\mu)\Gamma(1+\delta)}{\Gamma(-\delta/2)\Gamma(1+\delta/2)}.
\]

Due to the condition \(0 < \delta < \mu\), \(\langle |x|^\delta \rangle\) is always positive, as can be seen from the \(\Gamma\) functions in Eq. (69). Also, whereas \(\langle |x|^\delta \rangle \propto t^{\delta/\mu}\) is always sublinear in \(t\), the rescaling of this fractional moment leads to the pseudo mean squared displacement \([x^2] \propto t^{2/\mu}\), i.e., “superdiffusion”.

Consider two special cases. For \(\delta \to 0\), one proves from Eq. (69) the normalisation:
\[
\lim_{\delta \to 0} \langle |x|^\delta \rangle = 1,
\]
by help of \(1/\Gamma(z) \sim z\), for \(z < 1\). In the Gaussian limit \(\mu = 2\), the linear time dependence
\[
\lim_{\delta \to 2, \mu \to 2} \langle |x|^\delta \rangle = 2K_1 t
\]
of the mean squared displacement is recovered. Note that numerically obtained fractional moments have been used in single molecule spectroscopy for system characterisation [138,139].

For certain physical systems, for instance the diffusion in energy space encountered in single molecule spectroscopy [138,139] or the diffusion on a polymer chain in chemical space [255] the divergence (63) of the mean squared displacement does not violate physical principles. However, if dealing with massive particles in direct space, physics implies a finite velocity of propagation. The latter dilemma can be overcome by replacing Lévy flights by Lévy walks, a version of the CTRW with a spatiotemporal coupling, usually introduced through the delta coupling \( \psi(x, t) = \frac{1}{4}w(t)\delta(|x| - vt) \) [52,151–153], with a time cost penalising long jumps [49,52,151]. This is, for example, the case in the chaotic phase diffusion in Josephson junctions [256], in turbulent flow [257], or in chaotic Hamiltonian systems [49,258,259]. As we are mainly concerned with subdiffusive systems, Lévy walks are not considered here.

Boundary value problems for Lévy flights are more involved than for the subdiffusive case, as the long jumps make the definition of a boundary actually quite intricate. Such problems were discussed for Lévy flights in the half-space by Zumofen and Klafter [260] and for a box by Drysdale and Robinson [261].

3.6. The competition between long rests and long jumps

As noted, the problem of the diverging mean squared displacement encountered in the discussion of Lévy flights is often circumvented by the consideration of \((x-t)\) scaling relations, or measuring the width of the pdf \( W(x, t) \) rather than its variance. An alternative method was applied in Ref. [217], making use of the definition

\[
\langle x^2(t) \rangle_L \sim \int_{L_1 \to t^{1/\mu}} dx x^2 W(x, t) \sim t^{2/\mu}
\]

according to which the walker is considered in an imaginary, growing box, leading to the \( \sim t^{2/\mu} \) behaviour. The latter was verified by numerical simulations [217]. Note that the cut-offs of the integral in Eq. (72) are time dependent. The imaginary box spans the spatial interval \( \Delta(t) = (L_1 - L_2)t^{1/\mu} \) which grows in the course of time. It gives a measure, that a finite portion of the probability is gathered within the given interval \( \Delta(t) \).

By use of such relations, one can consider a random walk characterised through broad pdfs for both waiting time and jump length, thus leading to infinite \( T \) and \( \Sigma^2 \), and the FDE

\[
\frac{\partial W}{\partial t} = \sigma D_1^{1-\alpha} K_\Sigma^\alpha \nabla^\mu W(x, t)
\]

with \( K_\Sigma^\alpha \equiv \sigma^\mu/\tau^2 \). In this case, the appropriately defined quantity \( \langle x^2(t) \rangle_L \), which we call the pseudo or imaginary mean squared displacement, reveals the temporal form

\[
\langle x^2(t) \rangle_L \sim t^{2/\mu}.
\]
Consequently, the phase diagram displayed in Fig. 12 can be drawn. Note that in expression (74), the exponent \( \alpha \) enters in a “proper” way, i.e., it is already present for \( \Sigma^2 < \infty \), whereas the Lévy index \( \mu \) manifests the artifice of introducing the pseudo mean squared displacement. Conversely, the competition of laminar motion events (“flights”) and localisation (waiting) events in the Lévy walk picture is given through the relation [262]

\[
\langle x^2 \rangle \sim \begin{cases} 
 t^{2+\alpha-\mu} & \text{if } 1 < \mu < 2, \ 0 < \alpha < 1 , \\
 t^{\alpha-\mu} & \text{if } 1 < \mu < 2, \ \alpha > 1 , 
\end{cases}
\]  

(75)

whereby the time spans spent in laminar motion events is governed by the pdf \( \psi(t) \sim at^{-\mu-1} \), and the waiting times are drawn from the pdf \( \tilde{\psi}(t) \sim bt^{-\alpha-1} \). Eq. (75) for Lévy walks clearly differs from the multiplicative nature by which the exponents \( \alpha \) and \( \mu \) enter into Eq. (74).

3.7. What’s the course, helmsman?

Let us take stock at this point and address two important issues concerning the fractional diffusion concept.

3.7.1. The long time limit and its consequence for the fractional diffusion equation

In the derivation of both the subdiffusive FDE (34) and the Lévy flight FDE (58) from the CTRW scheme, the diffusion limit \((k,u) \to (0,0)\) was drawn. Equivalently, the diffusion limit corresponds to choosing \((\sigma,\tau) \to (0,0)\) with \( K_\alpha = \sigma^2/\tau^\alpha = const \) or \( K^\mu = \sigma^\mu/\tau = const \) which matches the limit
\[ K = \lim_{\Delta x \to 0, \Delta t \to 0} (\Delta x)^2 / (2 \Delta t) \] drawn in the Brownian case. In this sense, the FDE is valid in the diffusion limit \( t \gg \tau \). In this diffusion limit, the equivalence to the CTRW model holds true for all moments. Indeed, from the FDE (34) one obtains the general expression

\[ \langle x^{2n}(t) \rangle = (2n)! \frac{K^n t^n}{\Gamma(1 + nz)}. \quad (76) \]

From the CTRW theory with untruncated wave number expansion, one finds higher-order corrections of the form

\[ \langle x^{2n}(t) \rangle = (2n)! \frac{K^n t^n}{\Gamma(1 + nz)} (1 + O((K t^z)^{-1})) \quad (77) \]

which can be neglected in the drawn diffusion limit.

Differences with the CTRW model occur in details of the pdf \( W(x, t) \) where higher orders in the \( k \) expansion come into play [239], and in higher dimensions as discussed in Ref. [215].

### 3.7.2. Fractional diffusion equations and limit theorems

The Gaussian solution obtained for Brownian motion is a consequence of the central limit theorem. How should the results for the subdiffusive FDE (34) and its Lévy flight counterpart (58) be judged from the limit theorem point of view?

The subdiffusive process combines the long tailed waiting time process with a jump length distribution that possesses a finite characteristic variance. The resulting pdf \( W(x, t) \) which is the solution of the FDE (34) is not Gaussian. However, in the sense of Bouchaud and Georges [42], there exists a central limit theorem controlling this process as \( W(x, t) = \xi(t)f(x/\xi(t)) \) as can be seen from Eq. (42). Moreover, the temporal distribution of this subdiffusive process is a one-sided Lévy distribution characterised through its Laplace transform, the quantity \( W(x, u) \) as given in Eq. (44). The existence of such a central limit theorem for this type of random walk process is further corroborated by the properties of the Fox function in the solution (42) which guarantee a smooth transition to the classical Gaussian in the limit \( z \to 1 \).

Lévy flights are Markovian processes and they are governed by a pdf which is Lévy stable. Therefore, they are a direct result of the Lévy–Gnedenko generalised central limit theorem.

Thus, the FDE concept is a valid extension of the standard diffusion equation in the same sense as CTRW generalises Brownian motion.

### 4. Fractional diffusion–advection equations

Diffusion with an additional velocity field \( v \) and diffusion under the influence of a constant external force field are, in the Brownian case, both modelled by the diffusion–advection equation (DAE)

\[ \frac{\partial W}{\partial t} + v \frac{\partial W}{\partial x} = K_1 \frac{\partial^2}{\partial x^2} W(x, t) . \quad (78) \]
In the case of anomalous diffusion this is no longer true, i.e., the fractional generalisation may be different for the advection case and the transport in an external force field. We start with considering the external velocity field, the external force field is discussed in the following section.

In the stationary state, i.e., without inertial terms, described by the DAE (78), the problem is Galilei invariant: it is invariant under a transformation $x \rightarrow x - vt$. Requiring this property to carry over to the anomalous case, a straightforward extension of the CTRW scheme leads to a fractional diffusion–advection equation (FDAE). A Galilei variant model is discussed subsequently. Alternative formulations for advected Lévy flights are presented at the end of this section.

A more detailed summary of FDAEs including dispersive sedimentation processes and the partial sticking mechanism, as well as suggestions for the measurement, are given in Ref. [207].

### 4.1. The Galilei invariant fractional diffusion–advection equation

In the moving frame of the test particle which is dragged along the homogeneous velocity field $v$, the random walk is governed by the usual jump pdf $\psi(x, t)$. The corresponding jump pdf $\phi(x, t)$ in the laboratory frame is consequently obtained via the transformation

$$\phi(x, t) = \psi(x - vt, t) .$$

This carries over, Fourier–Laplace transformed, to the functional relation

$$\phi(k, u) = \psi(k, u + ivk)$$

between $\phi(k, u)$ and $\psi(k, u)$. For the case of infinite characteristic waiting time $T$, i.e., a broad waiting time pdf, and finite jump length variance $\Sigma^2$, the Fourier–Laplace form of the propagator,

$$W(k, u) = \frac{1}{u + ivk + K_x k^2 u^{1-2}} ,$$

can be directly deduced from the CTRW solution (25) for the diffusion limit $k \rightarrow 0$ and $u \rightarrow 0$. Proceeding along the same steps as outlined in Section 3, one is led to the FDAE [202]

$$\frac{\partial W}{\partial t} + v \frac{\partial W}{\partial x} = aD_t^{1-\sigma}K_x \frac{\partial^2}{\partial x^2} W(x, t) ,$$

by virtue of Eq. (32). Due to the required Galilei invariance, the solution for the propagator in $(x, t)$ space is given by the Galilei-shifted solution of the FDE (34), i.e.,

$$W(x, t) = W_{v=0}(x - vt, t)$$

where $W_{v=0}(x, t)$ denotes the free propagator according to Eqs. (42) and (43), for the sharp initial value $W_0(x) = \delta(x)$.
The moments of the FDAE (82) are readily calculated, and one obtains

\[ \langle x(t) \rangle = vt , \]  

\[ \langle x^2(t) \rangle = \frac{2K_x}{I(1 + \alpha)} t^\alpha + v^2 t^2 , \]  

\[ \langle (\Delta x(t))^2 \rangle = \frac{2K_x}{I(1 + \alpha)} t^\alpha . \]

Thus, the mean squared displacement \( \langle (\Delta x(t))^2 \rangle \) contains solely the molecular contribution, i.e., the relative mixing in the moving frame, whereas the first moment \( \langle x(t) \rangle \) accounts for the simple drag along the velocity field \( v \), as is to be expected from the ordinary drift term \( v \partial W/\partial x \) occurring in the FDAE (82). The assumed Galilei invariance thus carries over to the moments, as it should. This behavior is also found in the temporal evolution of the pdf which is depicted in Figs. 13 and 14.

A possible realisation of the Galilei invariant subdiffusion is the motion of a particle in a flow field where the flowing substance itself causes the occurrence of subdiffusion, like in the case of a polymer solution, and a small bead immersed in it.

4.2. The Galilei variant fractional diffusion–advection equation

Instead of assuming the Galilei invariance expressed in Eqs. (79) and (80) leading to the propagator (83), Compte [205], and Compte et al. [206], for an explicitly position-dependent velocity field \( v(x) \), assume the relation

\[ \phi(x, t; x_0) = \psi(x - \tau_a v(x_0), t) \]  

Fig. 13. Galilei invariant subdiffusive model. The propagator is shown for the dimensionless times \( t = 0.02, 0.2 \) and 2. The propagator is symmetric with respect to its maximum which is translated with velocity \( v = 1 \). The cusps marking the initial condition are distinct in comparison to the Brownian result shown in Fig. 14.
between the jump pdf $\phi$ and the jump pdf $\psi(x, t)$ of the free diffusion process. Thereby, $\phi$ depends on both the jump length $x$ and the starting point $x_0$. Additionally, a microscopic advection time $\tau_a$ is introduced. On that basis, the framework developed in Ref. [205] is applied in Ref. [206] to Taylor flows.

