

## 1. TASK C

In the current section we will prove that every function that can be computed in the **SFP** formalism can be expressed in the **POR** formalism, in the following section we will show that the converse holds too. In order to prove the former so we will proceed as follows:

- We will first give a formal definition of the **SFP** machines.
- Then we will define a set of the data structures and functions that can be used in order to encode a **SFP** and to simulate its execution.
- After that, we will prove that all the data structures and functions that we described can be defined in the **POR** formalism.
- Finally we will prove our result, namely that all the functions that are calculated by a **SFP** machine are **POR** functions, too.

### 1.1. Definition of the SFP formalism.

**Definition 1** (Stream machine). *A Stream Machine is a quadruple  $M_S := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$  where:*

- $\mathbb{Q}$  is a finite set of states ranged over by  $q_0, q_1, \dots, q_n$ .
- $\Sigma$  is a finite set of characters ranged over by  $c_0, c_1, \dots, c_n$ .
- $\delta_{\mathbf{SFP}} : \mathbb{Q} \times \hat{\Sigma} \times \{0, 1\} \longrightarrow \mathbb{Q} \times \hat{\Sigma} \times \{L, R\}$  is a transition function that describes the new configuration reached by a **SFP** machine.  $L, R$  are two fixed constants, and  $\hat{\Sigma} = \Sigma \cup \{*\}$   $\wedge * \neq 0 \wedge * \neq 1$ .
- $q \in \mathbb{Q}$  is an initial state.

In the former definition,  $*$  denotes a generic blank character that is not part of  $\Sigma$ ; mind that assuming  $\Sigma = \{0, 1\}$  would not be reductive. From now on we will denote the blank character  $*$  as  $c_{|\Sigma|+1}$ .

**Definition 2** (Configuration of a Stream Machine). *The configuration of a stream machine  $M$  is a quadruple  $\langle \sigma, q, \tau, \omega \rangle$  where:*

- $\sigma \in \hat{\Sigma}^*$  is the portion of the first tape to the left of the head.
- $q \in \mathbb{Q}$  is the current state of  $M$ .
- $\tau \in \hat{\Sigma}^*$  is the portion of the first tape to the right of the head.
- $\omega \in \{0, 1\}^{\mathbb{N}}$  is the portion of the second tape that hasn't been read yet.

**Definition 3** (Stream Machine Reachability function). *Given a stream machine  $M_S := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$ , we define the partial transition function  $\vdash_{\delta_{\mathbf{SFP}}} : \hat{\Sigma}^* \times \mathbb{Q} \times \hat{\Sigma}^* \times \{0, 1\}^{\mathbb{N}} \longrightarrow \hat{\Sigma}^* \times \mathbb{Q} \times \hat{\Sigma}^* \times \{0, 1\}^{\mathbb{N}}$  between two configuration of  $M_S$  as:*

$$\begin{array}{ll}
 \langle \sigma, q, c\tau, 0\omega \rangle \vdash_{\delta_{\mathbf{SFP}}} \langle \sigma c', q', \tau, \omega \rangle & \text{if } \delta_{\mathbf{SFP}}(q, c, 0) = \langle q', c', R \rangle \\
 \langle \sigma c_0, q, c_1\tau, 0\omega \rangle \vdash_{\delta_{\mathbf{SFP}}} \langle \sigma, q', c_0 c'_1 \tau, \omega \rangle & \text{if } \delta_{\mathbf{SFP}}(q, c_1, 0) = \langle q', c'_1, L \rangle \\
 \langle \sigma, q, c\tau, 1\omega \rangle \vdash_{\delta_{\mathbf{SFP}}} \langle \sigma c', q', \tau, \omega \rangle & \text{if } \delta_{\mathbf{SFP}}(q, c, 1) = \langle q', c', R \rangle \\
 \langle \sigma c_0, q, c_1\tau, 1\omega \rangle \vdash_{\delta_{\mathbf{SFP}}} \langle \sigma, q', c_0 c'_1 \tau, \omega \rangle & \text{if } \delta_{\mathbf{SFP}}(q, c_1, 1) = \langle q', c'_1, L \rangle
 \end{array}$$

**Definition 4.** *Given a stream machine  $M_S := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$ , we denote with  $\{\vdash_{\delta_{\mathbf{SFP}}}^n\}_n$  the smallest family of relations relation for which:*

$$\begin{array}{l}
 \langle \sigma, q, \tau, \omega \rangle \vdash_{\delta_{\mathbf{SFP}}}^0 \langle \sigma, q, \tau, \omega \rangle \\
 \langle \sigma, q, \tau, \omega \rangle \vdash_{\delta_{\mathbf{SFP}}}^n \langle \sigma', q', \tau', \omega' \rangle \wedge \langle \sigma', q', \tau', \omega' \rangle \vdash_{\delta_{\mathbf{SFP}}} \langle \sigma'', q', \tau'', \omega'' \rangle \rightarrow \langle \sigma, q, \tau, \omega \rangle \vdash_{\delta_{\mathbf{SFP}}}^{n+1} \langle \sigma'', q', \tau'', \omega'' \rangle
 \end{array}$$

**Lemma 1.**  $\forall n. \vdash_{\delta_{\mathbf{SFP}}}^n$  is a function.

*Proof.* By induction on  $n$

- 0 In this case the  $\vdash_{\delta_{\mathbf{SFP}}}^n$  is the identity function.  
 $n + 1$  As induction hypothesis we have that  $\vdash_{\delta_{\mathbf{SFP}}}^n$  is a function, then, since  $\vdash_{\delta_{\mathbf{SFP}}}$  is a function, and  $\vdash_{\delta_{\mathbf{SFP}}}^{n+1} = \vdash_{\delta_{\mathbf{SFP}}}^n \circ \vdash_{\delta_{\mathbf{SFP}}}$ , we have the thesis.

□

**Notation 1.** Given a stream machine  $M_S := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$  and a configuration  $\langle \sigma, q, \tau, \omega \rangle$  we denote with  $\langle \sigma, q, \tau, \omega \rangle \Vdash_{\delta_{\mathbf{SFP}}}$  the following condition:

$$\neg \exists \sigma', q', \tau', \omega'. \langle \sigma, q, \tau, \omega \rangle \vdash_{\delta_{\mathbf{SFP}}} \langle \sigma', q', \tau', \omega' \rangle$$

**Definition 5** (Value computed by a Stream Machine). Given a machine  $M_S := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$ , we say that  $M_S$  computes  $\gamma$  on input  $\sigma$  and oracle  $\omega$  if and only if:

$$\exists n. \langle \epsilon, q_0, \sigma, \omega \rangle \vdash_{\delta_{\mathbf{SFP}}}^n \langle \gamma, q', \tau, \psi \rangle \Vdash_{\delta_{\mathbf{SFP}}}$$

for some  $\tau, q', \psi$ . In that case, we write  $M_S(\sigma, \omega) = \gamma$ .

**Definition 6** (**SFP** Stream Machine). We say that a stream machine  $M_{\mathbf{SFP}} := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$  is a **SFP** Stream Machine if and only if:

$$\exists p \in \text{POLY}. \forall \sigma, \omega, n. \langle \epsilon, q_0, \sigma, \omega \rangle \vdash_{\delta_{\mathbf{SFP}}}^n \langle \gamma, q', \tau, \psi \rangle \Vdash_{\delta_{\mathbf{SFP}}} \rightarrow n \leq p(|\sigma|)$$

The result that we are addressing in the current section can be now restated as:

**Lemma 2** (Representation of Stream Machines in  $\mathcal{POR}$ ). For every deterministic **SFP** machine  $M_{\mathbf{SFP}} := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$ , there exists a  $\mathcal{POR}$  function  $f$  such that  $f(\sigma, \omega) = M_{\mathbf{SFP}}(\sigma, \omega)$ .

## 1.2. Encoding of a SFP Machine, Bottom-up.

1.2.1. *Basic data structures and functions.* The previous subsection of this work outlines some data structures and functions that we respectively need to represent and compute in order to simulate the execution of a **SFP** machines; for example, since we represented **SFP** machines as tuples of elements, we will need to represent such structures. Furthermore, since we defined the transition function by cases, we need some control structures that will allow us to implement that specific behaviour.

If we want to implement the **SFP** machine in a formalism that works on a domain  $\mathbb{D}$ , we need to implement in  $\mathbb{D}$  the following data structures:

- *Natural* numbers, that will allow us to define an encoding of characters, and states.
- *Boolean* values.
- *Strings* on a binary alphabet.
- *tuples*, by means of which we will encode tapes and configurations.

**Notation 2.** We will represent the data structures described above by means of the following notation:

- We will use the  $\mathbb{D}_S$  notation in order to denote the image of the set  $S$  modulo its encoding over  $\mathbb{D}$ .
- We will use the  $\bar{n} \in \mathbb{D}_{\mathbb{N}}$  symbol in order to denote the encoding of the number  $n \in \mathbb{N}$  in the domain  $\mathbb{D}$ .
- We will use the  $1 \in \mathbb{D}_{\{0,1\}}$  symbol in order to denote the encoding of the true boolean value and the  $0$  symbol for denoting the false boolean value as represented in  $\mathbb{D}$ .
- We will use the symbols  $\sigma, \tau \in \mathbb{D}_{\mathbb{S}}$  in order to denote the representation of a string over the alphabet  $\{0, 1\}$  (a string in  $\mathbb{S}$ ) in the domain  $\mathbb{D}$ .
- We will range the tuples on the following meta-variables:  $t_0, t_1, \dots, t_n$ .

On top of the data structures that we have introduced, we need some basic functions. To perform computations on natural numbers, we need at least the following functions:

- Addition, denoted with the  $+$  symbol.
- Subtraction, denoted with the  $-$  symbol.
- Multiplication, denoted with the  $\cdot$  symbol.
- Exponentiation, that we will represent with the common power notation.

The latter function will be useful for expressing the complexity bound of a **SFP** machine, that is a polynomial. All the functions above mentioned need to have the following signature:

$$\mathbb{D}_{\mathbb{N}} \times \mathbb{D}_{\mathbb{N}} \longrightarrow \mathbb{D}_{\mathbb{N}}$$

And, if  $*$  is a function of the above mentioned, it must hold that:

$$\forall n, m, o \in \mathbb{N}. n * m = o \rightarrow \bar{n} * \bar{m} = \bar{o}$$

Boolean values can be defined as a subset of natural numbers. It means that if we are able to define the set  $\mathbb{D}_{\mathbb{N}}$ , we will be able to define the set  $\mathbb{D}_{\{0,1\}}$  as a subset of the above mentioned set and to implement boolean values and boolean functions as a subset of the natural numbers and of the function defined on such values. For example assigning 0 to the false boolean value and 1 to the true boolean value will do the job. By means of booleans we will be able to define:

- A conditional function over a generical set  $S$ , i.e. the **if** :  $\mathbb{D}_{\{0,1\}} \times \mathbb{D}_S \times \mathbb{D}_S \longrightarrow \mathbb{D}_S$  function, which respects its commonly intended specification.
- Logical connectives.

The **if** function is an important control function that we will employ in order to determine the configuration that follows the current one. Together with the **if** structure, we will need another simple control structure: the bounded iteration, i.e. a **for** expression. It will allow us to simulate the execution of a **SFP** machine up to its polynomial bound. For all our data structures we need a binary function  $eq$  that returns 1 if its parameters are equal with respect to the identity and 0 otherwise.

