

Memory in Kardar-Parisi-Zhang growth: exact results via the replica Bethe ansatz, and experiments

P. Le Doussal (LPTENS)

- growth in plane, local stoch. rules => 1D KPZ class (integrability)
- discrete models in “KPZ class” => large time universality related to random matrix theory: Tracy Widom distributions of largest eigenvalue of GUE, GOE..
Airy process, determinantal structure **at level of one-time quantities**

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- continuum KPZ equation/equivalent continuum directed polymer problem:
possible to calculate **one-time distributions** (for some initial conditions)
for all times, and see convergence to TW

Replica Bethe Ansatz method: **in math: discrete models => rigorous replica**
integrable systems (Bethe Ansatz) +disordered systems(replica)

with : Pasquale Calabrese (SISSA) Alberto Rosso (LPTMS Orsay)

Thomas Gueudre (Torino) Andrea de Luca (Oxford)

Thimothee Thiery (Leuven)

large deviations KPZ/fermions

G. Schehr, S. Majumdar, D. Dean **Alexandre Krajenbrink (LPTENS)**

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- two-time observables for models in 1d KPZ class ?

Jacopo de Nardis, PLD, arXiv:1612.08695
J. Stat. Mech. (2017) 053212

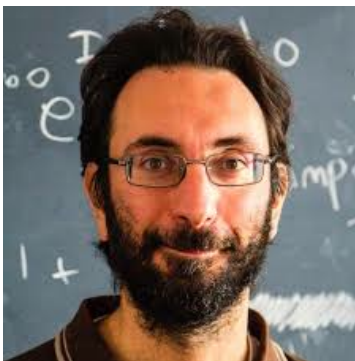
**Tail of the two-time height distribution
for KPZ growth in one dimension**

Jacopo de Nardis, PLD, K. Takeuchi,
Phys. Rev. Lett. 118, 125701 (2017)

Memory and universality in interface growth

PLD arXiv:1709.06264 (2017)

Maximum of an Airy process plus Brownian motion and memory in KPZ growth



Pasquale Calabrese
(SISSA)



Alberto Rosso
(LPTMS Orsay)



Thomas Gueudre
(U.Torino)



Marton Kormos
(U.Budapest)



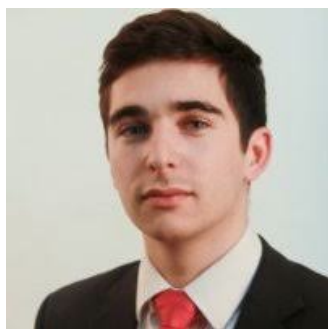
Thimothee Thiery
(U. Leuven)



Andrea de Luca
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Jacopo de Nardis
(ENS)



Alexandre Krajenbrink
(LPTENS)



Satya Majumdar + Gregory Schehr
(LPTMS Orsay)

David Dean
(Bordeaux Univ.)



Kazumaza Takeuchi (Tokyo U.)

Why is it interesting?

J. Baik, Z. Liu talk
multi-time
TASEP ring (2017)

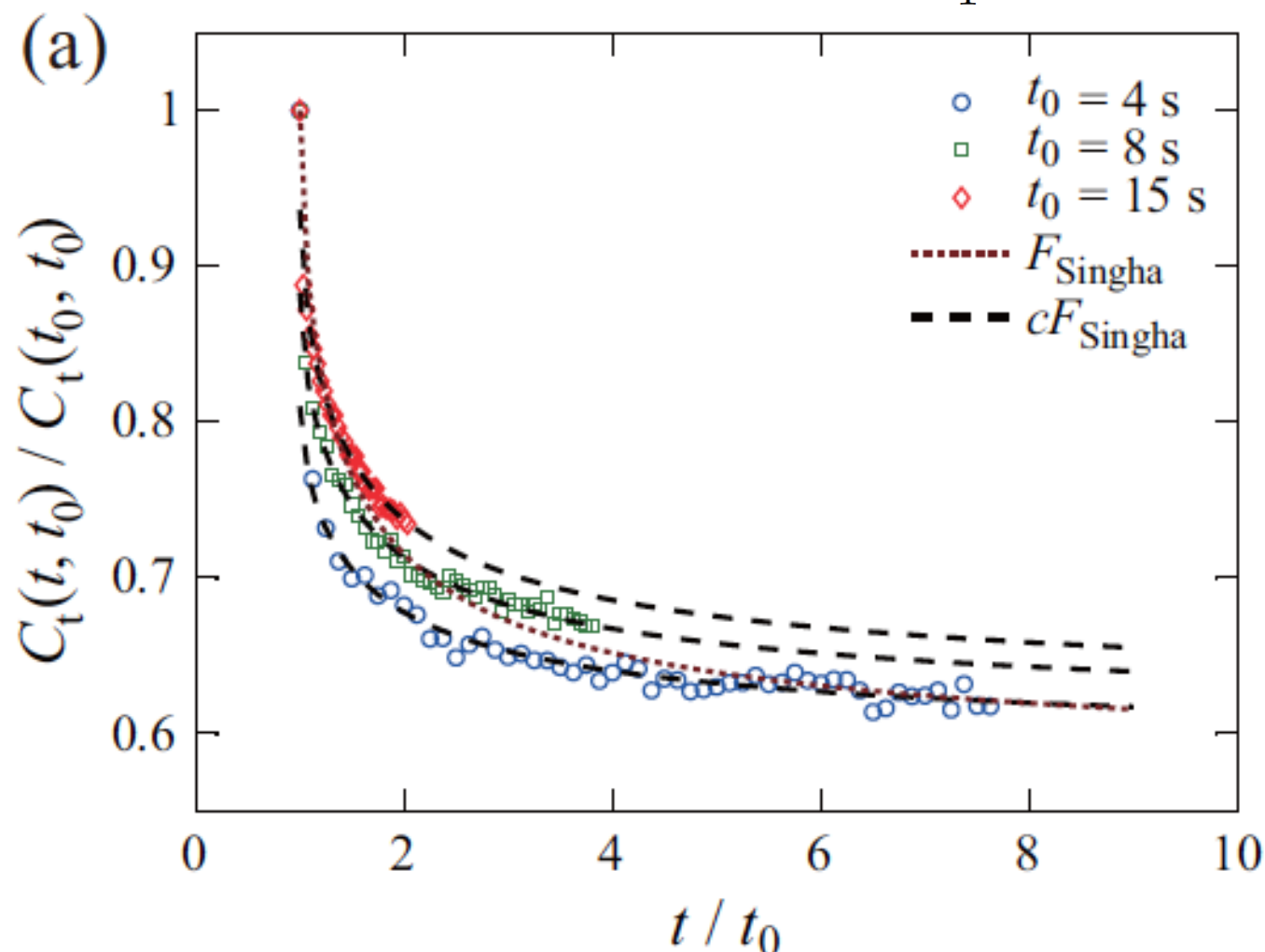
- what is large multi-time structure: determinantal?
- memory effect in time evolution: ergodicity breaking

droplet initial condition

two-time covariance ratio

$$\frac{\overline{h(0, t_1)h(0, t_2)}^c}{\overline{h(0, t_1)}^2} \xrightarrow[t_1, t_2 \rightarrow +\infty]{\frac{t_2 - t_1}{t_1} = \Delta} C_\Delta \xrightarrow[\Delta \rightarrow \infty]{} C_{+\infty} \begin{matrix} = 0 ? \\ > 0 ? \end{matrix}$$

persistent correlation



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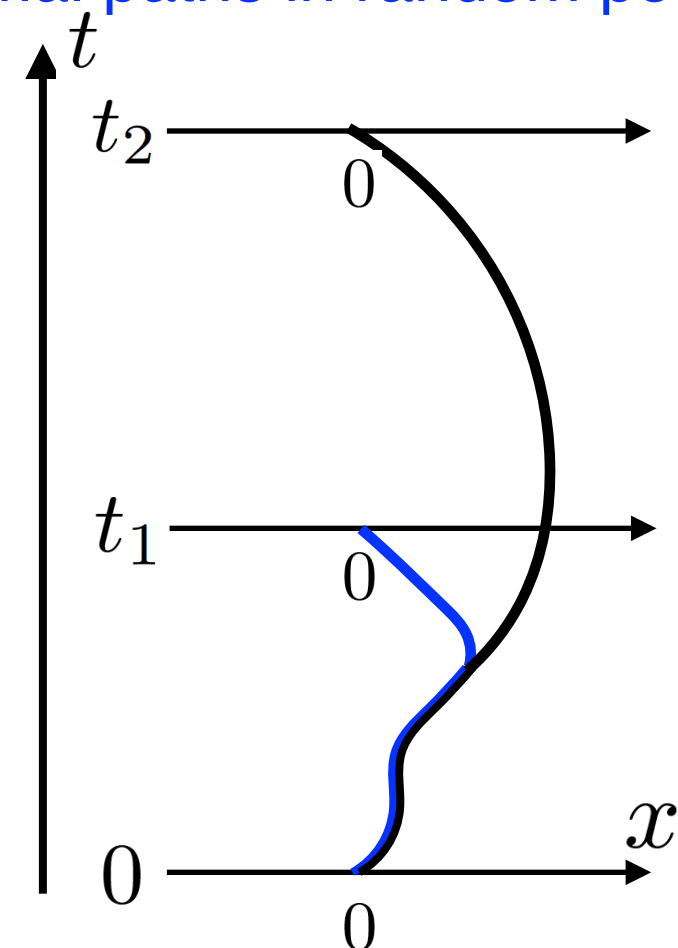
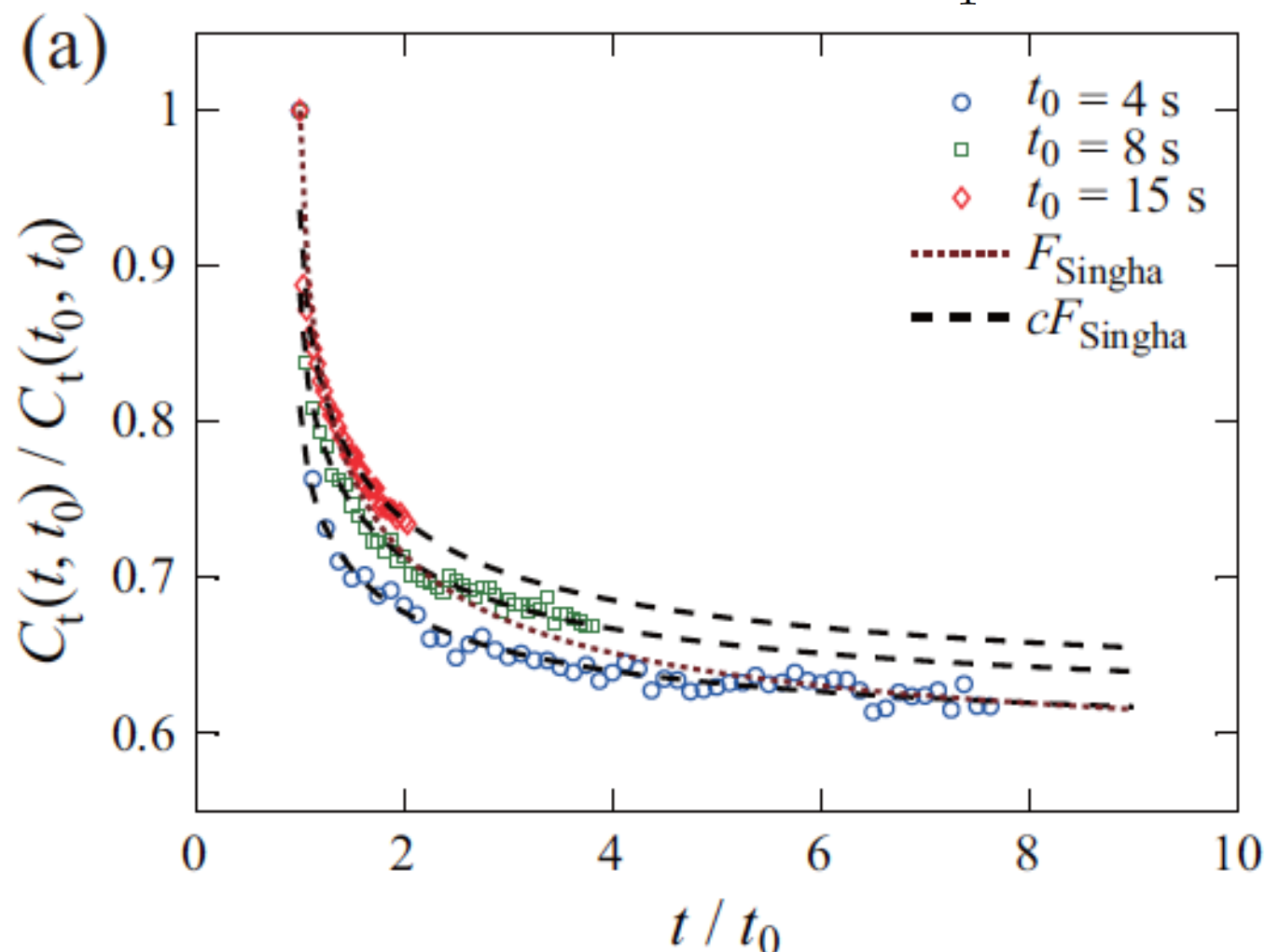
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$$\xrightarrow[\Delta \rightarrow \infty]{} C_{+\infty} \begin{matrix} = 0 ? \\ > 0 ? \end{matrix}$$

persistent correlation

droplet initial condition

directed polymer: overlap/coalescence
optimal paths in random potential



Part I : one-time KPZ/DP: Replica Bethe Ansatz (RBA)

- KPZ equation, KPZ class, random matrices, Tracy Widom distributions.
- solving KPZ at any time by mapping to directed paths
then using (imaginary time) quantum mechanics
attractive bose gas (integrable) \Rightarrow large time TW distrib. for KPZ height
- droplet initial condition \Rightarrow GUE
- flat initial condition \Rightarrow GOE
- half space initial condition \Rightarrow GSE
- stationary (Brownian) initial condition \Rightarrow Baik-Rains

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Part II: two-time KPZ via RBA joint PDF $h(0, t_1) h(0, t_2)$

II a) direct approach any $\frac{t_2 - t_1}{t_1} = \Delta$ exact tail of JPDF
droplet initial condition for “large” positive $h(0, t_1)$

=> height difference crossover from stationary to GUE

=> agrees with experiments/numerics in broad range of $h(0, t_1)$

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II b) limit $\Delta \rightarrow \infty$ exact JPDF of max of Airy process
—————> persistent correlations in KPZ

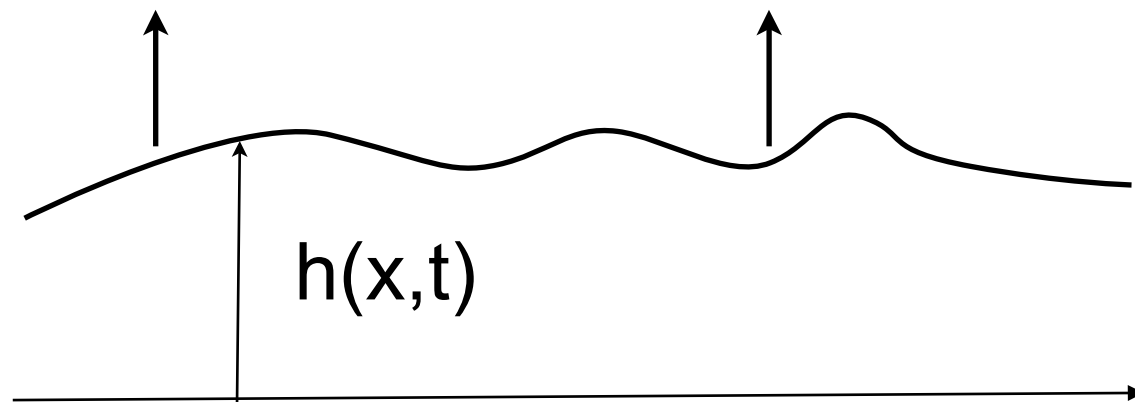
a) b) mutually agree in matching region !

both RBA+approximations (magic recipe)

how to model a growing interface ?

neglect overhangs

large scale effective model



Edwards-Wilkinson

$$\partial_t h = \nu \partial_x^2 h + \eta(x, t) + v$$

surface tension noise

$$h_{\omega, q} = \frac{\eta_{\omega, q}}{\nu q^2 + i\omega}$$

$$\overline{hh}(q, \omega) = \frac{D}{\nu^2 q^4 + \omega^2}$$

$$h \sim t^{1/4} \sim x^{1/2}$$

$$\overline{\eta(x, t)\eta(x', t')} = D\delta(x - x')\delta(t - t')$$

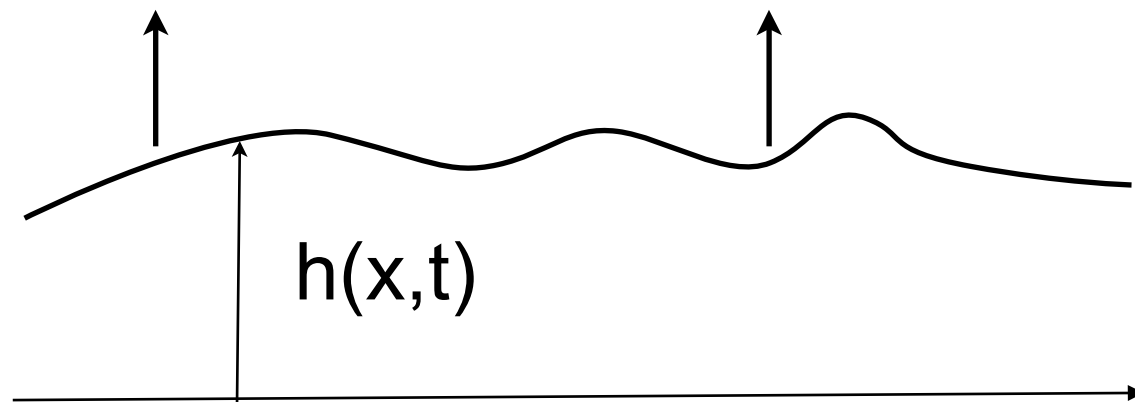
$$x \sim t^{1/2}$$

$P(h)$ is gaussian, simple diffusive dynamics

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Kardar Parisi Zhang equation

Phys Rev Lett 56 889 (1986)

non-linearity

A diagram showing a line with a slope in the \$x-h\$ plane. A vertical displacement \$\delta h\$ is indicated, and a dashed line shows the growth direction. The angle \$\theta\$ is shown between the growth direction and the vertical.

$$\begin{aligned} \frac{\partial h}{\partial t} &= \frac{v}{\cos \theta} \\ &= v \sqrt{1 + (\partial_x h)^2} \\ &\approx v + \frac{v}{2} (\partial_x h)^2 \end{aligned}$$

$$\partial_t h = \nu \partial_x^2 h + \frac{\lambda_0}{2} (\partial_x h)^2 + \eta(x, t)$$

noise

random deposition

Kardar Parisi Zhang equation

Phys Rev Lett 56 889 (1986)

growth of an interface of height $h(x,t)$

$$\partial_t h = \underbrace{\nu \partial_x^2 h}_{\text{diffusion}} + \frac{\lambda_0}{2} (\partial_x h)^2 + \underbrace{\eta(x,t)}_{\text{noise}}$$

$$\overline{\eta(x,t)\eta(x',t')} = D\delta(x-x')\delta(t-t')$$

- 1D scaling exponents $h \sim t^{1/3} \sim x^{1/2} \quad x \sim t^{2/3}$

- $P(h=h(x,t))$ non gaussian

even at large time PDF depends on
some broad features
of initial condition

flat $h(x,0)=0$

wedge $h(x,0) = -w|x|$ (droplet)

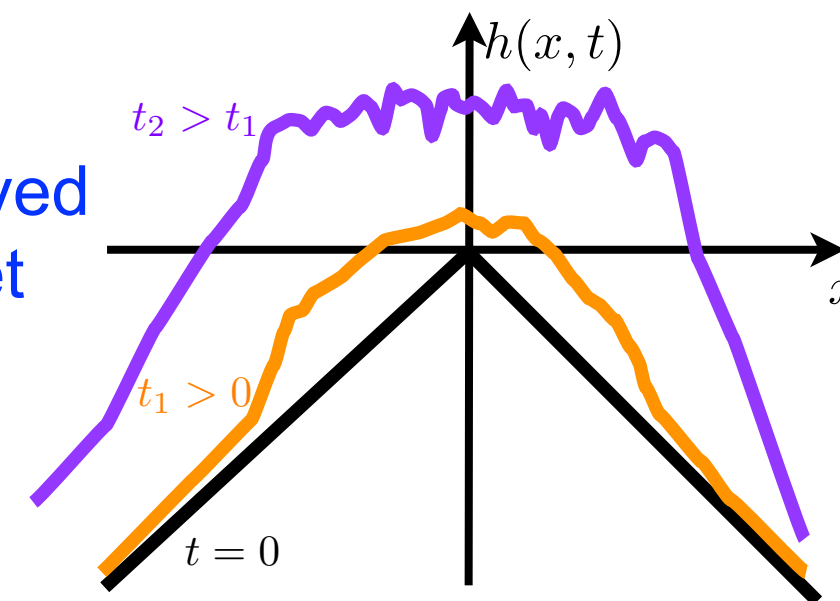
stationary $h(x,0)=B(x)$

related to different RMT ensembles !

