

# Ergodicité de certains automates cellulaires bruités

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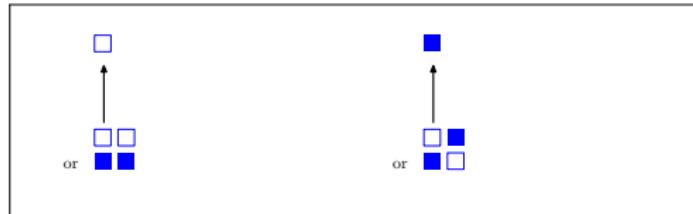
Séminaire Flajolet, Jeudi 1er juin 2017



# The XOR Cellular Automaton

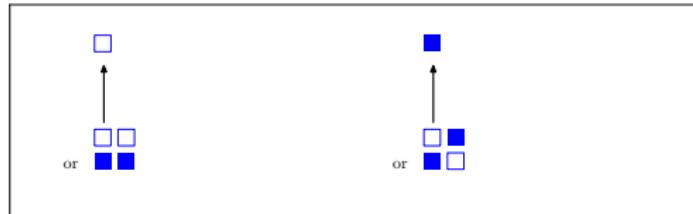
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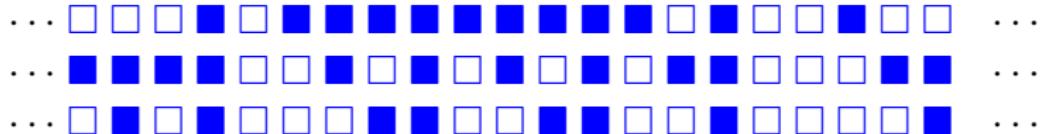
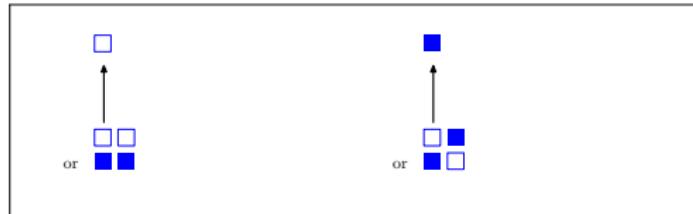
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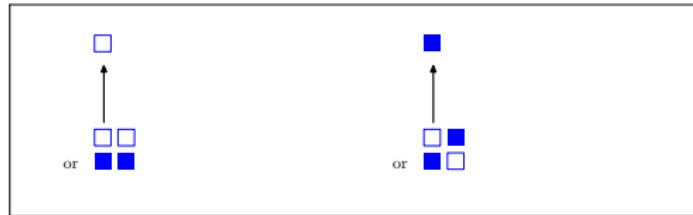


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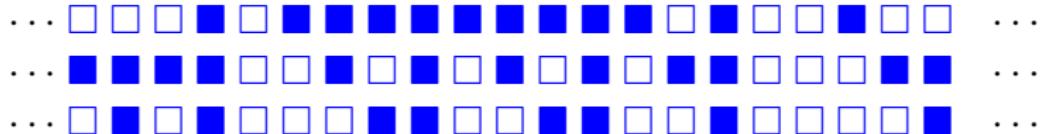
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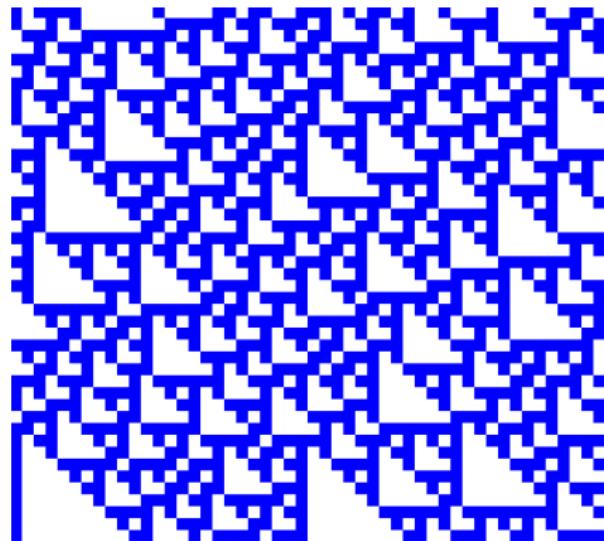
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Additive CA (XOR)

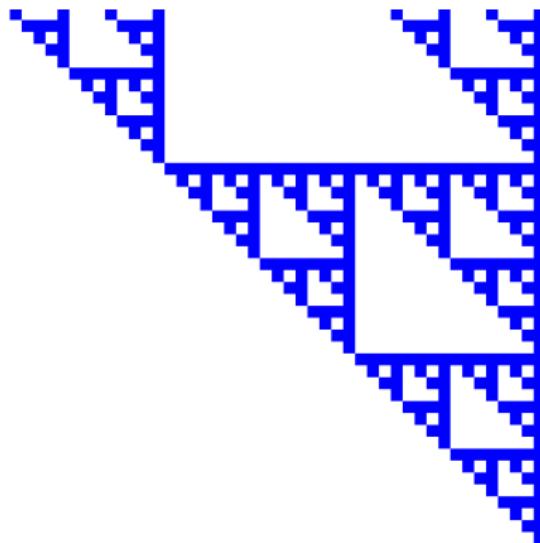


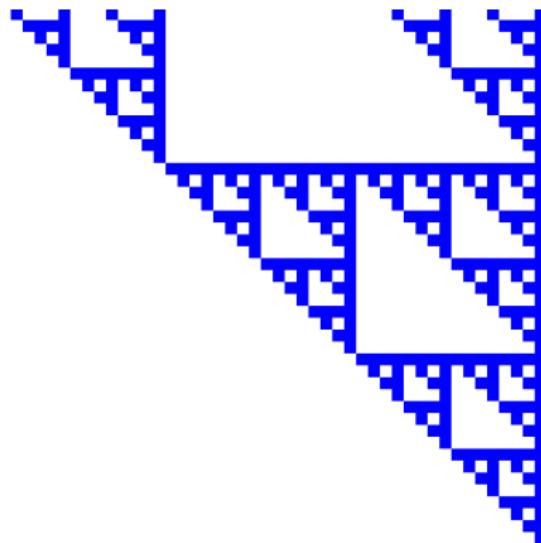
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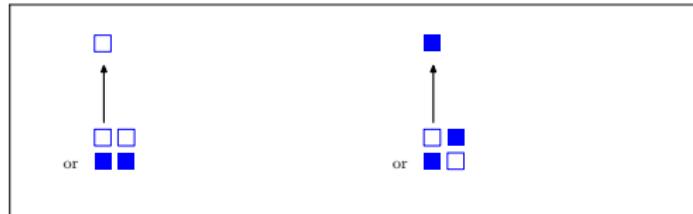
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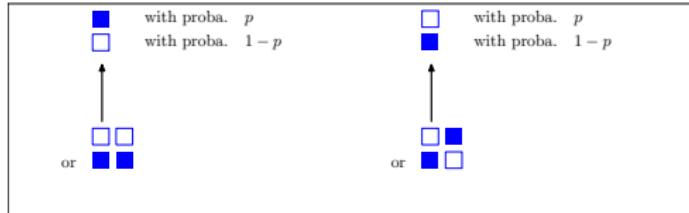
The system keeps some information on its initial configuration.

