

Comonical transformation

Equilibrium problem revisited:

$$P[q(x,0) = \delta(x); q(x,\infty) = P] = \int \mathcal{D}[P,q] e^{-L S[P,q]}$$

$$S[P,q] = \int_{-\infty}^{\infty} dt \left\{ \int dx \dot{q}^2 - H[P,q] \right\}$$

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

$$= \frac{\sigma(q)}{2} (\partial_x P) \cdot \left\{ \partial_x q - \frac{2D(q)}{\sigma(q)} \partial_x q \right\}$$

Comonical transformation

$$P(x,t) = -\hat{P}(x,t) + \int_P^{\hat{q}(x,t)} ds \frac{2D(s)}{\sigma(s)}$$

$$q(x,t) = \hat{q}(x,t)$$

$$\textcircled{1} \int_{-\infty}^{\infty} dt \int_0^1 dx P \dot{q} = \int_0^1 dx \int_{-\infty}^{\infty} dt \cdot \frac{d\hat{q}(x,t)}{dt} \left[\int_P^{\hat{q}(x,t)} ds \frac{2D(s)}{\sigma(s)} \right]$$

$$= \int_0^1 dx \int_{-\infty}^{\infty} dt \cdot \hat{q}(x,t) \left[\int_P^{\hat{q}(x,t)} ds \frac{2D(s)}{\sigma(s)} \right] - \int_{-\infty}^{\infty} dt \int_0^1 dx \hat{P}(x,t) \frac{d\hat{q}(x,t)}{dt}$$

$$= - \int_0^1 dx \int_{-\infty}^{\infty} dt \frac{d\hat{q}(x,t)}{dt} \left[\int_P^{\hat{q}(x,t)} ds \frac{2D(s)}{\sigma(s)} \right] + \int_{-\infty}^{\infty} dt \int_0^1 dx \hat{P}(x,t) \frac{d\hat{q}(x,t)}{dt}$$

$$H[P, q] = \frac{\sigma(\hat{q}(x, t))}{2} \cdot \left[\partial_x \hat{P}(x, t) - \frac{2D(\hat{q})}{\sigma(\hat{q})} \cdot \partial_x \hat{q} \right] \partial_x \hat{P}$$

New Action

$$S[P, q] \rightarrow \hat{S}[\hat{P}, \hat{q}]$$

$$= - \int_0^1 dx \int_0^\infty dt \cdot \dot{\hat{q}} \left[\int_P^{\hat{q}} ds \frac{2D(s)}{\sigma(s)} \right]$$

$$+ \int_0^\infty dt \int_0^1 dx \left\{ \hat{P} \dot{\hat{q}} - \underbrace{H[P, q]}_{\text{Same as H}} \right\}$$

$$\hat{S}[\hat{P}, \hat{q}]$$

First term:

$$- \int_0^1 dx \int_0^\infty dt \dot{\hat{q}} \left[\psi'(\hat{q}(x, t)) - \psi'(P) \right]$$

Boundary: $\hat{q}(x, 0) = \sigma(x)$
 $\hat{q}(x, \infty) = P$
 $\hat{P} = 0$ at boundary; $\hat{q} = P$ at boundary

$$= - \int_0^1 dx \left\{ \psi'(P) - \psi'(\sigma(x)) - \psi'(P)(P - \sigma(x)) \right\}$$

$$= \phi[\sigma(x)]$$

The new Action:

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

Downhill $\rightarrow \hat{S}_{min} = 0$

Explicit solution for SEP using Canonical Transformation. (69)

Introductory example: Equilibrium of general system.

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

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

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

$$= \frac{\sigma(q)}{2} \cdot \partial_x P \cdot \left\{ \partial_x P - \frac{2D(q)}{\sigma(q)} \cdot \partial_x q \right\}$$

Change of variable:

$$P(x,t) = \hat{P}(x,t) + \int_P^{q(x,t)} ds \frac{2D(s)}{\sigma(s)}$$

$$q(x,t) = q(x,t)$$

The Action

$$S[P,q] = \int_{-\infty}^0 dt \int_0^1 dx \left[\int_P^{q(x,t)} ds \frac{2D(s)}{\sigma(s)} \right] \dot{q} + \int_{-\infty}^0 dt \left\{ \int_0^1 dx \hat{P} \cdot \dot{q} - \hat{H}[\hat{P},q] \right\}$$

where $\hat{H}[\hat{P},q] = \frac{\sigma(q)}{2} \left[\partial_x \hat{P} + \frac{2D(q)}{\sigma(q)} \cdot \partial_x q \right] \cdot \partial_x \hat{P}$

(Different if $P=P(x)$
(and hence fails))

• \Rightarrow [Canonical transformation]

$$P(\sigma(x)) = e^{-L \int_{-\infty}^0 dt \int_0^1 dx \left[\int_P^{q(x,t)} ds \frac{2D(s)}{\sigma(s)} \right] \dot{q}}$$

$$\int \mathcal{Q}[\hat{P}, q] e^{-L \hat{S}[\hat{P}, q]}$$

• ~~Use the identity~~

• Use $\frac{2D(s)}{\sigma(s)} = f''(s)$

$$\Rightarrow \int dx \int_{-\infty}^0 dt \int_P^{q(x,t)} ds f''(s) \cdot \dot{q} = \int dx \int_{-\infty}^0 dt \dot{q} [f'(q) - f'(P)]$$

$$= \int dx [f(q) - f(P) - f'(P)(q-P)]$$

$$= \Phi[\sigma(x)]$$

• show $\boxed{\min \hat{S}[\hat{P}, q] = 0}$

$$\hat{H}[\hat{P}, q] = \frac{\sigma(q)}{2} (\partial_x \hat{P})^2 + D(q) \cdot \partial_x q \cdot (\partial_x \hat{P})$$

$$\Rightarrow \dot{q} = \frac{\delta \hat{H}}{\delta \hat{P}} = -\partial_x \left\{ \sigma(q) (\partial_x \hat{P}) + D(q) \partial_x q \right\}$$

$$\Rightarrow \boxed{\dot{q} + \partial_x (D(q) \partial_x q) = -\partial_x (\sigma(q) \partial_x \hat{P})}$$

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

$$\Rightarrow \boxed{\hat{P} - \partial_x (D(q) \partial_x \hat{P}) = -\frac{\sigma'(q)}{2} (\partial_x \hat{P})^2}$$

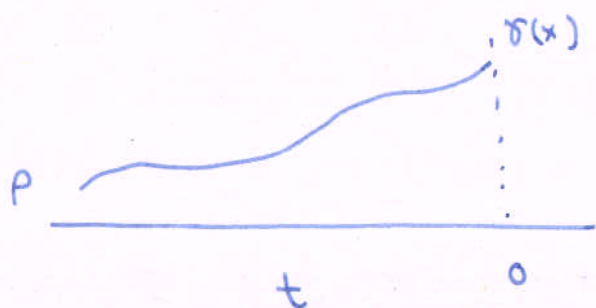
Boundary conditions:

$$\begin{array}{l|l} q(x, -\infty) = \rho & q(0, t) = P \\ q(x, 0) = \delta(x) & q(1, t) = P \end{array} \quad \left| \quad \begin{array}{l} \hat{P}(0, t) = 0 = \hat{P}(1, t) \end{array} \right.$$

Solution: "Downhill works"

$$\hat{P}(x, t) = 0$$

$$\hat{q}(x, t) = -\partial_x (D(q) \partial_x q)$$



$$\Rightarrow \hat{S}_{\min} = 0$$

$$q_R(x, t) = q_L(x, -t)$$

$$\hat{q}_R = \partial_x (D(q_R) \partial_x q_R)$$

Symmetric simple exclusion process

* we take similar strategy*

* Find a canonical transformation

such that downhill solution is the least action path*

Formulation: $D=1$; $\sigma = 2q(1-q)$

$$H[P, q] = \frac{\sigma(q)}{2} \cdot \partial_x P \cdot \left[\partial_x P - \frac{\partial_x q}{q(1-q)} \right] = \frac{\sigma(q)}{2} \cdot \partial_x P \cdot \partial_x \left[P - \ln \frac{q}{1-q} \right]$$

$$\neq \frac{\sigma(q)}{2} \cdot \partial_x P \cdot \partial_x \left[P - \frac{\sigma V}{8q} \right] \quad \text{where } V(q) = \int dx \left[A \ln q + (1-q) \ln(1-q) \right]$$