- short-time regime $\lambda_0 = 0$

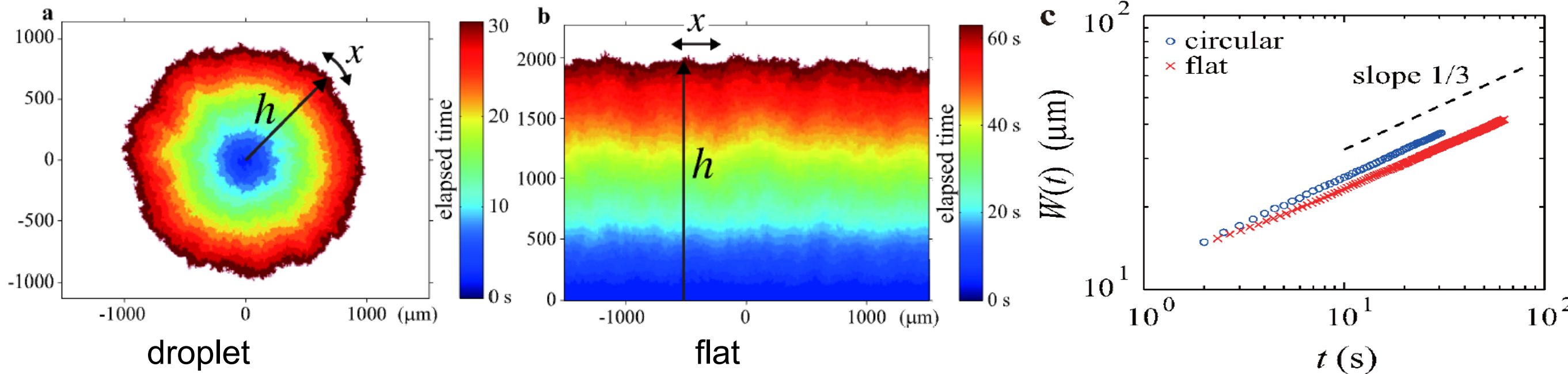
Edwards Wilkinson $P(h)$ gaussian

wedge=curved
=droplet



- Turbulent liquid crystals

Takeuchi, Sano PRL 104 230601 (2010)



$$W(t) \equiv \sqrt{\langle [h(x,t) - \langle h \rangle]^2 \rangle}$$

$$h(x, t) \simeq_{t \rightarrow +\infty} v_{\infty} t + \chi t^{1/3}$$

χ is a random variable

$$h \sim t^{1/3} \sim x^{1/2}$$

also reported in:

- slow combustion of paper

J. Maunuksela et al. PRL 79 1515 (1997)

- bacterial colony growth

Wakita et al. J. Phys. Soc. Japan. 66, 67 (1996)

- fronts of chemical reactions

S. Atis (2012)

- formation of coffee rings via evaporation

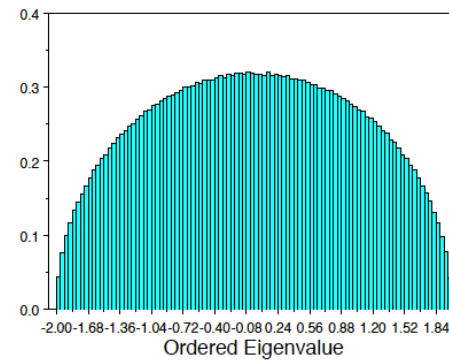
Yunker et al. PRL (2012)

Large N by N random matrices H, with Gaussian independent entries

eigenvalues λ_i $i = 1, ..N$

Universality large N :

- DOS: semi-circle law



$\beta =$ 1 (GOE) real symmetric
2 (GUE) hermitian
4 (GSE) symplectic

- distribution of the largest eigenvalue

Tracy Widom (1994)

$$H \rightarrow NH$$

$$\lambda_{max} = 2N + \chi N^{1/3}$$

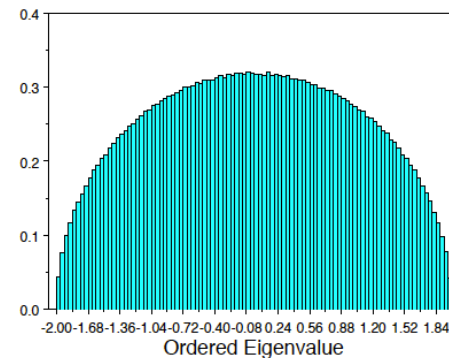
$$Prob(\chi < s) = F_{\beta}(s)$$

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$$H \rightarrow NH \quad \lambda_{max} = 2N + \chi N^{1/3} \quad Prob(\chi < s) = F_\beta(s)$$

CDF given by Fredholm determinants

$$\text{GUE} \quad F_2(s) = \text{Det}(I - K) \quad K(a, b) = K_{Ai}(a, b)\theta(a - s)\theta(b - s)$$

$$K_{Ai}(a, b) = \int_0^{+\infty} dv Ai(a + v) Ai(b + v) \quad \text{Airy kernel}$$

$$\text{GOE} \quad K_1(x, y) = \theta(x) Ai(x + y + s)\theta(y)$$

What is a Fredholm determinant?

$$K(x, y) = \theta(x)K(x, y)\theta(y)$$

$$\begin{aligned} \text{Det}[I - K] &= e^{\text{Tr} \text{Ln}(I-K)} = e^{-\sum_{p=1}^{\infty} \frac{1}{p} \text{Tr} K^p} \\ &= 1 - \text{Tr} K + \frac{1}{2}(\text{Tr} K^2 - (\text{Tr} K)^2) + .. \end{aligned}$$

$$\text{Det}[I - K] = \sum_p \frac{(-1)^p}{p!} \int_{v_1 > 0, \dots, v_p > 0} \det[K(v_i, v_j)]_{p \times p}$$

$$(I - K)\phi(x) = \phi(x) - \int_y K(x, y)\phi(y)$$

Calculation of F1(s)

$$d(z) = \det(I - zK|_{L^2(a,b)})$$

$$\sum_{j=1}^m w_j f(x_j) \approx \int_a^b f(x) dx$$

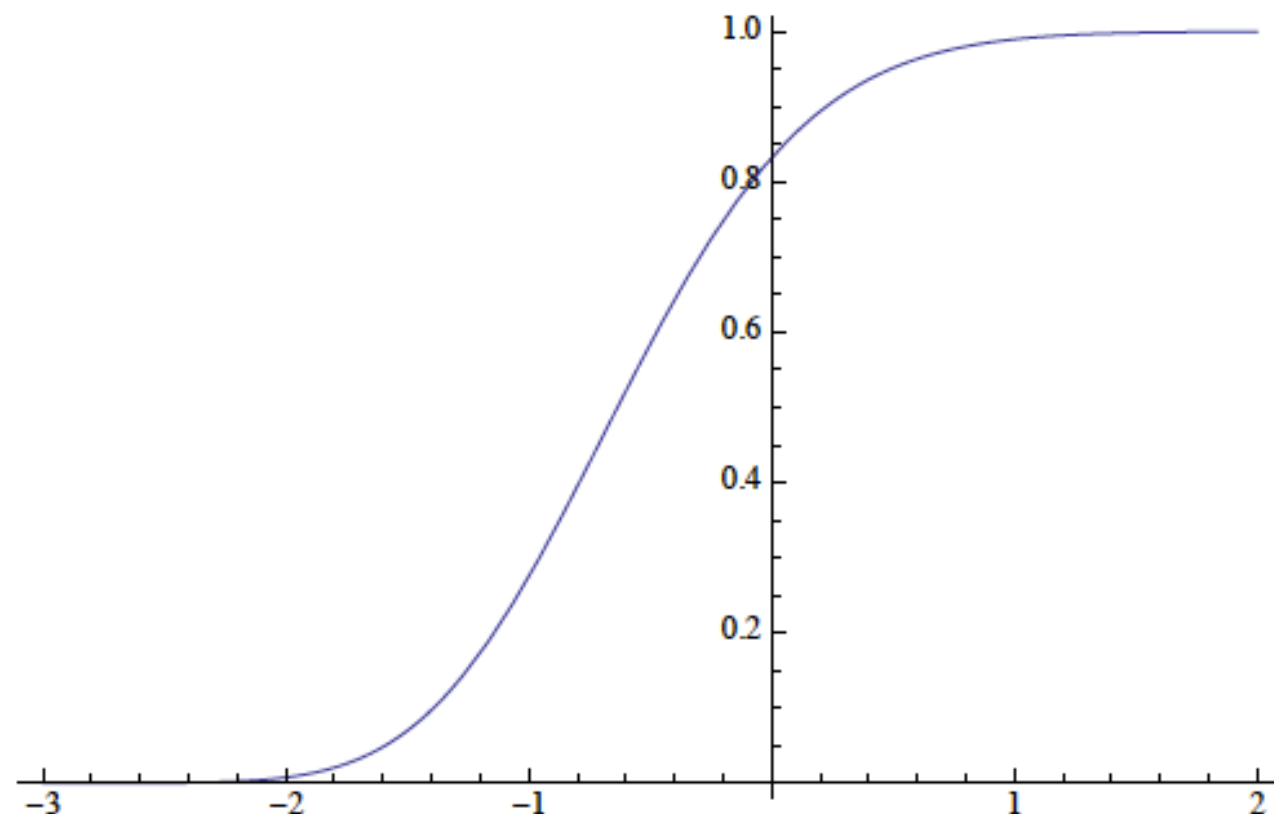
Gauss Legendre quadrature rule

$$d_m(z) = \det \left(\delta_{ij} - z w_i^{1/2} K(x_i, x_j) w_j^{1/2} \right)_{i,j=1}^m$$

```
In[1]:= Needs["NumericalCalculus`"]
GaussLegendre[a_, b_, m_] :=
Module[{beta, T, V, c, d, e}, beta = Table[i / Sqrt[(2 i - 1) (2 i + 1)], {i, 1, m - 1}];
T = DiagonalMatrix[beta, -1] + DiagonalMatrix[beta, 1];
V = Eigensystem[N[T, 10]]; e = V[[2]]; d = Table[e[[i, 1]], {i, 1, m}];
c = (V[[1]] + 1) / 2; {d^2 (b - a), (1 - c) a + b c}
FredholmDet[K_, z_, a_, b_, m_] := Module[{w, x}, {w, x} = GaussLegendre[a, b, m];
w = Sqrt[w]; Det[IdentityMatrix[m] + (Transpose[{w}] . {w}) Outer[K, x, x]]]
```

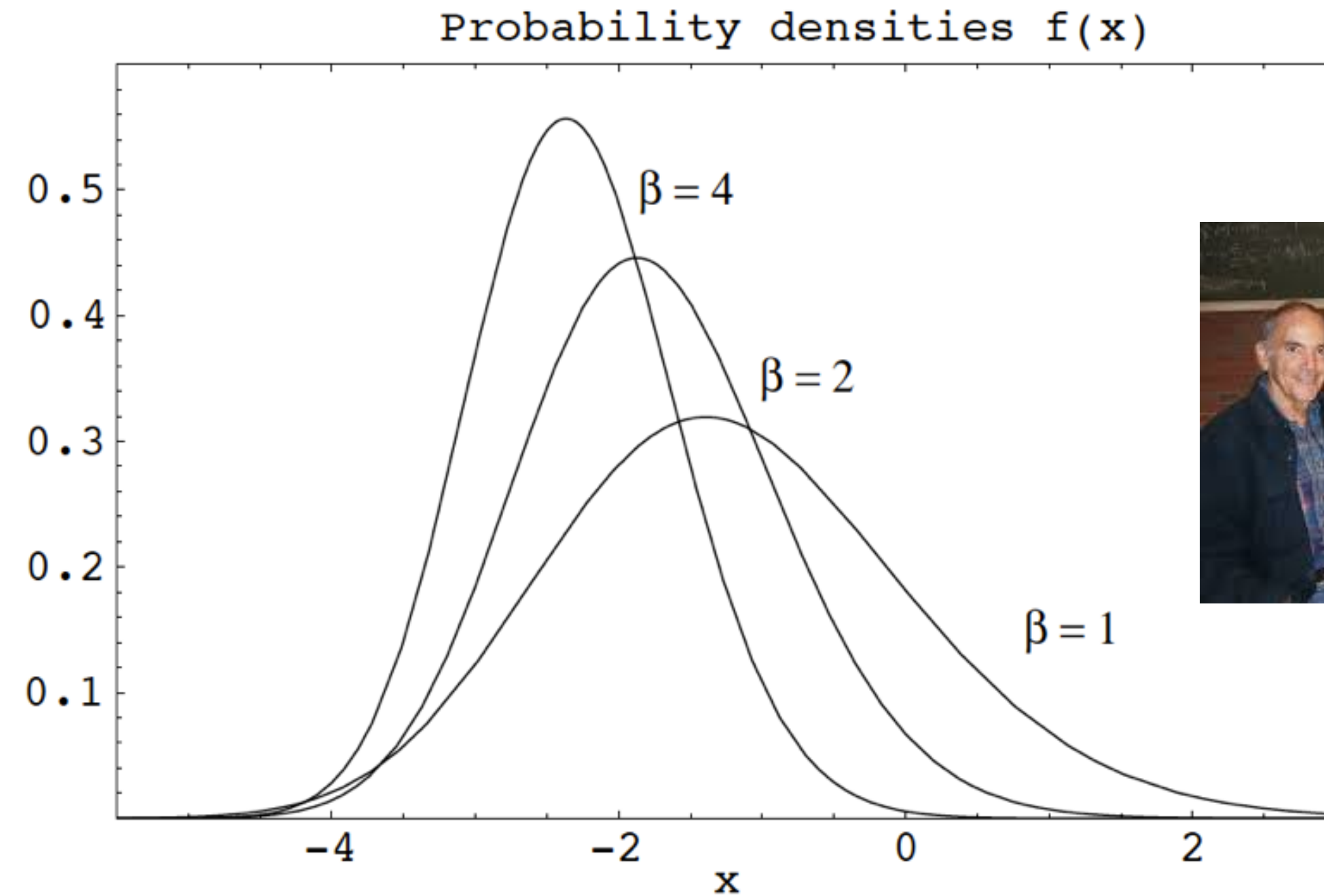
```
In[24]:= K[x_, y_] = -AiryAi[x + y];
g[x0_, b_, m_] := N[FredholmDet[K, x0, x0, b, m]]
g2 = Interpolation[Table[{x, g[x, 8, 20]}, {x, -10, 5, 0.2}]];
Plot[g2[s], {s, -3, 2}]
```

Out[27]=



Bornemann (2009)

Tracy Widom distributions Tracy Widom (1994)



Exact results for height distributions for some discrete models in KPZ class

- PNG model

$$h(0, t) \simeq_{t \rightarrow \infty} 2t + t^{1/3} \chi$$

Baik, Deift, Johansson (1999)

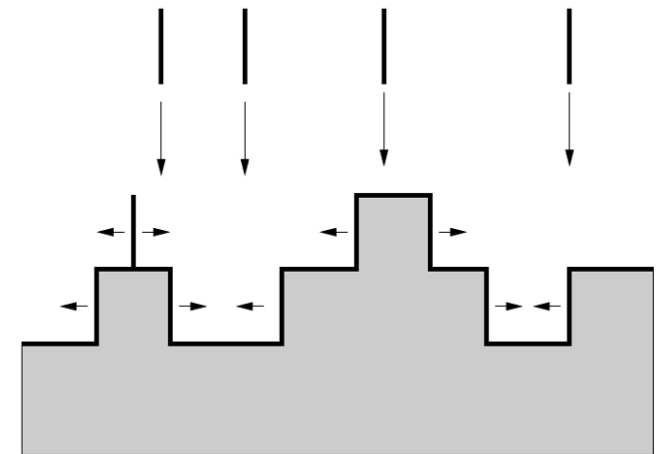
droplet IC

GUE

Prahofer, Spohn, Ferrari, Sasamoto,..
(2000+)

flat IC

GOE



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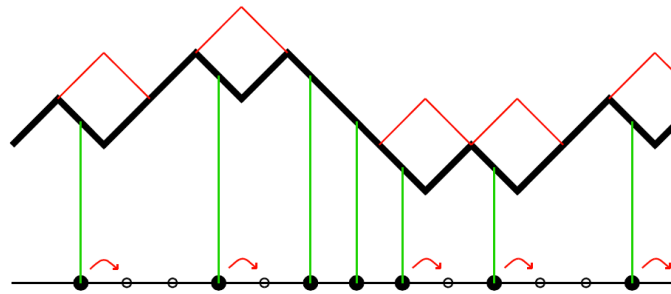
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- similar results for totally asymmetric exclusion process (TASEP)

Johansson
(1999), ...

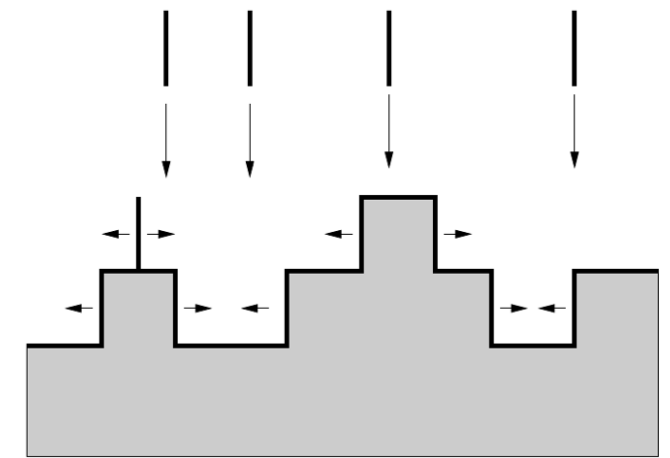
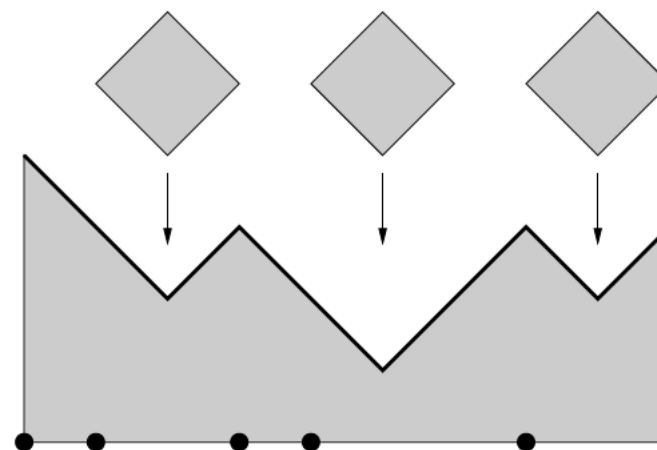


droplet IC

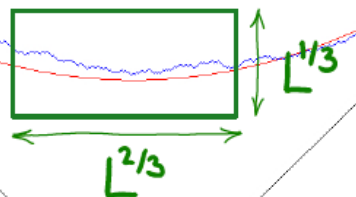
GUE

flat IC

GOE



time L



droplet:step initial data

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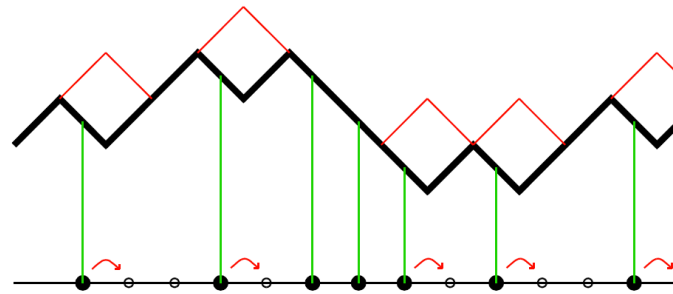
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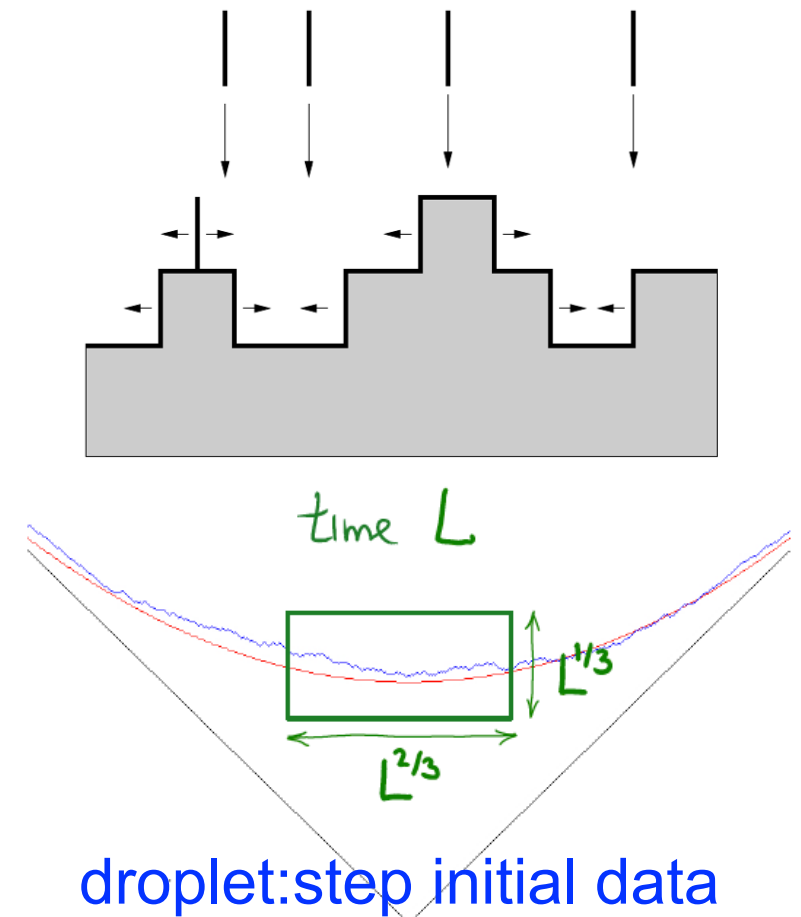
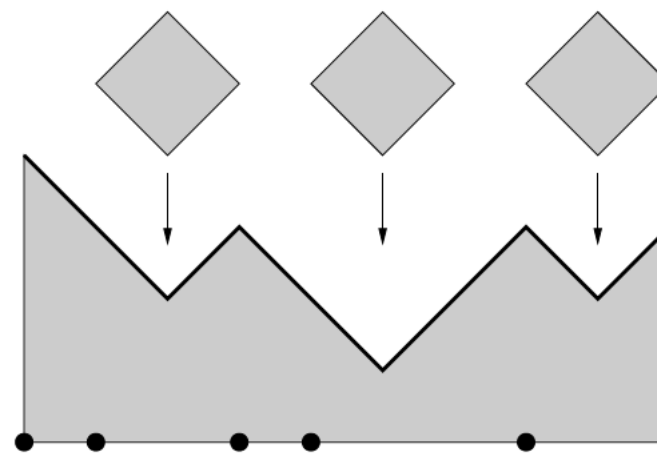


droplet IC

GUE

flat IC

GOE



- multi-space point correlations: Airy processes $A_2(y)$ $A_1(y)$ $\mathcal{A}_{\text{stat}}(\hat{x})$
GUE GOE BR

$$(\Gamma t)^{-\frac{1}{3}} (h_{\text{drop}}(x, t) - v_{\infty} t) \simeq \mathcal{A}_2(\hat{x}) - \hat{x}^2$$

$$\hat{x} = A \frac{x}{2t^{\frac{2}{3}}}$$

Airy2 process: stationary, reflection symmetric

(scaled centered) trajectory of largest eigenvalue in DysonBM

m-spacepoint joint CDF is m*m matrix Fredholm determinant (extended Airy kernel)

