

# Analyzing Flexible Timeline-based Plans

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**Abstract.** Timeline-based planners have been shown quite successful in addressing real world problems. Nevertheless they are considered as a niche technology in AI P&S research as an application synthesis with such techniques is still considered a sort of “black art”. Authors are currently developing a knowledge engineering tool around a timeline-based problem solving environment; in this framework we aim at integrating verification and validation methods. This work presents a verification process suitable for a timeline-based planner. It shows how a problem of flexible temporal plan verification can be cast as model-checking on timed game automata. Additionally it provides formal properties and checks the effectiveness of the proposed approach with a detailed experimental analysis.

## 1 INTRODUCTION

A key aspect for technological development of AI Planning and Scheduling (P&S) stems in the design and implementation of effective knowledge engineering environments to support application development. In the past, several planning systems were endowed with development environments to facilitate application design. More recent examples of such environments are EUROPA [5] and ASPEN [10]. Such environments can be enhanced in several ways. In a recent line of work we have envisaged the synthesis of knowledge engineering environments in which constraint-based and validation and verification (V & V) techniques both contribute to effective P&S. In particular, we are working on timeline-based planning. A known problem in timeline-based planning as used in [9, 5, 10] is the connection with plan execution which is instrumental in several real domains. Such architectures return an envelope of potential solutions in form of a flexible plan which is commonly accepted to be less brittle of a single plan when coping with execution. But the general formal properties of such a representation are far from being statically defined. Some aspects of such plans have been studied by working on the temporal network which is underlying the constraint based plan representation often used by such systems – see for example [11, 8]. We are currently tackling the more general problem of flexible temporal plan verification. This is an open issue in timeline-based planning and we are not aware about other approaches in the literature. In [4], we proposed the use of a model checker (UPPAAL-TIGA) based on the TGA formalism to verify flexible plans. Here, that initial work is put on a firmer ground by: (a) showing formal properties of the translation of the flexible plan verification problem as model-checking on time game automata; (b) introducing a benchmark problem which is realistic and rich enough to allow experiments along different directions; (c) presenting a first complete experimental analysis showing

that the approach based on model checking is feasible and requires time constants that are acceptable for static analysis.

## 2 PRELIMINARIES

This section shortly presents the two basic ingredients we combine in our knowledge engineering environment: timeline-based planning and timed game automata. In [1], the same ingredients are put together to propose a method for (non flexible) plan synthesis and not for flexible temporal plan verification. In [6], UPPAAL is considered for domain modeling and analysis, not addressing plan verification and not considering the TGA formalism.

### 2.1 Timeline-Based Planning and Execution

Timeline-based planning [9] is an approach to temporal planning which has been applied in the solution of several real world problems. The approach pursues a general idea that planning and scheduling consist in the synthesis of desired temporal behavior for complex physical systems. The set of features of a domain that needs control are modeled as a set of temporal functions whose values over a time horizon have to be planned for. Such functions are synthesized during problem solving by posting planning decisions. The evolution of a single temporal feature over a time horizon is called the *timeline* of that feature. In this paper, the time varying features are called multi-valued *state variables* as in [9]. As in classical control theory, the evolution of controlled features are described by causal laws which determine legal temporal evolution of timelines. Such causal laws are specified for the state variables in a *domain specification* which identifies the operational constraints in a given domain. In this context, the task of a planner is to find a sequence of control decisions that bring the variables into a final desired set of evolutions always satisfying the domain specification. We assume that the temporal features represented as state-variables have a finite set of possible values assumed over temporal intervals. The temporal evolutions are sequences of operational states – i.e., stepwise constant functions of time. Operational constraints specify which value transitions are allowed, the duration of each valued interval (i.e., how long a given operational status can be maintained) and synchronization constraints between different state variables.

More formally, a state variable is defined by a tuple  $\langle \mathcal{V}, \mathcal{T}, \mathcal{D} \rangle$  where: (a)  $\mathcal{V} = \{v_1, \dots, v_n\}$  is a finite set of *values*; (b)  $\mathcal{T} : \mathcal{V} \rightarrow 2^{\mathcal{V}}$  is the *value transition* function; (c)  $\mathcal{D} : \mathcal{V} \rightarrow \mathbb{N} \times \mathbb{N}$  is the *value duration* function, i.e. a function that specifies the allowed duration of values in  $\mathcal{V}$  (as an interval  $[lb, ub]$ ). (b) and (c) specify the operational constraints on the values in (a).

A *planning domain* is defined as a set of state variables  $\{\mathcal{SV}_1, \dots, \mathcal{SV}_n\}$ . They cannot be considered as reciprocally decoupled, but a set of additional relations exist, called *synchronizations*, modeling the existing temporal and causal constraints among the values taken by different state variable timelines (i.e., patterns of legal

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occurrences of the operational states across the timelines). More formally, a synchronization has the form

$$\langle \mathcal{T}\mathcal{L}, v \rangle \longrightarrow \langle \{ \langle \mathcal{T}\mathcal{L}'_1, v'_1 \rangle \dots, \langle \mathcal{T}\mathcal{L}'_n, v'_n \rangle \}, \mathcal{R} \rangle$$

where:  $\mathcal{T}\mathcal{L}$  is the reference timeline;  $v$  is a value on  $\mathcal{T}\mathcal{L}$  which makes the synchronization applicable;  $\{ \langle \mathcal{T}\mathcal{L}'_1, v'_1 \rangle \dots, \langle \mathcal{T}\mathcal{L}'_n, v'_n \rangle \}$  is a set of target timelines on which some values  $v'_j$  must hold; and  $\mathcal{R}$  is a set of relations which bind temporal occurrence of the reference value  $v$  with temporal occurrences of the target values.

