

III. General form of Langevin equations

so far we had equations of the form

$$\dot{\underline{v}} = -\gamma \underline{v} + m^{-1} \underline{f}(t)$$

Generalization:

Let $\{\underline{x}(t)\} = x_1(t), x_2(t), \dots, x_M(t)$
be a set of dynamical variables

$$\textcircled{*} \left(\dot{x}_i(t) = h_i(\{\underline{x}(t)\}, t) + \sum_{j=1}^M D_{ij}(\{\underline{x}(t)\}, t) f_j(t) \right)$$

where $\langle f_i \rangle = 0$

$$\langle f_i(t) f_j(t') \rangle = \Pi_{ij} \delta(t-t')$$

$D_{ij} = \text{const}$ "additive noise"

D_{ij} depends on \underline{x} : "multiplicative noise"

sometimes one needs the integral form of the equation (4)

$$\begin{aligned} \Rightarrow x_i(t+\tilde{\tau}) - x_i(t) &= \int_t^{t+\tilde{\tau}} dt' h_i(\{x_i(t')\}, t') \\ &+ \int_t^{t+\tilde{\tau}} \sum_{j=1}^M D_{ij}(\{x_i(t')\}, t') f_j(t') \end{aligned} \quad (**)$$

notice the (possible) problem in the second term:

$f_j(t')$ is highly irregular (random variable with zero correlation time)
 this irregularity transfers to $x(t)$

\Rightarrow which value of $\frac{f_j(t')}{x(t')}$ shall we use to evaluate the integral?

note: This problem does not occur for the first term since h_i is assumed to be a deterministic, regular function

\Rightarrow we can just approximate

$$\int_t^{t+\tilde{\tau}} dt' h_i(\{x_i(t')\}, t') \rightarrow h_i(\{x_i(t)\}, t) \cdot \tilde{\tau}$$

for small $\tilde{\tau}$

\Rightarrow Problem of Stochastic Integrals!

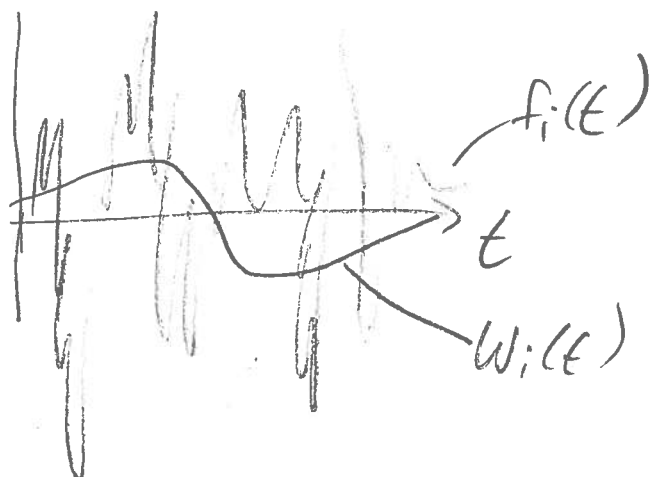
strategy:

introduce the "Wiener increment"

$$W_i(\tilde{t}) = \int_t^{\tilde{t}} dt' f_i(t')$$

$$dW_i(t) = f_i(t) dt$$

note : W_i behaves somewhat smoother than f_i



but : $\dot{W}_i = f_i(t)$ does not exist, strictly speaking.

insert into (**):

$$x_i(t+\tilde{t}) - x_i(t) = \int_t^{t+\tilde{t}} dt' h_i(\{x(t'), t'\}) + \int_t^{t+\tilde{t}} \sum_{j=1}^M D_{ij}(\{x(t'), t'\}) dW_j(t')$$

"Riemann - Stieltjes integral")

note: there are two ways to evaluate the stochastic integral in the second term!

- 1) Ho
- 2) Stratonovich

Consider integrals of the type $A = \int_t^{t+\hat{\tau}} D(x(t'), t') dW(t')$
(one-dimensional)

Discretization: Divide the time interval $\hat{\tau}$ into N sub-intervals

$$A^{Ho} = \lim_{N \rightarrow \infty} \sum_{m=0}^N D(x(t_m), t_m) (W(t_{m+1}) - W(t_m))$$

\Rightarrow evaluation of D
at the left boundary
of each subinterval

$$A^{Stratonovich} = \lim_{N \rightarrow \infty} \sum_{m=0}^N \left[\frac{1}{2} (D(x(t_{m+1}), t_{m+1}) - D(x(t_m), t_m)) \cdot (W(t_{m+1}) - W(t_m)) \right]$$

\Rightarrow evaluation via an average of D !

