# A Classification of the Expressive Power of Well-structured Transition Systems 

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#### Abstract

We compare the expressive power of a class of well-structured transition systems that includes relational automata, Petri nets, lossy channel systems, constrained multiset rewriting systems, and data nets. For each one of these models we study the class of languages generated by labelled transition systems describing their semantics. We consider here two types of accepting conditions: coverability and reachability of a fixed a priori configuration. In both cases we obtain a strict hierarchy in which constrained multiset rewriting systems is the the most expressive model.


Keywords. Expressiveness, well-structured systems, language theory.

## 1 Introduction

The theory of well-structured transition systems $[1,13]$ is a powerful tool for studying the decidability of verification problems of infinite-state systems. A system is well-structured when its transition relation is monotonic with respect to a well-quasi ordering defined over configurations. A well-known example of well-structured system is that of Petri nets [21] equipped with marking inclusion $[1,13]$. For a well-structured transition system, the coverability problem can be decided by the symbolic backward reachability algorithm scheme proposed in [1]. Since checking safety properties can be translated into instances of the coverability problem, an algorithm for coverability like that proposed in [1] can be used for automatic verification of an infinite-state system. This connection has been exploited in order to develop automatic verification procedures for several infinite-state models:

- relational automata (RA) [8], an abstract models of imperative programs with integer valued variables;
- Reset and transfer nets $[11,10]$, i.e. Petri nets extended with whole-place operations that atomically operate on the whole set of tokens in a place;
- lossy (FIFO) channel systems (LCS) [4, 7], an abstract models of unreliable communication systems;

[^0]- constrained multiset rewriting systems (CMRS) [2], an extension of Petri nets in which tokens are colored with natural numbers and in which transitions are guarded by conditions on colors;
- affine well-structured nets (aWSNs) [18] a generalization of Petri nets and transfer/resetnets in which the firing of a transition is split into three steps: subtraction, multiplication, and addition of black tokens. Multiplication is a whole-place operation that generalizes transfer and reset arcs;
- data nets (DNs) [19], a generalization of aWSNs in which subtraction, multiplication and addition are defined on tokens that carry data taken from an infinite, ordered domain. Conditions on data values can be used here to restrict the type of tokens on which apply whole-place operations. DNs are a natural extension of CMRS with whole-place operations on colored tokens.

Although several efforts have been spent on studying the expressive power of extensions of Petri nets like reset and transfer nets [11, 12, 14]), a comparison of the relative expressiveness of the class of well-structured transition systems is still missing. Such a comparison is a challenging research problem with a possible practical impact. Indeed, it can be useful to extend the applicability of a verification method (e.g. a particular instance of the scheme of [1]) to an entire class of models.

In this paper we apply tools of language theory to formally compare the expressive power of a large class of well-structured infinite-state systems that includes extensions of Petri nets, constrained multiset rewriting systems, lossy channel systems, relational automata, and data nets. To achieve the goal, for each one of these models we study the class of languages generated by labeled transition systems describing their semantics. We consider here two types of accepting conditions: coverability (with respect to a fixed ordering) and reachability of a given configuration. Two models are considered to be equivalent if they generate the same class of languages.

For coverability accepting conditions, we obtain the following classification.

- We show that, differently from nets with indistinguishable tokens, wholeplace operations do not augment the expressive power of models in which tokens carry data taken from an ordered domain. The proof is based on a weak, effectively constructible encoding of data nets into CMRS that can be used to reduce the coverability problem from one model to the other. As a corollary, we have that the symbolic backward reachability algorithm for solving the coverability problem in CMRS described in [2] can also be applied to data nets.
By slightly extending the CMRS encoding, we also prove that the coverability problem remains decidable for different extensions of data nets. In particular we consider data net transitions that select data that must be fresh (in [19] a transition selects values that may be fresh).
- We prove that lossy channel systems are equivalent to a syntactic fragment of constrained multiset rewriting, we named $\Gamma_{0}$. The fragment $\Gamma_{0}$ is obtained by restricting conditions of a rule in such a way that equalities cannot be
used as guards. Furthermore, we prove that lossy channel systems are strictly less expressive than the full model of constrained multiset rewriting systems. We then show that Petri nets are equivalent to a syntactic fragment of constrained multiset rewriting systems, we named $\Gamma_{1}$, obtained by considering nullary predicates only.
- We prove that aWSNs are strictly more expressive than Petri nets and strictly less expressive than LCSs, thus separating Petri nets from LCSs with respect to their relative expressive power. Furthermore, we prove that $a W S N s$ are as expressive as transfer/reset nets. This result show that the inclusion between the coverability languages of transfer/reset nets and LCS proved in [3] is strict.
- We prove that relational automata are equivalent to a syntactic fragment of constrained multiset rewriting, we named $\Gamma_{2}$, obtained by imposing an upper bound on the size (number of predicates) of reachable configurations.
- Finally, we prove that $\Gamma_{2}$ generates the class of regular languages. This implies that relational automata are strictly less expressive than Petri nets.

For reachability accepting conditions, we obtain a slightly different classification. First, we prove that $\Gamma_{0}$ is equivalent to constrained multiset rewriting systems and two counter machines. Thus, with reachability acceptance, $\Gamma_{0}$ and constrained multiset rewriting systems turn out to be strictly more expressive than lossy channel systems. On the contrary, $\Gamma_{1}$ is still equivalent to Petri nets and strictly less expressive than $\Gamma_{0}$ and $\Gamma_{2}$ is still equivalent to relational automata and to finite automata. Finally, we show that lossy channel systems and Petri nets define incomparable classes of languages.

Concerning related work, the relative expressiveness of well-structured systems has been investigated for a limited number of extensions of Petri nets with reset, transfer, and non-blocking arcs in [12, 14]. Classical results on finite and infinite languages generated by Petri nets can be found, e.g., in [15]. A classification of infinite-state systems in terms of structural properties and decidable verification problems is presented in [16]. The classification is extended to wellstructured systems in [6]. A classification of the complexity of the decision procedures for coverability is studied in [19]. In contrast with the aforementioned work, we provide here a strict classification of the expressive power of several well-structured transition systems built with the help of tools of language theory. An extended version of the present paper is available as technical report [3].

Outline In Section 2, we give some preliminary notions on well-structured transition systems. In Section 3, we give some first results on the class of languages accepted by constrained multiset rewriting systems. In Section 6, 7, and 8, we compare the class of languages recognized by constrained multiset rewriting systems and, respectively, lossy channel systems, Petri nets, and relational automata. Finally, in Section 9 we discuss some final remarks.

## 2 Wsts and Languages with Coverability Acceptance

In this section we recall some definitions taken from [1]. A transition system is a tuple $T=(S, R)$ where $S$ is a (possibly infinite) set of configurations, $R$ is a finite set of transitions where each $\xrightarrow{\sigma} \in R$ is a binary relation over $S$, i.e. $\xrightarrow{\sigma} \subseteq S \times S$. We use $\gamma \xrightarrow{\sigma} \gamma^{\prime}$ to denote $\left(\gamma, \gamma^{\prime}\right) \in \xrightarrow{\sigma}$, and $\gamma \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma^{\prime}$ to denote that there exist $\gamma_{1}, \ldots, \gamma_{k-1}$ such that $\gamma \xrightarrow{\rho_{1}} \gamma_{1} \ldots \xrightarrow{\rho_{k-1}} \gamma_{k-1} \xrightarrow{\rho_{k}} \gamma^{\prime}$. A quasiordering $(S, \preceq)$ is a well-quasi ordering if for any infinite sequence $s_{1} s_{2} \ldots s_{i} \ldots$ there exist indexes $i<j$ such that $s_{i} \preceq s_{j}$. A transition system $T=(S, R)$ is well-structured with respect to a quasi-order $\preceq$ on $S$ iff: $(i) \preceq$ is a well-quasi ordering; (ii) for any $\xrightarrow{\sigma} \in R$ and $\gamma_{1}, \gamma_{1}^{\prime}, \gamma_{2}$ s.t. $\gamma_{1} \preceq \gamma_{1}^{\prime}$ and $\gamma_{1} \xrightarrow{\sigma} \gamma_{2}$, there exists $\gamma_{2}^{\prime}$ s.t. $\gamma_{1}^{\prime} \xrightarrow{\sigma} \gamma_{2}^{\prime}$ and $\gamma_{2} \preceq \gamma_{2}^{\prime}$, i.e., $T$ is monotonic. We use $T=(S, R, \preceq)$ to indicate a well-structured transition system (wsts for short).

To formalize the comparison between models, a wsts $T=(S, R, \preceq)$ can be viewed as a language acceptor. For this purpose, we assume a finite alphabet $\Sigma$ and a labelling function $\lambda: R \mapsto \Sigma$ that associates to each transition of $R$ a symbol of $\Sigma \cup\{\epsilon\}$, where $\epsilon$ denotes the empty sequence $(w \cdot \epsilon=\epsilon \cdot w=w$ for any $\left.w \in \Sigma^{*}\right)$. In the following, we use $\gamma_{1} \xrightarrow{w} \gamma_{2}$ with $w \in \Sigma^{*}$ to denote that $\gamma_{1} \xrightarrow{\rho_{1} \cdots \rho_{k}} \gamma_{2}$ and $\lambda\left(\xrightarrow{\rho_{1}}\right) \cdots \lambda\left(\xrightarrow{\rho_{k}}\right)=w$. Furthermore, we associate to $T$ an initial configuration $\gamma_{\text {init }} \in S$ and a final configuration $\gamma_{a c c} \in S$ and assume an accepting relation $\bowtie: S \times S$. For a fixed accepting relation $\bowtie$, we define the language accepted (generated) by $T=\left(S, R, \preceq, \gamma_{i n i t}, \gamma_{a c c}\right)$ as:

$$
L(T)=\left\{w \in \Sigma^{*} \mid \gamma_{i n i t} \xrightarrow{w} \gamma \text { and } \gamma_{a c c} \bowtie \gamma\right\}
$$

In this paper we consider two types of accepting relations:

- Coverability: the accepting relation $\bowtie$ is defined as $\gamma_{a c c} \preceq \gamma$.
- Reachability: the accepting relation $\bowtie$ is defined as $\gamma_{a c c}=\gamma$.

Let $\mathcal{M}$ be a wsts model (e.g. Petri nets) and let $T$ be one of its instances (i.e. a particular net). We define $L_{c}(T)$, resp $L_{r}(T)$, as the language accepted by $T$ with accepting relation $\bowtie_{c}$, resp. $\bowtie_{r}$. We say that $L$ is a $c$-language, resp. $r$-language, of $\mathcal{M}$ if there is an instance $T$ of $\mathcal{M}$ such that $L=L_{c}(T)$, resp. $L=L_{r}(T)$. We use $L_{c}(\mathcal{M})$, resp. $L_{r}(\mathcal{M})$, to denote the class of $c$-languages, resp. $r$-languages, of $\mathcal{M}$. Finally, we use $\mathcal{L}_{1} \nsim \mathcal{L}_{2}$ to denote that $\mathcal{L}_{1}$ and $\mathcal{L}_{2}$ are incomparable classes of languages.
Given a wsts $T=(S, R, \preceq)$ with labels in $\Sigma \cup\{\epsilon\}$, a lossy version of $T$ is a wsts $T^{\prime}=\left(S, R^{\prime}, \preceq\right)$ for which there exists a bijection $h: R \mapsto R^{\prime}$ such that $\xrightarrow{\rho} \in R$ and $\xrightarrow{h(\rho)}$ have the same label, $\xrightarrow{\rho} \subseteq \xrightarrow{h(\rho)}$ and if $\gamma \xrightarrow{h(\rho)} \gamma^{\prime}$, then $\gamma \xrightarrow{\rho} \gamma^{\prime \prime}$ with $\gamma^{\prime} \preceq \gamma^{\prime \prime}$. In other words, in a lossy version of a wsts the set of reachable configurations contains configurations that are smaller than those of the original model. The next lemma states an important property used in the remainder of the paper.
Lemma 1. For any lossy version $T^{\prime}$ of a wsts $T$, we have that $L_{c}(T)=L_{c}\left(T^{\prime}\right)$.

## 3 Constrained Multiset Rewriting Systems (CMRS)

In this section we recall the main definitions and prove the first results for constrained multiset rewriting systems [2]. Let us first give some preliminary definitions. We use $\mathbb{N}$ to denote the set of natural numbers and $\bar{n}$ to denote the interval $[0, \ldots, n]$ for any $n \in \mathbb{N}$. We assume a set $\mathbb{V}$ of variables which range over $\mathbb{N}$, and a set $\mathbb{P}$ of unary predicate symbols. For a set $A$, we use $A^{*}$ and $A^{\otimes}$ to denote the sets of (finite) words and (finite) multisets over $A$ respectively. Sometimes, we write multisets as lists, so $[1,5,5,1,1]$ represents a multiset with three occurrences of 1 and two occurrences of 5 ; [ ] represents the empty multiset. We use the usual relations and operations such as $\leq$ (inclusion), + (union), and - (difference) on multisets. For a set $V \subseteq \mathbb{V}$, a valuation $V a l$ of $V$ is a mapping from $V$ to $\mathbb{N}$. A condition is a finite conjunction of gap order formulas of the forms: $x<_{c} y, x \leq y, x=y, x<c, x>c, x=c$, where $x, y \in \mathbb{V}$ and $c \in \mathbb{N}$. Here $x<_{c} y$ stands for $x+c<y$. We often use $x<y$ instead of $x<_{0} y$. Sometimes, we treat a condition $\psi$ as a set, and write e.g. $\left(x<_{c} y\right) \in \psi$ to indicate that $x<_{c} y$ is one of the conjuncts in $\psi$. We use true to indicate an empty set of conditions. A term is of the form $p(x)$ where $p \in \mathbb{P}$ and $x \in \mathbb{V}$. A ground term is of the form $p(c)$ where $p \in \mathbb{P}$ and $c \in \mathbb{N}$. We sometimes say that a predicate symbol is nullary to mean that its parameter is not relevant (hence may be omitted).

A constrained multiset rewriting system (CMRS) $\mathcal{S}$ consists of a finite set of rules each of the form $L \leadsto R: \psi$, where $L$ and $R$ are multisets of terms, and $\psi$ is a condition. We assume that $\psi$ is consistent (otherwise, the rule is never enabled). For a valuation $\operatorname{Val}$, we use $\operatorname{Val}(\psi)$ to denote the result of substituting each variable $x$ in $\psi$ by $\operatorname{Val}(x)$. We use $\operatorname{Val} \models \psi$ to denote that $\operatorname{Val}(\psi)$ evaluates to true. For a multiset $T$ of terms we define $\operatorname{Val}(T)$ as the multiset of ground terms obtained from $T$ by replacing each variable $x$ by $\operatorname{Val}(x)$. A configuration is a multiset of ground terms. Each rule $\rho=L \leadsto R: \psi \in \mathcal{S}$ defines a relation between configurations. More precisely, $\gamma \xrightarrow{\rho} \gamma^{\prime}$ if and only if there is a valuation Val s.t. the following conditions are satisfied: (i) Val $\models \psi,(i i) \gamma \geq \operatorname{Val}(L)$, and (iii) $\gamma^{\prime}=\gamma-\operatorname{Val}(L)+\operatorname{Val}(R)$.

Example 1. Consider the CMRS rule:

$$
\rho=[p(x), q(y)] \leadsto[q(z), r(x), r(w)]:\{x+2<y, x+4<z, z<w\}
$$

A valuation which satisfies the condition is $\operatorname{Val}(x)=1, \operatorname{Val}(y)=4, \operatorname{Val}(z)=8$, and $\operatorname{Val}(w)=10$.

A CMRS configuration is a multisets of ground terms, e.g., $[p(1), p(3), q(4)]$. Therefore, we have that $[p(1), p(3), q(4)] \xrightarrow{\rho}[p(3), q(8), r(1), r(10)]$.

Let us fix a CMRS $\mathcal{S}$ operating on a set of predicate symbols $\mathbb{P}$. Let cmax be the maximal constant which appears in the rules of $\mathcal{S} ; c \max$ is equal to 0 if there are no constant in $\mathcal{S}$. We now define an ordering $\preceq_{c}$ on configurations extracted from the ordering defined in [2] to solve the coverability problem.

Definition 1. Given a configuration $\gamma$, we define the index of $\gamma$, index $(\gamma)$, to be a word of the form $D_{0} \cdots D_{c \max } d_{0} B_{0} d_{1} B_{1} d_{2} \cdots d_{n} B_{n}$ where
$-D_{0}, \ldots, D_{c \max }, B_{0}, \ldots, B_{n} \in \mathbb{P}^{\otimes}$ and $d_{0}, d_{1}, \ldots, d_{n} \in \mathbb{N} \backslash\{0\} ;$

- $B_{i}$ must not be empty for $0 \leq i \leq n$;
- for each $p \in \mathbb{P}, D_{i}$ contains $k$ occurrences of predicate $p$ iff $p(i)$ occurs $k$ times in $\gamma$ for $0 \leq i \leq$ cmax;
- given $v_{0}=$ cmax $+d_{0}$, for each $p \in \mathbb{P}, B_{0}$ contains $k$ occurrences of predicate $p$ iff $p\left(v_{0}\right)$ occurs $k$ times in $\gamma$;
- given $v_{i+1}=v_{i}+d_{i+1}$, for each $p \in \mathbb{P}, B_{i+1}$ contains $k$ occurrences of predicate $p$ iff $p\left(v_{i+1}\right)$ occurs $k$ times in $\gamma$ for all $0 \leq i<n$;
- for all $p(v) \in \gamma$ with $v>$ cmax, there exists $i: 0 \leq i \leq n$ such that $v=c \max +d_{0}+d_{1}+\ldots+d_{i}$.

The ordering $\preceq_{c}$ is defined as follows.
Definition 2. Let $D_{0} D_{1} \cdots D_{\text {cmax }} d_{0} B_{0} d_{1} B_{1} d_{2} \cdots d_{n} B_{n}$ be the index of a configuration $\gamma_{1}$ and $D_{0}^{\prime} D_{1}^{\prime} \cdots D_{c \max }^{\prime} d_{0}^{\prime} B_{0}^{\prime} d_{1}^{\prime} B_{1}^{\prime} d_{2}^{\prime} \cdots d_{m}^{\prime} B_{m}^{\prime}$ be the index of a configuration $\gamma_{2}$. Then, $\gamma_{1} \preceq_{c} \gamma_{2}$ iff $D_{i} \leq D_{i}^{\prime}$ for $0 \leq i \leq c m a x$ and there exists a monotone injection $h: \bar{n} \mapsto \bar{m}$ such that $B_{0} \leq B_{h(0)}^{\prime}, B_{i} \leq B_{h(i)}^{\prime}$, $d_{0} \leq \sum_{k=0}^{h(0)} d_{k}^{\prime}$, and $d_{i} \leq \sum_{k=h(i-1)+1}^{h(i)} d_{k}^{\prime}$ for $1 \leq i \leq n$.
In the rest of the paper we assume that the values appearing in the initial configuration $\gamma_{\text {init }}$ and in the accepting configuration $\gamma_{a c c}$ are smaller or equal than cmax. The ordering $\preceq_{c}$ is obtained by composing string embedding and multiset inclusion. From standard properties of orderings, it follows that $\preceq_{c}$ is a well-quasi ordering. Furthermore, a CMRS is monotonic with respect to corresponding ordering $\preceq_{c}$. The following property then holds.

Proposition 1 ([2]). A CMRS $\mathcal{S}$ equipped with $\preceq_{c}$ is well-structured.

### 3.1 A Symbolic Algorithm for Testing Coverability

In this section we give an overview of the algorithm for solving the coverability problem based on the generic backward analysis algorithm presented in [1]. The difficult challenge in applying this methodology is to invent a symbolic representation (called constraints) which allows effective implementation of each step, and which guarantees termination of the algorithm.

The algorithm operates on constraints, where each constraint $\phi$ characterizes an infinite set $\llbracket \phi \rrbracket$ of configurations. A constraint $\phi$ is of the form $T: \psi$ where $T$ is a multiset of terms and $\psi$ is a condition. The constraint characterizes the (upward closed) set $\llbracket \phi \rrbracket=\{\gamma \mid \exists \operatorname{Val} .(\operatorname{Val} \models \psi) \wedge(\operatorname{Val}(T) \leq \gamma)\}$ of configurations. Notice that if $\psi$ is inconsistent, then $\llbracket \phi \rrbracket$ is empty. Such a constraint can be safely discarded in the reachability algorithm presented below. Therefore, we assume in the sequel that all conditions in constraints are consistent. We define $\operatorname{Var}(\phi)=\operatorname{Var}(T) \cup \operatorname{Var}(\psi)$. Observe that the coverability problem can be reduced to constraint reachability. More precisely, $\gamma_{i n i t} \xrightarrow{*} p_{f i n}$ is equivalent to $\gamma_{\text {init }} \xrightarrow{*} \phi_{\text {fin }}$ where $\phi_{\text {fin }}$ is the constraint $\left[p_{f i n}(x)\right]:$ true.

For constraints $\phi_{1}, \phi_{2}$, we use $\phi_{1} \sqsubseteq \phi_{2}$ to denote that $\phi_{1}$ is entailed by $\phi_{2}$, i.e., $\llbracket \phi_{1} \rrbracket \supseteq \llbracket \phi_{2} \rrbracket$. For a constraint $\phi$, we define $\operatorname{Pre}(\phi)$ to be a finite set
of constraints which characterize the configurations from which we can reach a configuration in $\phi$ through the application of a single rule. In other words $\bigcup_{\phi_{1} \in \operatorname{Pre}(\phi)} \llbracket \phi_{1} \rrbracket=\left\{\gamma \mid \exists \gamma^{\prime} \in \llbracket \phi \rrbracket . \gamma \longrightarrow \gamma^{\prime}\right\}$.
For instance, given $\phi_{1}=\left[p\left(x_{1}\right), q\left(x_{2}\right), q\left(x_{3}\right)\right]:\left\{x_{1}<_{2} x_{2}, x_{2}<_{1} x_{3}\right\}$, and the configurations $\gamma_{1}=[p(2), q(8), q(5), p(1)]$ and $\gamma_{2}=[p(2), q(2), q(5), p(1)]$. Then $\gamma_{1} \in \llbracket \phi_{1} \rrbracket$ and $\gamma_{2} \notin \llbracket \phi_{1} \rrbracket$. Consider now $\phi_{2}=\left[p\left(y_{1}\right), q\left(y_{2}\right)\right]:\left\{y_{1}<y_{2}\right\}$ and $\phi_{3}=\left[p\left(y_{1}\right), q\left(y_{2}\right)\right]:\left\{y_{1}<_{4} y_{2}\right\}$. Then $\phi_{2} \sqsubseteq \phi_{1}$ and $\phi_{3} \nsubseteq \phi_{1}$.
Given an instance of the coverability problem, defined by $\gamma_{\text {init }}$ and the constraint $\phi_{\text {fin }}$ corresponding to $p_{\text {fin }}$, the symbolic algorithm performs a fixpoint iteration starting from $\phi_{f i n}$ and repeatedly applying Pre on the generated constraints. The iteration stops if either (i) we generate a constraint $\phi$ with $\gamma_{\text {init }} \in \llbracket \phi \rrbracket$; or (ii) we reach a point where, for each newly generated constraint $\phi$, there is a constraint $\phi^{\prime}$ generated in a previous iteration with $\phi^{\prime} \sqsubseteq \phi$. We give a positive answer to the coverability problem in the first case, while we give a negative answer in the second case.

