

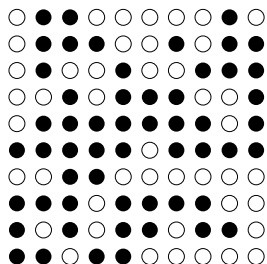
Restricted density classification in one dimension

Siamak Taati

Leiden University, The Netherlands

AUTOMATA 2015 — Turku, June 2015

Density classification task



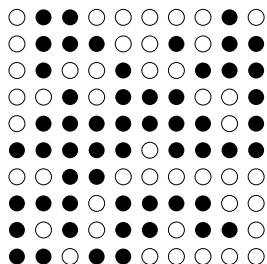
Given: an array of symbols

Task: Determine which symbol is in **majority**.

Requirements

- ▶ A cellular automaton algorithm [say, with periodic boundary]
[using no extra symbols!]
- ▶ Output: must reach consensus on the majority symbol
- ▶ Scalability: must work for arrays of arbitrary size

Density classification task



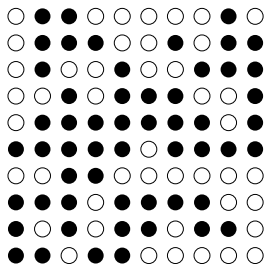
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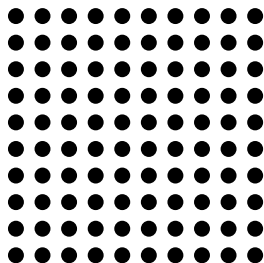
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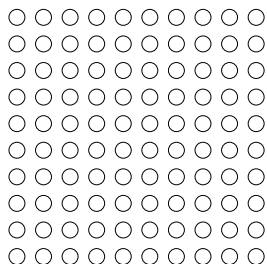
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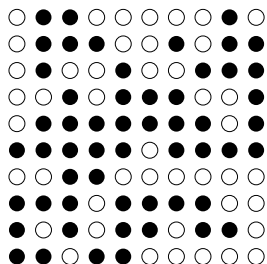
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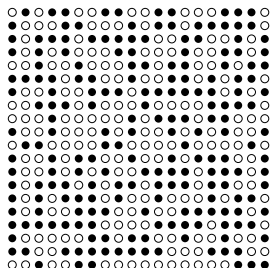
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Unfortunately ... (Land and Belew, 1995)

No perfect solution exists!

Natural relaxation

- ▶ classify correctly with “high probability”

Density classification task

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Density classification: infinite lattice

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Terminology

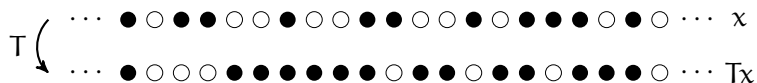
- ▶ A **configuration** of symbols: $x : \mathbb{Z} \rightarrow \{○, ●\}$
- ▶ A cellular automaton (CA): $T : \{○, ●\}^{\mathbb{Z}} \rightarrow \{○, ●\}^{\mathbb{Z}}$
- ▶ T classifies x according to density if

$$\begin{aligned} T^t x &\rightarrow \text{all-}● && \text{if } \text{density}_●(x) > 1/2, \\ T^t x &\rightarrow \text{all-}○ && \text{if } \text{density}_●(x) < 1/2. \end{aligned}$$

as $t \rightarrow \infty$.

[convergence \equiv site-wise eventual agreement]

Density classification: infinite lattice



Terminology

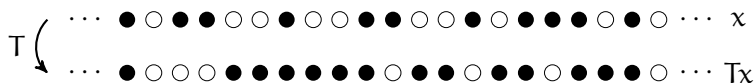
- ▶ A configuration of symbols: $x : \mathbb{Z} \rightarrow \{0, \bullet\}$
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- ▶ T classifies x according to density if

$$\begin{aligned} T^t x &\rightarrow \text{all-}\bullet && \text{if } \text{density}_{\bullet}(x) > 1/2, \\ T^t x &\rightarrow \text{all-}0 && \text{if } \text{density}_{\bullet}(x) < 1/2. \end{aligned}$$

