

A universality class in Markovian persistence

This article has been downloaded from IOPscience. Please scroll down to see the full text article.

2000 J. Phys. A: Math. Gen. 33 7025

(<http://iopscience.iop.org/0305-4470/33/40/301>)

View [the table of contents for this issue](#), or go to the [journal homepage](#) for more

Download details:

IP Address: 171.66.16.123

The article was downloaded on 02/06/2010 at 08:33

Please note that [terms and conditions apply](#).

A universality class in Markovian persistence

O Deloubrière

Laboratoire de Physique Théorique†, Bâtiment 210, Université de Paris-Sud, 91405 Orsay Cedex, France

Received 12 April 2000

Abstract. We consider the class of Markovian processes defined by the equation $dx/dt = -\beta x + \sum_k z_k \delta(t - t_k)$. Such processes are encountered in systems (such as coalescing systems) where dynamics creates discrete upward jumps at random instants t_k and of random height z_k . We observe that the probability for these processes to remain above their mean value during an interval of time T decays as $e^{-\theta T}$, defining θ as the persistence exponent. We show that θ takes the value β , which thereby extends the well known result of the Gaussian noise case to a much larger class of non-Gaussian processes.

1. Introduction

Let us consider a stationary random process $x(t)$ and suppose that at $t = 0$ this process is above a given threshold X . What is the probability $Q(T)$ that the process has never crossed the threshold X up to time $t = T$? In many examples of statistical mechanics we observe

$$Q(T) \sim e^{-\theta_X T} \quad T \rightarrow \infty \quad (1)$$

where θ_X is a function of the threshold. The question of determining θ_X is called the ‘persistence problem’ and has recently been the primary focus of many papers in the study of out-of-equilibrium systems. For a short review the reader can see [1] and references therein. In many persistence problems, only the case where $X = \langle x \rangle$ is considered and usually one writes θ for $\theta_{\langle x \rangle}$.

It often happens that stationary processes can be obtained from physical nonstationary systems by rescaling the space variable and taking the new time variable $t = \log \tau$ [1], where τ is the physical time. In terms of the latter the persistence probability decays as a power law, thus defining θ_X as a new independent exponent [2] called the persistence exponent.

In this paper we present a class of stationary Markov processes $x(t)$ that share a universal persistence exponent when $X = \langle x \rangle$. This class of processes is defined by the linear equation of motion

$$\frac{dx}{dt} = -\beta x + \eta(t) \quad (2)$$

where $\eta(t)$ is a noise term consisting of upward jumps uniformly distributed along the time axis with density ρ ,

$$\eta(t) = \sum_{k=-\infty}^{\infty} z_k \delta(t - t_k). \quad (3)$$

† Laboratoire associé au Centre National de la Recherche Scientifique—UMR 8627.

The jump heights z_k are independent identically distributed random variables of law $R(z)$ such that $\langle z \rangle$ exists. In other words the jump times simply define a Poisson process [6] and $\eta(t)$ is a succession of steps of random height and at random moments.

If the jump heights are all equal to a , the limit $a \rightarrow 0, \rho \rightarrow \infty$ with ρa^2 fixed transforms the shifted noise $\eta(t) - \rho a$ into Gaussian white noise. In this case equation (2) is the usual Langevin equation, which is therefore included as a limit in our class of processes.

The interest of this paper is to solve the persistence problem for the process defined by equation (2). Simulations have shown that for large T $Q(T)$ decays exponentially with an inverse decay rate θ_X depending on the threshold and on the shape of $R(z)$, such as a Dirac peak, an exponential or a power law. However, if $X = \langle x \rangle = \rho \langle z \rangle / \beta$ the result $\theta = \beta$ appears for any law $R(z)$. This equality was known, until now, only for the Langevin equation, with Gaussian $\eta(t)$ [2–4].

2. A simple method for the calculation of θ

In this section we show that θ_X is the lowest eigenvalue of a differential operator. Then when the threshold is set to $X = \langle x \rangle$, the persistence problem is solved straightforwardly.

