

# Using progress sets on non-deterministic transition systems for multiple UAV motion planning<sup>\*</sup>

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**Abstract:** This paper presents a new approach for control synthesis of non-deterministic transition systems under Linear Temporal Logic specifications with applications to multiple Unmanned Aerial Vehicles (UAV) motion planning problems. The consideration of such systems is motivated by the non-determinism possibly introduced while abstracting dynamical systems into finite transition systems. More precisely, we consider transition systems enhanced with a *progress set* describing the fact that the system cannot stay indefinitely in some subset of states. The control synthesis problem is firstly translated into a terminating planning problem. Then, a backward reachability strategy searches for a path from the initial set to the goal set. At each iteration, subsets of states contained in the progress set are added to the path, thus ensuring the reachability to the goal set in finite time. If a solution to the terminating problem is found, the obtained controller is translated back to the initial problem formulation. This approach is validated through an experiment involving two UAVs with a surveillance specification.

*Keywords:* Guidance, navigation and control of vehicles; Multi-vehicle systems; Formal control synthesis; Temporal logic specifications; Non-deterministic transition system.

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## 1. INTRODUCTION

Robot motion planning with temporal logic specifications has received a lot of attention in the control community. More recently, the expressiveness of temporal specifications coupled with automated control synthesis methods brought powerful tools. Such tools can be used for provably correct control and planning design of robotic systems under high-level specifications in the form of temporal logic formulas (Belta et al., 2007).

The Linear Temporal Logic (LTL) specification language provides an expressive framework where safety, reachability and reactive properties are suitably formulated (Baier and Katoen, 2008). More importantly, they can be translated to finite automata (Clarke et al., 1999; Babiak et al., 2012) which makes relevant the use of computer science tools (e.g., graph search and fix point algorithms). In this work, we use LTL formulas that can be expressed as deterministic automata (Alur and La Torre, 2004; Fainekos et al., 2006).

Common approaches for control synthesis under high-level specifications involve the construction of an abstraction of the system. This abstraction is supposed to be used in place of a dynamical system and aims at reducing the overall complexity of the initial synthesis problem. If the system and the abstraction verify some behavioural

relationship (see e.g., Tabuada, 2009), then the controller synthesized on the abstraction under the LTL specification can be refined into a controller such that the original system satisfies the same specification. Although some methods result in deterministic abstractions (Kloetzer and Belta, 2008b; Boskos and Dimarogonas, 2015), other abstraction approaches induce some non-determinism (i.e., a control input can induce several successors from the same initial state, see e.g., Moor and Raisch, 2002; Nilsson and Ozay, 2014). Non-deterministic transitions in the abstraction can arise due to the state space partitioning or because of initial disturbances of the continuous system. Methods to lower these effects exist, such as partitioning the state space according to the system flow (see e.g., Lafferriere et al., 2000; Tabuada, 2009) or designing local controllers that confine disturbances in each cell of the partition (see e.g., Kloetzer and Belta, 2008b). However, none of them can guaranty to get ride of the non-determinism (see e.g., Habets et al., 2006), and, in such cases, the non-deterministic nature of the system needs to be taken into account.

While control synthesis problems with deterministic models can easily be solved through a reformulation into a model checking problem (Clarke et al., 1999), the same approach cannot be applied with non-deterministic models (Kloetzer and Belta, 2008a). Fixed points methods have been widely used in correct-by-design control synthesis for non-deterministic models (see e.g., Cimatti et al., 2003; Kloetzer and Belta, 2008a). They determine the set of valid controllers by trying to maximize some winning region which corresponds to the subspace of states where there

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<sup>\*</sup> This work was supported by the H2020 ERC Starting Grant BUCOPHSYS, the EU H2020 AEROWORKS project, the EU H2020 Co4Robots project, the Swedish Foundation for Strategic Research, the Swedish Research Council and the KAW Foundation.

exists a control strategy solving the synthesis problem. As it has been highlighted in Cimatti et al. (2003), such planner prevents any cycle (path of state-transition that starts and ends at the same state) from being part of a solution as they might keep the state away from the goal set forever. Other approaches use a *fairness assumption*: for any infinite run, every transition will be taken an infinite number of times (Cimatti et al., 2003; Fu et al., 2011). A direct consequence of this assumption is that every trajectory cannot stay indefinitely in a cyclic path, and thus, these cycles can be part of the solution plan. This fairness assumption is justified in probabilistic models where the probability of any transition is not zero. For this reason, it has been used in action planning where any action can be assigned to a success probability, and an infinite number of attempts will almost surely result in a success. Such probabilistic approaches cannot always hold in the case of control synthesis of a dynamical system: any transition from one state of the abstraction to itself might correspond to an equilibrium point in the original continuous system, and any cycle in the abstraction might correspond to a stable orbit of the system. Therefore the global fairness property is not granted.

Models with local fairness property have been investigated in De Giacomo et al. (2010) where the fairness assumption is modelled as an LTL formula  $\Box\Diamond\varphi \Rightarrow \Box\Diamond\psi$  which stands for: “by trying  $\varphi$  infinitely often,  $\psi$  will happen infinitely often”. This formulation can then be integrated in a general reactivity framework. In this approach, and contrary to fixed point techniques where cycles are all eliminated, or all accepted (when fairness property stands), only non-blocking cycles identified by the model can be part of the solution plan. Other common approaches in control synthesis have been using the quotient transition system in order to deal with cycles in motion planning: this abstraction suppresses any self-transition of the model and only considers a fragment of the LTL formulas (Tůmová et al., 2010; Pappas, 2003).

