# Observability of Place/Transition Nets

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Abstract—In this paper, we discuss the problem of estimating the marking of a place/transition (P/T) net based on event observation. We assume that the net structure is known while the initial marking is totally or partially unknown. We give algorithms to compute a marking estimate that is a lower bound of the actual marking. The special structure of Petri nets allows us to use a simple linear algebraic formalism for estimate and error computation. The error between actual marking and estimate is a monotonically nonincreasing function of the observed word length, and words that lead to null error are said complete. We define several observability properties related to the existence of complete words, and show how they can be proved. To prove some of them, we also introduce a useful tool, the observer coverability graph, i.e., the usual coverability graph of a P/T net augmented with a vector that keeps track of the estimation error on each place of the net. Finally, we show how the estimate generated by the observer may be used to design a state feedback controller for forbidden marking speci-

Index Terms—Discrete-event systems, observers, Petri nets, state estimation.

#### I. INTRODUCTION

THIS PAPER deals with the problem of estimating the marking of a place/transition (P/T) net based on the observation of transition firings and presents a set of analytical tools to determine several observability properties. An observer constructed following this approach can also be used in a state-feedback control loop, as discussed in the final part of the paper.

This framework provides a useful paradigm that can be applied to different settings, from discrete-event control, to failure diagnosis and error recovery. The assumption that only event occurrences, i.e., transition firings, may be observed—while the plant state, i.e., the marking, cannot—is common in discrete event control. The assumption that the state of the plant is not known (or is only partially known) is natural during error recovery. Consider for instance the case of a plant remotely controlled: if the communication fails the state may evolve and when the communication is re-established the state will be at best partially known. In a manufacturing environment, one may consider the case in which resources (i.e., tokens) enter unobserved, or in which we know how many resources have entered the system but not their exact location.

When the structure and the initial marking of a P/T net is known, the knowledge of the transition firings is sufficient to

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reconstruct the marking that each new firing yields. In this work we assume that only the net structure is known and consider two cases: 1) the initial marking is not known and 2) the initial marking is know to belong to a "macromarking," i.e., we know the token content of subsets of places but not the exact token distribution. Case 2) can in effect be seen as a generalization of case 1), but we preferred to handle the two cases separately.

In both cases, we show how it is possible to estimate the actual marking of the net based on the observation of a word of events (i.e., transition firings) and we give algorithms for computing the estimates and, in case 2), bounds on the error. The estimate is always a lower bound of the actual marking. The system that computes the estimate is called an observer.

The special structure of Petri nets allows us to use a simple linear algebraic formalism for estimate and error computation. In particular, the set of markings consistent with an observed word, i.e., the set of markings in which the system may actually be given the observed word, can be easily characterized as a convex set of integers.

The error function between the actual marking and the estimate can be shown to be a monotonically nonincreasing function of the observed word length. Observed words that lead to a null error are said to be "complete." Complete observers are the discrete-event counterpart of asymptotic observers for time-driven systems.

In this paper, we define several observability properties and show that they are decidable. In particular, we consider two main properties. Marking observability (MO) means that there exists at least one word that is complete, while strong marking observability (SMO) means that all words can be completed in a finite number of steps into a complete word.

We set up a hierarchy considering the possibility that the two properties are satisfied by a net N starting from an initial marking  $M_0$ , by a net N starting from any marking M reachable from an initial marking  $M_0$  (uniform observability) or by a net N starting from any marking in  $\mathbb{N}^m$  (structural observability) where m is the number of places of the net.

To prove some of these properties we introduce a useful tool, the *observer coverability graph*, i.e., the usual coverability graph of a P/T net augmented with a vector that keeps track of the estimation error on each place of the net.

All the considered properties can be proven either by the use of the observer coverability graph, or by reducing them to other decision problems (e.g., home-space properties, marking reachability, existence of repetitive sequences) that can be checked using algorithms well known in the literature and that will not be further discussed in this paper. Although more efficient algorithms may exist to prove all observability properties, our main aim here is to characterize these properties and to show that they are decidable.

Finally, we show how the estimate generated by the observer may be used to design a state feedback controller, that ensures that the controlled system never enters a set of forbidden states. In the discussion, we assume that all events are controllable and we focus on a special class of specifications that limit the weighted sum of markings in subset of places. Clearly, the use of marking estimates (as opposed to the exact knowledge of the actual marking of the plant) leads to a worse performance of the closed-loop system in the sense that to rule out the possibility that the plant enter a forbidden marking, the controller may prevent the firing of transitions whose firing is perfectly legal given the actual marking of the plant.

# A. Relevant Literature

Observability is a fundamental property that has received a lot of attention in the framework of time-driven systems, given the importance of reconstructing plant states that cannot be measured. Although less popular in the case of discrete-event systems, the issue of state estimation and of control under partial state observation has been discussed in the literature.

For systems represented as finite automata, Ramadge [21] was the first to show how an observer could be designed for a partially observed system.

Caines *et al.* [3] showed how it is possible to use the information contained in the past sequence of observations (given as a sequence of observation states and control inputs) to compute the set of consistent states. In [4], the observer output is used to steer the state of the plant to a desired terminal state. The approach in [3] is based on the construction of an observer tree to determine the set of markings consistent with the observed behavior: the tree contains all consistent markings. A similar approach was also used by Kumar *et al.* [11] when defining observer based dynamic controllers in the framework of supervisory predicate control problems.

Özveren and Willsky [18] proposed an approach for building observers that allows one to reconstruct the state of finite automata after a word of bounded length has been observed, showing that an observer may have an exponential number of states.

The main drawback of the automata-based approach is the requirement that the set of consistent markings must explicitly be enumerated. It may be useful to pinpoint that on the contrary the procedure we present in this paper simply produces an estimate of the state, while the special structure of a Petri net permits to determine, using linear algebraic tools, if a given marking is consistent without the explicit enumeration of the (possibly infinite) consistent set.

Finally, as an example in which state estimators have been used for fault diagnosis in systems represented as finite state machines, we mention the work of Wang and Schwartz [25]. The purpose of the observer was in this framework not that of reconstructing the state of the system, but rather that of detecting if the system is faulty and recognizing the fault type.

Very few works dealt with observability in Petri nets. As far as we know, the first ones were [7] and [9] where the preliminary concepts discussed in this paper have been introduced.

Meda and Ramírez [15] used Interpreted Petri nets to model the system and the observer. The main idea is to start the observer with a marking bigger than the real one and then eliminate some tokens until the observer and system markings are equal. Interpreted Petri nets have also been used by Ramirez-Treviño *et al.* in [22] where it was shown that observability defined as in [15] is equivalent to observability in [9] and it was shown how to construct an observer for binary Interpreted Petri nets when the observability property is verified.

The issue of controlling a plant with incomplete (state or event) measurements has also been discussed in the discrete event control literature.

Zhang and Holloway [26] used a controlled Petri net model for forbidden state avoidance under partial *event* observation with the assumption that the initial marking be known.

The use of state-feedback control under partial *state* observation has been discussed by Li and Wonham [13], [14] and by Takai *et al.* [23]. In these papers, the partial observation is due to a static mask, that maps the plant state space into an observation space. The main focus was in finding necessary and sufficient conditions for the existence of "optimal" state feedback control laws given a mask (optimal means that the resulting closed-loop behavior is the same for the controller with mask and the controller with complete state observation).

Unlike the aforementioned approach, the setting we deal with in this paper assumes that the mask is induced by the computed estimate, and it changes as the plant evolves. Initially, when the estimate is crude, it is often the case that these restrictive "optimal" conditions are not verified. We propose a control scheme that tries to make the best use of the available estimate to ensure the correct behavior of the plant under control.

The notion of macromarking used in this paper is similar to the notion of *uncertain marking*, as described by Cardoso *et al.* [5]. These authors considered high-level nets as models of manufacturing systems and they assumed as theoretical basis of uncertain markings possibilistic logic. In our approach, on the contrary, we restrict our attention to purely logical models (P/T nets) and we assume that all possible markings have equal probability.

### II. BACKGROUND

In this section, we provide some basic definitions that will be used in the following of the paper. We first recall some basic terminology on Petri nets, we then provide the definition of both linear and semilinear sets and we recall the main results on decidability of home-space property.

### A. Petri Nets

In this section, we recall the Petri net formalism used in this paper. For a more comprehensive introduction to Petri nets see [17]. A P/T net is a structure N=(P,T,Pre,Post), where P is a set of m places; T is a set of n transitions;  $Pre:P\times T\to \mathbb{N}$  and  $Post:P\times T\to \mathbb{N}$  are the pre- and post-incidence functions that specify the arcs. The incidence matrix of the net is defined as C(p,t)=Post(p,t)-Pre(p,t).

We define  $p^{\bullet} = \{t \in T \mid Pre(p,t) > 0\}$  as the set of output transitions of place p.

A marking is a vector  $M: P \to \mathbb{N}$  that assigns to each place of a P/T net a nonnegative number of tokens, represented by black

dots. A P/T system or net system  $\langle N, M_0 \rangle$  is a net N with an initial marking  $M_0$ .

A transition t is enabled at M if  $M \ge Pre(\cdot, t)$  and may fire yielding the marking  $M' = M + C(\cdot, t)$ . We write M[w]M' to denote that the enabled sequence of transitions w may fire at Myielding M' or, equivalently, we use the notation M' = w(M)and  $M = w^{-1}(M')$ . Moreover, we denote  $w(M_0) = M_w$ . Finally, we denote as  $w_0$  the sequence of null length. The set of all sequences finable in  $\langle N, M_0 \rangle$  is denoted  $L(N, M_0)$  (this is also called the prefix-closed free language of the net). If the firing sequence w is enabled at  $M_0$ , we also say that w is a word in  $L(N, M_0)$ .

Let  $w = t_{\alpha_1}, t_{\alpha_2}, \dots, t_{\alpha_k}$  be a sequence in  $L(N, M_0)$ . The sequence  $w_i = t_{\alpha_1}, \dots, t_{\alpha_i}$  with  $i \in \mathbb{N}$  and  $i \leq k$  is a prefix of w of length i and we write  $w_i \leq w$ .

A marking M is reachable in  $\langle N, M_0 \rangle$  iff there exists a firing sequence w such that  $M_0[w]M$ . The set of all markings reachable from  $M_0$  defines the reachability set of  $\langle N, M_0 \rangle$  and is denoted  $R(N, M_0)$ .

A repetitive sequence w is such that M[w]M' with  $M' \ge M$ . Then,  $\forall i \geq 1, w^i$  is enabled at M. A repetitive sequence w is said to be nonstationary if M[w]M' with  $M' \geq M$ : such a sequence strictly increases the token count of one or more places.

