

# LTL<sub>f</sub> Synthesis on First-Order Agent Programs in Nondeterministic Environments

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## Abstract

We investigate the synthesis of policies for high-level agent programs expressed in Golog, a language based on situation calculus that incorporates nondeterministic programming constructs. Unlike traditional approaches for program realization that assume full agent control or rely on incremental search, we address scenarios where environmental nondeterminism significantly influences program outcomes. Our synthesis problem involves deriving a policy that successfully realizes a given Golog program while ensuring the satisfaction of a temporal specification, expressed in Linear Temporal Logic on finite traces (LTL<sub>f</sub>), across all possible environmental behaviors. By leveraging an expressive class of first-order action theories, we construct a finite game arena that encapsulates program executions and tracks the satisfaction of the temporal goal. A game-theoretic approach is employed to derive such a policy. Experimental results demonstrate this approach’s feasibility in domains with unbounded objects and non-local effects. This work bridges agent programming and temporal logic synthesis, providing a framework for robust agent behavior in nondeterministic environments.

## 1 Introduction

Agents operating in dynamic environments often need to react to changes beyond their control. For example, a service robot may be tasked with serving coffee to customers, who may place an order at any time. Also, some actions may have unexpected outcomes, e.g., while attempting to fulfill the task, the robot might accidentally drop the coffee. Such scenarios can be modeled as *fully observable nondeterministic* (FOND) planning tasks (Geffner and Bonet 2013; Ghallab, Nau, and Traverso 2016) where actions have multiple possible outcomes, or as reactive synthesis problem (e.g., based on *LTL on finite traces* (LTL<sub>f</sub>) (De Giacomo and Vardi 2015)), where certain propositions are controlled by the agent while others are governed by the environment. However, these approaches have some limitations. They assume a fixed, finite set of propositions, effectively imposing a closed-world assumption and requiring a completely known initial state. Furthermore, they rely on alternating actions between the agent and the environment, which fails to capture scenarios where the environment may

perform an arbitrary number of actions before the agent can respond. Additionally, existing solutions often generate arbitrary plans or policies without incorporating user-specified partial strategies unless explicitly encoded in the specification.

On the other hand, GOLOG (Levesque et al. 1997), a well-established agent programming language, offers significant flexibility. Based on the situation calculus (McCarthy and Hayes 1969; Reiter 2001a), GOLOG supports first-order reasoning over arbitrarily large or even infinite domains and accommodates incomplete information about the initial state. It also allows for the specification of partial strategies through nondeterministic programs. Given such a program, *program realization* is the task of resolving program nondeterminism to produce a successful program execution, e.g., by means of search, or in an online incremental fashion (De Giacomo et al. 2009). However, it is typically assumed that the agent is in complete control, even if it only has incomplete knowledge (Reiter 2001b; Claßen and Neuss 2016) or its actions are stochastic (Boutilier et al. 2000). Recently, the situation calculus has been extended with nondeterministic actions (De Giacomo and Lespérance 2021; Claßen and Delgrande 2021) similar to FOND planning, where the environment chooses an outcome. However, this still assumes that agent and environment act in turns.

To address these limitations, we propose an extension to GOLOG that partitions actions into agent actions and environment actions. In this framework, the agent selects among currently applicable agent actions, guided by the program and the basic action theory, but cannot constrain the environment. The environment may select any applicable environment action or any action chosen by the agent. This allows for arbitrary sequences of environment actions, similar to the *supervisory control* paradigm (Ramadge and Wonham 1989). We also propose to describe the agent’s goal as a temporal formula, which allows for formulating trajectory constraints such as safety and reachability on the program.

In this setting, program realization becomes a synthesis task. Given a GOLOG program and a temporal goal, the task is to synthesize a policy that executes the program while satisfying the temporal goal, independent of and reacting to all possible environment behaviors. In this paper, we focus on the decidable fragment of GOLOG with acyclic basic action theories restricted to C<sup>2</sup> (Zarriß and Claßen 2016) and tem-

poral goals given as  $LTL_f$  formulas (De Giacomo and Vardi 2013). We provide a decidable approach for this problem by constructing a finite game arena that captures all possible program executions while tracking the satisfaction of the temporal specification, and then applying a game-theoretic approach to synthesize a policy. Exploiting an encoding of  $LTL_f$  formulas that interprets temporal formulas as propositional atoms (Li et al. 2020), the construction works on-the-fly and avoids building irrelevant parts.

The remainder of this paper is structured as follows. After discussing related work in Section 2, we summarize GOLOG and introduce  $LTL_f$  in the context of GOLOG programs in Section 3. We describe the synthesis approach in Section 4 and evaluate it experimentally in Section 5, before concluding in Section 6.

## 2 Related Work

Verification of GOLOG programs has been explored in various contexts. Initially, verification efforts relied on manual proofs (De Giacomo, Ternovska, and Reiter 1997; Liu 2002; Shapiro, Lespérance, and Levesque 2002). Claßen and Lakemeyer (2008) describe a (possibly not terminating) system that is capable of automatically verifying properties of non-terminating GOLOG programs. Later research identified decidable fragments of GOLOG grounded in  $C^2$ , the decidable two-variable fragment of first-order logic with counting (Grädel, Otto, and Rosen 1997). Verification of GOLOG programs with *context-free* or *local-effect* basic action theories (BATs) in  $C^2$  and with pick operators restricted to finite domains is decidable for properties in CTL (Claßen et al. 2014), LTL (Zarriëß and Claßen 2014a), and  $CTL^*$  (Zarriëß and Claßen 2014b). Beyond local-effect BATs, verification remains decidable if the BAT is *acyclic*, i.e., there is no cyclic dependency between fluents in the effect descriptors, or *flat*, i.e., effect descriptors are quantifier-free (Zarriëß and Claßen 2016). *Bounded theories*, where the number of objects described by any situation is bounded, also results in decidable verification (De Giacomo, Lespérance, and Patrizi 2016). All these approaches rely on a finite abstraction of the infinite program configuration space, which yields decidability, and hence could be used as basis for our approach.

