

Weizmann 24.1.2017

KPZ Growing Interfaces: How Flat is Flat?

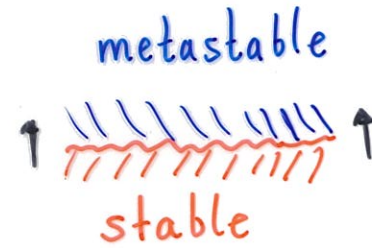
Herbert Spohn

TUMünchen

joint work

S. Chhita, P.L. Ferrari (Bonn)

KPZ growth



- independent nucleations
- diffusive relaxation
- sideways growth

HERE 1+1 dimensions

height function h

$$\partial_t h = \frac{1}{2} \lambda (\partial_x h)^2 + \frac{1}{2} \partial_x^2 h + \underbrace{W}_{\text{space-time white noise}}$$

space-time white noise

→ flat initial data $h(x,0) = 0$ ←

Conjecture:

$$h(0,t) \cong v_0 t + (Tt)^{1/3} \overset{\text{GOE}}{\underbrace{\text{universal}}}$$

↑ model



$$P(\sum_{GOE} \leq s) = \det(1 - P_s K^\circ P_s)$$

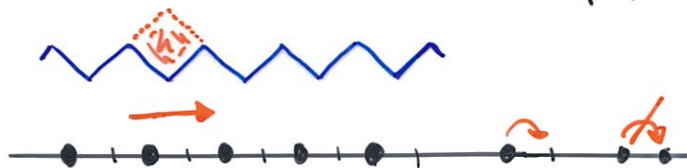
$$K^\circ(x,y) = \text{Ai}(x+y)$$

P_s projects on $[s, \infty)$

proved in discrete setting

- single step growth

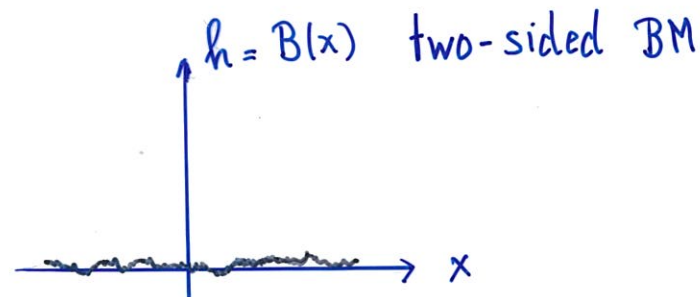
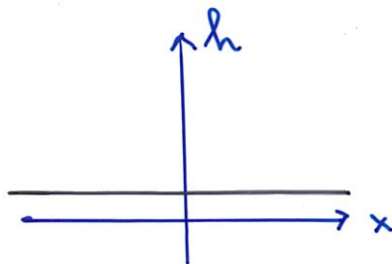
TASEP



fill minima with rate 1

Borodin, Ferrari, Prähofer, Sasamoto 2006

How flat is flat?



Theorem (Borodin, Corwin, Ferrari, Vetö 2014)

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

$$h(0, t) \cong v_0 t + (Tt)^{1/3} \underbrace{\zeta_{BR}}_{\text{Baik-Rains}}$$

What about

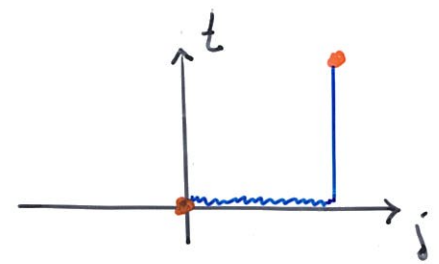
$$\langle h(x, 0) \rangle = 0$$

$$\langle h(x, 0)^2 \rangle \cong |x|^{2K} \quad K < 1 \quad ?$$

$\Rightarrow h'(x)$ stationary

single step

$\eta_j = \begin{cases} 0 \\ 1 \end{cases}, j \in \mathbb{Z}, \eta_j(t) \Rightarrow \text{height } h(j,t)$



• stationary $\{\eta_j\}$ Bernoulli $E(\eta_j) = p$

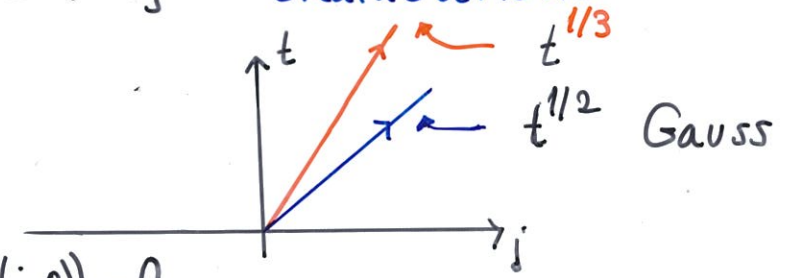
Ferrari, HS. 2006

• $p = \frac{1}{2}$ $h(0,t) \approx v_0 t + (Tt)^{1/3} \approx_{BR}$

• general p : current $j(p) = p(1-p), j'(p) = 1-2p$

characteristic

HERE



$j \rightarrow \eta_j$ stationary process $E(\eta_j) = \frac{1}{2}, E(h(j,0)) = 0$

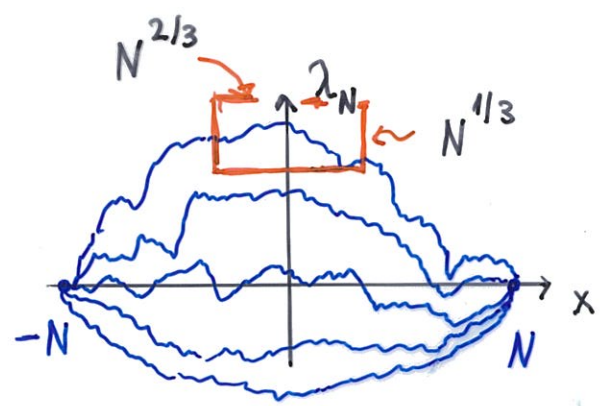
• CLT

$\lim_{l \rightarrow \infty} \frac{1}{\sqrt{l}} h([yl, lx], 0) = \sqrt{2} \circ B(x)$

two-sided Brownian motion

weak convergence of path measure on $C(\mathbb{R})$

a variational result (Johansson 2003)



Airy₂ process $A_i(x)$

$$\lim_{N \rightarrow \infty} N^{-1/3} (\lambda_N(N^{2/3}x) - 2N) = A_i(x) - x^2$$

N Brownian bridges

conditioned on nonintersection

$\Rightarrow A_i(x)$

continuous path

stationary

locally Brownian

fixed x : $A_i(x)$ is GUE Tracy-Widom

long-ranged

$$E(A_i(x) A_i(0))_c \cong \frac{1}{|x|^2}$$

Theorem

$\sup_x \{ A_i(x) - x^2 \}$ is GOE Tracy-Widom

stationary initial data

TASEP

$$F^{(\sigma)}(s) = \mathbb{P} \left(\sup_u \{ \sqrt{2} \sigma B(u) + Ai(u) - u^2 \} \leq s \right)$$

