

# Preparation of the exam of Stochastic processes

## List of questions on the lectures

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$\sim 1/3$  of the exam will come from the list of questions below

Questions are in black ;

answers are in the lectures or tutorials + colored additional remarks below

#### Chapter 2 : Probability

Q1: Give a brief demonstration of the **central limit theorem**.

→ recall the assumptions,

→ use of the generating function of the cumulants  $\hat{w}(k) = \ln \hat{p}(k) \underset{k \rightarrow 0}{\simeq} -i\kappa_1 k - \frac{1}{2}\kappa_2 k^2$  and recover the Gaussian distribution

Q2: Discuss the concept of **large deviation** on the simple case of the binomial distribution : show that the binomial distribution presents the large deviation form

$\mathcal{P}_N(n) = C_N^n p^n q^{N-n} \underset{N \rightarrow \infty}{\sim} \exp \left\{ -N \Phi\left(\frac{n}{N}\right) \right\}$  and derive the large deviation function  $\Phi(y)$ .

Recover the central limit theorem from it in this specific situation.

→ understand the difference between typical and atypical fluctuations

Using the Stirling formula we have

$$\ln \mathcal{P}_N(n) \simeq \underbrace{N \ln N - n \ln n - (N-n) \ln(N-n)}_{-n \ln(n/N) - (N-n) \ln[(N-n)/N]} + n \ln p + (N-n) \ln q + \mathcal{O}(\ln N)$$

where  $q = 1 - p$ . Therefore

$$\ln \mathcal{P}_N(n) \simeq N [-y \ln y - (1-y) \ln(1-y) + y \ln p + (1-y) \ln q] \quad (1)$$

i.e.

$$\Phi(y) = y \ln \left( \frac{y}{p} \right) + (1-y) \ln \left( \frac{1-y}{1-p} \right) \quad (2)$$

We see that  $\Phi'(y) = 0$  for  $y = y_* = p$ . At this point  $\Phi(y_*) = 0$  and  $\Phi''(y_*) = 1/(pq)$ .

- The quadratic behaviour of the LDF  $\Phi(y) \simeq \frac{1}{2pq}(y-p)^2$  for  $y \rightarrow y_* = p$  corresponds to the central limit theorem and the Gaussian distribution,  $\mathcal{P}_N(S) \sim \exp \left\{ \frac{1}{2Npq}(S-Np)^2 \right\}$ . The correspondence originates from the fact that  $\mathcal{P}_N(n) = \text{Proba}\{S_N = n\}$  is the distribution of the sum of  $N$  independent Bernoulli random variables ( $S_N = \sum_{i=1}^N \xi_i$  for  $\xi_i = 0$  with proba  $q$  and  $\xi_i = 1$  with proba  $p$ ).
- For  $|y - y_*| \sim \mathcal{O}(1)$  the LDF presents strong deviations to the quadratic form.

Q3: **Symmetric Lévy distribution.**— Consider the sum  $S_N$  of  $N$  i.i.d. random numbers with *symmetric* power law distribution  $p(x) \underset{x \rightarrow \infty}{\sim} |x|^{-1-\mu}$  for  $\mu \in ]0, 2[$ . Using that the characteristic function presents the limiting behaviour  $\hat{p}(k) \underset{k \rightarrow 0}{\simeq} 1 - c|k|^\mu$ , show that the distribution

of the sum  $S_N$  can be expressed in terms of the Lévy law  $\mathcal{L}_{\mu,0}(x) \stackrel{\text{def}}{=} \int_{\mathbb{R}} \frac{dk}{2\pi} e^{-|k|^\mu + ikx}$  and discuss the scaling of  $S_N$  with  $N$ .

→ cf. derivation in the lecture notes.

Q4: Consider  $N$  correlated Gaussian random variables  $x_1, \dots, x_N$  with distribution  $P(\mathbf{x}) \propto \exp\{-\frac{1}{2}\mathbf{x}^T A \mathbf{x}\}$ , where  $A$  is a positive definite real symmetric matrix.

a) Show that the correlations are  $\langle x_i x_j \rangle = (A^{-1})_{ij}$ .

b) Application : **discrete Ornstein-Uhlenbeck process**.— We consider random Gaussian variables  $\Phi = (\dots, \phi_n, \dots)^T$  with weight  $P(\Phi) \propto \exp[-S(\Phi)]$  where the action is

$$S(\Phi) = \frac{1}{2} \sum_{t \in \mathbb{Z}} [(\phi_{t+1} - \phi_t)^2 + \mu^2 \phi_t^2] \quad (3)$$

Write the action under the form  $S = \frac{1}{2}\Phi^T A \Phi$  and show that the matrix  $A$  involves the discrete Laplace operator  $\Delta_{n,m} = \delta_{n,m+1} - 2\delta_{n,m} + \delta_{n,m-1}$ .

c) Give the eigenvalues and the (normalised) eigenvectors of  $\Delta$  on the infinite line ( $n \in \mathbb{Z}$ ). Deduce the correlation function  $\langle \phi_t \phi_{t'} \rangle$ .

d) Discuss the limit  $\mu \rightarrow 0$ .

