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Breakdown of the hydrodynamic limit for extreme current fluctuations

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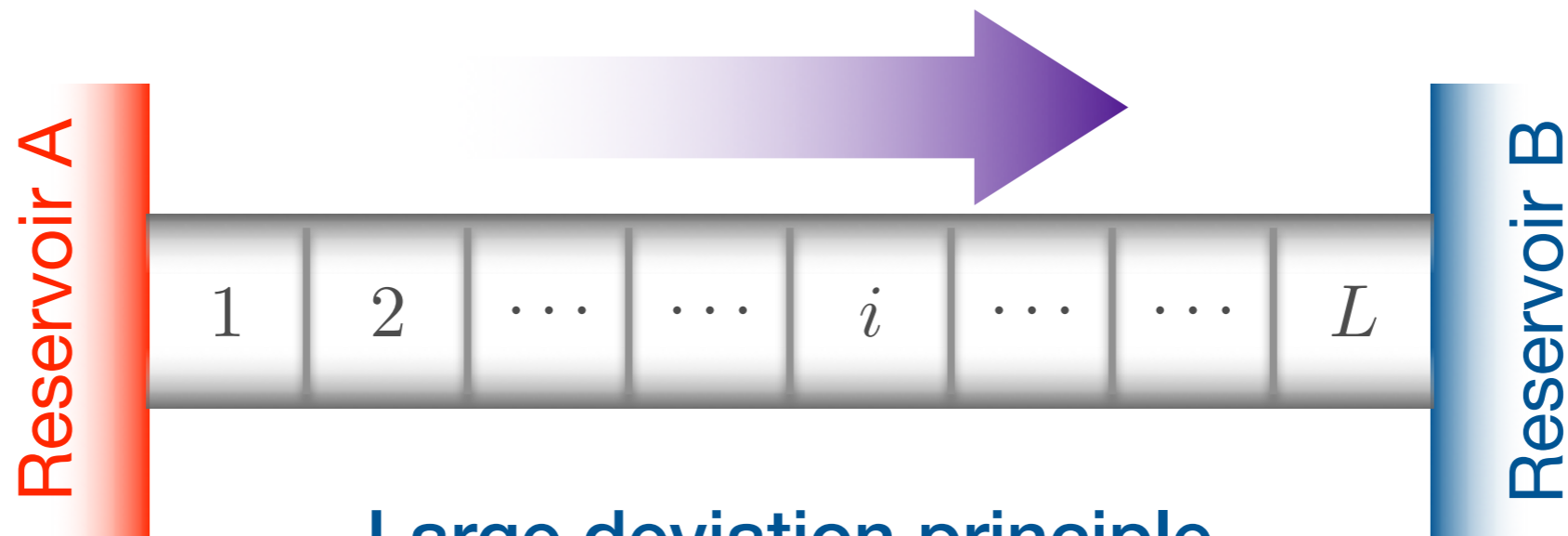
Introduction

Current fluctuations in the hydrodynamic limit

Boundary-driven systems

Time-averaged current

$$J = \frac{Q_B}{T} \quad \begin{array}{l} \text{(Total flux to B)} \\ \text{(Observation time)} \end{array}$$



Large deviation principle

For $T \gg 1$, $P(J) \sim e^{-T \Phi(J)}$ Large deviation function (LDF)

Cumulant generating function

$$\langle e^{\lambda Q_B} \rangle = e^{T \underline{\mu(\lambda)}}, \quad \Phi(J) = \sup_{\lambda} [\lambda J - \underline{\mu(\lambda)}]$$

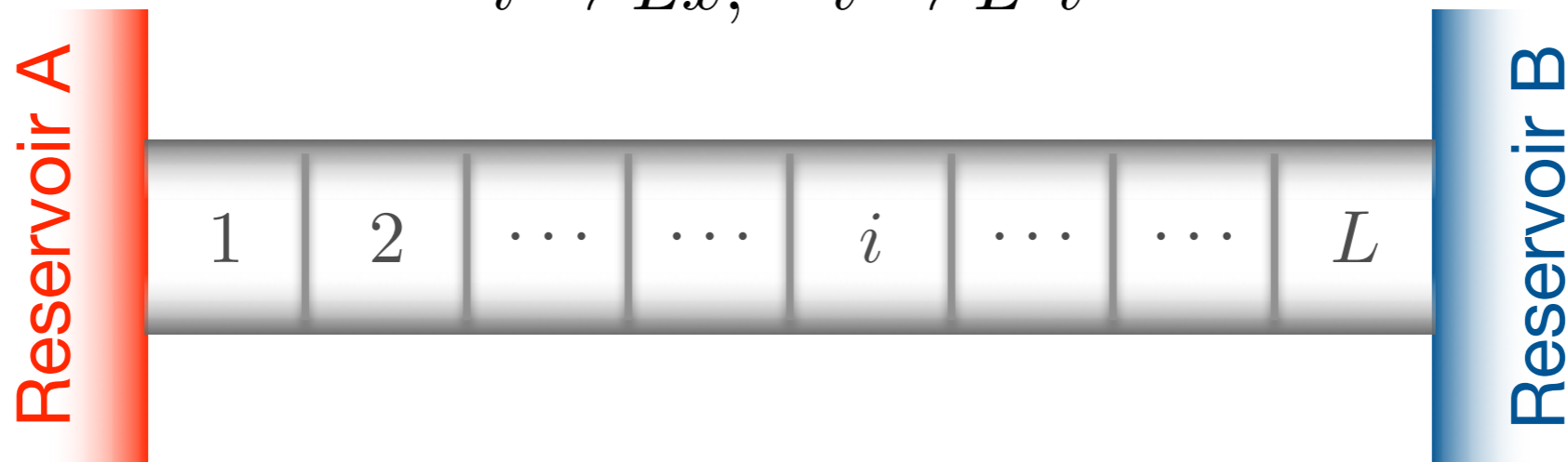
Hydrodynamic limit

Goal: obtain
 $\mu(\lambda)$ and $\Phi(J)$

$L \rightarrow \infty$, local equilibrium

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$



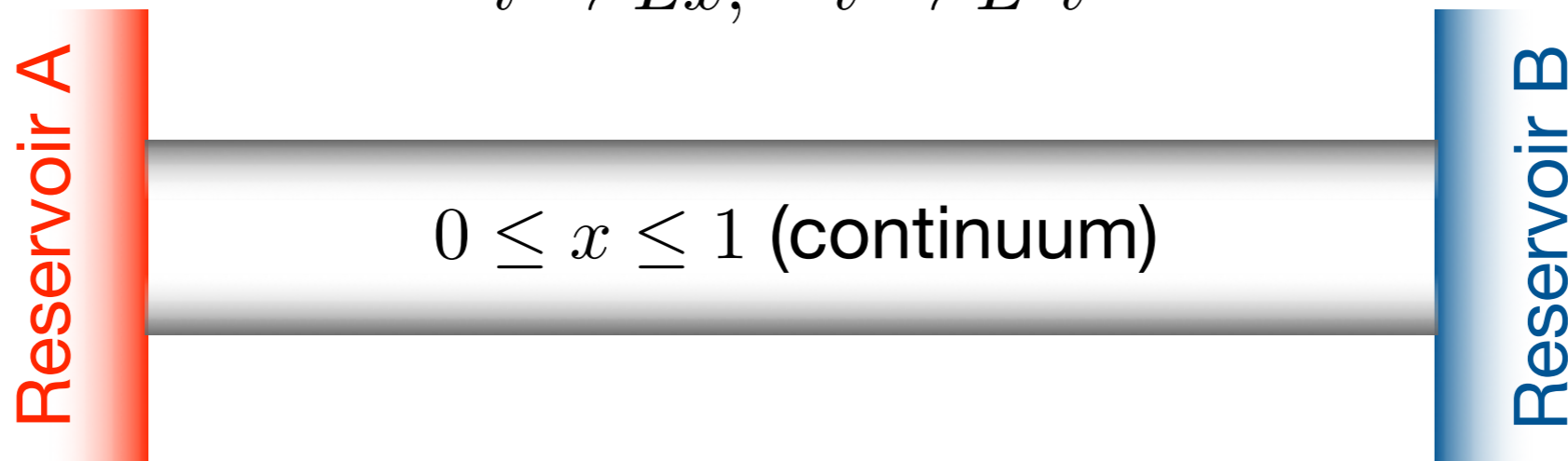
Hydrodynamic limit

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

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Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[-D(\rho) \partial_x \rho + \sqrt{\sigma(\rho)} \eta \right]$$

$$\langle \eta(x, t) \eta(x', t') \rangle = L^{-1} \delta(x - x') \delta(t - t')$$

Hydrodynamic limit

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

$L \rightarrow \infty$, local equilibrium

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Boundary
condition:
 $\rho(0) = \bar{\rho}_A$

$$0 \leq x \leq 1 \text{ (continuum)}$$

Boundary
condition:
 $\rho(1) = \bar{\rho}_B$

Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[\underbrace{-D(\rho)}_{\text{Diffusivity}} \partial_x \rho + \underbrace{\sqrt{\sigma(\rho)}}_{\text{Mobility}} \eta \right]$$

$$\langle \eta(x, t) \eta(x', t') \rangle = \underline{L^{-1}} \delta(x - x') \delta(t - t')$$

Weak noise
(microscopic origin)

Calculation of current LDFs

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

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Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[-D(\rho) \partial_x \rho + \sqrt{\sigma(\rho)} \eta \right]$$

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Microscopic approaches (take the limit later)

Derrida, Douçot, Roche, J. Stat. Phys. **115**, 717 (2004)

Bodineau & Derrida, Phys. Rev. Lett. **92**, 180601 (2004)

Macroscopic fluctuation theory (take the limit first)

Bertini et al., Phys. Rev. Lett. **94**, 030601 (2005)

Bertini et al., J. Stat. Phys. **123**, 237 (2006)

Simplification
by saddle-point
approximation

Macroscopic fluctuation theory

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[-D(\rho) \partial_x \rho + \sqrt{\sigma(\rho)} \eta \right]$$

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$$0 \leq x \leq 1 \text{ (continuum)}$$

Boundary condition:
 $\rho(1) = \bar{\rho}_B$

$= j$

1. Express the CGF in a path integral form.

$$\langle e^{\lambda Q_B} \rangle = \int \mathcal{D}\rho \left\langle \underbrace{\delta(\partial_t \rho + \partial_x j)}_{\text{Hydrodynamic equation}} e^{\frac{L \int_0^{T/L^2} dt \int_0^1 dx \lambda j}{= \lambda Q_B}} \right\rangle$$

Fluctuating current

j

Macroscopic fluctuation theory

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[\underbrace{-D(\rho)\partial_x \rho + \sqrt{\sigma(\rho)}\eta}_{=j} \right]$$

Boundary
condition:
 $\rho(0) = \bar{\rho}_A$

$$0 \leq x \leq 1 \text{ (continuum)}$$

Boundary
condition:
 $\rho(1) = \bar{\rho}_B$

2. Introduce the conjugate field $\hat{\rho}$.

$$\begin{aligned} \langle e^{\lambda Q_B} \rangle &= \int \mathcal{D}\rho \left\langle \underbrace{\delta(\partial_t \rho + \partial_x j)}_{=} e^{L \int_0^{T/L^2} dt \int_0^1 dx \lambda j} \right\rangle_j \\ &= \int \mathcal{D}\hat{\rho} e^{-L \int_0^{T/L^2} dt \hat{\rho}(\partial_t \rho + \partial_x j)} \end{aligned}$$

Macroscopic fluctuation theory

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[\underbrace{-D(\rho)\partial_x \rho + \sqrt{\sigma(\rho)}\eta}_{=j} \right]$$

Boundary condition:
 $\rho(0) = \bar{\rho}_A$
 $\hat{\rho}(0) = 0$

$$0 \leq x \leq 1 \text{ (continuum)}$$

Boundary condition:
 $\rho(1) = \bar{\rho}_B$
 $\hat{\rho}(1) = \lambda$

3. Average over the fluctuating current.

$$\langle e^{\lambda Q_B} \rangle = \int \mathcal{D}\rho \mathcal{D}\hat{\rho} \left\langle e^{-L \int_0^{T/L^2} dt \int_0^1 dx (\hat{\rho} \partial_t \rho - j \partial_x \hat{\rho})} \right\rangle_j$$

Macroscopic fluctuation theory

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[-D(\rho) \partial_x \rho + \sqrt{\sigma(\rho)} \eta \right]$$

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condition:
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 $\hat{\rho}(1) = \lambda$

4. Apply the saddle-point approximation.

$$e^{T\mu(\lambda)} = \int \mathcal{D}\rho \mathcal{D}\hat{\rho} e^{-\frac{L}{L \gg 1} \int_0^{T/L^2} dt \int_0^1 dx [\hat{\rho} \partial_t \rho - \underline{H(\rho, \hat{\rho})}]} \quad H(\rho, \hat{\rho}) = -D(\rho) (\partial_x \rho) (\partial_x \hat{\rho}) + \frac{1}{2} \sigma(\rho) (\partial_x \hat{\rho})^2$$

Macroscopic fluctuation theory

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[-D(\rho) \partial_x \rho + \sqrt{\sigma(\rho)} \eta \right]$$

