

KPZ Equation Limit of Interacting Particle Systems

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joint work with Ivan Corwin and with Amir Dembo

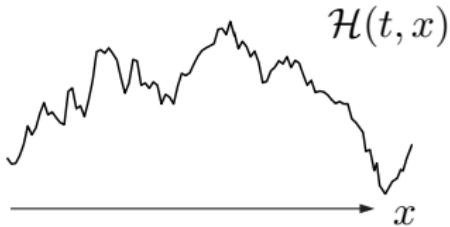
The KPZ equation

The KPZ (Kardar-Parisi-Zhang) equation

$$\partial_t \mathcal{H} = \frac{1}{2} \partial_{xx} \mathcal{H} - \frac{1}{2} (\partial_x \mathcal{H})^2 + \xi$$

is a paradigmatic equation describing random surface growth with smoothing effect and slope dependent growth.

- $\partial_{xx} \mathcal{H}$: smoothing;
- $(\partial_x \mathcal{H})^2$: slope depend. growth;
- ξ : spacetime white noise,
 $E(\xi(t,x)\xi(t',x'))=\delta(t-t')\delta(x-x')$



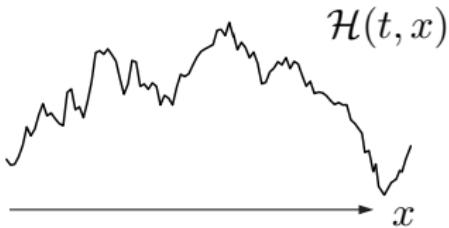
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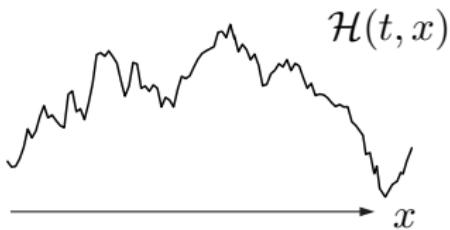
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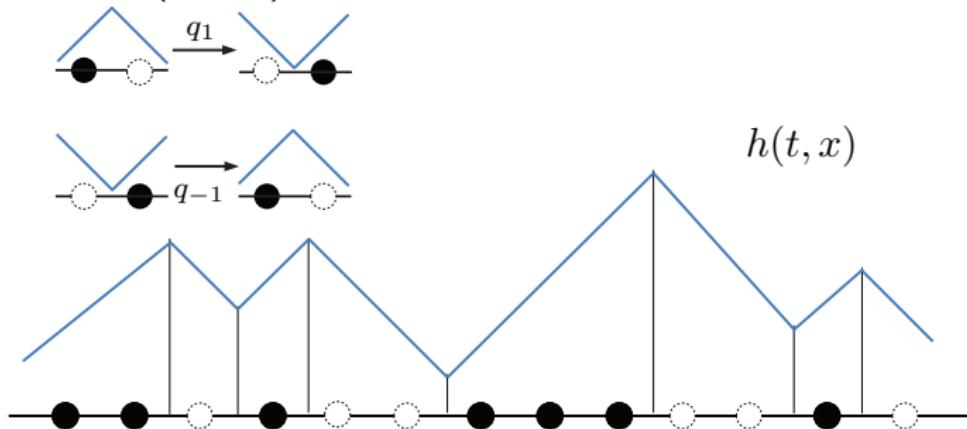
- Exhibit non-Gaussian distribution.
- Is an integrable model, with asymptotic ($t \rightarrow \infty$) distribution related to RMT.

A Discrete Example

- The corner growth model / Asymmetric Simple Exclusion Process (ASEP).