Universal

Theorem

$$\lim_{t \rightarrow \infty} \mathbb{P} \left(h(0,t) \geq \frac{7}{2}t - s(t/2)^{1/3} \right) = F^{(\sigma)}(s)$$

Flat is $\sigma = 0$ subdiffusive

deterministic initial data, KPZ
Quastel, Remenik 2016

special cases

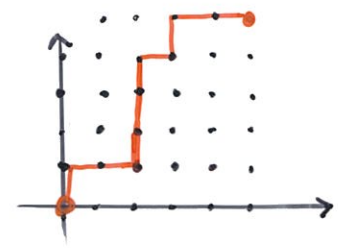
$\sigma = 0$ GOE Tracy-Widom

$\sigma = 1$ Baik-Rains this theorem + Ferrari, HS 2006

$\sigma = \infty$ $\mathbb{P} \left(\sup_u \{ B(u) - u^2 \} \leq s \right)$ Groeneboom 2010

Proof relies on

TASEP \iff last passage percolation



i.i.d. EXP up/right
optimal path

tightness
slow decorrelation

Cator, Pimentel 2015

Corwin, Ferrari, Peche 2012

Borodin, Peche 2008

Corwin, Liu, Wang 2016

HERE:

formal argument based on KPZ

sketch

$$\partial_t h = \frac{1}{2} (\partial_x h)^2 + \frac{1}{2} \partial_x^2 h + W \quad || \quad h(x, 0) = \sigma B(x)$$

Cole-Hopf

$$Z = e^h$$

stationary: $\sigma=1$

Funaki, Quastel 2015

$$\partial_t Z = \frac{1}{2} \partial_x^2 Z + WZ$$

directed polymer $b(t)$

$$Z(x, t) = \mathbb{E}_{b(t)=x} \left(e^{\int_0^t W(b(s), s) ds} e^{\sigma B(b(0))} \right)$$

|| finite temperature
last passage percolation

random

$$Z(0, t) = \int dx \mathbb{E}_{b(t)=0, b(0)=xt^{2/3}} \left(e^{\int_0^t W(b(s), s) ds} \right) e^{\sigma B(xt^{2/3})}$$

point-to-point

$$e^{t^{1/3} (A_i(x) - x^2) + v_0 t}$$

fixed x theorem



$$Z(0, t) \cong e^{v_0 t} \int dx e^{t^{1/3} (A_i(x) - x^2 + \sigma B(x))}$$

$$\lim_{t \rightarrow \infty} t^{-1/3} (\log Z(0, t) - v_0 t) = \sup_u \{ A_i(u) - u^2 + \sigma B(u) \}$$

→ properties of $F^{(\sigma)}$?

Monte-Carlo simulations

→ universality

conical (KPZ) point of 6-vertex model

M C of TASEP

$$\eta_i = \begin{cases} 0 \\ 1 \end{cases}$$

Markov chain

$$0 \leftrightarrow 1$$

probability $1 - \alpha$

α	σ
0	0
$\frac{1}{2}$	1
→ 1	→ ∞

NO adjustable parameters

- large σ approximation

$$F_{app}^{(\sigma)}(s) = \mathbb{P} \left(\underbrace{A_i(0)}_{GUE} + \sup_n \left\{ \sqrt{2} \sigma B(n) - n^2 \right\} \leq s \right)$$

GUE

from Groeneboom 2010

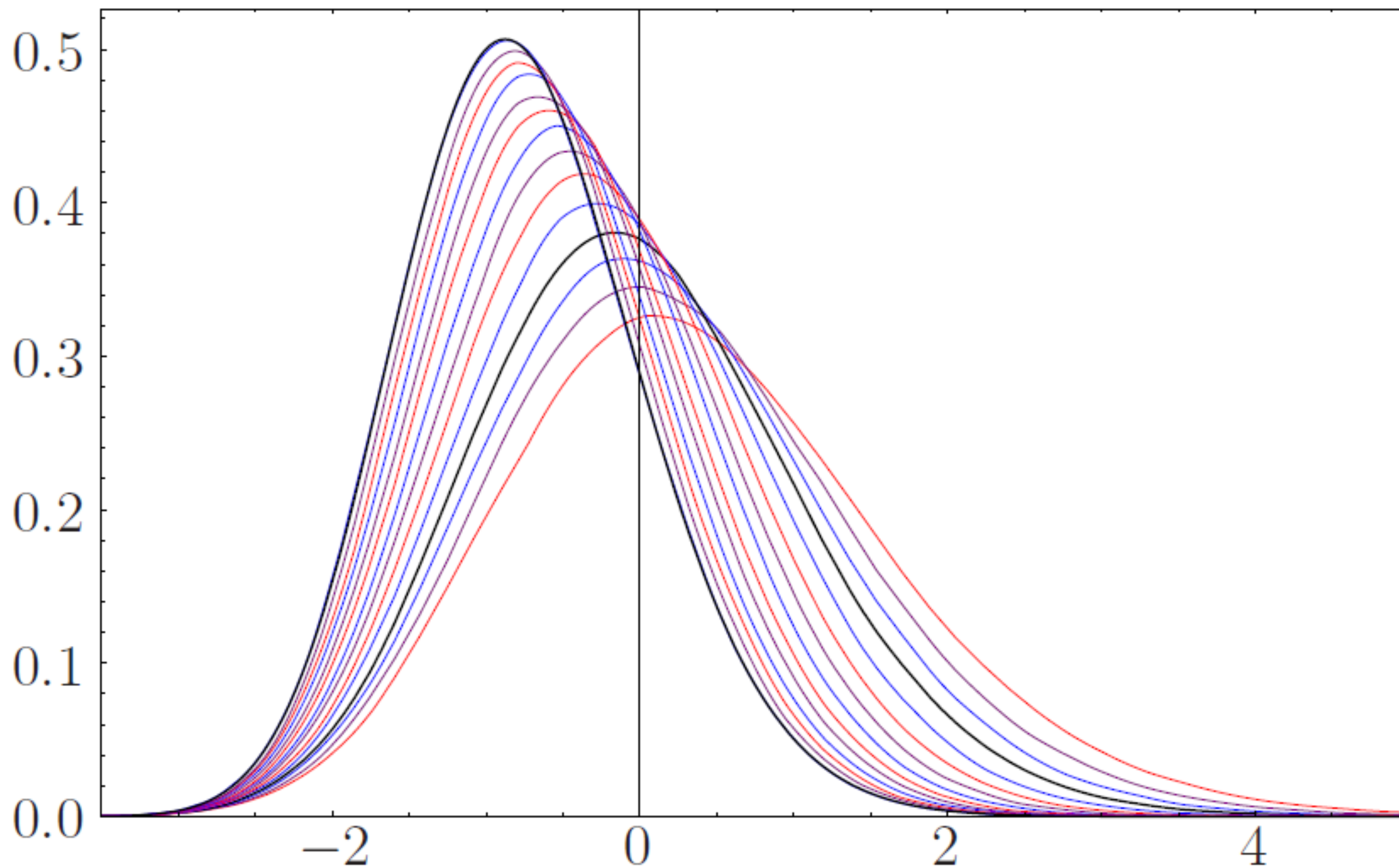


Figure 6: Probability densities of $F^{(\sigma)}(s)$ with $\sigma = \sqrt{\alpha/(1-\alpha)}$ from TASEP simulation until time $t_{\max} = 10^3$ and 10^6 runs. The different plots corresponds to the values $\alpha = 0, 0.05, 0.1, 0.15, 0.20, 0.25, 0.3, 0.35, 0.4, 0.45, 0.5, 0.54, 0.58, 0.62$. The left-most black line is the exact rescaled GOE distribution ($\sigma = 0$), which overlaps with $\alpha = 0$ from the simulations. The black line in the middle is the stationary case ($\sigma = 1$).

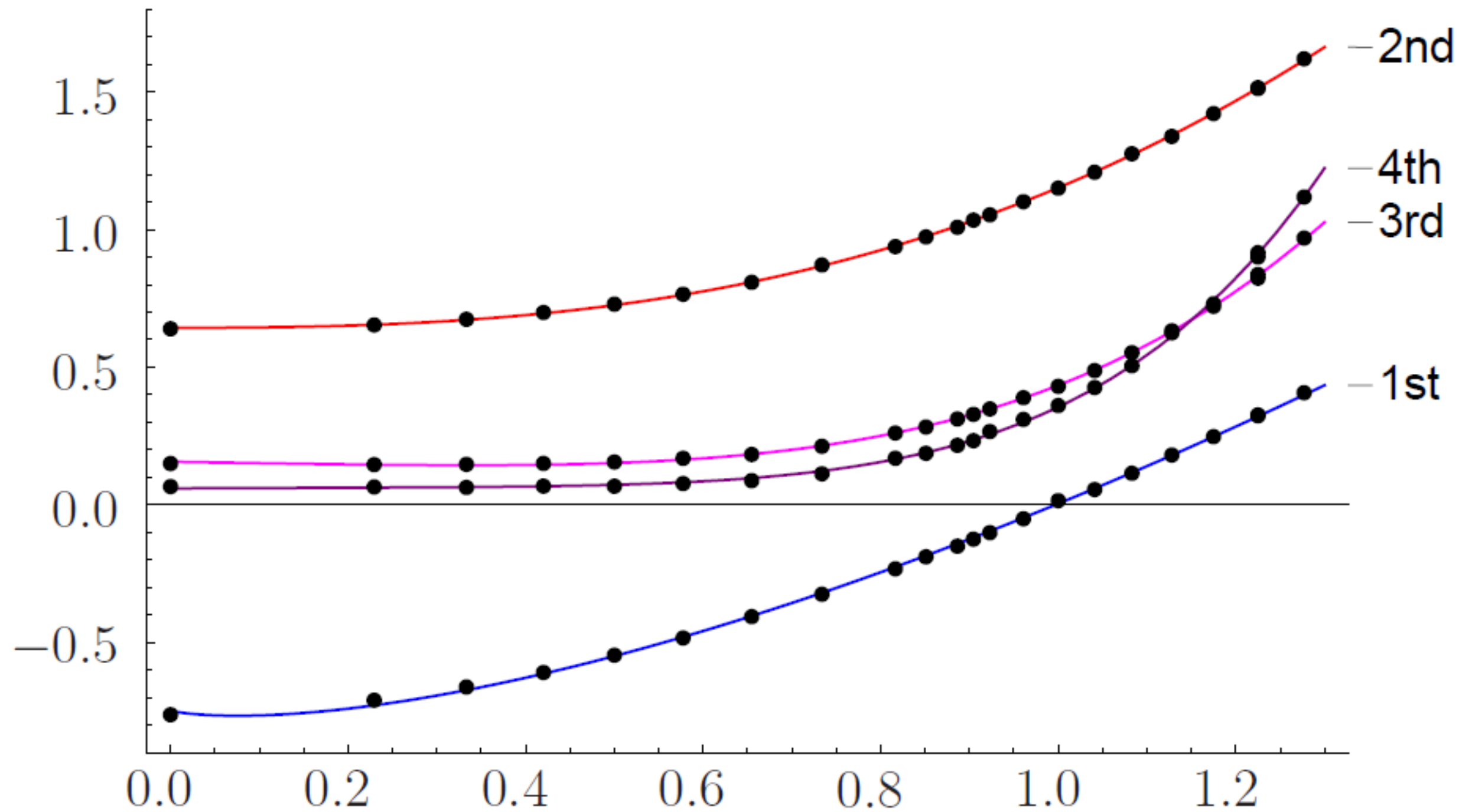


Figure 5: Empirical values for the first four cumulants of $F^{(\sigma)}$ as a function of σ . The empirical extrapolations are: $\kappa_1(\sigma) = -0.74795 - 0.96976\sigma + 1.7217\sigma^{4/3}$, $\kappa_2(\sigma) = 0.64268 + 0.0068163\sigma + 0.50202\sigma^{8/3}$, $\kappa_3(\sigma) = 0.15631 - 0.049956\sigma + 0.32731\sigma^4$, and $\kappa_4(\sigma) = 0.059476 + 0.012039\sigma + 0.28315\sigma^{16/3}$. The fitting made for the values shown in the figure, gives quite good results also for large values of σ

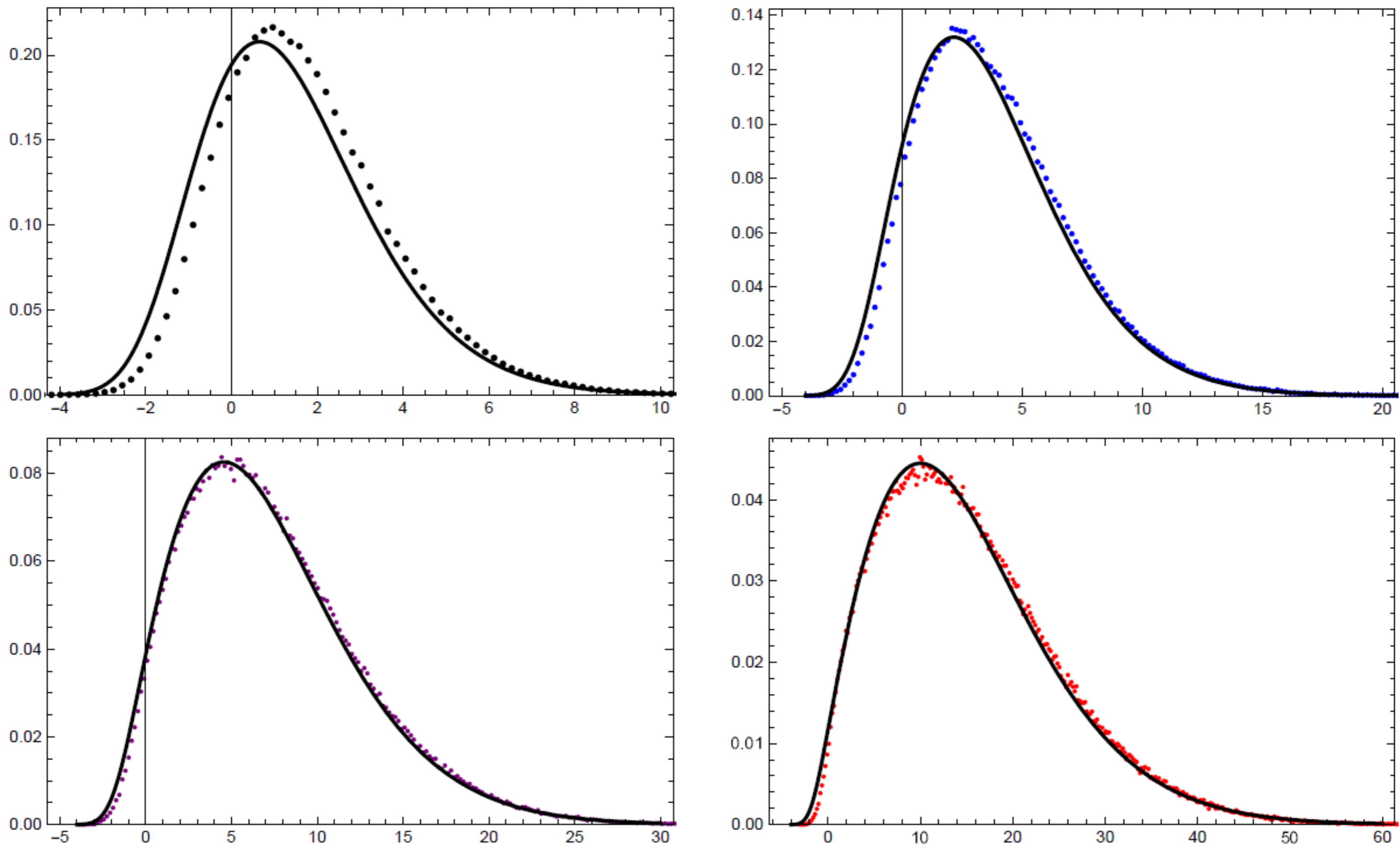


Figure 3: Comparison between the probability densities of $F^{(\sigma)}(s)$ and $F_{\text{appr}}^{(\sigma)}(s)$: top-left is $\sigma = 2.0$, top-right is $\sigma = 3.0$, bottom-left is $\sigma = 4.36$ and bottom-right is $\sigma = 7.0$. The simulation for $\sigma = 2.0$ uses maximal time $t_{\text{max}} = 2000$, while the other $t_{\text{max}} = 3000$. The number of runs are 5×10^5 for each case.