We need binary strings because we will need to access handle the  $\omega$  tape, in particular we will use a simple function of random access, with the following signature:

$$\cdot[\cdot] : \mathbb{D}_{\{0,1\}}^{\mathbb{N}} \times \mathbb{D}_{\mathbb{N}} \longrightarrow \{0, 1\}$$

such that:

$$\forall \sigma \in \mathbb{D}_{\{0,1\}}^{\mathbb{N}}. \forall n \in \mathbb{D}_{\mathbb{N}}. \sigma[\bar{n}] = 1 \leftrightarrow \text{the } n\text{-th bit of } \sigma \text{ is } 1$$

**Definition 7.** Let  $\mathbb{T}_S^n$  be the set of homogeneous tuples of elements in  $S$  with cardinality  $n$ .

Finally, we need the following functions for handling tuples:

- A family of constructors, which we will use in order to build tuples of finite dimension. We will represent this function putting its elements between angular brackets. These functions will have the following signature:  $\langle \cdot, \dots, \cdot \rangle : \mathbb{D}_S^n \longrightarrow \mathbb{D}_{\mathbb{T}_S^n}$ . For example  $\langle \bar{0}, \bar{1} \rangle$ , will be the instantiation of the tuple's constructor on the encoding of 1 and 0 as first and second argument.
- A function which computes the size of a tuple, that we denote with  $|\cdot| : \mathbb{D}_{\mathbb{T}_S^n} \longrightarrow \mathbb{D}_{\mathbb{N}}$ .
- A family of projectors, which we will use in order to extract values from a tuple. We will denote these unary functions with  $\pi_i : \mathbb{D}_{\mathbb{N}} \times \mathbb{D}_{\mathbb{T}_S^n} \longrightarrow \mathbb{D}_S$  where  $i$  is the position of the element returned by the projector. If the index of the element is greater than the tuple's size, we assume that the projection function will return a default value.
- Four manipulators:
  - An unary function which deletes the rightmost element of a tuple, which we call  $rmr : \mathbb{D}_{\mathbb{T}_S^n} \longrightarrow \mathbb{D}_{\mathbb{T}_S^{n-1}}$ .

- An unary function which deletes the leftmost element of a tuple, which we call  $rml : \mathbb{D}_{\mathbb{T}_S^n} \longrightarrow \mathbb{D}_{\mathbb{T}_S^{n-1}}$ .
- A binary function which inserts an element in the rightmost position of a tuple, which we call  $addr : \mathbb{D}_S \times \mathbb{D}_{\mathbb{T}_S^n} \longrightarrow \mathbb{D}_{\mathbb{T}_S^{n+1}}$ .
- A function that inserts an element in the leftmost position of a tuple, which we call  $addl : \mathbb{D}_S \times \mathbb{D}_{\mathbb{T}_S^n} \longrightarrow \mathbb{D}_{\mathbb{T}_S^{n+1}}$ .

The two last modifiers that we have presented will be useful when handling tapes and transition, since they will allow us to simulate the movement of the machine's head.

**Definition 8** (Correctness of an encoding of tuples). *We say that an implementation of tuples is correct if and only if:*

$$\begin{aligned} \forall n. \forall 1 \leq i \leq n. \pi_i(\langle x_1, \dots, x_n \rangle) &= x_i \\ \exists x. \forall i \geq n. \pi_i(\langle x_1, \dots, x_n \rangle) &= x \end{aligned}$$

1.2.2. *Complex data structures and functions.* On top of the data structures that we have recently defined, we can define the encoding of a **SFP** machine and an emulating interpreter as described below.

**Definition 9** (Encoding of the transition function). *The encoding of a machine's transition function is defined as above:*

$$enct(\{t_0, \dots, t_N\}) = \langle g(t_0), \dots, g(t_{|\delta_{\mathbf{SFP}}|}) \rangle$$

And  $g$  as:

$$g(\langle \langle q_i, c_j, b \rangle, \langle q_k, c_l, D \rangle \rangle) := \begin{cases} \langle \bar{i}, \bar{j}, \bar{k}, \bar{l}, 0, b \rangle & \text{if } D = L \\ \langle \bar{i}, \bar{j}, \bar{k}, \bar{l}, 1, b \rangle & \text{otherwise} \end{cases}$$

The encoding of the transition function  $\delta_{\mathbf{SFP}}$  is finite, since the domain of  $\delta_{\mathbf{SFP}}$  is so. Let  $\{t_0, \dots, t_N\}$  be the *finite* subset of the set  $(\mathbb{Q} \times \hat{\Sigma} \times \{0, 1\}) \times (\mathbb{Q} \times \hat{\Sigma} \times \{L, R\})$  that describes the function. It is possible to observe that the definition above is only given by means of data structures that we have defined in the previous section, which are natural values and tuples. Furthermore, we represent the transition functions extensively, enumerating all its members.

In the same way we define the representation of tapes and configurations:

**Definition 10** (Encoding of a portion of a tape). *We encode all the finite portions of a tape  $\sigma := c_i, \dots, c_k$  as:*

$$tenc(c_i, \dots, c_k) := \langle \bar{i}, \dots, \bar{k} \rangle$$

**Definition 11** (Representation of the configuration of a stream machine). *The representation of the configuration of a stream machine is defined as the 4-uple  $\langle \sigma, \bar{i}, \tau, \bar{k} \rangle$  where:*

- $\sigma = tenc(\sigma')$ , where  $\sigma'$  is the shortest portion of the tape that starts from the cell on the immediate left of the head is followed (on its left) by an infinite sequence of blank characters  $*$ .
- $\bar{i}$  is the encoding of the index of the current state  $q_i$ .
- $\tau = tenc(\tau')$ , where  $\tau'$  is the shortest portion of the tape that starts from the cell under the head, continues on its right and is followed (again on its right) by an infinite sequence of blank characters  $*$ .
- $\bar{k}$  is the encoding of the length of the prefix of the oracle tape that has already been consumed.

**Remark 1.** The encoding of the initial state of a stream machine  $\langle \epsilon, q_0, \sigma, \omega \rangle$  is  $\langle \langle \rangle, \bar{0}, \text{tenc}(\sigma), \bar{0} \rangle$ .

Now that we have described all the static aspects of a stream machine, we can define some functions that allow us to emulate the dynamic behaviour of a machine.

1.2.3. *A first result.* We are going to prove a lemma that states that the previously defined functions can be used to emulate the execution of a Stream Machine. Once this result will be proved, in order to demonstrate the lemma that we are addressing, i.e. 2, we will only need to represent in the  $\mathcal{POR}$  formalism all the functions that we have described before.

**Lemma 3** (Implementation of **SFP**). *Each formalism which works on a domain  $\mathbb{D}$  in which it's possible to express the data structures, functions and control primitives described in the subsections 1.2.1 and 1.2.2 and that is closed under composition is at least as expressive as the Stream Machines are.*

*Proof.* Let  $M_S := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$  be a stream machine and  $t := \text{enct}(\delta_{\mathbf{SFP}})$  the encoding of its transitions as defined in 9, let  $\langle \sigma, \bar{i}', \tau, \bar{p} \rangle$  be the current configuration encoded as defined in 11. We can define on  $t$  the function which computes the matching transition as follows:

$$\begin{aligned} \text{matcht}(\langle \rangle, \langle \sigma, \bar{i}', \tau, \bar{p} \rangle, \omega) &:= 0 \\ \text{matcht}(\langle \langle \bar{i}, \bar{j}, \bar{k}, \bar{l}, d, b \rangle, t_0, \dots, t_m \rangle, \langle \sigma, \bar{i}', \tau, \bar{p} \rangle, \omega) &:= \\ := \begin{cases} \langle \bar{i}, \bar{j}, \bar{k}, \bar{l}, d, b \rangle & \text{if } \bar{i} = \bar{i}' \wedge \bar{j} = \pi_1(\tau) \wedge \omega(\bar{p}) = b \\ \text{matcht}(\langle t_0, \dots, t_m \rangle, \langle \sigma, \bar{i}', \tau, \bar{p} \rangle, \omega) & \text{otherwise} \end{cases} \end{aligned}$$

Now we need to define a function that applies a transition to a state:

$$\begin{aligned} \text{apply}(0, \langle \sigma, \bar{i}', \tau, \bar{p} \rangle) &:= \langle \sigma, \bar{i}', \tau, \bar{p} \rangle \\ \text{apply}(\langle \bar{i}, \bar{j}, \bar{k}, \bar{l}, d, b \rangle, \langle \sigma c_1, \bar{i}', c_2 \tau, \bar{p} \rangle) &:= \begin{cases} \langle \text{addr}(\sigma c_1, \bar{l}), \bar{k}, \tau, \bar{p} + 1 \rangle & \text{if } d = R \\ \langle \sigma, \bar{k}, \text{addl}(c_1, \text{addl}(\bar{l}, \tau)), \bar{p} + 1 \rangle & \text{otherwise} \end{cases} \end{aligned}$$

Finally we need a function that emulates the execution of the machine for a fixed number of steps, passed as a parameter. We define it by induction on the number of steps.

$$\begin{aligned} \text{step}(t, s, 0, \omega) &:= s \\ \text{step}(t, s, n + 1, \omega) &:= \text{apply}(\text{matcht}(t, \text{step}(t, s, n, \omega), \omega), \text{step}(t, s, n, \omega), n) \end{aligned}$$

Finally we define  $\text{eval}_M(\sigma, \omega)$  as follows

$$\text{eval}_{\langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle, n}(\sigma, \omega) := \text{step}(\text{enct}(\delta_{\mathbf{SFP}}), \langle \langle \rangle, \bar{0}_\omega, \text{tenc}(\sigma), \bar{0} \rangle, \bar{n}, \omega)$$

If  $n$  is sufficiently big,  $\text{eval}_{\langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle, n}(\sigma, \omega) = M_S(\sigma, \omega)$

□

**1.3. Expressivity of the  $\mathcal{POR}$  formalism.** The aim of this section is to show that all the encodings and functions over the set  $\mathbb{D}$  that we introduced in Section 1.2 can be expressed in the  $\mathcal{POR}$  formalism. To do so, we need to define some even simpler functions that we will use to build the others.

Before doing so, we will give some intuitions about the actual implementation of the structures:

- The set  $\mathbb{D}$  consists in the set  $\mathbb{S}$ , which is the set of the binary strings.
- The strings over the set  $\{0, 1\}$  are native in  $\mathcal{POR}$ , so they won't need to be implemented, i.e.  $\mathbb{D}_{\mathbb{S}}$  is  $\mathbb{S}$  itself; furthermore, the only random access to such strings will be the reading of some bits of the oracle  $\omega$  for which  $\mathcal{POR}$  has a primitive function  $f_q$ .
- Numbers will be represented in unary notation, starting from the 1 string. This means that  $\mathbb{D}_{\mathbb{N}} = 1^+$ .

1.3.1. *Preliminaries on Strings.* Since the binary strings are the only datatype that is present in the  $\mathcal{POR}$  formalism, we would like to express some basic operations on such data that will turn out to be useful in the following parts.

Given a variable name  $x$ , and an oracle  $\omega$ , every constant can be represented in  $\mathcal{POR}$  as follows:

**Definition 12** (Constant Strings). *A constant string  $c_0c_1 \dots c_n$  can be computed in  $\mathcal{POR}$  by means of the following function:*

$$c_0c_1 \dots c_n := C_{c_n}(\dots (C_{c_1}(C_{c_0}(E(x, \omega), \omega), \omega), \dots), \omega)$$

Similarly, the concatenation between two generic strings can be defined as follows:

**Definition 13** (Concatenation of Strings).

$$\begin{aligned} \text{concat}(x, \epsilon, \omega) &:= x \\ \text{concat}(x, y0, \omega) &:= C_0(\text{concat}(x, y, \omega), \omega)|_{xy} \\ \text{concat}(x, y1, \omega) &:= C_1(\text{concat}(x, y, \omega), \omega)|_{xy} \end{aligned}$$

We also define the meta-language notation  $|\cdot| : \mathbb{S} \rightarrow \mathbb{N}$  that represents the size of a string, further we will introduce a  $\mathcal{POR}$  function that computes the size of a tuple  $|\cdot|_{\omega} : SS_{\mathbb{T}_{\mathbb{S}}}^n \rightarrow \mathbb{S}_{\mathbb{N}}$ . Those functions are not the same object.