2.1. Transformation into an eigenvalue problem

Let $P^*(x, t)$ be the probability distribution of x at time t . It is the solution of the master equation

$$\partial_t P(x, t) = \beta \partial_x (x P(x, t)) + \rho \int_0^\infty P(x - z, t) R(z) dz - \rho P(x, t). \quad (4)$$

For the persistence problem only the subclass of functions $x(t)$ that are greater than X during the time interval T is relevant. Let $P_+^*(x, t) dx$ be the probability that $x(t')$ has not crossed the level X for $t' \in [0, t]$ (given that $x(0) > X$) and that $x(t)$ is in the interval $[x, x + dx]$. $P_+^*(x, t)$ is then solution of the same master equation but with the absorbing boundary condition $P_+^*(x, t) = 0$ when $x < X$. Because of this condition, $P_+^*(x, t)$ will decay to zero as $t \rightarrow \infty$. We will write it as a superposition of decay modes $P_+^\lambda(x) e^{-\lambda x t}$, where $P_+^\lambda(x)$ is an eigenfunction satisfying the eigenvalue equation corresponding to equation (4)

$$-\lambda_X P_+^\lambda(x) = \beta \frac{d}{dx} (x P_+^\lambda(x)) + \rho \int_0^\infty P_+^\lambda(x - z) R(z) dz - \rho P_+^\lambda(x) \quad (5)$$

with the condition $P_+^\lambda(x) = 0$ when $x < X$. Let λ_X^* be the eigenvalue of the slowest decaying mode. Then we have

$$Q(T) \sim e^{-\lambda_X^* T} \int_X^\infty P_+^{\lambda^*}(x) dx \quad \text{as } T \rightarrow \infty \quad (6)$$

so that

$$\theta_X = \lambda_X^* \quad (7)$$

if $\int_X^\infty P_+^{\lambda^*}(x) dx$ exists.

Finally let us call $P_+^\theta(x) \equiv P_+^{\lambda^*}(x)$ the persistence mode. It is necessary to have

$$P_+^\theta(x) \geq 0 \quad \text{for } x \geq X \quad \text{and} \quad \int_X^\infty P_+^\theta(x) dx < \infty \quad (8)$$

in order to define $P_+^\theta(x, t)$ as a probability distribution. Moreover the definition of the persistence problem ensures that $x(0)$ is chosen within the stationary distribution of the process. This distribution is such that $\langle x \rangle$ exists. Therefore we will only focus on solutions $P_+^\theta(x)$ with

$$\int_X^\infty x P_+^\theta(x) dx < \infty. \quad (9)$$

As a result solving the persistence problem is now equivalent to calculating the lowest eigenvalue λ_x^* of equation (5) that satisfies (8) and (9).

2.2. Calculation of θ

We will now show that $\lambda_{\langle x \rangle}^* = \beta$ irrespective of the jump size distribution. To this end we define

$$f_{\lambda X}^{(n)} = \int_X^\infty x^n P_+^\lambda(x) dx \quad n = 0, 1, \dots \tag{10}$$

and we integrate equation (5) on x from X to ∞ . As $P_+^\theta(x) = 0$ when $x < X$, we obtain the first relation

$$\lambda_X f_{\lambda X}^{(0)} = \beta X P_+^\lambda(X). \tag{11}$$

We now multiply equation (5) by x and integrate as before. After one integration by parts we obtain

$$\lambda_X f_{\lambda X}^{(1)} = \beta X^2 P_+^\lambda(X) + (\beta + \rho) f_{\lambda X}^{(1)} - \rho \int_X^\infty dx \int_0^\infty dz x P_+^\lambda(x - z) R(z). \tag{12}$$