Related to verification is *synthesis* of temporal properties, which can be described as two-player games between the system and the environment (Abadi, Lamport, and Wolper 1989; Pnueli and Rosner 1989). Given a specification, e.g., in Linear Temporal Logic (LTL), and a partition of the symbols into controllable and uncontrollable ones, the players alternate selecting a subset of their symbols. LTL has also been used to describe temporally extended goals for planning (Bacchus and Kabanza 1998; De Giacomo and Vardi 2000; Geffner and Bonet 2013), possibly resulting in infinite plans (Patrizi et al. 2011). LTL can also be used to specify *conformant planning* problems with temporally extended goals (Calvanese, De Giacomo, and Vardi 2002) and synthesis is related to FOND planning (Camacho et al. 2017, 2018; De Giacomo and Rubin 2018) as a nondeterministic effect can be seen as an environment action. Moreover, there has been a particular interest in  $LTL_f$  (De Giacomo and Vardi 2013), where the synthesis problem can be solved

by transforming the  $LTL_f$  specification into a finite automaton (De Giacomo and Vardi 2015). Like LTL,  $LTL_f$  synthesis is 2EXPTIME-complete, although  $LTL_f$  synthesis tools usually perform better. Recently, several methods have been proposed to improve the performance of  $LTL_f$  synthesis, e.g., based on BDDs (Zhu et al. 2017) and on-the-fly forward search (Xiao et al. 2021; De Giacomo et al. 2022; Favorito 2023).

## 3 Preliminaries

We describe the logic  $\mathcal{ES}$  and an  $\mathcal{ES}$ -based variant of GOLOG and then introduce  $LTL_f$  in the context of GOLOG programs.

### The Logic $\mathcal{ES}$

The logic  $\mathcal{ES}$  (Lakemeyer and Levesque 2010) is a first-order modal variant of the situation calculus. Following (Zarriëß and Claßen 2016), we consider  $\mathcal{ES}$  formulas restricted to  $C^2$ .

**Syntax** *Terms* are of sort *object* or *action*. We use  $x, y, \dots$  (possibly with decorations) to denote object variables, and  $a$  for a variable of sort action.  $N_O$  is a countably infinite set of *object constant symbols*, and  $N_A$  a countably infinite set of *action function symbols* whose arguments are all of sort object. Let  $\mathcal{N}_O$  denote the set of all ground terms (called *standard names*) of sort object, and  $\mathcal{N}_A$  those of sort action. Formulas are constructed over equality atoms and *fluent* predicates with at most two arguments of sort object, using the usual Boolean connectives, quantifiers, counting quantifiers, as well as modalities  $\Box\phi$  (“ $\phi$  holds after any sequence of actions”), and  $[t]\phi$  (“ $\phi$  holds after executing action  $t$ ”). We call a formula *fluent* if it does not mention  $\Box$  or  $[ \cdot ]$ . A *sentence* is a formula without free variables. A  *$C^2$ -fluent formula* is a fluent formula without actions and with at most two variables.

**Semantics** A *trace* is a finite sequence of action standard names. When a trace represents a history of already executed actions, it is called a *situation*. For a trace  $z = \langle \alpha_1, \dots, \alpha_n \rangle \in \mathcal{Z}$ , we write  $|z|$  for the length  $n$  of  $z$ ,  $z \cdot \alpha$  for the concatenation  $\langle \alpha_1, \dots, \alpha_n, \alpha \rangle$  of  $z$  with an action  $\alpha$ ,  $z[i]$  for the  $i$ th action  $\alpha_i$ ,  $z[\cdot i]$  for the prefix  $\langle \alpha_1, \dots, \alpha_i \rangle$ , and  $z[i \cdot ]$  for the suffix  $\langle \alpha_i, \dots, \alpha_n \rangle$ . Let  $\mathcal{Z} = \mathcal{N}_A^*$  be the set of all traces, and  $\mathcal{P}_F$  the set of all *primitive formulas*  $F(n_1, \dots, n_k)$ , where  $F$  is a  $k$ -ary fluent with  $0 \leq k \leq 2$  and the  $n_i$  are object standard names. A *world*  $w$  maps primitive formulas and situations to truth values, i.e.,  $w : \mathcal{P}_F \times \mathcal{Z} \rightarrow \{0, 1\}$ . The set of all worlds is denoted by  $\mathcal{W}$ .

**Definition 1** (Truth of Formulas). *Let  $w \in \mathcal{W}$  be a world and  $\alpha$  an action standard name. We define for every  $z \in \mathcal{Z}$ :*

1.  $w, z \models F(n_1, \dots, n_k)$  iff  $w[F(n_1, \dots, n_k), z] = 1$ ;
2.  $w, z \models (n_1 = n_2)$  iff  $n_1$  and  $n_2$  are identical;
3.  $w, z \models \phi_1 \wedge \phi_2$  iff  $w, z \models \phi_1$  and  $w, z \models \phi_2$ ;
4.  $w, z \models \neg\phi$  iff  $w, z \not\models \phi$ ;
5.  $w, z \models \forall x.\phi$  iff  $w, z \models \phi_n^x$  for every  $n \in \mathcal{N}_x$ ;
6.  $w, z \models \exists^{\leq m} x.\phi$  iff  $|\{n \in \mathcal{N}_x \mid w, z \models \phi_n^x\}| \leq m$ ;
7.  $w, z \models \exists^{\geq m} x.\phi$  iff  $|\{n \in \mathcal{N}_x \mid w, z \models \phi_n^x\}| \geq m$ ;
8.  $w, z \models \Box\phi$  iff  $w, z \cdot z' \models \phi$  for every  $z \in \mathcal{Z}$ ;
9.  $w, z \models [\alpha]\phi$  iff  $w, z \cdot \alpha \models \phi$ .