Three useful elementary facts about Petri nets that will be used in this paper are the following.

Fact 1:

- i)  $M \leq M' \Longrightarrow L(N,M) \subseteq L(N,M')$ .
- ii) If w is enabled at M and M' then: M M' = w(M) w(M)
- iii) The reachability set  $R(N, M_0)$  is infinite iff there exists a nonstationary repetitive sequence in  $L(N, M_0)$ .

Finally, we denote  $\vec{0}_m$  ( $\vec{1}_m$ ) a  $m \times 1$  vector of zeros (ones).

## B. Home-Space Property

Linear and semilinear sets were first introduced in [19] in order to study some problems from formal language theory.

Definition 2: We say that  $\mathcal{E} \subseteq \mathbb{N}^m$  is a linear set if there exists some  $V \in \mathbb{N}^m$  and a finite set  $\{V_1, \dots, V_n\} \subseteq \mathbb{N}^m$  such

$$\mathcal{E} = \left\{ V' \in \mathbb{N}^m | V' = V + \sum_{i=1}^n k_i V_i \text{ with } k_i \in \mathbb{N} \right\}.$$

V is called the base of  $\mathcal{E}$ , and  $V_1, \ldots, V_n$  are called its periods. A semilinear set is the finite union of a family of linear sets. A first result regarding decidability is the following.

Theorem 3 [6]: Given a net system  $\langle N, M_0 \rangle$  and a semilinear set  $\mathcal{E}$  it is decidable if  $R(N, M_0) \cap \mathcal{E} = \emptyset$ .

Finally, we introduce the definition of home space [16] and an important theorem that will be used when proving some properties of estimates.

Definition 4 [16]: Let  $\mathcal{HS}$  be a set of markings. We say that  $\mathcal{HS}$  is a home space of a P/T net  $\langle N, M_0 \rangle$  iff  $\forall M \in R(N, M_0)$ ,  $\exists M' \in \mathcal{HS}$  such that  $M' \in R(N,M)$ . If  $\mathcal{HS}$  is a singleton, we call its unique element a home state.

Thus, a set of markings  $\mathcal{HS}$  is a home space if from any reachable marking it is possible to reach some marking in  $\mathcal{HS}$ . Furthermore, there exist special classes of sets for which the home state property is decidable.

Theorem 5 [16]: The property of being a home space for finite unions of linear sets having the same periods is decidable.

#### III. MARKING ESTIMATION

In this section, we present an algorithm for estimating the state of a net system  $\langle N, M_0 \rangle$  whose marking cannot be directly observed under the following assumptions:

- A2) the structure of the net N = (P, T, Pre, Post) is known, while the initial marking  $M_0$  is not;
- A3) the event occurrences (i.e., the transition firings) can be observed.

After w has been observed we define, the set  $\mathcal{M}(w)$  of w consistent markings as the set of all markings in which the system may be given the observed behavior.

Definition 6: Given an observed word w, the set of w consistent markings is  $\mathcal{M}(w) = \{M \mid \exists M' \in \mathbb{N}^m, M'[w] \}M\}$ .

Given an evolution of the net  $M_{w_0}[t_{\alpha_1}\rangle M_{w_1}[t_{\alpha_2}\rangle \cdots$ , we use the following algorithm to compute the estimate  $\mu_{w_i}$  of each actual marking  $M_{w_i}$  based on the observation of the word of events  $w_i = t_{\alpha_1} t_{\alpha_2} \cdots t_{\alpha_i}$ .

Algorithm 7 (Marking Estimation With Event Observation)

- 1. Let the initial estimate be  $\mu_{w_0} = \vec{0}_m$  .
- **2.** Let i = 1.
- **3.** Wait until  $t_{lpha_i}$  fires.
- **4.** Update the estimate  $\mu_{w_{i-1}}$  to  $\mu'_{w_i}$  with

$$\mu'_{w_i}(p) = \max\{\mu_{w_{i-1}}(p), Pre(p, t_{\alpha_i})\}.$$

- 5. Let  $\mu_{w_i} = \mu'_{w_i} + C(\cdot, t_{\alpha_i})$ . 6. Let i = i+1.
- 7. Goto 3.

Note that in step 4. of the algorithm we update the previously computed estimate  $\mu_{w_{i-1}}$ , since the firing of  $t_{\alpha_i}$  implies that  $M_{w_{i-1}} \geq Pre(\cdot, t_{\alpha_i})$ . In the following, we will always denote the estimate computed by this algorithm after having observed the word w as  $\mu_w$ .

The estimate computed by Algorithm 7 is a lower bound of the actual marking of the net.

Proposition 8: Let  $w = t_{\alpha_1} t_{\alpha_2} \cdots \in L(N, M_0)$  be an observed string and  $w_i$  its prefix of length i. Then

$$\forall i, \quad \mu_{w_i} \le \mu'_{w_{i+1}} \le M_{w_i}.$$

Proof: Clearly  $\mu_{w_i} \leq \mu'_{w_{i+1}}$  for all i. Also  $t_{\alpha_1}$  is enabled at  $M_0 \equiv M_{w_0}$ , hence  $M_{w_0} \geq Pre(\cdot, t_{\alpha_1}) = \mu'_{w_1}$ . By induction, assume  $\mu_{w_{i-1}} \leq \mu'_{w_i} \leq M_{w_{i-1}}$ . Then  $\mu_{w_i} = \mu'_{w_i} + C(\cdot, t_{\alpha_i}) \leq M_{w_{i-1}} + C(\cdot, t_{\alpha_i}) = M_{w_i}$ . Finally,  $t_{\alpha_{i+1}}$  is enabled at  $M_{w_i}$ , hence  $M_{w_i} \geq Pre(\cdot, t_{\alpha_{i+1}})$ . This implies that  $M_{w_i}(p) - \mu'_{w_{i+1}}(p) = M_{w_i}(p) - \mu_{w_i}(p) \ge 0 \text{ if } \mu'_{w_{i+1}}(p) = \mu_{w_i}(p) \text{ while } M_{w_i}(p) - \mu'_{w_{i+1}}(p) = M_{w_i}(p) - Pre(p, t_{\alpha_{i+1}}) \ge 0$ 0 if  $\mu'_{w_{i+1}}(p) > \mu_{w_i}(p)$ .

It is possible to give an easy characterization of the set of consistent markings in terms of estimate. Let us first consider

the following lemma that states that the minimal initial marking enabling a sequence w on a net N is  $w^{-1}(\mu_w)$ .

Lemma 9: Let  $w=t_{\alpha_1}t_{\alpha_2}\cdots t_{\alpha_k}$  be a sequence of transitions of a net N. Then,  $w\in L(N,M_0)$  if and only if  $M_0\geq w^{-1}(\mu_w)\equiv \mu_w-\sum_{i=1}^k C(\cdot,t_{\alpha_i})$ .

 $\begin{array}{lll} \textit{Proof:} & (\text{if) Let } \sigma_i &=& t_{\alpha_i}t_{\alpha_{i+1}}\cdots t_{\alpha_k} \text{ be the suffix} \\ \text{of } w \text{ of length } k+1-i. \text{ We will show that for all } \\ i, & \sigma_i^{-1}(\mu_w) &\equiv& \mu_w - \sum_{j=i}^k C(\cdot,t_{\alpha_j}) &\geq& Pre(\cdot,t_{\alpha_i}). \\ \text{This implies that } w &\in& L(N,w^{-1}(\mu_w)), \text{ hence (by fact } \\ 1.i), &\forall& M_0 &\geq& w^{-1}(\mu_w), &w &\in& L(N,M_0). \text{ Clearly,} \\ \sigma_k^{-1}(\mu_w) &=& \mu_w - C(\cdot,t_{\alpha_k}) &=& \mu'_{w_k} &\geq& Pre(\cdot,t_{\alpha_k}). \text{ By induction, assume now that } \\ \sigma_i^{-1}(\mu_w) &=& \sigma_{i+1}^{-1}(\mu_w) - C(\cdot,t_{\alpha_i}) &\geq& \mu'_{w_{i+1}} - C(\cdot,t_{\alpha_{i+1}}). \\ \text{Then, } \sigma_i^{-1}(\mu_w) &=& \sigma_{i+1}^{-1}(\mu_w) - C(\cdot,t_{\alpha_i}) &\geq& \mu'_{w_{i+1}} - C(\cdot,t_{\alpha_i}) &\geq& \mu_{w_i} - C(\cdot,t_{\alpha_i}) &=& \mu'_{w_i} &\geq& Pre(\cdot,t_{\alpha_i}). \end{array}$ 

(only if) By contradiction: Let  $w \in L(N, M_0)$  with  $M_0(p) < w^{-1}(\mu_w)(p)$ . Then (by fact 1.ii),  $w(M_0)(p) < \mu_w(p)$  and this violates Proposition 8.

This lemma leads to the following theorem.

Theorem 10: Given an observed word  $w \in L(N, M_0)$  and the corresponding estimated marking  $\mu_w$  computed by Algorithm 7, the set of w consistent markings is

$$\mathcal{M}(w) = \{ M \in \mathbb{N}^m \mid M \ge \mu_w \}.$$

It is also possible to define a meaningful measure of the place estimation error, as the token difference between a marking and its estimate in a given place.

Definition 11: Let us consider a place  $p \in P$  and an observed word  $w \in L(N, M_0)$ . Let  $M_w$  and  $\mu_w$  be the corresponding marking and its estimate. The place estimation error in p is  $e_p(M_w, \mu_w) = M_w(p) - \mu_w(p)$  and its update after the firing of t is  $e_p(M_w, \mu'_{wt}) = M_w(p) - \mu'_{wt}(p)$ .

Analogously, it is possible to define a measure of the estimation error, as the token difference between a marking and its estimate.

Definition 12: Given a marking  $M_w$  and its estimate  $\mu_w$ , the estimation error is  $e(M_w,\mu_w)=\sum_{p\in P}e_p(M_w,\mu_w)=\vec{1}_m^T\cdot (M_w-\mu_w)$  and its update after the firing of t is  $e(M_w,\mu'_{wt})=\vec{1}_m^T\cdot (M_w-\mu'_{wt})$ .

Note that the place estimation error is a monotonically non-increasing function of the observed word length.