In this paper we propose a new method for control synthesis of non-deterministic transition systems under LTL specifications. We consider a non-deterministic augmented Finite Transition System (FTS) (firstly introduced for switching systems by Nilsson and Ozay, 2014). This model extends the standard FTS structure with the knowledge of a *progress set* that identifies local control strategies that are terminating (trajectories cannot stay indefinitely in the subset of states using the associated control inputs). We use this model as an abstraction for a dynamical system where the size of the progress set was large, and for this reason, the approach of De Giacomo et al. (2010), that efficiently deals with small progress sets, was not appropriate. We first introduce the model and the problem to be solved (Section 2). Then the product automaton of the model and the Büchi Automaton is defined and translated to a terminating planning problem (Section 3.1). A backward reachability algorithm is used (Section 3.3) to find a path through a terminating control strategy (Section 3.2) bringing the initial state deterministically in finite time to the goal set. Correctness and termination of the solutions are investigated. This approach is then validated in an experiment (Section 4) involving multiple Unmanned Aerial Vehicles (UAV).

## 2.1 Definitions

Let  $\mathbb{N}$  denote the set of natural numbers and  $\mathbb{R}$  the set of real numbers. Given two sets  $X, Y$ , let  $|X| \in \mathbb{N}$  denote the cardinality of  $X$ ,  $2^X$  its power set and  $X^\omega$  the set of infinite words with elements chosen in  $X$ . For a word  $w$  of  $X^\omega$ ,  $Inf(w) \subseteq X$  denotes the set of elements appearing infinitely often in  $w$ . For  $Z \subseteq X \times Y$ ,  $Z|_X \subseteq X$  denotes the projection of the set  $Z$  on the set  $X$ . Let  $X \setminus Y$  be the set of elements of  $X$  not in  $Y$ . If  $\mathbb{K}$  is  $\mathbb{R}$  or  $\mathbb{N}$ , let  $a, b \in \mathbb{K}$ ,  $[a, b]_{\mathbb{K}}$  be the set  $\{x \in \mathbb{K} \mid a \leq x \leq b\} \subset \mathbb{K}$ .

We use a Finite Transition System (FTS) enhanced with a *progress set* adapted from Nilsson and Ozay (2014):

*Definition 1.* An Augmented Finite Transition System (AFTS) is defined by  $\mathcal{T} = \langle S, S_0, U_{\mathcal{T}}, \delta_{\mathcal{T}}, \mathcal{P}_{\mathcal{T}} \rangle$  where:  $S$  is the set of states;  $S_0 \subseteq S$  is the set of initial states;  $U_{\mathcal{T}}$  is the input alphabet;  $\delta_{\mathcal{T}} : S \times U_{\mathcal{T}} \rightarrow 2^S$  is the transition function;  $\mathcal{P}_{\mathcal{T}} \subseteq 2^{S \times U_{\mathcal{T}}}$  is the progress set. An *execution* of the AFTS  $\mathcal{T}$  is an infinite sequence  $r \in (S \times U_{\mathcal{T}})^\omega$  of state-control input pairs  $r = \{(s_k, u_k)\}_{k \in \mathbb{N}}$  such that:  $s_0 \in S_0$ ;  $\forall k \in \mathbb{N}, s_{k+1} \in \delta_{\mathcal{T}}(s_k, u_k)$ ; and  $\forall m \in \mathcal{P}_{\mathcal{T}}, Inf(r) \not\subseteq m$ .

The progress set  $\mathcal{P}_{\mathcal{T}}$  identifies local control strategies that are terminating, i.e., any execution reaching an element  $m \in \mathcal{P}_{\mathcal{T}}$  is guaranteed to exit  $m$  in finite time. Note that the condition  $\forall m \in \mathcal{P}_{\mathcal{T}}, Inf(r) \not\subseteq m$  does not forbid any execution  $r$  to visit infinitely often an element  $m \in \mathcal{P}_{\mathcal{T}}$  as long as the execution also leaves  $m$  infinitely often. A similar definition of the AFTS suitable for switching systems can be found in Nilsson and Ozay (2014) where the progress set is a subset of  $2^S \times U_{\mathcal{T}}$ .

For each execution  $r$  of  $\mathcal{T}$ , we call  $\rho = r|_{S^\omega} \in S^\omega$  its associated *run*. In the present work, we assume that the AFTS  $\mathcal{T}$  is *well-formed* meaning that we have:  $S_0 \neq \emptyset$ ; for every reachable state  $s \in S$  there exists at least one control action leading to another state, i.e.,  $\exists u \in U_{\mathcal{T}}, |\delta_{\mathcal{T}}(s, u)| \geq 1$ ; and all elements of  $\mathcal{P}_{\mathcal{T}}$  have an outgoing transition, i.e.,  $\forall m \in \mathcal{P}_{\mathcal{T}}, \exists (s, u) \in m, \delta_{\mathcal{T}}(s, u) \not\subseteq m|_S$ .