In [2] we show computability of membership, entailment, and define an effective predecessor operator for constraints. To give an idea of these definition, let $\mathcal{S}$ be a CMRS and $\phi_{2}$ be a constraint. We define $\operatorname{Pre}\left(\phi_{2}\right)=\bigcup_{\rho \in \mathcal{S}} \operatorname{Pr} e_{\rho}\left(\phi_{2}\right)$, where $\operatorname{Pre}_{\rho}\left(\phi_{2}\right)$ describes the effect of running the rule $\rho$ backwards from the configurations in $\phi_{2}$. Let $\rho=(L \leadsto R: \psi)$ and $\phi_{2}=\left(T_{2}: \psi_{2}\right)$. Let $W$ be any set of variables such that $|W|=\left|\operatorname{Var}\left(\phi_{2}\right) \cup \operatorname{Var}(\rho)\right|$. We define $\operatorname{Pre}_{\rho}\left(\phi_{2}\right)$ to be the set of constraints of the form $T_{1}: \psi_{1}$, such that there are renamings Ren, Ren ${ }_{2}$ of $\operatorname{Var}(\rho)$ and $\operatorname{Var}\left(\phi_{2}\right)$ respectively to $W$, and

$$
\text { - } T_{1}=\operatorname{Ren}_{2}\left(T_{2}\right) \ominus \operatorname{Ren}(R)+\operatorname{Ren}(L) \quad \bullet \psi_{1}=\operatorname{Ren}(\psi) \wedge \operatorname{Ren}_{2}\left(\psi_{2}\right)
$$

Example 2. For instance, consider the constraint $\phi=\left[q\left(x_{1}\right), s\left(x_{2}\right), r\left(x_{2}\right)\right]$ : $\left\{x_{1}<x_{2}\right\}$ and the rule $\rho=\left[p\left(y_{1}\right), p\left(y_{3}\right)\right] \leadsto\left[q\left(y_{2}\right), r\left(y_{3}\right)\right]:\left\{y_{3}<y_{2}\right\}$. Fix $W=\left\{w_{1}, w_{2}, w_{3}, w_{4}, w_{5}\right\}$, and define $\operatorname{Ren}_{2}=\left(x_{1} \mapsto w_{1}, x_{2} \mapsto w_{2}\right)$, and $\operatorname{Ren}=\left(y_{1} \mapsto w_{3}, y_{2} \mapsto w_{1}, y_{3} \mapsto w_{4}\right)$.
Then one member of $\operatorname{Pre}_{\rho}$ is given by
$\left[s\left(w_{2}\right), r\left(w_{2}\right), p\left(w_{3}\right), p\left(w_{4}\right)\right]:\left\{w_{1}<w_{2}, w_{4}<w_{1}\right\}$, which after normalization becomes $\left[s\left(w_{2}\right), r\left(w_{2}\right), p\left(w_{3}\right), p\left(w_{4}\right)\right]:\left\{w_{4}<_{1} w_{2}\right\}$.
The termination of the algorithm is obtained by a non-trivial application of a methodology based on the theory of well- and better-quasi orderings described in [2].

### 3.2 Three Interesting Fragments of CMRS: $\Gamma_{0}, \Gamma_{1}$ and $\Gamma_{2}$

In this section we defined three fragments of CMRS that we use as a technical tool for comparisons with other wsts.

The fragment $\Gamma_{0}$ In the fragment $\Gamma_{0}$ of CMRS every rule $L \leadsto R: \psi$ satisfies the following conditions: every variable $x$ occurs at most once in $L$ and at most once in $R$, and $\psi$ does not contain equality constraints. As an example, $[p(x), r(y)] \sim[q(x), r(z)]: x<y, y<z$ is a rule in $\Gamma_{0}$, whereas $[p(x), q(x)] \leadsto$ $[q(y)]:$ true and $[p(x)] \leadsto[q(y), r(y)]:$ true are not in $\Gamma_{0}$.

The fragment $\Gamma_{1}$ The fragment $\Gamma_{1}$ is obtained by restricting CMRS to nullary predicates only (i.e. predicates with no parameters).
The fragment $\Gamma_{2}$ The fragment $\Gamma_{2}$ is defined as follows. Let us first use $|B|$ to denote the cardinality of a multiset $B . \Gamma_{2}$ is the fragment of CMRS in which a rule $L \leadsto R: \psi$ satisfies the condition $|R| \leq|L|$. In other words, in $\Gamma_{2}$ the cardinality of a reachable configuration is always bounded by the cardinality of the initial configuration.

In the rest of the paper we show that these three fragments have the same expressive power resp. of Lossy FIFO Channel Systems, Petri nets, and Integral Relational Automata. To prove this statement, it is useful to isolate properties of CMRS and of these fragments with respect to coverability acceptance.

### 3.3 Properties of CMRS

In this section we prove some properties of CMRS needed in the rest of the paper.

We first introduce some new terminology. We say that a configuration $\gamma$ with index $(\gamma)=D_{0} \ldots D_{\text {cmax }} B_{0} d_{0} \ldots d_{n} B_{n}$ is linear if $B_{i}$ is a singleton multiset for $i: 0 \leq i \leq n$. We also say that an execution $\gamma_{0} \xrightarrow{\rho_{1}} \ldots \xrightarrow{\rho_{k}} \gamma_{k}$ is linear whenever $\gamma_{i}$ is linear for any $i: 0 \leq i \leq k$.
Furthermore, we say that $\gamma$ is $c m a x$-bounded if $\operatorname{index}(\gamma)=D_{0} \ldots D_{c \max }$, i.e. all the natural numbers in $\gamma$ are between 0 and cmax.

An important property of CMRS is related to the possibility of lifting an execution from an initial cmax-bounded configuration $\gamma_{i n i t}$ to a configuration $\gamma$ to a new execution leading from $\gamma_{\text {init }}$ to a configuration with larger "gaps" (for values greater than cmax) than those in $\gamma$. In the rest of the section we often refer to the index of a configuration $\gamma$, a notion defined in the first section of the appendix (definition of the wqo on CMRS configurations).
We first define a restriction $\prec$ of the relation $\preceq_{c}$ in which we require that the distribution of predicates in two configurations has the same structure but larger gaps.
Formally, $\gamma_{1} \prec \gamma_{2}$ holds iff the following conditions are satisfied:
$-\operatorname{index}\left(\gamma_{1}\right)=D_{0} \ldots D_{c \max } d_{0} B_{0} d_{1} \ldots d_{n} B_{n}$

- index $\left(\gamma_{2}\right)=D_{0} \ldots D_{c \max } d_{0}^{\prime} B_{0} d_{1}^{\prime} \ldots d_{n}^{\prime} B_{n}$
- $d_{i}^{\prime} \geq d_{i}$ for any $i: 0 \leq i \leq n$.

We say that an execution $\gamma_{0} \xrightarrow{\rho_{1}} \ldots \gamma_{1} \xrightarrow{\rho_{k}} \gamma_{k}$ subsumes an execution $\gamma_{0}^{\prime} \xrightarrow{\rho_{1}^{\prime}}$ $\gamma_{1}^{\prime} \ldots \xrightarrow{\rho_{k}^{\prime}} \gamma_{k}^{\prime}$ if for all $i: 0 \leq i \leq k, \gamma_{i}^{\prime} \prec \gamma_{i}$ and for all $i: 1 \leq i \leq k, \rho_{i}=\rho_{i}^{\prime}$.

The following property then holds. The proof is in appendix A.
Proposition 2. Let a CMRS with initial cmax-bounded configuration $\gamma_{\text {init }}$. For any execution $e=\gamma_{\text {init }} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma$, for any configuration $\gamma^{\prime}$ such that $\gamma \prec \gamma^{\prime}$, there exists an execution $\gamma_{\text {init }} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma^{\prime}$ that subsumes $e$.

Now, we introduce the notion of linearization of a configuration a technical definition we use to prove other interesting properties of the class of languages accepted by CMRS.

Given a configuration $\gamma$ with

$$
\operatorname{index}(\gamma)=D_{0} \ldots D_{c \max } d_{0} B_{0} \ldots d_{i} B_{i}+[p] d_{i+1} B_{i+1} \ldots d_{n} B_{n}
$$

where $B_{i}$ is not empty, we say that $\gamma^{\prime}$ is a linearization of $\gamma$ if

$$
\operatorname{index}\left(\gamma^{\prime}\right)=D_{0} \ldots D_{c \max } d_{0}^{\prime} B_{0} \ldots d_{i}^{\prime} B_{i} d[p] d_{i+1}^{\prime} B_{i+1} \ldots d_{n}^{\prime} B_{n}
$$

such that
$-\forall 0 \leq j \leq n: d_{j}^{\prime} \geq d_{j} ;$
$-d \geq 1$.
The following lemmas then hold. The proof is in appendix A.
Lemma 2. Let $\mathcal{S}$ be a $\Gamma_{0}$ model with initial cmax-bounded configuration $\gamma_{\text {init }}$. Suppose there exists a linear execution $\gamma_{\text {init }} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma_{k}, \gamma_{k} \xrightarrow{\rho} \gamma$ and $\gamma$ is not linear. Then, there exists a (possibly different) linear execution $\gamma_{\text {init }} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma_{k}^{\prime}$ such that $\gamma_{k}^{\prime} \xrightarrow{\rho} \gamma^{\prime}$ and $\gamma^{\prime}$ is a linearisation of $\gamma_{k}$.

Lemma 3. For a $\Gamma_{0} \mathcal{S}$, let $\gamma_{1}$ and $\gamma_{2}$ be two configurations such that $\gamma_{2}$ is a linearisation of $\gamma_{1}, \gamma_{1} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma_{3}$ implies there exists $\gamma_{4}$ such that $\gamma_{2} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma_{4}$.

Given a CMRS $\mathcal{S}$ of $\Gamma_{0}$ with initial cmax-bounded linear configuration $\gamma_{\text {init }}$ and accepting cmax-bounded linear configuration $\gamma_{a c c}$, we define

$$
L_{c}^{\text {lin }}(\mathcal{S})=\left\{w \mid \gamma_{i n i t} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma_{k} \text { is linear, } \gamma_{a c c} \preceq_{c} \gamma_{k}, \lambda\left(\rho_{1}\right) \cdots \lambda\left(\rho_{k}\right)=w\right\}
$$

From Lemma 2 and 3, we obtain the following proposition.
Proposition 3. For all $C M R S \mathcal{S}$ of $\Gamma_{0}$ with an initial cmax-bounded linear configuration $\gamma_{\text {init }}$ and cmax-bounded linear accepting configuration $\gamma_{a c c}$, we have $L_{c}(\mathcal{S})=L_{c}^{\text {lin }}(\mathcal{S})$.

## Proof.

〇: Immediate.
$\subseteq$ : To simplify the presentation, let us assume that $\forall L \leadsto R: \psi \in \mathcal{S}$ : for each variable $x$ that appears in $L+R$ : either $(x=c) \in \psi(0 \leq c \leq c \max )$ or $(x>c$ max $) \in \psi$. This assumption implies that the effect of a rule $\rho$ is constant if we only consider ground terms $p(x)$ with $x: 0 \leq x \leq c \max$.
Suppose that $\gamma_{\text {init }} \xrightarrow{\rho_{1}} \gamma_{1} \xrightarrow{\rho_{2}} \ldots \xrightarrow{\rho_{k}} \gamma_{k}$ with $\gamma_{a c c} \preceq_{c} \gamma_{k}$.
Now suppose that $\gamma_{1}$ is not linear. Applying Lemma 2, we have a linearization $\gamma_{1}^{\prime}$ of $\gamma_{1}$ such that $\gamma_{\text {init }} \xrightarrow{\rho_{1}} \gamma_{1}^{\prime}$. Furthermore, following Lemma 3, we have $\gamma_{1}^{\prime} \xrightarrow{\rho_{2} \ldots \rho_{k}} \gamma_{k}^{\prime}$. Iterating the reasoning, we obtain a linear configuration $\gamma_{1}^{\prime \prime}$ such that $\gamma_{\text {init }} \xrightarrow{\rho_{1}} \gamma_{1}^{\prime \prime}$ and $\gamma_{1}^{\prime \prime} \xrightarrow{\rho_{2} \ldots \rho_{k}} \gamma_{k}^{\prime \prime}$.

Repeating the reasoning for the other intermediate configurations, we conclude that there exists a linear execution $\gamma_{i n i t} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma_{k}^{\prime \prime \prime}$.

From our hypothesis, $\rho_{1} \ldots \rho_{k}$ has constant effect if we only consider ground terms $p(x)$ with $x: 0 \leq x \leq c m a x$. Hence, we have that

$$
\sum_{p(n) \in \gamma_{k}, 0 \leq n \leq c \max }[p(n)]=\sum_{p(n) \in \gamma_{k}^{\prime \prime \prime}, 0 \leq n \leq c \max }[p(n)]
$$

and $\gamma_{a c c} \preceq_{c} \gamma_{k}$ implies that $\gamma_{a c c} \preceq_{c} \gamma_{k}^{\prime \prime \prime}$ since $\gamma_{a c c}$ is cmax-bounded. We conclude that $\lambda\left(\rho_{1}\right) \cdots \lambda\left(\rho_{k}\right) \in L_{c}(\mathcal{S})$ implies $\lambda\left(\rho_{1}\right) \cdots \lambda\left(\rho_{k}\right) \in L_{c}^{\text {lin }}(\mathcal{S})$.

## 4 Expressive Power of CMRS

We are ready now to give a first characterization for the expressive power of CMRS. In [14, Prop. 4], the authors show that there exists a recursively enumerable (RE) language that cannot be recognized by any wsts with coverability acceptance. Hence, the following proposition holds.

Theorem 1. $L_{c}(C M R S) \subset R E$.
With reachability as accepting condition, CMRS recognize instead the class of recursively enumerable languages (RE).

Theorem 2. $L_{r}(C M R S)=R E$.
Proof. We prove that CMRS can weakly simulate 2-counter machines. In the proof we also show that repeated reachability is undecidable for CMRS. We recall the model of a 2-counter machine (CM) which is pair $(Q, \delta)$, where $Q$ is a finite set of states, and $\delta$ is the transition function. A transition is of the form ( $q_{1}, o p, q_{2}$ ), where $q_{1}, q_{2} \in Q$, and op is either an increment (of the form $c n t_{1}++$ or $c n t_{2}++$ ); a decrement (of the form $c n t_{1}--$ or $c n t_{2}--$ ); or a zero-testing (of the form $c n t_{1}=0$ ? or $c n t_{2}=0$ ?). Operations and tests on counters have their usual semantics, assuming that the values of counters are natural values. In particular, decrement on a counter equal to zero is blocking. A 2-counter machine accepts an execution if it ends into the state $q_{\text {fin }}$. A lossy 2-counter machine ( $L C M$ ) is of the same form as a counter machine. The difference in semantics is in the zero-testing operation. More precisely, the zero-testing of $c n t_{1}$ is simulated by resetting the value $c n t_{1}$ to zero, and decreasing the value of $c n t_{2}$ by an arbitrary natural number (possibly 0 ). The zero-testing of $c n t_{2}$ is performed in a similar manner.

Assume an LCM $\mathcal{M}=(Q, \delta)$. We shall construct a CMRS $\mathcal{S}$ which will simulate simulate $\mathcal{M}$. The simulation of $\mathcal{M}$ occurs in a sequence of phases. During each phase, $\mathcal{S}$ simulates increment and decrement transitions of $\mathcal{M}$. Each phase is indexed by a natural number which is incremented at the end of the phase. As soon as $\mathcal{M}$ performs a zero-testing of a counter, $\mathcal{S}$ enters an intermediate stage. After conclusion of the intermediate stage, a new phase is started and the index phase is increased.

The set of predicates symbols in $\mathcal{S}$ is divided into three groups:

- Two nullary predicate symbols $q$ and $q^{\prime}$ for each $q \in Q$. We use $q^{\prime}$ during the intermediate stages of the simulation;
- Two predicate symbols $c n t_{1}$ and $c n t_{2}$, which encode the values of $c n t_{1}$ and $\mathrm{cnt}_{2}$ respectively;
- A predicate phase whose argument carries the index of the current phase. Furthermore, we use a predicate symbol phase ${ }^{\prime}$ to store the index of the previous phase during the intermediate stages of the simulation.

A configuration of $\mathcal{S}$ contains, during a given phase of the simulation, the following ground terms:

- A term of the form $q$ which encodes the current state of $\mathcal{M}$.
- A term of the form phase (c) where $c$ is the index of the current phase.
- Terms of the form $c n t_{1}(c)$ where $c$ is the index of the current phase. The number of such terms encodes the current value of $\mathrm{cnt}_{1}$. There are also a number of terms of the form $c n t_{1}(d)$ where $d$ is stricly lesser than the index of the current phase. Such terms are redundant and do not affect the encoding. Similar terms exist to encode $c n t_{2}$.

An increment transition $\left(q, c n t_{1}++, q_{2}\right) \in \delta$ labeled with $a$ is simulated by a rule labeled with $a$ of the form

$$
\left[q_{1}, \operatorname{phase}(x)\right] \leadsto\left[q_{2}, \operatorname{phase}(x), \operatorname{cnt}_{1}(x)\right]: \text { true }
$$

We increase the value counter $c n t_{1}$ by adding one more term whose predicate symbol is $c n t_{1}$ and whose argument is equal to the index of the current phase.

A decrement transition $\left(q, c n t_{1}--, q_{2}\right) \in \delta$ labeled with $a$ is simulated by a rule labeled with $a$ of the form

$$
\left[q_{1}, \operatorname{phase}(x), \operatorname{cnt}_{1}(x)\right] \leadsto\left[q_{2}, \operatorname{phase}(x)\right]: \text { true }
$$

We decrease the value counter $\mathrm{cnt}_{1}$ by removing one of the corresponding terms from the configuration. Observe that terms whose arguments are less than index of the current phase are not used, and hence they do affect the encoding.

A transition $\left(q, c n t_{1}=0 ?, q_{2}\right) \in \delta$ labeled by $a$ is simulated by the following three rules (the two first are labeled with $\epsilon$ and the last one with $a$ ):

$$
\begin{aligned}
& {\left[q_{1}, \operatorname{phase}(x), \operatorname{phase}^{\prime}(x)\right] \underset{\sim}{\prime}\left[q_{1}^{\prime}, \operatorname{phase}(y), \operatorname{phase}^{\prime}(x)\right]: x<y} \\
& {\left[q_{1}^{\prime}, \operatorname{cnt}_{2}(x), \operatorname{phase}(y), \operatorname{phase}^{\prime}(x)\right]} \\
& \sim\left[q_{1}^{\prime}, \operatorname{cnt}_{2}(y), \operatorname{phase}(y), \operatorname{phase}^{\prime}(x)\right]: \text { true } \\
& {\left[q_{1}^{\prime}, \operatorname{phase}(y), \operatorname{phase}^{\prime}(x)\right] \sim\left[q_{2}, \operatorname{phase}(y), \operatorname{phase}^{\prime}(y)\right]: \text { true }}
\end{aligned}
$$

We enter the intermediate phase by changing from $q_{1}$ to $q_{1}^{\prime}$. We store the current index using phase ${ }^{\prime}$, and generate a new index which is strictly larger than the current one. This resets counter $\mathrm{cnt}_{1}$ since all terms in its encoding will now have too small arguments. Finally, we change the arguments of (some of) the terms encoding $\mathrm{cnt}_{2}$ to the new index. Here, not all such terms may receive new arguments, and hence the value $c n t_{2}$ may "unintentionally" be reduced. We use
redundant terms to refer to terms which have either $c n t_{1}$ or $c n t_{2}$ as predicate symbol, and whose arguments are smaller than the current index,

Mayr shows in [20] undecidability of the repeated state reachability problem for LCM: given a lossy counter machine, and two states $q_{\text {init }}$ and $q_{f i n}$, check whether there is a computation starting from $q_{\text {init }}$ (with both counter values being equal to zero) and visits $q_{f i n}$ infinitely often.

We can extend the proof to show Theorem 2 as follows. The key observation here is that redundant terms are not removed during the simulation procedure described above. This means that any reachable configuration which does not contain redundant terms are generated by computations of the perfect (nonlossy) counter machines. We add the nullary predicate $q_{\text {fin }}^{\prime \prime}$ and the following rules labeled by $\epsilon$ to our CMRS:

$$
\begin{aligned}
{\left[q_{f i n}, \text { cnt }_{1}(x)\right] } & \leadsto\left[q_{f i n}^{\prime \prime}\right] & : \text { true } \\
{\left[q_{\text {fin }}^{\prime \prime}, \operatorname{phase}(x), \operatorname{cnt}_{1}(x)\right] } & \leadsto\left[q_{\text {fin }}^{\prime \prime}, \operatorname{phase}(x)\right] & : \text { true } \\
{\left[q_{\text {fin }}^{\prime \prime}, \operatorname{phase}(x), \operatorname{cnt}_{2}(x)\right] } & \leadsto\left[q_{\text {fin }}^{\prime \prime}, \operatorname{phase}(x)\right] & : \text { true } \\
{\left[q_{\text {fin }}^{\prime \prime}, \operatorname{phase}(x)\right] } & \leadsto\left[q_{\text {fin }}^{\prime \prime}\right] & : \text { true }
\end{aligned}
$$

In other words, if we reach a configuration where $\mathcal{M}$ is in $q_{\text {in }}$, then $\mathcal{S}$ can initiate a procedure where it moves to $q_{f i n}^{\prime \prime}$ and then erases the ground terms corresponding to the counters. However, we only remove such ground terms if there arguments are equal to the index of the current phase, i.e. redundant ground terms are not removed. This implies that there exists an execution where $\mathcal{S}$ recognizes a word $w$ that reaches $\left[q_{f i n}^{\prime \prime}\right]$ iff there exists an execution where the perfect MC recognizes the word $w$ that reaches $q_{\text {fin }}$.

## 5 Data nets

Data nets [19] are an extension of Petri nets in which tokens are colored with data taken from an infinite domain $D$ equipped with a linear and dense ordering $\prec$. Due to lack of space, we present here only the key concepts needed in the rest of the paper (see [19] for formal definitions). A data net consists of a finite set of places $P$ and of a finite set of transitions. A data net marking $s$ is a multiset of tokens that carry data in $D$. Formally, a marking $s$ is a finite sequence of vectors in $\mathbb{N}^{P} \backslash\{\mathbf{0}\}$, where $\mathbf{0}$ is the vector that contains only 0 's. Each index $i$ in the sequence $s$ corresponds to some $d_{i} \in D$ such that $i \leq j$ if and only if $d_{i} \prec d_{j}$. For each $p \in P, s(i)(p)$ is the number of tokens with data $d_{i}$ in place $p$.

First of all, a data net transition $t$ has an associated arity $\alpha_{t}$ (a natural number greater than zero). The arity $\alpha_{t}=k$ is used to non-deterministically select $k$ distinct data $d_{1} \prec \ldots \prec d_{k}$ from the current configuration $s$. Some of the selected data may not occur in $s$ (they are fresh). This choice induces a finite and ordered partitioning of the data in $s$, namely $R\left(\alpha_{t}\right)=\left(R_{0}, S_{1}, R_{1}, \ldots, S_{k}, R_{k}\right)$, where $R_{0}$ contains all data $d: d \prec d_{1}$ in $s, S_{i}=\left\{d_{i}\right\}$ for $i: 1, \ldots, k, R_{i}$ contains all $d: d_{i} \prec d \prec d_{i+1}$ in $s$ for $i: 1, \ldots, k-1$, and $R_{k}$ contains all $d: d_{k} \prec d$ in $s$. Clearly, $R\left(\alpha_{t}\right)$ also induces a natural partitioning of the multiset of tokens in $s$ based on the attached data.

