

General idea: stochastic process $x(t) \Rightarrow$ time evolution of $P(x, t)$ the probab that $x(t) = x$.

I) Langevin and Fokker-Planck

$$\dot{x}_i(t) = f_i(\vec{x}) + \gamma_i(t) \quad \text{where } \gamma_i \text{ GWN, centered, } \langle \gamma_i(t) \gamma_j(t') \rangle = 2\sigma \delta(t-t') \delta_{ij}$$

as a useful trick

$$P(x, t) = \int dx_0 \delta(x - x_0) P(x_0, t)$$

$$= \langle \delta(x - x_0(t)) \rangle_{x_0(t)}$$

$$P(\vec{x}, t) = \langle \delta(\vec{x} - \vec{x}_0(t)) \rangle_{\vec{x}_0(t)}$$

$$\delta(\vec{x}) = \prod_i \delta(x_i)$$

b) Itô calculus

$$\frac{\partial}{\partial x} f(x-y) = -\frac{\partial}{\partial y} f(x-y)$$

$$\frac{d}{dt} P(\vec{x}, t) = \left\langle \frac{d}{dt} \delta(\vec{x} - \vec{x}_0(t)) \right\rangle_{\vec{x}_0(t)}$$

$$\frac{d}{dt} \prod_i \delta(y_i - x_i(t)) = \sum_{j=1}^N \left\{ \underbrace{\frac{\partial}{\partial x_j} \left[\prod_i \delta(y_i - x_i(t)) \right]}_{-\frac{\partial}{\partial y_j} \left[\prod_i \delta(y_i - x_i(t)) \right]} \cdot \dot{x}_j(t) + \frac{1}{2} 2\sigma \frac{\partial^2}{\partial x_j^2} \prod_i \delta(y_i - x_i(t)) \right\}$$

$$= - \sum_j \frac{\partial}{\partial y_j} \left\{ \underbrace{\prod_i \delta(y_i - x_i(t)) f_j(\vec{x})}_{\prod_i \delta(y_i - x_i(t)) f_j(\vec{y})} + \prod_i \delta(y_i - x_i(t)) \gamma_j(t) - \sigma \frac{\partial}{\partial y_j} \prod_i \delta(y_i - x_i(t)) \right\} \quad (1)$$

Note that $\langle \prod_i \delta(y_i - x_i(t)) \gamma_j(t) \rangle \stackrel{\text{Itô}}{=} \underbrace{\langle \prod_i \delta(y_i - x_i(t)) \rangle}_{P(\vec{y})} \underbrace{\langle \gamma_j(t) \rangle}_0$

From (1), taking an average over the noise realisations

$$\frac{d}{dt} \langle \prod_i \delta(y_i - x_i(t)) \rangle = - \sum_j \frac{\partial}{\partial y_j} \left\{ \langle \prod_i \delta(y_i - x_i(t)) \rangle f_j(\vec{y}) + 0 - \sigma \frac{\partial}{\partial y_j} \langle \prod_i \delta(y_i - x_i(t)) \rangle \right\}$$

$$\frac{d}{dt} P(\vec{y}) = - \sum_j \frac{\partial}{\partial y_j} \left[f_j(\vec{y}) P(\vec{y}) - \sigma \frac{\partial}{\partial y_j} P(\vec{y}) \right]$$

$$= - \vec{\nabla} \cdot [\vec{f}(\vec{q}) P(\vec{q}) - \sigma \vec{\nabla} P(\vec{q})]$$

Comment: if $\sigma_{ij} = \sigma \delta_{ij}$, then Itô formula can be written as

$$\begin{aligned} \frac{d}{dt} g(\vec{x}(t)) &= \sum_i \frac{\partial}{\partial x_i} (g(x)) \dot{x}_i + \frac{\partial^2}{\partial x_i^2} [\sigma g(\vec{x})] \\ &= \vec{\nabla} g \cdot \dot{\vec{x}} + \Delta [\sigma g] \end{aligned}$$

$$\frac{d}{dt} P(\vec{q}) = \frac{d}{dt} \langle \delta(\vec{q} - \vec{x}_\epsilon) \rangle_{\vec{x}_\epsilon} = \left\langle \underbrace{\vec{\nabla}_{\vec{x}} \cdot [\delta(\vec{q} - \vec{x}_\epsilon)] \dot{\vec{x}}_\epsilon}_{-\vec{\nabla}_{\vec{q}} \delta(\vec{q} - \vec{x}_\epsilon)} + \sigma \Delta \delta(\vec{q} - \vec{x}_\epsilon) \right\rangle$$

$$= \left\langle -\vec{\nabla}_{\vec{q}} \delta(\vec{q} - \vec{x}_\epsilon) \cdot \vec{f}(\vec{x}_\epsilon) + \sigma \Delta \delta(\vec{q} - \vec{x}_\epsilon) \right\rangle$$

