

Construction of μ -Limit Sets of Two-Dimensional Cellular Automata

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Abstract

We prove a characterization of μ -limit sets of two-dimensional cellular automata, extending existing results in the one-dimensional case. This sets describe the typical asymptotic behaviour of the cellular automaton, getting rid of exceptional cases, when starting from the uniform measure.

Introduction

Cellular automata are discrete dynamical systems defined by a local rule, introduced in the 40s by John von Neumann [12]. They model a large variety of discrete systems and are linked with various areas of mathematics or computer science, in particular computation theory, complex systems, ergodic theory and combinatorics.

One of the main catalysts of the study of cellular automata was their surprisingly complex and organised behaviours, even when iterated on configurations with no particular structure (e.g. chosen at random). To formalise these observations, many authors tried to describe their asymptotic behaviour by considering the limit set, which is the set of configurations that can be reached after arbitrarily many steps. These sets were shown to have potentially high computational complexity [11, 1], and any nontrivial property on them is undecidable [9]. Nevertheless, the problem of characterizing which subshifts can be limit sets of CA remains open.

In 2000, Kůrka and Maass argued that limit sets did not provide a good description of empirical observations and introduced instead a measure-theoretical version [10]. The idea of μ -limit sets is to choose the initial configuration at random, according to some probability measure μ , and to consider all patterns whose probability to appear does not tend to 0. In the one-dimensional case, similar results of high complexity and undecidability were found [4, 3, 6, 2]. Another approach was developed in [5], considering the limit probability measure, with similar results.

In this article, we consider the two-dimensional case and prove a characterization of all subshifts that can be μ -limit sets of CA for μ the uniform Bernoulli measure. The method is constructive and inspired by the one-dimensional constructions in [2, 5].

1 Definitions

1.1 Cellular automata on two dimensions

Definition (Configurations, patterns, cylinders).

Let \mathcal{A} be a finite alphabet. We introduce $\mathcal{A}^{\mathbb{Z}^2}$ the set of (two-dimensional) **configurations**. Denote \mathcal{A}^* the set of finite **patterns**, that is, any element of $\mathcal{A}^{\mathbb{U}}$ for some $\mathbb{U} \subset_{\text{finite}} \mathbb{Z}^2$ (denote $\mathbb{U} = \text{supp}(u)$ the **support** of the pattern u).

Such a pattern is said to be square or rectangular if its support is.

Given $u \in \mathcal{A}^*$ and $i, j \in \mathbb{Z}^2$, define the cylinder $[u]_{i,j} = \{x \in \mathcal{A}^{\mathbb{Z}^2} \mid x_{(i,j)+\text{supp}(u)} = u\}$.

Endowed with the product topology, $\mathcal{A}^{\mathbb{Z}^2}$ is a compact and metrisable space. A distance inducing this topology is:

$$\forall x, y \in \mathcal{A}^{\mathbb{Z}^2}, d_C(x, y) = 2^{-\Delta(x, y)} \quad \text{where } \Delta(x, y) = \min\{|i|+|j| \mid i, j \in \mathbb{Z}^2, x_{i,j} \neq y_{i,j}\}$$

The **frequency** of a pattern $u \in \mathcal{A}^*$ in another pattern $v \in \mathcal{A}^*$ is defined as:

$$\text{Freq}(u, v) = \frac{\#\left\{(i, j) \in \text{supp}(v) : \begin{array}{l} (i, j) + \text{supp}(u) \subseteq \text{supp}(v) \\ v_{(i,j)+\text{supp}(u)} = u \end{array} \right\}}{\#\{(i, j) \in \text{supp}(v) : (i, j) + \text{supp}(u) \subseteq \text{supp}(v)\}} \quad \text{and 0 if it is undefined.}$$

Definition (Shift actions).

Define the two shifts actions $\sigma_{\uparrow}, \sigma_{\rightarrow} : \mathcal{A}^{\mathbb{Z}^2} \rightarrow \mathcal{A}^{\mathbb{Z}^2}$ by:

$$\forall x \in \mathcal{A}^{\mathbb{Z}^2}, i, j \in \mathbb{Z}^2, \quad \sigma_{\rightarrow}(x)_{i,j} = x_{i-1,j} \quad \text{and} \quad \sigma_{\uparrow}(x)_{i,j} = x_{i,j-1}.$$

Definition (Cellular automata).