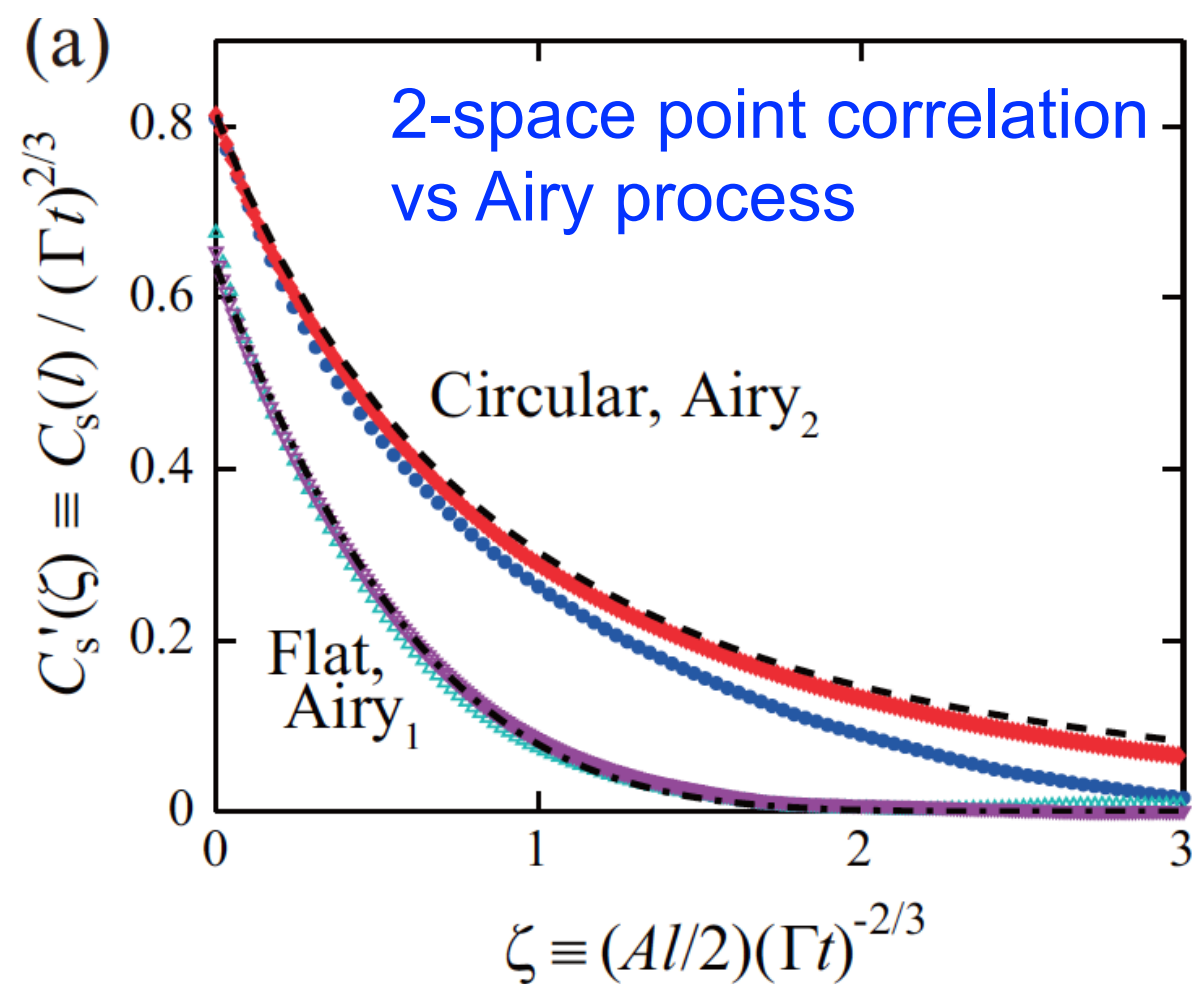
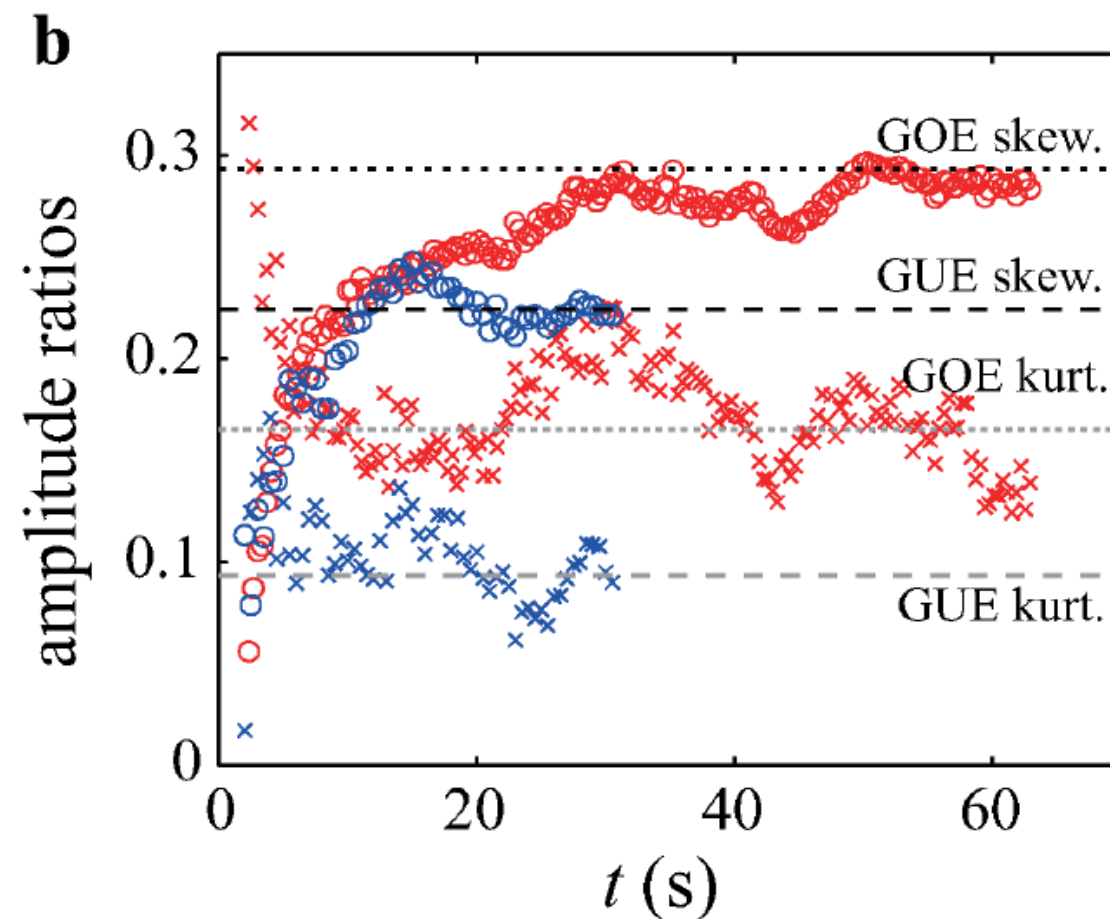
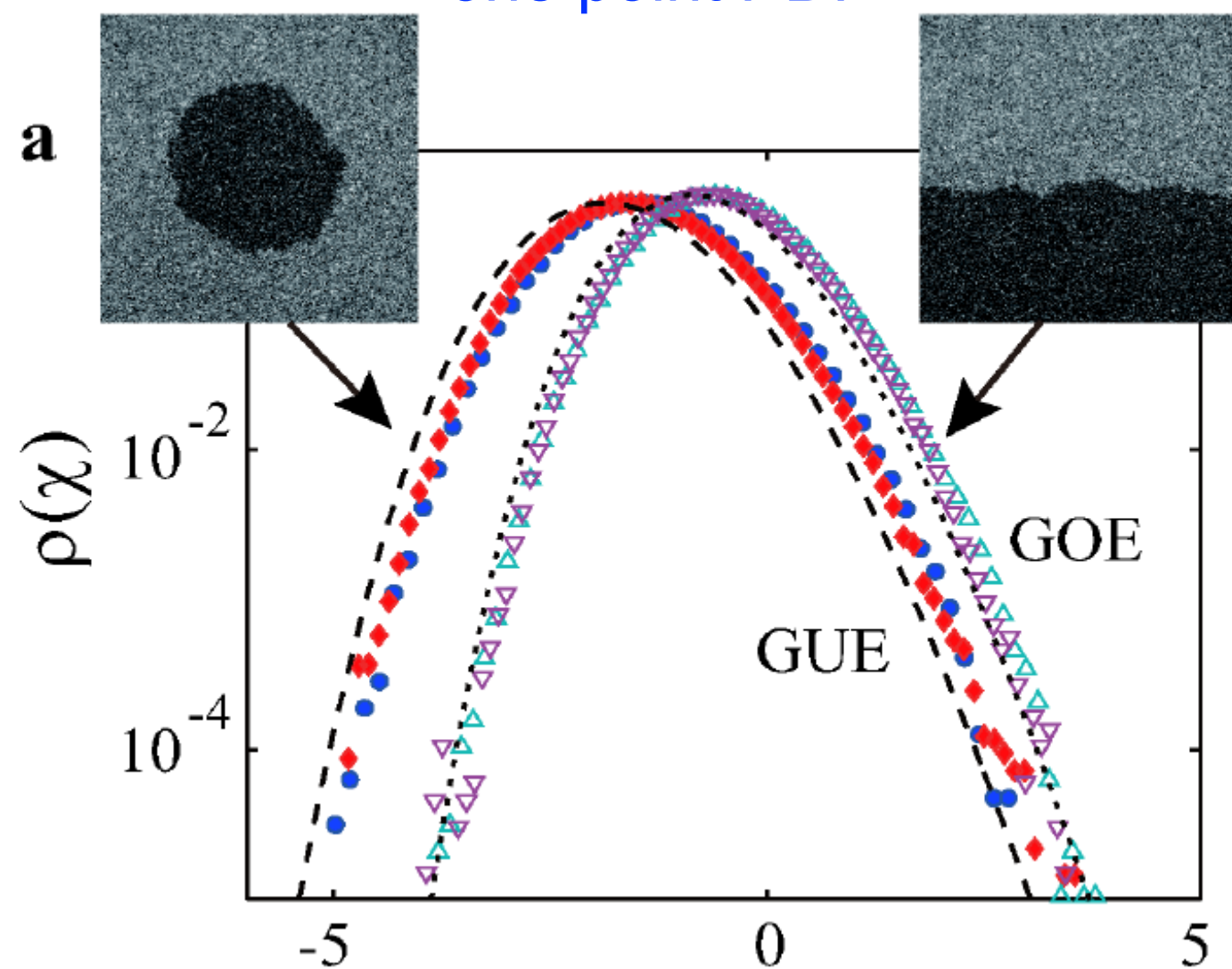
Takeuchi, Sano

skewness =

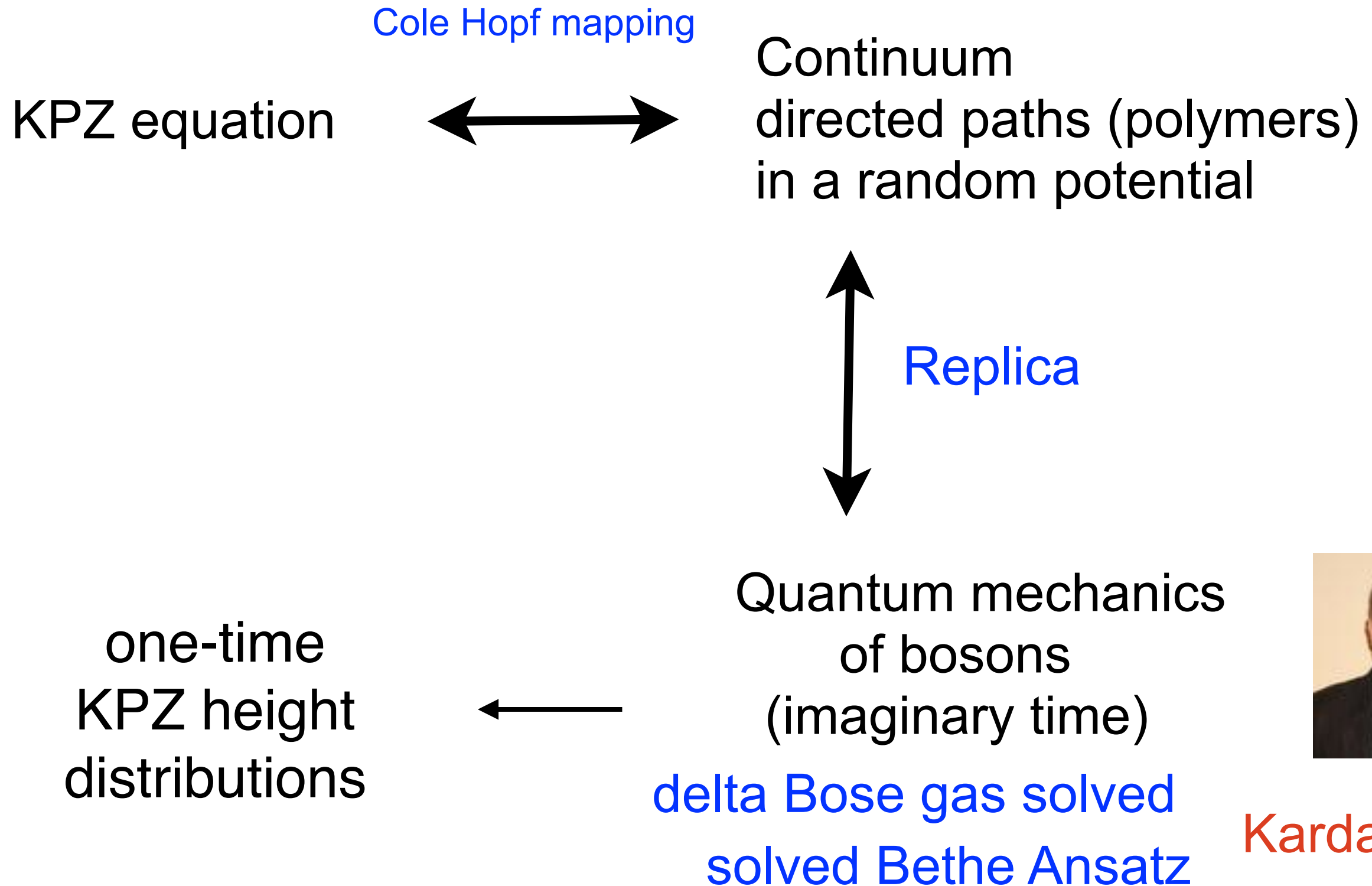
$$\frac{\langle (h - \langle h \rangle)^3 \rangle}{\langle (h - \langle h \rangle)^2 \rangle^{3/2}}$$

$$h \simeq v_\infty t + (\Gamma t)^{1/3} \chi,$$

one point PDF



part I

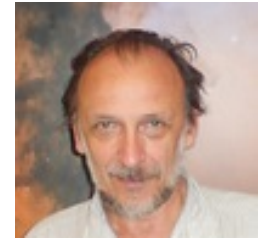


Kardar 87

- **Droplet** (Narrow wedge) KPZ/Continuum DP fixed endpoints

Replica Bethe Ansatz (RBA)

- P. Calabrese, P. Le Doussal, A. Rosso EPL 90 20002 (2010)
- V. Dotsenko, EPL 90 20003 (2010) J Stat Mech P07010



Weakly ASEP

- T Sasamoto and H. Spohn PRL 104 230602 (2010)
Nucl Phys B 834 523 (2010) J Stat Phys 140 209 (2010).
- G. Amir, I. Corwin, J. Quastel Comm. Pure. Appl. Math. 64 466 (2011)



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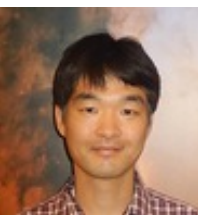
- **Flat** KPZ/Continuum DP one free endpoint

RBA: P. Calabrese, P. Le Doussal, PRL 106 250603 (2011) and J. Stat. Mech. P06001 (2012)

ASEP J. Ortmann, J. Quastel and D. Remenik, Comm. Pure App. Math. 70, 3 (2016), Annals of App. Prob. 16, 507 (2016)

- **Stationary** KPZ

RBA: T. Imamura, T. Sasamoto, Phys. Rev. Lett. 108, 190603 (2012) J. Stat. Phys. 150, 908-939 (2013).



Macdonald process: A. Borodin, I. Corwin, P. Ferrari and B. Veto, Math. Phys. Anal. Geom. 18(Art. 20), 95 (2015)



Cole Hopf mapping

KPZ equation: $\partial_t h = \nu \partial_x^2 h + \frac{\lambda_0}{2} (\partial_x h)^2 + \eta(x, t)$

define: $Z(x, t) = e^{\frac{\lambda_0}{2\nu} h(x, t)}$ $\lambda_0 h(x, t) = T \ln Z(x, t)$

it satisfies SHE: $\partial_t Z = \frac{T}{2} \partial_x^2 Z - \frac{V(x, t)}{T} Z$ $T = 2\nu$
 $\lambda_0 \eta(x, t) = -V(x, t)$

describes directed paths in random potential $V(x, t)$

Feynman Kac

$$Z(x, t | y, 0) =$$

$$\int_{x(0)=y}^{x(t)=x} Dx(\tau) e^{-\frac{1}{T} \int_0^t d\tau \frac{\kappa}{2} \left(\frac{dx(\tau)}{d\tau} \right)^2 + V(x(\tau), \tau)}$$

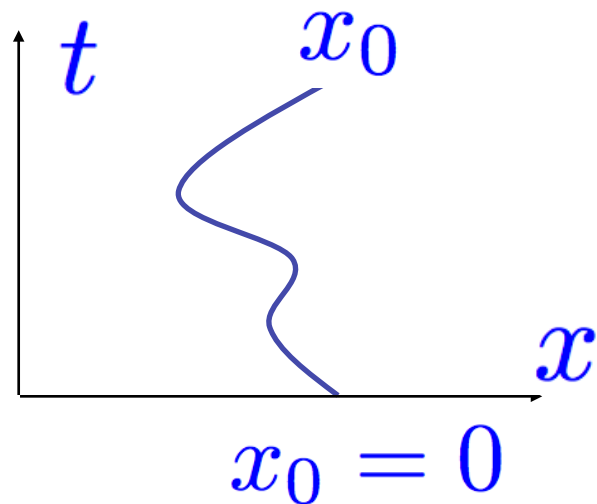
$$Z(x, y, t = 0) = \delta(x - y)$$

initial conditions

$$e^{\frac{\lambda_0}{2\nu} h(x,t)} = \int dy Z(x, t | y, 0) e^{\frac{\lambda_0}{2\nu} h(y, t=0)}$$

1) DP both fixed endpoints

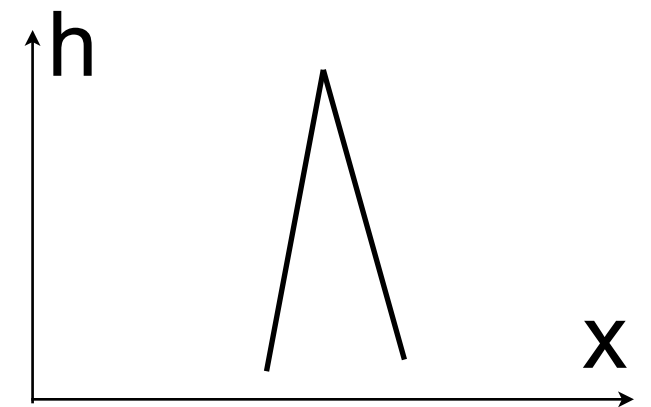
$$Z(x_0, t | x_0, 0)$$



KPZ: narrow wedge \Leftrightarrow droplet initial condition

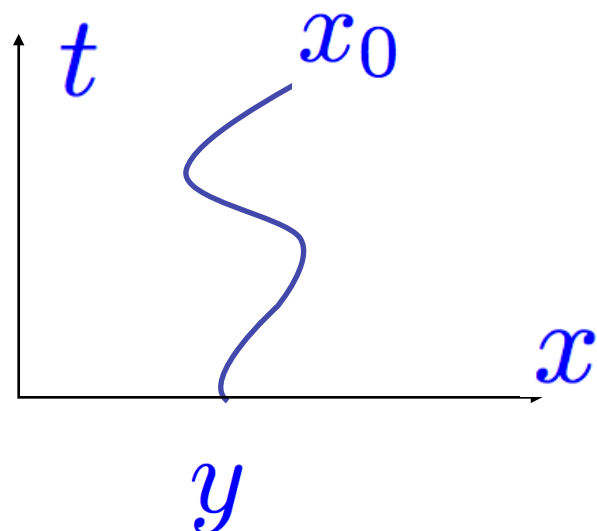
$$h(x, t = 0) = -w|x|$$

$$w \rightarrow \infty$$



2) DP one fixed one free endpoint

$$\int dy Z(x_0, t | y, 0)$$



KPZ: flat initial condition

$$h(x, t = 0) = 0$$

Schematically

$$Z = e^{\frac{\lambda_0 h}{2\nu}}$$

calculate $\overline{Z^n} = \int dZ Z^n P(Z) \quad n \in \mathbb{N}$

“guess” the probability distribution from its integer moments:

$$P(Z) \rightarrow P(\ln Z) \rightarrow P(h)$$

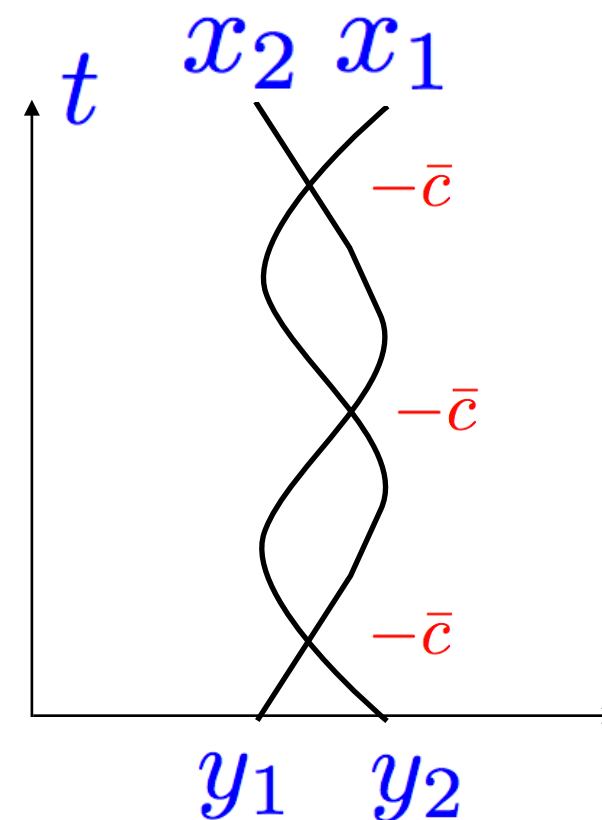
Quantum mechanics and Replica..

$$\mathcal{Z}_n := \overline{Z(x_1, t|y_1, 0) \dots Z(x_n, t|y_n, 0)} = \langle x_1, \dots, x_n | e^{-tH_n} | y_1, \dots, y_n \rangle$$

$$\partial_t \mathcal{Z}_n = -H_n \mathcal{Z}_n$$

$$x = T^3 \kappa^{-1} \tilde{x} \quad , \quad t = 2T^5 \kappa^{-1} \tilde{t}$$

drop the tilde..



$$H_n = - \sum_{i=1}^n \frac{\partial^2}{\partial x_i^2} - 2\bar{c} \sum_{1 \leq i < j \leq n} \delta(x_i - x_j)$$

Attractive Lieb-Liniger (LL) model (1963)

What do we need to solve KPZ with droplet initial condition ?