The reader is referred to the dedicated literature about formal methods (such as Baier and Katoen, 2008) for a formal definition of the LTL language. In summary, an LTL formula over a set  $S$  is defined inductively by  $\varphi ::= \top \mid a \mid \neg\varphi \mid \varphi_1 \wedge \varphi_2 \mid \bigcirc\varphi \mid \varphi_1 \text{U} \varphi_2$  where the logical operators  $\top$  (*true*),  $\neg$  (*negation*),  $\wedge$  (*conjunction*), the temporal operators  $\bigcirc$  (*next*) and  $\text{U}$  (*until*) are combined with elements  $a \in S$  and LTL formulas  $\varphi_1$  and  $\varphi_2$ . Useful operators  $\vee$  (*disjunction*),  $\Rightarrow$  (*implication*),  $\Box$  (*always*) and  $\Diamond$  (*eventually*) can be derived from the previous ones. These LTL formulas provide a user-friendly language to express specifications such as “avoid area  $d$ , visit area  $c$ , and avoid area  $a$  until you reach area  $b$ ” that can be expressed by the LTL formula  $\varphi = (\Box\neg d) \wedge (\Diamond c) \wedge ((\neg a) \text{U} b)$ . Every LTL formula over an input alphabet  $S$  can be translated into a Büchi Automaton (Clarke et al., 1999):

*Definition 2.* A Büchi Automaton is a tuple  $\mathcal{A}_\varphi = \langle Q_\varphi, Q_{\varphi_0}, S, \delta_\varphi, \mathcal{F}_\varphi \rangle$  where:  $Q_\varphi$  is the finite set of states;  $Q_{\varphi_0} \subseteq Q_\varphi$  is the set of initial states;  $S$  is the input alphabet;  $\delta_\varphi : Q_\varphi \times S \rightarrow 2^{Q_\varphi}$  is the transition function;  $\mathcal{F}_\varphi$  is the set of accepting states.

If  $Q_{\varphi_0}$  is a singleton and  $|\delta_{\varphi}(q, a)| \in \{0, 1\}$  for all  $a \in S$  and all  $q \in Q_{\varphi}$ , then  $\mathcal{A}_{\varphi}$  is called Deterministic Büchi Automaton (DBA). It is called a Non-deterministic Büchi Automaton (NBA) otherwise. In this paper, we use LTL specifications that can be represented with DBA. Such DBA can be generated using a NBA of a given LTL formula and applying determinization processes (Alur and La Torre, 2004; Babiak et al., 2012). However, this is not always achievable, and only a fragment of LTL formulas are translatable into DBA (see Theorem 4.50 in Baier and Katoen, 2008, pp. 190). Nonetheless, this fragment allows to express a wide range of specifications (Fainekos et al., 2006). For an infinite input word  $w = w_1w_2w_3\dots \in S^{\omega}$ , there exists a set of infinite state trajectories in  $\mathcal{A}_{\varphi}$  produced by  $w$  that we denote as  $\mathcal{R}_{\varphi}(w) \subseteq Q_{\varphi}^{\omega}$ .

*Definition 3.* The input word  $w \in S^{\omega}$  is said to be *accepted* by  $\mathcal{A}_{\varphi}$  if there exists  $r \in \mathcal{R}_{\varphi}(w)$  such that  $\text{Inf}(r) \cap \mathcal{F}_{\varphi} \neq \emptyset$ .

*Remark 4.* The input alphabet  $S$  of  $\mathcal{A}_{\varphi}$  is chosen as the set of states of  $\mathcal{T}$  so that a run of  $\mathcal{T}$  produces an input word for  $\mathcal{A}_{\varphi}$ . A run  $\mathcal{T}$  can then be accepted or not by  $\mathcal{A}_{\varphi}$ .

## 2.2 Problem statement

The considered problem is formulated as follows:

*Problem 5.* Given a non-deterministic and well-formed AFTS  $\mathcal{T} = \langle S, S_0, U_{\mathcal{T}}, \delta_{\mathcal{T}}, \mathcal{P}_{\mathcal{T}} \rangle$  and a DBA  $\mathcal{A}_{\varphi} = \langle Q_{\varphi}, Q_{\varphi_0}, S, \delta_{\varphi}, \mathcal{F}_{\varphi} \rangle$ , find a control strategy such that all runs of the controlled system are accepted by  $\mathcal{A}_{\varphi}$ .

To satisfy the acceptance condition of  $\mathcal{A}_{\varphi}$  (see Definition 3), the control strategy of the system  $\mathcal{T}$  should produce trajectories in  $\mathcal{A}_{\varphi}$  that reach the acceptance set  $\mathcal{F}_{\varphi}$  in finite time. Therefore, in the case of an FTS (AFTS with no progress set, i.e.,  $\mathcal{P}_{\mathcal{T}} = \emptyset$ ), every control strategy creating cyclic trajectories in  $\mathcal{A}_{\varphi}$  outside of the acceptance set  $\mathcal{F}_{\varphi}$  cannot be a solution of the problem. For the AFTS  $\mathcal{T}$ , if these corresponding cyclic executions are in one element of the progress set  $\mathcal{P}_{\mathcal{T}}$ , then these cycles are escaped in finite time, and the controller might be a solution of Problem 5.

## 3. SOLUTION

Problem 5 is solved in several steps. First, we create the product automaton of  $\mathcal{T}$  and  $\mathcal{A}_{\varphi}$ , and translate it into an equivalent formulation as a terminating planning problem (Section 3.1). Then, we identify control strategies on subsets of the state space which can reach, deterministically and in finite time, given sets of states (Section 3.2). Finally, we use these local terminating controllers to find a non-blocking control strategy starting the search from the goal set until the initial set is found (Section 3.3).

### 3.1 Terminating formulation

Let the product automaton of a DBA and an AFTS be:

*Definition 6.* The product automaton  $\mathcal{A}_p$  of the DBA  $\mathcal{A}_{\varphi}$  and of the AFTS  $\mathcal{T}$  is defined by  $\mathcal{A}_p = \mathcal{T} \otimes \mathcal{A}_{\varphi} = \langle Q_p, Q_{p_0}, U_p, \delta_p, \mathcal{F}_p, \mathcal{P}_p \rangle$  where:  $Q_p = S \times Q_{\varphi}$  is the set of states;  $Q_{p_0} = S_0 \times Q_{\varphi_0}$  is the set of initial states;  $U_p = U_{\mathcal{T}}$  is the input set;  $\delta_p : Q_p \times U_p \rightarrow 2^{Q_p}$  is the transition function defined by  $(s', q') \in \delta_p((s, q), u)$  iff  $s' \in \delta_{\mathcal{T}}(s, u)$ ,



Fig. 1: Construction of  $\mathcal{A}_t$  (right) from  $\mathcal{A}_p$  (left). Similar node shape (resp. line style) represents equivalent state sets (resp. transition function between sets).

