# A class of (2 + 1)-dimensional growth processes with explicit stationary measure

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- Dimer models (perfect matchings) and height function
- Irreversible dynamics: a (2+1)-d random growth model
- Speed and fluctuations

Perfect matchings of bipartite planar graphs



Perfect matchings of bipartite planar graphs



# Height function



Height function:

$$h(f') - h(f) = 4 \sum_{e \in C_{f \to f'}} \sigma_e(1_{e \in M} - 1/4)$$

where  $\sigma_e = +1/-1$  if *e* crossed with white on the right/left. Definition is path-independent.

# Ergodic Gibbs measures [Kenyon-Okounkov-Sheffield]

 Choose ρ = (ρ<sub>1</sub>, ρ<sub>2</sub>, ρ<sub>3</sub>) with ρ<sub>i</sub> ∈ (0, 1), ρ<sub>1</sub> + ρ<sub>2</sub> + ρ<sub>3</sub> = 1. There exists a unique translation invariant, ergodic Gibbs measure π<sub>ρ</sub> s.t. the density of horizontal, NW and NE lozenges are ρ<sub>1</sub>, ρ<sub>2</sub>, ρ<sub>3</sub>.

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- Dimer-dimer correlations decay algebraically:

$$\pi_{\rho}(1_{e\in M};1_{e'\in M})\approx |e-e'|^{-2}$$

• height function converges to GFF: if  $\int_{\mathbb{R}^2} \varphi(x) dx = 0$  then

$$\epsilon^2 \sum_{x} \varphi(\epsilon x) h_x \xrightarrow{\epsilon \to 0} \int \varphi(x) X(x) dx$$

with  $\langle X(x)X(y)\rangle = -\frac{1}{2\pi^2}\log|x-y|$ .

## Symmetric vs. asymmetric random dynamics



For d = 1: Symmetric vs. Asymmetric Simple Exclusion Process



In both SSEP/ASEP, Bernoulli( $\rho$ ) are invariant. For  $p \neq q$ , irreversibility (particle flux).







Asymmetric cube deposition/evaporation dynamics

- If p = q, Gibbs states are invariant (no surprise; reversibility)
- if  $p \neq q$ , stationary states presumably very different from  $\pi_{\rho}$ . Numerical simulations [Forrest-Tang-Wolf 1992] show  $\approx t^{0.24...}$  growth of height fluctuations.

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- large-scale dynamics should be described by "isotropic two-dimensional KPZ equation":

$$\partial_t h = \nu \Delta h + Q(\nabla h) +$$
white noise

with Q a positive-definite quadratic form (whatever mathematical sense this equation has...)

### Coupled simple exclusions with constraints



A two-dimensional generalization of Hammersley process



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For some slopes  $\rho$  (technical restrictions) I can actually prove better:

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#### • Generalization to domino tilings

### Comments

• A. Borodin, P. L. Ferrari [BF '08] study totally asymmetric case (q = 1, p = 0) and special (and deterministic) initial condition.

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Exact computations (explicit kernel for some time-space correlations)

 large-scale dynamics should be described by "anisotropic two-dimensional KPZ equation":

 $\partial_t h = \nu \Delta h + Q(\nabla h) +$ white noise

with Q a (+, -)-definite quadratic form. Physics literature [Wolf '91]: non-linearity irrelevant.

### Comments

• BF '08 obtain hydrodynamic limit and  $\sqrt{\log t}$  Gaussian fluctuations

$$\lim_{L\to\infty}\frac{1}{L}\mathbb{E}h(xL,yL,\tau L)=\mathbf{h}(x,y,\tau)$$

with

$$\partial_{\tau}\mathbf{h} = \mathbf{v}(\nabla\mathbf{h})$$

and

$$\frac{1}{\sqrt{\log L}}[h(xL, yL, \tau L) - \mathbb{E}(h(xL, yL, \tau L))] \Rightarrow \mathcal{N}(0, \sigma^2);$$

moreover, convergence of local statistics to that of a Gibbs measure.

#### Invariance on the torus

For simplicity, q = 1, p = 0. Stationary measure  $\pi_{\rho}^{L}$ : uniform measure with fraction  $\rho_{i}$  of lozenges of type i = 1, 2, 3.

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### From the torus to the infinite graph

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Key fact:

**Lemma**: The probability of seeing an inter-particle gap  $\geq \log R$  within distance R from the origin before time 1 is  $O(R^{-K})$  for every K.

# Comparison with the Hammersley process (HP)



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Lozenge dynamics  $\sim$  infinite set of coupled Hammersley processes. Comparison: lozenges move less than HP particles





Let 
$$Q_{\Lambda}(t) = \sum_{x \in \Lambda} (h_x(t) - h_x(0)).$$
  
 $\frac{d}{dt} \langle Q_{\Lambda}(t) \rangle = \langle K_{\Lambda}(\sigma_t) \rangle := \langle \sum_x |V(x,\uparrow) \cap \Lambda|(t) \rangle = v|\Lambda|$ 

Similarly, one can prove

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angle &= 2 \langle (Q_{\Lambda}(t) - \langle Q_{\Lambda}(t) 
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$$\begin{split} \frac{d}{dt} \langle (Q_{\Lambda}(t) - \langle Q_{\Lambda}(t) \rangle)^2 \rangle &= 2 \langle (Q_{\Lambda}(t) - \langle Q_{\Lambda}(t) \rangle) (K_{\Lambda}(\sigma_t) - \pi_{\rho}(K_{\Lambda})) \rangle \\ &+ \pi_{\rho} (\sum_{x} |V(x,\uparrow) \cap \Lambda|^2) \\ &\leq 2 \sqrt{\langle (Q_{\Lambda}(t) - \langle Q_{\Lambda}(t) \rangle)^2 \rangle} \sqrt{Var_{\pi_{\rho}}(K_1)} + O(L^2) \end{split}$$

Equilibrium estimate:

 $Var_{\pi_{\rho}}(K_1) = O(L^{2+\delta})$  or  $= O(L^2 \log L)$  for some slopes.

#### Therefore,

$$\frac{d}{dt}\langle (Q_{\Lambda}(t)-\langle Q_{\Lambda}(t)\rangle)^2\rangle \leq 2\sqrt{\langle (Q_{\Lambda}(t)-\langle Q_{\Lambda}(t)\rangle)^2\rangle}L^{1+\delta}+O(L^2)$$

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If L = 1, we get the (useless) bound  $\psi(T) = O(T)$ . If we choose L = T we get instead  $\psi(T) = O(T^{\delta})$  as wished.

#### Thanks!