

Physics 210b

- **Model Reduction** \leftrightarrow Statistical Dynamics
(Toward Renormalization)
- Motivation:
 - frequently of interest to reduce description of a system with many d.o.f.
 - some notion of / that some d.o.f.s
 - { relevant slow } of interest
 - { irrelevant fast }
 - often relevance based on space-time scales

e.g.

complex turbulent flow

= large scales - relevant
i.e. determine flow pattern macroscopics

+
small scales - irrelevant
i.e. don't need details (?)

Questions :

→ statistical description of small scales
 explicit description of large scales

- how, systematically?
- how couple ranges?

Goal of Mori-Zwanzig Theory (Model Reduction)
 is to project irrelevant d-o-f's onto relevant d-o-f's, and derive model of relevant d-o-f's.

How?

- derive Langevin equation for
 - relevant d-o-fs
 with (statistical) model of interaction with irrelevant d-o-fs
- key elements :
 - = memory kernel \rightarrow products
 - = noise

→ Memory kernel can have extended width \Rightarrow Non-Markovian

Gaveat Emptor :- Framework

- not convergent,
- porous.
- cumbersome
- continues in RG.

Sources : - Zwanzig } texts
 - Serra, et. al. }
 - Chorin }
 - Mori } porous.

also $\stackrel{\text{Goldenfeld}}{\text{FC}}$ - McComb
 texts.

Simple Examples

- (THE) Langevin Equation

$$\partial V + \frac{B}{m} V = \frac{\tilde{f}}{m}$$

$$\langle \tilde{f}(t) \tilde{f}(t') \rangle = |\tilde{f}|^2 \tau_{av} \delta(t-t')$$

- d-o-fs : - thermal fluctuations T_{ac}
- dynamics on many time scales enter
- simplest model !
- \rightarrow $\frac{dx}{dt} = \tilde{f}/\beta$
- $\Rightarrow \langle x^2 \rangle \sim 2 D t$
- $D = T/\beta$

Obviously,

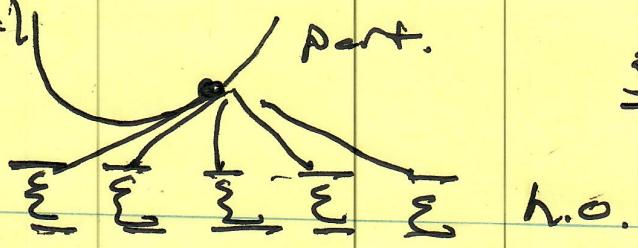
→ only particle motion is explicit

→ fast thermal d-o-fs \Rightarrow mean forcing

fluid dynamics (memory) \Rightarrow drag coefficient

Recall: 'Mechanical' model of thermal bath (Zwanzig)

potentially part.



c.e. $H = H_S + H_B$

$$H_S = \frac{p^2}{2m} + U(x) \rightarrow \text{"system relevant"}$$

$$H_B = \sum_j \left[\frac{p_j^2}{2} + \frac{1}{2} \omega_j^2 (q_j - \frac{\zeta_j}{\omega_j^2} x)^2 \right]$$

\rightarrow "bath"
irrelevant

\uparrow
coupling of
bath to system

sample!

- goal is to derive FOM for relevant variables in presence of bath

- * specify bath statistically

\rightarrow [~ coarse graining of irrelevant dynamics]

\Rightarrow irreversibility ?

Recall:

$$\frac{dp}{dt} = -\dot{U}(x(t)) - \int_0^t ds \ k(s) \rho \frac{(t-s)}{m} + F_p(t)$$

\downarrow
noise

Memory kernel

$$k(s) = \sum_j \frac{\delta_j^2}{\omega_j^2} \cos(\omega_j t) \rightarrow \text{kernel set}$$

$\left\{ \begin{array}{l} \text{couplings} \\ \text{free.} \end{array} \right. \begin{array}{l} \text{by} \\ \text{Bath} \\ \text{properties.} \end{array}$

$\rightarrow \sqrt{\text{self-energy}}$

\rightarrow Contrast:

$$\frac{dv}{dt} = -\frac{c}{m} v + \frac{F}{m} \rightarrow \text{local in time}$$

$$\int_0^t k(s) \rho \frac{(t-s)}{m} \rightarrow \text{non-local (Markovian)}$$

\rightarrow non-local
(Non-Markovian)