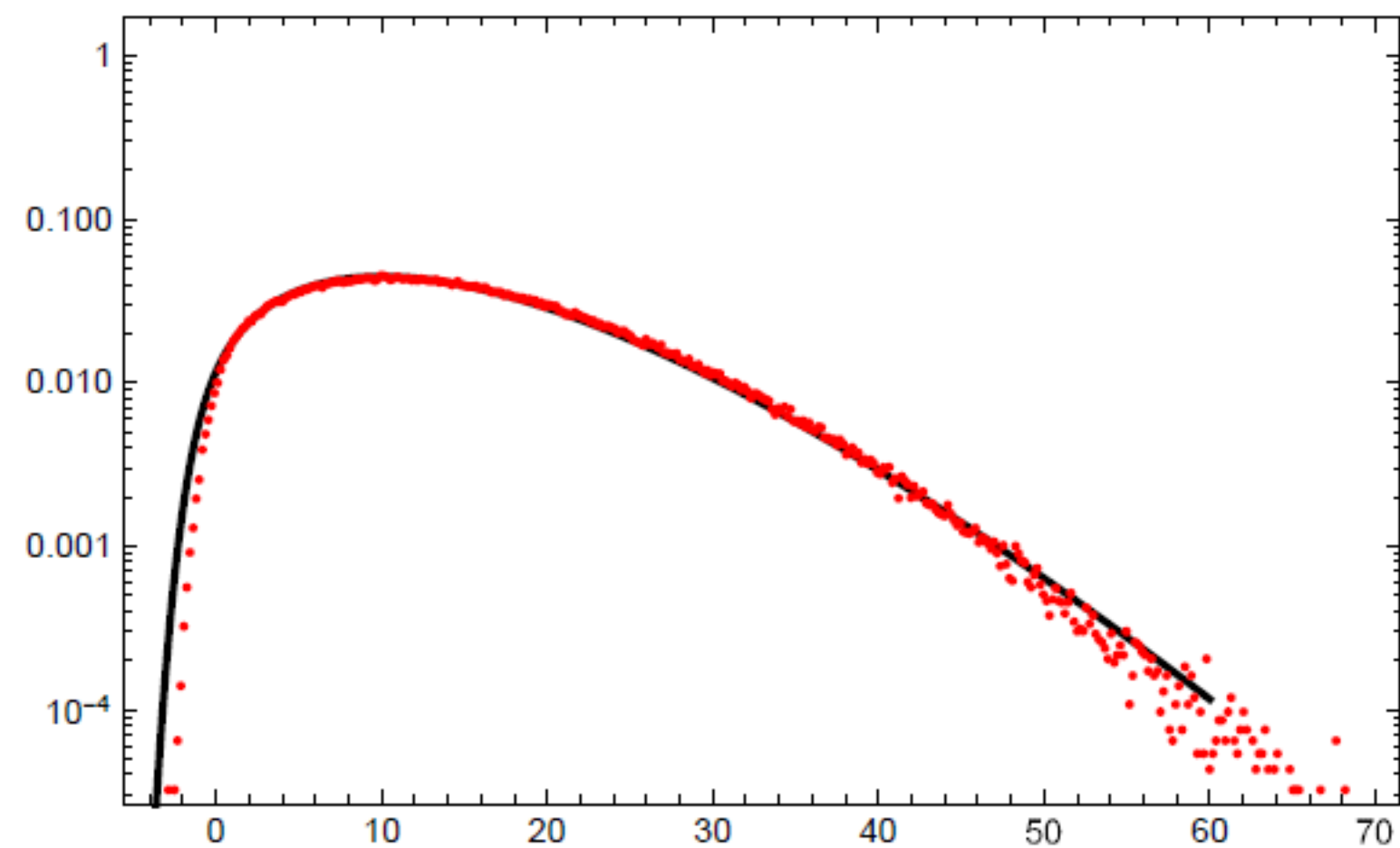
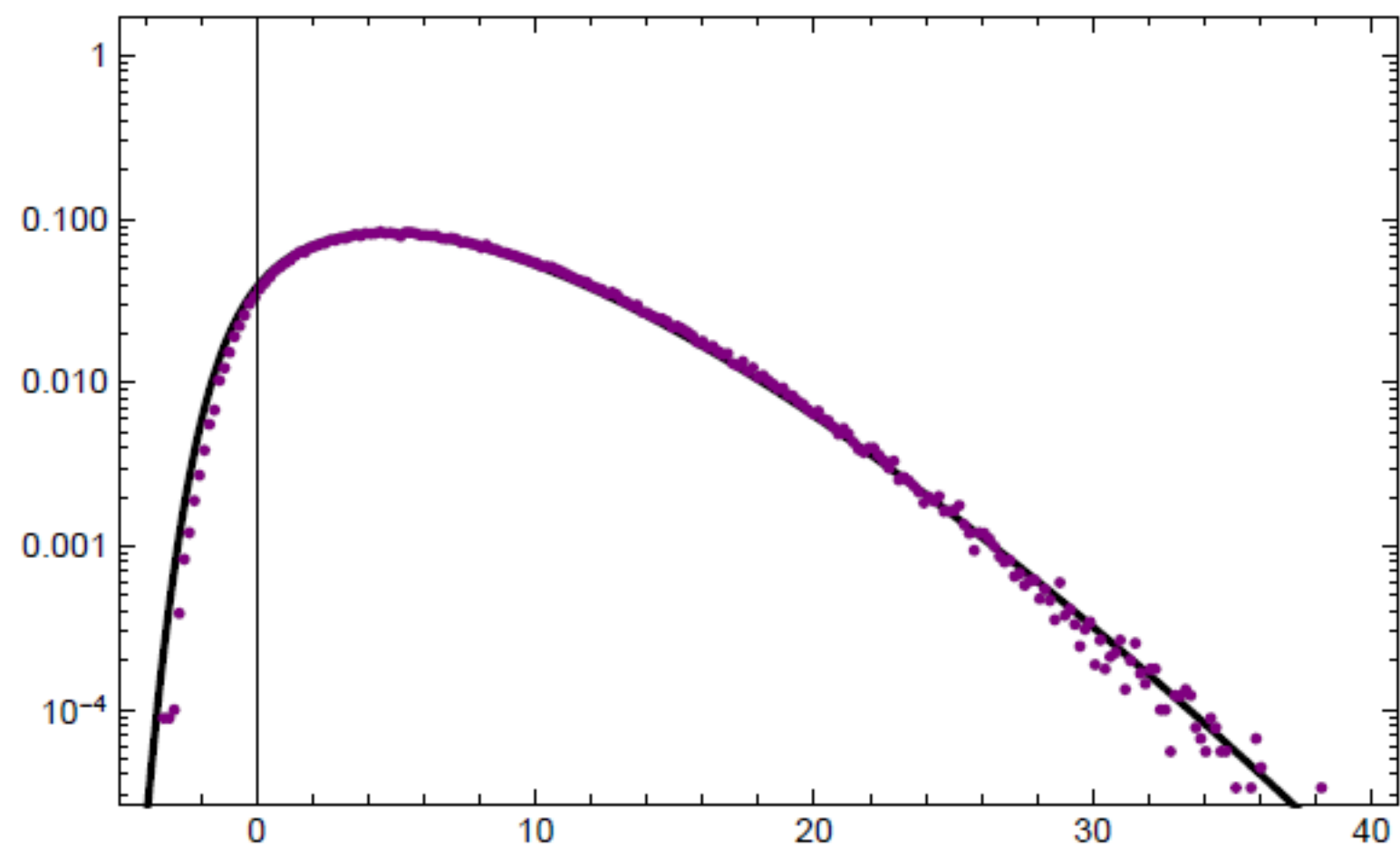
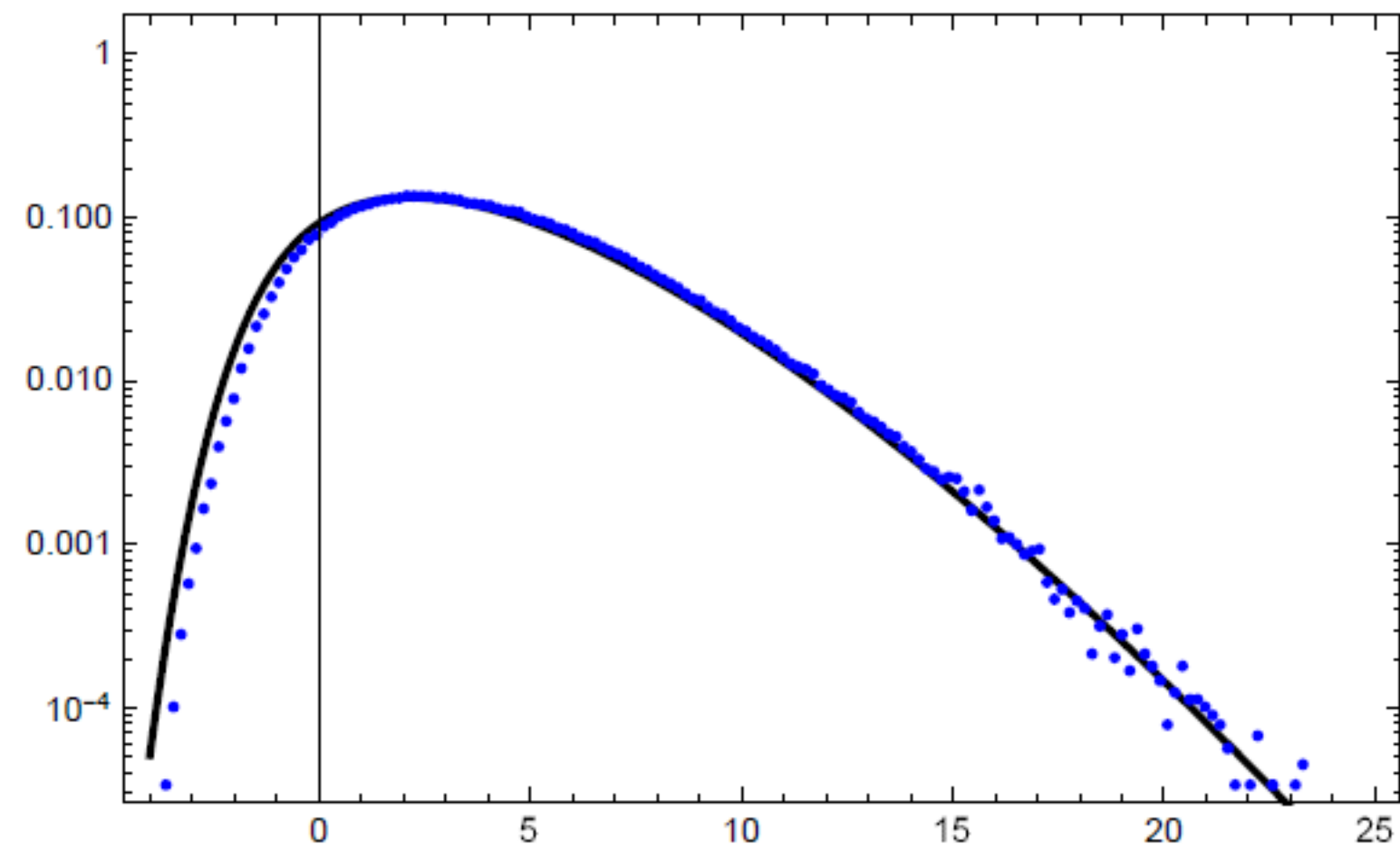
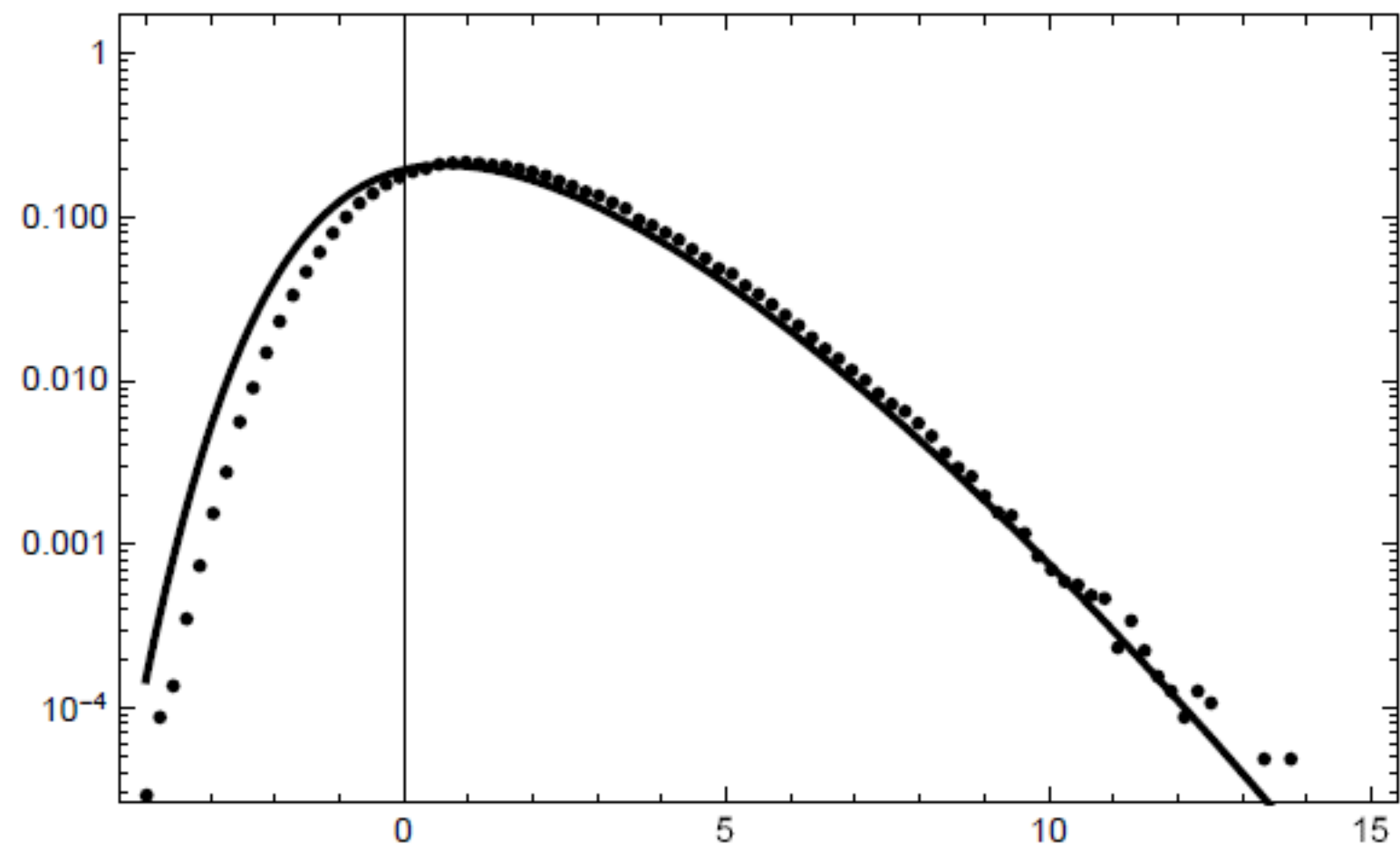
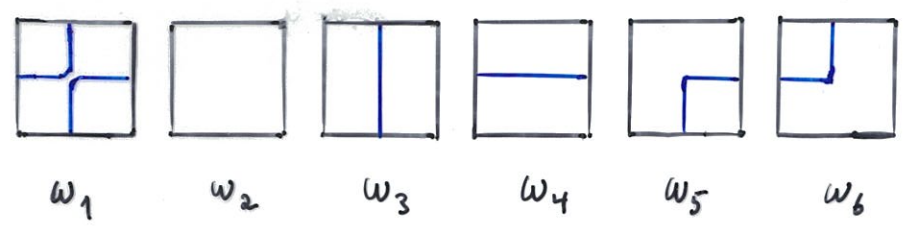


