

$$J[P_{eq}] = 0$$

\Rightarrow Reversibility .

[impossible to distinguish between a trajectory and its time reverse]

(explained next time) .

Use of systems at equilibrium.
How related to that?

Steady state $P_{st}(x)$:

$\rightarrow t \rightarrow \infty$
 if exists, will
 converge to it.

$$\partial_x [-F(x, t) P_{st}(x) + D \partial_x P_{st}(x)] = 0.$$

Ex: 1 particle on \mathbb{R} $F=0$:

$$P(x, t) = \frac{1}{\sqrt{4\pi Dt}} \exp\left(-\frac{1}{2} \frac{x^2}{Dt}\right) \xrightarrow{t \rightarrow \infty} 0$$

No steady
 state

② 1 particle on $[a, b]$, isolated.

$$\partial_x^2 P_{st}(x) = 0 \rightarrow P_{st}(x) = \frac{1}{b-a} \quad \left. \begin{array}{l} \text{uniform} \\ \text{distrib.} \end{array} \right\}$$

• Current of probabilities $J[P]$: FPEq:

$$\partial_t P = -\partial_x J[P]. \quad \text{with } J[P] = -[-FP + D \partial_x P].$$

• steady state: $\partial_x J[P_{st}] = 0$

• by def. on equilibrium steady state P_{eq} verifies:

$$J[P_{eq}] = 0, \text{ i.e. } F(x, t) P_{eq}(x) = +D \partial_x P_{eq}(x).$$

1D: $F(x) = -V'(x)$

$$V'(x) P_{eq}(x) = D \partial_x P_{eq}(x)$$

$$P_{eq}(x) = \exp\left(-\frac{1}{D} V(x)\right)$$

Boltzmann distribution in a
 potential V at "temperature" D .
 Thermodyn. results at long time.

$$= \int dx_1 P(x_1, t) [\varphi(x_1) + \delta t (F(x, t) \varphi' + D \varphi'')]]$$

$$\langle \varphi(x) \rangle_t$$

So we get

F supposed regular to have nice b.c.
but one can think to $\langle \dot{\varphi}(t) \rangle$ as a r.v.

$$\frac{\langle \varphi(x) \rangle_{t+\delta t} - \langle \varphi(x) \rangle_t}{\delta t} = \langle F(x, t) \varphi'(x) + D \varphi''(x) \rangle_t$$

(if F not regular, more complicated: no Brownian motion.
only solved recently ...)

eq:

$$\langle F(x, t) \varphi'(x) + D \varphi''(x) \rangle_t = \delta t \langle \varphi(x) \rangle_t$$

math. : φ is a test function - (if distributable) due to the noise - what is a
(going back for observable to try?) here?

3rd try: $\partial_t x_t = \eta + F$

informally:

$$\partial_t \varphi(x_t) = \varphi'(x_t) \partial_t x_t$$

$$\partial_t \langle \varphi(x) \rangle = \langle F(x, t) \varphi'(x) \rangle + \langle \eta \varphi'(x) \rangle$$

↳ PB to determine

= $D \langle \varphi''(x) \rangle$ from precedent calculation

NOT OBVIOUS -

Equation obtained by Ito formula.

Point of view of trajectories:

$$\partial_t x = F(x(t), t) + \eta(t), \quad \begin{cases} n=0 \\ \gamma=1 \\ \neq 0 \end{cases}$$

other method without discretizing space.

(\rightarrow Fokker-Planck - generalized F.P. from Langevin eq.)
[Kramers Eq.]

Let's discretize time & keep x continuous:

$$\int_t^{t+\delta t} \cdot$$

$$x_{t+\delta t} - x_t = \underbrace{\int_t^{t+\delta t} dt' F(x(t'), t')}_{\approx \delta t F(x_t, t)} + \underbrace{\int_t^{t+\delta t} dt' \eta(t')}_{\eta_t}$$

"local mean field" \approx η_t

\approx $O(\delta t)$ $O(\sqrt{\delta t})$

$$\eta_t = B(t+\delta t) - B(t).$$

$$\rightarrow B(t) = \int_0^t dt' \eta(t') : \text{Brownian motion}$$

$$\langle \eta_t \rangle = 0$$

$$\langle \eta_t \eta_t \rangle = \langle (B(t+\delta t) - B(t))^2 \rangle = C(\delta t) = 2D \delta t$$

$$\rightarrow \eta_t = O(\sqrt{\delta t}) : \text{"time increments of } B \text{ scales as } \sqrt{\delta t}$$

why important to have this scale property? \rightarrow see flier

Evaluation of $P(x, t)$ - 1st try:

$$P(x, t+\delta t) = \int dx_1 P(x_1, t) P(x, t+\delta t | x_1, t)$$

How depends on the discretisation we took?

in the limit:

$$\lim_{h \rightarrow 0} \sum_{k=1}^N dt \left(\frac{x_k - x_{k-1}}{dt} \right)^2 \xrightarrow{dt \rightarrow 0} \int_0^t dt' (\partial_{t'} x)^2$$

(Riemann Sum).

