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Continuous flow Systems and Control Methodology
Using Hybrid Petri nets

Latéfa GHOMRI*, Hassane ALLA**

* MELT, University of Tlemcen, Algeria, ghomri@mail.univ-tlemcen.dz
** GIPSA-LAB, University of Grenoble, France, Hassane.alla@gipsa-lab.grenoble-inp.fr

Abstract: In this paper, we consider the controller synthesis for continuous flow systems. These lasts are a sub-class of hybrid dynamic systems. Their main characteristics are positiveness and linearity. Transport, manufacturing, communication and biological systems are examples of continuous flow systems. Numerous tools and techniques exist in the literature for modelling and analyzing such systems. As positiveness is a hard constraint, an appropriate tool integrating naturally this constraint is strongly needed. Hybrid Petri Nets are an elegant modeling tool of positive systems, while Hybrid Automata are a powerful tool giving formally the reachable dynamic space. Combining these two tools aim to a sound approach for control synthesis of continuous flow systems. We start by considering the process to control and compute its reachable state space using specialized software like PHAVer. Algebraic inequalities define this reachable state space. The constrained behaviour is obtained by restricting this state space into a smaller desired space. This reduction is expressed in term of linear constraints only over the continuous variables; while the control is given by the discrete transitions (occurrence dates of controllable events). The controller synthesis methodology is based on the control of a hybrid system modelled by a D-elementary Hybrid Petri Net. The control consists in modifying the guard of the controllable transitions so as the reachable controlled state space is maximally permissive.

1. INTRODUCTION
Modelling and control of physical systems are crucial issues. In this work we are interested in a particular class of systems, such as transport, manufacturing, communication and biological systems. These systems have in common that they are positive dynamic systems, i.e., the state variables are positive. Often continuous and discrete event dynamics interact in these systems, so they can be considered as positive hybrid dynamic systems. This class of systems requires for their description, the use of continuous time models like differential equations, and discrete event models like finite state automata or Petri nets (PNs). In general, the state of a hybrid system is given by the discrete mode and the values of the continuous variables. This state may change either continuously, according to a differential equation or discretely by an instantaneous change of the discrete control mode.

The concept of controller synthesis, considered here, has for origin the work of Ramadge and Wonham (1989). The latter have introduced controller synthesis for discrete event systems. The process is a discrete event system described by a finite state automaton. The constraints imposed on its behaviour are modelled by any regular language. Both models (process and constraints) allow synthesizing a controller whose role is to prohibit some controllable events in order to always satisfy the specifications. The theory of Ramadge and Wonham (1989) has been extended in several directions; one of the major extensions of this theory is the controller synthesis for timed systems. In this case the process model considers time in an explicit manner. Time can either be discretized, in this case the process and its supervisor are modelled by finite state machines equipped with a discrete clock (Brandin and Wonham, 1997); or dense, and in this case the process and its supervisor are modelled by timed automata (Alur and Dill 1994). Several research studies have been devoted to the controller synthesis of timed systems (Altisen and Tripakis 1999, Asarin and Maler (1997), Asarin et al. 1998). The controller synthesis theory is well established for discrete event systems and timed systems. However, it has not yet an explicit solution for hybrid systems, although some studies have been devoted to this field (Wong-Toi, 1997). The difficulty of analysis, in general, and of controller synthesis, in particular, of hybrid systems, is due to the fact that restrictions are needed on the dynamics in order to have an algebraic characterization of the reachable state space. Another difficulty comes from the fact that the computation is not decidable in the general case.

Our aim is to develop a controller synthesis technique for hybrid dynamic systems modelled by a D-elementary Hybrid Petri Nets (HPNs). It is a formal tool with large description capacities and is well appropriate for positive hybrid systems modelling. Its dynamic analysis and performance calculation have been studied in (Ghomri and Alla, 2007). The
continuous part of a D-elementary HPN evolves in a piecewise linear manner, according to differential equations of the form: \( \dot{x} = k \), where \( k \) is constant. This continuous part is controlled by the discrete part which evolves in an independent manner (i.e. the discrete part evolution is completely independent from the continuous part evolution). Because of the strong interaction between the discrete and the continuous parts and the lack of analysis tools (software) for D-elementary HPNs, we have proposed to translate it into a linear hybrid automaton, which has a great analysis power. In this way, we associate the modelling power of HPNs to the analysis power of hybrid automata. This paper is focused on the control synthesis, for more details on HPNs and hybrid automata the reader may refer to the basic papers [DAV, 10], [ALU, 95].