**Timeline based planning.** The temporal evolutions of a state variable will be described by means of *timelines*, that is a sequence of state variable values, a set of ordered transition points between the values and a set of distance constraints between transition points. When the transition points are bounded by the planning process (lower and upper bounds are given for them) instead of being exactly specified, as it happens in case of a least commitment solving approach for instance, we refer to the timeline as *time flexible* and to the plan resulting from a set of flexible timeline as a *flexible plan*.

A *plan* is defined as a set of timelines  $\{ \mathcal{T}\mathcal{L}_1, \dots, \mathcal{T}\mathcal{L}_n \}$  over the same interval for each state variable. The process of *solution extraction* from a plan is the process of computing (if exists) a *valid* and completely specified set of timelines from a given set of time-flexible timelines. A solution is *valid* with respect to a domain theory if every temporal occurrence of a reference value implies that the related target values hold on target timelines presenting temporal intervals that satisfy the expected relations.

**Plan execution.** During plan execution the plan is under responsibility to an executive program that forces value transitions over timelines. A well known problem with execution is that not all the value transitions are under responsibility of the executive but event exists that are under control of *nature*. As a consequence, an executive cannot completely predict the behavior of the controlled physical system because the duration of certain processes or the timing of exogenous events is outside of its control. In such cases, the values for the state variables that are under the executive scope should be chosen so that they do not constrain uncontrollable events. This is the *controllability problem* defined, for example, in [11] where *contingent* and *executable* processes are distinguished. The contingent processes are not controllable, hence with uncertain durations, instead the executable processes are started and ended by the executive system. Controllability issues underlying a plan representation have been formalized and investigated for the Simple Temporal Problems with Uncertainty (STPU) representation in [11] where basic formal notions are given for *dynamic* controllability (see also [8]). In the timeline-based framework, we introduce the same controllability concept defined on STPU as follows. Given a plan as a set of flexible timelines  $\mathcal{P}\mathcal{L} = \{ \mathcal{T}\mathcal{L}_1, \dots, \mathcal{T}\mathcal{L}_n \}$ , we call *projection* the set of flexible timelines  $\mathcal{P}\mathcal{L}' = \{ \mathcal{T}\mathcal{L}'_1, \dots, \mathcal{T}\mathcal{L}'_n \}$  derived from  $\mathcal{P}\mathcal{L}$  setting to a fixed value the temporal occurrence of each uncontrollable timepoint. Considering  $N$  as the set of controllable flexible timepoints in  $\mathcal{P}\mathcal{L}$ , a *schedule*  $T$  is a mapping  $T : N \rightarrow \mathbb{N}$  where  $T(x)$  is called *time* of timepoint  $x$ . A *schedule* is *consistent* if all value durations and synchronizations are satisfied in  $\mathcal{P}\mathcal{L}$ . The *history* of a timepoint  $x$  w.r.t. a schedule  $T$ , denoted by  $T\{ \prec x \}$ , specifies the time of all uncontrollable timepoints that occur prior to  $x$ . An *execution strategy*  $S$  is a mapping  $S : \mathcal{P} \rightarrow \mathcal{T}$  where  $\mathcal{P}$  is the set of projections and  $\mathcal{T}$  is the set of schedules. An execution strategy  $S$  is *viable* if  $S(p)$  (denoted also  $S_p$ ) is consistent for each projection  $p$ . Thus, a flexible plan  $\mathcal{P}\mathcal{L}$  is *dynamically controllable* if there exists a viable execution strategy  $S$  such that  $S_{p1}\{ \prec x \} = S_{p2}\{ \prec x \} \Rightarrow S_{p1}(x) = S_{p2}(x)$

for each controllable timepoint  $x$  and projections  $p1$  and  $p2$ .

## 2.2 Timed Game Automata

Timed game automata (TGA) model have been introduced in [7] to model control problems on timed systems. Here, we briefly recall some of them that we shall use in the rest of the paper.

**Definition 1** A **Timed Game Automaton** is a tuple  $\mathcal{A} = (Q, q_0, \text{Act}, X, \text{Inv}, E)$  where:  $Q$  is a finite set of locations;  $q_0 \in Q$  is the initial location;  $\text{Act}$  is a finite set of actions split in two disjoint sets,  $\text{Act}_c$  the set of controllable actions and  $\text{Act}_u$  the set of uncontrollable actions;  $X$  is a finite set of a nonnegative, real-valued variables called *clocks*;  $\text{Inv} : Q \rightarrow B(X)$  is a function associating to each location  $q \in Q$  a constraint  $\text{Inv}(q)$  (the invariant of  $q$ );  $E \subseteq Q \times B(X) \times \text{Act} \times 2^X \times Q$  is a finite set of transitions. Where  $B(X)$  is the set of constraints in the form  $x \sim c$ , where  $c \in \mathbb{Z}$ ,  $x, y \in X$ , and  $\sim \in \{ <, \leq, \geq, > \}$ . We write  $q \xrightarrow{g, a, Y} q' \in E$  for  $(q, g, a, Y, q') \in E$ .

A *state* of a TGA is a pair  $(q, v) \in Q \times \mathbb{R}_{\geq 0}^X$  that consists of a discrete part and a valuation of the clocks (i.e., a value assignment for each clock in  $X$ ). An *admissible* state for a  $\mathcal{A}$  is a state  $(q, v)$  s.t.  $v \models \text{Inv}(q)$ . From a state  $(q, v)$  a TGA can either let time progress or do a discrete transition and reach a new state.