# The XOR Cellular Automaton



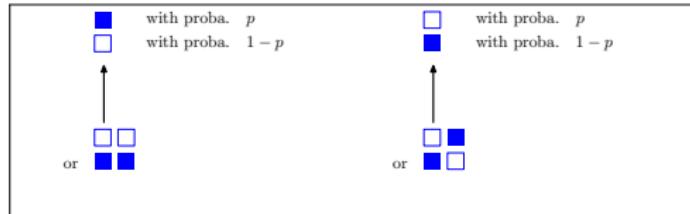
$\dots \square \dots$

# The XOR Cellular Automaton



$\dots \square \blacksquare \dots$

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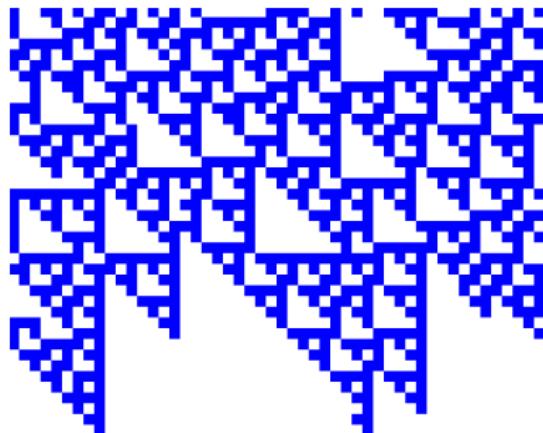
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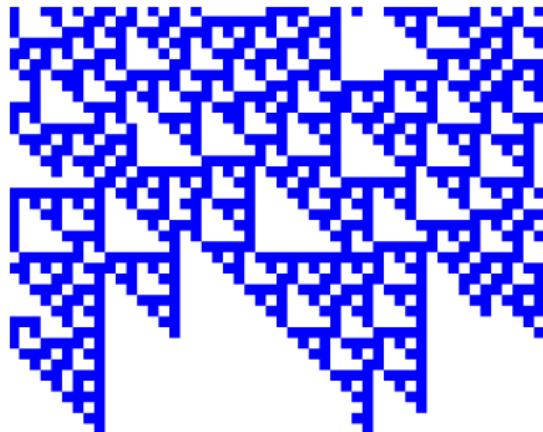
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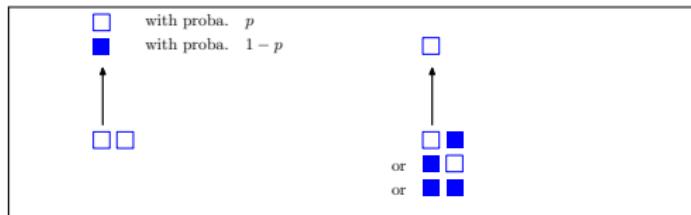


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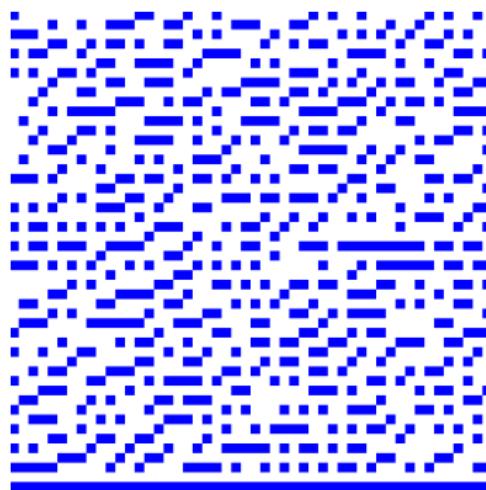
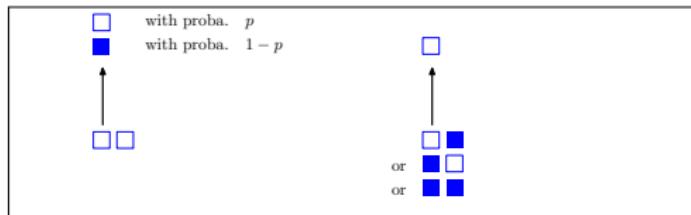


The system **forgets** its initial configuration.

# The noisy hardcore Cellular Automaton

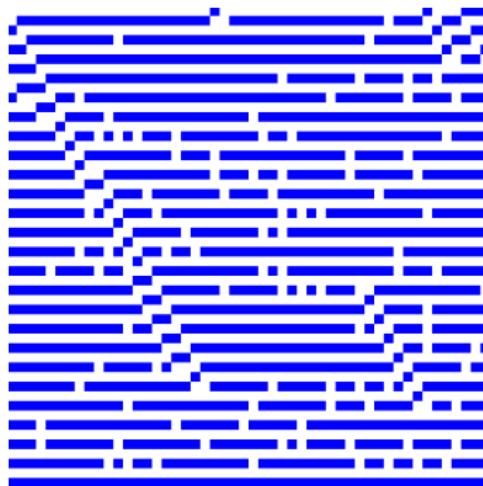
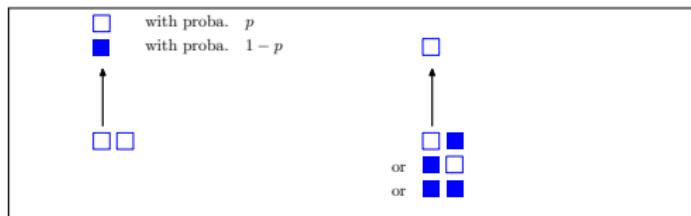


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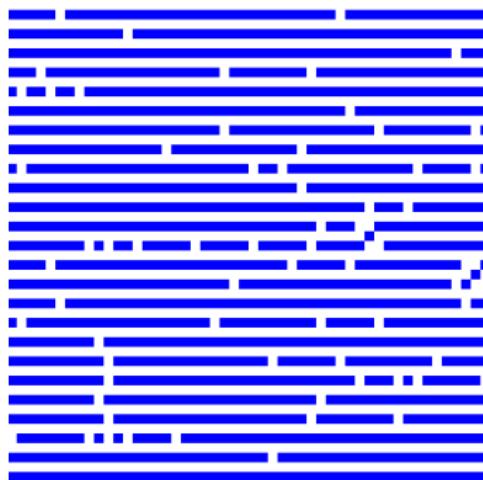
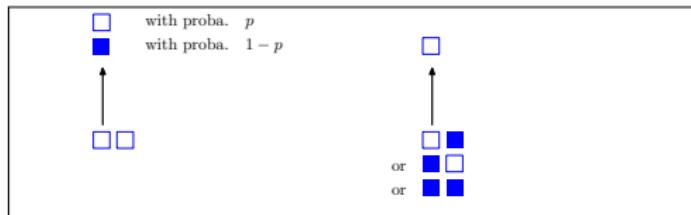
$$p = 0.5$$

# The noisy hardcore Cellular Automaton



$$p = 0.1$$

# The noisy hardcore Cellular Automaton



$$p = 0.05$$

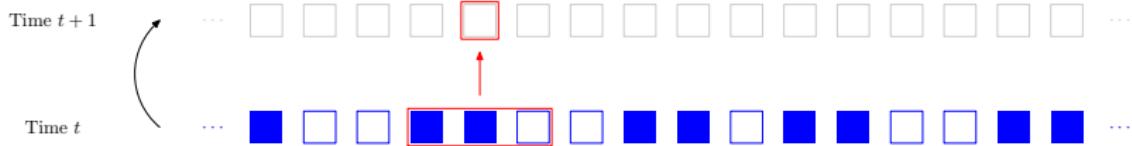
# Definition of Cellular Automata

A CA  $F$  on  $E = \mathbb{Z}^d$  is defined by:

- a finite **alphabet**  $\mathcal{A}$  (set of symbols)
- a finite **neighbourhood**  $\mathcal{N} \subset \mathbb{Z}^d$ ,
- a **local function**  $f : \mathcal{A}^{\mathcal{N}} \rightarrow \mathcal{A}$ .

From configuration  $(x_k)_{k \in \mathbb{Z}^d} \in \mathcal{A}^{\mathbb{Z}^d}$ , cell  $k$  is updated by symbol

$$f((x_{k+v})_{v \in \mathcal{N}})$$



# Definition of Cellular Automata

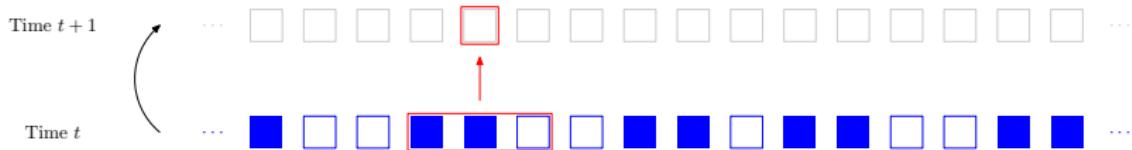
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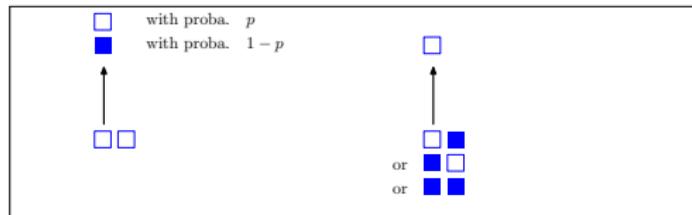
From configuration  $(x_k)_{k \in \mathbb{Z}^d} \in \mathcal{A}^{\mathbb{Z}^d}$ , cell  $k$  is updated by symbol  $y$  with probability:

$$f((x_{k+\nu})_{\nu \in \mathcal{N}})(y),$$

simultaneously and independently of the other cells.