The rest of the paper is organized as follows: Section 2 gives an intuitive presentation of the ideas developed in this paper. Section 3 will be devoted to modelling tools; namely: D-elementary HPNs and hybrid automata, as well as their use in the process description. Section 4 corresponds to the main contribution on this paper; it is divided into two parts. In the first part we will show how to model specifications imposed on the system, and in the second part we will explain the proposed approach to solve the control problem. The problem is formally solved for a location and some directions are given for the synthesis of the whole controller. Finally a conclusion and some perspectives will be given in Section 5.

2. INTUITIVE PRESENTATION

For the control approach that we propose here we start by considering a system whose abstraction is a hybrid dynamic system, and we want to restrict its continuous dynamics by acting on the discrete variables. These dynamics are expressed in term of constraints (also called specifications) over the continuous variables. Restricting the reachable state space needs variables such that the control of these variables will prohibit the state space to reach any undesirable value.

We set here the fundamental hypothesis: 1) the control points are the discrete variables, more precisely the occurrence dates of controllable events, and 2) there is no control in the continuous part. The controller synthesis will take advantage on the coupling between the discrete and continuous dynamics. This is illustrated in the following example.

**Example:** Consider a producer consumer system composed of a machine that supplies a buffer with a production rate of 20 parts/\( \text{mn} \) (Figure 1.a). The buffer is used to satisfy a demand of 13 parts/\( \text{mn} \). Stop or start the machine is effective after a delay of 2 mns. Initially the buffer contains 50 parts.

In this example, it is obvious that the number of parts in the buffer can be infinite, i.e. the firing of \( T_3 \) is infinitely delayed. Let us suppose that we want to impose to the buffer level to never exceed 100 parts (specification). In order to control the buffer level we must act on the stop date of the machine. Even for this simple case it is difficult to calculate the firing instants of the discrete transitions so that the specification given above is always verified, for any location. The specification presented in this example comes to limit the state space reached by the continuous variable. The translation of a D-elementary HPN in linear hybrid automata allows calculating the reachable space using specialized software like PHAVer (Figure 2.a). In this paper our control objective is to determine formally the new guards of discrete transitions, so that specifications are met. This control must be maximally permissive as we will show in the sequel; This is indicated in the unfolded hybrid automaton in Figure 2.b.

Our control synthesis approach is based on the following three steps:

1. **Modelling the system without constraints by a D-elementary HPN;**
2. **Translating the D-elementary HPN in a linear hybrid automaton;**
3. **Modelling the specifications and computing the new transitions guards that ensure the specifications compliance;**

These three steps are summarized in Figure 3 below. Each block in this Figure corresponds to a step. The two first steps correspond to previous results and the contribution of this paper corresponds to the third step.

3. MODELING OF POSITIVE HYBRID DYNAMIC SYSTEMS

This section corresponds to the two first steps of our approach, i.e. blocks 1 and 2 in Figure 3. As previously mentioned, we are interested here with positive hybrid dynamic systems. We use D-elementary HPNs for modelling
these systems. Not only this formalism inherits all the advantages of classical PNs, but also it is well suitable for the class of systems considered here.

The continuous dynamics are linear in the sense of the linear hybrid automata. HPN is a generic name for a set of formalisms used for modelling hybrid systems. This set has in common the integration of a discrete PN and a continuous PN. D-elementary HPN is a specific HPN which integrates a T-time PN that controls a constant speed continuous PN.