Recall:

$$\sum_j \rightarrow \int d\omega g(\omega)$$

NB: Statistics
Description of
density of states
for bath

$$\delta^2 = \delta^2(\omega)$$

and if: $\sigma(\omega) \sim \omega^2$
 $\gamma^2 \sim \text{const.}$

$$h(t) = \gamma^2 \sigma(t)$$

\downarrow
 Localized kernel \rightarrow Markovian

\rightarrow Noise?

$$F_p(t) = \sum_j \gamma_j P_j(\omega) \sin \frac{\omega_j t}{\omega_j}$$

$$+ \sum_j \gamma_j \left(q_j(\omega) - \frac{x_j}{\omega_j} x(\omega) \right) (\cos \omega_j t)$$

$P_j(\omega)$, $q_j(\omega)$ distributed according to:
 $f_{eq} \hat{=} \exp[-H_B/I]$

$$\left\langle \left(q_j(\omega) - \frac{x_j}{\omega_j} x(\omega) \right)^2 \right\rangle = \frac{T}{\omega_j^2} \quad \text{etc.}$$

and EDT \rightarrow (description builds on)

$$\langle F_p(t) F_p(t') \rangle = T \int_0^T h(t-t') \text{memory kernel.}$$

→ Why review?

MZT seeks to convert full problem to:

Non-Markovian Langevin Eqn.' for relevant variables in terms of

- memory kernel
 - noise
- } for irrelevant variables

⇒ includes statistical representation of irrelevant d.o.f's

⇒ reduced model.

→ Generic approach to model reduction ...

→ MST Formalism - General Theory

- N variables
 - modes
 - particles

$$\mathbf{z} = (z_1, \dots, z_N)$$

$$\dot{z}_j = h_j(z) \rightarrow \text{dynamical eqns.}$$

- then assume can linearize terms from EOM:

$$\dot{p}_j = -\gamma_j p_j + \Sigma_j(p) \quad j = 1 \dots N$$

- if concerned with evolution on time scale $\sim \bar{\tau}$, then can classify variables

- $\gamma_i \bar{\tau} > 1 \rightarrow$ "inertial" / fast variables
 i.e. have equilibrated on time scale $\bar{\tau}$
 (analogous to terminal velocity)

$\delta_i \tau < \pm \rightarrow$ "relevant" / slow variables
 i.e. evolving, not equilibrated
 on τ

Idea: Project irrelevant variables onto relevant variables.

For fast variables:

$$\dot{p}_j = -\delta_j p_j + z_j(\theta), \quad j = M+1, \dots, N$$

$(M < N)$

$$\dot{p}_j \approx 0 \Rightarrow p_j = \frac{z_j(\theta)}{\delta_j}$$

$$P_{\text{fast}} = Z_{\text{fast}}(\theta) / \delta_f$$

↓
 slower fast
 eliminated
 variables to slow
 variables

Can use to obtain dynamical
 equations in terms of slow variables
 only.

Now, for pdf evolution \rightarrow

Seek Master Egn. for Non-Markovian system.

\Leftrightarrow construct from Liouville

- Variables = $a \rightarrow$ slow, relevant
- $b \rightarrow$ fast, irrelevant

so for pdf:

$$\partial_t p(a, b, t) = L p(a, b, t)$$

\uparrow
Liouville (redu. Kubo)
operator
(differential) \rightarrow i.e. Poisson
bracket

Can define reduced pdf:

$$S(a, t) = \int_b^t p(a, b, t)$$

\int_b^t
integrates out irrelevant, fast
variables.

And, assume can decompose L as:

$$L = L_a + L_b + L_i$$

↓ ↓ ↓
 slow fast interaction

(fast-slow coupling)

Now: - assuming "equilibrium distribution" exists for b variables only.

[N.B. What does "equilibrium" mean?
Irreversibility?]

$$\rho_{\Sigma}(b)$$

$$- L_b \rho_{\Sigma}(b) = 0$$

$$\int db \rho_{\Sigma}(b) = 1$$

- can define projection \underline{P} onto
a variables via:

$$\begin{aligned} \underline{P} P(a, b, t) &= \rho_{\Sigma}(b) \int db p(a, b, t) \\ &= \rho_{\Sigma}(b) S(a, t) \end{aligned}$$