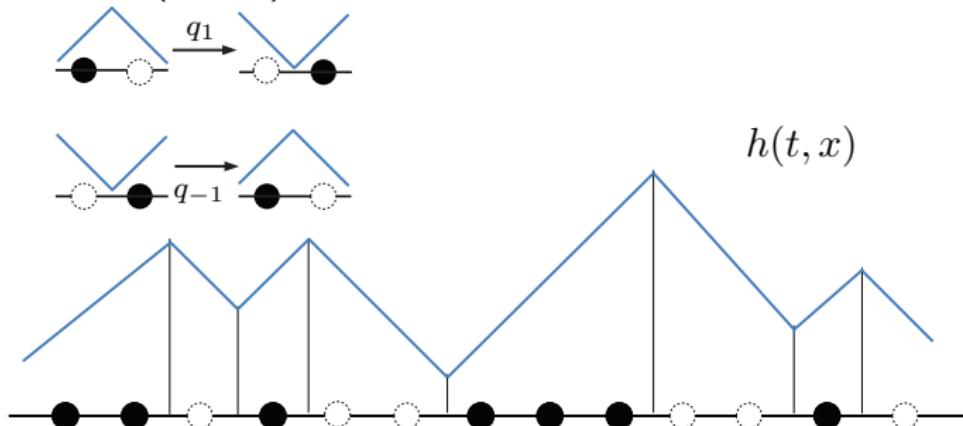
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- Under $\mathcal{H}_\varepsilon(t, x) := \varepsilon^b \mathcal{H}(\varepsilon^{-z}t, \varepsilon^{-a}x)$, the KPZ equation scales as

$$\partial_t \mathcal{H}_\varepsilon = \varepsilon^{2a-z} \frac{1}{2} \partial_{xx} \mathcal{H} - \varepsilon^{2a-z-b} \frac{1}{2} (\partial_x \mathcal{H})^2 + \varepsilon^{b-\frac{z}{2}+\frac{a}{2}} \xi.$$

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The weak asymmetry causes $\frac{1}{2} \mapsto \frac{\varepsilon}{2}$, so $(z, a, b) = (4, 2, 1)$ preserves the KPZ equation.

The Weak Universality

- For ASEP, under the weak asymmetry,

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- Such a convergence is conjectured to hold *universally*
 - for models in the KPZ universality class: e.g. directed polymer, last passage percolation, q -TASEP.
 - regardless of details of microscopic interaction, under scalings that preserve the KPZ equation.

The Hopf-Cole (HC) transformation

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- $\mathcal{Z}(t, x) := e^{-\mathcal{H}(t, x)}$ formally converts the KPZ equation

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into the Stochastic Heat Equation (SHE)

$$\partial_t \mathcal{Z} = \frac{1}{2} \partial_{xx} \mathcal{Z} + \mathcal{Z} \xi.$$

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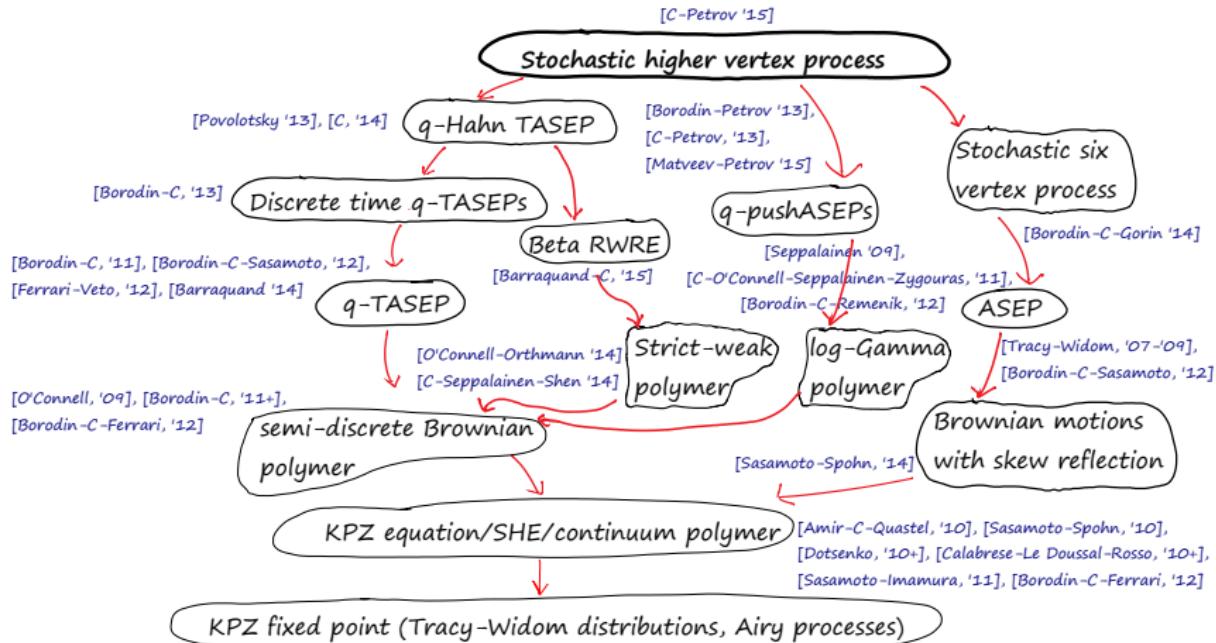
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- Q: when is a discrete HC transformation available?
- A: for *all* integrable models.