The resulting FDAE [204–206]

$$
\frac{\partial W}{\partial t} = 0D_1^{1-z} \left[ - A_x \frac{\partial}{\partial x} v(x) + K_x \frac{\partial^2}{\partial x^2} \right] W(x, t) \tag{88}
$$

has the same structure as the fractional Fokker–Planck equation. For a homogeneous velocity field, the resulting FDAE

$$
\frac{\partial W}{\partial t} = 0D_1^{1-z} \left[ - A_x \frac{\partial}{\partial x} v + K_x \frac{\partial^2}{\partial x^2} \right] W(x, t) \tag{89}
$$

has a constant drift coefficient. It can be proved that the fractional solution does not fulfil a generalised Galilei invariance of the form $W(x - v^* t^*, t)$. However, according to Section 5.4 the fractional solution $W_a(x, t)$ can be expressed in terms of the Brownian solution $W_1(x, t)$, the Galilei shifted Gaussian

$$
W_1(x, t) = \frac{1}{\sqrt{4\pi K t}} \exp\left( - \frac{(x - vt)^2}{4Kt} \right), \tag{90}
$$

for numerical purposes [207]. This way, Fig. 15 was obtained which is to be compared with Figs. 13 and 14 for the Galilei invariant case. Note the growing skewness of the solution. This persistence of the initial condition has already been celebrated by Scher and coworkers.
Fig. 15. Galilei variant subdiffusive model. The propagator is shown for the dimensionless times $t = 0.02, 0.2$ and 2. The propagator is asymmetric with respect to its maximum which stays fixed at the origin. The plume stretches more and more into the direction of the velocity.

The difference to the Galilei invariant model is also mirrored in the moments. From Eq. (89) one deduces

$$\langle x(t) \rangle = \frac{A_x v t^x}{\Gamma(1 + x)} ,$$  \hspace{1cm} (91)

$$\langle x^2(t) \rangle = \frac{2 A_x^2 v^2 t^{2x}}{\Gamma(1 + 2x)} + \frac{2K_x t^x}{\Gamma(1 + x)} ,$$  \hspace{1cm} (92)

where now the first moment increases sublinearly in time. This model describes physical systems where trapping occurs, compare Section 6, i.e., the particle gets repeatedly immobilised in the environment for a trapping time drawn from the waiting time pdf $w(t)$, before it gets dragged along the velocity stream again. Experimental realisations might be found in porous systems like the one displayed in Fig. 16 where the particle gets trapped in still regions off the velocity backbone, or the multiple trapping systems addressed in Section 6, or possibly in gel electrophoresis [263].

4.3. Alternative approaches for Lévy flights

For Lévy flights in an external velocity field $v$, i.e., for finite characteristic waiting time $T$ but infinite jump length variance $\Sigma^2$, the FDAE [202]

$$\frac{\partial W}{\partial t} + v \frac{\partial W}{\partial x} = K^\mu \nabla^\mu W(x,t)$$  \hspace{1cm} (93)

is recovered [202], which describes a Markovian process with diverging mean squared displacement, see the discussion in Section 3.5. The drift term $v \partial W/\partial x$ in Eq. (93) is the same as in the
Fig. 16. Swiss cheese percolation network, as used by Klemm et al. [106]. The image clearly shows the relatively sharp distinction between regions with high flow velocities (backbones), and still domains. A particle trapped in one of the side-pores off the backbones stays approximately stationary, until it rejoins the main stream backbones. Courtesy A. Klemm and R. Kimmich, Ulm University.

standard DAE (78), and thus the result is simply given in accordance to the subdiffusive case, Eq. (83), by \( W_v(x,t) = W_{v=0}(x - vt, t) \) where here \( W_{v=0}(x, t) \) is the Lévy stable solution (61). The latter is symmetric in \( x \).

There may occur situations where the resulting propagator is asymmetric. This case is modelled by the alternative approach in Refs. [208,209] where a skewed Lévy distribution is assumed instead of the symmetric one leading to the Riesz/Weyl operator \( \nabla^{\nu} \) introduced in the preceding parts. Consequently, the resulting FDAE possesses an additional parameter accounting for the asymmetry, in comparison to the FDAE (93).

5. The fractional Fokker–Planck equation: anomalous diffusion in an external force field

Many physical transport problems take place under the influence of an external force field: a constant electrical bias field exerting a force on charge carriers, a periodic potential encountered in certain problems in solid state physics or in the modelling of molecular motors, a bistable potential in reaction dynamics or molecular switching processes, or a harmonic potential describing a bound particle. In this section, a framework for the treatment of anomalous diffusion problems under the influence of an external force field is developed. A physical model based on the Langevin equation with Gaussian white noise for multiple trapping systems is introduced in the next section.
5.1. The Fokker–Planck equation

Normal diffusion in an external force field is often modelled in terms of the Fokker–Planck equation (FPE) [24,29,35–37,78–80,264–266]

\[
\frac{\partial W}{\partial t} = \left[ \frac{\partial}{\partial x} V(x) + K_1 \frac{\partial^2}{\partial x^2} \right] W(x,t)
\]  

(94)

where \( m \) is the mass of the diffusing test particle, \( \eta_1 \) denotes the friction constant characterising the interaction between the test particle and its embedding, and the force is related to the external potential through \( F(x) = -\frac{dV(x)}{dx} \). Let us first list some basic properties of the FPE (94) to which the fractional counterpart developed below can be compared.

(i) In the force-free limit, the FPE (94) reduces to Fick’s second law and thus the time evolution of the mean squared displacement follows the linear form (4).

(ii) Single modes of the FPE (94) relax exponentially in time,

\[ T_n(t) = \exp(-\lambda_{n,1} t), \]

(95)

where \( \lambda_{n,1} \) is the eigenvalue of the Fokker–Planck operator \( L_{FP} \) defined below.

(iii) The stationary solution

\[ W_{st}(x) \equiv \lim_{t \to \infty} W(x,t) \]

(96)

is given by the Gibbs–Boltzmann distribution

\[ W_{st}(x) = N \exp(-\beta V(x)) \]

(97)

where \( N \) is a normalisation constant and \( \beta \equiv (k_B T)^{-1} \) denotes the Boltzmann factor.

(iv) The FPE (94) further fulfills the Einstein–Stokes–Smoluchowski relation [38]

\[ K_1 = \frac{k_B T}{m \eta_1} \]

(98)

which is closely connected with the fluctuation–dissipation theorem.

(v) Finally, the second Einstein relation

\[ \langle x(t) \rangle_F = \frac{1}{2} \frac{F \langle x^2(t) \rangle_0}{k_B T}, \]

(99)

is recovered from Eq. (94) which connects the first moment in presence of the constant force \( F \), \( \langle x(t) \rangle_F \), with the second moment in absence of this force, \( \langle x^2(t) \rangle_0 \), with the second moment in absence of this force,

\[ \langle x^2(t) \rangle_0 = 2K_1 t. \]

(100)

The latter relationship, Eq. (99), is a consequence of linear response.

5.2. The fractional Fokker–Planck equation

The FPE (94) is well studied for a variety of potential types, and the respective results have found wide application. For the description of anomalous transport in the presence of an external field,
we introduce a fractional extension of the FPE, namely the fractional Fokker–Planck equation (FFPE) [212–215]
\[
\frac{\partial W}{\partial t} = \alpha D^{1-\alpha} \left[ \frac{\partial}{\partial x} V'(x) \right] - \frac{\partial^2}{\partial x^2} W(x,t). \tag{101}
\]

The FP-operator
\[
L_{\text{FP}} = \frac{\partial}{\partial x} V'(x) + K_z \frac{\partial^2}{\partial x^2}
\]

occurring in the FFPE (101) contains the generalised diffusion constant \(K_z\) and the generalised friction constant \(\eta_z\) of dimension \([\eta_z] = s^{z-2}\). For \(z \to 1\), the standard FPE (94) is recovered, and for \(V(x) = \text{const.}\), i.e. in the force-free limit, the FDE (34) emerges.

We will show that this equation fulfils the following requirements:

(i) The FFPE (101) describes subdiffusion in accordance to the mean squared displacement
\[
\langle x^2(t) \rangle_0 = \frac{2K_z}{T(1 + x)} t^x \tag{103}
\]

in the force-free limit. This is obvious as for \(V(x) = \text{const.}\), Eq. (101) reduces to the standard FDE, Eq. (34).

(ii) The relaxation of single modes is governed by a Mittag–Leffler pattern
\[
T_a(t) = E_a(-\lambda_a t^x). \tag{104}
\]

(iii) The stationary solution is given by the Gibbs–Boltzmann distribution (97).


(v) The second Einstein relation (99) can be shown to hold true for Eq. (101).

The FFPE (101) will be derived explicitly in the following subsection, and independently in the next section as the high friction limit of a fractional phase space model.

In order to derive the stationary solution \(W_{\text{st}}(x)\) of the FFPE (101), note that the right-hand side of the Eq. (101) can be rewritten as
\[
- \alpha D^{1-\alpha} \frac{\partial S(x,t)}{\partial x} \tag{105}
\]
in terms of the probability current [36]
\[
S(x,t) = \left( - \frac{V(x)}{m\eta_z} - K_z \frac{\partial}{\partial x} \right) W(x,t). \tag{106}
\]

If a stationary state is reached, \(S(x,t)\) must be a constant. Thus, if \(S_{\text{st}}(x_0) = 0\) at any point \(x_0\), it vanishes everywhere, and the stationary solution of the FFPE satisfies [36,212]
\[
\frac{V'(x)}{m\eta_z} W_{\text{st}}(x) + K_z \frac{d}{dx} W_{\text{st}}(x) = 0 \tag{107}
\]
from which the exponential result
\[ W_{st}(x) = N \exp \left( - \frac{V(x)}{m\eta_z K_x} \right) \] (108)
can be inferred. Requiring, in analogy to the standard case, that \( W_{st} \) is given by the Boltzmann distribution (97), the generalised Einstein–Stokes–Smoluchowski relation [42,203,212–215,267]
\[ K_x = k_B T/m\eta_z \] (109)
is readily recovered. Thus, the FFPE (101) obeys some generalised fluctuation–dissipation theorem. The generalised Einstein–Stokes–Smoluchowski relation (109) has recently been supported by findings of Amblard and coworkers [120,121]. In order to check whether the FFPE (101) satisfies the second Einstein relation (99), the first moment in presence of the constant force \( F \) is calculated under linear response conditions, obtaining the expression
\[ \langle x(t) \rangle_F = \frac{F}{m\eta_z \Gamma(1+\lambda)} t^\lambda . \] (110)
A comparison to the force-free result of the mean squared displacement, Eq. (103), leads to the second Einstein relation (99). The validity of this relation in the case of subdiffusion has found support experimentally in the work by Schiff et al. [103].

At this point it is worth digressing a bit and noting that the condition \( \partial W/\partial t = 0 \), i.e., the usual stationary condition, allows for a unique solution for the fractional equation (101), namely the Gibbs–Boltzmann form (97). On the one hand, \( \partial W/\partial t = 0 \) implies \( W = \text{const}(x) \). On the other hand, \( {}_0D_t^{1-\lambda} W(x,t) = 0 \) is fulfilled by \( W(x,t) = \text{const}(x) \) corresponding to the reasoning in the above derivation via the vanishing probability current, \( S_{st}(x) = 0 \), as well as by \( W(x,t) = \text{const}(x) t^{-\lambda} \), according to relation (37). This contradicts, however, the requirement \( W = \text{const}(x) \) imposed by the stationarity condition. Thus, the exponential form derived above is unique. This fact is mirrored in the stationary solution obtained by the method of separation of variables discussed below, which leads to a time-independent solution. Note further that a quasi-stationary form similar to \( \propto t^{-\lambda} \) was found for a fractional phase space equation by Hilfer in his thermodynamical derivation of fractional dynamics [235,268,269].