$q' \in \delta_{\varphi}(q, s)$ ;  $\mathcal{F}_p = S \times \mathcal{F}_{\varphi}$  is the acceptance set;  $\mathcal{P}_p \subseteq 2^{Q_p \times U_p}$  is the progress set defined for  $m \in 2^{Q_p \times U_p}$  by  $m \in \mathcal{P}_p \Leftrightarrow m|_{S \times U_{\mathcal{T}}} \in \mathcal{P}_{\mathcal{T}}$ .

$\mathcal{A}_p$  is a Büchi Automaton augmented with the progress set of  $\mathcal{T}$ . An execution  $r \in (Q_p \times U_p)^{\omega}$  of  $\mathcal{A}_p$  is *valid* if  $\forall m \in \mathcal{P}_p, \text{Inf}(r) \not\subseteq m$  (condition inherited from Definition 1). A valid execution  $r$  of  $\mathcal{A}_p$  is *accepted* if the corresponding state trajectory  $\rho = r|_{Q_p}^{\omega}$  satisfies  $\text{Inf}(\rho) \cap \mathcal{F}_p \neq \emptyset$ . An accepted execution of  $\mathcal{A}_p$  thus corresponds to a valid execution of  $\mathcal{T}$  that is accepted by  $\mathcal{A}_{\varphi}$ .

A controller of  $\mathcal{A}_p$  is a map  $\pi : Q_p \rightarrow U_p$  that associates to each state  $p \in Q_p$  an *available control action*  $u \in U_p(p)$  where  $U_p(p) = \{u \in U_p \mid |\delta_p(p, u)| \geq 1\}$ . The controller  $\pi$  is *closed* if every state reachable from the initial set  $Q_{p_0}$  using  $\pi$  is associated with a control action; and is *terminating* if all the reachable states can reach the acceptance set  $\mathcal{F}_p$  in finite time. Finding a controller for Problem 5 consists in finding a closed and terminating controller over  $\mathcal{A}_p$ . In what follows, this control problem is translated into a terminating formulation.

*Proposition 7.* (Theorem 4 in Patrizi et al., 2013). Let a non-deterministic product automaton  $\mathcal{A}_p = \langle Q_p, Q_{p_0}, U_p, \delta_p, \mathcal{F}_p, \mathcal{P}_p \rangle$ , where  $\mathcal{F}_p = \{g_1, \dots, g_n\}$ ,  $|\mathcal{F}_p| = n$ , and  $Q_p = \{p_1, \dots, p_m, g_1, \dots, g_n\}$ ,  $|Q_p| = n + m$ . Let  $G_t^0 = \{g_1^0, \dots, g_n^0\}$  be a duplicate definition of  $\mathcal{F}_p$  (with different names for the corresponding states so that  $G_t^0 \cap \mathcal{F}_p = \emptyset$ ). The *terminating formulation* of  $\mathcal{A}_p$  is denoted as  $\mathcal{A}_t = \langle Q_t, Q_{t_0}, U_t, \delta_t, G_t, \mathcal{P}_t \rangle$  such that:  $Q_t = Q_p \cup G_t^0$  is the set of states;  $Q_{t_0} = Q_{p_0} \cup G_t^0$  is the set of initial states;  $U_t = U_p$  is an input alphabet;  $\delta_t : Q_t \times U_t \rightarrow 2^{Q_t}$  is the transition function defined by  $p' \in \delta_p(p, u) \Leftrightarrow p' \in \delta_t(\sigma(p), u)$ ;  $G_t = \mathcal{F}_p$  is the goal set;  $\mathcal{P}_t = \mathcal{P}_p$  is the progress set. where  $\sigma : Q_p \rightarrow Q_t$  is a function mapping  $Q_p \setminus \mathcal{F}_p$  to  $Q_t \setminus (G_t \cup G_t^0)$  and  $\mathcal{F}_p$  to  $G_t^0$ , namely:  $\sigma(p) = p$  if  $p \in Q_p \setminus \mathcal{F}_p$  and  $\sigma(g_j) = g_j^0$  for  $j \in [1, n]_{\mathbb{N}}$ .  $\mathcal{A}_t$  is said to be *well-formed* if  $\mathcal{A}_p$  is well-formed. A controller  $\pi_t : Q_t \rightarrow U_t$  of  $\mathcal{A}_t$  is said to be closed (resp. terminating) if the associated controller  $\pi_p = \pi_t \circ \sigma$  of  $\mathcal{A}_p$  is closed (resp. terminating).