Hint : we give the integral  $\int_0^{2\pi} \frac{d\theta}{2\pi} \frac{\sinh \lambda}{\cosh \lambda + \cos \theta} e^{in\theta} = e^{-\lambda|n|}$ .

a) → use the generating function (no need to calculate an integral). Cf. lecture notes.

b) to d) → Cf. tutorial n°1

### Chapter 3 : Langevin equation

Q5: **Wiener process**.— Consider the Wiener process  $W(t) = \int_0^t du \eta(u)$  where  $\eta(t)$  is a normalised Gaussian white noise such that  $\langle \eta(t) \rangle = 0$  and  $\langle \eta(t) \eta(t') \rangle = \delta(t - t')$ .

a) Compute the correlator  $\langle W(t) W(t') \rangle$  and deduce  $\langle [W(t) - W(t')]^2 \rangle$ .

b) Give the distribution of the process  $P_t(W)$ .

a) easy to get  $\langle W(t) W(t') \rangle = \min(t, t')$  and  $\langle [W(t) - W(t')]^2 \rangle = |t - t'|$ .

b) The process is a sum of Gaussian variables, hence its distribution is also Gaussian. The knowledge of the two first cumulants is enough,  $\langle W(t) \rangle = 0$  and  $\langle W(t)^2 \rangle = t$ , thus  $P_t(W) = \frac{1}{\sqrt{2\pi t}} \exp[-W^2/(2t)]$ .

Q6: **Langevin equation (Ornstein-Uhlenbeck process)**.— consider a particle in a fluid with velocity  $v(t)$  obeying the Langevin equation  $\frac{d}{dt}v(t) = -\frac{1}{\tau}v(t) + \frac{1}{\tau}\sqrt{2D}\eta(t)$  where  $\eta(t)$  is a normalised Gaussian white noise such that  $\langle \eta(t) \rangle = 0$  and  $\langle \eta(t) \eta(t') \rangle = \delta(t - t')$ . The time  $\tau$  is related to the friction coefficient  $\gamma = m/\tau$ ,  $m$  being the mass.

a) Give the solution of the differential equation. Deduce the mean  $\langle v(t) \rangle$  and the correlator  $\langle v(t) v(t') \rangle_c$  for fixed  $v(0) = v_0$ .

b) Deduce the conditional probability  $P_t(v|v_0)$ .

c) Discuss the large time behaviour of  $P_t(v|v_0)$  and recover the Einstein relation between the diffusion constant  $D$ , the friction coefficient  $\gamma = m/\tau$  and  $k_B T$ .

d) Define the overdamped regime and write the SDE for the position  $x(t)$ .

→ cf. lecture notes.

### Chapter 4 : Markov processes

Q7: Give the definition of a Markov process in few words. Give an example of Markov process and an example of non Markovian process.

Q8: Consider a Markov process  $X(t)$  characterized by the probability  $P_t(x)$  and the conditional probability  $P_t(x|y)$ .

- Express  $\langle x(t) \rangle$  in terms of these distributions (i) assuming a fixed  $x(0) = x_0$ , (ii) assuming a random  $x(0)$ .
- Express  $\langle x(t)x(t') \rangle$  in terms of the conditional probability for a fixed  $x(0) = x_0$ .
- Application to the Poisson process  $\mathcal{N}(t) \in \mathbb{N}$  : the conditional probability is

$$P_t(n|m) = \begin{cases} \frac{(\lambda t)^{n-m}}{(n-m)!} e^{-\lambda t} & \text{for } n \geq m \\ 0 & \text{for } n < m \end{cases} \quad (4)$$

Express and compute  $\langle \mathcal{N}(t) \rangle$  for  $\mathcal{N}(0) = 0$ .

d) Express  $\langle \mathcal{N}(t)\mathcal{N}(t') \rangle$  as a double sum and compute it for  $\mathcal{N}(0) = 0$  when  $t' < t$ . Deduce the correlator  $\langle \mathcal{N}(t)\mathcal{N}(t') \rangle_c = \langle \mathcal{N}(t)\mathcal{N}(t') \rangle - \langle \mathcal{N}(t) \rangle \langle \mathcal{N}(t') \rangle$ .

Hint : the calculation involve sums of the form  $\sum_{n=0}^{\infty} n \frac{(\lambda t)^n}{n!}$  or  $\sum_{n=0}^{\infty} n(n-1) \frac{(\lambda t)^n}{n!}$  which are very easy to compute!

a) - c) Cf. lecture notes

d)  $\langle \mathcal{N}(t)\mathcal{N}(t') \rangle = \sum_{n=m}^{\infty} \sum_{m=0}^{\infty} n P_{t-t'}(n|m) m P_{t'}(m|0)$ . The two sums are easy to compute. Eventually, one finds  $\langle \mathcal{N}(t)\mathcal{N}(t') \rangle_c = \lambda \min(t, t')$ .

Q9: **Poisson process** : the Poisson process  $\mathcal{N}(t) \in \mathbb{N}$  counts the occurrences of *independent events* occurring with a constant rate  $\lambda$  on the interval  $[0, t]$ . Denote the probability  $P_n(t) = \text{Proba}\{\mathcal{N}(t) = n\}$ .

- Show that  $P_0(t)$  obeys the differential equation  $\partial_t P_0(t) = -\lambda P_0(t)$ . Derive a set of coupled differential equations for the probabilities  $P_n(t)$ .
- Introduce the generating function  $G(z; t) \stackrel{\text{def}}{=} \sum_{n=0}^{\infty} z^n P_n(t)$ . What is the value of  $G(z; 0)$ ? Get a differential equation for  $G(z; t)$  and solve it.
- Deduce the expression for  $P_n(t)$ .
- Determine the cumulants  $\langle \mathcal{N}(t)^k \rangle_c$  of the Poisson process.
- Argue that the distribution  $q(\tau)$  of the time separating two successive events is related to  $P_0(t)$  and give  $q(\tau)$ .

a) We have

$$P_n(t + \delta t) = P_n(t) \underbrace{(1 - \lambda \delta t - \dots)}_{\text{no event on } [t, t+\delta t]} + P_{n-1}(t) \underbrace{\lambda \delta t}_{\text{one event}} + \dots \quad (5)$$