# The noisy hardcore Cellular Automaton



Set of cells:  $E = \mathbb{Z}$   
Alphabet:  $\mathcal{A} = \{\blacksquare, \square\}$   
Neighbourhood:  $\mathcal{N} = \{0, 1\}$

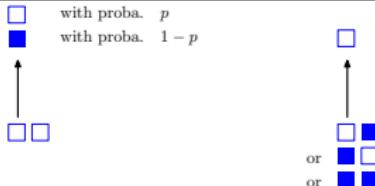
Local function:

$$f : \{\blacksquare, \square\}^2 \rightarrow \mathcal{M}(\{\blacksquare, \square\})$$

defined by:

$$f(\square\square) = (1 - p)\delta_{\blacksquare} + p\delta_{\square}$$

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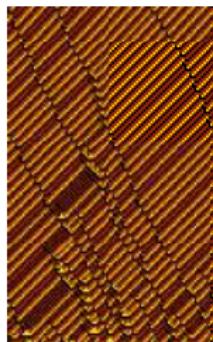
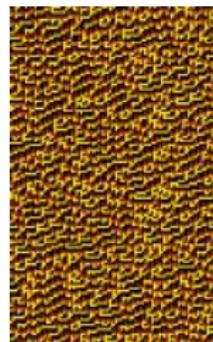
Global function:

$$F : \mathcal{M}(\{\blacksquare, \square\}^{\mathbb{Z}}) \rightarrow \mathcal{M}(\{\blacksquare, \square\}^{\mathbb{Z}})$$

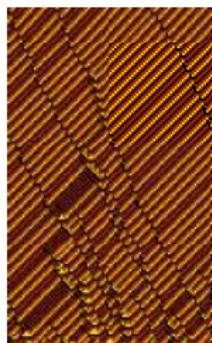
$$\mu \mapsto \mu F$$

- $F : \mathcal{A}^{\mathbb{Z}} \rightarrow \mathcal{A}^{\mathbb{Z}}$  is a CA  $\iff F$  continuous and  $F \circ \sigma = \sigma \circ F$
- Model of parallel computing
- Simple definition and wide variety of behaviour!

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- $F : \mathcal{A}^{\mathbb{Z}} \rightarrow \mathcal{A}^{\mathbb{Z}}$  is a CA  $\iff F$  continuous and  $F \circ \sigma = \sigma \circ F$
- Model of parallel computing **with noise**
- Simple definition and wide variety of behaviour!



# Notion of ergodicity

What are the CA that present some **robustness** to random errors?

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If a PCA forgets every bit of information on the initial configuration, we say it is ergodic.

## Ergodicity

A PCA  $F$  on  $\mathcal{A}^{\mathbb{Z}^d}$  is **ergodic** if:

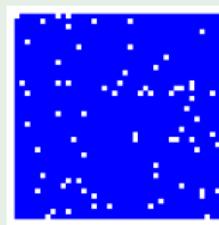
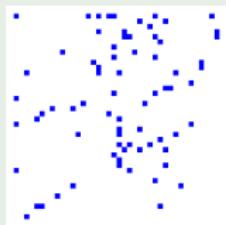
- it has a **unique invariant probability distribution**  $\pi \in \mathcal{M}(\mathcal{A}^{\mathbb{Z}^d})$ , such that  $F\pi = \pi$ ,
- for any initial measure  $\mu \in \mathcal{M}(\mathcal{A}^{\mathbb{Z}^d})$ , the sequence of iterates  $(F^n\mu)_{n \geq 0}$  **converges** weakly to  $\pi$ .

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- Remark: **false** for  $d \geq 2$

## Counter-example for $d = 2$ : Toom CA

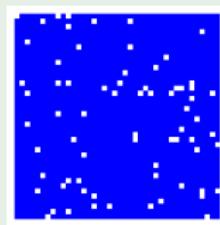
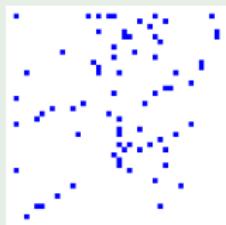
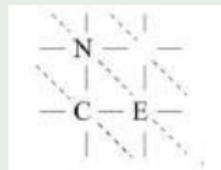
Majority rule:  $(\mathcal{T}(x))_{i,j} = \text{maj}(x_{i,j}, x_{i,j+1}, x_{i+1,j})$



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## Counter-example for $d = 1$ : very complicated! [Gács 2001]

# Question

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**Is the positive rate conjecture true for elementary PCA?**

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Elementary PCA:  $\mathcal{A} = \{0, 1\}$ ,  $\mathcal{N} = \{0, 1\}$ ,  $f(01) = f(10)$

**Is the positive rate conjecture true for elementary PCA?**

It has been shown that for 90% of the cube of parameters, the PCA is ergodic [Toom et al. 1990].

## Content

- The case of deterministic CA

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- The **envelope PCA** and a general criterion of ergodicity for the high-noise regime

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- The **envelope PCA** and a general criterion of ergodicity for the high-noise regime
- Some proofs of ergodicity for PCA **close to being deterministic**
  - ① The noisy hardcore CA
  - ② Noisy nilpotent CA
  - ③ Noisy spreading CA
  - ④ Noisy permutive CA

# Ergodicity of deterministic CA

A deterministic CA  $F : \mathcal{A}^{\mathbb{Z}^d} \rightarrow \mathcal{A}^{\mathbb{Z}^d}$  is **nilpotent** if there exists  $\alpha \in \mathcal{A}$  such that:  $\forall x \in \mathcal{A}^{\mathbb{Z}^d}, \exists n \in \mathbb{N}, F^n(x) = \alpha^{\mathbb{Z}^d}$ .

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**Proposition [Bušić-Mairesse-M., STACS 2011 & Adv. Appl. Proba]**

For deterministic CA, ergodicity  $\Leftrightarrow$  nilpotency.

*Proof.*  $\Leftarrow$  is easy ;  $\Rightarrow$  in two steps:

- ① the unique invariant measure has to be a measure concentrated on a monochromatic configuration  $\alpha^{\mathbb{Z}^d}$ ,
- ② the convergence properties then implies the nilpotency (using [Guillon & Richard 2008], and [Salo 2012] for  $d \geq 2$ ).

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**Corollary** (with [Kari 1992])

The ergodicity of one-dimensional deterministic CA (and hence of PCA) is undecidable.

# Update function of a PCA

A way to run a PCA (on  $\mathcal{A} = \{0, 1\}$ ) from configuration  $x \in \mathcal{A}^{\mathbb{Z}}$ :

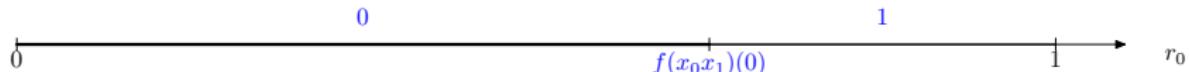
- generate for each cell  $k$  independently and uniformly a random number  $r_k$  in  $[0, 1]$ ,
- choose the new state of the cell  $k$  to be **0** if  $r_k < f((x_{k+v})_{v \in \mathcal{N}})(0)$ , and **1** otherwise.

$\dots \quad x_{-3} \quad x_{-2} \quad x_{-1} \quad x_0 \quad x_1 \quad x_2 \quad x_3 \quad \dots$

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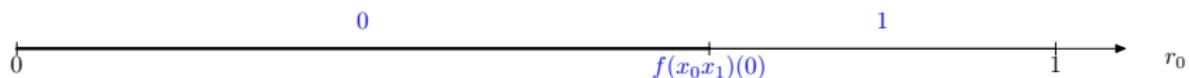


<b>y<sub>0</sub></b>											
...	$x_{-3}$	$x_{-2}$	$x_{-1}$	<b><math>x_0</math></b>	<b><math>x_1</math></b>	$x_2$	$x_3$	...			
...	$r_{-3}$	$r_{-2}$	$r_{-1}$	<b><math>r_0</math></b>	$r_1$	$r_2$	$r_3$	...			