A (two-dimensional) **cellular automaton** is a continuous action $F : \mathcal{A}^{\mathbb{Z}^2} \rightarrow \mathcal{A}^{\mathbb{Z}^2}$ that commutes with σ_{\rightarrow} and σ_{\uparrow} . Equivalently, it can be defined by a **local rule** $\bar{F} : \mathcal{A}^{\mathbb{U}_F} \rightarrow \mathcal{A}$, where $\mathbb{U}_F \subset \mathbb{Z}^2$ is a finite **neighbourhood**, in the sense that

$$\forall x \in \mathcal{A}^{\mathbb{Z}^2}, i, j \in \mathbb{Z}^2, F(x)_{i,j} = \bar{F}((x_{(i,j)+u})_{u \in \mathbb{U}_F}).$$

This equivalence is known as the Curtis-Hedlund-Lyndon theorem [7].

1.2 Probability measures

Definition (Probability measures on $\mathcal{A}^{\mathbb{Z}^2}$).

Let \mathfrak{B} be the Borel sigma-algebra of $\mathcal{A}^{\mathbb{Z}^2}$. Denote by $\mathcal{M}(\mathcal{A}^{\mathbb{Z}^2})$ the set of probability measures on $\mathcal{A}^{\mathbb{Z}^2}$ defined on the sigma-algebra \mathfrak{B} . Let $\mathcal{M}_{\sigma}(\mathcal{A}^{\mathbb{Z}^2})$ be the $\sigma_{\uparrow}, \sigma_{\rightarrow}$ -**invariant probability measures** on $\mathcal{A}^{\mathbb{Z}^2}$, that is to say the measures $\mu \in \mathcal{M}(\mathcal{A}^{\mathbb{Z}^2})$ such that $\mu(\sigma_{\uparrow}^{-1}(B)) = \mu(\sigma_{\rightarrow}^{-1}(B)) = \mu(B)$ for all $B \in \mathfrak{B}$. For a continuous application $F : \mathcal{A}^{\mathbb{Z}^2} \rightarrow \mathcal{A}^{\mathbb{Z}^2}$, denote $F\mu$ the image of the measure μ by F : $F\mu(X) = \mu(F^{-1}(X))$.

Definition (Bernoulli measure).

The Bernoulli measure $\mu_\lambda \in \mathcal{M}_\sigma(\mathcal{A}^{\mathbb{Z}^2})$ associated with $\lambda = (\lambda_a) \in [0; 1]^{\mathcal{A}}$ such that $\sum_{a \in \mathcal{A}} \lambda_a = 1$ is defined as:

$$\forall u \in \mathcal{A}^{\mathbb{U}}, \mu_\lambda([u]) = \prod_{(i,j) \in \mathbb{U}} \lambda_{u_{i,j}}.$$

Definition (μ -limit set).

Let $F : \mathcal{A}^{\mathbb{Z}^2} \rightarrow \mathcal{A}^{\mathbb{Z}^2}$ be a CA and μ an initial probability measure. The μ -limit set of F $L_\mu(F)$ is defined by:

$$u \in L_\mu(F) \iff F^t \mu([u]) \xrightarrow[t \rightarrow \infty]{} 0.$$

1.3 Computability

The standard Turing machine model has access to a one-dimensional working tape than can be infinite on one or both sides. We consider in this paper that the machines have access to a two-dimensional tape infinite in all directions, in order to simplify some constructions. The only difference is that the computing head, when reading the current state and the letter on the tape at its current location, has the ability to move in four different directions: $\uparrow, \downarrow, \rightarrow, \leftarrow$. This model remains exactly as powerful as a Turing machine.

Definition (Computable sequence of patterns).

A sequence of patterns $(u_n)_{n \in \mathbb{N}} \in (\mathcal{A}^*)^{\mathbb{N}}$ is computable if there exists a Turing machine that, given as input an integer n written in binary, stops and outputs u_n .

In the previous definition, the Turing machine's alphabet contains at least \mathcal{A} and $\{0, 1\}$. We can assume the input is written left to right on row 0 surrounding by a special blank state.

Proposition 1.1. *Let $F : \mathcal{A}^{\mathbb{Z}^2} \rightarrow \mathcal{A}^{\mathbb{Z}^2}$ be a CA and $\mu \in \mathcal{M}_\sigma(\mathcal{A}^{\mathbb{Z}^2})$ be the uniform Bernoulli measure. Then there is a computable sequence of square patterns $(w_i)_{i \in \mathbb{N}}$ such that*

$$u \in L_\mu(F) \iff \text{Freq}(u, w_i) \xrightarrow[i \rightarrow \infty]{} 0.$$

The sequence is built using de Bruijn tori, a combinatorial object constructed explicitly in [8]. Due to space constraints, the proof is in the appendix.

