

III.) Kinetic Equations

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- a.) Kinetic Equations - An Overview

- consider Langevin equation, for Brownian motion
stochastic E field

$$\frac{dv}{dt} = -\rho v + \tilde{\eta} \quad \text{i.e. } \frac{dv}{dt} = \frac{e}{m} \tilde{E}$$

really seek $P(x, v, t) \equiv$ probability to find the particle at (x, v) in phase space at time t .
 object of Kinetic Equation

Kinetic equations seek to evolve/determine $P(x, v, t)$ directly, rather than to solve Langevin equation and the average.

Two examples:

→ Boltzmann Egn.

→ Landau.

CB.E for small $|z|$

→ glancing collision

- Landau
 - Boltzmann equation is an example of a
kinetic equation

$$f(x_1, v_1, \dots, x_N, v_N, t) \xrightarrow{\text{BBGKY}} f(x, v, t) + \text{Boltzmann Eq} \quad \left. \begin{array}{l} \text{standard distribution eqn} \\ (\text{phase space density}) \end{array} \right\}$$

Liouvillean

e.g. involves $\left\{ \begin{array}{l} \text{coarse graining} \\ \text{averaging} \end{array} \right\}$, from $\Gamma_1^t \dots \Gamma_N^t \rightarrow x, v$.

- for stochastic processes can formulate hierarchy of equations:

① Master Equation

~~Master Equation~~

$P(n, t)$ = probability to find system in n^{th} state

then, "birth" "death"

$$\frac{\partial P(n, t)}{\partial t} = \underset{\substack{\text{in} \\ \text{transitions} \\ \text{in from} \\ \text{other states}}} \downarrow - \underset{\substack{\text{out} \\ \text{transitions} \\ \text{out from } n \\ \text{to other states } n'}} \uparrow$$

$$\frac{\partial P(n,t)}{\partial t} = \sum_{n'} \left[P(n',t) W(n',n) - P(n,t) W(n,n') \right]$$

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$n' \rightarrow n$ transition probability (rate)
 \downarrow
 in probability of state n'

$n \rightarrow n'$ transition probability (rate)
 \downarrow
 probability of state n

here: probability in $\sim (P \text{ of other states}) * (\text{transition probability})$
 probability out $\sim (P \text{ of } n) * (\text{transition probability})$

Master equation is splendid example of "garbage in, garbage out" nature of kinetic equations, in that Master Egn. is only as good as transition probabilities used to construct it!

- Master equation tacitly "coarse-grains" in that evolution slower than transition event rate



then $n \rightarrow n'$ event occurs faster than τ .

② Fokker-Planck Equation

Consider system with no memory, i.e. each step on \mathcal{T} independent of prior history.

So can write:

$$P(x_3, t_3 | x_1, t_1) = \int dx_2 P(x_3, t_3 | x_2, t_2) P(x_2, t_2 | x_1, t_1)$$

prob. of x_3 at t_3
 starting from x_1
 at t_1 .

integration
 over
 intermediate
 states

↓
 $2 \rightarrow 3$
 jump
 $1 \rightarrow 2$ jump.

$$\xrightarrow{3} = \sum_{1 \rightarrow 2 \rightarrow 3} \dots$$

and

- multiplicative, as independent steps
- sum over intermediate states.

above is Chapman-Kolmogorov Equation

Now can extend to where

transition probability, of X , of
↓ step ΔX in time τ

$$P(X_2, t_2 | X_1, t_1) = T(X, \Delta X, \tau)$$

i.e. $t_2 - t_1$ is jump time τ
 $X_2 - X_1$ is jump step ΔX

then Chapman - Kolmogorov Equation becomes:

$$P(X, t+\tau) = \int d(\Delta X) P(X-\Delta X, t) T(X, \Delta X, \tau)$$

and expansion (with T indep. X) \Rightarrow

i.e. ΔX small \Rightarrow small curremt;

$$\frac{\partial P}{\partial t} = - \frac{\partial}{\partial X} \left\{ \frac{\langle \Delta X \rangle P}{\tau} - \frac{\partial}{\partial X} \frac{\langle \Delta X \Delta X \rangle P}{2\tau} \right\}$$

$$= - \frac{\partial}{\partial X} F_p$$

↓
 probability
 flux

generic form of
 Fokker-Planck Equation.
 (F-P. E.)