Using the absorbing boundary condition, we can write

$$\int_X^\infty dx \int_0^\infty dz x P_+^\lambda(x - z) R(z) = \int_X^\infty dy P_+^\lambda(y) \int_0^\infty dz (y + z) R(z) \tag{13}$$

and the RHS of this equation simply reduces to $f_{\lambda X}^{(1)} + \langle z \rangle f_{\lambda X}^{(0)}$. Then equations (12) and (13) give the second relation

$$\lambda_X f_{\lambda X}^{(1)} = \beta X^2 P_+^\lambda(X) + \beta f_{\lambda X}^{(1)} - \rho \langle z \rangle f_{\lambda X}^{(0)}. \tag{14}$$

We then eliminate $P_+^\lambda(X)$ from equations (11) and (14) and obtain finally using $\rho \langle z \rangle = \beta \langle x \rangle$

$$(f_{\lambda X}^{(1)} - X f_{\lambda X}^{(0)}) \lambda_X = \beta (f_{\lambda X}^{(1)} - \langle x \rangle f_{\lambda X}^{(0)}) \tag{15}$$

which is valid for any λ_X for which $f_{\lambda X}^{(1)} < \infty$. One can remark that equations (11) and (14), and therefore equation (15), could not be obtained if the jump heights were not all positive. So this property is essential in this paper.

Let us now take X equal to the average of the process: $X = \langle x \rangle$. Equation (15) shows that we must have

$$\lambda_{\langle x \rangle} = \beta \quad \text{or} \quad f_{\lambda_{\langle x \rangle}}^{(1)} / f_{\lambda_{\langle x \rangle}}^{(0)} = \langle x \rangle. \tag{16}$$

If in equation (16) we set $\lambda_{\langle x \rangle} = \lambda_{\langle x \rangle}^* = \theta$, we see from equations (8) and (10) that $f_{\theta \langle x \rangle}^{(1)} / f_{\theta \langle x \rangle}^{(0)}$ is the average of the process x in the persistence mode, i.e. given that it remains always above $X = \langle x \rangle$. This quantity is therefore larger than $\langle x \rangle$ and the second one of equation (16) cannot be satisfied. Finally the only possibility is $\lambda_{\langle x \rangle}^* = \beta$ which proves the main result of this note.

For $\lambda_{\langle x \rangle} \neq \lambda_{\langle x \rangle}^*$ equation (16) shows that

$$\int_{\langle x \rangle}^\infty dx (x - \langle x \rangle) P_+^\lambda(x) = 0 \quad \lambda_{\langle x \rangle} > \lambda_{\langle x \rangle}^*. \tag{17}$$

It follows that for $\lambda_{\langle x \rangle} \neq \lambda_{\langle x \rangle}^*$ any eigenfunction $P_+^\lambda(x)$ with $f_{\lambda X}^{(1)} < \infty$ must have at least one zero for $\langle x \rangle < x < \infty$. One can also see equation (17) as an orthogonality relation between two eigenvectors of equation (5) that are $x - \langle x \rangle$ (with the eigenvalue -2β) and $P_+^\lambda(x)$.

One final remark is necessary. We have studied the persistence problem in the most natural situation, that is when the condition $f_{\lambda \langle x \rangle}^{(1)} < \infty$ is fulfilled. However, one could be interested in other cases where the initial distributions $P(x, 0)$ have a large tail for $x \rightarrow \infty$ that do not ensure

this condition. If one starts with such a distribution, one can expect the persistence problem to be governed by the shape of this initial distribution, with different answers depending on the choice of the latter. Therefore large initial distributions where $f_{\lambda X}^{(1)} = \infty$ are simply cases of no interest in our study.

