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# Definability by Horn formulas and linear time on cellular automata\*

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## Abstract

We establish an *exact* logical characterization of linear time complexity of cellular automata of dimension  $d$ , for any fixed  $d$ : a set of pictures of dimension  $d$  belongs to this complexity class *iff* it is definable in existential second-order logic restricted to *monotonic* Horn formulas with built-in successor function and  $d + 1$  first-order variables. This logical characterization is optimal modulo an open problem in parallel complexity. Furthermore, its proof provides a systematic method for transforming an inductive formula defining some problem into a cellular automaton that computes it in linear time.

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## Introduction

Descriptive complexity provides machine-independent views of complexity classes. Typically, Fagin’s Theorem [5] characterizes **NP** as the class of problems definable in existential second-order logic (ESO). Similarly, Immerman-Vardi’s Theorem [15] and Grädel’s Theorem [8, 9] characterize the class **P** by first-order logic plus least fixed-point, and second-order logic restricted to Horn formulas, respectively. The link between *computational* and *descriptive* complexity can be made as tight as possible [20, 14, 17, 11]. Two of the present authors have proved in [12, 13] that a problem is recognized in linear time on a non-deterministic cellular automaton of dimension  $d$  iff it is definable in ESO logic with built-in successor and  $d + 1$  first-order variables. Is there such a natural characterization in logic for the more interesting deterministic case? This question motivates the present paper.

A number of algorithmic problems (linear context-free language recognizability, product of integers, product of matrices, sorting, . . .) are computable in linear time on cellular automata of appropriate dimension. For each such problem, the literature describes the algorithm of the cellular automaton in an informal way [2, 16]. In parallel computational models, algorithms are often difficult to design. However, the problems they solve can often be simply

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defined inductively. For instance, the product of two integers in binary notation is inductively defined by the classical Horner rule.

The first contribution of this paper is the observation that those inductive processes are naturally formalized by Horn formulas [9]. As our second and main contribution, we notice that for every concrete problem defined by a Horn formula with  $d + 1$  first-order variables ( $d \geq 1$ ), this inductive computation by Horn rules has a precise *geometrical* characterization: It can be modeled as the displacement of a  $d$ -dimensional hyperplane  $H$  along some fixed line  $D$  in a space of dimension  $d + 1$ . Provided we interpret the line  $D$  as a temporal axis, the parallel displacement of  $H$  with respect to  $D$  coincides with a computation of a  $d$ -dimensional cellular automaton. The converse is obvious: a  $d$ -dimensional cellular automaton computation can be regarded as the parallel displacement of a  $d$ -dimensional hyperplane – its set of cells – along the time axis.

In the next section, a logic is designed which captures these inductive behaviors (see Def. 14). Roughly speaking, it is obtained from the logic ESO-HORN tailored by Grädel to characterize  $\mathbf{P}$ , by restricting both the number of first-order variables and the arity of second-order predicate symbols. Besides, it includes an additional restriction – the ‘monotonicity condition’ – that reflects the geometrical consideration above-mentioned. We denote this logic by  $\text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ .

Now we can quote the main result of the paper (Thm. 15): a set  $L$  of  $d$ -pictures can be decided in linear time by a deterministic cellular automaton – written  $L \in \mathbf{DLIN}_{\text{ca}}^d$  – if, and only if, it can be expressed in  $\text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ . For short:

$$\mathbf{DLIN}_{\text{ca}}^d = \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1}). \quad (1)$$

A noticeable interest of this result is the constructive method of its proof. In order to design a cellular automaton that computes a problem in linear time, one has to define inductively the problem with a monotonic Horn formula. The normalized form of the formula is automatically obtained: this *is* the program of the cellular automaton<sup>1</sup>.

The paper is structured as follows: The next section collects the preliminary definitions and gives a precise statement of our main result. In Sec. 2, we establish the left-to-right inclusion of the identity displayed in (1). The rest of the paper is devoted to the converse inclusion, whose proof is far more involved. In Sec. 3 we build a monotonic Horn formula expressing the language of palindromes (a “toy” example) and deduce from it a cellular automaton that recognizes this language in linear time. Sec. 4 generalizes this construction to any problem defined in  $\text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ , thus completing the proof of (1). In Sec. 5, we conclude by arguing for the optimality of our result.

## 1 Preliminaries

### 1.1 Cellular automata, picture languages, linear time

We essentially use the terminology of [10].

► **Definition 1.** A *cellular automaton of dimension  $d$*  ( $d$ -CA or CA, for short) is a quadruple  $\mathcal{A} = (d, \mathcal{Q}, \mathcal{N}, \delta)$ , where  $d \in \mathbb{N}$  is the *dimension* of the automaton,  $\mathcal{Q}$  is a finite set whose

<sup>1</sup> For lack of space, this paper gives only one example of this method on a “toy” problem. However, we explicitly describe two more significant examples in the Appendix: first, the product of boolean matrices; second, the product of integers by the Horner method.

elements are called *states*,  $\mathcal{N}$  is a finite subset of  $\mathbb{Z}^d$  called the *neighborhood* of the automaton, and  $\delta : \mathcal{Q}^{\mathcal{N}} \rightarrow \mathcal{Q}$  is the *local transition function* of the automaton.

► **Definition 2.** A *d-dimensional configuration*  $\mathcal{C}$  over the set of states  $\mathcal{Q}$  is a mapping from  $\mathbb{Z}^d$  to  $\mathcal{Q}$ . The elements of  $\mathbb{Z}^d$  will be referred to as *cells*.

► **Definition 3.** Given a cellular automaton  $\mathcal{A} = (d, \mathcal{Q}, \mathcal{N}, \delta)$ , a configuration  $\mathcal{C}$  and a cell  $c \in \mathbb{Z}^d$ , we call *neighborhood of c in C* the mapping  $\mathcal{N}_{\mathcal{C}}(c) : \mathcal{N} \rightarrow \mathcal{Q}$  defined by  $\mathcal{N}_{\mathcal{C}}(c)(v) = \mathcal{C}(c + v)$ .

From the local transition function  $\delta$  of  $\mathcal{A} = (d, \mathcal{Q}, \mathcal{N}, \delta)$ , we can define the *global transition function* of the automaton  $\Delta : \mathcal{Q}^{\mathbb{Z}^d} \rightarrow \mathcal{Q}^{\mathbb{Z}^d}$  obtained by applying the local rule on each cell, that means  $\Delta(\mathcal{C})(c) = \delta(\mathcal{N}_{\mathcal{C}}(c)) = \delta((\mathcal{C}(c + v))_{v \in \mathcal{N}})$ , for each cell  $c$ .

One identifies the CA  $\mathcal{A}$  with its global rule:  $\mathcal{A}(\mathcal{C}) = \Delta(\mathcal{C})$  is the image of a configuration  $\mathcal{C}$  by the action of  $\mathcal{A}$ . Moreover,  $\mathcal{A}^t(\mathcal{C})$  is the configuration resulting from applying  $t$  times the global rule of  $\mathcal{A}$  from the initial configuration  $\mathcal{C}$ .

► **Definition 4.** For a given cellular automaton: a state  $q$  is *permanent* if a cell in state  $q$  remains in this state *regardless* of the states of its neighbors; a state  $q$  is *quiescent* if a cell in state  $q$  remains in this state if *all* its neighbors are also in state  $q$ .

Cellular automata of dimension  $d$  operate on *d-pictures*.

► **Definition 5.** Let  $\Sigma$  be a finite alphabet. For integers  $d, n \geq 1$ , a *picture of dimension d* (*d-picture*) and *side n over*  $\Sigma$  is a mapping  $p : \llbracket 1, n \rrbracket^d \rightarrow \Sigma$ . We denote by  $\Sigma^{(d)}$  the set of *d-pictures* over  $\Sigma$ . Any subset of  $\Sigma^{(d)}$  is called a *d-picture language* over  $\Sigma$ .

► **Remark.** *d-picture* languages can capture a wide variety of decision problems if the alphabet  $\Sigma$  is sufficiently expressive. For instance, the product problem of boolean square matrices is a 2-picture problem over the three-part alphabet  $\Sigma = \{0, 1\}^3$  that consists of square pictures  $M$  such that the projection of  $M$  over the last part of the alphabet is equal to the product of its projections over the first two parts.

► **Definition 6.** Given a picture  $p : \llbracket 1, n \rrbracket^d \rightarrow \Sigma$ , we define the *picture configuration* associated with  $p$  with *permanent* or *quiescent* state<sup>2</sup>  $q_0 \notin \Sigma$  as the function  $\mathcal{C}_{p, q_0} : \mathbb{Z}^d \rightarrow \Sigma \cup \{q_0\}$  such that  $\mathcal{C}_{p, q_0}(\mathbf{x}) = p(\mathbf{x})$  if  $\mathbf{x} \in \llbracket 1, n \rrbracket^d$  and  $\mathcal{C}_{p, q_0}(\mathbf{x}) = q_0$  otherwise.

► **Definition 7.** Given a *d-picture language*  $L \subseteq \Sigma^d$ , we say that a cellular automaton  $\mathcal{A} = (d, \mathcal{Q}, \mathcal{N}, \delta)$  such that  $\Sigma \subseteq \mathcal{Q}$  with permanent states  $q_a$  and  $q_r$  (accepting and rejecting states) *recognizes*  $L$  with permanent state (quiescent state, respectively)  $q_0 \in \mathcal{Q} \setminus (\Sigma \cup \{q_a, q_r\})$  *in time*  $\tau : \mathbb{N} \rightarrow \mathbb{N}$  (for short,  $\tau(n)$ ) if for any picture  $p : \llbracket 1, n \rrbracket^d \rightarrow \Sigma$ , starting from the configuration  $\mathcal{C}_{p, q_0}$  at time 0, the state of cell  $\mathbf{n} = (n, \dots, n)$  of  $\mathcal{A}$ , called the *reference cell*, is  $q_a$  or  $q_r$  at time  $\tau(n)$  with  $\mathcal{A}^{\tau(n)}(\mathcal{C}_{p, q_0})(\mathbf{n}) = q_a$  if  $p \in L$  and  $\mathcal{A}^{\tau(n)}(\mathcal{C}_{p, q_0})(\mathbf{n}) = q_r$  if  $p \notin L$ .

► **Definition 8.** For  $d \geq 1$ , we call **DLIN<sub>ca</sub><sup>d</sup>** the class of *d-picture problems*  $L$  for which there exist a *d-CA*  $\mathcal{A}$  with quiescent state  $q_0$  and a function  $\tau(n) = O(n)$  such that  $L$  can be recognized by  $\mathcal{A}$  in time  $\tau(n)$ . Such a problem is said to be *recognizable in linear time*.

<sup>2</sup> The condition that each cell outside the input domain  $\llbracket 1, n \rrbracket^d$  remains in a permanent state (resp. quiescent state)  $q_0$  means that the computation space is exactly the set of input cells (resp. is not bounded).

The class  $\mathbf{DLIN}_{ca}^d$  is very robust: it is not modified under many changes: neighborhoods, precise time/space bounds, input presentation, etc. The proof of the first part of our main result will use the following restrictive characterization which is a consequence of a general linear acceleration theorem (see *e.g.* [10, 19]).

► **Lemma 9.** [10]  $\mathbf{DLIN}_{ca}^d$  is the class of  $d$ -picture problems that can be recognized in time  $n - 1$  by a  $d$ -CA of neighborhood  $\mathcal{N}_2 = \{-2, -1, 0, 1, 2\}^d$  with permanent state  $q_0$ .

## 1.2 Picture structures and monotonic Horn formulas

The local nature of cellular automata acting on pictures is captured by logical formulas acting on first-order structures, the so-called *picture structures*, that represent these pictures. Before defining picture structures, let us detail their signatures. Given a dimension  $d \geq 1$  and  $k$  alphabets  $\Sigma_1, \dots, \Sigma_k$ , we denote by  $\mathbf{sg}(d; \Sigma_1, \dots, \Sigma_k)$  the signature below:

$$\mathbf{sg}(d; \Sigma_1, \dots, \Sigma_k) = \{(Q_s^1)_{s \in \Sigma_1}, \dots, (Q_s^k)_{s \in \Sigma_k}, \min, \max, \text{suc}, \text{pred}\}.$$

Here, each  $Q_s^i$  is a  $d$ -ary relation symbol,  $\min$  and  $\max$  are unary relation symbols, and  $\text{suc}$  and  $\text{pred}$  are unary function symbols.