Note: For $D = \text{const}$ we have $A^{Ho} = A^{Stratonovich}$!!

Where is this issue important ??

E.g., for the relation between Langevin-like equations and the Fokker-Planck equation!

Consider the so-called Kramers-Moyal coefficients

$$K_{i_1 i_2 \dots i_n}^{(n)}(\underline{x}(t), t) = \frac{1}{n!} \lim_{\tau \rightarrow 0} \frac{1}{\tau} \langle (x_{i_1}(t+\tau) - x_{i_1}(t)) \cdot (x_{i_2}(t+\tau) - x_{i_2}(t)) \cdot \dots \cdot (x_{i_n}(t+\tau) - x_{i_n}(t)) \rangle$$

fixed $\underline{x}(t)$

Strategy to calculate the $K^{(n)}$'s =

$$\text{we } x_i(t+\tau) - x_i(t) = \int_t^{t+\tau} dt' \left[h_i(\underline{x}(t'), t') + \sum_j D_{ij}(\underline{x}(t'), t') f_j(t') \right]$$

$(i=1, \dots, M)$

and expand the functions h_i and D_{ij} around the (fixed) vector $\underline{x}(H)$

The averages $\langle \dots \rangle$ can then be calculated as time averages

result:

$$k_i^{(A)} = \lim_{\tilde{\tau} \rightarrow 0} \frac{1}{\tilde{\tau}} \langle x_i(t+\tilde{\tau}) - x_i(t) \rangle$$

$$\left. \begin{aligned} k_i^{(A)} &= h_i(\underline{x}(t), t) \\ &+ \frac{\pi}{2} \sum_{j,k} \frac{\partial D_{ij}}{\partial x_k}(\underline{x}(t), t) D_{kj}(\underline{x}(t), t) \end{aligned} \right\} \textcircled{*}$$

so-called "drift coefficient"

Note:

- The second term in $\textcircled{*}$ is the so-called "noise-induced" drift.

It arises only in the Stratonovich interpretation, not for Itô !!

- The noise-induced drift vanishes anyway if $D_{ij} = \text{const}$
- Example (simple Brownian motion)

$$\left. \begin{aligned} \dot{r} &= v \\ \dot{v} &= -\gamma v + m^{-1} f(t) \end{aligned} \right\} \Rightarrow \begin{aligned} h_r &= 0, \quad h_v = -\gamma v \\ D_{rr} &= D_{rv} = D_{vr} = 0 \\ D_{vv} &= +\frac{1}{m} \end{aligned}$$

$$\Rightarrow k_r^{(A)} = 0$$

$$k_v^{(A)} = -\gamma v$$

Second Kravars-Moyal coefficient

$$K_{ij}^{(2)}(\{x(t)\}, t)$$

$$= \frac{1}{2} \lim_{\tau \rightarrow 0} \frac{1}{\tau} \langle (x_i(t+\tau) - x_i(t))(x_j(t+\tau) - x_j(t)) \rangle$$

one finds:

$$K_{ij}^{(2)} = \frac{1}{2} \sum_k D_{ik}(\{x(t)\}, t) D_{kj}(\{x(t)\}, t)$$

often called "diffusion coefficient"
(in a generalized sense !!)

also note:

if the stochastic forces are distributed via a Gauss distribution,

$$K_{i_1 i_2 \dots i_n}^{(n)} = 0 \quad \text{for } n \geq 3 \quad !!$$

Examples for $K^{(2)}$: non-overdamped
Brownian particle

$$\dot{r} = 0$$

$$\dot{v} = -\gamma v + m^{-1} f(t)$$

$$D_{rr} = D_{rv} = D_{vr} = 0$$

$$D_{vv} = \frac{1}{m}$$

we already had:
 $K_v^{(1)} = -\gamma v$!

$$\Rightarrow K_{vv}^{(2)} = K^{(2)} = \frac{\Gamma}{2m^2}$$

$$= \gamma \frac{k_B T}{m} = \gamma^2 D$$

↑ ↑
equilibrium (equipartition) Einstaer

overdamped case:

$$-\gamma v = m^{-1} f(t)$$

$$v = -\frac{1}{\gamma m} f(t)$$

$$\Rightarrow h_v = 0, D_{vv} = \left(-\frac{1}{\gamma m}\right)$$

$$\Rightarrow K_v^{(1)} = 0$$

$$K_{vv}^{(2)} = \frac{\Gamma}{2} \left(+\frac{1}{\gamma^2 m^2}\right) = \gamma k_B T m \cdot \frac{1}{\gamma^2 m^2} = \frac{k_B T}{\gamma m} = \underline{\underline{D}}$$

normal diffusion coefficient!