A *time transition* for  $\mathcal{A}$  is 4-tuple  $(q, v) \xrightarrow{\delta} (q, v')$  where  $(q, v) \in S$ ,  $(q, v') \in S$ ,  $\delta \in \mathbb{R}_{\geq 0}$ ,  $v' = v + \delta$ ,  $v \models \text{Inv}(q)$  and  $v' \models \text{Inv}(q)$ . That is, in a time transition a TGA does not change location, but only its clock values. Note that all clock variables are incremented by the same amount  $\delta$  in valuation  $v'$ . This is why variables in  $X$  are named *clocks*. Accordingly,  $\delta$  models the *elapsed time* during the time transition.

A *discrete transition* for  $\mathcal{A}$  is 5-tuple  $(q, v) \xrightarrow{a} (q', v')$  where  $(q, v) \in S$ ,  $(q', v') \in S$ ,  $a \in \text{Act}$  and there exists a transition  $q \xrightarrow{g, a, Y} q' \in E$  s.t.  $v \models g$ ,  $v' = v[Y]$  and  $v' \models \text{Inv}(q')$ . In other words, there is a discrete transition (labeled with  $a$ ) from state  $(q, v)$  to state  $(q', v')$  if the clock values (valuation  $v$ ) satisfy the *transition guard*  $g$  and the clock values after resetting the clocks in  $Y$  (valuation  $v'$ ) satisfy the invariant of location  $q'$ . Note that an admissible transition always leads to an admissible state and that only clocks in  $Y$  (reset clocks) change their value (namely, to 0).

A *run* of a TGA  $\mathcal{A}$  is a finite or infinite sequence of alternating time and discrete transitions of  $\mathcal{A}$ . We denote with  $\text{Runs}(\mathcal{A}, (q, v))$  the set of runs of  $\mathcal{A}$  starting from state  $(q, v)$  and write  $\text{Runs}(\mathcal{A})$  for  $\text{Runs}(\mathcal{A}, (q, \vec{0}))$ . If  $\rho$  is a finite run, we denote with  $\text{last}(\rho)$  the last state of run  $\rho$  and with  $\text{Duration}(\rho)$  the sum of the elapsed times of all time transitions in  $\rho$ .

A *network* of TGA (nTGA) is a finite set of TGA evolving in parallel with a CCS style semantics for parallelism. Namely, at any time, only one TGA in the network can change location, unless a synchronization on labels takes place. In the latter case, the two automata synchronizing on the same label move together. Note that time does not elapse during synchronizations.

Given a TGA  $\mathcal{A}$  and three symbolic configurations *Init*, *Safe*, and *Goal*, the *reachability control problem* or reachability game  $RG(\mathcal{A}, \text{Init}, \text{Safe}, \text{Goal})$  consists in finding a *strategy*  $f$  such that  $\mathcal{A}$  starting from *Init* and supervised by  $f$  generates a winning run that stays in *Safe* and enforces *Goal*.

A *strategy* is a partial mapping  $f$  from the set of runs of  $\mathcal{A}$  starting from *Init* to the set  $\text{Act}_c \cup \{ \lambda \}$  ( $\lambda$  is a special symbol that denotes "do nothing and just wait"). For a finite run  $\rho$ , the strategy  $f(\rho)$  may say (1) no way to win if  $f(\rho)$  is undefined, (2) do nothing, just wait

in the last configuration  $\rho$  if  $f(\rho) = \lambda$ , or (3) execute the discrete, controllable transition labeled by  $l$  in the last configuration of  $\rho$  if  $f(\rho) = l$ .

The restricted behavior of a TGA  $\mathcal{A}$  controlled with some strategy  $f$  is defined by the notion of *outcome*. The outcome  $Outcome(q, f)$  is defined as the subset of  $Runs(\Pi, \mathcal{A})$  that can be generated from  $q$  executing the uncontrollable actions in  $Act_u$  or the controllable actions provided by the strategy  $f$ . A *maximal run*  $\rho$  is either an infinite run or a finite run that satisfies either i)  $last(\rho) \models Goal$  or ii) if  $\rho \xrightarrow{a}$  then  $a \in Act_u$  (i.e. the only possible next discrete actions from  $last(\rho)$ , if any, are uncontrollable actions). A strategy  $f$  is a *winning strategy* from  $q$  if all maximal runs in  $Outcome(q, f)$  are in  $WinRuns(q, \mathcal{A})$ . A state  $q$  in a TGA  $\mathcal{A}$  is *winning* if there exists a winning strategy  $f$  from  $q$  in  $\mathcal{A}$ .

### 3 USING nTGA TO MODEL TIMELINE-BASED PLANNING SPECIFICATIONS

In our approach, flexible timeline-based plan verification is performed by solving a Reachability Game using the UPPAAL-TIGA tool. To this end, this section describes how a flexible timeline-based plan, state variables and domain theory can be modeled using the TGA formalism. Our strategy is the following. First, timelines and state variables are mapped to TGA. Second, we model the flexible plan *view* of the world by partitioning state variables/timelines into two classes: controllable and uncontrollable. Finally, an *Observer* TGA is introduced in order to check for value constraints violations as well as synchronizations violations.