► **Definition 10.** Let  $p_1, \dots, p_k$  be pictures of respective alphabets  $\Sigma_1, \dots, \Sigma_k$ . We assume that the  $p_i$ 's have the *same dimension*  $d$  and the *same side*  $n$ . The *picture structure* of the  $k$ -tuple  $(p_1, \dots, p_k)$  is the structure of signature  $\mathbf{sg}(d; \Sigma_1, \dots, \Sigma_k)$  defined as follows:

$$\mathcal{S}(p_1, \dots, p_k) = \langle \llbracket 1, n \rrbracket, (Q_s^1)_{s \in \Sigma_1}, \dots, (Q_s^k)_{s \in \Sigma_k}, \min, \max, \text{suc}, \text{pred} \rangle.$$

Here,  $n$  is the common side of the  $p_i$ 's. Besides, symbols of  $\mathbf{sg}(d; \Sigma_1, \dots, \Sigma_k)$  are interpreted on  $\mathcal{S}(p_1, \dots, p_k)$  as follows, where we denote the same way a symbol and its interpretation:

- each  $Q_s^i$  is the set of cells of  $p_i$  labelled by  $s$ . Formally:  $Q_s^i = \{a \in \llbracket 1, n \rrbracket^d : p_i(a) = s\}$ ;
- $\min$  and  $\max$  are the singleton sets  $\{1\}$  and  $\{n\}$ , respectively;
- $\text{suc}$  and  $\text{pred}$  are the successor and predecessor functions: that is  $\text{suc}(n) = n$  and  $\text{suc}(a) = a + 1$  for  $a \in \llbracket 1, n - 1 \rrbracket$ ;  $\text{pred}(1) = 1$  and  $\text{pred}(a) = a - 1$  for  $a \in \llbracket 2, n \rrbracket$ .

In the following, we will freely use the natural notation  $x + i$ , for any fixed integer  $i \in \mathbb{Z}$ . It abbreviates  $\text{suc}^i(x)$  if  $i > 0$ , and  $\text{pred}^{-i}(x)$  if  $i < 0$ . For  $i = 0$ , it represents  $x$ .

We will use the usual definitions and notations in logic (see [4, 18, 9]). All formulas considered hereafter belong to *existential second-order logic*. More precisely, we shall focus on the following logic:

► **Definition 11.**  $\mathbf{ESO}^d(\forall^{d+1}, \text{arity}^{d+1})$  is the class of formulas of the form  $\exists \mathbf{R} \forall \mathbf{x} \psi$ , where  $\mathbf{R} = (R_1, \dots, R_r)$  is a tuple of  $(d + 1)$ -ary relation symbols,  $\mathbf{x} = (x_0, \dots, x_d)$  is a  $(d + 1)$ -tuple of first-order variables, and  $\psi$  is a quantifier-free first-order formula of signature  $\mathbf{sg}(d; \Sigma) \cup \mathbf{R}$  for some tuple of alphabets  $\Sigma = (\Sigma_1, \dots, \Sigma_k)$ .

Such a formula involves two sorts of predicate symbols: those of  $\mathbf{sg}(d; \Sigma)$  are called *input predicates* and those of  $\mathbf{R}$  are called *guessed predicates*.

It is proved in [13] that the above logic exactly characterizes  $\mathbf{NLIN}_{ca}^d$ , the non-deterministic counterpart of  $\mathbf{DLIN}_{ca}^d$ . The ‘inclusion’  $\mathbf{DLIN}_{ca}^d \subseteq \mathbf{ESO}^d(\forall^{d+1}, \text{arity}^{d+1})$  immediately follows, but the converse inclusion is quite unlikely, since it entails  $\mathbf{DLIN}_{ca}^d = \mathbf{NLIN}_{ca}^d$ , which in turn implies  $\mathbf{P} = \mathbf{NP}$ . Nevertheless, this engages us in looking for a logic characterizing  $\mathbf{DLIN}_{ca}^d$  inside the logic  $\mathbf{ESO}^d(\forall^{d+1}, \text{arity}^{d+1})$ . A first restriction of this logic is naturally suggested by the Grädel’s characterization of  $\mathbf{P}$  already mentioned in the introduction:

► **Definition 12.**  $\text{ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  brings together formulas  $\exists \mathbf{R} \forall \mathbf{x} \psi$  among  $\text{ESO}^d(\forall^{d+1}, \text{arity}^{d+1})$  whose quantifier-free part  $\psi$  is a conjunction of Horn clauses of the form<sup>3</sup>  $\alpha_1 \wedge \dots \wedge \alpha_m \rightarrow \alpha_0$  such that:

- each premise  $\alpha_1, \dots, \alpha_m$  is
  - either a *guessed atom*  $R(x_0 + i_0, \dots, x_d + i_d)$  with  $R \in \mathbf{R}$  and  $i_0, \dots, i_d \in \mathbb{Z}$ ,
  - or an *input literal*  $I(t_1 + i_1, \dots, t_q + i_q)$  or  $\neg I(t_1 + i_1, \dots, t_q + i_q)$ , with  $I \in \text{sg}(d; \Sigma)$ ,  $t_1, \dots, t_q \in \mathbf{x}$ , and  $i_0, \dots, i_q \in \mathbb{Z}$ ;
- the conclusion literal  $\alpha_0$  is either a ‘constant’ – the boolean  $\perp$  or an input literal – or a *guessed atom*<sup>4</sup> of the restricted form  $R(x_0, \dots, x_d)$  with  $R \in \mathbf{R}$ .

We will see that this new logic still contains  $\text{DLIN}_{ca}^d$  but that here again the converse inclusion is unlikely, as argued in Sec. 5. Whence the necessity of a further restriction of the logic, detailed in Def. 14. For now, let us give some motivation for this restriction.

The first-order part of an  $\text{ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ -formula can be viewed as a program whose execution, on a given picture structure taken as input, computes the guessed predicates from the input ones. Consider for instance the Horn clause  $R(x - 2, y - 1) \wedge R'(x + 1, y - 2) \rightarrow R(x, y)$  built on guessed predicates  $R$  and  $R'$ . It establishes a dependence between the values of the guessed predicates (taken as a whole) at place  $(x, y)$  and their values at place  $(x - 2, y - 1)$ , on one hand, and at place  $(x + 1, y - 2)$ , on the other hand. This notion is formalized by the definition below.

► **Definition 13.** Let  $\Phi = \exists \mathbf{R} \forall x_0, \dots, x_d \psi$  be in  $\text{ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ . A *nonzero* tuple  $(i_0, \dots, i_d) \in \mathbb{Z}^{d+1}$  is an *induction vector* of  $\Phi$  if there exists a Horn clause  $C$  in  $\psi$  and two guessed predicates  $R, R'$  in  $\mathbf{R}$  such that  $C$  includes  $R(x_0, \dots, x_d)$  as its conclusion and  $R'(x_0 + i_0, \dots, x_d + i_d)$  among its hypotheses. The set of induction vectors of  $\Phi$  is called its *induction system*.

The logic involved in the characterization of  $\text{DLIN}_{ca}^d$  that constitutes the core of this paper is defined as follows:

► **Definition 14.** A formula  $\Phi \in \text{ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  with induction system  $\mathcal{S}$  is said *monotonic* and we write  $\Phi \in \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  if there exist  $a_0, \dots, a_d \in \mathbb{Z}$  such that each induction vector  $(v_0, \dots, v_d) \in \mathcal{S}$  fulfils  $a_0 v_0 + \dots + a_d v_d < 0$ . This condition is called the *monotonicity condition*.

In other words, there exists a hyperplane  $a_0 x_0 + \dots + a_d x_d = 0$ , called a *reference hyperplane* of  $\mathcal{S}$ , such that each vector  $(v_0, \dots, v_d) \in \mathcal{S}$  belongs to the same strict half-space determined by this hyperplane, that means  $a_0 v_0 + \dots + a_d v_d < 0$ . One also says that  $\mathcal{S} \subset \mathbb{Z}^{d+1}$  *satisfies the monotonicity condition* w.r.t. the reference hyperplane.

We are now in a position to state formally the main result of the paper:

► **Theorem 15.** For  $d \geq 1$ ,  $\text{DLIN}_{ca}^d = \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ .

The two ‘inclusions’ underlying the above characterization are proved in Sec. 2 and 4.

<sup>3</sup> We will always assume that conjunction has priority over implication.

<sup>4</sup> Alternatively, in Horn formulas, ‘guessed’ predicates and ‘guessed’ atoms can be called more intuitively ‘computed’ predicates and ‘computed’ atoms.

## 2 DLIN<sub>ca</sub> $\subseteq$ mon-ESO-HORN

► **Proposition 16.** For  $d \geq 1$ ,  $\text{DLIN}_{ca}^d \subseteq \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ .

**Proof.** Let  $L \subseteq \Sigma^{(d)}$  be a  $d$ -picture language in  $\text{DLIN}_{ca}^d$ . By Lemma 9, there exists a CA  $\mathcal{A} = (d, \mathcal{Q}, \mathcal{N}_2, \delta)$  of neighborhood  $\mathcal{N}_2 = \{-2, -1, 0, 1, 2\}^d$  that recognizes  $L$  in time  $\tau(n) = n - 1$  with *permanent* state  $q_0$ . Let  $\llbracket 1, n \rrbracket$  denote the interval of the  $n$  instants of the computation of  $\mathcal{A}$  on a  $d$ -picture of side  $n$ ; in particular, the initial and final instants are numbered 1 and  $n$ , respectively. We are going to construct a formula  $\Phi \in \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  that defines  $L$ , i.e., that expresses that the computation of  $\mathcal{A}$  on a  $d$ -picture  $p$  accepts it. It is of the form  $\Phi \equiv \exists(R_s)_{s \in \mathcal{Q}} \forall t \forall \mathbf{c} \psi$  where  $\mathbf{c}$  denotes the  $d$ -tuple of variables  $(c_1, \dots, c_d)$  and, for  $s \in \mathcal{Q}$ , the intended meaning of the guessed atom  $R_s(t, \mathbf{c})$  is the following: at the instant  $t$ , the cell  $\mathbf{c}$  is in the state  $s$ . For simplicity of notation, let us assume  $d = 1$ . Also assume  $n \geq 5$ . The quantifier-free part  $\psi$  of  $\Phi$  is the conjunction of three kinds of Horn clauses:

1. the *initialization clauses*: for each  $s \in \Sigma$ , the clause  $\min(t) \wedge Q_s(c) \rightarrow R_s(t, c)$ ;
2. five kinds of *computation clauses* that compute the state at instant  $t > 1$  of any cell  $c$  according to its possible neighborhoods for  $\mathcal{N}_2 = \{-2, -1, 0, 1, 2\}$  :

(i)  $c = 1$ ; (ii)  $c = 2$ ; (iii) general case  $c \in \llbracket 3, n - 2 \rrbracket$ ; (iv)  $c = n - 1$ ; (v)  $c = n$ .

Let us consider the general case: for any 5-tuple of states  $(s_{-2}, s_{-1}, s_0, s_1, s_2) \in (Q - \{q_0, q_a, q_r\})^5$ , the clause

$$\left( \begin{array}{l} R_{s_{-2}}(t-1, c-2) \wedge R_{s_{-1}}(t-1, c-1) \wedge \\ R_{s_0}(t-1, c) \wedge R_{s_1}(t-1, c+1) \wedge R_{s_2}(t-1, c+2) \end{array} \right) \rightarrow R_{\delta(s_{-2}, s_{-1}, s_0, s_1, s_2)}(t, c)$$

computes the state at any instant  $t > 1$  of any cell  $c$  in the interval  $\llbracket 3, n - 2 \rrbracket$ , which can be tested by the use of  $\neg \min()$  and  $\neg \max()$  in the premises;

3. the *accepting clause*  $R_{q_r}(t, c) \rightarrow \perp$ , which says that the computation does not reject, and hence accepts, since by hypothesis each computation of  $\mathcal{A}$  either accepts or rejects.