IV. Fokker-Planck Equation

Starting point:

Master equation (for continuous variables, here in \mathbb{D})

$$\frac{\partial}{\partial t} P(x;t) = \int dx' \left[\underbrace{W(x; x't)}_{\text{gain}} P(x';t) - \underbrace{W(x'; x;t)}_{\text{loss}} P(x;t) \right]$$

\Rightarrow strictly speaking, one needs to know the transition rates over the entire space ($\int dx'$)!

We now assume that the transition rates are significant only if x' is close to x

\Rightarrow expand the transition rates in powers of $\Delta = x - x'$

"Kramers-Moyal expansion"

(here we will not do this expansion explicitly...)

Result:

$$\frac{\partial}{\partial t} P(x,t) = \sum_{n \geq 1} \left(-\frac{\partial}{\partial x}\right)^n \tilde{K}^{(n)}(x,t) P(x,t)$$

where $\tilde{K}^{(n)}(x,t) = \frac{1}{n!} \int_{-\infty}^{\infty} d\Delta (\Delta)^n W(x+\Delta, x, t)$

Note:
 One can further show that $\tilde{K}^{(n)}(x,t)$ is identical to the Kramers-Moyal coefficient defined in Chapter III !!

i.e. $\tilde{K}^{(n)}(x,t) = K^{(n)}(x,t)$
 $= \frac{1}{n!} \lim_{\tau \rightarrow 0} \frac{1}{\tau} \langle (x(t+\tau) - x(t))^n \rangle$

Specialize now to the case that only $K^{(1)} \neq 0$, $K^{(2)} \neq 0$, but $K^{(n)} = 0$ for $n \geq 3$

exactly valid for Gaussian random forces

(more generally: $K^{n \geq 3} \approx 0$ for systems where the transition probability depends on...

$$\Rightarrow \frac{\partial}{\partial t} P(x,t) = \left[-\frac{\partial}{\partial x} K^{(1)}(x,t) + \frac{\partial^2}{\partial x^2} K^{(2)}(x,t) \right] P(x,t)$$

note: The partial derivatives act both on $K^{(1)}, K^{(2)}$ and on $P(x,t)$!!

many variables:

$$\frac{\partial}{\partial t} P(\underline{x}, t) = \left[-\sum_{i=1}^M \frac{\partial}{\partial x_i} K_i^{(1)}(\underline{x}, t) + \sum_{i,j} \frac{\partial^2}{\partial x_i \partial x_j} K_{ij}^{(2)}(\underline{x}, t) \right] P(\underline{x}, t)$$

Fokker-Planck equation (FP)

e.g. Brownian particle, non-damped, with external force $F(x)$

- $x_1 = x$
- $x_2 = y$
- $x_3 = z$
- $x_4 = v_x$
- $x_5 = v_y$
- $x_6 = v_z$

note:

- The expression [...] is the FP operator

$$\frac{\partial}{\partial t} P = \hat{L}_{FP} P$$

- Rewrite the FP equation as continuity equation

define current: $J_i = K_i^{(1)} P - \sum_j \frac{\partial}{\partial x_j} K_{ij}^{(2)} P$



$$\Rightarrow \left[\frac{\partial}{\partial t} P(\underline{x}, t) + \underbrace{\sum_i \frac{\partial}{\partial x_i} J_i}_{\text{"}\nabla \cdot \underline{J}\text{"}} = 0 \right]$$

expresses conservation of the total probability!

$$\int d\underline{x} P(\underline{x}, t) = 1$$

stationary process:

$$\frac{\partial}{\partial t} P(\underline{x}, t) = 0$$

$$\Leftrightarrow \underline{J} = \text{const}$$

Stationary solution of the FP equation

Consider, for simplicity, a problem in 1D

$$\frac{\partial}{\partial t} P(x,t) = -\frac{\partial}{\partial x} J(x,t)$$

(eg. overdamped Brownian particle in 1D)

$$\text{where } J(x,t) = \left(K^{(1)}(x) - \frac{\partial}{\partial x} K^{(2)}(x) \right) P(x,t)$$

equilibrium stationary state $\Rightarrow J=0$ and $P(x,t) \Rightarrow P^{\text{eq stat}}(x)$

$$\underbrace{K^{(1)}(x) P^{\text{eq stat}}(x)} = + \frac{\partial}{\partial x} \left(K^{(2)}(x) P^{\text{eq stat}}(x) \right)$$

$$\frac{K^{(1)}(x)}{K^{(2)}(x)} K^{(2)}(x) P^{\text{eq stat}}(x) = \frac{\partial}{\partial x} \left(K^{(2)}(x) P^{\text{eq stat}}(x) \right)$$