We define $\bar{k}=\{1, \ldots, k\}$ and $[k]_{0}=\{0, \ldots, k\}$. Transition $t$ operates on the regions in the partitioning $R\left(\alpha_{t}\right)$ in three steps defined resp. by three matrices $F_{t}, H_{t} \in \mathbb{N}^{R\left(\alpha_{t}\right) \times P}$, and $G_{t} \in \mathbb{N}^{R\left(\alpha_{t}\right) \times P \times R\left(\alpha_{t}\right) \times P}$.
(1) Subtraction: $F_{t}$ specifies the number of tokens with data $d_{1}, \ldots, d_{k}$ that has to be removed from $s$. By definition, $F_{t}\left(R_{i}, p\right)=0$ for $i \in[k]_{0}$ and $p \in P$. The transition $t$ is enabled if the subtraction is possible on each place $P$. This step yields an intermediate configuration $s_{1}$ defined as follows:

For each $i \in \bar{k}$ and $p \in P, s_{1}\left(d_{i}\right)(p)=s\left(d_{i}\right)(p)-F_{t}\left(S_{i}, p\right)$
For each $j \in[k]_{0}, d \in R_{j}, p \in P, s_{1}(d)(p)=s(d)(p)$.
(2) Multiplication: $G_{t}$ specifies how many tokens are transferred from one place to another with possible multiplication of their occurrences and modification of their data $\left(G_{t}\left(\pi, p, \pi^{\prime}, p^{\prime}\right) \geq 0\right.$ and, by definition, $G_{t}\left(R_{i}, p, R_{j}, q\right)=0$ for any $i \neq j \in[k]_{0}$ and any $\left.p, q \in P\right)$. This step yields an intermediate configuration $s_{2}$ defined as follows:
For each $i \in \bar{k}$ and $p \in P$ :
$s_{2}\left(d_{i}\right)(p)=\sum_{j \in \bar{k}} \sum_{q \in P} s_{1}\left(d_{j}\right)(q) \cdot G_{t}\left(S_{j}, q, S_{i}, p\right)+\sum_{j \in[k]_{0}} \sum_{d \in R_{j}} \sum_{q \in P} s_{1}(d)(q) \cdot G_{t}\left(R_{j}, q, S_{i}, p\right)$
For each $i \in[k]_{0}, d \in R_{i}$, and $p \in P$ :

$$
s_{2}(d)(p)=\sum_{j \in \bar{k}, q \in P} s_{1}\left(d_{j}\right)(q) \cdot G_{t}\left(S_{j}, q, R_{i}, p\right)+\sum_{q \in P} s_{1}(d)(q) \cdot G_{t}\left(R_{i}, q, R_{i}, p\right)
$$

Notice that transfers of tokens from region $R_{i}$ to region $R_{j}$ with $i \neq j$ are forbidden.
(3) Addition: Finally, $H_{t}$ specifies the number of tokens that are added to each place in $P$. Its application yields the successor configuration $s^{\prime}$ such that:

For each $i \in \bar{k}$ and $p \in P, s^{\prime}\left(d_{i}\right)(p)=s_{2}\left(d_{i}\right)(p)+H_{t}\left(S_{i}, p\right)$
For each $j \in[k]_{0}, d \in R_{j}, p \in P, s^{\prime}(d)(p)=s(d)(p)+H_{t}\left(R_{j}, p\right)$.
As proved in [19], data nets are well-structured with respect to the well-quasi ordering $\preceq_{d}$ defined on markings as follows. Let $\operatorname{Data}(s)$ be the set of data values that occur in a marking $s$. Then, $s_{1} \preceq_{d} s_{2}$ iff there exists an injective function $h: \operatorname{Data}\left(s_{1}\right) \mapsto \operatorname{Data}\left(s_{2}\right)$ such that $(i) h$ is monotonic and $(i i) s_{1}(d)(p) \leq$ $s_{2}(h(d))(p)$ for each $d \in \operatorname{Data}\left(s_{1}\right)$ and $p \in P$.
Example 3. Consider a data net with $P=\{p, q\}$ and the transition in Fig. 1.
For a generic configuration $s$, the new configuration $s^{\prime}$ is such that:
$-s^{\prime}\left(d_{1}\right)(p)=s\left(d_{1}\right)(p)-1+\Sigma_{d \in R_{0}} 3 * s(d)(p)$ and $s\left(d_{1}\right)(q)^{\prime}=1 ;$

- For each $d \prec d_{1}, s^{\prime}(d)(p)=s(d)(p)+3 * s(d)(q)+2 * s\left(d_{1}\right)(q)$ and $s(d)(q)^{\prime}=$ $s(d)(q)$;
- For each $d \gg d_{1}, s(d)(p)^{\prime}=s(d)(p)$ and $s(d)(q)^{\prime}=s(d)(q)$.

For instance, let $e_{1} \prec e_{2} \prec e_{3} \prec e_{4} \in D$ and assume that the transition selects $e_{3}$ as index in $S_{1}$, then:

$$
\left(\right) \rightarrow\left(\right)
$$

Fig. 1. A data net transition with arity $\alpha_{t}=1$.

### 5.1 Datanets vs CMRS

In [19] the authors mention that it is possible to define an encoding of CMRS in the fragment of data net without whole place operations (Petri data nets) that preserves coverability. From this observation, it follows that $L_{c}(C M R S) \subseteq$ $L_{c}$ (Data nets).

In this section we tighten this relation and show that for each Data nets $\mathcal{D}$ we can effectively build a CMRS $\mathcal{S}$ such that $L_{c}(\mathcal{S})=L_{c}(\mathcal{D})$. In the following, given a multi-set $M$ with symbols in $P$ and a value or variable $x$, we use $M^{x}$ to denote the multi set of $P$-terms such that $M^{x}(p(x))=M(p)$ (=number of occurrences of $p$ in $M$ ) for each $p \in P$, and $M^{x}(p(y))=0$ for any $y \neq x$ and $p \in P$.

Configurations Assume an initial data net marking $s_{0}$ with data $d_{1} \prec \ldots \prec$ $d_{n}$. We build a CMRS representation of $s_{0}$ by non-deterministically selecting $n$ natural numbers $v_{1}<\ldots<v_{n}$ strictly included in some interval $[f, l]$. $P$-terms with parameter $v_{i}$ represent tokens with data $d_{i}$ in place $p$. Formally, we generate the representation of $s_{0}$ by adding to $\mathcal{S}$ a rule that rewrites an initial zerary term init as follows: ${ }^{4}$

$$
[\text { init }] \leadsto[\text { first }(f), \operatorname{last}(l)]+\sum_{i: 1, \ldots, n} M_{i}^{x_{i}}: f<x_{1}<\ldots<x_{n}<l \quad \text { (init) }
$$

where $M_{i}$ is the multiset $s_{0}\left(d_{i}\right)$ for each $i \in \bar{n}$. The non-determinism in the choice of $f, l, x_{1}, \ldots, x_{n}$ make the CMRS representation of $s_{0}$ independent from specific parameters assumed by terms.

Transitions are encoded by CMRS rules that operate on the values in $[f, l]$ used in the representation of a marking. Most of the CMRS rule are based on left-to-right traversals of $P$-terms with parameters in $[f, l]$.

Subtraction Consider a transition $t$ with $\alpha_{t}=k$. We first define a (silent) CMRSrule that implements the subtraction step of $t$ :

$$
\begin{aligned}
& {[\text { first }(f), \operatorname{last}(l)]+F_{t}\left(S_{1}\right)^{x_{1}}+\ldots+F_{t}\left(S_{k}\right)^{x_{k}} \leadsto} \\
& \quad\left[\imath_{0}(f), \imath_{1}\left(x_{1}\right), \ldots, \imath_{k}\left(x_{k}\right), \imath_{k+1}(l), \text { new }_{t}\right]: f<x_{1}<\ldots<x_{k}<l
\end{aligned}
$$

${ }^{4}$ We recall that $\left[t_{1}, \ldots, t_{n}\right]$ denotes a multisets of terms. Furthermore, $\sum_{i: 1, \ldots, k} M_{i}=$ $M_{1}+\ldots+M_{k}$, where + is multiset union.

In the subtract rule we non-deterministically associate a value $w_{i}$ to region $S_{i}$. The selection is performed by removing (from the current configuration) the multiset $F_{t}\left(S_{i}\right)^{x_{i}}$ that contains $F_{t}\left(S_{i}, p\right)$ occurrences of $p\left(x_{i}\right)$ for each $p \in P$. The association between value $x_{i}$ and region $S_{i}$ is maintained by storing $x_{i}$ in a $\imath_{i}$-term (introduced in the right-hand side of the rule). If $F_{t}\left(S_{i}, p\right)=0$ for any $p \in P$, then $v_{i}$ may be associated to a data $d_{i}$ not occurring in the current marking (i.e. selection of fresh data is a special case). Furthermore, by removing both the first- and the last-term, we disable the firing of rules that encode other data net transitions. Fig. 9 in appendix shows an example of application of the subtract rule.
In the rest of the section we refer to The values $x_{1}, \ldots, x_{k}$ stored in $\imath_{1}-, \ldots, \imath_{k^{-}}$ terms play the role of pointers to the regions $S_{1}, \ldots, S_{k}$ We refer to them as to the set of $\alpha_{t}$-indexes. The parameters of terms in $[f, l]$ associated to the other regions $R_{0}, \ldots, R_{k}$ are called region-indexes.

Multiplication To simulate the multiplication step we proceed as follows. We first make a copy of the multiset of $P$-terms with parameters $v_{1}, \ldots, v_{n}$ in $[f, l]$ by copying each $p$-term with parameter $v_{i}$ in a $\bar{p}$-term with parameter $w_{i}$ such that $f^{\prime}<w_{1}<\ldots<w_{n}<l^{\prime}$ and $\left[f^{\prime}, l^{\prime}\right]$ is an interval to the right of $[f, l]$, i.e., $l<f^{\prime}$. The new $w_{t}$-term in the subtract rule is used to enable the set of (silent) CMRS rules in Fig. 2 in appendix that create the copy-configuration. During the copy we add a $\checkmark$-term for any visited region index. These terms are used to remember region indexes whose corresponding $\bar{P}$-terms are all removed in the multiplication step (e.g. when all tokens with data $d \in R_{i}$ are removed).

For instance, $\left[p\left(v_{1}\right), p\left(v_{2}\right), p\left(v_{2}\right), q\left(v_{3}\right)\right]$ with $f<v_{1}<v_{2}<v_{3}<l$ is copied as $\left[\bar{p}\left(w_{1}\right), \checkmark\left(w_{1}\right), \bar{p}\left(w_{2}\right), \bar{p}\left(w_{2}\right), \checkmark\left(w_{2}\right), \bar{q}\left(w_{3}\right) \checkmark\left(w_{3}\right)\right]$ for some $w_{1}, w_{2}$, $w_{3}$ such that $f<l<f^{\prime}<w_{1}<w_{2}<w_{3}<l^{\prime}$. The CMRS rules of Fig. 2 use a special term as a pointer scan the indexes in $[f, l]$ from left to right and create new $\bar{P}$-term with parameters in the interval $\left[f^{\prime}, l^{\prime}\right]$. The pointer is non-deterministically moved to the right. Thus during the traversal we may forget to copy some token. This is the first type of loss we find in our encoding. Notice that lost tokens have parameters strictly smaller that $f^{\prime}$.

The simulation of the multiplication step operates on the copy-configuration only (that with $\bar{P}$-terms). The (silent) CMRS rules that implement this step are shown in Fig. 3 in appendix. The intuition behind their definition is as follows.

We first consider all $\alpha_{t^{\prime}}$-indexes of $\bar{P}$-terms from left to right. For each $\alpha_{t^{-}}$ index $v_{i}$, we proceed as follows. We first select and remove a term $\bar{p}\left(v_{i}\right)$ (encoding a given token). We compute then the effect of the whole-place operation on the entire set of $\alpha_{t}$-indexes (including $v_{i}$ itself). More specifically, for an $\alpha_{t}$-index $v_{j}$ we add $G_{t}\left(S_{i}, p, S_{j}, q\right)$ occurrences of the term $q\left(v_{j}\right)$ to the current CMRS configuration. The use of $P$ - and $\bar{P}$-terms with parameters in the same interval allows us to keep track of tokens still to transfer ( $\bar{P}$-terms) and tokens already transferred ( $P$-terms). We then consider all remaining indexes by means of a left-to-right traversal of region-indexes in the current configuration. During the traversal, we add new $P$-terms with region-indexes as parameters as specified by $G_{t}$. During this step, we may forget to transfer some $\bar{P}$-term. This is the second
type of loss we find in the encoding. After this step we either consider the next token with $\alpha_{t}$-index $v_{i}$ or we move to the next $\alpha_{t}$-index. Fig. 10(a) in appendix illustrates the simulation of this kind of transfers (i.e. from $S_{i}$ to $S_{j} / R_{j}$ ).

After the termination of the whole-place operations for terms with $\alpha_{t}$-indexes, we have to simulate the transfer of $\bar{P}$-terms with region-indexes. For each such an index, we transfer tokens within the same region-index or to an $\alpha_{t}$-index. To simulate these operations we scan region-indexes from left-to-right to apply the matrix $G_{t}$. Furthermore, we mark visited region-indexes using $\checkmark$-terms. The $\checkmark$-terms are used in the simulation of the addition step. The (silent) CMRS rules that implement this step (enabled by the by term $\operatorname{tr} R_{t}$ ) are shown in Fig. 4 in appendix. Fig. 10(b) in appendix illustrates the simulation of this type of whole-place operation.

As a last step we add tokens to $\alpha_{t}$-indexes and visited region indexes as specified by $H_{t}$. For $\alpha_{t}$-indexes, we need a single rule that applies the matrix $H_{t}$. For region-indexes, we traverse from left-to-right the current configuration and apply $H_{t}$ to each marked (with a $\checkmark$-term) region-index $w$. As mentioned before, the $\checkmark$-term allows us to apply $H_{t}$ to regions emptied by the multiplication step. The rules for this step (associated to terms $a d d_{t}$ and $a d d R_{t}$ ) are shown in Fig. 4. All the rules are silent except the last one whose label is the same as that of $t$. Fig. 10 in appendix shows an example of their application.

During the traversal, we may ignore some (marked) region-index. This is the last type of loss in our encoding. The new configuration is the final result of the simulation of the transition. Due to the possible losses in the different simulation steps, we may get a representation of a data net configuration smaller than the real successor configuration.

To formalize the relation between a data net $\mathcal{D}$ and its CMRS encoding $\mathcal{E}(\mathcal{D})$, for a configuration $s$ with data $d_{1} \prec \ldots \prec d_{k}$ we use $s^{\boldsymbol{v}}$ to denote the CMRS representation with indexes $\boldsymbol{v}=\left(v_{1}, \ldots, v_{k}\right)$.

Proposition 4. For configurations $s_{0}, s_{1}, s, s^{\prime}$, the following properties hold:
i) If $s_{0} \xrightarrow{w} s_{1}$ in $\mathcal{D}$, then there exists $\boldsymbol{v}$ such that $[$ init $] \stackrel{w}{\Longrightarrow} s_{1} \boldsymbol{v}$ in $\mathcal{E}(\mathcal{D})$;
ii) Furthermore, if $[$ init $] \stackrel{w}{\Longrightarrow} c$ in $\mathcal{E}(\mathcal{D})$ and $s^{\boldsymbol{v}} \preceq_{c} c$ for some $\boldsymbol{v}$, then there exists $s_{1}$ such that $s_{0} \xrightarrow{w} s_{1}$ in $\mathcal{D}$ with $s \preceq_{d} s_{1}$.

Finally, suppose that the accepting data net marking is a sequence $M_{1} \ldots M_{k}$ of $k$ vectors (multi-sets) over $\mathbb{N}^{P}$. Then, we add a silent CMRS rule

$$
[\operatorname{first}(f), \operatorname{last}(l)]+\sum_{i \in\{1, \ldots, k\}} M_{i}^{x_{i}} \leadsto[a c c]: f<x_{1}<x_{2}<\ldots<x_{k}<l, x=0
$$

where $a c c$ is a fresh (with arity zero) predicate. By adding this rule, the accepting CMRS configuration can be defined as the singleton [acc]. From Lemma 1 and Prop. 4, we have the following result.

Theorem 3. $L_{c}($ data nets $)=L_{c}(C M R S)$

### 5.2 Extensions of data nets

An encoding into CMRS can also be applied to extend the decidability result for coverability to different extensions of data nets. As a first extension, we consider freshness of data values. Let us consider a data net transition $t$ with $\alpha_{t}=k$. In the semantics of data nets, some of the $k$ data values selected by $t$ may be fresh, i.e., they don't have to occur in the current configuration. This definition can be extended by introducing the constraint that some of the selected data must be fresh for the transition to be fired. In the Petri net setting a similar operator has been considered in the $\nu$-nets of [17] to create new, unused identifier.

For simplicity, we consider here the extension of data net transitions in which we require that only one of the $\alpha_{t}$ data value must be fresh. This kind of transition can be modelled by extending the CMRS encoding of the subtraction step as follows. Before selecting the $\alpha_{t}$ data, we make a copy (in a new interval) of the current configuration. In the new configuration we non-deterministically mark using predicate new a value $x$ distinct from the values used to represent tokens. After this preliminary step, we apply the subtraction phase by requiring that the value $x$ is one of the selected ones (i.e. we need $\alpha_{t}$ rules for this last step). Formally, we use the rules in Fig. 6 in appendix. This extension provides a direct way to model freshness without need of ordering identifiers and of maintaining in a special place the last used one (the natural way of modelling $\nu$-nets in ordinary data nets).

Another possible extension concerns the relaxation of some of the restrictions in the definition of data nets in [19]. For instance, assume we allow transfers between regions $R_{i}$ and $R_{j}$ with $i \neq j$. The semantics of a transfer with $G_{t}\left(R_{i}, p, R_{j}, p^{\prime}\right)=m>0$ is the following. For each $d \in R_{i}$, place $p \in P$, and each token with $d$ in $p$, we add $m$ tokens with data $d^{\prime}$ to $p^{\prime}$ for each $d^{\prime} \in R_{j}$. Furthermore, we can also consider a new type of whole-place operation within the same region $R_{i}$ in which we can multiply the tokens with data $d$ for each data with value $d^{\prime} \in R_{i}$ with $d \neq d^{\prime}$. More formally, assume we add a new matrix $M_{t}\left(R_{i}, p, R_{i}, p^{\prime}\right)$ to specify, for each token in $p$ with data $d \in R_{i}$, how many tokens to add to place $p^{\prime}$ with data $d^{\prime}$ for each $d^{\prime} \in R_{i}, d^{\prime} \neq d$. These extensions of data net transitions are still monotonicity w.r.t. $\preceq_{d}$. Furthermore, they can be weakly simulated in CMRS as shown in Fig. 7 in appendix. Thus, we have that coverability remains decidable and, from Lemma 1, we have that $L_{c}($ extended data nets $)=L_{c}(C M R S)$.

## 6 Lossy FIFO Channel Systems

In this section we study the relationship between the fragment $\Gamma_{0}$ of CMRS defined in Section 3.2 and lossy (FIFO) channel systems (LCS) [4].

A Lossy FIFO Channel System (LCS) consists of an asynchronous parallel composition of finite-state machines that communicate through sending and receiving messages via a finite set of unbounded lossy FIFO channels (in the sense that they can non-deterministically lose messages). Formally, an LCS $\mathcal{F}$

For $k=\alpha_{t}, i \in\{0, \ldots, k\}$, and any $p \in P$ :
Copy of indexes in $\alpha_{t}$

$$
\begin{aligned}
& {\left[\imath_{0}\left(x_{0}\right), \ldots, \imath_{k+1}\left(x_{k+1}\right), n e w_{t}\right] \xrightarrow{\leadsto}} \\
& \quad\left[\imath_{0}\left(x_{0}\right), \ldots, \imath_{k+1}\left(x_{k+1}\right), \jmath_{0}\left(x_{0}^{\prime}\right), \ldots, \jmath_{k+1}\left(x_{k+1}^{\prime}\right), \uparrow\left(x_{0}\right), \uparrow\left(x_{0}^{\prime}\right)\right]: \\
& \quad x_{k+1}<x_{0}^{\prime}<\ldots<x_{k+1}^{\prime}
\end{aligned}
$$

Copy $p$ to $\bar{p}$ for $\alpha_{t}-$ indexes
$\left[\uparrow(x), \imath_{i}(x), \Uparrow(y), \jmath_{i}(y), p(x)\right] \leadsto\left[\uparrow(x), \imath_{i}(x), \Uparrow(y), \jmath_{i}(y), \bar{p}(y)\right]:$ true
Copy $p$ to $\bar{p}$ for region - indexes
$\left[u_{i}(x), \uparrow(u), p(u), l_{i+1}\left(x^{\prime}\right), \jmath_{i}(y), \Uparrow(v), \jmath_{i+1}\left(y^{\prime}\right)\right] \leadsto$ $\left[\imath_{i}(x), \uparrow(u), \imath_{i+1}\left(x^{\prime}\right), \jmath_{i}(y), \bar{p}(v), \Uparrow(v), \jmath_{i+1}\left(y^{\prime}\right)\right]: x<u<x^{\prime}, y<v<y^{\prime}$
Move pointers to the right
$\left[\uparrow(u), p\left(u^{\prime}\right), \imath_{k+1}(x), \Uparrow(v), \jmath_{k+1}(y)\right] \leadsto$
$\left[\uparrow\left(u^{\prime}\right), \imath_{k+1}(x), \Uparrow\left(v^{\prime}\right), p\left(v^{\prime}\right), \checkmark\left(v^{\prime}\right), \jmath_{k+1}(y)\right]: u<u^{\prime}<x, v<v^{\prime}<y$
Terminate copy, replace current conf with new one

$$
\begin{aligned}
& {\left[\imath_{0}(f), \imath_{1}\left(x_{1}\right), \ldots, \imath_{k}\left(x_{k}\right), \imath_{k+1}(l), \jmath_{0}\left(f^{\prime}\right), \jmath_{1}\left(x_{1}^{\prime}\right), \ldots ., \jmath_{k}\left(x_{k}^{\prime}\right), \jmath_{k+1}\left(l^{\prime}\right), \uparrow(u), \Uparrow(v)\right] \leadsto} \\
& \quad\left[\imath_{0}\left(f^{\prime}\right), \imath_{1}\left(x_{1}^{\prime}\right), \ldots ., \imath_{k}\left(x_{k}^{\prime}\right), \imath_{k+1}\left(l^{\prime}\right), \operatorname{tr}_{t}\right]: \text { true }
\end{aligned}
$$

