

(Many degrees of freedom)

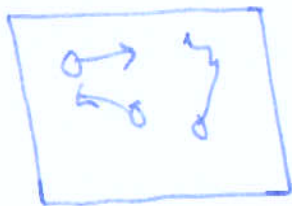
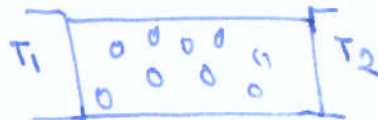
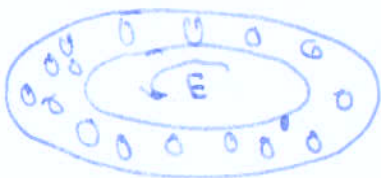
Ref: ① Kipnis and Landim: Scaling limits of interacting particle systems. Springer publication (2013).

② H. Spohn (1991): Large scale dynamics of interacting particles Springer.

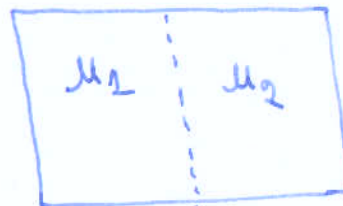
③ Eyink, Lebowitz, Spohn (1996): Hydrodynamics and fluctuations outside of local equilibrium: driven diffusive systems J. Stat. Phys. 83, 385, (1996)

Mention other references.

③ Systems with many degrees of freedom



Active system



relax.

Questions: ① What are average density, ^{Energy} ~~Temp~~ profile: $\langle \rho(x,t) \rangle$, $\langle E(x,t) \rangle$

② Average flow: $\langle J(x,t) \rangle$

③ Correlations: $\langle \rho(x,t) \rho(y,t') \rangle$; $\langle J(x,t) J(y,t') \rangle$

④ Stationary measure: $P(\rho(x))$, $P(J(x))$

⑤ Phase transition, spontaneous symmetry breaking, "collective behavior"

A description in Macroscopic (coarse-grain) scale

Hydrodynamics \rightarrow average + fluctuation

Micro \rightarrow Macro

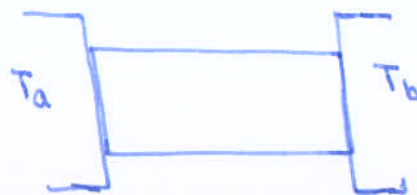
"Langevin equation for fields"

ONE-DIMENSION

Diffusive systems

"Predominant transport mechanism"

Fowler's law (1822)



large L
and T

$$\frac{\langle Q_T \rangle}{T} \approx \frac{D(T_a)}{L} \cdot (T_a - T_b)$$

\uparrow decays with system length.

Fick's law (1855)

large L
and T

$$\frac{\langle Q_T \rangle}{T} = \frac{D(P)}{L} \cdot (P_a - P_b)$$

Examples:

- ① ~~colloids~~
- ② Typically where internal dynamics is coupled with bath.
- ③ Colloids in fluids
- ④ ~~Metals~~ conductors.
- ⑤ Interacting particles
- ⑥ Active particles.
- ⑦ Traffic models

Remark: Anomalous transport in 1-d

Ex: Ideal gas
harmonic chain
anharmonic chain

$$\langle J \rangle \sim \frac{1}{L^{1-\alpha}} \quad 0 < \alpha < 1$$

Ref: Dhar. Heat transport in low-dimensional systems. Advances in Physics 2008

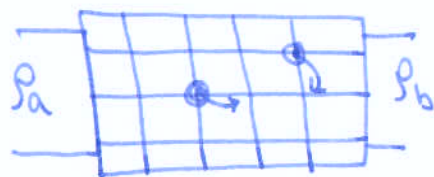
② Lepri, Livi, Politi. Thermal conduction in classical low-dimensional lattices. Physics Reports 2003.

Theoretical Models: (we shall consider)

① Interacting Langevin particles

$$\dot{x}_i(t) = F(x_i(t)) - \frac{d}{dx_i} \sum_j U(x_i - x_j) + \eta_i$$

② Particles on lattice



(a) Non-interacting particles:

(b) Simple exclusion Process:

Desoiza (2007)

Non equilibrium steady states

J. Stat. Mech. P07023

Evans, Hanney
JPA (2008)
38, R195

(c) Zero-range process:

(d) Ising model with spin exchange

$$E = -J \sum r_i n_i$$

exchange $\propto e^{-\beta E}$

* Same critical behavior of liquid-gas transition in real fluid.