μ eigenstates

E_μ eigen-energies

fixed endpoint DP partition sum

$$\nearrow e^{-tH} = \sum_{\mu} |\mu\rangle e^{-E_\mu t} \langle \mu|$$

$$\overline{Z(x_0 t | x_0 0)^n} = \langle x_0 \dots x_0 | e^{-tH_n} | x_0, \dots x_0 \rangle$$

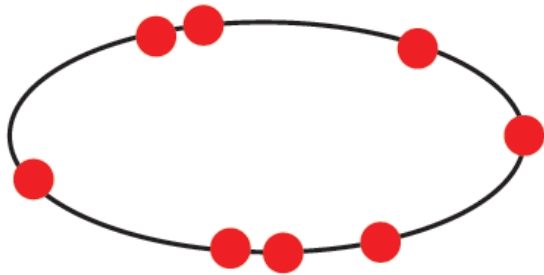
symmetric states = bosons

$$= \sum_{\mu} \Psi_{\mu}^*(x_0 \dots x_0) \Psi_{\mu}(x_0 \dots x_0) \frac{1}{||\mu||^2} e^{-E_{\mu} t}$$

we need:

- eigenfunctions at x_0, \dots, x_0
- norms of eigenstates
- energies of eigenstates
- perform summation over eigenstates

LL model: n bosons on a ring with local delta attraction



$$H_n = - \sum_{i=1}^n \frac{\partial^2}{\partial x_i^2} - 2\bar{c} \sum_{1 \leq i < j \leq n} \delta(x_i - x_j)$$

Bethe Ansatz:

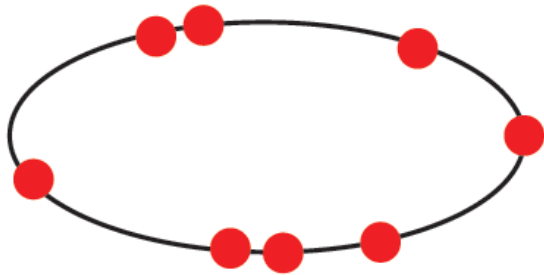
all (un-normalized) eigenstates are of the form (plane waves + sum over permutations)

$$\Psi_{\mu} = \sum_P A_P \prod_{j=1}^n e^{i\lambda_{P_j} x_j}$$

$$E_{\mu} = \sum_{j=1}^n \lambda_j^2 \quad A_P = \prod_{n \geq \ell > k \geq 1} \left(1 - \frac{ic \operatorname{sgn}(x_{\ell} - x_k)}{\lambda_{P_{\ell}} - \lambda_{P_k}} \right)$$

They are indexed by a set of rapidities $\lambda_1, \dots, \lambda_n$

LL model: n bosons on a ring with local delta attraction



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They are indexed by a set of rapidities $\lambda_1, \dots, \lambda_n$

which are determined by solving the N coupled Bethe equations (periodic BC)

$$e^{i\lambda_j L} = \prod_{\ell \neq j} \frac{\lambda_j - \lambda_{\ell} - i\bar{c}}{\lambda_j - \lambda_{\ell} + i\bar{c}}$$

n bosons+attraction => bound states

Bethe equations + large L => rapidities have imaginary parts

Derrida Brunet 2000

- ground state = a single bound state of n particles **Kardar 87**

$$\psi_0(x_1, \dots, x_n) \sim \exp\left(-\frac{\bar{c}}{2} \sum_{i < j} |x_i - x_j|\right) \quad E_0(n) = -\frac{\bar{c}^2}{12} n(n^2 - 1)$$

$$\overline{Z^n} = \overline{e^{n \ln Z}} \sim_{t \rightarrow \infty} e^{-t E_0(n)} \sim e^{\frac{\bar{c}^2}{12} n^3 t} \quad \text{exponent } 1/3$$

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can it be continued in n ? NO !

information about the LARGE DEVIATION tail
of the distribution of “free energy”

$$f = -\ln Z = -h$$

$$P(f) \sim_{f \rightarrow -\infty} \exp\left(-\frac{2}{3}(-f)^{3/2}\right)$$

PLD, S. Majumdar, G. Schehr, arXiv1606.08509

Large deviations for the height in 1D Kardar–Parisi–Zhang growth at late times

EPL 113, 60004 (2016)

n bosons+attraction => bound states

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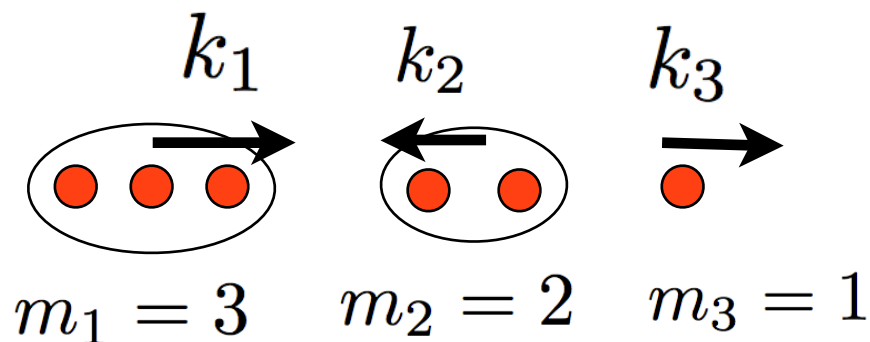
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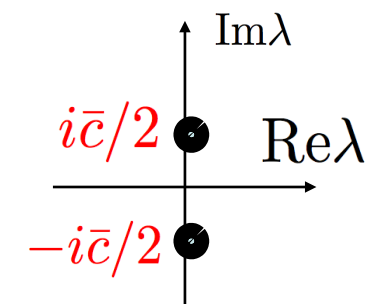
$$E_0(n) = -\frac{\bar{c}^2}{12}n(n^2 - 1)$$

need to sum over all eigenstates !

- all eigenstates are: All possible partitions of n into n_s "strings" each with m_j particles and momentum k_j



$$n = \sum_{j=1}^{n_s} m_j$$



$$\lambda_{j,a_j} = k_j + \frac{i\bar{c}}{2}(m_j + 1 - 2a_j) \quad \begin{matrix} a_j = 1, \dots, m_j \\ j = 1, \dots, n_s \end{matrix}$$

$$\Rightarrow E_\mu = \sum_{j=1}^{n_s} (m_j k_j^2 - \frac{\bar{c}^2}{12} m_j (m_j^2 - 1))$$

Integer moments of partition sum: fixed endpoints (droplet IC)

$$\overline{Z^n} = \sum_{\mu} \frac{|\Psi_{\mu}(0..0)|^2}{||\mu||^2} e^{-E_{\mu}t}$$

$$\Psi_{\mu}(0..0) = n!$$

norm of states: Calabrese-Caux (2007)

$$\overline{\hat{Z}^n} = \sum_{n_s=1}^n \frac{n!}{n_s! (2\pi\bar{c})^{n_s}} \sum_{(m_1, \dots, m_{n_s})_n} n = \sum_{j=1}^{n_s} m_j$$

$$\int \prod_{j=1}^{n_s} \frac{dk_j}{m_j} \Phi[k, m] \prod_{j=1}^{n_s} e^{m_j^3 \frac{\bar{c}^2 t}{12} - m_j k_j^2 t},$$

$$\Phi[k, m] = \prod_{1 \leq i < j \leq n_s} \frac{(k_i - k_j)^2 + (m_i - m_j)^2 c^2 / 4}{(k_i - k_j)^2 + (m_i + m_j)^2 c^2 / 4}$$

how to get $P(\ln Z)$ i.e. $P(h)$?

$$\ln Z = -\lambda f$$

$$\lambda = \left(\frac{\bar{c}^2}{4}t\right)^{1/3}$$

$$f = -\ln Z = -h \quad \text{random variable expected } O(1)$$

introduce generating function of moments $g(x)$:

$$g(x) = 1 + \sum_{n=1}^{\infty} \frac{(-e^{\lambda x})^n}{n!} \overline{Z^n} = \overline{\exp(-e^{\lambda(x-f)})}$$

so that at large time:

$$\lim_{\lambda \rightarrow \infty} g(x) = \overline{\theta(f-x)} = \text{Prob}(f > x)$$

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what we aim
to calculate=
Laplace transform of
 $P(Z)$

what we actually study

so that at large time:

$$\lim_{\lambda \rightarrow \infty} g(x) = \overline{\theta(f - x)} = \text{Prob}(f > x)$$

reorganize sum over number of strings

$$g(x) = 1 + \sum_{n_s=1}^{\infty} \frac{1}{n_s!} Z(n_s, x)$$

$$Z(n_s, x) = \sum_{m_1, \dots, m_{n_s}=1}^{\infty} \frac{(-1)^{\sum_j m_j}}{(4\pi\lambda^{3/2})^{n_s}}$$

$$\prod_{j=1}^{n_s} \int \frac{dk_j}{m_j} \prod_{1 \leq i < j \leq n_s} \frac{(k_i - k_j)^2 + (m_i - m_j)^2 \lambda^3}{(k_i - k_j)^2 + (m_i + m_j)^2 \lambda^3} \prod_{j=1}^{n_s} e^{\frac{1}{3} \lambda^3 m_j^3 - m_j k_j^2 + \lambda x m_j}$$



reorganize sum over number of strings

$$g(x) = 1 + \sum_{n_s=1}^{\infty} \frac{1}{n_s!} Z(n_s, x)$$

$$Z(n_s, x) = \sum_{m_1, \dots, m_{n_s}=1}^{\infty} \frac{(-1)^{\sum_j m_j}}{(4\pi\lambda^{3/2})^{n_s}} \int_{-\infty}^{\infty} dy Ai(y) e^{yw} = e^{w^3/3}$$

Airy trick

$$\prod_{j=1}^{n_s} \int \frac{dk_j}{m_j} \prod_{1 \leq i < j \leq n_s} \frac{(k_i - k_j)^2 + (m_i - m_j)^2 \lambda^3}{(k_i - k_j)^2 + (m_i + m_j)^2 \lambda^3} \prod_{j=1}^{n_s} e^{\frac{1}{3} \lambda^3 m_j^3 - m_j k_j^2 + \lambda x m_j}$$

double Cauchy formula

$$\det \left[\frac{1}{i(k_i - k_j) \lambda^{-3/2} + (m_i + m_j)} \right]$$

$$= \prod_{i < j} \frac{(k_i - k_j)^2 + (m_i - m_j)^2 \lambda^3}{(k_i - k_j)^2 + (m_i + m_j)^2 \lambda^3} \prod_{i=1}^{n_s} \frac{1}{2m_i}$$

$$\frac{1}{X} = \int_0^{\infty} dv e^{-vX}$$

Results: 1) $g(x)$ is a Fredholm determinant at any time t

$$Z(n_s, x) = \prod_{j=1}^{n_s} \int_{v_j > 0} dv_j \det[K(v_j, v_\ell)] \quad \lambda = \left(\frac{\bar{c}^2}{4}t\right)^{1/3}$$

$$K(v_1, v_2) = - \int \frac{dk}{2\pi} dy Ai(y + k^2 - x + v_1 + v_2) e^{-ik(v_1 - v_2)} \frac{e^{\lambda y}}{1 + e^{\lambda y}}$$

$$g(x) = 1 + \sum_{n_s=1}^{\infty} \frac{1}{n_s!} Z(n_s, x) = Det[I + K] \quad \text{by an equivalent definition of a Fredholm determinant}$$

$$K(v_1, v_2) \equiv \theta(v_1) K(v_1, v_2) \theta(v_2)$$

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$$K(v_1, v_2) \equiv \theta(v_1) K(v_1, v_2) \theta(v_2)$$

$$\text{2) large time limit} \quad \lambda = +\infty \quad \frac{e^{\lambda y}}{1 + e^{\lambda y}} \longrightarrow \theta(y)$$

Airy function identity

$$\int dk Ai(k^2 + v + v') e^{ik(v - v')} = 2^{2/3} \pi Ai(2^{1/3}v) Ai(2^{1/3}v')$$

$$g(x) = \text{Prob}(f > x = -2^{2/3}s) = \text{Det}(1 - P_s K_{Ai} P_s) = F_2(s)$$

$$K_{Ai}(v, v') = \int_{y>0} Ai(v + y) Ai(v' + y) \quad \text{GUE-Tracy-Widom distribution}$$

Summary: one-time observables for models in KPZ class and a tale of tails

droplet class initial condition, large time t

$$h(x=0, t) = -\frac{t}{12} + t^{1/3}\chi_2 + o(t^{1/3})$$

$$\text{Prob}(\chi_2 < \sigma) = F_2(\sigma)$$

GUE-TW distribution

$$F_2(\sigma) = \text{Det}[I - P_\sigma K_{\text{Ai}} P_\sigma]$$

$$K_{\text{Ai}}(v, v') = \int_0^{+\infty} dy \text{Ai}(y+v) \text{Ai}(y+v')$$

Tail ?

GUE-Tracy Widom distribution

$$F_2(\sigma) = \text{Det}[I - P_\sigma K_{\text{Ai}} P_\sigma]$$

$$F_2'(\sigma_1) \sim \frac{e^{-\frac{4}{3}\sigma_1^{3/2}}}{8\pi\sigma_1} \quad \sigma_1 \gg 1$$

GUE-Tracy Widom distribution

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Tail approximant: $F_2(\sigma) = F_2^{(1)}(\sigma) + O(e^{-\frac{8}{3}\sigma^{3/2}}) \quad \sigma \rightarrow +\infty$
neglects terms of order $O(K_{\text{Ai}}^2)$

$$F_2^{(1)}(\sigma) \equiv 1 - \text{Tr}[P_\sigma K_{\text{Ai}}] = 1 - \int_\sigma^{+\infty} dv K_{\text{Ai}}(v, v)$$

$$F_2^{(1)}(\sigma) - 1 = O(e^{-\frac{4}{3}\sigma^{3/2}})$$

GUE-Tracy Widom distribution

$$F_2(\sigma) = \text{Det}[I - P_\sigma K_{\text{Ai}} P_\sigma] \quad F'_2(\sigma_1) \sim \frac{e^{-\frac{4}{3}\sigma_1^{3/2}}}{8\pi\sigma_1} \quad \sigma_1 \gg 1$$

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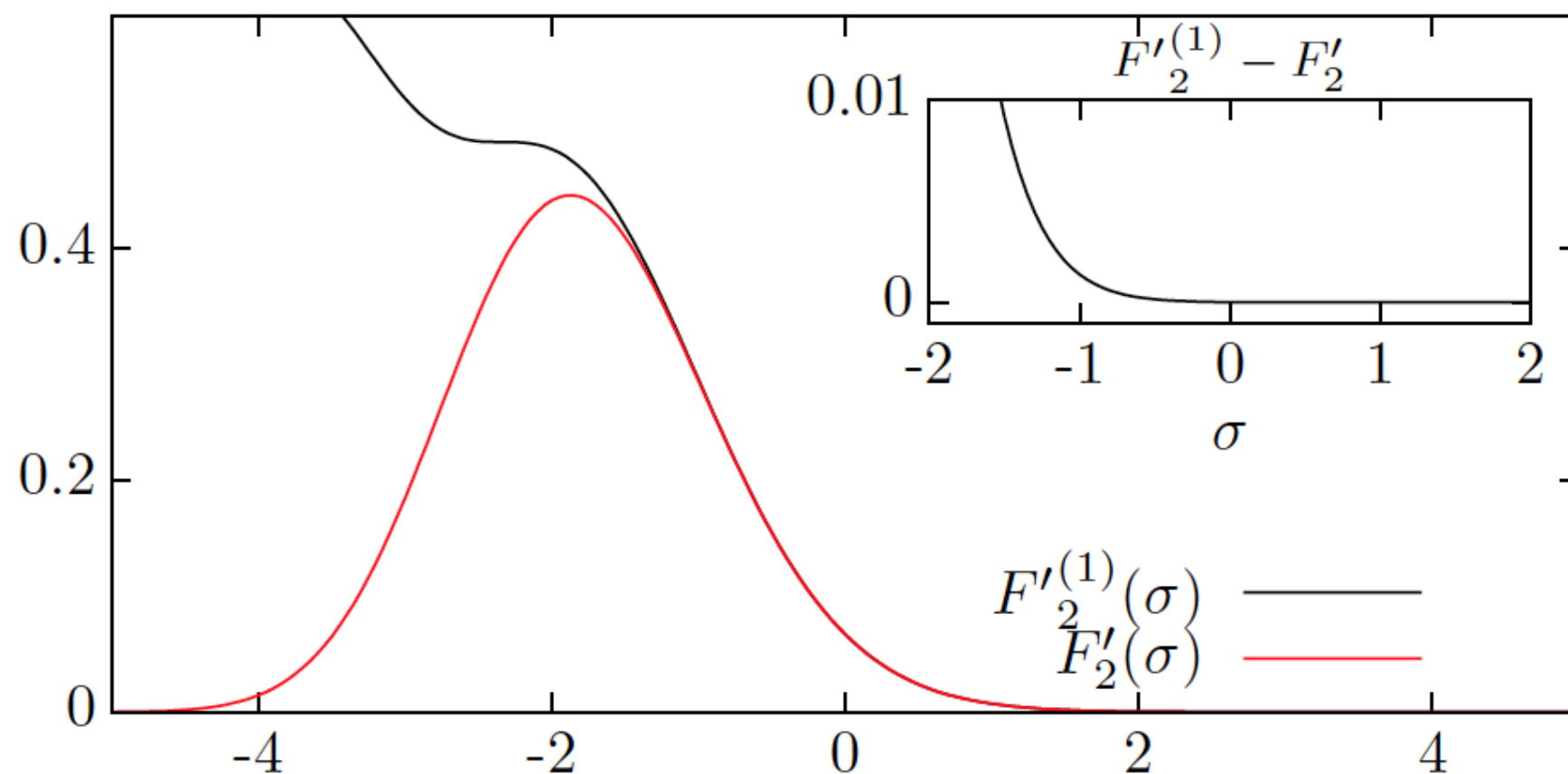


Figure 1. Plot of the Tracy-Widom distribution $F'_2(\sigma)$ (red line) compared with its tail for positive σ given by $F_2^{(1)}(\sigma)$ (black line). Inset: difference between the Tracy-Widom distribution $F'_2(\sigma)$ and its tail $F_2^{(1)}(\sigma)$, given in (12).

why is this tail approximant interesting ?

$$F_2^{(1)}(\sigma) \equiv 1 - \text{Tr}[P_\sigma K_{\text{Ai}}] = 1 - \int_\sigma^{+\infty} dv K_{\text{Ai}}(v, v)$$

it corresponds to keeping only contributions of **one n-string** when calculating generating function $ns=1$

\Leftrightarrow n particles all in a single bound state = the ground state of the Lieb Liniger model

neglects terms of order $O(K_{\text{Ai}}^2)$ contributions of two mj-strings, ..