Since constants and concatenations of strings are representable in  $\mathcal{POR}$ , we will use respectively their explicit representation and juxtaposition for representing those operations.

**Notation 3** (Constant Strings). *When introducing constant string we will use their explicit notation instead of writing their definition in  $\mathcal{POR}$ , for example:*

$$\text{concat}(C_0(C_1(E(x, \omega))), C_1(C_0(E(x, \omega))), \omega)$$

*Will be written as 1001 and  $E(x, \omega)$  will be written as  $\epsilon$ .*

Before going further with the encoding of numerals and arithmetical operations, we would like to point out the fact that all the  $\mathcal{POR}$  functions need to have an oracle as parameter: for example, we passed  $\omega$  to the *concat* function. When dealing with data structures, and manipulating functions, it will often be useless but, at least inside the definitions, we will use it.

The implementation of some functions will be easier if we can use a function that reverses the strings. Such a function can be defined in the  $\mathcal{POR}$  formalism as follows:

**Definition 14** (Reversing Function).

$$\begin{aligned} rv(\epsilon, \omega) &:= \epsilon \\ rv(y0, \omega) &:= 0rv(y, \omega)|_{y0} \\ rv(y1, \omega) &:= 1rv(y, \omega)|_{y0} \end{aligned}$$

1.3.2. *Natural Numbers.* All the finite natural numbers can be represented as follows:

**Definition 15** (Encoding of  $\mathbb{N}$  over  $\mathbb{S}$ ).

$$\begin{aligned}\bar{0}_\omega &:= C_1(\epsilon, \omega) \\ \overline{n+1}_\omega &:= C_1(\bar{n}_\omega, \omega)\end{aligned}$$

**Remark 2** (Set  $\mathbb{S}_\mathbb{N}$ ). *As we pointed out before, there's a bijection between  $1^+ \subset \mathbb{S}$  and  $\mathbb{N}$ , so  $\mathbb{S}_\mathbb{N} = 1^+$ .*

**Remark 3** (Size of numbers).

$$\forall n \in \mathbb{N}. |\bar{n}_\omega| = n + 1$$

**Remark 4** (Appropriateness of  $\mathbb{S}_\mathbb{N}$ ). *We say that  $\mathbb{S}_\mathbb{N}$  is an appropriate representation of  $\mathbb{N}$ , meaning that:*

- $\forall e \in \mathbb{S}_\mathbb{N}. \exists! n \in \mathbb{N}. \bar{n}_\omega = e$
- $\forall n \in \mathbb{N}. \exists! e \in \mathbb{S}_\mathbb{N}. \bar{n}_\omega = e$

*Proof.* The results can be respectively obtained by induction on the size of the element and on the value of  $n$ .  $\square$

The successor of a number  $n$ , passed to the formal parameter  $y$ , can be calculated simply adding an 1 at the end of  $n$ , i.e. as follows:

**Definition 16** (Successor function). *We define a function  $S : \mathbb{S}_\mathbb{N} \longrightarrow \mathbb{S}_\mathbb{N}N$  that computes the successor of a number:*

$$\begin{aligned}S(\epsilon, \omega) &:= \epsilon \\ S(y0, \omega) &:= C_0(\epsilon, \omega)|_{y11} \quad (*) \\ S(y1, \omega) &:= C_1(C_1(y), \omega)|_{y11}\end{aligned}$$

Since the representation of a number is only composed by 1s, the row marked with  $(*)$  is useless when dealing with natural numbers. All the other functions that we will implement for the manipulation of numbers will present a similar issue, but sometimes, those definition will turn out to be useful.

**Definition 17** (Predecessor of a Natural Number). *If  $y \in \mathbb{S}_\mathbb{N}$  is the encoding of a number, the  $pd$  function calculates its predecessor simply by removing its last digit.*

$$\begin{aligned}pd(\epsilon, \omega) &:= \epsilon \\ pd(y0, \omega) &:= y|_y \\ pd(y1, \omega) &:= y|_y\end{aligned}$$

**Remark 5.**

$$\forall \sigma, c_1, c_2. pd(pd(\sigma c_1 c_2)) = \sigma$$

*Proof.* The claim is a trivial consequence of the definition of  $pd$ .  $\square$

**Notation 4** (Oracle Included Function). *From now on, we will use the notation  $op_\omega$  to denote an operator  $op$  that uses  $\omega \in \mathbb{B}$  as oracle, i.e. the expression  $xop_\omega y$  will be a shorthand for  $op(x, y, \omega)$ .*

**Definition 18** (Sum of two Natural Numbers). *The sum of two numbers  $+_\omega$  can be easily implemented by using the operation of concatenation that we introduced in 1.3.1.*

$$\bar{n} +_\omega \bar{m} = pd(concat(\bar{n}, \bar{m}, \omega))$$

The  $\mathcal{POR}$  encoding of the difference between two numbers is quite cumbersome, because the only form of recursion that is allowed by such formalism is on the longest non trivial prefix of a single argument. Intuitively we can decrease the measure of both the numbers since the second is  $\epsilon$ ; at that time, we have decreased the first argument one too many, so we need to return its successor.

**Definition 19** (Difference of two Natural Numbers). *We define the function  $-_\omega$  which encodes the difference between two natural numbers as follows*

$$\begin{aligned} x -_\omega \epsilon &:= S(x, \omega) \\ x -_\omega y0 &:= pd(x, \omega) -_\omega y|_x \\ x -_\omega y1 &:= pd(x, \omega) -_\omega y|_x \end{aligned}$$

In order to multiply two values  $x$  and  $y$ , we can remove their last digit, so that their size is equal to the number that they encode, then concatenate  $x$  to itself  $y$  time and return the successor of the number that we get; formally:

**Definition 20** (Multiplication of two Natural Numbers). *We define the multiplication between two natural numbers  $\cdot_\omega$  as follows:*

$$\begin{aligned} x \cdot_\omega^* \epsilon &:= \epsilon \\ x \cdot_\omega^* y0 &:= ((x \cdot_\omega^* y) +_\omega x)|_{x \times y1} \\ x \cdot_\omega^* y1 &:= ((x \cdot_\omega^* y) +_\omega x)|_{x \times y1} \end{aligned}$$

$$x \cdot_\omega y := S(pd(x, \omega) \cdot_\omega^* pd(y, \omega))$$

With this encoding we cannot go much further since the computation of an exponential would require an exponential size for the representation of the output, but our iteration is bounded by a term in  $L_{\mathbb{W}}$ , but the size of a number in our encoding is linear in its value. However, we can still represent monomials, and so polynomials, as follows:

**Definition 21** (Monomials). *Given a  $k \in \mathbb{N}$ , we define the function that computes the value  $\bar{n}^k_\omega$  as follows:*

$$\forall k \in \mathbb{N}. \bar{n}^k_\omega := \left( \prod_{i=0}^k \right)_\omega \bar{n}$$

where:

$$\left( \prod_{i=0}^k \right)_\omega e_i := 1 \cdot_\omega e_0 \cdot_\omega e_1 \cdot_\omega \dots \cdot_\omega e_k$$

**1.3.3. Boolean algebra.** Before going further, it's time to define some predicates. As it happens in the definition of the  $\mathcal{POR}$  function  $Q$ , we say that a predicate  $P(\vec{x})$  is true if (and only if) it returns 1, it's false if it returns 0, otherwise it's undefined. Let's start with some zero-order logic and predicates.



Given that  $x_1$  and  $x_2$  are values, we can define a function that returns  $x_1$  if a condition  $y$  is met,  $x_2$  if such condition is false, and  $\epsilon$  otherwise. Such function behaves as an **if** expression:

**Definition 22** (**if** expression).

$$\begin{aligned}\mathbf{if}'(x_1, x_2, \epsilon, \omega) &:= \epsilon \\ \mathbf{if}'(x_1, x_2, y0, \omega) &:= x_2|_{x_1x_2} \\ \mathbf{if}'(x_1, x_2, y1, \omega) &:= x_1|_{x_1x_2} \\ \mathbf{if}(t, f, c, \omega) &:= \mathbf{if}'(t, f, c, \omega)\end{aligned}$$

**Definition 23** (Logical connectives). *Thanks to the  $\mathbf{if}$  function, we can define some basic (and complete) connectives for the propositional logic:*

$$\begin{aligned}(P_1 \wedge P_2)(\vec{x}, \omega) &:= \mathbf{if}(P_2(\vec{x}, \omega), 0, P_1(\vec{x}, \omega), \omega) \\ (P_1 \vee P_2)(\vec{x}, \omega) &:= \mathbf{if}(1, P_2(\vec{x}, \omega), P_1(\vec{x}, \omega), \omega) \\ (\neg P)(\vec{x}, \omega) &:= \mathbf{if}(0, 1, P(\vec{x}, \omega), \omega)\end{aligned}$$

Now we define some predicates that will help us to develop the tuple's encoding. In particular, we will represent such structures, representing their values with an encoding that prefixes a 1 to each bit of their binary representation. For this reason, when we will decode a tuple's value it will be useful to know whether the length of the remaining part of such value is even or odd in order to decide whether it's the case to keep or remove a certain bit.

**Definition 24** (Basic Logical predicates). *The basic logical predicates in  $\mathcal{POR}$  are the function defined below.*

$$\begin{aligned}\mathit{odd}(\epsilon, \omega) &:= 0 \\ \mathit{odd}(y0, \omega) &:= \neg(\mathit{odd}(y))|_0 \\ \mathit{odd}(y1, \omega) &:= \neg(\mathit{odd}(y))|_0 \\ \mathit{even}(x, \omega) &:= \neg\mathit{odd}(x, \omega) \\ \mathit{eq}(x, y, \omega) &:= Q(x, y, \omega) \wedge Q(y, x, \omega)\end{aligned}$$

It's important to observe that the *odd* and *even* predicates work as their opposites for the encoding of natural numbers, i.e. the following remark holds:

**Remark 6** (Even and odd's idiosyncrasy).

$$\begin{aligned}\forall \sigma \in \{0, 1\}^*. \forall b \in \{0, 1\}. \mathit{odd}(\sigma) &\leftrightarrow \mathit{even}(\sigma b) \\ \forall n \in \mathbb{N}. \mathit{odd}(\overline{n}_\omega) &\leftrightarrow n \text{ is even.}\end{aligned}$$

Before defining tuples, let's proceed with the definition of two string-specific predicates:

**Definition 25** (String specific Predicates). *For working with strings we define the following predicates which respectively extract the rightmost and the leftmost bit of the string.*

$$\begin{aligned}
lst(\epsilon, \omega) &:= \epsilon \\
lst(y0, \omega) &:= 0|_1 \\
lst(y1, \omega) &:= 1|_1 \\
\\ 
fst(\epsilon, \omega) &:= \epsilon \\
fst(y0, \omega) &:= If(0, fst(y, \omega), Q(y, \epsilon, \omega), \omega)|_1 \\
fst(y1, \omega) &:= If(1, fst(y, \omega), Q(y, \epsilon, \omega), \omega)|_1
\end{aligned}$$

**Remark 7.**

$$\begin{aligned}
\forall \sigma \in \{0, 1\}^*. lst(\sigma 1) &= 1 \wedge lst(\sigma 0) = 0 \\
\forall \sigma \in \{0, 1\}^*. fst(1\sigma) &= 1 \wedge fst(0\sigma) = 0
\end{aligned}$$

1.3.4. *tuples.* In order to represent the tuples, we use Odifreddi's notation as described in (vol 2 p.183). The encoding, that we have briefly introduced in 1.3.3 makes use of a couple of functions  $\mathcal{D}$  and its left inverse  $\mathcal{H}$ .

**Definition 26** (Encoding and Decoding Functions).

$$\begin{aligned}
\mathcal{D}(\sigma 0) &:= \mathcal{D}(\sigma)10 \\
\mathcal{D}(\sigma 1) &:= \mathcal{D}(\sigma)11 \\
\\ 
\mathcal{H}(\sigma 10) &= \mathcal{H}(\sigma)0 \\
\mathcal{H}(\sigma 11) &= \mathcal{H}(\sigma)1
\end{aligned}$$

Thanks to this simple encoding we can represent tuples simply by juxtaposing their values separated by a special character, for example 00, because such sequence can't be generated by  $\mathcal{D}$  (the proof of such result can be shown by induction on its first argument).