5.3. Separation of variables and the fractional Ornstein–Uhlenbeck process

5.3.1. The separation of variables

In order to determine the pattern according to which the stationary solution is approached, the separation ansatz
\[ W_n(x,t) = T_n(t) \phi_n(x) \] (111)
for a given mode \( n \) is introduced into the FFPE (101), leading to the decoupled set of eigen-equations
\[ \frac{dT_n(t)}{dt} = - \lambda_{n,\lambda} {}_0D_t^{1-\lambda} T_n(t) \] (112)
\[ L_{FP} \phi_n(x) = - \lambda_{n,\lambda} \phi_n(x) \] , (113)
featuring the fractional eigenvalues $\lambda_{n,a}$. The temporal eigenfunction $T_n(t)$ being governed by the fractional relaxation equation (112), it is described by the Mittag–Leffler pattern

$$T_n(t) = E_\alpha(-\lambda_{n,a} t^\alpha) \equiv \sum_{j=0}^{\infty} \frac{(-\lambda_{n,a} t^\alpha)^j}{\Gamma(1+\alpha j)}$$

with the choice $T_n(0) = 1$. For $\alpha = 1$, the standard exponential form $T_n(t) = \exp(-\lambda_{n,1} t)$ follows, whereas for $0 < \alpha < 1$, the initial stretched exponential behaviour

$$T_n(t) \sim \exp\left(-\frac{\lambda_{n,a} t^\alpha}{\Gamma(1+\alpha)}\right)$$

turns over to the power-law long-time behaviour

$$T_n(t) \sim \frac{1}{\Gamma(1-\alpha)\lambda_{n,a} t^\alpha}.$$  

This interpolation from stretched exponential to inverse power-law behaviour has been reported from the rheology of polymeric systems [20,21], and from protein re-binding [22]. Let us briefly examine the importance of the Mittag–Leffler function. In Refs. [180,270] it is shown, that the Mittag–Leffler function is the exact relaxation function for an underlying fractal time random walk process, and that this function directly leads to the Cole–Cole behaviour [270,271] for the complex susceptibility which is broadly used to describe experimental results. Furthermore, the Mittag–Leffler function can be decomposed into single Debye processes, the relaxation time distribution of which is given by a modified, completely asymmetric Lévy distribution [270]. This last observation is related to the formulation of Mittag–Leffler relaxation described in Refs. [21,180]. In Ref. [245], the significance of the Mittag–Leffler function was shown, where its Laplace transform was obtained as a general result for a collision model in the Rayleigh limit.

The full solution of the FFPE (101) is given by the sum over all eigensolutions, i.e., by

$$W(x,t|x',0) = e^{\Phi(x')/2-\Phi(x)/2}\sum_n \psi_{n}(x)\psi_{n}(x')E_\alpha(-\lambda_{n,a} t^\alpha)$$

for an initial distribution concentrated in $x'$. Here, the eigenfunctions $\psi_n(x) = e^{\Phi(x)/2}\phi_n(x)$ of the Hermitian operator $L$ are defined in terms of the $\phi_n(x)$, the eigenfunctions of the FP-operator $L_{FP}$, via the scaled potential $\Phi(x) = V(x)/[k_B T]$, $L$ and $L_{FP}$ sharing the same eigenvalues $\lambda_{n,a}$ [36]. On arranging these eigenvalues in increasing order, i.e., $0 \leq \lambda_{0,a} < \lambda_{1,a} < \lambda_{2,a} < \cdots$, the first eigenvalue is zero iff there exists a stationary solution, which is positive definite and defined through

$$W_{st}(x) = \lim_{t \to \infty} W(x,t).$$

This stationary solution is independent of the fractional index $\alpha$ and coincides with the required Boltzmann solution (97). However, the temporal relaxation of a single mode $n$ towards equilibrium is highly non-exponential.

The FFPE (101) obeys, in Laplace space, the functional relation [212]

$$W_\lambda(x,u) = \frac{\eta_2}{\eta_1} u^{\alpha-1} W_1\left(x, \frac{\eta_2}{\eta_1} u^\alpha\right)$$

for an initial distribution concentrated in $x'$. Here, the eigenfunctions $\psi_n(x) = e^{\Phi(x)/2}\phi_n(x)$ of the Hermitian operator $L$ are defined in terms of the $\phi_n(x)$, the eigenfunctions of the FP-operator $L_{FP}$, via the scaled potential $\Phi(x) = V(x)/[k_B T]$, $L$ and $L_{FP}$ sharing the same eigenvalues $\lambda_{n,a}$ [36]. On arranging these eigenvalues in increasing order, i.e., $0 \leq \lambda_{0,a} < \lambda_{1,a} < \lambda_{2,a} < \cdots$, the first eigenvalue is zero iff there exists a stationary solution, which is positive definite and defined through

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$$W_\lambda(x,u) = \frac{\eta_2}{\eta_1} u^{\alpha-1} W_1\left(x, \frac{\eta_2}{\eta_1} u^\alpha\right)$$
for the same initial condition $W_0(x) = \delta(x - x')$. The subscripts refer to the solutions of the FPE (94) for $z = 1$ and the FFPE (101) for arbitrary $0 < z < 1$, respectively. Thus, in the Laplace domain, the subdiffusive system passes through the same states as its normal counterpart, for a rescaled Laplace variable. In Section 5.4 we show that the solution $W_z$ of the FFPE (101) can be expressed in terms of $W_1$ through an integral relation from which the non-negativity of $W_z$ can be proved, as the Pawula theorem guarantees that the solution $W_1$ is a proper pdf. For the force–free FDE (34) describing subdiffusion, Schneider and Wyss proved directly that the propagator $W(x, t)$, Eq. (42), is a proper pdf.

5.3.2. The fractional harmonically bound particle

It is worthwhile considering the example of a subdiffusive, harmonically bound particle, i.e. the subdiffusive motion in the potential $V(x) = \frac{1}{2}m\omega^2x^2$ which exerts a restoring force on the test particle. We find the following solution [212]

$$W = \sqrt{\frac{m\omega^2}{2\pi k_B T}} \sum_{n=0}^{\infty} \frac{1}{2^n n!} E_n(- n\tilde{\tau}) H_n\left(\frac{\tilde{x}}{\sqrt{2}}\right) H_n\left(\frac{\tilde{x}}{\sqrt{2}}\right) e^{-\tilde{x}^2/2},$$

employing reduced coordinates $\tilde{t} = t/\tau$ and $\tilde{x} = x/\sqrt{m\omega^2/[k_B T]}$, as well as $\tau^{-z} \equiv \omega^2/\eta_z$. $H_n$ denotes the Hermite polynomials [236], and the eigenvalues here are $\lambda_{n,z} = m\omega^2/\eta_z$. This result is plotted in Figs. 17–19 in the course of time, for an asymmetric initial condition, and compared to the Brownian result. The stationary solution of this process is found to be [236]

$$W_{st}(x) = \sqrt{\frac{m\omega^2}{2\pi k_B T}} H_0\left(\frac{\tilde{x}}{\sqrt{2}}\right) H_0\left(\frac{\tilde{x}}{\sqrt{2}}\right) e^{-\tilde{x}^2/2} = \sqrt{\frac{m\omega^2}{2\pi k_B T}} \exp\left(\frac{-m\omega^2 x^2}{2k_B T} \right)$$

i.e., the Gibbs–Boltzmann distribution, as it should.

For a given potential $V(x)$, one can extract the moments of order $n$, $\langle x^n(t) \rangle$, directly from the FFPE (101), by integration over $\int_{-\infty}^{\infty} dx x^n$. This leads to a fractional differential equation for the respective moment. For instance, for $\langle x(t) \rangle$ one obtains the relation $(d/dt)\langle x(t) \rangle = \partial D_1^{1-z}(\omega^2/\eta_z)\langle x(t) \rangle$, i.e., the fractional relaxation equation, solved by the Mittag–Leffler pattern

$$\langle x(t) \rangle = x_0 E_z(- (t/\tau)^z).$$

Eq. (121) describes the relaxation of the mean of an off-centre initial distribution, sliding into the symmetric final state on the bottom of the potential valley, which is characterised by the symmetric limit $\lim_{t \to \infty} \langle x(t) \rangle = 0$. The temporal evolution of the second moment,

$$\langle x^2(t) \rangle = x_0^2 + [x_0^2 - x_{th}^2] E_z(- 2(t/\tau)^z)$$

also follows the Mittag–Leffler pattern, however characterised by the relaxation time scale $t/2^{1/z}$. Hereby, the thermal equilibrium is defined as $x_{th}^2 = k_B T/[m\omega^2]$ reached for $t \to \infty$. The second moment (122) is graphed for the subdiffusive case in Fig. 20, in comparison to the Brownian counterpart.

Note that this process discussed in position space is equivalent to the fractional version of the Ornstein–Uhlenbeck process. The Laplace transformed versions of Eqs. (121) and (122) were found in Ref. [245] to describe the relaxation of a heavy test particle immersed in a bath of light particles, in a generalised Rayleigh limit.
Fig. 17. Pdf \( W(x,t) \), Eq. (119), of the fractional Ornstein–Uhlenbeck process, for the anomalous diffusion exponent \( \alpha = 1/2 \). The initial value is chosen to be \( W_0(x) = \delta(x - 1) \). The maximum clearly slides towards the origin, acquiring an inversion symmetric shape. The curves are drawn for the times \( t = 0.02, 0.2, \) and \( 20 \), employing the integral relation with the Brownian solution. Note the distinct cusps around the initial position. Compare Fig. 18.

Fig. 18. Fractional Ornstein–Uhlenbeck process. Comparison of the numerical behaviour of the summation representation (dashed) with 151 summation terms and the integral representation (A transform). The latter is obtained by \texttt{Mathematica} employing the numerical integration command \texttt{NIntegrate}. The cusp which is a typical feature for subdiffusive processes is much more pronounced in the curve obtained through the integral transformation. The computation time for the latter is even shorter than for the calculation of the truncated sum so that this representation is preferable for numerical purposes.
5.4. The connection between the fractional solution and its Brownian counterpart

The solution of the FFPE, $W(x, t)$, follows the scaling relation given in Eq. (118). As shown by Barkai and Silbey [272], Eq. (118) can be rewritten in the form

$$W_2(x, t) = \int_0^\infty ds A(s, t)W_1(x, s)$$

which corresponds to a modified Laplace transformation from $t$ to $(\eta_2/\eta_1)u_s$. The kernel $A(s, t)$ is defined in terms of the inverse Laplace transformation

$$A(s, t) = \mathcal{L}^{-1}\left\{\frac{\eta_2}{\eta_1 u_s^{1-s}}\exp\left(-\frac{\eta_2}{\eta_1}u_s^s\right)\right\},$$

the result being the modified one-sided Lévy distribution $L_2^{+18}$

$$A(s, t) = \frac{t}{2s}L_2^\left(\frac{t}{(s^*)^{1/2}}\right), \quad s^* \equiv \eta_2 s/\eta_1 .$$

Consequently, the transformation (123) guarantees the existence and positivity of $W_2(x, t)$ if only the Brownian counterpart, $W_1(x, t)$, is a proper pdf.

---

18 $L_2^{+}(t)$ is defined on the positive real axis and shows the asymptotic power-law behaviour $L_2^{+} \sim t^{-1-s}$, compare Appendix C.
In the normalised version, the kernel $A$ is given through $A(s, u) = u^{1/2} e^{-su}$. Rewriting it in terms of a Fox function, one obtains the exact representation

$$A(s, t) = \frac{1}{(s^2 t)^{1/2}} H_1^{1,0} \left[ \frac{s^{1/2}}{t} \left( 1, 1/z \right) \right] \tag{126}$$

in terms of the Fox function $H_1^{1,0}$ whose series representation reads

$$A(s, t) = \sum_{s=0}^{\infty} \frac{(-1)^n}{\Gamma(1 - 2 - zn) \Gamma(1 + n)} \left( \frac{s}{t^2} \right)^{1+n} \tag{127}$$

From the properties of the Fox function, or by using Mathematica, one can derive the following results for some special cases:

for $\alpha = 1/2$:

$$A(s, t) = \frac{1}{\sqrt{\pi t}} \exp \left( -\frac{s^2}{4t} \right) ; \tag{128}$$

for $\alpha = 1/3$:

$$A(s, t) = \sqrt{\frac{s}{3t}} \left[ I_{-1/3} \left( \frac{2s^{3/2}}{3\sqrt{3t}} \right) - I_{1/3} \left( \frac{2s^{3/2}}{3\sqrt{3t}} \right) \right] ; \tag{129}$$
\( \alpha = 2/3: \)

\[
A(s, t) = \frac{1}{t^{\alpha/3}} \left[ \frac{1}{\Gamma(1/3)} \, \text{I} F_1 \left( \frac{5}{6}, \frac{2}{3}; -\frac{4s^3}{27t^2} \right) - \frac{s}{t^{\alpha/3} \Gamma(-1/3)} \, \text{I} F_1 \left( \frac{7}{6}, \frac{4}{3}; -\frac{4s^3}{27t^2} \right) \right].
\]  

(130)

With such expressions at hand, one has a very convenient method for plotting the fractional solutions via relation (123). This method shows a good performance when used in Mathematica with the NIntegrate command. Via this method, Figs. 15 and 17–19 have been obtained.

Note that the connection between \( = a \) and \( = 1 \) is related to the quantity \( s(s) \) that exactly \( s \) events occurred in time \( t \) known from a random walk characterised by the waiting time distribution \( w(t) \), \( \chi_s(u) = [w(u)]'[1 - w(u)]/w(u) \) [277]. For the special form of the waiting time distribution, \( w(u) = e^{-u} \), one finds in the long time limit \( \chi_s(u) = u^{s-1} e^{-su} \equiv A(s, u) \).

