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## 1 **From the central limit theorem to Stirling's formula... and back**

- 1) We take here advantage of the stability of Poisson random variables upon addition, to deduce Stirling's approximation for  $n!$  when  $n$  is large, from the central limit theorem.
  - a) We denote  $\mathcal{P}(\lambda)$  the law of a Poisson random variable with parameter  $\lambda$ . What is the law for the sum of two independent such variables? In the remainder, we take  $\lambda = 1$  for simplicity.
  - b) We define  $S_n = \sum_{i=1}^n x_i$  where the  $x_i$  are IID Poisson variables  $\mathcal{P}(\lambda = 1)$ . How does the central limit theorem constrain the distribution of  $S_n$  for  $n$  large?
  - c) By computing the probability that  $n - 1/2 < S_n < n + 1/2$  in two ways, show that when  $n \rightarrow \infty$

$$n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \quad (1)$$

- 2) We consider the simple random walk starting from the origin, where at each discrete time  $n = 1, 2, \dots$ , a jump  $\pm 1$  is made, with equal probability  $1/2$ . Take  $n$  even.
  - a) What is the probability to return to the origin after  $n$  steps?
  - b) Use Stirling's formula to show that this probability behaves for large  $n$  as  $\frac{\sqrt{2}}{\sqrt{\pi n}}$
  - c) What is the probability that the walker is at position  $m$  after  $n$  steps, with  $m = 0, \pm 2, \pm 4 \ll n$ ?
  - d) Recover the central limit theorem result for the above probability. Beware "the factor 2".

## 2 Consistency of Itō-Doblin and Stratonovich calculus

For a Langevin equation with additive noise, such as

$$\frac{dx}{dt} = \mu F(x) + \sqrt{2D} \eta(t) \quad \text{with} \quad \langle \eta(t) \rangle = 0 \quad \text{and} \quad \langle \eta(t) \eta(t + \tau) \rangle = \delta(\tau), \quad (2)$$

and an arbitrary function  $\varphi(x)$ , check the consistency of Itō-Doblin and Stratonovich calculus for computing

$$\left\langle \frac{d\varphi(x(t))}{dt} \right\rangle. \quad (3)$$

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## 3 Itō-Doblin, Stratonovich and Wick

We consider the Wiener process  $x(t)$  with diffusion coefficient  $D$ . Is this process Gaussian? Invoking Wick theorem, compute the explicit time dependence of  $\langle x^4 \rangle$  and  $\langle x^6 \rangle$ . Recover these results from both Itō-Doblin and Stratonovich routes, computing

$$\left\langle \frac{dx^4}{dt} \right\rangle \quad \text{and} \quad \left\langle \frac{dx^6}{dt} \right\rangle. \quad (4)$$

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## 4 The Wiener process from the scaling limit of a random walk

We consider the symmetric random walk on  $\mathbb{Z}$ , with probabilities  $1/2$  to jump left or right, at each step. Starting from the origin, the walker has position  $x_n$  after  $n$  steps. From  $x_n$ , we define by linear interpolation the continuous time process  $x_c(\tilde{n})$ , where  $\tilde{n}$  is now real, and  $x_c(\tilde{n}) = x_n$  when  $\tilde{n} = n$  is an integer. From this, we introduce the process

$$x_\varepsilon(t) = \varepsilon x_c\left(\frac{t}{\Delta t}\right). \quad (5)$$

What is the appropriate scaling limit, where  $\varepsilon \rightarrow 0$ ,  $\Delta t \rightarrow 0$ , such that  $x_\varepsilon(t) \rightarrow W(t)$ , the Wiener process?

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## 5 Markovian or non Markovian?

Our interest goes to a persistent random walk in 1D, where the probability to make a step in the same direction as the previous one is  $p > 1/2$ . With probability  $1 - p$ , the step is made in the reverse direction. Does the position of the walker after  $n$  steps,  $x_n$ , define a Markov process? How can define a Markov process here?