(e) Heat conduction: Kipnis, Marchioro, Presutti model

$$\{e_i, e_j\} \longrightarrow \left\{ (1-\sigma)(e_i + e_j), \overset{\text{random number}}{\sigma}(e_i + e_j) \right\}$$

Ref: Heat flow in an exactly soluble model
J. Stat. Phys. 27, 85 (1982)

"Start with examples"

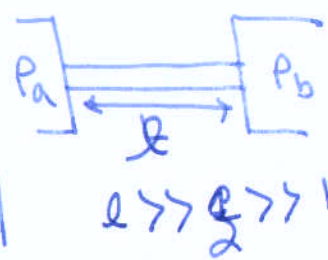
Transport coefficients: "One dimension"

flux t and l



$$\frac{\langle Q_t \rangle}{t} \approx \frac{D(P)}{L} (P_a - P_b)$$

Non-equilibrium



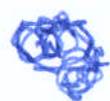
$$\frac{\langle Q_t^2 \rangle - \langle Q_t \rangle^2}{t} = \frac{\sigma(P)}{L}$$

Diffusivity and mobility

[Can verify for non-interacting particles]

Kubo

* Two are related by fluctuation-response relation.



$$\frac{\langle Q_t \rangle}{t(P_a - P_b)} \propto \frac{\langle Q_t^2 \rangle}{t}$$



local-equilibrium

* Idea is to write large-scale fluctuations ("slow modes")

$$y(x, T) \approx -D(P(x, T)) \frac{\partial P(x, T)}{\partial x} + \xi(x, T)$$

with $\langle \xi(x, T) \rangle = 0$

$$\langle \xi(x, T) \xi(x', T') \rangle = \sigma(P(x, T)) \delta(x-x') \delta(T-T')$$

~~check~~ $\langle J(x, T) \rangle = - \langle D(P(x, T)) \frac{\partial P}{\partial x} \rangle$ ~~check~~ In addition $\partial_T P = -\partial_x J$

Exercise: check: $Q_T = \frac{1}{l} \int_0^l dx \int_0^{\tilde{T}} dT J(x, T)$

① $\langle Q_T \rangle = -\frac{1}{l} \int_0^{\tilde{T}} dT \int_0^l D(P(x, T)) \frac{\partial P}{\partial x} dx \approx \frac{D(P(x, T)) \cdot \tilde{T} (P_2 - P_1)}{l}$

② $\langle Q_T^2 \rangle_c = \frac{1}{l^2} \int_0^l dx \int_0^l dx' \int_0^{\tilde{T}} dT \int_0^{\tilde{T}} dT' \sigma(P(x, T)) \cdot \delta(x-x') \delta(T-T')$

$$= \frac{1}{l^2} \int_0^l dx \int_0^{\tilde{T}} dT \sigma(P(x, T))$$

$$\approx \sigma(\bar{P}) \cdot \frac{\tilde{T}}{l}$$

Argument: $l \gg$ correlation length for fast modes continuity

* In addition: local equilibrium

$\Rightarrow D(P)$ and $\sigma(P)$ are related,

SKIP

Rescaled coordinates: coarse-graining

$x = \frac{x}{\ell}$; $t = \frac{T}{\ell^2}$ (Diffusive scaling only fluctuation)



$\odot P(x, T) \approx q(x, t) + \mathcal{O}(\frac{1}{\ell})$

$$\left[* \frac{n_{\ell}}{\ell} = \frac{1}{\ell} \int_{-1/2}^{1/2} dx P(x, T) \approx \int_{-1/2}^{1/2} q(x, t) dx \right]$$

$$\approx q(x, t) + \mathcal{O}(\frac{1}{\ell})$$

$$\lim_{\ell \rightarrow \infty} \int_{-1/2}^{1/2} dx \frac{P(x, T)}{\ell} = q(x, t)$$

$\odot \mathcal{J}(x, T) \approx \frac{1}{\ell} j(x, t) + \mathcal{O}(\frac{1}{\ell^2})$

because current decays with system length.

Using together:

~~original~~ $\mathcal{J}(x, T) = -D(P(x, T)) \partial_x P(x, T) + \xi(x, T)$

$$\Rightarrow \dot{q}(x, t) = -D(q(x, t)) \partial_x q(x, t) + \underbrace{\ell \xi(x, T)}_{\mathcal{J}(x, T)} \Rightarrow \xi(x, T) = \frac{1}{\ell} \mathcal{J}(x, T)$$

$\bullet \langle \mathcal{J}(x, t) \rangle = \ell \langle \xi(x, T) \rangle = 0$

$$\bullet \langle \mathcal{J}(x, t) \mathcal{J}(y, t') \rangle = \ell^2 \langle \xi(x, T) \xi(y, T') \rangle$$

$$= \ell^2 \sigma(P(x, T)) \delta(x-y) \delta(T-T')$$

$$= \frac{\ell^2}{\ell^3} \cdot \sigma(q(x, t)) \delta(x-y) \delta(t-t')$$

* Correlation of fast modes which as relaxed to local equilibrium"

$$\Rightarrow \langle \mathcal{J}(x, t) \mathcal{J}(y, t') \rangle = \frac{1}{\ell} \sigma(q(x, t)) \delta(x-x') \delta(t-t')$$

\bullet Continuity: $\partial_T P = -\partial_x \mathcal{J}$

$$\Rightarrow \partial_t q = -\partial_x j$$

\otimes [noise strength decrease] $\sim \frac{1}{\ell}$ weak noise

Together:

$$\partial_t q = \partial_x [D(q) \partial_x q - j] \text{ with } \langle \mathcal{J} \mathcal{J} \rangle = \frac{\sigma}{\ell} \delta(x-x') \delta(t-t')$$

⊙ A phenomenological approach.