$$\Rightarrow K^{(2)}(x) P^{\text{eq stat}}(x) = \alpha e^{\int_c^x dx' \frac{K^{(1)}(x')}{K^{(2)}(x')}}$$

$$\Rightarrow P^{\text{eq stat}}(x) = \frac{\alpha}{K^{(2)}(x)} e^{\int_c^x dx' \frac{K^{(1)}(x')}{K^{(2)}(x')}} - \Phi(x)$$

c is an irrelevant integration constant

$$\equiv \alpha e^{-\Phi(x)}$$

$$\text{where } \Phi(x) = \ln K^{(2)}(x) - \int_c^x dx' \frac{K^{(1)}(x')}{K^{(2)}(x')}$$

one sees: $K^{(2)}$ must be positive !!

e.g. non-overdamped Brownian particle, no external forces, 1D

$$x \rightarrow v$$

$$k^{(v)} = -\gamma v, \quad k^{(v)} = m^{-2} \frac{\pi}{2}$$

one finds:

$$\begin{aligned} \Phi(v) &= \ln \frac{\pi}{2} - \int_0^v dv' \frac{(-\gamma v')}{\frac{\pi}{2} m^{-2}} \\ &= \dots = \text{const} + \underbrace{\frac{\gamma m^2}{\frac{\pi}{2}}}_{\frac{m}{2} k_B T} v^2 \end{aligned}$$

$$\Rightarrow P^{\text{stat}}(v) \sim e^{-\frac{m}{2k_B} v^2}$$

as expected !!

Maxwell-Boltzmann distribution

$$\Rightarrow \frac{\partial}{\partial t} P(x,t) \approx 0$$

$$\Rightarrow J(x,t) \approx \text{const} = J$$

current density

constant in space and time!

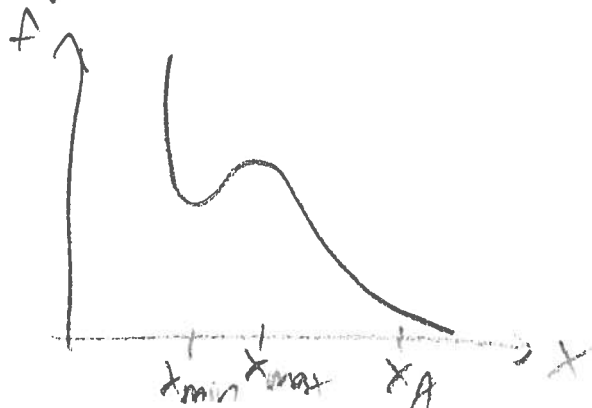
("quasi-stationary state")

\Rightarrow from $\textcircled{*}$

$$J e^{f(x)/D} = -D \frac{\partial}{\partial x} \left(e^{f(x)/D} \underbrace{P(x)}_{\text{assumed to be independent of } t!} \right)$$

integrate from x_{\min} to a point x_A outside the barrier

- and assume that $P(x_A) \approx 0$, since the particles are essentially trapped in the valley!!



$$\Rightarrow J = D e^{f(x_{\min})/D} \cdot P(x_{\min}) \cdot \left[\int_{x_{\min}}^{x_A} dx' e^{f(x')/D} \right]^{-1}$$

Consider now the probability density:

note: We have already assumed that we can work in the stationary limit !!

$$\rightarrow P(x) \rightarrow P^{\text{stat}}(x) = e^{-\Phi(x)}$$

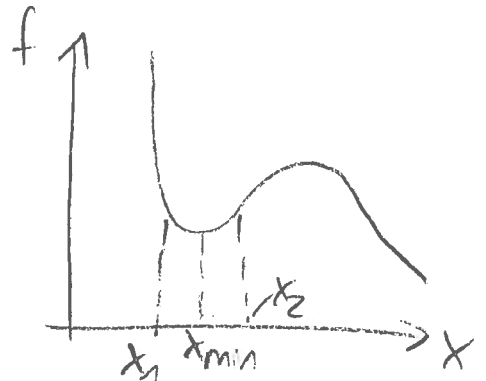
where $\Phi(x) = f(x)/D$

$$\Rightarrow \frac{P(x)}{P(x_{\min})} = e^{-[f(x) - f(x_{\min})]/D} \quad \text{for any point } x \text{ in the valley}$$

integrate within the valley to get the total probability to find the particle

$$\rho = \int_{x_1}^{x_2} dx P(x)$$

$$= P(x_{\min}) e^{f(x_{\min})/D} \int_{x_1}^{x_2} dx e^{-f(x)/D}$$



Combine this with our result for the current: