

§6 | $\frac{1}{h_0}$ vs. Stochastic

• Q: Given an ODE, e.g.

$$\dot{x} = f(x) :$$

What does it mean?

A: Constructive approach:
estimate ~~for~~ $x_i = x(i \cdot \Delta t)$, numerically
then let $\Delta t \rightarrow 0$.

• Possibility no 1: ^{explicit} Euler scheme

$$x_i = x_{i-1} + f(x_{i-1}) \cdot \Delta t.$$

• Possibility no 2: implicit

$$x_i = x_{i-1} + f(x_i) \Delta t.$$

• Possibility no 3: mixed

$$x_i = x_{i-1} + \frac{1}{2} [f(x_{i-1}) + f(x_i)] \Delta t.$$

\Rightarrow For deterministic ODEs
different schemes differ only
in terms of numerical
performance, stability, etc.
Not in the limit result. (14)

- What about SDEs ?

$$\dot{X} = f(x) + \sqrt{2D(x)} \dot{I}$$

$$\langle \dot{I}(t) \dot{I}(t') \rangle = \delta(t - t')$$

- Explicit Euler scheme:

$$X_i = X_{i-1} + f(X_{i-1}) \Delta t + \sqrt{2D(X_{i-1})} \mathcal{N} \cdot \sqrt{\Delta t}$$

\uparrow
 \equiv normally distributed normal variable
 mean 0, variance 1
 \equiv Wiener increment

\longrightarrow Ito

- Mixed scheme:

$$X_i = X_{i-1} + \frac{1}{2} [f(X_{i-1}) + f(X_i)] \Delta t + \frac{1}{2} [\sqrt{2D(X_{i-1})} + \sqrt{2D(X_i)}] \mathcal{N} \cdot \sqrt{\Delta t}$$

\longrightarrow Stratonovich

$$X_i = X_{i-1} + f(X_{i-1}) \cdot \Delta t + O(\Delta t^{3/2})$$

$$+ g(X_{i-1}) \cdot \mathcal{N} \cdot \sqrt{\Delta t}$$

$$+ g'(X_{i-1}) \cdot g(X_{i-1}) \cdot \mathcal{N}^2 \cdot \Delta t + O(\Delta t^{3/2})$$

$$g(x) = \sqrt{2D(x)}$$

(15)

$\Rightarrow \langle W^2 \rangle = 1 \neq 0$

\Rightarrow Ito vs. Stratonovich

interpretation give different drift terms.

Wrap-up

• Static & non-linear SDE

with local upscaling in

interpretation does not make sense

(it's like writing a text

and not performing an audit

language it was written)

many more interpretations:

• Ito-Stratonovich

• algebra-calculus by

Itô & Stratonovich, PDE

• A Stratonovich SDE

$SDE(S) \dot{X} = f(X) + g(X) \dot{W}$

can be rewritten as

Ito SDE (I) $\dot{X} = f(X) + g(X) \dot{W} + \frac{1}{2} g^2(X)$

and vice versa.

We use into notation

mostly suitable for a given task

Stratonovich : ordinary chain rule
 Ito chain rule $y(x) : \dot{y} = \frac{\partial y}{\partial x} \dot{x}$

$$(I) \quad \dot{x}_k = f_k + g_{ke} \dot{z}_e$$

$$\langle \dot{z}_e(t_1) \dot{z}_e(t_2) \rangle = \delta_{ee} \delta(t_1 - t_2)$$

$$y = y(x)$$

$$\dot{y} = \frac{\partial y}{\partial x_j} \dot{x}_j + \frac{1}{2} \frac{\partial^2 y}{\partial x_k \partial x_l} g_{km} g_{ml}$$

Switching

between Ito & Stratonovich

(S)

$$\dot{x}_k = f_k + g_{ke} \dot{z}_e$$

(I)

$$\dot{x}_k = f_k + g_{ke} \dot{z}_e + \frac{1}{2} \frac{\partial g_{ke}}{\partial x_m} g_{ml}$$

Fokker-Planck

equation

refers to entire drift term.

$$\dot{P} = \frac{\partial}{\partial x_k} \left[\underbrace{-(f_k + L \frac{\partial g_{ke}}{\partial x_m} g_{ml})}_{\text{det. drift}} P + \underbrace{\frac{1}{2} \frac{\partial}{\partial x_m} (g_{ke} g_{ml})}_{\text{noise induced drift}} P \right]$$