The terminating formulation  $\mathcal{A}_t$  of  $\mathcal{A}_p$  is built by mapping all transitions from  $\mathcal{F}_p$  to transitions from a duplicate set  $G_t^0$  of  $\mathcal{F}_p$  and by adding  $G_t^0$  to the initial set  $Q_{t_0}$ . Figure 1 illustrates the construction of the transition function  $\delta_t$  and of the set of states  $Q_t$  of  $\mathcal{A}_t$ .  $\mathcal{A}_p$  is supposed to be well-formed (see Problem 5), thus  $\mathcal{A}_t$  is also well-formed (meaning that  $\forall q \in Q_t \setminus G_t, \exists u \in U_t(q)$  where  $U_t(q) = \{u \in U_t \mid |\delta_t(q, u)| \geq 1\}$ ). Let an *execution* of  $\mathcal{A}_t$  be a finite or infinite sequence  $r = \{(q_k, u_k)\}_{k < I}$  of length  $I \in \mathbb{N} \cup \{\infty\}$  with elements chosen in  $Q_t \times U_t$  such that:  $q_0 \in Q_{t_0}$ ;  $\forall k < I, q_{k+1} \in \delta_t(q_k, u_k)$ ; and  $\forall m \in \mathcal{P}_t, \text{Inf}(r) \not\subseteq m$ . For a closed controller  $\pi_t$ , a  $\pi_t$ -execution corresponds to a sequence of  $\mathcal{A}_t$  such that  $r = \{(q_k, \pi_t(q_k))\}_{k < I}$  with  $I \in \mathbb{N} \cup \{\infty\}$ . From Proposition 7, we can now equivalently solve Problem 5 by trying to find a closed and terminating

controller  $\pi_t$  for the system  $\mathcal{A}_t$  such that all the  $\pi_t$ -executions of  $\mathcal{A}_t$  reach the goal set  $G_t$  in finite time.

### 3.2 Terminating modules search

Any control strategy in  $\mathcal{A}_t$  creating cyclic execution might produce runs that loop ad infinitum, hence keeping the state away from  $G_t$  forever. If, however, each of these potential cycles is included into elements of the progress set  $\mathcal{P}_t$ , then none of them can block the system indefinitely. Thus, if  $\pi_t$  brings the state from terminating modules (defined in the sequel) to other terminating modules strictly closer to the goal set  $G_t$ , then the termination property is ensured. This section details the search of a terminating module that reaches, deterministically and in finite time, a given set of states.

We first introduce some definitions. Let a *module*  $m \subseteq Q_t \times U_t$  of  $\mathcal{A}_t$  be a set of state-control input pairs such that  $\forall (q, u) \in m, u \in U_t(q)$  and every state is associated to a unique control action ( $\forall q \in m|_{Q_t}, |m \cap (\{q\} \times U_t)| = 1$ ). We call  $\mathcal{M}$  the set of modules of  $\mathcal{A}_t$ . For  $m \in \mathcal{M}$ , let  $Post(m) = \bigcup_{(q,u) \in m} \delta_t(q, u)$  be the set of successor states of  $m$  and let  $\overline{Post}(m) = Post(m) \setminus m|_{Q_t}$  be the set of *outgoing* successor states of  $m$ .

*Definition 8.*  $m \in \mathcal{M}$  is a *terminating module* of  $\mathcal{A}_t$  if all executions of  $\mathcal{A}'_t = \langle Q_t, m|_{Q_t}, U_t, \delta_t, G_t, \mathcal{P}_t \rangle$  exit  $m$  in finite time, i.e., for every execution  $r = \{(q_k, u_k)\}_{k < I}$ ,  $I \in \mathbb{N} \cup \{\infty\}$  of  $\mathcal{A}'_t$ , there exists  $i < I$  such that  $q_i \notin m|_{Q_t}$ , and  $\forall k < i, (q_k, u_k) \in m$ .

Note that each module  $m \in \mathcal{M}$  in the progress set  $\mathcal{P}_t$  or such that  $Post(m) \cap m|_{Q_t} = \emptyset$  (i.e., without self-transitions) is terminating. For a goal set  $g \subseteq Q_t$ , let  $PRECEDINGSINGLETONMODULES(g) \subseteq \mathcal{M}$  be defined as the set of singleton modules with states chosen in  $Q_t \setminus g$  that have one or more successors in  $g$ , namely:  $\{(q, u)\} \in PRECEDINGSINGLETONMODULES(g)$  iff  $q \in Q_t \setminus g$  and  $\delta_t(q, u) \cap g \neq \emptyset$ . For a module  $m \in \mathcal{M}$ , let  $EXPANDMODULE(g, m) \subseteq \mathcal{M}$  be defined by  $m' \in EXPANDMODULE(g, m)$  iff  $m'|_{Q_t} = m|_{Q_t} \cup (Post(m) \setminus g)$  and  $m \subset m'$ .  $EXPANDMODULE(g, m)$  is the set of possible expansions of  $m$  with states chosen in  $Q_t \setminus g$ .

For  $\mathcal{A}_t$  the terminating problem formulation,  $g \subseteq Q_t$  a set of states to be reached, and  $W \subseteq 2^{Q_t}$  a set of winning regions where a terminating control strategy exists (see Section 3.3), let  $m = PRECEDINGMODULE(\mathcal{A}_t, g, W)$  be the module in  $\mathcal{M}$  of smallest cardinality that verifies:

- $\overline{Post}(m) \subseteq g$ : all outgoing successor states are in  $g$ ,
- $m \in \mathcal{P}_t$  or  $Post(m) \subseteq g$ :  $m$  is a terminating module,
- $g \cup m|_Q \notin W$ : the set of winning regions.

If no module verifies the previous properties, then  $m = \emptyset$ .

### 3.3 Backward Reachability Algorithm

In this section,  $PRECEDINGMODULE$  is used to iteratively expand a winning region where a terminating and closed control strategy exists. Algorithm 1 implements this as a Depth-First Search (DFS) using a backward reachability strategy: preceding terminating modules are searched and added to a terminating controller starting from the goal set  $G_t$  until all initial states in  $Q_{t_0}$  are found.