**Modeling a Planning Domain as an nTGA.** Let  $\mathcal{PD} = \{SV_1, \dots, SV_n\}$  be the set of state variables defining our planning domain. We will model each  $SV \in \mathcal{PD}$  with a TGA  $\mathcal{A}_{SV} = (Q_{SV}, q_0, Act_{SV}, X_{SV}, Inv_{SV}, E_{SV})$ . Then the set  $SV = \{ASV_1, \dots, ASV_n\}$  represents our planning domain  $\mathcal{PD}$  as an nTGA. The TGA  $\mathcal{A}_{SV}$  is defined as follows. The set  $Q_{SV}$  of locations of  $\mathcal{A}_{SV}$  is just the set  $\mathcal{V}$  of values of  $SV$ . The initial state  $q_0$ , of  $\mathcal{A}_{SV}$  is the initial value in the timeline of  $SV$ . The set of clocks  $X_{SV}$  of  $\mathcal{A}_{SV}$  consists of just one local clock:  $c_{sv}$ . The set  $Act_{SV}$  of actions of  $\mathcal{A}_{SV}$  consists of the values  $\mathcal{V}$  of  $SV$ . If  $SV$  is controllable then the actions in  $Act_{SV}$  are controllable (i.e.,  $Act_{SV} = Act_{cSV}$ ), otherwise they are uncontrollable (i.e.,  $Act_{SV} = Act_{uSV}$ ). Location invariants  $Inv_{SV}$  for  $\mathcal{A}_{SV}$  are defined as follows:  $Inv_{SV}(v) := c_{sv} \leq ub$ , where:  $v \in Q_{SV} = \mathcal{V}$  and  $\mathcal{D}(v) = [lb, ub]$ . The set  $E_{SV}$  of transitions of  $\mathcal{A}_{SV}$  consists of transitions of the form  $v \xrightarrow{g, v', Y} v'$ , where:  $g = c_{sv} \geq lb$ ,  $Y = \{c_{sv}\}$ ,  $v \in Q_{SV} = \mathcal{V}$ ,  $\mathcal{D}(v) = [lb, ub]$ ,  $v' \in \mathcal{T}(v)$ .

**Modeling a Flexible Plan as an nTGA.** Let  $\mathcal{P} = \{\mathcal{TL}_1, \dots, \mathcal{TL}_n\}$  be a flexible plan for our planning domain  $\mathcal{PD}$ . We will model each  $\mathcal{TL} \in \mathcal{P}$  with a TGA  $\mathcal{A}_{\mathcal{TL}} = (Q_{\mathcal{TL}}, q_0, Act_{\mathcal{TL}}, X_{\mathcal{TL}}, Inv_{\mathcal{TL}}, E_{\mathcal{TL}})$ . Then,  $Plan = \{\mathcal{A}_{\mathcal{TL}_1}, \dots, \mathcal{A}_{\mathcal{TL}_n}\}$  represents  $\mathcal{P}$  as an nTGA. The TGA  $\mathcal{A}_{\mathcal{TL}}$  is defined as follows. The set  $Q_{\mathcal{TL}}$  of locations of  $\mathcal{A}_{\mathcal{TL}}$  consists of the value intervals (*plan steps*) in  $\mathcal{TL}$  along with a location  $l_{goal}$  modeling the fact that the plan has been completed. Thus,  $Q_{\mathcal{TL}} = \mathcal{TL} \cup \{l_{goal}\}$ . The initial state  $q_0$ , of  $\mathcal{A}_{\mathcal{TL}}$  is the first value interval  $l_0$  in  $\mathcal{TL}$ . The set of clocks  $X_{\mathcal{TL}}$  of  $\mathcal{A}_{\mathcal{TL}}$  consists of just one element: the *plan clock*  $c_p$ . Let  $SV$  be the state variable corresponding to the timeline  $\mathcal{TL}$  under consideration. The set  $Act_{\mathcal{TL}}$  of actions of  $\mathcal{A}_{\mathcal{TL}}$  consists of the values of  $SV$ . If  $SV$  is controllable then the actions in  $Act_{\mathcal{TL}}$  are controllable (i.e.,  $Act_{\mathcal{TL}} = Act_{c\mathcal{TL}}$ ), otherwise they are uncontrollable (i.e.,  $Act_{\mathcal{TL}} = Act_{u\mathcal{TL}}$ ). Location invariants  $Inv_{\mathcal{TL}}$  for  $\mathcal{A}_{\mathcal{TL}}$  are defined as follows. For each  $l = [lb, ub] \in \mathcal{TL}$  we define  $Inv_{\mathcal{TL}}(l) := c_p \leq ub$ . For the goal lo-

cation  $l_{goal}$  the invariant  $Inv_{\mathcal{TL}}(l_{goal})$  is identically true, modeling the fact that once plan is completed we can stay there as long as we like. The set  $E_{\mathcal{TL}}$  of transitions of  $\mathcal{A}_{\mathcal{TL}}$  consists of *intermediate* and *final* transitions. An intermediate transition has the form  $l \xrightarrow{g, v', Y} l'$ , where:  $g = c_p \geq lb$ ,  $Y = \emptyset$  with  $l$  and  $l'$  consecutive time intervals in  $\mathcal{TL}$ . A final transition has the form  $q \xrightarrow{\emptyset, \emptyset, \emptyset} q'$ , where:  $q = l_{pl}$  ( $pl$  is the plan length),  $q' = l_{goal}$ . Note that, by using state variable values as transitions label we implement the synchronization between state variables and planned timelines.

**Modeling Synchronizations with an Observer TGA.** We model synchronization between  $SV$  and  $Plan$  with an *Observer*, that is a TGA reporting an error when an illegal transition occurs.

The observer TGA  $\mathcal{A}_{Obs} = (Q_{Obs}, q_0, Act_{Obs}, X_{Obs}, Inv_{Obs}, E_{Obs})$  is defined as follows. The set of locations is  $Q_{Obs} = \{l_{ok}, l_{err}\}$  modeling *legal* ( $l_{ok}$ ) and *illegal* ( $l_{err}$ ) executions. The initial location  $q_0$  is  $l_{ok}$ . The set of actions is  $Act_{Obs} = \{a_{fail}\}$ . The set of clocks is  $X_{Obs} = \{c_p\}$ . There are no invariants, that is  $Inv_{Obs}(l)$  returns always the empty constraint. This models the fact that  $\mathcal{A}_{Obs}$  can stay in any location as long as it likes. The set  $E_{Obs}$  consists of two kind of uncontrollable transitions: *value transitions* and *sync transitions*. Let  $s_p \in \mathcal{TL}$  be a plan step and  $v_p \in SV$  its associated planned value. A value transition has the form  $l_{ok} \xrightarrow{g, a_{fail}, \emptyset} l_{err}$ , where:  $g = \mathcal{TL}_{s_p} \wedge \neg SV_{v_p}$ . Let  $\langle \mathcal{TL}, v \rangle \rightarrow \langle \{\mathcal{TL}'_1, \dots, \mathcal{TL}'_n\}, \{v'_1, \dots, v'_n\}, \mathcal{R} \rangle$  be a synchronization. A sync transition has the form  $l_{ok} \xrightarrow{g, a_{fail}, \emptyset} l_{err}$ , where:  $g = \neg \mathcal{R}(\mathcal{TL}_v, \mathcal{TL}'_{v'_1}, \dots, \mathcal{TL}'_{v'_n})$ . Note how, for each possible cause of error (illegal value occurrence or synchronization violation), a suitable transition is defined, forcing our Observer TGA to move to the error location which, once reached, cannot be left.

The nTGA  $\mathcal{PL}$  composed by the set of automata  $PL = SV \cup Plan \cup \{\mathcal{A}_{Obs}\}$  models Flexible plan, State Variables and Domain Theory descriptions.

**Theorem 1** *The nTGA  $\mathcal{PL}$  describes all and only the behaviors implemented by the flexible plan  $\mathcal{P}$ .*

**Sketch of Proof.** The network  $Plan = \{\mathcal{A}_{\mathcal{TL}_1}, \dots, \mathcal{A}_{\mathcal{TL}_n}\}$  represents all the possible planned temporal behaviors over all the timelines. In fact, each automaton  $\mathcal{A}_{\mathcal{TL}_i}$  describes the planned temporal sequence of values for the timeline  $\mathcal{TL}_i$  within the planning horizon  $\mathcal{H}$ . While, the automata in  $SV = \{\mathcal{A}_{SV_1}, \dots, \mathcal{A}_{SV_n}\}$  represent exactly the given state variables description. We recall that the use of input/output actions implements straightforward relations between allowed values and planned values for each timeline. By construction, we have a one-to-one mapping between flexible plan behaviors and automata evolutions: for each evolution in  $Plan \cup SV$ , a corresponding behavior in  $\mathcal{P}$  exists and vice versa. On one hand, any possible behavior in  $Plan \cup SV$  but not in a flexible plan, would violate temporal timepoint plan constraints. On the other hand, any possible flexible plan behavior in  $\mathcal{P}$  but not in  $Plan \cup SV$ , would violate automata guards or invariants. The Observer automaton checks for both values consistency (between planned timelines and state variables) and synchronizations satisfaction. While value consistency is trivial, again by construction, the Observer holds the error location whenever a transition guard is activated, that is, whenever the related flexible behavior violates the associated synchronization. On the opposite, whenever a flexible behavior violates a synchronization, the related guard is activated, hence enforcing the error location for the Observer. Thus,  $\mathcal{PL}$  describes all and only the possible flexible plan behaviors in  $\mathcal{P}$ .  $\square$

## 4 TIME FLEXIBLE PLAN VERIFICATION

In Theorem 1, we demonstrated by construction that we obtain a one-to-one mapping between flexible behaviors, defined by  $\mathcal{P}$ , and automata behaviors, defined by  $\mathcal{PL}$ , with the Observer automaton holding the error location if either an illegal value occurs or a synchronization is violated. Thus, the plan verification problem can be reduced to a Reachability Game by introducing a Reachability Game  $RG(\mathcal{PL}, Init, Safe, Goal)$  where  $Init$  represents the set of initial locations, one for each automaton in  $\mathcal{PL}$ ,  $Safe = \{l_{ok}\}$ , and  $Goal$  is for the set of goal locations, one for each  $\mathcal{TL}_i$  in  $\mathcal{PL}$ .

**Theorem 2** *Given  $RG(\mathcal{PL}, Init, Safe, Goal)$  defined considering  $Init = \{q \mid q \text{ is } q_0 \in Q_{\mathcal{TL}_i} \forall \mathcal{TL}_i \in Plan\} \cup \{q \mid q \text{ is } q_0 \in Q_{SV_i} \forall SV_i \in SV\} \cup \{q \mid q \text{ is } q_0 \in Q_{Obs}\}$ ,  $Safe = \{l_{ok}\}$  and  $Goal = \{l \mid l \text{ is } l_{goal} \in Q_{\mathcal{TL}_i} \forall \mathcal{TL}_i \in Plan\}$ , solving/winning the game implies plan validity for  $\mathcal{P}$ .*

**Sketch of Proof.** In Theorem 1, we show that  $\mathcal{PL}$  describes all and only the behaviors implemented by the flexible plan  $\mathcal{P}$ . If there exists a winning strategy  $f$  for  $RG(\mathcal{PL}, Init, Safe, Goal)$ , then the  $Outcome(Init, f)$  represents the subset of  $Runs(\mathcal{PL}) \subseteq WinRuns(Init, f)$  that guarantees that (i)  $Goal$  states are reached and (ii)  $Safe$  states are enforced. Then, each  $\rho \in Outcome(Init, f)$  reaches all the locations in  $\{l \mid l \text{ is } l_{goal} \in Q_{\mathcal{TL}_i} \forall \mathcal{TL}_i \in Plan\}$  while the observer holds  $l_{ok}$ . As a straightforward consequence we have that for each timeline  $\mathcal{TL}_i$ , all the transitions can be performed by maintaining allowed values (w.r.t. state variable definition) and without violating any synchronization. Thus, the plan is valid.  $\square$