Fig. 2. Silent CMRS rules for $n e w_{t}$ : generation of a new configuration with $\bar{p}$-terms.
is a tuple $(Q, C, N, \delta)$ where $Q$ is a finite set of control states (the Cartesian product of those of each finite-state machine), $C$ is a finite set of channels, $M$ is a finite set of messages, $\delta$ is a finite set of transitions, each of which is of the form $\left(q_{1}, O p, q_{2}\right)$ where $q_{1}, q_{2} \in Q$, and $O p$ is a mapping from channels to channel operations. For any $c \in C$ and $a \in M$, an operation $O p(c)$ is either a send operation ! $a$, a receive operation ? $a$, the empty test $\epsilon$ ?, or the null operation nop. A configuration $\gamma$ is a pair $(q, w)$ where $q \in Q$, and $w$ is a mapping from $C$ to $M^{*}$ giving the content of each channel. The initial configuration $\gamma_{\text {init }}$ of $\mathcal{F}$ is the pair $\left(q_{0}, \varepsilon\right)$ where $q_{0} \in Q$, and $\varepsilon$ denotes the mapping that assigns the empty sequence $\epsilon$ to each channel. The (strong) transition relation (that defines the semantics of machines with perfect FIFO channels) is defined as follows: $\left(q_{1}, w_{1}\right) \xrightarrow{\sigma}\left(q_{2}, w_{2}\right)$ if and only if $\sigma=\left(q_{1}, O p, q_{2}\right) \in \delta$ such that if $O p(c)=!a$, then $w_{2}(c)=w_{1}(c) \cdot a$; if $O p(c)=? a$, then $w_{1}(c)=a \cdot w_{2}(c)$; if $O p(c)=\epsilon$ ? then $w_{1}(c)=\epsilon$ and $w_{2}(c)=\epsilon$; if $O p(c)=n o p$, then $w_{2}(c)=w_{1}(c)$. Now let $\preceq_{l}$ be the quasi ordering on LCS configurations such that $\left(q_{1}, w_{1}\right) \preceq_{l}\left(q_{2}, w_{2}\right)$ iff $q_{1}=q_{2}$ and $\forall c \in C: w_{1}(c) \preceq_{w} w_{2}(c)$ where $\preceq_{w}$ indicates the subword relation. By Higman's theorem, we know that $\preceq_{l}$ is a well-quasi ordering. We introduce then the weak transition relation $\xlongequal{\sigma}$ that defines the semantics of LCS: we have $\gamma_{1} \stackrel{\sigma}{\Longrightarrow} \gamma_{2}$ iff there exists $\gamma_{1}^{\prime}$ and $\gamma_{2}^{\prime}$ s.t. $\gamma_{1}^{\prime} \preceq_{l} \gamma_{1}, \gamma_{1}^{\prime} \xrightarrow{\sigma} \gamma_{2}^{\prime}$, and $\gamma_{2} \preceq_{l} \gamma_{2}^{\prime}$. Thus, $\gamma_{1} \stackrel{\sigma}{\Longrightarrow} \gamma_{2}$ means that $\gamma_{2}$ is reachable from $\gamma_{1}$ by first losing messages from the channels and reaching $\gamma_{1}^{\prime}$, then performing a transition, and, thereafter losing again messages from channels. As shown in [4], an LCS is well-structured with respect to $\preceq_{l}$. Furthermore, as shown in [5], in presence of transitions labeled

For $k=\alpha_{t}, i \in\{1, \ldots, k\}, j \in\{0, \ldots, k\}$, and $p \in P$ :
Start from first index
$\left[t r_{t}\right] \leadsto\left[t r_{t, 1}\right]:$ true
Select a token from an index in $\alpha_{t}$, apply $G_{t}$ to other indexes:
$\left[\imath_{0}\left(x_{0}\right), \imath_{1}\left(x_{1}\right), \ldots, \imath_{i}\left(x_{i}\right), \ldots, \imath_{k}\left(x_{k}\right), \imath_{k+1}\left(x_{k+1}\right), \bar{p}\left(x_{i}\right), t r_{t, i}\right] \leadsto$
$\quad\left[\imath_{0}\left(x_{0}\right), \imath_{1}\left(x_{1}\right), \ldots, \imath_{k}\left(x_{k}\right), \imath_{k+1}\left(x_{k+1}\right), \operatorname{apply} y_{t, i, p}(x)\right]+\sum_{j=1}^{k} G_{t}\left(S_{i}, p, S_{j}\right)^{x_{j}}: x_{0}<x<x_{k+1}$
Apply $G_{t}$ to indexes inside regions, move to the right
$\left[\imath_{j}(v), \operatorname{apply}_{t, i, p}(u), \bar{p}(u), \imath_{j+1}\left(v^{\prime}\right)\right] \leadsto$
$\left[\imath_{j}(v)\right.$, apply $\left.y_{t, i, p}\left(u^{\prime}\right), \bar{p}(u), \imath_{j+1}\left(v^{\prime}\right)\right]+G_{t}\left(S_{i}, p, R_{j}\right)^{u}: v<u<v^{\prime}, u<u^{\prime}$
Terminate visit continue with next token
$\left[\operatorname{apply}_{t, i, p}(u)\right] \sim\left[\operatorname{tr}_{t, i}\right]:$ true
Move to next index
$\left[\operatorname{tr}_{t, j}\right] \leadsto\left[\operatorname{tr}_{t, j+1}\right]:$ true
Terminate transfer of tokens for indexes in $\alpha_{t}$, start transfer of tokens of regions $\left[\imath_{0}(f), \operatorname{tr}_{t, k}\right] \leadsto\left[\operatorname{tr} R_{t}(f)\right]:$ true

Fig. 3. Silent CMRS rules for simulation of transfer: $G_{t}\left(S_{i}, p, \pi\right)^{x}$ is the multiset that, for each $q \in P$, contains $G_{t}\left(S_{i}, p, \pi, q\right)$ occurrences of the term $q(x)$.
with $\epsilon$, we can restrict our attention to systems with only one channel. As a last remark, notice that for any model with lossy semantics like LCS, e.g. lossy vector addition systems [20], the class of $c$-languages coincide with the class of $r$-languages, i.e., $L_{r}(L C S)=L_{c}(L C S)$.

Our first result is that $\Gamma_{0}$ and LCS define the same class of $c$-languages.
Theorem 4. $L_{c}\left(\Gamma_{0}\right)=L_{c}(L C S)$.
To prove the previous result, we give separate proofs of the two inclusions.
Proposition 5. $L_{c}(L C S) \subseteq L_{c}\left(\Gamma_{0}\right)$
Proof. Assume an LCS $\mathcal{F}$. We build a $\Gamma_{0} \mathcal{S}$ that simulates $\mathcal{F}$. The set of predicate symbols in $\mathcal{S}$ consists of the following: For each $q \in Q$, there is a nullary predicate symbol $q$ in $\mathcal{S}$. For each channel $c_{i}$ we use the function symbols head ${ }^{i}$ and $t a i l^{i}$ as pointers to the head and tail of the queue $c_{i}$. For each channel $c_{i}$ and each message $a \in M$ we have the predicate symbol $a^{i}$ in $\mathcal{S}$.
If $C=\left\{c_{1}, \ldots, c_{n}\right\}$, then the initial configuration $\left(s_{0}, \epsilon\right)$ is represented as

$$
M_{0}=\left[s_{0}, \operatorname{head}^{1}\left(v_{0}\right), \operatorname{tail}^{1}\left(v_{0}+1\right), \ldots, \operatorname{head}^{n}\left(v_{0}\right), \operatorname{tail}^{n}\left(v_{0}+1\right)\right]
$$

for some $v_{0} \in \mathbb{N}$.
In order to represent the queue $c_{i}$ containing the word $a_{1} a_{2} \ldots a_{n}$, we will use the multiset

$$
\left[\operatorname{head}^{i}\left(v_{0}\right), a_{1}^{i}\left(v_{1}\right), \ldots, a_{n}^{i}\left(v_{n}\right), \operatorname{tail}^{i}\left(v_{n+1}\right)\right]
$$

For $\mathrm{k}=\alpha_{\mathrm{t}}, \mathrm{i} \in\{0, \ldots, \mathrm{k}\}$, and any $\mathrm{p} \in \mathrm{P}$ :
Remove token and apply $G_{t}$ to indexes inside regions

$$
\begin{aligned}
& {\left[\imath_{0}\left(x_{0}\right), \imath_{1}\left(x_{1}\right), \ldots, \imath_{i}\left(x_{i}\right), \imath_{i+1}\left(x_{i+1}\right), \ldots, \imath_{k+1}\left(x_{k+1}\right), \bar{p}(u), \operatorname{tr} R_{t}(u)\right] \sim} \\
& \quad\left[\imath_{0}\left(x_{0}\right), \imath_{1}\left(x_{1}\right), \ldots, \imath_{i}\left(x_{i}\right), \imath_{i+1}\left(x_{i+1}\right), \ldots, \imath_{k+1}\left(x_{k+1}\right), \operatorname{tr} R_{t}(u)\right] \\
& \quad+G_{t}\left(R_{i}, p, R_{i}\right)^{u}+\Sigma_{j=1}^{k} G_{t}\left(R_{i}, p, S_{j}\right)^{x_{j}}: x_{i}<u<x_{i+1} \\
& \text { Move pointers to the right } \\
& {\left[\text { tr } R_{t}(u), \imath_{k+1}(l)\right] \sim\left[\operatorname{tr} R_{t}\left(u^{\prime}\right), \imath_{k+1}(l),\right]: u<u^{\prime}<l} \\
& \text { Terminate visit, move to addition step } \\
& \left.\left[\text { tr } R_{t}(u)\right] \leadsto[\text { add }]\right]: \text { true }
\end{aligned}
$$

Fig. 4. Silent CMRS rules for $t r R$ : transfer inside a region-index and from a regionindex to $\alpha_{t}$-indexes: $G_{t}\left(R_{i}, p, \pi\right)^{x}$ is the multiset that, for each $q \in P$, contains $G_{t}\left(R_{i}, p, \pi, q\right)$ occurrences of the term $q(x)$.

$$
\begin{aligned}
& \text { For } k=\alpha_{t}, i \in\{0, \ldots, k\} \text {, and any } p \in P: \\
& \text { Apply } H_{t} \text { to indexes in } \alpha_{t} \\
& {\left[\imath_{0}\left(x_{0}\right), \imath_{1}\left(x_{1}\right), \ldots, \imath_{k}\left(x_{k}\right), \text { add } d_{t}\right] \leadsto} \\
& \quad\left[\imath_{0}\left(x_{0}\right), \iota_{1}\left(x_{1}\right), \ldots, \imath_{k}\left(x_{k}\right), \text { add } R_{t}\left(x_{0}\right)\right]+\sum_{j=1}^{k} H_{t}\left(S_{j}\right)^{x_{j}}: \text { true }
\end{aligned}
$$

Apply $H_{t}$ to an index inside a region and advance pointer

$$
\begin{aligned}
& {\left[\imath_{i}(v), \imath_{i+1}\left(v^{\prime}\right), a d d R_{t}(u), \vee(u)\right] \leadsto} \\
& {\left[\imath_{i}(v), \imath_{i+1}\left(v^{\prime}\right), \operatorname{add} R_{t}\left(u^{\prime}\right)\right]+H_{t}\left(R_{i}\right)^{u}: v<u<v^{\prime}, u<u^{\prime}}
\end{aligned}
$$

Terminate simulation of transition $t$

$$
\left[\imath_{0}\left(x_{0}\right), \imath_{1}\left(x_{1}\right), \ldots, \imath_{k+1}\left(x_{k+1}\right), \operatorname{add} R_{t}(u)\right] \stackrel{\lambda(t)}{\sim}\left[\text { first }\left(x_{0}\right), \text { last }\left(x_{k+1}\right)\right]: \text { true }
$$

Fig. 5. CMRS rules for $a d d_{t}$ and $a d d R_{t}$ (all silent except the last one): $H_{t}(\pi)^{x}$ is the multiset that, for each $p \in P$, contains $H_{t}(\pi, p)$ occurrences of the term $p(x)$ for any $\pi \in R\left(\alpha_{t}\right)$.
for some positive integers $v_{0}<v_{1}<\ldots<v_{n+1}$.
Since an LCS transition ( $q_{1}, O p, q_{2}$ ) operates simultaneously on all the queues, the corresponding CMRS rule (with the same label) has the following form:

$$
\left[q_{1}\right]+B_{1}+\ldots+B_{n} \leadsto\left[q_{2}\right]+B_{1}^{\prime}+\ldots+B_{n}^{\prime}: C_{1} \cup \ldots \cup C_{n}
$$

where $B_{i}, B_{i}^{\prime}$ and $C_{i}$ define the encoding of $O p\left(c_{i}\right)$ for $i: 1, \ldots, n$. The encoding of the operation is defined by atomic formulas defined on a distinct variables $x_{1}, y_{1}, \ldots, x_{n}, y_{n}$ as follows. For $O p\left(c_{i}\right)=!a$,

$$
B_{i}=\left[\operatorname{tail}^{i}\left(x_{i}\right)\right] \quad B_{i}^{\prime}=\left[a^{i}\left(x_{i}\right), \operatorname{tail}^{i}\left(y_{i}\right)\right] \quad C_{i}=\left\{x_{i}<y_{i}\right\}
$$

For $O p\left(c_{i}\right)=? a$,

$$
B_{i}=\left[\operatorname{head}^{i}\left(x_{i}\right), a^{i}\left(y_{i}\right)\right] \quad B_{i}^{\prime}=\left[\operatorname{head}^{i}\left(y_{i}\right)\right] \quad C_{i}=\left\{x_{i}<y_{i}\right\}
$$

```
Start copy :
    \([f \operatorname{first}(f), \operatorname{last}(l)] \leadsto\left[\operatorname{first}_{1}(f), \operatorname{scan}_{t}^{0}(f), \operatorname{last}_{1}(l), \operatorname{copy}_{t}^{0}\left(l^{\prime}\right)\right]: l<l^{\prime}\)
Copy a term:
For \(i \in\{0,1\}\)
    \(\left[\operatorname{scan}_{t}^{i}(x), p(x), \operatorname{copy}_{t}^{i}(y)\right] \leadsto\left[\operatorname{scan}_{t}^{i}(x), \operatorname{copy}_{t}^{i}(y), p(y)\right]:\) true
Move to the right :
For \(i \in\{0,1\}\)
\(\left[\operatorname{scan}_{t}^{i}(x)\right.\), last \(\left._{1}(l), \operatorname{copy}_{t}^{i}(y)\right] \leadsto\)
    \(\left[\operatorname{scan}_{t}^{i}\left(x^{\prime}\right), \operatorname{last}_{1}(l), \operatorname{copy}_{t}^{i}\left(y^{\prime}\right)\right]: x<x^{\prime}<l, y<y^{\prime}\)
Move to the right and reserve a fresh value in the copy tape:
    \(\left[\operatorname{scan}_{t}^{0}(x), \operatorname{last}_{1}(l), \operatorname{copy}_{t}^{0}(y)\right] \sim\)
        \(\left[\operatorname{scan}_{t}^{1}\left(x^{\prime}\right), \operatorname{last}_{1}(l), \operatorname{fresh}_{t}\left(y^{\prime}\right), \operatorname{copy}_{t}^{1}\left(y^{\prime \prime}\right)\right]: x<x^{\prime}<l, y<y^{\prime}<y^{\prime \prime}\)
Terminate copy (copy tape becomes current configuration) :
\(\left[\right.\) first \(\left._{1}(f), \operatorname{scan}_{t}^{1}(x), \operatorname{last}_{1}(l), \operatorname{copy}_{t}^{1}(y)\right] \leadsto\left[\right.\) first \(_{1}(l)\), last \(\left._{1}(y)\right]:\) true
Subtraction:
For \(j \in\{1, \ldots, k\}\)
    \(\left[\right.\) first \(_{1}(f)\), fresh \(\left._{t}(x), \operatorname{last}_{1}(l)\right]+\sum_{i=1}^{k} F_{t}\left(S_{i}\right)^{x_{i}} \leadsto\)
        \(\left[\imath_{0}(f), \imath_{1}\left(x_{1}\right), . ., \imath_{j}(x), . ., \imath_{k}\left(x_{k}\right), \imath_{k+1}(l), n e w_{t}\right]:\)
            \(f<x_{1}<. .<x_{j}<x<x_{j+1}<x_{k}<l\)
```

Fig. 6. Silent CMRS rules for the simulation of subtraction with selection of one fresh value.

For $O p\left(c_{i}\right)=$ empty?,

$$
B_{i}=\left[\operatorname{head}^{i}\left(x_{i}\right) \mid \operatorname{tail}^{i}\left(y_{i}\right)\right] \quad B_{i}^{\prime}=\left[\operatorname{head}^{i}\left(x_{i}^{\prime}\right) \mid \operatorname{tail}\left(y_{i}^{\prime}\right)\right] \quad C_{i}=\left\{y_{i}<x_{i}^{\prime}<y_{i}^{\prime}\right\}
$$

For $O p\left(c_{i}\right)=n o p$,

$$
B_{i}=B_{i}^{\prime}=[] \quad C_{i}=\text { true }
$$

Now let $\gamma_{a c c}=\left(s_{a c c}, \epsilon\right)$ be an accepting LCS configuration (i.e. all configurations with control state $s_{a c c}$ allow to accept words).
The accepting CMRS configuration is $\left[s_{a c c}\right]$. Let us consider a LCS with one channel. Note that $n$ channels can be encoded into one channel. Hence, considering a unique channel is not restrictive.
The following properties then hold. Given an LCS configuration $\gamma=(s, w)$, let $\gamma^{\bullet}$ be the corresponding CMRS encoding. Moreover, given $\gamma^{\bullet}$ containing head ${ }^{1}(c)$, let $\mathrm{G}\left(\gamma^{\bullet}\right)$ be the set of CMRS configurations built from $\gamma^{\bullet}$ by adding some ground terms $a^{1}\left(c^{\prime}\right)$ where $a \in M$ and $c^{\prime}<c$, i.e. by adding useless ground terms corresponding to lost messages.
It is easy to check that (1) if $\gamma_{1}^{\bullet} \xrightarrow{\rho} \eta$ with $\eta \in \mathbf{G}\left(\gamma_{2}^{\bullet}\right)$ in $\mathcal{S}$, then $\gamma_{1} \xrightarrow{\sigma_{\rho}} \gamma_{2}$ in $\mathcal{F}$ where $\rho$ is the CMRS rule corresponding to the LCS transition $\sigma_{\rho}$. Indeed, notice that in the CMRS implementation of the dequeue operation we move the head pointer to an arbitrary position within the queue and thus we perform a lossy step followed by a dequeue step. Similarly, the emptiness test is simulated by means of a lossy step in which all elements are removed from the queue (with

For $\mathrm{k}=\alpha_{\mathrm{t}}, \mathrm{i} \in\{0, \ldots, \mathrm{k}\}$, and any $\mathrm{p} \in \mathrm{P}:$
Remove token, apply $G_{t}$ to indexes inside region and to $\alpha_{t}$-indexes

$$
\begin{aligned}
& {\left[\imath_{0}\left(x_{0}\right), \imath_{1}\left(x_{1}\right), \ldots, \imath_{i}\left(x_{i}\right), \imath_{i+1}\left(x_{i+1}\right), \ldots, \imath_{k+1}\left(x_{k+1}\right), \bar{p}(u), \operatorname{tr} R_{t}(u)\right] \leadsto} \\
& {\left[\imath_{0}\left(x_{0}\right), \imath_{1}\left(x_{1}\right), \ldots, \imath_{i}\left(x_{i}\right), \imath_{i+1}\left(x_{i+1}\right), \ldots, \imath_{k+1}\left(x_{k+1}\right), \operatorname{apply} R_{i, p, t}(u)\right]} \\
& \quad+G_{t}\left(R_{i}, p, R_{i}\right)^{u}+\Sigma_{j=1}^{k} G_{t}\left(R_{i}, p, S_{j}\right)^{x_{j}}: x_{i}<u<x_{i+1}
\end{aligned}
$$

Apply $G_{t}$ to other region - indexes, move to the right For $\mathrm{i} \neq \mathrm{j}, \pi \in\{\overline{\mathrm{p}}, \sqrt{ }\}$ :
$\left[\imath_{j}(v), \operatorname{apply} R_{t, i, p}(u), \pi(u), \imath_{j+1}\left(v^{\prime}\right)\right] \leadsto$
$\left[\imath_{j}(v)\right.$, apply $\left.R_{t, i, p}\left(u^{\prime}\right), \pi(u), \imath_{j+1}\left(v^{\prime}\right)\right]+G_{t}\left(R_{i}, p, R_{j}\right)^{u}: v<u<v^{\prime}, u<u^{\prime}$
Apply $M_{t}$ to other index in the same region, move to the right

$$
\begin{aligned}
& {\left[v_{i}(v), \text { apply } R_{t, i, p}(u), \pi(z), \imath_{i+1}\left(v^{\prime}\right)\right] \leadsto} \\
& {\left[i_{i}(v), \operatorname{apply} R_{t, i, p}\left(u^{\prime}\right), \pi(z), \imath_{i+1}\left(v^{\prime}\right)\right]+M_{t}\left(R_{i}, p, R_{j}\right)^{z}: z \neq u, v<u<v^{\prime}, u<u^{\prime}}
\end{aligned}
$$

Terminate visit continue with next token
$\left[\operatorname{appl}_{t, i, p}(u)\right] \leadsto\left[\operatorname{tr} R_{t}\right]:$ true

Move $\operatorname{tr} R_{t}$ pointer to the right
$\left[\operatorname{tr} R_{t}(u), \imath_{k+1}(l)\right] \leadsto\left[\operatorname{tr} R_{t}\left(u^{\prime}\right), \imath_{k+1}(l),\right]: u<u^{\prime}<l$
Terminate visit, move to addition step

$$
\left[\operatorname{tr} R_{t}(u)\right] \sim\left[a d d_{t}\right]: \text { true }
$$

Fig. 7. Silent CMRS rules for the simulation of transfer between distinct regions, or distinct indexes inside the same region.
the weak reduction of LCS the emptiness test is always executable and it has the effect of emptying the queue). Finally, the enqueue operation is simulated in an exact way. We can also easily see that for all $\eta \in \mathrm{G}\left(\gamma_{1}^{\bullet}\right):(2)$ if $\eta \xrightarrow{\rho} \eta^{\prime}$, then $\gamma_{1}^{\bullet} \xrightarrow{\rho} \gamma_{2}^{\bullet}$ with $\eta^{\prime} \in \mathrm{G}\left(\gamma_{2}^{\bullet}\right)$. Hence, if $\gamma_{0}^{\bullet} \xrightarrow{\rho_{1}} \eta_{1} \ldots \xrightarrow{\rho_{n}} \eta_{n}$ with $\left[s_{a c c}\right] \preceq_{c} \eta_{n}$, then we deduce from (2) that for all $i: 1 \leq i \leq n$, there exists $\gamma_{i}^{\bullet}$ such that $\eta_{i} \in \mathrm{G}\left(\gamma_{i}^{\bullet}\right)$. Moreover, for all $i: 0 \leq i<n, \gamma_{i}^{\bullet} \xrightarrow{\rho_{i}} \eta_{i+1}^{\prime}$ with $\eta_{i+1}^{\prime} \in \mathrm{G}\left(\gamma_{i+1}^{\bullet}\right)$. Since $\left[s_{a c c}\right] \preceq_{c} \eta_{n}$ and $\eta_{n} \in \mathrm{G}\left(\gamma_{n}^{\bullet}\right)$ we also have $\left[s_{a c c}\right] \preceq_{c} \gamma_{n}^{\bullet}$. Following (1), we deduce $\gamma_{0} \stackrel{\sigma_{\rho_{1}}}{\Longrightarrow} \gamma_{1} \ldots \stackrel{\sigma_{\rho_{n}}}{\Longrightarrow} \gamma_{n}$ with $\gamma_{a c c} \preceq_{l} \gamma_{n}$.
Vice versa, suppose that $\gamma_{1} \xrightarrow{\sigma} \gamma_{2}$ in $\mathcal{F}$. Then, we have that there exists $\eta$ such that $\gamma_{1}^{\bullet} \xrightarrow{\rho_{\sigma}} \eta^{\bullet}$ and $\gamma_{2}^{\bullet} \preceq_{c} \eta^{\bullet}$ where $\rho_{\sigma}$ is the CMRS rule corresponding to $\sigma$. This is immediately verified for the enqueue operation and for the empty test (their simulation is exact, and thus returns a more precise representation of the queues). The same holds for the dequeue operation since we cannot forget elements to the right of the new position of the header.
Now let $\gamma_{1} \xlongequal{\rho_{0}} \gamma_{2} \ldots \xlongequal{\rho_{n}} \gamma_{n}$ with $\gamma_{a c c} \preceq_{l} \gamma_{n}$. Then, we know that there exist $\eta_{2}^{\prime}, \ldots, \eta_{n}^{\prime}$ such that $\gamma_{i}^{\bullet} \xrightarrow{\sigma_{\rho_{i-1}}} \eta_{i+1}^{\bullet}$ and $\gamma_{i+1}^{\bullet} \preceq_{c} \eta_{i+1}^{\bullet}$ for $i: 1, \ldots, n-1$. By the monotonicity of CMRS, we have that $\gamma_{1}^{\bullet} \xrightarrow{\sigma_{\rho_{0}} \ldots \sigma_{\rho_{n}}} \eta_{n}^{\bullet}$, and $\left[s_{a c c}\right] \preceq_{c} \gamma_{a c c}^{\bullet} \preceq_{c} \eta_{n}^{\bullet}$.