- $L=0$
- $L=1/2$
- $L=1$

: Ito
 : Stratonovich
 : Ito - formal

Lau-Lubensky
 PRE

$$\langle \gamma(t) \gamma(t') \rangle = \delta(t-t')$$

$$\Gamma \dot{z} = -z + \eta$$

Example of colored noise (Ornstein-Uhlenbeck process)

with Gaussian white noise ξ

Stokman's SDE

with yield a

then taking the limit

finite correlation time τ

colored noise of

is an SDE with

$$\dot{x} = f(x) + g(x)\xi$$

Theorem (Ito-Stratonovich):

applies:

Physics: often Stratonovich (weak), often Ito-Stratonovich theorem

the part

frome not looking back into

Example: rotational diffusion in 2D.

$$\dot{\varphi} = \xi$$



$$e_1 = \begin{pmatrix} \cos \varphi \\ \sin \varphi \end{pmatrix}, \quad e_2 = \begin{pmatrix} -\sin \varphi \\ \cos \varphi \end{pmatrix}$$

$$So(2): \quad \frac{d}{dt} [e_1, e_2] = [e_1, e_2] \cdot \xi \cdot \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

$$(S) \quad \begin{cases} \dot{e}_1 = \xi e_2 \\ \dot{e}_2 = -\xi e_1 \end{cases}$$

$$\downarrow \quad e_1 = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \quad e_2 = \begin{pmatrix} x_3 \\ x_4 \end{pmatrix}, \quad g = \sqrt{2D} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} \Rightarrow \sum_m \frac{\partial g^k}{\partial x_m} g^m = 2D \begin{pmatrix} x_1 \\ x_2 \\ -x_3 \\ -x_4 \end{pmatrix}$$

$$(I) \quad \begin{cases} \dot{e}_1 = \xi e_2 + \frac{1}{2} \left\langle \xi \frac{\partial e_2}{\partial e_1} \cdot \dot{e}_1 \right\rangle + \frac{1}{2} \left\langle \xi \frac{\partial e_2}{\partial e_2} \cdot \dot{e}_2 \right\rangle \\ \dot{e}_2 = -\xi e_1 + \frac{1}{2} (2D_{rot}) \cdot (-e_1) \end{cases}$$

correct but not needed.

$$= \xi e_2 + \frac{1}{2} - D_{rot} e_1$$

$$\dot{e}_2 = -\xi e_1 - D_{rot} e_2$$

Exercise:

Run simulations for

$$(I) \quad \dot{e}_1 = \xi e_2 - D_{rot} e_1$$

$$\dot{e}_2 = -\xi e_1 - D_{rot} e_2$$

and

$$(II) \quad \dot{e}_1 = \xi e_2$$

$$\dot{e}_2 = -\xi e_1$$

log example I (geometrische Brownian motion)

(5) $\dot{X} = X \xi - D X \quad \equiv \quad (I) \quad \dot{X} = X \tilde{\xi}$

$\langle \int(t) \int(t) \rangle = 2D \int(t-t)$

$m(t) = \langle X(t) \rangle$

$\frac{d}{dt} m = \langle \dot{X} \rangle = \langle X \tilde{\xi} \rangle$

$\langle X \rangle \cdot \langle \tilde{\xi} \rangle = 0$

$\rightarrow m(t) = m_0$

but variance grows exponentially

Note: $y = \ln x \Rightarrow y = \int - D \Rightarrow y \sim W(-D, D)$

log

example II

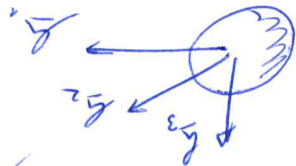
(5) $\dot{X} = X \xi \quad \equiv \quad (I) \quad \dot{X} = X \int + D X$

$\frac{d}{dt} m = \langle \dot{X} \rangle = \langle X \int + D X \rangle = 0 + D m$