Higher Spin Exclusion Processes (Courtesy of Corwin)

Degenerations to known integrable stochastic systems in KPZ class



Higher Spin Exclusion Processes (HSEP)

- A discrete time exclusion process

$$\{\dots < y_n(t) < y_{n-1}(t) < \dots \in \mathbb{Z}\}_{t \in \mathbb{N}},$$

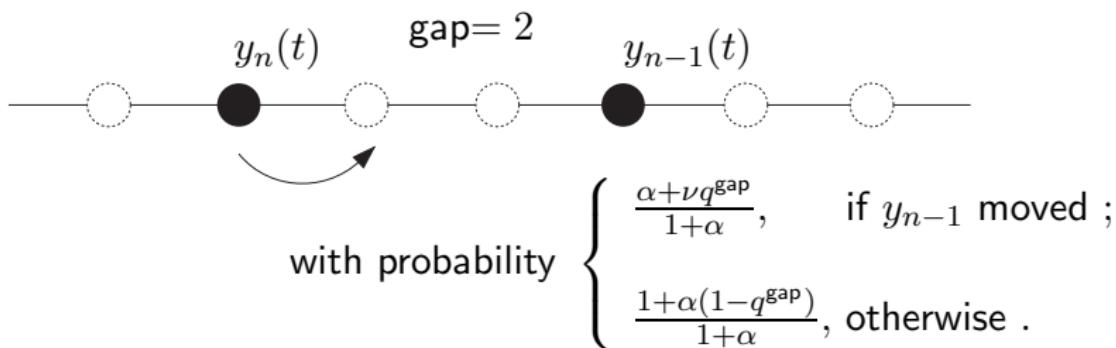
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$\{\dots < y_n(t) < y_{n-1}(t) < \dots \in \mathbb{Z}\}_{t \in \mathbb{N}}$, indexed by four parameters $(q, \alpha, \nu, J) \in (0, 1) \times [0, 1] \times [0, 1] \times \mathbb{N}_+$.

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- Consider only $J = 1$ for simplicity.



The HC transformation for HSEP

Fix $\rho \in (0, 1)$, which measures the average density.

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- Define $Z(t, x) := \lambda^t \rho^{x+\mu t} Q_{x+\mu t}(t)$, where $Q_n(t) := q^{y_n(t)+n}$,

$$\lambda := \frac{1+\alpha f}{1+\alpha qf} > 0, \quad \mu := \frac{a-a'}{b-b'} > 0;$$

$$a := \frac{\alpha f}{1+\alpha f}, \quad a' := \frac{\alpha qf}{1+\alpha qf}, \quad b := \frac{f}{1-f}, \quad b' := \frac{\nu f}{1-\nu f}, \quad f := \frac{1-\rho}{1-\nu\rho}.$$

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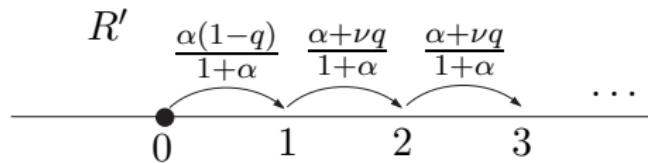
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- Let $p(y, x) := \mathbf{P}(R = x - y)$ and $L(y, x) := p(y, x) - \mathbf{1}_{y=x+\mu}$ be the semigroup and pseudo generator of the the following random walk.

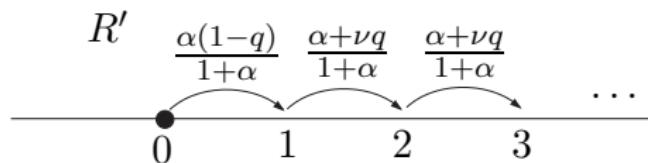


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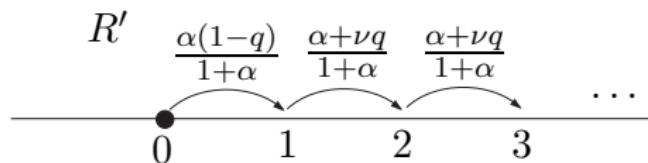
Lemma

$$\sum_{n=0}^{\infty} \lambda \rho^n \mathbf{P}(R' = n) = 1.$$

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Lemma

$$\sum_{n=0}^{\infty} \lambda \rho^n \mathbf{P}(R' = n) = 1. \quad \mathbf{E}(R) = 0.$$