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 $\hat{\rho}(0) = 0$

$$0 \leq x \leq 1 \text{ (continuum)}$$

Boundary condition:
 $\rho(1) = \bar{\rho}_B$
 $\hat{\rho}(1) = \lambda$

5. Derive Hamilton's equations.

$$T\mu(\lambda) = -L \min_{\rho, \hat{\rho}} \int_0^{T/L^2} dt \int_0^1 dx \left[\hat{\rho} \partial_t \rho - \underline{H(\rho, \hat{\rho})} \right]$$

$$H(\rho, \hat{\rho}) = -D(\rho) (\partial_x \rho) (\partial_x \hat{\rho}) + \frac{1}{2} \sigma(\rho) (\partial_x \hat{\rho})^2$$

Position Momentum



Macroscopic fluctuation theory

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[-D(\rho) \partial_x \rho + \sqrt{\sigma(\rho)} \eta \right]$$

Boundary condition:
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$$0 \leq x \leq 1 \text{ (continuum)}$$

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5. Derive Hamilton's equations. $H(\rho, \hat{\rho}) = -D(\rho)(\partial_x \rho)(\partial_x \hat{\rho}) + \frac{1}{2} \sigma(\rho)(\partial_x \hat{\rho})^2$

$$T\mu(\lambda) = -L \min_{\rho, \hat{\rho}} \int_0^{T/L^2} dt \int_0^1 dx [\hat{\rho} \partial_t \rho - H(\rho, \hat{\rho})]$$

$$\partial_t \rho = \frac{\partial H}{\partial \hat{\rho}} = -\partial_x [-D(\rho) \partial_x \rho + \sigma(\rho) \partial_x \hat{\rho}] \quad \partial_t \hat{\rho} = -\frac{\partial H}{\partial \rho}$$

$\partial_x \hat{\rho} = \text{noise realization}$

Macroscopic fluctuation theory

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

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Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[-D(\rho) \partial_x \rho + \sqrt{\sigma(\rho)} \eta \right]$$

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$$0 \leq x \leq 1 \text{ (continuum)}$$

Boundary
condition:
 $\rho(1) = \bar{\rho}_B$
 $\hat{\rho}(1) = \lambda$

6. Assume that the minimizing solution is stationary.

$$T\mu(\lambda) = -L \min_{\rho, \hat{\rho}} \int_0^{T/L^2} dt \int_0^1 dx [\hat{\rho} \partial_t \rho - H(\rho, \hat{\rho})]$$

$$\partial_t \rho = \frac{\partial H}{\partial \hat{\rho}} = 0, \quad \partial_t \hat{\rho} = -\frac{\partial H}{\partial \rho} = 0 \quad \longrightarrow \quad \rho(x, t) = \rho_\lambda^*(x), \quad \hat{\rho}(x, t) = \hat{\rho}_\lambda^*(x)$$

Bodineau & Derrida, Phys. Rev. Lett. **92**, 180601 (2004)

Macroscopic fluctuation theory

Goal: obtain $\mu(\lambda)$ and $\Phi(J)$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Fluctuating hydrodynamics

$$\partial_t \rho = -\partial_x \left[-D(\rho) \partial_x \rho + \sqrt{\sigma(\rho)} \eta \right]$$

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7. Evaluate the Hamiltonian at the “optimal profile”.

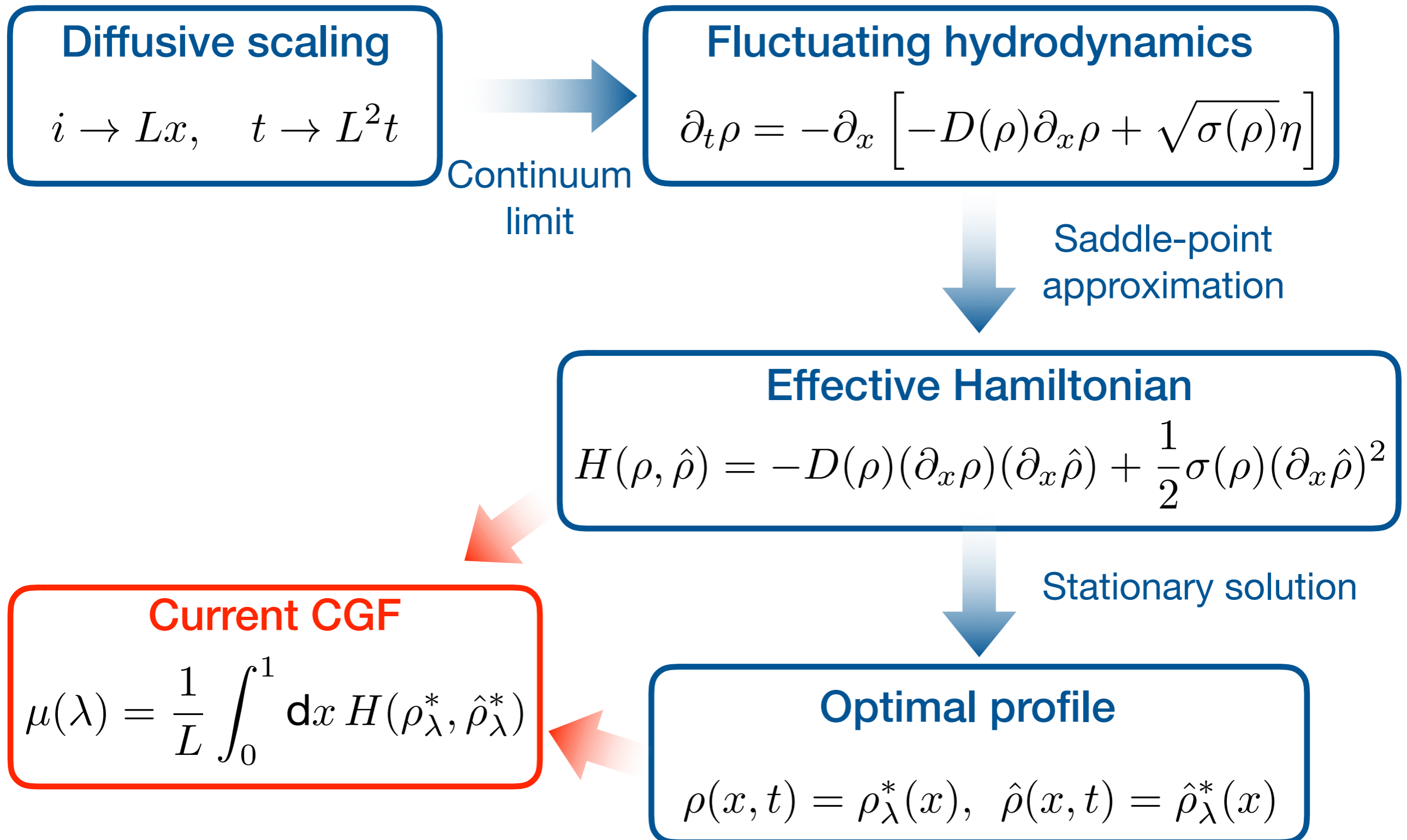
Current CGF

$$\mu(\lambda) = \frac{1}{L} \int_0^1 dx H(\rho_\lambda^*, \hat{\rho}_\lambda^*)$$

Optimal profile

$$\rho(x, t) = \rho_\lambda^*(x), \quad \hat{\rho}(x, t) = \hat{\rho}_\lambda^*(x)$$

Macroscopic fluctuation theory



Current fluctuations in the hydrodynamic limit

Current CGF

For $D(\rho) = 1$ and $\sigma(\rho) = c_2\rho^2 + c_1\rho$,

$$\mu(\lambda) = \frac{1}{L} \int_0^1 dx H(\rho_\lambda^*, \hat{\rho}_\lambda^*) = \begin{cases} -\frac{2}{Lc_2} (\operatorname{arcsinh} \sqrt{\omega})^2 & \text{for } \omega > 0 \\ +\frac{2}{Lc_2} (\operatorname{arcsin} \sqrt{-\omega})^2 & \text{for } \omega < 0 \end{cases}$$

$$\omega(\lambda, \bar{\rho}_A, \bar{\rho}_B) = \frac{c_2}{c_1^2} (1 - e^{-c_1\lambda/2}) [c_1(\bar{\rho}_B - e^{c_1\lambda/2}\bar{\rho}_A) - c_2(e^{c_1\lambda/2} - 1)\bar{\rho}_A\bar{\rho}_B]$$

Imparato, Lecomte, van Wijland, Phys. Rev. E **80**, 011131 (2009)

Current LDF $\Phi(J) = \sup_{\lambda} [\lambda J - \mu(\lambda)]$

Symmetric exclusion process

(SEP: $c_1 = 2, c_2 = -2$)

$\Phi(J) \sim J^2$ (Gaussian)

Kipnis–Marchioro–Presutti model

(KMP: $c_1 = 0, c_2 = 2$)

$\Phi(J) \sim J$ (Exponential)

Current fluctuations in the hydrodynamic limit

Current CGF

$$\mu(\lambda) = \frac{1}{L} \int_0^1 dx H(\rho_\lambda^*, \hat{\rho}_\lambda^*)$$

Current LDF

$$\Phi(J) = \sup_\lambda [\lambda J - \mu(\lambda)]$$

$$\text{In any case, } J(\lambda) = \frac{d\mu}{d\lambda} = O(L^{-1})$$

Interpretation

In the hydrodynamic limit, current fluctuations are always comparable to the **average density gradient**.

Questions

- ◆ **How** can we study even larger current fluctuations?
- ◆ Can extreme current fluctuations induce **non-hydrodynamic** tail behaviors?

Current fluctuations in the hydrodynamic limit

Questions

- ◆ **How** can we study even larger current fluctuations?
- ◆ Can extreme current fluctuations induce **non-hydrodynamic** tail behaviors?

Objectives

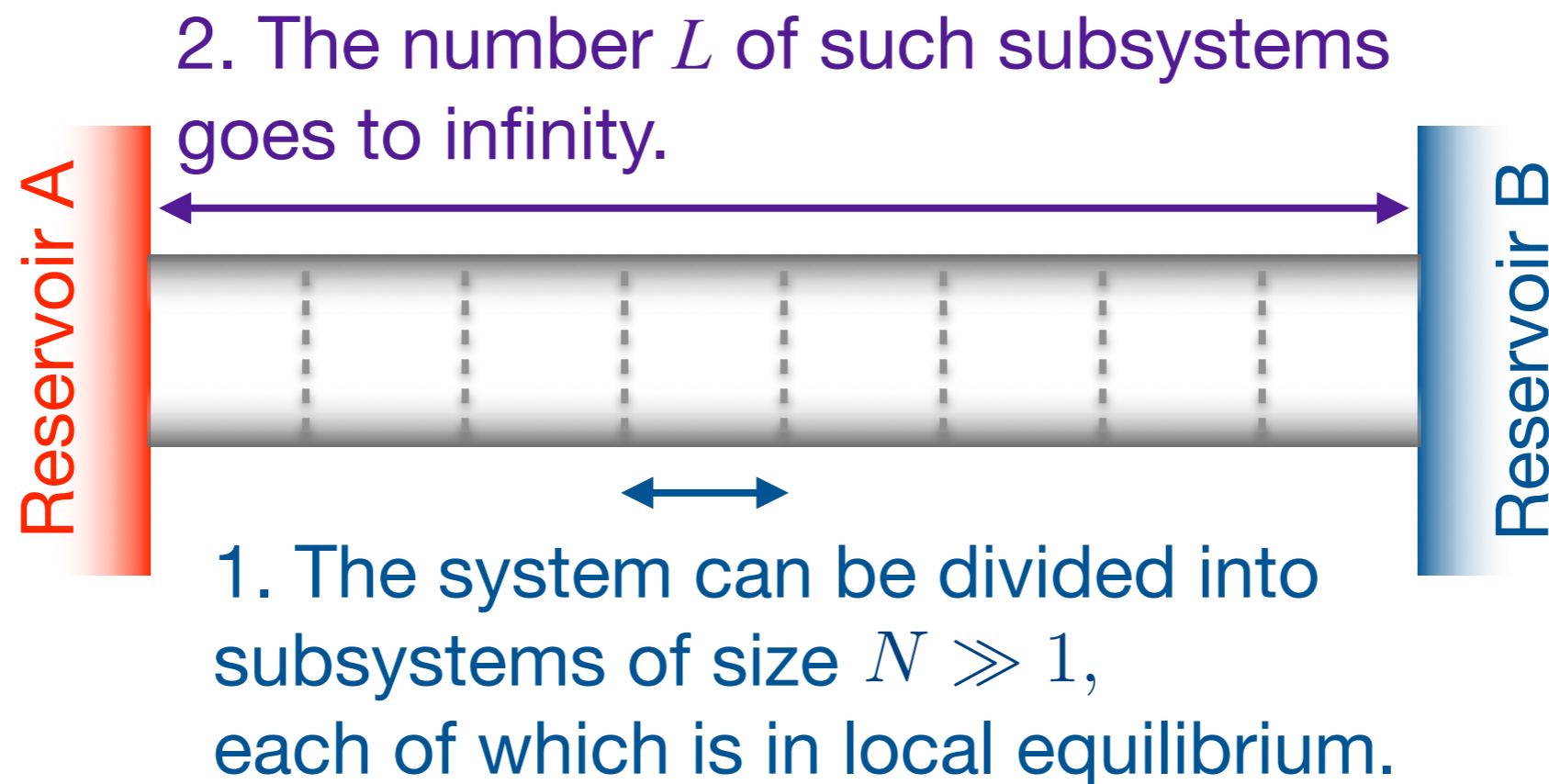
- ◆ Develop a **rescaling scheme** which enables **saddle-point techniques** but does not directly lead to the hydrodynamic limit.
- ◆ Check if the rescaling scheme can **recover the hydrodynamic limit** under appropriate conditions.
- ◆ Obtain current CGFs assuming **stationary saddle-point solutions**.