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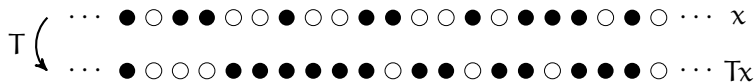
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Density classification: infinite lattice



Terminology

- ▶ A configuration of symbols: $x : \mathbb{Z} \rightarrow \{0, 1\}$
- ▶ A cellular automaton (CA): $T : \{0, 1\}^{\mathbb{Z}} \rightarrow \{0, 1\}^{\mathbb{Z}}$
- ▶ T classifies x according to density if

$$\begin{aligned} T^t x &\rightarrow \text{all-1} && \text{if } \text{density}_1(x) > 1/2, \\ T^t x &\rightarrow \text{all-0} && \text{if } \text{density}_1(x) < 1/2. \end{aligned}$$

as $t \rightarrow \infty$.

[convergence \equiv site-wise eventual agreement]

Task (relaxed)

Classify random configurations with high probability. [Random=?]

Classification of random configurations

... ● ○ ● ● ○ ○ ● ○ ○ ● ● ○ ○ ● ○ ● ● ● ○ ● ○ ... X

More specifically ...

Let X be a configuration chosen at random, using **independent biased coin flips**. $X_i = \begin{cases} \bullet & \text{with prob. } p, \\ \circ & \text{with prob. } 1 - p. \end{cases}$

Then, $\text{density}_\bullet(X) = p$ almost surely. [by the law of large numbers]

Task (relaxed)

$T^t X \rightarrow \text{all-}\bullet$ if $p > 1/2$,
 $T^t X \rightarrow \text{all-}\circ$ if $p < 1/2$.

with high probability.

Classification of random configurations

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$$\begin{aligned} T^t X \rightarrow \text{all-}\bullet & \quad \text{if } p > 1/2, \\ T^t X \rightarrow \text{all-}\circ & \quad \text{if } p < 1/2. \end{aligned}$$

with high probability: **probability 1**.

Note!

The distribution of X is shift-invariant and **ergodic**, and $\{T^t X \rightarrow \text{all-}\bullet\}$ and $\{T^t X \rightarrow \text{all-}\circ\}$ are **shift-invariant events**.

Classification of coin-flip configurations

Task (almost-sure classification)

$$T^t X \rightarrow \text{all-}\bullet \quad \text{if } p > 1/2,$$

$$T^t X \rightarrow \text{all-}\circ \quad \text{if } p < 1/2.$$

with probability 1.

Question

Is there a CA that classifies coin-flip configurations almost surely?

[i.e., for any $0 \leq p \leq 1$]

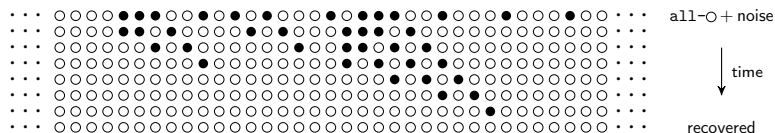
In ≥ 2 dimensions (Bušić, Fatès, Mairesse, Marcovici, 2013)

Perfect solution: Toom's NEC rule!

In 1 dimension

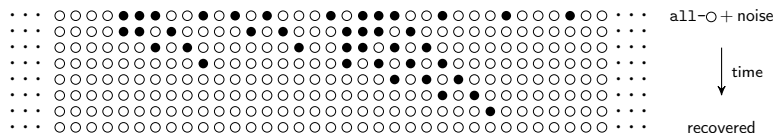
- ▶ Not known! But there are candidates ...
- ▶ Modest goal: when the bias is strong ...

