

Large deviations of the KPZ equation via the delta Bose gas

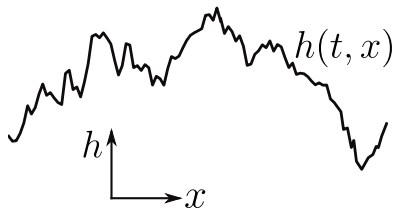
KPZ meets KPZ

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Kardar–Parisi–Zhang (KPZ)

$$\partial_t h = \frac{1}{2} \partial_{xx} h + \frac{1}{2} (\partial_x h)^2 + \eta$$



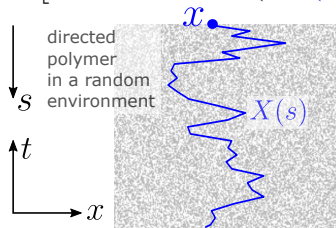
Stochastic Heat Equation (SHE)

$$\partial_t Z = \frac{1}{2} \partial_{xx} Z + \eta Z$$

Feynman–Kac:

$$Z(T, x) =$$

$$\mathbb{E}_{\text{BM}} \left[e^{\int_0^T ds \eta(T-s, X(s))} Z(0, X(T)) \right]$$



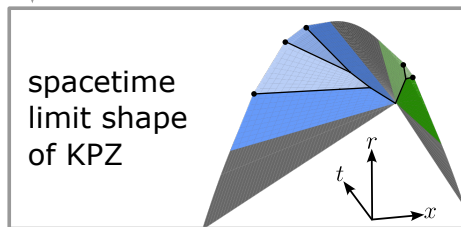
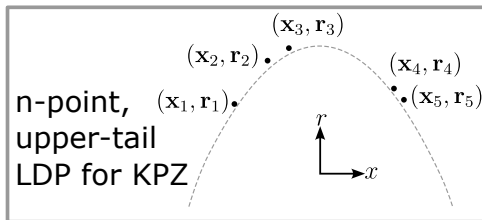
$$e^{h(t,x)} = Z(t, x)$$

Plan for this talk

$$Z(0, \cdot) = \delta_0 \text{ for most of the talk}$$

attractive
Brownian
Particles (PBs)

n-point
moments
of SHE



SHE moments \Leftrightarrow LDP of KPZ equation

$$\mathbf{E}\left[\left(Z(T,0)e^{\frac{T}{24}}\right)^m\right] = \mathbf{E}\left[e^{m(h(T,0)+\frac{T}{24})}\right] \approx e^{\frac{T}{24}m^3}, \quad m \in \mathbb{Z}_{>0}$$

- Proven by [Chen 15] by Feynman–Kac (for flat-like initial condition).
- Proven in [Corwin–Ghosal 20] by formulas.

[Das-T 21]

$$\mathbf{E}\left[\left(Z(T,0)e^{\frac{T}{24}}\right)^p\right] = \mathbf{E}\left[e^{p(h(T,0)+\frac{T}{24})}\right] = e^{\frac{T}{24}p^3}, \quad p \in \mathbb{R}_{>0}$$

which gives the upper tail LDP

$$\mathbf{P}\left[h(T,0) + \frac{T}{24} \approx \mathbf{r}\right] \approx e^{-T \frac{4\sqrt{2}}{3} \mathbf{r}^{3/2}}, \quad \mathbf{r} \in \mathbb{R}_{>0}$$

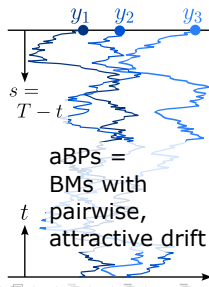
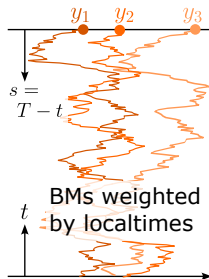
SHE moments \Leftrightarrow attractive BPs

By Feynman–Kac,

$$\begin{aligned} & \mathbf{E} [Z(T, y_1) \cdots Z(T, y_m)] \\ &= \mathbb{E}_{\text{BM}} \left[e^{\int_0^T ds \sum_{i < j} \delta_0(X_i - X_j)} \prod_{i=1}^m Z(0, X_i(T)) \right] \end{aligned}$$

By Tanaka + Girsanov,

$$\begin{aligned} & \mathbb{E}_{\text{BM}} \left[e^{\int_0^T ds \sum_{i < j} \delta_0(X_i - X_j)} (\cdot) \right] \\ &= e^{\frac{T}{24}(m^3 - m)} \\ & \mathbb{E}_{\text{aBP}} \left[e^{\sum_{i,j} \frac{1}{2} |X_i(0) - X_j(0)|} (\cdot) e^{-\sum_{i,j} \frac{1}{2} |X_i(T) - X_j(T)|} \right] \\ & dX_i(s) = \sum_{j=1}^m \frac{1}{2} \text{sgn}(X_j - X_i) ds + dB_i(s) \end{aligned}$$