Here,  $\mathcal{N}_x$  refers to the set of all standard names of the same sort as  $x$ , and  $\phi_n^x$  the result of simultaneously replacing all free occurrences of  $x$  in  $\phi$  by  $n$ . We understand  $\forall, \exists, \supset, \equiv, \top$  and  $\perp$  as the usual abbreviations. For a set of sentences  $\Sigma$  and a sentence  $\alpha$ , we write  $\Sigma \models \alpha$  (read:  $\Sigma$  entails  $\alpha$ ) to mean that for every  $w$ , if  $w, \langle \rangle \models \alpha'$  for every  $\alpha' \in \Sigma$ , then  $w, \langle \rangle \models \alpha$ . Finally, we write  $\models \alpha$  (read:  $\alpha$  is valid) to mean  $\{\} \models \alpha$ . Note that rule 2 above includes a unique names assumption for actions and objects into the semantics.

## Basic Action Theories

To encode a dynamic domain, we employ a *basic action theory* (BAT) (Reiter 2001a) with additional restrictions (Zarri  and Cla en 2016) for ensuring decidability:

**Definition 2** (Basic Action Theory). *A basic action theory (BAT)  $\mathcal{D} = \mathcal{D}_0 \cup \mathcal{D}_{post}$  is a set of axioms, where  $\mathcal{D}_0$  is a finite set of  $C^2$ -fluent sentences describing the initial state of the world, and  $\mathcal{D}_{post}$  is a finite set of successor state axioms (SSAs), one for each fluent, of the form<sup>1</sup>  $\Box[a]F(\vec{x}) \equiv \gamma_F^+ \vee F(\vec{x}) \wedge \neg\gamma_F^-$ , where the positive effect condition  $\gamma_F^+$  and the negative effect condition  $\gamma_F^-$  are disjunctions of formulas of the form  $\exists\vec{y}. (a = A(\vec{v}) \wedge \varepsilon \wedge \kappa)$  such that*

- the free variables of the formula  $\exists\vec{y}. (a = A(\vec{v}) \wedge \varepsilon \wedge \kappa)$  are among  $\vec{x}$  and  $a$ ,
- $A(\vec{v})$  is an action term and  $\vec{v}$  contains  $\vec{y}$ ,
- the effect descriptor  $\varepsilon$  is a fluent formula with no terms of sort action and the number of variables in  $\varepsilon$  that do not occur in  $\vec{v}$  or occur bound in  $\varepsilon$  is less than or equal to two,
- the context condition  $\kappa$  is a fluent formula with free variables among  $\vec{v}$ , no terms of sort action, and at most two bound variables.

Intuitively, the effect descriptor is the part of the effect condition that expresses *which objects* are affected, while the context condition encodes *whether* the effect takes place.

**Acyclic BATs** For a BAT  $\mathcal{D}$ , we can construct the *fluent dependency graph*  $\Delta_{\mathcal{D}}$ , which captures the dependencies between fluents in the effect descriptors. In  $\Delta_{\mathcal{D}}$ , each node is a fluent of  $\mathcal{D}$  and there is a directed edge  $(F, F')$  from fluent  $F$  to fluent  $F'$  if there exists a disjunct  $\exists\vec{y}. (a = A(\vec{v}) \wedge \varepsilon \wedge \kappa)$  in  $\gamma_F^+$  or  $\gamma_F^-$  such that  $F'$  occurs in  $\varepsilon$ . A BAT is *acyclic* if  $\Delta_{\mathcal{D}}$  is acyclic. Furthermore, the *fluent depth* of an acyclic BAT, denoted by  $\text{fd}(\mathcal{D})$ , is the length of the longest path in  $\Delta_{\mathcal{D}}$  and the *fluent depth of  $F$  w.r.t.  $\mathcal{D}$* , denoted by  $\text{fd}_{\mathcal{D}}(F)$ , is the length of the longest path in  $\Delta_{\mathcal{D}}$  starting in  $F$ .

## GOLOG Programs

We consider a set of program expressions that includes ground actions ( $\alpha$ ), tests for  $C^2$ -fluent sentences ( $\phi?$ ), sequence of subprograms  $(\delta_1; \delta_2)$ , nondeterministic choice  $(\delta_1 | \delta_2)$ , interleaved concurrent execution  $(\delta_1 \parallel \delta_2)$ , and non-deterministic iteration  $(\delta^*)$ . We write  $\text{nil} \doteq \top?$  for the empty program that always succeeds.

<sup>1</sup>The operator  $\Box$  has lowest precedence while  $[\cdot]$  has highest precedence and free variables are implicitly assumed to be universally quantified from the outside.

A GOLOG program  $\mathcal{G} = (\mathcal{D}, \delta)$  consists of a  $C^2$ -BAT  $\mathcal{D} = \mathcal{D}_0 \cup \mathcal{D}_{post}$  and a program expression  $\delta$ , where all fluents occurring in  $\mathcal{D}$  and  $\delta$  have a SSA in  $\mathcal{D}_{post}$ . For a program  $\mathcal{G} = (\mathcal{D}, \delta)$ , we write  $\mathcal{A}_{\mathcal{G}}$  for all action terms occurring in  $\delta$  and we may omit the subscript if  $\mathcal{G}$  is clear from context.

The semantics of GOLOG programs is based on transitions between configurations, where a configuration  $\langle z, \rho \rangle$  consists of a sequence of already performed actions  $z \in \mathcal{Z}$  and the remaining program  $\rho \in \text{sub}(\delta)$ . Given a world  $w \in \mathcal{W}$ , the transition relation  $\xrightarrow{w}$  among configurations is defined inductively. The set of final configurations  $\text{Fin}(w)$  defines the configurations where the program may terminate.

**Definition 3** (Program Transition Semantics). *For any world  $w$ , the set of final configurations  $\text{Fin}(w)$  is the smallest set such that*

$$\begin{aligned} \langle z, \phi? \rangle &\in \text{Fin}(w) \text{ if } w, z \models \phi \\ \langle z, \delta_1; \delta_2 \rangle &\in \text{Fin}(w) \text{ if } \langle z, \delta_1 \rangle \in \text{Fin}(w) \text{ and } \langle z, \delta_2 \rangle \in \text{Fin}(w) \\ \langle z, \delta_1 | \delta_2 \rangle &\in \text{Fin}(w) \text{ if } \langle z, \delta_1 \rangle \in \text{Fin}(w) \text{ or } \langle z, \delta_2 \rangle \in \text{Fin}(w) \\ \langle z, \delta_1 \parallel \delta_2 \rangle &\in \text{Fin}(w) \text{ if } \langle z, \delta_1 \rangle \in \text{Fin}(w) \text{ and } \langle z, \delta_2 \rangle \in \text{Fin}(w) \\ \langle z, \delta^* \rangle &\in \text{Fin}(w) \end{aligned}$$

For any world  $w$ , the transition relation  $\xrightarrow{w}$  among configurations is the least set satisfying

$$\begin{aligned} \langle z, \alpha \rangle &\xrightarrow{w} \langle z \cdot \alpha, \text{nil} \rangle \text{ if } \alpha \text{ is a ground action} \\ \langle z, \delta_1; \delta_2 \rangle &\xrightarrow{w} \langle z', \rho; \delta_2 \rangle \text{ if } \langle z, \delta_1 \rangle \xrightarrow{w} \langle z', \rho \rangle \\ \langle z, \delta_1 | \delta_2 \rangle &\xrightarrow{w} \langle z', \rho \rangle \text{ if } \langle z, \delta_1 \rangle \in \text{Fin}(w) \text{ and } \langle z, \delta_2 \rangle \xrightarrow{w} \langle z', \rho \rangle \\ \langle z, \delta_1 \parallel \delta_2 \rangle &\xrightarrow{w} \langle z', \rho \rangle \text{ if } \langle z, \delta_1 \rangle \xrightarrow{w} \langle z', \rho \rangle \text{ or } \langle z, \delta_2 \rangle \xrightarrow{w} \langle z', \rho \rangle \\ \langle z, \delta_1 \parallel \delta_2 \rangle &\xrightarrow{w} \langle z', \rho \parallel \delta_2 \rangle \text{ if } \langle z, \delta_1 \rangle \xrightarrow{w} \langle z', \rho \rangle \\ \langle z, \delta_1 \parallel \delta_2 \rangle &\xrightarrow{w} \langle z', \delta_1 \parallel \rho \rangle \text{ if } \langle z, \delta_2 \rangle \xrightarrow{w} \langle z', \rho \rangle \\ \langle z, \delta^* \rangle &\xrightarrow{w} \langle z', \rho; \delta^* \rangle \text{ if } \langle z, \delta \rangle \xrightarrow{w} \langle z', \rho \rangle \end{aligned}$$

We write  $\|\delta\|_w^z$  for the set of traces starting in configuration  $\langle z, \delta \rangle$  and ending in a final configuration.

**Situation-Determined Programs** Following (De Giacomo, Lesp rance, and Mui e 2012), we say that a program  $\mathcal{G} = (\mathcal{D}, \delta)$  is *situation-determined*, iff for all  $w \in \mathcal{W}$  with  $w \models \mathcal{D}$ , all  $z, z' \in \mathcal{Z}$ , and all program expressions  $\delta', \delta''$ :  $\langle z, \delta \rangle \xrightarrow{w^*} \langle z', \delta' \rangle$  and  $\langle z, \delta \rangle \xrightarrow{w^*} \langle z', \delta'' \rangle$  implies  $\delta' = \delta''$ . We assume that all programs are situation-determined.

## LTL<sub>f</sub>

For temporal properties, we define temporal formulas with the same syntax as LTL<sub>f</sub> formulas, but replacing propositions with  $C^2$ -fluent sentences  $\phi$ , i.e.,  $\Phi ::= \phi \mid \Phi \wedge \Phi \mid \mathcal{X}\Phi \mid \Phi \mathcal{U}\Phi$ . For a temporal formula  $\Phi$ , we denote the set of subformulas of  $\Phi$  with  $\text{cl}(\Phi)$ . For a set of formulas  $\Psi$ , we write  $\bigwedge \Psi$  for  $\bigwedge_{\Phi \in \Psi} \Phi$ . As usual, we define  $\mathcal{F}\Phi \doteq \top \mathcal{U}\Phi$  and  $\mathcal{G}\Phi \doteq \neg \mathcal{F} \neg \Phi$ , as well as  $\Phi_1 \vee \Phi_2 \doteq \neg(\neg\Phi_1 \wedge \neg\Phi_2)$ ,  $\mathcal{N}\Phi \doteq \neg \mathcal{X} \neg \Phi$ , and  $\Phi_1 \mathcal{R} \Phi_2 \doteq \neg(\neg\Phi_1 \mathcal{U} \neg\Phi_2)$ . We define the truth of a temporal formula  $\Phi$ , given a world  $w$  and traces  $z, z'$ :

- $w, z, z' \models \phi$  iff  $w, z \models \phi$ ,
- $w, z, z' \models \Phi_1 \wedge \Phi_2$  iff  $w, z, z' \models \Phi_1$  and  $w, z, z' \models \Phi_2$ ,
- $w, z, z' \models \mathcal{X}\Phi$  iff  $z' = \alpha \cdot z'' \neq \langle \rangle$  and  $w, z \cdot \alpha, z'' \models \Phi$ ,

- $w, z, z' \models \Phi_1 \mathcal{U} \Phi_2$  iff there exists  $k \leq |z'|$  such that  $w, z \cdot z'[..k], z'[k+1..] \models \Phi_2$  and for all  $0 \leq i < k$ ,  $w, z \cdot z'[..i], z'[i+1..] \models \Phi_1$ .

