

Ergodicity breaking and localization

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We study the joint action of the non-Poisson renewal events (NPR) yielding Continuous-time random walk (CTRW) with index $\alpha < 1$ and two different generators of Hurst coefficient $H \neq 0.5$, one generating fractional Brownian motion (FBM) and another scaled Brownian motion (SBM). We discuss the ergodicity breaking emerging from these joint actions and we find that in both cases the adoption of time averages leads to localization. In the case of the joint action of NPR and SBM, localization occurs when SBM would produce subdiffusion. The joint action of NPR and FBM, on the contrary, may lead to localization when FBM is a source of superdiffusion. The joint action of NPR and FBM is equivalent to extending the CTRW to the case where the jumps of the runner are correlated and we argue that the memory-induced localization requires a refinement of the theoretical perspective about determinism and randomness.

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I. INTRODUCTION

The proper evaluation of averages in statistical physics requires use of the Gibbs idealization of infinitely many copies of the same system, namely the Gibbs ensemble that is of fundamental importance for the theoretical predictions. In practice, when only the observation of a single system is possible, this ideal ensemble average is assumed to be identical to the more accessible average in time on the same system. This is the ergodic assumption [1] that the tracking of single molecules proves to be frequently violated [2], a phenomenon called *ergodicity breaking*. Establishing a good theoretical command of ergodicity breaking is of fundamental importance when we move from physics to biology, to physiology, and to sociology, since ergodicity breaking seems to be a general property of these complex systems.

A. Continuous time random walk

It has to be stressed, however, that one of the origins of the ergodicity breaking observed in molecular diffusion in biological cells is the occurrence of non-Poisson renewal (NPR) events, which generate nonstationary correlation functions, characterized by aging [2] in a form that is not to be confused with the nonstationarity of the slow approach to thermodynamical equilibrium.

This form of ergodicity breaking was very clearly illustrated in Refs. [3,4]. The diffusion process of a molecule that as a result of ensemble average would yield $\langle x^2(t) \rangle \propto t^\alpha$ with $\alpha < 1$, namely the subdiffusional scaling $\eta = \alpha/2$, was studied by the authors of Refs. [3,4], making averages in time rather than on infinitely many copies of the same system. These authors showed that the second moment evaluated in time does not yield the scaling $\alpha/2$. They got instead the

result [5]

$$\langle \overline{\delta^2(\Delta)} \rangle = \frac{1}{L - \Delta} \int_0^{L-\Delta} \langle [x(t) - x(t + \Delta)]^2 \rangle dt \propto \frac{\Delta}{L^{1-\alpha}}, \quad (1)$$

where L is the length of the time series analyzed. This impressive result indicates that the adoption of a time average turns a subdiffusion into a normal diffusion process with the surprising property, however, that the intensity of the second moment becomes weaker and weaker on increasing the length of the time series. The same kind of dependence on the length of the time series holds true for the power spectrum $P(f)$ of the renewal generators of subdiffusion, very well described [6,7] by

$$P(f) \propto \frac{1}{f^{2-\alpha}} \frac{1}{L^{1-\alpha}}. \quad (2)$$

Notice that we use the condition $\alpha < 1$ throughout the whole paper and that we adopt the notation $\langle O \rangle$ and \overline{O} to denote ensemble and time average, respectively. We shall adopt the notation $\langle \overline{O} \rangle$ to denote ensemble averages of time averages.

The result of Eq. (1) found in Refs. [3,4] is based on the model of anomalous diffusion, called continuous-time random walk (CTRW) [8]. The runner makes a sequence of jumps $\xi(1), \xi(2), \dots, \xi(n), \dots$. These are uncorrelated Gaussian fluctuations, with vanishing mean value $\langle \xi \rangle = 0$ and a finite second moment $\langle \xi^2 \rangle$. The number of events n is interpreted [9] as operational time. In the clock time the time interval between two operational events is described by the distribution density

$$\lim_{\tau \rightarrow \infty} \psi(\tau) \propto \frac{1}{\tau^{1+\alpha}}, \quad (3)$$

with $0 < \alpha < 1$. The authors of Ref. [10] derived this distribution density for neurophysiological events assuming a logarithmic relation between the clock time and the subjective or psychological time of the individuals. To make our picture as general as possible, including other possible sources of complexity, we use the term *operational time*, to denote the number of renewal events, rather than psychological time

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as it would be natural in the case of neurophysiological processes [10]. We shall focus on the condition $n \gg 1$ and for this reason we shall interpret n as being the continuous time, hereby denoted by the symbol τ_ψ . The clock time is continuous, but the numerical calculations convert it into a discrete time that can be interpreted again as continuous in the long-time limit. The theoretical interpretation of the clock-time region very close to the origin $t = 0$ forces us to consider again the discrete nature of operational time. This will be done hereby by introducing the parameter L_0 , of the order of the numerical time step Δt , below which no event occurs.

B. Extension of continuous-time random walk

We extend CTRW to the condition where in the operational time scale the fluctuation perceived by the runner is not random but correlated, and the decay of the correlation function of the fluctuation ξ is extremely slow. Let us consider the stationary and normalized correlation function

$$\Phi_\xi(|t_1 - t_2|) = \frac{\langle \xi(t_1)\xi(t_2) \rangle}{\langle \xi^2 \rangle}, \quad (4)$$

with $\langle \xi \rangle = 0$. We assign to the correlation function the analytical form

$$\Phi_\xi(\tau \equiv |t_1 - t_2|) = \left(\frac{T_B}{\tau + T_B} \right)^\nu. \quad (5)$$

The correlation time τ_c of the correlation function is established by evaluating the Laplace transform of $\Phi_\xi(\tau)$,

$$\hat{\Phi}_\xi(u) \equiv \int_0^\infty \exp(-ut) \Phi_\xi(\tau) d\tau, \quad (6)$$

which, after setting $u = 0$, yields

$$\tau_c = \begin{cases} \frac{T_B}{\nu-1} & \nu > 1 \\ \infty & 0 < \nu < 1. \end{cases} \quad (7)$$

We focus on the case of the very slow approach to equilibrium with $0 < \nu < 1$. This correlation function is not integrable, thereby making the correlation time infinite. This generalized form of CTRW, with the random fluctuation ξ of CTRW replaced by a correlated fluctuation, is equivalent to making the random walker run in the operational time regime, the well-known fractional Brownian motion (FBM), as discussed in Refs. [11,12], rather than the ordinary Brownian motion.

It is important to stress that FBM is ergodic in the sense that both the ensemble average and the time average yields $\langle x^2(t) \rangle \propto t^{2H}$ and $\langle \overline{\delta^2(\Delta)} \rangle \propto \Delta^{2H}$, with $0 < H < 1$. However, the authors of Ref. [13] found that this ergodic regime is realized after a very extended nonergodic regime that becomes perennial in the limit $H \rightarrow 1$. It is possible that the main result of this article, see Sec. IC, may have a connection with this extremely slow convergence to the ergodic regime.