Proposition 13: Let  $w = t_{\alpha_1} t_{\alpha_2} \cdots \in L(N, M_0)$  be an observed word and  $w_i$  its prefix of length i. Then,  $\forall i$  and  $\forall p$ 

$$e_p(M_{w_i}, \mu_{w_i}) \ge e_p(M_{w_i}, \mu'_{w_{i+1}}) = e_p(M_{w_{i+1}}, \mu_{w_{i+1}})$$
 (1)

and

$$e_p(M_{w_i}, \mu'_{w_{i+1}}) = \min \{e_p(M_{w_i}, \mu_{w_i}), M_{w_i} - Pre(p, t_{\alpha_{i+1}})\}.$$
 (2)

*Proof:* To prove the first statement, we observe that by Proposition 8,  $\mu_{w_i}(p) \leq \mu'_{w_{i+1}}(p) \leq M_{w_i}(p)$ , hence  $e_p(M_{w_i}, \mu_{w_i}) \geq e_p(M_{w_i}, \mu'_{w_{i+1}})$ . Also  $e_p(M_{w_i}, \mu'_{w_{i+1}}) = (M_{w_i}(p) - \mu'_{w_{i+1}}(p)) = (M_{w_i}(p) + C(p, t_{\alpha_{i+1}}) - \mu'_{w_{i+1}}(p) - C(p, t_{\alpha_{i+1}})) = (M_{w_{i+1}}(p) - \mu_{w_{i+1}}(p)) = e_p(M_{w_{i+1}}, \mu_{w_{i+1}})$ .

To prove the second statement, we observe that  $e_p(M_{w_i}, \mu'_{w_{i+1}}) = M_{w_i}(p) - \mu'_{w_{i+1}}(p) = M_{w_i}(p) - \max\{\mu_{w_i}(p), Pre(p, t_{\alpha_{i+1}})\} = \min\{M_{w_i}(p) - \mu_{w_i}(p), M_{w_i}(p) - Pre(p, t_{\alpha_{i+1}})\} = \min\{e_p(M_{w_i}, \mu_{w_i}), M_{w_i}(p) - Pre(p, t_{\alpha_{i+1}})\}.$ 

Thus, it follows that also the estimation error is a monotonically nonincreasing function of the observed word length.

Proposition 14: Let  $w=t_{\alpha_1}t_{\alpha_2}\cdots\in L(N,M_0)$  be an observed word,  $w_i$  the prefix of w of length i, and  $\mu_{w_i}$  and  $\mu'_{w_i}$  the estimate and the updated estimate of  $M_{w_i}$ . Then,  $\forall i$ 

$$e(M_{w_i}, \mu_{w_i}) \ge e(M_{w_i}, \mu'_{w_{i+1}}) = e(M_{w_{i+1}}, \mu_{w_{i+1}}).$$

*Proof:* This immediately follows from Proposition 13.  $\square$ 

## IV. PROPERTIES OF ESTIMATES

It is natural to ask under which conditions the estimated marking computed by Algorithm 7 converges to the actual marking. This motivated us to define the following properties.

Definition 15: Given a net system  $\langle N, M_0 \rangle$ , and a place  $p \in P$ , we say that a word  $w \in L(N, M_0)$  is

- p-complete if  $e_p(M_w, \mu_w) = 0$ , i.e., if  $\mu_w(p) = M_w(p)$ ;
- marking complete if w is p complete for all  $p \in P$ .

Thus, a marking-complete word allows one to reconstruct the actual marking of the net.

Based on this, we can define these properties of a net system. Definition 16: A net system  $\langle N, M_0 \rangle$  is:

- marking observable (MO) if there exists a marking complete word  $w \in L(N, M_0)$ ;
- strongly marking observable (SMO) in k steps if
  - 1)  $\forall w \in L(N, M_0)$  such that  $|w| \ge k$ , w is marking complete;
  - 2)  $\forall w \in L(N, M_0)$  such that |w| < k, either w is marking complete or  $\exists t \in T$  such that  $M_0[wt)$ .

In the aforementioned definitions, we note that the observability properties depend not only on the net structure N, but also on the initial marking  $M_0$ , that we assume is unknown. Thus, it may seem that those properties have little significance per se. In effect, we will use the characterization of MO and SMO to define two more general properties that have greater significance.

Definition 17: A net system  $\langle N, M_0 \rangle$  is

- uniformly marking observable (uMO) if  $\forall M \in R(N, M_0), \langle N, M \rangle$  is MO;
- uniformly strongly marking observable (uSMO) in k steps if  $\forall M \in R(N, M_0), \langle N, M \rangle$  is SMO in k steps.

The properties of uMO and uSMO are important if we consider the following problem: we consider a system whose initial marking  $M_0$  is known. Due to a communication failure the system evolves unobserved. When the communication is re-established, we can only be sure that the actual marking belongs to the set  $R(N, M_0)$ . We want to know if the marking can be reconstructed starting from any of these reachable markings.

Definition 18: A net N is

• structurally marking observable (sMO) if it is MO for any initial marking  $M_0 \in \mathbb{N}^m$ ;

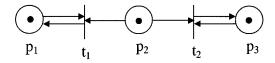


Fig. 1. A net system that is not MO, but whose places are MO.

• structurally strongly marking observable (sSMO) in k steps if  $\langle N, M_0 \rangle$  is SMO in a number of steps k (that depends on  $M_0 \in \mathbb{N}^m$ ).

The properties of sMO and sSMO are even more general and only depend on the net structure N.

The aforementioned properties are related as shown in the following partial-order diagram:

Here, sMO  $\longrightarrow$  uMO means that if a net N is sMO then  $\langle N, M_0 \rangle$  is uMO for all initial markings  $M_0$ . By means of simple counterexamples, it is possible to prove that properties not in a partial-order relationship are uncorrelated.

Note that sometimes, only the marking of a subset of places can be reconstructed, thus making it necessary to introduce the following definition.

Definition 19: Given a net system  $\langle N, M_0 \rangle$ , a place  $p \in P$  is

- MO if there exists a p-complete word  $w \in L(N, M_0)$ ;
- SMO in  $k_p$  steps (where  $k_p$  depends on the place p) if:
  - 1)  $\forall w \in L(N, M_0)$  such that  $|w| \ge k_p$ , w is p-complete;
  - 2)  $\forall w \in L(N, M_0)$  such that  $|w| < k_p$ , either w is p complete or  $\exists t \in T$  such that  $M_0[wt)$ .

Analogously, we can extend to a single place p all the properties of estimates previously defined for a net (system), namely, uMO, uSMO, sMO, and sSMO. For the sake of brevity, we omit formal definitions. We highlight, however, the following implications:

- $\forall p, p \text{ is MO} \iff \langle N, M_0 \rangle \text{ is MO};$
- $\forall p, p \text{ is SMO (uMO, uSMO, sMO, sSMO)} \iff \langle N, M_0 \rangle$  is SMO (uMO, uSMO, sMO, sSMO).

Note that the first one only holds in one sense. In fact, even if all places are observable, this does not imply that there exists one sequence that reconstructs the marking of all places.

Example 20: Let us consider the net system  $\langle N, M_0 \rangle$  in Fig. 1. All places are MO but the net system is not MO. In fact, if  $t_1$  fires, we reconstruct the marking of places  $p_1$  and  $p_2$ , but the net reaches a dead marking, thus making it impossible to reconstruct the marking of place  $p_3$ . Analogously, the firing of  $t_2$  enables us to reconstruct the actual marking of places  $p_2$  and  $p_3$ , but it produces a deadlock, thus not enabling the reconstruction of the marking in  $p_1$ .

### V. OBSERVER COVERABILITY GRAPH

In this section, we show how to construct an *observer coverability tree* and the corresponding *observer coverability graph* (*OCG*) to represent both the set of reachable markings of a net system and the error of the estimate computed in accordance

with Algorithm 7. More precisely, each node of the OCG contains a vector M covering a marking of the net and an upper bound error vector  $u \in \mathbb{N}^m$ . We will show that the OCG is useful to prove observability properties.

Algorithm 21 (Observer Coverability Tree) 1. Let  $u_0=M_0$ . Label the initial node  $(M_0/u_0)$  as the root and tag it "new".

- 2. If "new" nodes exist, select a new node (M/u) and:
- 2.1. If (M/u) is identical to a node labeled "old" then tag (M/u) "old" and go to step 2.
- 2.2. If no transitions are enabled at M, tag (M/u) "dead" and go to step 2.
- 2.3. For each transition t enabled at M do the following:
- 2.3.1.  $\forall~p~\in~P$ , if  $M(p)=\omega$  then let  $\tilde{M}(p)=M(p)$  and  $\tilde{u}(p)=u(p)$ , else let  $\tilde{M}(p)=M(p)+C(p,t)$  and  $\tilde{u}(p)=\min\{u(p),M(p)-Pre(p,t)\}$ ;
- 2.3.2. on the path from the root to (M/u) if there exists a marking  $\overline{M} \leq \tilde{M}$  and  $\tilde{M} \neq \overline{M}$ , i.e.,  $\overline{M}$  is covered by  $\tilde{M}$ , then let  $\tilde{M}(p) = \omega$  for each p such that  $\tilde{M}(p) > \overline{M}(p)$ ;
- 2.3.3. introduce  $(\dot{M}/\tilde{u})$  as a node, draw an arc with label t from (M/u) to  $(\tilde{M}/\tilde{u})$ , and tag  $(\tilde{M}/\tilde{u})$  "new".
- 2.4 Tag (M/u) "old" and go to step 2.

Note that its construction follows the well-known rules of a coverability tree for a P/T net [17]. Also, we note that the error-bound vector u is set to the actual error for the root node and then it is updated as we add new nodes. Note, however, that whenever we reach a marking whose component M(p) is  $\omega$ , the error bound u(p) is not updated any more (see step 2.3.1).

The observer coverability graph of a Petri net  $\langle N, M_0 \rangle$  is a labeled directed graph  $\mathcal{G} = (V, E)$  with transition function  $\delta: V \times E \to V$ . Its node set V is the set of all distinct labeled nodes in the observer coverability tree, and each arc in E is labeled with a transition t to represent a firing such that  $\delta((M/u),t)=(M'/u')$ , where (M/u) and (M'/u') are in V. Note that in the OCG all tags used in the construction of the observer coverability tree are omitted.

We will also represent the initial marking by a round corner box, while a thick box represents a marking whose estimation error bound vector is  $u = \vec{0}_m$ .

Example 22: Let us consider the net systems in Fig. 2 and their OCG. Since the two nets are unbounded, in both cases  $\omega$  appears. The OCG of a bounded net is shown in Fig. 3.

Let us demonstrate that the OCG of a P/T net has a finite number of nodes.