Rest of the talk

- ◆ Large N limit
- ◆ Current fluctuations of the SEP-like model
- ◆ Current fluctuations of the KMP-like model
- ◆ Properties of optimal profiles
- ◆ Conclusions and future works

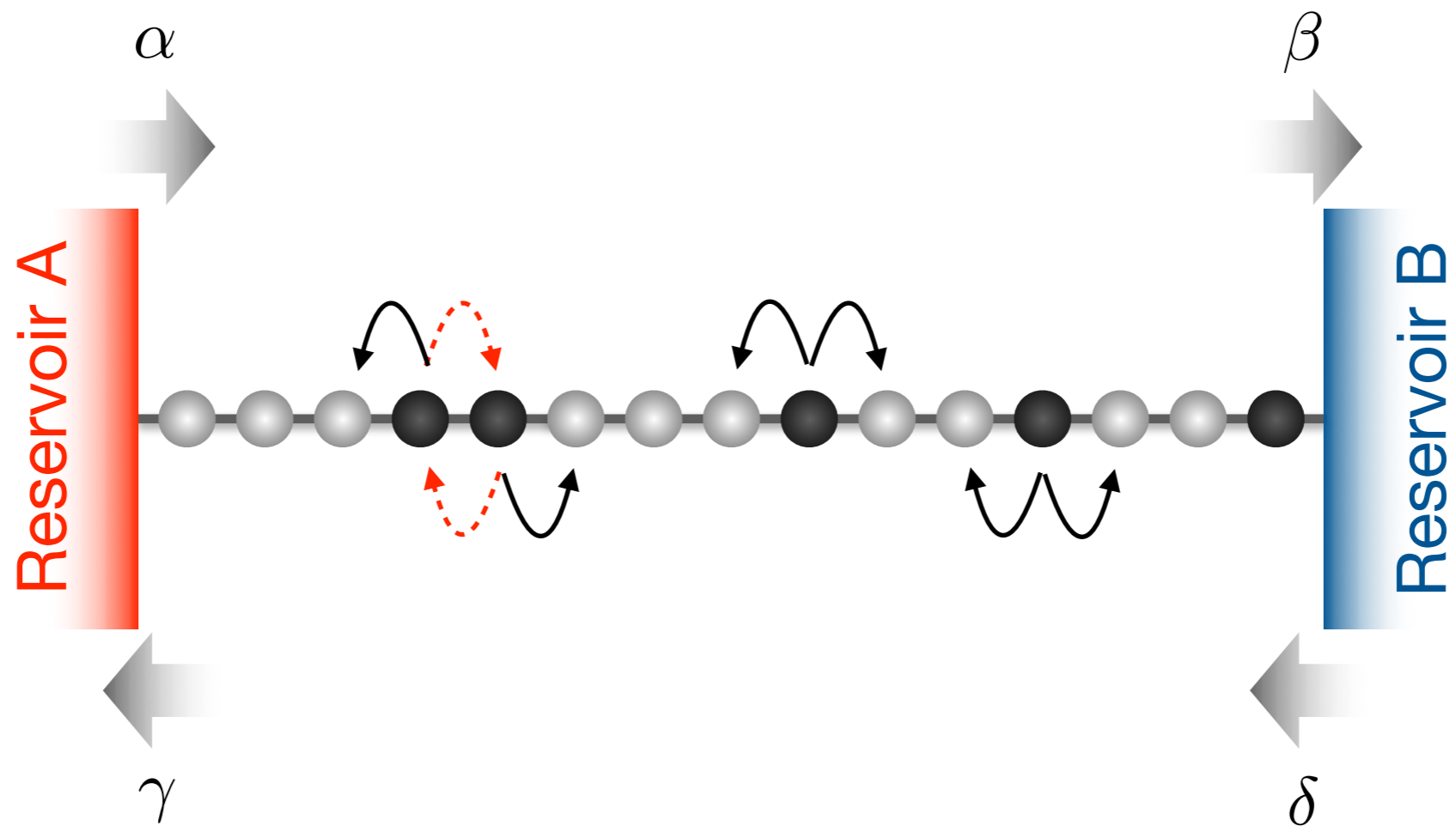
Revisiting the hydrodynamic limit

Hydrodynamic limit involves two limiting processes.

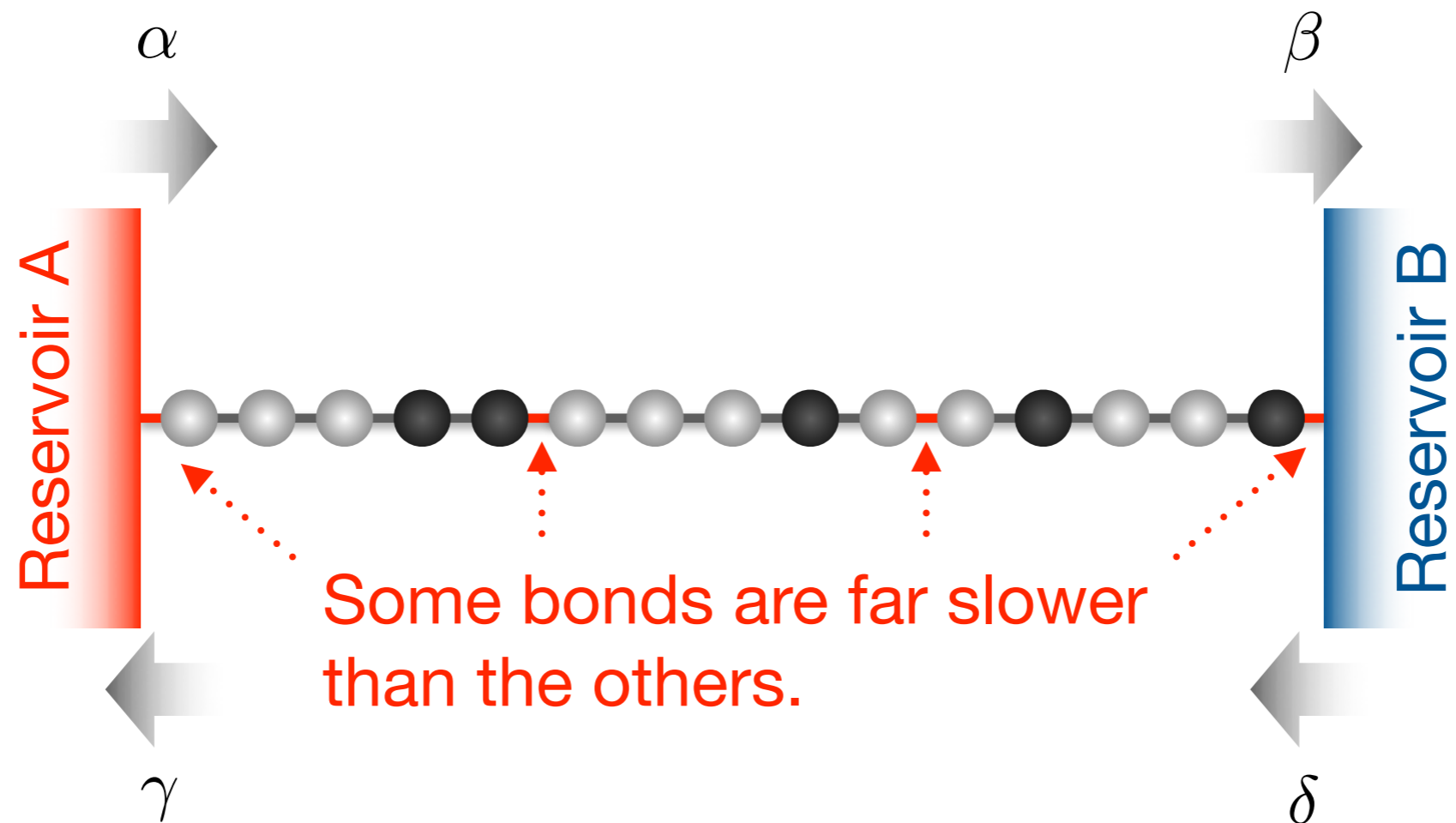


Idea: take only the first limit, leave the second limit untaken.