2 Main theorem

Theorem 2.1. *Let μ be the uniform Bernoulli measure over \mathcal{A} and $(w_i)_{i \in \mathbb{N}}$ a computable sequence of square patterns. Then there exists an alphabet $\mathcal{B} \supseteq \mathcal{A}$ and a cellular automaton F over \mathcal{B} such that:*

$$u \in L_\mu(F) \iff \text{Freq}(u, w_i) \xrightarrow[i \rightarrow \infty]{} 0.$$

This theorem along with Proposition 1.1 gives hence a characterization of all μ -limit sets for μ the uniform Bernoulli measure.

The proof of the theorem relies on an explicit construction and we will effectively prove the result by describing the CA.

Similarly to what was done for one-dimensional CA in [2, 5], the idea is, starting from some random configuration according to a measure μ , to build a partition of connex subsets of the plane using auxiliary states. In each subset, independently, each w_i is computed successively and concatenated copies of it are written over all the subset. To ensure the density of auxiliary states tends to 0, they merge progressively in a controlled manner, offering more space for computation.

3 Construction

3.1 Overview

First, we present a sketch of the different steps of the construction corresponding to a computable sequence of patterns $(w_i)_{i \in \mathbb{N}}$. The alphabet \mathcal{B} is the product of different layers, each layer being used for a different auxiliary process, plus two special states (seed and heart). The **main layer** is the writing layer whose alphabet is \mathcal{A} ; each other layer uses a different alphabet containing a blank symbol $\#$ corresponding to the absence of information. Hence we have $\mathcal{A} \subset \mathcal{B}$ up to the bijection $a \leftrightarrow (a, \#, \dots, \#)$.

- Colonising the space: Section 3.2.

Starting from a random configuration drawn according to μ , we first want to “clean” the randomly generated content of the auxiliary layers. \mathcal{B} contains a **seed** state $\boxed{*}$. Each seed, at time 1, erases the contents of a small area around it and give birth to membranes growing in every direction except when they meet other membranes. They erase all information contained in the auxiliary layers and membranes faking life which are recognized with the help of age counters.

- Internal metabolism: partitioning the cleaned space. Section 3.3.1.

Each seed gives birth to a heart $\boxed{\heartsuit}$ that will be the core of a living organism. Every organism owns an age counter making sure they are all synchronized. Regularly, the organism around each living heart will grow in each direction until it meets a fellow organism, thus claiming its territory.

- Internal metabolism: fighting for survival. Section 3.3.2.

Organisms need to become larger and larger through time, hence we regularly have to remove some of the hearts. When two hearts are too close, one of them is removed to ensure that the distance between hearts is large and tends to infinity.

- Internal metabolism: Computing and writing. Sections 3.3.4 and 3.3.5.

In each organism, when the territory is established, some w_n is computed and then written all over the territory. Copies of it will hence cover the cleaned surface.

Throughout this article, t refers to the number of steps since time 0.

3.2 Colonisation of the space

3.2.1 Growing squares

There is a particular **seed** state $\boxed{*}$ that can be present only in the initial configuration. It is the only relevant information in the initial configuration. Every occurrence of $\boxed{*}$ triggers the birth at time 1 and subsequent growth of a living square-shaped membrane (initially forming a 5×5 cells square).

If seeds are too close from each other and do not have enough space to form the initial organism, the northernmost seed is destroyed (westernmost in case of a tie).

A layer of the alphabet, called **cleaning layer** is dedicated to the membrane growth and cleaning process. The membrane spreads slowly to the outside, thanks to a respiration process that "pushes" the membrane to the outside. A membrane is a boundary between its inside and the outside, thus defining the direction in which it expands. To each point of the membrane is associated a binary counter that keeps track of its age (see Figure 1).

Definition (Redundant binary basis).

Let $c = c_{n-1} \dots c_0 \in \{0, 1, 2\}^n$ be a counter. The **value** of c is $\sum_{i=0}^{n-1} c_i 2^i$ (reverse order). Since $2 = 10$, 2 can be seen as a 0 with a carry.

At each step, the counters are incremented by adding one to the least significant bit and the carries are propagated along the counter, which can be done in a local manner ($02 \rightarrow 10, 12 \rightarrow 20$).

If the membrane has sides of length n , there are n such counters on each side with the same value, with superpositions of two of them in the cells near the corner. As they grow, they need more than one cell and form a band of growing width along the membrane as shown in Figure 1. For a living membrane, the counters are created with value 0 at step $t = 1$, ensuring their age is the current time minus 1. In the other cases, the membrane and counters already existed at time $t = 0$ (with value at least 0), which means they appear older than living membranes.

This counter is used to control the speed of the membrane. The respiration consists in taking a step forward (according to the direction of the membrane) each time the age of the counter is the exact square of an integer. The successive squares are computed under the counter, on the computation layer, using a space $O(\log t)$ if t is the age of the membrane.