Note :

- F-P. Equation - no memory on scales $\epsilon > \tau$
- F.P. Equation - "coarse-grains" out $\begin{cases} t < \tau \\ X < \Delta X \end{cases}$

- F-P Equation is less general, but more tractable than Master Equation.
- intimately connected to diffusion.

③ Zwanzig - Mori Equation is

F-P Egn. with Memory Kernel (Memory correction)

i.e. variables x_1, x_2, \dots, x_N

for t slower than some $\bar{\tau}_1$, separate into 'fast' and 'slow' variables

$$\begin{array}{c} x_1, x_2, \dots, x_p \} x_{p+1} \dots x_N \\ \downarrow \quad \downarrow \\ \text{slow} \qquad \qquad \qquad \text{fast} \\ \dot{x}_i/x_i < 1/\bar{\tau} \qquad \qquad \dot{x}_i/x_i > 1/\bar{\tau} \end{array}$$

Z-M theory :

- assumes fast variables come to \odot equilibrium on fine scales $\bar{\tau}$
- can describe evolution in terms of slow variables, only.

then:

- $\underline{P}(x_1, x_2, \dots, x_p) \rightarrow (x_1, \dots, x_p)$

\downarrow
Projection operator P projects evolution onto reduced # degrees of freedom, the slow variables.

- write projected Liouville equation, for slow variables $\Rightarrow Z\text{-M. Egn.}$
- not surprisingly, $Z\text{-M. Egn.}$ can reduce to F-P. Egn.
- $Z\text{-M.}$ clearly coarse-grains over fast variables
- $Z\text{-M.}$: projection procedure part, but not all of R.G. procedure (Renormalization Group) theory

(8) Fokker-Planck Theory

Extends basic ideas notes

- seek Pdf P of Markovian, stochastic variable
- Markovian \equiv stochastic process s/t $t + \Delta t$ determined by state at t , only.
 \leftrightarrow no memory

so, as in Brownian Motion

$$P(\underline{v}, t + \Delta t) = \int d(\Delta v) P(\underline{v} - \Delta v, t) T(\Delta v, \Delta t)$$

↑
state at
 $t + \Delta t$ ↑
state at
 t ↑
transition
probability

\Rightarrow expand

$$P(\underline{v}, t) + \Delta t \frac{\partial P}{\partial t} = \int d(\Delta v) \left\{ P(\underline{v}, t) T(\Delta v, \Delta t) \right.$$

$$\left. - \frac{\partial}{\partial \underline{v}} (\Delta v T(\Delta v, \Delta t) P(\underline{v}, t)) + \frac{1}{2} \frac{\partial^2}{\partial \underline{v}^2} (\Delta v \Delta v T(\Delta v, \Delta t) P(\underline{v}, t)) \right\}$$

now as T is transition probability, it is normalized,
 \Rightarrow

$$\underline{\underline{S}} \quad \int d\Delta V T(\Delta V, \Delta t) = 1$$

$$\int d\Delta V \Delta V T(\Delta V, \Delta t) = \langle \Delta V \rangle$$

expectation
(must exist)

$$\int d\Delta V \Delta V \Delta V T(\Delta V, \Delta t) = \langle \Delta V \Delta V \rangle$$

Variance
(must exist)

$$P(V, t) + \Delta t \frac{\partial P}{\partial t} = P(V, t) - \frac{\partial}{\partial V} \left(\langle \Delta V \rangle P(V, t) \right)$$

$$+ \frac{1}{2} \frac{\partial}{\partial V} \left[\frac{\partial}{\partial V} \left(\langle \Delta V \Delta V \rangle P(V, t) \right) \right]$$

$$\begin{cases} \frac{\partial P(V, t)}{\partial t} = - \frac{\partial}{\partial V} \cdot \left\{ \frac{\langle \Delta V \rangle P(V, t)}{\Delta t} - \frac{\partial}{\partial V} \cdot \frac{\langle \Delta V \Delta V \rangle P(V, t)}{2 \Delta t} \right\} \\ = - \frac{\partial}{\partial V} \cdot F_p \end{cases}$$

- Fokker-Planck Equation

Now, can note:

- $\frac{\partial P}{\partial t} = -D \cdot \nabla^2 P$ structure assures F-P. Egn. conserves probability. Derivative order matters!
- Obviously, in "small" kick limit, ~~can relate F-P. Egn. to Master Egn.~~
- as example, for Brownian Motion:

$$\frac{\partial v}{\partial t} = -\beta v + \tilde{q}(t)$$

$\tilde{q}(t)$ \rightarrow broadband noise

$$\therefore \frac{\langle \Delta v \rangle}{\Delta t} = -\beta v$$

$$\frac{\langle \Delta v \Delta v \rangle}{2\Delta t} = D_v \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$D_v = \tilde{q}_0^2 T_{\text{eff}}$$

(uncorrelated direction)

$$\frac{\partial P}{\partial t} = -\frac{\partial}{\partial v} \cdot \left\{ -\beta v P - \frac{\partial D_v}{\partial v} P \right\} \rightarrow \left\{ \begin{array}{l} \text{F-P. Egn.} \\ \text{for} \\ \text{Brownian} \\ \text{Motion} \end{array} \right.$$

in 1D,

$$\frac{\partial P}{\partial t} = +\frac{\partial}{\partial v} \left\{ \beta v P + D_v \frac{\partial P}{\partial v} \right\}$$

$$D_v = \frac{1}{2} V_{th}^2$$

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so, at equilibrium ($\partial P / \partial t = 0$)

$$P \approx \exp \left[- \beta v^2 / 2 D_v \right]$$

i.e. Gaussian p.d.f. formed by balance of drag with diffusion.

In the absence of drag, with $P(v, 0) = \delta(v - v_0)$

$$P(v, t) = \frac{1}{\sqrt{\pi D_v t}} \exp \left[-v^2 / 2 D_v t \right]$$

i.e. diffusion p.d.f.

- F-P. Equation structure (general) :

$$\begin{matrix} \text{drag/drift} \\ \text{term} \end{matrix} \rightarrow \frac{\langle \Delta v \rangle P}{\Delta t} = \underline{\nabla} P \quad \hookrightarrow \begin{matrix} \text{drift} \\ \text{velocity} \end{matrix}$$

$$\begin{matrix} \text{diffusion} \\ \text{term} \end{matrix} \rightarrow - \frac{\partial}{\partial v} \cdot \frac{\langle \Delta v \Delta v \rangle}{2 \Delta t} P = - \frac{\partial}{\partial v} \cdot \underline{\underline{D}_v} P$$

\downarrow
diffusion tensor

$$\text{and: } \frac{\partial P}{\partial t} + \frac{\partial}{\partial v} (\underline{\nabla} P) = + \underline{\underline{D}_v} \cdot \underline{\nabla}_v \underline{\underline{D}_v} P$$

$$\underline{\underline{D}_v} = - \underline{\nabla} P - \underline{\underline{D}_v} \cdot \underline{\underline{D}_v} P$$

drift \rightarrow deterministic part of motion

diffusion \rightarrow random part. (noise related)

- requirements for applicability of
Fokker Planck Theory

\rightarrow stochastic motion

\rightarrow step size

$\Delta V, \Delta P$

\rightarrow no memory ($t > \Delta t$)

and

$$\left. \begin{array}{l} \langle \Delta V \rangle < \infty \\ \langle \Delta V^2 \rangle < \infty \end{array} \right\} \rightarrow \text{convergence of lowest 2 moments}$$

at al Central Limit Theorem.

if $\langle \Delta V^2 \rangle \rightarrow \infty$, need turn to Fractional kinetics.
 CTRW
 \rightarrow Levy Flights, etc.

- Fokker-Planck equation \leftrightarrow Markov process or chain, which is gradient unfolding of transitions probability just \approx

conservative dynamical system is gradual unfolding of contact transformation.