**TNF and XNF** As we intend to track the satisfiability of the temporal formula  $\Phi$  over the traces of the program, we adapt Tail Normal Form (TNF) and neXt Normal Form (XNF) from (Li et al. 2020). TNF explicitly marks the end of satisfying traces, while XNF allows us to split the temporal formula into a local part, which can be evaluated at the current state, and a future part, which is evaluated against the remaining trace. First, we say a formula is in *Negated Normal Form* (NNF) if all negations are in front of only atoms. Each  $LTL_f$  formula can be transformed into NNF by using the dual operators to push negation inwards. Based on NNF, we define TNF, which marks the last state of satisfying traces:

**Definition 4.** Let  $\Phi$  be an  $LTL_f$  formula in NNF. Its TNF  $\text{tnf}(\Phi)$  is defined as  $\text{t}(\Phi) \wedge \mathcal{F} \text{Tail}$ , where *Tail* is a new atom to identify the last state of satisfying traces and  $\text{t}(\Phi)$  is an  $LTL_f$  formula defined recursively as follows:

1.  $\text{t}(\Phi) = \Phi$  if  $\Phi$  is  $\top, \perp$ , or a  $C^2$ -fluent sentence;
2.  $\text{t}(\mathcal{X}(\Psi)) = \neg \text{Tail} \wedge \mathcal{X}(\text{t}(\Psi))$ ;
3.  $\text{t}(\mathcal{N}(\Psi)) = \text{Tail} \vee \mathcal{X}(\text{t}(\Psi))$ ;
4.  $\text{t}(\Phi_1 \wedge \Phi_2) = \text{t}(\Phi_1) \wedge \text{t}(\Phi_2)$ ;
5.  $\text{t}(\Phi_1 \vee \Phi_2) = \text{t}(\Phi_1) \vee \text{t}(\Phi_2)$ ;
6.  $\text{t}(\Phi_1 \mathcal{U} \Phi_2) = (\neg \text{Tail} \wedge \text{t}(\Phi_1)) \mathcal{U} \text{t}(\Phi_2)$ ;
7.  $\text{t}(\Phi_1 \mathcal{R} \Phi_2) = (\text{Tail} \vee \text{t}(\Phi_1)) \mathcal{R} \text{t}(\Phi_2)$ .

When interpreting a TNF formula over a trace, *Tail* needs to be treated separately, as it is not a fluent sentence. We define:  $w, z, z' \models \text{Tail}$  iff  $z' = \langle \rangle$ . It can be shown that  $\Phi$  and  $\text{tnf}(\Phi)$  are equivalent:<sup>2</sup>

**Theorem 1.** Let  $\Phi$  be a temporal formula,  $w$  a world, and  $z$  and  $z'$  traces. Then  $w, z, z' \models \Phi$  iff  $w, z, z' \models \text{tnf}(\Phi)$ .

In the following, each  $LTL_f$  formula is assumed to be in TNF and we may omit the common part  $\mathcal{F} \text{Tail}$ .

We continue by interpreting temporal formulas as propositional formulas by treating sub-formulas with a temporal operator as outermost connective as if they were propositional atoms. For a temporal formula  $\Phi$ , we define the set of *propositional atoms*  $\text{PA}(\Phi)$  of  $\Phi$  inductively: (1)  $\text{PA}(\Phi) = \{\Phi\}$  if  $\Phi$  is an atom,  $\mathcal{X}$ ,  $\mathcal{U}$ , or  $\mathcal{R}$  formula; (2)  $\text{PA}(\Phi) = \text{PA}(\Psi)$  if  $\Phi = \neg\Psi$ ; and (3)  $\text{PA}(\Phi) = \text{PA}(\Phi_1) \cup \text{PA}(\Phi_2)$  if  $\Phi = \Phi_1 \wedge \Phi_2$  or  $\Phi = \Phi_1 \vee \Phi_2$ . For a temporal formula  $\Phi$ , let  $\Phi^p$  be  $\Phi$  understood as a propositional formula over  $\text{PA}(\Phi)$ . A propositional assignment  $P$  of  $\Phi^p$  is a partial function  $P : \text{PA}(\Phi) \rightarrow \{0, 1\}$  that assigns truth values to the propositional atoms  $\text{PA}(\Phi)$ . We write  $P \models \Phi^p$  if  $P$  satisfies  $\Phi^p$ . A propositional assignment  $P$  can also be understood as a set of literals  $\{p \in \text{PA}(\Phi) \mid P(p) = 1\} \cup \{\neg p \in \text{PA}(\Phi) \mid P(p) = 0\}$  and we use  $P$  to denote both interchangeably.

If  $\Phi$  is satisfiable, then there exists a corresponding propositional assignment:

**Lemma 2.** Let  $w$  be a world,  $\Phi$  an  $LTL_f$  formula, and  $z$  and  $z'$  traces. Then  $w, z, z' \models \Phi$  implies there exists a propositional assignment  $P$  with  $P \models \Phi^p$  and  $w, z, z' \models \bigwedge P$ .

<sup>2</sup>Proofs can be found in (Hofmann and Claßen 2024).

The converse is not necessarily true: Let  $\Phi = \mathcal{X}(a) \wedge \mathcal{X}(\neg a)$ . Clearly,  $\Phi$  is not satisfiable, but  $\{\mathcal{X}(a), \mathcal{X}(\neg a)\}$  is a satisfying propositional assignment of  $\Phi^p$ .

We now define XNF, where each  $\mathcal{U}$  and  $\mathcal{R}$  operator is pushed inwards such that the only outermost temporal connective is  $\mathcal{X}$ :

**Definition 5.** Let  $\Phi$  be a temporal formula. Its *neXt Normal Form* (XNF)  $\text{xnf}(\Phi)$  is defined recursively as follows:

1.  $\text{xnf}(\Phi) = \Phi$  if  $\Phi$  is  $\top, \perp$ , a  $C^2$ -fluent sentence, or  $\mathcal{X}\Psi$ ;
2.  $\text{xnf}(\Phi_1 \wedge \Phi_2) = \text{xnf}(\Phi_1) \wedge \text{xnf}(\Phi_2)$ ;
3.  $\text{xnf}(\Phi_1 \vee \Phi_2) = \text{xnf}(\Phi_1) \vee \text{xnf}(\Phi_2)$ ;
4.  $\text{xnf}(\Phi_1 \mathcal{U} \Phi_2) = \text{xnf}(\Phi_2) \vee (\text{xnf}(\Phi_1) \wedge \mathcal{X}(\Phi_1 \mathcal{U} \Phi_2))$ ;
5.  $\text{xnf}(\Phi_1 \mathcal{R} \Phi_2) = \text{xnf}(\Phi_2) \wedge (\text{xnf}(\Phi_1) \vee \mathcal{X}(\Phi_1 \mathcal{R} \Phi_2))$ .

