A SCALING METHOD FOR STOCHASTIC PROCESS

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In this work, we use the ideas of scaling to investigate stochastic process for asymptotic times, we pay particular attention to the phenomenon of anomalous diffusion. The combination of method of complex variables with scaling concepts allows us to investigate the mechanism of diffusion as well for intermediates times. We generalized the concept of the diffusion exponent to include other than the asymptotic power-law behaviour. A method is proposed to obtain the diffusion coefficient analytically through the introduction of a time scaling factor $\lambda$. We obtain also an exact expression for $\lambda$ for all kinds of diffusion. Moreover, we show that $\lambda$ is a universal parameter determined by the diffusion exponent. The results are then compared with numerical calculations and very good agreement is observed. We show the existence of two kinds of ballistic diffusion, one ergodic and another non-ergodic. The method is general and may be applied to many types of stochastic problem.

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1. Introduction

Scaling methods have been of large application in physics, in particular, in statistical mechanics. In a recent work [1], one scaling method to obtain asymptotic results for long time behaviour in anomalous diffusion was discussed. We revisit the method here and we call attention to its connexion with another stochastic phenomena, where memory is present. The study of systems with long-range memory reveals some physical phenomena that are still not well understood, especially in systems which are outside the state of equilibrium or those in which the existence of anomalous diffusion is verified [1–9]. Here, we show a simple analytical method which describes the behaviour of the diffusion for large and intermediate times. In order to
do that, we first generalize the concept of the diffusion exponent. Then, we present a conjecture to obtain, through the introduction of a time scaling factor $\lambda$, an analytical asymptotic result for the diffusion coefficient for long times. We obtain the scaling factor exactly and we show as well its universal behaviour. We derive a numerical method to obtain the correlation function of velocities for an ensemble of particles from any given memory. We compare both methods and obtain excellent agreement. The method has general application in the study of stochastic processes and it could be applied to several situations of physical interest.

2. Generalized Langevin’s equation

The generalized Langevin equation (GLE) is a stochastic differential equation which can be used to model systems driven by coloured random forces. For the velocity operator $v(t)$, this equation can be written as

$$m \frac{dv(t)}{dt} = -m \int_0^t \Gamma(t-t') v(t') \, dt' + \xi(t),$$

where $\Gamma(t)$ is the retarded friction kernel of the system, or the memory function. Here, $\xi(t)$ is a stochastic noise subject to the conditions $\langle \xi(t) \rangle = 0$, $\langle \xi(t) v(0) \rangle = 0$, and

$$C_\xi(t) = \langle \xi(t) \xi(0) \rangle = m^2 \langle v^2(t) \rangle \, \Gamma(t),$$

where $C_\xi(t)$ is the correlation function for $\xi(t)$, and the angular brackets denote an average over the ensemble of particles. Equation (2) is Kubo’s Fluctuation-Dissipation Theorem (FDT) \[10, 11\]. The presence of the kernel $\Gamma(t)$ allows us to study a large number of correlated processes. In the real world, the vast majority of problems are non-Markovian, i.e., there is correlation between the various stages of dynamic evolution. This property is what we call memory, and it makes remote events of the past important to dynamic events in the present time.

Using the GLEs it is possible to study the asymptotic behaviour of the second moment of the particle movement

$$\lim_{t \to \infty} \langle x^2(t) \rangle = 2D(t)t \sim t^\alpha,$$

which characterizes the type of diffusion presented by the system. Here, $D(t)$ is the diffusion coefficient as a function of time.

Moreover, for an asymptotic behaviour of the form

$$\lim_{t \to \infty} \langle x^2(t) \rangle \sim t^\alpha [\ln(t)]^\pm 1$$
we shall define respectively an $\alpha^\pm$ diffusive behaviour [1]. Here, the exponent $\alpha = \alpha^\pm$ arises in analogy with the critical exponents in a phase transition. For example, in the two-dimensional Ising model, the critical exponent for the specific heat is $\alpha = 0^+$ because it does not have a power law behaviour; rather it has $\ln |T - T_c|$ behaviour for temperatures $T$ close to the transition temperature $T_c$. This generalized nomenclature is pertinent here since there is quite a large number of possibilities of combinations for logarithmic and power-law behaviours.