Autonomous transitiont ${ }_{1}$ :

$$
F_{t_{1}}{ }^{x}+[\checkmark(x), \operatorname{last}(l)] \stackrel{\lambda\left(t_{1}\right)}{\sim} H_{t_{1}}{ }^{x}+[\checkmark(x), \operatorname{last}(l)]: x<l
$$

Synchronized transitions $t_{2}$ and $t_{3}$ :

$$
\begin{aligned}
& \quad F_{t_{2}}{ }^{x}+F_{t_{3}}{ }^{y}+[\sqrt{ }(x), \checkmark(y), \text { last }(l)] \stackrel{\lambda\left(t_{2}, t_{3}\right)}{\sim} \\
& H_{t_{2}}{ }^{x}+H_{t_{3}}{ }^{y}+[\checkmark(x), \checkmark(y), \text { last }(l)]: x \neq y<l \\
& \text { Replication rule } t_{4} \text { labelledwiththemarkingM }: \\
& \quad F_{t_{4}}{ }^{x}+[\checkmark(x), \text { last }(y)] \xrightarrow{\lambda\left(t_{4}\right)} H_{t_{4}}{ }^{x}+M^{y}+[\checkmark(x), \checkmark(y), \text { last }(z)]: z>y
\end{aligned}
$$

Fig. 8. CMRS rules that simulate autonomous, synchronized, and replication transitions.

Proposition 6. $L_{c}\left(\Gamma_{0}\right) \subseteq L_{c}(L C S)$
Proof. Consider a $\Gamma_{0} \mathcal{S}$ over the finite set of predicate symbols $\mathbb{P}$, an initial configuration $\gamma_{\text {init }}$ and an accepting configuration $\gamma_{a c c}$. W.l.o.g. let us assume that for each $p(v) \in \gamma_{i n i t}+\gamma_{a c c}: 0 \leq v \leq c \max$ (to satisfy this condition we can add a rule that is never fireable and in which there is a constant greater than all values in $\gamma_{\text {init }}+\gamma_{a c c}$ ).

The proof follows three steps: first, we show how to encode a configuration as words (i.e. contents of LCS queues). Second, we show how a rule $L \sim R: \psi$ can be applied to the word representation of configurations, and finally we show how to simulate such an application using an LCS.
$\Gamma_{0}$ configurations as words. $\Gamma_{0}$ configurations consisting of terms with strictly increasing parameters can be naturally viewed as words defined over the corresponding predicate symbols. The execution of a $\Gamma_{0}$ on such a configuration, however, may lead to a new configuration with two terms with the same value.

As an example, consider the rule $\rho$ defined as $[p(x), q(y)] \leadsto[p(x), r(z), q(y)]$ : $x<z<y$ and a configuration $\gamma=[p(0), t(3), q(6)]$. We notice here that the application of $\rho$ to $\gamma$ may lead to different results depending on the valuation of $z$. One of the possible successors is $\gamma^{\prime}=[p(0), r(3), t(3), q(6)] . \gamma^{\prime}$ obtained by applying the valuation $x \mapsto 0, z \mapsto 3, y \mapsto 6$. The question now is if we gain something in assigning to $z$ the same value of a parameter in another term. The answer is no. Indeed, since in $\Gamma_{0}$ we cannot test for $=$ in a rule, the effect of mapping $z$ to 3 only restricts the set of rules that can be fired at $\gamma^{\prime}$, i.e., this choice can lead to dead ends.

This intuition is made formal in Prop. 3. This lemma tells us that all strings in $L_{c}(\mathcal{S})$ can be recognized by an execution that passes through configurations where all the terms with a value greater than cmax are totally ordered on the values of their parameters w.r.t. $<$ (i.e. they can be viewed as words). Notice that this reasoning can be applied only to terms with values greater than cmax. Indeed, for this kind of terms Prop. 2 tells us that if we fire a sequence of transitions and reach a configurations $\gamma$ from $\gamma_{i n i t}$ then we can fire the same sequence of transitions from $\gamma_{i n i t}$ and reach a configuration with larger gap than in $\gamma$. Prop. 2 also implies that we do not have to retain gap between parameters
greater than cmax since it is always possible to increase them. Terms with values smaller than cmax must be treated in a special way.

More precisely, a configuration $\gamma$ is encoded as a word $w_{1} \cdot w_{2}$ where $w_{1}$ and $w_{2}$ are built as follows.

- Each ground term $p(c) \in \gamma$ with $0 \leq c \leq c m a x$ is encoded as a (message) symbol $(p, c)$. Thus, from $\gamma$ we first extract the word

$$
w_{1}=w_{1}^{0} \cdots w_{1}^{c \max }
$$

where $w_{1}^{i}$ has many occurrences of $(p, i)$ as those of $p(i)$ for any predicate $p \in \mathbb{P}$ and $0 \leq i \leq c m a x$ (multiple occurrence of the same term produce different symbols, to disambiguate the encoding we assume a total order on symbols in $\mathbb{P}$ ).

- Each ground term $p(c) \in \gamma$ with $c>c \max$ is encoded as a symbol $p$. Thus, from $\gamma$ we also extract the word

$$
w_{2}=p_{1} \cdots p_{k}
$$

where $p_{i}\left(c_{i}\right) \in \gamma, c_{i}>c \max$ and $c_{i}<c_{j}$ for $1 \leq i<j \leq k$. Here we assume that there cannot be two terms with the same value for parameters greater than cmax.
Applying rewriting rules to words. W.l.o.g. we assume that each rule $L \leadsto R: \psi$ in $\mathcal{S}$ with set of variables $V$ satisfies the following conditions.

- For each $x \in V$, either $(x=c) \in \psi$ and $0 \leq c \leq c m a x$ or $(x>c m a x) \in \psi$.
- Furthermore, we assume that for all pair of variables $x, y$ in $L+R$ such that $x>c \max \in \psi$ and $y>c \max \in \psi$ we have that $x \bullet y \in \psi$ with $\bullet \in\{=,>,<\}$.
Given a $\Gamma_{0}$ rule $\rho$, we can compile $\rho$ in a finite set of $\Gamma_{0}$ rules that satisfy the above mentioned conditions and that model the possible effects of applying $\rho$. The rules are obtained by completing the order in $\rho$ with all possible missing relations between variables. By Prop. 3, we can safely introduce new equality constraints only when the resulting rule respecting the restrictions of $\Gamma_{0}$ (i.e. we do not need to introduce equality constraints involving more than two variables). As an example, the effect of the rule $[p(x), q(y)] \leadsto[r(z)]: x<y$ on a configuration in $\Gamma_{0}$ is modelled by the rules $[p(x), q(y)] \leadsto[p(z)]: x<y<z,[p(x), q(y)] \leadsto$ $[p(z)]: x<z<y,[p(x), q(y)] \leadsto[p(z)]: z<x<y,[p(x), q(y)] \leadsto[p(x)]: x<y$, and $[p(x), q(y)] \leadsto[p(y)]: x<y$. Notice that in the last two rules we introduce an implicit equality between a variable in the rhs and a variable in the lhs.

Under these assumptions, a rule $\rho=L \leadsto R: \psi$ in $\mathcal{S}$ defined over the variables $V=\left\{x_{1}, \ldots, x_{m+r}\right\}$ can be represented by the word

$$
w^{\rho}=w_{1}^{\rho} \cdot w_{2}^{\rho}
$$

where $w_{1}^{\rho}$ describes the effect of $\rho$ on $w_{1}$, i.e. on ground terms with parameter smaller than cmax, and $w_{2}^{\rho}$ describes the effect of $\rho$ on $w_{2}$, i.e. on ground terms with parameter greater than cmax. More precisely,

$$
w_{2}^{\rho}=\left[\begin{array}{l}
\alpha_{1} \\
\beta_{1}
\end{array}\right] \ldots\left[\begin{array}{l}
\alpha_{r} \\
\beta_{r}
\end{array}\right]
$$

is the maximal sequence that satisfies the following conditions:

- For $1 \leq k \leq r$,
- $\alpha_{k}=p \in \mathbb{P}$, if $p\left(x_{j_{k}}\right) \in L$ and $\left(x_{j_{k}}>c \max \right) \in \psi$,
- $\beta_{k}=q \in \mathbb{P}$, if $q\left(x_{j_{k}}\right) \in R$ and $\left(x_{j_{k}}>c \max \right) \in \psi$,
- $\alpha_{k}$ and $\beta_{k}$ are equal to $\epsilon$ in all other cases.
- For $1 \leq k \leq r-1, x_{j_{k}}<x_{j_{k+1}}$ follows from $\psi$;

Pairs of the form $\left[\begin{array}{l}\epsilon \\ \epsilon\end{array}\right]$ are not included in $w_{2}^{\rho}$. The word $w_{2}^{\rho}$ specifies the order of terms in $\rho$ and how a single term of a configuration (element in a word) is modified (using the pair $\left[\begin{array}{l}\alpha \\ \beta\end{array}\right]$ ) by the rule. Notice that the syntactic restrictions of $\Gamma_{0}$ ensure that there cannot be elements $\left[\begin{array}{l}\alpha \\ \beta\end{array}\right]$ with more than one predicate in $\alpha$ or $\beta$. Furthermore, if $\alpha=\epsilon$ then $\rho$ adds a new occurrence of $\beta$, if $\beta=\epsilon$ then $\rho$ removes an occurrence of $\alpha$.

As an example, for $c m a x=2$, the rule $\rho$ defined as $[p(x), q(y)] \leadsto[q(x), r(z)]:$ $2<x<z<y$ is represented by the word

$$
w_{2}^{\rho}=\left[\begin{array}{l}
p \\
q
\end{array}\right] \cdot\left[\begin{array}{l}
\epsilon \\
r
\end{array}\right] \cdot\left[\begin{array}{l}
q \\
\epsilon
\end{array}\right]
$$

The word $w_{1}^{\rho}=w_{1,0}^{\rho} \cdots w_{1, \text { cmax }}^{\rho}$ is such that for all $i: 0 \leq i \leq \operatorname{cmax}, w_{1, i}^{\rho}=\varepsilon$ if there is no variable $x$ such that $x=i \in \psi$, otherwise $w_{1, i}^{\rho}$ is a sequence of elements of the form $\left[\begin{array}{c}(p, i) \\ \epsilon\end{array}\right]$ or $\left[\begin{array}{c}\epsilon \\ (p, i)\end{array}\right]$ that describe the modifications that concerns ground terms with parameter $x=i$ as mentioned above.

Given a word $w_{1} \cdot w_{2}$ associated to a configuration $\gamma$ and a word $w_{1}^{\rho} \cdot w_{2}^{\rho}$ associated to a rule $\rho$, it should be clear now that the application of $\rho$ to $\gamma$ can be simulated by rewriting $w_{1}$ according to the ordered pairs in $w_{1}^{\rho}$ and $w_{2}$ according to the ordered pairs in $w_{2}^{\rho}$. Clearly, the application of $w_{2}^{\rho}$ to $w_{2}$ has some non-determinism, since we only have to ensure that in the resulting string $w_{2}^{\prime}$ the order in the two strings is preserved.

Going back to our example, the application of $w_{2}^{\rho}$ to the word $w_{2}=q \cdot p \cdot s \cdot q \cdot t$ produces the strings $q \cdot q \cdot r \cdot s \cdot t$ and $q \cdot q \cdot s \cdot r \cdot t$. (we recall that Prop. 3 tells us that we can safely ignore configurations in which $r$ gets the same value as $s$ ).

Simulation in LCS. We are ready now to define the encoding of a $\Gamma_{0} \mathcal{S}$ into an LCS $\mathcal{F}$. The LCS $\mathcal{F}$ has one channel $c$ that contains the word encodings of configurations and one channel $c^{\prime}$ used as auxiliary memory. The control states of $\mathcal{F}$ are used for encoding different steps of simulation of a rule $\rho=L \leadsto R: \psi$ where we assume that all pair of variables $x, y$ in $\rho$ are in the relation $<\cup=$ induced by $\psi$. In particular, we will assume to have one distinct control state for each pair in $w^{\rho}$.

First, we simulate the effect on the terms with parameters less or equal than cmax. For a fixed $i 1 \leq i \leq c m a x$, the simulation consists in dequeuing
symbols of the form $(p, i)$ from $c$, and by copying them into $c^{\prime}$ after applying the transformations defined in $w_{1}^{\rho}$. Notice that the information on the structure of $w_{1}^{\rho}$ can be stored in the control states of $\mathcal{F}$. When there are no more symbols of the form $(p, i)$ in $c$, we moves to the value $i+1$.

Second, we simulate the effect of $w_{2}^{\rho}$ for ground terms $p(c)$ with $c>c \max$. Suppose that $w_{2}^{\rho}$ has $r$ pairs $\left[\begin{array}{l}\alpha \\ \beta\end{array}\right]$. Starting from the first pair in $w_{2}^{\rho}$, we define control states in which we either copy symbols from $c$ to $c^{\prime}$ or, non deterministically, decide to apply the current pair $\left[\begin{array}{l}\alpha \\ \beta\end{array}\right]$ to the head $p$ of the queue in $w_{2}^{\rho}$ :

- If $\alpha=p$ and $\beta=q$, then we remove $p$ from $c$, add $q$ to $c^{\prime}$, and move to the next pair in $w_{2}^{\rho}$;
- If $\alpha=p$ and $\beta=\epsilon$, then remove $p$ from $c$, and move to the next pair in $w_{2}^{\rho}$;
- If $\alpha=\epsilon$ and $\beta=q$, we add $q$ to $c^{\prime}$, and move to the next pair in $w_{2}^{\rho}$.

Note that since we non-deterministically choose the positions where modifications must be applied, the LCS $\mathcal{F}$ may get into a deadlock. Deadlocked computations do not influence the language $L_{c}(\mathcal{F})$.

Once the new word has been written into $c^{\prime}$ (and $c$ is empty), we copy the content of $c^{\prime}$ into $c$ and get ready for simulating the execution of another rule. In this last step we also recognize the symbols $\lambda(\rho)$ that labels $\rho$ (all the other transitions used to simulate $\rho$ are labelled by $\varepsilon$ ).

Finally, note that channels may lose messages. As a consequence, we encode lossy $\Gamma_{0}$ into LCS where ground terms may non-deterministically disappear during executions. However, $r$-language accepted by $\Gamma_{0}$ and lossy $\Gamma_{0}$ are the same (for coverability acceptance).

We show next that CMRS are strictly more expressive than LCS and $\Gamma_{0}$.
Theorem 5. $L_{c}(L C S) \subset L_{c}(C M R S)$.
Proof. We define a language $L_{e n t}$ which is accepted by a CMRS and that cannot be accepted by any LCS. Assume a finite alphabet $\Sigma$ such that $\{\$, \#\} \nsubseteq \Sigma$. For each $w=a_{1} \cdots a_{k} \in \Sigma^{*}$, we interpret $w$ in the following as the multiset $\left[a_{1}, \ldots, a_{k}\right]$. Hence, we do not distinguish words in $\Sigma^{*}$ from the multiset they represent, and vice versa. In particular, we will use the notation $a_{1} \cdots a_{k} \leq$ $a_{1}^{\prime} \cdots a_{l}^{\prime}$ to denote that $\left[a_{1}, \ldots, a_{k}\right] \leq\left[a_{1}^{\prime}, \ldots, a_{l}^{\prime}\right]$. Define $V$ to be the set of words of the form $w_{1} \# w_{2} \# \cdots \# w_{n}$ where $w_{i} \in \Sigma^{*}$ for each $i: 1 \leq i \leq n$. Consider $v=w_{1} \# w_{2} \# \cdots \# w_{m} \in V$ and $v^{\prime}=w_{1}^{\prime} \# w_{2}^{\prime} \# \cdots \# w_{n}^{\prime} \in V$. We write $v \sqsubseteq v^{\prime}$ to denote that there is an injection $h:\{1, \ldots, m\} \mapsto\{1, \ldots, n\}$ such that

1. $1 \leq i<j \leq m$ implies $h(i)<h(j)$ ( $h$ is monotonic) and
2. $w_{i} \leq w_{h(i)}^{\prime}$ ( $\leq$ is multiset inclusion) for each $i: 1 \leq i \leq m$.

We now define the language $L_{\text {ent }}=\left\{v \$ v^{\prime} \mid v^{\prime} \sqsubseteq v\right\} \subseteq(\Sigma \cup\{\#, \$\})^{*}$. As an example, given $\Sigma=\{a, b\}$, we have that $[a, b, b] \#[a, b, b] \#[a, a] \$[b, a] \#[a, a]$ is in $L_{\text {ent }}$, whereas $[a, b, b] \#[b, a, b] \#[a, a] \$[a, a] \#[a, b]$ is not in $L_{\text {ent }}$.

We now exhibit a CMRS $\mathcal{S}$ with $L_{c}(\mathcal{S})=L_{\text {ent }}$. The set of predicate symbols which appear in $\mathcal{S}$ consists of (i) a predicate symbol $a$ for each $a \in \Sigma$, and (ii) the symbols guess, check, sep\# and ok. The initial configuration $\gamma_{i n i t}$ is defined as $[$ guess $(0)]$. Furthermore, we have the following rules:
(1) For each $a \in \Sigma$, we have a rule labelled with $a$ and which is of the form

$$
[\operatorname{guess}(x)] \sim[\operatorname{guess}(x), a(x)]: \operatorname{true}
$$

Rules of this form are used to guess the letters in $w_{i}$ in the first part of a word in $L_{\text {ent }}$. We keep track of the symbols inside $w_{i}$ through their argument. These arguments are all the same by definition of the rule.
(2) A rule labelled with \# of the form:

$$
[\operatorname{guess}(x)] \leadsto\left[\operatorname{sep}_{\#}(x), \operatorname{guess}(y)\right]:\{x<y\}
$$

This rule is used to switch from the guessing of the part $w_{i}$ to the guessing of the next part $w_{i+1} \cdot \operatorname{sep}_{\#}(x)$ remembers the parameter on which the switch has been executed.
(3) A rule labelled with $\$$ of the form:

$$
[\operatorname{guess}(x)] \sim\left[\operatorname{check}(y), \operatorname{sep}_{\#}(x)\right]:\{y=0\}
$$

This rule is used to switch from the guessing of the part $w_{1} \# \ldots \# w_{n}$ to the selection of the second part of the word. The parameter of check is equal to the initial value of guess, i.e. to 0 . This way, we can scan the word stored in the first phase from left-to-right, i.e., working on the argument order we define a monotonic injective mapping $h$.
(4) For each $a \in \Sigma$, we have a rule labelled with $a$ which is of the form

$$
[\operatorname{check}(y), a(y)] \leadsto[\operatorname{check}(y)]: \text { true }
$$

This rule is used to read a word (multiset) $u_{i}$ contained in $w_{h(i)}$.
(5) A rule labelled with \# of the form:

$$
\left[\operatorname{check}(x), \operatorname{sep}_{\#}(x), \operatorname{sep}_{\#}(y)\right] \leadsto\left[\operatorname{check}(y), \operatorname{sep}_{\#}(y)\right]:\{x<y\}
$$

This rule is used to pass from $u_{i}$ to $u_{i+1}$ for $i \geq 1$.
(6) A rule labelled with $\epsilon$ of the form:

$$
[\operatorname{check}(x)] \leadsto[o k(y)]:\{y=0\}
$$

This rule is used to non-deterministically terminate the checking phase. The accepting configuration $\gamma_{a c c}$ is defined as [ok(0)].

Assuming that $\Sigma=\{a, b\}$, we now show that $L_{e n t}$ is not an LCS language. Suppose that $L_{c}(\mathcal{F})=L_{\text {ent }}$ for some LCS $\mathcal{F}=(Q,\{c\}, M, \delta)$. We show that this leads to a contradiction. Let $\gamma_{\text {init }}$ be the initial global state in $\mathcal{F}$ and $\gamma_{a c c}$ be the accepting global state. We use a binary encoding enc : $Q \cup M \mapsto \Sigma^{*}$ such that $e n c(m) \not \leq e n c\left(m^{\prime}\right)$ if $m \neq m^{\prime}$. We will also use a special word $v_{i n i t} \in \Sigma^{*}$ such
that $v_{\text {init }} \not \leq e n c(m)$ for each $m \in Q \cup M$. It is clear that such enc function and $v_{\text {init }}$ exist. As an example, if $|Q \cup M|=n$ then we define enc as an injective map from $Q \cup M$ to multisets of $n+1$ elements with $i+1$ occurrences of $a$ and $n-i$ occurrences of $b$ for $0 \leq i \leq n$, and we use the multiset with $n+1$ occurrences of $b$ for $v_{\text {init }}$. For instance, for $n=2$ we use $[a, a, a],[a, a, b],[a, b, b]$ for control states and messages and $[b, b, b]$ for $v_{\text {init }}$. We extend enc to global states such that if $\gamma=\left(q, m_{1} m_{2} \cdots m_{n}\right)$ then

$$
\operatorname{enc}(\gamma)=\operatorname{enc}(q) \# \operatorname{enc}\left(m_{1}\right) \# \operatorname{enc}\left(m_{2}\right) \# \cdots \# \operatorname{enc}\left(m_{n}\right)
$$

Observe that (i) enc $(\gamma) \in V$; (ii) for global states $\gamma_{1}$ and $\gamma_{2}$, it is the case that $\gamma_{1} \preceq_{l} \gamma_{2}$ iff $e n c\left(\gamma_{1}\right) \sqsubseteq e n c\left(\gamma_{2}\right)$; and (iii) $v_{\text {init }} \nsubseteq e n c(\gamma)$ for each global state $\gamma$.