Eventually, in the limit  $\delta t \rightarrow 0$  we get

$$\partial_t P_n(t) = -\lambda P_n(t) + \lambda P_{n-1}(t) \quad \text{for } n > 0 \quad (6)$$

and

$$\partial_t P_0(t) = -\lambda P_0(t) \quad \text{for } n = 0. \quad (7)$$

b) We multiply the differential equation by  $z^n$  and sum over  $n$  :  $\partial_t \sum_n z^n P_n(t) = -\lambda \sum_n z^n P_n(t) + \lambda \sum_n z^n P_{n-1}(t)$ , leading to

$$\partial_t G(z; t) = \lambda(z - 1) G(z; t) \quad (8)$$

We have  $G(z; 0) = 1$  (normalization). The differential equation for  $G(z; t)$  is straightforward to solve, we get

$$G(z; t) = \exp[\lambda t(z - 1)]. \quad (9)$$

c) Expansion of  $G(z; t)$  in powers of  $z$  gives  $P_n(t) = \frac{(\lambda t)^n}{n!} e^{-\lambda t}$

d) Direct calculation of the moments/cumulants is unpleasant... However the generating function of the moments is easy to obtain

$$\left\langle e^{k\mathcal{N}(t)} \right\rangle \left( = \sum_{n=0}^{\infty} \frac{k^n}{n!} \langle \mathcal{N}(t)^n \rangle \right) = \sum_{n=0}^{\infty} e^{kn} P_n(t) = G(e^k; t) = \exp[\lambda t(e^k - 1)] \quad (10)$$

The generating function of the cumulants is therefore

$$\sum_{n=1}^{\infty} \frac{k^n}{n!} \langle \mathcal{N}(t)^n \rangle_c = \ln G(e^k; t) = \lambda t (e^k - 1) \quad (11)$$

We deduce that **all cumulants are equal**

$$\langle \mathcal{N}(t)^n \rangle_c = \lambda t. \quad (12)$$

e) Denote  $\tau$  the time of occurrence of the first event (i.e. this is the time separating two consecutive events). We have

$$P_0(t) = \text{Proba}\{\text{no event on } [0, t]\} = \text{Proba}\{\tau > t\} = \int_t^{\infty} d\tau q(\tau) \quad (13)$$

where  $q(\tau)$  is the distribution of the time. As a result, the time intervals are exponentially distributed

$$q(\tau) = \lambda e^{-\lambda\tau} \quad (14)$$

**Q10: Compound Poisson process :** We consider the master equation

$$\frac{\partial P_t(x)}{\partial t} = \int dy [W(x|y)P_t(y) - W(y|x)P_t(x)] \quad (15)$$

for a translation invariant kernel  $W(x|y) = \lambda w(x - y)$  ; here  $\lambda$  is the rate of jumps and  $w(\eta)$  the distribution of the jump amplitudes.

a) show that the probability is conserved.

b) Introduce the Fourier transforms  $\hat{P}_t(k) = \int dx e^{-ikx} P_t(x)$  and  $\hat{w}(k) = \int d\eta e^{-ik\eta} w(\eta)$ . Deduce a differential equation for  $\hat{P}_t(k)$  and express the solution  $P_t(x)$  under the form of an integral, for initial condition  $P_0(x) = \delta(x)$ .

c) Assuming that we can write  $\hat{w}(k) \simeq 1 - \frac{1}{2}ck^2$  for  $k \rightarrow 0$ , deduce the distribution  $P_t(x)$ . What is the meaning of the parameter  $c$  ? How would you qualify the process in this limit ?

d) Same question for  $\hat{w}(k) \simeq 1 - c|k|$  for  $k \rightarrow 0$ .

a) obvious

b) The master equation involves a convolution, hence, after Fourier transform we have  $\partial_t \hat{P}_t(k) = \lambda [\hat{w}(k) - 1] \hat{P}_t(k)$ . Using  $\hat{P}_0(k) = 1$ , integration is elementary. Finally we get

$$P_t(x) = \int_{-\infty}^{+\infty} \frac{dk}{2\pi} e^{\lambda t [\hat{w}(k) - 1] + ikx}. \quad (16)$$

c) for high rate (or large time), the integral selects the  $k \rightarrow 0$  behaviour  $\hat{w}(k) \simeq 1 - \frac{1}{2}ck^2$ , where  $c$  is the variance of the jumps. Hence the integral is Gaussian

$$P_t(x) \simeq \frac{1}{\sqrt{2\pi\lambda ct}} \exp - \frac{x^2}{2\lambda ct} \quad (17)$$

i.e. we have the scaling  $x \sim \sqrt{t}$ . In the large time limit the random walk corresponds to the continuous Brownian motion, as a result of the central limit theorem.

d) If the jump distribution presents a symmetric power law tail  $w(\eta) \sim c/\eta^2$  we have  $\hat{w}(k) \simeq 1 - c|k|$  for  $k \rightarrow 0$  (for example  $w(\eta) = (a/\pi)(\eta^2 + a^2)^{-1}$  gives  $\hat{w}(k) = e^{-a|k|}$ ). The large time limit involves the  $k \rightarrow 0$  behaviour, thus

$$\hat{P}_t(x) \simeq \int_{-\infty}^{+\infty} \frac{dk}{2\pi} e^{-c\lambda t |k| + ikx} = \frac{c\lambda t}{\pi (x^2 + (c\lambda t)^2)} \quad (18)$$

corresponding to the scaling  $x \sim t$  (anomalous diffusion, motion is superdiffusive). This is a Lévy flight.

**Q11: Diffusion in a ring.**— Consider a particle on the  $N$  sites of a ring. We denote  $P_n(t)$  the probability to be on site  $n \in \{1, \dots, N\}$  at time  $t \in \mathbb{N}$ . At each time step ( $\Delta t = 1$ ) the particle jumps from site  $m$  to site  $n$  with probability  $M_{nm} \in [0, 1]$ .