**Verification in UPPAAL-TIGA.** In order to solve  $RG(\mathcal{PL}, Init, Safe, Goal)$ , we use UPPAAL-TIGA [2]. If there is no winning strategy, UPPAAL-TIGA gives a counter strategy for the opponent (environment) to make the controller lose. Given a nTGA, a set of goal states (*win*) and/or a set of bad states (*lose*), four types of winning conditions can be issued [2]. Then, to solve the reachability game, we ask UPPAAL-TIGA to check the formula  $\Phi = A [ Safe U Goal ]$  in  $\mathcal{PL}$ . In fact, this formula means that along all the possible paths,  $\mathcal{PL}$  remains in  $Safe$  states until  $Goal$  states are reached. Moreover, recalling the *dynamic controllability* definition for timelines given in Section 2.1, we may notice that each possible evolution of the uncontrollable automata corresponds to a timeline projection  $p$ . Each strategy/solution for the  $RG$  corresponds to a consistent schedule  $T$  and a set of strategy represents a viable execution strategy  $S$ . Thus, the winning strategies produced by UPPAAL-TIGA represents a viable execution strategy  $S$  for the flexible plan  $\mathcal{P}$ . Furthermore, the use of forward algorithms [2] guarantees that  $S$  is such that  $S_{p1}\{\prec x\} = S_{p2}\{\prec x\} \Rightarrow S_{p1}(x) = S_{p2}(x)$ , for each controllable timepoint  $x$  and projections  $p1$  and  $p2$ . As a consequence, we obtain the following Corollary.

**Corollary 1** *Given  $RG(\mathcal{PL}, Init, Safe, Goal)$  defined as above and using UPPAAL-TIGA to find a winning execution strategy  $S$ . If UPPAAL-TIGA solves  $RG$  then the flexible plan is dynamically controllable by means of  $S$ .*

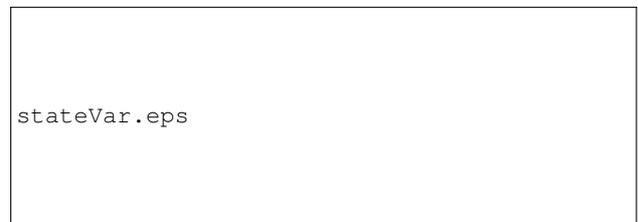
Notice that our approach to dynamic controllability checking relies on the fact that UPPAAL-TIGA works with forward algorithms.

## 5 A NEW BENCHMARK DOMAIN

In this section, we present a case study that we use in our experimental analysis. The domain is inspired by a Space Mission Long Term Planning problem as described in [3]. We consider a remote

space agent (RSA) that operates around a target planet. The RSA can either point to the planet and use its instruments to produce scientific data or point towards a communication station (Relay Satellite or Earth) and communicate previously produced data. The RSA is controlled by a planner and an executive system to accomplish the required tasks (scientific observations, communication, and maintenance activities). For each orbit followed by the RSA around the planet, the operations are split with respect to 3 orbital phases: (1) the pericentre (the orbital segment closest to the target planet); (2) the apocentre (the orbital segment farthest from the planet); (3) the orbital segments between the pericentre and apocentre. Around pericentre, the agent should point toward the planet, thus allowing observations of the planet surface (Science operations). Between pericentre and apocentre passages, the agent should point to Earth for transmitting data. Communication with Earth should occur within a ground-station availability window. Ground-station visibility can either partially overlap or fully contain a pericentre passage. Maintenance operations should occur around the apocentre passages. The RSA is also endowed with a set of scientific instruments or payloads (e.g., stereo cameras, altimeters, spectrometers, etc.) whose activities are to be planned for during the pericentre phase taking into account physical constraints. In particular, here we assume that instruments can be activated one at a time by following a fixed execution sequence of operations: warm-up, process, turn-off. Additionally, there are other constraints to be satisfied. Constraints on uplink windows frequency and duration require four hours uplink time for each 24 hours, and these uplink windows must be as regular as possible, one every about 20 hours. Apocentre slots for spacecraft maintenance windows must be allocated between 2 and 5 orbits apart, and the maintenance duration is of 90 minutes.

**Timeline-based Specification.** To obtain a timeline-based specification of the domain we use: *Planned State Variables* representing the timelines where there are activities under the agent control (they are planned for by the agent); *External State Variables*, representing values imposed over time which can only be observed



**Figure 1:** Transitions for the *planned state variables* describing the Spacecraft Operative Mode (left) and any of the Instruments (right).

*Planned State Variables.* A state variable *Spacecraft Operative Mode* specifies the observation, communication, and maintenance opportunities for the agent. In Fig. 1-left, we detail the values that can be taken by this state variable, their durations, and the allowed value transitions. Additional planned state variables, called *Instrument-1...*, *Instrument-n*, are introduced to represent the scientific payloads. For each variable *Instrument-i* we introduce four values: *Warmup*, *Process*, *Turnoff*, and *Idle* (see Fig. 1-right).

*External State Variables.* The *Orbit Events* state variable (Fig. 2, top) maintains the temporal occurrences of pericentres and apocentres represented by the values: *PERI* and *APO* (they have fixed durations). The *Ground Station Availability* state variables (Fig. 2, bottom) are a family of variables that maintain the visibility of various ground stations. The allowed values for these state variables are either *Available* or *Unavailable*.



