

Equation of evolution for the probability density $P(x,t)$:

1.10
Lecture on
Complex Systems

"deterministic force" "stochastic force"

$$\frac{dx(t)}{dt} = F(x(t), t) + \eta(t)$$

$F(x,t)$ is assumed to be regular enough

Remark: without forces ($F=0$) this was denoted $x(t)$ previously; now that $x(t)$ evolves with $F(x,t)$, we call this $B(t)$

Where we denote by definition

$$\int_t^{t+\delta t} dt F(x(t), t) \approx \delta t F(x(t), t)$$

$$\eta_\epsilon = \int_t^{t+\delta t} dt \eta(t) = B(t+\delta t) - B(t)$$

$B(t) = \int_0^t dt \eta(t)$: Brownian motion

we have seen that $\langle \eta_\epsilon^i \eta_\epsilon^j \rangle = \langle (B(t+\delta t) - B(t))^2 \rangle = C(\delta t) = \underline{\underline{2D \delta t}}$ (*)

↳ A) when expanding, η_ϵ is of order $\sqrt{\delta t}$.

$$x_{t+\delta t} - x_t = \delta t F(x_t, t) + \eta_\epsilon$$

Remark: one could take for η_ϵ : $\begin{cases} +\delta x & \text{prob } \delta \\ -\delta x & \delta \\ 0 & 1-2\delta \end{cases}$
As for the free case ($F=0$) one could take the limit: $\frac{\delta x \rightarrow 0}{\delta t \rightarrow 0} \frac{\delta x^2}{\delta t}$ fixed

1st try: directly expand $P(x, t+\delta t)$

$$P(x, t+\delta t) = \int dx_2 P(x_2, t) P(x, t+\delta t | x_2, t)$$

$$P(x, t+\delta t) - P(x, t) = \int dx_2 P(x_2, t) [P(x, t+\delta t | x_2, t) - \delta(x-x_2)]$$

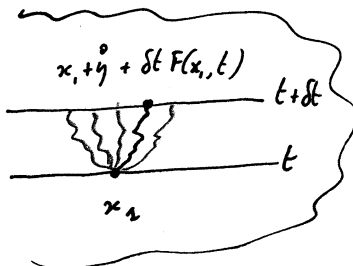
a small δt expansion would be difficult (it is difficult to expand around a Dirac delta) \rightarrow in such situations, integrate over a "test function":

2nd try: consider the time-evolution of the average of an observable $\varphi(x)$:

Consider an observable $\varphi(x)$, vanishing at $|x| \rightarrow \infty$ (a "test function" for mathematicians)

Average value at time $t+\delta t$

$$\langle \varphi(x) \rangle_{t+\delta t} = \int dx_2 \int d\eta \underbrace{P(\eta)}_{\text{average over the noise } \eta} \underbrace{P(x_2, t)}_{\text{distribution at time } t} \varphi(x_2 + \eta + \delta t F(x_2, t))$$



$\langle \eta \rangle = 0, \langle \eta^2 \rangle = 2D\delta t$
see (*)

The expansion at small δt is now easier

$$\approx \varphi(x_2) + \eta \varphi'(x_2) + \delta t F(x_2, t) \varphi'(x_2) + \frac{1}{2} \eta^2 \varphi''(x_2) + \mathcal{O}(\delta t^{3/2})$$

of order δt after averaging, see (*)

$$\langle \varphi(x) \rangle_{t+\delta t} = \int dx_2 P(x_2, t) \{ \varphi(x_2) + \delta t F(x_2, t) \varphi'(x_2) + D \delta t \varphi''(x_2) \} + \mathcal{O}(\delta t^{3/2})$$

$$\langle \varphi(x) \rangle_{t+\delta t} = \langle \varphi(x) \rangle_t + \delta t \langle F(x, t) \varphi'(x) + D \varphi''(x) \rangle_t + \mathcal{O}(\delta t^{3/2})$$