- What is the name of such a stochastic process ? What is the condition on the  $N \times N$  matrix  $M$  ? How do we call such a matrix ?
- Write the master equation. Express  $P_n(t+1) - P_n(t)$  in terms of  $M_{nm}P_m(t) - M_{mn}P_n(t)$ .
- What is the equation for the stationary state  $P_n^*$  ? Does it always exist for a finite number  $N$  of sites ? Support your statement by a proof.
- Recall the detailed balance condition. What is the nature of the stationary state in this case ? How do we call the stationary state if the detailed balance condition does not hold ? Give two (simple) examples corresponding to the two situations in the ring (be brief).
- Give a (simple) example of a situation where no stationary state exists. How do we call such a process ?

a), b), c)  $\rightarrow$  cf. lecture notes.

d) Detailed balance is the strong condition  $M_{nm}P_m^* - M_{mn}P_n^* = 0$ , corresponding to the *equilibrium state*.

If  $M_{nm}P_m^* - M_{mn}P_n^* \neq 0$  but  $\sum_m (M_{nm}P_m^* - M_{mn}P_n^*) = 0$  the stationary state is a NESS.

Example on the ring. Consider simple translation invariant transition probabilities between nearest-neighbour sites only,  $M_{n,n-1} = p$  and  $M_{n-1,n} = q = 1 - p$ . On the ring, the stationary solution is obviously the uniform distribution  $P_n^* = 1/N$ . We have  $M_{n,n-1}P_{n-1}^* - M_{n-1,n}P_n^* = (p - q)/N$ . Thus :

- for  $p = q$ , detailed balance holds and the stationary state is an *equilibrium*.
- for  $p \neq q$ , detailed balance does not hold and the stationary state is a NESS. In this case there exists a non vanishing steady current around the ring,  $J = (p - q)/N$ .

e) Only in the limit  $N \rightarrow \infty$  can  $P_n^*$  be non normalisable. Then no stationary state exists and the process is "transient". Example : the free diffusion for  $p = q = 1/2$ , then  $P_n(t) \simeq \frac{1}{\sqrt{\pi Dt}} \exp[-n^2/(4Dt)]$  for large  $n$  and large  $t$ .

## Chapter 5 : SDE

**Q12: Itô calculus.**— We denote by  $W(t)$  the Wiener process (a normalised BM).

- What is  $dW(t)^2$  ?
- Consider the SDE

$$dx(t) = a(x(t))dt + b(x(t))dW(t) \quad (\text{Itô}). \quad (19)$$

What is the main assumption in Itô calculus ?

- Recover the Itô formula for  $d\varphi(x(t))$  where  $x(t)$  solves the Itô equation and  $\varphi(x)$  a regular function.
- Itô SDE and FPE :** Deduce the FPE from the Itô SDE (19).

$\rightarrow$  details in lecture notes.

**Q13:** We recall that the Itô SDE

$$dx(t) = a(x)dt + b(x)dW(t) \quad (\text{Itô}) \quad (20)$$

is related to the PFE  $\partial_t P_t(x) = -\partial_x[a(x)P_t(x)] + \frac{1}{2}\partial_x^2[b(x)^2P_t(x)]$ , and the Stratonovich SDE

$$dx(t) = \phi(x)dt + b(x)dW(t) \quad (\text{Stratonovich}) \quad (21)$$

to the PFE  $\partial_t P_t(x) = -\partial_x [\phi(x) P_t(x)] + \frac{1}{2} \partial_x [b(x) \partial_x [b(x) P_t(x)]]$ .

a) What is the relation between  $a(x)$  and  $\phi(x)$  ensuring that the two SDE describe the same stochastic process ?

b) We consider the Stratonovich SDE

$$dx(t) = \phi(x) dt + \sqrt{2D(x)} dW(t) \quad (\text{Stratonovich}) \quad (22)$$

give the corresponding Itô SDE.

c) What is the drift  $\phi(x)$  leading to  $\frac{d}{dt} \langle x(t) \rangle = 0$  ?

a) the two FPE coincide for  $\phi(x) = a(x) - \frac{1}{2} b(x) b'(x)$ .

b) the corresponding Itô SDE is  $dx(t) = F(x) dt + \sqrt{2D(x)} dW(t)$  with  $F(x) = \phi(x) + \frac{1}{4} [b(x)^2]'$  with  $b(x) = \sqrt{2D(x)}$ , i.e.  $F(x) = \phi(x) + \frac{1}{2} D'(x)$ ,

$$dx(t) = \left[ \phi(x) + \frac{1}{2} D'(x) \right] dt + \sqrt{2D(x)} dW(t) \quad (\text{Itô}) \quad (23)$$

c) We compute easily  $\frac{d}{dt} \langle x(t) \rangle$  from the Itô SDE ( $x(t)$  and  $dW(t)$  independent) :  $\frac{d}{dt} \langle x(t) \rangle = \langle F(x(t)) \rangle$ . The mean velocity vanishes for the drift  $\phi(x) = -\frac{1}{2} D'(x)$  in the Stratonovich SDE.

## Chapter 6 : FPE

Q14: Consider the FPE

$$\frac{\partial P_t(x)}{\partial t} = -\frac{\partial}{\partial x} [F(x) P_t(x)] + \frac{\partial^2}{\partial x^2} [D(x) P_t(x)] \quad (24)$$

a) discuss the meaning of the two terms. Explain their effects (for uniform  $F(x) = F$  and  $D(x) = D$ ).

b) Give the expression of the probability current  $J_t(x)$ .

→ cf. lecture notes.