2.3. Perturbation for X close to $\langle x \rangle$

Very few cases in statistical physics offer the possibility of calculating the quantity

$$\langle x \rangle_X = f_{\theta X}^{(1)} / f_{\theta X}^{(0)} \quad (18)$$

which is the average of the process in the persistence mode. However as $f_{\theta \langle x \rangle}^{(n)} = f_{\beta \langle x \rangle}^{(n)}$ one can assume that

$$f_{\theta X}^{(n)} = f_{\beta \langle x \rangle}^{(n)} + \tilde{f}_{\theta X}^{(n)} \quad (19)$$

where $\tilde{f}_{\theta X}^{(n)} \rightarrow 0$ if $|X - \langle x \rangle| \rightarrow 0$. Rewriting equation (15) for $\lambda_X = \lambda_X^*$ one obtains

$$\theta_X = \beta \frac{f_{\theta X}^{(1)} - \langle x \rangle f_{\theta X}^{(0)}}{f_{\theta X}^{(1)} - X f_{\theta X}^{(0)}} \quad (20)$$

which can also be written

$$\theta_X = \beta \frac{1}{1 - \frac{X - \langle x \rangle}{\langle x \rangle_X - \langle x \rangle}}. \quad (21)$$

Finally one can suppose that X is close to $\langle x \rangle$ and perturbs equation (21) to find

$$\theta_X = \beta \left(1 + \frac{X - \langle x \rangle}{\langle x \rangle_{\langle x \rangle} - \langle x \rangle} + \dots \right) \quad |X - \langle x \rangle| \rightarrow 0. \quad (22)$$

Unfortunately we cannot go further here. We do not know how to calculate $\langle x \rangle_{\langle x \rangle}$ except in the Gaussian limit, which will be the focus of the next section.

3. The special case of Gaussian noise

Our goal here is to illustrate the method of section 2 in the special known case of Gaussian process. If one considers constant jumps with $R(z) = \delta(z - a)$ and sets $\rho a^2 = \Gamma$, one can take the limit $a \rightarrow 0$ with Γ fixed in equation (4) and find

$$\partial_t P(x, t) = \beta \partial_x ((x - \langle x \rangle) P(x, t)) + \frac{\Gamma}{2} \partial_{xx} P(x, t) + \mathcal{O}(a). \quad (23)$$

In this limit equation (23) leads to the Fokker–Planck equation for the zero-average process $x(t) - \langle x \rangle$, which hence becomes the stationary Gaussian–Markovian process, also called the Ornstein–Uhlenbeck process. This shows that the Gaussian case, which is of high interest in statistical physics, is included as a special limit in the class of processes studied in this paper.

3.1. Calculation of θ

For simplicity we will now write x instead of $x - \langle x \rangle$ and X instead of $X - \langle x \rangle$. We thus consider the zero-average Gaussian process $x(t)$ and we want to determine the persistence exponent θ_X related to the probability of having never crossed the level X up to time T . As in section 2 it is equivalent to finding the lowest eigenvalue λ_X^* of

$$-\lambda_X P_+^\lambda(x) = \beta \frac{d}{dx} (x P_+^\lambda(x)) + \frac{\Gamma}{2} \frac{d^2}{dx^2} P_+^\lambda(x) \quad P_+^\lambda(X) = 0. \quad (24)$$

The solution of this equation is

$$P_+^\lambda(x) = e^{-y^2/4}(A_1 D_{\lambda/\beta}(y) + A_2 D_{\lambda/\beta}(-y)) \tag{25}$$

where $y = \sqrt{2\beta/\Gamma}x$ is dimensionless and D_ν is the parabolic cylinder function of index ν [5]. The integration constants $A_{1,2}$ are such that

$$A_1 D_{\lambda/\beta}(Y) + A_2 D_{\lambda/\beta}(-Y) = 0 \tag{26}$$

with $Y = \sqrt{2\beta/\Gamma}X$. After one performs the same integrations as in section 2, equation (24) gives the two relations

$$\lambda_X f_{\lambda X}^{(0)} = \frac{\Gamma}{2} \frac{dP_+^\lambda}{dx} \Big|_X \tag{27}$$

$$\lambda_X f_{\lambda X}^{(1)} = \beta f_{\lambda X}^{(1)} + \frac{\Gamma}{2} X \frac{dP_+^\lambda}{dx} \Big|_X. \tag{28}$$

Eliminating $\frac{dP_+^\lambda}{dx} \Big|_X$ from equations (27) and (28) one finally obtains

$$(f_{\lambda X}^{(1)} - X f_{\lambda X}^{(0)})\lambda_X = \beta f_{\lambda X}^{(1)} \tag{29}$$

which is strictly equivalent to equation (15). In the same way as in section 2 it now follows that if $X = 0$ then $\theta = \lambda_0^* = \beta$, which is the known result of Gaussian persistence [2–4].

