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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 λ . 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 ± 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".

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2 Consistency of Itô-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 Itô-Doblin and Stratonovich calculus for computing

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

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3 Itô-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 Itô-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 we 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:

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

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 stationary process is Markovian iff its autocorrelation function is exponential*.

- 1) We consider first a Gaussian, stationary 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 γ is some parameter.

- f) Using previous results, compute the correlation function $\langle x(0)x(\tau) \rangle$ as a function of γ . What tis the physical significance of γ ?
 - 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, stationary, and with an exponential correlation function, is Markovian.

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8 Fokker-Planck from Itô-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 φ , 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 λ 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 μ ? 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

$$\boxed{\chi(\tau) = -\beta \theta(\tau) \frac{d}{d\tau} \langle x(\tau)x(0) \rangle_{\text{eq}}}. \quad (22)$$

- 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

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

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.

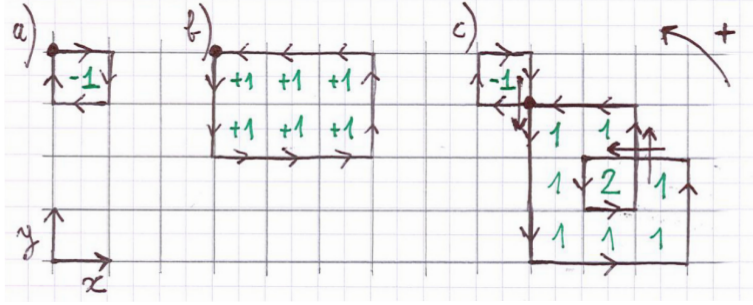


Figure 1: A few closed random walks on \mathbb{Z}^2 with: a) $n = 4$ steps and enclosed area $\mathcal{A} = -1$; b) $n = 10$ steps and $\mathcal{A} = 6$; c) $n = 20$ steps and $\mathcal{A} = -1 + 7 \times 1 + 2 = 8$. In panel c), the unit cell indicated with a 2 is enclosed twice by the walk and contributes 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 $\binom{4}{2} = 36$ 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).