**Function** BACKWARDSEARCH( $\mathcal{A}_t$ )

**Data:**  $\mathcal{A}_t$ : the terminating problem formulation

**Result:**  $K$  if a solution is found, *Fail* otherwise

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1  $L \leftarrow \{K_G\}$ ; // plans to visit
2  $L_v \leftarrow \emptyset$ ; // visited plans
3 repeat
4    $K \leftarrow \arg \max_{x \in L} (|x|)$ ; // biggest
   cardinality plan
5   if  $Q_{t_0} \subseteq \tilde{K}|_{Q_t}$  then // initial set found
6     return  $K$ ; // return valid plan
7    $L_v \leftarrow L_v \cup \{K\}$ ; // Add  $K$  to  $L_v$ 
8    $W \leftarrow \{\tilde{l}|_{Q_t} \mid l \in L \cup L_v\}$ ; // set of visited
   sets of states
9    $m \leftarrow PRECEDINGMODULE(\mathcal{A}_t, \tilde{K}|_{Q_t}, W)$ ;
10  if  $m \neq \emptyset$  then
11     $L \leftarrow L \cup \{K \cup \{m\}\}$ ; // add the new plan
   to  $L$ 
12  else
13     $L \leftarrow L \setminus \{K\}$ ; // remove the plan from  $L$ 
14  end
15 until  $|L| = 0$ ;
16 return Fail; // no valid plan found

```

**Algorithm 1.** Backward reachability algorithm (BRA)

Let a *plan*  $K \subseteq \mathcal{M} \cup \{m_0\}$  be a set of terminating modules of  $\mathcal{A}_t$  (see Definition 8) with an additional fake module  $m_0 = \{(g, \emptyset) \mid g \in G_t\}$  comprised of each goal state associated to none of the control actions. Let  $\tilde{K} = \bigcup_{k \in K} k$  be the corresponding set of state-control action pairs.  $L$  corresponds to the set of plans to visit and is initialized (line 1) with  $K_G = \{m_0\}$ . At each step, the highest cardinality plan  $K$  is selected from  $L$  (line 4). If the initial set belongs to  $\tilde{K}|_{Q_t}$ , then plan  $K$  is returned (line 6). Otherwise, a terminating module  $m$  preceding  $\tilde{K}|_{Q_t}$  (line 9) is returned by  $PRECEDINGMODULE$ , and the new plan  $K \cup \{m\}$  is added to  $L$  (line 11). Note that  $m$  is chosen such that  $m|_{Q_t} \cup \tilde{K}|_{Q_t} \notin W$  (see Section 3.2) where  $W \subset 2^{Q_t}$  is the set of winning regions. If no valid module is found (lines 9 and 10), then plan  $K$  is removed from  $L$  (line 13). When  $L$  is empty (line 15), that means that among all the possible configurations of reachable states comprised of terminating modules, none of them could find a solution and a *Fail* flag is returned (line 16).

*Theorem 9.* Let  $K$  be a valid plan returned by Algorithm 1 ( $K \neq Fail$ ) and  $\pi_K$  be the controller associated with  $K$  defined by  $\pi_K(q) = u$  for all  $(q, u) \in \tilde{K} = \bigcup_{k \in K} k$ . The controller  $\pi_K$  is closed and terminating<sup>1</sup>.

As the controller  $\pi_K$  is closed and terminating, the controller (after mapping with  $\sigma$  - previously defined by  $\forall p \in Q_p \setminus \mathcal{F}_p, \sigma(q) = q$  and  $\forall j \in [1, n]_{\mathbb{N}}, \sigma(g_j) = g_j^0$ ) is a valid solution of Problem 5.

*Corollary 10.* Let  $K$  be a valid plan returned by Algorithm 1 (i.e.,  $K \neq Fail$ ) and  $\pi_K$  be the controller associated to  $K$  defined by  $\pi_K(q) = u$  for all  $(q, u) \in \tilde{K}$ . The controller  $\pi_p = \pi_K \circ \sigma$  is a solution of Problem 5.

<sup>1</sup> The proof can be found in the extended version of this paper: <https://hal.archives-ouvertes.fr/hal-01502558>

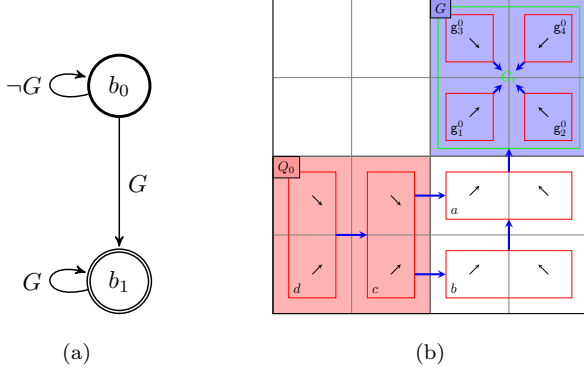


Fig. 2: (a) is a DBA that represents  $\varphi = (-G)\mathbf{U}(\square G)$  (with  $\mathcal{F}_\varphi = \{b_1\}$  and  $Q_{\varphi_0} = \{b_0\}$ ). (b) is an illustration of Algorithm 1. The agent goes from  $Q_0$  (in red) to reach and stay in  $G$  (in blue) following the control actions (black arrows). Four control actions are available ( $\nwarrow, \nearrow, \swarrow$  and  $\searrow$ ) and each of them points in the direction of the 3 successors of the considered state (square cell). Each red box corresponds to a terminating module and the green box corresponds to the goal set  $G_t$  of  $\mathcal{A}_t$ .