**\Rightarrow assume this property holds
for more complicated observables**

Part II: two-time KPZ via RBA

II a) direct approach

with Jacopo de Nardis (ENS)

any ratio of times BUT only partial tail

Jacopo de Nardis, PLD, [arXiv:1612.08695](https://arxiv.org/abs/1612.08695) J. Stat. Mech. (2017) 053212

Tail of the two-time height distribution for KPZ growth in one dimension

two-time problem: - KPZ equation w. droplet initial conditions
- 2 directed polymers with fixed endpoints

in same random potential

$$H_1 \equiv h(0, t_1) = \ln Z_\eta(0, t_1 | 0, 0)_{X=0}$$

$$H_2 \equiv h(X, t_2) = \ln Z_\eta(X, t_2 | 0, 0)$$

two-time problem: - KPZ equation w. droplet initial conditions
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in same random potential

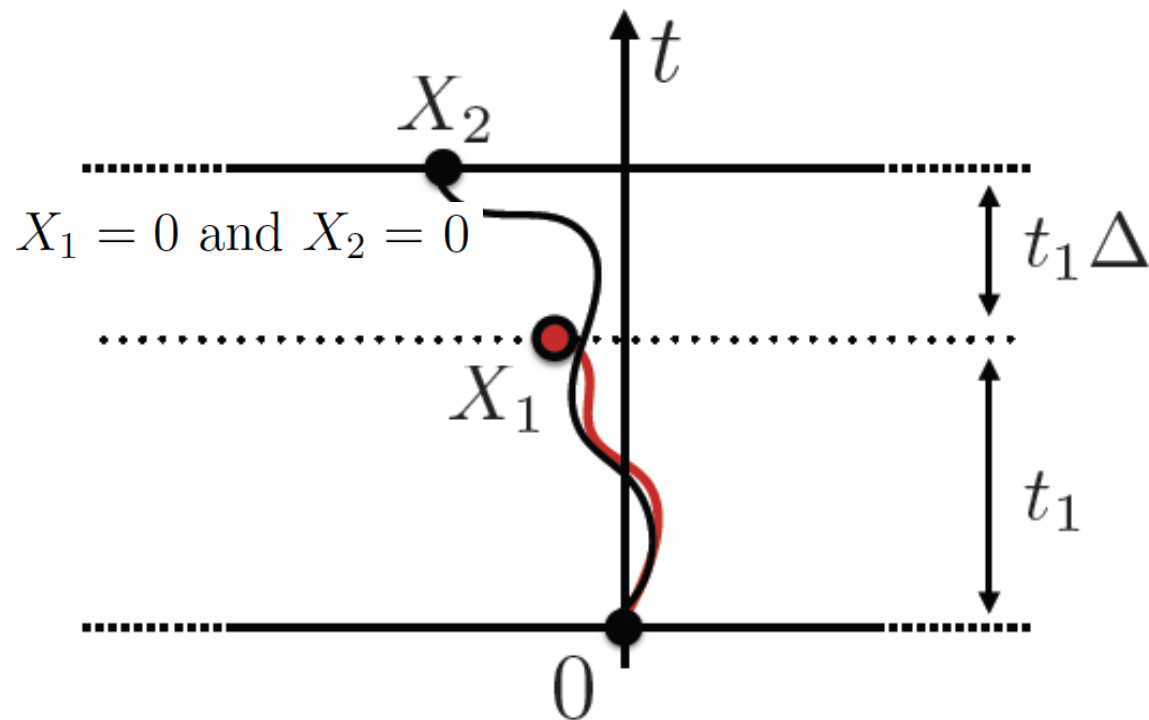
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$$H_2 \equiv h(X, t_2) = \ln Z_\eta(X, t_2 | 0, 0)$$

limit: $t_1 \rightarrow +\infty$

$t_2 \rightarrow +\infty$

$$\Delta = \frac{t_2 - t_1}{t_1} \quad \text{fixed}$$



two-time problem: - KPZ equation w. droplet initial conditions
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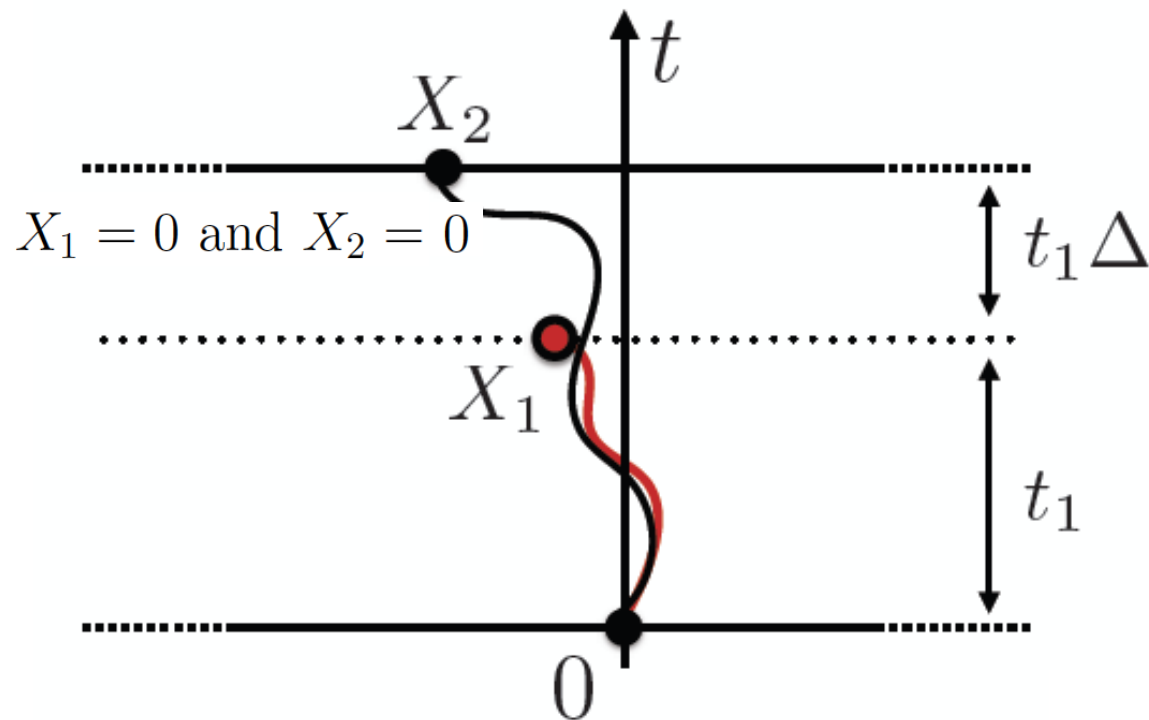
define rescaled heights h_1, h

$$H_1 = t_1^{1/3} h_1$$

$$H_2 - H_1 = (t_2 - t_1)^{1/3} h = \Delta^{1/3} t_1^{1/3} h$$

JPDF of rescaled heights

$$P_\Delta(\sigma_1, \sigma) = \lim_{t_1 \rightarrow +\infty} \overline{\delta(h_1 - \sigma_1) \delta(h - \sigma)}$$



Other works on two times

- **Victor Dotsenko** V. Dotsenko, arXiv:1304.0626, J. Stat. Mech. (2013) P06017.
V. Dotsenko, arXiv:1507.06135
V. Dotsenko, arXiv:1603.08945, J.Phys. A: Math. Theor. 49 (2016) 27 LT01

claim for the JPDF from continuum KPZ/DP model (droplet IC)

we believe: incorrect, many terms missing

- **Kurt Johansson, arXiv:1502.00941**

TWO TIME DISTRIBUTION IN BROWNIAN DIRECTED PERCOLATION

ABSTRACT. In the zero temperature Brownian semi-discrete directed polymer we study the joint distribution of two last-passage times at positions ordered in the time-like direction. This is the situation when we have the slow de-correlation phenomenon. We compute the limiting joint distribution function in a scaling limit. This limiting distribution is given by an expansion in determinants which is not a Fredholm expansion. A somewhat similar looking formula was derived non-rigorously in a related model by Dotsenko.



exact formula for the JPDF from semi-discrete DP model (droplet IC)

but VERY complicated formula!

- **Ivan Corwin and Alan Hammond (in progress)** asymptotics
- **Patrick Ferrari and Herbert Spohn, arXiv:1602.00486**
2-time second cumulant: asymptotics for droplet and flat, exact for stationary

- **J. Baik, Z. Liu, multi-time TASEP ring (talk)**



Main result: tail approximant of the joint PDF

$$t_1 \rightarrow +\infty$$

$$t_2 \rightarrow +\infty$$

$$P_{\Delta}(\sigma_1, \sigma) = \lim_{t_1 \rightarrow +\infty} \overline{\delta(h_1 - \sigma_1) \delta(h - \sigma)} \quad H_1 = t_1^{1/3} h_1 \quad H_2 - H_1 = (t_2 - t_1)^{1/3} h$$

$$P_{\Delta}(\sigma_1, \sigma) = P_{\Delta}^{(1)}(\sigma_1, \sigma) + O(e^{-\frac{8}{3}\sigma_1^{3/2}}) \quad \Delta = \frac{t_2 - t_1}{t_1}$$



$$O(e^{-\frac{4}{3}\sigma_1^{3/2}}) \text{ uniformly in } \sigma$$

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$$\downarrow$$

$$O(e^{-\frac{4}{3}\sigma_1^{3/2}}) \text{ uniformly in } \sigma$$

$$P_{\Delta}^{(1)}(\sigma_1, \sigma) = \left(\partial_{\sigma_1} \partial_{\sigma} - \Delta^{-1/3} \partial_{\sigma}^2 \right) f_{\Delta}(\sigma_1, \sigma)$$

$$f_{\Delta}(\sigma_1, \sigma) = \Delta^{1/3} F_2(\sigma) \text{Tr} \left[P_{\sigma} K_{\sigma_1}^{(4)} P_{\sigma} (I - P_{\sigma} K_{\text{Ai}} P_{\sigma})^{-1} \right] - F_2(\sigma) \text{Tr}[P_{\sigma_1} K_{\text{Ai}}]$$

$$K_{\sigma_1}^{(4)}(u, v) = \int_0^{\infty} dy_1 dy_2 \text{Ai}(-y_1 + u) K_{\text{Ai}}(y_1 \Delta^{1/3} + \sigma_1, y_2 \Delta^{1/3} + \sigma_1) \text{Ai}(-y_2 + v)$$

Limits of the JPDF

$$\Delta = \frac{t_2 - t_1}{t_1} > 0$$

From the formula we show

very separated times $\lim_{\Delta \rightarrow \infty} P_{\Delta}^{(1)}(\sigma_1, \sigma) = F_2^{(1)'}(\sigma_1) F_2'(\sigma)$
 $t_2/t_1 \rightarrow +\infty$

H1, H2 are two
independent GUE-TW

close times $\lim_{\Delta \rightarrow 0} P_{\Delta}^{(1)}(\sigma_1, \sigma) = F_2^{(1)'}(\sigma_1) F_0'(\sigma)$

two independent variables:
H1 is GUE-TW, H2-H1 is Baik-Rains
stationary KPZ

we obtain corrections

we argue exact (Airy processes)

$$\lim_{\Delta \rightarrow \infty} P_{\Delta, \infty}(\sigma_1, \sigma) = F_2'(\sigma_1) F_2'(\sigma)$$

$$\lim_{\Delta \rightarrow 0} P_{\Delta, \infty}(\sigma_1, \sigma) = F_2'(\sigma_1) F_0'(\sigma)$$

Crossover from Baik-Rains to GUE-Tracy Widom

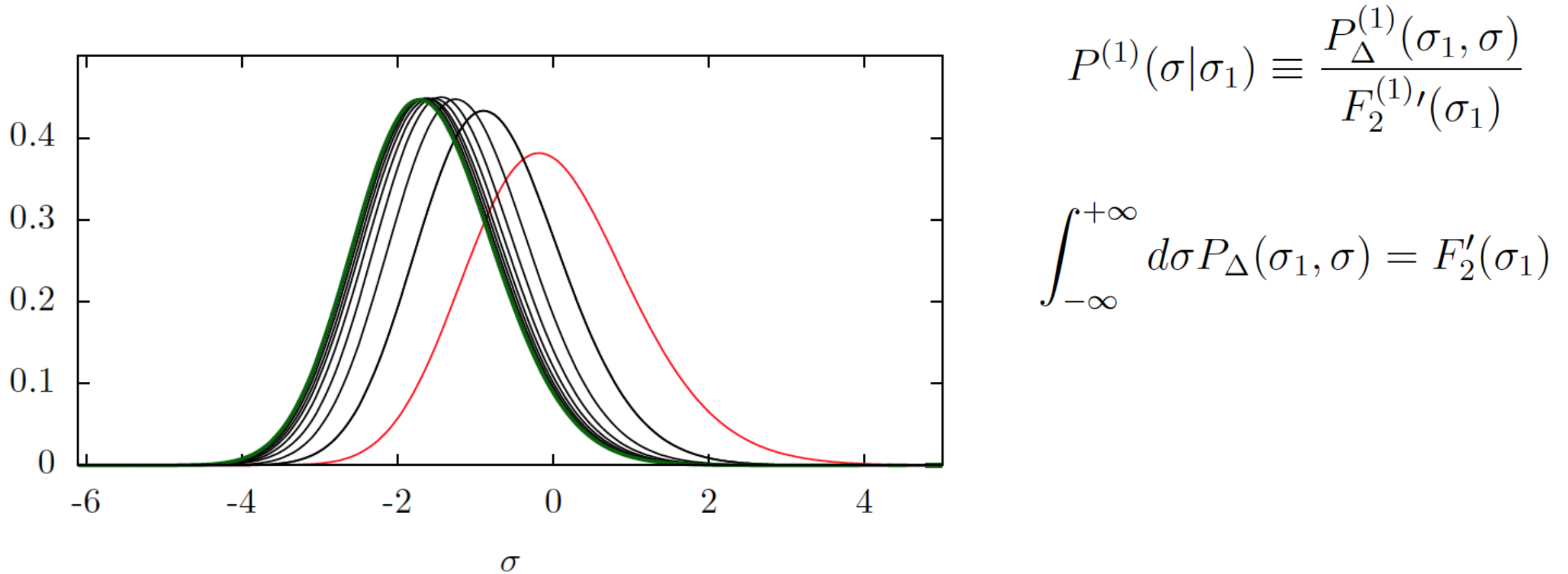
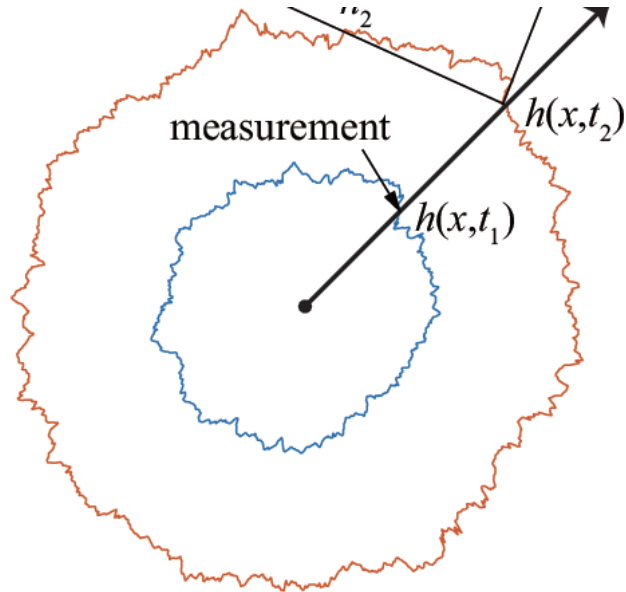


Figure 3. Plot of the conditional probability distribution $P^{(1)}(\sigma|\sigma_1)$, defined in (38), of the scaled height difference $h \equiv (H_2 - H_1)/t_1^{1/3} = \sigma$ for a fixed value of the height at the earlier time $h_1 \equiv H_1/t_1^{1/3} = \sigma_1 = 0$, as a function of σ . The various curves correspond to increasing values of $\Delta^{1/3} = (0.7k)$ with $k = 0, \dots, 10$ (from right to left). The functions interpolate between the $\Delta = 0$ point (red line) which coincides with the Baik-Rains probability distribution $F_0'(\sigma)$, and the $\Delta \rightarrow \infty$ (green line) which corresponds to the GUE Tracy-Widom probability distribution $F_2'(\sigma)$.



Memory and universality in interface growth

Jacopo De Nardis,^{1,*} Pierre Le Doussal,^{2,†} and Kazumasa A. Takeuchi^{3,‡}

arXiv1611.04756

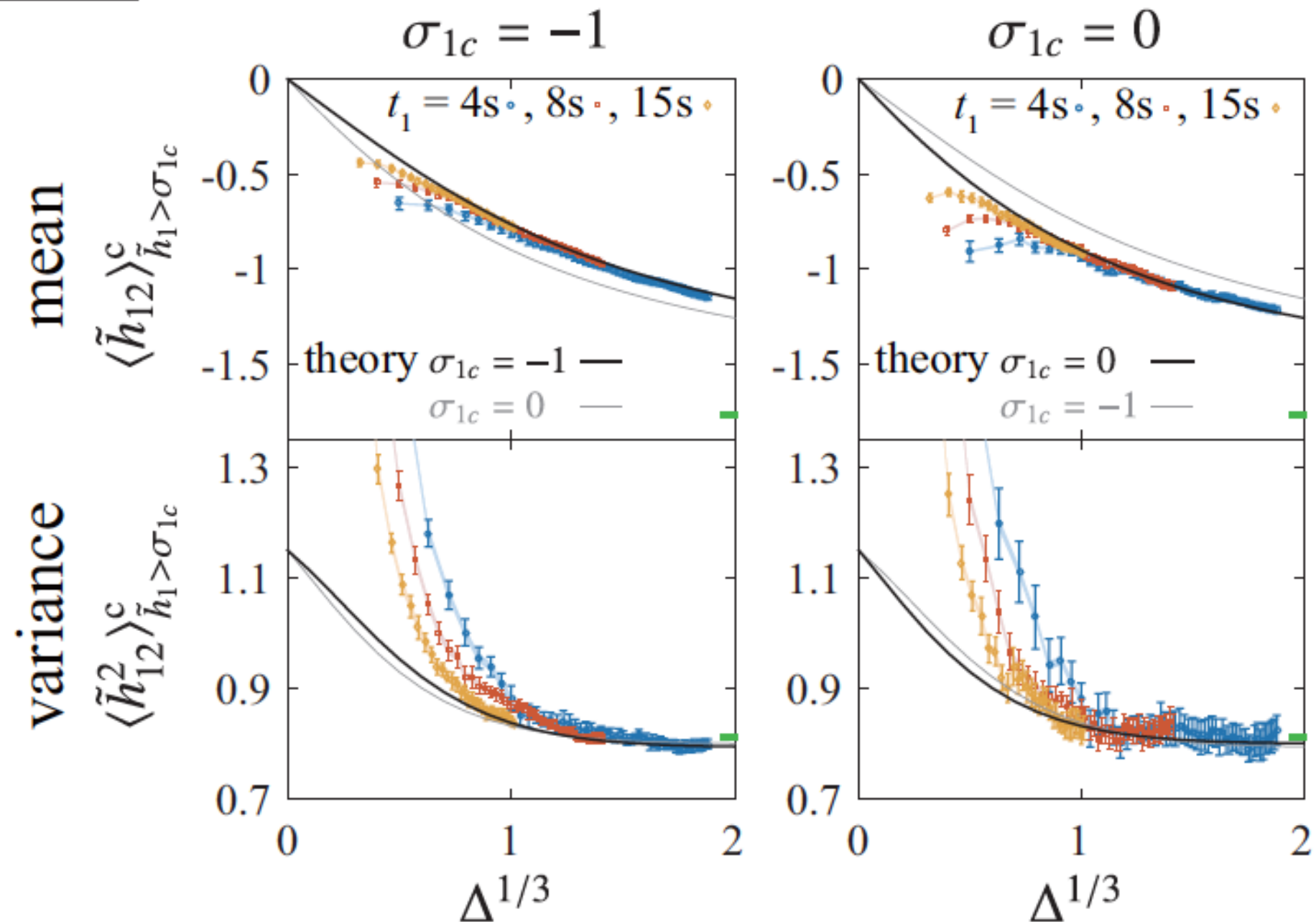
Phys. Rev. Lett. 118, 125701 (2017)

liquid-crystal experiment

$$\tilde{h}_{12} := \frac{h(0, t_2) - h(0, t_1) - v_\infty t_1 \Delta}{(\Gamma t_1 \Delta)^{1/3}}$$

scaled height difference

moments of height difference
conditioned to value of height
at the earlier time



want to calculate joint moments

$$Z_{n_1, n_2}(t_1, t_2) = \overline{Z_1^{n_1} Z_2^{n_2}} = \overline{Z_\eta(X_1, t_1 | 0, 0)^{n_1} Z_\eta(X_2, t_2 | 0, 0)^{n_2}}$$

want to calculate joint moments

$$Z_{n_1, n_2}(t_1, t_2) = \overline{Z_1^{n_1} Z_2^{n_2}} = \overline{Z_\eta(X_1, t_1 | 0, 0)^{n_1} Z_\eta(X_2, t_2 | 0, 0)^{n_2}}$$

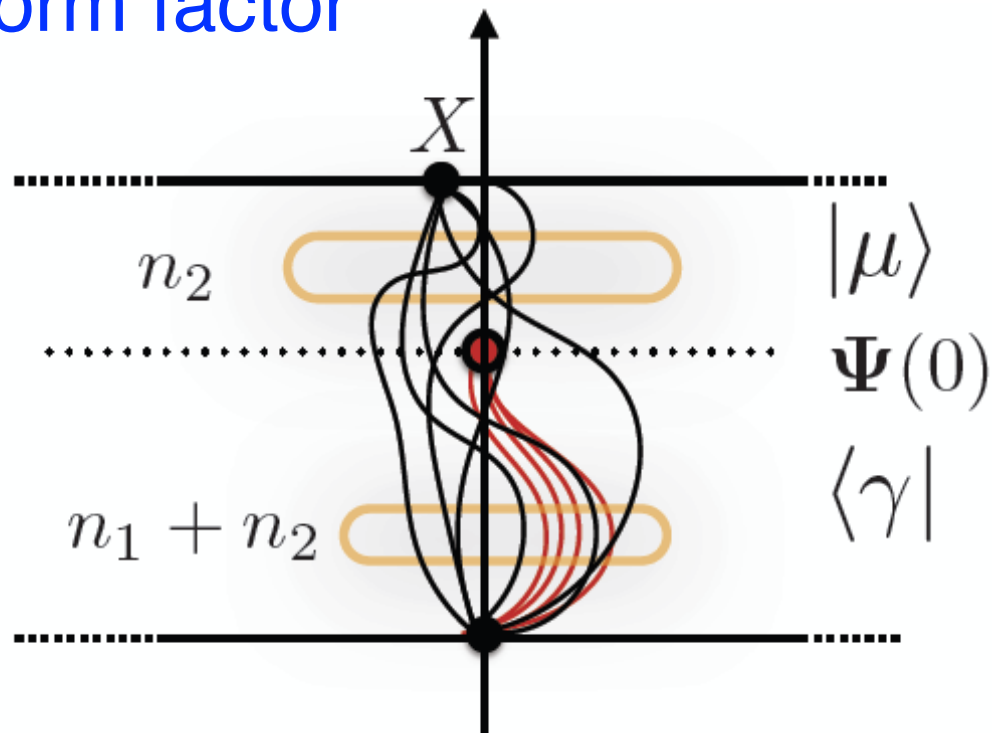
quantum mechanics (exact)

$$e^{-tH_n} = \sum_{\mu \in \Lambda_n} |\mu\rangle e^{-tE_\mu} \langle \mu|$$

$$Z_{n_1, n_2}(t_1, t_2) = \sum_{\mu \in \Lambda_{n_2}, \gamma \in \Lambda_{n_1+n_2}} \frac{\psi_\gamma^*(0, \dots, 0) \psi_\mu(X_2, \dots, X_2)}{||\gamma||^2 ||\mu||^2} e^{-\Delta t_1 E_\mu - t_1 E_\gamma} F_{\mu; \gamma}^{n_2; n_1+n_2}$$

$$F_{\mu; \gamma}^{n_2; n_1+n_2} \equiv \prod_{\alpha=1}^{n_2} \int dy_\alpha e^{-w|y_j - X_1|} \psi_\mu^*(y_1, \dots, y_{n_2}) \psi_\gamma(\underbrace{X_1, \dots, X_1}_{n_1}, y_1, \dots, y_{n_2})$$

form factor



$$n_1, n_2 \geq 0$$

want to calculate joint moments

$$Z_{n_1, n_2}(t_1, t_2) = \overline{Z_1^{n_1} Z_2^{n_2}} = \overline{Z_\eta(X_1, t_1|0, 0)^{n_1} Z_\eta(X_2, t_2|0, 0)^{n_2}}$$

quantum mechanics (exact)

$$e^{-tH_n} = \sum_{\mu \in \Lambda_n} |\mu\rangle e^{-tE_\mu} \langle \mu|$$

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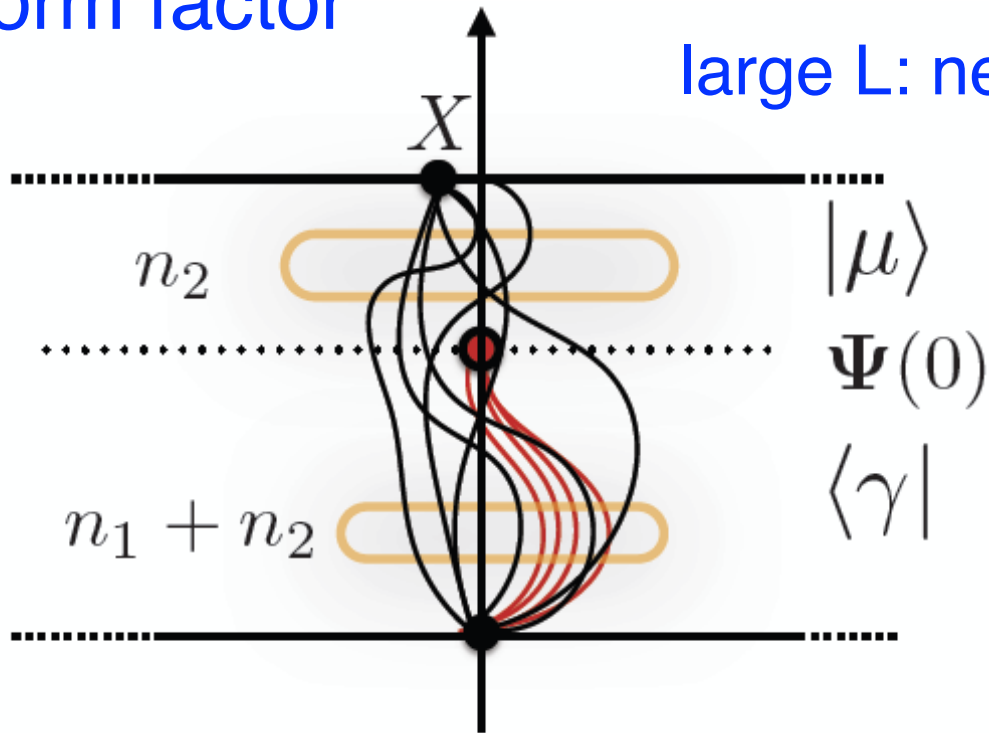
$$F_{\mu;\gamma}^{n_2;n_1+n_2} \equiv \prod_{\alpha=1}^{n_2} \int dy_{\alpha} e^{-w|y_j-X_1|} \psi_{\mu}^*(y_1, \dots, y_{n_2}) \psi_{\gamma}(\underbrace{X_1, \dots, X_1}_{n_1}, y_1, \dots, y_{n_2})$$

form factor

large L: need string form factors

$$F_{\mu;\gamma}^{n_2;n_1+n_2} = F_{\mathbf{p},\mathbf{m}^\mu;\mathbf{q},\mathbf{m}^\gamma}^{n_2;n_1+n_2}$$

difficult!