Can be rigorously proved for just few examples.

Example 1: ~~non-interacting~~ ^{Brownian} particles

Interacting Brownian particles.


$$\frac{dx_i(t)}{dt} = F(x_i) - \underbrace{\partial_{x_i} \sum_j V(x_i - x_j)}_{\text{we take zero}} + \xi_i(t)$$

Ref: D. Deam: Phys. A 29, 1996, L613

$$\frac{dx_i(t)}{dt} = \xi_i(t) \quad \text{with} \quad \langle \xi_i(t) \xi_j(t') \rangle = 2D \delta_{ij} \delta(t-t') \quad \text{Take } D=1$$

Density: $\rho(x, T) = \sum_i \delta(x - x_i(T))$ not smooth

$= \sum_i \rho_i(x, T)$



• Choose a test function: ~~...~~

⊙ $f(x_i) = \int dx \rho_i(x, T) f(x)$

$$\Rightarrow \frac{df}{dt} = \int dx \frac{\partial \rho_i(x, T)}{\partial T} \cdot f(x)$$

⊙ ~~...~~ ^{Ito choice} $\frac{df(x_i(t))}{dt} = f'(x_i(t)) \xi_i(t) + \frac{1}{2} f''(x_i(t)) \cdot 2D$

$$\begin{aligned} \Rightarrow \frac{df}{dt} &= \int dx \delta(x - x_i(t)) [f'(x) \xi_i(t) + D f''(x)] \\ &= \int dx f(x) \left[-\frac{\partial}{\partial x} (\delta(x - x_i(t)) \xi_i(t)) + \frac{\partial^2}{\partial x^2} \delta(x - x_i(t)) D \right] \end{aligned}$$

$$\Rightarrow \frac{\partial P_i(x,T)}{\partial T} = \frac{\partial^2}{\partial x^2} [D P_i(x,T)] - \frac{\partial}{\partial x} [\xi_i(T) P_i(x,T)]$$

Sum \sum_i

$$\Rightarrow \partial_T P(x,T) = \partial_x^2 [D P(x,T)] - \partial_x \left[\underbrace{\sum_i \xi_i(T) P_i(x,T)}_{Q(x,T)} \right]$$

$$Q(x,T) = \sum_i \xi_i(T) \delta(x - x_i(T))$$

$$\langle Q(x,T) \rangle = 0$$

$$\begin{aligned} \langle Q(x,T) Q(y,T') \rangle &= \sum_i \sum_j \langle \xi_i(T) \xi_j(T') \rangle \delta(x - x_i(T)) \delta(y - x_j(T')) \\ &= 2D \cdot \sum_i \sum_j \delta_{ij} \delta(T - T') \delta(x - x_i(T)) \delta(y - x_j(T')) \\ &= 2D \cdot \delta(T - T') \delta(x - y) \sum_i \delta(x - x_i(T)) \\ &= 2D P(x,T) \delta(T - T') \delta(x - y) \end{aligned}$$

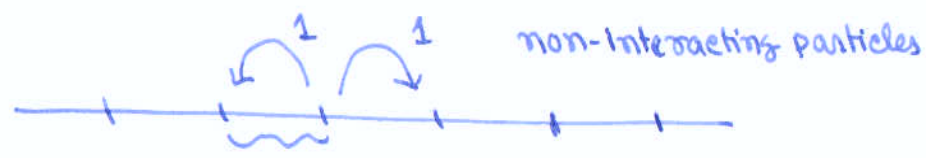
$$\Rightarrow \partial_T P(x,T) = \partial_x^2 [D P(x,T)] - \partial_x Q(x,T) = -\partial_x [-D \partial_x P + Q(x,T)]$$

where $\langle Q(x,T) Q(y,T') \rangle = \frac{2D P(x,T) \cdot \delta(T - T') \delta(x - y)}{\sigma(P)}$ → take D=1

Remark: ① $\frac{2D}{\sigma} = \frac{1}{P} = + \frac{d^2}{dP^2} [+ P \ln P - P]$

② Can be generalized for inter-particle interactions.