Q15: We consider the FPE  $\partial_t P_t(x) = -\partial_x [F(x) P_t(x)] + D \partial_x^2 [P_t(x)]$ .

a) Show that the "forward generator"  $\mathcal{G}^\dagger = -\partial_x F(x) + D \partial_x^2$  can be expressed in terms of  $\partial_x$  and  $e^{-V(x)/D}$  where  $F(x) = -V'(x)$ .

b) Deduce the expression of the equilibrium state. Under what condition on  $V(x)$  does the equilibrium state exists ?

c) What is the appropriate (nonunitary) transformation allowing to map the non self adjoint operator  $\mathcal{G}^\dagger$  onto a self adjoint one of the form  $H_+ = \mathcal{Q}^\dagger \mathcal{Q}$  ? Give  $\mathcal{Q}$  and  $\mathcal{Q}^\dagger$ .

→  $\mathcal{G}^\dagger = D \partial_x e^{-V(x)/D} \partial_x e^{+V(x)/D}$ , etc, cf. lecture notes.

Q16: **The Ornstein-Uhlenbeck process.**— Consider the SDE  $dx(t) = F(x)dt + \sqrt{2D} dW(t)$ .

We recall that the FPE  $\partial_t P_t(x) = \mathcal{G}^\dagger P_t(x)$  can be mapped onto the imaginary time Schrödinger equation  $-\partial_t \psi(x; t) = H_+ \psi(x; t)$  through  $P_t(x) = \psi(x; t) \phi_0(x)$  where  $\phi_0(x) = \exp[\frac{1}{2D} \int_0^x dy F(y)]$ . The Hamiltonian has the form  $H_+ = \mathcal{Q}^\dagger \mathcal{Q}$  where  $\mathcal{Q} = \sqrt{D} (-\partial_x + \frac{1}{2D} F(x))$  and  $\mathcal{Q}^\dagger = \sqrt{D} (\partial_x + \frac{1}{2D} F(x))$ .

We consider the Ornstein-Uhlenbeck process such that  $F(x) = -kx$ .

a) Compute the commutator  $[\mathcal{Q}, \mathcal{Q}^\dagger]$ .

b) Give the expression of  $\phi_0(x)$ . Compute  $\mathcal{Q} \phi_0(x)$  and deduce that  $\phi_0(x)$  is an eigenstate of  $H_+$ . Give the corresponding eigenvalue  $\lambda_0$ . What is the solution of the FPE related to this state ?

- c) Show that  $\mathcal{Q}^\dagger \phi_0(x)$  is also eigenstate of  $H_+$  and give the related eigenvalue  $\lambda_1$ .  
d) Using the same idea, deduce the full spectrum of eigenvalues of  $H_+$ .  
e) We recall that the conditional probability is expressed in terms of the spectrum of  $H_+$  as

$$P_t(x|x_0) = \frac{\phi_0(x)}{\phi_0(x_0)} \sum_{n=0}^{\infty} \phi_n(x) \phi_n(x_0) e^{-\lambda_n t} \quad (25)$$

Deduce the averaged return probability  $\int dx P_t(x|x)$  for the Ornstein-Uhlenbeck process (remember that  $\phi_n(x)$  are normalised).

- a) cf. correction of the test  
b) Action of  $\mathcal{Q} = \sqrt{2D}(-\partial_x + \frac{1}{2D}F(x))$  on  $\phi_0(x) = \exp[\frac{1}{2D} \int_0^x dy F(y)]$  is zero. Hence  $\phi_0(x) = e^{-kx^2/(4D)}$  has eigenvalue  $\lambda_0 = 0$ . This is the equilibrium state  $P_{eq}(x) = c \phi_0(x)^2 = c e^{-kx^2/(2D)}$   
c) use commutator :  $\lambda_1 = \lambda_0 + k = k$ .  
d) By recurrence,  $(\mathcal{Q}^\dagger)^n \phi_0(x)$  has an eigenvalue  $\lambda_n = nk$  for  $n \in \mathbb{N}$ .  
e)  $\int dx P_t(x|x) = \sum_{n=0}^{\infty} e^{-nkt} = (1 - e^{-kt})^{-1}$ .

**Q17: Construction of the conditional probability.**— Consider the SDE  $dx(t) = F(x)dt + \sqrt{2D}dW(t)$ . We recall that the FPE  $\partial_t P_t(x|x_0) = \mathcal{G}^\dagger P_t(x|x_0)$  can be mapped onto the imaginary time Schrödinger equation  $-\partial_t \psi(x;t) = H_+ \psi(x;t)$  through  $P_t(x|x_0) = \psi(x;t)\phi_0(x)$  where  $\phi_0(x) = \exp[\frac{1}{2D} \int_0^x dy F(y)]$ . For a confining drift, the Hamiltonian  $H_+$  has a spectrum of eigenvalues and eigenvectors  $(\lambda_n, \phi_n(x))$  with  $\lambda_0 = 0$ .

- a) An initial state can be decomposed over the basis of orthonormal eigenstates  $\phi_n(x)$  of  $H_+$  as  $\psi(x;0) = \sum_n c_n \phi_n(x)$ . If the initial condition for the FPE is  $P_0(x|x_0) = \delta(x - x_0)$ , give the coefficients  $c_n$ .  
b) What is the wave function  $\psi(x;t)$  at time  $t$ ?  
c) Deduce the corresponding solution of the FPE.  
d) Check that  $\int dx P_t(x|x_0) = 1$ . Analyze the  $t \rightarrow \infty$  behaviour of  $P_t(x|x_0)$  and discuss it.

a) Using orthonormalisation condition  $\int dx \phi_n(x)\phi_m(x) = \delta_{n,m}$ , we obtain  $c_n = \int dx \psi(x;0) \phi_n(x)$ . For  $P_0(x|x_0) = \delta(x - x_0)$ , we have  $\psi(x;0) = P_0(x|x_0)/\phi_0(x) = \delta(x - x_0)/\phi_0(x)$  thus  $c_n = \phi_n(x_0)/\phi_0(x_0)$ .