It can be shown that  $\Phi$  and  $\text{xnf}(\Phi)$  are equivalent:

**Theorem 3.** Let  $\Phi$  be a temporal formula,  $w$  a world, and  $z$  and  $z'$  finite traces. Then  $w, z, z' \models \Phi$  iff  $w, z, z' \models \text{xnf}(\Phi)$ .

For a propositional assignment  $P$  of  $\Phi^p$  in XNF, we define  $L(P) = \{l \mid l \in P \text{ is a literal other than } (\neg)\text{Tail}\}$ ,  $X(P) = \{\theta \mid \mathcal{X}\theta \in P\}$ , and  $T(P) = \top$  if *Tail*  $\in P$  and  $T(P) = \perp$  otherwise.

XNF allows us to track the partial satisfaction of a temporal formula over a trace. After each action, we will determine each satisfying assignment  $P$  such that  $L(P)$  is satisfied by the current state and we will track  $X(P)$  in the remaining trace. We will use this in the following to construct a game arena that tracks the satisfaction of a temporal formula  $\Phi$ .

## 4 Approach

Our goal is to determine an execution of a given GOLOG program that satisfies the given temporal formula, for all possible environment behaviors. The controller must determine which actions to execute; more specifically, which branch to follow in all nondeterministic choices of the program, while not restricting the environment in its actions. Formally, our goal is to find a successful policy, defined as follows:

**Definition 6 (Policy).** Let  $\mathcal{G} = (\mathcal{D}, \delta)$  be a GOLOG program and  $\mathcal{A} = \mathcal{A}_C \dot{\cup} \mathcal{A}_E$  a partition of the actions  $\mathcal{A}$  of  $\mathcal{G}$  into *controllable* and *environment actions*. A policy is a partial mapping  $\pi : \mathcal{W} \times \mathcal{Z} \times \text{sub}(\delta) \rightarrow 2^{\mathcal{A}}$  such that:

1. if  $w \models \mathcal{D}$ , then  $\pi$  is defined on  $(w, \langle \rangle, \delta)$ ;
2. if  $\alpha \in \pi(w, z, \rho)$ , then  $\langle z, \rho \rangle \xrightarrow{w} \langle z \cdot \alpha, \rho' \rangle$  for some  $\rho' \in \text{sub}(\delta)$ ;
3. if  $\alpha \in \pi(w, z, \rho)$  and  $\langle z, \rho \rangle \xrightarrow{w} \langle z \cdot \alpha, \rho' \rangle$ , then  $\pi$  is defined on  $(w, z \cdot \alpha, \rho')$ ;
4. if  $\alpha \in \mathcal{A}_E$  and  $\langle z, \rho \rangle \xrightarrow{w} \langle z \cdot \alpha, \rho' \rangle$  for some  $\rho' \in \text{sub}(\delta)$ , then  $\alpha \in \pi(w, z, \rho)$ ;
5. if  $\pi(w, z, \rho) = \emptyset$ , then  $\langle z, \rho \rangle \in \text{Fin}(w)$ .

Intuitively, a policy chooses a subset  $\pi(w, z, \rho)$  from all possible actions in the current configuration  $\langle z, \rho \rangle$  and world  $w$ . From this subset, the environment then chooses one action to be executed. The agent's choices are restricted: Every possible environment action must be selected, hence the agent can never limit the environment's choices.

A policy  $\pi$  induces a set of traces  $\|\pi\|_w$  in world  $w$ , where  $z = \langle \alpha_1, \dots, \alpha_n \rangle \in \|\pi\|_w$  if there are  $\rho_1, \dots, \rho_n$

such that (1)  $\langle \langle \rangle, \delta \rangle \xrightarrow{w} \langle z[..1], \rho_1 \rangle \xrightarrow{w} \dots \xrightarrow{w} \langle z, \rho_n \rangle$ ; (2)  $\alpha_{i+1} \in \pi(w, z[..i], \rho_i)$ ; and (3)  $\pi(w, z, \rho_n) \subseteq \mathcal{A}_E$  and  $\langle z, \rho_n \rangle \in \text{Fin}(w)$ . Hence, the environment may choose to terminate the execution if  $\langle z, \rho \rangle$  is a final configuration and the agent chose no further actions to execute. Note that by definition, a policy is a restriction of the program execution, i.e.,  $\|\pi\|_w \subseteq \|\delta\|_w$ . We call a policy *terminating* if for every infinite sequence of  $\pi$ -compatible configurations  $\langle \langle \rangle, \delta \rangle, \langle z_1, \rho_1 \rangle, \langle z_2, \rho_2 \rangle, \dots$  and for every  $i$ , there is a  $j \geq i$  such that  $\pi(w, z_j, \rho_j) \subseteq \mathcal{A}_E$  and  $\langle z_j, \rho_j \rangle \in \text{Fin}(w)$ . Intuitively, a terminating policy ensures that at any point of the execution trace, there is some future final configuration where the policy does not choose any agent actions and hence the environment may terminate. A policy may still result in an infinite trace if the environment continues to select actions indefinitely. However, we exclude those from consideration as we assume that the environment eventually stops. We can now formalize our goal:

**Definition 7** (Synthesis Problem). *Given a GOLOG program  $\mathcal{G} = (\mathcal{D}, \delta)$  and a temporal formula  $\Phi$ , find a terminating policy  $\pi$  for  $\mathcal{G}$  that satisfies  $\Phi$ , i.e., for every world  $w$  with  $w \models \mathcal{D}$  and every  $z \in \|\pi\|_w$ , it holds that  $w, \langle \rangle, z \models \Phi$ .*

We note that it is in general undecidable to determine whether a satisfying policy exists. In (Zarri  and Cla en 2014a, 2016) it was shown that the related verification problem (a special case of the synthesis problem) becomes decidable if (1)  $\mathcal{C}^2$  is used as base logic, (2) successor state axioms are acyclic, and (3) ‘‘pick operators’’ are disallowed, i.e., all actions in the program are ground. Furthermore, dropping any of these three restrictions while maintaining the other two immediately leads to undecidability: for (1) this is due to the undecidability of FOL, and for (2) and (3) due to the possibility of reducing the halting problem for Turing machines to the verification problem.