To make it possible for the readers to appreciate the role of FBM memory, we find it convenient to discuss also the joint effect of NPR events and another form of an ergodicity-breaking process called scaled Brownian motion (SBM) [14]. This form of anomalous diffusion is attracting increasing interest due to the researchers working on anomalous diffusion in biological cells [2,15–17] and granular gases [18]. The

SBM [15] generates the fluctuation $\xi(t)$ according to the prescription

$$\xi(t) = \sqrt{kt^{2H-1}} \xi_R, \quad (8)$$

where k is a constant with the proper dimension and ξ_R a random Gaussian noise, thereby yielding the time-dependent diffusion coefficient $D(t)$ given by

$$D(t) = kt^{2H-1} \quad (9)$$

and leading the second moment of the diffusing variable $x(t)$, defined by $\dot{x} = \xi$, to increase in time as

$$\langle x(t)^2 \rangle \propto \int_0^t dt' D(t') \propto t^{2H}. \quad (10)$$

C. Main result of this paper

The main result of this paper is the theoretical and numerical discovery that the joint action of FBM memory and NPR events yields the localization of single trajectories. When the correlation function of Eq. (5) is sufficiently slow, the fluctuations of single trajectories generated by the extended form of CTRW described in Sec. IB tend to vanish, thereby making the position $x(t)$ virtually time independent in the time asymptotic limit. This is equivalent, to some extent, which is explained in the text, to replacing the factor $1/L^{1-\alpha}$ in Eq. (1) with $1/L^\nu$ where $\gamma \equiv 2H(1-\alpha)$, with the effect of generating a form of ergodicity breaking similar to that of Eq. (1), in the sense that the intensity of $\overline{\delta^2(\Delta)}$ is a decreasing function of the total length L of the time series analyzed. The parameter γ is not confined to $\gamma < 1$ and for $\alpha < 0.5$ a critical value of ν , ν_c , exists yielding $\gamma > 1$ for $\nu < \nu_c$ and making the intensity of $\overline{\delta^2(\Delta)}$ vanish. We refer to this phenomenon as *memory-induced localization*. It is important to stress that this form of localization, being a property of single trajectories, is compatible with the spreading of different trajectories generating the nonvanishing scaling $\eta = \alpha H$.

Using the same theoretical approach, this article sheds light also on the formal result of the negative scaling coefficient η found in the earlier work of Ref. [19]. In this paper we see that this earlier result is explained by the joint action of NPR and SBM, with the effect of making $\overline{\delta^2(\Delta)}$ vanish. However, this form of joint action is less surprising than the main result of this article, because the joint action of SBM fluctuations and NPR events yields localization when both processes yield subdiffusion.

The outline of this paper is as follows. In Sec. II we review the connection between NPR processes and the out-of-equilibrium cascade of events produced by the preparation of these systems, with an event occurring at $t = 0$. Section III is devoted to the discussion of the joint action of NPR events and FBM fluctuations. In Sec. IV we discuss the joint action of NPR events and SBM fluctuations. Finally, we devote Sec. V to concluding remarks.

II. RENEWAL

To generate the cascade of renewal events in this paper we adopt an algorithm based on the idealized Manneville map [20] prescription. In this case, the survival probability, namely, the

probability that a new event occurs at a time interval larger than t from an earlier event is given by the function $\Psi(t)$ defined by

$$\Psi(t) = \left(\frac{T}{T+t} \right)^\alpha, \quad (11)$$

where the parameter T defines the short-time scale, not showing yet the asymptotic complexity. The survival probability $\Psi(t)$ fits the normalization condition

$$\Psi(0) = 1, \quad (12)$$

which implies that an event certainly occurs at a given $t > 0$. The corresponding waiting time distribution density is given by

$$\psi(t) = \frac{\alpha T^\alpha}{(t+T)^{1+\alpha}}, \quad (13)$$

which in the asymptotic time limit clearly shows its inverse power-law complexity with the analytical form

$$\psi(t) = \frac{\Lambda^\alpha}{t^{1+\alpha}}, \quad (14)$$

where

$$\Lambda \equiv \alpha^{1/\alpha} T. \quad (15)$$

Let us consider the case of infinitely many realizations with the condition that all of them have an event at time $t = 0$. The rate of events at time t is given by

$$R(t) = \sum_{n=1}^{\infty} \psi_n(t), \quad (16)$$

where $\psi_n(t)$ is the probability density that an event occurring at time t is the last of a series of $n - 1$ earlier events. The renewal condition leads to the mathematical definition of $\psi_n(t)$ through the iterative relation

$$\psi_n(t) = \int_0^t \psi_{n-1}(t') \psi_1(t-t'), \quad (17)$$

where $\psi_1(t) = \psi(t)$ with $\psi(t)$ given by Eq. (13). The time convolution of Eq. (17) makes it possible to derive the Laplace transform of the right-hand side of Eq. (16) and so of $R(t)$. By inverse-Laplace transforming this result, see, e.g., Ref. [4], we are led to

$$R(t) = \frac{A}{t^{1-\alpha}}, \quad (18)$$

with

$$A \equiv \frac{\alpha}{\Lambda^\alpha \Gamma(\alpha) \Gamma(1-\alpha)}. \quad (19)$$

The expression of Eq. (18) is the well-known cascade of Feller events [21], which is the clear sign of the fact that there is no characteristic time scale for the dynamics of the system, when $\alpha < 1$.

It is important to notice that, as mentioned earlier, n can be interpreted as a discrete time, usually called *operational time*. In the asymptotic limit the operational time $n \rightarrow \infty$ becomes identical to a continuous time that we denote as τ_ψ . The numerical simulation of this article is based on the adoption of the finite time step $\Delta t = 1$. The evaluation of

$\overline{\delta^2}(\Delta)$ according to the prescription of Eq. (1) is done with moving windows of size Δ . In Eq. (1) the left side of the window closest to the preparation of system is located at $t = 0$. To make it easier for the reader to understand the phenomenon of memory-induced localization, we modify this definition, adopting as the bottom limit of integration $L_0 > 0$ rather than $t = 0$. This definition rests on the interpretation of discrete operational time as counting the occurrence of events and on the fact that, in addition to the preparation event, no event can occur at time shorter than L_0 . The order of magnitude of L_0 is, of course, $\Delta t = 1$.