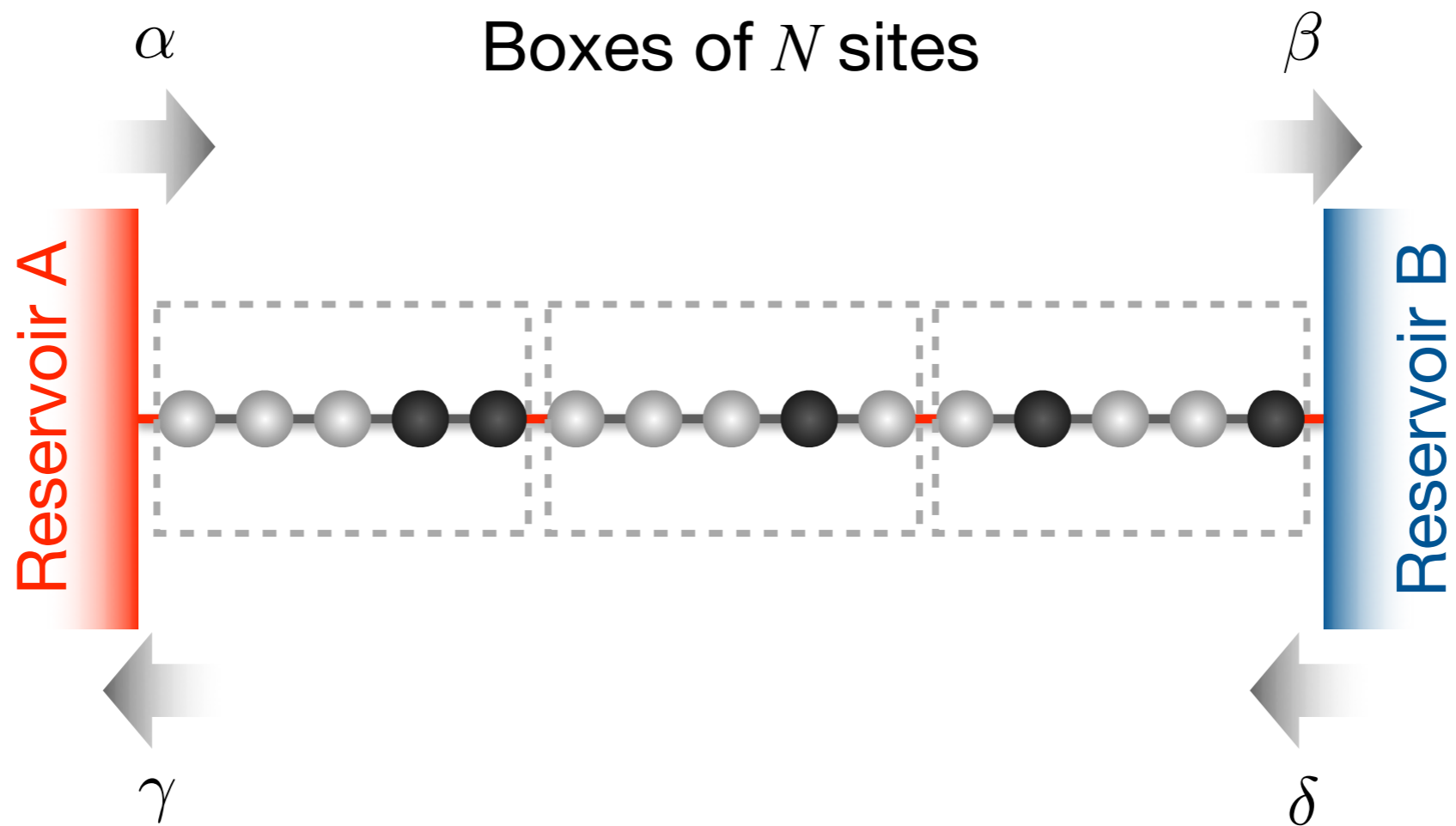
Example: coarse-grained SEP



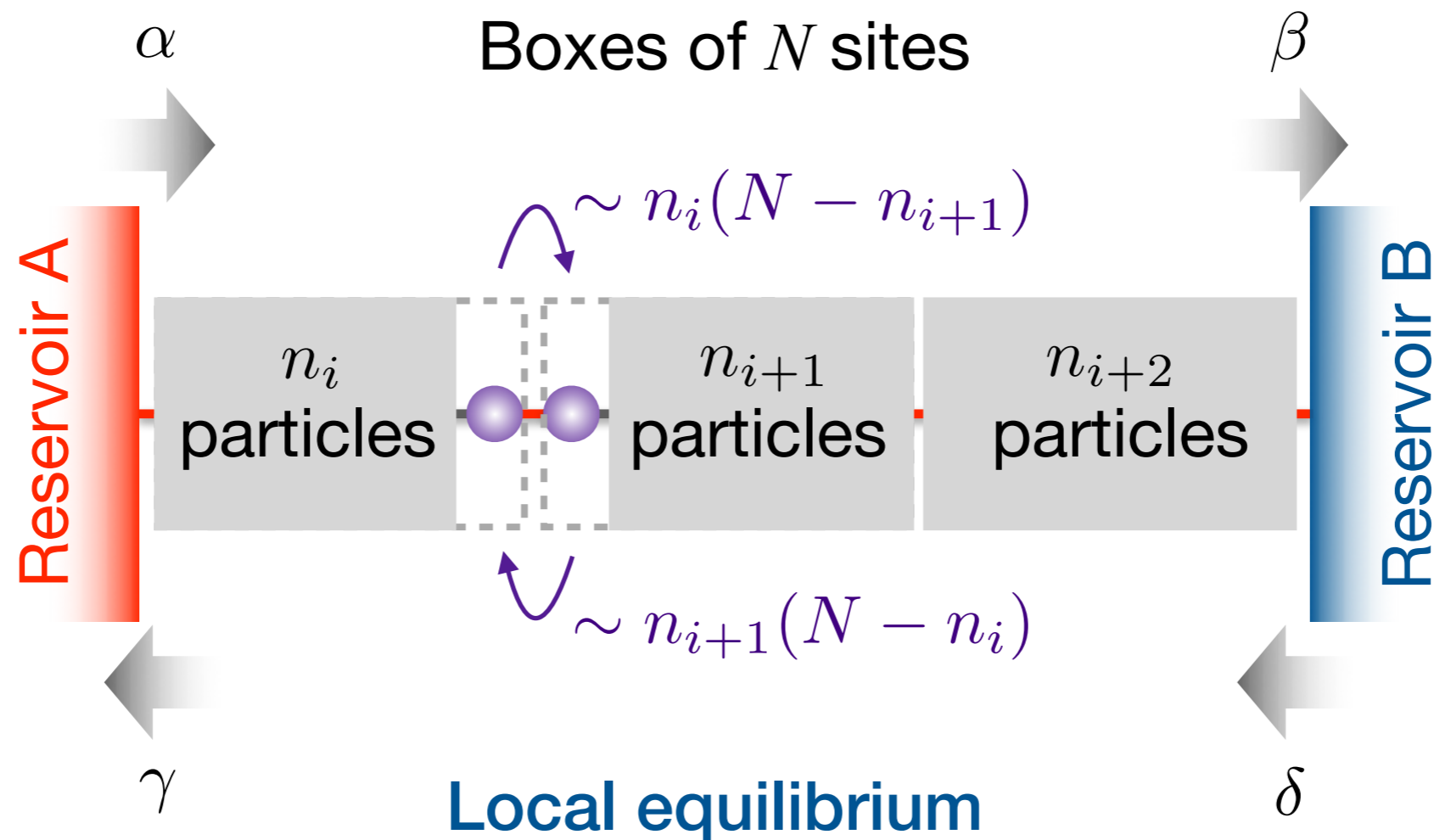
Example: coarse-grained SEP



Example: coarse-grained SEP

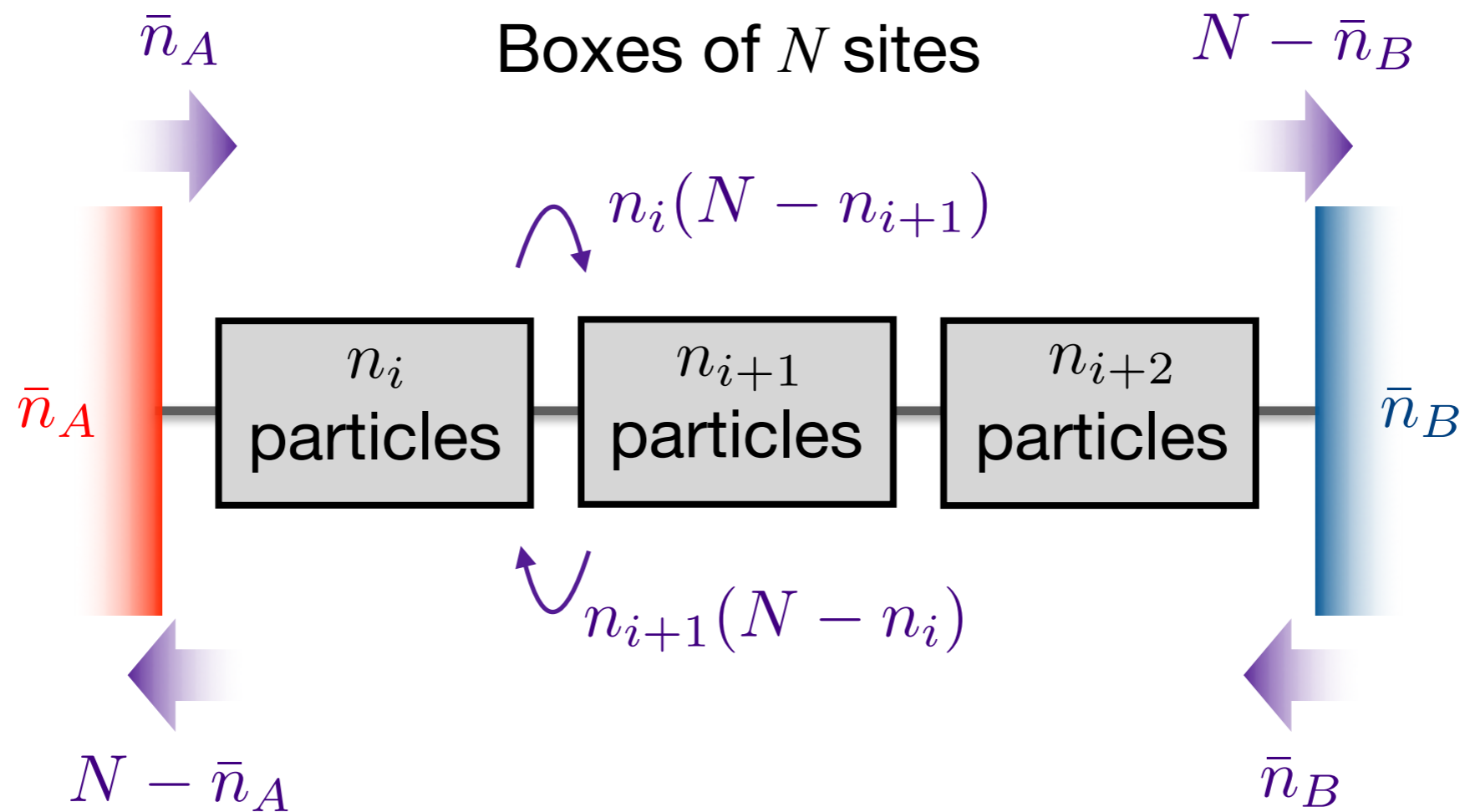


Example: coarse-grained SEP



For each box, any configuration with a fixed number of particles is equally likely.

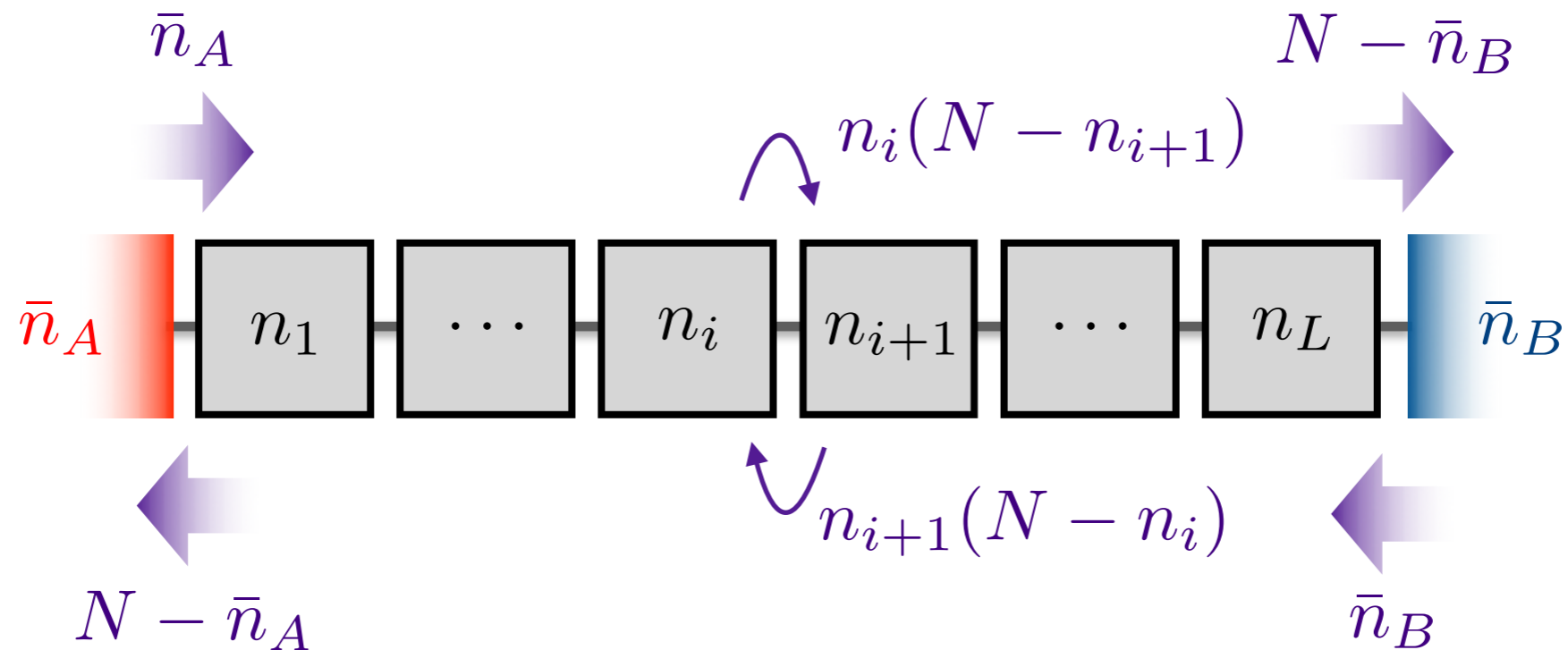
Example: coarse-grained SEP



Coarse-grained dynamics

(with appropriate boundary rates and rescaling of time)

Example: coarse-grained SEP



Propagator for the number configuration

$$P[\mathbf{n}_f, t_f | \mathbf{n}_i, t_i] = \int D\mathbf{n} D\hat{\mathbf{n}} \exp \left\{ - \int_{t_i}^{t_f} dt [\hat{\mathbf{n}} \cdot \dot{\mathbf{n}} - H(\mathbf{n}, \hat{\mathbf{n}})] \right\}$$

Propagator for the number configuration

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Rescaling to density

$$\bar{n}_{A,B} \rightarrow N \bar{\rho}_{A,B}$$

$$n_i \rightarrow N \rho_i$$

$$\hat{n}_i \rightarrow \hat{\rho}_i$$

$$t \rightarrow N^{-1} t$$

Propagator for the density configuration

$$P[\boldsymbol{\rho}_f, t_f | \boldsymbol{\rho}_i, t_i] = \int \mathcal{D}\boldsymbol{\rho} \mathcal{D}\hat{\boldsymbol{\rho}} \exp \left\{ - \underline{N} \int_{t_i}^{t_f} dt [\hat{\boldsymbol{\rho}} \cdot \dot{\boldsymbol{\rho}} - H(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}})] \right\}$$

$N \gg 1$ enables the saddle-point approximation

Pre-Hamiltonian

Large N limit

$$H(\mathbf{n}, \hat{\mathbf{n}}) \equiv \sum_{i=1}^{L-1} \left[n_i(N - n_{i+1}) (e^{\hat{n}_{i+1} - \hat{n}_i} - 1) + n_{i+1}(N - n_i) (e^{\hat{n}_i - \hat{n}_{i+1}} - 1) \right] \\ + n_1(N - \bar{n}_A) (e^{-\hat{n}_1} - 1) + \bar{n}_A(N - n_1) (e^{\hat{n}_1} - 1) \\ + n_L(N - \bar{n}_B) (e^{-\hat{n}_L} - 1) + \bar{n}_B(N - n_L) (e^{\hat{n}_L} - 1)$$

Rescaling to density

$$\bar{n}_{A,B} \rightarrow N \bar{\rho}_{A,B}$$

$$n_i \rightarrow N \rho_i$$

$$\hat{n}_i \rightarrow \hat{\rho}_i$$

$$t \rightarrow N^{-1} t$$

Effective Hamiltonian ($N \gg 1$)

$$H(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}}) = \sum_{i=1}^{L-1} \left[\rho_i(1 - \rho_{i+1}) (e^{\hat{\rho}_{i+1} - \hat{\rho}_i} - 1) + \rho_{i+1}(1 - \rho_i) (e^{\hat{\rho}_i - \hat{\rho}_{i+1}} - 1) \right] \\ + \rho_1(1 - \bar{\rho}_A) (e^{-\hat{\rho}_1} - 1) + \bar{\rho}_A(1 - \rho_1) (e^{\hat{\rho}_1} - 1) \\ + \rho_L(1 - \bar{\rho}_B) (e^{-\hat{\rho}_L} - 1) + \bar{\rho}_B(1 - \rho_L) (e^{\hat{\rho}_L} - 1)$$

Effective Hamiltonian ($N \gg 1$)

Large N limit

$$H(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}}) = \sum_{i=1}^{L-1} \left[\rho_i(1 - \rho_{i+1}) (e^{\hat{\rho}_{i+1} - \hat{\rho}_i} - 1) + \rho_{i+1}(1 - \rho_i) (e^{\hat{\rho}_i - \hat{\rho}_{i+1}} - 1) \right] \\ + \rho_1(1 - \bar{\rho}_A) (e^{-\hat{\rho}_1} - 1) + \bar{\rho}_A(1 - \rho_1) (e^{\hat{\rho}_1} - 1) \\ + \rho_L(1 - \bar{\rho}_B) (e^{-\hat{\rho}_L} - 1) + \bar{\rho}_B(1 - \rho_L) (e^{\hat{\rho}_L} - 1)$$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2t$$