5.5. The fractional analogue of Kramers escape theory from a potential well

As we have seen, it is an important feature of the fractional kinetic equations that its solution \( W_\alpha \) can be expressed in terms of its Brownian counterpart \( W_1 \). In fact, all kinetic processes associated with such a fractional equation are affected by scaling relations such as Eq. (118).

Let us recall that in the standard Kramers problem the escape of a scalar test particle subject to a Gaussian white noise over a potential barrier is considered in the limit of low diffusivity, \( \eta/\kappa < K \) where \( \Delta V \) is the barrier height and \( K \) the diffusion constant [36]. The temporal decay of the probability to still find the particle within the potential well is given by an exponential function

\[
p(t) = e^{-r_k t}
\]

(131)

where the Kramers rate in the overdamped limit is defined through [36]

\[
r_k = \frac{1}{2\pi m \eta} \sqrt{V''(x_{\text{min}})}|V''(x_{\text{max}})| \exp(-\beta \Delta V)
\]

(132)

with \( \Delta V = V(x_{\text{max}}) - V(x_{\text{min}}) \); compare Fig. 21. In Eq. (132), the exponential function contains the Boltzmann factor \( \beta \equiv (k_B T)^{-1} \) so that the inverse Kramers rate follows an Arrhenius activation \( r_k^{-1} \propto e^{c/T} \) [274].

Let us now derive the fractional counterpart to the exponential decay pattern from Eq. (131). Application of relation (118) to the Laplace transform \( p(u) = (r_k + u)^{-1} \) of the survival probability, Eq. (131), produces

\[
p_s(u) = \eta_s \frac{1}{\eta u + r_k^2 u^{1-\alpha}}
\]

(133)

with the generalised, fractional Kramers rate

\[
r_k^{(s)} = \frac{\eta}{\eta_s} r_k = \frac{1}{2\pi m \eta_s} \sqrt{V''(x_{\text{min}})}|V''(x_{\text{max}})| \exp(-\beta \Delta V).
\]

(134)
Note that consequently the Arrhenius form of the temperature activation is preserved. Via Laplace inversion of Eq. (133) one finds the fractional survival probability

\[ p_a(t) = E_a(-r_a^2t^z) \]  

in terms of the Mittag–Leffler function \( E_a \).

Often the notion of time-dependent rate coefficients is preferred, i.e., the survival probability is defined as \( p(t) = \exp(-k(t)t) \) in terms of the rate coefficient \( k(t) \). For the fractional Kramers model we therefore find \( k(t) = |\ln E_a(-r_a^2t^z)/t \) which leads to the two limiting cases

\[ k(t) \sim \frac{r_a^2}{t^{1-z}\Gamma(1+z)}, \quad t \ll (r_a^2)^{1/z} \]  

and

\[ k(t) \sim \frac{\zeta}{t} \ln(t[r_a^2\Gamma(1-z)]^{1/z}), \quad t \gg (r_a^2)^{1/z}. \]  

A detailed investigation of this topic is found in Ref. [273], including the discussion of a possible application to ligand rebinding in proteins, compare Refs. [200,275,276].

5.6. The derivation of the fractional Fokker–Planck equation

For the derivation of the subdiffusive FFPE (101), and a further generalisation accounting for a broad jump distance statistics introduced below, we again draw from the random walk
formulation and its duality with (generalised) diffusion equations. In order to derive the standard FPE, the random walk model employed in Section 3 has to be modified as to account for the broken spatial homogeneity, in comparison to the force-free diffusion under natural boundary conditions. A non-linear external force field $F(x)$ acting upon the system, the local probabilities to jump right or left, $A(j)$ and $B(j)$, explicitly depend on the position $j$. The corresponding discrete master equation therefore becomes

$$W_j(t + \Delta t) = A_{j-1} W_{j-1}(t) + B_{j+1} W_{j+1}(t) ,$$

(138)

compare to Eq. (10). Taylor expansions in time, analogous to those already introduced, and in space,

$$A_{j-1} W_{j-1}(t) = A(x) W(x, t) - \Delta x \frac{\partial A(x) W(x, t)}{\partial x} + \frac{(\Delta x)^2}{2} \frac{\partial^2 A(x) W(x, t)}{\partial x^2} + O([\Delta x]^3)$$

(139)

lead to the FPE (94), with the appropriate limits

$$\frac{V'(x)}{m \eta_1} \equiv \lim_{\Delta x \to 0, \Delta t \to 0} \frac{\Delta x}{\Delta t} [B(x) - A(x)] ,$$

(140)

$$K_1 \equiv \lim_{\Delta x \to 0, \Delta t \to 0} \frac{(\Delta x)^2}{2 \Delta t} .$$

(141)

For taking these limits, we impose the normalisation $A(x) + B(x) = 1$ and note that the inhomogeneity in jumping left or right, $A(x) - B(x)$, becomes small for $\Delta x \to 0$, according to a Boltzmann distribution for a system close to thermal equilibrium, so that

$$B(x) - A(x) \simeq \frac{\Delta x V'(x)}{2 k_B T} + O([\Delta x]^2) .$$

(142)

The master equation (138) involves local steps in time and position, thus enabling the Taylor expansions. Broad waiting time pdfs or jump length pdfs do not allow for such an expansion, as they are connected with long-range steps, and thus $\Delta x$ or $\Delta t$ cannot be considered small parameters.

An extension of above scheme for subdiffusion in an external force field has recently been presented by Barkai et al. [215], and is discussed below. Here, we prefer an alternative derivation which also accounts for a broad jump length statistics [213,214]. Note that for non-local jumps, already the simple master equation (138) has to be modified to the form

$$W_j(t + \Delta t) = \sum_{n=1}^{\infty} A_{j,n} W_{j-n}(t) + \sum_{n=1}^{\infty} B_{j,n} W_{j+n}(t)$$

(143)

allowing for jumps from any site $j \pm n$ to site $j$, involving the normalisation

$$\sum_{n=1}^{\infty} (A_{j,n} + B_{j,n}) = 1 ,$$

(144)
i.e., a step to site \( j \) can come from any site \( j \pm n \). In order to introduce the continuum limit of this random walk model, the transfer kernel \([213,214]\)

\[
A(x, x') \equiv \lambda(x - x')[A(x')\Theta(x - x') + B(x')\Theta(x' - x)]
\]

(145)

replaces the homogeneous and isotropic jump length pdf \( \lambda(x - x') = \lambda(|x - x'|) \) in the CTRW approach. Thus, the function \( A \) explicitly depends on both the departure site \( x' \) and the arrival site \( x \). \( A \) obeys the normalisation condition

\[
\int_{-\infty}^{\infty} d\delta A(x'||\delta) = 1
\]

(146)

where \( A(x'||x - x') \equiv A(x, x') \). Including an arbitrary waiting time pdf \( w(t) \), it was shown that the underlying inhomogeneous CTRW is governed by the generalised master equation \([213]\)

\[
\frac{\partial W}{\partial t} = \int_{-\infty}^{\infty} dx' \int_{0}^{t} dt' K(x, x'; t - t') W(x', t')
\]

(147)

with the kernel

\[
K(x, x'; u) \equiv uw(u) \frac{A(x', x) - \delta(x)}{1 - w(u)}
\]

(148)

given in Laplace space; or, equivalently, through the equation

\[
W(x, t) = \int_{-\infty}^{\infty} dx' \int_{0}^{t} dt' w(t - t') A(x, x') W(x', t') + \Psi(t) W_0(x)
\]

(149)

Transforming to Fourier–Laplace space, the FFPE (101) can be derived for a broad waiting time pdf with a diverging characteristic waiting time \( T \), Eq. (20), in combination with a finite jump length variance \( \Sigma^2 \), Eq. (21). Taking also non-local jump statistics into account, i.e., assuming a jump length pdf with infinite \( \Sigma^2 \), one recovers the FFPE

\[
\frac{\partial W}{\partial t} = \sigma_D^{-1} - \left[ \frac{\partial}{\partial x} \frac{V'(x)}{m\eta_x} + K^u \nabla^u \right] W(x, t)
\]

(150)

whereby the drift and diffusion coefficients are given by \([213,214]\)

\[
\frac{V'(x)}{m\eta_x} \equiv \frac{2\sigma}{\mu \tau_x}[B(x) - A(x)]
\]

(151)

\[
K^u \equiv \frac{\sigma^u}{\tau_u^2}
\]

(152)

The FFPE (150) thus describes the competition between subdiffusion and Lévy flights, as it was already encountered for the FDE (73).

An alternative derivation for the FFPE (101), from an asymmetric random walk scheme is given by Barkai et al. as follows \([215]\).
We consider an unbounded random walk on a one-dimensional lattice with a lattice spacing $a$. Individual lattice sites are denoted by $\{ \ldots, -1, 0, 1, \ldots, n, \ldots \}$. At time $t = 0$ the particle be located at site $n = 0$. Once the particle has arrived at site $n$ it is trapped there for some random time. These waiting times are given by $\{ \tau_i \}$ and $i = 1, 2, \ldots$. The $\{ \tau_i \}$ are independent random variables, identically distributed according to a pdf $w(\tau)$. It is assumed that $w(\tau)$ is independent of the location of the particle $n$ (i.e., it is independent of the external field). It is further assumed that the particle executes only nearest neighbour jumps. The probability of hopping from site $n$ to $n + 1$ is $A(n)$, and from site $n$ to site $n - 1$ it is $B(n)$, the normalisation condition being $A(n) + B(n) = 1$. $A(n)$ and $B(n)$ are time independent.

The probability that the random walker has jumped $i$ times in the interval $(0, t)$, $Q_i(t)$, is given in Laplace space through

$$Q_i(u) = \frac{1 - w(u)}{u} w(u)^i,$$  \hspace{1cm} (153)

using the Laplace transform $Y(u) = (1 - w(u))/u$ of the sticking probability (24). If $W_n(t)$ denotes the probability of finding the particle at site $n$ at time $t$, and $p_i(n)$ be the probability to be on site $n$ after step $i$, then

$$W_n(t) = \sum_{i=0}^{\infty} p_i(n) Q_i(t).$$ \hspace{1cm} (154)

Using Eq. (153),

$$W_n(u) = \frac{1 - w(u)}{u} \sum_{i=0}^{\infty} p_i(n) w(u)^i.$$ \hspace{1cm} (155)

The evolution of $p_i(n)$ is determined by the discrete time and space equation

$$p_{i+1}(n) = A(n-1)p_i(n-1) + B(n+1)p_i(n+1).$$ \hspace{1cm} (156)

In Eq. (156) we have used the assumption that the directional jump probabilities $A(n)$ and $B(n)$ are independent of the waiting times. The continuum limit of this equation is obtained by using the replacement $p_i(n) \to p_i(x)$, where $p_i(x) \, dx$ is the probability of finding the particle after the $i$th jump in the interval $(x, x + dx)$. Similarly, $A(n) \to A(x)$ and $B(n) \to B(x)$, with the normalisation $A(x) + B(x) = 1$. In addition we have $A(n+1) \to A(x+a)$ and $B(n+1) \to B(x+a)$. We now expand Eq. (156) in a Taylor series in $a$, a typical term being

$$A(n-1)p_i(n-1) \to A(x)p_i(x) + \frac{\partial}{\partial x} [A(x)p_i(x)](a) + \frac{\partial^2}{\partial x^2} [A(x)p_i(x)] a^2 + \cdots,$$ \hspace{1cm} (157)

where higher-order terms proportional to $a^3$, $a^4$, etc., are omitted.

