Critical percolation under conservative dynamics

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Joint work with Erik Broman (Uppsala University)
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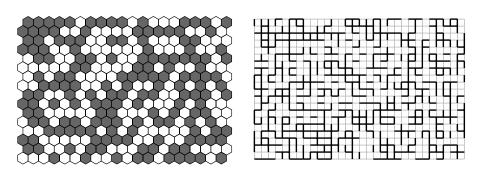
PASI conference, Buenos Aires, January 2012

Overview

- Dynamical percolation
- Conservative dynamics on percolation

"standard" dynamical percolation

Start with an initial configuration $\omega_{t=0}$ at $p = p_c(\mathbb{T}) = p_c(\mathbb{T}^2) = 1/2$.



And let evolve each edge (or site) **independently** at rate 1. This gives a **Markov** process $(\omega_t)_{t>0}$ on critical percolation configurations.

Main results known

Theorem (Schramm, Steif, 2005)

On the triangular lattice \mathbb{T} , there exist exceptional times t for which $0 \stackrel{\omega_t}{\longleftrightarrow} \infty$. Furthermore, a.s.

$$\dim_{\mathcal{H}}(\mathbf{Exc}) \in \left[\frac{1}{6}, \frac{31}{36}\right]$$

Main results (continued)

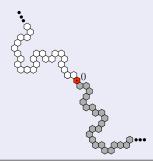
Theorem (G., Pete, Schramm, 2008)

• On the square lattice \mathbb{Z}^2 , there are exceptional times as well (with $\dim_{\mathcal{H}}(\mathbf{Exc}) \geq \epsilon > 0$ a.s.)

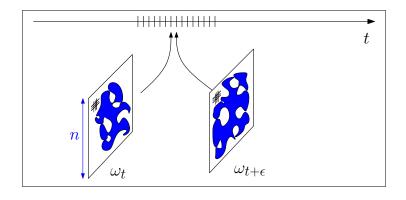
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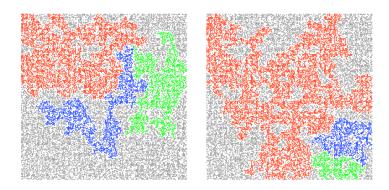
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- On the triangular lattice \mathbb{T}
 - a.s. $\dim_{\mathcal{H}}(\operatorname{Exc}) = \frac{31}{36}$
 - There exist exceptional times Exc⁽²⁾ such that



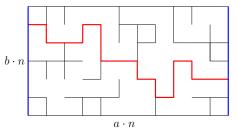
Strategy: noise sensitivity of percolation



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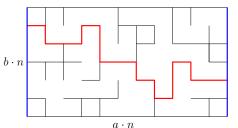
Large scale properties are encoded by Boolean functions of the 'inputs'



Let $f_n: \{-1,1\}^{O(1)n^2} \to \{0,1\}$ be the Boolean function defined as follows

$$f_n(\omega) := \left\{ egin{array}{ll} 1 & ext{if left-right crossing} \\ 0 & ext{else} \end{array}
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Theorem (Benjamini, Kalai, Schramm, 1998)

For any fixed t > 0:

$$\operatorname{Cov}\left[f_n(\omega_0), f_n(\omega_t)\right] \underset{n \to \infty}{\longrightarrow} 0$$

We say in such a case that $(f_n)_{n\geq 1}$ is noise sensitive.

Main tool to study noise sensitivity: Fourier analysis

Decompose $f: \{-1,1\}^m \to \{0,1\}$ into "Fourier" series

$$f(\omega) = \sum_{S} \hat{f}(S) \chi_{S}(\omega),$$

where
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where $\chi_S(x_1,\ldots,x_m):=\prod_{i\in S}x_i$.

$$\mathbb{E}[f(\omega_0) f(\omega_t)] = \mathbb{E}[\left(\sum_{S_1} \widehat{f}(S_1) \chi_{S_1}(\omega_0)\right) \left(\sum_{S_2} \widehat{f}(S_2) \chi_{S_2}(\omega_t)\right)]$$

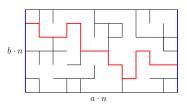
$$= \sum_{S} \widehat{f}(S)^2 \mathbb{E}[\chi_{S}(\omega_0) \chi_{S}(\omega_t)]$$

$$= \sum_{S} \widehat{f}(S)^2 e^{-t|S|}$$

Thus the covariance can be written:

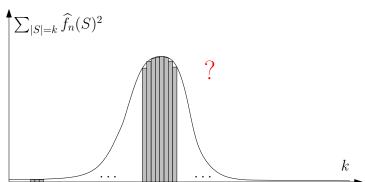
$$\mathbb{E}\big[f(\omega_0)\,f(\omega_t)\big] - \mathbb{E}\big[f(\omega)\big]^2 = \sum_{S\neq\emptyset} \widehat{f}(S)^2\,e^{-t\,|S|}$$

Fourier spectrum of critical percolation

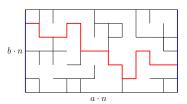


Let $f_n, n \ge 1$ be Boolean functions defined above.

