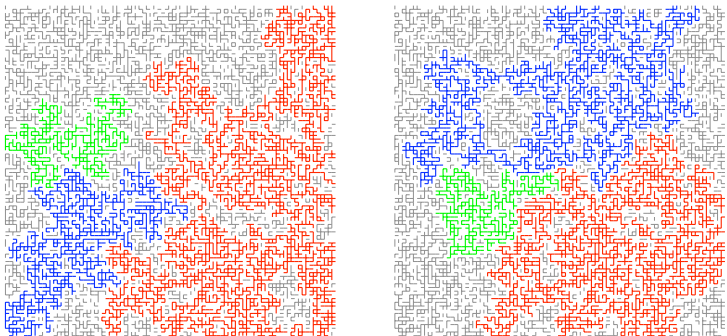


Lectures on noise sensitivity and percolation

Christophe Garban and Jeffrey E. Steif



Clay summer school, Buzios 2010

Boolean functions

Definition

A **Boolean function** is a function

$$f : \{-1, 1\}^n \rightarrow \{0, 1\} \text{ OR } \{-1, 1\}$$

Boolean functions

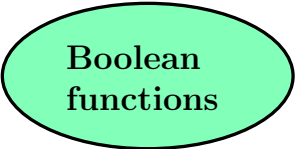
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Example: **Majority**

$$f(x_1, \dots, x_n) = \text{sign}\left(\sum_{i=1}^n x_i\right)$$



**Boolean
functions**

computer sci-
ence



**Boolean
functions**

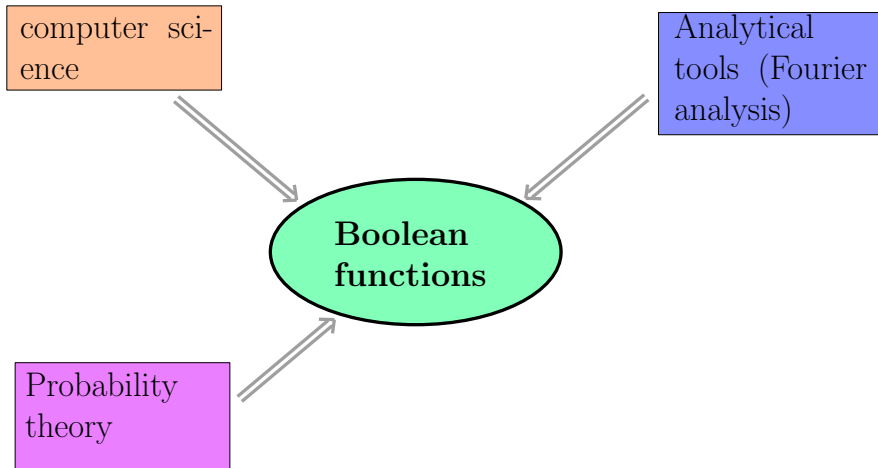
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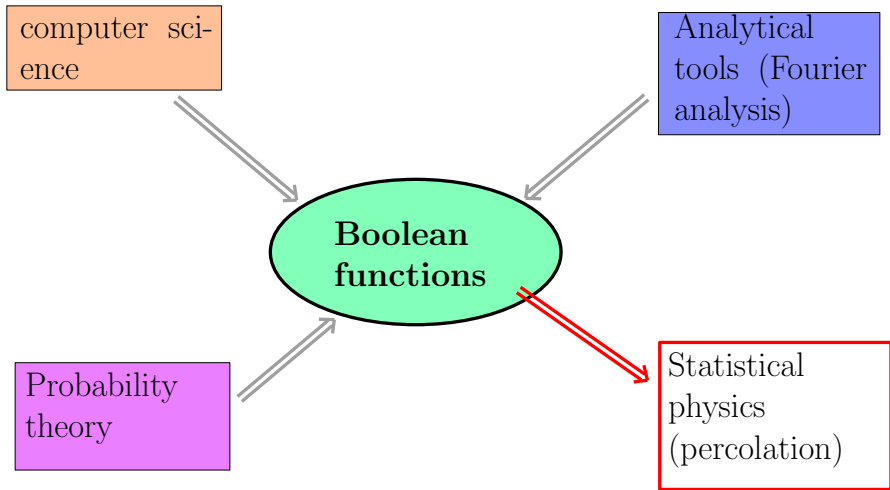


**Boolean
functions**

Probability
theory







A concrete situation : VOTING SCHEMES

Imagine one has n people labelled $1, \dots, n$ which are deciding between candidates A and B according to a certain procedure or *voting scheme*. This procedure can be represented by a Boolean function

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For instance, you may think of

$$\left\{ \begin{array}{l} A = \text{Al Gore} \\ B = \text{Bush} \\ n \approx 10^8 \end{array} \right.$$

Noise stability

Suppose the election is “well-balanced” between A and B . One may thus consider the actual configuration of votes as a **random**

$$\omega = (x_1, \dots, x_n) \in \{-1, 1\}^n,$$

sampled according to the **uniform measure**. The outcome of the election should be $f(\omega)$.