Another remark can be made here. The solution of equation (24) that vanishes at the origin is proportional to (with $A_1 = -A_2 = A$)

$$\tilde{P}(y) = Ae^{-y^2/4}(D_{\lambda/\beta}(y) - D_{\lambda/\beta}(-y)) \tag{30}$$

which can be reduced to

$$\tilde{P}(y) = A' ye^{-y^2/2} {}_1F_1((1 - \lambda/\beta)/2, 3/2, y^2/2) \tag{31}$$

where A' is simply a numerical constant [5]. Using the result [7]

$$\int_0^\infty e^{-u} u^{1/2} {}_1F_1((1 - \nu)/2, 3/2, u) du = 0 \quad \nu > 1 \tag{32}$$

one has immediately $f_{\lambda 0}^{(1)} = 0$ for any $\lambda > \beta$, which is a direct proof of equation (17) in the Gaussian case.

3.2. Perturbation for small X

Our method offers a faster way than the one presented in [3], in the Gaussian case, to calculate θ_X when X is close to $\langle x \rangle$ (which is zero here).

As $P_+^\beta(x) = A' ye^{-y^2/2}$ [3], one finds $\langle x \rangle_{(x)} = \sqrt{\pi\Gamma/4\beta}$ and equation (22) immediately gives

$$\theta_X = \beta \left(1 + 2\sqrt{\frac{\beta}{\pi\Gamma}} X + \dots \right) \quad X \rightarrow 0 \tag{33}$$

which is the well known perturbative result for the Gaussian persistence exponent [3].

4. Application

We give here some examples of Markovian but non-Gaussian processes defined by equation (2).

4.1. 1D Percolation

Let us consider an infinite 1D lattice. Each site can be occupied or empty. At an initial time, say $t = 0$, we start with a totally empty lattice. Then, at each time step, one site is chosen randomly among non-occupied ones and is then set occupied forever. The density of occupied sites p is therefore an increasing function of time. It goes from zero to unity as time goes from zero to infinity (in the limit of an infinite lattice, which is the case we consider here). We define clusters in a standard way, that is as sets of adjacent occupied sites. It is then possible to study percolation on this lattice as a dynamical process. De Freitas and dos Santos Lucena [10] have already performed this study numerically.

The interest here is to consider a semi-infinite lattice and the cluster that contains its origin. When p goes from zero to unity, the size $s(p)$ of this cluster will grow from zero to infinity. It is very easy to see that

$$\langle s \rangle = \frac{p}{1-p} \quad \text{and} \quad \sqrt{\Delta s^2} = \frac{\sqrt{p}}{1-p}. \quad (34)$$

The evolution equation for $s(p)$ is also quite simple. This random process is constant except when new sites are connected to the cluster, which creates upward jumps of height j . Hence

$$\frac{ds}{dp} = \sum_{k=0}^{\infty} j_k \delta(p - p_k) \quad (35)$$

where k is the jump index. If $\tilde{R}(j)\Delta p$ is the probability of having a jump j when the density of occupied sites goes from p to $p + \Delta p$ then

$$\tilde{R}(j)\Delta p = p^{j-1}(1-p) \frac{\Delta p}{1-p}. \quad (36)$$

The quantity $\Delta p/(1-p)$ is simply the probability of making a jump during Δp . The normalized distribution law for the jumps is then

$$\tilde{R}(j) = p^{j-1}(1-p) \quad (37)$$

and the density of jumps is $\rho(p) = 1/(1-p)$. As p tends to 1, the typical timescale of the dynamics of the process $s(p)$, which is $\sim 1/\rho$, goes to zero. The aim is now to deal with a stationary process. If the new timescale is