Continuum limit

$$X_{i+1} - X_i \sim L^{-1} \\ \rho_0 = \bar{\rho}_A, \quad \rho_{L+1} = \bar{\rho}_B \\ \hat{\rho}_0 = \hat{\rho}_{L+1} = 0$$

Hydrodynamic limit ($N \gg 1, L \gg 1$)

$$H(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}}) = \int_0^1 dx \left[-(\partial_x \rho)(\partial_x \hat{\rho}) + \rho(1 - \rho)(\partial_x \hat{\rho})^2 \right]$$

$D(\rho) = 1, \sigma(\rho) = \rho(1 - \rho)$: this model is indeed SEP-like.

$$\bar{n}_{A,B} \rightarrow N \bar{\rho}_{A,B}$$

$$n_i \rightarrow N \rho_i$$

$$\hat{n}_i \rightarrow \hat{\rho}_i$$

$$t \rightarrow N^{-1} t$$

$$i \rightarrow Lx$$

$$t \rightarrow L^2 t$$

Microscopic
box dynamics

Large N limit
(discrete space)

Hydrodynamic limit
(continuum)

Objectives

- ✓ Develop a **rescaling scheme** which enables **saddle-point techniques** but does not directly lead to the hydrodynamic limit.
- ✓ Check if the rescaling scheme can **recover the hydrodynamic limit** under appropriate conditions.
- ◆ Obtain current CGFs assuming **stationary saddle-point solutions**.

$$\bar{n}_{A,B} \rightarrow N \bar{\rho}_{A,B}$$

$$n_i \rightarrow N \rho_i$$

$$\hat{n}_i \rightarrow \hat{\rho}_i$$

$$t \rightarrow N^{-1} t$$

$$i \rightarrow Lx$$

$$t \rightarrow L^2 t$$

Microscopic
box dynamics

Large N limit
(discrete space)

Hydrodynamic limit
(continuum)

Other works using the large N limit

- ◆ Population dynamics

Meerson & Sasorov, Phys. Rev. E **83**, 011129 (2011)

- ◆ Spin- j representation of SEP and KMP ($j = N/2 \rightarrow \infty$)

Tailleur, Kurchan, Lecomte, J. Phys. A **41**, 505001 (2008)

Objectives

- ✓ Develop a rescaling scheme which enables saddle-point techniques but does not directly lead to the hydrodynamic limit.
- ✓ Check if the rescaling scheme can recover the hydrodynamic limit under appropriate conditions.
- ◆ Obtain current CGFs assuming stationary saddle-point solutions.

Calculation of the current CGF

Rescaling

$$\begin{aligned}\bar{n}_{A,B} &\rightarrow N\bar{\rho}_{A,B} \\ n_i &\rightarrow N\rho_i \\ \hat{n}_i &\rightarrow \hat{\rho}_i \\ t &\rightarrow N^{-1}t\end{aligned}$$

Saddle-point
approximation

Current CGF

$$\mu_L(\lambda) = H_\lambda(\boldsymbol{\rho}_\lambda^*, \hat{\boldsymbol{\rho}}_\lambda^*)$$

Optimal profile

$$\rho_i(t) = \rho_{\lambda,i}^*, \hat{\rho}_i(t) = \hat{\rho}_{\lambda,i}^*$$

Stationary solution

$$\begin{aligned}H_\lambda(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}}) &= \sum_{i=1}^{L-1} \left[\rho_i(1 - \rho_{i+1}) (e^{\hat{\rho}_{i+1} - \hat{\rho}_i} - 1) + \rho_{i+1}(1 - \rho_i) (e^{\hat{\rho}_i - \hat{\rho}_{i+1}} - 1) \right] \\ &+ \rho_1(1 - \bar{\rho}_A) (e^{-\hat{\rho}_1} - 1) + \bar{\rho}_A(1 - \rho_1) (e^{\hat{\rho}_1} - 1) \\ &+ \rho_L(1 - \bar{\rho}_B) (e^{-\hat{\rho}_L + \lambda} - 1) + \bar{\rho}_B(1 - \rho_L) (e^{\hat{\rho}_L - \lambda} - 1)\end{aligned}$$

Behaviors of the current CGF

Result for the large N limit

$$\longrightarrow \Phi(J) \sim J \ln J$$

$$\mu_L(\lambda) = \begin{cases} (L+1) \sinh^2 \left(\frac{1}{L+1} \operatorname{arcsinh} \sqrt{\omega} \right) & \text{for } \omega > 0 \\ -(L+1) \sin^2 \left(\frac{1}{L+1} \operatorname{arcsin} \sqrt{-\omega} \right) & \text{for } \omega < 0 \end{cases}$$

Hydrodynamic result

$$\longrightarrow \Phi(J) \sim J^2$$

$$\mu(\lambda) = \begin{cases} \frac{1}{L+1} \operatorname{arcsinh}^2 \sqrt{\omega} & \text{for } \omega > 0 \\ -\frac{1}{L+1} \operatorname{arcsinh}^2 \sqrt{-\omega} & \text{for } \omega < 0 \end{cases}$$

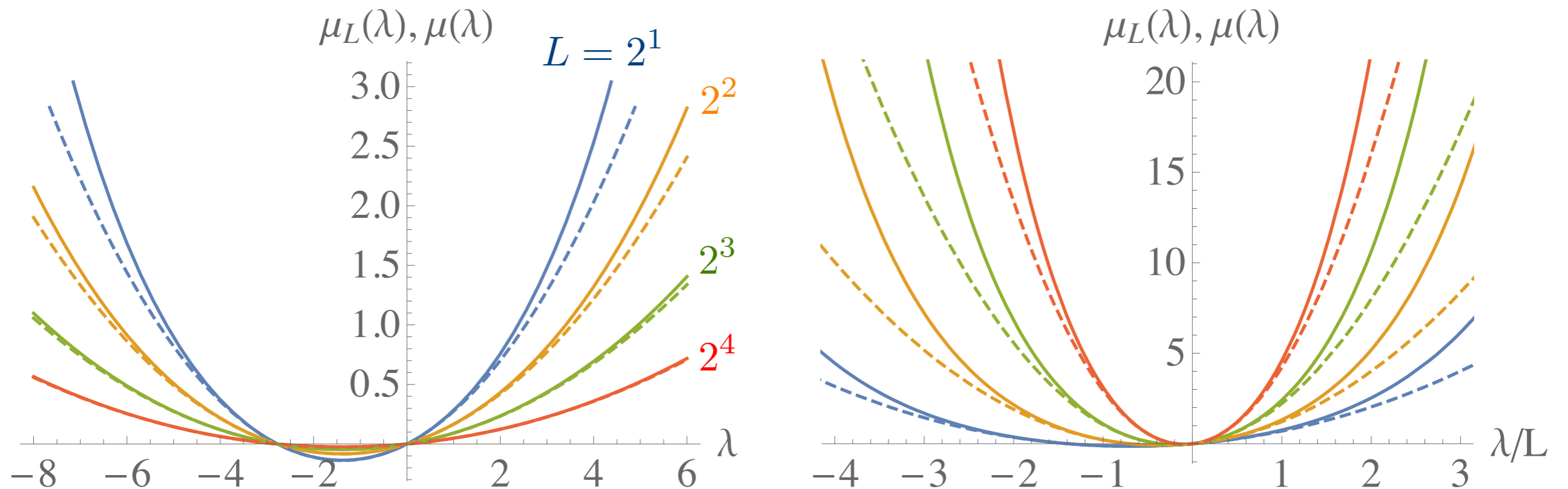
Auxiliary variable for both cases

$$\omega(\lambda, \bar{\rho}_A, \bar{\rho}_B) = (1 - e^{-\lambda})[\bar{\rho}_B - e^{\lambda} \bar{\rho}_A + (e^{\lambda} - 1) \bar{\rho}_A \bar{\rho}_B]$$

Behaviors of the current CGF

Comparison between the two limits

Dashed: hydrodynamic
Solid: large N limit

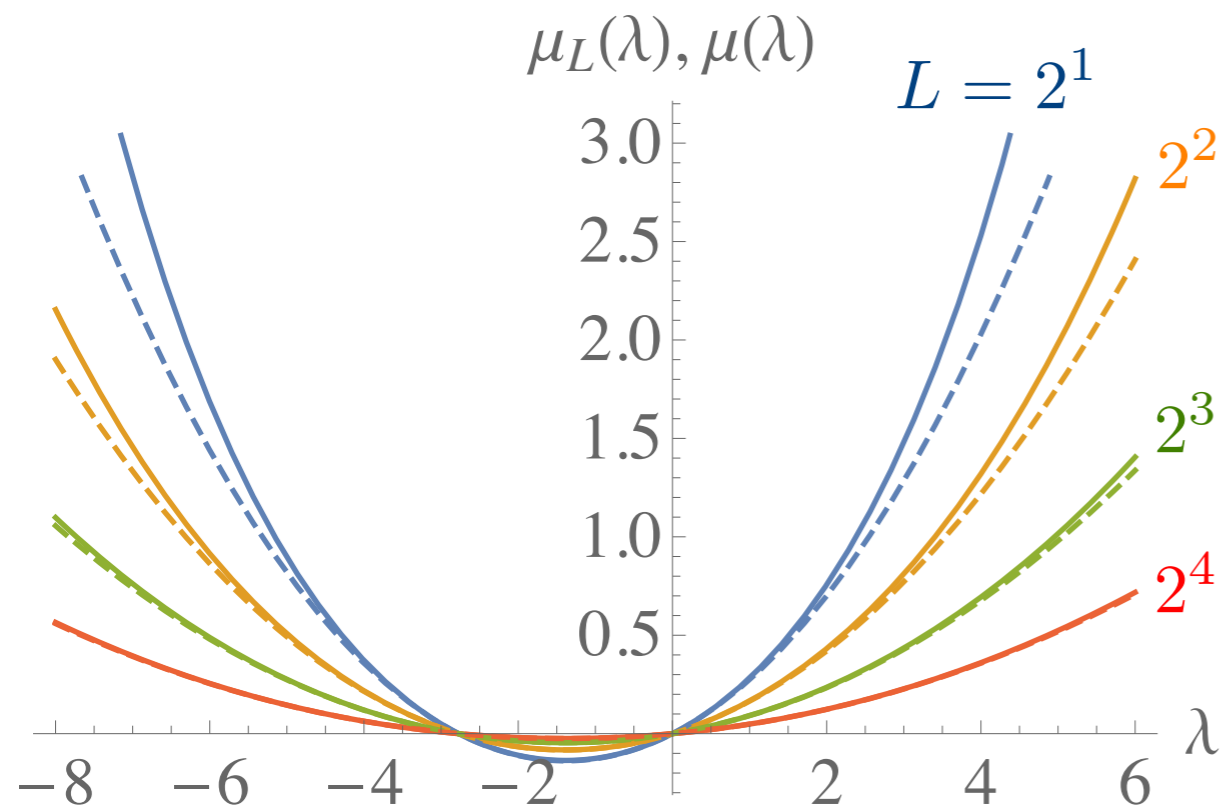


$$\mu_L(\lambda) - \mu(\lambda) = O(L^{-3})$$

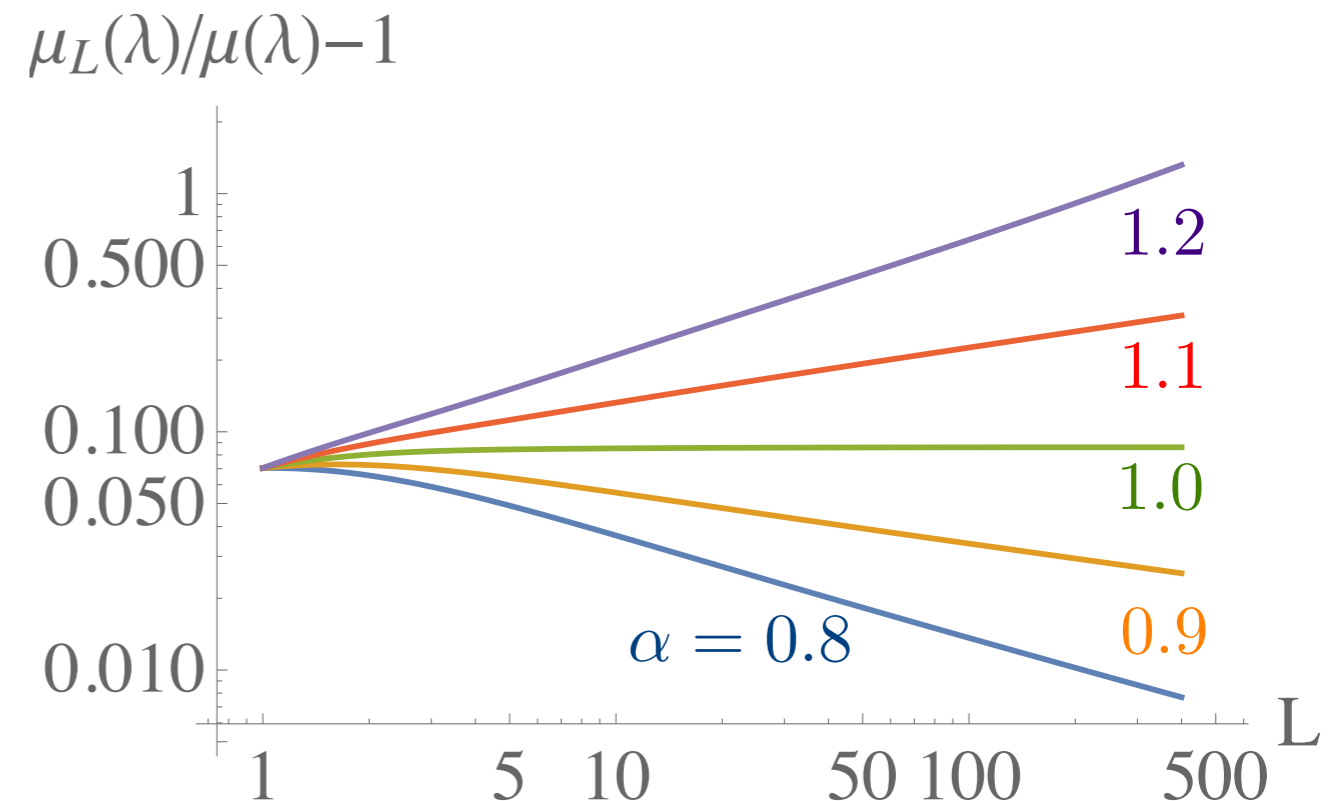
Behaviors of the current CGF

Comparison between the two limits

Dashed: hydrodynamic
Solid: large N limit



$$\mu_L(\lambda) - \mu(\lambda) = O(L^{-3})$$



$$\lim_{L \rightarrow \infty} \mu_L(\bar{\lambda} L^\alpha) \neq \lim_{L \rightarrow \infty} \mu(\bar{\lambda} L^\alpha) \text{ if } \alpha \geq 1$$