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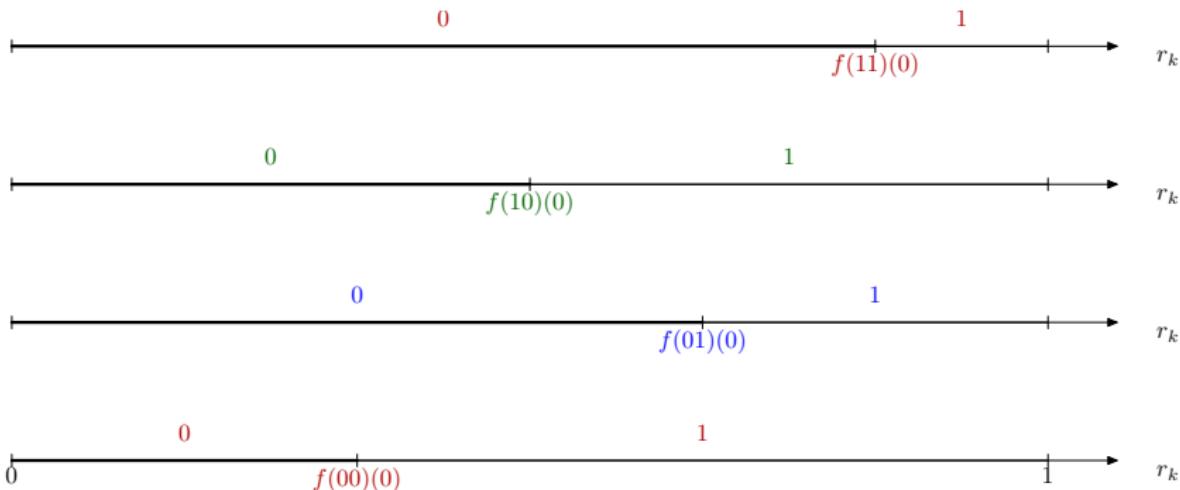
It defines an **update function** for  $F$ , given by:

$$\phi : \mathcal{A}^{\mathbb{Z}} \times [0, 1]^{\mathbb{Z}} \rightarrow \mathcal{A}^{\mathbb{Z}}$$

$$\phi(x, r)_k = \begin{cases} \mathbf{0} & \text{if } r_k < f((x_i)_{i \in k + \mathcal{N}})(0) \\ \mathbf{1} & \text{otherwise.} \end{cases}$$

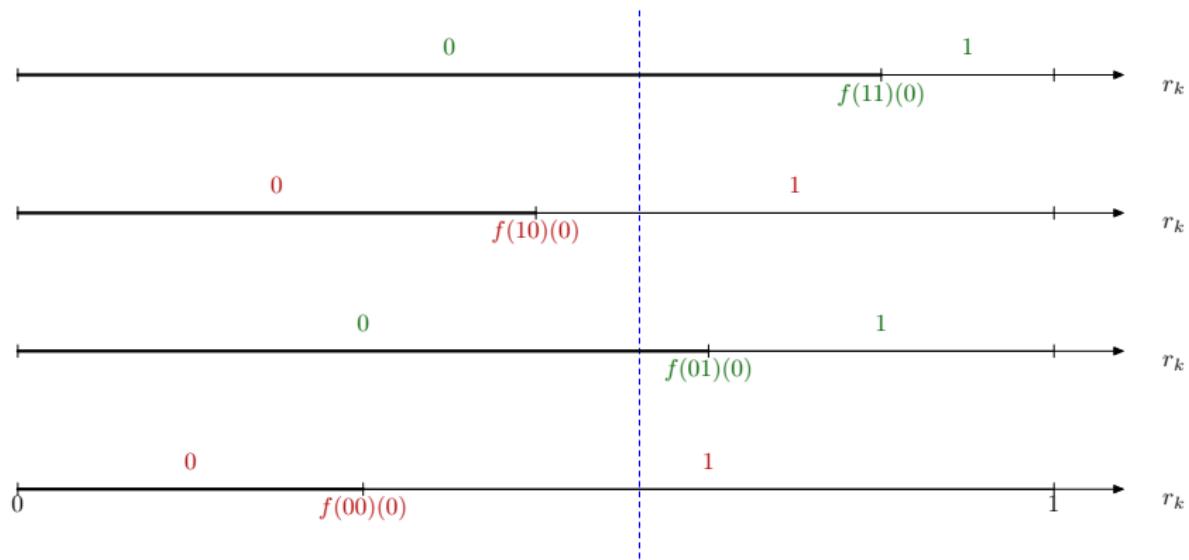
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Example:  $\mathcal{A} = \{0, 1\}$ , neighbourhood  $\mathcal{N} = \{0, 1\}$



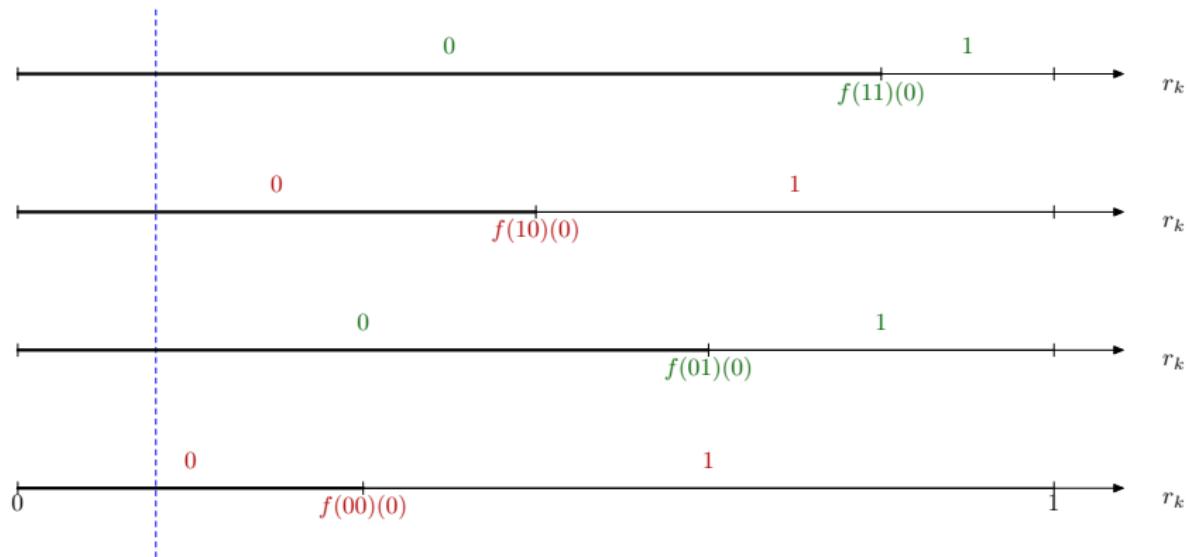
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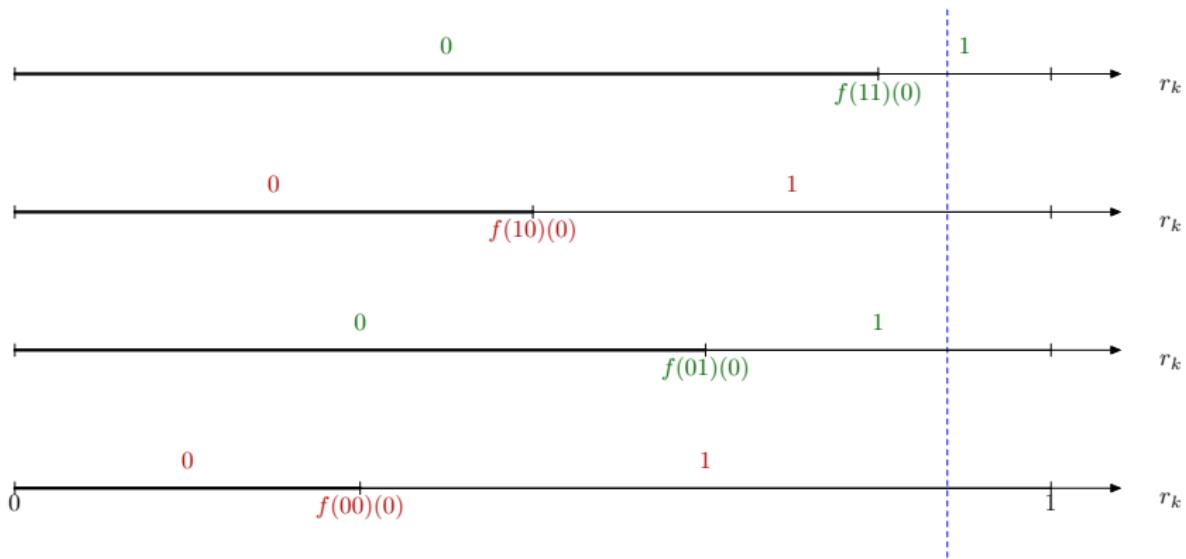
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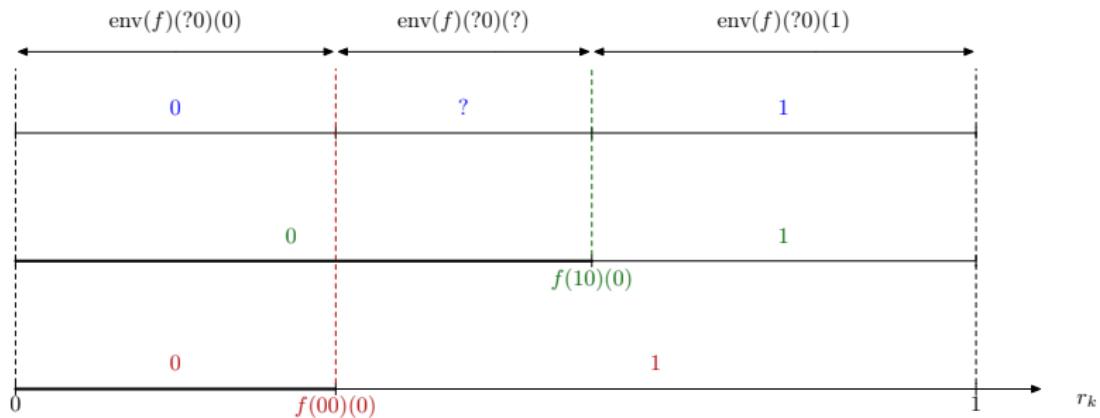