In the following, applying the same three restrictions, we describe a sound and complete method for determining a terminating policy  $\pi$  that satisfies  $\Phi$ . We will do so by constructing a finite game arena  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  that captures the possible program executions while tracking the satisfaction of  $\Phi$ . Once we have constructed  $\mathbb{A}_{\mathcal{G}}^{\Phi}$ , we can use a game-theoretic approach to determine a terminating policy that satisfies  $\Phi$ . However, as both the number of worlds satisfying  $\mathcal{D}$  and the number of reachable program configurations is generally infinite, we first need to construct a finite abstraction based on *characteristic graphs* and *types*.

## Characteristic Graphs

*Characteristic graphs* (Cla en and Lakemeyer 2008) provide a finite encoding of the reachable program configurations. In such a graph, the nodes correspond to programs  $\rho$ , intuitively representing what remains to be executed, while an edge  $\rho \xrightarrow{\alpha:\psi} \rho'$  encodes that a transition is possible from  $\rho$  to  $\rho'$  through action  $\alpha$ , if formula  $\psi$  holds. In addition, each program  $\rho$  has an associated *termination condition*  $\varphi(\rho)$ , in the form of a fluent formula.

**Definition 8** (Characteristic Graph). *Given a program expression  $\delta$ , the termination condition  $\varphi(\delta)$  of  $\delta$  is a fluent*

*formula inductively defined as follows:*

$$\begin{aligned} \varphi(\alpha) &= \perp \text{ if } \alpha \text{ is a ground action} & \varphi(\phi?) &= \phi \\ \varphi(\delta_1; \delta_2) &= \varphi(\delta_1) \wedge \varphi(\delta_2) & \varphi(\delta_1 \parallel \delta_2) &= \varphi(\delta_1) \vee \varphi(\delta_2) \\ \varphi(\delta_1 \parallel \delta_2) &= \varphi(\delta_1) \wedge \varphi(\delta_2) & \varphi(\delta^*) &= \top \end{aligned}$$

*For any program expression  $\delta$ , the set of outgoing edges  $\delta \xrightarrow{\alpha:\psi} \rho$  with action  $\alpha$  and guard condition  $\psi$  to resulting program  $\rho$  is defined inductively as follows:*

- $\alpha \xrightarrow{\top} \text{nil}$ , if  $\alpha$  is a primitive action;
- $(\delta_1; \delta_2) \xrightarrow{\alpha:\psi} (\rho; \delta_2)$ , if  $\delta_1 \xrightarrow{\alpha:\psi} \rho$ ;
- $(\delta_1; \delta_2) \xrightarrow{\alpha:\varphi(\delta_1) \wedge \psi} \rho$ , if  $\delta_2 \xrightarrow{\alpha:\psi} \rho$ ;
- $(\delta_1 \parallel \delta_2) \xrightarrow{\alpha:\psi} \rho$ , if  $\delta_1 \xrightarrow{\alpha:\psi} \rho$  or  $\delta_2 \xrightarrow{\alpha:\psi} \rho$ ;
- $(\delta_1 \parallel \delta_2) \xrightarrow{\alpha:\psi} (\rho \parallel \delta_2)$ , if  $\delta_1 \xrightarrow{\alpha:\psi} \rho$ ;
- $(\delta_1 \parallel \delta_2) \xrightarrow{\alpha:\psi} (\delta_1 \parallel \rho)$ , if  $\delta_2 \xrightarrow{\alpha:\psi} \rho$ ;
- $\delta^* \xrightarrow{\alpha:\psi} (\rho; \delta^*)$ , if  $\delta \xrightarrow{\alpha:\psi} \rho$ .

*For any program expression  $\delta$ , the corresponding characteristic graph is given by  $\mathcal{C}_{\delta} = \langle v_0, V, E \rangle$ , where  $v_0 = \delta$  (initial node), and the nodes  $V$  and edges  $E$  are the smallest sets such that (1)  $\delta \in V$ ; and (2) if  $\delta' \in V$  and  $\delta' \xrightarrow{\alpha:\psi} \delta''$ , then  $\delta'' \in V$  and  $\delta' \xrightarrow{\alpha:\psi} \delta'' \in E$ .*

We denote the set  $V$  with  $\text{sub}(\delta)$ , the *subprograms reachable from  $\delta$* . We note (Cla en and Lakemeyer 2008):

**Lemma 4.** *For any program  $\delta$ ,  $\mathcal{C}_{\delta}$  is finite, and for any world  $w$ , situation  $z$ , and  $\delta' \in \text{sub}(\delta)$ , it holds that (1)  $\langle z, \delta' \rangle \in \text{Fin}(w)$  iff  $w, z \models \varphi(\delta')$ ; and (2)  $\langle z, \delta' \rangle \xrightarrow{w} \langle z \cdot \alpha, \delta'' \rangle$  iff  $\delta' \xrightarrow{\alpha:\psi} \delta''$  and  $w, z \models \psi$ .*

Characteristic graphs therefore exactly capture the program transition semantics. We can hence use them as finite abstractions of the reachable program configurations. Also, using characteristic graphs, there is a (simple to test) sufficient condition for programs being situation-determined:

**Lemma 5.** *If every ground action  $\alpha$  occurs at most once among the outgoing edges of every node in  $\mathcal{C}_{\delta}$ , then  $\delta$  is situation-determined.*

## Types

With characteristic graphs, we already have a finite representation of the possible program configurations. However, there are additional sources of infiniteness. For one, during the execution of a program, we may accumulate infinitely many effects. Second, there are infinitely many possible worlds that satisfy the BAT  $\mathcal{D}$ . However, for acyclic BATs, it has been shown that the set of possible effects is finite, and that the set of worlds that satisfy  $\mathcal{D}$  can be represented by a finite set of equivalence classes, so-called *types of worlds* (Zarri  and Cla en 2016). We will now describe how to construct types for a given BAT  $\mathcal{D}$ .