In this way, the behaviour of $D(t)$ can be determined using

$$\lim_{t \to \infty} D(t) = \lim_{t \to \infty} \lim_{z \to 0} \int_0^t C_v(t') \exp(-zt') \, dt' = \lim_{z \to 0} \tilde{R}(z), \quad (5)$$

where $R(t) = C_v(t)/C_v(0)$, with $C_v(0) = 1$, and $\tilde{R}(z)$ is the Laplace transform of $R(t)$. For $t \to \infty$ and normal diffusion, this is the Kubo formula [11]. The limits can be justified using the final value theorem (FVT) for a Laplace transform, i.e., for any function $g(t)$ with Laplace transform $\tilde{g}(z)$ then

$$\lim_{t \to \infty} g(t) = \lim_{z \to 0} z \tilde{g}(z).$$

Now, a Laplace transform of the integral gives $\tilde{D}(z) = \tilde{R}(z)/z$, and we end up with the equation above.

Now we multiply Eq. (1) by $v(0)$ and take the average over the ensemble, with $\langle \xi(t)v(0) \rangle = 0$, to obtain a self-consistent equation for $R(t)$ in the form

$$\dot{R}(t) = - \int_0^t \Gamma(t - t') R(t') \, dt'. \quad (6)$$

We then Laplace transform Eq. (6) to get

$$\tilde{R}(z) = \frac{1}{z + \tilde{\Gamma}(z)}. \quad (7)$$

Time correlation functions play a central role in the non-equilibrium statistical mechanics in many areas, such as the dynamics of polymeric chains [13–19], metallic liquids [20], Lennard–Jones liquids [21], ratchet devices [22, 23], diffusion of spin waves in disordered systems [24], Heisenberg ferromagnets and dense fluids [25]. Consequently, to invert this transform, or a similar one, is crucial. Unfortunately, in most cases, it is not an easy task. In those situations, the use of numerical methods is an alternative to overcome this problem. Our main objective here is to show a process to obtain the asymptotic behaviour analytically. Although the method can be applied to several situations, we concentrate here on the analysis of diffusion.
3. Beyond the asymptotic method

We consider now the FVT for $D(t)$

$$\lim_{t \to \infty} D(t) = \lim_{z \to 0} z \tilde{D}(z) = \lim_{z \to 0} \tilde{R}(z).$$

We claim that after a “transient time” $\tau$, i.e., for $t > \tau$, the leading term for $D(t)$ will fulfil Eq. (5) within a given approximation. In this context, $t \to \infty$ is equivalent to $t \gg \tau$. Now, we imposed the scaling

$$z \to \lambda/t.$$  \hspace{1cm} (9)

In order to determine $\lambda$, we rewrite Eq. (5) as

$$\lim_{t \to \infty} D(t) = \lim_{t \to \infty} \tilde{R}(z = \lambda/t) = \lim_{t \to \infty} \frac{t}{f(t)},$$

where

$$f(t) = \lambda + t \tilde{\Gamma}(\lambda/t).$$  \hspace{1cm} (10)

The derivative of Eq. (10) yields

$$\lim_{t \to \infty} R_1(t) = \lim_{t \to \infty} \frac{d}{dt} D(t) = \lim_{t \to \infty} \left[ 1 - t \frac{d}{dt} \ln[f(t)] \right] / f(t),$$

while from the FVT for $R(t)$, we get

$$\lim_{t \to \infty} R_2(t) = \lim_{z \to 0} z \tilde{R}(z) = \lim_{t \to \infty} \frac{\lambda}{f(t)}.$$  \hspace{1cm} (11)

The relative difference

$$\Delta R(t) = \frac{R_2 - R_1}{R_2} = \left[ \lambda - 1 + t \frac{d}{dt} \ln[f(t)] \right] / \lambda$$

should evolve to zero as $t \to \infty$. For $\lambda \neq 0$, this yields the exact value

$$\lambda = 1 - \lim_{t \to \infty} t \frac{d}{dt} \ln[f(t)].$$  \hspace{1cm} (12)