# Envelope PCA

Introduction of an envelope PCA defined on the alphabet

$$\mathcal{B} = \{\mathbf{0} = \{0\}, \mathbf{1} = \{1\}, \mathbf{?} = \{0, 1\}\},$$

to handle configurations partially known.



The update function  $\tilde{\phi}$  of  $\text{env}(P)$  satisfies for  $x \in \mathcal{A}^{\mathbb{Z}}$  and  $y \in \mathcal{B}^{\mathbb{Z}}$ ,

$$x \in y \Rightarrow \forall r \in [0, 1]^{\mathbb{Z}}, \phi(x, r) \in \tilde{\phi}(y, r).$$

# Coupling from the past algorithm

Let  $F$  be a PCA on  $E = \mathbb{Z}$ ,  $\mathcal{A} = \{0, 1\}$ , with  $\mathcal{N} = \{0, 1\}$ .

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$a \quad b$

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$$\begin{matrix} ? & ? \\ ? & ? & ? \end{matrix} \quad (r_i^1)_{0 \leq i \leq 2}$$

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$$\begin{matrix} ? & \textcolor{blue}{1} \\ ? & ? & ? \end{matrix}$$

$$(r_i^1)_{0 \leq i \leq 2}$$

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$$\begin{matrix} ? & 1 \\ ? & ? & ? \\ ? & ? & ? & ? \end{matrix}$$

$$\begin{aligned} (r_i^1)_{0 \leq i \leq 2} \\ (r_i^2)_{0 \leq i \leq 3} \end{aligned}$$

# Coupling from the past algorithm

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? 1

? 0 1

? ? ? ?

? ? ? ? ?

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## Proposition

If this algorithm stops a.s. then the PCA is ergodic, and the algorithm samples perfectly its unique invariant distribution.

## Proposition

The algorithm stops a.s. if and only if the envelope PCA is ergodic.

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But the ergodicity of the envelope PCA is not equivalent to the ergodicity of the original PCA!

# Ergodicity criterion

Let  $F$  be a PCA on  $E = \mathbb{Z}$ ,  $\mathcal{A} = \{0, 1\}$ , with  $\mathcal{N} = \{0, 1\}$ .

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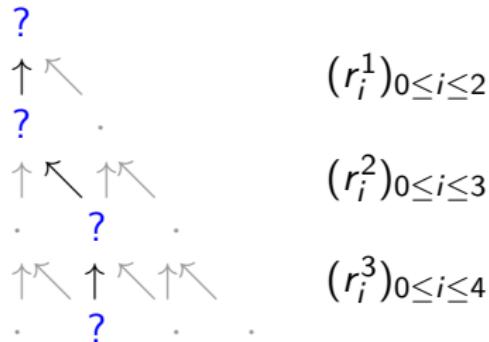
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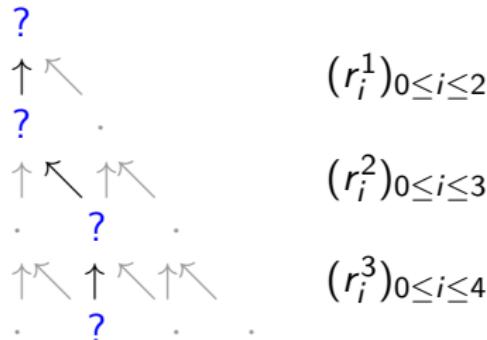
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$$\begin{array}{c}
 ? \\
 \uparrow \nwarrow \\
 ? \\
 \cdot \\
 \uparrow \nwarrow \uparrow \nwarrow \\
 \cdot \quad ? \quad \cdot \\
 \uparrow \nwarrow \uparrow \nwarrow \uparrow \nwarrow \\
 \cdot \quad ? \quad \cdot \quad \cdot
 \end{array}
 \qquad
 \begin{array}{l}
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$$\begin{aligned}
 \text{Let } p_? &= 1 - \min_{x \in \mathcal{A}^{\mathcal{N}}} f(x)(0) - \min_{x \in \mathcal{A}^{\mathcal{N}}} f(x)(1) \\
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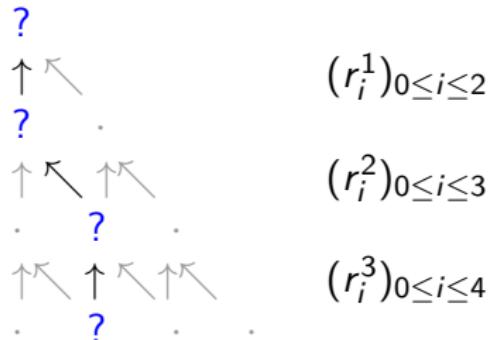


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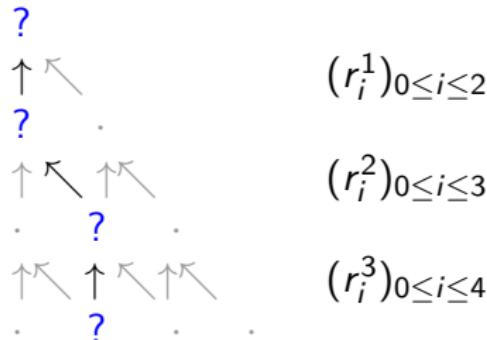
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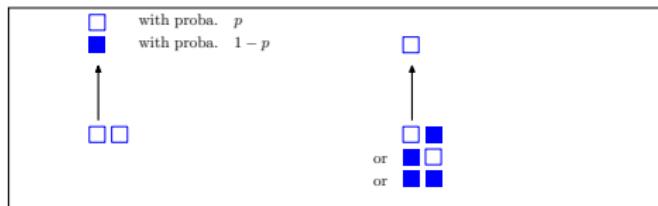
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Remark:  $\alpha^* \geq 1/|\mathcal{N}|$ .