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In fact due to inevitable errors in the recording of the votes, the outcome is $f(\omega^\epsilon)$ instead. Here ω^ϵ is a “slight **perturbation**” of ω .

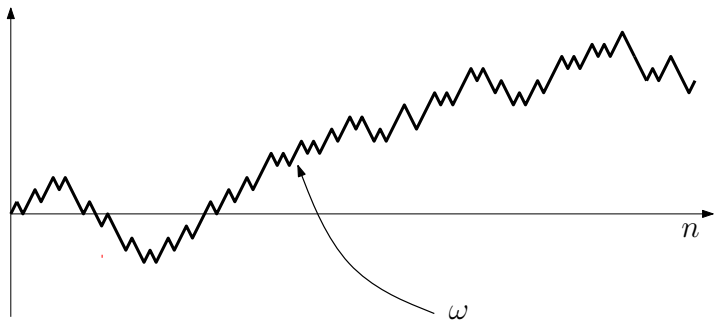
Informal definition

Noise stability corresponds to

$$\mathbb{P}[f(\omega) \neq f(\omega^\epsilon)] \text{ being “small”}.$$

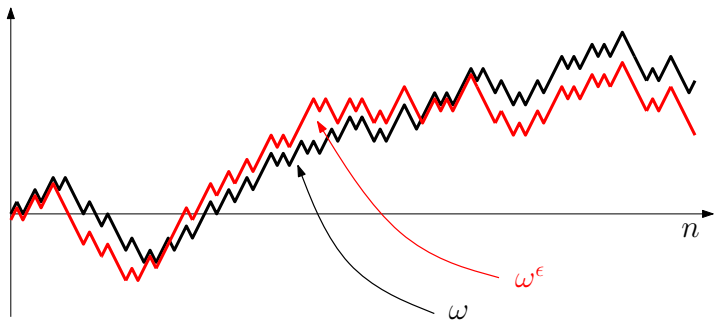
Case of the majority function

If $f(\omega) = \text{sign}(\sum x_i)$,



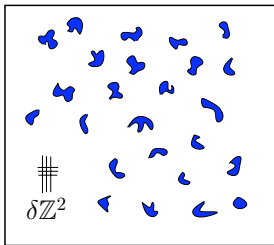
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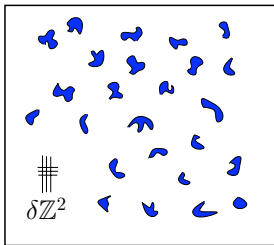
Percolation

Sub-critical ($p < p_c$)

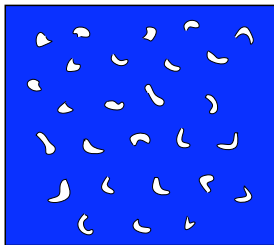


Percolation

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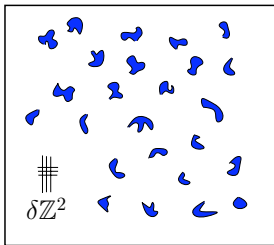


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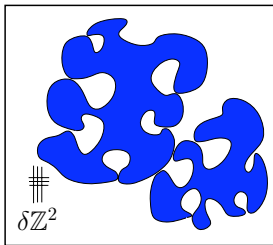


Percolation

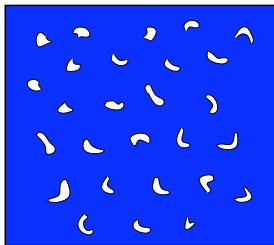
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Critical (p_c)