Now, for projection needs:

$$\underline{\underline{P}} \underline{\underline{P}} P = \underline{\underline{P}} P \quad (\text{idempotency})$$

$$\begin{aligned} \underline{\underline{P}}^2 P(a, b, t) &= \underline{\underline{P}} P_{\mathcal{Z}}(b) S(a, t) \\ &= P_{\mathcal{Z}} \int db P_{\mathcal{Z}}(b) S(a, t) \\ &= P_{\mathcal{Z}} S(a, t) = \underline{\underline{P}} P(a, b, t) \end{aligned}$$

$\underline{\underline{P}}$ is projection ✓

- And can further define:

$$P_1 = \underline{\underline{P}} P(a, b, t)$$

$$P_2 = (\underline{\underline{I}} - \underline{\underline{P}}) P(a, b, t) = \underline{\underline{Q}} P(a, b, t)$$

$$\underline{\underline{Q}} P = L P$$

and

$$\frac{dP_1}{dt} = \underline{P} L (P_1 + P_2) \quad (\text{oper. } \underline{P})$$

→ slow

$$\frac{dP_2}{dt} = \underline{Q} L (P_1 + P_2) \quad \rightarrow \text{fast}$$

Solving P_2 equation \Rightarrow

$$\begin{aligned} P_2(t) &= e^{\underline{Q}Lt} P_2(0) + e^{\underline{Q}Lt} \int_0^t ds e^{-\underline{Q}Ls} \underline{Q} L P_1(s) \\ &= e^{\underline{Q}Lt} P_2(0) + \int_0^t e^{\underline{Q}L(t-s)} Q L P_1(t-s) ds \end{aligned}$$

then can substitute into P_1 eqn.
to eliminate P_2 .
 \Rightarrow

\int_0^t prof i.c. $a \rightarrow$ slow
 $b \rightarrow$ fast

$$\frac{dP_1}{dt} = PL P_1(t) + PL e^{\underline{Q}Lt} \underbrace{Q p(b)}_{\text{fast} \rightarrow \text{slow}}$$

$$+ \int_0^t PL e^{\underline{Q}Ls} Q L \underbrace{P_1(t-s)}_{\phi(s)}$$

$\phi(s)$ - memory kernel

→ Similar structure to both problem (not surprisingly)

Salient features:

- memory kernel $\phi(s) \rightarrow$ from elimination of part b's in terms slow a's
- has form \rightarrow time propagation

$$PL e^{\int QLS} PL P$$

2 Liouville operators

key physics:
What controls time
history in
memory kernel?

- recall form of non-Markovian renormalized Vlasov eqn.

$$-i(\omega - kv) f_{k\omega} - \partial_v D_{k\omega} \partial_v f_{k\omega}$$

$$= -\frac{q}{m} E_{k\omega} \frac{\partial \langle f \rangle}{\partial v}$$

$$L e^{QLS} L \rightarrow \partial_v D_{k\omega} \partial_v$$

[similar structure]

$\hookrightarrow \tilde{T}_C$ with $\omega + \omega$

→ Simplifying the Non-Markovian Master Eqn.

$$\frac{d}{dt} P_t = PLP_p(t) + PLC e^{QLt} Q p(0)$$

$$+ \int_0^t \underbrace{\phi(s)}_{\text{kernel}} p(t-s) ds$$

$$L = L_a + L_b + L_c$$

$$= L_0 + L_i$$

$\overset{b}{\overset{a}{\rightarrow}}$ a, b interacting
zeroth order Liouville operators
decoupled Liouillian
operator

Then, like Schrödinger \rightarrow Heisenberg
(zweinzig states) $\overset{b}{\overset{a}{\rightarrow}}$ follows Kubo