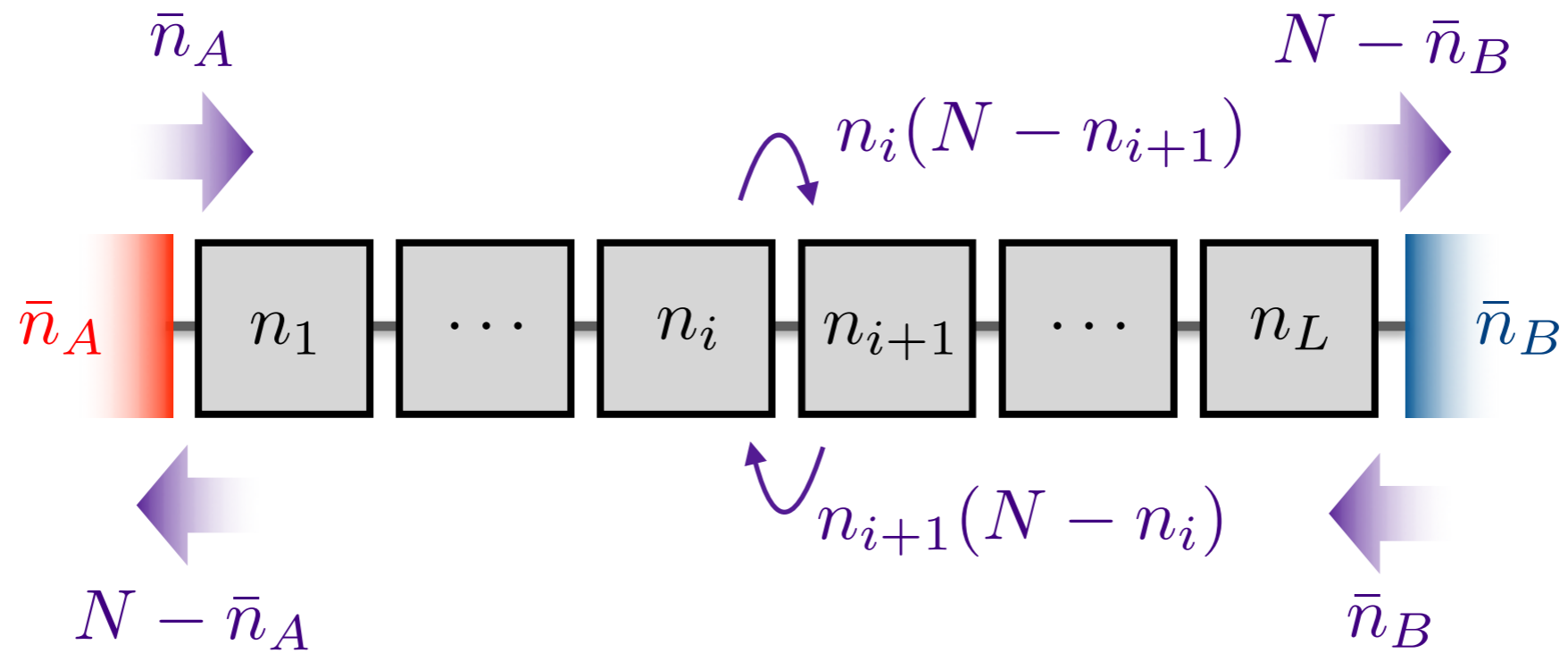
Since $J = \frac{\partial \mu_L}{\partial \lambda} \sim e^{\bar{\lambda} L^{\alpha-1}}$, non-hydrodynamic behaviors are observed for J stronger than $O(1)$.

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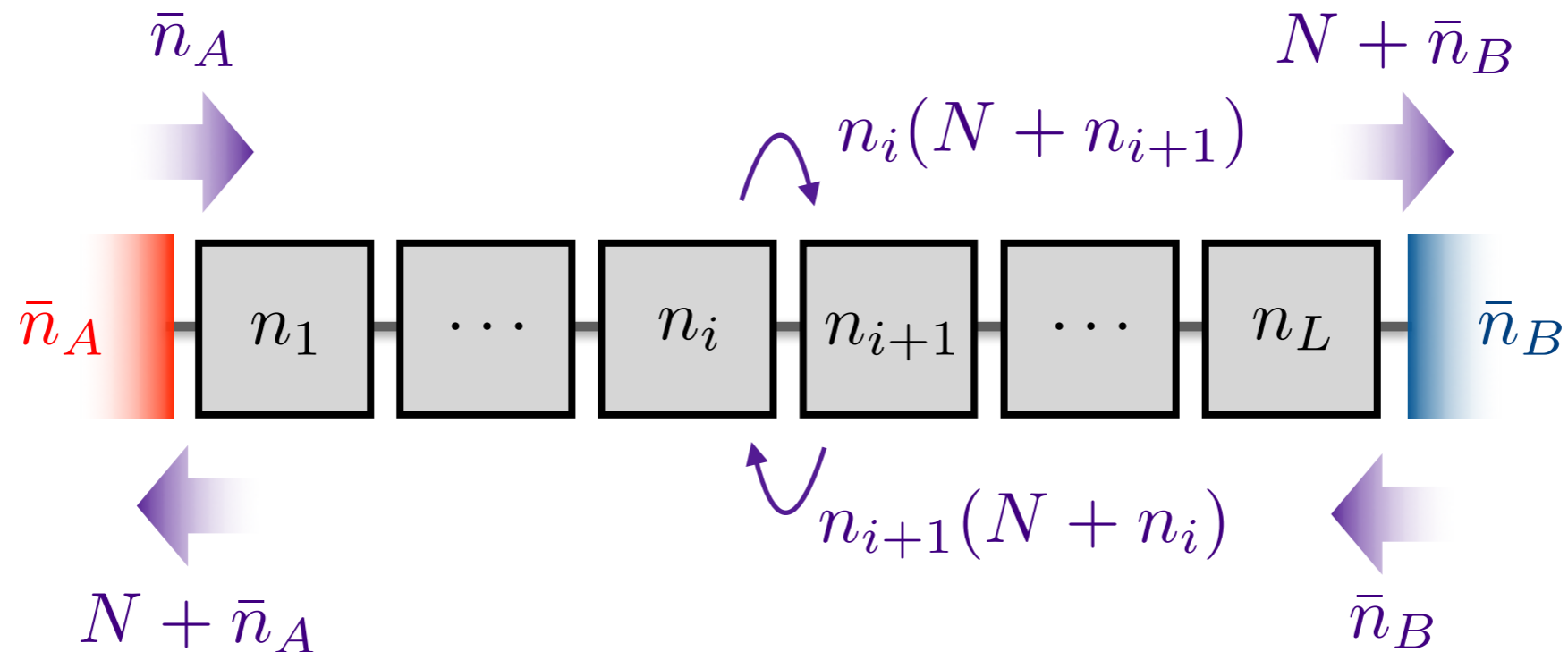
Why $O(1)$, instead of somewhere between $O(1/L)$ and $O(1)$?

Before trying to answer this question, let's check a different class of model first.

SEP-like dynamics



KMP-like dynamics



Features

- ◆ **Attraction** instead of excluded volume repulsion.
- ◆ N is no longer an upper bound on the number of particles.
- ◆ Depending on the reservoir conditions, **the occupancy of each site can be much larger than N .**

Idea: consider the case when $\bar{n}_{A,B} \sim N^2$.

Propagator for the number configuration

$$P[\mathbf{n}_f, t_f | \mathbf{n}_i, t_i] = \int D\mathbf{n} D\hat{\mathbf{n}} \exp \left\{ - \int_{t_i}^{t_f} dt [\hat{\mathbf{n}} \cdot \dot{\mathbf{n}} - H(\mathbf{n}, \hat{\mathbf{n}})] \right\}$$

Rescaling to density

$$\bar{n}_{A,B} \rightarrow N^2 \bar{\rho}_{A,B}$$

$$n_i \rightarrow N^2 \rho_i$$

$$\hat{n}_i \rightarrow N^{-1} \hat{\rho}_i$$

$$t \rightarrow N^{-1} t$$

Propagator for the density configuration

$$P[\boldsymbol{\rho}_f, t_f | \boldsymbol{\rho}_i, t_i] = \int \mathcal{D}\boldsymbol{\rho} \mathcal{D}\hat{\boldsymbol{\rho}} \exp \left\{ - \underline{N} \int_{t_i}^{t_f} dt [\hat{\boldsymbol{\rho}} \cdot \dot{\boldsymbol{\rho}} - H(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}})] \right\}$$

$N \gg 1$ enables the saddle-point approximation

$$\begin{aligned}
 H_\lambda(\mathbf{n}, \hat{\mathbf{n}}) \equiv & \sum_{i=1}^{L-1} \left[n_i(N + n_{i+1}) (e^{\hat{n}_{i+1} - \hat{n}_i} - 1) + n_{i+1}(N + n_i) (e^{\hat{n}_i - \hat{n}_{i+1}} - 1) \right] \\
 & + n_1(N + \bar{n}_A) (e^{-\hat{n}_1} - 1) + \bar{n}_A(N + n_1) (e^{\hat{n}_1} - 1) \\
 & + n_L(N + \bar{n}_B) (e^{-\hat{n}_L + \lambda} - 1) + \bar{n}_B(N + n_L) (e^{\hat{n}_L - \lambda} - 1)
 \end{aligned}$$

Rescaling to density

$$\bar{n}_{A,B} \rightarrow N^2 \bar{\rho}_{A,B}$$

$$n_i \rightarrow N^2 \rho_i$$

$$\hat{n}_i \rightarrow N^{-1} \hat{\rho}_i$$

$$t \rightarrow N^{-1} t$$

Effective Hamiltonian ($N \gg 1$)

$$\begin{aligned}
 H_\lambda(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}}) \equiv & \sum_{i=1}^{L-1} \left[(\hat{\rho}_{i+1} - \hat{\rho}_i)(\rho_i - \rho_{i+1}) + (\hat{\rho}_i - \hat{\rho}_{i+1})^2 \rho_i \rho_{i+1} \right] \\
 & + \hat{\rho}_1(\bar{\rho}_A - \rho_1) + \hat{\rho}_1^2 \rho_1 \bar{\rho}_A + (\hat{\rho}_L - \lambda)(\bar{\rho}_B - \rho_L) + (\hat{\rho}_L - \lambda)^2 \rho_L \bar{\rho}_B
 \end{aligned}$$

$$H_\lambda(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}}) \equiv \sum_{i=1}^{L-1} [(\hat{\rho}_{i+1} - \hat{\rho}_i)(\rho_i - \rho_{i+1}) + (\hat{\rho}_i - \hat{\rho}_{i+1})^2 \rho_i \rho_{i+1}]$$

$$+ \hat{\rho}_1(\bar{\rho}_A - \rho_1) + \hat{\rho}_1^2 \rho_1 \bar{\rho}_A + (\hat{\rho}_L - \lambda)(\bar{\rho}_B - \rho_L) + (\hat{\rho}_L - \lambda)^2 \rho_L \bar{\rho}_B$$

Diffusive scaling

$$i \rightarrow Lx, \quad t \rightarrow L^2 t$$

Continuum limit

$$X_{i+1} - X_i \sim L^{-1}$$

$$\rho_0 = \bar{\rho}_A, \quad \rho_{L+1} = \bar{\rho}_B$$

$$\hat{\rho}_0 = 0, \quad \hat{\rho}_{L+1} = \lambda$$

Hydrodynamic limit ($N \gg 1, L \gg 1$)

$$H(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}}) = \int_0^1 dx [-(\partial_x \rho)(\partial_x \hat{\rho}) + \rho^2 (\partial_x \hat{\rho})^2]$$

$D(\rho) = 1, \sigma(\rho) = \rho^2$: this model is indeed KMP-like.