Property 23: Let  $\mathcal{G}$  be the OCG of  $\langle N, M_0 \rangle$ . The number of nodes in  $\mathcal{G}$  is bounded by  $v = v' \cdot \prod_{p \in P} (M_0(p) + 1)$  where v' is the number of nodes in the usual coverability graph of  $\langle N, M_0 \rangle$ .

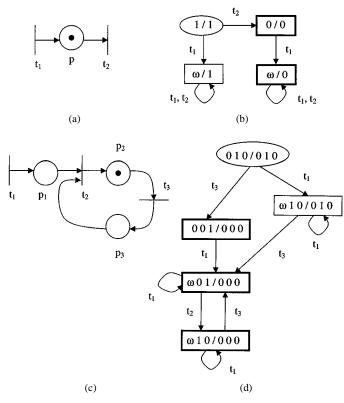


Fig. 2. Unbounded Petri nets and their observer coverability graphs.

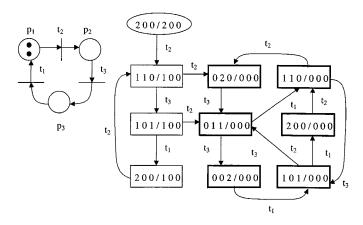


Fig. 3. Bounded Petri net and its observer coverability graph.

*Proof:* By virtue of Algorithm 21, the initial error bound vector is equal to the initial estimate, i.e.,  $u_0 = M_0$ . Moreover, by Proposition 14 the place estimation error is a monotonically nonincreasing function of the observed word length, thus, the estimation error in the generic place p may assume at most  $M_0(p) + 1$  different values. It follows that the number of nodes in  $\mathcal G$  is limited by the number of nodes v' in the coverability graph times  $\prod_{p \in P} (M_0(p) + 1)$ .

Apart from the well known properties that can be studied through the coverability graph of a P/T net [17], the OCG enables us to study some more interesting properties.

Proposition 24: Let  $\mathcal G$  be the OCG of  $\langle N, M_0 \rangle$ . Given  $w \in L(N, M_0)$  consider the node (M/u) reached on the graph executing w, i.e., let  $(M/u) = \delta((M_0/u_0), w)$ . The following holds.

i) The place estimation error  $e_p(M_w, \mu_w) \in [\ell(p), u(p)]$  where u(p) is the component of u corresponding to place p and

$$\ell(p) = \begin{cases} u(p), & \text{if } M(p) \neq \omega \\ 0, & \text{if } M(p) = \omega. \end{cases}$$

ii) The estimation error

$$e(M_w, \mu_w) \in \left[ \sum_{p \in P} \ell(p), \sum_{p \in P} u(p) \right].$$

*Proof:* We prove this by induction on the length of w.

i) When  $w\equiv w_0$ , i.e., w is a word of null length,  $(M/u)=(M_0/u_0)$ , and  $e_p(M_{w_0},\mu_{w_0})=M_0(p)-0=M_0(p)\equiv u_0(p)$ .

Assume that the property i) holds for a word  $w' \in L(N, M_0)$  and let  $\delta((M_0/u_0), w') = (M'/u')$ . Let t be an enabled transition at  $M_{w'}$  and w = w't: in the OCG there will be a transition  $\delta((M'/u'), t) = (M/u)$ . We can consider two cases.

If 
$$M(p) \neq \omega$$
, then  $M_{w'} = M'(p) \neq \omega$  and

$$e_p(M_w, \mu_w) = \min\{e_p(M_{w'}, \mu_{w'}), M_{w'}(p) - Pre(p, t)\}\$$
  
= \min\{u'(p), M'(p) - Pre(p, t)\} = u(p).

where the first equality derives from (1), the second one from the induction hypothesis, and the third one from step 2.3.1 of Algorithm 21.

If 
$$M(p) = \omega$$
, then

$$e_p(M_w, \mu_w) = \min\{e_p(M_{w'}, \mu_{w'}), M_{w'}(p) - Pre(p, t)\}\$$
  
 $\leq e_p(M_{w'}, \mu_{w'}) \leq u'(p) = u(p).$ 

where the last inequality derives from the induction hypothesis, and the last equality from step 2.3.1 of Algorithm 21.

ii) Immediately follows from the previous item.

Example 25: Let us consider again the net system in Fig. 2(a) and its OCG. The estimation error relative to the node labeled with  $(\omega/1)$  may be either null or unitary. If we consider  $w=t_1t_2t_2$  then  $e_p(M_w,\mu_w)=0$ , thus on the OCG we read an upper bound of the estimation error. On the contrary,  $e_p(M_w,\mu_w)=1$  is the exact estimation error for all words w such that  $\forall w' \leq w$ ,  $|w'|_{t_1} \geq |w|_{t_2}$ .

Now, let us consider the net system in Fig. 2(b). Here, every node with label  $M(p_1)=\omega$  is also characterized by  $u(p_1)=0$ , i.e., the upper bound on the place estimation error in  $p_1$  is null. Therefore, in this case in each node of the OCG we can read the actual place estimation error in  $p_1$ .

Finally, in the example in Fig. 3 no  $\omega$  appears in  $\mathcal G$  being the net bounded, thus, in each node u is the exact estimation error vector.

### VI. PROPERTIES ANALYSIS

In this section, we discuss in detail the observability problem. In particular, we provide necessary and sufficient conditions to characterize the properties defined above and we also prove that all these properties are decidable. The OCG is a useful tool when dealing with some analysis problems.

## A. Word Completeness

An elementary necessary and sufficient condition for completeness of a word is the following.

Proposition 26: A word  $w \in L(N, M_0)$  is: i) p-complete iff  $M_0(p) = w^{-1}(\mu_w)(p)$ ; ii) marking complete iff  $M_0 = w^{-1}(\mu_w)$ .

*Proof:* The word w is p-complete (resp., marking complete) if and only if  $M_w(p) = \mu_w(p)$  (respectively,  $M_w = \mu_w$ ), i.e., if and only if  $M_0(p) = w^{-1}(M_w)(p) = w^{-1}(\mu_w)(p)$  (resp.,  $M_0 = w^{-1}(M_w) = w^{-1}(\mu_w)$ ).

Example 27: Let us consider the net system in Fig. 3. The word  $w=t_2$  is not marking complete because  $\mu_{t_2}=\begin{bmatrix}0&1&0\end{bmatrix}^T$  and  $t_2^{-1}(\mu_{t_2})=\begin{bmatrix}1&0&0\end{bmatrix}^T\lneq M_0$ . On the contrary, the word  $w=t_2t_2$  is marking complete because  $\mu_{t_2t_2}=\begin{bmatrix}0&2&0\end{bmatrix}^T$  and  $\begin{bmatrix}t_2t_2\end{bmatrix}^{-1}(\mu_{t_2t_2})=\begin{bmatrix}2&0&0\end{bmatrix}^T=M_0$ .

Another semidecision procedure for completeness can be given using the OCG.

Proposition 28: Let us consider a net system  $\langle N, M_0 \rangle$  and its OCG  $\mathcal G$ . Let (M/u) be the node in  $\mathcal G$  reached executing  $w \in L(N, M_0)$ , i.e.,  $(M/u) = \delta((M_0/u_0), w)$  and let us consider a place  $p \in P$ .

- i) If u(p) = 0 (resp.,  $u = \vec{0}_m$ ), then w is p-complete (respectively, marking complete).
- ii) If  $M(p) \neq \omega$  and  $u(p) \neq 0$ , then w is not p complete, hence, it is not marking complete.

*Proof:* It follows from proposition 24.  $\square$ 

Note that the OCG provides necessary and sufficient conditions for the completeness of a word only in the case of bounded P/T nets, when  $\omega$  does not appear in the graph. On the contrary, it only provides two distinct sufficient or necessary conditions for the completeness of a word in the case of unbounded nets.

Example 29: Let us consider again the bounded net system in Fig. 3. The OCG allows one to say that the word  $w=t_2$  is not marking complete because its execution leads to  $(1\ 1\ 0/1\ 0\ 0)$ , while the word  $w=t_2t_2$  is marking complete because its execution leads to  $(0\ 2\ 0/0\ 0\ 0)$ .

Let us consider the unbounded net system in Fig. 2(a). If we consider  $w=t_1t_2t_2$ , w is complete but this is not deducible from the OCG because its execution leads to  $(\omega/1)$ .

Theorem 30: Let  $\langle N, M_0 \rangle$  be a net system and w a word in  $L(N, M_0)$ . It is decidable whether w is marking complete wrt to  $\langle N, M_0 \rangle$ .

*Proof:* It follows from Proposition 26, because it is sufficient to determine  $\mu_w$  using Algorithm 7, and then compute  $w^{-1}(\mu_w)$ .

# B. Observability

We first provide a necessary and sufficient condition for marking observability.

Proposition 31: The net system  $\langle N, M_0 \rangle$  is marking observable iff

$$L(N, M_0) \supsetneq \bigcup_{\overline{M}_0 \le M_0} L(N, \overline{M}_0).$$

*Proof:* In general,  $L(N, M_0) \supseteq \bigcup_{\overline{M}_0 \subseteq M_0} L(N, \overline{M}_0)$ . We prove that the system is not observable iff the equality holds.

In fact  $L(N, M_0) = \bigcup_{\overline{M}_0 \lneq M_0} L(N, \overline{M}_0) \iff \forall w \in L(N, M_0), \exists \overline{M}_0 \lneq M_0$  such that  $w \in L(N, \overline{M}_0) \iff$  (by Lemma 9)  $\forall w \in L(N, M_0), \exists \overline{M}_0$  such that  $M_0 \ngeq \overline{M}_0 \ge w^{-1}(\mu_w) \iff$  (by Proposition 26)  $\forall w \in L(N, M_0), w$  is not complete  $\iff \langle N, M_0 \rangle$  is not marking observable.

Checking for language inclusion is difficult (see [20]), thus, we look for simpler decision procedures. In particular, the OCG provides a simpler semi-decision (i.e., only sufficient) condition for the marking observability.

Proposition 32: Let us consider a net system  $\langle N, M_0 \rangle$  and its OCG  $\mathcal{G}$ .

- i) A place p is marking observable if there exists a node in  $\mathcal{G}$  such that u(p) = 0.
- ii) The net system is marking observable if there exists a node in  $\mathcal{G}$  such that  $u = \vec{0}_m$ .

*Proof:* It follows from the definition of marking observability and from Proposition 28.  $\Box$ 

On the contrary, the OCG provides necessary and sufficient conditions for strong marking observability. Let us first demonstrate, as an intermediate result, that the repeated firing of a repetitive sequence does not decrease the estimation error.