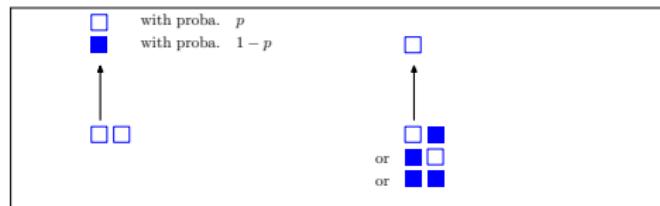
# The noisy Hardcore Cellular Automaton

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PCA  $F$

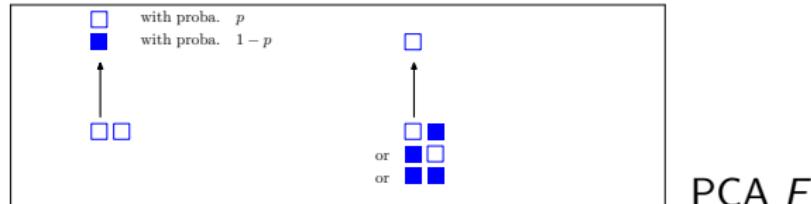
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PCA  $F$

$$p_{?} = 1 - p$$

# The noisy Hardcore Cellular Automaton



$$p? = 1 - p$$

By the ergodicity criterion,  $F$  is ergodic for  $p > 1 - \alpha^* \approx 0.295$ .

What can we say for small values of  $p$ ?

Theorem [Holroyd-M.-Martin, arXiv 2015]

For any parameter  $p \in (0, 1)$ , the noisy Hardcore CA is ergodic.

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Right-weight of a symbol “?” =

- 3 if it is followed by a 0, then by a 1,
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Total weight = left-weight + right-weight.

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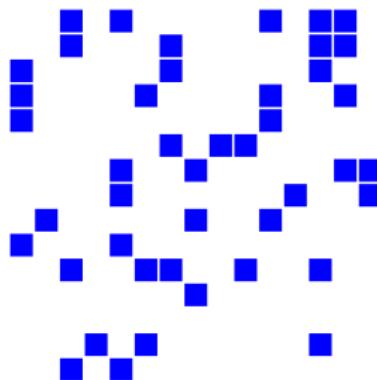
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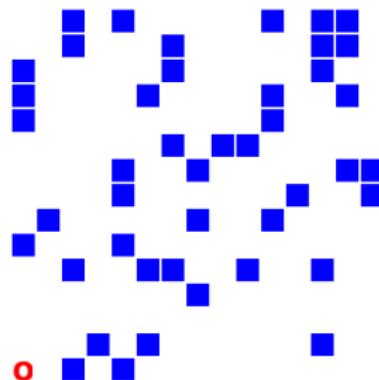
*Example:* in 10??10, the first "?" has a weight  $3+1=4$  and the second one a weight  $1+1=2$ .

# Definition of the percolation game

Grid  $\mathbb{N} \times \mathbb{N}$ , with each site colored in blue independently with probability  $p$  (here,  $p = 0.2$ ).

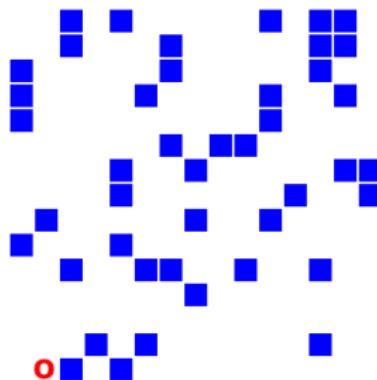


Grid  $\mathbb{N} \times \mathbb{N}$ , with each site colored in blue independently with probability  $p$  (here,  $p = 0.2$ ).



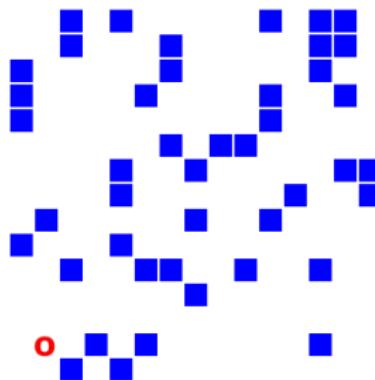
**One** token, that **two** players move alternatively, from position  $x$  to a white position among  $x + (0, 1)$  or  $x + (1, 0)$ .

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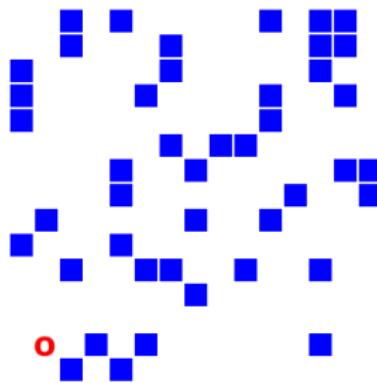
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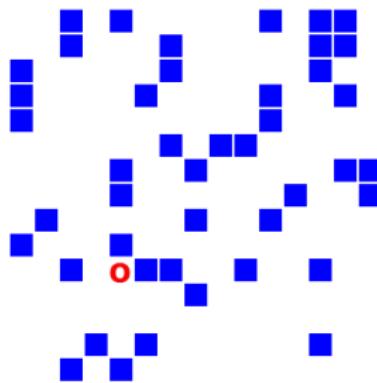
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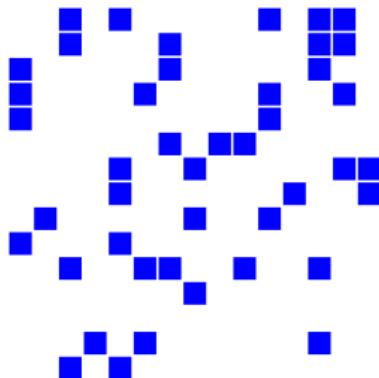
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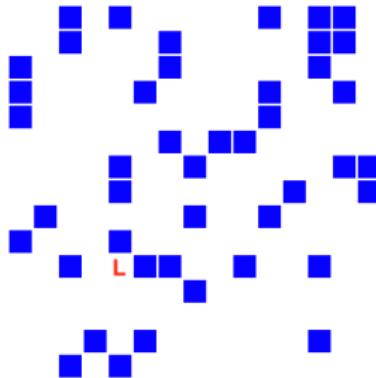
A position is:

- a win (**W**) if from this position, the player whose turn it is to play has a winning strategy,
- a loss (**L**) if from this position, the other player has a winning strategy,
- a draw (**D**) if neither player has a winning strategy, so that with “best play”, the game will continue for ever.



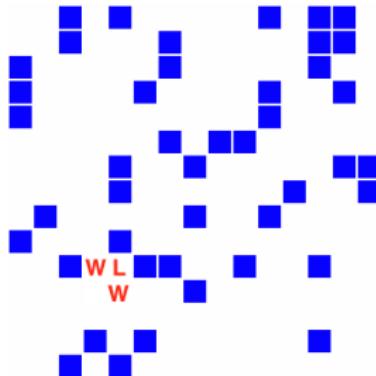
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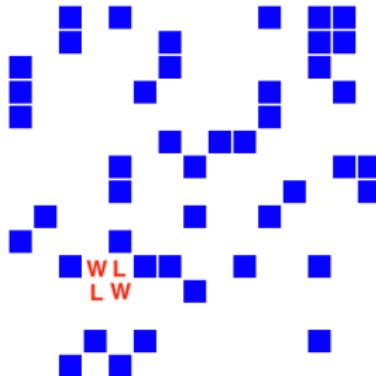
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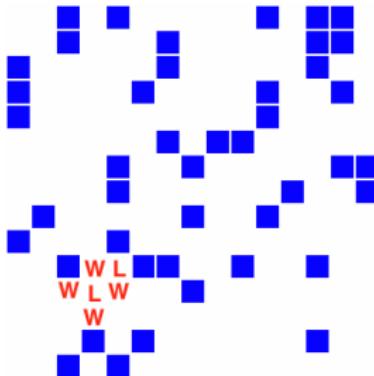
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**Question:** what can we say for the percolation game on  $\mathbb{N}^3$ ?