Another way to get the attractive BPs

Define $Q(t, y_1, \dots, y_m) := \mathbf{E}[Z(t, y_1) \cdots Z(t, y_m)]$. By Itô,

$$\partial_t Q = \underbrace{\left(\frac{1}{2} \sum_{i=1}^m \partial_{y_i}^2 + \sum_{i < j} \delta_0(y_i - y_j) \right)}_{:= -H} Q$$

The Hamiltonian H has the ground state

$$\psi(y_1, \dots, y_m) = \exp\left(-\frac{1}{2} \sum_{i < j} |y_i - y_j|\right)$$

Performing the ground-state transformation gives

$$\frac{1}{\psi} (-H) \psi = \frac{m^3 - m}{24} + \underbrace{\sum_{i \neq j} \frac{1}{2} \operatorname{sgn}(x_j - x_i) \partial_{x_i}}_{:= L, \text{ generator of attractive BPs}} + \frac{1}{2} \sum_{i=1}^m \partial_{x_i}^2$$

Multi-point moments

The goal of this talk is to

1. get

$$\mathbf{E} \left[\prod_{c=1}^n (Z(T, T\mathbf{x}_c) e^{\frac{T}{24}})^{m_c} \right] = \mathbf{E} \left[e^{\sum_{c=1}^n m_c (h(T, T\mathbf{x}_c) + \frac{T}{24})} \right] \approx e^{T L_{\text{SHE}}(\vec{\mathbf{x}}, \vec{\mathbf{m}})},$$

for $\vec{\mathbf{x}} = (\mathbf{x}_1, \dots, \mathbf{x}_n) \in \mathbb{R}^n$, $\vec{\mathbf{m}} \in (\mathbb{Z}_{>0})^n$ and an explicit $L_{\text{SHE}}(\vec{\mathbf{x}}, \vec{\mathbf{m}})$, and

2. then use the moments to get the LDP and limit shape (caveat to be explained).

[Lin 23]: Doing Step 1 by formulas.

High powers

We'll actually consider powers that grow to $+\infty$ with $T \rightarrow \infty$:

$$\mathbf{E} \left[\prod_{c=1}^n (Z(T, 0) e^{\frac{T}{24}})^{m_c N} \right] = \mathbf{E} \left[e^{\sum_{c=1}^n m_c N (h(T, \mathbf{x}_c) + \frac{T}{24})} \right] \approx e^{TN^3 L_{\text{SHE}}(\vec{\mathbf{x}}, \vec{\mathbf{m}})},$$

for $\vec{\mathbf{x}} \in \mathbb{R}^n$, $\vec{\mathbf{m}} \in (\frac{1}{N} \mathbb{R}_{>0})^n$, and $N \rightarrow \infty$ at arbitrary rate relative to $T \rightarrow \infty$.

Why doing this?

$$\mathbf{E} \left[(Z(T, 0) e^{\frac{T}{24}})^m \right] = \mathbf{E} \left[e^{m(h(T, 0) + \frac{T}{24})} \right] \approx e^{\frac{T}{24} m^3}, \quad m \in \mathbb{Z}_{>0}$$

High powers

We'll actually consider powers that grow to $+\infty$ with $T \rightarrow \infty$:

$$\mathbf{E} \left[\prod_{c=1}^n \left(Z(T, 0) e^{\frac{T}{24}} \right)^{m_c N} \right] = \mathbf{E} \left[e^{\sum_{c=1}^n m_c N \left(h(T, \mathbf{x}_c) + \frac{T}{24} \right)} \right] \approx e^{TN^3 L_{\text{SHE}}(\vec{\mathbf{x}}, \vec{\mathbf{m}})},$$

for $\vec{\mathbf{x}} \in \mathbb{R}^n$, $\vec{\mathbf{m}} \in (\frac{1}{N}\mathbb{R}_{>0})^n$, and $N \rightarrow \infty$ at arbitrary rate relative to $T \rightarrow \infty$.