First attempt: Try as before

$$p = \hat{p} + \int_{P(x)}^q \frac{1}{s(1-s)} ds = \hat{p} + \ln \frac{q}{1-q} - \ln \frac{P}{1-P}$$

$P(x) \leftarrow \otimes$

$$\begin{aligned} \circ \int_{-\infty}^0 dt \int p \dot{q} dx &= \int dx \int_{-\infty}^0 dt \left[\ln \frac{q}{1-q} - \ln \frac{P}{1-P} \right] \cdot \dot{q} + \int dx \int_{-\infty}^0 dt \hat{p} \dot{q} \\ &= \int dx \cdot \left[\sigma \ln \frac{\sigma}{P} + (1-\sigma) \ln \frac{1-\sigma}{1-P} \right] + \int_{-\infty}^0 dt \int dx \hat{p} \dot{q} \end{aligned}$$

$$\begin{aligned} \circ H[p, q] &\rightarrow \hat{H}[\hat{p}, q] \\ &= \frac{\sigma(q)}{2} \cdot \left[\partial_x \left(\hat{p} - \ln \frac{P}{1-P} \right) \right]^2 + \partial_x q \cdot \left[\partial_x \left(\hat{p} - \ln \frac{P}{1-P} \right) \right] \end{aligned}$$

Least Action paths

$$\dot{q} + \partial_{xx} q = - \partial_x \left[\sigma(q) \cdot \partial_x \left(\hat{p} - \ln \frac{P}{1-P} \right) \right]$$

$$\hat{p} - \partial_{xx} \left(\hat{p} - \ln \frac{P}{1-P} \right) = - \frac{\sigma'(q)}{2} \left[\partial_x \left(\hat{p} - \ln \frac{P}{1-P} \right) \right]^2$$

Boundary condition

$$\begin{array}{l|l|l} q(x, 0) = P(x) & q(0, t) = P_a & \hat{p}(0, t) = 0 \\ q(x, 1) = \sigma(x) & q(1, t) = P_b & \hat{p}(1, t) = 0 \end{array}$$

Down-hill solution: $\hat{p} = \ln \frac{P}{1-P}$ and $\dot{q} + \partial_{xx} q = 0$

\Rightarrow But contradicts $\hat{p}(0, t) = 0 = \hat{p}(1, t)$

(Comment: $\hat{p} = 0$ is not a solution as eq's are not satisfied)

Second attempt: Transformation

$$p = \ln \frac{1-F}{F} + \ln \frac{q}{1-q}$$

$$q = R \cdot F(1-F)$$

• Step 1:

$$\int_{-\infty}^0 dt P \dot{q} = \left[q \ln q + (1-q) \ln(1-q) + q \ln \frac{1-F}{F} \right]_{t=-\infty}^{t=0} + \int_{-\infty}^0 dt R \dot{F}$$

[Proof:

$$\text{L.H.S} = \int dt \cdot \ln \frac{q}{1-q} \cdot \dot{q} + \int dt \cdot \ln \frac{1-F}{F} \cdot \dot{q}$$

$$= \int dq \ln \frac{q}{1-q} + q \cdot \ln \frac{1-F}{F} \Big|_{t=-\infty}^{t=0} - \int_{-\infty}^0 dt \cdot q \cdot \frac{d}{dt} \ln \frac{1-F}{F}$$

$$= \left[q \ln q + (1-q) \ln(1-q) + q \ln \frac{1-F}{F} \right]_{t=-\infty}^{t=0} + \int_{-\infty}^0 dt \frac{q \cdot \partial_t F}{F(1-F)}$$

↓
R·F

• Step 2: $H[A, q] = -\partial_x R \cdot \partial_x F - R \cdot (\partial_x F)^2 \left[R - \frac{2}{1-F} \right] = \hat{H}[R, F]$

[Proof: $H = q(1-q) \partial_x P \cdot \partial_x [P - \ln \frac{q}{1-q}]$

$$= q(1-q) \left[-\frac{\partial_x F}{F(1-F)} + \frac{\partial_x q}{q(1-q)} \right] \left(-\frac{\partial_x F}{F(1-F)} \right)$$

$$= \frac{q(1-q)}{F^2(1-F)^2} (\partial_x F)^2 - \frac{(\partial_x q)(\partial_x F)}{F(1-F)}$$

$q = R F(1-F)$

→ $= \frac{1-q}{F(1-F)} R (\partial_x F)^2 - \frac{(\partial_x q)(\partial_x F)}{F(1-F)}$

WR $\frac{\partial_x q}{F(1-F)} = \partial_x R + \frac{1-2F}{F(1-F)} \cdot R \cdot \partial_x F$

$$\begin{aligned} \rightarrow R(\partial_x F)^2 \cdot \frac{1-q}{F(1-F)} - R(\partial_x F)^2 \cdot \frac{1-2F}{F(1-F)} - \partial_x R \cdot \partial_x F \\ = R(\partial_x F)^2 \cdot \frac{1}{F(1-F)} (1-q - 1 + 2F) - \partial_x R \cdot \partial_x F \\ = -R(\partial_x F)^2 \left[R - \frac{2}{1-F} \right] - \partial_x R \cdot \partial_x F \quad \text{Proved} \end{aligned}$$

Step 3 $R = \hat{F} + \frac{1}{1-F}$

$$\int_{-\infty}^0 dt R \dot{F} = \int_{-\infty}^0 dt \frac{\hat{F} \dot{F}}{1-F} + \int_{-\infty}^0 dt \hat{F} \cdot \dot{F} = -\ln(1-F) \Big|_{t=-\infty}^{t=0} + \int_{-\infty}^0 dt \hat{F} \cdot \dot{F}$$

$$\begin{aligned} \hat{H}[R, F] &= -\left(\hat{F} + \frac{1}{1-F}\right)\left(\hat{F} - \frac{1}{1-F}\right)(\partial_x F)^2 - \partial_x \hat{F} \cdot \partial_x F - \frac{\partial_x F}{(1-F)^2} \cdot \partial_x F \\ &= -\hat{F}^2 (\partial_x F)^2 - (\partial_x \hat{F})(\partial_x F) \\ &= K[\hat{F}, F] \quad \text{Next page} \end{aligned}$$

Finally

$$S[P, q] = \int dx \cdot \left[q \ln \frac{q}{F} + (1-q) \ln \frac{1-q}{1-F} \right]_{t=-\infty}^{t=0} + \int_{-\infty}^0 dt \cdot \left\{ \int dx \cdot \hat{F} \cdot \dot{F} - K[\hat{F}, F] \right\} \quad \left| \quad K = -\hat{F}^2 (\partial_x F)^2 - (\partial_x \hat{F}) \cdot (\partial_x F) \right.$$

Where

$$F = \frac{q}{q + (1-q)e^P} \quad \text{and} \quad \hat{F} = (1-q)(e^P - 1) - q(e^{-P} - 1)$$

[Proof of the last transformation relation

$$\bullet \quad P = \ln \frac{1-F}{F} + \ln \frac{q}{1-q} \Rightarrow F = \frac{q}{q + (1-q)e^P}$$

$$\begin{aligned} \bullet \quad \hat{F} &= R - \frac{1}{1-F} = \frac{q}{F(1-F)} - \frac{1}{1-F} = \frac{1}{1-F} \left[\frac{q}{F} - 1 \right] \\ &= \frac{1}{1-F} [q + (1-q)e^P - 1] = (1-q)(e^P - 1) \frac{q + e^P(1-q)}{(1-q)e^P} \\ &= \cancel{\frac{(1-q)(e^P - 1)}{(1-q)}} = (1-q)(e^P - 1) - q(e^{-P} - 1) \end{aligned}$$

where we used

$$1-F = \frac{(1-q)e^P}{q + (1-q)e^P} \quad]$$

Boundary conditions:

$$\left. \begin{aligned} F(0, t) &= P_a \\ F(1, t) &= P_b \end{aligned} \right| \begin{aligned} \hat{F}(0, t) &= 0 \\ \hat{F}(1, t) &= 0 \end{aligned}$$