By construction,  $\Phi$  belongs to  $\text{ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  and the induction system is  $\{-1\} \times \{-2, -1, 0, 1, 2\}^d$ , which has a reference hyperplane of equation  $t = 0$ . Hence,  $\Phi \in \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ , which completes the proof. ◀

The proof of the converse inclusion,  $\text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1}) \subseteq \text{DLIN}_{ca}^d$ , given in Sec. 4, is much more elaborate. In order to give its main ideas which are essentially of geometrical nature, we now present the inductive definition of a “toy” problem by a monotonic Horn formula from which we will derive a cellular automaton that recognizes the problem.

## 3 From the formula to the automaton: the example of palindromes

Let  $\text{PALINDROME}(\Sigma)$  denote the language of palindromes over a fixed alphabet  $\Sigma$ .

### 3.1 A monotonic Horn formula defining the language of palindromes

Let us prove that  $\text{PALINDROME}(\Sigma)$  is definable in  $\text{mon-ESO-HORN}^1(\forall^2, \text{arity}^2)$ . In addition to the set of input unary predicates  $\text{Input} = \{\min, \max, (Q_s)_{s \in \Sigma}\}$  involved in the picture structure that represents a word  $w = w_1 w_2 \dots w_n \in \Sigma^*$ , we need to consider three guessed binary predicates symbols  $R_=$ ,  $R_<$  and  $R_{\text{noPal}}$ . The first two are inductively enforced to encode, respectively, the equality relation and the usual strict linear order over the domain  $\llbracket 1, n \rrbracket$ . This is done with the clauses  $\theta_1, \dots, \theta_5$  below:

$$\begin{aligned}
\theta_1: & \min(x) \wedge \min(y) \rightarrow R_{=} (x, y); \\
\theta_2: & \neg \min(x) \wedge \neg \min(y) \wedge R_{=} (x-1, y-1) \rightarrow R_{=} (x, y); \\
\theta_3: & \neg \max(x) \wedge R_{=} (x+1, y) \rightarrow R_{<} (x, y); \\
\theta_4: & \neg \max(x) \wedge R_{<} (x+1, y) \rightarrow R_{<} (x, y); \\
\theta_5: & R_{<} (x, x) \rightarrow \perp.
\end{aligned}$$

The (set of) clauses  $\theta_6$  and  $\theta_7$  below inductively define  $R_{\text{noPal}}$  as the set of couples  $(x, y) \in \llbracket 1, n \rrbracket^2$  such that  $x < y$  and the factor  $w_x \dots w_y$  of the input word  $w$  is not a palindrome. Then the clause  $\theta_8$  forces  $w$  to be a palindrome:

$$\begin{aligned}
\theta_6: & R_{<} (x, y) \wedge Q_s(x) \wedge Q_{s'}(y) \rightarrow R_{\text{noPal}}(x, y), \text{ for all } s \neq s' \text{ in } \Sigma; \\
\theta_7: & R_{<} (x, y) \wedge R_{\text{noPal}}(x+1, y-1) \rightarrow R_{\text{noPal}}(x, y); \\
\theta_8: & \min(x) \wedge \max(y) \wedge R_{\text{noPal}}(x, y) \rightarrow \perp.
\end{aligned}$$

In conclusion,  $\text{PALINDROME}(\Sigma)$  is defined by the following formula  $\Phi_{\text{pal}}$ , over the structure  $\mathcal{S}(w) = \langle \llbracket 1, n \rrbracket, (Q_s)_{s \in \Sigma}, \min, \max, \text{succ}, \text{pred} \rangle$  associated with an input word  $w = w_1 \dots w_n$ :

$$\Phi_{\text{pal}} \equiv \exists R_{=}, R_{<}, R_{\text{noPal}} \forall x, y \bigwedge_{i \leq 8} \theta_i.$$

Moreover,  $\Phi_{\text{pal}}$  belongs to  $\text{ESO-HORN}^1(\forall^2, \text{arity}^2)$  and has  $\mathcal{S} = \{(-1, -1), (1, 0), (1, -1)\}$  as its induction system (see Def. 13). Clearly, the system  $\mathcal{S}$  satisfies the monotonicity condition of Def. 14 with the line of equation  $-x + 2y = 0$  as its reference hyperplane. It follows:

► **Proposition 17.**  $\text{PALINDROME}(\Sigma) \in \text{mon-ESO-HORN}^1(\forall^2, \text{arity}^2)$ .

### 3.2 From $\Phi_{\text{pal}}$ to $\mathcal{A}_{\text{pal}}$

It remains to transform the formula  $\Phi_{\text{pal}}$  above into a one-dimensional cellular automaton  $\mathcal{A}_{\text{pal}}$  that recognizes the language  $\text{PALINDROME}(\Sigma)$ . For sake of simplicity, we first ignore the input literals and only take account of guessed atoms in the Horn clauses  $\theta_i$ . Notice that in each clause whose conclusion is a guessed atom  $R(x, y)$ ,  $R \in \{R_{=}, R_{<}, R_{\text{noPal}}\}$ , the guessed atoms occurring as premises have one of the following forms:

$$R'(x, y), \quad R'(x+1, y), \quad R'(x+1, y-1), \quad R'(x-1, y-1).$$

Intuitively, if one regards the set  $D_t = \{(x, y) \in \llbracket 1, n \rrbracket^2 \mid -x + 2y = t\}$  as the line of cells of a one-dimensional CA at instant  $t$ , then the conjunction of the above clauses  $\theta_i$  can be regarded as the transition function of such a CA (see Fig. 1). More formally, in order to introduce the *time* parameter  $t$ , we eliminate one of the variables,  $x$  for example, and we regard the other variable,  $y$ , as the *space* variable  $c$ . That is, one makes the change of variables<sup>5</sup>:  $t = -x + 2y$ ;  $c = y$ .

Let us now explain how the automaton  $\mathcal{A}_{\text{pal}}$  to be constructed can take account of the input literals. For each point  $(x, y) \in \llbracket 1, n \rrbracket^2$ , we call  $\text{state}(x, y)$  the tuple of boolean values of all input and output atoms on  $x$  and  $y$ . That is,

$$\text{state}(x, y) = \left( \begin{array}{c} \min(x), \min(y), \max(x), \max(y), \\ (Q_s(x))_{s \in \Sigma}, (Q_s(y))_{s \in \Sigma}, \\ R_{=}(x, y), R_{<}(x, y), R_{\text{noPal}}(x, y) \end{array} \right),$$

<sup>5</sup> There is an analogy between our method and the so-called *loop-skewing* or *polytope/polyhedron* method in compilation and parallel algorithms [6, 1, 7].



where the values  $R_=(x, y)$ ,  $R_<(x, y)$ ,  $R_{\text{noPal}}(x, y)$ , are deduced by the Horn formula.

By the change of variables  $(x, y) \mapsto (t = -x + 2y, c = y)$  whose converse is the function  $(t, c) \mapsto (x = -t + 2c, y = c)$ , each input atom of the form  $I(x)$  becomes  $I(-t + 2c)$  and each input atom of the form  $I(y)$  becomes  $I(c)$ . The CA we construct has to memorize in each cell  $c$  at instant  $t$  the boolean values  $I(c)$  and  $I(-t + 2c)$ , for  $I \in \text{Input}$ . This can be realized as follows:

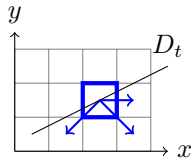
- (a) For each  $I(c)$  (former  $I(y)$ ): the CA *conserves* on cell  $c$  the boolean value  $I(c)$  from an instant  $t - 1$  to the next instant  $t$ ;
- (b) For each  $I(-t + 2c)$  (former  $I(x)$ ): because of the identity  $-t + 2c = -(t - 2) + 2(c - 1)$ , whence  $I(-t + 2c) \equiv I(-(t - 2) + 2(c - 1))$ , the CA only has to *move* to each cell  $c$  at instant  $t$  the boolean value  $I(-(t - 2) + 2(c - 1))$  that is present at instant  $t - 2$  on the cell  $c - 1$ .

All in all, the state of each point  $P = (t, c) = (-x + 2y, y)$  is determined by the states of the following points (as shown on Fig. 2):

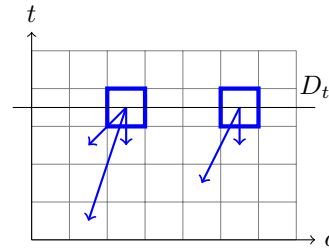
- $P_1 = (-(x - 1) + 2(y - 1), y - 1) = (t - 1, c - 1)$ ,  $P_2 = (-(x + 1) + 2y, y) = (t - 1, c)$  and  $P_3 = (-(x + 1) + 2(y - 1), y - 1) = (t - 3, c - 1)$ , because of guessed atoms, and
- $P_2 = (t - 1, c)$  and  $P_4 = (t - 2, c - 1)$  because of the above items (a) and (b), respectively, for input atoms.

Hence, the state of a cell  $c$  at instant  $t$  is determined by the states of: (i) cell  $c - 1$  at instant  $t - 1$ ; (ii) cell  $c$  at instant  $t - 1$ ; (iii) cell  $c - 1$  at instant  $t - 3$ ; (iv) cell  $c - 1$  at instant  $t - 2$ . Figures 1 and 2 below summarize the effects of the change of variables on the induction system.<sup>6</sup>

It seems that we have achieved the design of an automaton of neighborhood  $\{-1, 0\}$  that recognizes the language  $\text{PALINDROME}(\Sigma)$  in *linear time* since  $t = -x + 2y$  and  $x, y \in \llbracket 1, n \rrbracket$  imply  $-n + 2 \leq t \leq 2n - 1$ . However, it remains to describe both the initialization and the end of the computation.



■ **Figure 1** Induction system for guessed atoms before the change of variables.



■ **Figure 2** Induction system after the change of variables for guessed atoms (left) and input atoms (right).

**The result and the initialization of the computation:**

The result of the computation is **accept** or **reject** according to whether  $\mathcal{S}(w)$  does or does not satisfy the formula  $\Phi_{\text{pal}}$ , where  $\mathcal{S}(w)$  is the structure  $\langle \llbracket 1, n \rrbracket, (Q_s)_{s \in \Sigma}, \text{min}, \text{max}, \text{y}, \text{pred} \rangle$  associated with the input word  $w = w_1 \dots w_n$ . As this is testified by the clause  $\theta_8 = \text{min}(x) \wedge \text{max}(y) \wedge R_{\text{noPal}}(x, y) \rightarrow \perp$ , on the point of coordinates  $(x = 1, y = n)$  which become

<sup>6</sup> At first glance, Conditions (iii) and (iv) seem to contradict the requirement that the state of any cell  $c$  of a CA at instant  $t$  should be determined by the *only* states of its neighbour cells at the *previous* instant  $t - 1$ . However, we can overcome the problem by using the ability of a cell to memorize at any instant  $t$  its states at instants  $t - 1$  and  $t - 2$  with a finite number of states.

after the change of variables ( $t = -x + 2y = 2n - 1$ ,  $c = y = n$ ), the acceptance/rejection can be read on the state of cell  $c = n$  at the instant  $t = 2n - 1$  so that the final state  $q_a$  or  $q_r$  is obtained at the following instant  $2n$ .

The initialization of the computation requires some care in connection with the items (a) and (b) of the previous paragraph, about the input bits:

- (1) *Initializing each  $I(c)$  (former  $I(y)$ ):* The state of each cell  $c \in \llbracket 1, n \rrbracket$  at the instant just before  $-n + 2$ , i.e. at instant  $-n + 1$ , should store the boolean value  $I(c)$ , for each  $I \in \text{Input}$ .
- (2) *Initializing each  $I(-t + 2c)$  (former  $I(x)$ ):* Because of the correspondence  $x = -t + 2c$  or, equivalently,  $c = (x + t)/2$ , for all  $x \in \llbracket 1, n \rrbracket$ , the boolean value  $I(x)$  should be stored in the state of the cell  $c = (x + t)/2$  at the maximal instant  $t < -n + 2$  such that  $(x + t)/2$  is an *integer*; that is the cell  $c = (x - n)/2$  at instant  $-n$  if  $x - n \equiv 0 \pmod{2}$  and  $c = (x - n + 1)/2$  at instant  $-n + 1$  if  $x - n \equiv 1 \pmod{2}$ : see Fig. 3.