Why doing this?

$$\mathbf{E} \left[\left(Z(T, 0) e^{\frac{T}{24}} \right)^{Nm} \right] = \mathbf{E} \left[e^{Nm \left(h(T, 0) + \frac{T}{24} \right)} \right] \approx e^{\frac{TN^3}{24} m^3}, \quad m \in \frac{1}{N}\mathbb{Z}_{>0}$$

- $\frac{1}{N}\mathbb{Z}_{>0} \rightarrow \mathbb{R}_{>0}$ which gives the full upper-tail LDP.
- But doing this changes the scale of deviations:

$$\mathbf{P} \left[h(T, NT\mathbf{x}_c) + \frac{T}{24} \approx TN^2 \mathbf{r}_c, c = 1, \dots, n \right] \approx \exp \left(- TN^3 I_{\text{KPZ}}(\vec{\mathbf{x}}, \vec{\mathbf{r}}) \right)$$

LDP for the attractive BPs

$$\mu_N(s) := \frac{1}{N} \sum_{i=1}^{Nm} \delta_{X_i^N(s)}, \quad \mu_N \in \mathcal{C}([0, 1], \mathfrak{m}\mathcal{P}(\mathbb{R}))$$

Take any $T = T_N$ with $N^2T = N^2T_N \rightarrow \infty$.

Theorem (T 23)

As $N \rightarrow \infty$, the empirical measure μ_N satisfies an LDP on $\mathcal{C}([0, 1], \mathfrak{m}\mathcal{P}(\mathbb{R}))$ with speed N^3T and an explicit rate function \mathbb{I} .

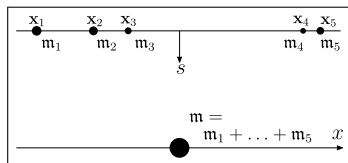
Remark

- Under the diffusive scaling, $N \rightarrow \infty$ and N^2T fixed, [\[Dembo–Shkolnikov–Varadhan–Zeitouni 16\]](#) proved the LDP for a general class of rank-based diffusions.
- The behavior under $N^2T \rightarrow \infty$ (considered here) is very different from that under the diffusion scaling (considered in [\[DSVZ 12\]](#)).

Back to moment Lyapunov exponents

Corollary

Under $Z(0, \cdot) = \delta_0$,
$$\mathbf{E} \left[\prod_{c=1}^n Z(T, NT\mathbf{x}_c)^{Nm_c} \right] \approx e^{N^3 T \cdot L_{\text{SHE}}(\vec{m})}$$



$$\begin{aligned} L_{\text{SHE}}(\vec{m}) &= L_{\text{SHE}}(m_1, \dots, m_n) \\ &:= \frac{m^3}{24} + \sum_{c, c'=1}^n \frac{1}{2} m_c m_{c'} |\mathbf{x}_c - \mathbf{x}_{c'}| - \mathbb{I}_* \end{aligned}$$

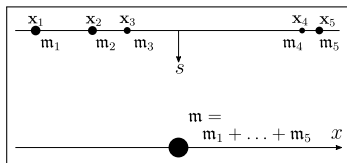
$$\mathbb{I}_* := \inf \left\{ \mathbb{I}(\mu) : \mu \in \mathcal{C}([0, 1], \mathfrak{m}\mathcal{P}(\mathbb{R})), \mu(0) = \sum_{c=1}^n m_c \delta_{\mathbf{x}_c}, \mu(1) = m\delta_0 \right\}$$

Remark. The initial condition should actually be: $Z(0, \cdot) = \mathbf{1}_{[-\alpha, \alpha]}$, with $N \rightarrow \infty$ first and $\alpha \rightarrow 0$ later. A separate argument in [Lin-T 23] shows that this initial condition approximates the true delta initial condition.

Back to moment Lyapunov exponents

Corollary

Under $Z(0, \cdot) = \delta_0$, $\mathbf{E} \left[\prod_{c=1}^n Z(T, NT\mathbf{x}_c)^{Nm_c} \right] \approx e^{N^3 T \cdot L_{\text{SHE}}(\vec{m})}$



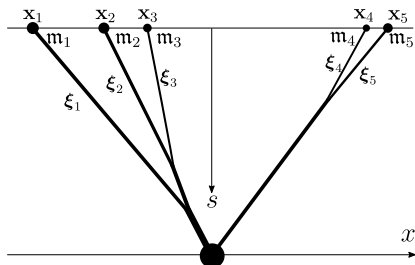
$$\begin{aligned} L_{\text{SHE}}(\vec{m}) &= L_{\text{SHE}}(m_1, \dots, m_n) \\ &:= \frac{m^3}{24} + \sum_{c, c'=1}^n \frac{1}{2} m_c m_{c'} |\mathbf{x}_c - \mathbf{x}_{c'}| - \mathbb{I}_* \end{aligned}$$

$$\mathbb{I}_* := \inf \left\{ \mathbb{I}(\mu) : \mu \in \mathcal{C}([0, 1], \mathbf{m}\mathcal{P}(\mathbb{R})), \mu(0) = \sum_{c=1}^n m_c \delta_{\mathbf{x}_c}, \mu(1) = m\delta_0 \right\}$$

Theorem (T 23)

The infimum has a unique minimizer $\mu = \xi$, the **optimal deviation**, which we describe next.