Interpretations & related problems



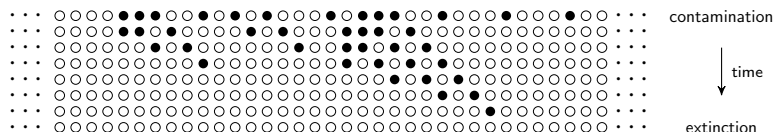
- ▶ Density classification as **recovery from noise**
- ▶ What if there is noise at every step?
 - ▶ Do all-○ and all-● remain stable?
 - In 2d: – Toom's NEC rule (Toom, 1974, 1980)
 - In 1d: – Same candidates ...
 - Gács's sophisticated construction (1986, 2001)
- ▶ Density classification as a percolation problem
 - ▶ Do contaminations survive (without escaping to infinity)?
[Contaminations can only spread through neighbours.]
- ▶ Sharp phase transition in the behaviour when varying p

Interpretations & related problems



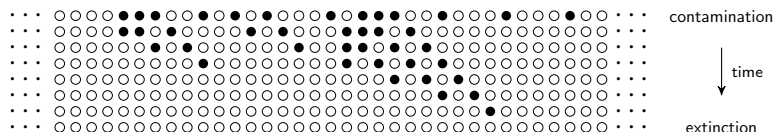
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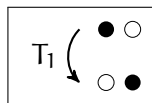
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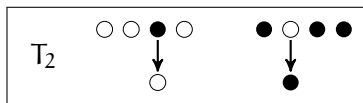
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Candidate I: modified traffic (Kari and Le Gloanec, 2012)

$$T = T_2 T_1$$



traffic



filter

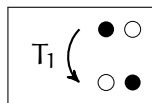


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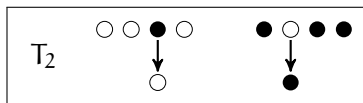
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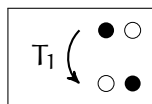


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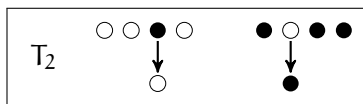
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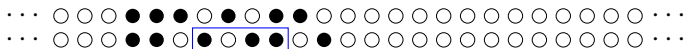
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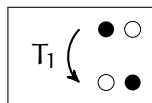


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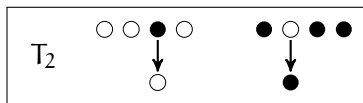
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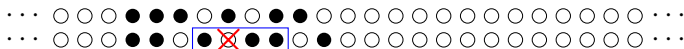
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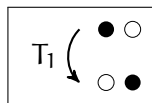


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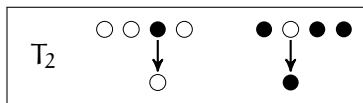
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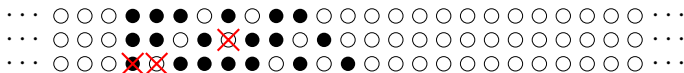
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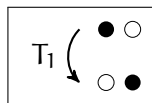


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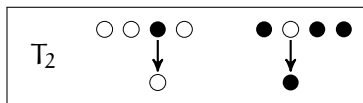
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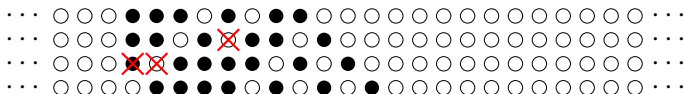
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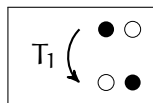


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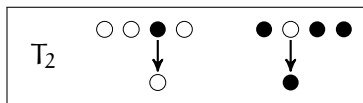
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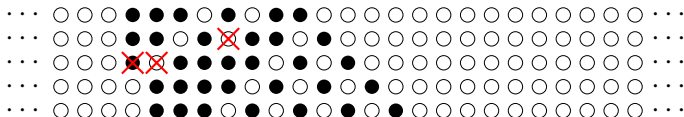
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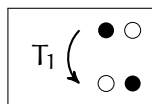


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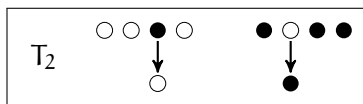
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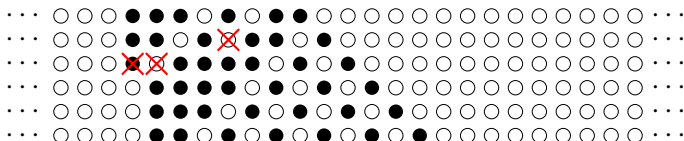
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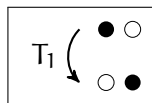