III. JOINT ACTION OF NPR EVENTS AND FBM FLUCTUATION

The first form of joint action of two different sources of anomalous diffusion discussed in this article is an extension of CTRW. In the ordinary CTRW, when the runner jumps, she makes jumps in the positive or negative direction according to the sign of the fluctuation ξ and this fluctuation is random. In the extended CTRW of this article, the runner sees in her operational time a correlated fluctuation ξ , as a consequence of the infinite memory of the FBM generating fluctuations. In other words, we assume that the runner in her operational time τ_ψ makes a diffusion described by FBM. In the clock time, in the extended time region between two events, the runner does not move. This has the effect of producing the time scale dilatation

$$t = \tau_\psi^{1/\alpha}, \quad (20)$$

thereby favoring subdiffusion. This problem has been already studied in the earlier work of Refs. [22–24].

However, we focus on the case $H > 0.5$, which in the operational time scale yields super-diffusion. For this reason we think that the localization effect revealed by the theoretical and numerical analysis of this article is a surprising property.

A. Review of the dynamical origin of FBM

To make this paper as self-contained as possible, let us review the derivation of FBM done in the earlier work of Refs. [11,12]. We move from the equation of motion

$$\dot{x} \equiv \frac{dx}{dt} = \xi(t). \quad (21)$$

In this paper, to generate the trajectory $x(t)$ yielding the fluctuation $\xi(t)$, we use the algorithm of Ref. [25]. This not only allows us to establish an approach equivalent to the dynamical approach to FBM of the earlier work of Refs. [11,12] but it also makes it possible for us to make a comparison with SBM of Eq. (8). After recording $\xi(t)$ we evaluate its correlation function using the time average and, resting on the ergodicity property, we establish numerically the parameter T_B of Eq. (4) and the mean quadratic value $\langle \xi^2 \rangle$.

Notice that although the prescriptions of this subsection will be applied to the operational time τ_ψ , for notational simplicity we adopt the conventional symbol t for time. The fluctuation $\xi(t)$ is characterized by the stationary correlation function of Eqs. (4) and (5). We integrate Eq. (21), square the result, make the ensemble average, and we use the property that the

correlation function depends on $|t_1 - t_2|$ [see Eq. (4)] to get

$$\langle x^2(t) \rangle = 2\langle \xi^2 \rangle \int_0^t dt' \int_0^{t'} dt'' \Phi_\xi(t''). \quad (22)$$

Thus we obtain that for $t \rightarrow \infty$

$$\langle x^2(t) \rangle = ct^{2H}, \quad (23)$$

where

$$H \equiv 1 - \frac{\nu}{2} \quad (24)$$

and

$$c \equiv \begin{cases} \frac{2\langle \xi^2 \rangle T_B^\nu}{(1-\nu)(2-\nu)} & H \neq 0.5 \\ \langle \xi^2 \rangle & H = 0.5. \end{cases} \quad (25)$$

B. Time averages

The evaluation of the second moment through time averages is given by

$$\overline{\delta^2}(\Delta) = \frac{1}{L - \Delta} \int_{L_0}^{L - \Delta} [x(t) - x(t + \Delta)]^2 dt, \quad (26)$$

where L is the total length of the time series of the diffusional variable $x(t)$. Note that, as stated earlier, to take into account the discrete nature of the operational time the bottom limit of time integration is given by L_0 , which is of the order of 1.

We adopt the approximation of keeping the length Δ of the mobile window much smaller than the total length L , thereby replacing Eq. (26) with

$$\overline{\delta^2}(\Delta) = \frac{1}{L} \int_{L_0}^L [x(t) - x(t + \Delta)]^2 dt. \quad (27)$$

After evaluating the time average over a single trajectory, we make the time average over another trajectory, and so on. Then we evaluate an ensemble average over the time averages and we rewrite Eq. (27) as

$$\langle \overline{\delta^2}(\Delta) \rangle = \frac{1}{L} \int_{L_0}^L \langle [x(t) - x(t + \Delta)]^2 \rangle dt. \quad (28)$$

Note that $x(t)$ changes only when an event occurs in the operational time. As a consequence,

$$\langle [x(t) - x(t + \Delta)]^2 \rangle = cn(t + \Delta, t)^{2H}, \quad (29)$$

where c is given by Eq. (25) and $n(t + \Delta, t)$ is the number of events occurring between $t + \Delta$ and t .

This number is evaluated by adopting the Feller prescription [4,21]. We assume that the system is prepared at time $t = 0$, we use for $R(t)$ Eq. (18), and, as a consequence, the total number of events produced moving from 0 to $\tau > 0$, $m(\tau)$, is

$$m(\tau) = \int_0^\tau \frac{A}{t^{1-\alpha}} dt' = \frac{A}{\alpha} \tau^\alpha. \quad (30)$$

The number $n(t + \Delta, t)$ is given by

$$n(t + \Delta, t) = \frac{A}{\alpha} [(t + \Delta)^\alpha - t^\alpha]. \quad (31)$$

We make the assumption $\Delta \ll t$. With this assumption we are allowed to use a Taylor series expansion and to neglect the

higher-order terms, thereby getting

$$n(t + \Delta, t) \sim At^{\alpha-1} \Delta. \quad (32)$$

By plugging Eq. (32) into Eq. (28) we get, after a straightforward time integration,

$$\langle \overline{\delta^2}(\Delta) \rangle = \frac{B \Delta^{2H}}{L(1-\gamma)} [L^{1-\gamma} - L_0^{1-\gamma}], \quad (33)$$

where

$$\gamma \equiv 2H(1-\alpha). \quad (34)$$

Note that

$$B = cA^{2H}. \quad (35)$$

We introduce this factor to recover the key result of Eq. (28) when $H = 0.5$.

When $\gamma < 1$ and $L \gg L_0$ Eq. (33) yields

$$\langle \overline{\delta^2}(\Delta) \rangle = \frac{B}{L^{2H(1-\alpha)}} \frac{1}{1 - 2H(1-\alpha)} \Delta^{2H}. \quad (36)$$

This equation is a generalization of the important result of Eq. (1). In fact, when $H = 0.5$ Eq. (36) becomes identical to Eq. (1), which is the main result of the theoretical proposal of Ref. [4]. Another interesting property of Eq. (36) is that for $\alpha = 1$ the second moment $\overline{\delta^2}(\Delta)$ becomes independent of the length L of the time series and identical to the second moment evaluated with the conventional ensemble average, thereby showing that FBM is ergodic. If $\alpha < 1$, then the cascade of Feller events makes the intensity of $\overline{\delta^2}(\Delta)$, namely the factor of δ^{2H} in Eq. (36), depend on the time length of the sequence. This form of ergodicity breaking does not affect the FBM scaling.