One can now take the limit $\delta t \rightarrow 0$ by isolating $\frac{\langle \varphi(x) \rangle_{t+\delta t} - \langle \varphi(x) \rangle_t}{\delta t}$ (1.11)

which yields $\partial_t \langle \varphi(x) \rangle_t = \langle F(x,t) \varphi'(x) + D \varphi''(x) \rangle_t$ (*)

Lecture on
Complex System

Remark: from the Langevin equation $\partial_t x = F(x,t) + \eta$
 one is tempted to write $\partial_t \varphi(x_t) = \partial_t x_t \varphi'(x_t) = (F(x_t,t) + \eta_t) \varphi'(x_t)$
 and to average as $\partial_t \langle \varphi(x) \rangle_t = \underbrace{\langle F(x,t) \varphi'(x) \rangle_t}_{\text{"deterministic term"}} + \underbrace{\langle \eta \varphi'(x) \rangle_t}_{\text{"stochastic term"}}$

the "deterministic term" is the first term of (*) and is easy to see.
 the "stochastic term" $\langle \eta \varphi'(x) \rangle_t$ is not easy to determine and is equal to $D \langle \varphi'' \rangle_t$ (see (*)).
 \rightarrow this is due to the fact that $\eta(t)$ is a very irregular function "white noise".

Remark: (*) is also true without averaging.

This is the Ito formula $\partial_t [\varphi(x,t)] = \partial_t \varphi(x,t) + F(x,t) \partial_x \varphi(x,t) + D \partial_x^2 \varphi(x,t)$

Remark: time-derivative of averages: point of view of trajectories and distributions

\rightarrow From the point of view of trajectories: difference of values of the function φ on the same trajectory x_t . This difference is rather "singular" since x_t is "rough".

$$\langle \varphi(x_{t+\delta t}) - \varphi(x_t) \rangle = \underbrace{\int dx P(x, t+\delta t) \varphi(x)}_{\text{same as}} - \underbrace{\int dx P(x, t) \varphi(x)}_{\text{at time } t}$$

Here x is a dummy variable and represents all possible values of x at time $t+\delta t$, of probability $P(x, t+\delta t)$

$$\langle \varphi(x_{t+\delta t}) - \varphi(x_t) \rangle = \int dx [P(x, t+\delta t) - P(x, t)] \varphi(x)$$

$\downarrow \delta t \rightarrow 0$

$$\partial_t \langle \varphi(x) \rangle_t = \int dx \partial_t P(x, t) \varphi(x)$$

Viewpoint of trajectories
 (see computation p 10-11) or
 viewpoint of distributions, depending
 on what is useful.

Viewpoint of probability distributions
 [Again, more simple to tackle as regards the
 regularity of functions & their expansion, since
 the prob. distribution $P(x,t)$ is "smooth".]

x Equation on the probability distribution:

Lecture on (1.12)
Complex Systems

$$\forall \varphi \quad \partial_t \langle \varphi(x) \rangle_t = \langle F(x,t) \varphi'(x) + D \varphi''(x) \rangle_t$$

$$\int dx \partial_t P(x,t) \varphi(x) = \int dx \left\{ P(x,t) F(x,t) \varphi'(x) + D P(x,t) \varphi''(x) \right\}$$

$$\int dx \partial_t P(x,t) \varphi(x) = \int dx \left\{ -\partial_x [P(x,t) F(x,t)] + D \partial_x^2 P(x,t) \right\} \varphi(x)$$