### 3.4 Illustration

Figure 2b illustrates how Algorithm 1 finds a solution plan. An agent starting in  $Q_0$  needs to verify  $\varphi = (-G)\mathbf{U}(\square G)$ , namely “reach and stay in  $G$ ”. A DBA of  $\varphi$  is represented in Figure 2a. The progress set  $\mathcal{P}_\mathcal{T}$  is partially described by  $\{a, b, c, d\} \subset \mathcal{P}_\mathcal{T}$ .  $\mathcal{T}$  is informally defined in Figure 2b. Due to the special form of the LTL formula  $\varphi$ ,  $\mathcal{A}_t$  can be defined equivalently than in Section 3.1 by:  $G_t = G$ ;  $G_t^0 = \{g_1^0, \dots, g_4^0\}$  duplicate of  $G = \{g_1, \dots, g_4\}$ ;  $Q_t = S \cup G_t^0$ ;  $Q_{t_0} = Q_0 \cup G_t^0$ ;  $\mathcal{P}_t = \mathcal{P}_\mathcal{T}$ ; and  $s' \in \delta_\mathcal{T}(s, u) \Leftrightarrow s' \in \delta_t(\sigma(s), u)$  (where  $\sigma : S \rightarrow Q_t$  is defined by:  $\sigma(s) = s$  iff  $s \in S \setminus G_t$  and  $\sigma(g_i) = g_i^0$  for  $i \in [1, 4]_{\mathbb{N}}$ ). Note that in Figure 2b,  $G_t$  and  $G_t^0$  are superimposed however, we remind that  $G_t \cap G_t^0 = \emptyset$ .

Algorithm 1 is initialized with a plan  $K_0 = \{m_0\}$  where  $m_0$  is the fake goal module introduced in Section 3.3 (i.e.,  $m_0|_{Q_t} = G_t$ , green box in Figure 2b). At the first iteration, the plan  $K_0$  is selected from the set  $L = \{K_0\}$  of plans to visit. PRECEDINGMODULE searches for a terminating module preceding the set of states  $G_t = \tilde{K}_0|_{Q_t}$ : e.g.,  $m_1 = g_1^0 = \{(g_1^0, \nearrow)\}$  is returned as all successors  $Post(m_1)$  of  $m_1$  are in  $\tilde{K}_0|_{Q_t}$ . The new plan  $K_1 = K_0 \cup \{m_1\}$  is added to  $L$ . At the next iteration of Algorithm 1 (and at each iteration in this example), the highest cardinality plan of  $L$  is the last one added to  $L$ . For the same reasons than for module  $m_1$  with plan  $K_0$ , modules  $m_2 = g_2^0$ ,  $m_3 = g_3^0$  and  $m_4 = g_4^0$  are successively added to plans  $K_2 = K_1 \cup \{m_2\}$ ,  $K_3 = K_2 \cup \{m_3\}$  and  $K_4 = K_3 \cup \{m_4\}$ . At the 5<sup>th</sup> iteration of Algorithm 1,  $K_4$  is the biggest plan of  $L = \{K_0, \dots, K_4\}$  and PRECEDINGMODULE searches for a terminating module going to  $\tilde{K}_4|_{Q_t} = G_t \cup G_t^0$  deterministically and in finite time; there is no singleton terminating module, the smallest valid ones are composed of at least 2 elements, e.g.,  $a$  is returned as it belongs to  $\mathcal{P}_t$  and goes to  $G_t \subset \tilde{K}_4|_{Q_t}$ .  $m_5 = a$  is added to  $K_5 = K_4 \cup \{m_5\}$ .

In the next iteration, the terminating module  $b$  is returned as it goes to  $a|_{Q_t} \subset \tilde{K}_6|_{Q_t}$  and as  $b \in \mathcal{P}_t$ , then  $m_6 = b$  is

added to the plan  $K_6 = K_5 \cup \{m_6\}$ .  $m_7 = c$  is added to  $K_7 = K_6 \cup \{m_7\}$  as it goes deterministically to  $a|_{Q_t} \cup b|_{Q_t} \subset \tilde{K}_6|_{Q_t}$  and as  $c \in \mathcal{P}_\mathcal{T}$ . Note that the transition between the modules is not deterministic ( $Post(c) \subseteq (a|_{Q_t} \cup b|_{Q_t})$ ). Finally,  $m_8 = d$  is added to  $K_8 = K_7 \cup \{m_8\}$ , and as  $d|_{Q_t} \cup c|_{Q_t} = Q_{t_0} \subset \tilde{K}_8|_{Q_t}$ ,  $K_8$  is a plan that brings all the trajectories starting in  $Q_{t_0}$  to the goal set in finite time. Algorithm 1 returns the solution plan  $K = K_8$ . Every trajectory of the system  $\mathcal{T}$  controlled by  $\pi_p = \pi_K \circ \sigma$  (Theorem 9 and Corollary 10) verifies the specification  $\varphi$ .

## 4. EXPERIMENT

Multiple quadricopters (Iris Plus, 3DRobotics) are tracked with a motion capture system (Qualisys) and controlled from an offboard computer. Algorithms are implemented in python, and the LTL translation to DBA is done using `ltl3ba` (Babiak et al., 2012).

Let the discrete time dynamical system  $\mathcal{S}_d$  (sampling time of 1s) of 2 quadricopters be defined for  $i \in \mathbb{N}$  by:  $x_{i+1} = x_i + u + w$ , where  $x = [x_1, x_2, x_3, x_4]^\top \in \mathbb{R}^4$  is the state, with  $[x_1, x_2]^\top$  (resp.  $[x_3, x_4]^\top$ ) the position of agent 1 (resp. agent 2),  $u \in \mathcal{U} = \{-0.2, 0.2\}^4$  is the control input, and  $w \in \mathcal{W} = [-0.1, 0.1]_{\mathbb{R}}^4$  is a disturbance. For  $x, y \in \mathbb{R}^4$ , we define the closed interval  $[x, y] = \{z \in \mathbb{R}^{2n} \mid x \leq z \wedge z \leq y\} \subset \mathbb{R}^4$  (where  $\leq$  is the component-wise inequality). Let the following intervals of  $\mathbb{R}^4$  be defined by:  $X = [[-1, -1, -1, -1]^\top, [1, 1, 1, 1]^\top]$ ,  $X_a = [[-1, -1, 0.2, 0.2]^\top, [-0.2, -0.2, 1, 1]^\top]$ , and  $X_b = [[0.2, 0.2, -1, -1]^\top, [1, 1, -0.2, -0.2]^\top]$ .