The system is supposed to be close to thermal equilibrium defined by a temperature $T$, $A(x) \simeq B(x) \simeq 1/2$, and to be according to detailed balance $A(x) - B(x) \simeq a F(x)/(2k_B T)$. In this case we obtain from Eqs. (157)–(158) in the continuum limit

$$p_{i+1}(x) = p_i(x) + \frac{a^2}{2} \left[ \frac{\partial^2}{\partial x^2} p_i(x) - \frac{\partial}{\partial x} F(x) \right] p_i(x) + \cdots.$$ \hspace{1cm} (158)
Rewriting Eq. (155) leads to

\[
W(x, u) = \frac{1 - w(u)}{u} p_0(x) + \frac{1 - w(u)}{u} \sum_{i=1}^{\infty} p_i(x) w(u)^i
\]

(159)

where the continuum approximation \( W_n(u) \to W(x, u) \) has been made. Inserting Eq. (158) into Eq. (159), and using \( p_0(x) = \delta(x) \), one finds

\[
W(x, u) = \frac{1 - w(u)}{u} \delta(x) + \frac{1 - w(u)}{u} \sum_{i=1}^{\infty} \left\{ p_{i-1}(x) + \frac{a^2}{2} \partial^2_{x^2} p_{i-1}(x) - \frac{a^2}{2} \partial_x \left[ p_{i-1}(x) \frac{F(x)}{k_B T} \right] + \cdots \right\} w(u)^i
\]

(160)

Notice that according to Eq. (155)

\[
\frac{1 - w(u)}{u} \sum_{i=1}^{\infty} p_{i-1}(x) w(u)^i = W(x, u) w(u)
\]

(161)

and hence from Eq. (160) one obtains

\[
W(x, u) = \frac{1 - w(u)}{u} \delta(x) + w(u) \left\{ W(x, u) + \frac{a^2}{2} \partial^2_{x^2} W(x, u) - \frac{a^2}{2} \partial_x \left[ W(x, u) \frac{F(x)}{k_B T} \right] + \cdots \right\}
\]

(162)

Introducing the waiting time probability density function, which for small \( u \) behaves as

\[
w(u) = 1 - (u \tau)^x + c_1 (u \tau)^{2x} + \cdots
\]

(163)

with \( 0 < x < 1 \), the first moment of the waiting times diverges. Inserting Eq. (163) into Eq. (162) one is led to

\[
W(x, u) = \frac{(u \tau)^x - c_1 (u \tau)^{2x} + \cdots}{u} \delta(x) + \left[ 1 - (u \tau)^x + c_1 (u \tau)^{2x} + \cdots \right] x \left\{ W(x, u) + \frac{a^2}{2} \partial^2_{x^2} W(x, u) - \frac{a^2}{2} \partial_x \left[ W(x, u) \frac{F(x)}{k_B T} \right] + \cdots \right\}
\]

(164)

Consider the limit \( a \to 0 \). In the standard diffusion approximation (i.e., \( x = 1 \)) such a limit is meaningful only when both the mean waiting time and the lattice spacing \( a \) approach zero. For those cases where the mean waiting time diverges, the standard limit of the diffusion approximation breaks down. We take \( a \to 0 \) and \( \tau \to 0 \), while the ratio

\[
\lim_{a^2 \to 0, \tau \to 0} \frac{a^2}{2 \tau^2} = K_x
\]

(165)
is kept finite. When $x = 1$, $K_1 = a^2/(2\langle \tau \rangle)$ and $\langle \tau \rangle = \tau$ is the finite mean waiting time, as expected for this normal case. Multiplying Eq. (164) by $\tau^{-\zeta}u^{1-\zeta}$ and using the limiting procedure defined in Eq. (165), we find that

$$uW(x, u) - \delta(x) = u^{1-\zeta}L_{FP}W(x, u), \quad (166)$$

where

$$L_{FP} \equiv K_s \left( \frac{\partial^2}{\partial x^2} - \frac{\partial}{\partial x} \frac{F(x)}{k_B T} \right)$$

is the well known Fokker–Planck operator. Eq. (166) can be rewritten in $t$ space in terms of the fractional Riemann–Liouville operator [232] as

$$\frac{\partial W}{\partial t} = D_0^{1-\zeta}L_{FP}W(x, t). \quad (168)$$

5.7. A fractional Fokker–Planck equation for Lévy flights

The FFPE (150) for $\mu = 2$, i.e., for the case with finite jump length variance $\Sigma^2$, reduces to the subdiffusive FFPE (101) discussed in Section 5.2. Here we briefly discuss the opposite case of finite characteristic waiting time $T$, but diverging $\Sigma^2 \to \infty$. This Markovian analogue of the FFPE (150),

$$\frac{\partial W}{\partial t} = \left[ \frac{\partial}{\partial x} \frac{V'(x)}{m\eta_1} + K_1^\mu - D_x^\mu \right]W(x, t) \quad (169)$$

describes Lévy flights in the force field $F(x)$. It can be independently derived from the Langevin equation [159,160,216,217]

$$\frac{d}{dt}x(t) = \frac{F(x)}{m\eta_1} + \Gamma(t) \quad (170)$$

with the Lévy noise $\Gamma(t)$ the distribution of which, $p(\Gamma)$, is given through $p(k) = \exp( - K^\mu |k|^\mu )$ in Fourier space. The FFPE was investigated in some detail by Jespersen et al. [217].

The basic feature of the FFPE (169) is that it describes systems far off thermal Boltzmann equilibrium. Thus, for a harmonically bound particle underlying the potential $V(x) = \frac{1}{2}m\omega^2 x^2$, the stationary solution

$$W_{st}(x) = \frac{1}{|x|} H_{\frac{1}{2}, \frac{1}{2}} \left[ \frac{|x|^{\mu}s_\omega^2}{K^\mu_1 \eta_1 m^\mu} \right] (1, 1, 1, \frac{\mu}{2}) \quad (171)$$

is Lévy stable, its Fourier transform being

$$W_{st}(k) = \exp\left( - \frac{\eta_1 mK^\mu_1 |k|^\mu}{\mu s^2} \right), \quad (172)$$
Consequently, the asymptotic behaviour

\[
W_{st}(x) \sim \frac{K^2 \eta_1 m}{\mu \omega^2 |x|^{1+\mu}}
\]

is enforced for large $|x|$. Especially, the mean squared displacement diverges, even for the stationary case.

Making use of the method of characteristics, it can be shown that the solution of the FFPE (169) for the force $F(x) = -\omega^2 x + F$ is given by

\[
W_{\omega,F} = W_0 \left( x - \frac{F}{\omega^2} \left[ 1 - \exp\left\{ -\frac{\omega^2 t}{m \eta_1} \right\} \right] \right) \left[ 1 - \exp\left\{ -\frac{\omega^2 t}{m \eta_1} \right\} \right]
\]

where $W_0(x,t)$ denotes the force-free solution

\[
W_0(x,t) = \frac{1}{\mu |x|} H^{1;1}_{2;2} \left[ \frac{|x|}{(K_t^2 t)^{1/\mu}} \right] (1,1/\mu),(1,1/2)
\]

which is again Lévy stable [217].

5.8. A generalised Kramers–Moyal expansion

Taking into account higher-order terms in the Taylor expansions of the type (139), leading to the FPE (94), the Kramers–Moyal (KM) expansion

\[
\frac{\partial W}{\partial t} = \sum_{n=1}^{\infty} \left( -\frac{\partial}{\partial x} \right)^n D^{(n)}(x)W(x,t)
\]

can be obtained, where the KM-coefficients are defined through

\[
D^{(n)}(x) = \frac{(\Delta x)^n}{n! \Delta t} [A(x) + (-1)^n B(x)]
\]

Alternatively, the KM-expansion foots on an expansion of the distribution function [36]

\[
P(x,t + \tau|x,t) = \int dy \delta(x - y)P(y,t + \tau|x',t),
\]

$P$ denoting the transition probability from $x'$ to $x$ during the time span $\tau$, in combination with the formal expansion

\[
\delta(x - y) = \sum_{n=0}^{\infty} \frac{(y - x')}{n!} \left( -\frac{\partial}{\partial x} \right)^n \delta(x' - x).
\]

Note that in the full KM-expansion (176) no limits have to be taken as the full Taylor expansion is included. This is connected with the Pawula theorem [36] as either the Taylor expansion has to be terminated after the second order, or no proper limit can be defined. A truncation of the KM-expansion after the $n$th term, $n > 2$, may lead to negative solutions for the pdf $W(x,t)$ [36]. In the random walk derivation, the lack of appropriate limits is obvious, as the limit is only properly defined for the quotient $(\Delta x)^2/\Delta t$. All higher terms are, however, of order $(\Delta x)^2+n/\Delta t$. 


A generalised KM-expansion has been obtained recently for systems underlying Lévy jump statistics with index $1 < \mu < 2$ [214]:

$$\frac{\partial W}{\partial t} = \sum_{n=1}^{\infty} -D_n^{\mu} D(x) + \sum_{n=1}^{\infty} (-1)^n \frac{\partial}{\partial x} D_n^{\mu} \bar{D}^{(n)}(x)$$

$$+ \sum_{n=0}^{\infty} (-1)^{1+n} \left( \frac{\partial}{\partial x} \right)^{1+2n} \bar{D}^{(n)}(x) \right] W(x, t)$$

where the generalised KM-coefficients are defined via

$$D^{(n)}(x) = i^{n-1} \frac{\sigma^{n\mu}}{\tau^x},$$

$$\bar{D}^{(n)}(x) = i^{n-1} (1 - \delta_{2,\mu}) \frac{\sigma^{1+n\mu}}{\tau^x} \left[ A(x) - B(x) \right] \sin \left( \frac{n \mu \pi}{2} \right) \frac{\sin \left( \frac{\pi}{2} [1 + n \mu] \right)}{2n! \sin \left( \frac{\pi}{2} [1 + n \mu] \right)},$$

$$\bar{D}^{(n)}(x) = \sigma^{1+2n} \left[ A(x) - B(x) \right] \left( \frac{1+2n}{\mu} \right) \frac{\sin \left( \frac{n \mu \pi}{2} \right)}{\pi \mu}.$$  

Note that the occurrence of possibly imaginary coefficients in higher-order terms in the generalised Kramers–Moyal expansion (180) is due to the differentiation theorem of the Fourier transformation (compare footnote 16); thus, the Riesz–Weyl operator occurs to orders $n\mu$, $n = 1, 2, \ldots$.

A crucial feature of the generalised KM-expansion is that the lowest-order contribution involves a first-order spatial derivative $\partial/\partial x$, and thus preserves the physical drift character. That means that, to lowest order, the external force in the generalised KM-expansion leads to a translation of the pdf $W(x, t)$. The latter statement is different from the findings of Zaslavsky et al. [187,193,211] who assume a generalisation of relation (179) in their description of chaotic Hamiltonian systems.

6. From the Langevin equation to fractional diffusion: microscopic foundation of dispersive transport close to thermal equilibrium

In this final section we briefly review a physical scenario giving some insight into the origin of the fractional Fokker–Planck equation for multiple trapping systems. From the continuous time version of the Chapman–Kolmogorov equation combined with the Markovian Langevin equation of a damped particle in an external force field, a fractional Klein–Kramers equation is derived whose velocity averaged high-friction limit reproduces the fractional Fokker–Planck equation, and explains the occurrence of the generalised transport coefficients $K_z$ and $\eta_z$. 

6.1. Langevin dynamics and the three stages to subdiffusion

In his treatment of the Brownian motion of a scalar test particle in a bath of smaller atoms or molecules exerting random collisions upon that particle, Langevin [64] amended Newton’s law of motion with a fluctuating force. On the basis of the resulting, stochastic, Langevin equation, the corresponding phase space dynamics is governed by the deterministic Klein–Kramers equation [35,36,81,82,278,279]. Its solution, the pdf \( W(x,v,t) \) to find the test particle at the position \( x, \ldots, x + dx \) with the velocity \( v, \ldots, v + dv \), at time \( t \), describes the macroscopic dynamics of the system. Thereby, two limiting cases can be distinguished, these being the Rayleigh equation controlling the velocity distribution \( \rho(v,t) \) in the force-free limit, and the Fokker–Planck equation from which the pdf \( \rho(x,t) \) can be derived.

Fractional Fokker–Planck and Klein–Kramers equations have been derived and discussed for Lévy flights which are Markovian but possess a diverging mean squared displacement. Although the fractional Fokker–Planck equation can be derived from the generalised master equation or continuous time random walk models as shown in the preceding section, a foundation on microscopic dynamics within the Langevin picture sheds some light on the coming into existence of fractional dynamics as is briefly shown in this section.

The three stages of this model comprise the following steps: Firstly, the Newtonian motion of the scalar test particle experiencing a random force, in accordance to the Langevin equation

\[
m \frac{d^2x}{dt^2} = -m \eta v + F(x) + m \Gamma(t), \quad v = \frac{dx}{dt}
\]

with the \( \delta \)-correlated Gaussian noise \( \Gamma(t) \). Secondly, its combination with kinetic energy-conserving trapping events which are ruled by the broad waiting time statistics according to \( w(t) \sim \tau^\alpha/t^{1+\alpha} \). During a trapping event the particle is temporarily immobilised. And, thirdly, the macroscopic average in which the long-tailed trapping events win out in the competition with the Langevin motion events of average duration \( \tau^\star \), in the spirit of the generalised central limit theorem. This model offers some physical insight into the origin of fractional dynamics for systems which exhibit multiple trapping such as the charge carrier transport in amorphous semiconductors [95–97], or the phase space dynamics of chaotic Hamiltonian systems [280].