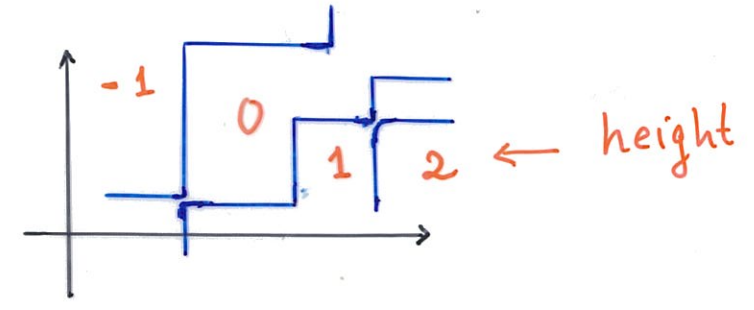
Figure 4: Logarithmic plots of the probability densities of Figure 3.

KPZ point of 6-vertex model



- tiling of the plane
- up-right paths, touching - no crossing

⇒ height function
 // SM of elastic membrane //



free energy

Bethe ansatz

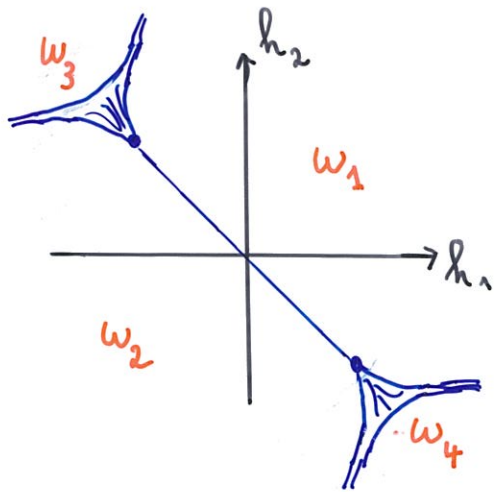
Lieb, Yang+Yang

⇒ ferroelectric ($\Delta > 1$)

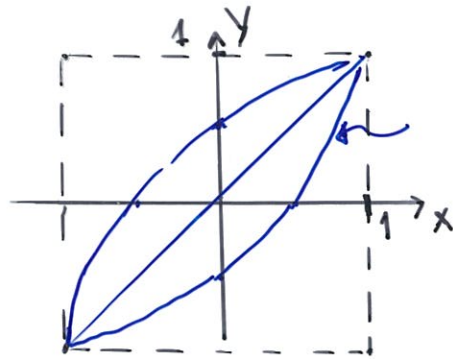
Bukman, Shore 1995

4 independent parameters

2 temperature-like, vertical/horizontal electric field \Leftrightarrow polarization