Since $L_{\text {ent }}=L_{c}(\mathcal{F})$ and $v \$ v \in L_{\text {ent }}$ for each $v \in V$, it follows that for each $v \in V$, there is a global state $\gamma$ such that $\gamma_{\text {init }} \xrightarrow{v} \gamma \xrightarrow{\$ v} \gamma^{\prime}$ with $\gamma_{a c c} \preceq_{l} \gamma^{\prime}$. We use reach $(v)$ to denote $\gamma$. We define two sequences $\gamma_{0}, \gamma_{1}, \gamma_{2}, \ldots$ of global states, and $v_{0}, v_{1}, v_{2}, \ldots$ of words in $V$ such that $v_{0}=v_{i n i t}, \gamma_{i}=\operatorname{reach}\left(v_{i}\right)$, and $v_{i+1}=\operatorname{enc}\left(\gamma_{i}\right)$ for each $i \geq 0$. By Higman's theorem we know that there is a $j$ such that $\gamma_{i} \preceq_{l} \gamma_{j}$ for some $i<j$. Let $j$ be the smallest natural number satisfying this property. First, we show that $v_{i} \nsubseteq v_{j}$. There are two cases: if $i=0$ then $v_{i} \not \equiv v_{j}$ by $(i i i)$; if $i>0$ then we know that $\gamma_{i-1} \preceq_{l} \gamma_{j-1}$ and hence, following (ii), $v_{i}=\operatorname{enc}\left(\gamma_{i-1}\right) \nsubseteq \operatorname{enc}\left(\gamma_{j-1}\right)=v_{j}$. Since $\gamma_{j}=\operatorname{reach}\left(v_{j}\right)$, we know that $\gamma_{i n i t} \xrightarrow{v_{j}} \gamma_{j}$. By monotonicity, $\gamma_{i} \xrightarrow{\$ v_{i}} \gamma_{i}^{\prime}, \gamma_{a c c} \preceq_{l} \gamma_{i}^{\prime}, \gamma_{i} \preceq_{l} \gamma_{j}$ implies $\gamma_{j} \xrightarrow{\$ v_{i}} \gamma_{j}^{\prime}$ with $\gamma_{a c c} \preceq_{l} \gamma_{i}^{\prime} \preceq_{l} \gamma_{j}^{\prime}$. We conclude that $\gamma_{i n i t} \xrightarrow{v_{j}} \gamma_{j} \xrightarrow{\$ v_{i}} \gamma_{j}^{\prime}$ with $\gamma_{a c c} \preceq_{l} \gamma_{j}^{\prime}$. Hence, $v_{j} \$ v_{i} \in L_{c}(\mathcal{F})=L_{\text {ent }}$ which is a contradiction since $v_{i} \nsubseteq v_{j}$.

Let us now consider $r$-languages. As mentioned at the beginning of the section, the expressive power of LCS remains the same as for coverability accepting conditions, However, this property does not hold anymore for $\Gamma_{0}$.

Proposition 7. $L_{c}\left(\Gamma_{0}\right) \subset L_{r}\left(\Gamma_{0}\right)=L_{r}(C M R S)=R E$.

Proof. It is well known that perfect FIFO channel systems with reachability accepting condition recognize the class RE. We prove that perfect channel systems accept the same languages as $\Gamma_{0}$ with reachability accepting condition. Given an LCS $\mathcal{F}$, let $\mathcal{S}$ be the $\Gamma_{0}$ used to encode an LCS in the proof of Theorem 4 . In each step of a run $\sigma$ in $\mathcal{S}$ the head and tail delimiters are moved to the right of their current positions. Thus, a "lost" ground term to left of the head delimiter, i.e. with parameter smaller than that of $h_{c}$, can never be removed in successive steps of $\sigma$. This implies that an accepting configuration in which all ground terms have parameters strictly greater than the parameter of the head delimiter characterize reachable configurations of a perfect FIFO channel system.

Hence, we have the following property.
Corollary 1. $L_{r}(L C S) \subset L_{r}(C M R S)$.

## 7 Petri Nets Extensions

Petri nets (PN), a well-known model of concurrent computation [21], can naturally be reformulated in a multiset rewriting system operating on nullary predicates only (i.e. predicates with no parameters). This class of rewriting rules corresponds to those in the fragment $\Gamma_{1}$ of CMRS defined in Section 3.2. It is easy to see that, if we associate a predicate symbol to each place of a net, configurations and rules of a $\Gamma_{1}$ model are just alternative representations of markings and transitions of a Petri net. As an immediate consequence of this connection, we have that $L_{c}\left(\Gamma_{1}\right)=L_{c}(P N)$ and $L_{r}\left(\Gamma_{1}\right)=L_{r}(P N)$. To formally compare $\Gamma_{1}$ with the other models, we use some known results on languages accepted by extensions of Petri nets. A lossy Petri net with inhibitor arcs (LN) is a Petri net in which it is possible to test if a place has no tokens and in which tokens may get lost before and after executing a transition. A transfer net [11] (TN) is a Petri net extended with transfer arcs.

Transfer nets (TN) are Petri nets extended with transfer arcs. A transfer arc is a pair $T \hookrightarrow q$ where $T$ is a set of places of the net and $q$ is a place. Given a set of places $P$, let us consider a transition $t=\left(I_{t}, O_{t}\right)$ with transfer $T \hookrightarrow q$. Given a marking $M, t$ is fireable if $I_{t} \leq M$. Its firing leads to the new marking $M^{\prime}$ computed in three steps: we first compute $M_{1}=M-I_{t}$, then we move all tokens in the places in $T$ to the place $q$ obtaining $M_{2}$; finally, we compute $M^{\prime}$ as $M_{2}+O_{t}$.
Reset nets (RN) are Petri nets extended with reset arcs, i.e., with a transfer arc $T \hookrightarrow \perp$ where $\perp$ is a special place used only to reset places.

As an example, let $P=\{p, q, r, s\}$ and consider a transition $t$ with $I=\{p, q\}$, $O=\{p, s\}$, and transfer $\operatorname{arc}\{p, q\} \hookrightarrow r$. Now, consider the marking $M=$ $[p, p, q, q, q]$. Then, the execution of $t$ leads to the marking $M^{\prime}=[p, s, r, r, r]$ (we first compute $M-I=[p, q, q]$, then execute the transfer obtaining $[r, r, r]$, and, finally, add $O$ ). If the transfer arc is instead $\{p, q\} \hookrightarrow \perp$, the execution of $t$ leads to the marking $M^{\prime}=[p, s]$ (marking do not refer to tokens in the special place $\perp)$.

We first notice that the lossy version of RN(TN) (i.e. where tokens can be lost before and after applying the effect of transitions)define the same $c$-languages as RN(TN). We now prove that Lossy TN, Lossy RN and LN recognize the same class of $c$-languages.

- Lossy TN as Lossy RN: Let $P$ and $T$ be the set of places and transitions of a Lossy TN $\mathcal{N}$. We build a lossy RN with places $P$ augmented by place $n$ and $s_{t}$ for each transition $t \in T$. The new places are used to distinguish normal transitions from simulations of the transfer of transition $t$. Consider now a transition $t$ with label $\ell, I_{t}=\left\{p_{1}, \ldots, p_{m}\right\}, O_{t}=\left\{q_{1}, \ldots, q_{n}\right\}$ and transfer $S \hookrightarrow q$ with $S=\left\{r_{1}, \ldots, r_{k}\right\}$. Transition $t$ is simulated via the following set of transitions:
- A transition $t_{0}$ labelled with $\epsilon$ such that $I_{t_{1}}=I \cup\{n\}$ and $O_{t_{1}}=\left\{s_{t}\right\}$. This transition checks if $t$ is fireable and then activates the simulation of its transfer by adding a token to $s_{t}$.
- A set of transitions $t_{1}, \ldots, t_{k}$ labelled with $\epsilon$ such that $I_{t_{i}}=\left\{s_{t}, r_{i}\right\}$ and $O_{t_{i}}=\left\{s_{t}, q\right\}$ for $i: 1, \ldots, k$. Each such transition moves a single token from a places in $S$ to $q$.
- A transition $t^{\prime}$ labelled with $\ell$ such that $I_{t^{\prime}}=\left\{s_{t}\right\}, O_{t^{\prime}}=O_{t} \cup\{n\}$, and with the reset $\operatorname{arc} S \hookrightarrow \perp$. This transition non-deterministically terminates the simulation of the transfer.
- Lossy RN as LN: Given a Lossy RN $\mathcal{N}$, we can build a LPN+0 $\mathcal{N}^{\prime}$ that accepts the same $c$-language simply by replacing each reset arc $T \hookrightarrow \perp$ of a transitions $t$ with an inhibitor $\operatorname{arc} Z_{t}=T$. Indeed, notice that the firing of a transition $t$ with reset $\operatorname{arc} T \hookrightarrow \perp$ in $\mathcal{N}$ at a marking $M$ has the effect of forcing all places in $T$ to be empty in the successor marking of $M$. Now, the corresponding transition $t^{\prime}$ in $\mathcal{N}^{\prime}$ can be fired at $M$ only if each place in $T$ is empty in $M$. However, since $\mathcal{N}^{\prime}$ is lossy this condition can always be verified (all tokens in places in $T$ may get lost) and it has the same effect on $M$ as $t$. Vice versa, if all tokens in places in $T$ get lost, then we can fire $t^{\prime}$ and its firing as the same effect of $t$.
- LN as Lossy TN: Given a LN $\mathcal{N}$, we build a Lossy TN $\mathcal{N}^{\prime}$ that accepts the same $c$-language simply by replacing each inhibitor arc $Z_{t}=T$ of a transitions $t$ with a transition $t^{\prime}$ with a transfer $\operatorname{arc} T \hookrightarrow p_{t}$ where $p_{t}$ is a new place. We assume that tokens in $p_{t}$ can never be re-used (i.e. $p_{t}$ cannot occur in the preset of a transition in $\mathcal{N}^{\prime}$ ). Indeed, notice that, since $\mathcal{N}$ is lossy, the inhibitor $\operatorname{arcs} Z_{t}=T$ in $\mathcal{N}$ are enabled if we first lose all tokens in places in $T$. Thus, the inhibitor arcs have the same effect of a transfer to the new place $p_{t}$ from which tokens can never be re-used. Vice versa, a transition $t$ with transfer of all tokens of places in $T$ to place $p_{t}$ can be simulated by its corresponding transition with inhibitor arc $t^{\prime}$. Indeed, in a lossy step all tokens in places in $T$ may get lost thus enabling the inhibitor arc $Z_{t}$.
- LN as LCS: Given a LN $\mathcal{N}$ with places $P$ and transitions in $T$, we build an LCS $\mathcal{F}$ that accepts the same $c$-language as follows. The LCS $\mathcal{F}$ has messages defined over the singleton set of symbols $\{t\}$. Furthermore, it uses a distinguished channel $c_{p}$ to model each place $p \in P$. Thus, we use a queue $c_{p}$ with $k$ occurrences of $t$ to simulate a place $p$ with $k$ tokens. Notice that we do not need to exploit the FIFO ordering of channels. Based on this idea, the simulation of a transition becomes straightforward. The consumption of a token from place $p$ is simulated by a dequeue operation of message $t$ executed on channel $c_{p}$, the production of a token in place $p$ is simulated by an enqueue operation on channel $c_{p}$, and an inhibitor arc on place $p$ is modelled by the empty test on channel $c_{p}$.

Thus, we have that $L_{c}(L N)=L_{c}(R N)=L_{c}(T N), L_{c}(L N) \subseteq L_{c}(L C S)$, and, as for LCS, $L_{r}(L N)=L_{c}(L N)$.

Furthermore, in [14] the authors proved that $L_{c}(P N) \subset L_{c}(T N)$. From all these properties, we obtain the following result.

Theorem 6. $L_{c}\left(\Gamma_{1}\right) \subset L_{c}\left(\Gamma_{0}\right)$.
For $r$-languages, the classification changes as follows.

Theorem 7. $L_{r}\left(\Gamma_{1}\right) \nsim L_{r}(L C S), L_{r}\left(\Gamma_{1}\right) \nsim L_{r}(L N)$, and $L_{r}\left(\Gamma_{1}\right) \subset L_{r}\left(\Gamma_{0}\right)$.
Proof. We first prove that $L_{r}\left(\Gamma_{1}\right)=L_{r}(P N) \nsubseteq L_{c}(L C S)=L_{r}(L C S)$, hence $L_{r}\left(\Gamma_{1}\right) \nsubseteq L_{c}(L N)=L_{r}(L N)$ since $L_{c}(L N) \subseteq L_{c}(L C S)=L_{r}(L C S)$. Consider the language $L=\left\{a^{n} b^{n} \mid n \geq 0\right\}$. It is easy to verify that there exists a Petri net $\mathcal{N}$ such that $L_{r}(\mathcal{N})=L$. We now prove that $L \notin L_{r}(L C S)$. Per absurdum, suppose there exists an LCS $\mathcal{F}$ such that $L_{c}(\mathcal{F})=L$. For any $k \geq 1$, let $\gamma_{k}$ and $\gamma_{k}^{\prime}$ be two global states s.t. $\gamma_{\text {init }}$ leads to $\gamma_{k}$ by accepting the word $a^{k}, \gamma_{k}$ leads to $\gamma_{k}^{\prime}$ by accepting the word $b^{k}$, and $\gamma_{a c c} \preceq_{l} \gamma_{k}^{\prime}$. Since $\preceq_{l}$ is a well-quasi ordering, there exists $i<j$ such that $\gamma_{i} \preceq_{l} \gamma_{j}$. By monotonicity of $\mathcal{F}$, we have $\gamma_{j}$ leads to $\gamma^{\prime \prime}$ by accepting the word $b^{i}$ and $\gamma_{a c c} \preceq_{l} \gamma_{i}^{\prime} \preceq_{l} \gamma^{\prime \prime}$. We conclude that $a^{j} b^{i} \in L_{c}(\mathcal{F})$ with $i<j$, which gives us a contradiction.

We now prove that $L_{c}(L N) \nsubseteq L_{r}\left(\Gamma_{1}\right)$, hence $L_{c}(L C S) \nsubseteq L_{r}\left(\Gamma_{1}\right)$. Let $\Sigma=$ $\{a, b\}$ and let $L_{p a r}$ be the language over the alphabet $\Sigma \cup\{\#\}$ that contains all the words $w_{1} \# \ldots \# w_{n}$ with $n \geq 0$ such that $w_{i} \in \Sigma^{*}$ and there is no prefix of $w_{i}$ that contains more occurrences of symbol $b$ than those of symbol $a$, for $i: 1 \leq i \leq n$. Notice that the number of occurrences of symbols $a$ and $b$ in $w_{i}$ may be different. The language can be accepted by a $L N$ defined as follows. When we accept the symbol $a$ we add one token in a special place $p_{a}$. To accept the symbol $b$, we remove one token from $p_{a}$. To pass from $w_{i}$ to $w_{i+1}$, we accept symbol \# whenever $p_{a}$ is empty (in LN the empty test is just a reset).
We now show that $L_{p a r}$ cannot be recognized by a Petri net. Suppose that there exists a Petri net $\mathcal{N}$ such that $L_{r}(\mathcal{N})=L_{\text {par }}$. Starting from $\mathcal{N}$, we build a net $\mathcal{N}_{1}$ by adding a new place $d$ that keeps track of the difference between the number of occurrences of symbols $a$ and $b$ in the prefix of the word that is being processed in $\mathcal{N}$. Furthermore, we add the condition that $d$ is empty to the accepting marking of $\mathcal{N}$. It is easy to verify that $\mathcal{N}_{1}$ accepts the language $L_{b a l}$ consisting of words of the form $w=w_{1} \# \cdots \# w_{n}$ where $w_{i}$ belongs the the language of balanced parentheses on the alphabet $\Sigma$ for $i: 1 \leq i \leq n$. We exploit now [15, Lemma 9.8] that states that $L_{b a l}$ cannot be recognized by a Petri net with reachability accepting condition, which gives us a contradiction.

Finally, the property $L_{r}\left(\Gamma_{1}\right)=L_{r}(P N) \subset L_{r}\left(\Gamma_{0}\right)$ follows from [15, Lemma 9.8] and Prop. 7, Indeed, we have that $L_{b a l} \in L_{r}\left(\Gamma_{0}\right)=R E$ and $L_{b a l} \notin L_{r}\left(\Gamma_{1}\right)$.

Finally, we observe that we can use an argument similar to that used in the proof of Theorem 7 to show that $L_{r}(P N) \nsim L_{c}(C M R S)$.

### 7.1 Affine well-structured nets

Affine well-structured nets (aWSNs) [18] are a generalization of Petri nets with black tokens and whole-place operations like reset and transfer arcs [11]. They can also be viewed as a subclass of data nets in which a configuration $s$ is such that $s(d)(p)>0$ only for a specific data $d$ chosen a priori from $D$. Furthermore, all transitions have arity 1 and we can remove from $F_{t}, H_{t}$ and $G_{t}$ all the components in regions different from $S_{1}$, i.e., $F_{t}$ and $H_{t}$ are vectors in $\mathbb{N}^{P}$, and
$G_{t}$ is a matrix in $\mathbb{N}^{P} \times \mathbb{N}^{P}$. $a W S N$ configurations, called markings, are vectors in $\mathbb{N}^{P}$, i.e., finite multisets with symbols in $P$. In the rest of the paper we use $\left[a_{1}, \ldots, a_{n}\right]$ to indicate a multiset with elements $a_{1}, \ldots, a_{n}$. Furthermore, for a multiset $M$, we use $M(a)$ to denote the number of occurrences of $a$ in $M$. An aWSN-transition $t$ is enabled at marking $M$ if $F_{t} \leq M$ where $\leq$ denotes multiset inclusion. The firing of $t$ at $M$ produces a new marking $M^{\prime}=\left(M-F_{t}\right) G_{t}+H_{t}$ where - and + denote resp. multiset difference and union. aWSN are wellstructured with respect to inclusion of markings (vectors in $\mathbb{N}^{P}$ ).

Example 4. The projection of $F_{t}, H_{t}$ and $G_{t}$ in Fig. 1 on $S_{1}$ (i.e. restricted to the single data $d_{1}$ ) gives us the aWSN-transition $t$ with $\alpha_{t}=1$ defined as

$$
F_{t}=\left(\begin{array}{ll}
1 & 0
\end{array}\right) \quad H_{t}=\left(\begin{array}{ll}
0 & 1
\end{array}\right) \quad G_{t}=\left(\begin{array}{ll}
1 & 0 \\
0 & 0
\end{array}\right)
$$

This transition removes a token from $p$ and resets the number of tokens in $q$ to 1, i.e. for $M=\left(m_{1}, m_{2}\right)$ with $m_{1} \geq 1$, it yields $M^{\prime}=\left(m_{1}-1,1\right)$.

We compare now aWSNs and LCSs.
Theorem 8. $L_{c}(a W S N) \subset L_{c}(L C S)$.
Proof. (1) We first prove the inclusion $L_{c}(a W S N) \subseteq L_{c}(L C S)$. Assume an aWSN $W=(P, T, F, G, H)$ with $P=\left\{p_{1}, \ldots, p_{n}\right\}$. We build an LCS $\mathcal{F}=$ $(Q, C, N, \delta)$ such that $L_{c}(W)=L_{c}(\mathcal{F})$. The set of channels is defined as $C=$ $P \cup P^{\prime}$ where $P^{\prime}$ (auxiliary channels) contains a primed copy of each element in $P$. The set of messages $N$ contains the symbol • (a representation of a black token).
Assume that $q_{0} \in Q$ is the initial state of $\mathcal{F}$. Then, a marking $M=\left(m_{1}, \ldots, m_{n}\right)$ is encoded as an LCS configuration $\operatorname{enc}(M)$ with state $q_{0}$ and in which channel $p_{i} \in P$ contains a word $\bullet{ }^{m_{i}}$ containing $m_{i}$ occurrences of symbol $\bullet$ for $i \in \bar{n}$.
For each transition $t$ with label $\ell$, we need to simulate the three steps (subtraction, multiplication, and addition) that correspond to $F_{t}, G_{t}$ and $H_{t}$. Subtraction and addition can be simulated in a straightforward way by removing/adding the necessary number of tokens from/to each channel. The multiplication step is simulated as follows. For each $i \in \bar{n}$, we first make a copy of the content of channel $p_{i}$ in the auxiliary channel $p_{i}^{\prime}$. Each copy is defined by repeatedly moving a symbol from $p_{i}$ to $p_{i}^{\prime}$ and terminates when $p_{i}$ becomes empty. After the copy is terminated for all channels, we start the multiplication step. For each $i \in \bar{n}$, we remove a symbol from $p_{i}^{\prime}$ and add as many symbol to channel $p_{j}$ as specified by $G_{t}\left(p_{i}, p_{j}\right)$ for $j \in \bar{n}$. The analysis terminates when the channels $p_{1}^{\prime}, \ldots, p_{n}^{\prime}$ are all empty. The following properties then hold:
i) We first notice that $M \leq M^{\prime}$ iff $\operatorname{enc}(M) \preceq_{l} \operatorname{enc}\left(M^{\prime}\right)$;
ii) Furthermore, if $M_{0} \stackrel{w}{\Longrightarrow} M_{1}$ in $W$, then $\operatorname{enc}\left(M_{0}\right) \stackrel{w}{\Longrightarrow} \operatorname{enc}\left(M_{1}\right)$ in $\mathcal{F}$;
iii) Finally, since • symbols may get lost in $\mathcal{F}$, if $\operatorname{enc}\left(M_{0}\right) \stackrel{w}{\Longrightarrow} \operatorname{enc}\left(M_{1}\right)$ then there exists $M_{2}$ such that $M_{0} \xrightarrow{w} M_{2}$ and $M_{1} \leq M_{2}$.

If the accepting marking is $M_{f}=\left(m_{1}, \ldots, m_{k}\right)$ then the accepting LCS configuration contains the control state $q_{0}$, the channel $p_{i} \in P$ contains $m_{i}$ symbols $\bullet$, and the channels $p^{\prime} \in P^{\prime}$ are empty. Since we consider languages with coverability acceptance, $L_{c}(W)=L_{c}(\mathcal{F})$ immediately follows from properties (i), (ii), (iii) and Lemma 1.
(2) We prove now that $L_{c}(L C S) \nsubseteq L_{c}(a W S N)$. For this purpose, we exhibit a language in $L_{c}(L C S)$ and prove that it cannot be recognized by any aWSN.

Fix a finite alphabet $\Sigma=\{a, b, \sharp\}$ and let $\mathcal{L}=\left\{w \sharp w^{\prime} \mid w \in\{a, b\}^{*}\right.$ and $w^{\prime} \preceq w$ $w\}$. It is easy to define a LCS that accepts the language $L$ : we first put $w$ in a lossy channel and then remove one-by-one all of its messages. Thus, we have that $\mathcal{L} \in L_{c}(L C S)$. We now prove that there is no aWSN that accepts $\mathcal{L}$. Suppose it is not the case and there exists a aWSN $N$, with (say) $n$ places, that recognizes $\mathcal{L}$ with initial marking $M_{\text {init }}$ and accepting marking $M_{f}$.