**Figure 2:** An example of complete plan for the Remote Space Agent domain. The synchronizations among timelines are highlighted.

*Synchronizations constraints.* Any valid temporal plan needs synchronizations among the planned timelines (see Fig. 2, middle) and the external timelines (dotted arrows in Fig. 2). They represent how (a) science operations must occur during pericentres, i.e., the *Science* value must start and end during a *Peri* value; (b) maintenance operations must occur in the same time interval as apocentres, i.e., the *Maint* value is required to start and end exactly when the *Apo* value starts and ends; (c) communications must occur during ground station visibility windows, i.e., the *Comm* value must start and end during an *Available* value on any of the ground stations. As for scientific instruments, we introduce the following constraints: (d) if *Instrument-i* is not in *Idle* then the other instruments need to be in *Idle*; (e) the *Warmup* is before *Process* which is before *Turnoff*; (f) these activities are allowed only when *Science* is active along the *Operative Mode* timeline.

*Relaxed constraints.* Besides synchronization constraints, we need to consider other constraints which cannot be naturally represented in the planning model as structural constraints, but rather treated as meta-level requirements to be enforced by the planner heuristics and optimization methods. In our case study, we consider the following relaxed constraints: (g) *Maint* must be allocated between 2 and 5 orbits apart with duration of about 90 minutes (to be centered around the apocentre event); (h) science activities must be maximized, i.e., during each pericentre phase a *Science* event should occur.

## 6 EXPERIMENTAL EVALUATION

In this section, we analyze the plan verification performances with respect to temporal *flexibility* and execution *controllability*. We deploy our flexible time plan verification method to flexible plans automatically generated for our real world case study in different scenarios and execution contexts checking for *dynamic controllability* and *relaxed constraints* satisfaction. More specifically, we analyze the performances of our tool varying the following settings: *State variables*. Here, we consider three possible configurations: the RSA endowed with zero, one, or two scientific instruments. This affects the number of state variables (and synchronization constraints). *Flexibility*. For each scientific instrument activity (i.e., warm-up, process, turn-off), we set a minimal duration (i.e. about 2 minutes), but we allow temporal flexibility on the activity termination, namely, the end of each activity has a tolerance ranging from 5 to 10 seconds. E.g. if we set 5 seconds of flexibility, we introduce an uncertainty on the activity terminations, for instance, the warm-up activity can take from 120 to 125 seconds. This temporal interval represents the degree of temporal flexibility that we introduce in the system. *Hori-*

*zon*. We consider flexible plans with a horizon length ranging from 3 to 10 mission days. *Controllability*. We consider four different execution contexts: 1) all the instruments activities are controllable; 2) for each instrument the warm-up termination is not controllable; 3) for each instrument, warm-up and process terminations are not controllable; 4) for each instrument warm-up, process, and turn-off are not controllable. Note that the higher is the degree of flexibility/uncontrollability, the larger is the space of allowed behaviors to be checked, thus, the harder is flexible plan verification. In these settings, we analyze the performance of our tool considering the following issues: model generation, dynamic controllability checking, domain requirements checking. We run our experiments on a Linux workstation endowed with a 64-bit AMD Athlon CPU (3.5GHz) and 2GB RAM. In the following we illustrate the collected empirical results (the reported timings are in seconds).

**Model Generation.** A first, preliminary, analysis concerns the model generation process and the dimension of the generated UPPAAL-TIGA specification. This analysis is needed because the complexity of the generated UPPAAL-TIGA models can affect the scalability of the overall verification method. In fact, for this purpose, we developed a tool that automatically builds the UPPAAL-TIGA model given the description of the *planning domain* and the *flexible temporal plan* to be checked. Here, we want to assess the size of the generated model and the generation time with respect to the dimension of the planning domain and of the plan (state variables and plan length). In our experimental setting, we consider domain models with an incremental number of state variables (from 3 to 5) and plans with an incremental number of mission days (from 3 to 10). For each possible configuration, we consider the dimension of the generated model and the time elapsed for the generation. For all these configurations, the generation process is very fast and takes less than 200ms: the dimension of the generated model gradually grows with respect to the dimension of the flexible plan (both in terms of number of timelines and plan length) from 60 up to (about) 600 automata states with file size growing from 23kb to 147kb the dimension – in the case of 5 timelines and 10 mission days. In conclusion, the process of model generation is fast and the generated model grows linearly with the dimension of the plan, therefore, here the encoding phase is not a critical step.

**Flexible Plan Verification against Controllable Execution.** Here, we collect the time performances (CPU time) of plan verification in different scenarios (changing the degree of plan flexibility) and execution contexts (changing the plan controllability). Here, we analyze the plan verification performances in checking dynamic controllability in the easiest condition of controllability. Indeed, in this initial experimental setting, we consider fully controllable plans assuming all the scientific tasks to be controllable.

In Fig. 3(a) and Fig. 4(a), we illustrate the results gathered in the case of one and two instruments, respectively, considering the verifier performances under different plan length and flexibility conditions. For all the cases, verification succeeded.