Calculation of the current CGF

Rescaling

$$\bar{n}_{A,B} \rightarrow N^2 \bar{\rho}_{A,B}$$

$$n_i \rightarrow N^2 \rho_i$$

$$\hat{n}_i \rightarrow N^{-1} \hat{\rho}_i$$

$$t \rightarrow N^{-1} t$$

Saddle-point
approximation

Current CGF

$$\mu_L(\lambda) = H_\lambda(\boldsymbol{\rho}_\lambda^*, \hat{\boldsymbol{\rho}}_\lambda^*)$$

Optimal profile

$$\rho_i(t) = \rho_{\lambda,i}^*, \hat{\rho}_i(t) = \hat{\rho}_{\lambda,i}^*$$

Stationary solution

$$H_\lambda(\boldsymbol{\rho}, \hat{\boldsymbol{\rho}}) \equiv \sum_{i=1}^{L-1} [(\hat{\rho}_{i+1} - \hat{\rho}_i)(\rho_i - \rho_{i+1}) + (\hat{\rho}_i - \hat{\rho}_{i+1})^2 \rho_i \rho_{i+1}]$$

$$+ \hat{\rho}_1(\bar{\rho}_A - \rho_1) + \hat{\rho}_1^2 \rho_1 \bar{\rho}_A + (\hat{\rho}_L - \lambda)(\bar{\rho}_B - \rho_L) + (\hat{\rho}_L - \lambda)^2 \rho_L \bar{\rho}_B$$

Behaviors of the current CGF

Result for the large N limit

$$\mu_L(\lambda) = \begin{cases} -(L+1) \sinh^2 \left(\frac{1}{L+1} \operatorname{arcsinh} \sqrt{\omega} \right) & \text{for } \omega > 0 \\ (L+1) \sin^2 \left(\frac{1}{L+1} \operatorname{arcsin} \sqrt{-\omega} \right) & \text{for } \omega < 0 \end{cases}$$

Hydrodynamic result

$$\mu(\lambda) = \begin{cases} -\frac{1}{L+1} \operatorname{arcsinh}^2 \sqrt{\omega} & \text{for } \omega > 0 \\ \frac{1}{L+1} \operatorname{arcsinh}^2 \sqrt{-\omega} & \text{for } \omega < 0 \end{cases}$$

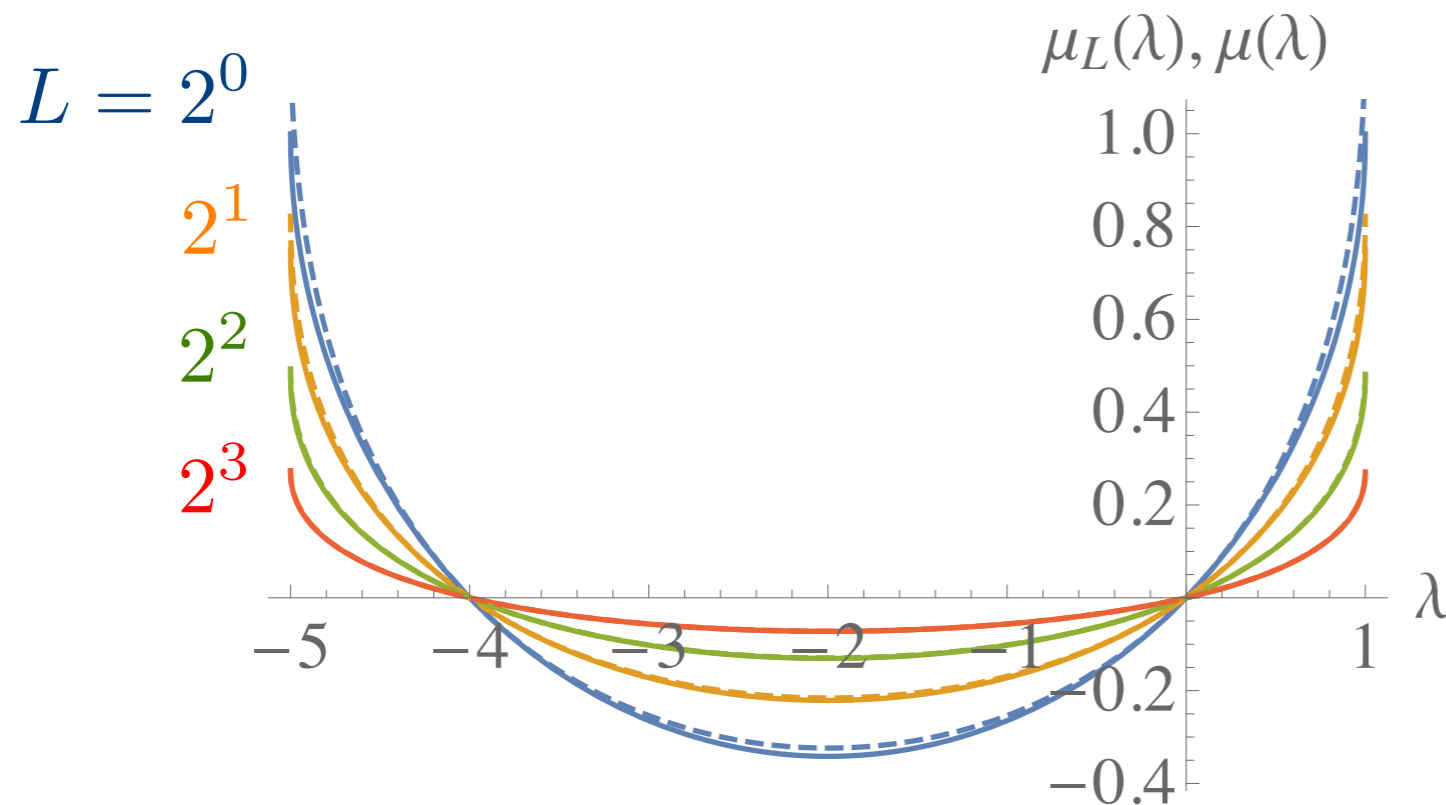
Auxiliary variable for both cases

$$\omega(\lambda, \bar{\rho}_A, \bar{\rho}_B) = \lambda(\bar{\rho}_B - \bar{\rho}_A) - \lambda^2 \bar{\rho}_A \bar{\rho}_B$$

Behaviors of the current CGF

Comparison between the two limits

Dashed: hydrodynamic
Solid: large N limit



The domain is bounded by

$$-\frac{1}{\bar{\rho}_B} < \lambda < \frac{1}{\bar{\rho}_A}$$



Each pair converge like L^{-3} over the entire domain.



Same exponential tails are observed for both cases.
Hydrodynamic and non-hydrodynamic regimes are undistinguishable.

SEP-like model

Since $J = \frac{\partial \mu_L}{\partial \lambda} \sim e^{\bar{\lambda} L^{\alpha-1}}$,
non-hydrodynamic behaviors
are observed for
 J stronger than $O(1)$.

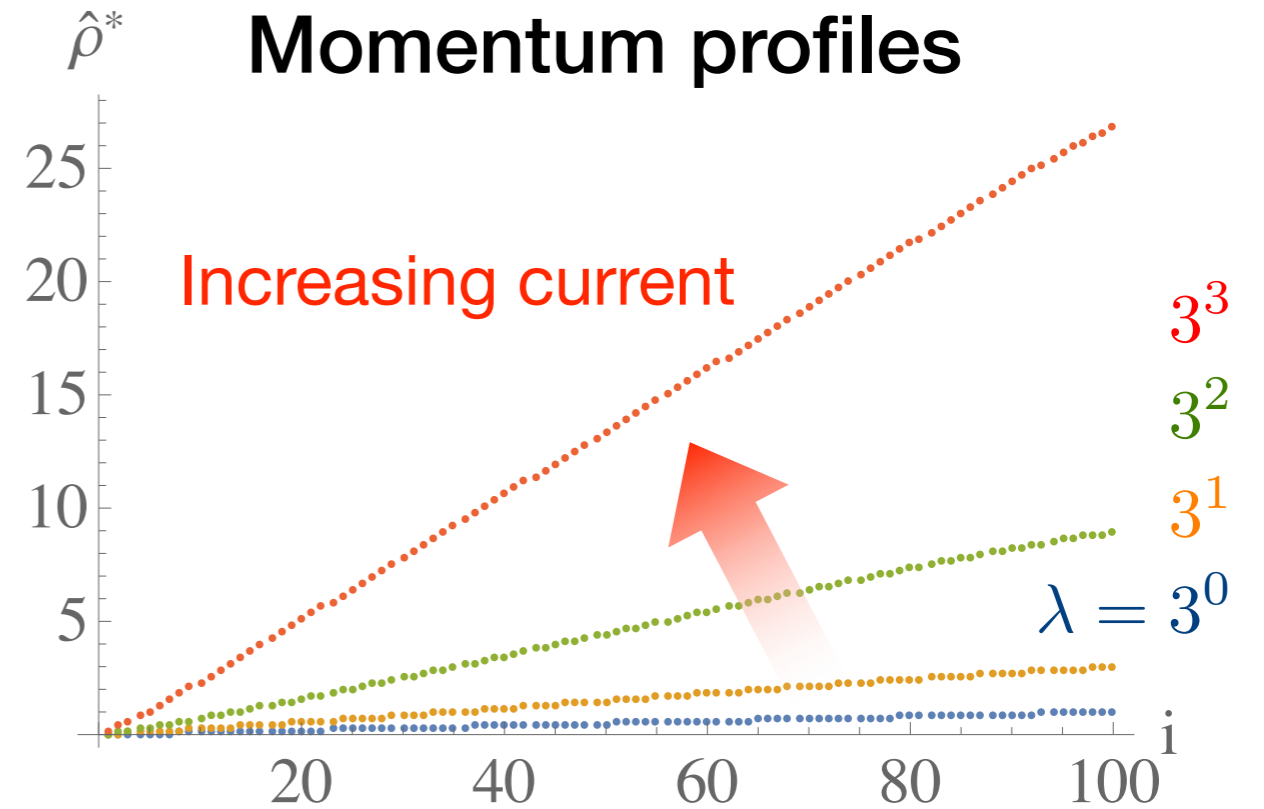
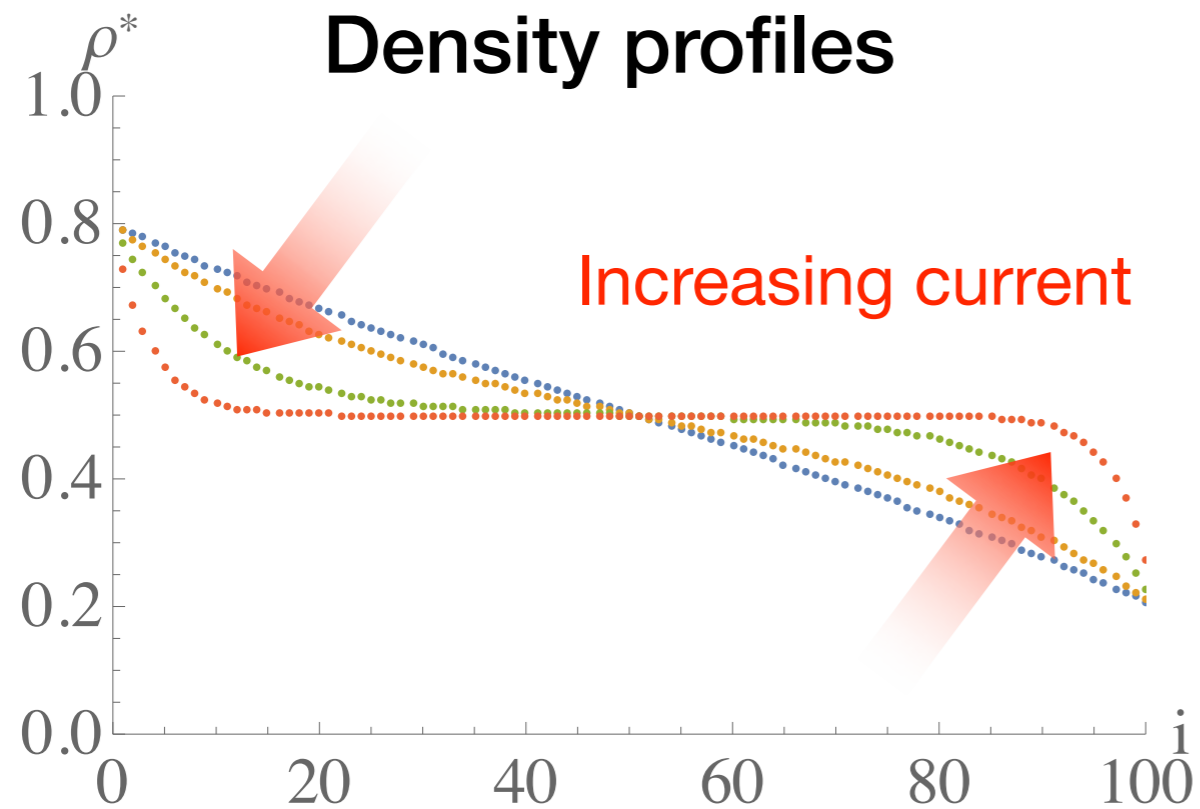
KMP-like model

Hydrodynamic and
non-hydrodynamic regimes
are undistinguishable.

