

Lecture 4 summary

Ergodicity: for stationary process

ensemble average = time average

$$\langle x \rangle = \bar{x}(T)$$

$$\bar{x}(T) = \frac{1}{2T} \int_{-T}^T dt x(t), \quad T \rightarrow \infty$$

Autocorrelation function $G(\tau) = \langle x(t)x(t+\tau) \rangle$

Power spectrum $S(\omega) = \lim_{T \rightarrow \infty} \frac{\pi}{T} |\hat{x}(\omega; T)|^2$

Wiener-Khinchin theorem

$$\begin{aligned} S(\omega) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} d\tau e^{-i\omega\tau} G(\tau) \\ G(\tau) &= \int_{-\infty}^{\infty} d\omega e^{i\omega\tau} S(\omega) \end{aligned} \quad \left| \begin{array}{l} \text{relation} \\ \text{between ACF} \\ \text{and power spectrum} \\ \text{of a stoch. process} \end{array} \right.$$

1.3 Differential Chapman-Kolmogorov equation

- (i) $\lim_{\Delta t \rightarrow 0} p(x, t + \Delta t | z, t) = W(x | z, t)$ transition probability per unit time
 $z \rightarrow x$ (jump from z to x)
- (ii) $A_i(z, t)$ drift
- (iii) $B_{ij}(z, t)$ diffusion

$$\begin{aligned} \frac{\partial}{\partial t} p(z, t | y, t') &= - \sum_i \frac{\partial}{\partial z_i} [A_i(z, t) p(z, t | y, t')] + \\ &+ \sum_{ij} \frac{1}{2} \frac{\partial^2}{\partial z_i \partial z_j} [B_{ij}(z, t) p(z, t | y, t')] + \\ &+ \int dx \left[\underset{\substack{\uparrow \\ \text{prob. of trans. per unit time}}}{W(z|x, t)} p(x, t | y, t') - W(x|z, t) p(z, t | y, t') \right] \end{aligned}$$

differential Chapman-Kolmogorov equation

Each of conditions (i), (ii), (iii) can now be seen to give rise to a distinctive part of the equation, whose interpretation can be provided. We can identify three processes taking place, which are known as jumps, drift and diffusion.

(a) jump processes: $A_i(z, t) = B_{ij}(z, t) = 0$

$$\frac{\partial}{\partial t} p(z, t | y, t') = \int dx [W(z|x, t) p(x, t | y, t') - W(x|z, t) p(z, t | y, t')] \quad \text{Master equation}$$

To first order in Δt we can solve approx-ly.

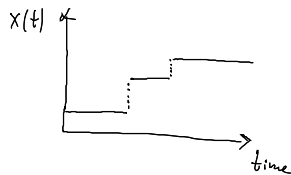
We notice that

$$p(z, t | y, t) = \delta(y - z). \quad \text{Therefore,}$$

$$p(z, t + \Delta t | y, t) = \delta(y - z) \left[1 - \int dx W(x | y, t) \Delta t \right] + \underbrace{W(z | y, t) \Delta t}$$

We see that for any Δt there is a finite probability given by the coefficient of the $\delta(y - z)$, for the particle to stay at the original position y . And the distribution of particles which do not remain at y is given by $W(z | y, t)$.

\Rightarrow a typical path $x(t)$ will consist of sections of straight lines $x(t) = \text{const}$ combined with discrete jumps whose distribution is given by $W(z | y, t)$.



(b) diffusion processes (continuous transitions)

$$W(z | x, t) = 0, \quad A_i(z, t) = 0$$

$$\frac{\partial}{\partial t} p(z, t | y, t') = \frac{1}{2} \sum_{ij} \frac{\partial^2}{\partial z_i \partial z_j} \{ B_{ij}(z, t) p(z, t | y, t') \}$$

B_{ij} is diffusion matrix

The diffusion matrix B_{ij} is positive semi-definite and symmetric as a result of its definition (see assumption (ii))

$$B_{ij} = B_{ji}$$

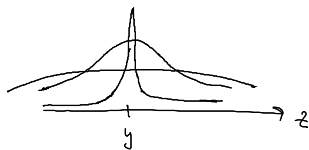
For one-dim. $\frac{1}{2} B_{ij} = \mathcal{D}$
diffusion constant

$$\boxed{\frac{\partial}{\partial t} p = \mathcal{D} \frac{\partial^2}{\partial z^2} p} \quad \text{diffusion equation}$$

Solution for 2.C. $p(z, t | y, t) = \delta(z - y)$ and small Δt :

$$p(z, t + \Delta t | y, t) = \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{2\mathcal{D}\Delta t}} \exp \left\{ -\frac{(z-y)^2}{4\mathcal{D}\Delta t} \right\}$$

Gaussian (normal) distribution with variance $\sigma^2 = 2\mathcal{D}\Delta t$ and mean y .



(c) Drift (contin.)

The first term is non-zero, so we are led to the special case of a Liouville equation:

$$\frac{\partial p(z, t | y, t')}{\partial t} = - \sum_i \{ A_i(z, t) p(z, t | y, t') \}$$

which occurs in classical mechanics; $A_i(z, t)$ is a drift vector.