![Diagram](image-url)

Figure 2. Hybrid automaton modelling the producer consumer system; a) before control; b) After control;

We have supposed that all the discrete transitions are associated with controllable events, and the continuous part is uncontrollable. This represents the current cases met in real life systems, however taking into account uncontrollable events in the discrete part is a direct extension of the proposed approach and constitutes a future research direction.
Definition 1 (D-elementary hybrid Petri nets):

A D-elementary HPN is a structure \( PN = (P, T, h, E, \Sigma, Pre, Post, U, V, m_0) \) such that:

1. \( P = \{P_1, P_2, \ldots, P_n\} \) is a finite set of \( n \) places; \( P = P^D \cup P^C \), such that \( P^D = \{P_1, P_2, \ldots, P_m\} \) is the set of \( m \) continuous places, and \( P^C = \{P_{nC+1}, P_{nC+2}, \ldots, P_n\} \) is the set of discrete places.
2. \( T = \{T_1, T_2, \ldots, T_m\} \) is a finite set of \( m \) transitions; \( T = T^C \cup T^D \), such that \( T^C = \{T_1, T_2, \ldots, T_{nc}\} \) is the set of \( nc \) continuous transitions, and \( T^D = \{T_{nc+1}, T_{nc+2}, \ldots, T_m\} \) is the set of discrete transitions.
3. \( h): P \cup T \rightarrow \{C, D\} \) defines the set of continuous nodes, \( h(x) = C \) and discrete nodes, \( h(x) = D \).
4. \( E \) is a set of events.
5. \( \Sigma: T^D \rightarrow E \) is a function that associates to each discrete transition an event from \( E \);
6. \( Pre: P \times T \rightarrow \mathcal{N} \) and \( Post: P \times T \rightarrow \mathcal{N} \) are the backward and forward incidence mappings. These mappings are such that:
   \[
   \forall (P_i, T_j) \in P^D \times T^D, \ Pre(P_i, T_j) = Post(P_i, T_j) = 0 \quad (1)
   \]
   \[
   \forall (P_i, T_j) \in P^D \times T^C, \ Pre(P_i, T_j) = Post(P_i, T_j) \quad (2)
   \]
7. \( U): E \rightarrow \mathcal{R} \times (\mathcal{R}^+ \cup \{\infty\}) \) associates each event \( e_j \) with its firing interval \( [a_j, b_j] \).
8. \( V): T^C \rightarrow \mathcal{R} \) associates a maximal firing speed \( V_j \) to each C-transition \( T_j \).
9. \( m_0 = \left( \begin{array}{c} m_{0C} \\ m_{0D} \end{array} \right) \) is the initial marking vector; it is the concatenation of \( m_{0C} \), the real-valued continuous palaces initial marking, and \( m_{0D} \) the natural-valued discrete places initial marking.

The condition (1) of point 6, in the above definition means that no arcs connect continuous places to discrete transitions. Physically, that means that the continuous dynamics has no influence on the discrete dynamics. This last has a fully independent behaviour. The condition (2) in the same point means and if an arc connects a discrete Place \( P_i \) to a continuous transition \( T_j \), the arc connecting \( T_j \) to \( P_i \) must exist. This appears graphically as loops connecting discrete places to continuous transitions. Physically this means that the discrete dynamics controls the evolution of the continuous dynamics. In a D-elementary hybrid Petri net, there is no transformation of tokens, neither from the continuous state to the discrete state, nor in the other sense.

A HPN combines a discrete and a continuous PN, its state at time \( \theta \) is given by the state of the two models. This strong coupling makes it hard to analyse the hybrid model. The translation of HPNs in hybrid automata allows using formal tools developed for the latter; and it is possible thereby to combine the description power of HPNs to the analysis power of hybrid automata.

Hybrid automata were introduced by Alur et al. (1995). They are the most general formalism to model hybrid dynamic systems, in the sense that they can model the largest range of this class of systems. They combine: differential equations, which are the basic description model of continuous dynamic systems, and finite state automata that represent the basic model for describing discrete event systems. This is why the hybrid automata are the most used formalism in model-checking and controller synthesis algorithms for hybrid dynamic systems.

To be able to take advantage both from of HPNs and hybrid automata, a translation approach was developed in (Ghomari and Alla, 2007). This approach allows translating D-elementary HPNs into hybrid automata. The translation approach has been proven formally in (Ghomari and Alla, 2013). It is presented in appendix A.

The translation of D-elementary HPN in figure 1.b gives the hybrid automaton of figure 4, that we label elementary hybrid automaton in this paper. It has particular properties, as detailed in the following definition. For a general definition of a linear hybrid automaton, the reader may refer to (Henzinger, 1996).