The two configurations at the successive instants  $-n$  and  $-n + 1$  described in items (1) and (2) are called *initialization* configurations. By construction, the space of both configurations – their informative cells, i.e. those in non-quiescent states – is included in the interval  $\llbracket -\lceil n/2 \rceil + 1, n \rrbracket$ .

According to our conventions, the initial configuration of the automaton should be the configuration  $\mathcal{C}_{w,q_0}$  associated with the input word  $w$ , as defined in Def. 6. However, one can design a routine which, starting from configuration  $\mathcal{C}_{w,q_0}$  (with quiescent state  $q_0$ ), computes the two initialization configurations by using the classical technique of signals in CA's (see [16]) as shown on Fig. 4. By a careful examination of this figure, we precisely observe that this precomputation is performed on the interval of cells  $\llbracket -\lceil n/2 \rceil + 1, n + 1 \rrbracket$  during the time interval  $\llbracket -3n, -n + 1 \rrbracket$ .<sup>7</sup>

We have now achieved the design of a cellular automaton  $\mathcal{A}_{\text{pal}}$  that recognizes in *linear time* the language  $\text{PALINDROME}(\Sigma)$  from the monotonic Horn formula  $\Phi_{\text{pal}}$  that defines it.

#### 4 mon-ESO-HORN $\subseteq$ DLIN<sub>ca</sub>

The main problem we have to deal with in the general case as in the previous example is the integration of the input to the computation of the CA to be constructed. For that purpose, we will need the following technical lemma whose proof is given in Appendix A.2:

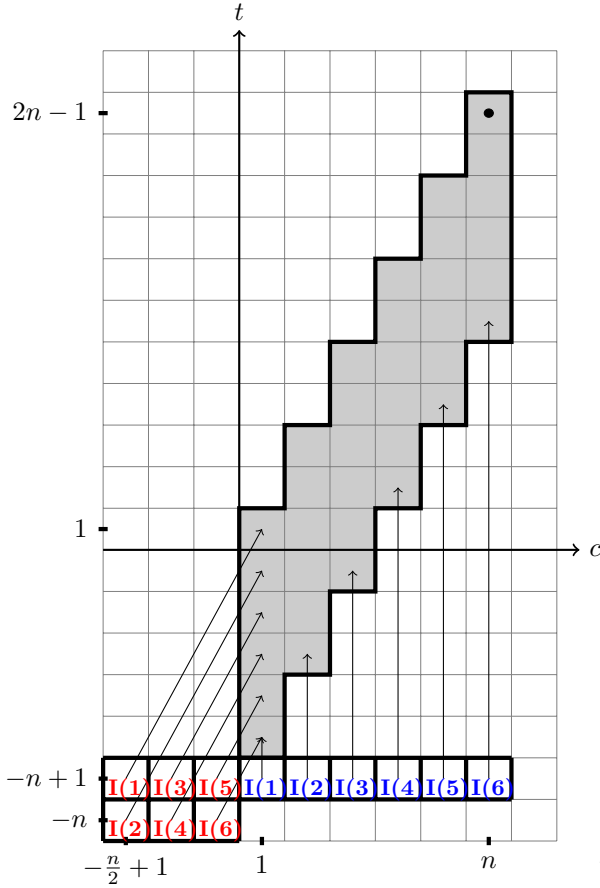
► **Lemma 18.** *Let  $\mathcal{S} \subset \mathbb{Z}^{d+1}$  be an induction system satisfying the monotonicity condition w.r.t. some reference hyperplane. Then,  $\mathcal{S}$  has another reference hyperplane of equation  $a_0x_0 + \dots + a_dx_d = 0$  where each coefficient  $a_i$  ( $i \in \llbracket 0, d \rrbracket$ ) is a non-zero integer.*

We are now ready to prove the most difficult inclusion of Thm. 15:

► **Proposition 19.** *For each  $d \geq 1$ ,  $\text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1}) \subseteq \text{DLIN}_{ca}^d$ .*

**Proof.** Let  $L$  be a  $d$ -picture language defined by a formula  $\Phi \equiv \exists R_1 \dots \exists R_r \forall x_0 \dots \forall x_d \psi$  in  $\text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  with an induction system  $\mathcal{S}$  and a reference hyperplane

<sup>7</sup> Notice that our numbering of instants is not canonical. It is only a convenient time scale for describing our algorithm. In particular, the initial instant of the (pre)-computation of the upper part of Fig. 4 is  $-2n$  when  $n$  is *even* and  $-2n - 1$  when  $n$  is *odd*, and the initial instant of the two signals of the lower part of Fig. 4 is  $-3n$ . We let the reader imagine the variants of Fig. 3 and Fig. 4 for the *odd* case.



■ **Figure 3** Initial positions and translation vectors for  $I(x) = I(-t+2c)$  (in red) and  $I(y) = I(c)$  (in blue) when  $n$  is even (here  $n = 6$ ). The gray parallelogram is where the induction actually happens. The result of the computation lies at the upper right cell.

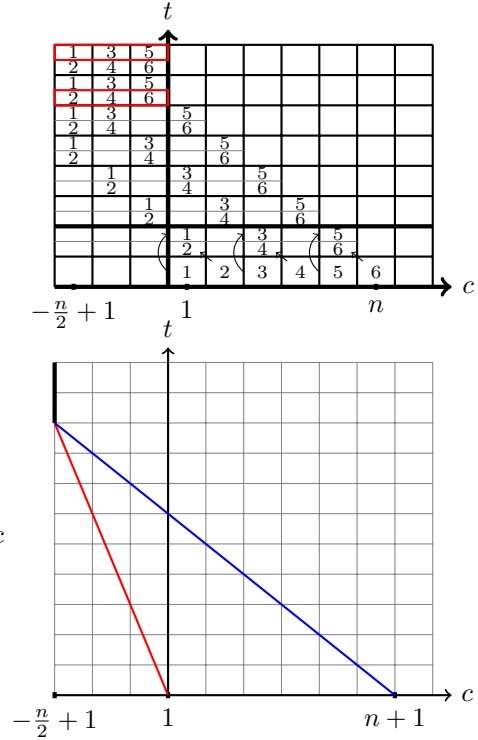
$a_0x_0 + a_1x_1 + \dots + a_dx_d = 0$ , with coefficients  $a_i \in \mathbb{Z}^*$  for all  $i \in \llbracket 0, d \rrbracket$ , as justified by Lemma 18. For simplicity of notation, assume all the coefficients  $a_i$  are positive.

First, for sake of simplicity, let us ignore the input literals and take account of the only guessed atoms  $R_i(x_0 + i_0, \dots, x_d + i_d)$  in the Horn clauses. Intuitively, if one regards the hyperplane  $H_t = \{(x_0, x_1, \dots, x_d) \in \llbracket 1, n \rrbracket^{d+1} \mid a_0x_0 + a_1x_1 + \dots + a_dx_d = t\}$  as the set of cells of a  $d$ -dimensional CA at instant  $t$ , then the conjunction of Horn clauses  $\psi$  can be regarded as the transition function of such a CA. More formally, in order to introduce the *time* parameter  $t$ , we eliminate one of the variables,  $x_0$  for example, and we regard the other variables,  $x_1, \dots, x_d$ , as the *space* variables, i.e. the respective  $d$  coordinates  $c_1, \dots, c_d$  of a cell. More precisely, one makes the following change of variables:

$$\left( \begin{array}{l} t = a_0x_0 + \dots + a_dx_d, \\ c_1 = x_1, \dots, c_d = x_d \end{array} \right), \text{ whose converse is } \left( \begin{array}{l} x_0 = (t - a_1c_1 - \dots - a_dc_d)/a_0, \\ x_1 = c_1, \dots, x_d = c_d \end{array} \right).$$

As in Section 3.2, we associate with each point  $\mathbf{x} = (x_0, \dots, x_d) \in \llbracket 1, n \rrbracket^{d+1}$ , the tuple  $\text{state}(\mathbf{x})$  of boolean values of all input and guessed atoms on  $\mathbf{x}$ . That is,

$$\text{state}(\mathbf{x}) = \left( (I(\mathbf{u}))_{\substack{I \in \text{Input} \\ \mathbf{u} \subseteq \mathbf{x}}}, (R(\mathbf{x}))_{R \in \text{Guess}} \right).$$



■ **Figure 4** The linear precomputation of  $I(x)$  can be done by stacking the information of the cells in the odd columns, then packing it to the left against a “wall” at  $c = -\frac{n}{2} + 1$  ( $n$  even). The bottom figure shows how the wall can be constructed in linear time with two signals of slope  $-\frac{1}{3}$  (resp.  $-1$ ) starting in cell 1 (resp.  $n + 1$ ) at instant  $-3n$ .

Here, we denote by **Input** (resp. **Guess**) the set of input (resp. guessed) predicates occurring in the formula. Furthermore,  $u \in \mathbf{x}$  means that  $u$  is any variable among  $x_0, \dots, x_d$ , while  $\mathbf{u} \subseteq \mathbf{x}$  means that  $\mathbf{u}$  is any  $m$ -tuple built from those variables, where  $m \leq d$  is the arity of  $I$ . Besides, the values of the guessed literals  $R(\mathbf{x})$ ,  $R \in \mathbf{Guess}$ , are deduced by the Horn formula  $\forall \mathbf{x}\psi$ .

If one ignores the input literals, the state of each point

$P = (t, c_1, \dots, c_d) = \left( \sum_{j=0}^d a_j x_j, x_1, \dots, x_d \right)$  is determined by the states of the points

$$P_v = \left( \sum_{j=0}^d a_j (x_j + v_j), x_1 + v_1, \dots, x_d + v_d \right) = \left( t + \sum_{j=0}^d a_j v_j, c_1 + v_1, \dots, c_d + v_d \right)$$

for each vector  $v = (v_0, \dots, v_d)$  of the induction system  $\mathcal{S}$ . In other words the state of the cell  $(c_1, \dots, c_d)$  at instant  $t$  is determined by the states of the cells  $(c_1 + v_1, \dots, c_d + v_d)$  at the respective *previous* instants  $t + \sum_{j=0}^d a_j v_j$  for the vectors  $v = (v_0, \dots, v_d) \in \mathcal{S}$ . (Recall that  $\sum_{j=0}^d a_j v_j < 0$ , by hypothesis.)

Let us now explain how the CA we construct can take account of the input atoms, i.e. let us describe how the CA moves the input bits. The crucial point is that at least one of the  $d + 1$  variables  $x_0, \dots, x_d$  does *not* occur in each input atom because the arity of each input predicate is at most  $d$ . This missing variable is used as a ‘transport variable’ of the values of the concerned input atom. As a *generic* example, let us consider the input atom  $I(x_0, x_2, \dots, x_d)$  where  $I$  is an input predicate of arity  $d$  and where the variable  $x_1$  does not occur<sup>8</sup>. After the above-mentioned change of variables, this atom becomes

$$I\left(\frac{1}{a_0}(t - a_1 c_1 \cdots - a_d c_d), c_2, \dots, c_d\right).$$

Because of the identity  $(t - a_1) - a_1(c_1 - 1) - a_2 c_2 \cdots - a_d c_d = t - a_1 c_1 - a_2 c_2 \cdots - a_d c_d$  the automaton only has to *move* to each cell  $(c_1, c_2, \dots, c_d)$  at instant  $t$  the boolean value

$$I\left(\frac{1}{a_0}((t - a_1) - a_1(c_1 - 1) - a_2 c_2 \cdots - a_d c_d), c_2, \dots, c_d\right)$$

which is stored at instant  $t - a_1$  in the state of cell  $(c_1 - 1, c_2, \dots, c_d)$ . In terms of cellular automaton, the values of the input atom  $I(x_0, x_2, \dots, x_d)$  are moved/transmitted by linear parallel ‘signals’ which cover all the inductive space  $\llbracket 1, n \rrbracket^{d+1}$ .

*Time and initialization of the computation* : Since the  $d + 1$  original variables  $x_0, \dots, x_d$  lie in  $\llbracket 1, n \rrbracket$ , the domain of the time variable  $t = a_0 x_0 + \dots + a_d x_d$  is  $\llbracket A, An \rrbracket$ , where  $A = a_0 + \dots + a_d$ . As a consequence, the equation of the cell hyperplane at the *initial* instant (resp. *final* instant) in the space-time diagram is  $t = A$  (resp.  $t = An$ )<sup>9</sup>.