The HC transformation for HSEP

Proposition (Corwin and Tsai, 2015)

$$\begin{aligned} & Z(t+1, x) - Z(t, x + \mu) \\ &= [LZ(t, \cdot)](x) + Z(t, x + \mu)W(t, x + \mu), \end{aligned}$$

where $W(t, y)$ is an (explicit) \mathcal{F} -martingale,

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$$\begin{aligned} Z(t, y)Z(t, y') \mathbf{E}(W(t, y)W(t, y)|\mathcal{F}(t)) \\ = (\frac{\nu+\alpha}{1+\alpha})^{|y-y'|} \Theta_1(t, y_1 \wedge y_2)\Theta_2(t, y_1 \wedge y_2), \end{aligned}$$

$\Theta_1(t, y) := q\lambda Z(t, y) - [pZ(t, \cdot)](t)$ and

$\Theta_2(t, y) := -\lambda Z(t, y) + [pZ(t, \cdot)](t).$

KPZ equation limit for HSEP, scaling

Scaling

- fix $(\rho, \alpha, \nu) \in (0, 1) \times [0, 1]^2$, and let $q = q_\varepsilon := e^{-\varepsilon}$.

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$$t_\varepsilon(t) := t_\varepsilon^* \varepsilon^{-3} t, \quad x_\varepsilon(t, x) := r_* \varepsilon^{-1} x + \frac{\mu_\varepsilon}{\varepsilon} \varepsilon^{-2} t.$$

Here $r_* := (b - b')^{-1} > 0$, $\frac{\mu_\varepsilon}{\varepsilon} = \mu + O(\varepsilon)$, and
 $t_\varepsilon^* := \varepsilon^{-1} (a^2 - {a'_\varepsilon}^2 - (a - a'_\varepsilon)(b + b'))^{-1} = t^* + O(\varepsilon)$.

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Assume the initial condition $Z_\varepsilon(0, \cdot)$ satisfies the following moment conditions: for all $n \in \mathbb{N}$ and $v \in (0, \frac{1}{2})$, there exists $C = C(v, n) < \infty$ such that

$$\mathbf{E}(Z_\varepsilon(0, x)^n) \leq C_n e^{C_n|x|},$$

$$\mathbf{E} |Z_\varepsilon(0, x) - Z_\varepsilon(0, x')|^n \leq C_n |x - x'|^{nv} e^{C_n(|x|+|x'|)}.$$

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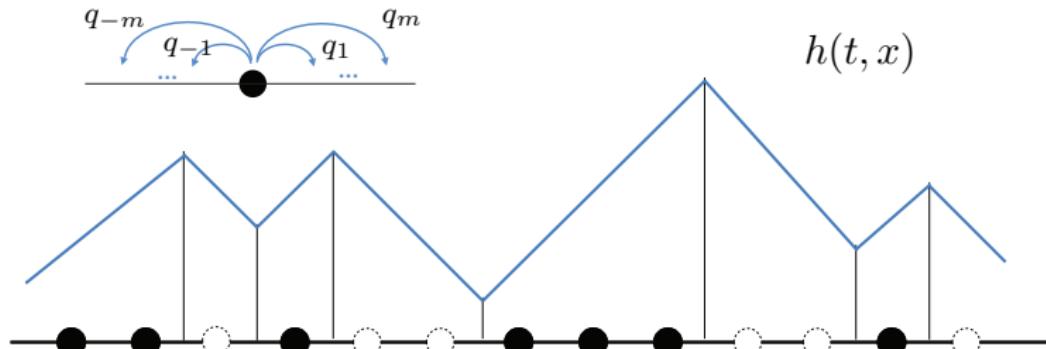
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Theorem (Corwin and Tsai, 2015)