![Figure 4: Hybrid automaton resulting from the translation of the D-elementary HPN in figure 1.b.](image)

**Definition 3 (Elementary hybrid automata):** An elementary hybrid automaton \( A = (\text{Loc}, X, E, \delta, F, \text{Inv}) \) such that:

1. \( \text{Loc} = \{q_1, q_2, \ldots\} \) is a finite set of locations;
2. \( X = \left( \begin{array}{c} X_C \\ X_D \end{array} \right) \) is the continuous state space. It is composed of two vectors: \( X_C = (m_1, m_2, \ldots, m_{nc})^T \) is the vector of \( nc \) real-valued variables modelling the continuous places marking; and \( X_D = \{t_1, t_2, \ldots, t_k\} \) is the vector of clocks corresponding to enabled transitions. A valuation \( v \) is a function that assigns a real-valued \( v(x) \in \mathcal{R}^n \) to each variable \( x \in X \).
3. $E$ is a set of events;
4. $\delta$ is a finite set of transitions, each transition is a quintuple $T = (q, a, g, \gamma, \alpha)$ such that:
   - $q \in \text{Loc}$ is the source location;
   - $a \in E$ is the event associated to the firing of $T$;
   - $g$ is the transition guard, it is a linear predicate on $X$; a transition can be fired whenever its guard is satisfied.
   - $\text{Init}$ is a reset function that affects a linear expression to variables of $X$ when taking the corresponding transition;
   - $q' \in \text{Loc}$ is the target location;
5. $F$ is a function that assigns to each location a continuous linear vector field on $X$. While in discrete location $q$, the continuous variables $m_i \in \mathbf{m_c}$ evolve according to a differential equation of the form $\dot{m}_i = B_i$, where $B_i$ is a the dynamic balance of the continuous place $P_i$ and the clocks $t_i \in T_0$ evolves according to the differential equation $t_i' = 1$.
6. $\text{Inv}$ is a function that affects to each location $q$, a linear predicate $\text{Inv}(q)$ that must be satisfied by the continuous variables in order to stay in the location $q$.

A state of a HA is a pair $(q, v)$ consisting of a location $q$ and a valuation of continuous variables $v$. This state can evolve in two manners: either continuously; in this case the continuous variables $v$. This state can evolve in transition in the HPN. They have guards of the form $\alpha \leq t_i \leq \beta_i$, where $\alpha_i$ and $\beta_i$ are real constants. Uncontrollable transitions correspond to emptying a continuous place marking. They have guards of the form $m_i = 0$.

The controller synthesis consists first in solving the problem for a location $q$ and then to iterate the procedure for the whole elementary hybrid automaton. Figure 5.a gives the structural representation of a location; we suppose, in a first time, that $q$ has only one output transition, which is controllable. The guard of a controllable transition is only function of one clock $t_i$. This guard is of the form: $\alpha_i \leq t_i \leq \beta_i$, where $\alpha_i$ and $\beta_i$ are real constants. Figure 5.b schematizes $R(A, q)$ the reachable state space in location $q$, which depends on the input state space $R(A, q)$, the vector Field $F(q)$ and on the invariant $\text{Inv}(q)$. We will resume the location $q$ throughout this paper to explain the controller synthesis procedure and to indicate the consequences of each step of the procedure on the reachable state space in $q$.

![Diagrams](image)

Figure 5.a) a location $q$ from the elementary hybrid automaton, b) reachable state space in location $q$.

4. CONTROLLER SYNTHESIS

The controller synthesis consists in realizing the last step described in block 3 of Figure 3. It will be solved in this paper for a location; therefore we will develop in this section the calculation of the new guards so that the location state space verifies the specifications. Some directions will be given to extend this hard computation problem in order to determine the final controller.

4.1 Control Specifications Modelling

In hybrid dynamic systems, specifications can be imposed on the discrete part or on the continuous part. We consider here specifications on the continuous behaviour; we think that it is the most currently met cases in real life systems. It means that specifications are only related to continuous variables $m_i$ and never on clocks. They have the form of linear inequalities.

**Definition 3 (Specification):** Let $S = (s_1, s_2, \ldots, s_m)^T$ be a real-valued constant vector and $b$ be a real constant. A
specification Spec on the continuous behaviour of the elementary hybrid automaton is a predicate of the form:

\[ S_1m_1 + S_2m_2 + \ldots + S_{nc}m_{nc} \leq b \]

Then \( S^T M_c \leq b \)

Let us recall that \( M_c \) is a vector of \( nc \) real-valued variables modelling the continuous places marking (Definition 2).