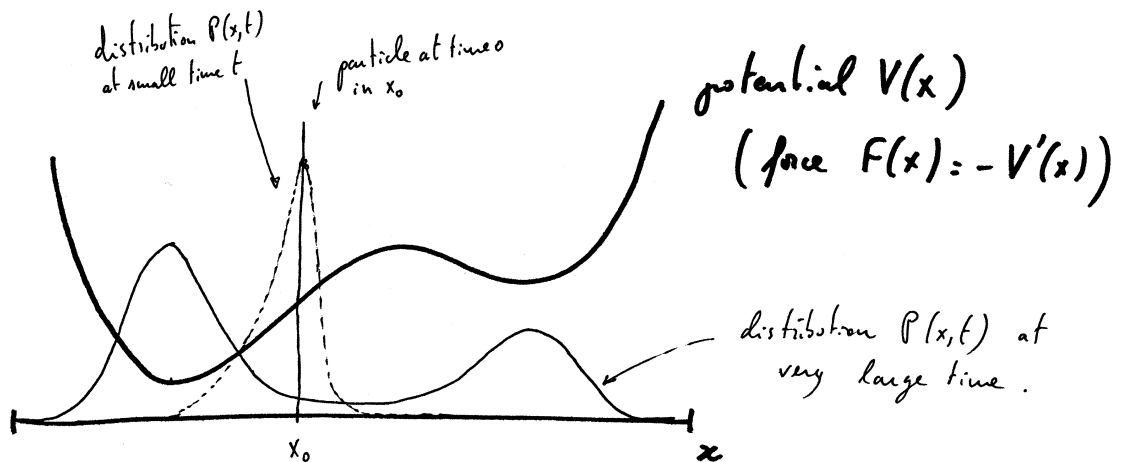
integrations by part, using $|\varphi| \xrightarrow{x \rightarrow \pm \infty} 0$

This equation is valid for all functions $\varphi(x)$ decreasing fast enough at infinity

Hence: $\partial_t P(x,t) = -\partial_x [F(x,t) P(x,t)] + D \partial_x^2 P(x,t)$

This is the Fokker-Planck equation for a 1D particle in a force $F(x,t)$.

x Example:



• Steady state: limit distribution $P_{st}(x) = \lim_{t \rightarrow \infty} P(x,t)$ if it exists

It verifies $0 = \partial_x [-F(x) P_{st}(x) + D \partial_x P_{st}(x)]$ for a time-independent force $F(x,t)$

x Remark: for the free particle ($F=0$) one has $\partial_t P = D \partial_x^2 P$ (diffusion equation)

⚠ on the real line: $P(x,t) = \frac{1}{\sqrt{4Dt}} e^{-\frac{x^2}{4Dt}} \xrightarrow{t \rightarrow \infty} 0$ there is no steady state
[long-time properties are caught here through scaling: $P(x,t) \sim \frac{1}{\sqrt{t}} \hat{P}(x/\sqrt{t})$]

• on an interval $[a,b]$: the steady state verifies $\begin{cases} \partial_x^2 P_{st}(x) = 0 \\ \partial_x P_{st}|_a = \partial_x P_{st}|_b = 0 \end{cases}$: no current on boundaries
(for an isolated system) $P_{st}(x) = \frac{1}{b-a}$ uniform distribution

• Equilibrium steady state:

in a potential $V(x)$
at temperature T

(1.13)
Lecture on
Complex Systems

* remark: if $P_{eq}(x) \propto e^{-\frac{1}{T}V(x)}$ (Boltzmann distribution)

one has $T \partial_x P_{eq}(x) = -V'(x) P_{eq}(x)$

hence: The Boltzmann distribution $P_{eq}(x) \propto e^{-\frac{1}{T}V(x)}$ is a steady-state solution of the Fokker-Planck equation $\partial_x [+V'(x) P_{eq}(x) + T \partial_x P_{eq}(x)] = 0$

This provides a link between thermodynamics (Boltzmann canonical ensemble)

where $D = T$ (hence the interpretation of $\eta(t)$ as a thermal noise)
stochastic dynamics (Langevin dynamics)

* interpretation in terms of a probability current

The FP equation writes $\partial_t P + \partial_x J(P) = 0$ with $J(P) = FP - D \partial_x P$

A steady-state distribution verifies $\partial_x J(P) = 0 \Rightarrow J(P) \equiv j$ (constant)

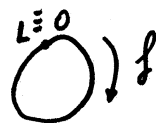
An equilibrium st. st. distribution — $J(P) = 0$ [specific of $j=0$]
When $j \neq 0$, we speak of a non-equilibrium steady state (NESS). $j \neq 0$