One is interested in the shape of their Fourier spectrum.

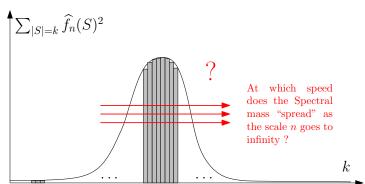


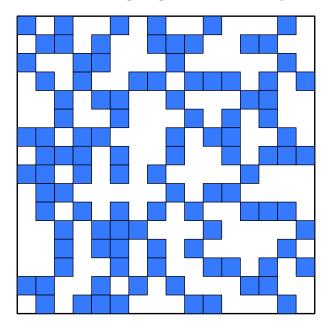
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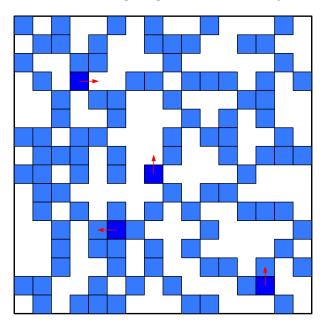


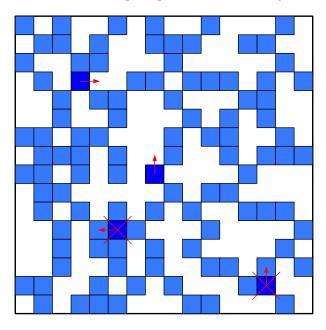
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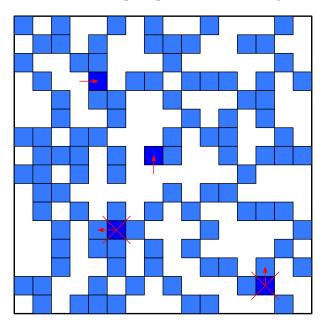
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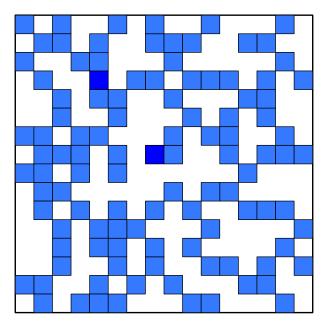












The system evolves according to the symmetric exclusion process

Let $(\omega_t^P)_{t\geq 0}$ be a sample of a symmetric exclusion process with symmetric kernel P(x,y), $(x,y)\in\mathbb{Z}^2\times\mathbb{Z}^2$ or $(x,y)\in\mathbb{T}\times\mathbb{T}$

We distinguish 2 cases:

(a) Nearest neighbor dynamics:

$$P(x,y) = \frac{1}{\text{degree}} \, 1_{x \sim y}$$

(b) Medium-range dynamics:

$$P(x,y) \approx \frac{1}{\|x-y\|^{2+\alpha}}$$
 for some exponent $\alpha > 0$

What we can and cannot :-(prove about these dynamics

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- 1. Let's start with the bad news: we don't know if there are exceptional times for ω_t^P .
- 2. If the dynamics is **medium-range** with exponent $\alpha > 0$ (recall $P(x,y) \asymp \|x-y\|^{-2-\alpha}$), then we get quantitative bounds on the noise sensitivity of the crossing events f_n under ω_t^P . More precisely:

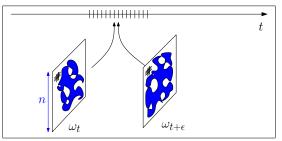
Theorem (Broman, G., Steif, 2011)

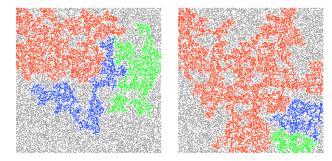
If P is any transition kernel with exponent $\alpha > 0$, then on $\mathbb{Z}^{2, \text{site}}$, $\mathbb{Z}^{2, \text{bond}}$ or \mathbb{T} , at the critical point, one has

$$\operatorname{Cov}(f_n(\omega_0^P), f_n(\omega_t^P)) \xrightarrow[n \to \infty]{} 0$$

Furthermore, one can choose $t = t_n \ge n^{-\beta(\alpha)}$.

In other words, for medium-range exclusion dynamics $(\alpha > 0)$, we also obtain this "picture"





Which approach for this problem ?