Example 2: On lattice



Conservation of pty

$$n_i(\tau+dt) - n_i(\tau) = \mathcal{J}_i(\tau) dt - \mathcal{J}_{i+1}(\tau) dt$$

net current between site i, i+1

$$\Rightarrow \frac{d n_i(\tau)}{d\tau} = \mathcal{J}_i(\tau) - \mathcal{J}_{i+1}(\tau)$$

Average

$$\langle \mathcal{J}_i(\tau) \rangle = \lim_{dt \rightarrow 0} \frac{\langle \# \text{ Ptl} \rangle \text{ flow in } dt}{dt}$$

$$= \lim_{dt \rightarrow 0} \frac{\langle n_i(\tau) \rangle dt - \langle n_{i+1}(\tau) \rangle dt}{dt}$$

$$\langle \mathcal{J}_i(\tau) \rangle = \langle n_i(\tau) \rangle - \langle n_{i+1}(\tau) \rangle$$

~~Fluctuation~~

Fluctuation: $\mathcal{J}_i(\tau) = \underbrace{n_i(\tau) - n_{i+1}(\tau)}_{\text{Fick's law}} + \gamma_i(\tau)$

One can show:

$$\langle \gamma_i(\tau) \rangle = 0$$

$$\langle \gamma_i(\tau) \gamma_j(\tau') \rangle = [\langle n_i(\tau) \rangle + \langle n_{i+1}(\tau) \rangle] \delta_{ij} \delta(\tau - \tau')$$

Ref: Sachdev, Derrida 2016, Jstat mech.

Rescaling: $x = \frac{i}{L} ; t = \frac{\tau}{L^2} ; n_i(\tau) \Rightarrow \rho(x,t)$

$$\mathcal{J}_i(\tau) \Rightarrow \frac{1}{L} j(x,t)$$

$$\gamma_i(\tau) \Rightarrow \frac{1}{L} \eta(x,t)$$

$$\Rightarrow \dot{j}(x,t) = -\partial_x \rho(x,t) + \eta(x,t)$$

Covariance:

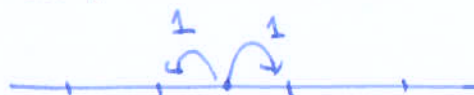
$$\langle \eta(x,t) \eta(y,t') \rangle = \frac{\sigma(a)}{2} \cdot \delta(x-y) \delta(t-t')$$

where $\sigma(a) = 2a$

* Same as Langevin equation case *

* D is absorbed in the time scale *

Similar proof for symmetric exclusion process:



$$\bullet \frac{d\eta_i(\tau)}{d\tau} = J_i(\tau) - J_{i+1}(\tau)$$

$$\bullet \langle \eta_i(\tau) \eta_j(\tau) \rangle = \left[\langle \eta_i^2 \rangle + \langle \eta_i(\tau) \rangle - 2 \langle \eta_i(\tau) \eta_{i+1}(\tau) \rangle \right] \delta_{ij} \delta(\tau - \tau')$$

$$\Rightarrow \langle \eta(x,t) \eta(y,t') \rangle = 2 P(x,t) (1 - P(x,t)) \cdot \delta(x-y) \delta(t-t')$$

For other systems no rigorous proof

Starting point:

$$\partial_t \rho(x,t) = -\partial_x J(x,t)$$

$$J(x,t) = -D(\rho) \partial_x \rho(x,t) + \eta(x,t)$$

$$\langle \eta(x,t) \eta(y,t') \rangle = \frac{\sigma(\rho)}{L} \delta(x-y) \delta(t-t')$$

We want to calculate:

$$\langle \rho(x,t) \rangle, \langle \rho(x,t) \rho(y,t) \rangle e^{-\lambda \phi(\rho(x,t))}$$

$$P(\rho(x)) \sim e^{-\lambda \phi(\rho(x))}$$

* also applies for other fields, energy density, Angle!

In the hydrodynamic scale microscopic details are in the two transport coefficients $D(\rho)$ and $\sigma(\rho)$.

How does one calculate them?



Approach 1:

$$\frac{\langle Q_t \rangle}{t} = \frac{D(\rho)}{l} (P_a - P_b) \quad \text{stationary state}$$

$$\frac{\langle Q_t^2 \rangle - \langle Q_t \rangle^2}{t} = \frac{\sigma(\rho)}{l}$$

Approach 2:

~~fluctuation~~ fluctuation dissipation relation

Difficulty is usually $\sigma(\rho)$

$$\frac{2D(\rho)}{\sigma(\rho)} = \frac{f''(\rho)}{kT} = \frac{1}{\rho^2 \chi \cdot kT} \leftarrow \text{set } kT = 1$$

where $f(\rho) = \frac{1}{L} \log Z(N,L)$

Canonical free energy (Helmholtz)

Proof: later

Examples: Non interacting particles:



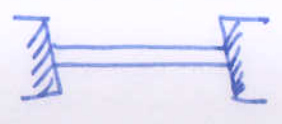
$$\langle J_i^{(+)} \rangle = \langle n_i^{(+)} \rangle - \langle n_{i+1}^{(+)} \rangle$$

$$\rightarrow \sum_i \langle Q_T \rangle = \int_0^T dt \sum_i \langle J_i^{(+)} \rangle \approx T \cdot (P_a - P_b)$$