The results in Fig. 3(a) and Fig. 4(a) show that an increment of temporal flexibility has a limited impact on the performances of the verification tool. This is particularly evident in the case of a single instrument, where the performances of the verification process seems not affected by the degree of temporal flexibility (Fig. 3(a)). On the other hand, in the case of 2 scientific instruments (Fig. 3(a)), we can observe a smooth growth of the verification time with respect to the allowed temporal flexibility. Of course, this is mainly due to the fact that in this case the verification process is to check all the synchro-

Full Controllability				1 Uncontrollable Task			
days	0s flex	5s flex	10s flex	days	0s flex	5s flex	10s flex
3	0.198	0.202	0.254	3	0.189	0.165	0.193
4	0.254	0.301	0.320	4	0.227	0.234	0.238
5	0.300	0.344	0.328	5	0.276	0.296	0.264
6	0.192	0.208	0.184	6	0.172	0.160	0.168
7	0.248	0.240	0.248	7	0.212	0.220	0.208
8	0.292	0.300	0.284	8	0.268	0.248	0.252
9	0.348	0.332	0.364	9	0.308	0.336	0.336
10	0.392	0.364	0.401	10	0.356	0.364	0.379

(a)

2 Uncontrollable Tasks				3 Uncontrollable Tasks			
days	0s flex	5s flex	10s flex	days	0s flex	5s flex	10s flex
3	0.189	0.192	0.188	3	0.198	0.221	0.212
4	0.246	0.237	0.245	4	0.267	0.283	0.267
5	0.296	0.324	0.288	5	0.304	0.288	0.288
6	0.156	0.164	0.164	6	0.188	0.172	0.176
7	0.212	0.216	0.212	7	0.212	0.208	0.220
8	0.260	0.263	0.264	8	0.252	0.236	0.248
9	0.316	0.288	0.336	9	0.312	0.300	0.332
10	0.345	0.321	0.335	10	0.367	0.353	0.379

(c)

Figure 3: One instrument varying flexibility and controllability.

nization constraints among the instruments, which are not considered in the case of a single instrument. However, even though the increment of temporal flexibility enlarges the number of behaviors to be checked, in the presence of fully controllable activities a single execution trace is sufficient to show plan controllability, hence the verification task is reduced to correct plan termination checking.

**Flexible Plan Verification against Partially Controllable Execution.** In the following, we consider the verifier performances in checking *dynamic controllability* in the presence of uncontrollable activities. Interestingly, also in this setting the execution time for verification grows in a gradual manner. In the case of a single scientific instrument, the gathered results (see Fig. 3b-c-d) are comparable with the ones collected in the fully controllable case. Even when we consider a setting where all the tasks are uncontrollable, our verification tool can easily accomplish plan verification for all the flexibility and plan length configurations (see Fig. 3(d)). In the case of 2 instru-

Full Controllability				1 Uncontrollable Task			
days	0s flex	5s flex	10s flex	days	0s flex	5s flex	10s flex
3	0.899	2.010	2.673	3	1.784	2.998	3.021
4	1.123	3.101	3.200	4	2.132	3.156	3.103
5	1.664	3.508	3.312	5	2.784	3.280	3.248
6	2.756	3.780	3.396	6	2.892	3.252	3.312
7	3.704	4.368	4.528	7	3.664	4.384	4.500
8	4.492	5.080	5.088	8	4.232	5.096	5.212
9	5.300	5.896	6.724	9	5.492	6.492	6.716
10	5.934	6.234	7.243	10	6.357	7.093	7.732

(a)

2 Uncontrollable Tasks				3 Uncontrollable Tasks			
days	0s flex	5s flex	10s flex	days	0s flex	5s flex	10s flex
3	2.022	3.105	3.227	3	2.243	3.143	3.004
4	2.214	3.326	3.339	4	2.527	3.340	3.122
5	2.444	3.452	3.548	5	2.880	3.528	3.052
6	2.652	3.212	3.328	6	2.628	3.404	3.704
7	3.612	4.412	4.464	7	3.604	4.252	4.284
8	4.200	4.879	5.208	8	4.212	4.668	4.98
9	5.300	5.876	6.812	9	5.176	6.088	6.384
10	6.604	7.012	8.002	10	6.392	7.478	8.244

(c)

Figure 4: Two instruments varying flexibility and controllability.

ments (5 timelines), the increment of flexibility gradually increments the time needed by the verification tool to verify the plans (see Fig. 4b-c-d). A similar increment can be observed when we increase the number of uncontrollable activities. If we keep constant the uncontrollable activities, the performances trend appears similar to the one of the fully controllable case. Nevertheless, even if we consider the worst case, i.e. all the activities uncontrollable and maximal temporal flexibility, the performances of the UPPAAL-TIGA verification tool are still very satisfactory: given flexible plans with horizon length up to 10 mission days and 5 timelines, plan verification can be successfully accomplished within few seconds (see Fig. 4(d)).

#### Flexible Plan Verification against Relaxed Domain Constraints.

We also perform tests to verify other domain-dependent constraints, namely, the two *relaxed constraints* on maintenance and science activities introduced in Section 5. In this experimental setting, we assume the system endowed with 2 scientific instruments (5 timelines). In Fig. 5, we report the experimental results collected increasing the degree of uncontrollability on the considered flexible plans. Chang-

ing the plan flexibility, the verifier presents performances that are analogous to the ones reported in the previous case. Thus, the additional properties to be checked provide a low additional overhead to the verification process.

experiment1.eps

Figure 5: Relaxed constraints: changing flexibility and controllability.

## 7 CONCLUSION

This work presents a verification process suitable for a timeline-based planner and shows how a temporally flexible plan verification problem can be cast as model-checking on timed game automata. A formal account has been provided to demonstrate that our method is appropriate to both represent and verify flexible plans using TGA and UPPAAL-TIGA. Then, we have introduced a realistic benchmark. The experimental results collected in this domain demonstrate the feasibility of our method and the effectiveness of UPPAAL-TIGA in a real world setting. Despite the increasing complexity of the verification configurations, the execution time gradually grows with the complexity of the task. Furthermore, the concurrent increase of temporal flexibility and plan uncontrollability does not determine the expected computational overhead. The UPPAAL-TIGA verifier can effectively handle the flexible plan verification task in all the considered configurations.

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