We can define three kinds of membranes:

Living membranes which were created by a seed, and whose counters all have value $t - 1$;

Dead membranes which have some incoherence (not closed, different counter values, no square computation...) and self-destructs when realising this;

Zombi membranes which are perfectly coherent despite not being created by a seed, and whose counters all have the same value $t' > t - 1$.

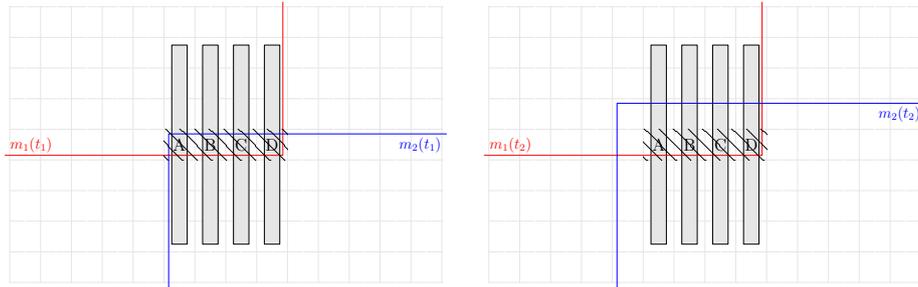


Figure 2: At time t_1 , the membranes m_1 and m_2 meet on cells A , B , C and D . The counters are represented by grey areas. At t_2 , when the comparison is finished, one of the squares may have grown (here m_2).

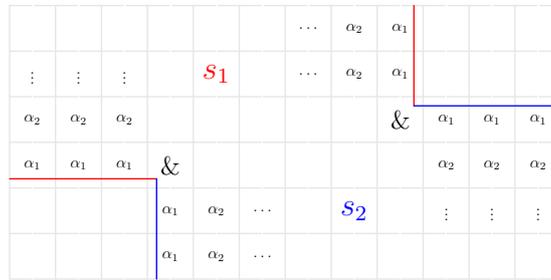


Figure 3: At the end of the comparison, if membrane counters share a common value, the common part of their boundaries is erased and $\&$ symbols mark the corners.

The membranes have the same age: they are both alive or both zombi. In any case, both membranes turn into a single one as shown in Figure 3. Some $\&$ symbols are written at the corners, so that, when both sides will grow again, they remember they are part of the same membrane.

A membrane is younger: the oldest one is zombi and can be safely destroyed. A death signal $\&$ spread in both directions along the oldest membrane, erasing it. The surviving membrane resumes its growth, with its age counters still accurate. The same happens if a membrane grows twice, disrupting the comparison process.

Notice that only the membrane and not the "insides" of the zombi are cleaned since it can contain other living membranes.

None of the signals or processes described in the following sections can enter or leave a membrane, or interact with it or counters, except if explicitly mentioned.

For $t \in \mathbb{N}$, denote

$$Pr(t) = \{F^t(c) \mid c \in \mathcal{B}^{\mathbb{Z}^2}, \exists(i, j) \in \mathbb{Z}^2, d_\infty((i, j), (0, 0)) \leq \lfloor \sqrt{t} \rfloor, c_{ij} = \& \}$$

the set of images of configurations containing a seed $\&$ at distance $\lfloor \sqrt{t} \rfloor$ at most of $(0, 0)$. As μ is the uniform Bernoulli measure, the following lemma is clear:

Lemma 3.2. $F^t \mu(Pr(t)) = 1 - (1 - \mu(\boxed{*}))^{(2\sqrt{t}+1)^2} \rightarrow_t 1$

This means that, with probability 1, for almost any configuration the central cell eventually belongs to the insides of a living membrane.

3.3 Working in the clean surface

We now consider only the protected area, which is the union of all insides of living membranes. Thus every construction presented in this section remains inside this area and stops if it reaches the membrane. They will take place on four new layers: the **age**, **partitioning**, **computating** and **writing** layers.

At some time $t_n = K^{2n}$, $n \in \mathbb{N}$ for some integer K that will be specified later, various operations are performed simultaneously inside all membranes. First, a simulated Turing machine computes w_n . Then, repeated copies of w_n are copied everywhere inside the membrane. Meanwhile, the heart checks that it is not too close to a neighbour, and one of them is deleted if it is the case.

These operations all happen in the time between t_n and $t_{n+1} - 1$, which is called the n^{th} generation.

3.3.1 Claiming its territory

At time 1, while creating a membrane, each seed $\boxed{*}$ transforms itself into a **heart** $\boxed{\heartsuit}$. Any heart is the centre of an **organism** to which it provides life. At the same time, a binary counter is given to each heart, thus giving it the knowledge of its age. This age is exactly the same for any heart inside a living membrane. This counter is the only thing contained in the age layer.