$\Rightarrow m = m_0 \exp(Dt)$

→ tutoren 7

Rotational diffusion in 3D



$$D^{rot} = \frac{8\pi^2 \eta^{1/3}}{k_B T}$$

Spherical coordinates

$$g_3 = (\cos \varphi, \sin \varphi \cos \theta, \sin \varphi \sin \theta)$$

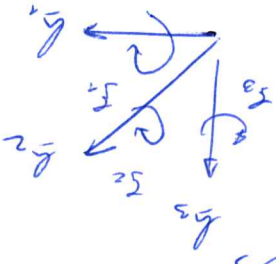
$$g_1 = g_3 = (\cos \varphi, \sin \varphi \cos \theta, \sin \varphi \sin \theta)$$

$$g_2 = (0, -\sin \theta, \cos \theta) = k_3 \times g_1$$

$$k_1 = \cos \varphi g_1 + \sin \varphi g_2$$

$$k_2 = k_3 \times k_1$$

Translational - rotational equations

$$\begin{cases} \dot{k}_3 = \dot{\varphi} k_1 - \dot{\theta} k_2 \\ \dot{k}_1 = -\dot{\varphi} k_3 + \dot{\theta} k_2 \\ \dot{k}_2 = -\dot{\varphi} k_1 - \dot{\theta} k_3 \end{cases}$$


$$\Rightarrow \dot{\varphi} = \sin \varphi \dot{\varphi}_1 + \cos \varphi \dot{\varphi}_2$$

$$\Rightarrow \dot{\varphi} = \sin \varphi \dot{\varphi}_1 + \cos \varphi \dot{\varphi}_2$$

Equivalent to Gaussian (think vector $S(t)$) with $\langle S(t_1) \cdot S(t_2) \rangle = 2 D^{rot} S(t_1 - t_2)$

$$\boxed{\varphi = D^{rot} \cos \varphi + \dot{\varphi}}$$

Steady state distribution
 Must be isotropic \Rightarrow check
 Potts



• What is $P^*(\psi)$?

$$L = 1 - \cos \psi$$

$$A = 2\pi \cdot r \cdot L, \quad r = 1$$

$$dA = 2\pi \cdot dL$$

$$\Rightarrow P^*(L) = \frac{1}{2}$$

$$P^*(L) dL = P^*(\psi) d\psi$$

$$dL = \sin \psi d\psi$$

$$\Rightarrow P^*(\psi) = \frac{1}{2} \sin \psi$$

$$\psi = \arccos(1 - L)$$

$$= -\frac{1}{2} \frac{\partial}{\partial L} (1 - L) = \frac{1}{2}$$

$$\Rightarrow P^*(\psi) \sim \exp\left(-\frac{D_{\text{rot}}}{U} \psi\right)$$

$$\sim \exp(L \sin \psi) \sim \sin \psi$$

[S] $\frac{1}{K}$, [L] T , [D_{rot}] $\frac{1}{s}$, [T] N_{rot}

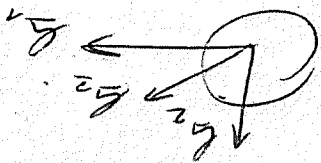
• Interpretation of $U(\psi)$

• entropy $S = k_B \ln \Omega(\psi)$

• free energy $F = -TS$

$$= -D_{\text{rot}} \ln \sin \psi = U_i$$

Rotational Diffusion in 3D



$$\tau_{rot} = \frac{LRT}{D_{rot}}$$

spherical coordinates
 $\alpha_3 = (\cos \alpha_1, \sin \alpha_1, 0)$
 $\alpha_2 = (\cos \alpha_1, \sin \alpha_1, 0)$
 $\alpha_1 = \alpha_2 \times \alpha_3$

tensor-dot equations

$$\left. \begin{aligned} \alpha_3 &= \alpha_2 \alpha_1 - \alpha_1 \alpha_2 \\ \alpha_2 &= -\alpha_2 \alpha_3 + \alpha_3 \alpha_2 \\ \alpha_1 &= -\alpha_1 \alpha_3 - \alpha_3 \alpha_1 \end{aligned} \right\} (5)$$

(5) \rightarrow
 (5) \rightarrow
 (5) \rightarrow

$$\alpha_1 = \alpha_2 \alpha_3 + \alpha_3 \alpha_2$$

$$\alpha_1 = \alpha_2 \alpha_3 + \alpha_3 \alpha_2 + \tau_{rot} \alpha_1$$

equivalent to $\alpha_1(t) > \alpha_2(t) > \alpha_3(t)$
 Note that $\alpha_1(t) > \alpha_2(t) > \alpha_3(t)$
 $\alpha_1(t) > \alpha_2(t) > \alpha_3(t)$

$$\alpha_1 = \alpha_2 + \alpha_3$$

Extended example:

Persistent random walk (2D).