For each $w \in\{a, b\}^{*}$, there is a marking $M_{w}$ such that $M_{i n i t} \xlongequal{w \sharp} M_{w} \stackrel{w}{\Longrightarrow} M$ and $M_{f} \leq M$ (otherwise $w \sharp w$ would not be in $L_{c}(N)$ ). Consider the sequences $w_{0}, w_{1}, w_{2}, \ldots$ and $M_{w_{0}}, M_{w_{1}}, M_{w_{2}}, \ldots$ of words and markings defined as follows:
$-w_{0}:=b^{n}$;

- If $M_{w_{i}}=\left(m_{1}, \ldots, m_{n}\right)$ then $w_{i+1}:=a^{m_{1}} b a^{m_{2}} b \cdots b a^{m_{n}}$, for $i=0,2, \ldots$

We observe that $(a) w_{0} \not \varliminf_{w} w_{i}$ for all $i>0$, since $w_{0}$ contains $n$ occurrences of $b$, while $w_{i}$ contains only $n-1$ occurrences of $b$; and (b) for any $i<j, M_{w_{i}} \leq M_{w_{j}}$ iff $w_{i+1} \preceq_{w} w_{j+1}$. By Dickson's lemma [9], there are $i<j$ such that $M_{w_{i}} \leq M_{w_{j}}$. Without loss of generality, we can assume that $j$ is the smallest natural number satisfying this property. Remark that we have that $w_{i} \npreceq_{w} w_{j}$. Indeed, $w_{0} \npreceq{ }_{w} w_{j}$ for any $j>0$ by $(a)$, and in the case of $i>0$ we have by $(b)$ that $w_{i} \not \nwarrow_{w} w_{j}$ since $M_{w_{i-1}} \not \leq M_{w_{j-1}}$. Since $M_{w_{i}} \leq M_{w_{j}}$, by monotonicity of aWSNs, we have that $M_{w_{i}} \xrightarrow{w_{i}} M$ with $M_{f} \leq M$ implies that $M_{w_{j}} \xlongequal{w_{i}} M^{\prime}$ with $M_{f} \leq M \leq M^{\prime}$. Hence, $M_{\text {init }} \stackrel{w_{j} \sharp w_{i}}{\Longrightarrow} M^{\prime}$ and $w_{j} \sharp w_{i} \in L_{c}(N)=\mathcal{L}$, which is a contradiction.

It is interesting to notice that aWSNs can also be simulated by reset nets by using an encoding similar to the one based on LCSs. The emptiness test on a channel is replaced by a reset on the corresponding place. From this observation and from the results in [3], we have the following classification

Proposition 8. $L_{c}($ Petri nets $) \subset L_{c}($ transfer $/$ reset nets $)=L_{c}(a W S N) \subset$ $L_{c}(L C S)$.

This result shows that $c$-language recognized by reset/transfer nets are strictly included in those recognized by LCSs.

## 8 (Integral) Relational Automata

In this section we compare the class of languages accepted by the fragment $\Gamma_{2}$ of CMRS defined in Section 3.2 with those accepted by relational automata [8].

An (integral) relational automaton (RA) operates on a finite set $X$ of positive integer variables, and is of the form $(Q, \delta)$ where $Q$ and $\delta$ are finite sets of control states and transitions respectively. A transition is a triple $\left(q_{1}, o p, q_{2}\right)$ where $q_{1}, q_{2} \in Q$ and $o p$ is of one of the following three operations: (i) reading: read $(x)$ reads a new value of variable $x$ (i.e. assigns a non-deterministically chosen value to $x$ ), (ii) assignment: $x:=y$ assigns the value of variable $y$ to $x$; (ii) testing: $x<y, x=y, x<c, x=c$, and $x>c$ are guards which compare the values of variables $x, y$ and the natural constant $c$. Assume a RA $\mathcal{A}=(Q, \delta)$. A valuation $v$ is a mapping form $X$ to $\mathbb{N}$. A configuration is of the form $(q, v)$, where $q \in Q$ and $v$ is a valuation. We define $\gamma_{\text {init }}$ to be $\left(q_{\text {init }}, v_{\text {init }}\right)$ where $q_{\text {init }} \in Q$ and $v_{\text {init }}(x)=0$ for all $x \in X$. For a transition $\rho \in \delta$ of the form $\left(q_{1}, o p, q_{2}\right)$, we let $\gamma_{1} \xrightarrow{\rho} \gamma_{2}$ if and only if $\gamma_{1}=\left(q_{1}, v_{1}\right), \gamma_{2}=\left(q_{2}, v_{2}\right)$, and one of the following holds: op $=\operatorname{read}(x)$ and $v_{2}(y)=v_{1}(y)$ for each $y \in X-\{x\} ;$ op $=(y:=x)$, $v_{2}(z)=v_{1}(z)$ for each $z \in X-\{y\}$, and $v_{2}(y)=v_{1}(x) ;$ op $=(x<y), v_{2}=v_{1}$, and $v_{1}(x)<v_{1}(y)$. Other testing operations are defined in a similar manner. In [8] Cerans has shown that RA equipped with the sparser-than order of tuples of natural numbers are well-structured. In the case of RA with that order, the coverability accepting condition is equivalent to the control state acceptance, i.e., a word is accepted if it is recognized by an execution ending in a particular control state $q_{a c c} \in Q$.

As stated in the following propositions, RA and $\Gamma_{2}$ define the same class of $c$ - and $r$-languages.

Proposition 9. $L_{c}\left(\Gamma_{2}\right)=L_{c}(R A)$.
Proof. Given an RA $\mathcal{A}=(Q, \delta)$ over the set of variables $X$, we can build the $\Gamma_{2}$ $\mathcal{S}$ defined below. The set of predicate symbols in $\mathcal{S}$ consists of the following: (i) for each $q \in Q$, there is a predicate symbol $q$ in $\mathcal{S}$; and (ii) for each variable $x$ in $X$, there is a predicate symbol $q_{x}$ in $\mathcal{S}$. Transitions in $\delta$ are encoded via the following CMRS rules (with the same labels)

$$
\begin{array}{lll}
\left(q_{1}, \text { read }(x), q_{2}\right) & \Rightarrow & {\left[q_{1}, p_{x}(z)\right]} \\
\left(q_{1}, x:=y, q_{2}\right) & \Rightarrow & {\left[q_{1}, p_{x}(z), p_{y}(w)\right] \sim\left[q_{2}, p_{x}(w)\right]: \operatorname{true}} \\
\left(q_{1}, x<y, q_{2}\right) & \Rightarrow & {\left[q_{1}, p_{x}(z), p_{y}(w)\right] \sim\left[q_{2}, p_{x}(z), p_{y}(w)\right]: \operatorname{true}} \\
& (w)]:\{z<w\}
\end{array}
$$

For $X=\left\{x_{1}, \ldots, x_{n}\right\}$, the initial configuration is $\gamma_{i n i t}=\left[q_{0}, p_{x_{1}}(0), \ldots, p_{x_{n}}(0)\right]$. The accepting configuration $\gamma_{a c c}$ is the multiset [ $q_{a c c}$ ].

For the other inclusion, by using Prop. 2, we assume w.l.o.g. that there is no gap order formula $x<_{c} y$ with $c>0$ in $\mathcal{S}$ and that $\gamma_{a c c}=[o k]$. To justify the second assumption, notice that we can always introduce new predicate symbols $k o$ and $o k$ and a new rule that can be executed only on a configuration $\gamma$ with $\gamma_{a c c} \preceq_{c} \gamma$ and that replace $k o$ with $o k$. All other rules are modified to be enabled only at configurations containing $k o$. Finally, we also observe that we can assume that all configurations of $\mathcal{S}$ have the same size (the size of the initial configuration of the $\Gamma_{2}$ model). Thus, we associate a variable of $X$ to each ground term of the initial CMRS configuration and compose the predicate symbols in a CMRS configuration to form a single control state. CRMS rules can then be simulated
in several steps by operations on variables and updates of control states. To each control state containing $o k$, we add a transition labeled with $\epsilon$ to the accepting control state $q_{a c c}$.

We now prove that $L_{c}\left(\Gamma_{2}\right)$ is the class of regular languages. For this purpose, we first need some preliminary definitions. Given a configuration $\gamma$ with

$$
\operatorname{index}(\gamma)=D_{0} \ldots D_{c \max } d_{0} B_{0} \ldots d_{n} B_{n}
$$

we define

$$
i n d e x^{\prime}(\gamma)=D_{0} \ldots D_{c \max } \sharp B_{0} \ldots B_{n} .
$$

Let us now consider an initial cmax-bounded configuration $\gamma_{i n i t}$ and a final cmax-bounded configuration $\gamma_{a c c}$. The symbolic graph $\mathcal{G}_{\mathcal{S}}$ associated to a $\Gamma_{2}$ specification $\mathcal{S}$ is an automaton $\left(V, \rightarrow_{\mathcal{G}_{\mathcal{S}}}, c_{0}, F\right)$ where
$-V=\left\{\right.$ index $^{\prime}(\gamma)| | \gamma\left|\leq\left|\gamma_{i n i t}\right|\right\} ;$
$-\rightarrow \mathcal{G}_{\mathcal{S}} \subseteq V \times \mathcal{S} \times V$ such that $\forall c_{1}, c_{2} \in V:\left(c_{1}, \rho, c_{2}\right) \in \rightarrow \mathcal{G}_{\mathcal{S}}$ iff there exist two configurations $\gamma_{1}$ and $\gamma_{2}$ such that index $\left(\gamma_{1}\right)=c_{1}, \operatorname{index}^{\prime}\left(\gamma_{2}\right)=c_{2}$ and $\gamma_{1} \xrightarrow{\rho} \gamma_{2} ;$
$-c_{0}=$ index $^{\prime}\left(\gamma_{\text {init }}\right)$;
$-F=\left\{\right.$ index $\left.^{\prime}(\gamma) \in V \mid \gamma_{a c c} \preceq_{c} \gamma\right\}$.
We easily see that $\mathcal{G}_{\mathcal{S}}$ is a finite automata since the number of predicate symbols that appears in states is bounded by the size of $\gamma_{\text {init }}$.
In the following, we use $c \xrightarrow{\rho} \mathcal{G}_{\mathcal{S}} c^{\prime}$ to denote that $\left(c, \rho, c^{\prime}\right) \in \rightarrow \mathcal{G}_{\mathcal{S}}$.
Moreover, given a sequence of rules $w=\rho_{1} \ldots \rho_{l}, c \xrightarrow{w} \mathcal{G}_{\mathcal{S}} c^{\prime}$ denotes that there exists $c_{1}, \ldots, c_{l-1}$ in $V$ such that $c \xrightarrow{\rho_{1}} \mathcal{G}_{\mathcal{S}} c_{1} \ldots \xrightarrow{\rho_{l}} \mathcal{G}_{\mathcal{S}} c_{l-1} \xrightarrow{\rho_{l}} \mathcal{G}_{\mathcal{S}} c^{\prime}$ and $w=\rho_{1} \cdots \rho_{l+1}$.
The next lemma states the main property of $\mathcal{G}_{\mathcal{S}}$ : all the executions of $\mathcal{G}_{\mathcal{S}}$ corresponds to an execution in $\mathcal{S}$ starting from $\gamma_{\text {init }}$.

Lemma 4. If index $\left(\gamma_{\text {init }}\right) \xrightarrow{w} \mathcal{G}_{\mathcal{S}} c$, then $\gamma_{\text {init }} \xrightarrow{w} \gamma$ such that index ${ }^{\prime}(\gamma)=c$.
Proof. The proof is by induction on the number of transitions to reach $c$.
$n=0$ : Immediate.
$n>0$ : Let

$$
\text { index }^{\prime}\left(\gamma_{\text {init }}\right) \xrightarrow{w} \mathcal{G}_{\mathcal{S}} c{\xrightarrow{\rho} \mathcal{G}_{\mathcal{S}}} c^{\prime}
$$

and suppose that $c=D_{0} \ldots D_{c \max } \sharp B_{0} \ldots B_{n}$ and $c^{\prime}=D_{0}^{\prime} \ldots D_{c m a x}^{\prime} \sharp B_{0}^{\prime} \ldots B_{n^{\prime}}^{\prime}$. By ind. hypothesis, we have $\gamma_{\text {init }} \xrightarrow{w} \gamma$ with index ${ }^{\prime}(\gamma)=c$.
Since $c \xrightarrow{\rho} \mathcal{G}_{\mathcal{S}} c^{\prime}$, we know from definition of symbolic graph that there exist two configurations $\gamma_{1}, \gamma_{2}$ such that index $\left(\gamma_{1}\right)=c$, index $\left(\gamma_{2}\right)=c^{\prime}$ and $\gamma_{1} \xrightarrow{\rho} \gamma_{2}$.
Since

$$
i n d e x^{\prime}(\gamma)=i n d e x^{\prime}\left(\gamma_{1}\right)=D_{0} \ldots D_{c \max } \sharp B_{0} \ldots B_{n}
$$

we have

$$
\operatorname{index}(\gamma)=D_{0} \ldots D_{c \max } d_{0} B_{0} \ldots d_{n} B_{n}
$$

and

$$
\operatorname{index}\left(\gamma_{1}\right)=D_{0} \ldots D_{c \max } d_{0}^{\prime} B_{0} \ldots d_{n}^{\prime} B_{n}
$$

Moreover, suppose that

$$
\operatorname{index}\left(\gamma_{2}\right)=D_{0}^{\prime} \ldots D_{c \max }^{\prime} b_{0} B_{0}^{\prime} \ldots b_{n^{\prime}} B_{n^{\prime}}
$$

Following Prop. 2, we have $\gamma_{\text {init }} \xrightarrow{w} \gamma^{\prime}$ such that

$$
\operatorname{index}\left(\gamma^{\prime}\right)=D_{0} \ldots D_{c \max } d_{0}^{\prime \prime} B_{0} \ldots d_{n}^{\prime \prime} B_{n}
$$

and for any $i: 0 \leq i \leq n d_{i}^{\prime \prime} \geq d_{i}+d_{i}^{\prime}$.
Since CMRS are monotonic (so do $\Gamma_{0}$ ) $\gamma_{1} \xrightarrow{\rho} \gamma_{2}$ and $\gamma_{1} \preceq_{c} \gamma^{\prime}$, we have that $\gamma^{\prime} \xrightarrow{\rho}$ $\gamma^{\prime \prime}$ with $\gamma_{2} \preceq_{c} \gamma^{\prime \prime}$. Since the number of ground terms in configurations is bounded, $\gamma_{2} \preceq_{c} \gamma^{\prime \prime}$ implies $\gamma_{2} \prec \gamma^{\prime \prime}$. Thus, we have that index $\left(\gamma^{\prime \prime}\right)=\operatorname{index}^{\prime}\left(\gamma_{2}\right)=c^{\prime}$.

Theorem 9. $L_{c}\left(\Gamma_{2}\right)=$ Regular Languages.
Proof. We first show how to encode a finite automata in $\Gamma_{2}$. The encoding of a finite automaton is direct: each state corresponds to a predicate and CMRS rules mimic the transition relation. Acceptance of words is simulated as follows: for any final state $c$ we have a rule $\{c(x)\} \leadsto\left\{p_{\text {acc }}(x)\right\}:$ true and the final configuration is $\left\{p_{a c c}(0)\right\}$. Finally, the initial configuration is $\left\{c_{0}(0)\right\}$ where $c_{0}$ is the initial state of the automaton.

We now show that all the $c$-languages accepted by a $\Gamma_{2}$ are regular. Consider a $\Gamma_{2} \mathcal{S}$ with an initial cmax-bounded configuration and a final cmax-bounded final configuration. Note that considering cmax-bounded configurations is not restrictive because we can always add a rule $\{p(x)\} \leadsto\{p(x)\}:(x=c)$, i.e. a rule that has no effect, where $c$ is greater than the maximal constant that appears in the initial and final configurations. From Lemma 4 we have that a word accepted by the symbolic graph $\mathcal{G}_{\mathcal{S}}$ corresponds to sequence of rules corresponding to a word accepted by $\mathcal{S}$ (following definition of $\mathcal{G}_{\mathcal{S}}, \gamma_{a c c} \preceq_{c} \gamma$ iff index $\left.(\gamma) \in F\right)$. Moreover, from the definition of $\mathcal{G}_{\mathcal{S}}$ we have

$$
\gamma_{\text {init }} \xrightarrow{\rho_{1}} \gamma_{1} \xrightarrow{\rho_{2}} \ldots \xrightarrow{\rho_{l-1}} \gamma_{l}
$$

implies

$$
\text { index }^{\prime}\left(\gamma_{\text {init }}\right) \xrightarrow{\rho_{1}} \mathcal{G}_{\mathcal{S}} \text { index }^{\prime}\left(\gamma_{1}\right) \xrightarrow{\rho_{2}} \mathcal{G}_{\mathcal{S}} \ldots \xrightarrow{\rho_{l-1}} \mathcal{G}_{\mathcal{S}} \text { index }^{\prime}\left(\gamma_{l}\right)
$$

Furthermore, from definition of accepting states $F, \gamma_{a c c} \preceq_{c} \gamma_{l}$ if and only if index ${ }^{\prime}\left(\gamma_{l}\right) \in F$. Hence, if we replace symbols $\rho$ in $\mathcal{G}_{\mathcal{S}}$ by $\lambda(\rho)$ we conclude that a words $w$ is accepted by $\mathcal{S}$ if and only if $w$ is accepted by $\mathcal{G}_{\mathcal{S}}$.

We are ready now to compare $\Gamma_{2}$ (hence RA) with the other models studied in this paper. For this purpose, we first observe that Petri nets can accept regular languages (finite automata can be encoded as Petri nets). Furthermore, it is straightforward to build a Petri net that accepts a non-regular language like $L=\left\{a^{n} \# b^{m} \mid n \geq m\right\}$. As a consequence of this observation and of Theorem 9, we have the following result.

Corollary 2. $L_{c}\left(\Gamma_{2}\right) \subset L_{c}\left(\Gamma_{1}\right)$.
Let us now consider the reachability accepting condition. We first notice that $L_{c}\left(\Gamma_{2}\right)=L_{r}\left(\Gamma_{2}\right)=L_{c}(R A)=L_{r}(R A)$. Indeed, in both cases of $\Gamma_{2}$ and $R A$ we can encode the reachability acceptance into the coverability acceptance by adding transitions (labelled with $\epsilon$ ) that can be fired only from the accepting configuration and leads to a configuration with control state $q_{a c c}$ in the case of $R A$ and a configuration containing a special accepting predicate symbol in the case of $\Gamma_{2}$. Thus, we have the following property.

Theorem 10. $L_{r}\left(\Gamma_{2}\right) \subset L_{r}\left(\Gamma_{1}\right)$.

## 9 Conclusions

In this paper we have compared wsts by using languages with coverability acceptance as a measure of their expressinevess. From our results we obtain the following classification:

$$
\left(\begin{array}{c}
L_{c}(F A) \\
= \\
L_{c}(R A)
\end{array}\right) \subset L_{c}(P N) \subset\left(\begin{array}{c}
L_{c}(a W S N) \\
= \\
L_{c}(T N) \\
= \\
L_{c}(R N) \\
= \\
L_{c}(L N)
\end{array}\right) \subset L_{c}(L C S) \subset\left(\begin{array}{c}
L_{c}(C M R S) \\
= \\
L_{c}(\text { data nets })
\end{array}\right)
$$

Furthermore, since CMRS and Petri data nets (data nets without whole-place operations) recognize the same class of $c$-languages (coverability in CMRS can be reduced to coverability in Petri data nets [19]) we have that data nets, Petri data nets, and transfer data nets (another subclass of data nets with restrictions on the type of transfers) all define the same class of $c$-languages as CMRS.

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## A Appendix: Technical Lemmas and Propositions

## A. 1 Proof of Proposition 2

The proof is by induction on the size $k$ of the sequence of rules $\rho_{1} \ldots \rho_{k}$.
Base case: $(k=0)$ Since we assume that all the constants that appear in $\gamma_{\text {init }}$ are lesser or equal than cmax, we have that $\operatorname{index}\left(\gamma_{i n i t}\right)$ is of the form $D_{0} \ldots D_{c m a x}$. Hence, the lemma trivially holds.

Induction step: $(k>0)$ Suppose that we have $\gamma_{\text {init }} \xrightarrow{\rho_{1} \ldots \rho_{k-1}} \gamma_{1} \xrightarrow{\rho_{k}} \gamma$ with

$$
\operatorname{index}\left(\gamma_{1}\right)=E_{0} \ldots E_{c m a x} e_{0} F_{0} \ldots e_{m} F_{m}
$$

Suppose that $\gamma$ is built from $\gamma_{1}$ by applying the instance $\rho=L_{1} \leadsto R_{1}$ of $\rho_{k}=L \leadsto R: \psi$. This means that there exists a multi-set of ground terms $\eta$ such that $\gamma_{1}=L_{1}+\eta$ and $\gamma=R_{1}+\eta$.
Under this hypothesis, the multisets in $\operatorname{index}\left(\gamma_{1}\right)$ satisfy the following conditions:

- For any $i: 0 \leq i \leq c m a x$, we have that $E_{i}=G_{i}+E_{i}^{L}$ where
- $E_{i}^{L}$ is the maximal (possibly empty) multiset of predicate symbols with parameter equal to $i$ that occur in $L_{1}$;
- $G_{i}$ is the maximal multiset of predicates with parameter $i$ that are not consumed by $\rho$ (i.e. they also occur in index $(\gamma)$ ).
- For any $i: 0 \leq i \leq m$ and given $v_{i}=c m a x+\Sigma_{0 \leq j \leq i} e_{j}$, we have that $F_{i}=H_{i}+F_{i}^{L}$ where
- $F_{i}^{L}$ is the maximal (possibly empty) multiset of predicate symbols with parameter equal to $v_{i}$ that occur in $L_{1}$;
- $H_{i}$ is the maximal multiset of predicates with parameter $v_{i}$ that are not consumed by $\rho$ (i.e. they also occur in index $(\gamma)$ ).
Let us now suppose that instead of removing $L_{1}$ from $\gamma_{1}$, we add $R_{1}$ to $\gamma_{1}$. The resulting configuration $\gamma_{2}=\gamma_{1}+R_{1}$ has index

$$
\operatorname{index}\left(\gamma_{2}\right)=E_{0}^{\prime} \ldots E_{c \max }^{\prime} \sigma_{0} c_{0} F_{0}^{\prime} \ldots \sigma_{m} c_{m} F_{m}^{\prime} \sigma_{m+1}
$$

with

$$
\forall 0 \leq i \leq m+1: \sigma_{i}=c_{0}^{i} K_{0}^{i} \ldots c_{n_{i}}^{i} K_{n_{i}}^{i}
$$

where (assuming $\forall 0 \leq k \leq m: e_{k}^{\prime}=c_{k}+\Sigma_{0 \leq j \leq n_{k}} c_{j}^{k}$ )

- For any $i: 0 \leq i \leq c m a x$, we have that

$$
E_{i}^{\prime}=G_{i}+E_{i}^{L}+E_{i}^{R}
$$

where $E_{i}^{R}$ is the maximal (possibly empty) multiset of predicate symbols with parameter equal to $i$ that occur in $R_{1}$;

- for any $i: 0 \leq i \leq m+1$, for any $j: 0 \leq j \leq n_{i}, K_{j}^{i}$ is the maximal (possibly empty) multiset of predicate symbols that occur in $R_{1}$ with parameter equal to
- Furthermore, for any $i: 0 \leq i \leq m$ :

$$
F_{i}^{\prime}=G_{i}+F_{i}^{L}+F_{i}^{R}
$$

where $F_{i}^{R}$ is the maximal (possibly empty) multiset of predicate symbols with parameter equal to $c \max +\Sigma_{0 \leq j \leq i} e_{i}^{\prime}$ that occur in $R_{1}$.
Intuitively, $\sigma_{i}$ represent the structure added to $\operatorname{index}\left(\gamma_{1}\right)$ by $R_{1}$ for what concerned all predicate symbols with a parameter $v$ not directly represented in index $\left(\gamma_{1}\right)$, i.e., such that

$$
c \max +\Sigma_{0 \leq j \leq i-1} e_{j}<v<c \max +\Sigma_{0 \leq j \leq i} e_{j} .
$$

The sequence

$$
\varsigma=\left(E_{0}^{L}+E_{0}^{R}\right) \ldots\left(E_{c \max }^{L}+E_{c \max }^{R}\right) \sigma_{0} c_{0}\left(F_{0}^{L}+F_{0}^{R}\right) \ldots \sigma_{m} c_{m}\left(F_{m}^{L}+F_{m}^{R}\right) \sigma_{m+1}
$$

can be transformed into index $\left(L_{1}+R_{1}\right)$ by removing all empty multisets and summing up constants in order to correctly maintain gaps between non-empty multisets of predicates.