In the rest of this section, only the partitioning layer is concerned.

At time t_n , every heart send signals at speed 1 in each direction until they meet a fellow signal, in which case they disappear and the symbol $\boxed{\#}$ is written where they met. These signals erase everything on the partitioning and computating layers but disappear if they reach a membrane. In this case, $\boxed{\#}$ is written along the membrane. The **territory** of heart $H \in \mathbb{Z}^2$ is the largest set of 4-connected cells containing H that does not contain the symbol $\boxed{\#}$. An organism is composed of a heart and its territory.

Simultaneously, at $t = t_n$, signals leave H and draw the **body** of H : a square of size $2n+1$ centered in H . The body is supposed to be entirely in the territory of H ; if not, the organism is in conflict with every other organism whose body superimpose with its own. At the end of each generation, we will make sure there does not remain any conflict by removing some of the hearts.

Thus, the global dynamics partitions the protected space by redefining territories during each generation, then resolves conflicts: during the n^{th} generation, the distance between two surviving hearts is at least $2n - 1$ (remember we use the distance d_∞).

3.3.2 Choosing its destiny

In this section, we describe conflicts. To get organisms larger and larger through time, we want them to contain at least their entire body, whose size depends of the current generation. We need as well to control the growth of the organisms to avoid the case of too large ones. Indeed, we have to write the computed pattern all over the organism before the beginning of the next generation. Thus, if at

some step a chain of conflicts between organisms appear, we do not want to erase all hearts simultaneously.

To avoid this, we add an algorithmic device and give to each heart some bit of information with the constraint that these bits have to be mutually independent at any given time. Then, for each conflict between two organisms, we choose the one to delete thanks to the sum of their two random bits.

First, we use two versions of the state $\boxed{*}$ in the initial configuration: $\boxed{*}_0$ and $\boxed{*}_1$. This bit is transmitted to $\boxed{\heartsuit}$ which has two versions $\boxed{\heartsuit}_0$ or $\boxed{\heartsuit}_1$. In both cases, we keep the notations $\boxed{*}$ and $\boxed{\heartsuit}$ when the value of the bit does not matter. The bit is also known by the whole boundary of the corresponding organism.

Second, note that, given some heart H living at generation n , the conflicting hearts are at distance $2n - 1$ or $2n$ or they would have conflicted before. Thus, they all belong to a square of side $4n + 1$ centered in H . The distance between each other is also $2n - 1$ or $2n$, hence there are at most 8 simultaneous conflicts, one at most in each eighth part of the plane centered in H : NNE, ENE, ESE, SSE, SSW, WSW, WNW and NNW.

To ensure that independence remains true, each time a heart is deleted, it should give some new information to its killer. Hence, we give 8 other binary bits to each seed, and therefore to each heart. Each eighth part of the territory's boundary carries one of these reserve bits alongside with the main one.

During the n^{th} generation, when two organisms O and O' of hearts respectively $\boxed{\heartsuit}_b$ at (x, y) and $\boxed{\heartsuit}_{b'}$ at (x', y') meet, the sum $\beta = b \oplus b'$ is computed where the boundaries meet. If $\beta = 0$ then the northernmost heart wins (westernmost in case of a tie) and the other way around if $\beta = 1$. Then the boundary of the killed organism (say O') offers its reserve bit b_r to the winner whose main bit becomes $b \oplus b_r$. If some organism kills many others simultaneously (at most 8), it sums all the reserve bits that are given to its. The key point is that all main bits are and remain independent. This is ensured since the reserve bits are not used until they pass to the winner.

On the other hand, a death signal is sent to the heart of the loser, which dies at the reception. This does not interrupt the processes of computation or copying that will be described later, but the organism will not grow again and signals from other hearts will erase it during the next generation.

Definition. Define the *radius* r of an organism as the largest distance from a cell inside its territory to its heart. The territory of the organism is hence bounded by $4r^2$.

Lemma 3.3. *There exists a constant K , such that $p_n \rightarrow_n 1$, where p_n is the probability that at least one living heart remains in a square of radius K^n during the n^{th} generation.*

Proof. Denote $q_n, n \in \mathbb{N}$ the probability for a cell to be a living heart during generation n . For $n = 0$, $q_0 > 0$ is a constant given by μ . Then, during each generation $k \leq n$, a heart survives with probability at least $(1/2)^8$ ($1/2$ for each conflict). Hence $q_n \geq q_0 * (1/2)^{8n}$.

Two different cells have each independently probability q_n to be a heart as long as there is no chain of conflicts between them. At generation n , they have been affected only by hearts at distance n^2 at most. So there are $d_n = \lfloor (2K^n + 1)/(2n^2 + 1) \rfloor^2$ independent cells in a square of radius K^n .