Lemma 33: Let  $\langle N, M_0 \rangle$  be a net system and let us assume that there exists a firing sequence w' that enables a repetitive sequence w, i.e.,  $M_0[w'\rangle M_{w'}[w\rangle M_{w'w}$  with  $M_{w'w}\geq M_{w'}$ . Then  $\forall \, p\in P$  and  $\forall \, i>1$ ,  $e_p(M_{w'w^i},\mu_{w'w^i})=e_p(M_{w'w},\mu_{w'w})$ .

*Proof:* While observing a sequence w, the error may decrease only during step 4 of Algorithm 7, i.e., when we compute the updating estimate.

Let t be the first transition in the sequence w. If t fires after  $w'w^i$ , in step 4 of Algorithm 7 we have  $\mu'_{w'w^it} \geq \operatorname{Pre}(\cdot,t)$ . Using Proposition 13, it is easy to show that for all  $i \geq 1(M_{w'w^{i+1}} - \mu_{w'w^{i+1}}) \leq (M_{w'w^i} - \mu'_{w'w^it})$ . thus  $\mu_{w'w^{i+1}} \geq (M_{w'w^{i+1}} - M_{w'w^i}) + \mu'_{w'w^it} \geq \mu'_{w'w^it} \geq \operatorname{Pre}(\cdot,t)$ . Therefore,  $\mu'_{w'w^{i+1}t} = \mu_{w'w^{i+1}t}$ , i.e., the estimate is not updated and the error remains constant each time w is repeated after it has fired once.

Proposition 34: Let us consider a net system  $\langle N, M_0 \rangle$ , its OCG  $\mathcal{G}$ , and a place p of N. The place p (resp., the net system) is strongly marking observable in  $k_p$  steps iff the error bound vector is such that u(p)=0 (resp.,  $u=\vec{0}_m$ ) for each node (M/u) in  $\mathcal{G}$  such that: a) the node (M/u) is in a cycle and b) the node (M/u) is dead. Moreover, if a) and b) are satisfied, it is possible to compute  $k_p$  as the length of the longest directed path that starts from the root, contains only intermediate nodes with u(p)>0 (respectively,  $u \geq \vec{0}_m$ ), and ends on a node with u(p)=0 (respectively,  $u=\vec{0}_m$ ).

*Proof:* We prove the property for a single place p; the proof for the net system trivially follows.

(if) By Proposition 23, the number of nodes in  $\mathcal G$  is finite and equal to v. Thus, any word w of length greater or equal to v must pass through a cycle in  $\mathcal G$ , hence, w is p complete by assumption a). Any word of length less that v that leads to a dead marking is also p-complete, by assumption b). This is sufficient to show that the place is SMO in  $k_p$  steps with  $k_p \leq v$ . The actual value of  $k_p$  may be computed as suggested in the statement.

<sup>1</sup>The length of a path is given by the number of edges along the path.

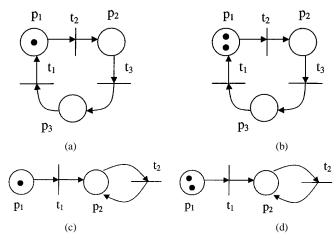


Fig. 4. P/T nets used in the examples.

(only if) We show this by contradiction, proving that if any of the two conditions are violated the place cannot be SMO. Clearly, if condition b) is violated, the place is not strongly marking observable by definition. Now, let us assume that a) is violated. We now consider two subcases.

- i) Assume there exists a node (M/u) along a cycle  $\gamma$  of  $\mathcal{G}$ with  $M(p) \neq \omega$  and u(p) > 0. Then, there exists w' such that  $M_{w'}=M$  and  $e_p(M_{w'},\mu_{w'})=u(p)>0$ . The cycle  $\gamma$  corresponds to a word w such that  $M_{w'}[w]M_{w'}$ , i.e., by Lemma 33 the infinite length sequence  $w'w^i$  may be fired for all i > 0 without reducing the estimation error and the place is not SMO.
- ii) Assume that there exists a node (M/u) with  $M(p) = \omega$ and u(p) > 0 (we do not even need to assume it is along a cycle). Then, consider the path along the observer coverability tree that reaches (M/u) from  $(M_0/u_0)$  and let  $(M, \tilde{u})$  be the first node we encounter along this path with  $M(p) = \omega$ . Then, at step 2.3.2 of Algorithm 21, we have identified a marking M such that  $M_0[w'] \overline{M}[w] M_{w'w}$  and  $M_{w'w} \geq M_{w'}$  ( $\tilde{M}$  is obtained from  $M_{w'w}$  by changing in  $\omega$  the components greater that the corresponding components of  $\overline{M}$ ). Also,  $e_p(M_{w'w}, \mu_{w'w}) = \tilde{u}(p) \ge u(p) > 0$ . Thus, by Lemma 33 the infinite length sequence  $w^i w^i$  may be fired for all i > 0 without reducing the estimation error and the place is not SMO.

Example 35: All net systems in Figs. 2 and 3 are MO but not SMO. On the contrary, one example of strong marking observability (in one step) can be obtained if we consider the net in Fig. 3 with initial marking  $M_0 = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix}^T$  [see Fig. 4(a)].

Analogously, the net systems in Fig. 4(c) and (d) are MO, but only the net system in Fig. 4(c) is SMO (in one step). In the case of the net system in Fig. 4(d), there exist arbitrarily long sequences that are enabled at the initial marking, and that are not complete. In fact,  $\forall i \in \mathbb{N}$ ,  $w \in t_1(t_2)^i$  is not complete.

These results can also be read in Table I that summarizes all observability properties of P/T nets in Fig. 4.

Finally, let us discuss the decidability of these properties.

Theorem 36: It is decidable whether the net system  $\langle N, M_0 \rangle$ is marking and strongly marking observable.

TABLE I OBSERVABILITY PROPERTIES OF THE NETS IN FIG. 4

	MO	SMO	uMO	uSMO	sMO	sSMO
(a)	×	×	×	×	×	_
(b)	×		×	_	×	
(c)	×	×	×	×	_	_
(d)	×		_	_	_	

*Proof:* Decidability of the MO property follows from Proposition 31 and the fact that language inclusion for free Petri nets languages is decidable [8], [20]. Decidability of the SMO property follows from Proposition 34, since the OCG is finite.

## C. Uniform Observability

In this section, we first provide necessary and sufficient conditions for both uniform MO and uniform SMO. Then, we prove the decidability of both these properties.

Let us first demonstrate an important lemma.

Lemma 37: Let  $\langle N, M_0 \rangle$  be a net system. A place  $p \in P$  is observable in  $\langle N, M_0 \rangle$  iff at least one element in the semilinear

$$\mathcal{A}_{p} = \{ M \in \mathbb{N}^{m} \mid M(p) = 0 \} \cup \left( \bigcup_{t \in p^{\bullet}} \{ M \in \mathbb{N}^{m} \mid M(p) \right)$$
$$= Pre(p, t), M \ge Pre(\cdot, t) \}$$
(3)

is reachable.

*Proof:* (if) Let w be a word in  $L(N, M_0)$ . Let us consider two subcases.

- i) If  $M_w \in \{M \in \mathbb{N}^m \, | \, M(p) = 0\}$ , then  $0 = M_w(p) \ge 1$
- $\begin{array}{l} \mu_w(p)\geq 0\text{, thus, }M_w(p)=\mu_w(p).\\ \text{ii) If }M_w\in \ \{M\in \ \mathbb{N}^m\mid \ M(p)=\ Pre(p,t),M\ \geq \ \mathbb{N}^m \mid M(p)=\ \mathbb{N}^m \mid M($  $Pre(\cdot,t)$ } where  $t \in p^{\bullet}$ , then t may fire at  $M_w$  and since  $M_w(p) = Pre(p,t)$  the updated estimate is  $\mu'_w(p) =$  $M_w(p)$ , hence,  $M_{wt}(p) = \mu_{wt}(p)$ .

(only if) We prove this by contradiction.

If no marking with M(p) = 0 is reachable, then  $M_w(p) >$  $0 \ \forall \ w \in L(N, M_0)$ , thus, the initial place estimation error is strictly positive. It may decrease only during step 4. of Algorithm 7. However, if  $\forall w$  and  $\forall t \in p^{\bullet}$ ,  $M_w(p) > Pre(p,t)$ , then  $\mu'_w(p) < M_w(p)$ , thus. the place estimation error keeps positive.

By virtue of the previous lemma, the study of uniform marking observability reduces to the study of m home space problems.

Proposition 38: A net system  $\langle N, M_0 \rangle$  is uniformly marking observable iff the semi-linear set  $A_p$  given by (3) is a home space  $\forall p \in P$ .

*Proof:* It follows from the previous lemma and the fact that a net system  $\langle N, M_0 \rangle$  is uniformly marking observable iff each place  $p \in P$  is observable in  $\langle N, M \rangle$ ,  $\forall M \in R(N, M_0)$ , i.e., iff the semi-linear set (3) is a home space  $\forall p \in P$ .

Let us now consider the uniform SMO property.

Proposition 39: A net system  $\langle N, M_0 \rangle$  is uniformly strongly marking observable only if its reachability set is finite.

Proof: If the reachability set is not finite, then (by fact 1.iii) there exist words w' and w such that  $M_0[w'\rangle M_{w'}[w\rangle M_{w'w}$  with  $M_{w'w} \geq M_{w'}$ . This means that  $w \in L(N, M_{w'}) \cap L(N, M_{w'w})$ , and (by Lemma 9)  $M_{w'w} \geq M_{w'} \geq w^{-1}(\mu_w)$ , thus (by proposition 26) the word w is not marking complete wrt  $\langle N, M_{w'w} \rangle$ . Also, we can have words of infinite length  $w^i$  (for all i > 1) that are not marking complete (by Lemma 33) thus the system  $\langle N, M_{w'w} \rangle$  is not SMO and finally  $\langle N, M_0 \rangle$  is not uSMO.