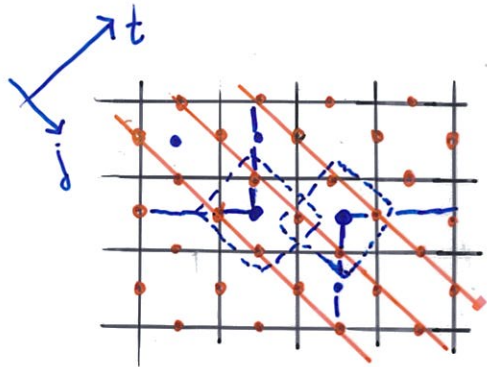
Legendre



extremal Gibbs measures

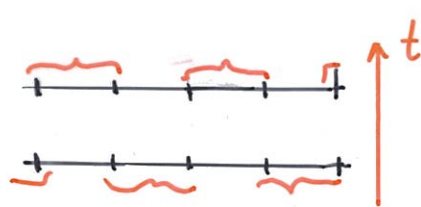
- KPZ point $\vec{w} = (1, 1, q, p, 1-q, 1-p)$ $0 \leq p, q \leq 1$

diagonal transfer matrix



$$\eta_{ij} = \begin{cases} 0 & \text{no path} \\ 1 & \text{path} \end{cases}$$

edge



$$\begin{array}{l} 10 \rightsquigarrow 01 \quad \text{probability } p \\ 01 \rightsquigarrow 10 \quad \text{probability } q \end{array}$$

implies

independent update
for each block

$$\begin{array}{l} 00 \rightsquigarrow 00 \quad 1 \\ 11 \rightsquigarrow 11 \quad 1 \\ 10 \rightsquigarrow 10 \quad 1-p \\ 01 \rightsquigarrow 01 \quad 1-q \end{array}$$

Markov chain on $\{0,1\}^{\mathbb{Z}}$

$$T = T_{\text{odd}} T_{\text{even}}$$

- invariant measure: product period 2 $\mu_{\text{even}}(1) = a, \mu_{\text{odd}}(1) = b$

stationarity $\parallel (1-q)a(1-b) = (1-p)(1-a)b \parallel$

$\eta_{i,j}$ stationary, mixing Markov chain \iff extremal Gibbs measure

- initial measure 2-periodic product general a, b
density $\frac{1}{2}$, implies $v=0$

\Rightarrow non-universal are computed NO adjustable parameters

SET $p = 0.6, q = 0.2, \rho = \frac{1}{2}$

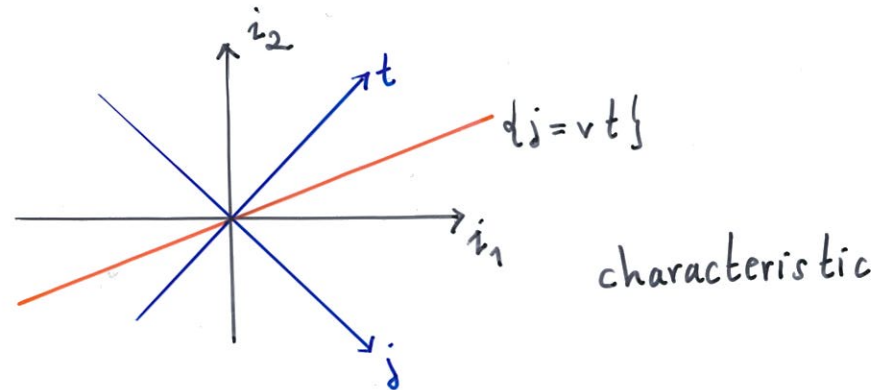
$$\begin{aligned} \sigma &= 1 && \text{stationary} \\ \sigma &= 0.65 \end{aligned}$$

\Rightarrow half-infinite 6-vertex \iff

stationary 6-vertex

Aggarwal 2016

current-density $j(p)$, $v = j'(p)$



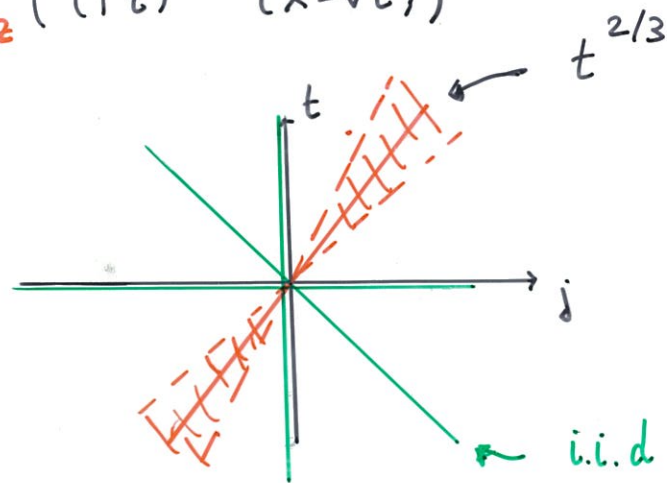
Theorem:

$$h(vt, t) - h(0, 0) \cong v_0 t + (\Gamma t)^{1/3} \zeta_{BR}$$

and $t^{2/3}$ sideways

implies two-point

$$S(j, t) = \langle \eta_{jt} \eta_{00} \rangle_c \cong A (\Gamma t)^{-2/3} f_{KPZ}((\Gamma t)^{-2/3} (x - vt))$$