Condition on time:

$$\left. \begin{aligned} q(x, -\infty) &= P(x) \\ q(x, 0) &= \delta(x) \end{aligned} \right\} \begin{aligned} &\text{Explicit transformation} \\ &\text{is difficult} \end{aligned}$$

Ansatz solution :

~~$\hat{F} = \dots$~~

Least Action paths

$$\dot{\hat{F}} = \frac{\delta K}{\delta \hat{F}} = \partial_{xx} F - 2 \hat{F} (\partial_x F)^2$$

$$\dot{\hat{F}} = - \frac{\delta K}{\delta F} = - \partial_{xx} \hat{F} - 2 \partial_x (\hat{F}^2 \partial_x F)$$

Claim Solution

$$\hat{F}_d = \frac{\partial_{xx} F_d}{(\partial_x F_d)^2} \quad \text{and} \quad \dot{\hat{F}}_d = - \partial_{xx} F_d$$

* Important to derive classical

check :

$$\dot{\hat{F}}_d = \partial_{xx} F_d - 2 \frac{\partial_{xx} F}{(\partial_x F)^2} (\partial_x F)^2 = - \partial_{xx} F$$

$$\dot{\hat{F}} = - \partial_x [\partial_x \hat{F} + 2 \partial_x F \cdot \hat{F}^2]$$

$$\begin{aligned} \text{L.H.S} &= \frac{d}{dt} \left(\frac{\partial_{xx} F}{(\partial_x F)^2} \right) = -2 \cdot \frac{\partial_x \partial_x F}{(\partial_x F)^3} \cdot \partial_{xx} F + \frac{\partial_{xx} \dot{F}}{(\partial_x F)^2} \\ &= + \frac{2 \partial_{xx}^3 F}{(\partial_x F)^3} \cdot \partial_{xx} F - \frac{\partial_x^4 F}{(\partial_x F)^2} \end{aligned}$$

$$\begin{aligned} \text{R.H.S} &= - \partial_x \left[-2 \frac{\partial_{xx} F}{(\partial_x F)^3} \cdot \partial_{xx} F + \frac{\partial_x^3 F}{(\partial_x F)^2} + 2 \cdot \partial_x F \cdot \frac{(\partial_x F)^2}{(\partial_x F)^4} \right] \\ &= - \frac{d}{dx} \left(\frac{\partial_x^3 F}{(\partial_x F)^2} \right) = 2 \cdot \frac{\partial_{xx} F}{(\partial_x F)^3} \cdot \partial_x^3 F - \frac{\partial_x^4 F}{(\partial_x F)^2} \end{aligned}$$

Proved

⊙ Boundary condition :

$$\hat{F}_a(0,t) = 0 = \hat{F}_a(1,t)$$

$$\Rightarrow \boxed{\partial_{xx} F_a(0,t) = 0} \text{ at } x=0,1$$

⊙ Condition on k_{int} : First Show

$$\boxed{F_a + \frac{F_a(1-F_a)}{a} \frac{\partial_{xx} F_a}{(\partial_x F_a)^2} = q(x,t)} \quad \leftarrow \begin{array}{l} \text{comes} \\ \text{from} \end{array} \quad \hat{J}_a = \frac{\partial_{xx} F_a}{(\partial_x F_a)^2}$$

Proof : $\hat{J}_a = \frac{\partial_{xx} F_a}{(\partial_x F_a)^2}$

$$\left. \begin{array}{l} \rightarrow = (1-q)(e^p - 1) + q(e^{-p} - 1) \\ \hookrightarrow \text{from } F = \frac{q}{q + (1-q)e^p} \end{array} \right\}$$

This implies

~~Boundary condition~~

$$F_a + \frac{F_a(1-F_a)}{a} \frac{\partial_{xx} F_a}{(\partial_x F_a)^2} = p(x) \text{ at } t = -\infty$$

$$F_a + \frac{F_a(1-F_a)}{a} \frac{\partial_{xx} F_a}{(\partial_x F_a)^2} = \delta(x) \text{ at } t = 0$$

~~Take choice :~~ ~~$F_a(x, \infty) = p(x)$~~

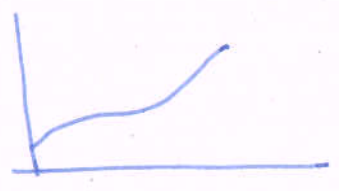
Works if $F(x, \infty) = p(x)$

Corresponding ~~minimal action~~ $K[\] = 0$ [integration by parts]

minimal action $\hat{S}_{\text{min}} = \int_{-\infty}^0 dt \int dx \frac{\partial_{xx} F}{(\partial_x F)^2} \cdot F = \int_{-\infty}^0 dt \int dx \frac{\delta \ln \partial_x F}{\delta F} = \int_{-\infty}^0 dt \int dx \frac{\delta \ln \partial_x F}{\delta F} \Big|_{-\infty}^0$

Hamilton - Jacobi approach

$$P_T[\sigma(x), p(x)] \sim e^{-L \cdot \Phi_T[\sigma(x), p(x)]}$$



One gets

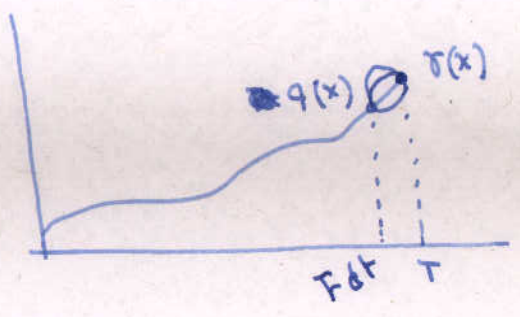
$$\boxed{\frac{d}{dt} \Phi_T(\sigma, p) = -H\left[\frac{\delta \Phi_T}{\delta \sigma}, \sigma\right]}$$

↑
p

Proof 1: generalize the approach given for Langevin equation.

Proof 2:

$$P_T[\sigma(x)] = \int \mathcal{D}[q] P_{T-dt}[q(x)]$$



$$W_{dt}(\sigma(x), q(x))$$

$$\text{one } [\partial_t A = -\partial_x J]$$

$$\longrightarrow \sigma(x) - q(x) = -dt \partial_x J$$

$$\Rightarrow W_{dt}(\sigma, q) \approx \int \mathcal{D}[J] \cdot e^{-L dt \int dx \frac{(J + D(\sigma) \partial_x \sigma)^2}{2\sigma(\sigma)}} \cdot \delta(\sigma - q + dt \partial_x J)$$

$$\Rightarrow P_T[\sigma(x)] = \int \mathcal{D}[J] P_{T-dt}(\sigma + dt \partial_x J) e^{-\frac{L}{2} dt \int dx \frac{(J + D(\sigma) \partial_x \sigma)^2}{\sigma(\sigma)}}$$

Substitute $P_T[\sigma] \sim e^{-L \Phi_T[\sigma]}$

$$\Rightarrow e^{-L \Phi[\sigma]} \equiv \int \mathcal{D}[J] \cdot e^{-L \Phi_{T-dt}(\sigma + dt \partial_x J) - L dt \int dx \frac{(J + D(\sigma) \partial_x \sigma)^2}{2\sigma(\sigma)}}$$

Final result

(75)

$$\begin{aligned}\phi[\sigma(x)] &= \int_0^1 dx \left[\sigma(x) \ln \frac{\sigma(x)}{F(x)} + (1-\sigma(x)) \ln \frac{1-\sigma}{1-F} + \ln \frac{\partial_x F}{p_a - p_a} \right]_{t=-\infty}^{t=0} \\ &= \int_0^1 dx \left[\sigma(x) \ln \frac{\sigma(x)}{F(x)} + (1-\sigma(x)) \ln \frac{1-\sigma}{1-F} + \ln \frac{\partial_x F}{p_a - p_a} \right]\end{aligned}$$

where

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

* Remark :

$$\frac{d}{dt} F_a(x,t) = - \frac{d^2}{dx^2} F_a(x,t)$$

with $F_a(x, -\infty) = p(x)$

$$F_a(x, 0) = \text{solution of } \left\{ F_a + F_a(1-F_a) \frac{\partial_{xx} F_a}{\partial_x F_a} = \sigma \right\}$$

Solution exist because we know F_{TR} exist.