Example 40: The net systems in Fig. 4(a)–(c) are uMO. In fact, in the first two cases,  $\forall p \in P$  the set  $\{M \in \mathbb{N}^m \mid M(p) = 1\}$ 0} is always a home space. In the third case, the set  $\{M \in \mathbb{N}^m \mid$  $M(p_1) = 0$ } and the set  $\{M \in \mathbb{N}^m \mid M(p_2) = Pre(p_2, t_2) = 0\}$ 1,  $M > Pre(\cdot, t_2)$  are home spaces. On the contrary, the net system in Fig. 4(d) is not uMO because  $\{M \in \mathbb{N}^m \mid M(p_2) = 0\}$  $0\} \cup \{M \in \mathbb{N}^m \mid M(p_2) = Pre(p_2, t_2) = 1, M \ge Pre(\cdot, t_2)\}\$ is not a home space and thus  $p_2$  is not uMO; in fact the net system  $\langle N, M \rangle$  is not MO at  $M = \begin{bmatrix} 0 & 2 \end{bmatrix}^T \in R(N, M_0)$ .

The net systems in Fig. 4(a) and (c) are uSMO. Obviously, the net system in Fig. 4(b) is not uSMO, being not SMO. Analogously, the net system in Fig. 4(d) not being uMO, it is also not uSMO.

Theorem 41: It is decidable if a net system  $\langle N, M_0 \rangle$  is uniformly and strongly uniformly marking observable.

*Proof:* Let us first prove the decidability of uniform marking observability. Because of Proposition 38 it is sufficient to prove that the home-space property for the set  $\mathcal{A}_p$  is decidable. Let us observe that  $\forall p \in P$  the semilinear set in (3) is given by the finite union of linear sets having the same periods. In fact, if we consider a generic place  $p_k \in P$ 

$$\begin{split} \{M \in \mathbb{N}^m \mid M(p_k) = 0\} &= \left\{ \sum_{i \neq k} a_i \vec{\varepsilon}_i \mid a_i \in \mathbb{N} \right\} \\ \{M \in \mathbb{N}^m \mid M(p_k) = Pre(p_k, t), \ M \geq Pre(\cdot, t)\} \\ &= \left\{ Pre(\cdot, t) + \sum_{i \neq k} b_i \vec{\varepsilon}_i \mid b_i \in \mathbb{N} \right\} \end{split}$$

where  $\vec{\epsilon_i}$  is the *i*th-canonical basis vector of dimension m. Thus, the decidability of the home-space property for  $A_p$  immediately follows by Theorem 5.

Second, let us prove the decidability of strong uniform marking observability. Let us observe that if the necessary requirement stated by Proposition 39 is satisfied, then the reachability set is finite and the uniform strong marking observability can be verified by proving the strong marking observability—that is decidable [9]—for a finite set of initial markings.

# D. Structural Observability

In this section, we provide necessary and sufficient conditions for both structural and strong structural marking observability and we prove the decidability of these properties.

Proving structural observability, requires the study of the system properties for all possible initial markings. The next two lemmas show that to prove that a place is observable for all initial markings in  $\mathbb{N}^m$ , just a finite subset of  $\mathbb{N}^m$  needs to be checked.

Lemma 42: If a place  $p \in P$  is observable in  $\langle N, M \rangle$  then it is also observable in  $\langle N, \overline{M} \rangle \ \forall \ \overline{M} \geq M \ \text{with} \ \overline{M}(p) = M(p).$ 

*Proof:* A place p is observable in  $\langle N, M \rangle$  if and only if  $\exists w \in L(N,M)$  such that M[w]M' and  $\mu_w(p) = M'(p)$ . In this case  $\forall \overline{M} \geq M$  with  $\overline{M}(p) = M(p), w \in L(N, \overline{M})$  (by fact 1.i) and  $\overline{M}[w]\overline{M}'$  with  $\overline{M}'(p) = M'(p) = \mu_w(p)$  (by fact 1.ii), i.e., p is also observable in  $\langle N, \overline{M} \rangle$ .

Lemma 43: Let N be a Petri net and let  $r_p$  $\max_{t \in T} Pre(p, t)$ . Let

$$M_i^p = \begin{cases} M_i^p(p') = 0, & \text{if } p' \neq p \\ M_i^p(p) = i. \end{cases}$$
 (4)

A place  $p \in P$  is observable in  $\langle N, M_i^p \rangle \ \forall \ i \in \mathbb{N}$ , iff p is

observable in  $\langle N, M_i^p \rangle$  for  $i=1,\dots,r_p+1$ . Proof: If p is observable in  $\left\langle N, M_{r_p+1}^p \right\rangle$  then  $\exists w$  and  $t \in p^{\bullet}$  such that  $M_{r_n+1}^p[w)\overline{M}$ ,  $\overline{M}(p) = Pre(p,t)$ , i.e., the firing of the word w reduces the number of tokens in p. This implies that for all  $M_i^p$  with  $i > r_p + 1$ , the word w may also fire until we reach a marking  $\overline{M}'$  such that  $\overline{M}' \geq M_{\varrho_i}^p$  and  $\overline{M}'(p) = \varrho_i \leq r_p$ . Since p is observable in  $\langle N, M_{\rho_i}^p \rangle$ , then it is also observable in  $\langle N, \overline{M}' \rangle$  by Lemma 42.

Proposition 44: A Petri net N is structurally marking observable iff  $\forall p \in P$ , p is observable in  $\langle N, M_i^p \rangle$ , where  $M_i^p$  is defined as in (4) and  $i = 1, \ldots, r_p + 1$ .

*Proof:* By definition, a Petri net N is sMO iff  $\forall p \in P, p$ is observable in  $\langle N, M \rangle \, \forall \, M \in \mathbb{N}^m$ . By Lemma 42 and Lemma 43 it is however sufficient to check that each p is observable for the finite number of initial makings given in the statement.  $\Box$ 

Proposition 45: A Petri net N is strongly structurally marking observable iff

- a) N has no repetitive sequences;
- b)  $\forall p \in P, \exists t \in T \text{ such that }$

$$Pre(p',t) = \begin{cases} 1, & \text{if } p' = p \\ 0, & \text{if } p' \neq p. \end{cases}$$

*Proof:* (if) We will prove that a) and b) imply that for any initial marking  $M_0$  in finite number of steps the net looses all its tokens: this is a sufficient condition for SMO of  $\langle N, M_0 \rangle$ by Lemma 37. In fact, if no repetitive sequences exist, for any initial marking the length of all words firable is bounded, i.e., after a finite number of firings the net reaches a dead marking. Furthermore, if assumption b) is verified, for each place  $p \in$ P there exists a transition t whose single input is p and the corresponding arc weight is unitary, i.e., if t cannot fire then place p must be empty. Thus the dead marking must be the zero marking.

(only if) We prove this by contradiction.

Let us first assume that a) is violated, and let w be a repetitive sequence. Clearly, for any  $M \geq w^{-1}(\mu_w)$ , the word w is not marking complete wrt  $\langle N, M \rangle$  (by Proposition 26). Also, we can have words of infinite length  $w^i$  (for all i > 1) that are not marking complete (by Lemma 33) thus the system  $\langle N, M \rangle$  is not SMO.

Second, we assume that a) is verified while  $\exists p \in P$  such b) is violated. We first observe that we can exclude the existence of transitions with no input arcs, because this would violate condition (a). Then it is obvious that given the marking  $M_1^p$  as in (4) (that contains one token in p and zero tokens elsewhere) no transition is enabled, thus the marking of p cannot be observed.

Theorem 46: It is decidable if a Petri net N is structurally marking observable and structurally strongly marking observ-

*Proof:* To prove that N is sMO it is sufficient to prove (by Proposition 44) that all places are observable in  $\langle N, M_0 \rangle$  for a finite number of initial markings  $M_0$ . The property of being observable for a place is decidable because of Theorem 3 and of the characterization given by Lemma 37.

To prove that N is sSMO it is sufficient to check by Proposition 45 that no repetitive sequences exist (and this may be checked with linear algebraic tools given the net incidence matrix) and that the net structure satisfies condition b) (this may be checked by inspection).

Example 47: Being sMO and sSMO structural properties of the net, the same conclusions can be drawn for nets in Fig. 2(a)–(d), respectively.

In particular, the net in Fig. 4(a) is sMO by proposition 44. On the contrary, it is not sSMO. In fact, if we consider the initial marking in Fig. 4(b) the net system is not SMO.

The net N in Fig. 4(c) is not sMO (thus, it is also not sSMO). In fact, if we consider  $M_0 = \begin{bmatrix} 0 & 2 \end{bmatrix}^T$ ,  $\langle N, M_0 \rangle$  is not MO.

A final remark regards the classes of nets that are sSMO. Although this property is rather easy to prove, the class of nets that satisfy this property is of little practical interest (they must become empty and deadlock in a finite number of steps). The property of structural MO, on the contrary, is more difficult to prove, but is satisfied by a wider (more interesting) class of nets.

#### VII. MARKING ESTIMATION WITH MACROMARKING

In Sections II–VI, no information was assumed on the initial marking  $M_0$  that originates the observed transition firings. It is often the case, however, that partial information about this marking is available.

As an example, let us assume that the net starts its evolution at a given time instant  $\tau_{-}$  from a known marking  $M_{-}$  (called start marking). After having evolved unobserved for some time, the net reaches a marking  $M_0$  (called initial marking) from which we begin the observation of the transition firings. Now, we know that  $M_0 \in R(N, M_{-})$  and we could use this information to better characterize the set of markings consistent with an observed word w given the information on the start marking as

$$\mathcal{M}(w \mid M_{-}) = \{M \mid \exists M_{0} \in R(N, M_{-}), M_{0}[w\rangle M\}.$$

The main problem with this is that this characterization is given in terms of Petri net reachability (the initial marking must be reachable from the start marking) that is hard to solve. Looking for simpler structures, we consider the case in which the knowledge of  $M_0$  can be written as  $M_0 \in \mathcal{V}(V,b)$ , where  $\mathcal{V}$  is a macromarking defined as follows.

Definition 48: We assume that the set of places P can be written as the union of r+1 subsets:  $P = P_0 \cup P_1 \cup \cdots \cup P_r$  such that  $P_0 \cap P_j = \emptyset$ , for all j > 0. The number of tokens contained in  $P_i$  (j > 0) is known to be  $b_i$ , while the number of tokens in  $P_0$  is unknown. For each  $P_j$ , let  $\vec{v}_j$  be its characteristic vector, i.e.,  $v_j(p) = 1$  if  $p \in P_j$ , else  $v_j(p) = 0$ . Let  $V = [\vec{v}_1, \dots, \vec{v}_r]$ and  $\vec{b} = [b_1, \dots, b_r]$ . We define the macromarking  $\mathcal{V}(V, \vec{b})$  as defined by the set  $\{M \in \mathbb{N}^m \mid V^T M = \vec{b}\}.$ 

Note that, as a special case, if V is a matrix of P-invariants, then by definition

$$R(N, M_{-}) \subseteq \{M \in \mathbb{N}^m \mid V^T M = V^T M_{-}\}$$

where  $V^T M_- = \vec{b}$  is known, thus a macromarking can also approximate the info about the start marking  $M_{-}$ .