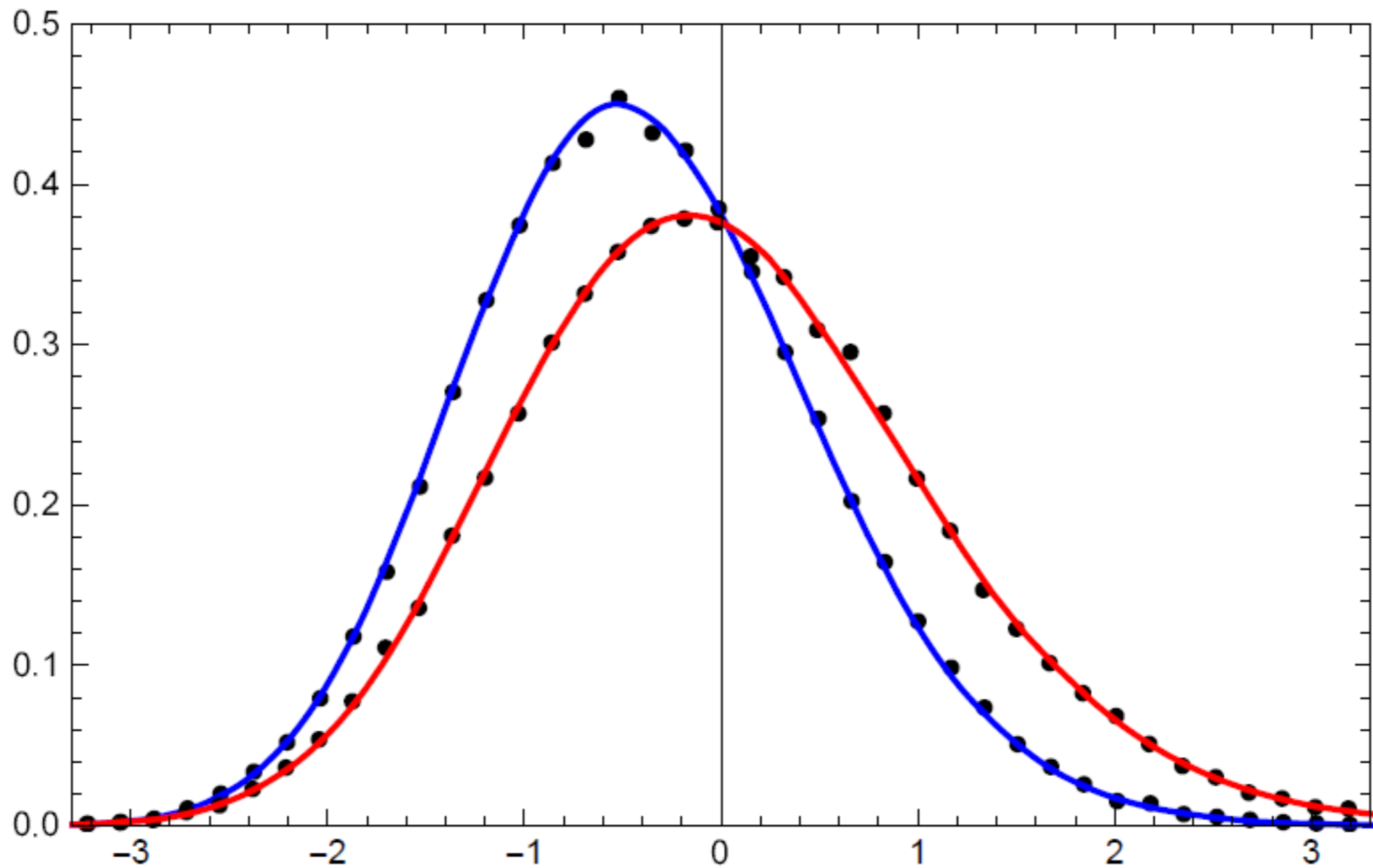


Figure 7: Illustration of the universality: for $\sigma^2 = \alpha/(1 - \alpha)$ with $\alpha = 0.3$ (blue) and the stationary benchmark $\alpha = 0.5$ (red), the dots are the probabilities for the current fluctuations of the stochastic six-vertex model. Here we choose generic values, $p = 0.6$, $q = 0.2$, and particle density $\rho = 1/2$. The data are scaled according to KPZ scaling theory.

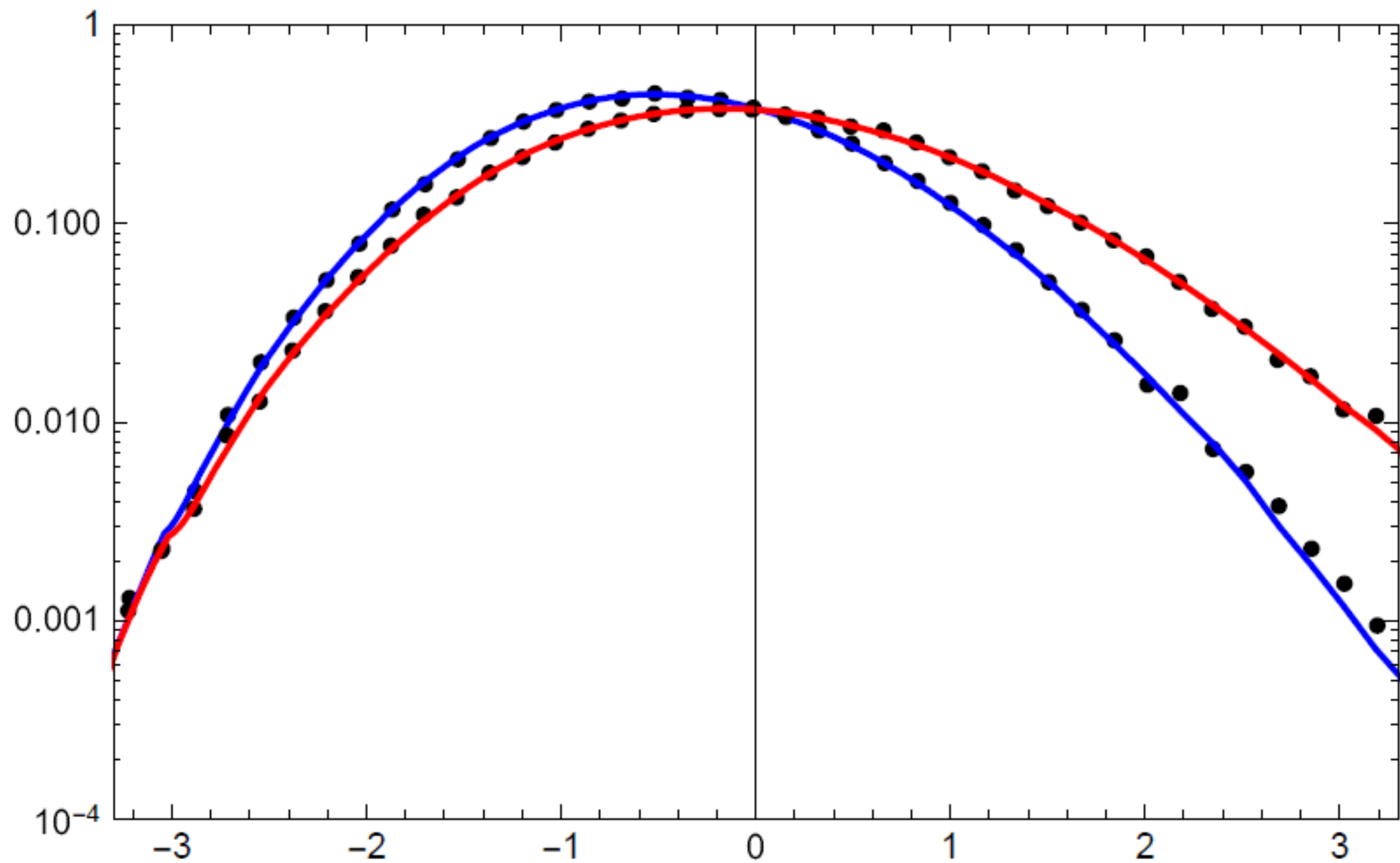


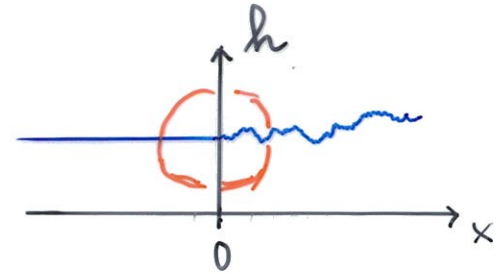
Figure 8: Logarithmic plot of the probability densities of Figure 7.

outlook / discussion

⇒ dependence on initial conditions

so far: flat, curved, stationary

and crossover



HERE: initial slope is stationary

OPEN PROBLEM

|| analytic formula for $F^{(\sigma)}$, $\sigma \neq 0, 1$ ||

KPZ fixed point

$$\lim_{\epsilon \rightarrow 0} \epsilon^{1/2} h(\epsilon^{-1} x, \epsilon^{-3/2} t) = ??$$

variational characterization ?