Optimal deviation



Optimal clusters ξ_1, \dots, ξ_n and optimal deviation ξ

• $\xi_c(s) :=$

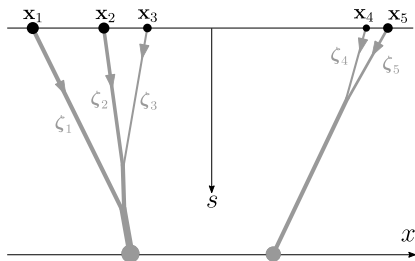
$$\xi(s) = \sum_{c=1}^n m_c \delta_{\xi_c(s)}$$

Optimal deviation

$$dX_i^N = \frac{1}{N} \sum_{j=1}^{Nm} \frac{1}{2} \operatorname{sgn}(X_j^N - X_i^N) ds + \frac{1}{\sqrt{N^2 T}} dB_i(s)$$

Inertia clusters, ζ_1, \dots, ζ_c

- ζ_c has mass m_c .



Optimal clusters ξ_1, \dots, ξ_n and optimal deviation ξ

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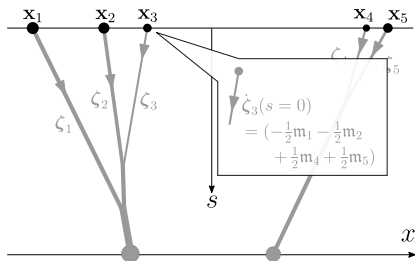
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Inertia clusters, ζ_1, \dots, ζ_c

- ζ_c has mass m_c .
- Start with velocity
($\dots - \frac{1}{2}m_{c-1} + \frac{1}{2}m_{c+1} + \dots$).



Optimal clusters ξ_1, \dots, ξ_n and optimal deviation ξ

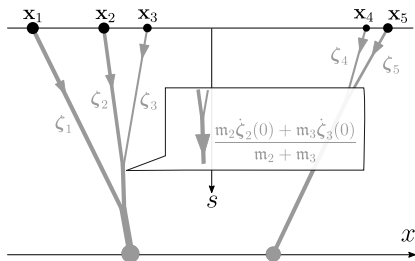
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- ζ_c has mass m_c .
- Start with velocity $(\dots - \frac{1}{2}m_{c-1} + \frac{1}{2}m_{c+1} + \dots)$.
- Merge according to conservation of momentum.



Optimal clusters ξ_1, \dots, ξ_n and optimal deviation ξ

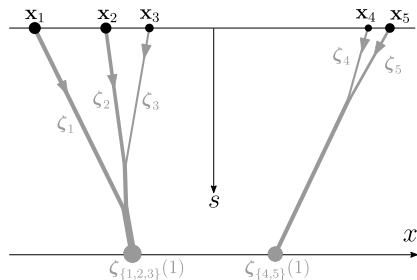
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Inertia clusters, ζ_1, \dots, ζ_c

- ζ_c has mass m_c .
- Start with velocity $(\dots - \frac{1}{2}m_{c-1} + \frac{1}{2}m_{c+1} + \dots)$.
- Merge according to conservation of momentum.



Branches, b : $c, c' \in b$ if and only if $\zeta_c(1) = \zeta_{c'}(1)$

Optimal clusters ξ_1, \dots, ξ_n and optimal deviation ξ

- $\xi_c(s) :=$

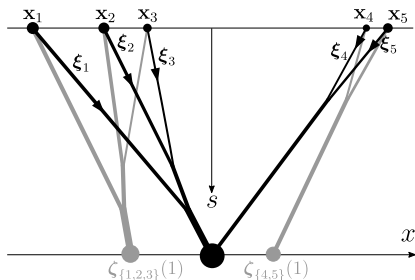
$$\xi(s) = \sum_{c=1}^n m_c \delta_{\xi_c(s)}$$

Optimal deviation

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Inertia clusters, ζ_1, \dots, ζ_c

- ζ_c has mass m_c .
- Start with velocity $(\dots - \frac{1}{2}m_{c-1} + \frac{1}{2}m_{c+1} + \dots)$.
- Merge according to conservation of momentum.