(stationary state)

$$\langle Q_T \rangle = \frac{T}{L} (P_a - P_b) \Rightarrow D = 1$$

• What is f(P):



all configurations are equally probable

$$\Rightarrow Z(N, L) = \frac{L^N}{N!}$$

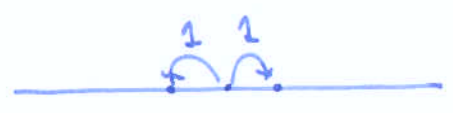
free energy $f = - \ln Z = - [N \ln L - \ln N!]$

$$\approx L \{ P \ln P - P \}$$

$$\Rightarrow f(P) = P \ln P - P$$

$$\Rightarrow \sigma = \frac{2D}{f''(P)} = 2P$$

② Symmetric exclusion process:



$$\begin{aligned} \text{A) } \langle J_i \rangle &= \langle n_i (1 - n_{i+1}) \rangle - \langle (1 - n_i) n_{i+1} \rangle \\ &= \langle n_i \rangle - \langle n_{i+1} \rangle \end{aligned}$$

$$\rightarrow D = 1$$

$$\text{B) } Z(N, L) = \binom{L}{N} \Rightarrow f(P) = P \ln P + (1-P) \ln(1-P)$$

$$\text{C) } \sigma = 2P(1-P)$$

Ex 3 % Exercise:

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next nearest neighbors ~~exclusion~~ exclusion.

$$\langle J_i \rangle = \langle n_i (1 - n_{i+1}) (1 - n_{i+2}) \rangle - \langle n_{i+1} (1 - n_i) (1 - n_{i-1}) \rangle$$

$$D(p) = \frac{1}{(1-p)^2}$$

$$f(p) = p \ln p + (1-2p) \ln(1-2p) - (1-p) \ln(1-p)$$

$$\sigma(p) = \frac{2p(1-2p)}{1-p}$$

Ex 4 % Zero range process

Ex 5 % Boundary driven Ising model

Ex 6 % KMP model

$$D(p) = 1$$

$$\sigma(p) = 2p^2$$

Variational formulation and proof of $\frac{\partial D}{\partial \sigma} = f''(P)$

Path measure:

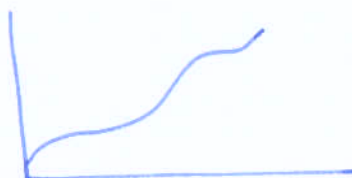
$$\text{Prob}(\text{Traj}) \sim e^{-L \int dt \int dx \frac{[\dot{\varphi} + D(\varphi) \partial_x \varphi]^2}{2\sigma(\varphi)}}$$

$$\partial_t P = -\partial_x J$$

$$J = -D \partial_x P + Q$$

Remark: Ito-stratonovich choice do not matter

Hamiltonian formulation:



$$P[\varphi(x,T) = \delta(x), \varphi(x,0) = \rho(x)]$$

$$= \int_{P(\varphi)} \omega[\varphi, J] e^{-L \int_0^T dt \int_0^1 dx \frac{[\dot{\varphi} + D(\varphi) \partial_x \varphi]^2}{2\sigma(\varphi)}} \delta(\partial_t \varphi + \partial_x J)$$

$$= \int \omega[\varphi, J, P] e^{-L \int_0^T dt \int_0^1 dx \frac{(\dot{\varphi} + D(\varphi) \partial_x \varphi)^2}{2\sigma(\varphi)}} - L \int_0^T dt \int_0^1 dx P(\partial_t \varphi + \partial_x J)$$

$$= \int \omega[\varphi, J, P] e^{-L \int dt \int dx \left\{ P \dot{\varphi} - J \cdot \partial_x P + \frac{(\dot{\varphi} + D \partial_x \varphi)^2}{2\sigma(\varphi)} \right\}}$$

~~assuming~~ Using $P(0) = 0$ } no condition
 $P(1) = 0$ } at the boundary

$$= \int \omega[\varphi, P] e^{-L \int dt \int dx P \dot{\varphi}} \int \omega[J] \cdot e^{-L \int dt \int dx \left\{ \frac{[J + (D \dot{\varphi} - \sigma P)']^2}{2\sigma} \right\}}$$

$$e^{-L \int dt \int dx \frac{D^2 (\partial_x \varphi)^2 - (D \dot{\varphi} - \sigma P)^2}{2\sigma}}$$

$$\approx \int \mathcal{D}[q, p] e^{-L \int dt \int dx \left\{ p \dot{q} - \frac{\sigma^2 p^2}{2} - 2\sigma D p q' \right\}}$$

$$\approx \int \mathcal{D}[p, q] e^{-L \int dt \int dx \left\{ p \dot{q} - \underbrace{\left[\frac{\sigma^2}{2} (\partial_x p)^2 - D(q) \partial_x q \cdot \partial_x p \right]}_{H[p, q]} \right\}}$$

$$\text{Prob}[\sigma(x), p_0(x)] \approx \int \mathcal{D}[p, q] e^{-L S[p, q]}$$

$$\text{with } S[p, q] = \int dt \left\{ \int dx p \dot{q} - H[p, q] \right\}$$

where $H[p, q] = \frac{\sigma^2(q)}{2} (\partial_x p)^2 - D(q) \cdot (\partial_x q) \cdot (\partial_x p)$