⊕ Remark 2 : The optimal trajectory of density is

$$q_a(x,t) = F_a(x,t) + F_a(x,t) (1-F_a(x,t)) \frac{\partial_{xx} F_a(x,t)}{\partial_x F_a(x,t)}$$

Taylor expand

$$= \int \rho(\sigma) e^{-L\phi_{T-dt}[\sigma]} \cdot L dt \cdot \int dx \cdot \frac{\delta \phi_{T-dt}}{\delta \sigma} \cdot \partial_x \mathcal{J} - L dt \int dx \frac{(\mathcal{J} + D \partial_x \sigma)^2}{2\sigma(\sigma)}$$

$$\Rightarrow e^{-L\phi_T[\sigma]} = e^{-L\phi_{T-dt}[\sigma]} \int \rho(\sigma) \cdot e^{-L dt \int dx \left[\frac{\delta \phi}{\delta \sigma} \cdot \partial_x \mathcal{J} + \frac{(\mathcal{J} + D \partial_x \sigma)^2}{2\sigma(\sigma)} \right]}$$

[First take $L \rightarrow \infty$ then $dt \rightarrow 0$]

$$\Rightarrow \frac{d\phi_T}{dt} = + \min_{\mathcal{J}} \int_0^1 dx \left[\frac{\delta \phi}{\delta \sigma} \cdot \partial_x \mathcal{J} + \frac{(\mathcal{J} + D \partial_x \sigma)^2}{2\sigma(\sigma)} \right]$$

$$= \min_{\mathcal{J}} \int_0^1 dx \left[-\partial_x \left(\frac{\delta \phi}{\delta \sigma} \right) \cdot \mathcal{J} + \frac{(\mathcal{J} + D \partial_x \sigma)^2}{2\sigma} \right] + \mathcal{J} \cdot \frac{\delta \phi}{\delta \sigma} \Big|_{x=0}^{x=1}$$

$$= \int_0^1 dx \left\{ \frac{[D(\sigma) \partial_x \sigma]^2}{2\sigma(\sigma)} - \frac{[D(\sigma) \partial_x \sigma - \sigma(\sigma) \partial_x \frac{\delta \phi}{\delta \sigma}]^2}{2\sigma(\sigma)} \right\} + \mathcal{J} \cdot \frac{\delta \phi}{\delta \sigma} \Big|_{x=0}^{x=1}$$

$$= - \int_0^1 dx \left\{ \frac{\sigma(\sigma)}{2} \left(\partial_x \frac{\delta \phi}{\delta \sigma} \right)^2 - D(\sigma) \cdot \partial_x \sigma \cdot \partial_x \frac{\delta \phi}{\delta \sigma} \right\} + \mathcal{J} \cdot \frac{\delta \phi}{\delta \sigma} \Big|_{x=0}^{x=1}$$

$$= -H \left[\frac{\delta \phi}{\delta \sigma}, \sigma \right] + \mathcal{J}(x) \cdot \frac{\delta \phi}{\delta \sigma(x)} \Big|_{x=0}^{x=1}$$

[Why $\int \frac{\delta \phi}{\delta \sigma} = 0$? self-consistent solution?]

⊗ choose a boundary condition $\frac{\delta \phi}{\delta \sigma} = 0$?

STATIONARY STATE

Example 1: Equilibrium:

~~$\phi_T[\sigma(x), p]$~~ $\phi_T[\sigma(x), p] = \int dx \cdot \{ f(\sigma) - f(p) - f'(p)(\sigma-p) \}$ indep of time.

$$\Rightarrow \frac{\delta \phi}{\delta \sigma} = f'(\sigma) - f'(p)$$

$$\Rightarrow \partial_x \frac{\delta \phi}{\delta \sigma} = f''(\sigma) \cdot \partial_x \sigma = \frac{2D(\sigma)}{\sigma(\sigma)} \cdot \partial_x \sigma$$

In stationary state $\frac{\partial \phi}{\partial \tau} = 0$

$$\Rightarrow H \left[\frac{\delta \phi}{\delta \sigma}, \tau \right] = \int dx \left\{ \underbrace{\frac{\sigma(\sigma)}{2} \cdot \partial_x \frac{\delta \phi}{\delta \sigma} - D(\sigma) \cdot \partial_x \sigma}_0 \right\} \cdot \partial_x \frac{\delta \phi}{\delta \sigma} = 0$$

Example 2: Non-interacting particles [Exercise]

Example 3: SEP: $\phi[\sigma] = \int dx \cdot \left\{ \sigma \ln \frac{\sigma}{F} + (1-\sigma) \ln \frac{1-\sigma}{1-F} + \ln \frac{\partial_x F}{P_b - P_a} \right\}$

$$\sigma = F + F(1-F) \cdot \frac{\partial_x F}{(\partial_x F)^2} \quad \left| \begin{array}{l} F(0) = P_a \\ F(1) = P_b \end{array} \right.$$

$$\circ \frac{\delta \phi}{\delta \sigma} = \ln \frac{\sigma(1-F)}{(1-\sigma)F} = \ln \left(1 + \frac{\partial_x F}{(\partial_x F)^2 - F \cdot \partial_x F} \right)$$

(substitute

$$\circ H = -2 \cdot \int_0^1 dx \cdot \frac{d}{dx} \left[\frac{\partial_x F}{\partial_x F} \right] = 2 \frac{\partial_x F}{\partial_x F} \Big|_{x=0} - 2 \frac{\partial_x F}{\partial_x F} \Big|_{x=1}$$

$$= 2 \cdot \frac{[\gamma(x) - F(x)] \partial_x F}{F(x)(1-F(x))} \Bigg|_{x=0}^{x=1}$$

= 0 because at boundary $\gamma(x) = F(x)$

Ref: Bertini, De solis, Gabrielli, Jona-Lasinio and Landim 2001, PRL, 87, 40601
 *Solving for general is difficult and not known yet 2002, JSP, 107, 635

Correlations

© Fax

Remark %

Solving for general system is difficult and an open problem!

Other known model % KMP : $D=1$; $\sigma = 2P^2$

Ref Bertini et al

Bertini, Gabrielli and Lebowitz (2005)
 JSP, 121, 843.

© A general solution and correlation!