$$(P \equiv C P(a, b, t))$$

$$L_0 = L_a + L_b$$

$$L_i(t) = e^{-L_0 t} L_i e^{L_0 t}$$

$$\frac{\partial_t}{\partial_t} P(t) = L_i(t) P(t)$$

L.R

$$\frac{\partial_t}{\partial_t} \left(e^{-L_0 t} P(s, t) \right) =$$

$$e^{-L_0 t} \left(-L_0 P + \frac{\partial P}{\partial t} \right) =$$

$$(e^{-L_0 t} L_i e^{L_0 t}) (e^{-L_0 t} P)$$

$$-L_0 P + \frac{\partial P}{\partial t} = L_i P$$

$$\frac{\partial P}{\partial t} = (L_0 + L_i) P$$

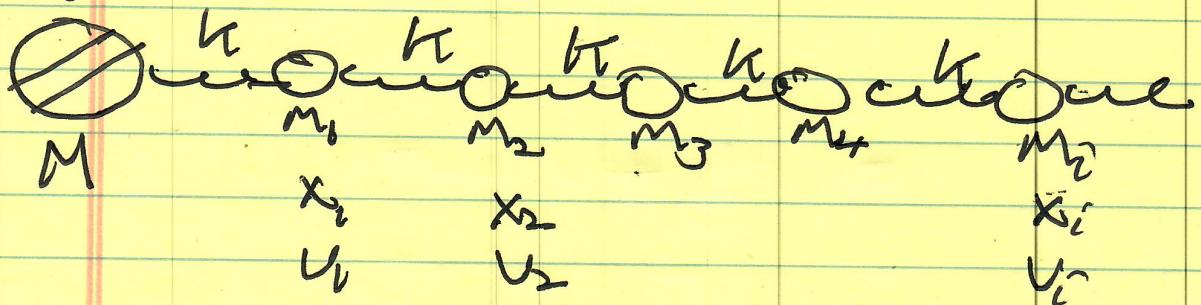
Note: Understood that exponentials to be calculated ala'

$$\exp \left[\int_0^t A(s) ds \right] = 1 + \sum_{r=1}^{\infty} \frac{(-1)^r}{r!} \int_0^t ds_1 \int_0^{s_1} ds_2 \dots \int_0^{s_{r-1}} ds_r +$$

$$\left[A(s_1) \dots A(s_r) \right]$$

→ Linear Chain → A Case Study

- Chain of springs with constant k and masses $M \gg m_1, m_2, \dots, m_N$
 \times
 \checkmark (molecular model)

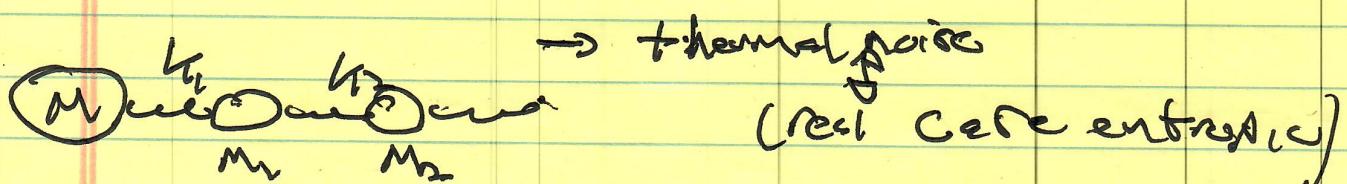


$$\omega_{\text{end}}^2 = \frac{k}{M} \ll \omega_i^2 = \frac{k}{m_i}$$

slow
relevant

fast
irrelevant

- Application - Linear Chain



Now → tweek with a laser

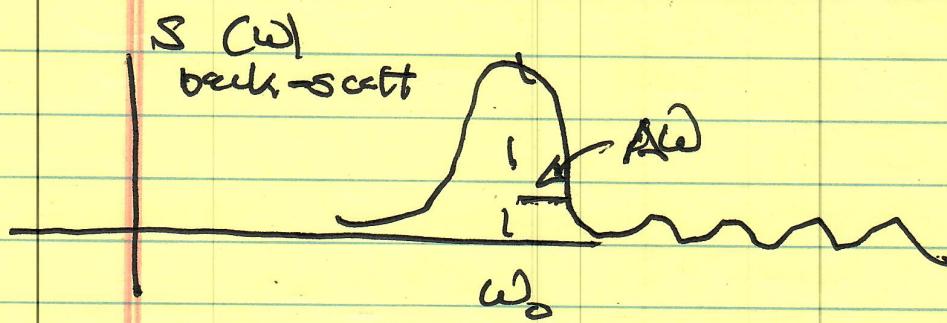
⇒ output is speckle ⇒ mode frequencies.

but recall:

$$\omega_0^2 = \frac{k}{M} \ll \omega_c^2$$

↓ ↓
relevant, fast
slow irrelevant

→ system's small
masses motion
constitute noise
for slow mode



→ What is $\Delta\omega$? → Relaxation rate of
of low frequency mode

→ But relaxation rate set by kicks due
fast modes → background effective noise
calculate

⇒ MZT useful in calculating relaxation
rates for slow modes in complex
systems.

Aside: How ~~do we~~ calculate
line width?