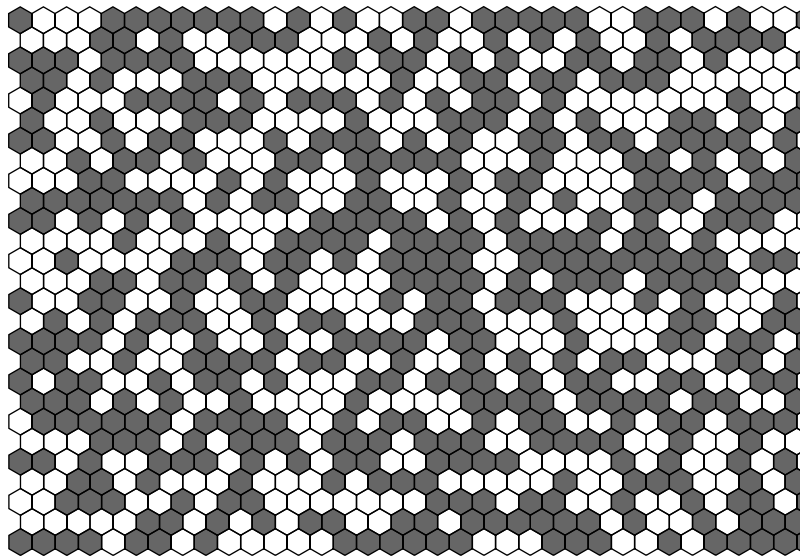
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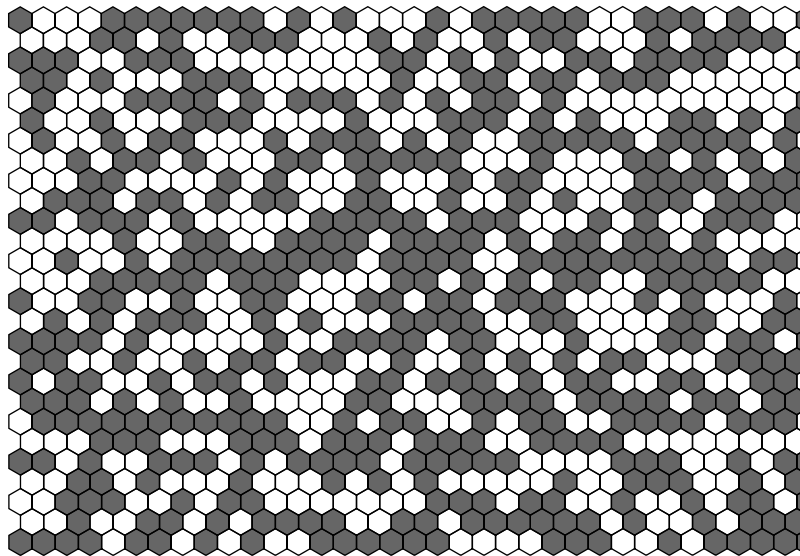
Question

How does critical percolation “react” to perturbations ?

ω :

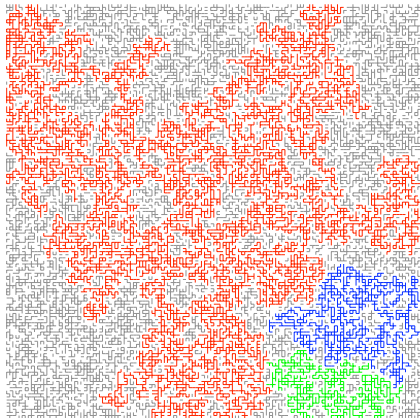
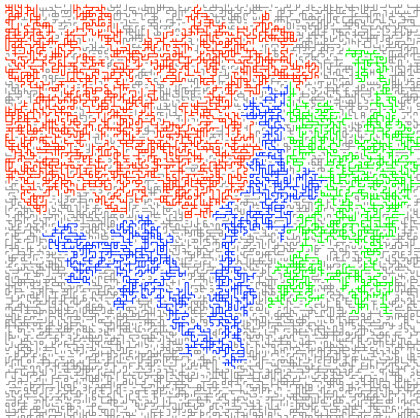


$\omega \rightarrow \omega^\epsilon$:



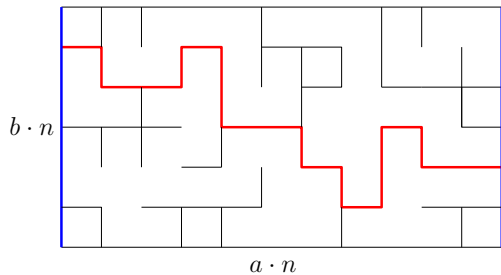
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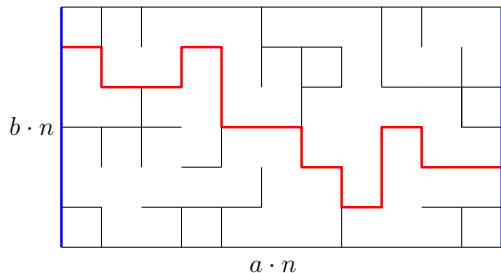


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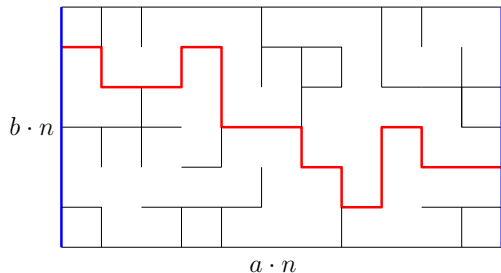
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Let $f_n : \{-1, 1\}^{O(1)n^2} \rightarrow \{0, 1\}$
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Informal definition

Noise sensitivity corresponds to $f_n(\omega)$ and $f_n(\omega^\epsilon)$ being very little correlated (i.e. $\text{Cov}(f_n(\omega), f_n(\omega^\epsilon))$ being very small).