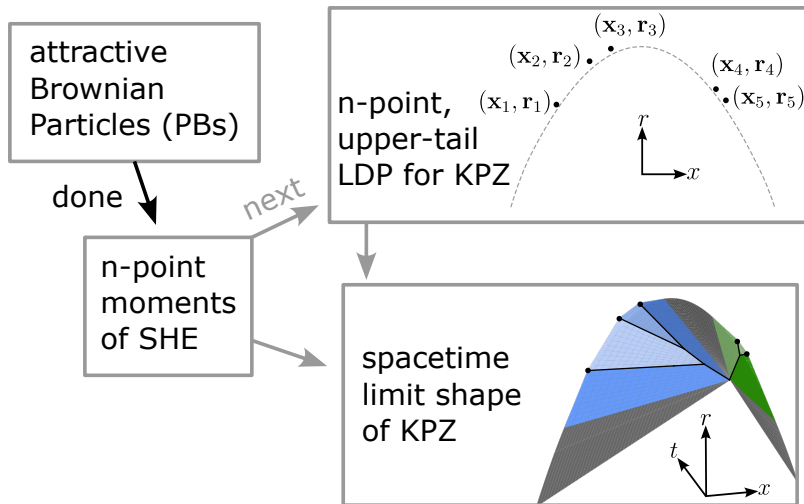


Branches, \mathfrak{b} : $c, c' \in \mathfrak{b}$ if and only if $\zeta_c(1) = \zeta_{c'}(1)$

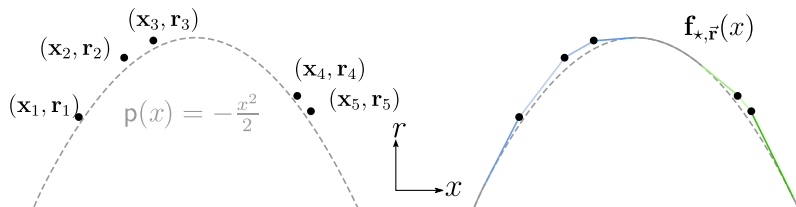
Optimal clusters ξ_1, \dots, ξ_n and optimal deviation ξ

- $\xi_c(s) := \zeta_c(s) + (-\zeta_{\mathfrak{b}}(1))s, \quad c \in \mathfrak{b} \quad \xi(s) = \sum_{c=1}^n m_c \delta_{\xi_c(s)}$

So far and what's next



Moments of SHE \rightarrow LDP for KPZ: Legendre transform

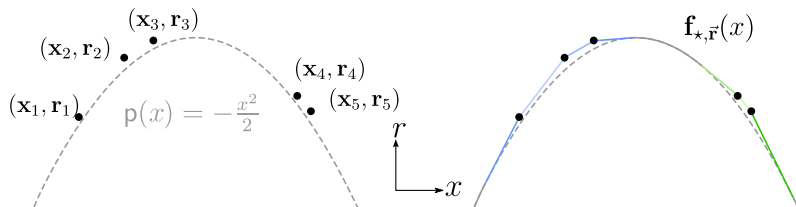


n -point, upper-tail rate function

$$I_{\text{KPZ}}(\vec{r}) = I_{\text{KPZ}}(\mathbf{r}_1, \dots, \mathbf{r}_n) := \int_{\mathbb{R}} dx \left(\frac{1}{2} (\partial_x f_{*, \vec{r}})^2 - \frac{1}{2} (\partial_x p)^2 \right)$$

Gibbs line ensembles [Corwin–Hammond 14, 16] and [Ganguly–Hegde 22].

Moments of SHE \rightarrow LDP for KPZ: Legendre transform



n -point, upper-tail rate function

$$I_{\text{KPZ}}(\vec{\mathbf{r}}) = I_{\text{KPZ}}(\mathbf{r}_1, \dots, \mathbf{r}_n) := \int_{\mathbb{R}} dx \left(\frac{1}{2} (\partial_x f_{*, \vec{\mathbf{r}}})^2 - \frac{1}{2} (\partial_x p)^2 \right)$$

Theorem (T 23)

Let $\mathcal{R}_{\text{conc}} := \{ \vec{\mathbf{r}} : f_{*, \vec{\mathbf{r}}} \geq p, f_{*, \vec{\mathbf{r}}} \text{ is concave} \}$. The functions

$$L_{\text{SHE}}(\vec{\mathbf{m}}) : [0, \infty)^n \rightarrow [0, \infty) \quad I_{\text{KPZ}}(\vec{\mathbf{r}}) : \mathcal{R}_{\text{conc}} \rightarrow [0, \infty)$$

are strictly convex and the Legendre transform of each other.

