Tutorial sheet 8

Discussion topic: Langevin model of Brownian motion

16. Stochastic processes

Let Y(t) be a stochastic process and $p_{Y,n}$ resp. $p_{Y,n|m}$ its *n*-point resp. conditional *n*-point densities.

i. Starting from the expression of Bayes' theorem, show that for all integers $n \ge 2$ one can write

where the *n* instants t_1, t_2, \ldots, t_n are all different, but otherwise arbitrary.

ii. Consider the previous identity for n = 3. Using the consistency condition which expresses $p_{Y,m}$ as an integral of $p_{Y,n}$ with n > m, here with m = 2, show that the single-time conditional density $p_{Y,1|1}$ obeys the integral-functional equation

$$p_{Y,1|1}(t_3, y_3|t_1, y_1) = \int p_{Y,1|2}(t_3, y_3|t_1, y_1; t_2, y_2) p_{Y,1|1}(t_2, y_2|t_1, y_1) \,\mathrm{d}y_2. \tag{1}$$

Generalize this relation to an equation for $p_{Y,1|n}$ involving $p_{Y,1|n+1}$.

17. Examples of Markov processes

In the lecture, we shall introduce a specific, important class of stochastic processes, the so-called *Markov processes*. These are entirely determined by their single-time density $p_{Y,1}$ and their conditional probability density $p_{Y,1|1}$, which is referred to as *transition probability* and obeys the *Chapman–Kolmogorov equation*

$$p_{Y,1|1}(t_3, y_3 | t_1, y_1) = \int p_{Y,1|1}(t_3, y_3 | t_2, y_2) p_{Y,1|1}(t_2, y_2 | t_1, y_1) \, \mathrm{d}y_2 \quad \text{for } t_1 < t_2 < t_3, \tag{2}$$

to be compared with Eq. (1) in exercise **16.ii**.

i. Wiener process

The stochastic process defined by the "initial condition" $p_{Y,1}(t=0, y) = \delta(y)$ for $y \in \mathbb{R}$ and the transition probability $(0 < t_1 < t_2)$

$$p_{Y,1|1}(t_2, y_2 \,|\, t_1, y_1) = \frac{1}{\sqrt{2\pi(t_2 - t_1)}} \, \exp\left[-\frac{(y_2 - y_1)^2}{2(t_2 - t_1)}\right]$$

is called Wiener process.

Check that this transition probability obeys the Chapman–Kolmogorov equation, and that the probability density at time t > 0 is given by

$$p_{Y,1}(t,y) = \frac{1}{\sqrt{2\pi t}} e^{-y^2/2t}.$$

Remark: Note that the above single-time probability density is solution of the diffusion equation

$$\frac{\partial f}{\partial t} = \frac{1}{2} \frac{\partial^2 f}{\partial y^2}$$

with diffusion coefficient $D = \frac{1}{2}$.

ii. Ornstein–Uhlenbeck process

The so-called *Ornstein–Uhlenbeck process* is defined by the time-independent single-time probability density

$$p_{Y,1}(y) = rac{1}{\sqrt{2\pi}} \, \mathrm{e}^{-y^2/2}$$

and the transition probability $(\tau > 0)$

$$p_{Y,1|1}(t+\tau, y \,|\, t, y_0) = \frac{1}{\sqrt{2\pi(1-\mathrm{e}^{-2\tau})}} \, \exp{\left[-\frac{(y-y_0\mathrm{e}^{-\tau})^2}{2(1-\mathrm{e}^{-2\tau})}\right]}.$$

a) Check that this transition probability fulfills the Chapman–Kolmogorov equation, so that the Ornstein–Uhlenbeck process is Markovian. Show that the process is also Gaussian, stationary, and that its autocorrelation function is $\kappa(\tau) = e^{-\tau}$.

b) What is the large- τ limit of the transition probability? And its limit when τ goes to 0^+ ?

c) Viewing the above transition probability as a function of τ and y, can you find a partial differential equation, of which it is a (fundamental) solution?

Hint: Let yourself be inspired(?) by the remark at the end of question **i**.

18. Characteristic functional of a stochastic process

In the lecture, the characteristic functional associated with a stochastic process $Y_X(t)$ has been defined as

$$G_Y[k(t)] \equiv \left\langle \exp\left[i\int k(t) Y(t) dt\right] \right\rangle,$$

with k(t) a test function.

Expand the exponential in power series of k and express the characteristic functional in terms of the n-time moments. How would you write the moment $\langle Y(t_1)Y(t_2)\cdots Y(t_n)\rangle$ as function of $G_Y[k(t)]$?