$$t = \log \frac{1}{1-p} \quad (38)$$

the jump times t_k are now uniformly distributed with unit density along the t -axis. Then we rescale the process $s(p)$, setting

$$x(t) = \frac{s(p)}{\sqrt{\Delta s^2}} \quad \text{and} \quad z_k = \frac{j_k}{\sqrt{\Delta s^2}} \quad (39)$$

and we take the limit $p \rightarrow 1$ or equivalently $t \rightarrow \infty$. The jumps z have now the probability density $R(z)$ so $\tilde{R}(j)\Delta j = R(z)\Delta z$. In the limit $p \rightarrow 1$

$$R(z) = e^{-z}. \quad (40)$$

If we write equation (35) in terms of x , t and z , the evolution equation is finally

$$\frac{dx}{dt} = -x(t) + \sum_{k=0}^{\infty} z_k \delta(t - t_k) \quad (41)$$

in the limit $t \rightarrow \infty$. This gives a second example of noise of equation (3) with an exponential distribution law for the jump height. The value of the persistence exponent is then $\theta = 1$.

4.2. Other examples

A second example of noise defined by equation (3) is given by the study of random walks. The time $t(S)$ needed for the walker to visit S distinct sites on a 1D lattice is a random process which, under the same kind of scaling as in the last subsection, satisfies equation (2) [8]. The jump height distribution $R(z)$ is then a power law with a cutoff. In this case the value of θ is 2. We will not go further about this result in this paper and the reader can see [8] and [9] for more details.

Finally the process $x(t)$ defined by equation (2) is simply the electric signal that one would observe in an RC circuit receiving random charges due to, for instance, the presence of radioactive elements. The persistence exponent is then $\theta = 1/RC$, that is the inverse relaxation time of the circuit.

5. Conclusion and discussion

With a very simple calculation we are able to find the persistence exponent of a large class of stationary Markov processes which includes the Gaussian case. As this quantity is universal within this entire class, it demonstrates the existence of universality classes related to non-Gaussian persistence exponents.

To go further one should study the properties of the master equation (4) because the level-dependent exponent θ_X is entirely determined by the knowledge of $\langle x \rangle_X$. However, the nontrivial asymptotic behaviour $\theta_X \sim X \log X + X^{1/3}(\log X)^{2/3} + \dots$, $X \rightarrow \infty$, found in [9] when the jumps are constant shows that this investigation is likely to be very difficult.

Acknowledgments

The author thanks Professor H J Hilhorst for numerous fruitful discussions about this paper and also Professor L dos Santos Lucena of the Universidade Federal do Rio Grande do Norte, Natal, Brazil, for pointing out the relation between 1D percolation and noise with discrete upward jumps.

References

- [1] Majumdar S N 1999 *Curr. Sci.* **77** 370
- [2] Majumdar S N, Bray A J, Cornell S J and Sire C 1996 *Phys. Rev. Lett.* **77** 2867
- [3] Sire C, Majumdar S N and Rüdinger A 2000 *Phys. Rev. E* **61** 1258
- [4] Oerding K, Cornell S J and Bray A J 1997 *Phys. Rev. E* **56** R25
- [5] Abramowitz M and Stegun I A 1965, 1972 *Handbook of Mathematical Methods* 9th edn (New York: Dover)
- [6] Van Kampen N G 1992 *Stochastic Processes in Physics and Chemistry* (Amsterdam: North-Holland)
- [7] Gradshteyn I S and Ryzhik I M 1965, 1980 *Table of Integrals, Series and Products* 4th edn (New York: Academic)
- [8] Gomes S R Jr, Deloubrière O and Hilhorst H J unpublished
- [9] Deloubrière O and Hilhorst H J 2000 *J. Phys. A: Math. Gen.* **33** 1993
- [10] de Freitas J E and dos Santos Lucena L 1999 *Physica A* **266** 81