The scaling works as long as the GLE, Eq. (7), works. To obtain $\lambda$, we need more information about $\tilde{\Gamma}(z)$, which may be different for every system. However, since our interest is in the asymptotic behaviour, we can expand $\tilde{\Gamma}(z)$, in Taylor or Laurent series around $z = 0$, in the form

$$\tilde{\Gamma}(z) \sim z^\nu [a - b \ln(z) - c/ \ln(z)],$$

where

$$f(t) = \lambda + t \tilde{\Gamma}(\lambda/t).$$

This allows us to determine $\lambda$ for each system. The above approach can be applied to a wide range of systems, with a variety of scaling laws, and is a powerful tool for understanding the asymptotic behaviour of dynamical systems.
where $a$, $b$, and $c$ are positive constants. Note that we give especial attention to $\ln(z)$, since it will give us the behaviour pointed out in Eq. (4). For $b = 0$, this gives a diffusion with exponent $\alpha$; for $b \neq 0$, this gives an $\alpha^-$, and for $a = b = 0$ and $c \neq 0$, we get an $\alpha^+$ diffusion. If $\tilde{I}(z)$ has another contribution, besides $\ln(z)$, that cannot be expanded at the origin we keep it and expand the other parts. However, most of the memories in the literature can be cast in the form of Eq. (16) for small $z$. Now, we introduce Eq. (16) into Eq. (15) to obtain $\lambda = \nu$ for $\nu < 1$, and $\lambda = 1$ for $\nu \geq 1$. Notice that it does not depend on $a$, $b$, or $c$, which suggests a universal behaviour.

In our conjecture, some points deserve attention: First, we are considering integrals, of the form of Eq. (5), where the function $R(t)$ is well behaved, and limited to $-1 < R(t) < 1$, since $C_v(t) \leq C_v(0)$. $R(t)$ is such that it always has a well-defined behaviour for finite $t$, even when the integral diverges as $t \to \infty$, as in superdiffusion. Second, $D(t)$ must have a leading term as $t \to \infty$, which determines the diffusion. For example, the inverse Laplace transform of $\tilde{R}(z)$ is

$$R(t) = \frac{1}{2\pi i} \int_{-i\infty + \eta}^{+i\infty + \eta} \tilde{R}(z) \exp(zt) dz. \quad (17)$$

Here, the real number $\eta$ is such that all the singularities lie at the left of the line joining the limits. Consider now Eq. (16) with $b = c = 0$, and $\nu \leq 1$; then $\lim_{z \to 0} \tilde{R}(z) \sim z^{-\nu}$, and

$$\lim_{t \to \infty} R(t) \propto t^{\nu-1} \int_{-i\infty + \eta'}^{+i\infty + \eta'} s^{-\nu} \exp(s) ds \propto t^{\nu-1}, \quad (18)$$

where we have done the transformations $s = zt$ and $\eta' = \eta/t$. For $\nu > 0$, the only pole is at $s = 0$, and the condition in $\eta'$ will be automatically satisfied. Now, by direct integration on Eq. (5), we obtain $D(t) \propto t^\nu$. From the scaling, we get the equivalent result

$$\lim_{t \to \infty} D(t) = \lim_{z \to 0} \tilde{R}(z = \lambda/t) \sim \lim_{t \to \infty} \tilde{R}(\lambda/t) \sim t^\nu. \quad (19)$$

Note that the above exact result is not only for power laws, but for any function behaving as a power law for large $t$. We confirm as well the relation $\alpha = \nu + 1$, obtained by Morgado et al. [3]. Our results can be readily expressed as

$$\lambda = \alpha - 1 = \alpha^\pm - 1 = \begin{cases} 
\nu, & -1 < \nu < 1, \\
1, & \nu \geq 1.
\end{cases} \quad (20)$$
The factor $\lambda$ depends only on the diffusion exponent $\alpha$, consequently it is universal. Moreover, it will be the same for $\alpha$ or $\alpha^\pm$. For normal diffusion $\alpha = 1$, or for $\alpha = 1^\pm$, $\lambda = 0$. However, we still can obtain the final value. Consider as an example the Langevin equation without memory; for that we have $R(t) = \exp(-\gamma t)$ and $\tilde{R}(z) = (\gamma + z)^{-1}$. From Eq. (10), we get

$$\lim_{t \to \infty} D(t) = \lim_{t \to \infty} \tilde{R}(\lambda/t) = \frac{t}{\gamma t + \lambda} = \gamma^{-1},$$

while direct integration gives

$$\lim_{t \to \infty} D(t) = \lim_{t \to \infty} \int_0^t R(t')dt' = \gamma^{-1}.$$

In this case, the scaling yields correctly the wanted final value.