Correlation

⊛ Cumulant generating function: "Recap of earlier lecture"

$$a[h] = \max_{\sigma(x)} \left[\int dx h(x) \sigma(x) - \phi[\sigma(x)] \right] \quad \text{Legendre transform}$$

cumulants

$$\langle \sigma(x_1) \dots \sigma(x_k) \rangle_c = \frac{1}{L^{k-1}} \frac{\delta^k a}{\delta h(x_1) \dots \delta h(x_k)} \Big|_{h=0}$$

⊛ What is $a[h]$ for NI particles?

$$\phi[\sigma] = \int dx \cdot \left\{ \sigma \ln \frac{\sigma}{P(x)} - \sigma + P \right\}$$

Legendre transform

$$\Rightarrow a[h] = \int dx \cdot P(x) \left(e^{h(x)} - 1 \right)$$

$$= \int dx h(x) \cdot P(x) + \frac{1}{2!} \int dx h(x)^2 P(x) + \dots$$

⇒ Average

$$\langle \sigma(x) \rangle = P(x)$$

$$\langle \sigma(x) \sigma(y) \rangle = \frac{1}{L} P(x) \delta(x-y) \quad \text{[short-range correlation]}$$

⊛ Symmetric exclusion process

$$a[h] = \int dx \left[\log(1-F+Fe^h) - \ln \frac{\partial_x F}{P_b - P_a} \right] \equiv \int dx \left[\ln(1+F(e^h-1)) - \ln \frac{F'}{P_b - P_a} \right]$$

$$\text{with } F'' + \frac{(F')^2 (1-e^h)}{1-F+Fe^h} = 0$$

$$\equiv F'' = \frac{(F')^2 (e^h-1)}{1+F(e^h-1)}$$

Expanding

$$G[h] = \int dx h(x) P(x) = \frac{1}{2} \int dx \int dy \cdot [x(1-y)\Theta(y-x) + y(1-x)\Theta(x-y)] h(x)h(y) \cdot (p_a - p_b)^2$$

$$+ \frac{1}{2} \int dx P(x)(1-P(x)) h(x)^2 + \dots$$

Average:

$$\langle \sigma(x) \rangle = P(x) = p_a(1-x) + p_b x$$

$$\langle \sigma(x)\sigma(y) \rangle = \frac{1}{L} \left\{ P(x)(1-P(x)) \delta(x-y) - [x(1-y)\Theta(y-x) + y(1-x)\Theta(x-y)] (p_a - p_b)^2 \right\}$$

(Long-range correlation)

↓
vanishes for $p_a = p_b$

~~How to derive for general case?~~

Back ground: long-range correlation at generic parameter value.

Mention other references.
"Sriram Ramaswami"
People from Bangalore?

Ref: Garrido, Lebowitz, Maes, Spohn: PRA 42, 1554 (1990)

Groisman, Lee, Sachdev PRL 64, 1927 (1990)

Dorfman, Kirkpatrick, Sengers Ann. Rev. Physics Chem 45, 213 (1994)

Garrido

Correlation for general case: A perturbative solution of H-J equation.

$$H\left[\frac{\delta\phi}{\delta\sigma}, \sigma\right] = 0$$

$$\Rightarrow \int dx \left\{ \frac{\sigma(x)}{2} \cdot \left(\partial_x \frac{\delta\phi}{\delta\sigma}\right)^2 - \partial_x \sigma \cdot D(x) \cdot \left(\partial_x \frac{\delta\phi}{\delta\sigma}\right) \right\} = 0$$

The cumulant generating function

$$G[h] = \int dx \cdot h(x) \sigma(x) - \phi[\sigma(x)]$$

where $h(x) = \frac{\delta\phi}{\delta\sigma(x)}$ (To be consistent $h=0$ at $x=0$ and $x=L$)

Also $\sigma(x) = \frac{\delta G}{\delta h(x)}$ (* Legendre transform is invertible * why?)

→ leads to

$$\int dx \cdot \left\{ \frac{1}{2} \cdot \sigma \left(\frac{\delta G}{\delta h} \right) \left(\partial_x h \right)^2 - \partial_x \left(\frac{\delta G}{\delta h} \right) \cdot D \cdot \left(\frac{\delta h}{\delta h} \right) \cdot \left(\partial_x h \right) \right\} = 0$$

H-J equation for $G[h]$

Cumulants:

$$G[h] = \int dx h(x) \langle \sigma(x) \rangle + \frac{1}{2!} \int dx \int dy \overbrace{\langle \sigma(x) \sigma(y) \rangle_c}^{h(x)h(y)} + \frac{1}{3!} \int dx_1 dx_2 dx_3 \dots \overbrace{\langle \sigma(x_1) \sigma(x_2) \sigma(x_3) \rangle_c}^{h(x_1)h(x_2)h(x_3) \dots}$$

~~Expand in powers of h~~

$$\Rightarrow \frac{\delta G}{\delta h(x)} = \underbrace{\langle \sigma(x) \rangle}_{P(x)} + \int dy h(y) \underbrace{\langle \sigma(x) \sigma(y) \rangle_c}_{C(x,y)} + \frac{1}{2} \int dy \int dz h(y) h(z) \underbrace{\langle \sigma(x) \sigma(y) \sigma(z) \rangle_c}_{C(x,y,z)} + \dots$$

Substitutz

$$\sigma\left(\frac{\delta G}{\delta h}\right) = \sigma(P(x)) + \sigma'(P) \cdot \int dy h(y) c(x,y) + \dots$$

similar for $D\left(\frac{\delta G}{\delta h}\right)$

The H-J equation

$$\int dx \cdot \left\{ \frac{1}{2} \cdot \left[\sigma(P) + \sigma'(P) \cdot \int dy h(y) c(x,y) + \dots \right] (\partial_x h)^2 - \left[\partial_x P + \int dy h(y) \partial_x c(x,y) + \dots \right] \left[D(P) + D'(P) \cdot \int dy h(y) c(x,y) + \dots \right] \partial_x h \right\} = 0$$

⇒ linear order

$$\boxed{- \int dx \cdot \partial_x P \cdot D(P) \cdot \partial_x h = 0} \Rightarrow \int dx \cdot h(x) \cdot \partial_x [D(P) \partial_x P] = 0$$

⇒ quadratic order

$$\int_0^1 dx \cdot \left\{ \frac{\sigma(P)}{2} \cdot (\partial_x h)^2 - (\partial_x P) \cdot D'(P) \cdot \int_0^1 dy h(y) c(x,y) \cdot \partial_x h(x) - D(P) \cdot \int_0^1 dy h(y) \partial_x c(x,y) \cdot \partial_x h \right\} = 0$$

$$\Rightarrow \int dx dy \cdot \left\{ \frac{\sigma(P(x))}{2} \cdot \delta(x-y) \cdot \partial_x h \cdot \partial_y h(y) - \partial_x (D(P) \cdot c(x,y)) \cdot \partial_x h(x) \cdot h(y) \right\} = 0$$

$$\boxed{\int dx dy \cdot h(x) h(y) \cdot \left\{ \frac{d^2}{dx dy} \left[\frac{\sigma(P(x)) \cdot \delta(x-y)}{2} \right] + \frac{d^2}{dx^2} [D(P) c(x,y)] \right\} = 0}$$

Because $h(x)$ is an arbitrary function.

① $\partial_x [D(\rho) \partial_x \rho] = 0 \Rightarrow D(\rho) \partial_x \rho = \langle J \rangle = \text{constant in } x$

determines the average

② Symmetrize the equation and get

$$\frac{d^2}{dx^2} [D(\rho(x)) c(x, y)] + \frac{d^2}{dy^2} [D(\rho(y)) c(y, x)] = -\frac{1}{2} \partial_x \partial_y [\sigma(\rho(x)) \delta(x-y) + \sigma(\rho(y)) \delta(y-x)]$$

\downarrow
 $c(x, y)$

$$= \partial_x [\sigma(\rho(x)) \delta'(x-y)]$$

Poisson-Type eqⁿ

③ $\mathcal{L}^{(3)} c(x, y, z) = - \sum_{\text{Perm } (x, y, z)} \left\{ \partial_x [\sigma'(x) \delta'(x-y) c(x, y)] + \partial_x^2 [D'(x) c(x, y) c(x, z)] \right\}$

General structure:

$$\mathcal{L}^{(n)} c(x_1 \dots x_n) = \Lambda \text{ (lower point correlations)}$$

Longrange nature:

$$c(x, y) = \underbrace{\frac{\sigma(\rho(x))}{2D(\rho(x))} \delta(x-y)}_{\text{equilibrium part } C_{eq}(x, y)} + B(x, y)$$

z dependence! ← long-range part

"Short range" $\left[\sim e^{-\frac{x}{\xi}} \right]$
 $\xi \gg a$

$$\partial_x^2 [D(p(x)) B(x,y)] + \partial_y^2 [D(p(y)) B(x,y)] = \delta(x-y) \left[\langle j \rangle \cdot \frac{d}{dx} \left(\frac{\sigma'(p)}{2D(p)} \right) \right]$$

change.