b) Therefore  $\psi(x;t) = \sum_n c_n \phi_n(x) e^{-\lambda_n t}$

c) We get

$$P_t(x) = \psi(x;t)\phi_0(x) = \frac{\phi_0(x)}{\phi_0(x_0)} \sum_n \phi_n(x) \phi_n(x_0) e^{-\lambda_n t} \equiv P_t(x|x_0) \quad (26)$$

d)  $P_t(x|x_0) \simeq \phi_0(x)^2 e^{-\lambda_0 t} + \frac{\phi_0(x)}{\phi_0(x_0)} \phi_1(x)\phi_1(x_0) e^{-\lambda_1 t} + \dots$ . The first term is the equilibrium distribution  $P_{eq}(x) = \phi_0(x)^2$ . The correction decays as  $\sim e^{-\lambda_1 t}$  hence  $\lambda_1$  is a relaxation rate toward equilibrium. Only the correction has the memory of the initial position  $x_0$  :

$$P_t(x|x_0) \simeq P_{eq}(x) + \frac{\phi_0(x)}{\phi_0(x_0)} \phi_1(x)\phi_1(x_0) e^{-\lambda_1 t} + \dots \quad (27)$$

**Q18: Diffusion in a ring.**— We consider the FPE  $\partial_t P_t(x|x_0) = \mathcal{G}^\dagger P_t(x|x_0)$  in a ring of perimeter  $L$  (i.e. in the interval  $[0, L]$  with periodic boundary conditions) for a uniform drift  $F(x) = F_0$ . The "forward generator" is  $\mathcal{G}^\dagger = D\partial_x^2 - F_0\partial_x$ .

- a) Argue that the eigenfunctions of  $\mathcal{G}^\dagger$  are plane waves  $\Phi^R(x) \propto e^{ikx}$ . Argue that the boundary conditions lead to quantify  $k \rightarrow k_n$  and give  $k_n$ .  
b) Give the corresponding eigenvalue such that  $\mathcal{G}^\dagger \Phi_n^R(x) = -\lambda_n \Phi_n^R(x)$ . Give the generator  $\mathcal{G}$  and deduce the left eigenfunction  $\Phi_n^L(x)$ . Normalisation is chosen such that  $\Phi_0^R(x)$  is the stationary distribution and  $\Phi_0^L(x) = 1$ . Check  $\int_0^L dx \Phi_n^L(x)\Phi_m^R(x) = \delta_{n,m}$ .



c) Decompose  $P_t(x|x_0)$  over the spectrum. Express it as a *real* series. Analyze the  $t \rightarrow \infty$  behaviour (identify a characteristic time  $\tau_D$ ). Plot  $P_t(x|x_0)$  for  $t \gg \tau_D$  (the dominant term plus the first  $x$ -dependent correction).

d) Replace the sum by an integral in the  $L \rightarrow \infty$  limit and deduce  $P_t(x|x_0)$ .

a) The eigenvectors of  $\partial_x$  are plane waves  $e^{ikx}$ . These are also the eigenvectors of the diffusion operator  $\mathcal{G}^\dagger = D\partial_x^2 - F_0\partial_x$ . For periodic boundary conditions,  $\Phi^R(x) = \Phi^R(x+L)$ , the wave vector is quantized as  $k_n = 2\pi n/L$  with  $n \in \mathbb{Z}$ .

b)  $\mathcal{G}^\dagger \Phi_n^R(x) = -\lambda_n \Phi_n^R(x)$  for

$$\lambda_n = Dk_n^2 + ik_n F_0 \quad \text{for } n \in \mathbb{Z}. \quad (28)$$

Note that  $k_n \rightarrow -k_n$  corresponds to  $\lambda_n \rightarrow \lambda_n^* = \lambda_{-n}$  and  $\Phi_n^R(x) \rightarrow \Phi_n^R(x)^*$ .

The generator is  $\mathcal{G} = D\partial_x^2 - F_0\partial_x$ , hence, here it is related to  $\mathcal{G}^\dagger$  by  $F_0 \rightarrow -F_0$ . Thus  $\Phi_n^L(x) \propto \Phi_n^R(x)|_{k_n \rightarrow -k_n}$ .

The choice of normalisation is

$$\Phi_n^R(x) = \frac{1}{L} e^{ik_n x} \quad \text{and} \quad \Phi_n^L(x) = e^{-ik_n x}. \quad (29)$$

We have indeed  $\int_0^L dx \Phi_n^L(x) \Phi_m^R(x) = \int_0^L \frac{dx}{L} e^{-ik_n x + ik_m x} = \delta_{n,m}$ .

c) The conditional probability is

$$\begin{aligned} P_t(x|x_0) &= \sum_{n \in \mathbb{Z}} \Phi_n^R(x) \Phi_n^L(x_0) e^{-\lambda_n t} = \frac{1}{L} \sum_{n \in \mathbb{Z}} e^{2i\pi n(x-x_0-F_0 t)/L - Dk_n^2 t} \\ &= \frac{1}{L} + \frac{2}{L} \sum_{n=1}^{\infty} \cos(2\pi n(x-x_0-F_0 t)/L) e^{-n^2 t/\tau_D} \quad \text{where } \tau_D = \frac{L^2}{(2\pi)^2 D} \end{aligned} \quad (30)$$

The lowest e.v.  $\lambda_0 = 0$  corresponds to the *stationary state*  $P_{st}(x) = \Phi_0^R(x) = 1/L$ . We identify a relaxation rate  $\text{Re}(\lambda_1) = 1/\tau_D$ . The time  $\tau_D \sim L^2/D$ , known as the “*Thouless time*”, is the typical time needed to explore the size of the system thanks to the diffusion.