After straightforward calculations basing on the continuous time version of the Chapman–Kolmogorov equation [164,219] which are valid in the long-time limit \( t \gg \max\{\tau, \tau^\star\} \), one obtains the fractional Klein–Kramers equation [219,220]

\[
\frac{\partial W}{\partial t} = D_1^{1-\alpha} \left[ -v^* \frac{\partial}{\partial x} + \frac{\partial}{\partial v} \left( \eta^* v - F^*(x) \right) \frac{m}{m} \right] + \frac{\eta^* k_B T}{m} \frac{\partial^2}{\partial v^2} W(x,v,t).
\]

Hereby, the Klein–Kramers operator in the square brackets has the same structure as in the Brownian case, except for the occurrence of the starred quantities which are defined through \( v^* \equiv \partial \eta, \eta^* \equiv \eta \partial \), and \( F^*(x) \equiv F(x) \partial \) whereby the factor \( \partial \) is the ratio \( \partial \equiv \tau^\star/\tau^\star \) of the inter-trapping time scale \( \tau^\star \) and the internal waiting time scale \( \tau \). This rescaling automatically reveals the generalised Einstein relation \( K_s = k_B/\eta^* \) [219,220].
6.2. The fractional Klein–Kramers equation and related transport equations

Integration of Eq. (185) over velocity, and of $v$ times Eq. (185) over velocity results in two equations whose combination leads to the fractional equation [219,281]

$$\frac{\partial W}{\partial t} + D_1^{1-\alpha} \frac{1}{\eta^*} W = D_1^{1-\alpha} \left[ -\frac{\partial}{\partial x} F(x) + K x \frac{\partial^2}{\partial x^2} \right] W(x,t) .$$ (186)

Eq. (186) is of the generalised Cattaneo equation type [283–286] and reduces to the telegrapher’s type equation found in the Brownian limit $\alpha = 1$ [281]. In the usual high-friction or long-time limit, one recovers the fractional Fokker–Planck equation (101).

The integration of the fractional Klein–Kramers equation (185) over the position coordinate, leads in the force-free limit to the fractional Rayleigh equation [282]

$$\frac{\partial W}{\partial t} = D_1^{1-\alpha} \eta^* \left[ \frac{\partial}{\partial v} + \frac{k_B T}{m} \frac{\partial^2}{\partial v^2} \right] W(v,t) .$$ (187)

Its solution, the pdf $W(v,t)$, describes the equilibration of the velocity distribution towards the Maxwell distribution

$$W_{st}(v) = \frac{\beta m}{\sqrt{2\pi}} \exp \left( - \frac{\beta m}{2} v^2 \right) .$$ (188)

The presented model for subdiffusion in the external force field $F(x) = -V'(x)$ provides a basis for fractional kinetic equations, starting from Langevin dynamics which is combined with long-tailed trapping events possessing a diverging characteristic waiting time $T$. This combined process is governed by the long-tailed form of the waiting time pdf, manifested in the fractional nature of the associated Eq. (185). Physically, this causes the rescaling of the fundamental quantity $\eta$ by the scaling factor $\partial$, to result in the generalised friction constant

$$\eta_\alpha = \eta/\partial .$$ (189)

It is interesting to note that a similar process depicting a force-free trapping-walk scenario on a kinematics level was described in Ref. [287] in $(x,t)$-coordinates, revealing the subdiffusive mean squared displacement $\langle x^2(t) \rangle \propto t^\alpha$.

The Langevin picture rules the Markov motion parts in between successive trapping states. On this stage the test particle consequently obeys to Newton’s law, in the noise-averaged sense defined above. Conversely, averaging the fractional Klein–Kramers equation (185) over velocity and position coordinates, one recovers the memory relation $(d/dt) \langle x(t) \rangle = \partial_0 D_1^{1-\alpha} \langle v(t) \rangle$ between the mean position $\langle x(t) \rangle$ and the mean velocity $\langle v(t) \rangle$. This “violation” is only due to the additional waiting time averaging which camouflages the Langevin-dominated motion events.

7. Conclusions

Roughly a hundred years have elapsed since the advent of random walk and diffusion theory. The success of the framework developed by its most important contributor, Albert Einstein, obtained an additional thrust when the experimentalist Perrin came up so successfully with his
determination of the Avogadro–Loschmidt number. The ideas of how random walks can be used as a model for the transport dynamics in physical systems are still the same, and have become a joint venture of mathematicians, physicists, chemists, engineers, earth and life scientists. The extension of random walk theory to incorporate generalised statistics which no longer follow the central limit theorem, and memory effects violating the Markovian nature of early days random walks, has created a very rich tool, rich enough to be able to describe certain features of complex systems.

Random walks are a very convenient tool in drawing a physical picture of the processes underlying the dynamics of systems which are of a probabilistic nature. To deal with problems involving boundary values or external fields, the differential equation approach is more convenient, however. In this report we have demonstrated the versatility of fractional equations of the diffusion, diffusion–advection, and Fokker–Planck type, generalising their standard counterparts. Appropriate methods of solutions such as the Fourier–Laplace technique, the method of images, and the method of separation of variables have been discussed. Mellin transformation techniques, term-by-term inversion, as well as the method of characteristics have been discussed elsewhere. We presented an integral transformation between the Brownian solution and its fractional counterpart. Moreover, a phase space model was discussed which explains the genesis of fractional dynamics in trapping systems. These issues make the fractional equation approach a powerful instrument.

Note that whereas we showed the solution of the FDE for simple boundary values, the opposite problem, a standard equation with fractal boundary conditions, has also received some interest [288–293]. With the fractional approach it should be possible to discuss such type of boundary value problems also for anomalous diffusion.

The main emphasis in the present work has been laid on subdiffusion. A basic feature arising in that context is the replacement of the exponential decay of modes by the Mittag–Leffler pattern. This feature involves the slow decay of the initial condition, slower dispersion, memory effects, and consequently a relatively slow approaching of the stationary state. Such strange dynamics have recently been discovered in protein systems [294]. In fact, it is supposed that dispersive kinetics are responsible for relaxation processes in protein systems [295, 296].

A further advantage of the fractional approach is the relatively straightforward way of calculating the moments. Being very simple for the CTRW approach in force-free diffusion, this advantage is obvious in such cases where a non-linear external force acts upon the test particle. Thereby, integration over the underlying fractional equation produces an ordinary fractional differential equation from which the moments can be inferred.

It has been shown that subdiffusive phenomena described by the FFPE are close to thermal equilibrium which is reached via the aforementioned Mittag–Leffler pattern. The calculation of the generalised Einstein–Stokes–Smoluchowski relation, and the validity of the second Einstein relation in a way justify the use of the subdiffusive FFPE as direct generalisation of the Brownian analogue. Both phenomena have been corroborated experimentally for subdiffusive systems.

Note added in proof

Recently, Barkai and Silbey have presented the fractional Klein–Kramers equation [66]

\[
\frac{\partial W(x, v, t)}{\partial t} + v \frac{\partial W}{\partial x} + \frac{F(x)}{m} \frac{\partial W}{\partial v} = \eta_0 D_{t^{1-a}} L_{FP} W(x, v, t)
\] (190)
for $0 < \alpha < 1$, with the Fokker–Planck operator

$$L_{FP} \equiv \frac{\partial}{\partial v} v + \frac{K_B T}{m} \frac{\partial^2}{\partial v^2}$$

(191)

and $[\eta_s] = s^{-\alpha}$. This equation corresponds to the force-free mean squared displacement

$$\langle x^2(t) \rangle = \frac{2k_B T}{m} t - t^2 E_{\alpha,3} (1 - \eta_s t^2) \sim \begin{cases} \frac{k_B T}{m} t^2, & \eta_s t^2 \ll 1 \\ \frac{2k_B T}{\eta_s m} t^{2-\alpha}, & \eta_s t^2 \gg 1 \end{cases}$$

(192)

and thus describes the transition from ballistic to sub-ballistic superdiffusion. Further studies of this equation, especially on the non-negativity of the pdf are under way.

We would further like to point out that a fractional Fokker–Planck equation was proposed by Jumarie [231], and that non-integer order spatial derivatives were presented by Onuki [246] and Suzuki [306]. A Lévy approach to quantum pdfs has been reported by West [307]. A first passage time study in anomalous diffusion on the basis of the FFPE has been accomplished by Rangarajan and Ding [308]; we thank these authors for sending us a set of unpublished manuscripts.

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Appendix A. Fractional differentiation and integration

As early as in 1695, when Leibniz and Newton had just been establishing standard calculus, Leibniz wrote in a letter to de l'Hospital: “Thus it follows that $d^{1/2}x$ will be equal to $x^{2/\sqrt{dx}}$ an apparent paradox, from which one day useful consequences will be drawn.” In the course of time, many famous mathematicians worked on this and related questions, creating the field which is known as fractional calculus today. The list of prominent names includes Laplace, Lacroix, Cayley, Pincherle, Holmgren, Grünwald, Krug, Heaviside, Laurent, Riemann, Liouville, Hardy, Hadamard, Lévy, Weyl, Erdélyi and Kober, thus indicating the relevance of the topic in the mathematicians’ point of view.
Throughout this report, we employ the so-called Riemann–Liouville fractional integral defined through [226,232–235]

\[ t_0 D^\alpha_t f(t) = \frac{1}{\Gamma(\alpha)} \int_{t_0}^t \frac{f(t')}{(t - t')^{1-\alpha}} \, dt' , \]  
\[ \text{(A.1)} \]

a direct extension of Cauchy’s multiple integral for arbitrary complex \( \alpha \) with \( \text{Re}(\alpha) > 0 \). A fractional derivative is then established via a fractional integration and successive ordinary differentiation according to

\[ t_0 D^\beta_t f(t) = \frac{d^n}{dt^n} t_0 D^\beta n f(t) \]  
\[ \text{(A.2)} \]

with \( \text{Re}(\beta) > 0 \), and the natural number \( n \) satisfies the inequality \( n \geq \text{Re}(\beta) > n - 1 \). Through this definition it is clear that the fractional differentiation is non-local. As in standard calculus, integration and differentiation are not commutative, i.e.,

\[ \frac{d}{dx} \int_0^x dt f(t) = f(t) , \quad \int_0^x dt \frac{d}{dx} f(t) = f(x) + c , \]  
\[ \text{(A.3)} \]

certain composition rules for successive fractional operations have to be formulated, see below.

In respect to the multiple Cauchy integral, a fractional integral may be viewed as a non-integer multiple integral, i.e., an \( \alpha \)-fold integral with \( \alpha \) a real, or even complex number. This might seem an odd statement; however, the same paradox is encountered in the notion of a fractal dimension which is a generally accepted mathematical notion. Indeed, there exist tight relations between fractional differentials and fractal geometry dimensions, as can be anticipated by the relation (A.16) established below, which demonstrates that \( t_0 D^\alpha_t \) is a map relating power-laws of different indices. Fractional differentials even serve as sensors for the fractal graph dimensions of certain functions [297,298].

Two special cases are of importance in this report, these being the Riemann–Liouville operator \( t_0 D^\alpha_t \) for \( t_0 = 0 \), and the Weyl operator \( -\infty D^\alpha_x \) for \( t_0 = -\infty \) in Eq. (A.1). Mathematically, the expression \( t_0 D^\alpha_t f(t) = (1/\Gamma(\alpha))t^{\alpha-1}f(t) \) is a Laplace convolution, whereas \( -\infty D^\alpha_x f(x) = (1/\Gamma(\mu))x^{\mu-1}f(x) \) represents a Fourier convolution. Therefore, Laplace and Fourier transformations will be a useful tool in solving fractional order differential and integrodifferential equations.

### A.1. The Riemann–Liouville fractional operator

The Riemann–Liouville operator, \( t_0 D^\alpha_t \), defined through

\[ t_0 D^\alpha_t f(t) = \frac{1}{\Gamma(n - p)} \frac{d^n}{dt^n} \int_{t_0}^t \frac{f(t')}{(t - t')^{1-n+p}} \, dt' , \]  
\[ \text{(A.4)} \]

for \( n \geq \text{Re}(p) > n - 1 \), obeys the following theorem for Laplace transformation:

\[ \mathcal{L}\{ t_0 D^\alpha_t f(t) \} = u^{-p}f(u) , \]  
\[ \text{(A.5)} \]
$q > 0$, which is a direct generalisation of the integral theorem for integer-order integrals. Conversely, the differentiation theorem is modified to the form [232]

$$
\mathcal{L}'\{ 0D_q^p f(t) \} = u^p f(u) - \sum_{j=0}^{n-1} u^j c_j
$$

(A.6)

which involves the quasi initial value terms

$$
c_j = \lim_{t \to 0} 0D_q^{p-1-j}f(t).
$$

(A.7)

Finally, the Riemann–Liouville operator obeys the following composition rules [232]:

$$
0D_q^p 0D_q^Q f(t) = 0D_q^{p+Q} f(t) \quad \text{if} \quad Q < 0 \vee Q < 1 \quad \text{and} \quad f(0) \text{ bounded} , \quad (A.8)
$$

$$
0D_q^p 0D_q^Q f(t) = 0D_q^{p+Q} \left( f(t) - \sum_{j=1}^{m} c_j t^{Q-k} \right), \quad Q > 0
$$

(A.9)

with

$$
c_j = \lim_{t \to 0} 0D_q^{Q-j}f(t) / \Gamma(Q-j+1).
$$

(A.10)

A.2. The Riesz/Weyl fractional operator

The Weyl fractional operator $-\infty D_x^u$ has a simpler behaviour under transformations, as due to $t_0 = -\infty$, no initial values come into play. Thus, its Fourier transform is

$$
\mathcal{F} \{-\infty D_x^u f(x)\} = (ik)^u f(k), \quad (A.11)
$$

and the composition rules become

$$
-\infty D_x^\mu -\infty D_x^\nu = -\infty D_x^{\mu+\nu} \quad \forall \mu, \nu. \quad (A.12)
$$

Note that, in the main text, we preferred the simpler notation suppressing the imaginary unit, defined through

$$
\mathcal{F} \{-\infty D_x^u f(x)\} \equiv -|k|^u f(k)
$$

(A.13)

which has somehow established in fractional applications, for instance compare [192]. In one dimension, the Weyl operator is equivalent to the Riesz operator which preserves the property (A.13) to higher dimensions. We refer to the symbol $-\infty D_x^u$ as the Riesz/Weyl operator.