* mean velocity in the system: the example of $V = -fx$ i.e. $F(x) = f$

① $\langle \partial_t x \rangle = \langle f + \eta \rangle \Rightarrow \langle \partial_t x \rangle \equiv v = f$

② for distribution and for the NESS:

one considers periodic boundary conditions



$\langle \partial_t x \rangle = \int dx x \partial_t P(x,t) = - \int dx x \partial_x J(P)$ from Fokker-Planck

$\stackrel{\text{iff}}{=} \int dx J(P)$ but $J(P) = j$ is a constant j

Hence: the velocity is $v \equiv \langle \partial_t x \rangle = j = f$

The system is in a true NESS ($j \neq 0$) iff $f \neq 0$.

It is not obvious to determine $P_{st}(x)$ for $f \neq 0$. See exercises.

↳ indeed a try as $P_{st}(x) \propto e^{-\frac{1}{T}fx}$ does not respect boundary conditions

• "Equilibrium dynamics" and reversibility: independent of time
↑
(Lecture on 1.14
Complex Systems)

• if there exists a potential $V(x)$ such that $F(x) = -V'(x)$ and that $P_{eq}(x) = \frac{1}{Z} e^{-\frac{1}{T} V(x)}$, $Z = \int dx e^{-\frac{1}{T} V(x)}$ is the stationary distribution [$J(P) = 0$]

then we say that the dynamics is an "equilibrium dynamics" and that $P_{eq}(x)$ is an "equilibrium steady state" (or "reversible measure").

• otherwise, the dynamics is a "non-equilibrium dynamics". [$J(P_{st}) \neq 0$]

→ There is a physical interpretation in terms of reversibility. Let us fix the final time t and consider all trajectories $x(\tau)$ for $0 \leq \tau \leq t$.

• probability density of a trajectory $(x(\tau))_{0 \leq \tau \leq t}$:

simplest way of obtaining it:

$$\text{Prob}[x(\tau)] = \int \mathcal{D}\eta \delta(\partial_t x - F(x) - \eta) e^{-\frac{1}{4T} \int_0^t d\tau \eta(\tau)^2}$$

This is justified e.g. from the expansion of the average of an observable depending on $[x(\tau)]_{0 \leq \tau \leq t}$:
 $\langle O[x(\tau)] \rangle = \int \mathcal{D}x O[x(\tau)] \text{Prob}[x]$

$$\text{Prob}[x(\tau)] = \exp\left\{-\frac{1}{4T} \int_0^t d\tau (\partial_t x - F(x))^2\right\}$$

• time-reversed trajectory: one defines $x^R(\tau) = x(t-\tau)$ ($0 \leq \tau \leq t$) which verifies $x^R(t) = x(0)$, $x^R(0) = x(t)$, $\partial_\tau x^R = -\partial_t x$

• joint probability of history & initial condition in the equilibrium dynamics:

$$P_{eq}(x(0)) \text{Prob}[x(\tau)] = e^{-\frac{1}{T} V(x(0))} \exp\left\{-\frac{1}{4T} \int_0^t d\tau (\partial_t x + V'(x))^2\right\}$$

$x \mapsto x^R$
 $\tau \mapsto t-\tau$

$$= e^{-\frac{1}{T} V(x^R(t))} \int_0^t d\tau (-\partial_\tau x^R + V'(x^R))^2 = \int_0^t d\tau \left[(\partial_\tau x^R + V'(x^R))^2 - 4 \partial_\tau x^R V'(x^R) \right]$$

no as to recover a $\partial_\tau(V(x^R))$

$$= -4 \left(\frac{V(x^R(t)) - V(x^R(0))}{t} \right) + \int_0^t d\tau (\partial_\tau x^R + V'(x^R))^2$$

these two terms cancel

$$= e^{-\frac{1}{T} V(x^R(0))} \exp\left\{-\frac{1}{4T} \int_0^t d\tau (\partial_\tau x^R + V'(x^R))^2\right\}$$

this one remains

There is an equivalence btw

- $J(P) = 0$ (no prob. current)
- $P(x) \propto e^{-\frac{1}{T} V(x)}$
- reversibility in that sense