We note by \( \text{Spec(q)} \) the general form of a specification, it is the conjunction of all the specifications imposed on the state space in location \( q \):

\[ \text{Spec(q)} = \text{Spec}_1 \wedge \text{Spec}_2 \wedge \ldots \wedge \text{Speck} \]

A continuous specification \( \text{Spec}(q) \) is a set of constraints on the continuous space reached by the HA in any discrete location \( q \). The automaton must stay in \( q \) if \( \text{Spec}(q) \) is satisfied, and must leave \( q \) by firing a transition before the violation of \( \text{Spec}(q) \). Of course, a specification can be different from a location to another one.

An example is given for location \( q_1 \) described in Figure 6. Thanks to PHAVer software, the computation of the reachable space in this location is obtained. It is determined by the following set of inequalities:

\[
R(A, q_1) = \begin{cases} 
  m_1 - 3t \geq 39 \\
  m_2 - t \geq 50 \\
  3m_2 - m_1 \geq 108 \\
  m_2 \geq 53 \\
  m_1 - 3m_2 \geq 114 \\
  m_1 \geq 48 \\
  m_2 - t \leq 53 \\
  t \geq 1 \\
  m_1 - 3t \leq 48 \\
  m_2 \leq 100
\end{cases}
\]

Let us add the following specification: \( m_1 - m_2 \leq 20 \), which must be verified at any time. The state space above contains a subset of values that violate this specification. The goal of the next section is to compute the new output guard so that this specification is always verified.

### 4.2 Control computation

The controller is obtained by modifying the guards on transitions of the elementary hybrid automaton such as the specification are verified in the maximally permissive way. This corresponds to block 3 of Figure 3. This calculation is made using the reachable state space of automaton \( A \).

Let us consider the general case of a location \( q \) from \( A \) (Figure 7.a) and the controllable output transition \( T_i \) whose guard is function of the clock \( t_i \): \( a_i \leq t_i \leq b_i \). The addressed problem is to calculate the largest interval \( [a^{'}, b^{'}, \ldots] \subseteq \{a, b\} \) in the dynamic behaviour of the elementary hybrid automaton, i.e. the guard \( a^{'}, b^{'}, \ldots \) that allows to meet the specifications.

For any location \( q \) in \( A \). We note by:

\[ C = (c_1, c_2, \ldots, c_{nc})^T \]

the vector of the \( m_i \) variables slopes (dynamics of the markings),

\[ D_{\text{linmin}} = (d_{\text{linmin}_1 \text{linmin}_2 \ldots \text{linmin}_{nc})}^T \]

the vector of minimal input values in \( R(A, q) \),

\[ D_{\text{linmax}} = (d_{\text{linmax}_1 \text{linmax}_2 \ldots \text{linmax}_{nc})}^T \]

the vector of maximal input values in \( R(A, q) \),

\[
S = \begin{pmatrix} 
\min(s_{1, 0}) \\
\min(s_{2, 0}) \\
\vdots \\
\min(s_{nc, 0})
\end{pmatrix} \quad \text{and} \quad S^* = \begin{pmatrix} 
\max(s_{1, 0}) \\
\max(s_{2, 0}) \\
\vdots \\
\max(s_{nc, 0})
\end{pmatrix}
\]

Where \( S \) is the vector of specification decomposed into 2 sub vectors \( S' \) and \( S^* \).

Such that: \( S = S' + S^* \);

\( d_{\text{max}} \) is the maximal sojourn time in \( R(A, q) \).
Such that:

The constrained behaviour

\[ \text{Spec}(q) \land R(A, q) \]

\[ S \]

\[ R \]

\[ \beta \]

\[ T \]

\[ D \]

\[ \alpha \]

\[ t \]

\[ \beta_i \]

\[ \alpha' \]

\[ \beta' \]

\[ t_{\text{min}} \]

\[ t_{\text{max}} \]

\[ \text{Spec}(q) \]

\[ R(A, q) \]

\[ \text{Inv}(q) \]

\[ \hat{X} = F(X) \]

\[ T_b \alpha_i \leq t_i \leq \beta_i \]

\[ \\text{Proof:} \]

The value of \( t_{\text{min}} \) is obvious; it is given by the reachable space from location \( q \) in \( A \).