# Noisy Nilpotent CA

# Nilpotent CA with noise

Let  $F : \mathcal{A}^{\mathbb{Z}} \rightarrow \mathcal{A}^{\mathbb{Z}}$  be a **nilpotent** CA.

There exists  $\alpha \in \mathcal{A}$  and  $s \geq 1$  such that:  $\forall x \in \mathcal{A}^{\mathbb{Z}}, F^s(x) = \alpha^{\mathbb{Z}}$ .

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Let  $R$  be any probabilistic cellular automaton on  $\mathcal{A}^{\mathbb{Z}}$ .

We denote by  $F_{\varepsilon}$  the PCA that consists in, for each cell independently, choosing to apply

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- the local rule of  $R$  with probability  $\varepsilon$ .

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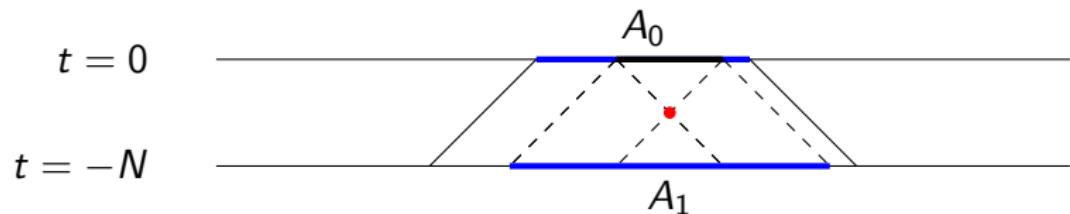
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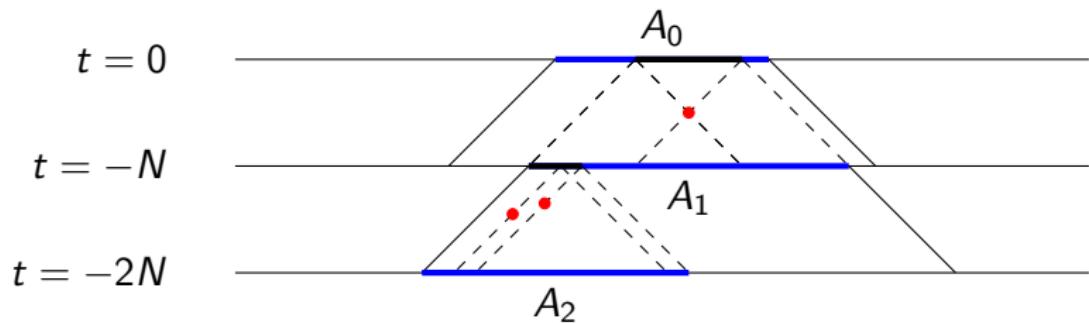
## Proposition [M.-Sablik-Taati]

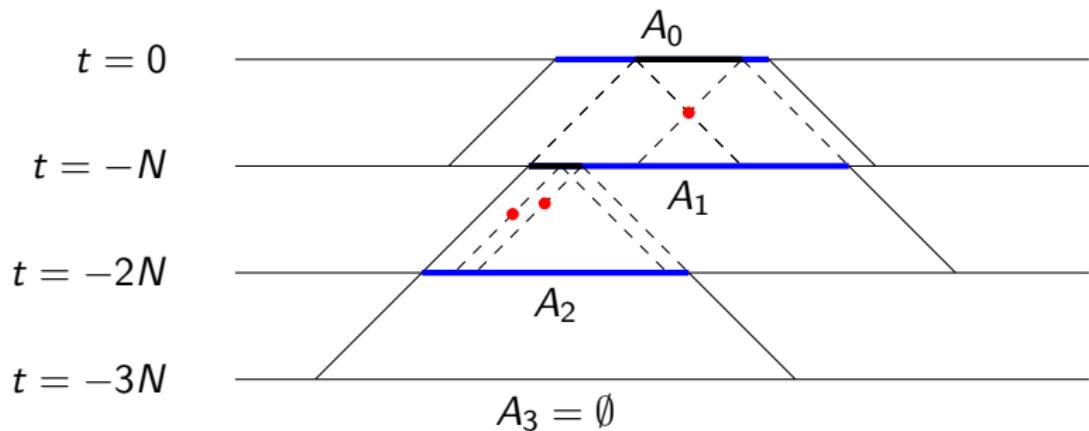
If  $\varepsilon$  is small enough, then the PCA  $F_{\varepsilon}$  is ergodic.

# Nilpotent CA with noise









# Spreading CA with noise

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$$(x_{k+n} = \alpha \text{ for some } n \in \mathcal{N}) \implies F(x)_k = \alpha.$$

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We denote by  $F_{\varepsilon}$  the PCA that consists in, for each cell independently,

- applying the local rule of  $F$  with probability  $1 - \varepsilon$ ,
- choosing a symbol according to a given distribution  $p$  with probability  $\varepsilon$ .

# Spreading CA with noise

Let  $F : \mathcal{A}^{\mathbb{Z}} \rightarrow \mathcal{A}^{\mathbb{Z}}$  be a deterministic CA. We say that the letter  $\alpha \in \mathcal{A}$  is a **spreading symbol** for  $F$  if

$$(x_{k+n} = \alpha \text{ for some } n \in \mathcal{N}) \implies F(x)_k = \alpha.$$

We denote by  $F_{\varepsilon}$  the PCA that consists in, for each cell independently,

- applying the local rule of  $F$  with probability  $1 - \varepsilon$ ,
- choosing a symbol according to a given distribution  $p$  with probability  $\varepsilon$ .

## Proposition [M.-Sablik-Taati]

If  $\varepsilon > 0$  and  $p(\alpha) > 0$ , then the PCA  $F_{\varepsilon}$  is ergodic.

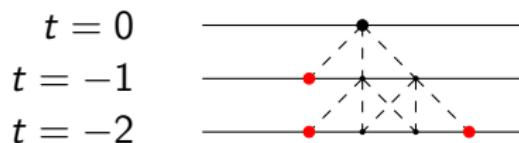
# Spreading CA with noise

$t = 0$       —————•————

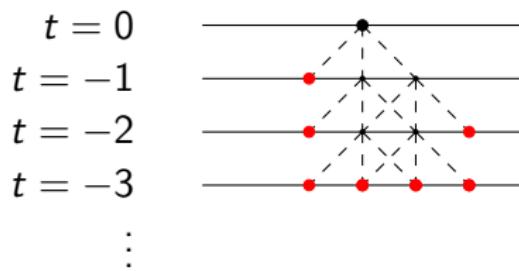
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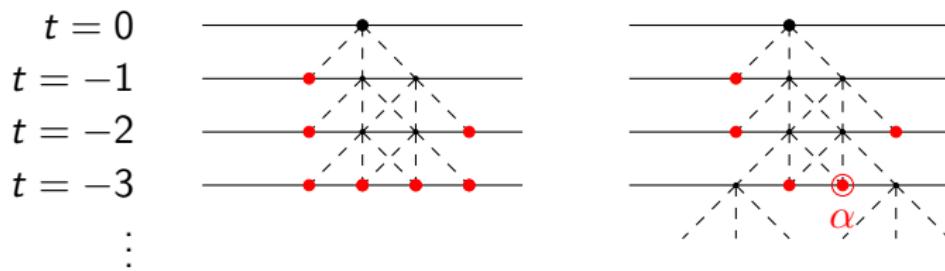
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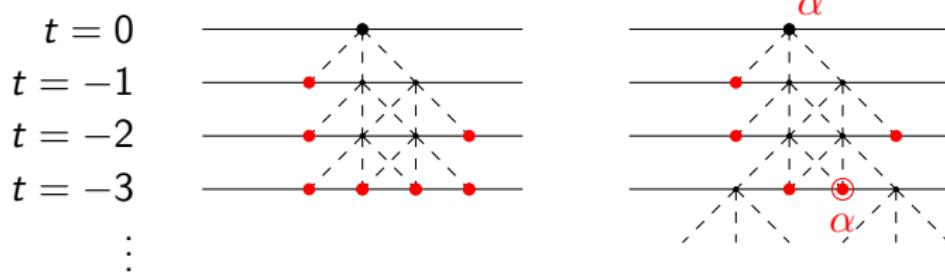
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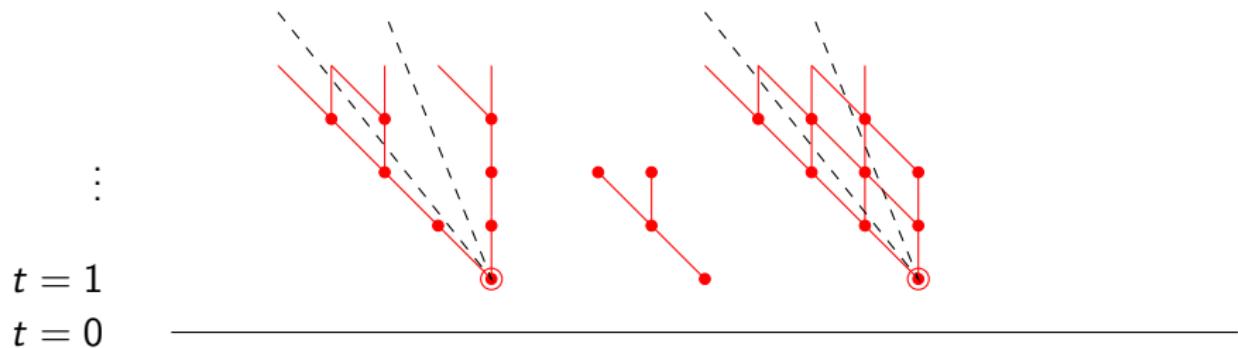