→ Then, write down equations of motion, each - mass:

$$\dot{x} = v$$

$$\ddot{x} = -\frac{k}{m}(x - x_0)$$

⋮

$$x_i = v_i, \quad \ddot{x}_i = -\frac{k}{m_i}(x_i - x_{i-1}) - \frac{k}{m_i}(x_i - x_{i+1})$$

Fall k
same

$$\dot{x}_i = v_i$$

$$\ddot{v}_i = -\frac{k}{m_i}(x_i - x_{i-1}) - \frac{k}{m_i}(x_i - x_{i+1})$$

⋮

To simplify, work in relative coordinates only:

$$\delta x_0 = x - x_0$$

$$\delta x_1 = x_1 - x_0$$

$$\delta x_i = x_i - x_{i-1}$$

so, EOMs in simplest forms:

$$\dot{\delta x}_0 = v$$

$$\ddot{v} = -\frac{k}{M}\delta x_0$$

$$\dot{\delta x}_i = v_i - v_{i-1}$$

$$\ddot{v}_i = \frac{k}{m_i}\delta x_0 - \frac{k}{m_i}\delta x_1$$

+ from $-(x_i - x)$

$$\dot{\delta x}_1 = v - v_0$$

$$\ddot{v}_i = \frac{k}{m_i}\delta x_{i-1} - \frac{k}{m_i}\delta x_i$$

So now construct Liouvilian:

$$L = \underbrace{\frac{dx}{dt} \frac{\partial}{\partial x} + \frac{dv}{dt} \frac{\partial}{\partial v}}_{M} + \sum_j \left(\frac{d\delta x_j}{dt} \frac{\partial}{\partial \delta x_j} + \frac{d\delta v_j}{dt} \frac{\partial}{\partial \delta v_j} \right).$$

where: $\rightarrow \frac{d\delta x_j}{dt} = (v_j - v_{j+1})$

$$\frac{dv_j}{dt} = \frac{k_e}{m_j} \delta x_{j+1} - \frac{k_e}{m_j} \delta x_j$$

$$\rightarrow \ddot{v} = - \frac{k_e}{m_j} \delta x_j$$

\times variable absorbed into δx_j

so

$$L = \left[- \frac{k_e}{m_1} \delta x_1 \frac{\partial}{\partial v} + \left[(v - v_1) \frac{\partial}{\partial \delta x_1} \right] \right]$$

$$+ \left[\left(\frac{k_e}{m_1} \delta x_1 - \frac{k_e}{m_2} \delta x_2 \right) \frac{\partial}{\partial v_1} \right] + \left[(v_1 - v_2) \frac{\partial}{\partial \delta x_2} \right]$$

$$+ \left(\frac{k_e}{m_2} \delta x_2 - \frac{k_e}{m_3} \delta x_3 \right) \frac{\partial}{\partial v_2}$$

+ ...

$$+ (V_C - V_{C(i)}) \frac{\partial}{\partial x_i} + \left(\frac{k}{m_i} \partial x_{i-1} - \frac{k}{m_i} \partial x_i \right) \frac{\partial}{\partial V_{C(i)}}$$

This yields Liouville equation:

$$\frac{\partial P}{\partial t} + D \cdot [V_F P] = 0$$

$$\nabla \cdot V_F = 0 \quad \text{timeless}$$

$$\frac{\partial P}{\partial t} + L P = 0, \quad L = \nabla_{V_F} \cdot D_F$$

remaining to decompose L

As before:

- need decompose into $\left\{ \begin{array}{l} \text{relevant (E)} \\ \text{slow} \\ \text{irrelevant (I)} \\ \text{fast} \\ \text{interaction (C)} \end{array} \right.$ parts
- $\omega_D^2 = \frac{k}{M} \ll \omega_j^2 = \frac{k}{m_j}$
slow (variable) frequency

but

- using relative coordinates \Rightarrow no isolated slow variables i.e.