The value of \( t_{\text{max}} \) is derived from the specification:

\[ S^T M_C \leq b \]

As (from property 1):

\[ M_C = C(t - t_0) + D_0 \]

We can write:

\[ S^T (C(t - t_0) + D_0) \leq b \]

The sign of the scalar \( S^T C \) is very important in the calculation of the maximal value of the time upper bound. It combines the weights in the specification with the slopes of the marking. Depending on the sign of \( S^T C \), three cases must be distinguished:

1. (case): \( S^T C > 0 \) \( \Rightarrow \) \( t = \frac{b + s^T C t_0 - s^T D_0}{s^T C} \)

is the condition verifying the specification.

And the more constraining bound for \( t \) is:

\[ t_i = \frac{b + s^T C t_0 - s^T D_{\text{min}} - s^T D_{\text{max}}}{s^T C} \]

- If \( t_i < \alpha' \), then the new guard is empty and location \( q \) is forbidden.

- If \( t_i \geq \alpha' \), then \( t_{\text{max}} = t_i \) and \( \beta'_i = \min (\beta_i, t_{\text{max}}) \)

2. (case): \( S^T C < 0 \) \( \Rightarrow \) \( t = \frac{b + s^T C t_0 - s^T D_0}{s^T C} \)

is the condition verifying the specification.

And the more constraining bound for \( t \) is:

\[ t_i = \frac{b + s^T C t_0 - s^T D_{\text{min}} - s^T D_{\text{max}}}{s^T C} \]

- If \( t_i > t_0 \), then the specification is not verified for the entrance space and location \( q \) is forbidden.

- If \( t_i \leq t_0 \), then \( t_{\text{max}} = d_{\text{max}} \) (maximal sojourn time for location \( q \) in \( A \))

and \( \beta'_i = \min (\beta_i, t_{\text{max}}) \)

3. (case): \( S^T C = 0 \)

- If \( S^T D_0 \leq b \), then the specification is always verified

- If \( S^T D_0 > b \), then the specification is not verified for some values of the entrance marking in location \( q \) which is forbidden.

\[ q \]

\[ T_b \alpha_i \leq t_i \leq \beta_i \]
Figure 7 illustrates the two first cases of Theorem 1. In Figure 8.a, the specification:

\[ m_1 - m_2 \leq 20 \]

\[ S^T = (1 -1), \quad S^T.C = (1 -1) \left( \frac{3}{1} \right) = 2 \]

Then \( t_{\text{max}} = \frac{20+2-51+53}{2} = 12 \) and the new guard is \([3, 12]\).

In Figure 8.b, the specification \( m_1 - m_2 \leq 20 \) gives

\[ S^T = (-1 1), \quad S^T.C = (-1 1) \left( \frac{3}{1} \right) = -2 \]

Then \( t_{\text{max}} = \frac{20-2-53+51}{-2} = -8 \); in that case, it is necessary to compute the maximal sojourn time in the location, \( d_{\text{max}} = \text{Min} \left( \frac{100-48}{3}, 100 - 53 \right) = 17.33 \). The new guard is \([3, 17.33]\).

\[ \begin{array}{c|c|c}
\text{Spec} & \text{Precision} & \text{Guard} \\
\hline
m_1 \in [48 \, 51] & q_1 & m_1 = 3, m_2 = 1, t = 1, m_1 \leq 100, m_2 \leq 100 \\
\hline
m_2 \in [53 \, 54] & \text{} & \end{array} \]

\[ 3 \leq t \leq 12 \]

\[ \begin{array}{c|c|c}
\text{Spec} & \text{Precision} & \text{Guard} \\
\hline
m_1 \in [48 \, 51] & q_1 & m_1 = 3, m_2 = 1, t = 1, m_1 \leq 100, m_2 \leq 100 \\
\hline
m_2 \in [53 \, 54] & \text{} & \end{array} \]

\[ 3 \leq t \leq 17.33 \]

**Example:**

Let us consider again the producer consumer system and the specification that imposes to never exceed 100 parts in the buffer. The controller guarantying to meet this specification is schematized in Figure 9. This controller is built location by location applying Theorem 1. A control is computed for one input state pace, called a visit. It is why he final controller given in Figure 9 is an unfolded automaton. This model is only timed since the control points are given by the discrete timed transitions. The continuous dynamic does not appear, its influence is taken into account in the computation of the guard. By acting on the discrete controls Start and Stop in the control timed intervals, the respect of the specification is guarantied.