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## 6 Position and velocity processes in the Langevin equation

We start with the 1D Langevin equation:

$$m \frac{d^2x}{dt^2} = -m\gamma \frac{dx}{dt} + R(t), \quad (6)$$

where  $R(t)$  is a Gaussian process with  $\langle R(t) \rangle = 0$  and  $\langle R(t_1)R(t_2) \rangle = \Gamma m^2 \delta(t_1 - t_2)$ . The goal is to study here the position and velocity processes that ensue,  $x(t)$  and  $v(t)$ . Two types of initial conditions will be addressed:

(i) at  $t = 0$ ,  $x(0) = 0$  and  $v(0) = v_0$  (fixed initial velocity);

(ii) at  $t = 0$ ,  $x(0) = 0$  and  $v(0)$  are distributed according to the equilibrium Maxwellian.

1) Velocity process.

- a) Express  $v(t)$  as a function of  $R(t)$ , for a given realization of  $R(t)$ .
  - b) Characterize explicitly the process, ie write the  $n$ -times probability densities for all  $n$ , for each of the initial conditions (i) and (ii). Is  $v(t)$  Markovian? Gaussian? Stationary?
- 2) Position process.
- a) Express  $x(t)$  as a function of  $v(t)$ .
  - b) Characterize explicitly the process  $x(t)$ , for each of the initial conditions (i) and (ii). Is  $x(t)$  Markovian? Gaussian? Stationary?

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## 7 Doob's theorem (1942)

The goal here is to show that *a Gaussian stationnary process is Markovian iff its autocorrelation function is exponential.*

- 1) We consider first a Gaussian, stationnary and Markovian process. For the sake of simplicity, we take the mean value to vanish and the variance to be unity, without loss of generality.
- a) Write explicitly the one-time probability distribution function  $p(x)$ .
  - b) Write the transition probability  $p(y, \tau|x, 0)$ . It involves four parameters that we set out to determine in the following questions.
  - c) Express the normalization condition of the transition probability.
  - d) Write the consistency condition fulfilled by  $p(x)$  and  $p(y, \tau|x, 0)$ .
  - e) Show then that

$$p(y, \tau|x, 0) = \frac{1}{\sqrt{2\pi(1-\gamma^2)}} e^{-\frac{(y-\gamma x)^2}{2(1-\gamma^2)}}, \quad (7)$$

where  $\gamma$  is some parameter.

- f) Using previous results, compute the correlation function  $\langle x(0)x(\tau) \rangle$  as a function of  $\gamma$ . What is the physical significance of  $\gamma$ ?
  - g) Write Chapman-Kolmogorov equation for the process, and show that the correlation function is exponential.
- 2) Show the reciprocal statement: a process that is Gaussian, stationnary, and with an exponential correlation function, is Markovian.

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## 8 Fokker-Planck from Itō-Doblin calculus

We consider the Langevin equation  $\dot{x} = \mu F(x) + \sqrt{2D}\eta(t)$  where  $\eta(t)$  is a Gaussian white noise with correlation  $\langle \eta(t)\eta(t') \rangle = \delta(t-t')$ . Take an arbitrary test function  $\varphi(x)$  and compute  $d\langle \varphi(x) \rangle / dt$ . Using the fact that

$$\langle \varphi(x(t)) \rangle = \int dx P(x, t) \varphi(x) \quad (8)$$

where  $P(x, t)$  is the position p.d.f at time  $t$ , derive the Fokker-Planck equation for  $P(x, t)$ .

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## 9 Several solutions to the diffusion equation

We aim here at solving the diffusion equation

$$\frac{\partial P}{\partial t} = D \frac{\partial^2 P}{\partial x^2}, \quad t \geq t_0, \quad \text{with the initial condition } P(x, t_0 | x_0, t_0) = \delta(x - x_0). \quad (9)$$

We thus work with the conditional density  $P(x, t | x_0, t_0)$ .

- 1) A first approach is to search for a scaling solution, with a form that would exhibit diffusive scaling. This amounts to choosing

$$P(x, t | x_0, t_0) = \psi(t) \varphi\left(\frac{x - x_0}{\sqrt{t - t_0}}\right). \quad (10)$$

Show that normalization of  $P$  specifies the time dependence of  $\psi(t)$ . Injecting the above form into (9), obtain and solve the ordinary differential equation fulfilled by  $\varphi$ , to get finally

$$P(x, t | x_0, t_0) = \frac{1}{\sqrt{4\pi D(t - t_0)}} \exp\left[-\frac{(x - x_0)^2}{4D(t - t_0)}\right]. \quad (11)$$