$$\delta x_0 = x - x_i$$

slow fast

- $L_a = 0$

$$L_i = -\frac{k}{m} \delta x_0 \frac{\partial}{\partial v} + v \frac{\partial}{\partial x_0}$$

b
interaction

$$\delta x_0 = x - x_i$$

$\underbrace{}_{\text{S}}$ $\underbrace{}_{\text{F}}$

- L_b \equiv irrelevant (fast) fast variable
Liouvillean
above

$$\begin{aligned}
 L_b &= (\cancel{x - v_i}) \frac{\partial}{\partial x_0} + \left(\frac{k}{m_i} \delta x_0 - \frac{k}{m} \delta x_i \right) \frac{\partial}{\partial v} \\
 &+ (v - v_2) \frac{\partial}{\partial x_1} + \left(\frac{k}{m_1} \delta x_1 - \frac{k}{m_2} \delta x_2 \right) \frac{\partial}{\partial v_2} \\
 &\vdots \\
 &(v_c - v_{c+1}) \frac{\partial}{\partial x_0} + \left(\frac{k}{m_1} \delta x_{c+1} - \frac{k}{m} \delta x_c \right) \frac{\partial}{\partial v_{c+1}}
 \end{aligned}$$

$$\Leftrightarrow L = \underbrace{L_a}_{\uparrow} + L_i + L_b$$

$$\frac{\partial P}{\partial t} = L^P$$

↑
↓

$$\Rightarrow \frac{\partial P}{\partial t} + V_F \cdot \nabla_F P = 0 \quad \Leftrightarrow \frac{\partial}{\partial t} V_F = 0$$

(Hamiltonian)

Now:

→ integrate out the fast variables

→ need $P_{eq}(b)$ → fast equilibrium distribution

$$\rightarrow \text{Recall } \underline{\underline{P}} = P_{eq}(b) \sqrt{db}$$

→ can write:

$$P_{eq}(dx_i, v_i) = \prod_{i=1}^N \exp \left[-\frac{d x_i^2}{2 \Delta_i^2} \right] * \exp \left[-\frac{v_i^2}{2 \langle v_i^2 \rangle} \right]$$

factorize, for each drop.

Gaussian convenient (moments converge)

→ thermokinetic assumption

For extremi:

- $\int_b P_{\Sigma}(b) = 0$

\Leftrightarrow constant FDT
(noise + damping)

- Form: CLT

Then,

$$\int_b P_{\Sigma}(b) = 0$$

then

$$(V - V_1) \frac{dx}{\Delta_0^2} + \left(\frac{k}{m_1} \frac{dx_0}{\Delta_0^2} - \frac{k}{m_1} \frac{dx_1}{\Delta_1^2} \right) \frac{V_1}{\langle V^2 \rangle}$$

$$+ (V_1 - V_2) \frac{dx_1}{\Delta_1^2} + \left(\frac{k}{m_2} \frac{dx_2}{\Delta_2^2} - \frac{k}{m_2} \frac{dx_1}{\Delta_1^2} \right) \frac{V_2}{\langle V^2 \rangle}$$

$$\therefore \frac{1}{\Delta_0^2} = \frac{k_1}{m_1} \frac{\langle V_1 \rangle^2}{\langle V^2 \rangle} \Rightarrow m \langle V^2 \rangle = \frac{c \Delta_0^2}{2}$$

$$V_1 \frac{dx_1}{\Delta_1^2} = \frac{k_1}{m_1} \frac{dx_1}{\Delta_1^2} \frac{\langle V_1 \rangle^2}{\langle V^2 \rangle}$$

$$m \frac{\langle V_2 \rangle^2}{2} = \frac{k_2 \Delta_0^2}{2}$$

⋮

27.

⇒ sequential equipartition thru chain!

$$\frac{k}{2} \Delta_j^2 = \frac{m}{2} \langle v_{j+1}^2 \rangle$$

then $\int L_b \rho_{eq}(b) = 0$

Recall: $L_i = -\frac{k}{m} \frac{\partial x_i}{\partial v} + v \frac{\partial}{\partial x_i}$

For total equilibrium:

$$P = C \exp \left[-\frac{v^2}{2 \langle v^2 \rangle} \right] \rho_{eq}(b)$$

$$M \langle v^2 \rangle = k \langle \Delta_0^2 \rangle = \dots m \langle v_i^2 \rangle$$

so, for projection,

$$\underline{P} (\sigma(v, b, t)) = \rho_{eq}(b) \int db \rho(v, b, t)$$

→ Now, to construct Master equation :

$$\mathcal{L} = \mathcal{L}_i$$

- recall derived :

i.e.

$$\frac{\partial}{\partial t} P_i^P = P_i^P (\Delta P_i^P(t) + \overline{P_i^P} \exp \left[\int_0^t ds Q_i^P(s) \right] * Q_i^P(0))$$

$$+ \int_0^t ds \overline{P_i^P(s)} \exp \left[\int_s^t ds' Q_i^P(s') \right] \overline{Q_i^P(s)} \overline{P_i^P(s)}$$