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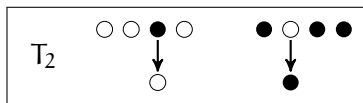
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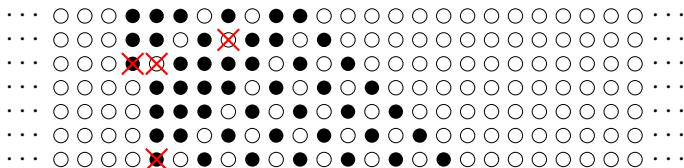
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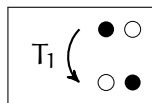


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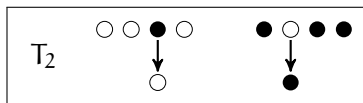
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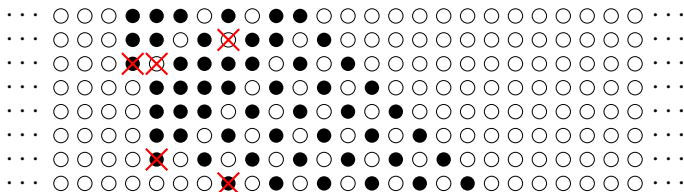
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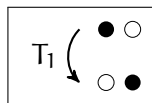


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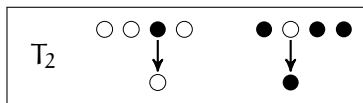
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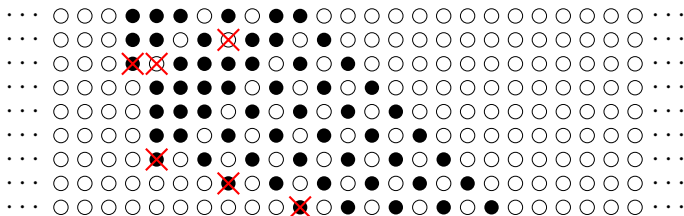
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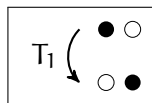


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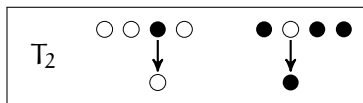
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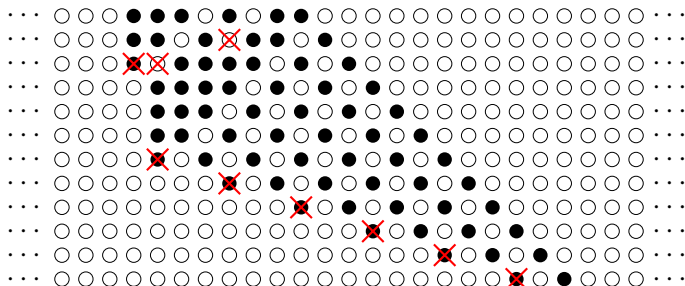
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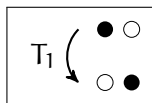


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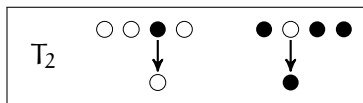
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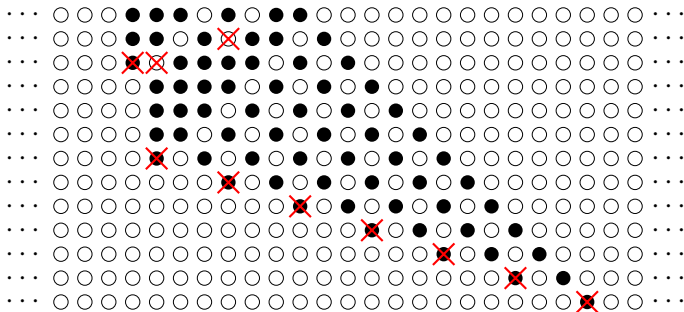
$$T = T_2 T_1$$



traffic



filter

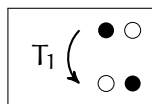


The island is washed out!