Equation (6) imposes also some requirements on $R(t)$. First, its derivative must be null at the origin, i.e., the integral in the right-hand side must be null at $t = 0$. This is true except for non-analytical memories, such as $\delta$ functions. Indeed, we do not expect an exponential behaviour of the form of $R(t) = \exp(-\gamma|t|)$ with a discontinuous derivative at the origin [26, 27]. Second, in Eq. (1), for a bath of harmonic oscillators the noise can be obtained as [26]

$$\xi(t) = \int \sqrt{2k_B Tg(\omega)} \cos[\omega t + \phi(\omega)] d\omega,$$

where $0 < \phi(\omega) < 2\pi$ are random phases and $g(\omega)$ is the noise spectral density. The FDT yields

$$\Gamma(t) = \int g(\omega) \cos(\omega t) d\omega.$$

This shows that the memory is an even function of $t$. An analytical extension of $\tilde{\Gamma}(z)$ in the whole complex plane has the property $\tilde{\Gamma}(-z) = -\tilde{\Gamma}(z)$. Consequently, from Eq. (7), $\tilde{R}(-z) = -\tilde{R}(z)$, or $R(-t) = R(t)$. In short, it requires well-behaved functions and derivatives. Even functions have zero derivatives at the origin as required before.

We shall call attention here that for finite time one can as well obtain $\lambda(t)$ using Eq. (15) as a map in the form

$$\lambda_{n+1}(t) = F_l(\lambda_n(t)),$$

where

$$F_l = 1 - t \frac{d}{dt} \ln[f(t)] = t \frac{d}{dt} \ln \left[\tilde{R}(\lambda_n/t)\right].$$

For a given memory, this map converges readily for a final value of $\lambda(t)$ after few iterations.
4. Ergodicity

Diffusion is a very good lab for the study of ergodicity because we can obtain simple relations which allows us to get clean conclusions without doubt. Let us consider the spectral density

\[ g(\omega) = \begin{cases} \ b\omega^{1-\beta} \omega^\beta, & \omega \leq \omega_s, \\ 0, & \omega > \omega_s. \end{cases} \] (27)

This is a generalization of the Debye density of states. Here, \( b > 0 \) is a dimensionless constant, and \( \omega_s \) is a cutoff frequency. For \( \beta \neq 0 \) we get anomalous diffusion. In particular, for \( \beta = 1 \) we introduce Eq. (27) into Eq. (24) to obtain

\[ \Gamma(t) = b\omega_s^2 \left( \frac{\sin(\omega_s t)}{\omega_s t} + \frac{\cos(\omega_s t) - 1}{(\omega_s t)^2} \right), \] (28)

with the Laplace transform

\[ \tilde{\Gamma}(z) = \frac{b z}{2} \ln \left[ 1 + \left( \frac{\omega_s}{z} \right)^2 \right]. \] (29)

First, we have the analytical function \( D(t) = \tilde{\Gamma}(z = \lambda/t) \); second, from Eq. (15) we obtain \( \lim_{t \to \infty} \lambda = 1 \), exactly. This is a ballistic diffusion of the form of \( \alpha = 2^- \).

In figure 1, we plot \( \lambda(t) \) as a function of \( t \). We use the map Eq. (25) and the Laplace transform of the memory Eq. (29). After 20 iterations, the difference \( |\lambda_{n+1} - \lambda_n| \) becomes less than \( 10^{-12} \). For both curves, the plot shows the evolution of \( \lambda(t) \) towards 1. The convergence is faster as the ratio \( \omega_s/b \) increases.

Now, we compare the analytical asymptotics with a numerical solution of Eq. (6). To do this, we rewrite this equation in a discrete form, and then we expand it up to terms of the order of \( \Delta t^{2n} \) to obtain

\[ R(t + \Delta t) = R(t - \Delta t) + 2 \sum_{k=0}^{n} R^{(2k-1)}(t) \frac{\Delta t^{2k-1}}{(2k-1)!}, \] (30)

where \( R^{(n)}(t) \) is the time derivative of \( R(t) \) of the order of \( n \). Note that this expansion eliminates all the even derivatives. Now, we can obtain all \( R(t + \Delta t) \) from the sequence of the previous value of \( R(t) \), starting from \( R(0) = 1 \). From these values, it is possible to get the diffusion coefficient through direct integration of Eq. (5).
Fig. 1. Function $\lambda(t)$ as a function of time $t$. Time in arbitrary units. We use the map (25) and the memory (29), after 50 iterations, we get convergence. Curve a, $\omega_s = 1$, and $b = 1$; curve b, $\omega_s = 5$, and $b = 1/2$.