**Remark:** the hypothesis on the noise is not the same as for nilpotent CA.

Small  $\varepsilon$ -perturbation *versus* memoryless zero-range noise given by the distribution  $p$ .

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# Noisy Permutive CA

A deterministic CA  $F$  of neighb.  $\mathcal{N} = \{0, 1\}$  is **left-permutive** if:

$\forall b \in \mathcal{A}, \exists$  permutation  $\sigma_b$  of  $\mathcal{A}$  such that  $f(ab) = \sigma_b(a)$ .

$$\begin{matrix} \sigma_b(a) \\ \backslash \\ a \quad b \end{matrix}$$

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## Example

Additive CA:  $\mathcal{A} = \mathbb{Z}/n\mathbb{Z}$  and  $f(ab) = a + b$ .

# Permutive CA with noise

Let  $F$  be a left-permutive CA.

We denote by  $F_\varepsilon$  the PCA that consists in, for each cell independently,

- applying the local rule of  $F$  with probability  $1 - \varepsilon$ ,
- choosing uniformly a symbol in  $\mathcal{A}$  with probability  $\varepsilon$ .

## Proposition [M.-Sablik-Taati]

For any  $\varepsilon \in (0, 1)$ , the PCA  $F_\varepsilon$  is ergodic.

Each  $a \in \mathcal{A}$  induces a permutation  $\sigma_a$  on  $\mathcal{A}^n$ :

$$y_1 \dots y_n = \sigma_a(x_1 \dots x_n).$$

$$\begin{array}{cccccc} t = 1 & y_1 & y_2 & \dots & y_n \\ t = 0 & x_1 & x_2 & \dots & x_n & a \end{array}$$

# Permutive CA with noise

When adding the noise  $R$ , each  $a \in \mathcal{A}$  induces a Markov chain  $P_a$  on  $\mathcal{A}^n$  :

$$\tilde{y}_1 \dots \tilde{y}_n \sim P_a(x_1 \dots x_n, \bullet).$$

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$t = 3$	.	.	...	.	$a$
$t = 2$	$\tilde{z}_1$	$\tilde{z}_2$	...	$\tilde{z}_n$	$a$
$t = 1$	$\tilde{y}_1$	$\tilde{y}_2$	...	$\tilde{y}_n$	$a$
$t = 0$	$x_1$	$x_2$	...	$x_n$	$a$

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$t = 2$	$\tilde{z}_1$	$\tilde{z}_2$	$\dots$	$\tilde{z}_n$	$a_3$
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$P_a$  is irreducible and aperiodic **for any  $a \in \mathcal{A}$** .

The **uniform measure on  $\mathcal{A}^n$**  is invariant under  $P_a$  **for any  $a \in \mathcal{A}$** .

For each  $a \in \mathcal{A}$ , there exists  $\theta_a < 1$  such that:

$$\|P_a\mu - P_a\nu\|_1 \leq \theta_a \|\mu - \nu\|_1.$$

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$$\|P_{a_t} \dots P_{a_2} P_{a_1} \mu - P_{a_t} \dots P_{a_2} P_{a_1} \nu\|_1 \leq \theta^t \|\mu - \nu\|_1.$$

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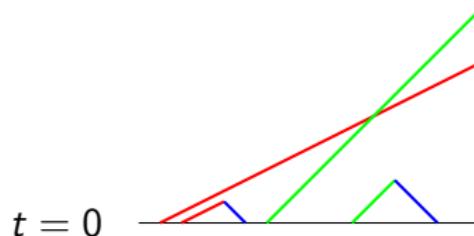
$$\|P_{a_t} \dots P_{a_2} P_{a_1} \mu - P_{a_t} \dots P_{a_2} P_{a_1} \nu\|_1 \leq \theta^t \|\mu - \nu\|_1.$$

In particular, for  $\nu = \lambda_n$  (uniform measure on  $\mathcal{A}^n$ ), we obtain that for any distribution  $\mu$  on  $\mathcal{A}^n$  and any  $a_1, \dots, a_t \in \mathcal{A}$ ,

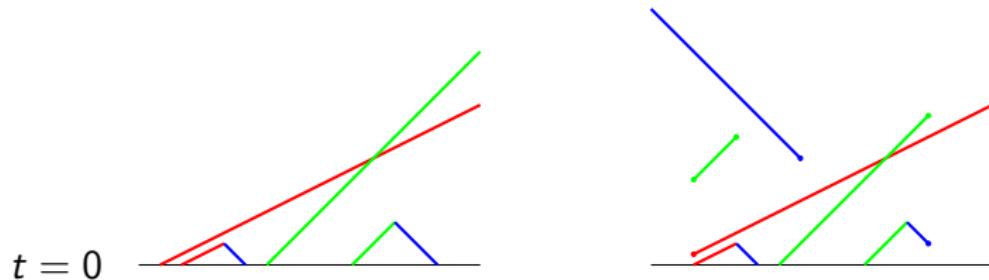
$$\|P_{a_t} \dots P_{a_2} P_{a_1} \mu - \lambda_n\|_1 \leq \theta^t \|\mu - \lambda_n\|_1 \leq 2\theta^t.$$

# Interacting gliders

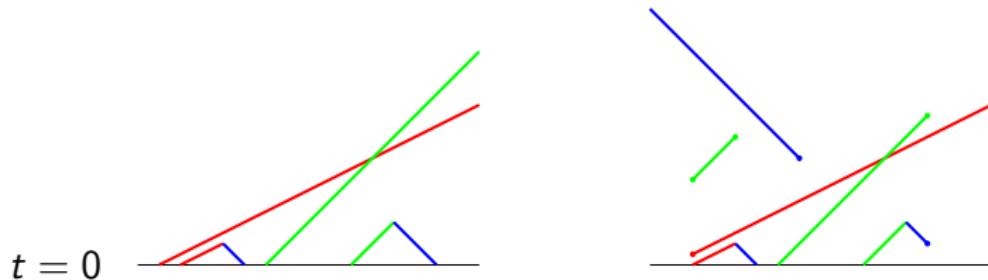
- Gliders with annihilation rules



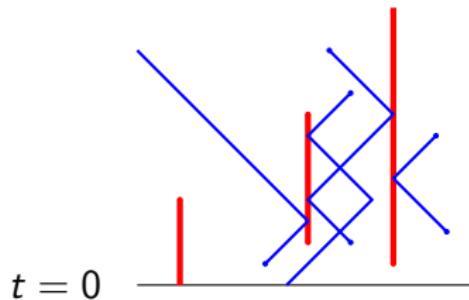
- Gliders with annihilation rules



- Gliders with annihilation rules



- Gliders with reflecting walls



- Finding good **weight systems** for other families of PCA.
- **Elementary PCA** (neighbourhood of size 2, binary states):  
is it true that if all the probability transitions are in  $(0, 1)$ ,  
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Thank you for your attention!