$$\Gamma = v_0 \dot{\underline{e}}_1$$

$$\underline{e}_1 = \begin{pmatrix} \cos \varphi \\ \sin \varphi \end{pmatrix}, \quad \dot{\varphi} = \Gamma, \quad \langle \underline{e}_1(t_1) \cdot \underline{e}_1(t_2) \rangle = 2D \cos \varphi(t_1 - t_2)$$

Proposition

$$C(t) = \langle \underline{e}_1(0) \cdot \underline{e}_1(t) \rangle = \frac{v_0}{\Gamma} \exp(-D \Gamma t)$$

$$t_p = \frac{1}{D \Gamma} \equiv \text{persistence time}$$

$$l_p = v_0 t_p \equiv \text{persistence length}$$

Proof:

$$\begin{aligned} \frac{d}{dt} C(t) &= \langle \underline{e}_1(0) \cdot \dot{\underline{e}}_1(t) \rangle \\ &= \langle [\underline{e}_1(0) \cdot \underline{e}_1(t)] [\underline{e}_1(t) \cdot \dot{\underline{e}}_1(t)] \rangle \\ &\quad + \langle [\underline{e}_1(0) \cdot \underline{e}_2(t)] [\underline{e}_2(t) \cdot \dot{\underline{e}}_1(t)] \rangle. \end{aligned}$$

+ add intermediate step.

1/20 Calculus: factors independent

$$\begin{aligned} &= \langle \underline{e}_1(0) \cdot \underline{e}_1(t) \rangle \cdot \langle \underline{e}_1(t) \cdot \dot{\underline{e}}_1(t) \rangle \\ &\quad + \langle \underline{e}_1(0) \cdot \underline{e}_2(t) \rangle \cdot \langle \underline{e}_2(t) \cdot \dot{\underline{e}}_1(t) \rangle \end{aligned}$$

$$= C(t) \cdot (-D \Gamma) + 0$$

$$\frac{d}{dt} C(t) = -D \Gamma C(t)$$

$$= -D \Gamma C(t)$$

g.e.d. (JBB)

Outlook: + ext. field.

$$= u \quad i$$

$$= -D_{\text{ext}} \quad \text{in } \text{ext. field}$$

$$F = -TS$$

• free energy

• entropy $S = k_B \ln \Omega$

[R] $\frac{1}{\text{mole}}$

[Dose] $\frac{1}{\text{mole}}$

[k_BT] = Nm

[S] $\frac{1}{\text{mole}}$

partition function $Z(\mu)$

$$\sim \exp(-\beta \langle E \rangle) \sim \Omega$$

$$p^*(\mu) \sim \exp\left(-\frac{\beta \langle E \rangle}{\Omega}\right)$$

$$\beta = \frac{1}{k_B T} = \frac{1}{\text{energy}}$$

$$\psi = -\frac{1}{\beta} \ln Z + \dots$$

$$\psi = D_{\text{ext}} \psi + \dots$$

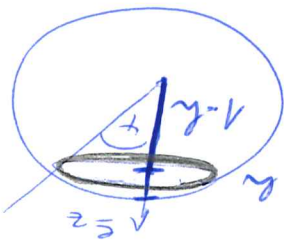
$$A = -D_{\text{ext}} \psi \quad \text{in } \text{ext. field}$$

$$dA = -\psi d\mu$$

$$P(A) dA = P(\mu) d\mu$$

$$P(\mu) = \frac{1}{Z} \Omega$$

$$P(A) = \frac{1}{Z}$$



$$k = 1 - \cos \theta$$

isotropic

mean $\langle \mu \rangle$

disorder

steady state

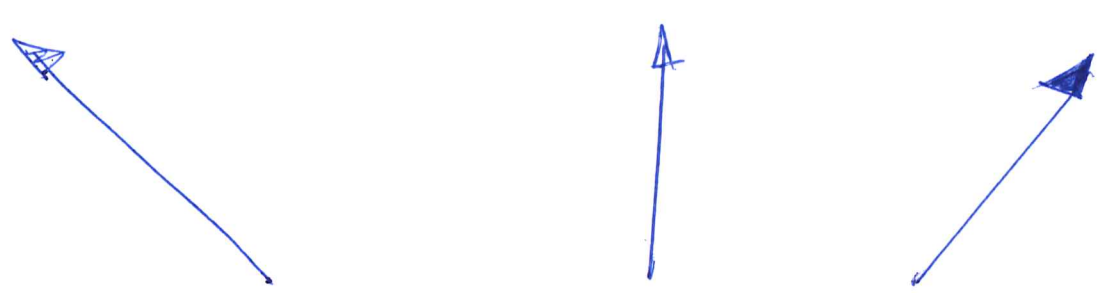
$$A = 2\psi k$$

How to derive a correct Langevin eqn?