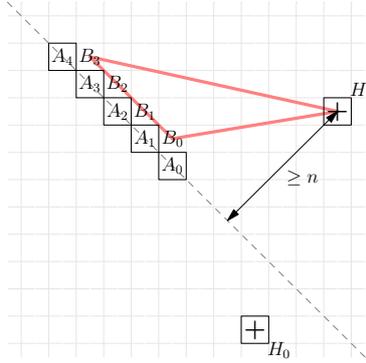


Figure 4: Two hearts H_1 and H_2 are conflicting. Cells A_0 to A_4 form the common boundary of their territory. The red triangle is a set of cells inside the territory of H_1 .

Now we have $1 - p_n \leq (1 - q_n)^{dn}$. This tends to 0 for $K \geq 17$. □

This lemma means that we only need to consider organisms of radius less than K^n . The other ones are sufficiently sparse.

Definition. An organism is said to be *healthy* during the n^{th} generation when its radius is less than K^n (K being given in the previous lemma).

3.3.3 Shape of organisms

Lemma 3.4. *If a cell A is in the organism of heart H , then each cell B such that $d_\infty(B, H) \leq d_\infty(A, H) - d_\infty(A, B)$ is in the same organism.*

Proof. The triangle inequality gives the result automatically, for any other heart H' :

$$d_\infty(B, H) \leq d_\infty(A, H) - d_\infty(A, B) \leq d_\infty(A, H') - d_\infty(A, B) \leq d_\infty(B, H').$$

□

Lemma 3.5. $F^{t_n} \mu(\llbracket \# \rrbracket) \cap Pr(t_n) = O(1/n)$

Proof. Given $n \in \mathbb{N}$, consider the set of cells containing state $\llbracket \# \rrbracket$ at time t_{n+1} within the protected area. It is possible to cut this set into horizontal, vertical or diagonal segments such that each one of them is the common boundary of two specific hearts. When two hearts claim their territory, they send signals in every direction at speed one. These signals may eventually cross to give birth to the boundary. Except if they cross exactly in their corners (hence four cells for each organism, which is negligible), the length of their common boundary is at least 2. Consider one of these boundary segments containing cells $\{A_0, A_1, \dots, A_k\}$ and denote H_0 and H_1 the associated hearts.

The proof is illustrated on Figure 4 in the case of a diagonal segment. Denote d the line supporting the segment, as $d_\infty(H_0, H_1) \geq 2n$, $\exists j \in \{0, 1\}$ such that $d_\infty(H_j, d) \geq n$. Denote O_j the organism centered in H_j . Since A_0, A_1, \dots, A_k

are on the boundary of O_j , there exist distinct points B_0, B_1, \dots, B_{k-1} adjacent to A_0, A_1, \dots, A_k and inside O_j .

Claim 3.6. *Every cell inside the triangle $B_0B_{k-1}H_j$ is inside O_j .*

Proof. For any such cell x , there exists $l \in [0, k-1]$ such that $d_\infty(H_j, B_l) = d_\infty(H_j, x) + d_\infty(x, B_l)$, hence, using Lemma 3.4, x belongs to O_j . \square

There are $\lfloor (k-1)(n-1)/2 \rfloor$ cells in the triangle $B_0B_{k-1}H_j$, which means that for each cell of the boundary segment, we produced $O(n)$ cells inside an organism.

Any cell inside an organism can be attached this way at two segments at most (the border of the triangle can be shared). Thus, for any cell containing $\boxed{\#}$, there are at least $\Theta(n)$ cells that do not contain $\boxed{\#}$, hence $F^{t_n} \mu(\boxed{\#}) \cap Pr(t_n) = O(1/n)$. \square

3.3.4 Computing

In this section, we deal only with the computing layer. At time $t_n, n \in \mathbb{N}$, at each heart of an organism, the same computation starts. While signals leave the heart to determine the boundaries of their territory, other signals draw the limits of a square of side \sqrt{n} whose downleft corner is the heart. This is the space allowed for computation. The heart creates a Turing machine head and the computation starts. It has to remain in this space and halt in less than K^{2n} .

Without loss of generality, we can choose the computable sequence of patterns $(w_i)_{i \in \mathbb{N}}$ such that w_n is the pattern computed during the n^{th} generation. Indeed, we can transform the original sequence by repeating each pattern until there is enough space and time to compute the following one. Denote \mathbb{U}_n the support of w_n and l_n its size: $\mathbb{U}_n = \text{supp}(w_n) = [0, l_n] \times [0, l_n]$. Considering the space allowed for computation, we have that $l_n \leq \sqrt{n}$.

3.3.5 Copying

Finally, our concern is the copying layer. After computing a pattern on the computing layer of an organism, we will write copies of it over the whole territory of this organism.