Hence:

$$P_{eq}(x(0)) \text{Prob}[x(\tau)] = P_{eq}(x^R(0)) \text{Prob}[x^R(\tau)] \text{, i.e.}$$

there is reversibility (trajectories and their time-reversed have the same probability, including initial eq. distrib.) when the dynamics is an "equilibrium dynamics".

otherwise not.

• Reversible dynamics, Doob transform and Schrödinger equation

Lecture on (4.15)
Complex Systems
D=T from now on

Consider the "equilibrium dynamics" (also termed "reversible dyn.")

$$\partial_t x = -V'(x) + \eta \quad (\Leftrightarrow) \quad \partial_t P(x,t) = +\partial_x [V'(x) P(x,t)] + T \partial_x^2 P(x,t)$$

This equation for $P(x,t)$ is linear, first-order in time, and looks like Schrödinger's equation $-i\partial_t \Psi(x,t) = [\partial_x^2 - V_{\text{quant}}(x)] \Psi(x,t)$ on the wavefunction Ψ without however taking the same form.

Let's introduce $P_{\text{sym}}(x,t) = e^{\frac{1}{2T}V(x)} P(x,t) = P_{\text{eq}}^{-1/2}(x) P(x,t)$ (This is the Doob transform)

One easily checks that the Fokker-Planck equation is equivalent to:

$$\partial_t P_{\text{sym}}(x,t) = [T \partial_x^2 - V_{\text{eff}}(x)] P_{\text{sym}}(x,t)$$

$$V_{\text{eff}}(x,t) = \frac{1}{4T} (V'(x))^2 - \frac{1}{2} V''(x) \quad \text{effective potential}$$

Hence: the FP equation for P_{sym} is equivalent to an [imaginary time] Schrödinger equation for P_{sym} in a potential $V_{\text{eff}}(x)$.

Again, this only works for "equilibrium dynamics".

Such a correspondence can be established in a much more general context, see later.

• Remark: correct form of the probability density of trajectories

The (quantum-like) Feynman path-integral would yield for the probability of histories:

$$P_{\text{sym}}(x,t) = \frac{1}{Z} e^{-\frac{1}{2T}V(x(t))} e^{-\frac{1}{4T} \int_0^t \left\{ \underbrace{(\partial_\tau x)^2 + (V'(x))^2}_{\text{up to time-boundary terms, corresponds to } (\partial_\tau x + V'(x))^2} - \underbrace{2TV''(x)}_{\text{this term is absent from the expression p. 14}} \right\} d\tau}$$

→ This comes from a different choice of time-discretization of the Langevin equation (Itô vs Stratonovitch) → $e^{V''}$ can be seen as a Jacobian.

See AWC. Lau & T.C. Lubensky PRE 76 011123 (2007) for details

• Generalization to several interacting particles:
 $x_i(t)$ each of them in a noise $\eta_i(t)$ $1 \leq i \leq N$
 global interactions $F_i(\{x_j\}, t)$

$$\partial_t x_i = F_i(\{x_j\}, t) + \eta_i(t)$$

$$\langle \eta_i(t) \eta_j(t') \rangle = 2 \Delta_{ij} \delta(t-t')$$

with $\{\eta_j(t)\}$ white noises
 Δ_{ij} invertible symmetric matrix
 η_j 's are uncorrelated if $\Delta_{ij} = D \delta_{ij}$