The initialization of the input values (input ‘signals’) before the instant  $t = A$  is the most delicate/technical part of the computation. It is sufficient to describe the initialization of the values of the input ‘signals’ for our generic example<sup>10</sup> of input atom  $\alpha \equiv I(x_0, x_2, \dots, x_d)$

<sup>8</sup> As we have seen in previous examples the case where one variable among  $x_2, \dots, x_d$  does not occur in an input atom is similar; the case where  $x_0$  does not occur or the case where the arity of the input predicate is less than  $d$  are easier to deal with as we have also seen.

<sup>9</sup> In the sequel,  $A$  always denote the sum  $a_0 + \dots + a_d$ . Also, recall that each  $a_i$  is positive.

<sup>10</sup> Here again, all the other examples have either exactly the same treatment or a simpler one.

or, equivalently,  $\alpha \equiv I(\frac{1}{a_0}(t - a_1c_1 - \dots - a_dc_d), c_2, \dots, c_d)$ , for which  $c_1$  (that is the missing variable  $x_1$  of  $\alpha$ ) is the “transport” variable. To give the reader the geometric intuition of the following construction in the general case we invite her to consult Fig. 3 and 4 of Sec. 3.2 in the particular case of atom  $\alpha \equiv I(x) \equiv I(-t + 2c)$  of the formula that defines PALINDROME.

Because of the correspondence  $t = a_0x_0 + a_1c_1 + a_2x_2 \dots + a_dx_d$  or, equivalently,  $c_1 = (t - a_0x_0 - a_2x_2 \dots - a_dx_d)/a_1$ , with  $c_2 = x_2, \dots, c_d = x_d$ , for all  $(x_0, x_2, \dots, x_d) \in \llbracket 1, n \rrbracket^d$ , the boolean value  $I(x_0, x_2, \dots, x_d)$  should be stored – for the initialization of its input “signal” – in the state of the cell  $(c_1, x_2, \dots, x_d)$  such that  $c_1 = (t_0 - a_0x_0 - a_2x_2 \dots - a_dx_d)/a_1$  at the maximal instant  $t_0 < A$  (depending on the tuple  $(x_0, x_2, \dots, x_d)$ ) such that the quotient  $(t_0 - a_0x_0 - a_2x_2 \dots - a_dx_d)/a_1$  is an *integer*. Let  $i$  be the integer in  $\llbracket 0, a_1 - 1 \rrbracket$  such that  $A - a_0x_0 - a_2x_2 \dots - a_dx_d \equiv -i \pmod{a_1}$ . It is easy to verify that  $t_0 = A - a_1 + i$ . So, the boolean value  $I(x_0, x_2, \dots, x_d)$  should be stored/initialized at the instant  $t_0 = A - a_1 + i$  in (the state of) the cell  $(c_1, x_2, \dots, x_d)$  where  $c_1 = (A - a_1 + i - a_0x_0 - a_2x_2 \dots - a_dx_d)/a_1$ : see Fig. 3.

Note that for the atom  $\alpha \equiv I(x_0, x_2, \dots, x_d)$ , there are  $a_1$  distinct “initialization” configurations in the respective  $a_1$  hyperplanes  $H_{t_0}$ , where  $t_0 = A - a_1 + i$  with  $i \in \llbracket 0, a_1 - 1 \rrbracket$ , according to the possible values of the function  $f(x_0, x_2, \dots, x_d) = A - a_0x_0 - a_2x_2 \dots - a_dx_d$  modulo  $a_1$ . Furthermore, one can verify that, by construction, the space of the “initialization” configurations – their informative cells, in non-quiet states – is included in a hypercube of the form  $\llbracket -bn, bn \rrbracket^d$ , for some constant integer  $b > 0$ .

*Pre-computation and end of computation:* The initial configuration of a  $d$ -CA that recognizes the  $d$ -picture language  $L$  should be the *picture configuration*  $\mathcal{C}_{p,q_0}$  where  $p$  is the input picture. Therefore, there should be a *pre-computation* starting from  $\mathcal{C}_{p,q_0}$  that computes the “initialization” configurations of the input atoms of  $\Phi$ . By the classical technique of signals of CA’s (see [16]) we have exemplified above in the case of PALINDROME (see Fig. 4), this can be done in linear space and linear time.

Similarly, the result of the computation should be given by the accept/reject state,  $q_a$  or  $q_r$ , in the *reference* cell  $\mathbf{n} = (n, \dots, n)$ . This is realized in linear time by gathering in the reference cell the possible contradictions deduced in cells for Horn clauses.

For lack of space, we have omitted to deal with loops in Horn clauses: the possible presence of guessed atoms of the form  $R(t, \mathbf{c})$ , i.e. without predecessor/successor functions, *both as conclusions and as hypotheses* of clauses of monotonic Horn formulas seemingly contradicts the “strict monotonicity” of the induction. We cope with this point in Appendix A.3. This achieves the proof of Prop. 19 and Thm. 15. ◀

## 5 Optimality of our main result

It is natural to ask whether the monotonicity condition can be removed or weakened in our main result. It is unlikely because it would imply (as proved in Appendix B.2) the following time-space trade-off which would be a breakthrough in computational complexity:

► **Proposition 20.** *If we had  $\mathbf{DLIN}_{ca}^d = \text{ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  or the weaker equality  $\mathbf{DLIN}_{ca}^d = \text{weak-mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  for a given  $d > 1$ , then any set of words recognizable by a 1-CA in time  $n^d$  on  $n$  cells would be recognizable by a  $d$ -CA in time  $O(n)$  on  $O(n^d)$  cells.*

Here,  $\text{weak-mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  denotes the variant of the class  $\text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  where the strict inequality  $a_0x_0 + \dots + a_dx_d < 0$  of the monotonicity condition is replaced by the non-strict inequality  $a_0x_0 + \dots + a_dx_d \leq 0$ .

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## Appendix

This appendix is divided into three sections. In Section A, we provide the referees with some details omitted in proofs leading to the main result (Thm. 15). In Section B, we argue the optimality of our result: a natural extension of the logic involved in the characterization of  $\mathbf{DLIN}_{ca}$  is considered, for which Thm. 15 still holds. In contrast, removing or weakening the monotonicity condition would threaten the validity of this theorem, as announced in the conclusion (Sec. 5). Section C insists on the algorithmic scope of the main result. We illustrate the whole process described before — expressing a given problem  $\pi$  with a monotonic Horn formula  $\Phi_\pi$ , and then extracting from  $\Phi_\pi$  a linear time automaton  $\mathcal{A}_\pi$  for  $\pi$  — with two natural problems more enlightening than the pedagogical case study (the set of palindromes) used in the paper. More precisely,

- Subsection A.1 gives a complete proof of the inclusion  $\mathbf{DLIN}_{ca} \subseteq \text{mon-ESO-HORN}$  (Prop. 16). In order to complete the proof of the converse inclusion (Prop. 19), Subsection A.2 presents the proof of the technical Lemma 18. Subsection A.3 explains how to adapt the program of the cellular automaton for coping with loops in Horn clauses.
- Subsection B.1 proves the conservation of Thm. 15 for a natural generalization of monotonic Horn formulas which allows permutations of variables in guessed atoms. In contrast, Subsection B.2 establishes that the *strict* monotonicity feature of formulas characterizing  $\mathbf{DLIN}_{ca}$ -languages is probably not optional (Prop. 20).
- Section C presents the inductive definition of two new classical problems: the product of boolean matrices by the usual inductive definition (C.1) and the product of integers by Horner's rule (C.2). In both cases, we derive from the monotonic Horn formula that defines the problem a cellular automaton that decides it.

## A Complements of proofs

### A.1 Proposition 16: The first inclusion

We present here a more complete version of the proof of Proposition 16. In particular, we present additional examples of *computation clauses* and *accepting clauses* for less general cases, that were omitted in the core of the article due to space limitations.

**Proposition 16.** For  $d \geq 1$ ,  $\mathbf{DLIN}_{ca}^d \subseteq \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ .

**Proof.** Let  $L \subseteq \Sigma^{(d)}$  be a  $d$ -picture language in  $\mathbf{DLIN}_{ca}^d$ . By Lemma 9, there exists a CA  $\mathcal{A} = (d, \mathcal{Q}, \mathcal{N}_2, \delta)$  of neighborhood  $\mathcal{N}_2 = \{-2, -1, 0, 1, 2\}^d$  that recognizes  $L$  in time  $\tau(n) = n - 1$  with *permanent* state  $q_0$ . Let  $\llbracket 1, n \rrbracket$  denote the interval of  $n$  instants of the computation of  $\mathcal{A}$  on a  $d$ -picture of side  $n$ ; in particular, the initial and final instants are numbered 1 and  $n$ , respectively. We are going to construct a formula  $\Phi \in \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  that defines  $L$ , i.e., expresses that the computation of  $\mathcal{A}$  on a  $d$ -picture  $p$  accepts it. It is of the form  $\Phi \equiv \exists(R_s)_{s \in \mathcal{Q}} \forall t \forall \mathbf{c} \psi$  where  $\mathbf{c}$  denotes the  $d$ -tuple of variables  $(c_1, \dots, c_d)$  and, for  $s \in \mathcal{Q}$ , the intended meaning of the guessed atom  $R_s(t, \mathbf{c})$  is the following: at the instant  $t$ , the cell  $\mathbf{c}$  is in the state  $s$ . For simplicity of notation, let us assume  $d = 1$ . The quantifier-free part  $\psi$  of  $\Phi$  is the conjunction of three kinds of Horn clauses:

- (1) *initialization clauses*: for each  $s \in \Sigma$ , the clause  $\min(t) \wedge Q_s(c) \rightarrow R_s(t, c)$ ;



- (2) five kinds of *computation clauses* that compute the state at instant  $t > 1$  of any cell  $c$  according to its possible neighborhoods for  $\mathcal{N}_2 = \{-2, -1, 0, 1, 2\}$ , i.e. the following five possible cases<sup>11</sup> for  $n \geq 4$ :

(i)  $c = 1$ ; (ii)  $c = 2$ ; (iii) general case  $c \in \llbracket 3, n - 2 \rrbracket$ ; (iv)  $c = n - 1$ ; (v)  $c = n$ .

– *general case (iii)*  $c \in \llbracket 3, n - 2 \rrbracket$  : for any 5-tuple of states  $(s_{-2}, s_{-1}, s_0, s_1, s_2) \in (Q - \{q_0, q_a, q_r\})^5$ , the following clause whose premises imply  $n \geq 5$ :

$$\left( \begin{array}{l} \neg \min(t) \wedge \neg \min(c) \wedge \neg \min(c - 1) \wedge \neg \max(c) \wedge \neg \max(c + 1) \wedge \\ R_{s_{-2}}(t - 1, c - 2) \wedge R_{s_{-1}}(t - 1, c - 1) \wedge \\ R_{s_0}(t - 1, c) \wedge R_{s_1}(t - 1, c + 1) \wedge R_{s_2}(t - 1, c + 2) \end{array} \right) \rightarrow R_{\delta(s_{-2}, s_{-1}, s_0, s_1, s_2)}(t, c)$$

– *case (i)*  $c = 1$  : for any 3-tuple of states  $(s_0, s_1, s_2) \in (Q - \{q_0, q_a, q_r\})^3$ , the following clause whose premises imply  $n \geq 4$ :

$$\left( \begin{array}{l} \neg \min(t) \wedge \min(c) \wedge \neg \max(c) \wedge \neg \max(c + 1) \wedge \neg \max(c + 2) \wedge \\ R_{s_0}(t - 1, c) \wedge R_{s_1}(t - 1, c + 1) \wedge R_{s_2}(t - 1, c + 2) \end{array} \right) \rightarrow R_{\delta(q_0, q_0, s_0, s_1, s_2)}(t, c)$$

– *case (ii)*  $c = 2$  : for any 4-tuple of states  $(s_{-1}, s_0, s_1, s_2) \in (Q - \{q_0, q_a, q_r\})^4$ , the following clause whose premises imply  $n \geq 4$ :

$$\left( \begin{array}{l} \neg \min(t) \wedge \neg \min(c) \wedge \min(c - 1) \wedge \neg \max(c) \wedge \neg \max(c + 1) \wedge \\ R_{s_{-1}}(t - 1, c - 1) \wedge R_{s_0}(t - 1, c) \wedge \\ R_{s_1}(t - 1, c + 1) \wedge R_{s_2}(t - 1, c + 2) \end{array} \right) \rightarrow R_{\delta(q_0, s_{-1}, s_0, s_1, s_2)}(t, c)$$

– clauses of cases (v)  $c = n$  and (iv)  $c = n - 1$  which are symmetrical to clauses of cases (i) and (ii) above.