During the n^{th} generation, the computation takes less than K^{2n} steps, which leaves $K^{2n+2} - K^{2n}$ steps before t_{n+1} . We will show that this is enough to periodically write copies of the result all over the organism, as long as the organism is healthy.

Consider an organism of heart $H = (x_H, y_H)$ during generation n . We first write 4 copies of w_n around H at $(x_H - l_n, y_H - l_n) + \mathbb{U}_n$, $(x_H - l_n, y_H) + \mathbb{U}_n$, $(x_H, y_H - l_n) + \mathbb{U}_n$ and $H + \mathbb{U}_n$. To copy a square, a machine copies all the states sequentially. First, the sides of the squares are marked on the copying layer with a state $\boxed{\square}$ (this takes $O(l_n)$ steps using counters initialised with value l_n), then the machine needs $2l_n$ steps to go to the copy emplacement, make the copy and come back. There are l_n^2 cells to copy, hence the whole process of copying a square takes $O(l_n^3)$ steps.

Starting with these 4 copies of w_n , 4 different copying processes will take place, each one in its quarter of the plane: north-east, north-west, south-west and south-east. We only detail the process in the north-east quarter.

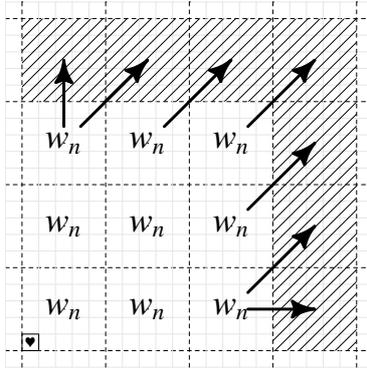


Figure 5: The square pattern is copied all over the whole territory both on axes and along diagonals, starting from the heart.

The base square is copied along the vertical and horizontal axes until it reaches the limit of the territory. Simultaneously, each of these copies replicates itself in diagonal towards the north-east. This way, the whole territory is eventually covered with copies of the computed pattern w_n . The set of states \square draw a grid of step l_n . The copying process is actually a wave starting at the heart of the organism and extending the area where the pattern w_n is written. See Figure 5.

Lemma 3.7. *For any healthy organism, copying takes less than $O(nK^n)$ timesteps during the n^{th} generation.*

Proof. Consider a healthy organism, as the radius is bounded by K^n and the grid step is l_n , there are sequences of at most K_n/l_n square copies to do in each quarter. Each one of these copies requires $O(l_n^3)$ steps, hence the total copy time is $O(nK^n)$ (recall $l_n \leq \sqrt{n}$). \square

Lemma 3.8. *During the n^{th} generation, any cell in a healthy organism that was not reached by the copying process is at distance \sqrt{n} or less of the boundary of the territory.*

Proof. Again, we prove it in the north-east quarter, the proof is symmetric in the other cases. Take a cell A in the territory of a healthy organism and at distance more than l_n of the boundary of the territory. A is in a square S of the \square grid (or would be by extending the grid). Thanks to the hypothesis we know that S entirely belongs to the organism. The copy process reached S , arriving from a square S' at the south, east or south-east of S depending of the position of S . Now, according to Lemma 3.4, S' entirely belongs to the organism.

We can this way go recursively all the way back to the heart, and the copy process is necessarily successful at each step. \square

4 Proof of the main theorem

We saw in previous sections, that a configuration tends to contain only healthy organisms, and that in a healthy organism, computing and copying can be both

achieved in less than $t_{n+1} - t_n$ timesteps. Hence we can conclude.

Proof. Given a sequence $(w_n)_n$, we build the cellular automaton F over the alphabet \mathcal{B} as described in the previous sections.

Suppose $t = t_{n+1} - 1, n \in \mathbb{N}$. First, if $s \in \mathcal{B} \setminus \mathcal{A}$, a cell can have state s if it is:

- outside the protected area, use Lemma 3.2;
- outside a healthy organism, use Lemma 3.3;
- in the border of a healthy organism, use Lemma 3.5;
- in the computation area of an organism, which are negligible since this area is a square of side \sqrt{n} in territories that contain a square of side n ;
- in the grid drawn in each territory (states $\overline{\mathcal{G}}$), negligible as well since the grid occupies less than $4l_n$ cells in each square of side l_n .

Therefore $L_\mu(F) \subseteq \mathcal{A}^*$.

Now, we show that we only need to consider the squares of the grid entirely located in a healthy organism. As said before, it is enough to restrict ourselves to healthy organisms. Every square that is only partially inside a healthy field is located into a band of width less than \sqrt{n} adjacent to the boundary of the field, hence there are at most $O(1/\sqrt{n})$ such cells thanks to Lemma 3.2. As we forced $i \leq \sqrt{n}$, we can effectively neglect those partial squares. In any other square, thanks to Lemma 3.8, we know that the copy was achieved successfully.