(can be generalized for bulk-driven @ax.

In equilibrium: $\langle j \rangle = 0 \Rightarrow B(x,y) = 0$

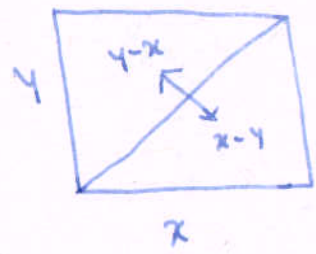
Non-interacting ptly: $\frac{\sigma(p)}{2D} = p \Rightarrow \frac{\sigma'(p)}{2D} = 1$
 $\Rightarrow \partial_x [] = 0 \Rightarrow B = 0$

Zero-range process: $\sigma'(p) = 2D \Rightarrow B = 0$

CAN
BE
GENERALIZED
TO
ANY
DIMENSION

When D = constant: $B(x,y) \equiv$ Potential due to ^{line} localized charge.

$B(x,y) \sim 1 - |x-y|$ [Dimension=1]



* In general d

$B(x,y) \sim \frac{1}{|x-y|^{d-2}}$ ~~Potential due to quadrupolar charge~~
Inverse Laplacian.

- ① Spohn 1983 J.P.A 16, 4275
- ② Garrido, ... PRA (1990)

* In general depends on multipole moments

[Sahu, Majumdar, Mukamel 2014
PRE 90, 012109]

Exercise: verify for SEP ($D=1, \sigma=2p(1-p)$)

$$B(x,y) = -(pa+pb)^2 [x(1-y)\theta(y-x) + y(1-x)\theta(x-y)]$$

What about higher point correlation & general solution?

Green's function:

$$\partial_t g(x,y,t) = \partial_x^2 [D(p(x)) g(x,y,t)]$$

with $g(x,y,0) = \delta(x-y)$

Any correlation can be written in terms of $g(x,y,t)$

Claim:

$$C(x_1, \dots, x_n) = - \int_0^\infty dt \int_0^1 dy_1 \dots \int_0^1 dy_n g(x_1, y_1, t) \dots g(x_n, y_n, t) \Lambda(y_1, \dots, y_n)$$

Proof: $\partial_{x_i}^2 D(x_i) C(x_1, \dots, x_n) + \dots = - \int_0^\infty dt \dots \cdot \frac{d}{dt} [g(x_1, y_1, t) \dots] \Lambda(y_1, \dots, y_n)$
 $= + \Lambda(x_1, \dots, x_n)$

① For $D=1$: $a = 2 \sum_{n=1}^\infty e^{-n^2 t} \sin(n\pi x) \sin(n\pi y)$

Is there any compact formula for general system?

① Verify for SEP. $\partial_t F(x,t) = \partial_x^2 F(x,t)$

② F in terms of g .

② What happens for non-interacting particles?

③ In-equilibrium?

③ The correlation is order $\frac{1}{L}$.

Can it be significant in any practical case.

Example: Fluctuation of total number of particles.

$$N = L \int dx \cdot \sigma(x)$$

For SEP

$$\bullet \langle N \rangle = L \cdot \frac{p_a + p_b}{2}$$

$$\bullet \langle N^2 \rangle - \langle N \rangle^2 = L \left[\int dx \cdot p(x)(1-p(x)) - 2(p_a - p_b)^2 \int_0^1 dx \int_0^1 dy x(1-y) \right]$$

significant difference due to this

④ Any example where the correlation ~~is not significant?~~ does not vanish at $L \rightarrow \infty$

* correlation is $\propto \langle J \rangle$

⑤ Bulk-driven model, system with localized drive.

Statistics of current

* Current is a characteristic of Non-equilibrium system.

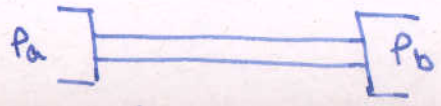
* Green-Kubo formula

$$\text{Conductivity} = \frac{1}{2P^2X} \int_{-\infty}^{\infty} dt \langle J(0)J(t) \rangle_{eq.}$$

* Gallavotti-Cohen symmetry

~~...~~

$$\frac{P\left(\frac{Q_t}{t} = -J\right)}{P\left(\frac{Q_t}{t} = J\right)} = e^{-t \cdot J \cdot (\mu_a - \mu_b)}$$



* Keeping average density

fixed \equiv fixed chemical potential μ

$$\Rightarrow G = \mu N - F$$

$$\Rightarrow \mu = \frac{dF}{dN} = f'(P)$$

[Original statement in terms of entropy production]

Ref: ① Gallavotti and Cohen, J. Stat. Phys. 80, 931 (1995)
② Evans and Searles, Adv. Phys. 51, 1529 (2002)
③ Kurchan J. Phys. A 31, 3719 (1998)]

* Transport properties, heat conduction, ...

① A. Dhar. Adv. Phys. ...

Large deviation of integrated current:

$$Q_T = \frac{1}{L} \int_0^L dx \int_0^T dt \mathcal{Y}(x,t)$$

$$\approx \frac{1}{L} \cdot L \cdot T \int_0^1 dz \int_0^{\tau} dt \frac{j(x,t)}{L} \quad (t,x) \equiv \left(\frac{T}{L}z, \frac{T}{L}t\right)$$

$$\approx L \int_0^1 dx \int_0^{\tau} dt j(x,t) \xrightarrow{\text{large } \tau} L \tau j + \text{subleading} \rightarrow \frac{\tau}{L} j$$

$$\mathcal{P}\left(\frac{Q_T}{\tau} = \frac{j}{L}\right) \sim e^{-\tau L \phi(j)} \equiv e^{-\frac{\tau}{L} \Phi(j)}$$

what is $\phi(j)$?

Proof and analysis:

How does cumulants scale:

$$\ln \langle e^{\lambda Q_T} \rangle = \lambda \langle Q_T \rangle + \frac{\lambda^2}{2} \langle Q_T^2 \rangle_c + \dots$$

$$\ln \int dj e^{\lambda \tau L j - \tau L \phi(j)}$$

$$\tau L \max_j \{ \lambda j - \phi(j) \} = \tau L a(\lambda) = \lambda \cdot \tau \cdot L \cdot a(0) + \lambda \cdot \tau \cdot L \cdot a'(0) + \dots$$

Comparing: $\langle Q_T \rangle = \tau L j \equiv \frac{\tau}{L} \cdot a(0)$
 $\langle Q_T^2 \rangle = \frac{\tau}{L} a'(0)$

all cumulants $\sim \frac{\tau}{L}$
 we are ignoring subleading term $\sim \tau/L^2$

① There is an expression of $\phi(j)$ for general D and σ

"Additivity principle"

Proof: $P\left(\frac{Q_T}{T} = \frac{j}{L}\right) \approx \int \mathcal{D}[j, q] e^{-L \int_0^T dt \int_0^1 dx \frac{[\dot{j} + D(q)\partial_x q]^2}{2\sigma(q)}}$

$\delta\left(\int_0^T dt \int_0^1 dx J(x,t)\right) = \frac{j}{L}$

for large L the optimal history $\circ \equiv (j, q(x,t)) = (j, q(x))$

Time independent

$e^{-L \cdot \mathcal{L} \cdot \min_{q(x)} \int_0^1 dx \frac{(\dot{j} + D(q)\partial_x q)^2}{2\sigma(q)}}$

$\Rightarrow \phi(j) = \min_{q(x)} \int_0^1 dx \frac{(\dot{j} + D(q)\partial_x q)^2}{2\sigma(q)}$

Solution: $\mathcal{L} = \min_{q(x)} \int_0^1 dx \left\{ \dot{j}^2 \left(\frac{1}{2\sigma}\right) + \dot{j} \cdot \frac{D\partial_x q}{\sigma} + \frac{D^2}{2\sigma} \cdot (\partial_x q)^2 \right\}$