This equation describes a completely deterministic motion, i.e., if $x(y, t)$ is the solution of the $\mathcal{D}\mathcal{D}E$:

$$\textcircled{*} \quad \frac{dX(t)}{dt} = A[X(t), t] \text{ with } X(y, t') = y$$

then the solution of the Liouville eq. with d.c.

$$p(z, t' / y, t') = \delta(z - y)$$

$$\text{is } p(z, t / y, t') = \delta[z - x(y, t)]$$

The proof of this statement is best obtained by direct substitution:

$$- \sum_i \frac{\partial}{\partial z_i} \{ A_i(z, t) \delta[z - x(y, t)] \} =$$

$$= - \sum_i \frac{\partial}{\partial z_i} \{ A_i[x(y, t), t] \delta[z - x(y, t)] \}$$

$$= - \sum_i \left\{ A_i[x(y, t), t] \frac{\partial}{\partial z_i} \delta[z - x(y, t)] \right\} \quad (**)$$

$$\text{and } \frac{\partial}{\partial t} \delta[z - x(y, t)] = - \sum_i \frac{\partial}{\partial z_i} \delta[z - x(y, t)] \frac{dx_i(y, t)}{dt} \quad (***)$$

and by use of $\textcircled{*}$ we see that $(**)$ and $(***)$ are equal \square .

Thus, if the particle is in a well-defined position (state) y at time t' , it stays on the trajectory obtained by solving the DDE $\textcircled{*}$ (deterministic motion of a particle).

Combination (b) and (c)

We assume that $W(z, t)$ is zero. In this case the diff. Chapman-Kolmogorov equation reduces to the Fokker-Planck equation:

$$\frac{\partial p(z, t / y, t')}{\partial t} = - \sum_i \frac{\partial}{\partial z_i} \left\{ A_i(z, t) p(z, t / y, t') \right\} + \frac{1}{2} \sum_{ij} \frac{\partial^2}{\partial z_i \partial z_j} \left\{ B_{ij}(z, t) p(z, t / y, t') \right\}$$

Hence, the FPE describes a process in which $X(t)$ has continuous sample paths. Let us consider computing $p(z, t + \Delta t / y, t)$, given $p(z, t / y, t) = \delta(z - y)$ $(*)$ d.c.

For small Δt , the solution of the FPE will still be on the whole sharply peaked, and hence derivatives of $A_i(z, t)$ and $B_{ij}(z, t)$ will be negligible compared to those of p . Therefore,

$$\frac{\partial p(z, t / y, t')}{\partial t} = - \sum_i A_i(y, t) \frac{\partial p(z, t / y, t')}{\partial z_i} + \sum_{ij} \frac{1}{2} B_{ij}(y, t) \frac{\partial^2 p(z, t / y, t')}{\partial z_i \partial z_j} \quad (***)$$

where we have also neglected the time dependence of A_i and B_{ij} for small $t - t'$. Equation $(***)$ can now be solved, subject to d.c. $(*)$ and we get:

$$p(z, t + \Delta t / y, t) = \left(\frac{1}{2\pi} \right)^{\frac{N}{2}} \left\{ \det [B(y, t)] \right\}^{-\frac{1}{2}} [\Delta t]^{-\frac{1}{2}} \times \\ \times \exp \left\{ -\frac{1}{2} \frac{[z - y - A(y, t) \Delta t]^T [B(y, t)]^{-1} [z - y - A(y, t) \Delta t]}{\Delta t} \right\},$$

that is, a Gaussian distribution with variance matrix $B(y, t)$ and mean $y + A(y, t) \Delta t$. We get the picture of a system moving with a systematic drift, whose velocity is $A(y, t)$, on which is superimposed a Gaussian fluctuation with covariance matrix $B(y, t) \Delta t$, that is, we can write

$$y(t + \Delta t) = y(t) + A(y(t), t) \Delta t + \eta(t) \Delta t^{\frac{1}{2}}$$

where $\langle \eta(t) \rangle = 0$

$$\langle \eta(t) \eta(t)^T \rangle = B(y, t)$$

i) sample paths which are always continuous — for, clearly, as $\Delta t \rightarrow 0$, $y(t + \Delta t) \rightarrow y(t)$;

ii) sample paths which are nowhere differentiable, because of the $\Delta t^{\frac{1}{2}}$

2. Classical statistics in non-equilibrium

2.1 Master equation

For Markov processes there is Chapman-Kolmogorov equation

$$p(x, t'' / x', t') = \int dz p(x, t'' / z, t) p(z, t / x', t') \quad (1) \\ t'' \geq t \geq t'$$

For jump processes

$$W(x/z, t) = \lim_{\Delta t \rightarrow 0} \frac{p(x, t + \Delta t / z, t)}{\Delta t}$$

probability of transition per unit time

We assume Δt is small

$$p(x, t + \Delta t / z, t) = \delta(x - z) \left[1 - \int dx_2 W(x_2 / z, t) \Delta t \right] + W(x / z, t) \Delta t$$

probability of transition per unit time from z to some state x_2

probability that no transition will take place within interval $[t, t + \Delta t]$