- (3) the *accepting clause*  $R_{q_r}(t, c) \rightarrow \perp$  which says that the computation does not reject, and hence accepts, since by hypothesis each computation of  $\mathcal{A}$  either accepts or rejects.

Recall that the computation clauses of above Item (2) say nothing in case  $n < 4$ . That is why specific clauses are needed for each of the cases  $n = 1$ ,  $n = 2$  and  $n = 3$ . For example, for  $n = 3$ , add for each input word  $w = w_1 w_2 w_3 \in \Sigma^3 \setminus L$  the clause<sup>12</sup>

$$\min(c) \wedge \neg \max(c) \wedge \neg \max(c + 1) \wedge \max(c + 2) \wedge Q_{w_1}(c) \wedge Q_{w_2}(c + 1) \wedge Q_{w_3}(c + 2) \rightarrow \perp$$

which “eliminates” the word  $w$ .

This ends the list of clauses of our formula. We let the reader convince herself or himself that the formula  $\Phi$  so obtained correctly defines the language  $L$  in case  $d = 1$ . It is easy but tedious – because of notation – to generalize the formula to any dimension  $d > 1$ . In particular, the number of kinds of computation clauses (Item (2) above) is  $5^d$  in the general case.

Finally, notice that by construction  $\Phi \in \text{ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  and that its induction system is  $\{-1\} \times \{-2, -1, 0, 1, 2\}^d$ , which has the reference hyperplane  $t = 0$ . This implies  $\Phi \in \text{mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$  and concludes the proof of Prop. 16. ◀

<sup>11</sup> Case (iii) does not occur when  $n = 4$ .

<sup>12</sup> Those clauses can be regarded as *accepting clauses* since *none* mentions the words  $w \in L \cap \Sigma^3$  and hence their conjunction globally *defines*  $L \cap \Sigma^3$ .

## A.2 Lemma 18: Existence of all-non-zero reference hyperplane

**Lemma 18.** *Let  $\mathcal{S} \subset \mathbb{Z}^{d+1}$  be an induction system satisfying the monotonicity condition w.r.t. some reference hyperplane. Then,  $\mathcal{S}$  has another reference hyperplane of equation  $a_0x_0 + \dots + a_dx_d = 0$  where each coefficient  $a_i$  ( $i \in \llbracket 0, d \rrbracket$ ) is a non-zero integer.*

**Proof.** Without loss of generality, assume that the reference hyperplane of the induction system  $\mathcal{S}$  is of the form  $f(x_0, \dots, x_k) = a_0x_0 + \dots + a_kx_k = 0$ , for  $0 \leq k < d$ , with the integer coefficients  $a_i \neq 0$ ,  $i \in \llbracket 0, k \rrbracket$ . Then,  $f(v_0, \dots, v_k) = a_0v_0 + \dots + a_kv_k < 0$  holds for each vector  $(v_0, \dots, v_d) \in \mathcal{S}$ , that means  $f(v_0, \dots, v_k) \leq -1$  since all the values involved are integers. Define the positive integer

$$A = 1 + \max_{(v_0, \dots, v_d) \in \mathcal{S}} (|v_{k+1}| + |v_{k+2}| + \dots + |v_d|)$$

and consider the hyperplane of equation

$$g(x_0, \dots, x_d) = A(a_0x_0 + \dots + a_kx_k) + x_{k+1} + \dots + x_d = 0,$$

where each variable  $x_0, \dots, x_d$  has a non-zero coefficient. We claim that  $g(v_0, \dots, v_d) < 0$  holds, for each vector  $(v_0, \dots, v_d) \in \mathcal{S}$ . By hypothesis,  $a_0v_0 + \dots + a_kv_k \leq -1$  holds; hence,  $A(a_0v_0 + \dots + a_kv_k) \leq -A$ . This implies  $g(v_0, \dots, v_d) \leq -A + |v_{k+1}| + \dots + |v_d|$ . Then,  $g(v_0, \dots, v_d) \leq -A + \max_{(v_0, \dots, v_d) \in \mathcal{S}} (|v_{k+1}| + \dots + |v_d|) = -1$  holds, by definition of  $A$ , and  $g(v_0, \dots, v_d) < 0$  as claimed.  $\blacktriangleleft$

## A.3 Proposition 19: Coping with loops in Horn clauses

The possible presence of guessed atoms of the form  $R(t, \mathbf{c})$ , i.e. without predecessor/successor functions, both as conclusions and as hypotheses of clauses of monotonic Horn formulas seems to contradict with the “strict monotonicity” of the induction. Let us now precisely explain how to cope with this problem. Without loss of generality, one can assume – as it is easily justified – that each clause of a monotonic Horn formula  $\Phi$  is of either of the two following forms (a) or (b):

- (a) *inductive clause*  $\alpha_1 \wedge \dots \wedge \alpha_m \wedge R_1(t, \mathbf{c}) \wedge \dots \wedge R_p(t, \mathbf{c}) \rightarrow R_0(t, \mathbf{c})$ ,  $p \geq 0$ , where the  $R_i(t, \mathbf{c})$  are guessed atoms over the same  $(d+1)$ -tuple of variables  $(t, \mathbf{c})$  and the  $\alpha_j$ ’s are either input literals or guessed atoms of the form  $R(t - u, c_1 + i_1, \dots, c_d + i_d)$  for some induction vector  $(-u, i_1, \dots, i_d)$ ,  $u > 0$ , of  $\Phi$ ;
- (b) *contradiction clause*  $\beta_1 \wedge \dots \wedge \beta_m \rightarrow \beta_0$  whose conclusion  $\beta_0$  is an input literal or the constant  $\perp$ , and the hypotheses  $\beta_1, \dots, \beta_m$  are either input literals or guessed atoms of the form  $R(t - u, c_1 + i_1, \dots, c_d + i_d)$  for some integers  $u, i_1, \dots, i_d$ , with  $u \geq 0$ , including at least one guessed atom of the form  $R(t, \mathbf{c})$ .

Notice that each inductive clause can be equivalently rewritten as

$$(a) \alpha_1 \wedge \dots \wedge \alpha_m \rightarrow \gamma(t, \mathbf{c})$$

where  $\gamma(t, \mathbf{c})$  is the Horn clause  $R_1(t, \mathbf{c}) \wedge \dots \wedge R_p(t, \mathbf{c}) \rightarrow R_0(t, \mathbf{c})$ , the atoms of which are guessed atoms over the same  $(d+1)$ -tuple  $(t, \mathbf{c})$ .

In essence, a cell of the automaton that simulates  $\Phi$  can evaluate “on the fly” the premises of Horn clauses while updating its own state if said premises are of the form  $R(t, \mathbf{c})$ . In order to accomplish this, at each instant  $t - 1$ , the next step of  $\mathcal{A}$  is divided into three successive substeps:

- *First substep:* For each inductive clause  $\alpha_1 \wedge \dots \wedge \alpha_m \rightarrow \gamma(t, \mathbf{c})$ , evaluate its premise  $\alpha_1, \dots, \alpha_m$ . Let  $\theta(t, \mathbf{c})$  denote the conjunction of the conclusions  $\gamma(t, \mathbf{c}) = R_1(t, \mathbf{c}) \wedge \dots \wedge R_p(t, \mathbf{c}) \rightarrow R_0(t, \mathbf{c})$  of the inductive clauses whose all the premises  $\alpha_1, \dots, \alpha_m$  evaluate to TRUE;
- *Second substep:* Compute<sup>13</sup> the set of guessed atoms  $(R(t, \mathbf{c}))_{R \in \text{Guess}}$  that are consequences of  $\theta(t, \mathbf{c})$ : by convention, the state of the cell  $\mathbf{c}$  as instant  $t$  is

$$\text{state}(t, \mathbf{c}) = \left( (I(\mathbf{u}))_{\substack{I \in \text{Input} \\ \mathbf{u} \subseteq \{t\} \cup \{\mathbf{c}\}}}, (R(t, \mathbf{c}))_{R \in \text{Guess}} \right);$$

- *Third substep:* For each contradiction clause  $(b)$ , evaluate its literals; reject if all the premises of the clause evaluate to TRUE whereas its conclusion evaluates to FALSE.

## B Generalizations

### B.1 Theorem. 15: Allowing permutations of variables

In Def. 12 that defines ESO-HORN<sup>d</sup>( $\forall^{d+1}$ , arity<sup>d+1</sup>)-formulas, we have assumed that variables occur in the standard order  $x_0, x_1, \dots, x_d$  in guessed atoms. In fact, guessed atoms with variables in *any order* can be allowed: the definition of an ESO-HORN<sup>d</sup>( $\forall^{d+1}$ , arity<sup>d+1</sup>)-formula and of its induction system (Def. 12 and 13) can be naturally extended to “non ordered” guessed atoms.

In the scope of this extension, a non-ordered guessed atom is a guessed atom of the form  $R(x_{\pi(0)} + i_{\pi(0)}, \dots, x_{\pi(d)} + i_{\pi(d)})$ , where  $\pi$  is a permutation of  $\llbracket 0, d \rrbracket$ . Likewise, the *induction vector* associated with such a non-ordered atom is the tuple  $(i_0, \dots, i_d) \in \mathbb{Z}^{d+1}$ , if this tuple is *nonzero*.

We continue to assume that the only allowed form of a guessed atom that occurs as the conclusion of a Horn clause is  $R(x_0, \dots, x_d)$ . However, it should be clear that this convention is not a real restriction: any Horn clause with non-ordered guessed atoms occurring both in its hypotheses and its conclusion can be equivalently rewritten with an ordered atom as its conclusion.

We claim that in this extended framework, Thm. 15 still holds.

**Proof.** (proof of the generalization of Thm. 15). Let us succinctly explain how general guessed atoms with variables in any order can be replaced by “ordered” guessed atoms, i.e., with variables in the standard order:  $x_0, x_1, \dots, x_d$ . The trick consists in associating to each guessed predicate  $R$  a new guessed predicate  $R_\pi$ , for each permutation  $\pi$  of  $\llbracket 0, d \rrbracket$ , and replacing each guessed atom  $R(x_{\pi(0)}, \dots, x_{\pi(d)})$  by the “ordered” atom  $R_\pi(x_0, \dots, x_d)$ . More generally, each occurrence of an atom  $R(x_{\pi(0)} + i_{\pi(0)}, \dots, x_{\pi(d)} + i_{\pi(d)})$  is replaced by  $R_\pi(x_0 + i_0, \dots, x_d + i_d)$ . We also have to add some Horn clauses to ensure the pairwise coherence of the  $(d+1)!$  predicates  $R_\pi$  associated with the same guessed predicate  $R$ . Let us explain in detail how to do this in the simple case  $d = 1$ . Let  $\bar{R}$  denote the binary guessed predicate associated with  $R$  for the inversion of its two arguments. Clearly, the mutual coherence of  $R$  and  $\bar{R}$  is ensured by the implication  $x_0 = x_1 \rightarrow (R(x_0, x_1) \leftrightarrow \bar{R}(x_0, x_1))$ , which should be formally replaced by the conjunction of the two following Horn clauses where

<sup>13</sup>This can be realized by pre-computing a table by the classical deduction algorithm on propositional Horn formulas (see e.g. [3]). Recall that the formula  $\theta$  is shorter than our fixed formula  $\Phi$ .

guessed atoms are “ordered”:

$$x_0 = x_1 \wedge \overline{R}(x_0, x_1) \rightarrow R(x_0, x_1) \quad \text{and} \quad x_0 = x_1 \wedge R(x_0, x_1) \rightarrow \overline{R}(x_0, x_1).$$

Notice that we have used the equality relation which can be inductively defined as a guessed predicate in either of the two following symmetrical ways :

- in the *increasing* induction given by the Horn clauses  $\min(x) \wedge \min(y) \rightarrow R_{=}(x, y)$  and  $\neg \min(x) \wedge \neg \min(y) \wedge R_{=}(x-1, y-1) \rightarrow R_{=}(x, y)$  with the induction vector  $(-1, -1)$  ;
- in the *decreasing* induction given by the Horn clauses  $\max(x) \wedge \max(y) \rightarrow R_{=}(x, y)$  and  $\neg \max(x) \wedge \neg \max(y) \wedge R_{=}(x+1, y+1) \rightarrow R_{=}(x, y)$  with the induction vector  $(1, 1)$  which is the opposite of the previous vector.