$$= -\vec{\nabla}_{\vec{q}} \cdot \left\langle \delta(\vec{q} - \vec{x}_\epsilon) \vec{f}(\vec{x}_\epsilon) \right\rangle + \sigma \Delta \langle \delta(\vec{q} - \vec{x}_\epsilon) \rangle$$

$$= -\vec{\nabla}_{\vec{q}} \cdot \underbrace{\left\langle \delta(\vec{q} - \vec{x}_\epsilon) \vec{f}(\vec{q}) \right\rangle}_{\langle \delta(\vec{q} - \vec{x}_\epsilon) \rangle \cdot \vec{f}(\vec{q})} + \sigma \Delta \underbrace{\langle \delta(\vec{q} - \vec{x}_\epsilon) \rangle}_{P(\vec{q})}$$

$$\frac{d}{dt} P(\vec{q}) = -\vec{\nabla}_{\vec{q}} \cdot [\vec{f}(\vec{q}) P(\vec{q})] + \sigma \Delta P(\vec{q})$$

c) Application to ABPs

$$\dot{\vec{r}} = v_0 \vec{u}(\theta) - \mu \vec{\nabla} V(\vec{r}) + \sqrt{2D_\epsilon} \vec{\gamma} \quad ; \quad \dot{\theta} = \sqrt{2D_\theta} \gamma$$

$$\dot{x} = v_0 \cos \theta - \mu \partial_x V + \sqrt{2D_\epsilon} \gamma_x$$

$$\dot{y} = v_0 \sin \theta - \mu \partial_y V + \sqrt{2D_\epsilon} \gamma_y$$

where $\gamma, \gamma_x, \gamma_y$ are 3 independent centered GWN of unit variance.

Fokker-Planck formula:

$$\frac{d}{dt} P(x, y, \theta, t) = - \frac{\partial}{\partial x} \left[(v_0 \cos \theta - \mu \partial_x V) P \right] + D_\epsilon \frac{\partial^2}{\partial x^2} P$$

$$\begin{aligned}
& - \frac{\partial}{\partial y} \left[(v_0 \sin \theta - \mu \partial_y V) P \right] + D_t \frac{\partial^2}{\partial y^2} P \\
& + D_n \frac{\partial^2}{\partial \theta^2} P \\
& = - \vec{\nabla}_{\vec{r}} \cdot \left[(v_0 \vec{u}(\theta) - \mu \vec{\nabla}_n V) P \right] + D_t \Delta_n P + D_n \Delta_\theta P
\end{aligned}$$

II Temp process & Master equations

II.1) discrete - state space

N configurations $\{q\}$; Note $w(q \rightarrow q')$

prob to go from q to q' in $[t, t+dt]$ $\sim \frac{w(q \rightarrow q') dt}{dt \rightarrow 0}$ don't care

Master equation $\frac{d}{dt} P(q, t) = \dots \Rightarrow P(q, t+dt) - P(q, t) = dt [\dots] + o(dt)$

in $[t, t+dt]$ $P(q \rightarrow q' \rightarrow q'') = w(q \rightarrow q') dt w(q' \rightarrow q'') dt \sim dt^2$

$\Rightarrow$ never have to go consider two processes.

(1) $P(q, t+dt) = [\text{Prob to be in } q \text{ at } t \text{ and to stay there}]$

+ $[\text{Prob to be in any } q' \neq q \text{ and to go from } q' \text{ to } q]$ (2)

$$(2) = \sum_{q' \neq q} P(q', t) w(q' \rightarrow q) dt$$

$$(1) = P(q, t) \times (1 - \text{prob to go from } q \text{ to any } q' \neq q) = P(q, t) \left[1 - \sum_{q'} w(q \rightarrow q') dt \right]$$

$$P(q, t+dt) - P(q, t) = dt \sum_{q' \neq q} w(q' \rightarrow q) P(q') - w(q \rightarrow q') P(q)$$

$$\frac{d}{dt} P(q, t) = \sum_{q' \neq q} w(q' \rightarrow q) P(q') - w(q \rightarrow q') P(q)$$