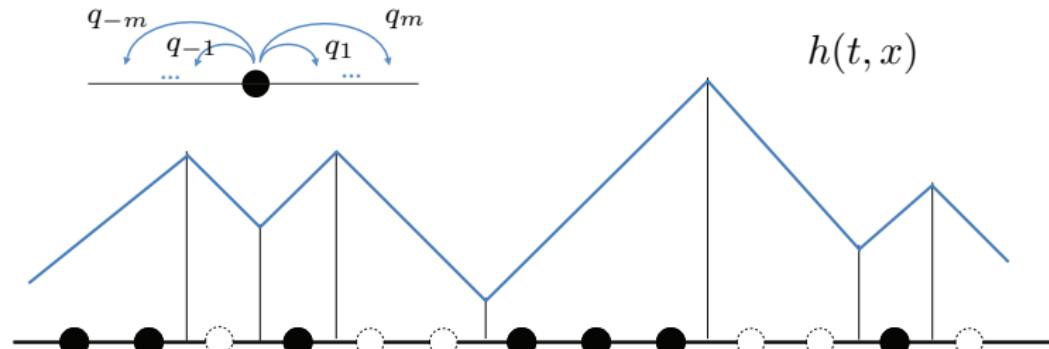
Let $\mathcal{Z}(t, x)$ be the $C(\mathbb{R}_+, \mathbb{R})$ -valued solution of the SHE starting from $\mathcal{Z}_{ic}(x) \in C(\mathbb{R})$. If $Z_\varepsilon(0, \cdot) \Rightarrow \mathcal{Z}_{ic}(\cdot)$, then

$$Z_\varepsilon(\cdot, \cdot) \Rightarrow \mathcal{Z}(\cdot, \cdot).$$

Non-nearest Neighbor Exclusion Processes



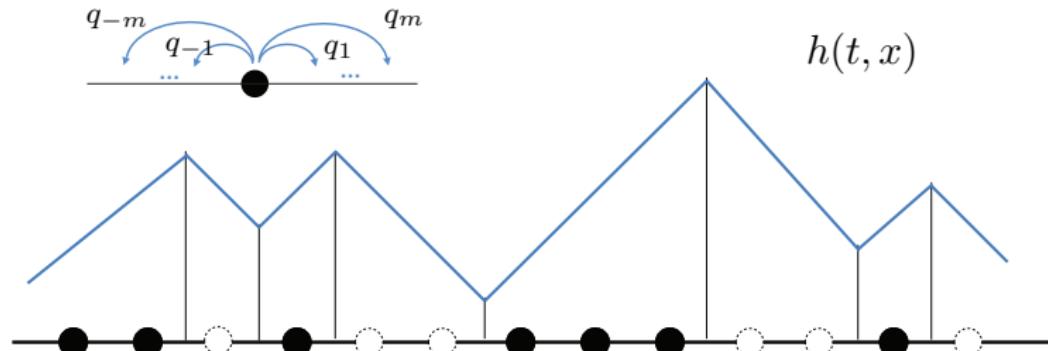
Non-nearest Neighbor Exclusion Processes



- Height function $h(t, x)$ defined for $x \in \mathbb{Z}$;
- Occupation variable $\eta(y)$ defined for $y \in \mathbb{Z} + \frac{1}{2}$, as

$$\eta(y) := \begin{cases} 1 & , \text{if occupied,} \\ -1 & , \text{if empty.} \end{cases}$$

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- Weak asymmetry:

$$q_i := \frac{1}{2}(\kappa_i - \varepsilon^1 \gamma_i), \quad q_{-i} := \frac{1}{2}(\kappa_i + \varepsilon^1 \gamma_i), \quad i = 1, \dots, m.$$

Non-nearest Neighbor Exclusion Processes

Assume $m = 3$.

Theorem (Dembo and Tsai, 2013)

Given any $\kappa_1, \kappa_2, \kappa_3 \in (0, 1)$ with $\kappa_1 + \kappa_2 + \kappa_3 = 1$ and $\gamma > 0$, for the following choice of asymmetry

$$\gamma_j = \gamma \left(\sum_{i=j}^3 \frac{2(i-j)}{j} \kappa_i + \kappa_j \right) + O(\varepsilon), \quad j = 1, 2, 3. \quad (\star)$$

Let $Z(t, x) := \exp(-\gamma\varepsilon^1 h(t, x) + c_\varepsilon t)$ and $Z_\varepsilon(t, x) := Z(\varepsilon^{-4}t, \varepsilon^{-2}x)$. If $Z_\varepsilon(0, \cdot)$ satisfies the preceding moment conditions and if $Z_\varepsilon(0, \cdot) \Rightarrow \mathcal{Z}_{ic}(\cdot)$, then