Note also that, as mentioned in Sec. I, to maintain the structure of Eq. (1) we have to assign to the slowness index ν of Eq. (5) a value larger than a critical value ν_c . We are now in a position to define this critical value of ν . First, we notice that $\gamma > 1$ is possible for $\alpha < 0.5$. In fact, the critical value of H , H_c , at which $\gamma = 1$, is given by

$$H_c = \frac{1}{2(1-\alpha)}. \quad (37)$$

To realize the condition $\gamma > 1$, we need to make H larger than H_c and this is impossible to do with $\alpha = 0.5$, a value of α assigning to H its maximal value $H = 1$. Larger values of α would make H exceed this maximal value. The adoption of $\alpha < 0.5$ generates $H_c < 1$, and, consequently, according to Eq. (34), $\gamma < 1$. The violation of the condition $\gamma < 1$ requires $H > H_c$ and a slowness parameter ν small enough. Using Eq. (24), we are led to define ν_c as

$$\nu_c = \frac{1 - 2\alpha}{1 - \alpha}. \quad (38)$$

The condition $\gamma < 1$ corresponds to $\nu > \nu_c$. Adopting a correlation function with a slower decay, $\nu < \nu_c$, is equivalent to setting $\gamma > 1$, and this has the dramatic effect of making $\overline{\delta^2}(\Delta)$ vanish. In fact, with $\gamma > 1$ Eq. (33) becomes

$$\langle \overline{\delta^2}(\Delta) \rangle = \frac{B \Delta^{2H}}{L(\gamma - 1)} \left[\frac{1}{L_0^{\gamma-1}} - \frac{1}{L^{\gamma-1}} \right]. \quad (39)$$

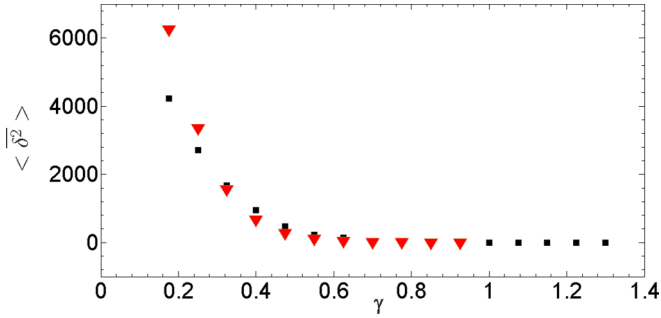


FIG. 1. The red triangles denote $\overline{\delta^2}(\Delta)$ of Eq. (36) as a function of γ given by Eq. (34). The ensemble average is done on 10 single trajectory realizations. We set $H = 0.75$, $\Delta = 10^3$, $L = 10^6$. To define Eq. (36) we use Eqs. (24), (25), (19), (35), and (15). Equation (24) with $H = 0.75$ yields $\nu = 0.5$. For Eq. (25) we use $\langle \xi^2 \rangle = 0.0657$, $T_B = 0.129$, yielding $c = 0.0627$. Equation (19) depends on Eq. (15) with α changing with γ and $T = 0.091$. The black squares show the result of numerical treatment and for $\gamma > 1$ they agree with the vanishing value predicted by Eq. (40).

It is convenient to write this expression in the form

$$\overline{\delta^2}(\Delta) = \frac{L_0}{L} \frac{B \Delta^{2H}}{L_0^\gamma (\gamma - 1)} \left[1 - \left(\frac{L_0}{L} \right)^{\gamma-1} \right], \quad (40)$$

which shows that for $L_0/L \rightarrow 0$ the second moment $\overline{\delta^2}(\Delta)$ vanishes, yielding a perfect localization, in the same way as Eq. (1) does for $\alpha \rightarrow 0$, as the readers can easily realize from Ref. [5] and from the property $\lim_{\alpha \rightarrow 0} \Gamma(\alpha) = \infty$. In the case of ordinary CTRW this localization is a natural consequence of the fact that with $\alpha \rightarrow 0$ for the entire time interval L no new event may occur in addition to the preparation event. The same effect emerges from the joint action of NPR events and FBM fluctuation, and this is the reason why we use the term localization to define it. However, the localization of this paper is not a consequence of lack of events, since, as shown in Sec. III C, the ensemble average approach yields a second moment of the diffusion process increasing in time, although with a very small scaling. The localization effect of this paper, produced by the joint action of NPR events and FBM fluctuation, is determined by the FBM infinite memory transmitted to the NPR events. The rare occurrence of NPR events makes the slow decay of the correlation function of Eq. (5) become infinitely slow and the single trajectories become equivalent to deterministic and ballistic trajectories, with the effect of annihilating $\overline{\delta^2}(\Delta)$. This is the reason why we refer to this effect as *memory-induced localization*. The theoretical prediction of memory-induced localization is satisfactorily supported by the numerical calculations of Fig. 1.

It is important to point out that the theoretical prediction of this section can be interpreted as being based on the assumption

$$\langle [x(t) - x(t + \Delta)]^2 \rangle \propto \langle n(t + \Delta, t) \rangle^{2H}. \quad (41)$$

As pointed out by the authors of Ref. [23] the number of events between $t + \Delta$ and t is not fixed and can be evaluated

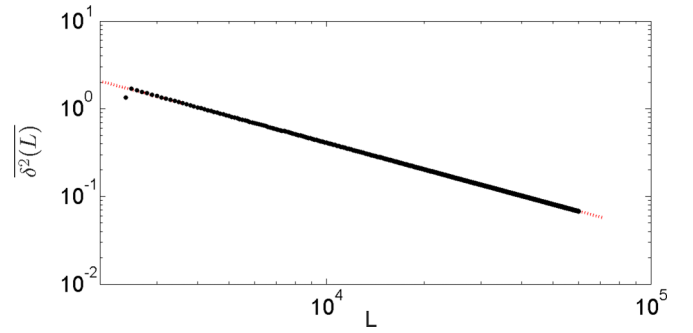


FIG. 2. The single trajectory time average evaluation of the second moment of Eq. (26) as a function of L . $H = 0.75$, $\alpha = 0.35$, $\Delta = 100$. The black dots denote the numerical results and the dashed red line is the fitting yielding the slope 1.0057. According to Eq. (36), the theoretical slope of the straight line is $2H(1 - \alpha) = 0.975$.

by adopting the following prescription:

$$\langle (n(t + \Delta) - n(t))^{2H} \rangle = \int_0^\Delta d\tau \psi_t(\tau) n(t + \tau)^{2H}, \quad (42)$$

where $\psi_t(\tau)$ is the waiting time distribution density when we begin waiting at time t far from the occurrence of the preparation event. Following the calculations done in Ref. [23] we get

$$\langle \overline{\delta^2}(\Delta) \rangle \propto \frac{1}{L^{1-\alpha}} \Delta^{1-\alpha+2H\alpha}. \quad (43)$$

As we see, with the adoption of the prescription of Eq. (42) the signature of the joint action of NPR events and FBM generating fluctuations is given by the power index of Δ which is $1 - \alpha + 2H\alpha$ rather than $2H$. There is no memory-induced localization.