Applications to dynamical percolation

Informal definition

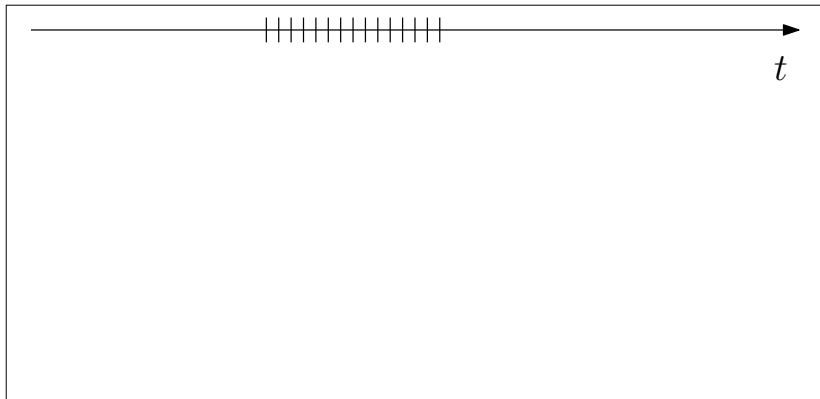
*This is a very simple (stationary) **dynamics** on percolation configurations.*

Applications to dynamical percolation

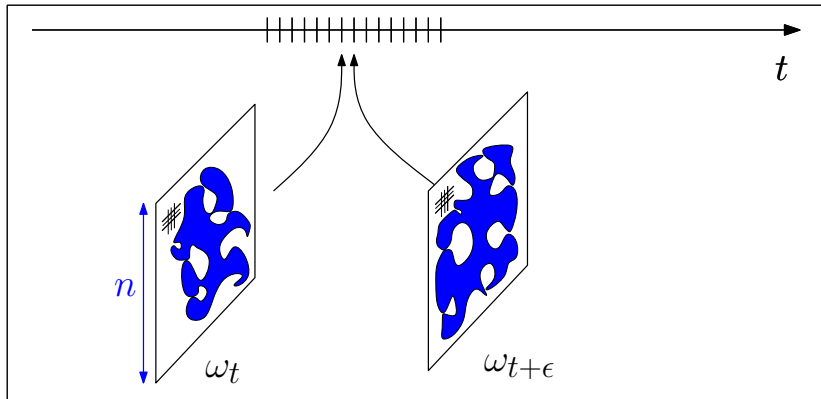
Informal definition

*This is a very simple (stationary) **dynamics** on percolation configurations. Each hexagon (or edge) switches color at the times of a Poisson Point Process.*

How is it related to Noise Sensitivity ?



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Applications to Sub-Gaussian fluctuations

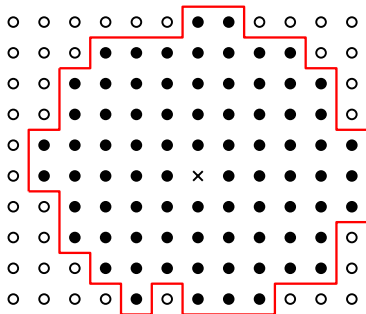
Informal definition (First Passage Percolation)

Let $0 < a < b$. Define the **random metric** on the graph \mathbb{Z}^d as follows: for each edge $e \in \mathbb{E}^d$, fix its length τ_e to be a with probability $1/2$ and b with probability $1/2$.

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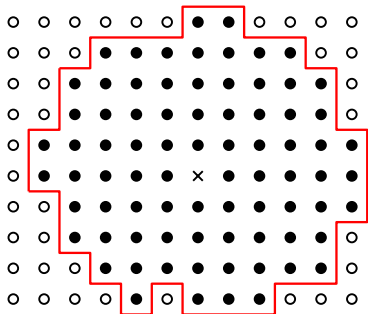
$$B_\omega(R) := \{x \in \mathbb{Z}^d, \text{dist}_\omega(0, x) \leq R\}$$

has an **asymptotic shape**.

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Question

What are the fluctuations around this asymptotic shape ?

What will be our main tools ?

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Fact

f being **noise sensitive** will correspond to f being of “High frequency”.

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- Some concepts which arised in computer science: **influence** of a variable, etc
- **Discrete Fourier analysis**
- **Hypercontractivity**
- **Randomized algorithms**
- Viewing the “frequencies of percolation” as **random fractals** of the plane.