Gaussian distribution of the noise:

$$\text{Prob}[\{\eta_i(t)\}] \propto \exp\left(-\frac{1}{2} \int_0^t d\tau \sum_{ij} \eta_i(\tau) (\Delta^{-1})_{ij} \eta_j(\tau)\right)$$

if $\Delta_{ij} = D \delta_{ij}$

$$\propto \exp\left(-\frac{1}{4D} \int_0^t d\tau \sum_i \eta_i(\tau)^2\right)$$

* Fokker-Planck equation: notation $\partial_i = \frac{\partial}{\partial x_i}$

$$\partial_t P(\{x_j\}, t) = - \sum_i \partial_i (F_i(x, t) P(x, t)) + \sum_{ij} \Delta_{ij} \partial_i \partial_j P(x, t)$$

deterministic contribution (uncorrelated) diffusive contribution
 in matrix notation: matrix Δ_{ij}

$$\partial_t P = - \vec{\nabla} \cdot (\vec{F}(x, t) P(x, t)) + \vec{\nabla} \cdot (\Delta \vec{\nabla} P(x, t))$$

if the noise is uncorrelated ($\Delta_{ij} = D \delta_{ij}$)

$$\partial_t P = - \vec{\nabla} \cdot (\vec{F}(x, t) P(x, t)) + D \vec{\nabla}^2 P(x, t)$$

* Itô formula:

$$\partial_t \varphi(x, t) = \sum_i F_i(x, t) \partial_i \varphi(x, t) + \sum_{ij} \Delta_{ij} \partial_i \partial_j \varphi(x, t) + \sum_i \eta_i(x, t) \partial_i \varphi(x, t)$$

* Equilibrium:

When $\forall i, F_i(x, t) = -\partial_i V(x)$

forces acting on
particle i

common potential to all particles
↑
describing interactions between particles
and also with the environment

i.e. when "forces are conservative"

i.e. when "forces derive from a potential"

then the Boltzmann-Gibbs measure is an equilibrium steady state

① For non-correlated noises $\langle \eta_i \eta_j \rangle = 2D \delta_{ij} \delta(t-t')$; denoting $D=T$

$$P_{eq}(x) = \frac{1}{Z} e^{-\frac{1}{T} V(x)}$$

with $Z = \int dx e^{-\frac{1}{T} V(x)}$

verifier $\partial_i P_{eq}(x) = -\frac{1}{T} \partial_i V(x) P_{eq}(x)$ hence $\partial_i^2 V P_{eq} + T P_{eq} = 0$

but the FP equation writes $\partial_t P = \sum_i \partial_i \left[\partial_i V P + T P \right]$

hence $P_{eq}(x)$ is an equilibrium distribution all components of the current are 0

② For correlated noises $\langle \eta_i(t) \eta_j(t') \rangle = 2 \Delta_{ij} \delta(t-t')$:

$$P_{eq}(x) = \frac{1}{Z} \exp(-W(x)) \Rightarrow \partial_i P_{eq} = -\partial_i W P_{eq}$$

searching for a solution of $\partial_i^2 V P_{eq} + \sum_j \Delta_{ij} \partial_j^2 P_{eq} = 0$ is equivalent to

$$\forall i, \partial_i^2 V = \sum_j \Delta_{ij} \partial_j^2 W \quad \text{ie } \forall i, \partial_j^2 W = \sum_k (\Delta^{-1})_{jk} \partial_k^2 V$$

A necessary and sufficient condition for this equation to have a solution (using Poincaré's lemma) is: $\forall i, j, \sum_k (\Delta^{-1})_{ki} \partial_k^2 V = \sum_k (\Delta^{-1})_{kj} \partial_k^2 V$

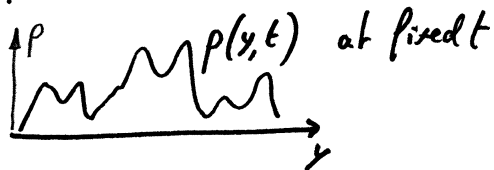
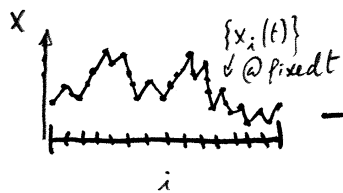
A possible solution is for quadratic potentials: $V = \sum_{ij} \Delta_{ij} x_i x_j$

|| But otherwise there is no generic characterisation of equilibrium with correlated noise.