It is possible to show that the doubling function  $\mathcal{D}$  is a  $\mathcal{POR}$  function

$$\begin{aligned}
\mathcal{D}(\epsilon, \omega) &:= \epsilon \\
\mathcal{D}(y1, \omega) &:= C_1(C_1(\mathcal{D}(y, \omega), \omega))|_{(11) \times (y1)} \\
\mathcal{D}(y0, \omega) &:= C_0(C_1(\mathcal{D}(y, \omega), \omega))|_{(11) \times (y1)}
\end{aligned}$$

It is easy to see that, given any string  $c_0 c_1 \dots c_n$ ,  $\mathcal{D}(c_0 c_1 \dots c_n) = 1c_0 1c_1 \dots 1c_n$ . The function  $\mathcal{H}$  is in  $\mathcal{POR}$  as well as the doubling function  $\mathcal{D}$ . We can define it as follows:

$$\begin{aligned}
\mathcal{H}(\epsilon, \omega) &:= \epsilon \\
\mathcal{H}(y0, \omega) &:= concat(\mathcal{H}(y, \omega), If(0, \epsilon, odd(y, \omega), \omega))|_{y0} \\
\mathcal{H}(y1, \omega) &:= concat(\mathcal{H}(y, \omega), If(1, \epsilon, odd(y, \omega), \omega))|_{y0}
\end{aligned}$$

**Lemma 4** ( $\mathcal{D}$ 's left-inverse).  $\forall \sigma \in \{0, 1\}^*, \omega, \mathcal{H}(\mathcal{D}(\sigma, \omega), \omega) = \sigma$

*Proof.* By (right) induction on  $\sigma$ :

$\epsilon$  The thesis comes from a trivial rewriting of the two functions' bodies.

$\tau c$  The thesis is

$$\begin{aligned}\mathcal{H}(\mathcal{D}(\tau c, \omega), \omega) &= \tau c \\ \mathcal{H}(\mathcal{D}(\tau, \omega)1c, \omega) &= \tau c\end{aligned}$$

By induction on  $\sigma$  we can also prove that  $\forall \sigma. \text{odd}(\mathcal{D}(\sigma)) = 0$ , so we can simplify our claim as follows:

$$\begin{aligned}\mathcal{H}(\mathcal{D}(\tau, \omega)1c, \omega) &= \tau c \\ \mathcal{H}(\mathcal{D}(\tau, \omega)1, \omega)c &= \tau c \\ \mathcal{H}(\mathcal{D}(\tau, \omega)1, \omega) &= \tau\end{aligned}$$

We have argued that  $\forall \sigma. \text{odd}(\mathcal{D}(\sigma)) = 0$ , so we can state the claim as:

$$\begin{aligned}\mathcal{H}(\mathcal{D}(\tau, \omega)1, \omega) &= \tau \\ \mathcal{H}(\mathcal{D}(\tau, \omega), \omega) &= \tau\end{aligned}$$

Which is the induction hypothesis, so we proved our lemma. □

We can finally define the encoding of tuples as follows:

**Definition 27** (Tuple Constructors). *We define the family of tuple constructors as the family of function defined as below and indexed by  $n$ :*

$$\langle x_0, x_1, \dots, x_n \rangle_\omega := 00\mathcal{D}(x_n)00 \dots 00\mathcal{D}(x_1)00\mathcal{D}(x_n)00\mathcal{D}(\bar{n}_\omega)00$$

We represent tuples of string by encoding all the possible values with sequences of two characters, and using 00 as separators. We now implement a function that allows us to remove the initial separators. The function(s)  $\langle \cdot \rangle$  is in  $\mathcal{POR}$  because they are defined by means of composition of concatenation and  $\mathcal{D}$ , that are both in  $\mathcal{POR}$ .

The definition of the tuple's constructor introduces a countable set of function, rather than a single function. Further we will show how to parameterize in  $\mathcal{POR}$  such functions.

**Remark 8** (tuple's size). *the size of a tuple  $\langle x_0, \dots, x_n \rangle_\omega$  is  $O(n + \max(|x_0|, \dots, |x_n|))$ .*

*Proof.* the size of a tuple can be expressed as  $\sum_{i=0}^n 2 \cdot |x_n| + 3n$  that is in  $O(n + \max(|x_0|, \dots, |x_n|))$ . □

Now, it's time to introduce the projectors, we will build them by many step. The first consists in the definition of a function that removes the separators (namely 00) from the encoding of a tuple.

**Definition 28** (Remover of the separator). *We define  $rmsep$ , i.e. the function which is intended to remove a separator in a tuple as the double nesting of the  $pd$  function, namely:*

$$rmsep(x) := pd(pd(x))$$

We start describing how to extract the right-most component. To do so, we define the following functions:

- $sz$  which returns 1 if ad only if the rightmost element of its first argument is 0, and otherwise it returns 0.
- $rc'$  which extracts the rightmost element of a tuple, without decoding it.
- $rc$  which is basically the function obtained wrapping  $rc'$  with  $\mathcal{H}$  in order to decode the tuple's encoding.

**Definition 29** (Functions for values' extraction). *The three functions that we have described above are implementable in  $\mathcal{POR}$  as follows:*

$$\begin{aligned}
sz(x, \omega) &:= eq(lst(x), 0) \\
rc'(\epsilon, \omega) &:= \epsilon \\
rc'(y0, \omega) &:= \text{if}(\epsilon, \text{concat}(rc'(y, \omega), 0), sz(y), \omega)|_{y0} \\
rc'(y1, \omega) &:= \text{concat}(rc'(y, \omega), 1, \omega)|_{y0} \\
rc(t, \omega) &:= \mathcal{H}(rc'(rmsep(t), \omega), \omega)
\end{aligned}$$

**Remark 9** (Correctness of  $rc$ ). *the following statements are valid:*

- $\forall x_0, x_1, \dots, x_n, \omega. rc'(00\mathcal{D}(x_0, \omega)00\mathcal{D}(x_1, \omega)00 \dots 00\mathcal{D}(x_n, \omega)) = \mathcal{D}(x_n)$
- $\forall x_0, x_1, \dots, x_n, \omega. rc(00\mathcal{D}(x_0, \omega)00\mathcal{D}(x_1, \omega)00 \dots 00\mathcal{D}(x_n, \omega)00) = x_n$

The only non-trivial statement is the first, that comes from the fact that 00 cannot appear inside  $\mathcal{D}(y, \omega)$ .

We can define a function which extracts the left sub-tuple of a tuple in a similar fashion to how we defined the  $rc$  function. This function is aimed to compute the part of a tuple that isn't returned by  $rc$ . Such value isn't actually a tuple, because, our definition records the cardinality of the tuple in its right-most element; so the values returned by  $lc$  aren't tuples because their rightmost element doesn't necessarily encode the tuple's cardinality.

The  $lc$  function is in  $\mathcal{POR}$ , indeed:

**Definition 30** (Left sub-tuple). *The function  $lc$  which computes the left sub-tuple of a tuple  $t$  is defined as follows:*

$$\begin{aligned}
lc'(\epsilon, \omega) &:= \epsilon \\
lc'(y0, \omega) &:= \text{if}(y0, lc'(y, \omega), sz(y) \wedge_{\omega} odd(t), \omega)|_y \\
lc'(y1, \omega) &:= lc'(y)|_y \\
lc(t, \omega) &:= lc'(rmsep(t), \omega)
\end{aligned}$$

Please observe that if the function  $lc$  is applied on tuples, the condition  $odd(t)$  is not required because when reading the encoding of a tuple from right to left, if we find a sequence 00, it's due to the presence of a separator. Differently, if we apply  $lc$  to that value obtained reversing a tuple, we can find the sequence 000, in that case we need to stop after that we have read the first tho 0. For doing that, we leverage the  $odd$  predicate which is true if the remaining part of the encoding has odd length. More formally we can state the correctness of our definition throughout the following remarks.

**Remark 10** (Correctness of  $lc$ ). *The following statement holds:*

$$\forall \sigma, \tau, x, \omega. lc'(\sigma 00\mathcal{D}(\tau)) = \sigma 00$$

**Remark 11** (Left component of the reverse of a tuple).

$$rv(lc(rv(00\mathcal{D}(x_0)00\mathcal{D}(x_1)00 \dots 00\mathcal{D}(x_n)00))) = 00\mathcal{D}(x_1)00 \dots 00\mathcal{D}(x_n)00$$

For every  $n \in \mathbb{N}$ , we define the  $n$ -th projector of a tuple by nesting  $n$  calls to  $lc$ , the resulting value will have the  $n$ -th element of the starting tuple as its rightmost element, we recall that the values inside a tuple are stored in decreasing left to right order.

**Definition 31** (Family of projectors). *we define the family of projectors  $\pi_n$  as below. We also overload the symbol  $\pi$  with the definition of a function which takes  $\bar{n}_\omega$  and behaves the same as  $\pi_n$ .*

$$\begin{aligned}\pi'(t, \epsilon, \omega) &:= t \\ \pi'(t, y0, \omega) &:= lc(\pi'(t, y, \omega), \omega)|_t \\ \pi'(t, y1, \omega) &:= lc(\pi'(t, y, \omega), \omega)|_t\end{aligned}$$

$$\begin{aligned}\pi_n(t, \omega) &:= rc(\pi'(t, pd(\bar{n}_\omega), \omega)) \\ \pi(t, x, \omega) &:= rc(\pi'(t, pd(x), \omega))\end{aligned}$$

We intentionally overloaded the  $\pi$  in order to increase the readability of future definitions.

**Remark 12** (Correctness of the tuple's encoding).

$$\begin{aligned}\pi_n(t, \omega) &= rc(lc^n(t, \omega)) \\ \forall n, \omega. \forall x_1, \dots, x_{n-1}, x_n. \forall 1 \leq k \leq n. \pi_k(\langle x_1, \dots, x_{n-1}, x_n \rangle_\omega) &= x_k \\ \forall n, \omega. \forall x_1, \dots, x_{n-1}, x_n. \forall k > n. \pi_k(\langle x_1, \dots, x_{n-1}, x_n \rangle_\omega) &= \epsilon \\ \forall n, \omega. \forall x_1, \dots, x_{n-1}, x_n. \pi_0(\langle x_1, \dots, x_{n-1}, x_n \rangle_\omega) &= \bar{n}_\omega \\ \forall t, n, \omega. \pi(t, \bar{n}_\omega, \omega) &= \pi_n(t, \omega) \\ \forall n \geq 1, m. \pi_n(1^m) &= \epsilon\end{aligned}$$

As a corollary of the previous remark, we have can define the function which computes the size of a tuple as follows:

$$|\langle x_0, x_1, \dots, x_n \rangle_\omega|_\omega := \pi_0(\langle x_0, x_1, \dots, x_n \rangle_\omega, \omega)$$

Now we need to define the modifiers that we introduced in 1.2.1; Let us start with the modifiers which removes the right- or the left-most element of a tuple.

**Definition 32** (Element Removers). *We define the removers that we introduced in 1.2.1 as follows:*

$$\begin{aligned}rmr(t, \omega) &:= lc(lc(t))\mathcal{D}(pd(|t|_\omega, \omega), \omega)00 \\ rml(t, \omega) &:= lc(rv(lc(rv(t, \omega), \omega), \omega), \omega)\mathcal{D}(pd(|t|_\omega, \omega), \omega)00\end{aligned}$$

The  $rmr$  function applies two times  $lc$  in order to remove the length of the tuple and the last element, after that, it appends the decreased and re-encoded length of the tuple.