- 2) A second method consists in taking the Laplace transform of Eq. (9). With  $\tau = t - t_0$ , we introduce

$$\tilde{P}(x, s) = \mathcal{L}_s [P(x, \tau)] = \int_0^\infty P(x, \tau) e^{-s\tau} d\tau. \quad (12)$$

Solve for  $\tilde{P}(x, s)$ . Using the inverse Laplace transform

$$\mathcal{L}_s^{-1}\left[\frac{e^{-a\sqrt{s}}}{\sqrt{s}}\right] = \frac{e^{-a^2/(4\tau)}}{\sqrt{\pi\tau}}, \quad (13)$$

recover Eq. (11). For completeness, show relation (13), e.g. with a complex analysis approach.

- 3) A third method bears some similarities with quantum mechanics. We consider the operator  $\hat{H} \equiv -D \partial^2 / \partial x^2$ . Assume first that the spectrum of  $\hat{H}$  is discrete, with a set of eigenvalues  $\lambda$  and associated eigenfunctions  $\psi_\lambda(x)$ . We assume that the latter form a complete set:

$$\sum_{\lambda} \psi_{\lambda}(x) \psi_{\lambda}^*(x_0) = \delta(x - x_0), \quad (14)$$

where the star denotes complex conjugate. Write the orthogonality condition between the eigenfunctions. Show that

$$P(x, t | x_0, t_0) = \sum_{\lambda} e^{-\lambda(t-t_0)} \psi_{\lambda}(x) \psi_{\lambda}^*(x_0). \quad (15)$$

Show that in the present case though, the spectrum is continuous, indexed by wavenumber  $k$  such that  $\lambda = Dk^2$ , with (normalized) eigenfunctions

$$\psi_k(x) = \frac{1}{\sqrt{2\pi}} e^{ikx}. \quad (16)$$

Rewriting Eq. (15) in the present situation as

$$P(x, t | x_0, t_0) = \int_{-\infty}^{\infty} e^{-Dk^2 t} e^{ik(x-x_0)} \frac{dk}{2\pi}, \quad (17)$$

recover the solution (11). This method leads here to a Fourier Transform approach (check this).

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## 10 Ornstein-Uhlenbeck, linear response, fluctuation-dissipation theorem

The Ornstein-Uhlenbeck process describes the velocity relaxation of a Brownian object subject to viscous friction, in the absence of an external force. Alternatively, the same framework also applies to an overdamped dynamics in a harmonic potential

$$U(x) = \frac{\kappa}{2} x^2. \quad (18)$$

We will assume that a small  $x$ -independent force is applied to the “particle”, so that

$$\frac{dx}{dt} = -\mu\kappa x + \mu f(t) + \sqrt{2D} \xi(t), \quad (19)$$

with  $\xi(t)$  a Gaussian white noise:

$$\langle \xi(t) \rangle = 0 \quad ; \quad \langle \xi(t) \xi(t') \rangle = \delta(t - t'). \quad (20)$$

- 1) What is the physical meaning of  $D$  and  $\mu$ ? How are they related at equilibrium? Equilibrium refers here to the suspending fluid in which the particle is immersed.
- 2) Solve Eq. (19) for the trajectory  $x(t)$ , with initial conditions  $x = x_0$  at  $t = t_0$ . Compute  $\langle x(t) \rangle$  for  $f(t) \neq 0$ .
- 3) Express  $\langle x(t)x(t') \rangle_{\text{eq}}$  for  $f = 0$ , *i.e.* at equilibrium. Check that the results only depends on  $t - t'$ .

On general grounds, the fluctuation-dissipation theorem relates a correlation function at equilibrium to the response function associated to a small externally applied force. Here, we are interested in the response of  $x(t)$  to  $f(t)$ , and the response function (which actually is an  $xx$  response function) is defined as

$$\langle x(t) \rangle = \int_{-\infty}^t \chi(t - t') f(t') dt'. \quad (21)$$

The fluctuation-dissipation theorem states (with  $\beta^{-1} = kT$  and  $\theta(t)$  the Heaviside function) that

$$\boxed{\chi(\tau) = -\beta \theta(t) \frac{d}{d\tau} \langle x(\tau)x(0) \rangle_{\text{eq}}}.$$

(22)