Equilibrium:



$$P[q(x) = \sigma(x)] \sim e^{-L \phi[\sigma(x)]}$$

Stat Mech where $\phi[\sigma] = \int_0^1 dx \left\{ f(\sigma) - f(P) - f'(P) [\sigma - P] \right\}$

Derivation using the Action formulation:

Similar approach as used for Langevin equation

$$P[\sigma(x)] = P[q(x, 0) = \sigma(x); q(x, \infty) = P]$$

$$= \int_{q(x, 0) = \sigma(x)}^{q(x, \infty) = P} \mathcal{D}[p, q] e^{-L S[p, q]}$$

$$\Rightarrow \boxed{\phi[\sigma(x)] = \min S[p, q]}$$

Variational calculus:

$$S[P, q] = \int_{-\infty}^0 dt \left\{ \int_0^1 dx p \dot{q} - H[P, q] \right\}$$

$\hookrightarrow H = \int dx \{ \dots \}$

δK/P

$$\Rightarrow \delta S[P, q] = \int_{-\infty}^0 dt \left\{ \int_0^1 dx \left(\dot{q} - \frac{\delta H}{\delta p} \right) \delta p(x, t) + \left(\dot{p} + \frac{\delta H}{\delta q} \right) \delta q \right\}$$

$$+ \int_0^1 dx p(x, T) \underbrace{\delta q(x, T)}_0$$

$$- \int_0^1 dx p(x, 0) \underbrace{\delta q(x, 0)}_0$$

Least Action paths

$$\dot{q} = \frac{\delta H}{\delta p} = - \partial_x \left[\frac{\sigma(q)}{2} \cdot \partial_x p - D(q) \partial_x q \right]$$

$$\Rightarrow \partial_t q - \partial_x (D(q) \partial_x q) = - \partial_x (\sigma(q) \partial_x p)$$

$$\dot{p} = - \frac{\delta H}{\delta q} = - \left\{ \frac{\sigma'(q)}{2} (\partial_x p)^2 - D'(q) \cdot \partial_x q \cdot \partial_x p + \partial_x (D(q) \partial_x p) \right\}$$

$$\Rightarrow \partial_t p = - \frac{\sigma'(q)}{2} (\partial_x p)^2 + \partial_x D(q) \cdot \partial_x p - \partial_x (D(q) \partial_x p)$$

$$\Rightarrow \partial_t p + D(q) \partial_x^2 p = - \frac{\sigma'(q)}{2} (\partial_x p)^2$$

With boundary conditions: $q(x, 0) = q(x)$ and $q(x, -\infty) = p$

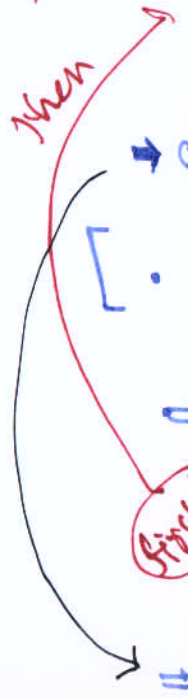
In addition:

$$\begin{cases} q(0, t) = p \text{ and } q(1, t) = p \\ p(0, t) = 0 \text{ and } p(1, t) = 0 \end{cases}$$

Solution:

$$P(x,t) = \int_p^{q(x,t)} ds \frac{2D(s)}{\sigma(s)} \quad (\text{ansatz})$$

* why lower limit?
because we want $p=0$ at $x=0,1$



→ Check one basis to solve only one equation.

$$\left[\cdot \partial_x P = \frac{2D}{\sigma} \cdot \partial_x q \Rightarrow \boxed{\sigma \partial_x P = 2D \partial_x q} \right]$$

$$\Rightarrow \partial_t q = \partial_x (D \partial_x q) - \partial_x (\sigma \partial_x P) \quad \leftarrow 2 \partial_x (D \partial_x q)$$

$$\boxed{\partial_t q = - \partial_x (D \partial_x q)}$$

time reversed equation.

$$\Rightarrow \partial_t P = \frac{2D}{\sigma} \cdot \partial_t q = - \frac{2D}{\sigma} \partial_x [D \partial_x q]$$

$$= - \frac{2D}{\sigma} \cdot \frac{1}{2} \partial_x (\sigma \partial_x P)$$

$$= - D \partial_{xx} P - \frac{D}{\sigma} \sigma' \partial_x q \cdot \partial_x P$$

$$= - D \partial_{xx} P - \frac{\sigma'}{\sigma} \left[\frac{\sigma}{2} \partial_x P \right] \cdot \partial_x P$$

$$\Rightarrow \boxed{\partial_t P = - D \partial_{xx} P - \frac{\sigma'}{2} (\partial_x P)^2}$$

automatically satisfied.]