The tuple obtained by removing the leftmost element can be obtained by reversing the tuple, using  $lc$  and then reversing the tuple again, as a consequence of remark 11. This part of the task is accomplished thanks to the three inner nested function calls to  $rc$  and  $lv$ , then the last value (the length) is removed and the length of the tuple is updated.

Similarly we can add a new element to a tuple following the same pattern that we used above: we remove the old length, perform the modification that we need (i.e. we add the element) and then we append the updated length. Finally, we can add an element to the left of a tuple, reversing the tuple, computing the encoding of the value throughout  $\mathcal{D}$ , appending it to the tuple

followed by a new separator and then reversing again the tuple. Then we only need to update the tuple's length.

**Definition 33** (Element Appenders). *We define the appenders that we introduced in 1.2.1 as follows:*

$$\begin{aligned} addr(t, x, \omega) &:= lc(t, \omega) \mathcal{D}(x, \omega) 00 \mathcal{D}(S(|t|_\omega, \omega), \omega) 00 \\ addl(t, x, \omega) &:= lc(rv(rv(t, \omega) rv(\mathcal{D}(x, \omega), \omega) 00, \omega), \omega) S(|t|_\omega) 00 \end{aligned}$$

Now we can also give an inductive definition of the tuple constructors:

**Definition 34** (Inductive definition of tuple's constructors).

$$\begin{aligned} \langle \rangle_\omega &:= 001100 \\ \langle x_0, \dots, x_{n-1}, x_n \rangle_\omega &:= addr(\langle x_0, \dots, x_{n-1} \rangle_\omega, x_n, \omega) \end{aligned}$$

**1.4. The first inclusion.** Now, all the basic functions introduced in Section 1.2.1 have been defined in  $\mathcal{POR}$ . Now we should proceed by showing the  $\mathcal{POR}$  implementation of the functions in Section 1.2.2.

**Lemma 5.** *The functions  $matcht$ ,  $apply$ ,  $step$  and  $eval$  as defined in lemma 3 are in  $\mathcal{POR}$ .*

*Proof.* We can define the function  $matcht$  by induction on the cardinality of the tuple. Given  $t$  the encoding of the transition function  $\delta$  by means of tuples, as defined in 9, and the current configuration  $c$ , as defined in 11. The function  $matcht$  analyses all the elements of the encoding of  $\delta$  and checks:

- Whether the current element of  $\omega$  corresponds to the value in the transition.
- Whether the current state matches with the one reported in the transition.
- Whether the value on the main tape corresponds to the value reported in  $t$ .

Its implementation under the  $\mathcal{POR}$  formalism is the following.

$$\begin{aligned} matcht'(t, \epsilon, c, \omega) &:= 0 \\ matcht'(t, yb, c, \omega) &:= \text{if}(\pi(t, yb, \omega), matcht'(t, y, c, \omega), \\ &\quad eq(\pi_1(\pi(t, yb, \omega)), \pi_2(c, \omega), \omega) \wedge_\omega \\ &\quad eq(\pi_2(\pi(t, yb, \omega)), \pi_1(\pi_3(c, \omega), \omega), \omega) \wedge_\omega \\ &\quad eq(query(pd(\pi_4(c, \omega)), \omega), \pi_6(\pi(t, yb, \omega), \omega), \omega), \omega)|_t \\ matcht(t, c, \omega) &:= nextt'(t, |t|_\omega, c, \omega) \end{aligned}$$

The application of a transition to the current configuration acts as follows:

- (1) It checks if the transition is 0, if it is true, the configuration is unchanged.
- (2) It checks if the transition moves the head to the right or to the left by means of the condition  $eq(\pi_5(t, \omega), \bar{0}_\omega)$ .
- (3) It computes builds a new tuple that encodes the resulting configuration

$$\begin{aligned}
\text{apply}(t, s, \omega) := & \text{if}(s, \text{if}(\langle \text{rmr}(\pi_1(s, \omega), \omega), \\
& \pi_3(t, \omega), \\
& \text{addl}(\text{addl}(\pi_3(s, \omega), \pi_4(t), \omega), \text{rc}(\text{lc}(\pi_1(s, \omega), \omega), \omega), \omega), \\
& S(\pi_4(s, \omega), \omega))_\omega, \\
& \langle \text{addr}(\text{rmr}(\pi_1(s, \omega), \omega), \pi_4(t), \omega), \pi_3(t, \omega), \text{rml}(\pi_3(s, \omega), \omega), S(\pi_4(s, \omega), \omega))_\omega, \\
& \text{eq}(\pi_5(t, \omega), \bar{0}_\omega), \omega), \text{eq}(t, 0, \omega), \omega)
\end{aligned}$$

Finally the function  $\text{step}$  of lemma 3 can be easily translated in  $\mathcal{POR}$ .

$$\begin{aligned}
\text{step}'(t, s, \epsilon, \omega) &:= s \\
\text{step}'(t, s, yb, \omega) &:= \text{apply}(\text{nextt}(t, \text{step}'(s, y, \omega), \omega), \text{step}'(s, y, \omega), \omega)
\end{aligned}$$

$$\text{step}(t, s, x, \omega) := \text{step}(t, s, \text{pd}(x, \omega), \omega)$$

Finally we define  $\text{eval}_\mu(\sigma, \omega)$  as follows

$$\text{eval}_{\mu, n}(\sigma, \omega) := \text{step}(\pi_3(\mu, \omega), \langle \langle \rangle_\omega, \pi_4(\mu, \omega)_\omega, \text{tenc}(\sigma), \bar{0}_\omega \rangle_\omega, \bar{n}_\omega, \omega)$$

□

Finally we can prove that that every function that can be expressed by a **SFP** machine can be expressed in  $\mathcal{POR}$ , too.

**Proposition 1.** *For each **SFP** machine  $M_{\mathbf{SFP}} := \langle \mathbb{Q}, \Sigma, \delta_{\mathbf{SFP}}, q_0 \rangle$ , there exists a  $\mathcal{POR}$  function  $f$  such that  $f(\sigma, \omega) = M_{\mathbf{SFP}}(\sigma, \omega)$ .*

*Proof.* Summing together lemma 3 and 5, we get the result. □

**Lemma 6.** *The encoding of the initial state of a **SFP** machine  $M(\sigma, \omega)$  in  $\mathcal{POR}$  is polynomial in the size  $\sigma$ .*

*Proof.* The size of a tuple is given by the remark 8, and thanks to it we can prove that the encoding of the initial state of the machine is  $O(4 + \max(6 + |\text{tenc}(\sigma)|))$ . The size of the representation of  $\text{tenc}(\sigma)$  in  $\mathcal{POR}$  is linear in the size of  $\sigma$ :  $\Sigma$  has a constant number of characters  $k$ , so the size of the encoding of a string  $\sigma$  is  $O(|\sigma| + 2k)$ , that is  $O(|\sigma|)$ , for that reason the encoding of the initial state of the machine is polynomial in the size of the initial value  $\sigma$ . □

**Lemma 7.** *The representation in  $\mathcal{POR}$  of the complexity bound of a **SFP** machine  $M(\sigma, \omega)$  in  $\mathcal{POR}$  is polynomial in the size  $\sigma$ .*

*Proof.* By definition of **SFP** machine, there exists a polynomial  $p$ , that expresses the bound in the size of the encoding of  $\sigma$ , i.e.  $|\sigma|$ , the size of  $\bar{p}(|\sigma|)_\omega$  is still polynomial in  $|\sigma|$ , because of remark 3. □

**1.5. The other direction.** It is interesting to observe that the previous encoding was loseless: we showed that a machine which uses random bits in a sequential way can naturally be represented by a function which is allowed to access random pieces of information in a *random* mode. The encoding was strong and preserved the identity of the  $\omega$  function from the **SFP** machine to the  $\mathcal{POR}$  function. Unfortunately (or not), the converse does not hold, too. The reason lies in the fact that a  $\mathcal{POR}$  function can query the oracle  $\omega$  on any value. This means that if  $n$  is the length

of an  $s \in \mathbb{S}$  computed by a  $\mathcal{POR}$  function, which we will show being polynomial in its inputs, the position in the tape in which the value  $\omega(s)$  is stored can grow exponentially in  $n$ .

Even intuitively, this should not be too much of a problem: the *amount* of randomness which is used by a  $\mathcal{POR}$  function will be shown to be polynomial in the size of its inputs, so a **SFP** machine can access the same amount of randomness but in a different order and in a sequential way. So the  $\omega$  function cannot be preserved during the encoding.

This is similar to what happens when attempting to encode a RAM into an ordinary TM: the complexity is preserved, but modulo a polynomial overhead. Similar, or even worse, the output of the destination **SFP** won't be preserved strongly, instead *the measure of the cylinders* leading the encoded and the encoding functions will be preserved.

This section will be organized as follows:

- We will first introduce the  $\mathcal{SIFP}$  formalism with its syntax and operational semantics.
- We will encode each  $\mathcal{POR}$  function in a correct  $\mathcal{SIFP}$  program.
- We will implement an interpret for the  $\mathcal{SIFP}$  on a **SFP** machine with a polynomial overhead and modulo the measure of the  $\omega$ s mapping the inputs to the outputs.
- Finally we will reduce the multi-tape **SFP** machines to canonical **SFP** machines, modulo a constant overhead in complexity, and a "stretching" of the  $\omega$  tape, i.e. the measure of the oracles diving an input to produce the same output in both the formalisms.

**1.5.1. The  $\mathcal{SIFP}$  formalism.** The **SFP** paradigm, which is a subclass of the Turing Machines' model, is far being a functional paradigm, while  $\mathcal{POR}$  is fully functional. For this reason, a direct encoding of the  $\mathcal{POR}$  (or even of the  $\mathcal{POR}^-$ ) formalism in **SFP** would be too much complicated because of the radically different natures of the two formalisms.

In order to simplify a little bit the whole encoding, we will pass through an intermediate imperative paradigm, the String's Imperative and Flipping Paradigm  $\mathcal{SIFP}$ .

The  $\mathcal{SIFP}$  paradigm is defined by an enumerable set of correct programs and an operational semantics.

**Definition 35** (Correct programs of  $\mathcal{SIFP}$ ). *The language of the  $\mathcal{SIMP}$  programs is  $\mathcal{L}(\text{Stm})$ , i.e. the set of strings produced by the non-terminal symbol **Stm** defined by:*

$$\begin{aligned} \text{Id} &::= X_i \mid Y_i \mid S_i \mid R \mid Q \mid Z \mid T \quad i \in \mathbb{N} \\ \text{Exp} &::= \epsilon \mid \text{Exp}.0 \mid \text{Exp}.1 \mid \text{Id} \mid \text{Exp} \sqsubseteq \text{Exp} \mid \text{Exp} \wedge \text{Exp} \mid \neg \text{Exp} \\ \text{Stm} &::= \text{Id} \leftarrow \text{Exp} \mid \text{Stm}; \text{Stm} \mid \text{while}(\text{Exp})\{\text{Stm}\} \mid \text{FlipExp} \mid \text{skip}; \end{aligned}$$

**Definition 36** (Store). *A store is a partial function  $\Sigma : \text{Id} \rightarrow \{0, 1\}^*$ .*

**Definition 37** (Empty store). *An empty store is a store that is undefined on all its domain. We represent such object with  $\square$ .*

**Definition 38** (Store updating). *We define the updating of a store  $\Sigma$  with a mapping from  $y \in \text{Id}$  to  $\tau \in \{0, 1\}^*$  as the store  $\Sigma_1$  defined as:*

$$\Sigma_1(x) := \begin{cases} \tau & \text{if } x = y \\ \Sigma(x) & \text{otherwise} \end{cases}$$