- 4) Before checking explicitly the above relation between correlation and response functions, we consider a static force  $f(t) = f_0$ , to realize that Eq. (22) hides the equipartition theorem. First, what is  $\langle x \rangle$  in this case? No calculation is necessary, just a sensical remark on the behavior of springs under constant tension. Check that you recover this expectation from the results of question 2 (take  $t_0 \rightarrow -\infty$  since we look at the static response). Second, making use of Eqs. (21) and (22), together with an integration by parts, show that  $\langle x^2 \rangle_{\text{eq}} = kT/\kappa$ .
- 5) Compute  $\chi(\tau)$  for an equilibrium system, that thus has had enough time to relax. Check that the fluctuation-dissipation theorem (22) is obeyed.
- 6) Discuss the Brownian limit  $\kappa \rightarrow 0$ .

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## 11 Two generalized Feynman-Kac relations

We consider the overdamped Langevin process

$$\frac{dx(t)}{dt} = \mu F(x, t) + \sqrt{2D} \xi(t) \quad (23)$$

where  $\xi(t)$  is a Gaussian white noise with  $\langle \xi(t) \rangle = 0$  and  $\langle \xi(t) \xi(t') \rangle = \delta(t - t')$ .

1) Two arbitrary functions  $V(x, t)$  and  $f(x, t)$  being chosen, we define the functional of the trajectory:

$$Q(x_0, t, t_0) = \left\langle f(x(t), t) e^{-\int_{t_0}^t V[x(\tau), \tau] d\tau} \right\rangle \quad (24)$$

where  $t_0 \leq t$  and the average over trajectories, represented by the bracket above, is performed for a fixed initial condition  $x(t_0) = x_0$ . Show that

$$\frac{\partial Q}{\partial t_0} + D \frac{\partial^2 Q}{\partial x_0^2} + \mu F(x_0, t_0) \frac{\partial Q}{\partial x_0} - V(x_0, t_0) Q = 0 \quad (25)$$

with the condition  $Q(x_0, t_0, t_0) = f(x_0, t_0)$  (sometimes called the terminal condition, since it corresponds to  $t_0 = t$ ). To this end, one may introduce the auxiliary propagator

$$p(x, \Omega, t | x_0, \Omega_0, t_0) \quad \text{where} \quad \Omega(t, t_0) = \Omega_0 + \int_{t_0}^t V[x(\tau), \tau] d\tau, \quad (26)$$

obeying a backwards Fokker-Planck equation.

In the case where  $V = 0$ , what evolution equation do we get?

2) We consider a variant, with the functional

$$\mathcal{Q}(x_0, t, t_0) = \int_{t_0}^t dt' \left\langle f(x(t'), t') e^{-\int_{t_0}^{t'} V[x(\tau), \tau] d\tau} \right\rangle, \quad (27)$$

where the bracket is again an average over trajectories starting from  $x(t_0) = x_0$ . Show that

$$\frac{\partial \mathcal{Q}}{\partial t_0} + D \frac{\partial^2 \mathcal{Q}}{\partial x_0^2} + \mu F(x_0, t_0) \frac{\partial \mathcal{Q}}{\partial x_0} - V(x_0, t_0) \mathcal{Q} + f(x_0, t_0) = 0. \quad (28)$$

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## 12 Martingales for the asymmetric random walk

We are interested in the asymmetric random walk on  $\mathbb{Z}$ , with probabilities  $p$  and  $q = 1 - p$  to jump right and left respectively. The walker starts at  $X_0 = 0$  and its position after  $n$  steps is denoted  $X_n$ . Show that

$$M_n \equiv \left( \frac{q}{p} \right)^{X_n} \quad \text{and} \quad M'_n \equiv X_n - n(p - p) \quad (29)$$

are martingales. Knowing this, use Doob's stopping time theorem to compute the probability that the walker leaves a predefined interval  $[-a, b]$  by the right boundary at  $b$  (we take  $a > 0$ ). Same question with the left boundary at  $-a$ . What is finally the mean exit time from the interval? Check that you recover previously known results.

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## 13 Algebraic area enclosed by a random walk

Among all possible symmetric random walks on the square lattice ( $\mathbb{Z}^2$ ), we restrict to those forming closed paths (i.e. ending at their starting point, see Fig. 1).