In Fig. 2, we plot the correlation function $R(t)$ as a function of time $t$. The curves correspond to the numerical solution, and are calculated using Eq. (30), and Eq. (28) with $\Delta t = 10^{-5}$. For curve a, $\omega_s = 1$, and $b = 1$; for curve b, $\omega_s = 5$ and $b = 1/2$.

Fig. 2. Correlation function $R(t)$ as a function of time $t$. We use the map (25) and the memory (28). Curve a, $\omega_s = 1$, and $b = 1$; curve b, $\omega_s = 5$ and $b = 1/2$.

In Fig. 3, we plot the diffusion coefficient $D(t)$ as a function of time $t$. The oscillatory curves corresponds to the numerical solution and are calculated from the data of Fig. 1. The curves without oscillations correspond to the analytical asymptotic limit, Eq. (10), with memory Eq. (29). Here, we see that the asymptotic curves are mean values of the oscillatory ones. In this range, the fit yields for curve a, $\lambda = 0.928 \pm 0.002$, and for curve b, $\lambda = 0.94822 \pm 0.00001$. We see in curve b that the two curves collapse onto a single one. Here, the transient time $\tau$ to which we refer before Eq. (10) is a
decreasing function of $b/\omega_s$. The value of $\lambda$ approaches the exact value 1 as the ratio $b/\omega_s$ decreases, or as time increases. This shows the efficiency of the scaling; even before convergence is fully established, curve a, the asymptotic curve gives us an average value that can be used to understand the main characteristics of the process.

![Diagram of diffusion coefficient $D(t)$ as a function of time $t$. Curve a, $\omega_s = 1$, and $b = 1$; curve b, $\omega_s = 5$, and $b = 0.5$. The oscillatory curves are the numerical result. The curves without oscillations are the analytical asymptotic limit.](image)

Consider now $\tilde{\Gamma}(z) = az$, exactly. That means $\tilde{R}(z) = [(1 + a)z]^{-1}$ or $R(t) = [1 + a]^{-1}$, and by direct integration, we get $D(t) = t/(1 + a)$ exactly. This is a ballistic $\alpha = 2$ diffusion. If we apply Eq. (10), we obtain the same result with $\lambda = 1$. Since from the relations (16) and (20) the value of $\lambda$ does not depend on $\ln (z)$, this result is exactly what we get from Eq. (29). There are important differences between the $\alpha = 2^-$ diffusion which, according to the Khinchin theorem [7, 28], is ergodic, and the $\alpha = 2$ diffusion, which does violate ergodicity. This distinction was not possible before the generalization of the diffusion exponent we present here.

### 5. Perspective and conclusion

In this work, we generalize the concept of the diffusion exponent and we propose a conjecture to investigate the asymptotic limits of anomalous diffusion through the introduction of a time scaling factor $\lambda$. We obtain the scaling parameter exactly and we show that it is universal and depends only on the diffusion exponent. We analyse the ballistic diffusions $\alpha = 2^-$ and $\alpha = 2$, both analytically and numerically. The method can be useful as well to analyse large amounts of data in stochastic processes [5], and in
different fields of science where it is necessary to inverse a Laplace transform of the form of Eq. (7). The phenomenon of diffusion also poses challenges in the understanding of fundamental concepts in statistical physics, such as entropy [6, 34] general properties as the correlation function [26], ergodicity [7, 9, 28–32], the Khinchin theorem [7, 28], and FDT [33].

In biological systems where motion [35] and pattern formation [36, 37] is a prime, diffusion has still an important contribution to do. A very broad and growing area is that of synchronization [38–40], where we expect the scaling may yield more analytical results.

In nonlinear phenomena, such as growth and etching [41–44], analytical results are rather difficult to obtain. For example, the KPZ equation has an exact solution for one dimension. However, no solutions for higher dimensions have been found, as in many other areas of non-equilibrium physics where even not exact solutions can be considered major results. In this way, we hope that this work may inspire research into similar asymptotic limits.

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