To create the FTS structure of  $\mathcal{T}$ , the state space  $\mathbb{R}^4$  of  $\mathcal{S}_d$  is partitioned (uniformly on  $X$  with a step  $x_d = 0.4$  for every dimension of  $\mathbb{R}^4$ ,  $\mathbb{R}^4 \setminus X$  is mapped to a single symbol), let  $\Xi : X \rightarrow S$  be the resulting partitioning function. As in Meyer (2015), a non-deterministic abstraction is created after the computation of reachable set over-approximations of  $\mathcal{S}_d$ . The set  $\mathcal{P}_\mathcal{T}$  is defined by  $m \in \mathcal{P}_\mathcal{T}$  iff the set of continuous states covering the symbols of  $m|_{Q_t}$  ( $X_m = \{x \in X \mid \Xi(x) \in m|_S\} \subseteq X$ ) is bounded and  $\min_{dx \in \mathcal{C}_m} \|dx\| > 0$  where  $\mathcal{C}_m$  is the convex hull of  $m|_{U_\mathcal{T}} + \mathcal{W}$  (with + the Minkowski sum). For  $m \in \mathcal{P}_\mathcal{T}$ , as  $X_m$  is bounded and as the time average of the state displacement has a strictly positive norm, every trajectory of  $\mathcal{S}_d$  starting in  $X_m$  will escape the set  $X_m$  in finite time (details are omitted due to space limitation). This guarantees the existence of some behavioural relationship (see Nilsson and Ozay, 2014) between the abstraction and the system  $\mathcal{S}_d$ .

Let the atomic propositions be:  $out \Leftrightarrow x \notin X$ ,  $a \Leftrightarrow x \in X_a$ ,  $b \Leftrightarrow x \in X_b$ ,  $collide \Leftrightarrow \max_{i \in \{1, 2\}} |x_i - x_{i+2}| < 0.4$ . Label  $a$  (resp.  $b$ ) corresponds to agent 1 in region blue (resp. in region red) and agent 2 in region red (resp. in region blue; see Figure 3), The system must verify the following specification:  $\varphi = (\square \neg out) \wedge (\square \neg collide) \wedge (\square \diamond a) \wedge (\square \diamond b)$ .

The controller  $\pi_p$  solution of Problem 5 is found with Algorithm 1 and Corollary 10. Figure 3 shows a few steps of the experiment and a video is available at <sup>2</sup>. All states of  $\mathcal{T}$  have non-deterministic transitions with 16 possible successors per control input. Modules of the solution plan

<sup>2</sup> [https://youtu.be/yj0\\_o1U1tYI](https://youtu.be/yj0_o1U1tYI)

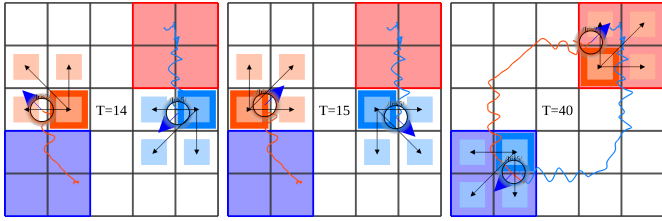


Fig. 3: Agent 1 (in blue) and agent 2 (in red) need to exchange position infinitely often. At timestep  $T$ , the current state of the agents (cell in plain color) is associated to a control input (thick blue arrows) and have multiple successors (thin black arrows and cells in light color).

$K$  (see Algorithm 1) have a cardinality of 8 when one of the agents has already reached its region and must wait for the other one to reach its own region, and a cardinality of 4 otherwise. Agents start in  $b$ . The discrete state  $p$  of  $\mathcal{A}_p$  is initialized by measuring the starting position  $x_0$ :  $p = (\Xi(x_0), p_0) \in Q_{p_0}$ . At each step, the state  $x$  of  $\mathcal{S}_a$  is measured and mapped to a symbol  $s = \Xi(x)$  of  $S$ , then the current discrete state  $p \in Q_p$  of  $\mathcal{A}_p$  is updated with the new state  $p' = (t, b)$ , unique element of  $\delta_p(p, \pi_p(p))$  such that  $t = s$ . Steps  $T \in \{14, 15\}$  highlight the cyclic transitions of the AFTS in one of the terminating module of the plan  $K$ . Step  $T = 40$  shows the end of the experiment where the agents reached  $a$ .

## 5. CONCLUSION

We proposed a solution to the control synthesis problem of a non-deterministic transition system under Linear Temporal Logic specifications that can be represented by Deterministic Büchi Automata. The considered system is modelled as a finite transition system enhanced with a progress set. Elements of the progress set identify local control strategies that are terminating (the state will escape a given set of states in finite time). Experiments with multiple UAVs show that this approach is relevant for real world applications where the non-determinism of the system cannot be narrowed after the abstraction of it.

Only a fragment of LTL formulas are translatable into DBA, and if there is no deterministic translation possible, then the product automaton of the FTS and the NBA is not fully observable (2 successors of the same state and input control might produce the same observation), and our approach is not valid anymore. How to handle such situations will be the topic of future works.

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