Memory kernel

higher order interaction effect

Now - ignore h.o. interactions, implicit
in exponentials

$$\exp [] \approx 1. \rightarrow \text{short } T_{\text{rec}}$$

$$\frac{\partial}{\partial t} P_i^P = \int_0^t ds \overline{P_i^P(s)} \overline{F_i^P(s)} \overline{A_i^P(s)}$$

$$\frac{\partial}{\partial t} \tilde{P}_i(t) = \int_0^t d\sigma \stackrel{P}{=} \mathcal{L}_i(t) \tilde{P}_i(\sigma)$$

or Kernel

$$\tilde{\Phi} = e^{-\mathcal{L}_B(t-s)} \tilde{P}_B(t-s)$$

truncates (slow)

$$\left\{ \begin{array}{l} A = \mathcal{L}_B / P_{eq}(v) \\ \text{absorb } P_{eq}(v) \\ \text{in } P_{eq}(b) \end{array} \right.$$

$$\tilde{P}_B(t-s) = \stackrel{P}{=} \mathcal{L}_i e^{-(\mathcal{L}_B + \mathcal{L}_i)(t-s)} \stackrel{P}{=} \mathcal{L}_i P$$

Thus can finally write non-Markovian Master equation for reduced system as:

$$\frac{\partial P(y,t)}{\partial t} = \frac{1}{(P_{eq}(b))} \int_0^t \tilde{\Phi}_B(t-s) \Delta_{eq}(b) P(y,s) ds$$

Kernel

to cancel piece of P

$$\tilde{\Phi}_B = P \mathcal{L}_i e^{-\mathcal{L}_B(t-s)} \mathcal{L}_i^{-1}$$

Now:

- time scale for $\partial C/V/t$

$$\frac{1}{P} \frac{\partial P}{\partial t} \sim \frac{1}{T_{slow}}$$

$T_{slow} \gg T_{slow-fast}$ → characteristic time of memory kernel

- so, take Markovian limit -

$$\frac{\partial P}{\partial t} = L_{eff} P \quad * \quad P = C(V, t)$$

$$L_{eff} = \frac{1}{P_{eff}(b)} \sum_i \Phi_B^{(i)} G^{(i)} P_{eff}^{(i)}$$

$$\Phi_B^{(i)} = P_i \delta_i e^{-\frac{P_i}{T_B}(t-\tau)} \cdot \sum_j P_j \quad \text{kernel}$$

i.e. $\frac{\partial P}{\partial t} = -\frac{1}{T_{eff}} P$

$$1/T_{eff} = |L_{eff}| \quad *$$

→ An Explicit Evaluation

$$\mathcal{L}_i = -\frac{k}{M} \frac{\partial x_i}{\partial v} + v \frac{\partial}{\partial x_i}$$

$$\Gamma = P_{eq}(b) \int_{a,b} p(a, b, t)$$

Also useful to recall:

$$P_{eq}(b) = (\text{const}) \prod_{i=1}^N \exp \left[-\frac{\Delta x_i^2}{2 \Delta_i^2} \right] \exp \left[-\frac{V_{ci}^2}{2 \Delta_{ci}^2} \right]$$

So

$$\Gamma \frac{\partial}{\partial v} = \int dx_0 v \frac{\partial}{\partial x_0}$$

$$= 0$$

≈ only $\partial/\partial v$ piece \mathcal{L}_i survives!

b includes
 $\partial x_0 \rightarrow$
relative
separation

→ should recall $\frac{\partial}{\partial V}$ acts on $P_{\text{eq}}(v, b)$

$$P = P_{\text{eq}} \frac{C}{P_{\text{eq}}}$$

$$\begin{aligned} \text{LHS} \\ f_i &= \frac{k}{M} \partial x_0 \left(\frac{\partial}{\partial V} + \frac{V}{k^2} \right) \\ &\quad - V \left(\cancel{\frac{\partial}{\partial x_0}} + \frac{k \partial x_0}{T} \right) \\ &\quad \text{sign absorbed in } f_i^2 \downarrow \end{aligned}$$

so for L_{eff} :

$$L_{\text{eff}} = \frac{1}{P_{\text{eq}}(b)} \left(\int ds \underline{P} f_i e^{-\underline{L}_b(t-s)} \underline{f_i P} \right) P_{\text{eq}}(b)$$