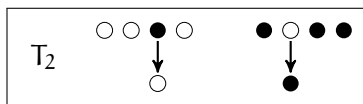
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Theorem (Kari and Le Gloanec, 2012)

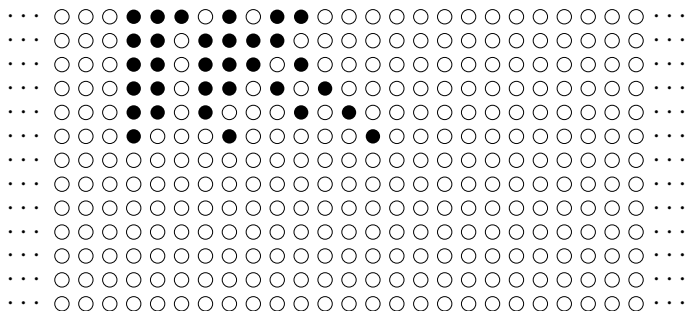
Every finite island is eventually washed out!

An island of length l is washed out within $2l$ steps.

Symmetry: $\circ \leftrightarrow \bullet$ and left \leftrightarrow right

Candidate II: GKL (Gács, Kurdyumov, Levin, 1978)

$$(\mathbb{T}x)_i \triangleq \begin{cases} \text{maj}(x_{i-3}, x_{i-1}, x_i) & \text{if } x_i = \circ, \\ \text{maj}(x_i, x_{i+1}, x_{i+3}) & \text{if } x_i = \bullet, \end{cases}$$

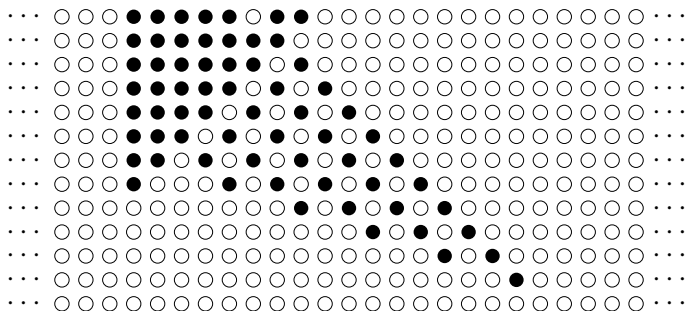


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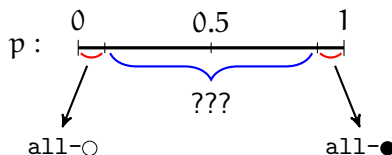
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Restricted classification

Theorem (T., 2014)

Let T be **modified traffic** or **GKL**. Then, T classified a biased coin flip configuration almost surely correctly provided **the bias is strong**.

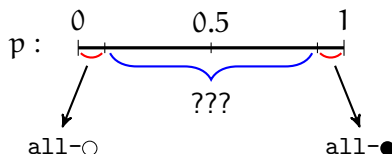


Recall: $p \triangleq \mathbb{P}(X_i = \bullet)$

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Remark 1

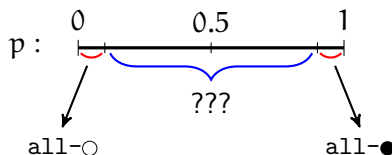
This shows a **phase transition** when p is varied.

However, it doesn't rule out **other phases in between**.

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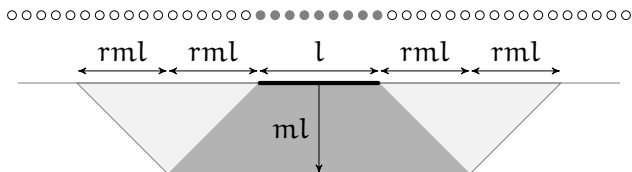


Recall: $p \triangleq \mathbb{P}(X_i = \bullet)$

Remark II

By **symmetry**, we can focus on p close to 0.