To support our arguments on the memory-induced localization, in spite of the striking difference with the rigorous result of Ref. [23], we make the numerical calculation of Fig. 2. This figure refers to the case where γ is very close to the border limit $\gamma = 1$ and we think that the agreement between numerical results and theoretical prediction is satisfactory. However, Fig. 3 shows that the theoretical prediction of Ref. [23] on

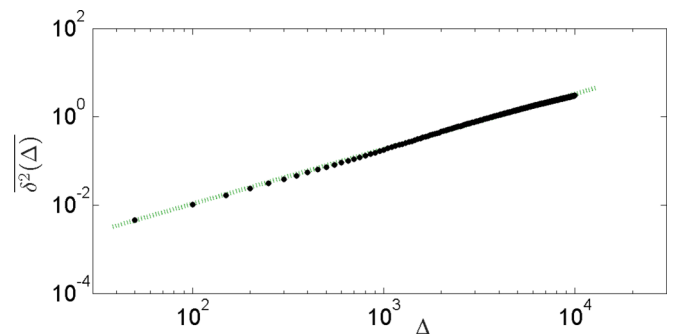


FIG. 3. The single trajectory time average evaluation of the second moment of Eq. (26) as a function of Δ . $H = 0.75$, $\alpha = 0.4$, $L = 10^6$. The black dots denote the numerical results and the dashed green line is the fitting yielding the slope 1.244. According Eq. (43), the slope of the straight line is $1 - \alpha + 2\alpha H = 1.2$. Note that $2H = 1.5$.

the power index of Δ is correct. In the concluding remarks we shall come back to discuss the conflict between the predictions of this article and the earlier work of Ref. [23].

C. Ensemble average

In the operational time scale the diffusion process is described by

$$p_{\tau_\psi}(x, \tau_\psi) = \frac{1}{\sqrt{2\pi k \tau_\psi^{2H}}} \exp\left(-\frac{x^2}{2k \tau_\psi^{2H}}\right). \quad (44)$$

Moving from the operational to the clock time, we get

$$p_C(x, t) = \int_0^\infty d\tau_\psi \beta(t, \tau_\psi) \frac{1}{\sqrt{2\pi k \tau_\psi^{2H}}} \exp\left(-\frac{x^2}{2k \tau_\psi^{2H}}\right), \quad (45)$$

where $\beta(t, \tau_\psi)$ denotes the distribution density of time t interpreted as the diffusion variable of the discrete time n that in the asymptotic limit becomes the continuous variable τ_ψ [26,27]. The clock time t is related to the operational time τ_ψ by the relation of Eq. (20), which is a consequence of the fact that t plays the role of diffusional variable. This yields

$$\tau_\psi \propto t^\alpha. \quad (46)$$

We can find the scaling η generated by Eq. (45) with the following intuitive argument. The distribution $\beta(t, \tau_\psi)$ sets the constraint $\tau_\psi = t^\alpha$. As a consequence, the variable x , which is proportional to τ_ψ^{2H} , turns out to be proportional to t^η , with

$$\eta = \alpha H. \quad (47)$$

This result can be obtained in a more rigorous way by noticing that the scaling of Eq. (46) corresponds to

$$\beta(t, \tau_\psi) = \frac{1}{t^\alpha} F\left(\frac{\tau_\psi}{t^\alpha}\right). \quad (48)$$

By plugging $\beta(t, \tau_\psi)$ of Eq. (48) into Eq. (45) and replacing the integration variable τ_ψ with $y \equiv \tau_\psi/t^\alpha$, we obtain

$$p_C(x, t) = \frac{1}{t^{\alpha H}} g_H\left(\frac{x^2}{t^{2H\alpha}}\right), \quad (49)$$

where

$$g_H(z) \equiv \int_0^\infty dy F(y) \frac{1}{\sqrt{2\pi y^{2H}}} \exp\left[-\frac{z^2}{2ky^{2H}}\right], \quad (50)$$

thereby yielding a more rigorous derivation of the scaling of Eq. (47).

Note that thanks to the earlier work of Ref. [28] the Mittag-Leffler function is connected to $\beta(t, \tau_\psi)$ through

$$E_\alpha(-t^\alpha s) = \int_0^\infty d\tau_\psi \beta(t, \tau_\psi) \exp(-\tau_\psi s) \quad (51)$$

for $s \rightarrow 0$. As a consequence, the double Laplace transform of $\beta(t, \tau_\psi)$, $\hat{\beta}(u, s)$, reads

$$\hat{\beta}(u, s) = \frac{1}{u + u^{1-\alpha} s}. \quad (52)$$

In the special case $\alpha = 0.5$ inverse-Laplace transforming Eq. (52) gives

$$\beta(t, \tau_\psi) = \frac{\exp\left(-\frac{\tau_\psi^2}{4t}\right)}{\sqrt{(\pi t)}}. \quad (53)$$

Using Eq. (45) we get

$$p_C(x, t) = \int_0^\infty \frac{\exp\left(-\frac{\tau_\psi^2}{4t}\right) \exp\left[-\frac{x^2}{2\tau_\psi^{2H}}\right]}{\sqrt{\pi t} \sqrt{2\pi \tau_\psi^{2H}}} d\tau_\psi. \quad (54)$$

On the basis of earlier arguments it is straightforward to prove that this equation can be written under the form

$$p_C(x, t) = \frac{1}{t^{\alpha H}} F\left(\frac{x}{t^{\alpha H}}\right). \quad (55)$$

For example, in the case $H = 1$ we get

$$p_C(x, t) = \frac{1}{\pi\sqrt{2t}} K_0\left(\frac{|x|}{\sqrt{2t}}\right), \quad (56)$$

where $K_0(z)$ is the modified Bessel function of the second kind. Note that Eq. (56) refers to the case $\alpha = 0$ and $H = 1$, namely the case where γ of Eq. (34) is $\gamma = 2$. This is a condition where $\langle \delta^2(\Delta) \rangle = 0$, according to the theoretical and numerical results of Sec. III B. In fact, as stressed in Sec. I C, localization occurs when $\gamma > 1$. The localization of a single trajectory is a clear manifestation of ergodicity breaking, because the localization of a single trajectory is derived from the statistical analysis in time, departing from the ensemble average analysis yielding no localization. Ergodicity breaking can also be realized as the ensemble and time analysis bringing two different scalings. We notice, in fact, that the scaling of Eq. (56), ensemble analysis, is $\eta = 1/2$, in accordance with Eq. (47), whereas the localization $\langle \delta^2(\Delta) \rangle = 0$, observed doing time analysis, suggests that the single trajectories have the scaling $\eta = 0$.