• Generalisation to several coupled noises: continuum

$$x_i(t) \longrightarrow p(y, t) \quad \begin{matrix} x \equiv p \\ y \equiv i \end{matrix}$$

$$\eta_i(t) \longrightarrow \eta(y, t)$$



one will call $p(y, t)$ a "profile", but the description holds in generality

$$\langle \eta_i(t) \eta_j(t') \rangle = 2D_{ij} \delta(t-t') \longrightarrow \langle \eta(y, t) \eta(y', t') \rangle = \underbrace{R(y'-y)}_{\text{describes correlations in direction } y} \delta(t-t')$$

($R(y) = 2D \delta(y'-y)$ for uncorrelated noise)

$$P(x, t) \longrightarrow P[p, t] \quad (\text{functional}) \text{ probability density of the profile } p(y), \text{ at time } t$$

* Langevin equation: from $\partial_t x_i = F_i(x, t) + \eta_i$ (see p. 1.16) one goes to

$$\partial_t p(y, t) = \mathcal{F}[p; y, t] + \eta(y, t)$$

* Fokker-Planck equation: from $\partial_t P(x, t) = - \sum_i \partial_i (F_i(x, t) P(x, t)) + \sum_{ij} \Delta_{ij} \partial_i \partial_j P(x, t)$ makes

$$\partial_t P[p, t] = - \int dy \frac{\delta}{\delta p(y)} (\mathcal{F}[p; y, t] P[p, y]) + \frac{1}{2} \int dy dy' R(y'-y) \frac{\delta^2 P[p, t]}{\delta p(y) \delta p(y')}$$

* Itô formula:

$$\partial_t (\varphi(p, t))^{p(y,t)} = \partial_t \varphi(p, t) + \int dy \mathcal{F}[p; y, t] \frac{\delta \varphi(p, t)}{\delta p(y)} + \frac{1}{2} \int dy dy' R(y'-y) \frac{\delta^2 \varphi(p, t)}{\delta p(y) \delta p(y')}$$

* Example of equilibrium (i.e. reversible) dynamics: $\mathcal{F}[p; y, t] = \frac{\delta \mathcal{H}[p]}{\delta p(y)}$

Then for an uncorrelated noise $R(y) = 2T \delta(y)$

$$P_{eq}[p] \propto \exp\left(-\frac{1}{T} \mathcal{H}[p]\right) \quad \text{is a solution of the FP equation, with 0 current:}$$

$$-\mathcal{F}[p; y, t] P_{eq}[p] + \frac{1}{2T} \frac{\delta P_{eq}[p]}{\delta p(y)} = 0$$

• Generalization to position-dependent noise:

But writing:

$$\boxed{\partial_t x = F(x, t) + \sqrt{2D(x, t)} \eta(t)} \quad \text{with } \langle \eta(t) \eta(t') \rangle = \delta(t-t')$$

Fokker-Planck: $\text{Prob}[x|q] = \exp\left(-\frac{1}{2} \int_0^t dt \frac{(\partial_t x - F(x, t))^2}{2D(x, t)}\right)$

$$\boxed{\partial_t P(x, t) = -\partial_x (F(x, t) P(x, t)) + \partial_x^2 (D(x, t) P(x, t))}$$

• With several coupled noises:

$$\boxed{\partial_t x_i = F_i(x, t) + \sqrt{2D(x, t)} \eta_i(t)} \quad \langle \eta_i(t) \eta_j(t') \rangle = \Delta_{ij} \delta(t-t')$$

$$\boxed{\partial_t P(x, t) = -\sum_i \partial_i (F_i(x, t) P(x, t)) + \sum_{ij} \partial_i \partial_j (\Delta_{ij} D(x, t) P(x, t))}$$