Washing out finite islands (in linear time)



Washing out an island of length l in ml steps

r : neighbourhood radius of the CA

Examples: – modified traffic and GKL with $m = 2$
– (also Toom's NEC rule)

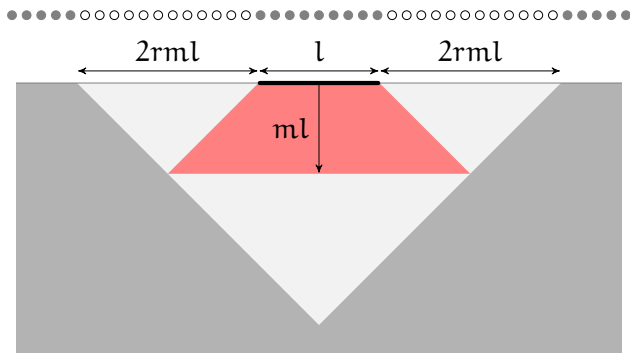
Claim

Suppose T washes out finite islands of \bullet in background of \circ in linear time. Then,

$$T^t X \rightarrow \text{all-}\circ \quad \text{almost surely}$$

if X is a coin flip configuration with p close to 0.

Isolated islands



An isolated island has a sufficiently wide margin of \circ

Observation

An **isolated island** disappears before sensing or affecting the rest of the configuration.

\Rightarrow removing an isolated island from a configuration x does not affect whether $T^t x \rightarrow \text{all-}\circ$ or not.

Isolated islands



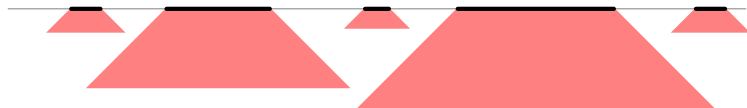
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Washing out sparse sets



Cleaning isolated islands makes larger islands isolated!

Cleaning procedure [Gács, 1986, 2001, Durand, Romashchenko, Shen, 2012]

Clean isolated islands recursively.

We call a configuration **sparse** if the cleaning procedure eventually cleans every \bullet .

→ Decompose a sparse configuration x
into a family of islands $\mathcal{I}(x)$.

Question

Is sparseness enough for $T^t \rightarrow \text{all-0}$?

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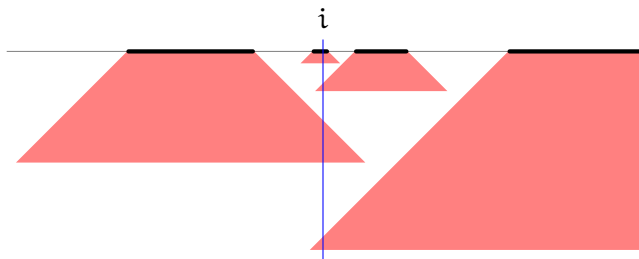
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Site i may be changed infinitely often!

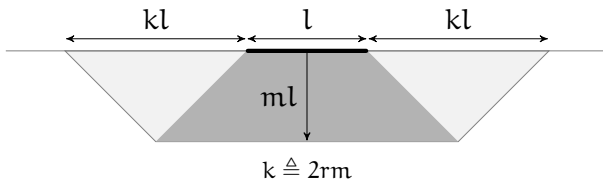
We call a sparse configuration x **strongly sparse** if every site is in the territory of at most finitely many islands in $\mathcal{I}(x)$.

[territory = island and its margin]

Observation

$T^t x \rightarrow \text{all-0}$ if x is strongly sparse.

Sparseness of random configurations



Theorem (Gács, 1986, 2001, Durand, Romashchenko, Shen, 2012)

A biased coin flip configuration X is almost surely strongly sparse if the parameter p is sufficiently close to 0.

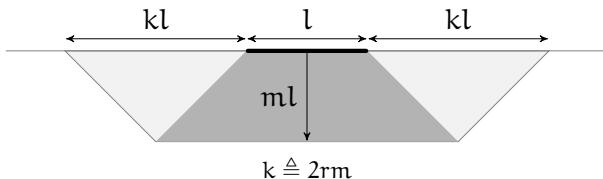
More precisely: it is enough that $p < (2k)^{-2}$

Examples

For modified traffic and GKL, $m = 2$ and $r = 3$, so $k = 12$.