At large time we have

$$P_t(x|x_0) \simeq \frac{1}{L} + \frac{2}{L} \cos(2\pi(x-x_0-F_0 t)/L) e^{-t/\tau_D} \quad (31)$$

which presents a (small) bump for  $x \sim x_0 + F_0 t$ , the position of the drifted particle after time  $t$ .

d) in the  $L \rightarrow \infty$  limit, we can simply replace the sum by an integral

$$P_t(x|x_0) \underset{L \rightarrow \infty}{\simeq} \int_{\mathbb{R}} \frac{dk}{2\pi} e^{-(Dk^2 + ikF_0)t + ik(x-x_0)} = \frac{1}{\sqrt{4\pi Dt}} \exp \left\{ -\frac{(x-x_0-F_0 t)^2}{4Dt} \right\}, \quad (32)$$

which describes the free diffusion on  $\mathbb{R}$  with a drift.

**Q19: Boundary conditions.**— Consider the FPE  $\partial_t P_t(x) = -\partial_x [F(x)P_t(x)] + D\partial_x^2 [P_t(x)]$  on  $\mathbb{R}_+$  for a mixed boundary condition at  $x = 0$

$$\tilde{\lambda} P_t(0) = \partial_x P_t(x)|_{x=0} \quad \forall t. \quad (33)$$

a) Give the expression of the probability current  $J_t(x)$  related to  $P_t(x)$

b) Show that  $\partial_t \int_0^\infty dx P_t(x)$  is expressed in term of the current. Show that for the mixed boundary condition, one finds

$$\partial_t \int_0^\infty dx P_t(x) = -\lambda P_t(0) \quad (34)$$

and express  $\lambda$ . What is its meaning ?

c) Deduce the condition for a reflecting boundary, such that the probability is conserved. How should one choose the constant  $\tilde{\lambda}$  in this case ?

d) What is an absorbing boundary ? What are  $\lambda$  and  $\tilde{\lambda}$  in this case ?

→ cf. lecture notes.



**Q20: Persistence of the free BM.**— We consider a free BM starting at  $x(0) = x_0$ .

- a) solve the FPE  $\partial_t P_t(x|x_0) = D\partial_x^2 P_t(x|x_0)$  for the conditional probability for a Dirichlet boundary condition at  $x = 0$ . What is the meaning of this boundary condition ?
- b) Show that  $S_{x_0}(t) = \int_0^\infty dx P_t(x|x_0)$  can be expressed in terms of the error function  $\text{erf}(z) \stackrel{\text{def}}{=} \frac{2}{\sqrt{\pi}} \int_0^z dt e^{-t^2}$ . What does  $S_{x_0}(t)$  represents ? We recall that  $\text{erfc}(z) = 1 - \text{erf}(z) \simeq \frac{1}{z\sqrt{\pi}} e^{-z^2}$  for  $z \rightarrow +\infty$ . Plot neatly  $S_{x_0}(t)$  as a function of  $t$ . Explain.
- c) We denote by  $T_{x_0}$  the first passage time at  $x = 0$ . Give the relation between  $S_{x_0}(t)$  and the distribution of the first passage time  $\mathcal{P}_{x_0}(T)$ . Give the expression of the distribution. Discuss the  $T \rightarrow \infty$  behaviour.

→ cf. lecture notes.

**Q21: First passage time.**— We consider the SDE  $dx(t) = F(x)dt + \sqrt{2D}dW(t)$ . The corresponding FPE is  $\partial_t P_t(x) = \mathcal{G}^\dagger P_t(x)$  where the "forward generator" is  $\mathcal{G}_x^\dagger = -\partial_x F(x) + D\partial_x^2$

- a) give the "generator"  $\mathcal{G}_x$ .
- b) We consider the FPE for the conditional probability  $P_t(x|x_0)$  with some reflection boundary condition at  $x = a$ ,  $\partial_{x_0} P_t(x|x_0)|_{x_0=a} = 0$  and some absorbing boundary condition at  $x = b > a$ ,  $P_t(x|x_0)|_{x_0=b} = 0$ . Show that the survival probability  $S_{x_0}(t) = \int_a^b dx P_t(x|x_0)$  obeys an equation similar to the FPE. What is the initial condition  $S_{x_0}(0)$  ?
- c) Give the relation between the survival probability and the distribution of the first passage time  $\mathcal{P}_{x_0}(T)$ .
- d) We recall that the moments  $T_n(x_0) = \int_0^\infty dT T^n \mathcal{P}_{x_0}(T)$  of the first passage time obey the recurrence

$$\mathcal{G}_{x_0} T_n(x_0) = -n T_{n-1}(x_0) \quad (35)$$

(with  $T_0(x) = 1$ ). What are the boundary conditions at  $x = a$  and  $x = b$  for  $T_n(x)$  ?

Show that  $T_1'(x)$  obeys a first order differential equation and solve it (introduce  $V(x) = -\int_0^x dy F(y)$ ).

Impose the boundary condition for  $T_1'(x)$  at  $x = a$ .

Deduce a formula for  $T_1(x_0)$ .

- e) Consider the situation where  $F(x) = -\mu$ , when the reflection is at  $x = 0$ . Compute  $T_1(x_0)$ . Discuss the result : consider limiting cases (i)  $\mu b/D \ll 1$ , (ii)  $\mu b/D \gg 1$  for  $\mu > 0$ , (iii)  $|\mu|b/D \gg 1$  for  $\mu < 0$ .

a) to d) → cf. lectures.

$$T_1(x_0) = \frac{1}{D} \int_{x_0}^b dx e^{V(x)/D} \int_a^x dx' e^{-V(x')/D} . \quad (36)$$

e) integration is easy if  $V(x) = \mu x$  :

$$T_1(x_0) = \frac{D}{\mu^2} \left[ e^{\mu b/D} - \frac{\mu b}{D} - e^{\mu x_0/D} + \frac{\mu x_0}{D} \right] \quad (37)$$

It vanishes at the absorbing boundary as it should.