**Definition 39** (semantics of  $\mathcal{SIFP}$ 's expressions). *The semantics of an expression  $E \in \mathcal{L}(\text{Exp})$  is the smallest function  $\rightarrow : \mathcal{L}(\text{Exp}) \times (\text{Id} \rightarrow \{0, 1\}^*) \times \mathbb{O} \rightarrow \{0, 1\}^*$  closed under the following rules:*

$$\frac{}{\langle \epsilon, \Sigma, \omega \rangle \rightarrow \epsilon} \quad \frac{\langle e, \Sigma, \omega \rangle \rightarrow \sigma}{\langle e.0, \Sigma, \omega \rangle \rightarrow \sigma \frown 0} \quad \frac{\langle e, \Sigma, \omega \rangle \rightarrow \sigma}{\langle e.1, \Sigma, \omega \rangle \rightarrow \sigma \frown 1}$$



$$\frac{\langle e, \Sigma, \omega \rangle \rightarrow \sigma \quad \langle f, \Sigma, \omega \rangle \rightarrow \tau \quad \sigma \subseteq \tau}{\langle e \sqsubseteq f, \Sigma, \omega \rangle \rightarrow 1} \quad \frac{\langle e, \Sigma, \omega \rangle \rightarrow \sigma \quad \langle f, \Sigma, \omega \rangle \rightarrow \tau \quad \sigma \not\subseteq \tau}{\langle e \sqsubseteq f, \Sigma, \omega \rangle \rightarrow 0}$$

$$\frac{\Sigma(Id) = \sigma}{\langle Id, \Sigma, \omega \rangle \rightarrow \sigma}$$

$$\frac{\langle e, \Sigma, \omega \rangle \rightarrow 0}{\langle \neg e, \Sigma, \omega \rangle \rightarrow 1} \quad \frac{\langle e, \Sigma, \omega \rangle \rightarrow \sigma \quad \sigma \neq 0}{\langle \neg e, \Sigma, \omega \rangle \rightarrow 0}$$

$$\frac{\langle e, \Sigma, \omega \rangle \rightarrow 1 \quad \langle f, \Sigma, \omega \rangle \rightarrow 1}{\langle e \wedge f, \Sigma, \omega \rangle \rightarrow 1} \quad \frac{\langle e, \Sigma, \omega \rangle \rightarrow \sigma \quad \langle f, \Sigma, \omega \rangle \rightarrow \tau \quad \sigma \neq 1 \wedge \tau \neq 1}{\langle e \wedge f, \Sigma, \omega \rangle \rightarrow 0}$$

**Definition 40** (Operational semantics of *SLFP*). *The semantics of a program  $P \in \mathcal{L}(\text{Stm})$  is the smallest function  $\triangleright : \mathcal{L}(\text{Stm}) \times (\text{Id} \rightarrow \{0, 1\}^*) \rightarrow (\text{Id} \rightarrow \{0, 1\}^*)$  closed under the following rules:*

$$\begin{array}{c} \frac{}{\langle \text{skip};, \Sigma \rangle \triangleright \Sigma} \quad \frac{\langle e, \Sigma \rangle \rightarrow \sigma}{\langle Id \leftarrow e, \Sigma \rangle \triangleright \Sigma[Id \leftarrow \sigma]} \quad \frac{\langle s, \Sigma \rangle \triangleright \Sigma' \quad \langle t, \Sigma' \rangle \triangleright \Sigma''}{\langle s; t, \Sigma \rangle \triangleright \Sigma''} \\[10pt] \frac{\langle e, \Sigma \rangle \rightarrow 1 \quad \langle s, \Sigma \rangle \triangleright \Sigma' \quad \langle \text{while}(e)\{s\}, \Sigma' \rangle \triangleright \Sigma''}{\langle \text{while}(e)\{s\}, \Sigma \rangle \triangleright \Sigma''} \quad \frac{\langle e, \Sigma \rangle \rightarrow \sigma \quad \sigma \neq 1}{\langle \text{while}(e)\{s\}, \Sigma \rangle \triangleright \Sigma} \\[10pt] \frac{\langle e, \Sigma, \omega \rangle \rightarrow \sigma \quad \omega(\sigma) = b}{\langle \text{Flip}(e), \Sigma, \omega \rangle \triangleright \Sigma[R \leftarrow \omega(\sigma)]} \end{array}$$

**Notation 5.** We will use the notation  $E \sqsubseteq F$  as a shorthand (syntactic sugar) for  $\neg(\neg E.0 \sqsubseteq F \wedge \neg E.1 \sqsubseteq F)$ .

**Remark 13.** We will use the pattern  $B \leftarrow \epsilon.1; \text{while}(c \wedge B)\{\text{Stm}; B \leftarrow \epsilon.0\}$  as an implementation of the if statement: it's indeed true that

- The statement *Stm* is executed if and only if  $c$  holds.
- The statement *Stm* is executed only one time.

For increasing the readability of our proof, we will introduce the following notation:

**Notation 6** (pseudo-procedure). A pseudo-procedure is a syntactic sugar for the *SLFP*'s language, which consists in a pseudo-procedure's name, a body a list of formal parameters. A call to such expression must be interpreted as the inlining of the pseudo-procedure's body in the place of the call in which the names of the formal parameters by the actual parameters and all the free names of the body are substituted by fresh names.

**Lemma 8** (Term representation in *SLFP*). *All the terms of  $L_{\mathbb{P}\mathbb{W}}$  can be represented in *SLFP*. Formally:  $\forall t \in L_{\mathbb{P}\mathbb{W}}. \exists \mathfrak{M} \in \mathcal{L}(\text{Stm}). \text{Vars}(t) = \{x_1, \dots, x_n\} \rightarrow \mathfrak{M}(x_1, \dots, x_n) = t(x_1, \dots, x_n)$*

*Proof.* We proceed by induction on the syntax of  $t$ . The correctness of such implementation is given by the following invariant properties:

- The result of the computation is stored in  $R$ .
- The inputs are stored in the registers of the group  $X$ .
- The function  $\mathfrak{M}$  doesn't write the values it accesses as input.

$\mathfrak{M}$  is defined as follows:

- $\mathfrak{M}(\epsilon) := R \leftarrow \epsilon;$
- $\mathfrak{M}(0) := R \leftarrow \epsilon.0;$

- $\mathfrak{M}(1) := R \leftarrow \epsilon.1;$
- $\mathfrak{M}(Id) := R \leftarrow Id;$

For sake of readability, we define the following program that copies the  $|Z|$ -th bit of  $S$  at the end of  $R$ , given that  $Z$  contains the  $|Z|$ -th prefix of  $S$ .

$$\begin{aligned}
 copyb(Z, S, R) &:= B \leftarrow \epsilon.1; \\
 &\quad \mathbf{while}(Z.0 \sqsubseteq S \wedge B) \{ \\
 &\quad \quad Z \leftarrow Z.0; \\
 &\quad \quad R \leftarrow R.0; \\
 &\quad \quad B \leftarrow \epsilon.0; \\
 &\quad \} \\
 &\quad \mathbf{while}(Z.1 \sqsubseteq S \wedge B) \{ \\
 &\quad \quad Z \leftarrow Z.1; \\
 &\quad \quad R \leftarrow R.1; \\
 &\quad \quad B \leftarrow \epsilon.0; \\
 &\quad \}
 \end{aligned}$$

**Lemma 9** (Complexity of *copyb*). *The pseudo-procedure copyb requires a number of steps which is a polynomial in the sizes of its arguments.*

*Proof.* The two **while**( ) { }s are used for implementing an **if** construct as described in Remark 13, so they cause no iteration. Moreover, the two statements are mutually exclusive, so this pseudo-procedure requires at most 5 steps.  $\square$

**Lemma 10** (Correctness of *copyb*). *After an execution of copyb:*

- *If the first argument is a strong prefix of the second, the size of the first argument ( $Z$ ) increases by one, and is still a prefix of the second argument ( $S$ ).*
- *Otherwise, the values stored in the first two registers don't change.*
- *Each bit which is stored at the end of  $Z$  is stored at the end of  $R$ .*

*Proof.* Suppose that the value stored in  $Z$  is a strong prefix of the value which is stored in  $S$ . Clearly it's true that  $Z.0 \sqsubseteq S \vee Z.1 \sqsubseteq S$ . In both case *copyb* increases the length of the portion of  $Z$  which is a prefix of  $S$ . If  $Z$  is not a prefix of  $S$  none of the two **ifs** is executed. The last conclusion comes from the observation that each assignment to  $Z$  is followed by a similar assignment to  $R$ .  $\square$

This pseudo-procedure will turn out to be useful in both the encodings of  $\frown$  and  $\times$ . For the  $\frown$  operator, we proceed with the following encoding:

$$\begin{aligned}
\mathfrak{M}(t \frown s) &:= \mathfrak{M}(s) \\
&\quad S \leftarrow R; \\
&\quad \mathfrak{M}(t) \\
&\quad Z \leftarrow \epsilon; \\
&\quad \mathbf{while}(Z \sqsubseteq S) \{ \\
&\quad \quad B \leftarrow \epsilon.1; \\
&\quad \quad \mathit{copyb}(Z, S, R) \\
&\quad \}
\end{aligned}$$

The encoding of the  $\times$  operator is the following:

$$\begin{aligned}
\mathfrak{M}(t \times s) &:= \mathfrak{M}(t) \\
&\quad T \leftarrow R; \\
&\quad \mathfrak{M}(s) \\
&\quad S \leftarrow R; \\
&\quad Z \leftarrow \epsilon; \\
&\quad R \leftarrow \epsilon; \\
&\quad Q \leftarrow \epsilon; \\
&\quad \mathbf{while}(Z \sqsubseteq S) \{ \\
&\quad \quad B \leftarrow \epsilon.1; \\
&\quad \quad \mathbf{while}(Z.0 \sqsubseteq S \wedge B) \{ \\
&\quad \quad \quad Z \leftarrow Z.0; \\
&\quad \quad \quad \mathbf{while}(Q \sqsubseteq T) \{ \\
&\quad \quad \quad \quad \mathit{copyb}(Q, T, R) \\
&\quad \quad \quad \} \\
&\quad \quad \quad Q \leftarrow \epsilon; \\
&\quad \quad \quad B \leftarrow \epsilon.0; \\
&\quad \quad \} \\
&\quad \quad \mathbf{while}(Z.1 \sqsubseteq S \wedge B) \{ \\
&\quad \quad \quad Z \leftarrow Z.1; \\
&\quad \quad \quad \mathbf{while}(Q \sqsubseteq T) \{ \\
&\quad \quad \quad \quad \mathit{copyb}(Q, T, R) \\
&\quad \quad \quad \} \\
&\quad \quad \quad Q \leftarrow \epsilon; \\
&\quad \quad \quad B \leftarrow \epsilon.0; \\
&\quad \quad \} \\
&\quad \}
\end{aligned}$$

□

**Lemma 11** (Complexity of  $\mathfrak{M}$ ).  $\forall t \in L_{\text{PW}}. \mathfrak{M}(t)$  can be computed in number of steps which is polynomial in the size of the variables in  $t$ .

*Proof.* We proceed by induction on the syntax of  $t$ .

- $\epsilon$  If the term is  $\epsilon$ ,  $\mathfrak{M}(t)$  consists in two steps.
- $0, 1$  If the term is a digit,  $\mathfrak{M}(t)$  consists in two steps.
- $x$  If the term is a variable,  $\mathfrak{M}(t)$  consists in two steps.
- $t \frown s$  From the Lemmas 10 and 9 we know that *copyb* requires a constant number of steps, and that each time *copyb* is executed,  $Z$  grows by one. Moreover, we know that the function respects the fact that  $Z$  is a prefix of  $S$ . For this reason, the complexity of the **while**( ) { } statement is linear in the size of  $Z$ . Finally, the complexity takes in account two polynomial due the recursive hypothesis on  $\mathfrak{M}$  and two steps for the two assignments before the **while**( ) { }. The overall sum of these complexities is still a polynomial.
- $t \times s$  We will distinguish the three levels of **while**( ) { }s by calling them *outer*, *middle* and *inner*. For Lemmas 9 and 10, the inner **while**( ) { }s take at most  $|T|$  steps, such value is a polynomial over the free variables of  $t$  according to Lemma ???. The value of  $T$  is kept constant after its first assignment, for this reason all the inner cycles require a polynomial number of steps. The middle cycles, are an implementation of the **if** construct according to Remark 13, so they are executed only once per each outer cycle. Moreover, they add a constant number of steps to the complexity of the inner cycles. This means that, modulo an outer cycle, the complexity is still polynomial. For the same argument of Lemma 9, the outer cycle takes at most  $|S|$  steps which is a polynomial according to ??.