$$n_1, n_2 \geq 0$$

want to calculate joint moments

$$Z_{n_1, n_2}(t_1, t_2) = \overline{Z_1^{n_1} Z_2^{n_2}} = \overline{Z_\eta(X_1, t_1 | 0, 0)^{n_1} Z_\eta(X_2, t_2 | 0, 0)^{n_2}}$$

quantum mechanics (exact)

$$e^{-tH_n} = \sum_{\mu \in \Lambda_n} |\mu\rangle e^{-tE_\mu} \langle \mu|$$

$$Z_{n_1, n_2}(t_1, t_2) = \sum_{\mu \in \Lambda_{n_2}, \gamma \in \Lambda_{n_1+n_2}} \frac{\psi_\gamma^*(0, \dots, 0) \psi_\mu(X_2, \dots, X_2)}{||\gamma||^2 ||\mu||^2} e^{-\Delta t_1 E_\mu - t_1 E_\gamma} F_{\mu; \gamma}^{n_2; n_1+n_2}$$

$$F_{\mu; \gamma}^{n_2; n_1+n_2} \equiv \prod_{\alpha=1}^{n_2} \int dy_\alpha e^{-w|y_j - X_1|} \psi_\mu^*(y_1, \dots, y_{n_2}) \psi_\gamma(\underbrace{X_1, \dots, X_1}_{n_1}, y_1, \dots, y_{n_2})$$

form factor

large L: need string form factors

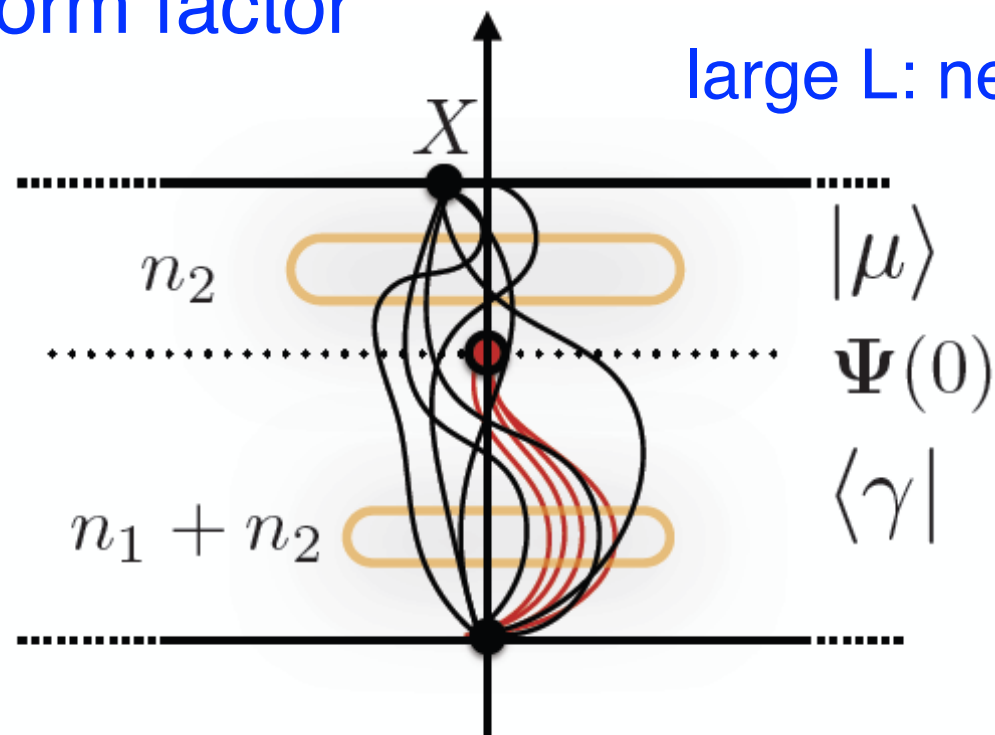
$$F_{\mu; \gamma}^{n_2; n_1+n_2} = F_{\mathbf{p}, \mathbf{m}^\mu; \mathbf{q}, \mathbf{m}^\gamma}^{n_2; n_1+n_2}$$

difficult!

in the sum keep only:

$|\mu\rangle = |\mathbf{p}, \mathbf{m}\rangle$ arbitrary string state with any number of strings

$|\gamma\rangle = |q, n_1 + n_2\rangle$ ONE n_1+n_2 -string state



$n_1, n_2 \geq 0 \Rightarrow$ 1) form factor simplifies, possible to carry the sum exactly !

\Rightarrow 2) gives the TAIL of JPDF for large positive h_1 , arbitrary h

Part II: two-time KPZ via RBA

II b) using Airy process

no restriction on $h(0, t_1)$ $h(0, t_2)$ BUT only at infinite time separation

$$\Delta \rightarrow +\infty$$

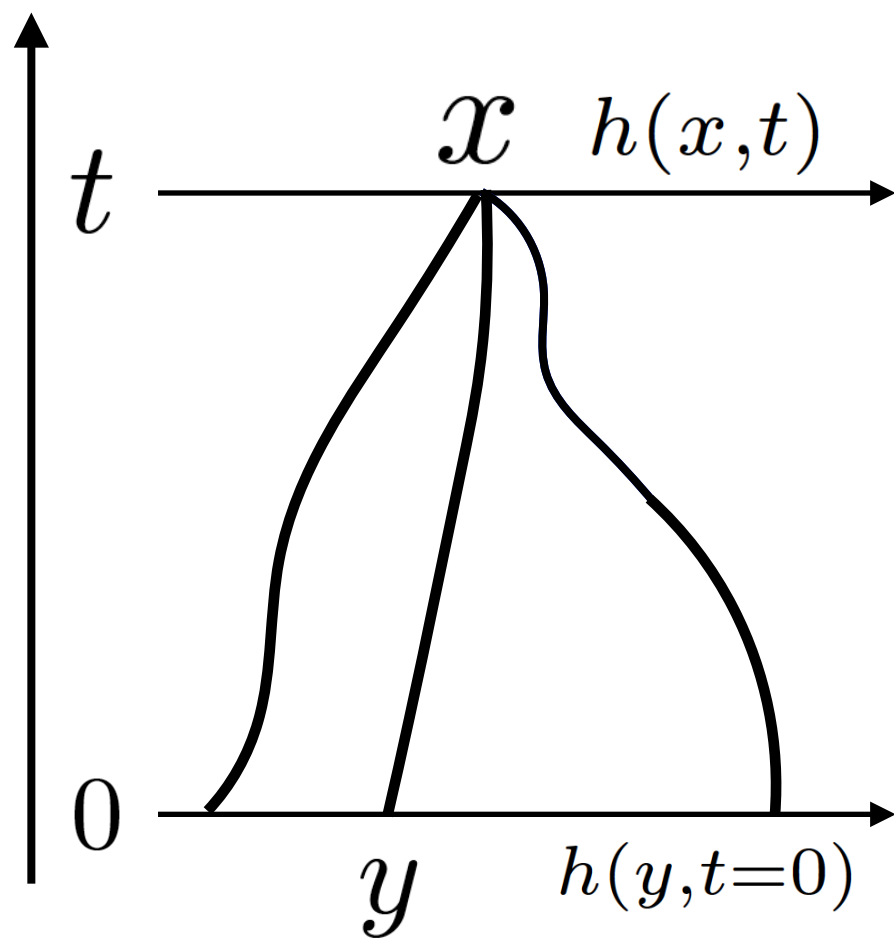
[PLD arXiv:1709.06264 \(2017\)](#)

Maximum of an Airy process plus Brownian motion and memory in KPZ growth

Large time limit of KPZ height for general initial condition



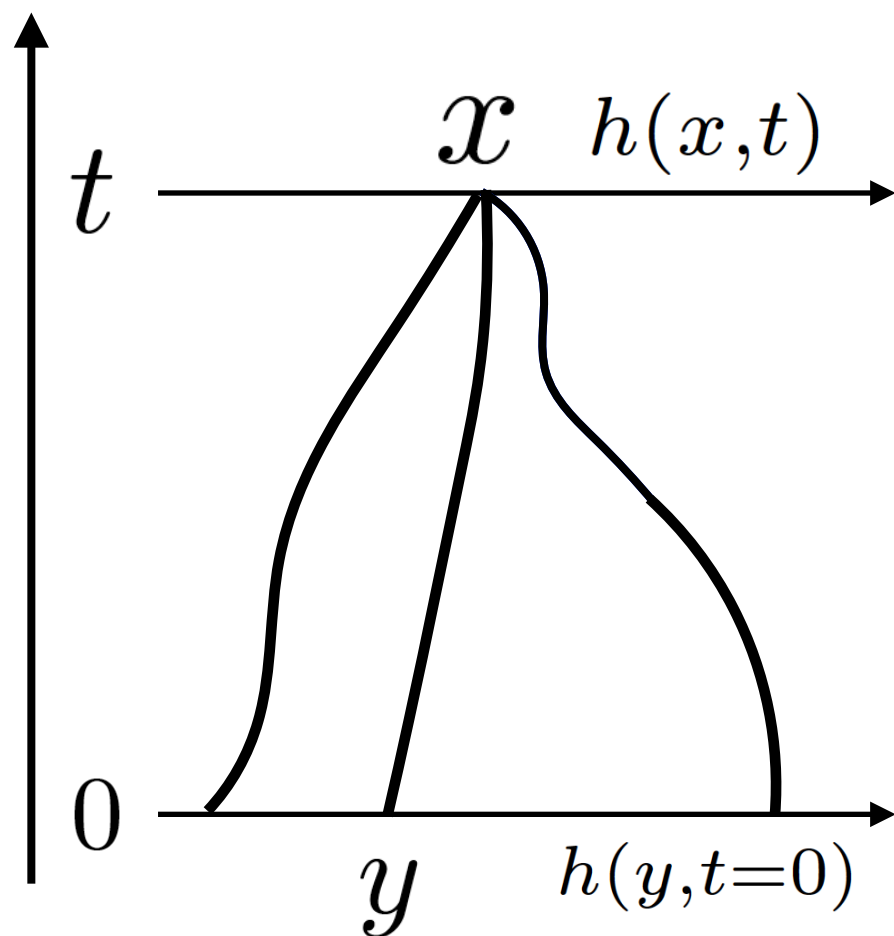
Variational problem using Airy process



general initial condition

$$e^{h(x,t)} = \text{sum over paths} \\ \int dy Z_\eta(x, t | y, 0) e^{h(y, t=0)}$$

general initial condition



$$e^{h(x,t)} = \text{sum over paths}$$

$$\int dy Z_\eta(x, t | y, 0) e^{h(y, t=0)}$$

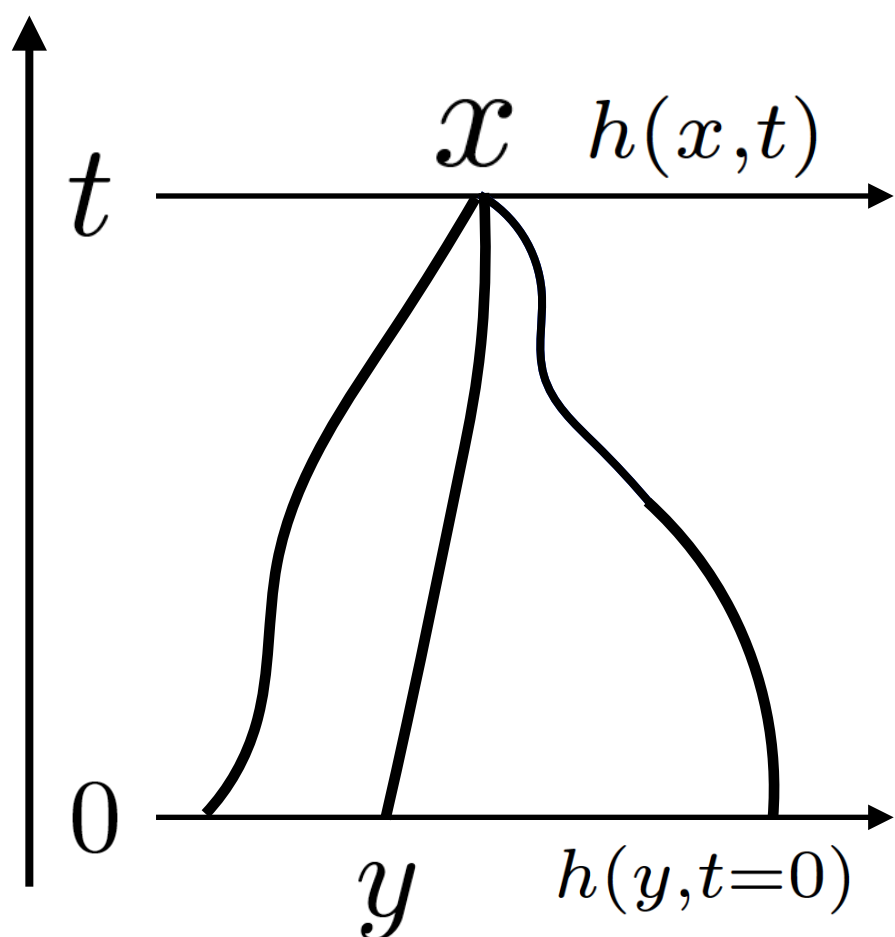
\downarrow \downarrow choose

$$e^{t^{1/3}(\mathcal{A}_2(\hat{x}-\hat{y}) - (\hat{x}-\hat{y})^2)} \quad t^{1/3} h_0\left(\frac{y}{2t^{1/3}}\right)$$

$$t^{-1/3} h(0, t) \simeq \max_{\hat{y}} \left(\mathcal{A}_2(\hat{y}) - \hat{y}^2 + h_0(\hat{y}) \right)$$

$$h_0(\hat{y}) \simeq t^{-1/3} h(2t^{2/3}\hat{y}, 0)$$

$$\hat{y} = \frac{y}{2t^{2/3}}$$



general initial condition

$$e^{h(x,t)} = \text{sum over paths} \int dy Z_\eta(x, t | y, 0) e^{h(y, t=0)}$$

\downarrow \downarrow choose

$$e^{t^{1/3}(\mathcal{A}_2(\hat{x}-\hat{y}) - (\hat{x}-\hat{y})^2)} \quad t^{1/3} h_0\left(\frac{y}{2t^{1/3}}\right)$$

$$t^{-1/3} h(0, t) \simeq \max_{\hat{y}} (\mathcal{A}_2(\hat{y}) - \hat{y}^2 + h_0(\hat{y}))$$

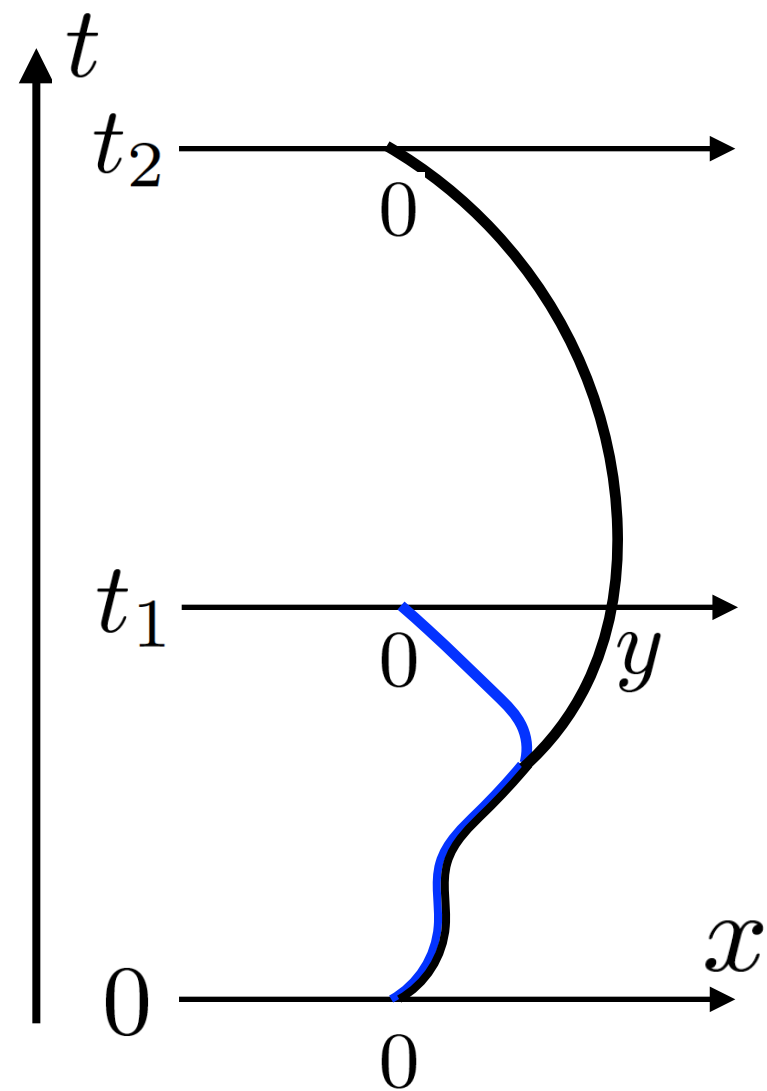
$$h_0(\hat{y}) \simeq t^{-1/3} h(2t^{2/3}\hat{y}, 0)$$

droplet $h_0(0) = 0 \quad h_0(\hat{y}) = -\infty$
 $\hat{y} \neq 0$

flat $h_0(\hat{y}) = 0$

stationary $h(x, t = 0) = B_0(x) \longrightarrow h_0(\hat{y}) = \sqrt{2}B(\hat{y})$ $\langle dB(x)^2 \rangle = dx$
 $B(0) = 0$

$$\hat{y} = \frac{y}{2t^{2/3}}$$



Two time KPZ via Airy processes

$$t_1, t_2 \rightarrow +\infty \quad \Delta = \frac{t_2 - t_1}{t_1} \quad \text{fixed}$$

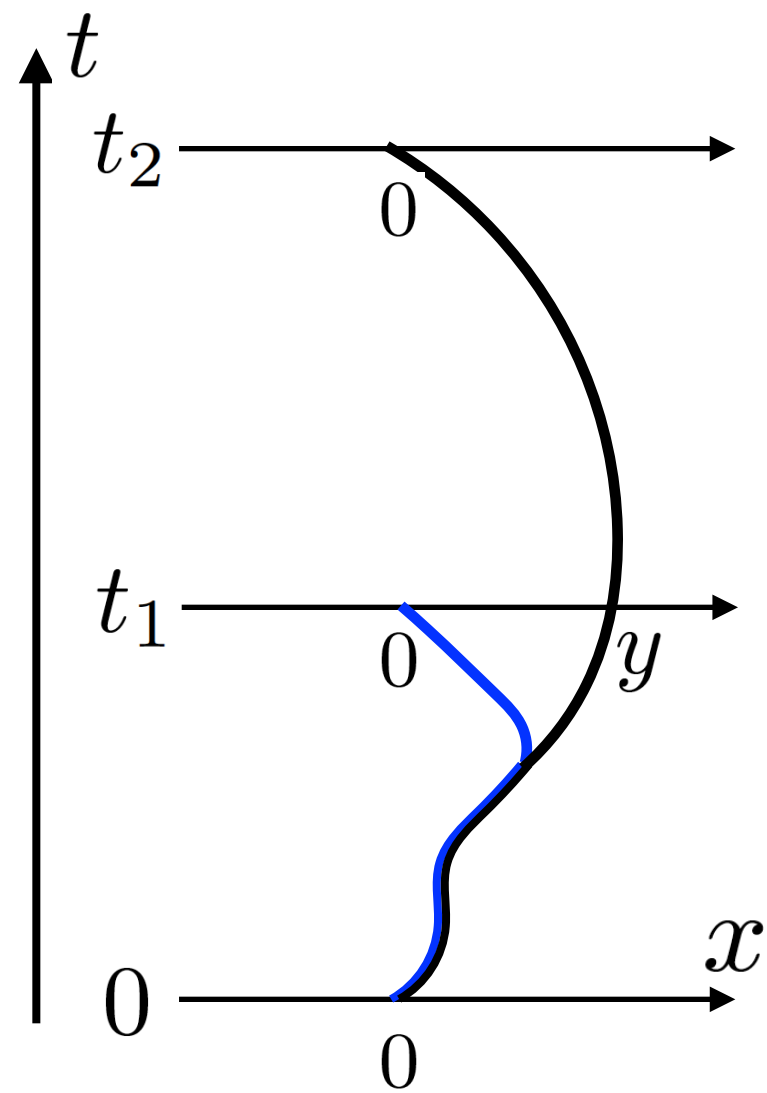
$$t_1^{-1/3} h(0, t_1) \simeq \mathcal{A}_2(0)$$

$$t_1^{-1/3} h(0, t_2) \simeq$$

$$\hat{y} = \frac{y}{2t_1^{2/3}}$$

$$\max_{\hat{y} \in \mathbb{R}} \left(\mathcal{A}_2(\hat{y}) - \hat{y}^2 + \Delta^{\frac{1}{3}} \left(\tilde{\mathcal{A}}_2\left(\frac{\hat{y}}{\Delta^{\frac{2}{3}}}\right) - \frac{\hat{y}^2}{\Delta^{\frac{4}{3}}} \right) \right)$$