(i)  $\mu b/D \ll 1$  : this is equivalent to send  $\mu \rightarrow 0$ . We find  $T_1(x_0) \simeq \frac{b^2 - x_0^2}{2D}$ . For  $x_0 \sim 0$  we get the typical time  $b^2/D$  to diffuse over a region of size  $b$ .

(ii)  $\mu b/D \gg 1$  for  $\mu > 0$  : we recover the Arrhenius behaviour due to the potential barrier  $T_1(x_0) \simeq \frac{D}{\mu^2} e^{\mu b/D} \sim \exp \left\{ \frac{1}{D} [V(b) - V(0)] \right\}$

(iii)  $|\mu|b/D \gg 1$  for  $\mu < 0$  : the time is dominated by the drift  $T_1(x_0) \simeq (b - x_0)/\mu$ .

## Chapter 7 : Functionals

Q22: We consider the free BM (the Wiener process)  $x(\tau)$ , defined for  $0 \leq \tau \leq t$ , issuing from  $x(0) = 0$ . We denote by  $\mathcal{A}[x(\tau)] = \int_0^t d\tau x(\tau)$  the area between the Brownian curve and the real axis. We study its distribution  $\mathcal{P}_t(A)$ .

- Write its characteristic function  $\tilde{\mathcal{P}}_t(p) = \langle e^{-p\mathcal{A}[x(\tau)]} | x(0) = 0 \rangle$  in terms of a path integral.
- Here, the path integral can be calculated easily, using that the integral is Gaussian : we have  $\int_{x(0)=0}^{x(t)=x} \mathcal{D}x(\tau) e^{-S[x(\tau)]} = \frac{1}{\sqrt{2\pi t}} e^{-S[x_{cl}(\tau)]}$ , where  $x_{cl}(\tau)$  is the solution of  $\frac{\delta S}{\delta x(\tau)} = 0$  for  $x_{cl}(0) = 0$  and  $x_{cl}(t) = x$ . Find  $x_{cl}(\tau)$ .
- Some calculation gives  $S[x_{cl}(\tau)] = -\frac{1}{6}p^2t^3 + \frac{1}{2t}(x + pt^2/2)^2$ . Compute  $\tilde{\mathcal{P}}_t(p)$ . Deduce  $\langle \mathcal{A}[x] \rangle$  and  $\langle \mathcal{A}[x]^2 \rangle$ .
- Give the distribution (it may be helpful to consider the Fourier transform  $\hat{\mathcal{P}}_t(k) = \langle e^{-ik\mathcal{A}[x]} \rangle = \tilde{\mathcal{P}}_t(ik)$ ).

a)

$$\tilde{\mathcal{P}}_t(p) = \overbrace{\int dx \int_{x(0)=0}^{x(t)=x} \mathcal{D}x(\tau) e^{-\frac{1}{2} \int_0^t d\tau \dot{x}(\tau)^2}}^{\text{sum over paths}} e^{-p\mathcal{A}[x(\tau)]}. \quad (38)$$

weight of a path

- $S[x(\tau)] = \int_0^t d\tau [\frac{1}{2} \dot{x}(\tau)^2 + p x(\tau)]$ , thus  $\frac{\delta S}{\delta x(\tau)} = -\ddot{x}(\tau) + p = 0$  has solution  $x_{cl}(\tau) = \frac{1}{2}p\tau^2 + v_0\tau + x_0$ . Imposing the boundary conditions we get  $x_0$  and  $v_0$ . The solution is  $x_{cl}(\tau) = \frac{1}{2}p\tau(\tau - t) + x\tau/t$ .
- It is easy (but a bit lengthy... not asked) to compute  $S[x_{cl}(\tau)] = \int_0^t d\tau [\frac{1}{2} \dot{x}_{cl}(\tau)^2 + p x_{cl}(\tau)] = -\frac{1}{6}p^2t^3 + \frac{1}{2t}(x + pt^2/2)^2$ . The path integral has a simple form, leading to

$$\tilde{\mathcal{P}}_t(p) = \int dx \frac{1}{\sqrt{2\pi t}} e^{\frac{1}{6}p^2t^3 - \frac{1}{2t}(x + pt^2/2)^2} = e^{\frac{1}{6}p^2t^3} \quad (39)$$

The  $p \rightarrow 0$  expansion of the characteristic function gives the moments  $\tilde{\mathcal{P}}_t(p) = 1 - p \langle \mathcal{A}[x] \rangle + \frac{p^2}{2} \langle \mathcal{A}[x]^2 \rangle + \dots$ , hence  $\langle \mathcal{A}[x] \rangle = 0$  (obvious, by symmetry) and  $\langle \mathcal{A}[x]^2 \rangle = \frac{1}{3}t^3$ .

d) The distribution is obviously Gaussian. If you did not noticed this by inspection of  $\tilde{\mathcal{P}}_t(p)$ , you can consider the Fourier transform

$$\hat{\mathcal{P}}_t(k) = \tilde{\mathcal{P}}_t(ik) = e^{-\frac{1}{6}k^2t^3} \quad (40)$$

which is a Gaussian with inverse Fourier transform

$$\mathcal{P}_t(A) = \sqrt{\frac{3}{2\pi t^3}} e^{-\frac{3}{2t^3}A^2}. \quad (41)$$

In fact, the result could have been obtained by a simpler method :  $x(\tau)$  is Gaussian thus  $\mathcal{A}[x(\tau)] = \int_0^t d\tau x(\tau)$  is a Gaussian random variable. Given the known correlation function  $\langle x(\tau)x(\tau') \rangle = \min(\tau, \tau')$  we easily get  $\langle \mathcal{A}[x]^2 \rangle = \frac{1}{3}t^3$ .