Thus, we add to assumptions A1 and A2 given in Section III the following assumption:

A3) the initial marking  $M_0$  belongs to the macromarking  $\mathcal{V}(V, \vec{b})$ , i.e., it satisfies the equation  $V^T M_0 = \vec{b}$ .

Given an evolution of the net  $M_0[t_{\alpha_1}\rangle M_1[t_{\alpha_2}\rangle \cdots$ , we use the following algorithm to compute estimate  $\mu_{w_i}$  and bound  $B_{w_i}$  of each actual marking  $M_{w_i}$  based on the observation of the word of events  $w_i=t_{\alpha_1}t_{\alpha_2}\cdots t_{\alpha_i}$  , and of the knowledge of the initial macromarking  $\mathcal{V}(V, \vec{b})$ .

Algorithm 49: Marking Estimation with Event Observation and Initial Macromarking

- 1. Let the initial estimate be  $\mu_{w_0} = \vec{0}_m$ .
- 2. Let the initial bound be  $B_{w_0} = \vec{b}$ .
- 3. Let i = 1.
- 4. Wait until  $t_{\alpha_i}$  fires.
- 5. Update the estimate  $\mu_{w_{i-1}}$  to  $\mu'_{w_i}$  with

$$\mu'_{w_i}(p) = \max\{\mu_{w_{i-1}}(p), Pre(p, t_{\alpha_i})\}.$$

- 6. Let  $\mu_{w_i} = \mu'_{w_i} + C(\cdot, t_{\alpha_i})$ . 7. Let  $B_{w_i} = B_{w_{i-1}} V^T \cdot (\mu'_{w_i} \mu_{w_{i-1}})$ .
- 8. Let i = i + 1.
- 9. Goto 4.

Note that the estimate  $\mu$  computed using this algorithm is the same of the estimate computed with Algorithm 7 and, thus, all observability properties already discussed do not change.

What is new is the additional information given by the bounds that will be used to characterize the set of consistent markings.

Theorem 50: Given an observed word  $w \in L(N, M_0)$  with initial macromarking  $\mathcal{V}(V,b)$ , the corresponding estimated marking  $\mu_w$  and bound  $B_w$  computed by Algorithm 49, the set of w consistent markings is  $\mathcal{M}(w \mid V, \vec{b}) = \{M \in \mathbb{N}^n \mid$  $V^T \cdot M = V^T \cdot \mu_w + B_w, M \ge \mu_w \}.$ 

*Proof:* Let  $w_0$  be the empty word. Then,  $\{M \in \mathbb{N}^n \mid$  $V^{T} \cdot M = V^{T} \cdot \mu_{w_{0}} + B_{w_{0}}, M \ge \mu_{w_{0}} \} = \{ M \in \mathbb{N}^{n} \mid V^{T} \cdot M = V^{T} \cdot \mu_{w_{0}} + \vec{b} - V^{T} \cdot \mu_{w_{0}}, M \ge \mu_{w_{0}} \} = \{ M \in \mathbb{N}^{n} \mid V^{T} \cdot M = \vec{b}, M \ge \mu_{w_{0}} \} = \{ M \in \mathbb{N}^{n} \mid V^{T} \cdot M = \vec{b} \} \equiv \mathcal{M}(w_{0} \mid V, \vec{b}).$ 

By induction, let us show that  $\mathcal{M}(w \mid V, \vec{b}) = \{M \in \mathbb{N}^n \mid$  $M \ge \mu_w, V^T \cdot M = V^T \cdot \mu_w + B_w \} \Longrightarrow \mathcal{M}(wt \mid V, \vec{b}) = \{M \in \mathbb{N}^n \mid M \ge \mu_{wt}, V^T \cdot M = V^T \cdot \mu_{wt} + B_{wt} \}.$ 

In fact,  $\mathcal{M}(wt \mid V, \vec{b}) \equiv \{M \in \mathbb{N}^n \mid \exists M' \in \mathcal{M}(w)\}$  $(V,b), M' \geq Pre(\cdot,t), M = M' + C(\cdot,t) = \{M \in \mathcal{M} \in \mathcal{M}$   $\mathbb{N}^n \mid \exists M', V^T \cdot M' = V^T \cdot \mu_w + B_w, M' \geq \mu_w, M' \geq Pre(\cdot, t), M = M' + C(\cdot, t)$ .

Now, let  $\mu'_{wt}$  be the updated estimate of  $M_w$  after t fires. Then,  $[M' \geq \mu_w] \wedge [M' \geq Pre(\cdot,t)] \iff M' \geq \mu'_{wt}$ . Furthermore, with the notation of Algorithm 49,  $B_w + V^T \cdot \mu_w = B_{wt} + V^T \cdot \mu'_{wt}$ , and  $\mu_{wt} = \mu'_{wt} + C(\cdot,t)$ . Hence,  $\mathcal{M}(wt \mid V, \vec{b}) = \{M \in \mathbb{N}^n \mid \exists M', V^T \cdot M' = V^T \cdot \mu'_{wt} + B_{wt}, M' \geq \mu'_{wt}, M = M' + C(\cdot,t)\} = \{M \in \mathbb{N}^n \mid V^T \cdot M = V^T \cdot \mu_{wt} + B_{wt}, M \geq \mu_{wt}\}$ .

The previous theorem allows us to write a general optimization problem of the form

$$\max \vec{c}^T \cdot M$$
s.t.  $M \in \mathcal{M}(w \mid V, \vec{b})$ 

as a linear integer programming problem (IPP)

$$\max \vec{c}^T \cdot M$$
s.t.  $V^T \cdot M = V^T \cdot \mu_w + B_w$ 

$$M \ge \mu_w.$$
 (5)

As an example, appropriately choosing the value of  $\vec{c}$ , such an IPP can be used to compute the maximum over all consistent markings of the tokens in the net (if  $\vec{c} = \vec{1}$ ), and of the tokens in a generic place  $p_i$  (if  $\vec{c} = \vec{e_i}$ ).

Note that if we do not want to solve an integer linear programming problem, it is possible to give ranges on the estimation errors by simple inspection of B.

Theorem 51: Consider an observed word  $w \in L(N, M_0)$  with initial macromarking  $\mathcal{V}(V, \vec{b})$ , the corresponding estimated marking  $\mu_w$  and bound  $B_w$  computed by Algorithm 49, and  $P = P_0 \cup P_1 \cup \cdots \cup P_r$  with the notation of Definition 48.

- 1)  $\forall M \in \mathcal{M}(w \mid V, \vec{b}), l \leq e(M, \mu_w) \leq u \text{ where } l = \max_j B_w(j), \text{ and } u = \vec{1}_T^T \cdot B_w \text{ if } P_0 = \emptyset, \text{ else } u = +\infty.$
- 2)  $\forall M \in \mathcal{M}(w \mid V, \vec{b}), e_{p_i}(M, \mu_w) \leq u_{p_i} \text{ where } u_{p_i} = \min_{j \mid p_i \in P_j} B_w(j) \text{ if } p_i \in P \setminus P_0, \text{ else } u_{p_i} = +\infty.$ Proof:
- 1) The first inequality immediately follows by definition of  $B_w$ . Moreover,  $V^TM = V^T\mu_w + B_w$ , thus  $V^T(M \mu_w) = B_w$  and  $\vec{1}_r^T \cdot V^T(M \mu_w) = \vec{1}_r^T \cdot B_w$ . If  $P_0 = \emptyset$ , then  $\vec{1}_r^T \cdot V^T \geq \vec{1}_m^T$ , thus  $e(M, \mu_w) = \vec{1}_m^T \cdot (M \mu_w) \leq \vec{1}_r^T \cdot V^T(M \mu_w) = \vec{1}_r^T \cdot B_w$ . On the contrary, if  $P_0 \neq \emptyset$ , then an arbitrarily large number of tokens can be added to  $P_0$ , thus,  $u = +\infty$ .
- 2) By definition, each subset  $P_j$  such that  $p_i \in P_j$  imposes a constraint of the form  $e_{p_i}(M,\mu_w) \leq B_w(j)$  on the place estimation error. When  $p_i$  belongs to more than one subset of places, the resulting constraints should be satisfied simultaneously, thus providing the above statement. On the contrary, if  $p_i \in P_0$ , no limit exists on the number of tokens that can be added to  $\mu_w$ .

Remark 52: In the case of disjoint subsets  $P_j$ 's,  $l=u=\overline{1}_r^T \cdot B_w$  if  $P_0=\emptyset$ , else  $l=\overline{1}_r^T \cdot B_w$ .

From Theorem 51, we have the following corollary that shows how the bound  $B_w$  may be used to prove that a word w is complete.

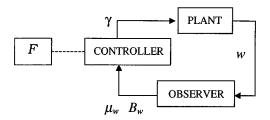


Fig. 5. State feedback control loop with observer.

Corollary 53: With the notation of Algorithm 49

- 1) if  $P_0 = \emptyset$ , then w is marking complete if and only if  $B_w = \vec{0}_r$ ;
- 2) if  $P_0 \neq \emptyset$ , then w is marking complete only if  $B_w = \vec{0}_r$ .

We conclude this section with the following observation. In Algorithm 49 by construction we are sure that for all w it holds  $\mu_w \leq M_w$ . However, if we are willing to pay the extra cost of solving m optimization problems of the form (5) at each iteration, we may be able to update each component of  $\mu_w$  in step 1. and step 6. of the algorithm as follows:

$$\tilde{\mu}_w(p) = \min\{M(p) \mid M \in \mathcal{M}(w \mid V, \vec{b})\}.$$

It is easy to show that this updated estimate is such that

$$\mu_w \leq \tilde{\mu}_w \leq M_w$$
.

#### VIII. CONTROL USING OBSERVERS

In this section, we show how the marking estimate constructed with the formalism discussed in Section VII can be used by a control agent to enforce a given specification on the plant behavior.

We make several assumptions that are briefly discussed here.

- The specification is given as a set of forbidden markings  $\mathcal{F}$ . The set of legal markings is  $\mathcal{L} = \mathbb{N}^m \mathcal{F}$ .
- The controller may disable transitions to prevent the plant from entering a forbidden marking. From the knowledge of  $\mu_w$  and  $B_w$ , the controller computes a control pattern  $\gamma: T \to \{0,1\}$ . If  $\gamma(t) = 0$  then t is disabled by the controller.
- All transitions are controllable, i.e., can be disabled by the controller.