Two strategies:

- 1. Either the noise sensitivity results for the iid case transfer to these conservative dynamics ?
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strategy 1. is "hopeless" since

Fact

There exist Boolean functions $(f_n)_n$ which are highly noise sensitive to i.i.d. noise but which are stable to symmetric exclusion P- dynamics.

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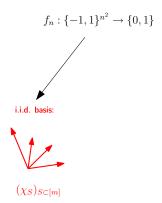
But there are difficulties:

- 1. In the finite volume case, such a basis obviously exists, but it highly depends on *P* and it is not very "explicit".
- In the infinite volume case, L_P is of course non-compact and it seems that it does not have pure-point spectrum.

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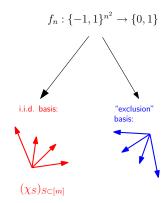
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We decompose f on the classical "i.i.d." basis even though it does not diagonalize our exclusion process:

$$\mathbb{E}[f(\omega_0^P) f(\omega_t^P)] = \mathbb{E}[\left(\sum_{S} \widehat{f}(S) \chi_S(\omega_0^P)\right) \left(\sum_{S'} \widehat{f}(S') \chi_{S'}(\omega_t^P)\right)]$$

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$$= \sum_{|S|=|S'|} \widehat{f}(S) \widehat{f}(S') \mathbf{P_t}(S,S')$$

where $P_t(S, S')$ is the probability that the set S travels in time t towards the set S' under the exclusion process.

$$\mathbb{E}\left[f_{n}(\omega_{0}^{P}) f_{n}(\omega_{t}^{P})\right] = \sum_{|S|=|S'|} \widehat{f}_{n}(S) \widehat{f}_{n}(S') \mathbf{P_{t}(S,S')}$$

$$" = " \left\langle \widehat{f}_{n}, P_{t} \star \widehat{f}_{n} \right\rangle$$

We would like to prove that for large scale n, the vectors $\{\hat{f}_n(S)\}_S$ and $\{P_t \star \hat{f}_n(S)\}_S$ are almost orthogonal.

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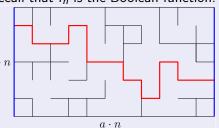
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Unfortunately, we know much more on the vector $\{\hat{f}_n(S)^2\}$ than on $\{\hat{f}_n(S)\}$:

The spectral measure ν_{f_n}

Definition

Recall that f_n is the Boolean function:



Define the spectral measure of f_n as follows:

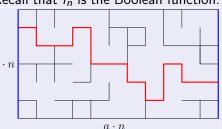
$$\nu_{f_n}(\mathscr{S}=S):=\hat{f}_n(S)^2$$

In particular \mathscr{S} can be considered as a **random** subset of $[0, an] \times [0, bn]$.

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We can prove the following:

Proposition (asymptotic singularity)

For any medium-range exponent $\alpha > 0$ and any fixed t > 0: as $n \to \infty$, the measures ν_{f_n} and $P_t \star \nu_{f_n}$ are asymptotically mutually singular

Why does this imply noise sensitivity?

Fact

• If ϕ^2 and ψ^2 are the densities of two probability measures on $\mathbb R$, then

$$\int \phi \ \psi \le 2 \sqrt{\int \phi^2 \wedge \psi^2}$$

• In particular, if the two corresponding probability measures are almost singular with respect to each other then $\int \phi \psi$ is small.

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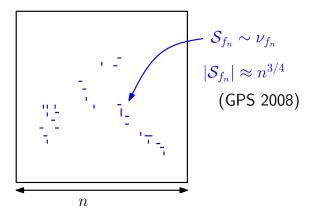
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Take
$$\begin{cases} \phi^2 \equiv \hat{f}_n(S)^2 \\ \psi^2 \equiv P_t \Big[(\hat{f}_n)^2 \Big] (S) \end{cases}$$
 this gives that
$$\left\langle \sqrt{\hat{f}_n(S)^2}, \sqrt{P_t \Big[(\hat{f}_n)^2 \Big] (S)} \right\rangle \text{ is small.}$$

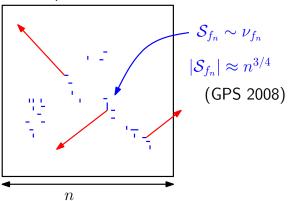
By Cauchy-Schwartz, one concludes that $\langle \hat{f}_n(S), P_t \star \hat{f}_n(S) \rangle$ is small.

Singularity in the medium-range case $(\alpha > 0)$ (Recall $P(x,y) \simeq \frac{1}{\|x-y\|^{2+\alpha}}$)



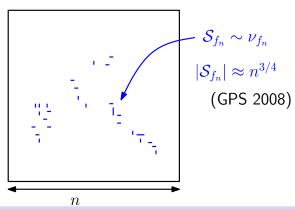
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If the exponent α is large...



Question

What about the nearest-neighbor case ?