⇒ X is classified almost surely correctly
if $p < 0.0017$ or $p > 0.9983$.

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Sparseness of random configurations: proof

[Gács, 1986, 2001, Durand, Romashchenko, Shen, 2012]

Choose an appropriate sequence $l_1 < l_2 < l_3 < \dots$ of lengths.

[to be determined ...]

Cleaning procedure: Version II

1. Clean all isolated islands of length $\leq l_1$.
2. Clean all isolated islands of length $\leq l_2$.
3. ...

Note: The notion of sparseness does not change!

Question

What is the probability that site u has state \bullet after n cleaning steps?

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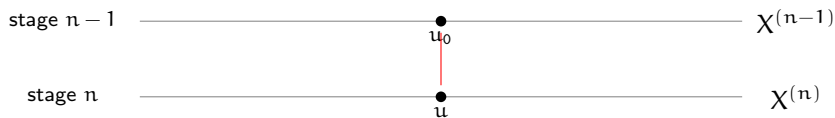
Explanation tree

stage n  $X^{(n)}$

If site u has state \bullet at stage n , then

Sparseness of random configurations: proof

Explanation tree

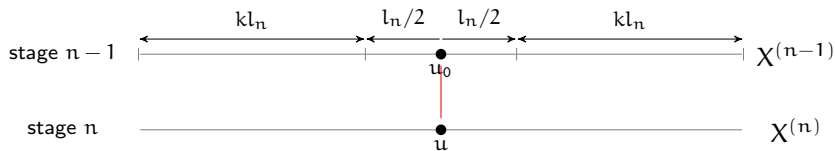


If site u has state \bullet at stage n , then

- ▶ u must have state \bullet at stage $n - 1$, and

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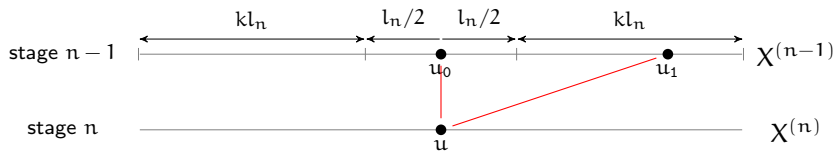


If site u has state \bullet at stage n , then

- ▶ u must have state \bullet at stage $n-1$, and
- ▶ u must not be inside an isolated island of length $\leq l_n$.

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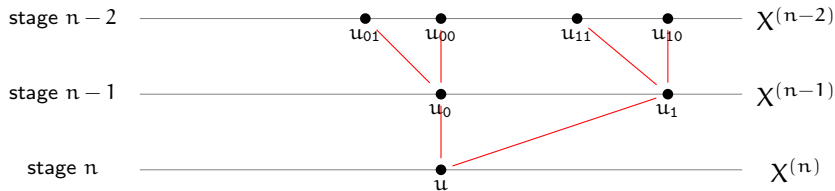


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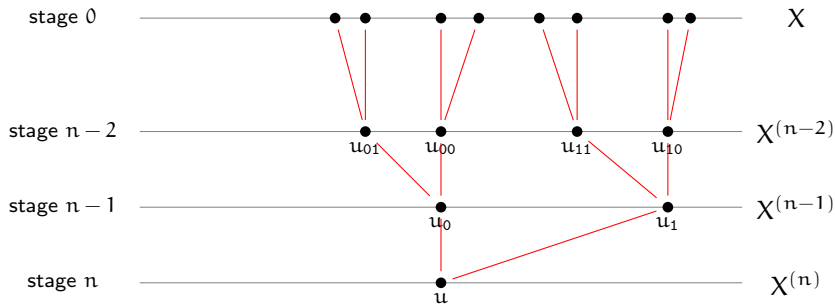