- far - Hamiltonian system \leftrightarrow Liouville Thm.
- no systematic bias

can show: (HW)

$$\frac{1}{2} \frac{\partial}{\partial V} \cdot \langle \underline{\Delta V} \underline{\Delta V} \rangle = \langle \underline{\Delta V} \rangle$$

i.e. partial cancellation of diffusion and drag / drift

$$\text{i.e. } \frac{\partial P}{\partial t} = - \frac{\partial}{\partial V} \left(\frac{\langle \underline{\Delta V} \rangle P}{\Delta t} - \frac{\partial}{\partial V} \cdot \frac{\langle \underline{\Delta V} \underline{\Delta V} \rangle P}{2 \Delta t} \right)$$

$$= - \frac{\partial}{\partial V} \left(\frac{\langle \underline{\Delta V} \rangle}{\Delta t} - \left(\frac{\partial}{\partial V} \cdot \frac{\langle \underline{\Delta V} \underline{\Delta V} \rangle}{2 \Delta t} \right) P \right)$$

$$- \frac{\langle \underline{\Delta V} \underline{\Delta V} \rangle}{2 \Delta t} \cdot \frac{\partial P}{\partial V} \right)$$

$$= \frac{\partial}{\partial V} \cdot P_V \cdot \frac{\partial P}{\partial V} \rightarrow \text{Form of diffusion equation for Hamiltonian system}$$

(note order of derivatives!)

Here $\langle \underline{A}V \rangle = \frac{1}{2} \frac{\partial}{\partial V}$. $\langle \underline{A}V \underline{A}V \rangle$ is analogue

of incompressibility of phase space flow for stochastic system.

→ Now can extend Fokker-Planck theory to bivariate evolution.

i.e. consider Brownian Motion in External Force Field ...

$$\frac{\partial \underline{v}}{\partial t} = -\beta \underline{v} + \underline{q}_{ext} + \tilde{\underline{q}}$$

\downarrow

↳ Brownian force

$$\frac{\underline{f}_{ext}}{m_p} = -\beta \underline{\phi}$$

\rightarrow potential (i.e. spring gravity)

$$\frac{d\underline{x}}{dt} = \underline{v}$$

so obviously particle random walks in \underline{x} and \underline{v} .
For phase space pdf:

$$P(\underline{x}, \underline{v}, t + \Delta t) = \int d(\Delta \underline{x}) \int d(\Delta \underline{v}) \left\{ P(\underline{x} - \Delta \underline{x}, \underline{v} - \Delta \underline{v}, t) T(\Delta \underline{x}, \Delta \underline{v}, \Delta t) \right\}$$

Furthermore, Brownian kick applied only in \underline{v}
 \Rightarrow \underline{x} kinematic

\Rightarrow

$$T(\underline{\Delta x}, \underline{\Delta v}, \Delta t) = \delta(\underline{\Delta x} - \underline{v} \Delta t) T(\underline{\Delta v}, \Delta t)$$

\therefore

$$\begin{aligned} P(\underline{x}, \underline{v}, ++\Delta t) &= \int d(\underline{\Delta x}) \int d(\underline{\Delta v}) P(\underline{x} - \underline{\Delta x}, \underline{v} - \underline{\Delta v}, +) * \\ &\quad \delta(\underline{\Delta x} - \underline{v} \Delta t) T(\underline{\Delta v}, \Delta t) \\ &= \int d(\underline{\Delta v}) P(\underline{x} - \underline{v} \Delta t, \underline{v} - \underline{\Delta v}, +) T(\underline{\Delta v}, \Delta t) \end{aligned}$$

so can re-write:

$$P(\underline{x} + \underline{v} \Delta t, \underline{v}, ++\Delta t) = \int d \underline{\Delta v} P(\underline{x}, \underline{v} - \underline{\Delta v}, +) T(\underline{\Delta v}, \Delta t)$$

and now expand, as before:

$$+ \underline{q}_{ext} \cdot \partial P / \partial \underline{v}$$

$$\frac{\partial P}{\partial t} + \underline{v} \cdot \nabla_x P = - \frac{\partial}{\partial \underline{v}} \left[\frac{\langle \underline{\Delta v} \rangle}{\Delta t} P - \frac{\partial}{\partial \underline{v}} \cdot \frac{\langle \underline{\Delta v} \underline{\Delta v} \rangle}{2 \Delta t} P \right]$$