5. CONCLUSION

In this paper, we have presented a controller synthesis technique for hybrid dynamic systems modelled by D-elementary HPNs. This model is first translated in a hybrid automaton before controller synthesis. We have highlighted our contribution on the computation of the control for a location. Specifications are added to the hybrid automaton in order to limit its reachable state space to a desired one.

This is obtained thanks to the computation of new guards associated with the controllable discrete transitions. We have determined algebraic formulas giving the new time bounds that allow respecting the specifications. This computation needs the bounds of the different variables for the input state space, which are obtained by linear programming. The controller is a timed automaton; it is optimal in the sense that it gives the most permissive state space of the system.
guarantying the specifications. This optimality is obtained via an algebraic calculation.

Figure 9. Controller of the consumer producer system.

The obtained timed automaton is the model of the closed loop system. It represents the system coupled to its controller, and has a nondeterministic behaviour. It is possible to choose one deterministic behaviour from an infinite number. It is also possible to calculate an optimal behaviour according to a given criterion.

Our future work consists in: 1) establishing a general algorithm for the automatic determination of the controller, starting from any D-elementary HPN; 2) Taking into account uncontrollable events in the discrete part; and 3) characterizing formally the sub class of HPNs such that the translation algorithm terminates always.

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Appendix A: Translation approach.

The translation approach can be summarized in the following three steps:

1. Isolate the discrete PN of the hybrid model and construct its equivalent timed automaton: Several algorithms permitting the translation of a Time PN in a timed automaton exist in the literature. We reuse the existing algorithms for this first step of our approach. Locations of the resulting timed automaton are said macro-locations in the following.

2. Construct the hybrid automaton corresponding to each macro-location of the timed automaton resulting from the previous step: Indeed, a macro-location represents a marking of the discrete PN, and therefore a configuration (set of locations) of the system. In this second step we determine the continuous PN corresponding to each macro-location and we translate it in a hybrid automaton. At the end of this second step we have a hierarchical model containing macro-locations, each of which comprises a hybrid automaton.

3. Replace transitions between macro-locations by transitions between internal locations: This last step transforms the hierarchical model of the previous step in a linear hybrid automaton.

To illustrate the different steps of the translation approach, let us consider again the D-elementary HPN representing the producer consumer system in Figure A1.a.

In Figure A1.b below we isolate the discrete part of the HPN of Figure A1.a. The time PN obtained has two reachable markings and only one transition enabled at a time. It is modelled by a timed automaton with two locations, that represent the markings, and one clock, representing the time during which a transition is enabled (Figure A1.c).

To each marking of the discrete part (time Petri net) corresponds a configuration of the system. Each configuration corresponds to one or several locations. The running of a constant speed continuous PN can be modelled by a linear hybrid automaton where the state variables are the continuous places marking.

To illustrate the different steps of the translation approach, let us consider again the D-elementary HPN representing the producer/consumer system in Figure A1.a.

In Figure A1.b below we isolate the discrete part of the HPN of Figure A1.a. The time PN obtained has two reachable markings and only one transition enabled at a time. It is modelled by a timed automaton with two locations, that represent the markings, and one clock, representing the time during which a transition is enabled (Figure A1.c).

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Figure A1.a. HPN of the producer-consumer system; b. T-Time Petri net of the discrete part; c. Equivalent hybrid automaton.

Only one type of events can change the continuous variables derivative, this autonomous event is the emptiness of a continuous place marking. This is always an uncontrollable event. For the HPN in Figure A1.a, the discrete part has 2 reachable markings, then 2 configurations for the continuous part. Figure A2 represents the hierarchical form of the hybrid automaton.

The final form of the hybrid automaton is obtained by replacing each transition between macro-locations by transitions between internal locations. A reachability analysis can be useful to eliminate the unreachable locations and non fireable transitions. The hybrid automaton resulting from the translation of the HPN in figure A1.a is represented in Figure A4.
Figure A2. Continuous behaviour corresponding to each discrete marking.

Figure A3. Hierarchical form of the hybrid automaton

Figure A4. Hybrid automaton resulting from the translation approach