II.2) Tumble dynamics, the limit of continuous space

$$\theta \in [0, 2\pi]$$

Path 1: prob to be in $[0, 0+do] = \underbrace{p(\theta)}_{\text{density of prob}} do$

$$p(\theta, t+dt) d\theta = \dots$$

Path 2: "Be wise, discretize"

$$Q_h = \frac{2\pi}{N} \cdot h \Rightarrow N \text{ values} \Rightarrow \frac{d}{dt} P(Q_h) = \sum_{m \neq h} P(Q_m) W(Q_m \rightarrow Q_h) - P(Q_h) W(Q_h \rightarrow Q_m)$$

uniform, isotropic tumbles $W(Q_m \rightarrow Q_h) = \alpha \cdot \frac{1}{N}$

$$\frac{d}{dt} P(Q_h) = \sum_m \frac{1}{N} [\alpha P(Q_m) - \alpha P(Q_h)]$$

$$P(Q_h) \rightarrow P(\theta) \frac{d\theta}{2\pi}$$

Now $\left(\frac{1}{2\pi} \sum_m \frac{2\pi}{N} \xrightarrow{N \rightarrow \infty} \frac{1}{2\pi} \int d\theta' \right)$

$$\frac{d}{dt} P(\theta) = \frac{\alpha}{2\pi} \int d\theta' [P(\theta') - P(\theta)] = -\alpha P(\theta) + \frac{\alpha}{2\pi} \int d\theta' P(\theta')$$

III A mixture of both world

. diffusive AND tumble

$$\textcircled{1} \text{ Langevin dynamics} \Rightarrow \partial_t P = \frac{\partial}{\partial x} \left[\frac{\partial}{\partial x} Q_1 - f(x) \right] P \equiv -H_{FP} P \quad (*)$$

$$H_{FP} = -\frac{\partial^2}{\partial x^2} Q_1 + \frac{\partial}{\partial x} f(x)$$

$$(*) \Leftrightarrow P(x, t+dt) = P(x, t) - dt H_{FP}^x P + o(dt^\alpha) \quad \alpha > 1$$

$$\dot{x} = [\text{Langevin eq}^0]; \quad \theta \xrightarrow{\alpha} \theta$$

$$P(x, Q_h, t+dt) = \int dx' P(x', Q_h, t) P[(x', Q_h, t) \rightarrow (x, Q_h, t+dt)] \quad (1)$$

$$+ \sum_{m \neq h} \int dx' P(x', Q_m, t) P[(x', Q_m, t) \rightarrow (x, Q_h, t+dt)] \quad (2)$$

in dt , $p(\text{two tumbles}) \propto dt^2 \Rightarrow$ disregarded $\Rightarrow$ in (1) $\Rightarrow$ no tumble

$$P[(x', Q_h, t) \rightarrow (x, Q_h, t+dt)] = (\text{prob no tumbles}) \times (\text{prob that Langevin takes the system from } x' \rightarrow x)$$

$$= (1 - \alpha dt) \underbrace{P_{FP}(x, Q_h, t+dt | x', Q_h, t)}_{P_{FP}(x, Q_h, t | x', Q_h, t) + dt \frac{\partial}{\partial t} P_{FP}(x, Q_h, t | x', Q_h, t)}$$

$$= (1 - \alpha dt) \left[\delta(x - x') - dt H_{FP}^{x, \theta_n} \delta(x - x') \right] + o(dt)$$

$$= (1 - \alpha dt - dt H_{FP}^{x, \theta_n}) \delta(x - x') + o(dt)$$

$$(1) \int dx' P(x', \theta_n, t) \left[(1 - \alpha dt - dt H_{FP}^{x, \theta_n}) \delta(x - x') \right] + o(dt)$$

$$= (1 - \alpha dt - dt H_{FP}^{x, \theta_n}) P(x, \theta_n, t)$$

$$(2) \quad x', \theta_n \rightarrow g, \underbrace{\theta_n \rightarrow g, \theta_n}_{\text{noise} \sim \alpha dt} \rightarrow x, \theta_n$$

$$\int (1 - \alpha dt - dt H_{FP}^{g, \theta_n}) \delta(g - x') \frac{\alpha}{N} dt \quad (1 - \alpha dt - dt H_{FP}^{x, \theta_n}) \delta(y - x) dg$$

$$\sim 1 \times \frac{\alpha}{N} dt \times 1 \times \delta(x' - x)$$

$$(2): \sum_{n \neq h} \int dx' P(x', \theta_n) \frac{\alpha}{N} dt \delta(x - x') \sim \frac{dt}{2\bar{c}} \alpha \int d\theta' P(x, \theta')$$

All in all:

$$P(x, \theta, t+dt) = (1 - \alpha dt - dt H_{FP}^{x, \theta}) \underbrace{P(x, \theta, t)} + \frac{\alpha dt}{2\pi} \int d\theta' P(x, \theta')$$

$$\frac{d}{dt} P(x, \theta, t) = -H_{FP}^{x, \theta} P(x, \theta, t) - \alpha P(x, \theta, t) + \frac{\alpha}{2\pi} \int d\theta' P(x, \theta')$$

with the proper choices for x, y

$$\frac{d}{dt} P(x, \theta, t) = -\frac{\partial}{\partial x} [v_0 \cos \theta P] - \frac{\partial}{\partial y} [v_0 \sin \theta P] - \alpha P + \frac{\alpha}{2\pi} \int d\theta' P(x, \theta')$$