D. Concluding remarks on this first form of joint action

We see that this form of joint action generates ergodicity breaking. In fact, the ensemble average analysis yields Eq. (36) and, consequently, the scaling of Eq. (47), $\eta = \alpha H$. The time average analysis, on the contrary, generates a scaling dependent on whether the prescription of Eq. (41) or the prescription of Eq. (42) is adopted. In the former case

$$\eta = H \quad (57)$$

and in the latter

$$\eta = H\alpha + \frac{1-\alpha}{2}, \quad (58)$$

thereby yielding ergodicity breaking regardless of what prescription is adopted. The three scalings are identical only in the absence of the Feller cascade, $\alpha = 1$, a property reminiscent of the results of Refs. [3,4].

From a theoretical point of view, averaging on infinitely many single trajectory realization casts some doubts on the prescription of Eq. (41) and especially on the scaling of Eq. (57). However, the observation of single trajectories, with the adoption of time averages, shows that the dependence on L of Eq. (36) is a correct prediction, thereby supporting

our conclusion that $\gamma > 1$ generates localization, whereas the ensemble average generates a sub-diffusion process more pronounced than that of CTRW, with nevertheless the nonvanishing scaling $\eta = \alpha H$.

IV. JOINT ACTION OF NPR EVENTS AND SBM FLUCTUATIONS

The second form of joint action is realized by adopting SBM rather than FBM and by assuming that SBM is activated in the clock time. The runner, however, spends a large part of her time sleeping. She jumps only when she is awake.

The ensemble average is established through the formula

$$\langle x^2(t) \rangle = \int_{L_0}^t dt_1 \int_{L_0}^t dt_2 \langle \xi^*(t_1) \xi^*(t_2) \rangle, \quad (59)$$

where $\xi^*(t)$ denotes the fluctuation perceived by the runner. This fluctuation differs from the vanishing value only when the runner is awake. The sleeping condition is equivalent to setting $\xi^*(t) = 0$. In this case, the joint action of the Feller cascade and the SBM diffusion time dependence yields

$$\langle \xi^*(t) \xi^*(t + \tau) \rangle = \frac{2kA}{t^{1-\alpha}} t^{2H-1} \delta(\tau), \quad (60)$$

with the constant k being determined by the prescription of Eq. (8) and the constant A by the prescription of Eq. (19). In this section we focus only on the scaling at the level of ensemble statistics and at the level of time statistics. Therefore, for simplicity's sake, we set $2kA = 1$. By plugging Eq. (60) with $2kA = 1$ into Eq. (59), we obtain

$$\langle x^2(t) \rangle = \frac{1}{2H - 1 + \alpha} [t^{2H-1+\alpha} - L_0^{2H-1+\alpha}]. \quad (61)$$

Notice that

$$2H - 1 + \alpha = 2\eta, \quad (62)$$

where

$$\eta = \frac{2H - 1 + \alpha}{2}. \quad (63)$$

This is the scaling generated by the generalized diffusion equation proposed by the authors of Ref. [19]. When $\eta > 0$ and $t \gg L_0$, Eq. (61) becomes

$$\langle x^2(t) \rangle = \frac{1}{2\eta} t^{2\eta}. \quad (64)$$

It is worth noting that Ref. [19] does not discuss the dynamics of this process when $\eta < 0$. As a result of the theoretical approach illustrated in this paper, we reach the compelling conclusion that this negative scaling leads to an exact localization. In fact, when $\eta < 0$, Eq. (61) yields

$$\langle x^2(t) \rangle = \frac{1}{2|\eta|} \left[\frac{1}{L_0^{2|\eta|}} - \frac{1}{t^{2|\eta|}} \right]. \quad (65)$$

This prediction is supported by the numerical results of Fig. 4. Notice that both Eq. (64) and Eq. (65) yield for $\eta \rightarrow 0$ the logarithmic scaling $x^2 \sim \ln t$. Figure 5 supports this prediction in the case $\eta < 0$.

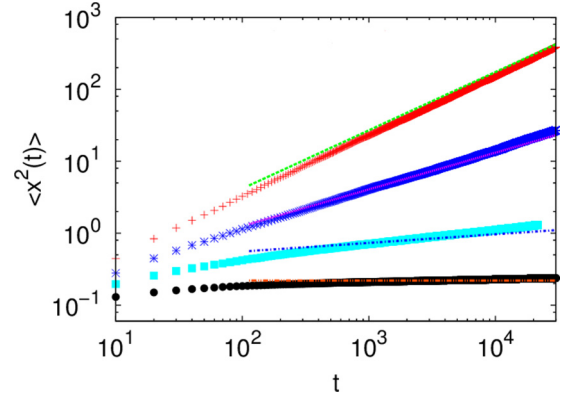


FIG. 4. The ensemble average of the second moment $\langle x^2(t) \rangle$ for different values of H and α . All the curves were generated from the average of 10^4 realizations. From the top to the bottom: $H = 0.45, \alpha = 0.9$; $H = 0.45, \alpha = 0.6$; $H = 0.10, \alpha = 0.9$; $H = 0.10, \alpha = 0.5$. All the straight lines but the bottom one are guidelines indicating the scaling $2\eta = (2H - 1 + \alpha)$. Note that for the bottom curve, corresponding to $2\eta = -0.2$, the numerical result fits the guideline straight curve with vanishing slope.

What about the analysis of single trajectories in this case? To do the time average analysis we must adopt

$$\begin{aligned} \overline{\delta^2(\Delta)} &= \frac{1}{L} \int_{L_0}^L \langle [x(t) - x(t + \Delta)]^2 \rangle dt \\ &= \frac{1}{L} \int_{L_0}^L dt C(\Delta, t), \end{aligned} \quad (66)$$

where the function $C(\Delta, t)$ reads

$$C(\Delta, t) \equiv \int_t^{t+\Delta} dt_1 \int_t^{t+\Delta} dt_2 \langle \xi^*(t_1) \xi^*(t_2) \rangle, \quad (67)$$

with the correlation function $\langle \xi^*(t_1) \xi^*(t_2) \rangle$ given by

$$\langle \xi^*(t_1) \xi^*(t_2) \rangle = \frac{1}{t_1^{2-\mu}} t_1^{2H-1} \delta(|t_1 - t_2|). \quad (68)$$

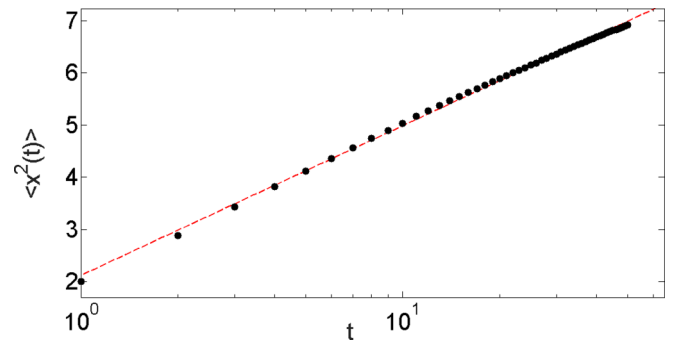


FIG. 5. Numerical evidence that $\eta < 0$ makes x proportional to $\ln t$. $H = 0.1, \alpha = 0.6$. The black dots are the numerical results and the red dashed line is a linear-log fitting with the slope 1.23. Note that with these values the scaling of Eq. (63) becomes negative, $\eta = -0.1$.