To simplify the presentation, let us assume that $\varsigma$ coincides with index $\left(L_{1}+\right.$ $R_{1}$ ). The following can be easily extended to the general case. We now observe that $\rho$ corresponds to an instance of a specialization $\rho^{\prime}$ of $\rho_{k}$ in which the variables in $\rho_{k}$ are totally ordered w.r.t. $<\cup=$. In other words, from $\varsigma$ we can reconstruct the constraint $\psi^{\prime}$ of $\rho^{\prime}$ as follows:

- To each non-empty multiset $M$ in $\varsigma$ we associate a distinct variable $x_{M}$, each predicate in $M$ takes $x_{M}$ as parameter in $\rho^{\prime}$;
- For each non empty multiset $M=E_{i}^{L}+E_{i}^{R}$ with $0 \leq i \leq c m a x$, we associate the condition $x_{M}=i$ in $\psi^{\prime}$;
- If $M$ and $M^{\prime}$ are two consecutive multisets in $\varsigma ~(a n d ~ o c c u r s ~ b e f o r e ~ M ') ~$ with a constant $c$ between them then $\rho^{\prime}$ contains the gap-order constraint $x_{M}<{ }_{c} x_{M^{\prime}}$.

Since the condition $\psi^{\prime}$ of $\rho^{\prime}$ corresponds to one of the possible linearizations of the condition of $\rho_{k}$, every instance of $\rho^{\prime}$ is also an instance of $\rho_{k}$. Furthermore, $\psi^{\prime}$ in $\rho^{\prime}$ represents the minimal gap-order constraints extracted from $\varsigma$ which is compatible with $\psi$ (i.e. $\psi \wedge \psi^{\prime} \equiv \psi^{\prime}$ ). This implies that any other instance $L_{2} \leadsto R_{2}$ of $\rho^{\prime}$ can be represented by a sequence
$\varsigma_{1}=\left(E_{0}^{L}+E_{0}^{R}\right) \ldots\left(E_{c m a x}^{L}+E_{c m a x}^{R}\right) \sigma_{0}^{\prime} f_{0}\left(F_{0}^{L}+F_{0}^{R}\right) \ldots \sigma_{m}^{\prime} f_{m}\left(F_{m}^{L}+F_{m}^{R}\right) \sigma_{m+1}^{\prime}$
where the following conditions are satisfied:

- for any $i: 0 \leq i \leq m$ we have $f_{i} \geq c_{i}$,
- for any $i: 0 \leq \bar{i} \leq m+1, \sigma_{i}^{\prime}=f_{0}^{i} K_{0}^{i} \ldots f_{n_{i}}^{i} K_{n_{i}}^{i}$ such that $f_{j}^{i} \geq c_{j}^{i}$ for all $j: 0 \leq j \leq n_{i}$.

Fixed a given instance $L_{2} \leadsto R_{2}$ of $\rho^{\prime}$ with associated sequence $\varsigma_{1}$, we define the new sequence

$$
\varsigma_{2}=S_{0} \ldots S_{c m a x} \sigma_{0}^{\prime} f_{0} T_{0} \ldots \sigma_{m}^{\prime} f_{m} T_{m} \sigma_{m+1}
$$

with $S_{i}=E_{i}^{L}+E_{i}^{R}+G_{i}$ for $i: 0 \leq i \leq c m a x$ and $T_{i}=F_{i}^{L}+F_{i}^{R}+H_{i}$ for $i: 0 \leq i \leq m$.
Following the definition of index and $\prec$, this sequence corresponds to the index of a configuration $\gamma_{2}^{\prime}$ such that $\gamma+R_{1}=\gamma_{2} \prec \gamma_{2}^{\prime}$. Now let us define the values

$$
e_{i}^{\prime \prime}=f_{i}+\Sigma_{0 \leq j \leq n_{i}} f_{j}^{i}
$$

for $i: 0 \leq i \leq m$. Furthermore, let us define the sequence

$$
\varsigma_{3}=\left(E_{0}^{L}+G_{0}\right) \ldots\left(E_{c \max }^{L}+G_{0}\right) e_{0}^{\prime \prime}\left(F_{0}^{L}+H_{0}\right) \ldots e_{m}^{\prime \prime}\left(F_{m}^{L}+H_{m}\right)
$$

Again, following definition of index and $\prec$, there exists $\gamma_{1}^{\prime}$ such that $\varsigma_{3}=$ index $\left(\gamma_{1}^{\prime}\right)$ and $\gamma_{1} \prec \gamma_{1}^{\prime}$.

Now we note that $\gamma_{2}^{\prime}$ corresponds to $\gamma_{1}^{\prime}+R_{2}$. This implies that the instance $L_{2} \leadsto R_{2}$ of $\rho$ can be applied at $\gamma_{1}^{\prime}$.

If we now define $\gamma=\gamma_{2}-L_{1}$ and $\gamma^{\prime \prime}=\gamma_{2}^{\prime}-L_{2}$, then we have $\gamma \prec \gamma^{\prime \prime}$. Indeed, $\operatorname{index}(\gamma)$ and $\operatorname{index}\left(\gamma^{\prime \prime}\right)$ are obtained by removing predicate symbols in multiset occurring in the same position in index $\left(\gamma_{2}\right)$ and $\operatorname{index}\left(\gamma_{2}^{\prime}\right)$, respectively.
Furthermore, we have that $\gamma_{1}^{\prime} \xrightarrow{\rho^{\prime}} \gamma^{\prime \prime}$. Finally, note that for any sequence $\varsigma_{1}$ there exists an instance $\rho^{\prime \prime}$ of $\rho^{\prime}$ (the specialization of $\rho_{k}$ we consider). Hence, there exists an instance $\rho^{\prime \prime}$ such that $\gamma^{\prime \prime}=\gamma^{\prime}$.

By applying the inductive hypothesis, we have that there exists an execution $\gamma_{\text {init }} \xrightarrow{\rho_{1} \ldots \rho_{k-1}} \gamma_{1}^{\prime}$ that subsumes $\gamma_{i n i t} \xrightarrow{\rho_{1} \ldots \rho_{k-1}} \gamma_{1}$ such that $\gamma_{1}^{\prime} \xrightarrow{\rho_{k}} \gamma^{\prime}$. We conclude that there exists an execution $\gamma_{\text {init }} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma^{\prime}$ that subsumes $\gamma_{\text {init }} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma$.

## A. 2 Proof of Lemma 2

If the length of the execution is one the thesis trivially holds because $\gamma_{\text {init }}$ is cmax-bounded. Now suppose that

$$
\gamma_{i n i t} \xrightarrow{\rho_{1}} \gamma_{1} \ldots \xrightarrow{\rho_{k}} \gamma_{k} \xrightarrow{\rho} \gamma
$$

and $\gamma$ is not linear. Suppose that $L_{1} \leadsto R_{1}$ is the instance of $\rho$ applied to $\gamma_{k}$ to obtain $\gamma$. As in the proof of Prop. 2, we have that

$$
\operatorname{index}\left(\gamma_{k}\right)=E_{0} \ldots E_{c \max } e_{0}\left(F_{0}^{L}+G_{0}\right) \ldots e_{m}\left(F_{m}^{L}+G_{m}\right)
$$

where $F_{i}^{L}$ are the predicate symbols of terms in $L_{1}$ with parameter cmax + $\Sigma_{0 \leq j \leq i} e_{j}$ for $i: 0 \leq i \leq m$.
Now let us consider $\gamma_{k}+R_{1}$. Then we have that

$$
\operatorname{index}\left(\gamma_{k}+R_{1}\right)=E_{0}^{\prime} \ldots E_{c m a x}^{\prime} \sigma_{0} c_{0} T_{0} \ldots \sigma_{m} c_{m} T_{m} \sigma_{m+1}
$$

where

- $E_{i}^{\prime}=E_{i}+E_{i}^{R}$ where $E_{i}^{R}$ are the predicate symbols of terms in $R_{1}$ with parameter $i$ for $i: 0 \leq i \leq$ cmax;
$-T_{i}=F_{i}^{L}+F_{i}^{R}+G_{i}$ where $F_{i}^{R}$ are the predicate symbols of terms in $R_{1}$ with parameter $c \max +\Sigma_{0 \leq j \leq i} e_{j}$ for $i: 0 \leq i \leq m$; and
$-\sigma_{i}$ is a sub-sequence

$$
c_{0}^{i} K_{0}^{i} \ldots c_{n_{i}}^{i} K_{n_{i}}^{i}
$$

that represents terms with new values added by $R_{1}$ for $i: 0 \leq i \leq m+1$.
Since $\gamma$ is not linear, there is a multiset $F_{r}^{R}+G_{r}$ or a multiset $K_{j}^{i}$ that contains at least two predicates $p$ and $q$ with the same parameter say $v$. Let us suppose such a multiset $F_{r}^{R}+G_{r}$. The case where a multi-set $K_{j}^{i}$ contains at least two predicates is treated is a similar way. Since by hypothesis $F_{r}^{L}+G_{r}$ contains at most one symbol and by the syntactic restriction of $\Gamma_{0}$, we have that (at least) one between $p(v)$ and $q(v)$ is produced by a valuation to a variable in $\rho$ which is not involved in $=$ constraints.

Following Prop. 2, for any $\gamma_{k}^{\prime}$ with $\gamma_{k} \prec \gamma_{k}^{\prime}$ we know that there exists a linear execution from $\gamma_{\text {init }}$ to $\gamma_{k}^{\prime}$ with the same rules $\rho_{1}, \ldots, \rho_{k}$ and passing through $\gamma_{1}^{\prime}, \ldots, \gamma_{k-1}^{\prime}$ such that $\gamma_{i} \prec \gamma_{i}^{\prime}$ for $i: 1 \leq i \leq k$. This implies that we can choose $\gamma_{k}^{\prime}$ and instance $L_{2} \leadsto R_{2}$ of $\rho_{k}$ such that
$\operatorname{index}\left(\gamma_{k}^{\prime}+R_{2}\right)=E_{0}^{\prime} \ldots E_{c m a x}^{\prime} \sigma_{0} c_{0} T_{0} \ldots \sigma_{r} c_{r} T_{r} \sigma_{r+1}^{\prime} c_{r+1}^{\prime} T_{r+1} \ldots \sigma_{m}^{\prime} c_{m}^{\prime} T_{m} \sigma_{m^{\prime}+1}$
where for any $j>r c_{j}^{\prime}$ and the constants in $\sigma_{j}^{\prime}$ are strictly greater than $c_{j}$ and the values in $\sigma_{j}$, respectively (i.e. we "shift to the right" all values greater than $v)$.
Now notice that in a $\Gamma_{0}$ rule it is not possible to impose the equalities over more than two parameters. Furthermore, when imposing equality of two paramters of ground terms, one ground term is removed by the rule and the second one is
added to configurations by the rule. Hence, there is no constraints that impose that the parameter of $p(v)$ and $q(v)$ must be equal. W.l.o.g. we assume that tere is no constraint that impose that the parameter $v$ of $p(v)$ must be equal to another parameter. This means that $\rho$ remains applicable to $\gamma_{k}^{\prime}$ whenever the evaluation for the argument of predicate $p$ is the value $v^{\prime}=v+1$.
With this new instance $L_{3} \leadsto R_{3}$ of $\rho$ we have that
$\operatorname{index}\left(\gamma_{k}^{\prime}+R_{3}\right)=E_{0}^{\prime} \ldots E_{c m a x}^{\prime} \sigma_{0} c_{0} T_{0} \ldots \sigma_{r} c_{r} T_{r} 1[p] \sigma_{r+1}^{\prime \prime} c_{r+1}^{\prime} T_{r+1} \ldots \sigma_{m}^{\prime} c_{m}^{\prime} T_{m} \sigma_{m^{\prime}+1}$
where $\sigma_{r+1}^{\prime \prime}$ is obtained from $\sigma_{r+1}^{\prime}$ by decrementing by 1 the first constant that appears.

We conclude by noticing that from $\gamma_{k}^{\prime}+R_{3}$ we can compute $\gamma^{\prime}$ by removing $L_{3}$. This operation maintains the same structure of the index of $\gamma_{k}^{\prime}+R_{3}$ for what concerns predicate $[p]$, i.e. assuming that $v=c \max +\sum_{j=1 . . i} d_{i}$ (remember $d_{i}$ are constants of index $(\gamma)$ ) we have

$$
\operatorname{index}\left(\gamma^{\prime}\right)=D_{0} \ldots D_{c \max } d_{0} B_{0} \ldots d_{i} B_{i} 1[p] d_{i+1}^{\prime} B_{i+1} \ldots d_{n}^{\prime} B_{n}
$$

such that $\forall i+1 \leq j \leq n: d_{j}^{\prime} \geq d_{j}$.

## A. 3 Proof of Lemma 3

We prove by induction on the number of transitions that $\gamma_{1} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma_{3}$ implies there exists $\gamma_{4}$ such that $\gamma_{2} \xrightarrow{\rho_{1} \ldots \rho_{k}} \gamma_{4}$ with either $\gamma_{3} \prec \gamma_{4}$ or $\gamma_{4}$ is a linearisation of $\gamma_{3}$.

Base case: $(k=1)$ Suppose that the lemma does not hold. Let $L_{1} \leadsto R_{1}$ be the instance of $\rho_{1}$ that allows to build $\gamma_{3}$ from $\gamma_{1}$, i.e. $\gamma_{3}=\gamma_{1}-L_{1}+R_{1}$. Suppose that $\gamma_{1}=\gamma_{1}^{\prime}+[p(v)], \gamma_{2}=\gamma_{2}^{\prime}+\left[p\left(v^{\prime}\right)\right]$ and $\gamma_{1}^{\prime} \prec \gamma_{2}^{\prime}$. In other words, the predicate $p$ is "isolated" in the index of $\gamma_{2}$. We consider two cases: either $L_{1} \leq \gamma_{1}^{\prime}$ or not, i.e. the instance $L_{1} \leadsto R_{1}$ does not remove $p(v)$ or it does.

In the case of $L_{1} \leq \gamma_{1}^{\prime}$, let

$$
\operatorname{index}\left(L_{1}+R_{1}\right)=E_{1} \ldots E_{\text {cmax }} e_{0} H_{0} \ldots e_{r} H_{r}
$$

To any sequence,

$$
\varsigma=E_{1} \ldots E_{c \max } e_{0}^{\prime} H_{0} \ldots e_{r}^{\prime} H_{r}
$$

such that $\forall 0 \leq i \leq r: e_{i}^{\prime} \geq e_{r}^{\prime}$ correspond another instance $L_{2} \leadsto R_{2}$ of $\rho_{1}$. Furthermore, since $\gamma_{2}$ is a linearization of $\gamma_{1}$, all the constants that appear in index $\left(\gamma_{2}\right)$ are greater than the corresponding ones in index $\left(\gamma_{1}\right)$. Hence, there exists an instance $L_{3} \leadsto R_{3}$ of $\rho_{1}$ that has the same effect on the structure of index $\left(\gamma_{2}\right)$ than the instance $L_{1} \leadsto R_{1}$ on $\operatorname{index}\left(\gamma_{1}\right)$, i.e. predicates are removed from and added to the same multi-sets and the same sequences of multi-sets (interleaved with constants) are added at the same point into index $\left(\gamma_{2}\right)$. Hence, $\gamma_{2} \xrightarrow{\rho_{1}} \gamma_{4}=\gamma_{2}-L_{3}+R_{3}$ and $\gamma_{4}$ is a linearisation of $\gamma_{3}$.

In the second case, i.e. $L_{1} \not \leq \gamma_{1}^{\prime}$ and $p(v)$ is removed from $\gamma_{1}$ when applying $L_{1} \leadsto R_{1}$, let

$$
\operatorname{index}\left(L_{1}+R_{1}\right)=E_{1} \ldots E_{c \max } e_{0} H_{0} \ldots e_{i} H_{i}+[p] e_{i+1} H_{i+1} \ldots e_{r} H_{r}
$$

such that $v=c \max +\sum_{j=0 . . i} e_{i}$. Following the syntactic restriction of $\Gamma_{0}$, either (i) $\rho_{1}$ imposes no equality constraint between the parameter of $p(v)$ and the parameter of another ground term $q\left(v^{\prime}\right)$, or (ii) $\rho_{1}$ imposes such an equality constraint on the parameters of $p(v)$ and $q\left(v^{\prime}\right)$ which is added by $\rho_{1}$. In case $(i)$, the sequence

$$
\varsigma=E_{1} \ldots E_{c \max } e_{0} H_{0} \ldots e_{i} H_{i} 1[p] e_{i+1} H_{i+1} \ldots e_{r} H_{r}
$$

corresponds to another instance $L_{2} \leadsto R_{2}$ of $\rho_{1}$, i.e. $\operatorname{index}\left(L_{2}+R_{2}\right)=\varsigma$, since the gap-orders between predicates defined by $L_{1}+R_{1}$ are not violated. Furthermore, if we increase the gap orders defined by $\varsigma$ we still obtain a sequence that corresponds to an instance of $\rho_{1}$. Hence, any sequence

$$
\varsigma^{\prime}=E_{1} \ldots E_{c m a x} e_{0}^{\prime} H_{0} \ldots e_{i}^{\prime} H_{i} e[p] e_{i+1}^{\prime} H_{i+1} \ldots e_{r}^{\prime} H_{r}
$$

such that $e \geq 1$ and $e_{i}^{\prime} \geq e_{i}$ for any $i: 0 \leq i \leq r$ corresponds to an instance of $\rho_{1}$. Furthermore, since $\gamma_{2}$ is a linearisation of $\gamma_{1}$, all the constants that appear in index $\left(\gamma_{2}\right)$ are greater than the corresponding ones in index $\left(\gamma_{1}\right)$. Hence, there exists a instance $L_{3} \leadsto R_{3}$ of $\rho_{1}$ that has the same effect on the structure of index $\left(\gamma_{2}\right)$ than $L_{1} \leadsto R_{1}$ on index $\left(\gamma_{1}\right)$, except that $L_{3} \leadsto R_{3}$ removes the multiset $[p]$ that corresponds to the ground term $p(v)$. We conclude that $\gamma_{2} \xrightarrow{\rho_{1}} \gamma_{4}=$ $\gamma_{2}-L_{3}+R_{3}$ and $\gamma_{3} \prec \gamma_{4}$.

In case (ii), the sequence

$$
\varsigma=E_{1} \ldots E_{c \max } e_{0} H_{0} \ldots e_{i} H_{i} 1[p, q] e_{i+1} H_{i+1} \ldots e_{r} H_{r}
$$

corresponds to another instance $L_{2} \leadsto R_{2}$ of $\rho_{1}$, i.e. $\operatorname{index}\left(L_{2}+R_{2}\right)=\varsigma$, since the gap-orders (and equality between parameters of $p$ and $q$ ) defined by $L_{1}+R_{1}$ are not violated. Again, any sequence

$$
\varsigma^{\prime}=E_{1} \ldots E_{c \max } e_{0}^{\prime} H_{0} \ldots e_{i}^{\prime} H_{i} e[p, q] e_{i+1}^{\prime} H_{i+1} \ldots e_{r}^{\prime} H_{r}
$$

such that $e \geq 1$ and $e_{i}^{\prime} \geq e_{i}$ for any $i: 0 \leq i \leq r$ corresponds to an instance of $\rho_{1}$. Furthermore, since $\gamma_{2}$ is a linearisation of $\gamma_{1}$, all the constants that appear in index $\left(\gamma_{2}\right)$ are greater than the corresponding ones in index $\left(\gamma_{1}\right)$. Hence, there exists a instance $L_{3} \leadsto R_{3}$ of $\rho_{1}$ that has the same effect on the structure of index $\left(\gamma_{2}\right)$ than $L_{1} \leadsto R_{1}$ on index $\left(\gamma_{1}\right)$, except that $L_{3} \leadsto R_{3}$ replaces the multi-set $[p]$ that corresponds to the ground terms $p\left(v^{\prime}\right)$ by the multi-set $[q]$. We conclude that $\gamma_{2} \xrightarrow{\rho_{1}} \gamma_{4}=\gamma_{2}-L_{3}+R_{3}$ and $\gamma_{4}$ is a linearisation of $\gamma_{3}$.

Induction Step: $(k>1)$ By induction Hypothesis, we have that $\gamma_{1} \xrightarrow{\rho_{1} \ldots \rho_{k-1}}$ $\gamma_{3}^{\prime}$ implies there exists $\gamma_{4}^{\prime}$ such that $\gamma_{2} \xrightarrow{\rho_{1} \ldots \rho_{k-1}} \gamma_{4}^{\prime}$ with either $\gamma_{3}^{\prime} \prec \gamma_{4}^{\prime}$ or $\gamma_{4}^{\prime}$ is a linearisation of $\gamma_{3}^{\prime}$.

In the case where $\gamma_{4}^{\prime}$ is a linearisation of $\gamma_{3}^{\prime}$, we apply the same resoning than in the base base.

In the case where $\gamma_{3}^{\prime} \prec \gamma_{4}^{\prime}$, let $\gamma_{3}^{\prime} \xrightarrow{\rho_{k}} \gamma_{3}$ and $L_{1} \leadsto R_{1}$ be the instance of $\rho_{k}$ used to build $\gamma_{3}$ from $\gamma_{3}^{\prime}$, i.e. $\gamma_{3}=\gamma_{3}^{\prime}-L_{1}+R_{3}$. Consider

$$
\operatorname{index}\left(L_{1}+R_{1}\right)=E_{1} \ldots E_{c m a x} e_{0} H_{0} \ldots e_{r} E_{r}
$$

To any sequence

$$
\varsigma=E_{1} \ldots E_{c \max } e_{0}^{\prime} H_{0} \ldots e_{r}^{\prime} E_{r}
$$

with $\forall 0 \leq i \leq r: e_{i} \leq e_{i}^{\prime}$ corresponds to another instance of $\rho_{k}$ where the gap-orders between parameters of ground terms increase. Hence, there exists an instance $L_{2} \leadsto R_{2}$ of $\rho_{k}$ such that $\gamma_{4}=\gamma_{4}^{\prime}-L_{2}+R_{2}$ and $\gamma_{3} \prec \gamma_{4}$.
(a)Subtraction

(b) Copy


Fig. 9. Simulation of: (a) subtraction and selection of indexes; (b) creation of a new configuration.
(a) Transfer from Si to $\mathrm{Si} / \mathrm{Rj}$

(b) Transfer from Ri to Ri/Sj

(c) Addition


Fig. 10. Simulation of: (a) transfer from $S_{i}$ to $R_{j} / S_{j}$; (b) transfer from $R_{i}$ to $R_{i} / S_{j}$; (c) addition.


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