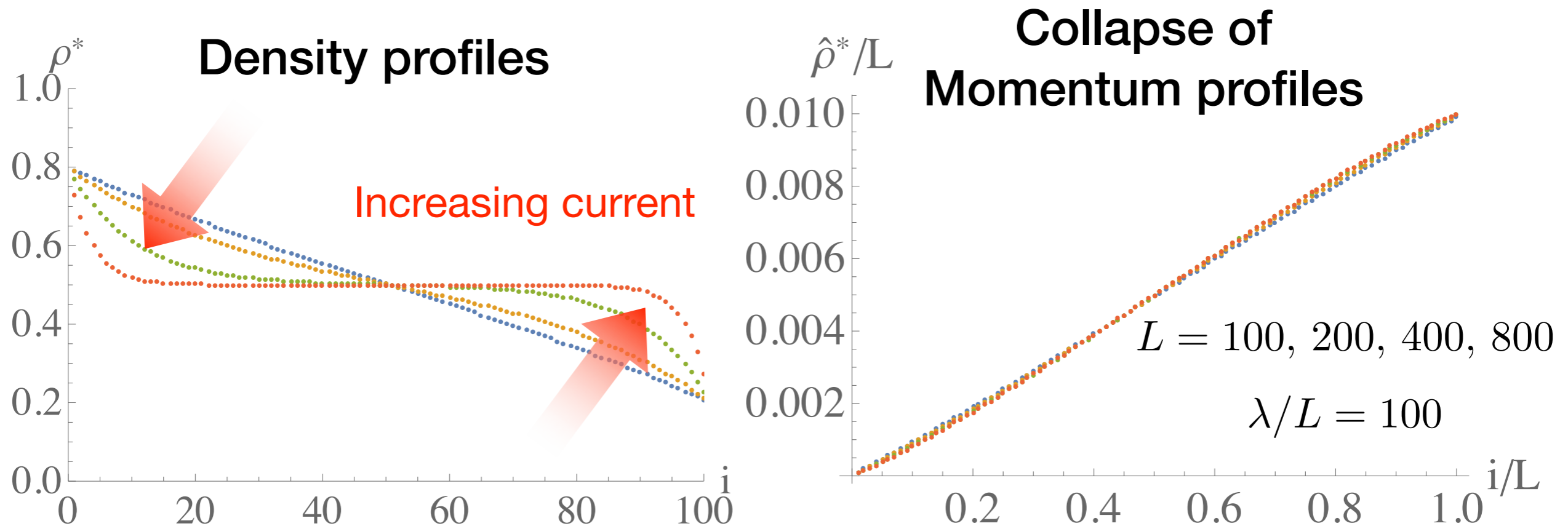
Why so different?

Let's check optimal profiles supporting large current fluctuations.

Optimal profiles of the SEP-like model



Optimal profiles of the SEP-like model



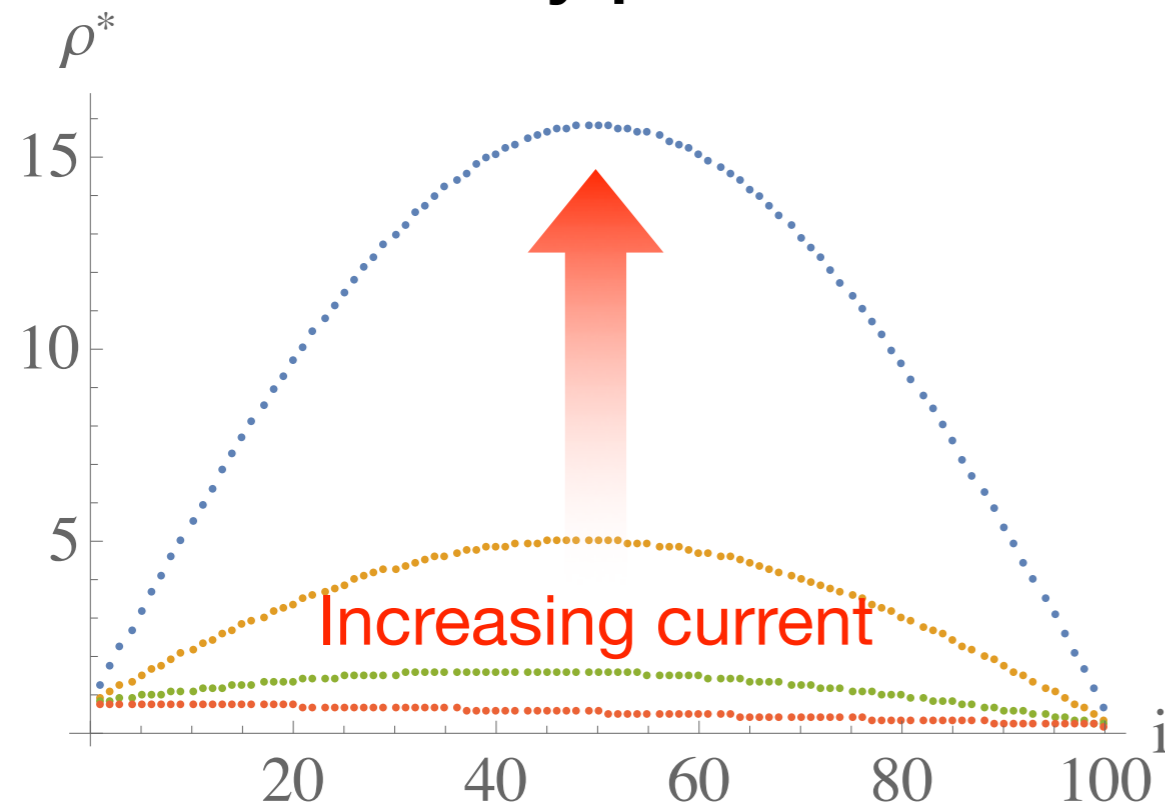
$$j = -\partial_x \rho + 2\rho(1 - \rho) \partial_x \hat{\rho}$$

Large current supported by large momentum gradient

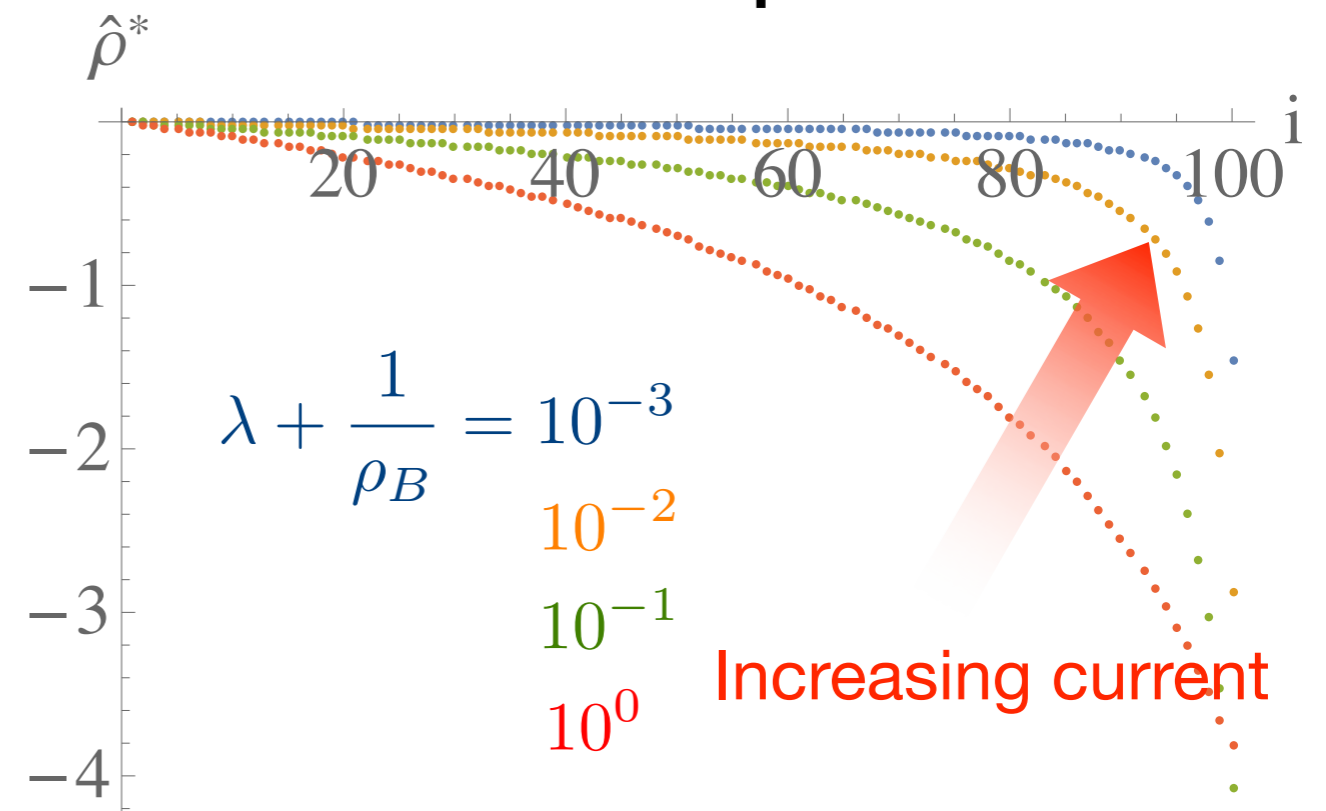
Since $\partial_x \hat{\rho} \sim L^{\alpha-1}$ for $\lambda \sim L^\alpha$,
 $\alpha = 1$ is the borderline for small gradient.

Optimal profiles of the KMP-like model

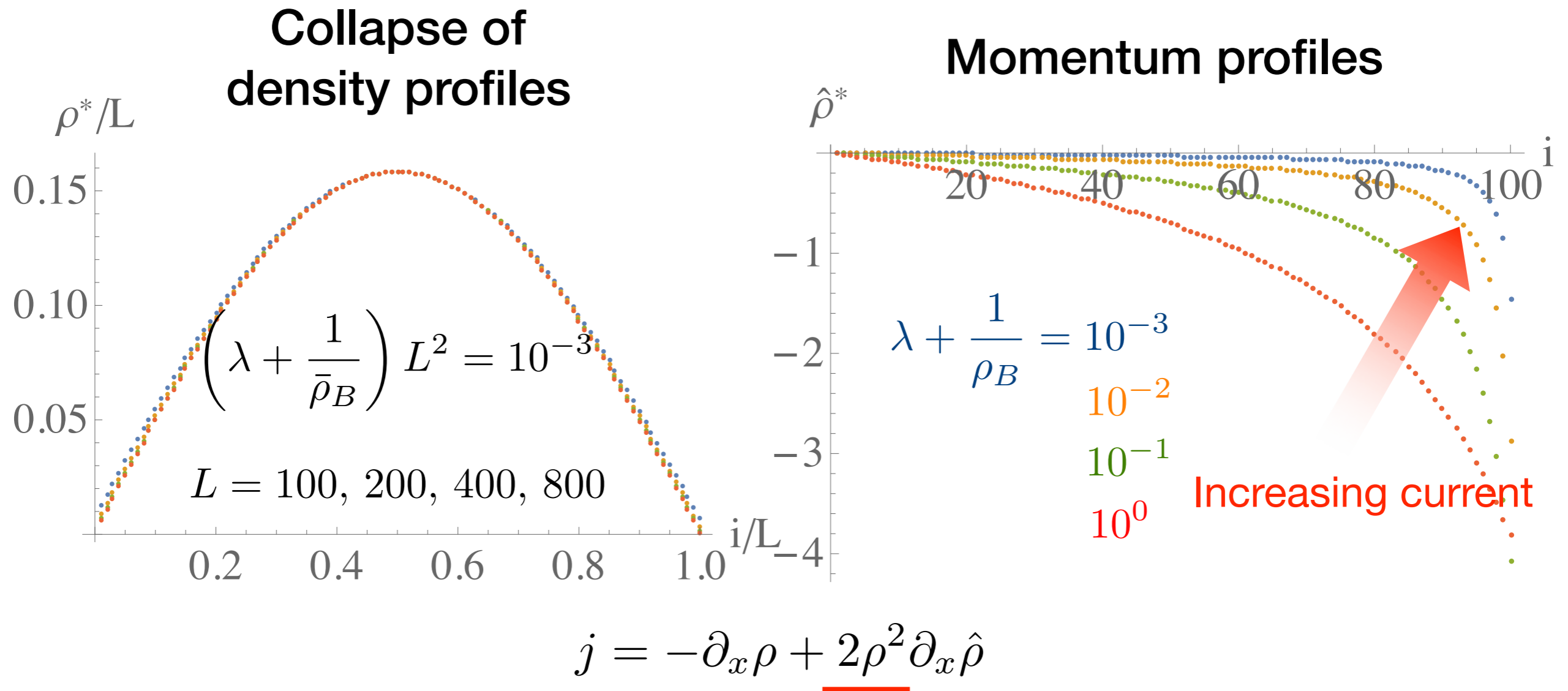
Density profiles



Momentum profiles



Optimal profiles of the KMP-like model



Large current supported by large density values

The large current is dominated by the non-gradient second term and is blind to the diverging density gradient.

“Deterministic component”

“Stochastic component”

SEP-like model

$$j = -\partial_x \rho + \underline{2\rho(1-\rho)} \partial_x \hat{\rho}$$

Large current supported by large momentum gradient

KMP-like model

$$j = -\partial_x \rho + \underline{2\rho^2} \partial_x \hat{\rho}$$

Large current supported by large density values

Stochastic component plays a dominant role in both cases.

Hydrodynamic description breaks down only when the stochastic component requires a large gradient to support a large current fluctuation.

Summary and future works

Summary

- ◆ We formulated **the large N limit** to investigate the possibility of non-hydrodynamic tail behaviors of current fluctuations.
- ◆ Under the assumption of stationary saddle-point solutions, we obtained **expressions for current CGFs of SEP-like and KMP-like models**.
- ◆ The hydrodynamic description **breaks down for the SEP-like model**, but it **remains valid for the KMP-like model**.

Future works

- ◆ The **validity of the stationary saddle-point solution** remains to be established more rigorously.
- ◆ The large N limit might be useful for making **minimal models** of nonequilibrium systems.