□

**Definition 41** (Function described by a *SLFP* program). We say that the value described by a correct *SLFP* program  $P$  is  $\mathcal{F}(P) : \mathcal{L}(\text{Stm}) \longrightarrow (\mathbb{S}^n \times \mathbb{O} \longrightarrow \mathbb{S})$ , where  $F$  is defined as above<sup>1</sup>:

$$F := \lambda x_1, \dots, x_n, \omega. \triangleright (\langle P, [] [X_1 \leftarrow x_1], \dots, [X_n \leftarrow x_n], \omega \rangle)(R)$$

In order to present in a more compact way the last translation, we need to introduce a pseudo-procedure which truncates a register to the length of another one.

**Definition 42** (Truncating pseudo-procedure). The *trunc*( $T, R$ ) pseudo-procedure is a *SLFP* program with free names  $T$  and  $R$ , defined as follows:

---

<sup>1</sup>Instead of the infix notation for  $\triangleright$ , we will use its prefixed notation in order to express the store associated to the program and the starting store that are between the curly brackets.

```

trunc( $T, R$ ) :=  $Q \leftarrow R$ ;
                $R \leftarrow \epsilon$ ;
                $Z \leftarrow \epsilon$ ;
                $Y \leftarrow \epsilon$ ;
               while( $Z \sqsubseteq T$ ) {
                  $B \leftarrow 1$ ;
                 while( $Z.0 \sqsubseteq T \wedge B$ ) {
                   copyb( $R, Q, Y$ )
                    $Z \leftarrow Z.0$ ;
                    $B \leftarrow 0$ ;
                 }
                 while( $Z.1 \sqsubseteq T \wedge B$ ) {
                    $B \leftarrow 1$ ;
                   copyb( $R, Q, Y$ )
                    $Z \leftarrow Z.1$ ;
                    $B \leftarrow 0$ ;
                 }
               }

```

**Lemma 12** (Complexity of truncation). *The pseudo-procedure trunc requires a number of steps which is at most polynomial in the sizes of its free names.*

*Proof.* By Lemma 9 we know that the pseudo-procedure requires a constant number of steps, furthermore, the inner cycles are the implementation of an **if** according to Remark 13, so they are executed only once per outer cycle. Finally the number of outer cycles is bounded by the  $|T|$ , so the whole complexity of the pseudo-procedure is polynomial (linear) in  $|T|$ .  $\square$

**Lemma 13** (Correctness of truncation). *The pseudo-procedure trunc truncates the register  $R$  to its  $|T|$ -th prefix.*

*Proof.* We proceed by induction on  $T$ .

$\epsilon$  Trivially we have  $R = \epsilon$  since the cycle isn't executed.

$\sigma b$  In this case, only one of the sub-cycles is executed (they are mutually-exclusive), a single more of  $Q$  is stored in  $R$  according to Lemma 10, and  $Q$  is unchanged after the execution of *copyb*. These arguments prove the claim. The register  $Y$  has no practical implications since it's only used in order to leverage the lemmas on *copyb*.  $\square$

**Lemma 14** (Implementation of  $\mathcal{POR}$  in  $\mathcal{SLFP}$ ).  $\forall f \in \mathcal{POR}^- . \exists P \in \mathcal{L}(Stm)$ .  
 $\forall x_1, \dots, x_n. F(P)(x_1, \dots, x_n, \omega) = f(x_1, \dots, x_n, \omega)$

*Proof.* For each function  $f \in \mathcal{POR}^-$  we define a program  $\mathfrak{L}(f)$  such that  $F(\mathfrak{L}(f))(x_1, \dots, x_n) = f(x_1, \dots, x_n)$  by induction on the syntax of  $f$ . The correctness of such implementation is given by the following invariant properties:

- The result of the computation is stored in  $R$ .
- The inputs and the sub-oracle are stored in the registers of the group  $X$ .

- The function  $\mathcal{L}$  doesn't change the values it accesses as input.

We define the function  $\mathcal{L}$  as follows.

- $\mathcal{L}(E) := R \leftarrow \epsilon; \mathbf{skip};$
- $\mathcal{L}(S_0) := R \leftarrow X_0.0; \mathbf{skip};$
- $\mathcal{L}(S_1) := R \leftarrow X_0.1; \mathbf{skip};$
- $\mathcal{L}(P_i^n) := R \leftarrow X_i; \mathbf{skip};$
- $\mathcal{L}(C) := R \leftarrow X_1 \sqsubseteq X_2; \mathbf{skip};$
- $\mathcal{L}(Q) := \text{Flip}(X_1); \mathbf{skip};$

Finally the encoding of the composition and of the bounded iteration:

$$\begin{aligned}
\mathcal{L}(f(h_1(x_1, \dots, x_n, \eta), \dots, h_k(x_1, \dots, x_n, \eta), \eta)) &:= \mathcal{L}(h_1) \\
&S_1 \leftarrow R; \\
&\dots \\
&\mathcal{L}(h_k) \\
&S_k \leftarrow R; \\
&Y_1 \leftarrow X_1; \\
&\dots \\
&Y_{\max(n+1, k+1)} \leftarrow X_{\max(n+1, k+1)}; \\
&X_1 \leftarrow S_1; \\
&\dots \\
&X_k \leftarrow S_k; \\
&X_{k+1} \leftarrow Y_{n+1}; \\
&\mathcal{L}(f) \\
&X_1 \leftarrow Y_1 \\
&\dots \\
&X_{\max(n+1, k+1)} \leftarrow Y_{\max(n+1, k+1)}; \\
&\mathbf{skip};
\end{aligned}$$

Supposing that  $g$  takes  $n$  parameters, the bounded iteration is computed as follows:

```

 $\mathfrak{L}(\text{ite}(g, h_1, h_2, t)) := Z \leftarrow X_{n+1};$ 
 $X_{n+1} \leftarrow \epsilon;$ 
 $\mathfrak{L}(g(x_1, \dots, x_n))$ 
 $Y \leftarrow X_{n+2};$ 
while( $X_{n+1} \sqsubset Z$ ){
   $B \leftarrow \epsilon.1;$ 
  while( $X_{n+1}.0 \sqsubseteq Z \wedge B$ ){
     $X_{n+1} \leftarrow X_{n+1}.0;$ 
     $\mathfrak{M}(t)$ 
     $T \leftarrow R;$ 
     $\mathfrak{L}(h_0);$ 
     $X_{n+1} \leftarrow R;$ 
     $\text{trunc}(T, R);$ 
     $X_{n+2} \leftarrow R;$ 
     $B \leftarrow \epsilon.0;$ 
  } while( $X_{n+1}.1 \sqsubseteq Z \wedge B$ ){
     $X_{n+1} \leftarrow X_{n+1}.1;$ 
     $\mathfrak{M}(t)$ 
     $T \leftarrow R;$ 
     $\mathfrak{L}(h_0);$ 
     $X_{n+1} \leftarrow R;$ 
     $\text{trunc}(T, R);$ 
     $X_{n+2} \leftarrow R;$ 
     $B \leftarrow \epsilon.0;$ 
  }
} $X_{n+2} \leftarrow Y;$ 
skip;

```

□

**Proposition 2** (Complexity of  $SIFP$ ).  $\forall f \in \mathcal{POR}^- . \mathfrak{L}(f)$  takes a number of steps which is polynomial in the size of the arguments of  $f$ .

*Proof.* We proceed by induction on the proof of the fact that  $f$  is indeed in  $\mathcal{POR}^-$ . All the base cases  $(E, S_0, S_1, P_i^n, C, Q)$  are trivial. The inductive steps follow easily:

- In the case of composition, we know that the thesis holds for all the pseudo-procedures  $\mathfrak{L}$ . The program requires a finite number of assignments more, so its complexity is still polynomial.
- In the case of iteration we can move the same argument of Lemma 9 for proving that the outer cycle is executed only  $|Z|$  times, which is a polynomial in an argument of the encoded function. Moreover, the inner cycles are an implementation of the **if** construct according to Remark 13, so they are executed only once per outer cycle. So the

thesis comes from Lemma 12, from Lemma 11, from the fact that the composition of polynomials is still polynomial and from the inductive hypothesis.  $\square$

**Remark 14.** *The number of registers used by  $\mathfrak{L}(f)$  is finite.*

*Proof.* Such value can be expressed by the function  $\#_r^\Sigma$  described below:

$$\begin{aligned}
\#_r^\mathfrak{M}(\epsilon) &:= 1 \\
\#_r^\mathfrak{M}(0) &:= 1 \\
\#_r^\mathfrak{M}(1) &:= 1 \\
\#_r^\mathfrak{M}(x) &:= 2 \\
\#_r^\mathfrak{M}(t \frown s) &:= 4 + \#_r^\mathfrak{M}(t) + \#_r^\mathfrak{M}(s) + 1 \\
\#_r^\mathfrak{M}(t \times s) &:= 7 + \#_r^\mathfrak{M}(t) + \#_r^\mathfrak{M}(s) + 1 \\
\\ 
\#_r^\Sigma(E) &:= 2 \\
\#_r^\Sigma(S_i) &:= 2 \\
\#_r^\Sigma(P_i^n) &:= 2 \\
\#_r^\Sigma(C) &:= 3 \\
\#_r^\Sigma(Q) &:= 2 \\
\#_r^\Sigma(f(h_1(x_1, \dots, x_n, \eta), \dots, h_k(x_1, \dots, x_n, \eta), \eta)) &:= 2k + \max(k + 1, h + 1) + \#_r^\Sigma(f) \\
\#_r^\Sigma(ite) &:= (n + 2) + 4 + \#_r^\mathfrak{M}(t) + 6 + \#_r^\Sigma(g)
\end{aligned}$$

The inductive cases are correct because as:

- $t \frown s$  The value takes account of the inductive calls, the four names used by the function and the register used by *copyb*.
- $t \times s$  The value takes account of the inductive calls, the seven names used by the function and the register used by *copyb*.
  - The value for the concatenation takes account of the  $X_i$  registers, the  $Y_i$ s, the  $S_i$ s and of the recursive calls.
  - The value for the bounded iteration takes account of the  $X_i$  registers, the  $Y$ ,  $Z$ ,  $B$ , and  $R$  register, the registers used by the function  $\mathfrak{M}$ , the six registers used by the *trunc* pseudo-procedure the  $S_i$ s and of the recursive calls.

The correctness of the definition can be obtained by induction on the syntax of the function  $f \in \mathcal{POR}$  that is being translated. The fact that  $\#_r^\Sigma$  is finite is trivial by induction on the definition of such function.  $\square$

Finally we need to state one of our last results, which now should be intuitive. Indeed, we can implement a *SIFP* program by means of a multi-tape **SFP** machine which uses a tape for storing the values of each register, plus two additional tapes for keeping the partial results during the evaluation of the expressions and another tape for simulating the access to the oracle.

The machine works thanks to some invariant properties:

- On each tape the meaningful values are stored on the immediate right of the head.
- The result of the last expression evaluated is stored on the  $e_0$  tape on the immediate right of the head.



- After any assignment operation, the values on the  $e_0$  and  $e_1$  tape are not meaningful anymore.
- The  $o$  tape contains a sequence of string shaped like  $00\mathcal{D}(b)00\sigma 00$ , i.e. a pseudo-encoding of a tuple in  $\mathcal{POR}$  which contains the coordinate of the access to the simulated oracle,  $\sigma$ , and the value.