Pre-factor? \rightarrow both π and h of normalisation.

Continuous Time notations:

$$\int \frac{dx_0 \dots dx_N}{[4\pi D dt]^{N/2}} \stackrel{(\text{def})}{=} \int Dx \quad \leftarrow \text{path integral over all possible paths}$$

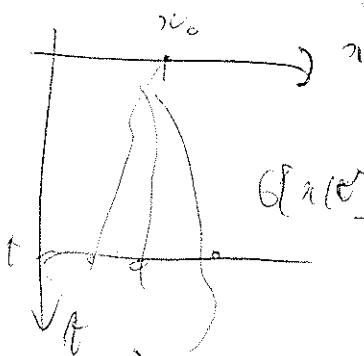
$$\langle G[x(t)] \rangle = \int Dx \cdot G[x(t)] e^{-\frac{1}{2} \int_0^t dt' \frac{(\partial_{t'} x)^2}{2D}} P(x_0)$$

depends not of the discretisation. (if G reasonable!).

In a quantum system \rightarrow "path integral".

Describe which depends on the full trajectory.

e.g. $G[x] = \int_0^t dt' x^2(t') \dots$



$$G[x(t)] = \int dt' E(x(t'), t')$$

Wiener Integration...

What about the white noise?

Goal: How can we describe systems with noise?
what type of noise?

$$x_{t+\delta t} = x_t + \dot{\eta}_t \delta t \quad \text{with } \dot{\eta}_t = \begin{cases} \delta x & \delta \\ -\delta x & \delta \\ 0 & 1-2\delta \end{cases}$$

$$\left| \begin{array}{l} \delta x \rightarrow 0 \\ \delta t \rightarrow 0 \end{array} \right. \quad \delta \frac{\delta x^2}{\delta t} \rightarrow D$$

$$\boxed{\partial_t P = D \partial_x^2 P} \quad \eta(t) = \frac{\delta x}{\delta t}$$

$\dot{x}(t) = \eta(t)$ $\xrightarrow{\text{continuum space}}$ white noise.

$x(t)$: Brownian motion.

$$P(x, t) = \frac{1}{\sqrt{4\pi Dt}} \exp\left(-\frac{1}{2} \frac{x^2}{2Dt}\right) \quad \text{with } x(0)=0$$

Properties of Brownian and White Noise.

$$\left[\begin{array}{l} \langle [x(t_2) - x(t_1)]^2 \rangle = 2D|t_2 - t_1| \\ \langle \eta(t_2) \eta(t_1) \rangle = 2D \delta(t_2 - t_1) \end{array} \right]$$

Fully describe Brownian
because Gaussian process.

Now more powerful tool to describe stochastic processes -

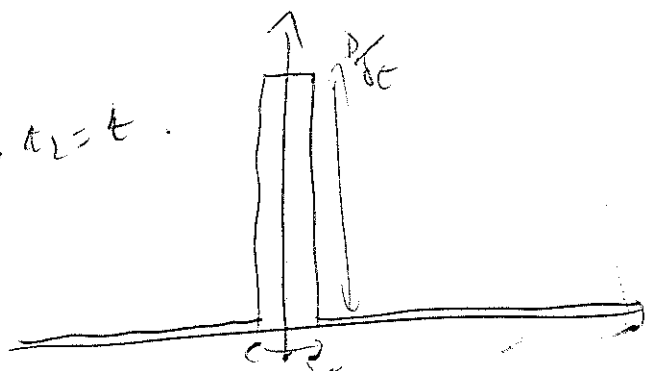
Remark: (from micro description)

$$t_1 \neq t_2: \underbrace{\langle \eta(t_2) \eta(t_1) \rangle}_{\text{independent}} = 0$$

$$t_1 = t_2 = t = \frac{1}{\delta t} \langle \dot{\eta}^2 t \rangle = \frac{1}{\delta t} \left(\delta x^2 \overset{\text{prob.}}{\delta} + \delta x^2 \delta + 0 \right)$$

$$= 2\delta \frac{\delta x^2}{\delta t} = \boxed{\frac{2D}{\delta t}}$$

$$t_1=0, t_2=t$$



$$\xrightarrow{\delta t \rightarrow 0} 2D \delta(t_2 - t_1)$$

$\delta t \rightarrow 0$ is limiting width.

other way
to obtain
the result.