Limit case of colored noise
 ↑
 Weyl-Liouville

Small-number fluctuations
 (e.g. N particles)
 $\frac{1}{\sqrt{N}}$
 ↑
 Continuum limit of master equation

Only Markov fluctuations, $I = \text{const.}$
 ↑
 use $P^* \sim \exp(-p \cdot u)$





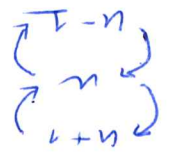
Master equation for two-state system



$P(u) \equiv$ prob. n entries in state u

• $\dot{P}(u, t) = w_1(u+1)P(u+1, t) - w_1 u P(u, t) + w_2(u-1)P(u-1, t) - w_2 u P(u, t)$

• $\dot{P}(u, t) = w_1 (E+ - 1) u P + w_2 (E- - 1) (N-u) P$



$E_{\pm} \equiv$ step operators (shift operators)

$(E+ f)(u) = f(u+1)$

$(E- f)(u) = f(u-1)$

• Let $X = u/N$, treat X as continuous variable

• Taylor expansion

$(E_{\pm} f)(x) = f(x) \pm f'(x) \cdot \frac{N}{1} + \frac{1}{2} f''(x) \cdot \left(\frac{N}{1}\right)^2 + \dots$

\Rightarrow

$\dot{P}(x, t) = w_1 \frac{\partial}{\partial x} (xP) + \frac{w_1}{2N} \frac{\partial^2}{\partial x^2} (x^2 P) - w_2 \frac{\partial}{\partial x} [(1-x)P] + \frac{w_2}{2N} \frac{\partial^2}{\partial x^2} [(1-x)^2 P]$

$= (w_1 + w_2) \frac{\partial}{\partial x} [(x-x^*)P] + \frac{1}{2} \frac{\partial^2}{\partial x^2} [2N \alpha x^2 - (w_1 + w_2)x^2 + (w_1 - w_2)x^2] P$

• Steady state? $X^* = \frac{w_2}{w_1 + w_2}$

Special case $w_1 = w_2 = w_0$

Conservation equation $\dot{P} = -\Delta P$

\Rightarrow at equilibrium $\dot{P} = 0$

$\Rightarrow P(x) \sim \exp(-\frac{(x - 1/2)^2}{2 \sigma^2})$, $\sigma^2 = \frac{1}{4N}$

Langevin equation

$$(I) \quad \dot{X} = (v_1 + v_2) (X^* - X) + \underbrace{\frac{g(x)}{2N}}_{\text{noise}}$$

$$(S) \quad \dot{X} = (v_1 + v_2) (X^* - X) + \underbrace{-\frac{1}{2} \frac{\partial g}{\partial x}}_{\text{noise}}$$

$$N \gg 1 \Rightarrow X^* \approx X \Rightarrow g(x) \approx g(x^*) \Rightarrow \dot{X} = v^*(x) = N(x^*, \sigma^2), \quad \sigma^2 = \frac{1}{4N} \frac{N \cdot (v_1 + v_2)^2}{v_1 v_2}$$

Applications

- Chemical reaction kinetics
- Small number fluctuations
- e.g. finding function of molecules (method)

• Euler-Schema für Ito SDEs

(I)

$$\dot{X} = f(X) + g(X) \dot{W}(t)$$

$$\langle \dot{X}(t), \dot{X}(t) \rangle = \sigma^2(t)$$

$$X_{t+\Delta t} = X_t + f(X_t) \Delta t + g(X_t) \dot{W}_t$$

$\dot{W}_t \sim W(0, \Delta t)$

• Euler-Schema für Stratonovich SDEs

(S)

$$\dot{X} = f(X) + g(X) \dot{W}(t)$$

$$\langle \dot{X}(t), \dot{X}(t) \rangle = \sigma^2(t)$$

$$X_{t+\Delta t} = X_t + f(X_t) \Delta t + \frac{1}{2} [g(X_t)]^2 \Delta t + g(X_t) \dot{W}_t$$

$\dot{W}_t \sim W(0, \Delta t)$

$$\dot{X}_t = X_t + g(X_t) \dot{W}_t$$