A.3. Differintegrable functions and an equivalent definition

Following Oldham and Spanier [232], we define the class of differintegrable functions as all functions $f(t)$ which can be expanded as a differintegrable series according to

$$
f(t) = (t - \tau)^p \sum_{j=0}^{\infty} f_j(t - \tau)^{j/n}
$$

(A.14)
where $\tau \not= 0$, $n$ is a natural number, and $p > -1$. Most of the special functions used in mathematical physics are subsumed under this definition. The definition (A.14) of differintegrable functions actually is very strict, and many functions or distributions like the Heaviside jump distribution or the Dirac delta distribution can be differintegrated to fractional order.

A definition equivalent to the Riemann–Liouville fractional integral (A.1) for $\text{Re}(\alpha) > 0$ is given through the limit [232]

$$
_0D_t^{-\alpha}f(t) \equiv \lim_{N \to \infty} \left[ \frac{(t - t_0)^{\alpha N - 1}}{N} \sum_{j=0}^{\alpha N - 2} \frac{\Gamma(j + \alpha)}{\Gamma(\alpha)j!}f\left(\frac{Nt - j\tau + j\tau}{N}\right)\right].
$$

(A.15)

Differently to Eq. (A.1), this definition also holds for fractional differentiation, i.e., $\text{Re}(\alpha) < 0$. Again, the non-local character becomes clear.

A.4. Examples

The fractional Riemann–Liouville differintegration of an arbitrary power for $t_0 = 0$ is given by

$$
_0D_t^{\mu}t^\mu = \frac{\Gamma(1 + \mu)}{\Gamma(1 + \mu - \nu)} t^{\mu - \nu}
$$

(A.16)

which coincides with the heuristic generalisation of the standard differentiation

$$
\frac{d^mt^m}{dt^n} = \frac{m!}{(m-n)!} t^{m-n}
$$

(A.17)

by introduction of the Gamma function. An interesting consequence of Eq. (A.16) is the non-vanishing fractional differintegration of a constant:

$$
_0D_t^11 = \frac{1}{\Gamma(1 - \nu)} t^{-\nu}.
$$

(A.18)

The Riemann–Liouville differintegration of the exponential function leads to

$$
_0D_t^{\nu}e^t = \frac{t^{-\nu}}{\Gamma(1 - \nu)} \, _1F_1(1, 1 - \nu, t)
$$

(A.19)

involving the confluent hypergeometric function \( _1F_1 \) [236]. This result can be found easily by differentiating term by term in the exponential series according to Eq. (A.16). On the other hand, for the Weyl fractional operator \(-_\alpha D_t^\nu\), the fundamental property of the exponential function, i.e. \((e^t)^{\nu} = e^\nu\) carries over to fractional orders:

$$
_\alpha D_t^\nu e^t = e^\nu.
$$

(A.20)

A.5. The singular nature of the fractional operator

Some caution has to be paid to the singularity at $t = t'$. Consider the derivation of Eq. (A.16). With the definition

$$
B(b + 1, d + 1) x^{b+d+1} \equiv \frac{\Gamma(b + 1)\Gamma(d + 1)}{\Gamma(b + d + 2)} x^{b+d+1} = \int_0^x dt (x - t)^b t^d = \int_0^x dt t^d (x - t)^d
$$

(A.21)
of Euler’s Beta function $B$, the Riemann–Liouville differintegration of the power-law $t^n$ can be achieved. The definition of the Beta function is valid for $b, d > 1$ only. The fractional differintegration of order $1 - \gamma$, $0 < \gamma < 1$, of the power $x^p$ is then

$$
o D_x^{1-\gamma} x^p = \frac{d}{dx} \frac{1}{\Gamma(\gamma)} \int_0^x dt (x - t)^{-1+\gamma} t^p = \frac{d}{dt} \frac{1}{\Gamma(\gamma)} B(\gamma, p + 1) x^{1-\gamma + p}
$$

$$
= \frac{\Gamma(p + 1)}{\Gamma(p + \gamma)} x^{p-1+\gamma}.
$$

(A.22)

It is tempting to evaluate the parametric differintegration $d/dx$ explicitly, according to

$$
\frac{d}{dx} \int_0^x dt f(x, t) = f(x, x) + \int_0^x dt \frac{\partial f(x, t)}{\partial x}.
$$

(A.23)

This leads to a term with a pole and a second term containing an integral which violates the definition of the Beta function. On closer inspection, this integral itself has a pole which in some way compensates the pole of the first term. In fact, a similar situation arises for Abel’s integral equation:

$$
x f(x) = \int_0^x dt (x - t)^{-1/2} f(t)
$$

(A.24)

solved by $f(x) = x^{-3/2} e^{-x/x}$. Whereas the differentiation of the left-hand side is no problem, on the right two singularities “compensate” each other. The definition of fractional differintegrals thus involves singular integrals. However, the mathematical framework is well defined if caution according to above considerations is paid.

Appendix B. Special functions: Mittag–Leffler and Fox functions

B.1. The Mittag–Leffler function

The Mittag–Leffler function [299,300] is the natural generalisation of the exponential function. Being a special case of the Fox function introduced below, it is defined through the inverse Laplace transform

$$
E_a( - (t/\tau)^\alpha) = \mathcal{L}^{-1} \left\{ \frac{1}{u + \tau^{-\alpha} u^{1-\alpha}} \right\},
$$

(B.1)

from which the series expansion

$$
E_a( - (t/\tau)^\alpha) = \sum_{n=0}^\infty \frac{( - (t/\tau)^\alpha)^n}{\Gamma(1 + \alpha n)}
$$

(B.2)

can be deduced. The asymptotic behaviour is

$$
E_a( - (t/\tau)^\alpha) \sim (t/\tau)^\alpha \Gamma(1 - \alpha)^{-1}
$$

(B.3)
Fig. 22. Mittag-Leffler relaxation. The full line represents the Mittag-Leffler function for index 1/2. The dashed lines demonstrate the initial stretched exponential behaviour and the final inverse power-law pattern.

for $t \gg \tau$, $0 < \alpha < 1$. Special cases of the Mittag-Leffler function are the exponential function

$$E_1(-t/\tau) = e^{-t/\tau} \tag{B.4}$$

and the product of the exponential and the complementary error function

$$E_{1/2}(-(t/\tau)^{1/2}) = e^{t/\tau} \text{erfc}(t/\tau)^{1/2}). \tag{B.5}$$

We note in passing that the Mittag-Leffler function is the solution of the fractional relaxation equation [20,21,180]

$$\frac{d\Phi(t)}{dt} = -\tau\, D_t^{\alpha} \Phi(t). \tag{B.6}$$

As already mentioned in the main text, and by Glöckle and Nonnenmacher [21], the Mittag-Leffler function interpolates between the initial stretched exponential form

$$E_\alpha(-(t/\tau)^\alpha) \sim \exp\left(-\frac{(t/\tau)^\alpha}{\Gamma(1+\alpha)}\right) \tag{B.7}$$

and the long-time inverse power-law behaviour (B.3). The Mittag-Leffler function $E_{1/2}(-(t/\tau)^{1/2})$ is displayed in Fig. 22.

B.2. The Fox function

The Fox function, also referred to as Fox’s $H$-function, $H$-function, generalised Mellin–Barnes function, or generalised Meijer’s $G$-function, has been applied in statistics before it was introduced
According to Ref. [243], these functions have been known since at least 1868, and Fox rediscovered them in his studies. Their introduction into physics is due to Bernasconi et al. [155] in the study of conductivity in disordered systems. Schneider [301] demonstrated that Lévy stable densities can be expressed analytically in terms of Fox functions. Wyss [174], and Schneider and Wyss [175] use Fox functions for the solution of the fractional diffusion equation.

In 1961 Fox defined the H-function in his studies of symmetrical Fourier kernels as the Mellin–Barnes type path integral [237, 240–243]:  

\[ H_{pq}^m(z) = H_{pq}^m \left[ z \left( \frac{(a_p, A_p)}{(b_q, B_q)} \right) \right] = H_{pq}^m \left[ z \left( (a_1, A_1), (a_2, A_2), \ldots, (a_p, A_p) \right) \right] = \frac{1}{2\pi i} \int_L ds \chi(s) z^s \]  

with the integral density:

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

Note that the path integral in Eq. (B.8) represents just the inverse Mellin transform of \( \chi(s) \) [237].

Due to the structure of the defining integral kernel \( \chi(s) \) from Eq. (B.9), the Fox functions fulfil several convenient properties, three of which we list here:

**Proposition B.1.** For \( k > 0 \)

\[ H_{pq}^m \left[ x \left( \frac{(a_p, A_p)}{(b_q, B_q)} \right) \right] = k H_{pq}^m \left[ x \left( \frac{(a_p, kA_p)}{(b_q, kB_q)} \right) \right] . \]

**Proposition B.2.**

\[ x^\alpha H_{pq}^m \left[ x \left( \frac{(a_p, A_p)}{(b_q, B_q)} \right) \right] = H_{pq}^m \left[ x \left( \frac{(a_p + \sigma A_p, A_p)}{(b_q + \sigma B_q, B_q)} \right) \right] . \]

**Proposition B.3.** The fractional differintegral of the Fox function is a map into the function class [180, 237]:

\[ (az)^\beta D^{\alpha\beta}_z \left[ z \left( \frac{(a_p, A_p)}{(b_q, B_q)} \right) \right] = z^{\alpha - \nu} H_{pq}^{m,n+1} \left[ \left( \frac{(a_p, A_p)}{(b_q, B_q)} \right) \right] . \]
An $H$-function can be expressed as a computable series in the form [237,242]

$$
H_{pq}^{mn}(z) = \sum_{k=1}^{m} \sum_{v=0}^{\infty} \left[ \prod_{j=1, j \neq k}^{m} \Gamma(b_j - B_j(b_h + v)/B_h) \prod_{j=m+1}^{n} \Gamma(1 - b_j + B_j(b_h + v)/B_h) \prod_{j=m+1}^{n} \Gamma(a_j - A_j(b_h + v)/B_h) \times \frac{(-1)^v z^{(b_h + v)/B_h}}{v! B_h} \right] (B.13)
$$

which is an alternating series and thus shows slow convergence. For large argument $|z| \to \infty$, the Fox functions can be expanded as a series over the residues [241]

$$
H_{pq}^{mn}(z) \sim \sum_{v=0}^{\infty} \text{res}(\chi(z)z^v)
$$

(B.14)

to be taken at the points $s = (a_j - 1 - v)/A_j$, for $j = 1, \ldots, n$.

Some special functions and their Fox function representation:

$$
z^b e^{-z} = H_0^1,0 \left[ z \left| \frac{1}{b, 1} \right. \right];
$$

(B.15)

$$
\frac{1}{1 + z} = H_1^1,1 \left[ \frac{1}{1 - r, 1} \right];
$$

(B.16)

$$
\frac{z^\beta}{1 + az^x} = a^{-\beta/2} H_1^1,1 \left[ az^x \left| \frac{\beta / x, 1}{(\beta / x, 1)} \right. \right].
$$

(B.17)

Maitland’s generalised hypergeometric or Wright’s function:

$$
p_{\psi q} \left[ (a_1, A_1), \ldots, (a_p, A_p), (b_1, B_1), \ldots, (b_q, B_q) \right] - z \right] = H_{1:p}^{1:p+1} \left[ z \left| (1 - a_1, A_1), \ldots, (1 - a_p, A_p) \right| (0, 1), (1 - b_1, B_1), \ldots, (1 - b_q, B_q) \right].
$$

(B.18)

Generalised Mittag–Leffler function ($E_{x,1}(z) = E_z(z)$):

$$
E_{x,\beta}( - z) = H_1^1,1 \left[ z \left| (0, 1) \right| (0, 1), (1 - \beta, x) \right] = \sum_{j=0}^{\infty} \frac{z^j}{\Gamma(\beta + xj)}.
$$

(B.19)

Appendix C. Some remarks on Lévy distributions and their exact representation in terms of Fox functions

The fundamental importance of the normal distribution is due to the Central Limit Theorem which, within the history of probability theory, is a consequence of the inequality of Bienaymé, the theorems of Bernoulli and de Moivre–Laplace, and the law of large numbers. These concepts were extended by the works of Paul Lévy, after which the generalised normal distributions are named,
As remarked by Takayasu [91], this property may be viewed as some kind of self-similarity.