At minimum $\frac{\delta \mathcal{L}}{\delta q} = 0$

$\Rightarrow \dot{j}^2 \left(\frac{1}{2\sigma}\right)' + \dot{j} \left(\frac{D}{\sigma}\right)' \partial_x q + \left(\frac{D^2}{2\sigma}\right)' (\partial_x q)^2 - \dot{j} \partial_x \left(\frac{D}{\sigma}\right) - \partial_x \left(2\partial_x q \cdot \frac{D}{2\sigma}\right) = 0$

cancels

$- 2 \cdot \partial_x q \cdot \left(\frac{D^2}{2\sigma}\right)' - 2 \cdot \left(\frac{D^2}{2\sigma}\right)' \cdot (\partial_x q)^2$

$$\Rightarrow \delta^2 \left(\frac{1}{2\sigma} \right)' - \left(\frac{D^2}{2\sigma} \right)' (\partial_x \eta)^2 - 2 \left(\frac{D^2}{2\sigma} \right) \partial_x \eta = 0$$

Multiply $\partial_x \eta$

$$\frac{d}{dx} \left[\delta^2 \left(\frac{1}{2\sigma} \right) \right] - \frac{d}{dx} \left[\frac{D^2}{2\sigma} \right] (\partial_x \eta)^2 - \left(\frac{D^2}{2\sigma} \right) \cdot \frac{d}{dx} (\partial_x \eta)^2 = 0$$

$$\Rightarrow \frac{d}{dx} \left[\delta^2 \left(\frac{1}{2\sigma} \right) - \frac{D^2}{2\sigma} (\partial_x \eta)^2 \right] = 0$$

Integrate

$$\delta^2 \left(\frac{1}{2\sigma} \right) - \frac{D^2}{2\sigma} (\partial_x \eta)^2 = k \cdot \delta^2$$

↖ a hole (constant)
↖ integration constant

$$D^2 (\partial_x \eta)^2 = \delta^2 [1 + 2k\sigma(\eta)]$$

$$\Rightarrow D \partial_x \eta = - \delta \sqrt{1 + 2k\sigma(\eta)} \quad \text{optimal profile}$$

↖ why -ve (for positive δ , $\partial_x \eta < 0$)

↖ ~~optimal profile~~

$$\delta = - \int_0^1 dx \frac{D \partial_x \eta}{\sqrt{1 + 2k\sigma}}$$

$$\Rightarrow \delta = \int_{P_b}^{P_a} dx \frac{D(\eta)}{\sqrt{1 + 2k\sigma}}$$

↖ ~~optimal profile~~
↖ k is determined from this!

What is $\phi(\dot{\theta})$?

$$\begin{aligned} \phi(\dot{\theta}) &= \int dx \cdot \frac{\dot{\theta}^2}{2\sigma} \cdot \left(\cancel{1+2K\sigma} 1 - \sqrt{1+2K\sigma} \right)^2 \\ &= \dot{\theta} \int dx \cdot \frac{\dot{\theta}}{2\sigma} \cdot [1 + 1 + 2K\sigma - 2\sqrt{1+2K\sigma}] \\ &= \dot{\theta} \int dx \cdot \frac{\dot{\theta}}{\sigma} \cdot [1 + K\sigma - \sqrt{1+2K\sigma}] \\ &= -\dot{\theta} \int dx \cdot \frac{d\sqrt{1+2K\sigma}}{\sigma} \left[1 - \frac{1+K\sigma}{\sqrt{1+2K\sigma}} \right] \\ &= \dot{\theta} \int dx \cdot \frac{D \cdot 2x^9}{\sigma} \left[1 - \frac{1+K\sigma}{\sqrt{1+2K\sigma}} \right] \end{aligned}$$

$$\Rightarrow \phi(\dot{\theta}) = \dot{\theta} \int_{P_a}^{P_b} dq \cdot \frac{D}{\sigma} \cdot \left[1 - \frac{1+K\sigma}{\sqrt{1+2K\sigma}} \right]$$

where K is given by

$$\dot{\theta} = \int_{P_b}^{P_a} dq \cdot \frac{D}{\sqrt{1+2K\sigma}}$$

Parametric solution

Remark 1

Galilean-Cohen symmetry: For $\dot{\theta} \rightarrow -\dot{\theta}$

$$-\dot{\theta} = \int_{P_b}^{P_a} dq \cdot \frac{D(q)}{\sqrt{1+2K^*\sigma}} = - \int_{P_b}^{P_a} dq \cdot \frac{D(q)}{\sqrt{1+2K\sigma}}$$

A simple choice $\frac{1}{\sqrt{1+2K^*\sigma}} = - \frac{1}{\sqrt{1+2K\sigma}}$

⊗ $K^* = K$ if one squares; it is a choice of which sign to take

$$\Rightarrow \phi(-j) = -j \int_{P_a}^{P_b} d\varphi \frac{D}{\sigma} \left[1 - \frac{1+K^*\sigma}{\sqrt{1+2K^*\sigma}} \right]$$

$$= -j \int_{P_a}^{P_b} d\varphi \frac{D}{\sigma} \left[1 + \frac{1+K\sigma}{\sqrt{1+2K\sigma}} \right]$$

$$\Rightarrow \phi(j) - \phi(-j) = j \int_{P_a}^{P_b} d\varphi \frac{2D}{\sigma} = j \int_{P_a}^{P_b} d\varphi f''(\varphi) = j [f'(P_b) - f'(P_a)]$$

$$= j(\mu_b - \mu_a)$$

$$\Rightarrow \boxed{\phi(j) - \phi(-j) = -j(\mu_a - \mu_b)}$$

$$\equiv \frac{P(-j)}{P(j)} = e^{-\tau \cdot L \cdot [\phi(-j) - \phi(j)]} = e^{-\tau \cdot L \cdot j(\mu_a - \mu_b)}$$

$$= e^{-\frac{\tau}{L} \cdot j \cdot (\mu_a - \mu_b)}$$

Remark 2 : $Q(\lambda) = \max_j \{ \lambda j - \phi(j) \}$

$$= -K \left[\int_{P_b}^{P_a} d\varphi \frac{D(\varphi)}{\sqrt{1+2K\sigma}} \right]^2$$

with $\lambda = \int_{P_b}^{P_a} d\varphi \frac{D}{\sigma} \left[\frac{1}{\sqrt{1+2K\sigma}} - 1 \right]$

* Check that :

$$\frac{\langle Q_T \rangle}{T} = \frac{1}{L} \int_{P_a}^{P_b} d\varphi D(\varphi) \approx \frac{1}{L} D(P)(P_a - P_b) \quad \left| \quad \frac{\langle Q_T^2 \rangle}{T} = \frac{1}{L} \frac{\int_{P_b}^{P_a} d\varphi \frac{D}{\sigma}}{\int_{P_a}^{P_b} d\varphi D} = \frac{\sigma(P)}{L}$$

Remark 3: For Non-Interacting particles: $D=1$; $\sigma=29$ (1991)

$$Q(\lambda) = P_a (e^\lambda - 1) + P_b (\bar{e}^\lambda - 1)$$

[check this]

For SEP: $D=1$; $\sigma=29(1-9)$

$$Q(\lambda) = [\ln(\sqrt{1+\omega} + \sqrt{\omega})]^2$$

where $\omega = P_a(e^\lambda - 1) + P_b(\bar{e}^\lambda - 1) - P_a P_b (e^\lambda - 1)(\bar{e}^\lambda - 1)$

- * This feature is found for other cases. and has a symmetry reason.
- * Same found in transport of weakly interacting Fermion between two baths

Remark 4

Ref: Lee, Levitov, Yakovets PRB 51, 4073 (1995)

Beenakker and Büttiker PRB 46, 1889 (1992)

Remark 4: ~~non-Gaussian~~ equilibrium

The distribution is non-Gaussian even at equilibrium. ($P_a = P_b = P$)

* Only in equilibrium at $P = P^*$ where $\sigma(P^*)$ is maximum it is Gaussian, i.e., all cumulant $\langle Q^n \rangle_c$ with $n > 2$ = 0.