One chooses either the increasing induction or the decreasing induction according to either vector  $(-1, -1)$  or  $(1, 1)$  belongs to the (strict) half-space chosen for the induction system of the original Horn formula: exactly one of them belongs to this half-space<sup>14</sup>.

To suggest the general case, let us exhibit in the case  $d = 5$  the following “Horn clause” which ensures the coherence of a guessed predicate  $R$  (of arity 6) with its permuted predicate  $R_{\pi}$  for the permutation  $\pi = (0 \mapsto 1, 1 \mapsto 2, 2 \mapsto 0, 3 \mapsto 4, 4 \mapsto 3, 5 \mapsto 5)$ , which is the composition of the disjoint circular permutations  $(0, 1, 2)$  and  $(3, 4)$  :

$$x_0 = x_1 = x_2 \wedge x_3 = x_4 \rightarrow (R(x_0, x_1, x_2, x_3, x_4, x_5) \leftrightarrow R_{\pi}(x_0, x_1, x_2, x_3, x_4, x_5)).$$

It is easy to generalize this example. However, here again we have introduced *equalities* (or multiple equalities, e.g.  $x_0 = x_1 = x_2$ ), which are *not* allowed. They can be defined as guessed predicates by an increasing or decreasing induction similar as above for the case  $d = 1$ . We leave the details as an exercise to the reader. ◀

## B.2 Proposition 20: Weak monotonicity is not an option

**Proposition 20.** *If we had  $\mathbf{DLIN}_{ca}^2 = \text{weak-mon-ESO-HORN}^2(\forall^3, \text{arity}^3)$ , then any language  $L$  recognizable by a 1-CA in time  $n^2$  on  $n$  cells would be recognizable by a 2-CA in time  $O(n)$  on  $O(n^2)$  cells. More generally, for all integer  $d > 1$ , if we had  $\mathbf{DLIN}_{ca}^d = \text{weak-mon-ESO-HORN}^d(\forall^{d+1}, \text{arity}^{d+1})$ , then any language  $L$  recognizable by a 1-CA in time  $n^d$  on  $n$  cells would be recognizable by a  $d$ -CA in time  $O(n)$  on  $O(n^d)$  cells.*

**Proof.** Assume  $\mathbf{DLIN}_{ca}^2 = \text{weak-mon-ESO-HORN}^2(\forall^3, \text{arity}^3)$ . Let  $L \subseteq \Sigma^*$  be a language recognized by a 1-CA  $\mathcal{A} = (2, \mathcal{Q}, \mathcal{N}, \delta)$  of neighborhood  $\mathcal{N} = \{-1, 0, 1\}$  in time  $n^2$  on  $n$  cells where  $n$  is the length of the input word. Let us associate to each word  $w = w_1 \dots w_n \in \Sigma^*$  the 2-picture  $p_w$  defined as follows:

$$p_w : \begin{cases} \llbracket 1, n \rrbracket^2 \rightarrow \Sigma \cup \{q_0\} \\ (i, j) \mapsto \begin{cases} w_i & \text{if } j = 1 \\ q_0 & \text{otherwise} \end{cases} \end{cases}$$

That is,  $p_w$  is the  $n \times n$  picture that is empty with the exception of its first column which contains  $w$ . Define the 2-picture language  $L^{\text{square}} = \{p_w : w \in L\}$ . The idea consists in folding the space-time diagram  $n^2 \times n$  of the accepting computation of  $\mathcal{A}$  of any word  $w \in L$  by folding

<sup>14</sup>Notice the particular case where the reference hyperplane is the line of equation  $x - y = 0$  to which the vectors  $(-1, -1)$  and  $(1, 1)$  are collinear: then one can slightly modify the reference hyperplane by using a similar technique as in the proof of Lemma 18.

the time  $n^2$  as an “accordion” of  $n$  slices of length  $n$ : the set of instants of the computation is the cartesian power  $\llbracket 1, n \rrbracket^2$  ordered as follows in a “accordion”, i.e. alternate increasing-decreasing, mode:  $(1, 1), (1, 2), \dots, (1, n), (2, n), (2, n - 1), \dots, (2, 1), (3, 1), (3, 2), \dots, (3, n), (4, n), (4, n - 1), \dots$ . For defining such a linear order, use the monadic predicate **odd**. (The important thing is that the corresponding successor function on pairs in  $\llbracket 1, n \rrbracket^2$  is defined “locally” using the “local” **suc** and **pred** functions and the **odd**, **min** and **max** predicates.) With that idea it is easy but tedious to define  $L^{\text{square}}$  by a formula in  $\Phi \in \text{ESO-HORN}^2(\forall^3, \text{arity}^3)$  that describes an accepting computation of  $w \in L$  corresponding to  $p_w \in L^{\text{square}}$ . More precisely,  $\Phi$  has 3 first-order variables  $t_1, t_2, c$  on the domain  $\llbracket 1, n \rrbracket$ : the values of the couple of variables  $(t_1, t_2)$  represent the  $n^2$  instants of the computation and the values of  $c$  represent the  $n$  cells of the input (the workspace). It should be clear that the induction system of  $\Phi$  is  $\mathcal{S} = \{(0, -1), (0, 1), (-1, 0)\} \times \{-1, 0, 1\}$ . It satisfies the weak monotonicity condition  $v_0 \leq 0$ , for each  $(v_0, v_1, v_2) \in \mathcal{S}$ . Then,  $\Phi \in \text{weak-mon-ESO-HORN}^2(\forall^3, \text{arity}^3)$ , and, by our assumption, it implies  $L^{\text{square}} \in \text{DLIN}_{\text{ca}}^2$ . That means that  $L^{\text{square}}$  is recognizable by a 2-CA in time  $O(n)$ . Equivalently,  $L$  is recognizable by a 2-CA in time  $O(n)$  on  $O(n^2)$  cells as claimed.

The proof of the general case, for any integer  $d > 1$ , is similar: use a generalization of our accordion-folding technique. ◀

### C Illustrations: Two other classical problems

In this section, we present the inductive definitions of the following decision problems by monotonic Horn formulas in order to convince the reader that such a definition is a natural step to design cellular automata that recognize/compute these problems:

#### MATRICES-PRODUCT

*Input* : Three boolean matrices  $n \times n$ , i.e. 2-pictures,  $A, B, C$ , of alphabet  $\{0, 1\}$ ;

*Question* : Is  $C$  the product  $A \times B$  ?

#### INTEGERS-PRODUCT

*Input* : three positive integers  $a, b, c$  in binary notation<sup>15</sup>, such that  $a, b \leq c$ .

*Question* : Have we  $a \times b = c$  ?

### C.1 Product of matrices

► **Proposition 21.** *MATRICES-PRODUCT is definable in  $\text{mon-ESO-HORN}^2(\forall^3, \text{arity}^3)$ .*

**Proof.** The input is represented by the picture structure

$$\mathcal{S}(A, B, C) = \langle \llbracket 1, n \rrbracket, A_0, A_1, B_0, B_1, C_0, C_1, \text{min}, \text{max}, \text{suc}, \text{pred} \rangle,$$

where  $A_0, A_1, B_0, B_1, C_0, C_1$  are binary relations: the matrix  $A$  is represented by  $A_i = \{(x, y) \in \llbracket 1, n \rrbracket^2 : A(x, y) = i\}$ ,  $i \in \{0, 1\}$ , and similarly for  $B$  and  $C$ . The conjunction of the following clauses (1-6) define by induction on the variable  $y$  the ‘cumulative’ ternary predicates  $T_1$  and  $T_0$  whose intended meaning is

$$T_1(x, y, z) \equiv \exists y' \leq y : A_1(x, y') \wedge B_1(y', z) \text{ and } T_0(x, y, z) \equiv \forall y' \leq y : A_0(x, y') \vee B_0(y', z).$$

<sup>15</sup> We will make no distinction between a positive integer  $a = \sum_{i=1}^n a_i 2^{i-1}$ ,  $a_i \in \{0, 1\}$ , and its binary notation  $a_n \dots a_1$ .

For the basic case  $y = 1$  when it gives 0, the clauses

- (1)  $\min(y) \wedge A_0(x, y) \rightarrow T_0(x, y, z)$  and
- (2)  $\min(y) \wedge B_0(y, z) \rightarrow T_0(x, y, z)$ .

To inductively maintain the value 0, the clauses

- (3)  $\neg \min(y) \wedge T_0(x, y - 1, z) \wedge A_0(x, y) \rightarrow T_0(x, y, z)$  and
- (4)  $\neg \min(y) \wedge T_0(x, y - 1, z) \wedge B_0(y, z) \rightarrow T_0(x, y, z)$ .

To introduce the value 1, the clause

- (5)  $A_1(x, y) \wedge B_1(y, z) \rightarrow T_1(x, y, z)$ .

To inductively maintain the value 1, the clause

- (6)  $\neg \min(y) \wedge T_1(x, y - 1, z) \rightarrow T_1(x, y, z)$ .

For the final result  $C = A \times B$ , the clauses

- (7)  $\max(y) \wedge T_i(x, y, z) \rightarrow C_i(x, z)$ , for  $i \in \{0, 1\}$ .

Clearly, the formula  $\Phi_{\text{mp}} \equiv \exists T_0, T_1 \forall x, y, z \varphi_{\text{mp}}$  where  $\varphi_{\text{mp}}$  is the conjunction of clauses (1), ..., (7) correctly defines MATRICES-PRODUCT and belongs to  $\text{mon-ESO-HORN}^2(\forall^3, \text{arity}^3)$  since its induction system  $\{(0, -1, 0)\}$  trivially satisfies the monotonicity condition. ◀

### Deriving the CA from the inductive formula

Now, in order to take account of the set of input predicates

$$\text{Input} = \{A_0, A_1, B_0, B_1, C_0, C_1, \min, \max\}$$

of the formula  $\Phi_{\text{mp}}$ , let us make the following change of variables presented in the proof of Lemma 18:

$$(t = x + y + z, c_1 = x, c_2 = z), \text{ whose converse is } (x = c_1, y = t - c_1 - c_2, z = c_2).$$

For each point  $(x, y, z) \in \llbracket 1, n \rrbracket^3$ , let us call  $\text{state}(x, y, z)$  the tuple consisting of

- the tuple of boolean values of the input atoms of  $\varphi_{\text{mp}}$ :  $\min(y)$ ,  $\max(y)$ ,  $A_i(x, y)$ ,  $B_i(y, z)$ ,  $C_i(x, z)$ , for  $i \in \{0, 1\}$ , that are true in the structure  $\mathcal{S}(A, B, C)$ ,
- *completed* by the guessed atom  $T_0(x, y, z)$  or  $T_1(x, y, z)$  deduced by the Horn formula  $\forall x, y, z \varphi_{\text{mp}}$ .

By our change of variables, each input atom  $A_i(x, y)$ ,  $B_i(y, z)$ ,  $C_i(x, z)$ ,  $\min(y)$ ,  $\max(y)$  becomes  $A_i(c_1, t - c_1 - c_2)$ ,  $B_i(t - c_1 - c_2, c_2)$ ,  $C_i(c_1, c_2)$ ,  $\min(t - c_1 - c_2)$ ,  $\max(t - c_1 - c_2)$ , respectively. The 2-dimensional CA we construct has to memorize in each cell of coordinates  $(c_1, c_2) \in \llbracket 1, n \rrbracket^2$  at instant  $t$  the tuples of boolean values of those atoms. This can be realized as follows:

- (a) For each  $A_i(c_1, t - c_1 - c_2)$  (former  $A_i(x, y)$ ): because of the identity  $t - c_1 - c_2 = (t - 1) - c_1 - (c_2 - 1)$ , the CA only has to *move* to each cell  $(c_1, c_2)$  at instant  $t$  the boolean value  $A_i(c_1, (t - 1) - c_1 - (c_2 - 1))$  which is stored at instant  $t - 1$  in the cell  $(c_1, c_2 - 1)$ .
- (b) For each  $B_i(t - c_1 - c_2, c_2)$  (former  $B_i(y, z)$ ): similarly, the CA has to *move* to each cell  $(c_1, c_2)$  at instant  $t$  the boolean value  $B_i((t - 1) - (c_1 - 1) - c_2, c_2)$  which is stored at instant  $t - 1$  in the cell  $(c_1 - 1, c_2)$ . Do similarly for  $\min(t - c_1 - c_2)$  and  $\max(t - c_1 - c_2)$ .