By plugging Eq. (68) into the right-hand side of Eq. (67) we get

$$C(\Delta, t) = \frac{1}{2\eta} [(t + \Delta)^{2\eta} - t^{2\eta}]. \quad (69)$$

Doing the usual Taylor series expansion with the assumption $\Delta < t$ and neglecting the terms of higher order in Δ/t , we obtain

$$C(\Delta, t) = \Delta t^{2\eta-1}. \quad (70)$$

Plugging $C(\Delta, t)$ of Eq. (70) into the right-hand side of Eq. (66) and integrating over t we get

$$\langle \overline{\delta^2}(\Delta) \rangle = \frac{\Delta}{2\eta L} [L^{2\eta} - L_0^{2\eta}]. \quad (71)$$

When $\eta > 0$ and $L \gg L_0$, we get

$$\langle \overline{\delta^2}(\Delta) \rangle = \frac{1}{L^{1-2\eta}} \frac{\Delta}{2\eta}. \quad (72)$$

When $\eta < 0$ and $L \gg L_0$ Eq. (71) becomes

$$\langle \overline{\delta^2}(\Delta) \rangle = \Delta \left(\frac{L_0}{L} \right)^{2|\eta|} \left[1 - \left(\frac{L_0}{L} \right)^{2|\eta|} \right], \quad (73)$$

yielding a vanishing second moment for $L_0/L \rightarrow 0$.

V. CONCLUDING REMARKS

This article affords a solution to the question raised by the authors of Ref. [19] on the interpretation of the scaling generated by the diffusion equation

$$\frac{\partial^\alpha}{\partial t^\alpha} p(x, t) = D(t) \frac{\partial^2}{\partial x^2} p(x, t), \quad (74)$$

where the fractional derivative rests on the Caputo prescription [29] and, according to Eq. (9),

$$D(t) = ct^{2H-1}. \quad (75)$$

The authors of Ref. [19] proved that this diffusion equation generates the scaling of Eq. (63) but they were unable to establish the nature of the individual stochastic trajectories. An important result of this article is the discovery that the single trajectories corresponding to this scaling are those illustrated in Sec. IV. The research that we did to answer this question led us to establish that the condition $\eta \rightarrow 0$ yields the logarithmic scaling $x^2 \sim \ln t$, regardless of whether the vanishing value of η is realized moving from $\eta > 0$ or from $\eta < 0$. This process is characterized by ergodicity breaking and it is remarkable that the condition $\eta < 0$ corresponds to a perfect localization, as proved by Eq. (73).

Much more surprising is the result that we obtain thanks to the joint action of NPR and FBM, as illustrated in Fig. 1. In this case, the localization induced by ergodicity breaking is realized with $H > 0.5$ corresponding to superdiffusion. In the whole region of the memory-induced localization $\gamma > 1$, the ensemble scaling $\eta = \alpha H$ applies and $\alpha < 0.5$ makes $\eta < 0.5$, even in the condition of maximal FBM superdiffusion, $H = 1$. Thus, the whole region $\gamma > 1$ corresponds to ensemble

subdiffusion. However, the memory-induced localization of this article is not a trivial consequence of analyzing the time average of single trajectories when we have ensemble subdiffusion. In fact, the conditions $\alpha = 0.6$, for instance, with $H = 1$ yields $\eta = 0.6$, namely superdiffusion, and with $H = 0.75$ yields $\eta = 0.45$, namely subdiffusion. Yet, both conditions are characterized by $\gamma < 1$, $\gamma = 0.8$, the former, and $\gamma = 0.6$, the latter, and, consequently, in neither conditions memory-induced localization is produced. In conclusion, we may have ensemble subdiffusion and no localization with time averages on the single trajectories.

The adoption of the prescription of Eq. (42) adopted by the authors of Ref. [23] casts some doubts on this effect, insofar as it yields the theoretical prediction of Eq. (43), with no localization effect. However, we note that this theoretical prediction is based on the assumption that the ensemble average commutes with the time integration, making Eq. (27) turn into Eq. (28). This property is correct if the ensemble average of the time averages is done on infinitely many realizations. The numerical results of Fig. (1) are obtained making an ensemble average over 10 single-trajectory realizations and the results of Fig. (2) and of Fig. (3) are based on single trajectories. They show that the scaling prediction of Eq. (57) is incorrect and that probably averaging over a suitably large number of realizations would lead to the scaling of Eq. (58). We believe that the correct theoretical prediction of Eq. (43) corresponds to the superposition of infinitely many single realizations, each of which has an almost fixed nonvanishing value $\overline{\delta^2}$, namely a localized single trajectory.

In conclusion, the memory-induced localization is a single-trajectory property and we are convinced that this result may lead to a new vision of memory and renewal processes. To shape this vision, it may be convenient to consider FBM with the second moment increasing as t^{2H} as an indication of randomness quite distinct from that signaled by the occurrence of renewal events. The correlation function of Eq. (5) is compatible with the derivation from the Hamilton formalism and, consequently, with the Laplace determinism [30,31]. This suggests that the entropy increase in this case is due to the lack of information generated by a contraction over the irrelevant variables that, in spite of being ignored, are responsible for the slow decay of this correlation function. The second moment increasing as t^α , on the contrary, has a completely different meaning. In the operational time, in fact, any time step is characterized by the action of a random event, thereby forcing the second moment to increase linearly with τ_ψ . As a matter of fact, as we have seen, $t^\alpha = \tau_\psi$, in line with the extended definition of Lyapunov coefficient adopted by Korabel and Barkai to adapt it to the occurrence of rare events of the $\alpha < 1$ condition [32]. In this case, when we depart from the singularity condition $H = 0.5$, where ξ is completely uncorrelated, and we move towards $H = 1$ with the decay of the correlation function of Eq. (5) becoming slower and slower, the correlated fluctuation ξ has the effect of completely quenching the action of NPR randomness at the critical value $H_c = 1/[2(1 - \alpha)]$, corresponding to the critical slowness of Eq. (38). Although we have been referring ourselves to this phenomenon as memory-induced localization, imagining the FBM memory as the cause of the effect, this may be probably

equivalent to interpret the rare events of the condition $\alpha < 1$ as the cause of the effect, being a physical condition that makes the decay of correlation function of Eq. (5) infinitely slow.

The settlement of these problems require further studies and a promising research direction is suggested by the recent work of Marzen and Crutchfield [33] where infinite memory and renewal events are discussed using the mutual information

between past and future and the amount of information from the past required to exactly predict the future.