n -point, upper-tail LDP for the KPZ equation

$$h_N(t, x) := \frac{1}{N^2 T} (h(Tt, NTx) - \log \sqrt{T})$$

$$\mathcal{E}_{N, \delta}(\vec{\mathbf{r}}) := \{ |h_N(1, \mathbf{x}_c) - \mathbf{r}_c| \leq \delta, \mathbf{c} = 1, \dots, n \}$$

Corollary (T 23 & Lin–T 23)

Under delta initial condition $Z(0, \cdot) = \delta_0$, for any $\vec{\mathbf{r}} \in \mathcal{H}_{\text{conc}}^\circ$,

$$\mathbf{P}[\mathcal{E}_{N, \delta}(\vec{\mathbf{r}})] \approx e^{-N^3 T \cdot I_{\text{KPZ}}(\vec{\mathbf{r}})}$$

$N \rightarrow \infty$ and $N^2 T = N^2 T_N \rightarrow \infty$ first; $\delta \rightarrow 0$ later.

Related results

First, when $n = 1$ and $\mathbf{x}_1 = 0$, we recover $I_{\text{KPZ}}(\mathbf{r}) = \frac{4\sqrt{2}}{3}\mathbf{r}^{3/2}$.

One point, upper-tail LDPs

- Hyperbolic scaling regime

$$\mathbf{P}\left[\frac{1}{T}h(Tt, 0) \approx -\frac{1}{24} + \mathbf{r}\right] \approx e^{-T \frac{4\sqrt{2}}{3}\mathbf{r}^{3/2}}, \quad T \rightarrow \infty, \mathbf{r} > 0$$

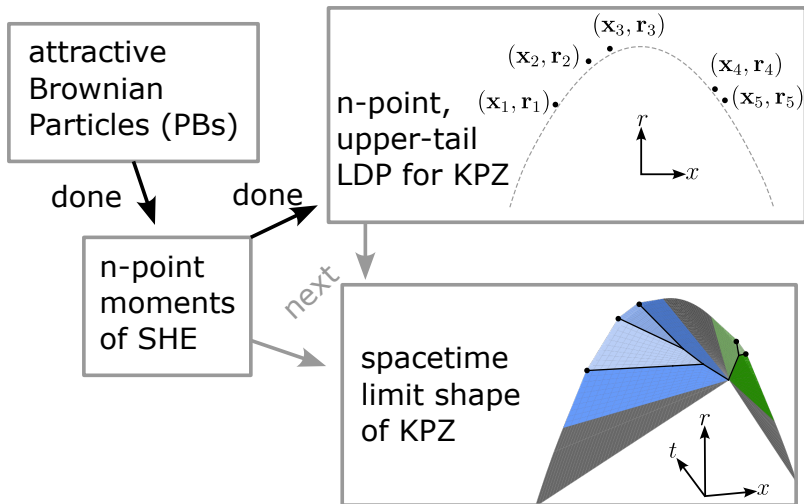
- Predicted in [Le Doussal–Majumdar–Schehr 16]; proven in [Das–T 21].
- Other scaling regimes and/or other initial conditions
 - Physics: Asida, Hartman, Janas, Kolokolov, Korshunov, Katzav, Krajenbrink, Le Doussal, Majumdar, Livne, Meerson, Prolhac, Rosso, Sasorov, Schmidt, Smith, Vilenkin, ...
 - Math rigorous: Corwin, Das, Gaudreau Lamarre, Ghosal, Lin, Tsai, ...

n -point upper tails and terminal-time limit shape

n -point upper tails and terminal-time limit shape

- [Ganguly–Hegde 22]
 - Detailed and optimal n -point bounds that hold for all $t > t_0$.
 - When specialized onto the hyperbolic scaling regime: the n -point LDP and the terminal-time limit shape $f_{\star, \mathbf{r}}$.

So far and what's next



Spacetime limit shape

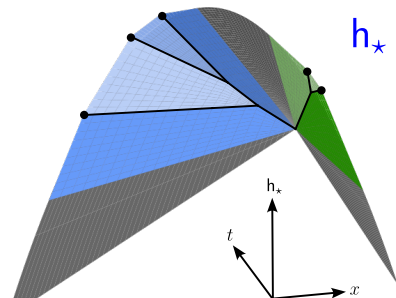
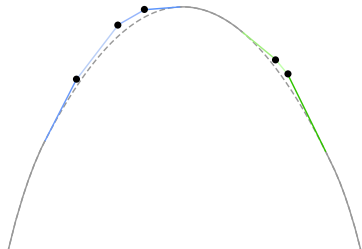
Theorem (Lin-T 23)