For all these reasons, for a square pattern u , $F^{t_n} \mu([u]) \sim_{n \rightarrow \infty} \text{Freq}(u, w_n)$.

Moreover, during the n^{th} generation, while the copying process is engaged but not finished, some part of the main layer contains copies of w_n and the rest is still filled with copies of w_{n-1} , hence, for some $0 \leq \alpha \leq 1$:

$$F^t \mu([u]) \sim_{n \rightarrow \infty} (\alpha F^{t_n} \mu([u]) + (1 - \alpha) F^{t_n} \mu([u]))$$

□

Perspectives

As for the one-dimensional case, we have a characterization of all subshifts that are μ -limit sets of CA. Some corollaries can be derived from this result, but the main open problem is to generalize it to larger classes of measures. In dimension 1, the difference is that there is no need for a trick such as the one used in Section 3.3.2 to resolve conflicts while avoiding erasing too many hearts. As this trick only works with the uniform Bernoulli measure, hence, a better understanding of the dynamics of disparition of the hearts should allow to generalize the result.

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A Proof for Proposition 1.1

Proof. In [8], Hurlbert and Isaak provided a construction of $(k^{2n^2}, k^{2n^2}, n, n)_k$ -de Bruijn tori for every integers k and n , that is, square patterns $T_n \in \{0, \dots, k\}^{2^{2n^2} \times 2^{2n^2}}$ such that all square subpatterns of side length at most $2n$ are present with the same frequency (assuming the subpattern are allowed to “wrap around” the border of the bigger pattern). Since the construction in [8] is explicit, one can devise an algorithm that, on input $n \in \mathbb{N}$, stops and outputs T_n .

Assume for clarity that the neighbourhood of F is $\{-1, 0, 1\}^2$ and let $f : \mathcal{A}^{\{-1, 0, 1\}^2} \rightarrow \mathcal{A}$ be its local rule. Define the image of a square pattern u under F , as an abuse of notation:

$$F : \begin{array}{ccc} \mathcal{A}^{n \times n} & \rightarrow & \mathcal{A}^{n-2 \times n-2} \\ u = (u_{i,j})_{0 \leq i,j \leq n} & \mapsto & (F(u_{i \pm 1, j \pm 1}))_{1 \leq i,j \leq n-1} \end{array}$$

By definition, it is clear that for any square pattern $u \in \mathcal{A}^{n \times n}$,

$$F\mu([u]) = \sum_{\substack{v \in \mathcal{A}^{n+2 \times n+2} \\ F(v)=u}} \mu([v]) = \sum_{\substack{v \in \mathcal{A}^{n+2 \times n+2} \\ F(v)=u}} \frac{1}{|\mathcal{A}|^{n^2}}.$$

Since all square patterns u of side length at most $2n$ are present with uniform frequency $\mu([u]) = \frac{1}{|\mathcal{A}|^{n^2}}$ in T_n , all square patterns of side length at most $2n-2$ are present with frequency $F\mu([u])$ in $F_{tor}(T_n)$ (where F_{tor} corresponds to the previous function applied in a toroidal manner, i.e., “wrapping around at the border”). This process can be iterated.

Now define $(w_n)_{n \in \mathbb{N}} = (F_{tor}^n(T_n))_{n \in \mathbb{N}}$. The sequence w_n is computable by computing u_{2n} and applying the local rule of F . We show that this sequence satisfies the requested property.

Let $u \in \mathcal{A}^{n \times n}$.

$$\begin{aligned} u \in L_\mu(F) &\Leftrightarrow \exists (t_i)_{i \in \mathbb{N}}, \exists d > 0, F^{t_i} \mu([u]) \rightarrow d \\ &\Leftrightarrow \exists (t_i)_{i \in \mathbb{N}}, \exists d > 0, \Rightarrow Freq(u, w_{t_i}) \rightarrow d, \end{aligned}$$

the second line being obtained by considering that, as soon as $t_i \geq n$, u is present with frequency $F^{t_i} \mu([u])$ in $F_{tor}^{t_i}(T_{t_i})$. Actually, this is not exactly true since our definition of frequency does not allow for “wrapping around” the border as in [8]. However, when determining the frequency of a pattern of side length n in $w_{t_i} \in \mathcal{A}^{2t_i^2 \times 2t_i^2}$, this removes only $n \times 4t_i^2$ possible positions out of $4t_i^4$, and thus does not affect the density by more than $\frac{n}{t_i^2} \rightarrow 0$. \square