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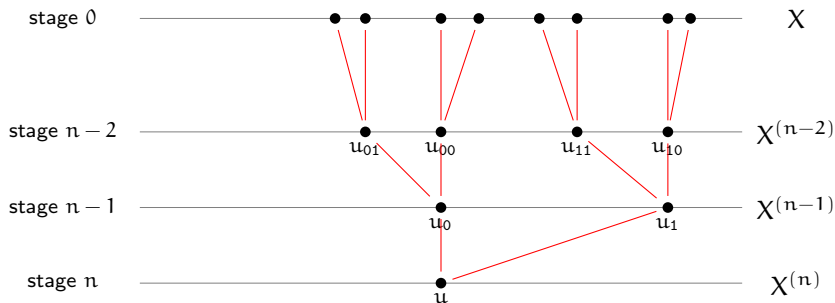


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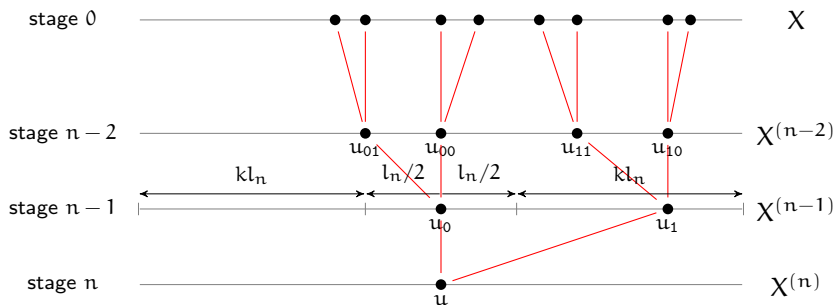
Probability of a tree

Choose $l_n \triangleq (4k+3)^{n-1}$ to sure the leaves are distinct!

$$\mathbb{P}(\text{a given tree is an explanation}) = p^{2^n}$$

Sparseness of random configurations: proof

Explanation tree



Number of a tree

Recursive inequality: $f_n \leq 2kl_n \times f_{n-1}^2$.

$$\# \text{ of trees} \leq (2k)^{2^{n+1}}$$

Sparseness of random configurations: proof

Probability of a tree

$$\mathbb{P}(\text{a given tree is an explanation}) = p^{2^n}$$

Number of a tree

$$\# \text{ of trees} \leq (2k)^{2^{n+1}}$$

Probability of survival after n cleaning steps

At least one explanation tree must exist!

$$\mathbb{P}(X_u^{(n)} = \bullet) \leq (2k)^{2^{n+1}} p^{2^n} = ((2k)^2 p)^{2^n}$$

which $\rightarrow 0$ as long as $p < (2k)^{-2}$.

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Probability of survival after n cleaning steps

$$\mathbb{P}(X_u^{(n)} = \bullet) \leq ((2k)^2 p)^{2^n}$$

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Corollary: sparseness

X is **sparse** as long as $p < (2k)^{-2}$.

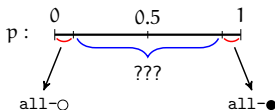
Corollary: strong sparseness

X is **strongly sparse** as long as $p < (2k)^{-2}$. [by Borel-Cantelli ...]

Q.E.D.

Remarks and open problems

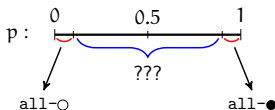
- ▶ Does an **intermediate phase** exist?



- ▶ Random configurations with other distributions
Sparseness? Classification? [Markov, Gibbs, ...]
- ▶ Would sparseness approach work for probabilistic CA?
 - Fatès, 2013: majority-traffic
 - Noisy version of majority
- ▶ Noise at each step: modified traffic and GKL
- ▶ Washing out errors on tilings and subshifts of finite type
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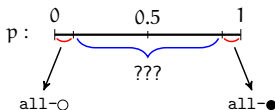
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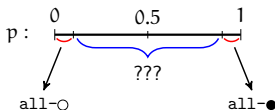
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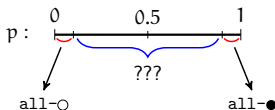
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Thank you for your attention!