⊙ Boundary condition:

$$\partial_t q = - \partial_x (D \partial_x q)$$

with $q(x, -\infty) = p_0$ and $q(x, 0) = \gamma(x)$

• check: $H[P, q] = \int \left[\frac{\sigma(q)}{2} (\partial_x P)^2 - \underbrace{D(q) \cdot \partial_x q \cdot \partial_x P}_{\frac{\sigma'}{2} \partial_x P} \right] dx$
 $= 0$

Then

$$\Phi[\sigma(x)] = \int_{-\infty}^0 dt \int_0^1 dx p \dot{q}$$

(6)

lets write

$$\frac{2D(s)}{\sigma(s)} = f_1''(s) \Rightarrow p(x,t) = \int_P^{q(x,t)} ds f_1''(s) \\ = f_1'(q(x,t)) - f_1'(P)$$

Substitute

$$\Phi[\sigma(x)] = \int_0^1 dx \int_{-\infty}^0 dt [f_1'(q) - f_1'(P)] \dot{q} \\ = \int_0^1 dx \int_{-\infty}^0 dt \left[\frac{d}{dt} f_1(q(x,t)) - f_1'(P) \cdot \dot{q} \right] \\ = \int_0^1 dx \cdot [f_1(\sigma(x)) - f_1(P) - f_1'(P)(\sigma - P)]$$

Compare with result from ~~stat~~ equilibrium statistical mechanics
 $\Rightarrow f(s) = \text{equilibrium free energy.}$

$$\Rightarrow \boxed{\frac{2D(s)}{\sigma(s)} = f''(s)}$$

Q: Why does this break down for non-equilibrium?

$P \leftrightarrow P(x)$ solution does not work

* Only case it works \equiv Non-interacting particles.

Non-equilibrium, Non-interacting Particles

(62)



$$D(\varphi) = 1$$

$$\sigma(\varphi) = 2\varphi$$

$$\bar{P}(x) = P_a(1-x) + P_b x$$

(Solve: $\frac{\partial P}{\partial t} = \partial_{xx} P$)

with $P(0,t) = P_a; P(1,t) = P_b$)

To show:

$$P(\sigma(x)) \approx e^{-L F[\sigma(x)]}$$

$$F[\sigma(x)] = \int_0^1 dx \left\{ \sigma(x) \ln \frac{\sigma(x)}{P(x)} - \sigma(x) + P(x) \right\}$$

Solution of the Hamilton's equation

~~$$\partial_t P + \partial_{xx} P = -(\partial_x P)^2$$~~

$$\partial_t \varphi - \partial_{xx} \varphi = -2 \partial_x (\varphi \partial_x P)$$

Solution:

$$P(x,t) = \int_{P(x)}^{\varphi(x,t)} \frac{1}{s} ds = \ln \varphi(x,t) - \ln P(x)$$



[leave as an exercise to show

$$\partial_t \varphi = -\partial_{xx} \varphi + 2 \partial_x \left[\varphi \cdot \frac{\partial_x P}{P} \right]$$

$\partial_t P$ eqⁿ is automatically satisfied!

check: $\partial_x P = \frac{\partial_x \varphi}{\varphi} - \frac{\partial_x P}{P} \Rightarrow \varphi \partial_x P = \partial_x \varphi - \varphi \cdot \frac{\partial_x P}{P}$

$$\begin{aligned} \odot \partial_t q &= \partial_{xx} q - 2 \partial_x (q \partial_x p) \\ &= \partial_{xx} q - 2 \partial_{xx} q + 2 \partial_x \left(q \frac{\partial_x p}{p} \right) \end{aligned}$$

$$\Rightarrow \boxed{\partial_t q + \partial_{xx} q = 2 \partial_x \left(q \frac{\partial_x p}{p} \right)}$$

$$\begin{aligned} \odot \partial_t p &= \frac{\partial_t q}{q} = \frac{1}{q} \left\{ -\partial_{xx} q + 2 \partial_x \left(q \frac{\partial_x p}{p} \right) \right\} \\ &= \frac{1}{q} \left\{ -\partial_x \left(q \partial_x p \right) + \partial_x \left(q \frac{\partial_x p}{p} \right) \right. \\ &\quad \left. + 2 \partial_x \left(q \frac{\partial_x p}{p} \right) \right\} \end{aligned}$$

$$\left. \begin{array}{l} \text{use} \\ q \partial_x p = \partial_x q - q \frac{\partial_x p}{p} \end{array} \right\}$$

$$= -\partial_{xx} p - \frac{1}{q} (\partial_x q) (\partial_x p) + \frac{1}{q} \partial_x \left(q \frac{\partial_x p}{p} \right)$$

$$\Rightarrow \partial_t p + \partial_{xx} p = -\frac{1}{q} (\partial_x q) (\partial_x p) + \frac{\partial_x q \cdot \partial_x p}{q} - \frac{q \cdot (\partial_x p)^2}{q p^2} + \frac{q \cdot (\partial_x p)^2}{q p}$$