Two time KPZ via Airy processes



$$t_1, t_2 \rightarrow +\infty \quad \Delta = \frac{t_2 - t_1}{t_1} \quad \text{fixed}$$

$$t_1^{-1/3} h(0, t_1) \simeq \mathcal{A}_2(0)$$

$$t_1^{-1/3} h(0, t_2) \simeq$$

$$\hat{y} = \frac{y}{2t_1^{2/3}}$$

$$\max_{\hat{y} \in \mathbb{R}} \left(\mathcal{A}_2(\hat{y}) - \hat{y}^2 + \Delta^{\frac{1}{3}} \left(\tilde{\mathcal{A}}_2\left(\frac{\hat{y}}{\Delta^{\frac{2}{3}}}\right) - \frac{\hat{y}^2}{\Delta^{\frac{4}{3}}} \right) \right)$$

$$\Delta \xrightarrow{\simeq} +\infty \quad \Delta^{\frac{1}{3}} \tilde{\mathcal{A}}_2(0) + \max_{\hat{y} \in \mathbb{R}} \left(\mathcal{A}_2(\hat{y}) - \hat{y}^2 + \sqrt{2} B(\hat{y}) \right) + O\left(\frac{1}{\Delta^{\frac{1}{3}}}\right)$$

in large time separation limit $\Delta \rightarrow +\infty$

correlations between $h(0, t_1)$ and $h(0, t_2)$ are contained in the joint distribution

$$\underset{\text{JCDF}}{G(\sigma_1, \sigma_2)} = \text{Prob}(\mathcal{A}_2(0) < \sigma_1, \max_{u \in \mathbb{R}} (\mathcal{A}_2(u) - u^2 + \sqrt{2} B(u)) < \sigma_2)$$

Explicit formula for JCDF

$$G(\sigma_1, \sigma_2) = \text{Prob}(\mathcal{A}_2(0) < \sigma_1, \max_{u \in \mathbb{R}} (\mathcal{A}_2(u) - u^2 + \sqrt{2}B(u)) < \sigma_2)$$

$$\begin{aligned} &= F_2(\sigma_1)Y_0(\sigma_1)\text{Tr}[(I - P_{\sigma_1}K_{\text{Ai}})^{-1}P_{\sigma_1}\text{Ai}_{\sigma_2-\sigma_1}\text{Ai}_{\sigma_2-\sigma_1}^T] \\ &\quad + F_2(\sigma_1)(\text{Tr}[(I - P_{\sigma_1}K_{\text{Ai}})^{-1}P_{\sigma_1}\text{Ai}_{\sigma_2-\sigma_1}\mathcal{B}_0^T] - 1)^2 \end{aligned}$$

$$\text{Ai}_{\sigma}(u) = \text{Ai}(u + \sigma)$$

marginals are

$$Y_{\hat{x}}(\sigma) := 1 + \mathcal{L}_{\hat{x}}(\sigma) - \text{Tr}[P_{\sigma}K_{\text{Ai}}(I - P_{\sigma}K_{\text{Ai}})^{-1}P_{\sigma}\mathcal{B}_{-\hat{x}}\mathcal{B}_{\hat{x}}^T]$$

$$\begin{aligned} \sigma_1 & \text{ GUE Tracy-Widom} \\ & F_2(\sigma_1) \end{aligned}$$

$$\mathcal{L}_{\hat{x}}(\sigma) = \sigma - 1 - \hat{x}^2 + \int_{\sigma}^{+\infty} dv(1 - \mathcal{B}_{\hat{x}}(v)\mathcal{B}_{-\hat{x}}(v))$$

$$\begin{aligned} \sigma_2 & \text{ Baik-Rains} \\ & F_0(\sigma_2) \end{aligned}$$

$$\mathcal{B}_w(v) = e^{\frac{1}{3}w^3 - vw} - \int_0^{+\infty} dy \text{Ai}(v + y)e^{wy}$$

$$G(\sigma_1, \sigma_2) = F_0(\sigma_2) \quad \sigma_1 \geq \sigma_2$$

$$F_0(\sigma) = \partial_{\sigma}(F_2(\sigma)Y_0(\sigma))$$

Consequence for 2-time KPZ: persistent correlations

$$C(t_1, t_2) = \frac{\overline{h(0, t_1)h(0, t_2)}^c}{\overline{h(0, t_1)^2}^c} \xrightarrow{t_1, t_2 \rightarrow +\infty} C_\Delta$$
$$C_{+\infty} = \frac{\langle \sigma_2 \sigma_1 \rangle^c}{\langle \sigma_1^2 \rangle}$$
$$\approx 0.6225 \pm 0.0015$$

Consequence for 2-time KPZ: persistent correlations

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$$C_{+\infty} = \frac{\langle \sigma_2 \sigma_1 \rangle^c}{\langle \sigma_1^2 \rangle}$$

$$\approx 0.6225 \pm 0.0015$$

$$0.626 \pm 0.003$$

T. Halpin-Healy
num.sim.DP (2017)

Consequence for 2-time KPZ: persistent correlations

$$C(t_1, t_2) = \frac{\overline{h(0, t_1)h(0, t_2)}^c}{\overline{h(0, t_1)^2}^c} \xrightarrow{t_1, t_2 \rightarrow +\infty} C_\Delta$$
$$C_{+\infty} = \frac{\langle \sigma_2 \sigma_1 \rangle^c}{\langle \sigma_1^2 \rangle}$$

conditional mean

$$\approx 0.6225 \pm 0.0015$$

$$h := (h(0, t_2) - h(0, t_1)) / (t_2 - t_1)^{1/3}$$

$$0.626 \pm 0.003$$

$$\bar{h}_{h_1=\sigma_1} \simeq \kappa_1^{\text{GUE}} + \frac{1}{\Delta^{1/3}} \langle \sigma_2 - \sigma_1 \rangle_{\sigma_1} + o\left(\frac{1}{\Delta^{1/3}}\right)$$

T. Halpin-Healy
num.sim.DP (2017)

check partial tail $\sigma_1 \gg 1$

$$p(\sigma_1, \sigma_2) \simeq -2\partial_{\sigma_2} K_{\text{Ai}}(\sigma_1, \sigma_2) - \text{Ai}(\sigma_2)^2$$

$$\langle \sigma_2 - \sigma_1 \rangle_{\sigma_1} \simeq \frac{[\int_{\sigma_1}^{+\infty} dy \text{Ai}(y)]^2 - \int_{\sigma_1}^{+\infty} dy K_{\text{Ai}}(y, y)}{K_{\text{Ai}}(\sigma_1, \sigma_1)}$$

exactly the result of the part II a) paper with de Nardis

$$h := (h(0, t_2) - h(0, t_1)) / (t_2 - t_1)^{1/3}$$

conditional mean

$$\overline{h}_{h_1 > \sigma_{1c}} = \kappa_1^{\text{GUE}} + \Delta^{-1/3} \langle \sigma_2 - \sigma_1 \rangle_{\sigma_1 > \sigma_{1c}}$$

$$\kappa_1^{\text{GUE}} = -1.771$$

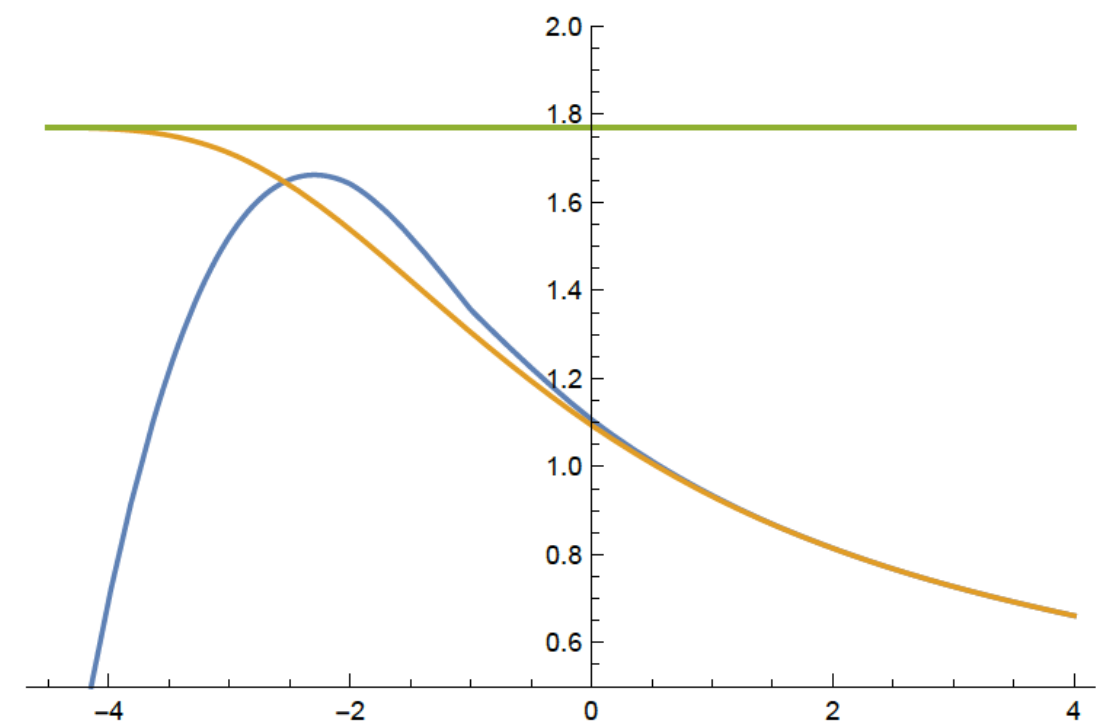
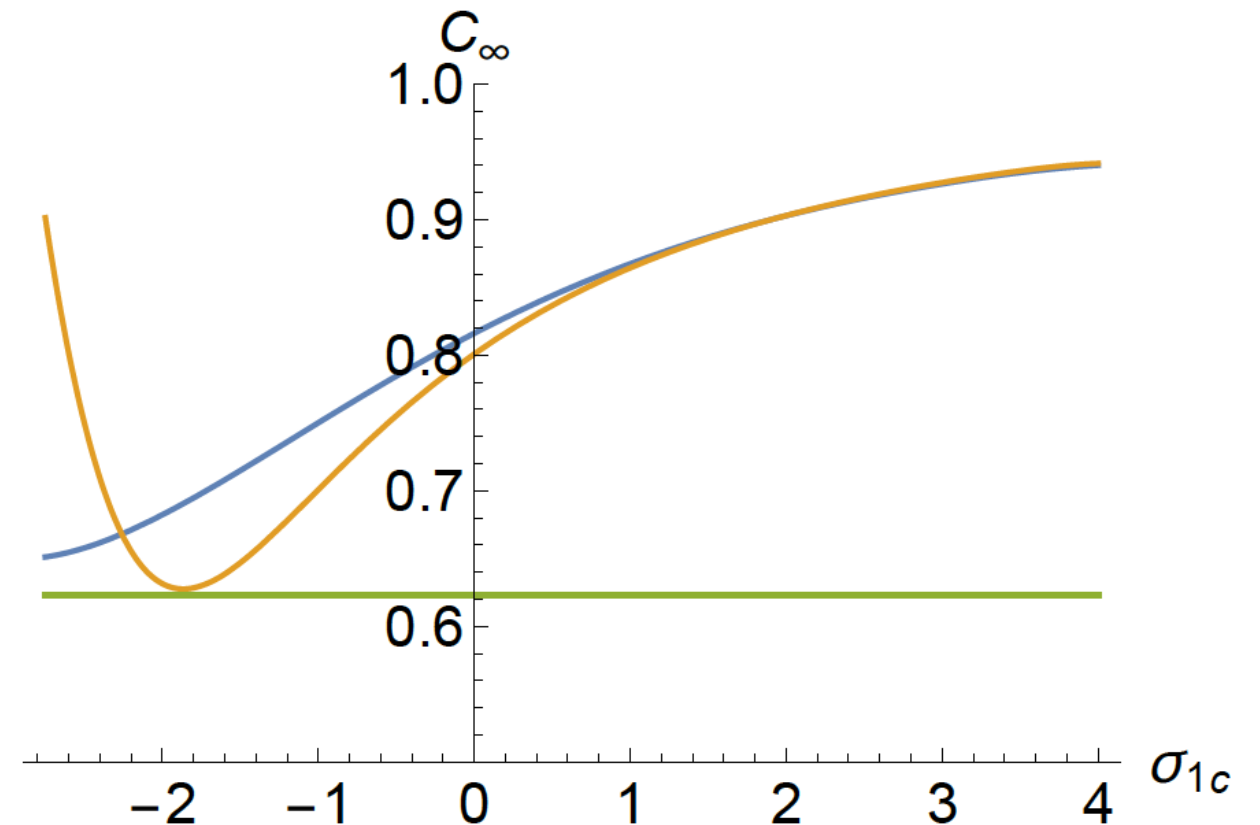


FIG. 2. Conditional average $\langle \sigma_2 - \sigma_1 \rangle_{\sigma_1 > \sigma_{1c}}$ (y axis) as a function of σ_{1c} (x axis), which described the averaged scaled KPZ

more detailed prediction
conditional covariance ratio

$$C_\infty(\sigma_{1c}) = \lim_{\Delta \rightarrow \infty} \frac{\overline{h_1 h_2}_{h_1 > \sigma_{1c}}^c}{\overline{h_1^2}_{h_1 > \sigma_{1c}}^c}$$

$$C_\infty \approx 0.6225 \pm 0.0015$$



$$C_{\Delta}(\sigma_{1c}) := \frac{\overline{h_1 h_2}_{h_1 > \sigma_{1c}}^c}{\overline{h_1^2}_{h_1 > \sigma_{1c}}^c}$$

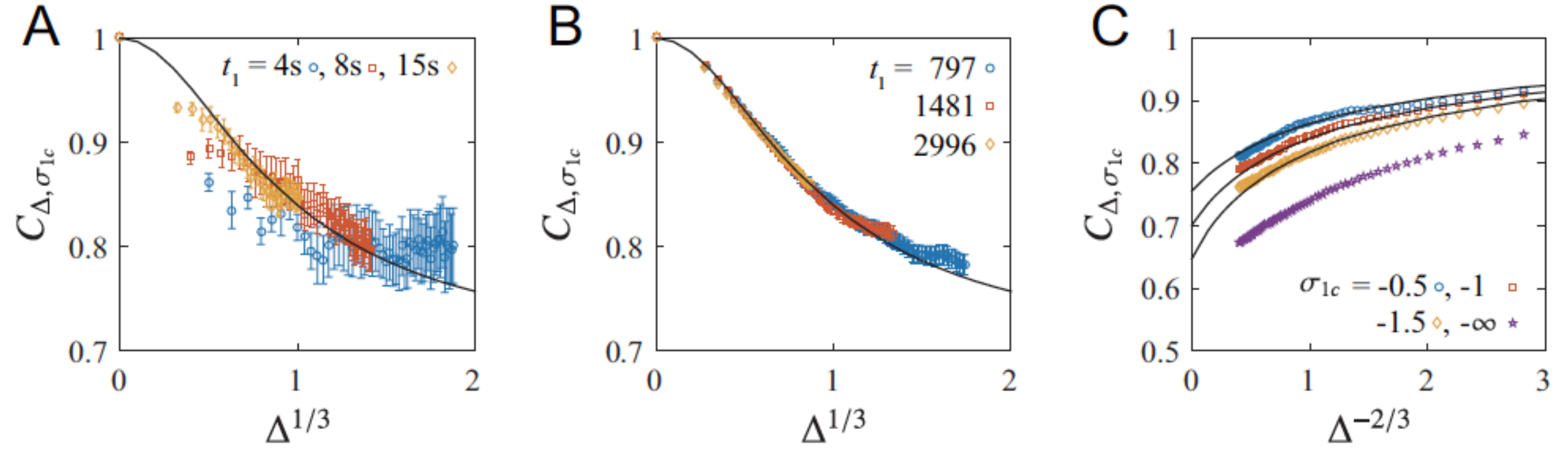


FIG. 4: The conditional covariance $C_{\Delta, \sigma_{1c}} = C(t_1, t_2)_{\tilde{h}_1 > \sigma_{1c}} / C(t_1, t_1)_{\tilde{h}_1 > \sigma_{1c}}$ ((3)). a,b, Experimental (A) and numerical (B) results for $\sigma_{1c} = -1$ and with varying t_1 (symbols), compared with the theoretical prediction (black line). The error bars indicate the standard errors. To reduce the effect of finite-time corrections, here we used such realizations that satisfy $\tilde{h}_1 > \tilde{h}_{1c}$ with $\text{Prob}[\tilde{h}_1 \geq \tilde{h}_{1c}] = 1 - F_2(\sigma_{1c})$. c, Numerical data for $t_1 = 1008$ and for $\sigma_{1c} = -0.5, -1, -1.5$ and $-\infty$ (unconditioned). Error bars are omitted here for the sake of visibility. The black lines indicate the theoretical predictions for finite σ_{1c} . At large Δ and for any σ_{1c} they converge to their asymptotic values as $C_{\Delta \rightarrow \infty, \sigma_{1c}} + A_{\sigma_{1c}} \Delta^{-2/3} + B_{\sigma_{1c}} \Delta^{-1} + \dots$. For $\sigma_{1c} = -\infty$ (the unconditioned case), the theory suggests a strictly positive asymptotic value, specifically $C_{\infty, -\infty} \approx 0.6$, which is consistent with the trend of the unconditioned data set in the panel c (purple stars).

J. de Nardis, PLD, K. Takeuchi
Phys. Rev. Lett. 118, 125701 (2017)

complicated exact expressions for some joint moments

$$\overline{Z_1^{n_1} Z_L^{n_L} Z_R^{n_R}} \longrightarrow \text{generating function}$$

↓

large time + “decoupling assumption”

magic trick

$$\text{Prob}(\mathcal{A}_2(-\hat{x}) < \sigma_1, \hat{h}_L(\hat{x}) + \hat{x}^2 < \sigma_L, \hat{h}_R(\hat{x}) + \hat{x}^2 < \sigma_R)$$

$$\hat{h}_{L,R}(\hat{x}) = \max_{\hat{y} < 0, \hat{y} > 0} (\mathcal{A}_2(\hat{y} - \hat{x}) - (\hat{y} - \hat{x})^2 + \sqrt{2}\hat{B}(\hat{y}))$$

Distribution of argmax of Airy minus parabola plus Brownian

$$H(\hat{x}) = \text{Prob}(\hat{y}_m > \hat{x}) \qquad \hat{y}_m = \text{argmax}_{\hat{y} \in \mathbb{R}} \left(\mathcal{A}_2(\hat{y}) - \hat{y}^2 + \sqrt{2}B(\hat{y}) \right)$$

application: midpoint probability
of very long DP

$$\begin{aligned} H(-\hat{x}) = & \int d\sigma F_2(\sigma) \left(Y_{\hat{x}}(\sigma) \right. \\ & \times \text{Tr}[(I - P_\sigma K_{\text{Ai}})^{-1} P_\sigma (\text{Ai}' + \hat{x} \text{Ai}) \text{Ai}^T] \\ & + (\text{Tr}[(I - P_\sigma K_{\text{Ai}})^{-1} P_\sigma \text{Ai} \mathcal{B}_{\hat{x}}^T] - 1) \\ & \left. \times \text{Tr}[(I - P_\sigma K_{\text{Ai}})^{-1} P_\sigma (\text{Ai}' + \hat{x} \text{Ai}) \mathcal{B}_{-\hat{x}}^T] \right) \end{aligned}$$

directed polymer from $(0, 0)$ to $(0, t_2)$

position $x(t_1) = y$ at intermediate time t_1

$$\overline{P_{t_1, t_2}(y)} dy = \frac{\overline{Z(0, t_2|y, t_1) Z(y, t_1|0, 0)}}{Z(0, t_2|0, 0)} dy \rightarrow P_\Delta(\hat{y}) d\hat{y}$$

$$P_{+\infty}(\hat{y}) d\hat{y} = \text{Prob}(\hat{y}_m \in [\hat{y}, \hat{y} + dy])$$

Distribution of argmax of Airy minus parabola plus Brownian

$$H(\hat{x}) = \text{Prob}(\hat{y}_m > \hat{x}) \qquad \hat{y}_m = \operatorname{argmax}_{\hat{y} \in \mathbb{R}} \left(\mathcal{A}_2(\hat{y}) - \hat{y}^2 + \sqrt{2}B(\hat{y}) \right)$$

$$\begin{aligned}
 H(-\hat{x}) = & \int d\sigma F_2(\sigma) \left(Y_{\hat{x}}(\sigma) \right. \\
 & \times \text{Tr}[(I - P_\sigma K_{\text{Ai}})^{-1} P_\sigma (\text{Ai}' + \hat{x} \text{Ai}) \text{Ai}^T] \\
 & + (\text{Tr}[(I - P_\sigma K_{\text{Ai}})^{-1} P_\sigma \text{Ai} \mathcal{B}_{\hat{x}}^T] - 1) \\
 & \left. \times \text{Tr}[(I - P_\sigma K_{\text{Ai}})^{-1} P_\sigma (\text{Ai}' + \hat{x} \text{Ai}) \mathcal{B}_{-\hat{x}}^T] \right)
 \end{aligned}$$

application:midpoint probability
of very long DP

directed polymer from (0,0) to (0,t₂)
position x(t₁) = y at intermediate time t₁

$$\overline{P_{t_1,t_2}(y)} dy = \frac{\overline{Z(0,t_2|y,t_1)Z(y,t_1|0,0)}}{Z(0,t_2|0,0)} dy \rightarrow P_\Delta(\hat{y}) d\hat{y}$$

$$P_{+\infty}(\hat{y}) d\hat{y} = \text{Prob}(\hat{y}_m \in [\hat{y}, \hat{y} + dy])$$

C. Maes and T. Thiery, arXiv:1704.06909

Fluctuation Dissipation relations stationary Burgers
<=> variance of the height in stationary KPZ

$$\mathcal{P}(\hat{y}) = f_{KPZ}(\hat{y})$$

$$f_{KPZ}(\hat{y}) = \frac{1}{4} g''(\hat{y})$$

I checked that

$$H(\hat{x}) = \frac{1}{2} - \frac{1}{4} g'(\hat{x})$$

$$g'(\pm\infty) = \pm 2$$

$$g(\hat{x}) = \langle \sigma^2 \rangle_{F_0,\hat{x}} - \langle \sigma \rangle_{F_0,\hat{x}}^2$$

Conclusions

- used replica Bethe ansatz (RBA), well-tested method to calculate 1-time observables for the KPZ equation (large time convergence to TW)
- partial solution of the 2-times JPDF of heights (h_1, h_2) for the KPZ equation at large times
 - I) obtained $P(h_1, h_2 - h_1)$ for large $h_1 > 0$, any t_2/t_1
 - => predicted conditional cumulants of $h_2 - h_1$ and conditional covariance ratio for large h_1 universal functions across KPZ class
 - found to agree with experiments in broad range of h_1
 - calculated with RBA various joint distributions of max of Airy plus Brownian
 - II) obtained 2 time covariance ratio (any h_1) for large t_2/t_1
 - agrees with I: RBA magic tricks work!
- => Memory effect for growth in expanding geometry
 - still no answer if simple multi-time structure will emerge
 - compare with other results: semi-discrete DP (Johansson)
TASEP ring (Baik, Liu) , e.g. tails?
 - push these RBA methods to get other multi-time observables

Next?