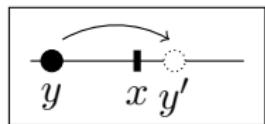
$$Z_\varepsilon(\cdot, \cdot) \Rightarrow \mathcal{Z}(\cdot, \cdot).$$

Microscopic eq'n for $Z(t, x) := \exp(-\gamma\varepsilon^1 h(t, x) + C_\varepsilon t)$

$$dZ(t, x) =$$

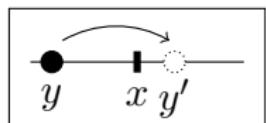
Microscopic eq'n for $Z(t, x) := \exp(-\gamma\varepsilon^1 h(t, x) + C_\varepsilon t)$

$$dZ(t, x) = \sum_{(y, y') \ni x} \left((e^{-2\gamma\varepsilon} - 1) Z(t, x) \right.$$



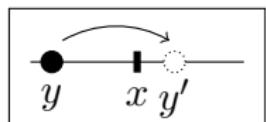
Microscopic eq'n for $Z(t, x) := \exp(-\gamma\varepsilon^1 h(t, x) + C_\varepsilon t)$

$$dZ(t, x) = \sum_{(y, y') \ni x} \left((e^{-2\gamma\varepsilon} - 1) Z(t, x) \frac{1+\eta(y)}{2} \frac{1-\eta(y')}{2} \right)$$



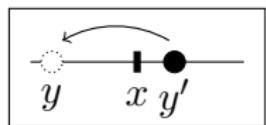
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Microscopic eq'n for $Z(t, x) := \exp(-\gamma\varepsilon^1 h(t, x) + C_\varepsilon t)$

$$dZ(t, x) = (\varepsilon \mathcal{L} + \varepsilon^2 \mathcal{Q}) dt + d\mathbf{M}\mathbf{G},$$

where

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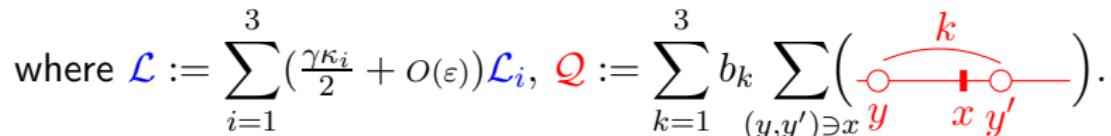
where $\mathcal{L} := \sum_{i=1}^3 (\frac{\gamma\kappa_i}{2} + O(\varepsilon)) \mathcal{L}_i$, $\mathcal{Q} := \sum_{k=1}^3 b_k \sum_{(y,y') \ni x} (\eta(y)\eta(y') Z(t, x))$.

$$\mathcal{L}_i := (-\eta(x+\tfrac{1}{2}-i) - \dots - \eta(x-\tfrac{1}{2}) + \eta(x+\tfrac{1}{2}) + \dots + \eta(x-\tfrac{1}{2}+i)) Z(t, x).$$

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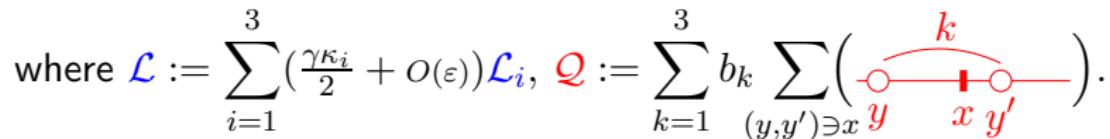


$$\mathcal{L}_i := (-\eta(x + \frac{1}{2} - i) - \dots - \eta(x - \frac{1}{2}) + \eta(x + \frac{1}{2}) + \dots + \eta(x - \frac{1}{2} + i)) Z(t, x).$$

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- Wish to match $(\varepsilon \mathcal{L} + \varepsilon^2 \mathcal{Q})$ to

$$\text{Laplacian} = (Z(t, x+i) + Z(t, x-i) - 2Z(t, x))$$

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$$\text{Laplacian} = \frac{1}{2} \sum_{i=1}^3 \tilde{\kappa}_i (Z(t, x+i) + Z(t, x-i) - 2Z(t, x))$$