$$\begin{aligned} &= \frac{1}{P_{\text{eq}}(b)} \left(\int ds P_{\text{eq}}(b) \int db \left(\frac{k}{M} \partial x_0 \left(\frac{\partial}{\partial V} + \frac{V}{k^2} \right) \right. \right. \\ &\quad \left. \left. - V k \partial x_0 \right) e^{-\underline{L}_b(t-s)} \left(\frac{k \partial x_0}{M} \left(\frac{\partial}{\partial V} + \frac{V}{k^2} \right) - \frac{V k \partial x_0}{T} \right) \times \right. \\ &\quad \left. P_{\text{eq}} \int db P_{\text{eq}}(b) \right) \end{aligned}$$

$$\text{As: } \langle \delta X_0 \rangle = 0$$

$$\rho \nabla \cdot \rho = 0 \quad (\text{obv.})$$

Have:

$$f_{\text{eff}} = \left(\frac{k}{M} \right) \sum_i \left(\langle v^2 \rangle \frac{\partial^2}{\partial v^2} + \frac{\partial}{\partial v} v \right)$$

$$\Phi(z) = \int_0^\infty e^{-zt} \frac{\langle \delta X_0 \delta X_0(t) \rangle_{\text{ex}}}{\cancel{Z}} \, dt$$

$$\Phi(0) = \frac{k}{M} \Phi(z) \Big|_{z=0} \quad \rightarrow \text{Kubo!}$$

and

dim v

$$\text{if } \rho(v,t) = f_{\text{eff}} \rho(v,t)$$

$$= \frac{k}{M} \Phi(0) \left[\langle v^2 \rangle \frac{\partial^2}{\partial v^2} + \frac{\partial}{\partial v} v \right] \rho(v,t)$$

\rightsquigarrow as Markovian \leftrightarrow Fokker-Planck form!

→ Note can re-write in Fokker-Planck Form:

$$\frac{\partial P}{\partial t} = \frac{\partial^2}{\partial V^2}(DP) - \frac{\partial}{\partial V}(BV P)$$

where

$$D = \int_{-\infty}^{\infty} P_{\Sigma}(h) \frac{k^2}{M} \frac{P \delta x_0}{\delta x_0} e^{\frac{h(t-T)}{\delta x_0}} \delta x_0 P_{\Sigma}(h) d(t-T)$$

$$B = P \int_{-\infty}^{-1} P_{\Sigma}^{-1}(h) \frac{k}{M} \delta x_0 e^{\frac{h(t-T)}{\delta x_0}} \frac{\partial}{\partial \delta x_0} P_{\Sigma} d(t-T)$$

Some observations:

- $F = B$ (eg) recovered by

→ weak interaction limit $\exp[] \approx 1$
 → Markovian

→ Predictable

→ could go directly, by kinetic formalism approach

→ simplification of complex problem.

→ Major assumption:

- Φ_{eq} thermal distribution

- i.e. eq. for irrelevant

$$\rightarrow D_{\text{eff}} = \frac{k}{M} \bar{\Phi}(0) \langle v^2 \rangle$$

{ Velocity
space
diffn }

$$\bar{\Phi}(0) = \lim_{T \rightarrow \infty} \frac{\int_0^\infty dt e^{-2t} \langle \delta x_0 \delta x(t) \rangle_{\text{eff}}}{\langle \delta x(0)^2 \rangle_{\text{eff}}}$$

aka Kubo.

- Questions:

- higher order \rightarrow D_{eff} renormalized

$$\frac{k}{M} \bar{\Phi}(0) = \gamma_v \Rightarrow \gamma_v + \delta \gamma_v$$

\downarrow
friction

i.e. consider simple case

$$\bar{\Phi}(t) = \frac{\langle \delta x \delta x(t) \rangle}{\langle \delta x^2 \rangle} = e^{-\beta m t} \quad (\text{Boltz})$$

$$\text{L.O. } \delta_V = \frac{k/M}{\delta_C}$$

$$\delta^2 \delta_V \sim \frac{(k/M)^2}{\delta_C^3}$$

\rightarrow left out?

- time scale separation arbitrary \Rightarrow scalings, etc.

- adiabatic variation of fast variables

adiabatic elimination \Rightarrow adiabatic evolution

slow feedback in modulations.

etc.