More generally have shown can write:

$$\left\{ \frac{dP}{dt} \right\} = F.P. \text{ Operator} = \beta \underbrace{\frac{\partial}{\partial V}}_{\text{deterministic orbits}} \cdot (\underline{V} P) + \alpha \underbrace{\frac{\partial^2 P}{\partial V^2}}_{\text{randomly fluctuating orbits}}$$

where "deterministic orbits" means:

$$\frac{dx}{dt} = v, \quad \frac{dv}{dt} = q_{ext}$$

$$\rightarrow \text{Now } P = P(x, v, t).$$

Often seek only $P(x, t)$, so ... can obtain full $P(x, v, t)$ and integrate over v , which is laborious

or derive moment equations of F.P. Equation in Γ , yield "fluid equations" in x !

obviously

akin to deriving fluid equations from Boltzmann equation

i.e. from F-P eqn. for $P(x, v, t)$

derive equations for:

$$n(x, t) = \int dv P(x, v, t) \rightarrow \text{density}$$

$$v(x, t) = \int dv v P(x, v, t) / n(x, t) \rightarrow \text{Eulerian velocity}$$

n -equation \leftrightarrow Schmoluchowski Equation

Now have: (for Brownian particle)

$$\frac{\partial P}{\partial t} + \underline{v} \cdot \nabla_x P + \underline{q}_{\text{ext}} \cdot \nabla_v P$$

$$= \beta \frac{\partial}{\partial v} (\underline{v} P) + D_v \frac{\partial^2 P}{\partial v^2}$$

which can be re-written as:

$\underbrace{\hspace{10em}}$
in a superficially very complicated form; as...

$$\frac{\partial \rho}{\partial t} = \beta \left(\frac{\partial}{\partial v} - \frac{1}{\beta} \frac{\partial}{\partial x} \right) \cdot \left(v \rho + \frac{\partial v}{\partial v} \frac{\partial \rho}{\partial v} - \frac{g_{ext}}{\beta} \rho + \frac{\partial v}{\beta} \frac{\partial \rho}{\partial x} \right)$$

$$+ \frac{\partial}{\partial x} \cdot \left(\frac{\partial v}{\beta} \frac{\partial \rho}{\partial x} - \frac{g_{ext}}{\beta} \rho \right)$$

$$\text{Now: } n(x, t) = \int dv \rho(x, v, t)$$

$$x + \frac{v}{\beta} = x_0$$

i.e. integrate along line stt $\dot{x} = -\dot{v}/\beta$

→ This annihilates term #①!

i.e. $x + \frac{v}{\beta} = \text{const} \Rightarrow \frac{\partial}{\partial v} - \frac{1}{\beta} \frac{\partial}{\partial x} = 0$

so obtain:

$$\frac{\partial n(\underline{x}, t)}{\partial t} = \frac{\partial}{\partial \underline{x}} \left(\frac{\partial v}{\partial^2} \frac{\partial n}{\partial \underline{x}} - \frac{q_{ext}(t)}{\rho} n \right)$$

- the Smoluchowski eqn. for $n(\underline{x}, t) \rightarrow$
spatial pdf

Observe:

- can short-circuit complicated derivation by simply going to "terminal velocity" limit.
i.e. eqns of motion:

$$\frac{\partial v}{\partial t} = -\beta v + q_{ext} + \tilde{q}$$

$$\frac{dx}{dt} = v$$

at terminal velocity

$$v = \frac{q_{ext}}{\beta} + \frac{\tilde{q}}{\beta}$$

$$\frac{dx}{dt} = \frac{q_{ext}}{\beta} + \frac{\tilde{q}}{\beta}$$

↳ deterministic

↳ random.

$$\left. \begin{aligned} \frac{\partial n}{\partial t} + \frac{\partial}{\partial x} \cdot \left(\frac{dx}{dt} n \right) &= D_{xx} \frac{\partial^2 n}{\partial x^2} \\ D_{xx} &= Dv / \beta^2 \end{aligned} \right\} \text{deterministic}$$

\Rightarrow Schmaluchowski Eqn.

- still conservative:

$$\frac{\partial n}{\partial t} = - \frac{\partial}{\partial x} \Gamma_n$$

$$\Gamma_n = \left(\frac{q_{\text{ext}}}{\beta} n - \frac{Dv}{\beta^2} \frac{\partial n}{\partial x} \right)$$

\downarrow
connection
velocity

\downarrow
Diffn