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- Wish to match $(\varepsilon \mathcal{L} + \varepsilon^2 \mathcal{Q})$ to

$$\text{Laplacian} = \sum_{i=1}^3 \varepsilon \frac{\gamma \tilde{\kappa}_i}{2} \mathcal{L}_i + \sum_{k=1}^2 \varepsilon^2 \tilde{b}_k \left(\sum_{y,y' \in (x-3,x)} \begin{array}{c} k \\ \text{---} \\ \textcircled{y} \quad y' \quad x \end{array} \right)$$

$$+ \sum_{y,y' \in (x,x+3)} \begin{array}{c} k \\ \text{---} \\ x \quad \textcircled{y} \quad y' \end{array} \Big) + \varepsilon^3 \text{cubic} + \dots$$

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- Match linear terms by choosing $\tilde{\kappa}_i = \kappa_i + O(\varepsilon)$.

Gradient Condition

- How to match quadratic terms?

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Definition

We call $f(x)$ a **gradient term** if

$$f(x) = \sum_{|j| \leq 3} (g_j(x + j) - g_j(x)),$$

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- How to match quadratic terms?

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We call $f(x)$ a **gradient term** if

$$f(x) = \sum_{|j| \leq 3} (g_j(x + j) - g_j(x)),$$

for some $g_j(x)$ of the form $g_j(x) = \eta(x + \ell_1) \cdots \eta(x + \eta_n) Z(t, x)$.

Gradient Condition

- How to match quadratic terms?

Proposition

$$\begin{aligned} \text{Diagram 1: } & \text{A horizontal line with two open circles at } y \text{ and } y'. \text{ A red vertical bar is at } x. \text{ A red curved arrow labeled } k \text{ connects the two circles.} \\ & = \text{Diagram 2: } \text{The same setup as Diagram 1, but the red bar } x \text{ is now between the two circles. The red curved arrow } k \text{ connects the circles.} + \text{grad. term} \\ & = \text{Diagram 3: } \text{A horizontal line with two open circles at } y \text{ and } y'. \text{ A red vertical bar is at } x. \text{ The red curved arrow } k \text{ connects the circles.} + \text{grad. term.} \end{aligned}$$

Gradient Condition

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Proposition

$$\begin{aligned} \text{Diagram 1: } & \text{A horizontal red line with two white circles at } y \text{ and } y'. \text{ A red arc labeled } k \text{ connects the top of } y \text{ to the top of } y'. \text{ A red vertical bar is at } x. \\ & = \text{Diagram 2: } \text{The same horizontal line with circles } y \text{ and } y' \text{ and bar } x. \text{ The arc } k \text{ now connects } y \text{ to } y' \text{ through the bar } x. \text{ Label: } + \text{grad. term.} \\ & = \text{Diagram 3: } \text{A horizontal red line with white circles at } y \text{ and } y'. \text{ A red vertical bar is at } x. \text{ The arc } k \text{ connects the top of } y \text{ to the top of } y' \text{ through the bar } x. \text{ Label: } + \text{grad. term.} \end{aligned}$$

- Matching the quadratic terms amounts to matching the coefficients: $kb_k = 2(3 - k)\tilde{b}_k$, which gives the condition (\star) .

Gradient Condition

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Proposition

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- Cubic terms: use $m \leq 3$.

Gradient Condition

- How to match quadratic terms?

Proposition

$$\begin{aligned} \text{Diagram 1: } & \text{Two circles } y \text{ and } y' \text{ connected by a horizontal line segment. A red arc labeled } k \text{ connects } y \text{ and } y'. \text{ A red vertical bar is at } x. \\ & = \text{Diagram 2: } \text{The same setup as Diagram 1, but the red bar } x \text{ is moved to the right between } y' \text{ and } x. \text{ Red arc } k \text{ connects } y \text{ and } y'. \\ & \quad + \text{grad. term} \\ \text{Diagram 3: } & \text{A red vertical bar } x \text{ is at } y. \text{ A red arc labeled } k \text{ connects } y \text{ and } y'. \\ & = \text{Diagram 4: } \text{The same setup as Diagram 3, but the red bar } x \text{ is moved to the right between } y \text{ and } y'. \text{ Red arc } k \text{ connects } y \text{ and } y'. \\ & \quad + \text{grad. term.} \end{aligned}$$

- Matching the quadratic terms amounts to matching the coefficients: $kb_k = 2(3 - k)\tilde{b}_k$, which gives the condition (\star) .
- Cubic terms: use $m \leq 3$.
- higher order terms: hydrodynamic limit argument.