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- [1] J. L. Lebowitz and O. Penrose, Modern ergodic theory, *Phys. Today* **26**(2), 23 (1973).
- [2] R. Metzler, J.-H. Jeon, A. G. Cherstvy, and E. Barkai, Anomalous diffusion models and their properties: And non-stationarity, non-ergodicity, ageing at the centenary of single particle tracking, *Phys. Chem. Chem. Phys.* **16**, 24128 (2014).
- [3] A. Lubelski, I. M. Sokolov, and J. Klafter, Nonergodicity Mimics Inhomogeneity in Single Particle Tracking, *Phys. Rev. Lett.* **100**, 250602 (2008).
- [4] Y. He, S. Burov, R. Metzler, and E. Barkai, Random Time-Scale Invariant Diffusion and Transport Coefficients, *Phys. Rev. Lett.* **101**, 058101 (2008).
- [5] We note that the complete expression for Eq. (1) is $\overline{\delta^2}(\Delta) = \{(\xi^2)\Delta\}/\{\alpha T^\alpha \Gamma(\alpha)\Gamma(1-\alpha)L^{1-\alpha}\}$.
- [6] G. Margolin and E. Barkai, Nonergodicity of a time series obeying Lévy statistics, *J. Stat. Phys.* **122**, 137 (2006).
- [7] M. Lukovic and P. Grigolini, Power specr for both interrupted and perennial aging processes, *J. Chem. Phys.* **129**, 184102 (2008).
- [8] E. W. Montroll, and G. H. Weiss, Random walks on lattices. II, *J. Math. Phys.* **6**, 167 (1965).
- [9] I. M. Sokolov, Lévy flights from a continuous-time process, *Phys. Rev. E* **63**, 011104 (2000).
- [10] P. Grigolini, G. Aquino, M. Bologna, and B. J. West, A theory of $1/f$ noise in human cognition, *Physica A* **388**, 4192 (2009).
- [11] R. Cakir, P. Grigolini, and A. A. Krokhin, Dynamical origin of memory and renewal, *Phys. Rev. E* **74**, 021108 (2006).
- [12] R. Cakir, A. Krokhin, and P. Grigolini, From the trajectory to the density memory, *Chaos Solitons Fract.* **34**, 19 (2007).
- [13] W. Deng and E. Barkai, Ergodic properties of fractional Brownian-Langevin motion, *Phys. Rev. E* **79**, 011112 (2009).
- [14] S. C. Lim and S. V. Muniandy, Self-similar Gaussian processes for modeling anomalous diffusion, *Phys. Rev. E* **66**, 021114 (2002).
- [15] J.-H. Jeon, A. V. Chechkin, and R. Metzler, Scaled Brownian motion: A paradoxical process with a time dependent diffusivity for the description of anomalous diffusion, *Phys. Chem. Chem. Phys.* **16**, 15811 (2014).
- [16] H. Safdari, A. G. Cherstvy, A. V. Chechkin, F. Thiel, I. M. Sokolov, and R. Metzler, Quantifying the non-ergodicity of scaled Brownian motion, *J. Phys. A* **48**, 375002 (2015).
- [17] A. Bodrova, A. V. Chechkin, A. G. Cherstvy, and R. Metzler, Ultraslow scaled Brownian Motion, *New J. Phys.* **17**, 063038 (2015).
- [18] A. Bodrova, A. V. Chechkin, A. G. Cherstvy, and R. Metzler, Quantifying non-ergodic dynamics of force-free granular gases, *Phys. Chem. Chem. Phys.* **17**, 21791 (2015).
- [19] M. Bologna, B. J. West, and P. Grigolini, Renewal and memory origin of anomalous diffusion: A discussion of their joint action, *Phys. Rev. E* **88**, 062106 (2013).
- [20] P. Allegrini, V. Benci, P. Grigolini, P. Hamilton, M. Ignaccolo, G. Menconi, L. Palatella, G. Raffaelli, N. Scafetta, M. Virgilio, and J. Yang, Compression and diffusion: A joint approach to detect complexity, *Chaos Solitons Fract.* **15**, 517 (2003).
- [21] W. Feller, *An Introduction to Probability Theory and Its Applications*, 2nd ed. (Wiley, New York, 1971), Vol. II.
- [22] F. Thiel and I. M. Sokolov, Weak ergodicity breaking in an anomalous diffusion process of mixed origins, *Phys. Rev. E* **89**, 012136 (2014).
- [23] Y. Meroz, I. M. Sokolov, and J. Klafter, Subdiffusion of mixed origins: When ergodicity and nonergodicity coexist, *Phys. Rev. E* **81**, 010101(R) (2010).
- [24] The very interesting work of Ref. [9] refers to the case where the diffusion process in the operational time scale occurs in a fractal environment where the second moment $\langle x^2(t) \rangle \propto t^\beta$. However, the theoretical results of this paper can be fruitfully adopted in the present article by replacing β with $2H$, noticing that the authors of Ref. [23] discussed the case where the diffusion process in the operational time is subdiffusional, $\beta < 1$ and, consequently, $H < 0.5$.
- [25] P. Abry and F. Sellan, The wavelet-based synthesis for the fractional Brownian motion proposed by F. Sellan and Y. Meyenr: Remarks and fast implementation, *Appl. Comp. Harmonic Anal.* **3**, 377 (1996).
- [26] B. Dybiec and E. Gudowska-Nowak, Subordinated diffusion and continuous time random walk asymptotics, *Chaos* **20**, 043129 (2010).
- [27] V. V. Uchaikin, Self-similar anomalous diffusion and Lévy-stable laws, *Phys. Usp.* **46**, 821 (2003).
- [28] P. Pramukul, A. Svenkeson, P. Grigolini, M. Bologna, and B. J. West, Complexity and the Fractional Calculus, *Adv. Math. Phys.* **2013**, 498789 (2013).
- [29] F. Mainardi and R. Gorenflo, On Mittag-Leffler-type functions in fractional evolution processes, *J. Comput. Appl. Math.* **118**, 283 (2000).
- [30] P. Grigolini, Emergence of biological complexity: Criticality, renewal and memory, *Chaos Solitons Fract.* **81**, 575 (2015).
- [31] M. Annunziato, P. Grigolini, and J. Riccardi, Fluctuation-dissipation process without a time scale, *Phys. Rev. E* **61**, 4801 (2000).
- [32] N. Korabel and E. Barkai, Pesin-Type Identity for Intermittent Dynamics with a Zero Lyapunov Exponent, *Phys. Rev. Lett.* **102**, 050601 (2009).
- [33] S. E. Marzen, J. P. Crutchfield, Statistical signatures of structural organization: The case of long memory in renewal processes, *Phys. Lett. A* **380**, 1517 (2016).