• Generalization to fields: $x_i(t) \mapsto \rho(y, t)$

$$\partial_t \rho(y, t) = F(\rho(y, t), t) + \sqrt{2D(\rho(y, t))} \eta(y, t) \quad \langle \eta(y, t) \eta(y', t') \rangle = R(y-y') \delta(t-t')$$

$$\partial_t P[\rho(y), t] = -\int dy \frac{\delta}{\delta \rho(y)} \left(F(\rho(y), t) P[\rho(y), t] \right) + \int dy dy' \frac{\delta^2}{\delta \rho(y) \delta \rho(y')} \left[R(y-y') P[\rho, t] \right]$$

Remark: Backward Fokker-Planck equation:

We have seen that $P(x, t | x_0, t_0) = P(x, t)$ verifies the forward FP equation

$$\partial_t P(x, t) = -\partial_x (F(x, t) P(x, t)) + D \partial_x^2 P(x, t)$$

We might be interested in the derivatives $\partial_{x_0}, \partial_{t_0}$ w.r.t. the initial conditions (e.g. to tackle 1st passage problems).

As on page 110, we integrate the Langevin equation $\dot{x} = F(x, t) + \eta$ $\delta t > 0$ between $t_0 - \delta t$ and t_0 to write $x(t_0) = x(t_0 - \delta t) + \delta t F(x_{-1}, t - \delta t) + \overset{\circ}{\eta}$ with $\overset{\circ}{\eta}$ Gaussian, $\langle \overset{\circ}{\eta} \rangle = 0$ $\langle \overset{\circ}{\eta}^2 \rangle = 2D \delta t$

$$x_{-1} = x_0 - \delta t F(x_0, t) - \overset{\circ}{\eta}$$

Then, for any function $f(x)$ one has:

$$\int dx f(x) P(x, t | x_0, t_0) = \int dx f(x) \int d\overset{\circ}{\eta} P(\overset{\circ}{\eta}) P(x, t | x_0 - \overset{\circ}{\eta} - \delta t F(x_0, t), t_0 - \delta t)$$

or $F(x_{-1}, t - \delta t)$ but this is the same at the order δt we are interested in

we perform the average $\int d\overset{\circ}{\eta} P(\overset{\circ}{\eta})$ through $\langle \overset{\circ}{\eta} \rangle = 0, \langle \overset{\circ}{\eta}^2 \rangle = 2D \delta t$

$$\approx P(x, t | x_0, t_0 - \delta t) - \overset{\circ}{\eta} \partial_{x_0} P(x, t | x_0, t_0) - \delta t \partial_{x_0}^2 P(x, t | x_0, t_0) - \frac{1}{2} \overset{\circ}{\eta}^2 \partial_{x_0}^2 P(x, t | x_0, t_0) + O(\delta t^{3/2})$$

one has to expand up to order δt

Collecting:

$$\int dx f(x, t) \frac{P(x, t | x_0, t_0) - P(x, t | x_0, t_0 - \delta t)}{\delta t} = \int dx f(x, t) \left[-F(x_0, t) \partial_{x_0} P(x, t | x_0, t_0) - D \partial_{x_0}^2 P(x, t | x_0, t_0) \right]$$

And finally:

$$-\partial_{x_0} P(x, t | x_0, t_0) = F(x_0, t) \partial_{x_0} P(x, t | x_0, t_0) + D \partial_{x_0}^2 P(x, t | x_0, t_0)$$

Note the change of sign & the position of ∂_{x_0}

This is the backward Fokker-Planck equation, which describes the variations of the probability density with respect to its initial condition. $P(x, t | x_0, t_0) = e^{\int_{t_0}^t \dots} P_0(x_0)$ indep. of time

For homogeneous processes where $P(x, t | x_0, t_0)$ is a function of $t - t_0$ only, e.g. if $F(x, t) = F(x)$ one has $\partial_{t_0} = -\partial_t$ and hence

$$\partial_t P(x, t | x_0, t_0) = F(x_0) \partial_{x_0} P(x, t | x_0, t_0) + D \partial_{x_0}^2 P(x, t | x_0, t_0)$$