- (c) For each  $C_i(c_1, c_2)$  (former  $C_i(x, z)$ ): the CA *conserves* on cell  $(c_1, c_2)$  the boolean value  $C_i(c_1, c_2)$  from an instant  $t - 1$  to the next instant  $t$ .

All in all, the state of each point  $P = (t, c_1, c_2) = (x + y + z, x, z)$  is determined by the state of the point  $P_1 = (x + (y - 1) + z, x, z) = (t - 1, c_1, c_2)$  because of guessed atoms and Item (c), and the states of the points  $P_2(t - 1, c_1, c_2 - 1)$  and  $P_3(t - 1, c_1 - 1, c_2)$  because of items (a) and (b), respectively.

Hence, the state of a cell  $(c_1, c_2)$  at instant  $t$  is determined by its state and the states of cells  $(c_1, c_2 - 1)$  and  $(c_1 - 1, c_2)$  at instant  $t - 1$ . It seems that we have achieved the design of a CA of neighborhood  $\{(0, 0), (0, -1), (-1, 0)\}$  that recognizes MATRICES-PRODUCT in *linear time* since  $t = x + y + z$  and  $x, y, z \in \llbracket 1, n \rrbracket$  imply  $3 \leq t \leq 3n$ . However, we have not described the “initialization” configuration, i.e. the configuration at instant  $t = 2$  (the instant just before instant 3) and the end of the computation. Designing the “initialization” configuration requires some care in connection with the above points (a), (b) and (c) about the input bits:

- *Initializing*  $A_i(c_1, t - c_1 - c_2)$ ,  $B_i(t - c_1 - c_2, c_2)$ ,  $\min(t - c_1 - c_2)$  and  $\max(t - c_1 - c_2)$ : At the instant  $t = 2$ , for all  $c_1 \in \llbracket 1, n \rrbracket$  and  $c_2 \in \mathbb{Z}$  (for all  $c_2 \in \llbracket 1, n \rrbracket$  and  $c_1 \in \mathbb{Z}$ , respectively) such that  $1 \leq 2 - c_1 - c_2 \leq n$ , the boolean values  $A_i(c_1, 2 - c_1 - c_2)$  ( $B_i(2 - c_1 - c_2, c_2)$ ,  $\min(2 - c_1 - c_2)$  and  $\max(2 - c_1 - c_2)$ , respectively) should be stored in the state of the cell  $(c_1, c_2)$ .
- *Initializing*  $C_i(c_1, c_2)$ : At the instant  $t = 2$ , for all  $(c_1, c_2) \in \llbracket 1, n \rrbracket^2$ , the boolean values  $C_i(c_1, c_2)$  should be stored in the state of the cell  $(c_1, c_2)$ .

One easily observes that, by construction<sup>16</sup>, the space of the “initialization” configuration is included in the square  $\llbracket 2 - 2n, n \rrbracket^2$ . This is linear space.

According to our conventions, the initial configuration of the CA should be the configuration  $\mathcal{C}_{p, q_0}$  associated with the input picture  $p = (A, B, C)$ , as defined in Def. 6. However, once again, we can design a routine which, starting from configuration  $\mathcal{C}_{p, q_0}$  (with quiescent state  $q_0$ ), computes the “initialization” configuration in linear space and linear time by using the classical technique of signals in CA’s (see [16]).

The result of the computation (*reject*, i.e. existence of *some* contradiction, or *accept*, i.e. *no* contradiction) can be read in the conclusion of clause (7)  $\max(y) \wedge T_i(x, y, z) \rightarrow C_i(x, z)$ , for  $y = n$  and  $x, z \in \llbracket 1, n \rrbracket$ , i.e. in the cells  $(c_1, c_2) \in \llbracket 1, n \rrbracket^2$  at the respective instants  $t = n + c_1 + c_2 \leq 3n$ . Here again, those informations can be communicated to (gathered in) only one cell, that is  $(c_1 = n, c_2 = n)$  at instant  $3n$ , to get in this reference cell the state  $q_r$  or  $q_a$  at instant  $3n + 1$ .

We have now achieved the design of a CA that recognizes in *linear time* the problem MATRICES-PRODUCT from the monotonic Horn formula that defines it.

## C.2 Product of integers

The problem INTEGERS-PRODUCT also has a natural inductive definition by a monotonic Horn formula from which one similarly derives a CA that decides the problem in linear time:

► **Proposition 22.** INTEGERS-PRODUCT is definable in mon-ESO-HORN<sup>1</sup>( $\forall^2$ , arity<sup>2</sup>) and hence belongs to DLIN<sub>ca</sub><sup>1</sup>.

<sup>16</sup>For example, the initialization of  $A_i(c_1, 2 - c_1 - c_2)$  is done on all cells  $(c_1, c_2)$  such that  $c_1 \in \llbracket 1, n \rrbracket$  and  $1 \leq 2 - c_1 - c_2 \leq n$ , that means  $2 - n - c_1 \leq c_2 \leq 1 - c_1$ , which implies  $2 - 2n \leq c_2 \leq 0$ .

**Proof.** The input is represented by the picture structure

$$\mathcal{S}(a, b, c) = \langle \llbracket 1, n \rrbracket, A_0, A_1, B_0, B_1, C_0, C_1, \min, \max, \text{succ}, \text{pred} \rangle,$$

where  $A_0, A_1, B_0, B_1, C_0, C_1$  are unary relations encoding the input tuple  $a = a_n \dots a_1$  ( $a = \sum_{i=1}^n a_i 2^{i-1}$ ,  $a_i \in \{0, 1\}$ ),  $b = b_n \dots b_1$ ,  $c = c_n \dots c_1$  ( $c_n = 1$ ) in a natural manner: for each  $i = 0, 1$ , we set  $A_i = \{x \in \llbracket 1, n \rrbracket : a_x = i\}$ , and similarly for the  $B_i$ 's and the  $C_i$ 's. Besides, for each  $y \in \llbracket 1, n \rrbracket$  we denote by  $\beta_y$  the integer with binary representation  $b_n \dots b_{y+1} b_y$  (prefix of  $b$ ). That is,  $\beta_y = b_n 2^{n-y} + \dots + b_{y+1} 2 + b_y$ . Thus we get the relations:

$$\beta_n = b_n, \quad \beta_1 = b, \quad \text{and} \quad \beta_y = 2\beta_{y+1} + b_y$$

that allow an inductive computation of  $ab = a\beta_1$  (the so-called Horner rule):

$$\text{Basic case } (y = n) : \quad a\beta_n = \begin{cases} 0 & \text{if } b_n = 0; \\ a & \text{if } b_n = 1. \end{cases}$$

$$\text{For any } y \in \llbracket 1, n-1 \rrbracket : \quad a\beta_y = 2(a\beta_{y+1}) + \begin{cases} 0 & \text{if } b_y = 0; \\ a & \text{if } b_y = 1. \end{cases}$$

In order to express the inductive computation in our logical framework, we need to define four guessed binary predicates  $S_0, S_1, Q_0, Q_1$ . The intended meaning of  $S_0, S_1$  is the following. For  $i \in \{0, 1\}$  and  $x, y \in \llbracket 1, n \rrbracket$ ,  $S_i(x, y)$  holds if  $i$  is the bit of index  $x$  of the product  $a\beta_y$ . The induction above mentioned can be displayed as follows:

The basic case  $y = n$  is expressed by the clauses

- (1)  $\max(y) \wedge B_0(y) \rightarrow S_0(x, y)$  and
- (2)  $\max(y) \wedge B_1(y) \wedge A_i(x) \rightarrow S_i(x, y)$ , for  $i = 0, 1$ .

The general case  $1 \leq y < n$  with  $b_y = 0$  is expressed by the clauses

- (3)  $\neg \max(y) \wedge B_0(y) \wedge \min(x) \rightarrow S_0(x, y)$  and
- (4)  $\neg \max(y) \wedge B_0(y) \wedge \neg \min(x) \wedge S_i(x-1, y+1) \rightarrow S_i(x, y)$  for  $i = 0, 1$ .

The general case  $1 \leq y < n$  with  $b_y = 1$  is more elaborate because of the addition  $a\beta_y = 2a\beta_{y+1} + a$ . It uses the guessed predicates  $Q_1, Q_0$  to encode the presence/absence of a carry at some index in this addition. More precisely,  $Q_1(x, y)$  (resp.  $Q_0(x, y)$ ) holds if this addition has a carry (resp. no carry) at index  $x$ . The following clauses<sup>17</sup> ( $i = 0, 1$ ) say that the least significant bit of the integer  $a\beta_y = 2a\beta_{y+1} + a$  is the least significant bit of  $a$ ; they also express that there is no carry at the least index 1:

- (5)  $\neg \max(y) \wedge B_1(y) \wedge \min(x) \wedge A_i(x) \rightarrow S_i(x, y) \wedge Q_0(x, y)$ .

The eight following clauses<sup>18</sup> that correspond to the possible triples  $(h, i, j) \in \{0, 1\}^3$  describe each bit of index  $x \in \llbracket 2, n \rrbracket$  of the sum  $a\beta_y = 2a\beta_{y+1} + a$ , and the corresponding carry:

$$(6) \quad \left( \begin{array}{l} \neg \max(y) \wedge B_1(y) \wedge \neg \min(x) \wedge \\ Q_h(x-1, y) \wedge S_i(x-1, y+1) \wedge A_j(x) \end{array} \right) \rightarrow Q_{i_1}(x, y) \wedge S_{i_0}(x, y)$$

<sup>17</sup> Here, for concision, conjunctions of atoms are allowed as conclusions of ‘‘Horn clauses’’.

<sup>18</sup> The clauses (6) express the following equality: the two bits integer (carry, bit) of index  $x$  of the sum  $a\beta_y = 2a\beta_{y+1} + a$  is the sum of the following three bits: the carry of index  $x-1$  of the sum  $a\beta_y = 2a\beta_{y+1} + a$ ; the bit of index  $x-1$  of  $a\beta_{y+1}$ ; the bit of index  $x$  of  $a$ .



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where  $i_1i_0$  is the 2 bits integer that is the sum  $h + i + j$ : which is 00, 01, 10 or 11. Finally, the clauses that define the result  $c = a \times b = a \times \beta_1$  are, for  $i \in \{0, 1\}$ ,

$$(7) \quad \min(y) \wedge S_i(x, y) \rightarrow C_i(x).$$

Clearly, INTEGERS-PRODUCT is defined by  $\Phi_{ip} \equiv \exists S_0, S_1, Q_0, Q_1 \forall x, y \varphi_{ip}$ , where  $\varphi_{ip}$  is the conjunction of the Horn clauses above. Moreover,  $\Phi_{ip}$  belongs to  $\text{ESO-HORN}^1(\forall^2, \text{arity}^2)$  and has  $\mathcal{S} = \{(-1, 1), (-1, 0)\}$  as its induction system, by clauses (4) and (6). (Notice that the other clauses give no other induction vector.) This system trivially satisfies the monotonicity condition with the line of equation  $x = 0$  as its reference hyperplane. However, noticing that here again as in the previous example (MATRICES-PRODUCT), one of the variables, which is  $y$  here, misses in this equation, one needs to modify the reference hyperplane so that its equation involves *all* the variables. We can choose for that the line of equation  $2x + y = 0$  as constructed in the proof of Lemma 18, or alternatively, the line of equation  $x - y = 0$  which is simpler to manage since the absolute value of each of its coefficients is 1.

From the monotonic Horn formula  $\Phi_{ip}$  and the line of equation  $x - y = 0$  as its reference hyperplan it is quite easy to design a CA that recognizes in *linear time* the problem INTEGERS-PRODUCT with the general method. ◀