Under $Z(0, \cdot) = \delta_0$, for any $\vec{\mathbf{r}} \in \mathcal{R}_{conc}^\circ$ and $R < \infty$,

$$\mathbf{P} \left[\|h_N - \mathbf{h}_\star\|_{\mathcal{L}^\infty([\frac{1}{R}, 1] \times [-R, R])} < \frac{1}{R} \mid \mathcal{E}_{N, \delta}(\vec{\mathbf{r}}) \right] \longrightarrow 1$$

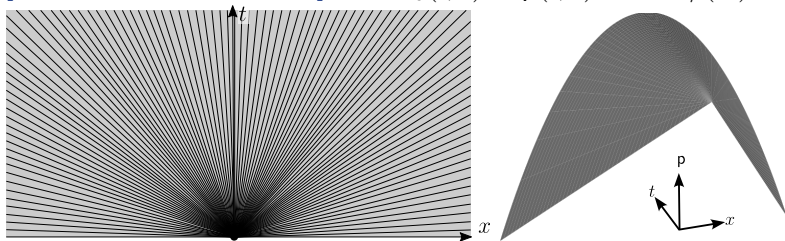
$N \rightarrow \infty$ and $N^2 T = N^2 T_N \rightarrow \infty$ first; $\delta \rightarrow 0$ later.

$$f_{\star, \vec{\mathbf{r}}}(x) = \mathbf{h}_\star(1, x)$$



Hydrodynamic limit (without conditioning)

- [Janjigian–Rassoul-Agha–Seppäläinen 22] The hydrodynamic limit h_0 is the entropy solution of $\partial_t h_0 = \frac{1}{2}(\partial_x h_0)^2$.
- [Amir–Corwin–Quastel 11] Here $h_0(t, x) = p(t, x) := -x^2/(2t)$.

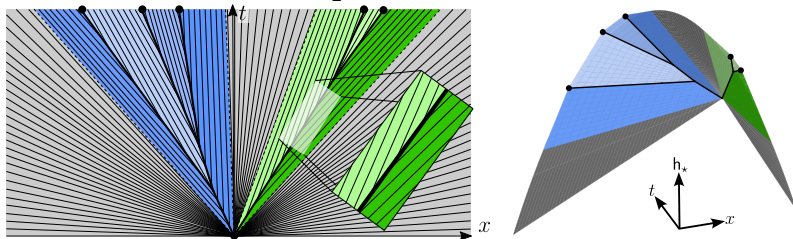


Limit shape (with conditioning)

- $h_\star(t, x)$ also solves $\partial_t h_\star = \frac{1}{2}(\partial_x h_\star)^2$, but is a *non-entropy* solution.

Limit shape (with conditioning)

- $h_*(t, x)$ also solves $\partial_t h_* = \frac{1}{2}(\partial_x h_*)^2$, but is a *non-entropy* solution.



- How to describe h_* ?

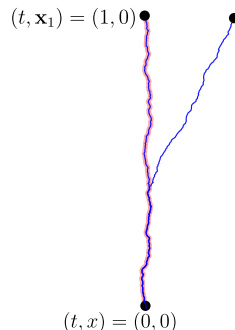
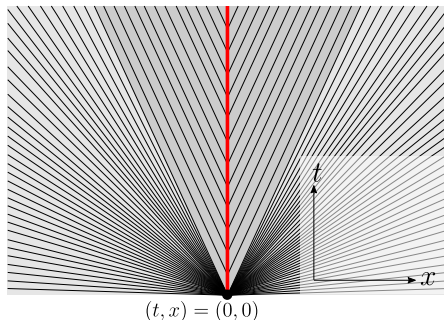
$h_*(1-s, x)$ is the entropy solution of the *backward* equation
 $-\partial_s h_* = \frac{1}{2}(\partial_x h_*)^2$.

Consistent with [Jensen 00] [Varadhan 04]

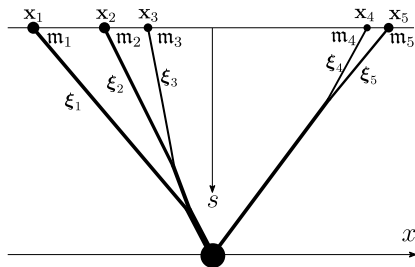
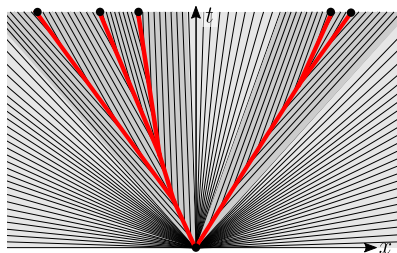
Mechanism of the deviations

$$e^{h(T, NTx)} = Z(T, NTx) = \mathbb{E}_{\text{BM}} \left[e^{\int_0^T ds \eta(T-s, X(s))} \delta_0(X(T)) \right]$$

Consider $n = 1$ and $\mathbf{x}_1 = 0$.