Perspectives/other works

- replica BA method

stationary KPZ	Sasamoto Inamura	$t \rightarrow \infty$	Airy process $A_2(y)$
2 space points	$Prob(h(x_1, t), h(x_2, t))$	Prohlac-Spohn (2011), Dotsenko (2013)	
2 times	$Prob(h(0, t), h(0, t'))$	Dotsenko (2013)	
endpoint distribution of DP	Dotsenko (2012)	Schehr, Quastel et al (2011)	

- rigorous replica..

Borodin, Corwin, Quastel, O Connell, ..

q-TASEP	$q \rightarrow 1$	avoids moment problem	$\overline{Z^n} \sim e^{cn^3}$
WASEP	Bose gas	moments as nested contour integrals	

- sine-Gordon FT

P. Calabrese, M. Kormos, PLD, EPL 10011 (2014)

- Integrable lattice directed polymers

geometric $T=0$, log-gamma, beta $T>0$

Johansson (2000) Seppalainen (2012)
COSZ(2011) BCR(2013), Thiery, PLD(2014)
Barraquand, Corwin(2014) T. Thiery, PLD(2015)
connect to RWtimedepRE

KPZ at finite time and fermions at finite temperature in a trap

■ KPZ equation with “droplet” initial condition

units: $t^* = 2(2\nu)^5 / D^2 \lambda_0^4$ scaled height: $\tilde{h}(0, t) = \frac{h(0, t) + \frac{t}{12}}{t^{1/3}}$
 $h^* = 2\nu / \lambda_0$

Calabrese, Le Doussal, Rosso '10/Dotsenko '10/
Sasamoto, Spohn '10/Amir, Corwin, Quastel '11

$$g_t(s) := \langle \exp(-e^{t^{1/3}}(\tilde{h}(0, t) - s)) \rangle = \text{Det}(1 - P_s K_{KPZ} P_s)$$

$t \rightarrow +\infty$ TW distrib

$$\text{Proba}(\tilde{h}(0, t) < s) = F_2(s)$$

$$K_{KPZ}(a, b) = \int_{-\infty}^{+\infty} dv \frac{\text{Ai}(a + v) \text{Ai}(b + v)}{e^{-t^{1/3}v} + 1}$$

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■ N non-interacting fermions in harmonic trap near edge

$$T \sim N^{1/3} \rightarrow +\infty \text{ with fixed } b = \frac{\hbar \omega N^{1/3}}{T}$$

position of rightmost fermion \Leftrightarrow KPZ height

Dean, PLD Majumdar,
 Schehr PRL (2015)

$$\frac{x_{max}(T) - r_{edge}}{w_N} \stackrel{\text{inlaw}}{=} \frac{h(0,t) + \frac{1}{12}t + G}{t^{1/3}} \qquad P(G) = e^{-G-e^{-G}}$$

$$w_N = \frac{N^{-1/6}}{\sqrt{2} \alpha} \qquad b = t^{1/3}$$

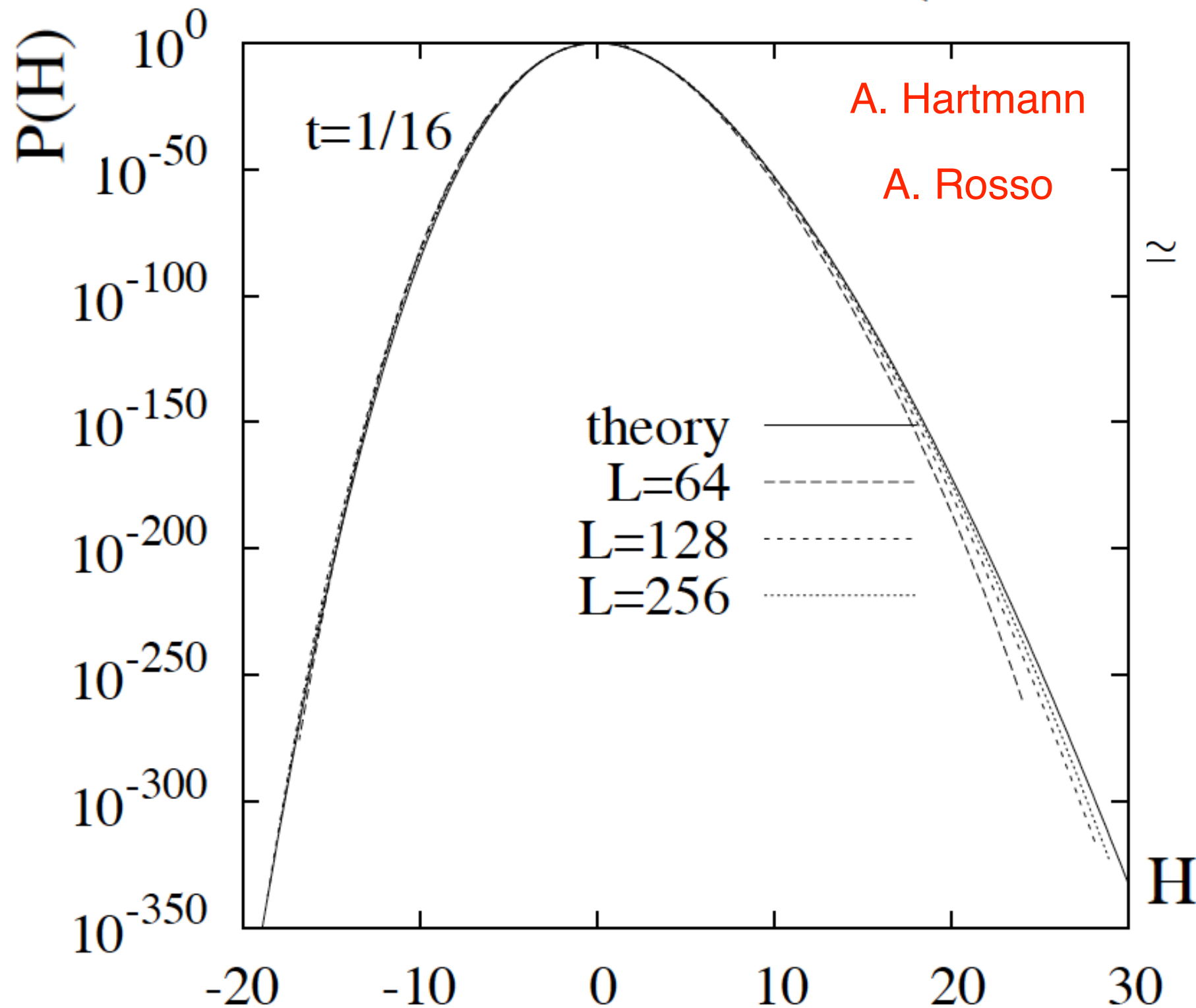
G is random variable with Gumbel distribution independent of h(0,t)

Two works I will NOT talk about !

Large deviations for KPZ (droplet IC) at short time

$$P(H, t) \sim \exp \left(-\frac{\Phi_{\text{drop}}(H)}{\sqrt{t}} \right)$$

$$\Phi_{\text{drop}}(H) = \begin{cases} \frac{-1}{\sqrt{4\pi}} \min_{z \in [-1, +\infty[} [ze^H + Li_{\frac{5}{2}}(-z)], & H \leq H_c \\ \frac{-1}{\sqrt{4\pi}} \min_{z \in [-1, 0[} [ze^H + Li_{\frac{5}{2}}(-z) \\ - \frac{8\sqrt{\pi}}{3} (-\ln(-z))^{\frac{3}{2}}], & H \geq H_c \end{cases} \quad (22)$$



A. Hartmann

A. Rosso

$$\simeq \begin{cases} \frac{4}{15\pi} |H|^{5/2} & , \quad H \rightarrow -\infty \\ \frac{H^2}{\sqrt{2\pi}} & , \quad |H| \ll 1 \quad \text{EW (typical)} \\ \frac{4}{3} H^{3/2} & , \quad H \rightarrow +\infty . \end{cases}$$

PLD, Majumdar, Rosso, Schehr,
PRL 117, 070403 (2016)

N mutually avoiding paths in random potential

$\hat{\mathcal{Z}}_1(t)$ continuum partition sum of one directed polymer w. fixed endpoints at 0

$$\ln \hat{\mathcal{Z}}_1(t) \simeq -t/12 + \hat{\gamma}_1 t^{1/3}$$

largest eigenvalues of a GUE
random matrix

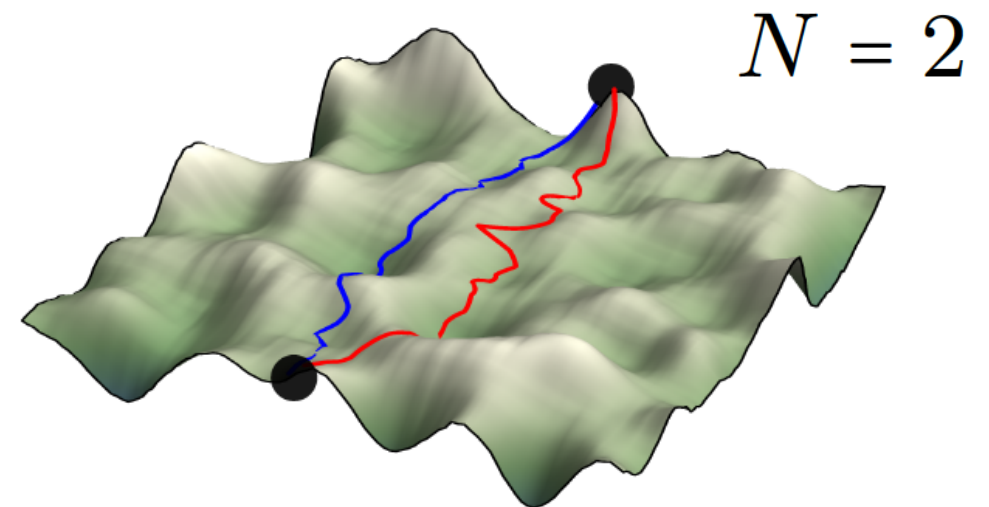
$\hat{\mathcal{Z}}_N(t)$ continuum partition sum of N non-crossing DP w. fixed endpoints at 0
in same random potential

CONJECTURE

$$\ln \hat{\mathcal{Z}}_N(t) \simeq -Nt/12 + t^{1/3} \hat{\zeta}^{(N)}$$

$$\hat{\zeta}^{(N)} \stackrel{\text{in law}}{\equiv} \sum_{i=1}^N \hat{\gamma}_i =: \hat{\gamma}$$

$\hat{\gamma}_1, \dots, \hat{\gamma}_N$ N largest eigenvalues of a GUE random matrix



T=0 semidiscrete DP model

Yor, O'Connell, Doumerc (2002)

T>0 continuum (KPZeq)

Andrea de Luca, PLD, arXiv1606.08509,
Phys. Rev. E 93, 032118 (2016) and 92, 040102 (2015)

Generalized Bethe Ansatz

Universal distribution of conductance in 2D localized phase

Somoza, Ortuno, Prior (2007)

$$\ln g = -\frac{2L}{\xi} + \alpha \left(\frac{L}{\xi} \right)^{1/3} \chi_2$$

ξ localization length

L system size

χ random variable with Tracy Widom distribution

$$H = \sum_i \epsilon_i c_i^\dagger c_i - t \sum_{\langle ij \rangle} c_i^\dagger c_j + c_j^\dagger c_i$$

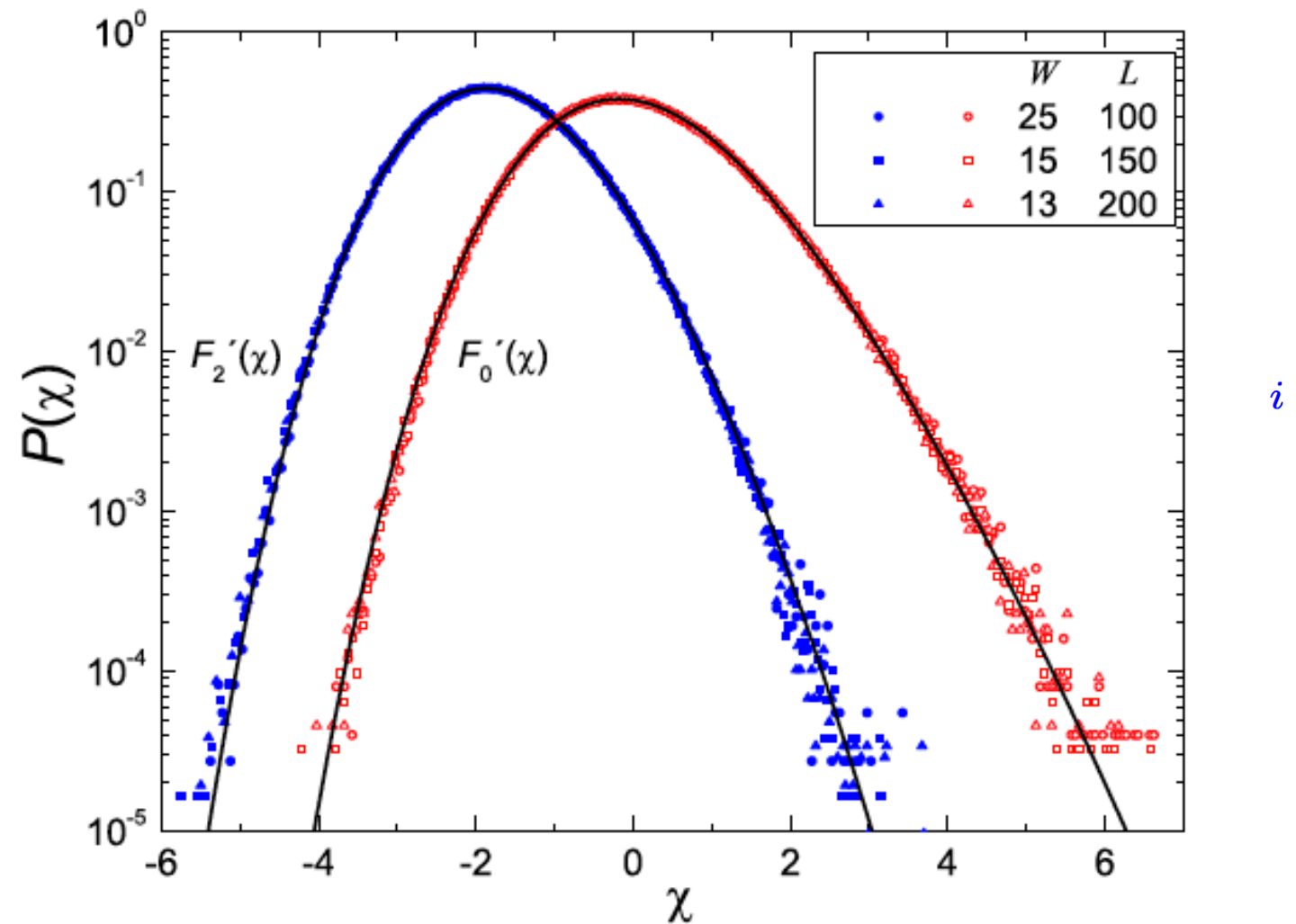


FIG. 1 (color online). Histograms of $\ln g$ versus the scaled variable χ for several sizes and disorders of the Anderson model with narrow (solid symbols) and wide (empty symbols) leads. The continuous lines correspond to $F'_2(\chi)$ and $F'_0(\chi)$.

Summary:

for droplet initial conditions $\frac{\lambda_0 h}{2\nu} \equiv \ln Z = v_\infty t + 2^{2/3} \left(\frac{t}{t^*}\right)^{1/3} \chi$

χ at large time has the same distribution
as the largest eigenvalue of the GUE

for flat initial conditions $\frac{\lambda_0 h}{2\nu} \equiv \ln Z = v_\infty t + \left(\frac{t}{t^*}\right)^{1/3} \chi$
similar (more involved)

χ at large time has the same distribution
as the largest eigenvalue of the GOE $t^* = \frac{8(2\nu)^5}{D^2 \lambda_0^4}$

in addition: $g(x)$ for all times
 $\Rightarrow P(h)$ at all t (inverse LT)

describes full crossover from
Edwards Wilkinson to KPZ

t^* is crossover time scale

GSE ? KPZ in half-space

Integrable directed polymer (DP) on square lattice

$$Z_t(x) = \sum_{\pi: (0,0) \rightarrow (x,t)} \prod_{(x',t') \in \pi} w_{x',t'}$$

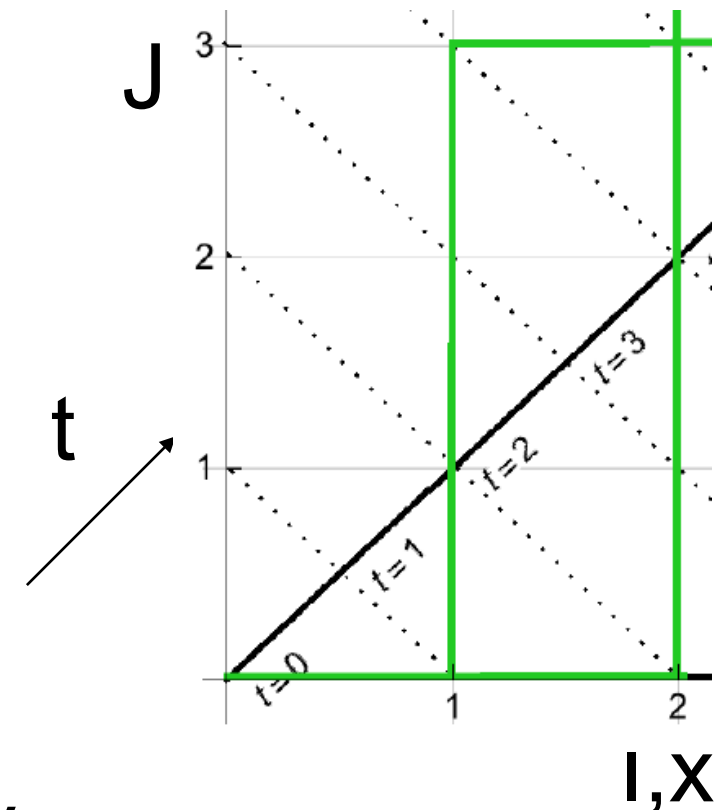
- log-Gamma DP on-site weights

inverse Gamma distribution

Seppalainen (2012) Brunet
COSZ(2011) BCR(2013), Thiery, PLD(2014)

$$w \in [0, +\infty[$$

$$P(w) = \frac{1}{\Gamma(\gamma)} w^{-1-\gamma} e^{-1/w}$$

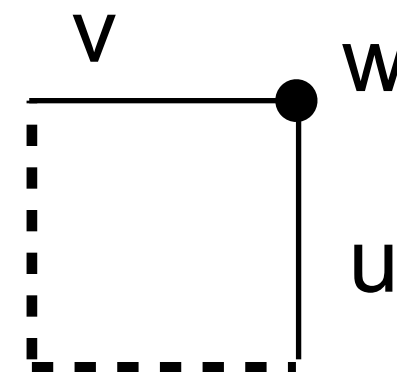


- Strict-Weak DP $u \in [0, +\infty[$ $v = 1$

Gamma distribution

Corwin, Seppalainen, Shen(2014)
O'Connell, Ortmann(2014)

$$P(u) = \frac{u^{\alpha-1}}{\Gamma(\alpha)} e^{-u}$$



- Beta DP $u, v \in [0, 1]$ $v = 1 - u$

Beta distribution

Barraquand, Corwin(2014)

$$p_{\alpha,\beta}(u) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} u^{\alpha-1} (1 - u)^{\beta-1} \quad \alpha > 0 \text{ and } \beta > 0$$

- Inverse-Beta DP

inverse-Beta distribution

$$u \in [1, +\infty[\quad v \in [0, +\infty[\quad v = u - 1$$

$$\tilde{p}_{\gamma,\beta}(u) = \frac{\Gamma(\gamma + \beta)}{\Gamma(\gamma)\Gamma(\beta)} \frac{1}{u^{1+\gamma}} \left(1 - \frac{1}{u}\right)^{\beta-1} \quad \gamma := 1 - (\alpha + \beta)$$

T. Thiery, PLD(2015)