When simulating a query, the machine can look up for the queried value on the  $o$  tape and, if it's not present, associate a new value to such coordinate according to the character that is read on the tape, and write that value on the tape associated to the  $R$  register. If the queried value is on the tape, the machine returns the value it has associated to that coordinate, namely  $b$ .

The encoding of a  $SLFP$  expression can be easily computed on the  $e_0$  tape: accesses to the identifiers basically consist in copy of tapes, which is a simple operation, due to the invariants properties mentioned above. Moreover, when we defined the  $SLFP$  we have used a restricted set of expressions: concatenations are easily implemented by the addition of a character at the end of the  $e_0$  tape which contains the value of the last expression computed, as stated by the induction hypothesis on the invariant properties. The only non-trivial operations are the binary ones: they require to implement a stack on  $e_0$ . This can be done by adding a blank character after the end of the last expression evaluated, followed by the next value on the stack. Then when the machine needs to combine the two values, the top of the stack is erased and copied on  $e_1$ , compared with the top of  $e_0$  and the result copied in place of  $e_0$ .

**Proposition 3.** *Each  $P \in SLFP$  program which is polynomial and uses  $k$  registers:*

- (1) *Can be simulated on a  $k + 3$ -tape Turing Machine  $M$  which uses an  $\Sigma = \{0, 1\}$  and  $*$  as blank character.*
- (2) *The simulation requires a polynomial complexity.*
- (3)  $\forall x, y. \forall \Sigma. \Sigma$  is defined on  $x \rightarrow \mu(\{\omega \in \mathbb{O} \mid \langle P, \sigma, \omega \rangle \triangleright y\}) = \mu(\{\omega \in \mathbb{O} \mid M(x, \omega) = y\})$ .

*Proof.* We won't give a formal proof of such proposition because it would require an extremely complex and almost uninformative construction of the machine, but we will describe its functioning by cases, showing that the overhead is polynomial by induction.

Our machine stores the values of each register in a specific tape, plus two additional tapes for keeping the values of the expressions,  $e_0$  and  $e_1$ , and an additional tape  $o$  for keeping track of the simulated random accesses. This is why our machine is  $k + 3$ -taped.

### Expressions

- This expression can be computed overwriting each 0 or 1 symbol on the  $e_0$  tape with  $*$ , and finally making the head go back to its starting position. This takes a polynomial number of steps because we know that  $p$  is polynomial in the size of its inputs, so the size of each tape is polynomial in those values, too.
- Id This expression can be computed by deleting the whole  $e_0$  tape, copying each 0 or 1 of the tape associated to the register  $Id$  in the  $e_0$  tape, and finally making the two heads go back to their initial position. This takes a polynomial number of steps because we know that  $p$  is polynomial in the size of its inputs, so the size of each tape is polynomial in those values, too.
- Exp.0 By induction hypothesis we know that the machine can compute Exp in a polynomial number of steps, and store its value in the  $e_0$  tape, then we need to advance the head to the last bit of such tape, add 0, and make the head go back to its initial position. The cost of this operation is polynomial because the size of any register is, and thanks to the inductive hypothesis.
- Exp.1 As above.

□ By induction hypothesis we know that the machine can compute the left and the right **Exp** in a polynomial number of steps. In this case the machine computes the left **Exp**, then it moves the head on the right of the value computed, leaving a blanc character in the middle, it computes the second expression and copies the rightmost value from  $e_0$  to  $e_1$  following the same procedure that we described for *Id*, but erasing it from  $e_0$  once finished. According to the inductive hypothesis it requires a polynomial number of steps, too. Now the machine has both the heads at the beginning of  $e_0$  and  $e_1$ . Now it proceeds left-to-right on  $e_1$  and from the rightmost element of  $e_0$ , until one of the following conditions is met:

- (1) A \* character is met in  $e_0$ .
- (2) The value read in  $e_0$  is different from \* and is different from the value read on  $e_1$ .
- (3) A \* character is met in  $e_1$  but not in  $e_0$ .

For each case it behaves as follows:

- (1) It overwrites the rightmost element of the  $e_0$  tape with \*, and writes 1.
- (2) It overwrites the rightmost element of the  $e_0$  tape with \*, and writes 0.
- (3) It overwrites the rightmost element of the  $e_0$  tape with \*, and writes 0.

The overall complexity is polynomial because the overhead is linear in the size of the two starting expression, which are polynomial for induction hypothesis. After that the machine erases the whole content of  $e_0$ , which requires a polynomial number of steps, because of the induction hypothesis on the second expression.

- ¬ By induction hypothesis, we know that the machine can compute the value of **Exp** in a polynomial number of steps. In this case the machine computes the **Exp**, then the machine checks whether it is a correct Boolean value (checking its size) and then it negates such value. Now the machine steps back the heads in two a single step. In this case, the complexity is trivially polynomial. If the value isn't a the machine overwrites the tape and writes 0. Also in this case, the overall complexity is polynomial.
- ∧ By induction hypothesis we know that the machine can compute the left and the right **Exp** in a polynomial number of steps. In this case the machine computes the left **Exp**, and copies the value from  $e_0$ , then it checks whether the value on  $e_1$  is a Boolean value; if it is not, the machine deletes the rightmost expression on the  $e_0$  tape, it writes 0, and passes to the next step. This requires a polynomial number of steps. Otherwise, it moves to the right of it leaving a blanc character in the middle, it computes the second expression and copies the rightmost value from  $e_0$  to  $e_1$  following the same procedure that we described for *Id*, but erasing it from  $e_0$  once finished. According to the inductive hypothesis it requires a polynomial number of steps, too. Now the machine checks whether the expression on  $e_1$  is a Boolean value or not and it behaves as above if the condition is not met. Finally the machine can compute the logical conjunction between the two values, store it in  $e_0$  and step back the heads in two steps. In this case the complexity is trivially polynomial. After that the machine erases the whole content of  $e_0$ , which requires a polynomial number of steps, because of the induction hypothesis on the second expression.

Finally we can describe how our machines implements the statements of *SIMP*.

#### Statements

- skip**; The machine executes the next instruction, if it exists, otherwise it terminates. Point (3) holds because the two sets are  $\mathbb{O}$ .
- ; The machine executes the statements in their order. This requires a polynomial complexity for inductive hypothesis and because the sum of polynomials is a polynomial. Point (3) holds because of the two induction hypotheses.

- ← The machine computes the values of the expression, then copies in the tape corresponding to the identifier in the same fashion as it copies the value of an identifier in the expression tape. For the same argument that we showed above, these operations require a polynomial number of steps. After that the machine erases both the  $e_0$  and the  $e_1$  tapes. Point (3) holds because both the sets are  $\mathbb{O}$ .

**while** The machine behaves in the following way:

- (1) It computes the values of the expression.
- (2) If such value (the rightmost expression stored in  $e_0$ ) is 0, the machine continues to the next instruction.
- (3) If the expression (as above) is 1, the machine evaluates the statement, it goes back to step (1).

These steps have the following complexities:

- (1) Polynomial, for induction hypothesis.
- (2) Polynomial because  $p$  is polynomial and so the size of the register is polynomial, too.
- (3) Polynomial, for induction hypothesis.

All the previous steps can be executed at most a polynomial number of times because  $P$ 's complexity is itself polynomial, so the execution of the while statement requires a polynomial number of steps. It's possible to see that point (3) holds in this case too by induction on the proof of the reduction. If the condition is false, i.e. reduces to 0, both the sets are  $\mathbb{O}$ , otherwise, the thesis can be obtained by applying the inductive argument for the while's body to the one obtained by the induction on the proof. This leads to our conclusion.

**Flip(Exp)** By induction hypothesis, we know that the machine can compute the value of **Exp** in a polynomial number of steps and store it in  $e_0$ . After that, the machine computes the stores the value obtained encoding  $e_0$  through  $\mathcal{D}$  in the tape  $e_1$ , later it copies this value in  $e_0$  an steps back both the heads on the expression tapes. Then it checks whether the  $o$  tape contains a substring shaped like  $00\mathcal{D}(b)00\sigma 00$ , if it happens the machine erases the  $e_0$  tape and writes  $b$  on the tape corresponding to  $R$ , erasing its previous value, otherwise it appends the string  $00\mathcal{D}(b)00\sigma 00$  at the end of the  $o$  tape, where  $b$  is obtained reading by reading on the oracle tape and copies  $b$  on the tape corresponding to  $R$ , after having erased its previous content. This suffices to show points (1) and (2). For point (3), it's enough to see that in both cases we are interested checking the same number (either 0 or 1) of additional bits from the oracle.

This machine is correct by construction.  $\square$

**Corollary 1.** *Each function in  $\mathcal{POR}$  can be executed on a multi-tape Stream Machine which uses an  $\Sigma = \{0, 1\}$  and  $*$  as blank character with a polynomial complexity. It also holds that  $\forall \vec{x}, y. \mu(\{\omega \in \mathbb{O} \mid f(\vec{x}, \omega) = y\}) = \mu(\{\omega \in \mathbb{O} \mid M(\vec{x}, \omega) = y\})$*

*Proof.* From Lemma 14 we showed that each  $\mathcal{POR}$  function can be implemented in a  $\mathcal{SIFP}$  program which reduces in a polynomial number of steps. In Proposition ?? we showed that this program can be executed on a Stream Machine with a polynomial overhead. The result concerning the measure of the cylinders comes from the fact that the translation of  $\mathcal{POR}$  is  $\mathcal{SIFP}$  preserves the identity of the output of a function given its inputs and an oracle, since the translation of  $\mathcal{SIFP}$  on multitape Stream Machine is such that  $\forall x, y. \forall \Sigma. \Sigma$  is defined on  $x \rightarrow \mu(\{\omega \in \mathbb{O} \mid \langle P, \sigma, \omega \rangle \triangleright y\}) = \mu(\{\omega \in \mathbb{O} \mid M(x, \omega) = y\})$ , we can instantiate this proposition on the input configuration and on the translation of a  $f \in \mathcal{POR}$  in  $\mathcal{SIFP}$ , obtaining that  $\forall x, y. x \rightarrow \mu(\{\omega \in \mathbb{O} \mid f(x, \omega) = y\}) = \mu(\{\omega \in \mathbb{O} \mid M(x, \omega) = y\})$ . We can also deduce that it won't be problematic to generalize the  $x$  to  $\vec{x}$ .  $\square$

**Corollary 2.** *Each polynomial  $\mathcal{POR}$  function can be executed on a single-tape Stream Machine which uses an  $\Sigma = \{0, 1\}$  and  $*$  as blank character with a polynomial overhead, preserving that  $\forall \vec{x}, y. \mu(\{\omega \in \mathbb{O} \mid f(\vec{x}, \omega) = y\}) = \mu(\{\omega \in \mathbb{O} \mid M(\vec{x}, \omega) = y\})$ .*

*Proof.* First observe that it's well known that we can aggregate  $k$  tapes enlarging the alphabet from 2 characters to  $2^k$  characters, which is constant in the size of the input, moreover we can reduce the alphabet of any Turing Machine representing  $2^k$  characters on  $k$  cells of a 2 tape over two characters with a polynomial overhead, because  $k$  (linear in the number of registers) is polynomial in the size of the input. If we apply this reasoning to a **SFP**, which is an ordinary TM plus one tape oracle tape, it's easy to see that we can apply the reduction to the working tapes only. This procedure will shift the position of the characters of  $\omega$  which are used as source of randomness, but won't change the number of checked bits and the identity relations between them, which is enough for preserving the measure property.  $\square$

**Theorem 1.**  $\mathbf{SFP} \subseteq \mathcal{POR} \wedge \forall f \in \mathcal{POR}. \exists M \in \mathbf{SFP}. \forall \vec{x}. \mu(\{\omega \in \mathbb{O} \mid f(\vec{x}, \omega) = y\}) = \mu(\{\omega \in \mathbb{O} \mid M(\vec{x}, \omega) = y\})$

*Proof.* The thesis is a conjunction of Proposition 1 and corollary 2.  $\square$