Noise corridors = optimal clusters in aBPs



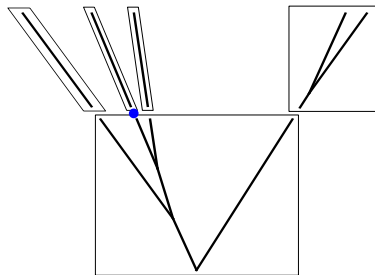
Proposition (T 23)

(Noise corridors in KPZ) = (optimal clusters in attractive BPs)

Elements of the proof

Given $\mathcal{E}_{N,\delta}(\vec{\mathbf{r}})$, we want to argue $h_N \approx h_\star$.

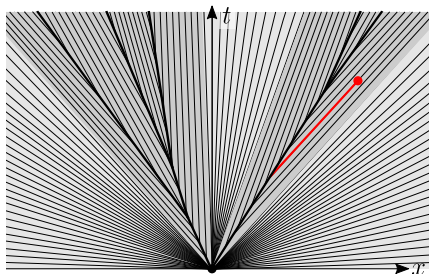
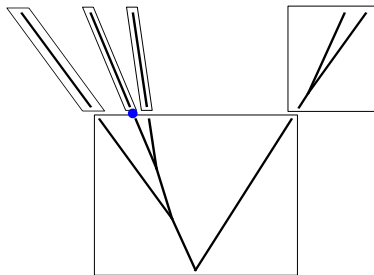
- Use a tree structure to show that $h_N(t, x) \approx h_\star(t, x)$ at any point (t, x) on the noise corridors / shocks.



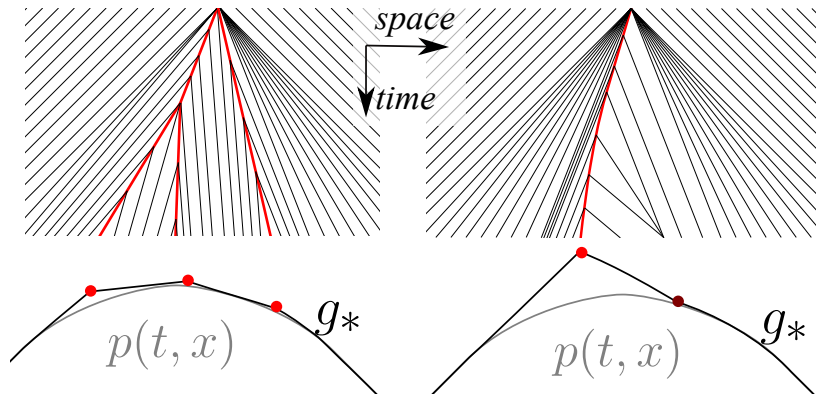
Elements of the proof

Given $\mathcal{E}_{N,\delta}(\vec{\mathbf{r}})$, we want to argue $h_N \approx h_\star$.

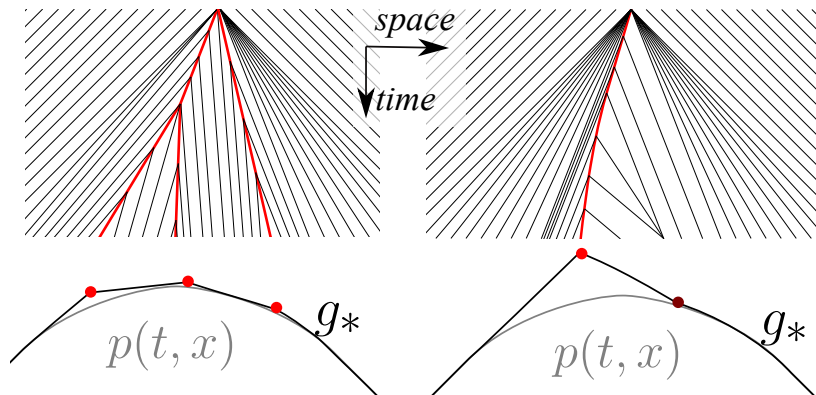
- Use a tree structure to show that $h_N(t, x) \approx h_\star(t, x)$ at any point (t, x) on the noise corridors / shocks.
- Once $h_N \approx h_\star$ holds along the noise corridors / shocks, the rest can be obtained by analyze the **increments of h_N along characteristics**.



Food for thought?



Food for thought?



Thank you for the attention and thanks to the organizers!