1) Show that there are  $\binom{n}{n/2}^2$  such walks. To this end, one can partition these walks according to  $k$ , the number of steps made along the  $x$  axis. There are conversely  $n - k$  steps made along the  $y$  axis. Note that  $n$  and  $k$  should be even. It may be useful here to use the relation

$$\sum_{j=0}^p \binom{p}{j} \binom{p}{n-j} = \sum_{j=0}^p \binom{p}{j}^2 = \binom{2p}{p} \quad (30)$$

which stems from counting in two ways the number of teams of  $p$  players one can form out of a group of  $2p$  players. Among the  $2p$  players,  $p$  have a cap, and  $p$  do not. Thus, in the team formed, there can be  $j = 0, 1, \dots$  up to  $j = p$  players with a cap.

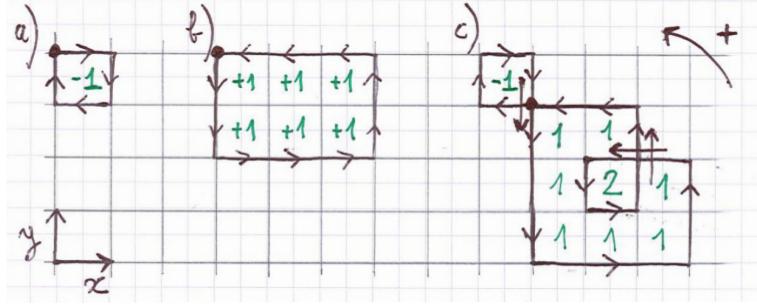


Figure 1: A few closed random walks on  $\mathbb{Z}^2$  with: a)  $n = 4$  steps and enclosed area  $\mathcal{A} = -1$ ; b)  $n = 10$  steps and  $\mathcal{A} = 6$ ; c)  $n = 20$  steps and  $\mathcal{A} = -1 + 7 \times 1 + 2 = 8$ . In panel c), the unit cell indicated with a 2 is enclosed twice by the walk and contributes a  $+2$  to  $\mathcal{A}$ ; 7 cells are enclosed once and clockwise, 1 cell is enclosed once counterclockwise.

2) For each closed walk, we define the algebraic area  $\mathcal{A}$  as the total oriented area spanned by the walk, as it traces the lattice. A unit lattice cell enclosed counterclockwise (resp. clockwise) counts  $+1$  (resp.  $-1$ ). Besides, if the cell is enclosed more than once, its area is counted with its multiplicity, see Fig 1. We denote  $C_n(\mathcal{A})$  the number of walks of  $n$  steps that have algebraic area  $\mathcal{A}$  ( $\mathcal{A} = 0, \pm 1, \pm 2 \dots$  and  $C_n(\mathcal{A}) = C_n(-\mathcal{A})$  by symmetry). The trick is to introduce two hopping operators,  $r$  (one step to the right, along  $x$ ) and  $u$  (one step up, along  $y$ ), which provides a natural way to represent a given walk:  $ru^{-1}r^{-1}u$  for the path in Fig. 1-a). The operators do not commute and it is convenient to declare  $ru = Qur$ , where  $Q \neq 1$  is some arbitrary number. With this rule, check that a closed walk, defined by a sequence of  $r$ ,  $r^{-1}$ ,  $u$  and  $u^{-1}$  yields  $Q^{\mathcal{A}}$ . From this remark, it follows that when selecting the  $u$  and  $r$  independent part in  $(r + r^{-1} + u + u^{-1})^n$ , we obtain  $C_n(\mathcal{A})$ :

$$(r + r^{-1} + u + u^{-1})^n = \sum_{\mathcal{A}=-\infty}^{\infty} C_n(\mathcal{A}) Q^{\mathcal{A}} + \dots \quad (31)$$

where  $\dots$  is for terms that explicitly depend on the operators (such as  $r^n$ ,  $r^{n-1}u$  etc.) and are associated to non-closed walks. For instance,  $(r + r^{-1} + u + u^{-1})^4 = 28 + 4Q + 4Q^{-1} + \dots$  meaning that among the  $\binom{4}{2}^2 = 36$  closed walks of 4 steps,  $C_4(0) = 28$  enclose a vanishing area,  $C_4(1) = 4$  enclose an area  $+1$  and  $C_4(-1) = 4$  enclose an area  $-1$  (you may check explicitly the latter two results).