Remark 50: On periodic boundary



Equilibrium

$$\ln \left\langle e^{\lambda Q_T} \right\rangle = a(\lambda) = \frac{\sigma(\rho)}{L} \cdot \frac{\lambda^2}{2} + \frac{D}{L^2} \cdot \mathcal{F} \left(\frac{\sigma \sigma'}{16 D^2} (\lambda L)^2 \right)$$

↓ \nearrow Bernoulli numbers

$$\mathcal{F}(x) = \sum_{k \geq 2} \frac{B_{2k-2}}{\Gamma(k) \Gamma(k+1)} (-2x)^k$$

$$[B_0 = 1, B_2 = \frac{1}{6}, B_4 = -\frac{1}{30}]$$

Cumulants


$$\frac{\langle Q^2 \rangle}{T} \sim \frac{1}{L} ; \quad \frac{\langle Q^4 \rangle}{T} \sim L^2$$


$$\frac{\langle Q^6 \rangle}{T} \sim L^4$$

- * More fluctuations
- * Can have traveling wave

[* Fluctuations depend on Boundary condition

Non surprising even in equilibrium

P.  $\phi[r] = \int dx \{ f(x) - f(\rho) - f'(\rho)(x-\rho) \}$

 $\phi[r] = \int dx \{ f(x) - f(\rho) \}$

Refs ① Appert-Rolland, Debnath, Lecomte, van Wijland

PRE 78, 021122 (2008)

② Lecomte, Imperato, van Wijland

Prog. Theo. Phys. Suppl. 184, 276 (2016)

Tutorial : ~~Verify the result~~

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Remark 6 : On an infinite line



$$P\left(\frac{Q_T}{\sqrt{T}} = i\right) \propto e^{-\sqrt{T} \phi(i)}$$

* A different regime !!

Equivalently

~~$$\ln \langle e^{\lambda Q_T} \rangle \sim \sqrt{T} \mu(\lambda)$$~~

$$\ln \langle e^{\lambda Q_T} \rangle \sim \sqrt{T} \cdot \mu(\lambda)$$

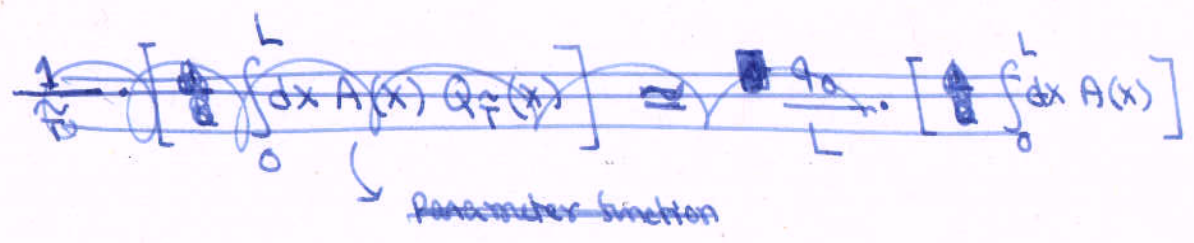
$$\Rightarrow \boxed{\langle Q^n \rangle_c \sim \sqrt{T}}$$

will do an explicit calculation

Briefly discuss joint distribution

(Joint distribution of current and density)

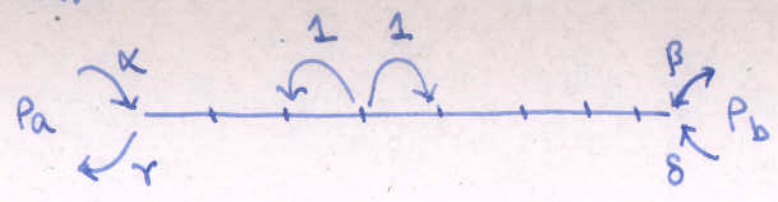
* The density profile is sensitive to where one measures current.



~~Joint probability~~ Joint Probability

~~Diagram~~ $P[Q_0, n(x)] = ?$

Result: A solvable case



$\alpha = \gamma P_a \Rightarrow P_a = \frac{\alpha}{\gamma}$

$\log \left\langle e^{\sum_{i=1}^{L-1} A_i Q_i(\tilde{\tau}) + \sum_{i=1}^L h_i n_i(\tilde{\tau})} \right\rangle$

$\xrightarrow[\substack{\text{Large } T \gg 1 \\ L \gg 1}]{\tilde{\tau}} \mu(A_1, \dots, A_{L-1}) + \alpha(A_1, \dots, A_{L-1}; h_1, \dots, h_L)$

where

$\mu = \frac{1}{L} [P_a (e^{\sum_i A_i} - 1) + P_b (e^{-\sum_i A_i} - 1)]$

$\alpha = \sum_i (e^{h_i} - 1) \left[P_a \left(1 - \frac{i}{L}\right) e^{\sum_{k=1}^{i-1} \lambda_k} + P_b \frac{i}{L+1} e^{-\sum_{k=i}^{L-1} \lambda_k} \right]$

Hydrodynamic limit :

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$$i \rightarrow xL$$

$$T \rightarrow tL^2$$

$$n_i(T) \rightarrow \theta(x,t)$$

$$Q_i(T) \rightarrow L \cdot \int_0^T dt \dot{f}(x,t)$$

$$A_i \rightarrow \alpha(x)/L$$

$$h_i \rightarrow h(x)$$

$$\mu \rightarrow \frac{1}{L} G\left(\int_0^1 dx \alpha(x)\right) = \frac{1}{L} \left[P_a \left(e^{\int_0^1 dx \alpha(x)} - 1 \right) + P_b \left(e^{-\int_0^1 dx \alpha(x)} - 1 \right) \right]$$

$$\alpha \rightarrow L \int_0^1 dx \cdot (e^{h(x)} - 1) \left[P_a (1-x) e^{\int_0^x dy \alpha(y)} + P_b \cdot x \cdot e^{-\int_x^1 dy \alpha(y)} \right]$$

$$= L \Psi[h(x), \alpha(x)]$$

$$\Rightarrow \ln \langle e^{\sum A_i Q_i + \sum h_i n_i} \rangle \longrightarrow \ln \left\langle e^{L \int_0^1 dx \alpha(x) \int_0^T dt \dot{f}(x,t) + L \int dx h(x) \theta(x)} \right\rangle$$
$$= e^{\mu \cdot L G(\int dx \alpha(x)) + L \Psi(h, \alpha)}$$

⊙ Remark : for $h=0$, one gets back ~~an~~ additivity result

for $\alpha=0$, one gets back density LDF result.

Remark: Taking Legendre transform

$$\frac{\frac{1}{\tau} \int_0^1 dx \alpha(x) \int_0^\tau dt j(x,t)}{\int_0^1 dx \alpha(x)} = q_0$$

$$P[q_0, \tau(x)] \propto e^{-\tau \cdot L \cdot \phi(q_0) - L \cdot \Psi[\tau(x), q_0, \alpha(x)]}$$

Consistency

Conditional probability

$$P[q_0, \tau(x)] = P[\tau(x) | q_0] P(q_0) \hookrightarrow e^{-\tau L \phi(q_0)}$$

$$\Rightarrow P[\tau(x) | q_0] \propto e^{-L \Psi[\tau(x), q_0, \alpha(x)]}$$

Large-deviation function condition on an empirical measure!

Remark: one can write H-J eqⁿ for $\Psi[]$

$$\int_0^1 dx \left\{ \frac{\sigma(\tau)}{2} \left(\partial_x \frac{\delta \Psi}{\delta \sigma} + \frac{d(x)}{dx} \cdot \phi'(q_0) \right)^2 - D(\tau) \partial_x \tau \cdot \left(\partial_x \frac{\delta \Psi}{\delta \sigma} + \frac{\alpha(x)}{\int dx \alpha} \cdot \phi'(q_0) \right) \right\} = q_0 \cdot \phi'(q_0) - \phi(q_0)$$

* Verify that NE satisfies

* Verify at $\phi'(q_0) = 0$ one gets back H-J equation for density.