According to Lévy [57], a distribution \( F \) is stable iff for the two positive constants \( c_1 \) and \( c_2 \) there exists a positive constant \( c \) such that \( X \) given by

\[
c_1 X_1 + c_2 X_2 = cX
\]

is a random variable following the same distribution \( F \) as the independent, identically distributed (iid) random variables \( X_1 \) and \( X_2 \). Alternatively, if

\[
\varphi(z) \equiv \langle e^{ixz} \rangle = \int_{-\infty}^{\infty} e^{ixz} dF(X)
\]

(C.2)

denotes the characteristic function of the distribution \( F \), then \( F \) is stable iff

\[
\varphi(c_1 z)\varphi(c_2 z) = \varphi(cz).
\]

(C.3)

A more general definition is given by Feller [88]. Let \( X, X_1, X_2, \ldots, X_n \) be iid random variables with a common distribution \( F \). Then \( F \) is called stable iff there exist constants \( c_n > 0 \) and \( \gamma_n \) such that

\[
Y_n \equiv \sum_i X_i \overset{d}{=} c_n X + \gamma_n
\]

(C.4)

where \( \overset{d}{=} \) indicates that the random variables of both sides follow the same distribution, \( F \). Consequently, the characteristic function according to definition (C.3) fulfills the functional relation

\[
\varphi^n(z) = \varphi(c_n z)e^{i\gamma_n z},
\]

(C.5)

which can be solved exactly. The result is

**Proposition C.1.**

\[
\psi(z) = \log \varphi(z) = iz\gamma - c|z|^\alpha \left\{ 1 + i\beta \frac{z}{|z|} \omega(z, \alpha) \right\},
\]

where \( \alpha, \beta, \gamma, c \) are constants (\( \gamma \) is any real number, \( 0 < \alpha \leq 2, -1 < \beta < 1, \) and \( c > 0 \)), and

\[
\omega(z, \alpha) = \begin{cases} \frac{\tan \frac{\pi \alpha}{2}}{2} & \text{if } \alpha \neq 1, \\ \frac{\log|z|}{\pi} & \text{if } \alpha = 1. \end{cases}
\]

(C.7)

\( \alpha \) is called the Lévy index or characteristic exponent. From Eq. (C.6) it can be shown that the normalisation factor \( c_n \) in Eq. (C.3) is \( n^{1/\alpha} \). The limiting case \( \alpha = 2 \) corresponds to the Gaussian

\[\text{77}\]

\[\text{65}\]
normal distribution governed by the central limit theorem. For \( \beta = 0 \), the distribution is symmetric. \( \gamma \) translates the distribution, and \( c \) is a scaling factor for \( X \). Thus, \( \gamma \) and \( c \) are not essential parameters, and disregarding them, the characteristic function satisfies the following proposition.

**Proposition C.2.** \( |\varphi(z)| = e^{-|z|^\alpha}, \quad \alpha \neq 1 \). \hspace{1cm} (C.8)

Thus, one can write

\[
\psi(z) = -|z|^\alpha \exp\left\{i \frac{\pi \beta}{2} \operatorname{sign}(z) \right\}
\]

with the new centring constant \( \beta \) which is restricted in the following region:

\[
|\beta| \leq \begin{cases} 
\alpha, & \text{if } 0 < \alpha < 1 \\
2 - \alpha, & \text{if } 1 < \alpha < 2
\end{cases}
\] \hspace{1cm} (C.10)

The resulting allowed parameter space is portrayed in Fig. 23.

**Proposition C.3.** The pdf \( f_{\alpha,\beta}(x) \) is the Fourier transform of \( \varphi(z) \), defined by Eq. (C.9):

\[
f_{\alpha,\beta}(x) = \frac{1}{\pi} \operatorname{Re} \int_0^\infty \exp\left( -ixz - z^\alpha \exp\left\{i \frac{\pi \beta}{2}\right\} \right) dz.
\] \hspace{1cm} (C.11)
Thus,
\[ f_{a,b}(x) = f_{a,-b}(-x) \]  
(C.12)
and consequently
\[ f_{a,0}(x) = f_{a,0}(-x) \]  
(C.13)
is symmetric in $x$.

**Proposition C.4** (Lévy–Gnedenko generalised central limit theorem). For iid random variables $X_1, X_2, \ldots$ let $Y_n = \sum_{i=1}^{n} X_i$. If the distribution of $Y_n$, with an appropriate normalisation, converges to some distribution $F$ in the limit $n \to \infty$, $F$ is stable. In particular, if its variance is finite, $F$ is Gaussian, and obeys to the central limit theorem.

**Proposition C.5.** The asymptotic behaviour of a Lévy stable distribution follows the inverse power-law
\[ f_{a,b}(x) \sim \frac{A_{a,b}}{|x|^{1+\alpha}}, \quad \alpha < 2. \]  
(C.14)

**Proposition C.6.** For all Lévy stable laws with $0 < \mu < 2$, the variance diverges:
\[ \langle x^2 \rangle \to \infty \]  
(C.15)

**Proposition C.7.** Conversely, the fractional moments of the absolute value of $x$,
\[ \langle |x|^\delta \rangle < \infty \]  
(C.16)
exist for any $0 \leq \delta < \mu \leq 2$.

**Proposition C.8.** The analytic form of a stable law is given through the Fox function [301]
\[ f_{a,b}(x) = \varepsilon x^{H_{\frac{1}{2}, \frac{1}{2}} \left[ 1 - \frac{1}{2}, 1 - \gamma, \gamma \right] \left[ 0, 1, 1 - \gamma, \gamma \right]} \]  
for $\alpha > 1$, and with the abbreviations $\varepsilon = 1/\alpha$ and $\gamma = (\alpha - \beta)/2\alpha$; for $\alpha < 1$, one obtains the result
\[ f_{a,b}(x^{-1}) = \varepsilon x^{2} H_{\frac{1}{2}, \frac{1}{2}} \left[ 1 - \frac{1}{2}, 1 - \gamma, \gamma \right] \left[ -1, 1, 1 - \gamma, \gamma \right] \]  
(C.18)

Some reductions of these Fox representations for special cases of $\beta$ are discussed by Schneider [301]. From the known theorems of the Fox function, the series representations and the asymptotic behaviour can be determined, thus, for $\alpha > 1$, one obtains
\[ f_{a,b}(x) = \frac{1}{\pi} \sum_{n=1}^{\infty} \frac{I(1 + n\varepsilon)}{n!} \sin(n\pi\gamma)(-x)^{n-1}, \]  
(C.19)
This fact, as well as the striking similarity between results (C.17) and (C.18) are based on the fundamental property

\[ H_{M,H} \left( \frac{(a_p,A_p)}{(b_q,B_q)} \right) = H_{Q,P} \left[ \frac{1}{x} \right] \left( 1 - b_q, B_q \right) \left( 1 - a_p, A_p \right). \]

and the asymptotic behaviour

\[ f_{x,\beta}(x) \sim \frac{1}{\pi} \sum_{n=1}^{\infty} \frac{\Gamma(1 + nz)}{n!} \sin(\pi nx) |x|^{-1-nz} \]  \hspace{1cm} (C.20)

for \( |\beta| > \alpha - 2 \). An exception is the case \( \beta = \alpha - 2 \). Conversely, for \( \alpha < 1 \), the expansions for large and small \( x \) are given by Eqs. (C.19) and (C.20), respectively, except the case \( |\beta| = \alpha \).

**Proposition C.9.** For \( \alpha = 2 \), \( \beta = 0 \), and the stable density is identical to the Gaussian normal distribution.

**Proposition C.10.** For \( \alpha = 1 \) and \( \beta = 0 \), the stable density is identical to the Cauchy or Lorentz distribution

\[ f_{1,0}(x) = \frac{a}{\pi(a^2 + x^2)} . \]  \hspace{1cm} (C.21)

**Proposition C.11.** If \( 0 < \alpha < 1 \) and \( \beta = -\alpha \), the pdf \( f_{x,-\alpha}(x) = 0 \ \forall x < 0 \) is one-sided. For instance, the one-sided stable density for \( \alpha = 1/2 \) and \( \beta = -1/2 \) is given by

\[ f_{1/2,-1/2}(x) = \frac{1}{2\sqrt{\pi}} x^{-3/2} e^{-1/4x} . \]  \hspace{1cm} (C.22)

**Proposition C.12.** For not too small and not too large \( x \), as well as \( \alpha \approx 0 \), the pdf \( f_{x,-\alpha}(x) \) can be approximated by a log-normal distribution [302]:

\[ f_{x,-\alpha}(x) \propto \frac{1}{x} \exp \left( -\frac{x^2}{2} (\log x)^2 \right) . \]  \hspace{1cm} (C.23)

For \( \alpha = 3/2 \) and \( \beta = 0 \), one recovers the Holtsmark distribution which is of some use in cosmology [88]. The above-mentioned special cases are included in the phase diagram, Fig. 23. Further examples can be found in the article of Schneider [301], and in the book of Feller [88].

In Fig. 24, the Gaussian normal distribution is compared to the Lévy stable law \( f_{1,0}(x) \), the Cauchy distribution.

A more recent monograph dealing with stable distributions is the book by Samorodnitsky and Taqqu [303]. Some additional historical remarks are to be found in the textbook by Johnson et al.  

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21 This fact, as well as the striking similarity between results (C.17) and (C.18) are based on the fundamental property [237,242]
Fig. 24. Comparison of the Gaussian normal pdf \( \pi^{-1/2}e^{-x^2} \) (dashed) and the Cauchy pdf \( \pi^{-1}(1 + x^2)^{-1} \), the Lévy stable law for Lévy index \( \mu = 1 \). Both are normalised to unity. Note that the concentration around zero is much more pronounced for the Gaussian. The insert shows the double-logarithmic plot of the same functions, pronouncing the slower power-law decay of the Cauchy pdf which turns over to the straight line in this presentation.

Fig. 25. A roaming Brownian particle P moves along a line, and radiates light. On the moving film F, the particle leaves a mark each time it passes the slit. The time intervals between single marks on the film is given by the one-sided distribution \( f_{1/2,-1/2}(\tau) \).

[304]. An interesting and readable summary of stable laws is given by Takayasu [91]. He also mentions a physical example, see Fig. 25, equivalent to the Pólya problem [38,305]. It leads to the one-sided pdf \( f_{1/2,-1/2}(\tau) \), Eq. (C.22) for the distribution of the time span \( \tau \) between individual signals. An application might be in single molecule spectroscopy [138,139].

Appendix D. Abbreviations used

pdf \hspace{1cm} \text{probability density function}

CTRW \hspace{1cm} \text{continuous time random walk}
FDE fractional diffusion equation
DAE diffusion–advection equation
FDAAE fractional diffusion–advection equation
FPE Fokker–Planck equation
FFPE fractional Fokker–Planck equation

KM-expansion Kramers–Moyal expansion
FKKE fractional Klein–Kramers equation

\( \psi(x, t) \) pdf of just arriving at position \( x \) at time \( t \)
\( W(x, t) \) pdf of being at \( x \) at time \( t \)
\( \psi(x, t) \) jump pdf, considered in the decoupled form \( \psi(x, t) = w(t)\hat{\lambda}(x) \)
\( w(t) \) waiting time pdf
\( \hat{\lambda}(x) \) jump length pdf
\( \Psi(t) \) sticking probability, Eq. (24)
\( K_{x} \) generalised diffusion coefficient, subdiffusion
\( K^{a} \) generalised diffusion coefficient, Lévy flights
\( \Lambda(x', x) \) transfer kernel
\( W_{0}(x) \) initial condition \( W_{0}(x) \equiv \lim_{t \to 0^{+}} W(x, t) \)
\( W_{st}(x) \) stationary solution \( W_{st}(x) \equiv \lim_{t \to \infty} W(x, t) \)
\( E_{a}(z) \) Mittag–Leffler function
\( H_{p,q}^{m,n}(z) \) Fox’s \( H \)-function
\( \beta \equiv (k_{B} T)^{-1} \) Boltzmann factor
\( m \) mass of test particle
\( \eta_{z} \) generalised friction coefficient

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[152] E. Barkai, J. Klafter, in [55].