$$= -\frac{1}{q} \partial_x p \cdot \left(q \partial_x p + q \frac{\partial_x p}{p} \right) + \frac{\partial_x q \cdot \partial_x p}{q p} - \frac{(\partial_x p)^2}{p^2}$$

$$= -(\partial_x p)^2 - (\partial_x p) \cdot \frac{\partial_x p}{p} + \frac{\partial_x q \cdot \partial_x p}{q p} - \frac{(\partial_x p)^2}{p^2}$$

$$\Rightarrow \left[-\frac{1}{q} \cdot \frac{\partial_x q \cdot \partial_x p}{p} + \left(\frac{\partial_x p}{p} \right)^2 + \frac{\partial_x q \cdot \partial_x p}{q p} - \frac{(\partial_x p)^2}{p^2} = 0 \right]$$

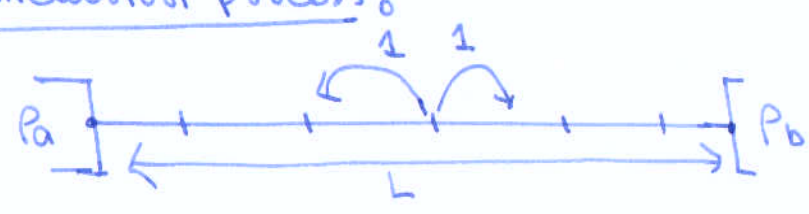
Then follow the earlier procedure.

$$\begin{aligned} \mathbb{P}[\sigma(x)] &= \int dx \left\{ f(\sigma) - f(p) - f'(p) \cdot (\sigma - p) \right\} \text{ with } f(\sigma) = \sigma \ln \sigma - \sigma \\ &= \int dx \left\{ \sigma \ln \frac{\sigma}{p} - \sigma + p \right\} \end{aligned}$$

\otimes local functional

out-side equilibrium

① Symmetric exclusion process



$$D(p) = 1 ; \sigma(p) = 2p(1-p)$$

Average profile:

$$P(x) = P_a(1-x) + x P_b$$

Profile at stationary state:

$$P(\sigma(x)) \propto e^{-L \Phi[\sigma(x)]}$$

Equilibrium: $P_a = P_b = p$

$$\Phi[\sigma(x)] = \int_0^1 dx \left\{ \sigma(x) \ln \frac{\sigma(x)}{p} + (1-\sigma(x)) \ln \frac{1-\sigma(x)}{1-p} \right\}$$

$$f(p) = p \ln p + (1-p) \ln(1-p)$$

"local functional"

Non-equilibrium: $P_a \neq P_b$

$$\Phi[\sigma(x)] = \min_{F(x)} \int_0^1 dx \left\{ \sigma(x) \cdot \ln \frac{\sigma(x)}{F(x)} + (1-\sigma(x)) \ln \frac{1-\sigma(x)}{1-F(x)} + \ln \frac{\partial_x F(x)}{P_b - P_a} \right\}$$

where $F(0) = P_a$ and $F(1) = P_b$

~~Comment: $P_a = P_b \Rightarrow F(x) = p \Rightarrow$ one gets back equilibrium.~~

Comment:

$$F(x) + \frac{F(x)(1-F(x))}{(\partial_x F)^2} \cdot (\partial_x^2 F) = \sigma(x)$$

Comment 2: For small $P_a - P_b$ one solves $F(x)$ perturbatively and get

$$F(x) \approx P(x) - \frac{(P_a - P_b)^2}{P_a(1 - P_a)} \left[(1-x) \int_0^x dy (\sigma(y) - P(y)) + x \int_x^1 dy (1-y) (\sigma(y) - P(y)) \right] + \dots$$

and one gets

$$\begin{aligned} \Rightarrow \Phi[\sigma(x)] = & \int_0^1 dx \left\{ \sigma(x) \ln \frac{\sigma(x)}{P(x)} + (1-\sigma(x)) \ln \frac{1-\sigma(x)}{1-P(x)} \right\} \\ & + \frac{(P_a - P_b)^2}{[P_a(1 - P_a)]^2} \int_0^1 dx \int_0^1 dy [x(1-y) (\sigma(x) - P(x)) (\sigma(y) - P(y))] \\ & + (P_a - P_b)^3 \# \dots \end{aligned}$$

⊗ Non-local function: \Rightarrow correlations are long-ranged.

How to solve this

⊗ ~~Microscopic~~ Microscopic solution: Matrix product Ansatz

① Derrida, Lebowitz, Speer: PRL 87, 150601 (2001)

② " " " : JSP 107, 599 (2002)

⊗ Using variational formulation

① Bertini, De Sole, Gabrielli, Jona-Lasinio, Landim

PRL 87, 040601 (2001)

JSP 107, 635 (2002)

② Tailleur, Kurchan, Leconte PRL 99, 150602 (2007)

JPA 41, R05001 (2008)