The considered control scheme is shown in Fig. 5.

Under the assumption that the initial marking  $M_0 \in \mathcal{L}$ , the following algorithm may be used by the controller at each step to ensure that markings in  $\mathcal{F}$  are not reached.

Algorithm 54: Let w be the observed word, and  $\mathcal{M}(w \mid V, \vec{b}) = \{M \in \mathbb{N}^n \mid V^T \cdot M = V^T \cdot \mu_w + B_w, M \geq \mu_w\},$  where  $\mu_w$  and  $B_w$  are computed by the observer.

for all  $t \in T$  begin  $\gamma(t) := 1;$  if  $\exists M \in \mathcal{M}(w \mid V, \vec{b}) \cap \mathcal{L} \text{ such that } M[t\rangle M', M' \in \mathcal{F}$  then  $\gamma(t) := 0;$  end.

Clearly, this algorithm prevents all transition firings that lead from  $\mathcal{L}$  to  $\mathcal{F}$  but is not necessarily optimal, in the sense that it may also prevent transition firings that lead from  $\mathcal{L}$  to  $\mathcal{L}$ . A similar algorithm was also discussed in [11, Alg. 5.3] to ensure predicate invariance using state estimates computed by a dynamic observer.

In general, it may be difficult to check the condition of the **if** statement of the algorithm. However, when  $\mathcal F$  is a finite set the observer estimate may be used to verify this condition. In fact, we simply have to check whether there exists a marking  $M \in \mathcal M(wt \mid V, \vec b) \cap \mathcal F$  such that  $t^{-1}(M) \in \mathcal L$ , and this is trivial given the characterization of Theorem 50.

We also would like to consider a special case in which a control law different from the one presented above may be suitable. Let the specification on the legal states be given by  $\mathcal{L} = \{M \in \mathbb{N}^m \mid S^T \cdot M \leq \vec{k}\}$  where  $S = [\vec{s_1} \cdots \vec{s_q}]$  with  $\vec{s_j} \in \mathbb{Z}^n$  and  $\vec{k} = [k_1 \cdots k_q]$  with  $k_j \in \mathbb{Z}$ . This kind of specifications, that we call *generalized mutual exclusion constraints* have been considered by various authors [10], [12], [24].

Assume that the initial marking  $M_0$  of the plant does not necessarily belong to  $\mathcal{L}$  (this is a natural assumption when considering error recovery problems). Then, given a marking M we may want to prevent the firing of transition t such that M[t)M' when the following two conditions are verified:

- a) there exists  $\vec{s}_j$  with  $\vec{s}_j \cdot M' > k_j$ , i.e.,  $M' \in \mathcal{F}$ ;
- b)  $\vec{s_j} \cdot M' > \vec{s_j} \cdot M$ , i.e., the firing of t either leads to a violation of the constraint (if  $M \in \mathcal{L}$ ) or to a "worse" violation of the constraint (if  $M \in \mathcal{F}$ ).

In this case, the following algorithm may be used to compute the control pattern  $\gamma$  at each step.

Algorithm 55: Let w be the observed word, and  $\mathcal{M}(w \mid V, \vec{b}) = \{M \in \mathbb{N}^n \mid V^T \cdot M = V^T \cdot \mu_w + B_w, M \geq \mu_w\}$ , where  $\mu_w$  and  $B_w$  are computed by the observer. Let  $\mathcal{L} = \{M \in \mathbb{N}^n \mid S^T \cdot M \leq \vec{k}\}$ .

```
\begin{array}{l} \text{for } all \ t \in T \\ \text{begin} \\ \gamma(t) := 1; \\ j := 1; \\ \text{while } j \leq q \ \text{and } \gamma(t) = 1 \ \text{do} \\ \text{begin} \\ \Delta := \vec{s}_j^T \cdot C(\cdot, t); \\ \text{if } \Delta > 0 \ \text{then} \\ \text{begin} \\ \overline{m} := \max \left\{ \vec{s}_j^T \cdot M \mid M \in \mathcal{M}(wt \mid V, \vec{b}) \right\}; \\ \text{if } \overline{m} > k_j \ \text{then } \gamma(t) := 0; \\ \text{end}; \\ j := j + 1; \\ \text{end}; \\ \text{end}. \end{array}
```

Thus, a transition is disabled at M only if its firing leads to a marking M' such that for at least one constraint  $j\colon \vec{s}_j^T\cdot M'>\vec{s}_j^T\cdot M$  (i.e.,  $\Delta>0$ ) and there exists a consistent marking M'' in  $\mathcal{M}(wt\mid V,\vec{b})$  that violates the constraint (i.e.,  $\vec{s}_j^T\cdot M''>k_j$ ).

The methodology developed in this paper is applied to a simple manufacturing example. Other examples can be found in [7].

Example 56: Let us consider the net in Fig. 3 with initial marking  $M_0 = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}^T$ . This system may represent a pool of three machines. Each token represents a machine that may be in any of three states: working (token in place  $p_1$ ), idle (token in place  $p_2$ ), loading (token in place  $p_3$ ). We assume that the specification on the system behavior requires that at most two machines may be simultaneously working, i.e., the set of forbidden states is  $\mathcal{F} = \{M \in \mathbb{N}^3 \mid M(p_1) > 2\}$ .

The initial macromarking  $M(p_1)+M(p_2)+M(p_3)=3$  captures our knowledge that there are three machines in the pool. Their initial state is, however, unknown.

To represent the global behavior of the plant with observer under control using Algorithm 55, we have represented the observer reachability graph of the controlled plant with observer in Fig. 6. The observer reachability graph has been constructed following the same rules of Algorithm 21. We have also introduced a new label at each node so as to better highlight the effect of the control pattern  $\gamma$ . Each node is now labeled (M/u/B) where M is the real marking, u is a vector whose components, being the net bounded, coincide with the place estimation errors, and B is the resulting bound.

Note that, given  $u = M - \mu$ , the bound B can be immediately computed as  $B = V^T u$ , thus the addition of the new label does not introduce significant variations on Algorithm 21.

Let us briefly discuss the graph in Fig. 6. The initial marking is represented by a round corner box. A dashed box represents a marking that cannot be reached because the transition firing leading to it is disabled by the controller (the corresponding edge is dashed). A thick box represents a marking reached by a complete word w, i.e.,  $u_w = \vec{0}$  and  $B_w = 0$ : the future evolution from such a marking is not shown.

Note that the transition firings disabled by the controller using Algorithm 55 in reality do not lead to forbidden markings: they are disabled because there exist markings consistent with the observation from which these transition firings would lead to forbidden markings. This can be easily verified by looking at the nodes within dashed boxes. In all these cases the value of  $\overline{m}$  in Algorithm 55 is equal to

$$\overline{m} = V^T \cdot \mu + B = V^T \cdot (M - u) + B = M(p_1) - u(p_1) + B = 3.$$

On the contrary, if the real marking would have been used to determine the control pattern, such a node would have been reachable, being  $V^T \cdot M \leq 2$ .

Let us finally observe that, since the controller may prevent the firing of transitions whose firing is perfectly legal, it may also be the case that the controlled system is blocking.

A preliminary solution to this problem has been presented in [7] and consists in the introduction of suitable recovery mechanisms with an  $ad\ hoc$  reasoning. A more general procedure to automatically recover the net from a blocking condition is given in [1]. This approach is essentially based on a linear algebraic characterization of deadlock markings, that reveal to be useful to derive additional information on the actual marking of the net, so as to improve the marking estimate, thus restricting the set of w consistent markings.

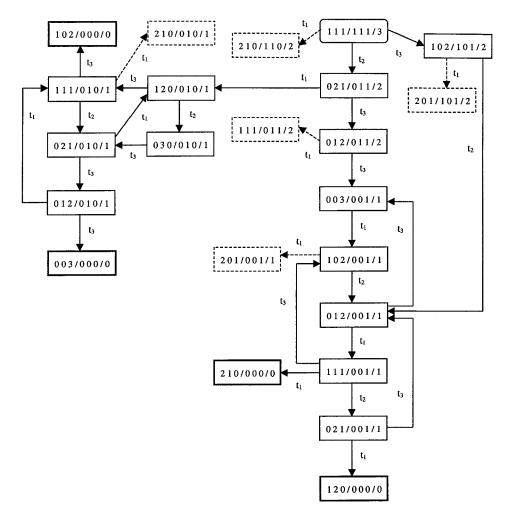


Fig. 6. Observer reachability graph of the controlled net system in example 56.

## IX. CONCLUSION

In this paper, we dealt with the problem of estimating the marking of a P/T net based on event observation, assuming that the net structure is known. We considered two cases: 1) the initial marking is not known and 2) the initial marking is known to belong to a *macromarking*, i.e., we know the token contents of subsets of places but not the exact token location.

We defined several observability properties and showed that they are decidable. In particular, we considered two main properties: *marking observability* and *strong marking observability*. The first one means that there exists at least one word that is complete, while the second one means that all words can be completed in a finite number of steps to a complete word.

We investigated the possibility that the two properties above are satisfied by a net N starting from an initial marking  $M_0$ , by a net N starting from any marking M reachable from an initial marking  $M_0$  (uniform observability) or by a net N starting from any marking in  $\mathbb{N}^m$  (structural observability) where m is the number of places of the net.

We also introduced the *observer coverability graph*, i.e., the usual coverability graph of a P/T net augmented with a vector that keeps track of the estimation error on each place of the net. We proved that it can be a useful tool when proving some of the above properties. We also showed that many observability prop-

erties can be proved by reducing them to other decision problems (e.g., home-space properties, marking reachability, existence of repetitive sequences) that can be checked using algorithms well known from the literature.

Finally, we showed how the estimate generated by the observer may be used to design a state feedback controller, that ensures that the controlled system never enters a set of forbidden states.

There are several ways in which this research may be extended.

First, we may consider the case in which not all transition firings are observable, or there may be transitions that do not generate distinct events. This may destroy the framework used in this paper because in this case the marking estimate is not anymore a lower bound of the actual marking. However, in some restricted case, e.g., when the firing of all unobservable transitions does not change the total number of tokens, we believe it may be possible to extend the approach used in this paper.

Second, we may associate a probabilistic structure to the transition firings. Then, given an initial marking  $M_0$ , we may define a function  $\Pi: \mathbb{N} \to [0,1]: \Pi(k)$  denotes the probability that, after having observed k event firings, we obtain a complete word. It would be interesting to study under which conditions the limit of  $\Pi(k)$  goes to one as k goes to infinity.

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