As our programs may only mention finitely many ground actions, we can rewrite the SSAs of an acyclic BAT by grounding the effects. This is done by replacing each SSA for a fluent  $F(\vec{x})$  by a set of instantiated formulas, one for each  $\alpha \in \mathcal{A}$ , of the form  $\Box[\alpha]F(\vec{x}) \equiv (\gamma_F^{\alpha}) \vee F(\vec{x}) \wedge$



A state  $s = (\tau, E, A, \rho)$  is final if  $(\varphi(\rho), E) \in \tau$  and accepting if  $(\emptyset, \top) \in A$ . We denote the set of all final states with  $\mathcal{S}_F$  and the set of all accepting states with  $\mathcal{S}_A$ . We also write  $\text{type}(s) = \tau$  for the type of the world in  $s$ .

Each state consists of (1) a type  $\tau$ , representing an equivalence class of worlds; (2) a node  $\rho$  from the characteristic graph, capturing the remaining program and its termination condition; (3) a set of accumulated effects  $E$ ; and (4) a set of temporal formulas  $A$ , which must be satisfied in the remaining execution to fulfill the specification  $\Phi$ . The initial states are those states with the initial program expression and no accumulated effects. Furthermore, regarding the temporal formula  $\Phi$  and  $A$  of an initial state, we first compute all the propositional assignments of  $\text{xnf}(\Phi)^p$ . For each assignment  $P$ , we check whether the local part  $L(P)$  is satisfied by the state. If so, the pair  $(\chi, \theta) = (X(P), T(P))$  is added to  $A$ , which intuitively states that  $\chi$  must be satisfied in the future and the program should terminate if  $\theta$  is true. For transitions, we first check whether there is an edge in the characteristic graph that allows the execution of the next action. If so, we accumulate the effects and check whether there is a propositional assignment of  $\text{xnf}(\bigwedge \chi_1^p)$  for some  $(\chi_1, \theta_1) \in A_1$  that allows the satisfaction of the temporal formulas in  $A_2$ . Similar to the initial states, we do so by checking whether the local part  $L(P)$  is satisfied by the current state and tracking  $X(P)$  and  $T(P)$  in the future.

By definition, a state is final if the program may terminate and it is accepting if  $\Phi$  is satisfied. Also note that  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  is finite as both types and reachable sub-programs are finite. It is also deterministic, as  $\mathcal{G}$  is situation-determined and for action successors, the satisfying assignments of  $\text{xnf}(\Phi)^p$  are collected in a single successor state.

We can show that  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  indeed corresponds to the executions of  $\mathcal{G}$  while tracking the satisfaction of  $\Phi$ :

**Theorem 7.** *Every execution of  $\mathcal{G} = (\mathcal{D}, \delta)$  satisfies  $\Phi$  iff every reachable final state of  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  is accepting.*

This provides us a decidable method for verifying an  $\text{LTL}_f$  property  $\Phi$  against a GOLOG program  $\mathcal{G}$ . However, the goal is to determine a policy that executes  $\mathcal{G}$  while satisfying  $\Phi$ .

## Synthesis

Above, we have described a finite game arena  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  that captures the executions of a program  $\mathcal{G}$  while tracking the satisfaction of a given  $\text{LTL}_f$  formula  $\Phi$ . In the following, we use a game-theoretic approach on  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  to determine a policy that successfully executes  $\mathcal{G}$  while satisfying  $\Phi$ . We do so by defining a game between two players, the system and the environment, that play on  $\mathbb{A}_{\mathcal{G}}^{\Phi}$ . We start by defining a strategy, which intuitively translates the conditions on a policy to the game arena  $\mathbb{A}_{\mathcal{G}}^{\Phi}$ :

**Definition 14 (Strategy).** *Let  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  be the game arena for some GOLOG program  $\mathcal{G}$  and temporal formula  $\Phi$ . Let  $s \in \mathcal{S}$  be a state of  $\mathbb{A}_{\mathcal{G}}^{\Phi}$ . A set of actions  $U \subseteq \mathcal{A}$  is valid in  $s$  under the following conditions:*

1. if  $\alpha \in U$ , then there is an edge  $s \xrightarrow{\alpha} s'$  for some  $s' \in \mathcal{S}$
2. if  $s \xrightarrow{\alpha} s'$  for some  $\alpha \in \mathcal{A}_E$  and  $s' \in \mathcal{S}$ , then  $\alpha \in U$
3. if  $U = \emptyset$ , then  $s$  is a final state

---

## Algorithm 1: Computing a strategy from $\mathbb{A}_{\mathcal{G}}^{\Phi}$

---

```

1: for all  $H \in 2^{\mathcal{S}_F \cap \mathcal{S}_A}$  do
2:    $G \leftarrow H$ ;  $R \leftarrow \{s \in G \mid \text{Succ}_E(s) = \emptyset\}$ ;  $\sigma \leftarrow \emptyset$ 
3:    $Q \leftarrow \{s \in \mathcal{S} \mid \text{Succ}(s) \cap G \neq \emptyset\}$ 
4:   while  $Q \neq \emptyset$  do
5:      $s \leftarrow \text{POP}(Q)$ 
6:     if  $s \in \mathcal{S}_F \setminus \mathcal{S}_A \wedge \text{Succ}_C(s) = \emptyset$  then continue
7:     if  $s \in R$  then continue
8:     if  $\text{Succ}_E(s) \neq \emptyset \wedge \forall s' \in \text{Succ}_E(s) : s' \in G \vee$ 
        $\text{Succ}_E(s) = \emptyset \wedge \exists s' \in \text{Succ}_C(s) : s' \in G$  then
9:        $G \leftarrow G \cup \{s\}$ ;  $R \leftarrow R \cup \{s\}$ 
10:      if  $s \in \mathcal{S}_F \cap \mathcal{S}_A$  then
11:         $\sigma(s) \leftarrow \{\alpha \mid \exists s' \in \text{Succ}_E(s). s \xrightarrow{\alpha} s'\}$ 
12:      else  $\sigma(s) \leftarrow \{\alpha \mid \exists s' \in G. s \xrightarrow{\alpha} s'\}$ 
13:       $Q \leftarrow Q \cup \{s' \mid s \in \text{Succ}(s')\}$ 
14:   if  $H \cup \mathcal{S}_0 \subseteq R$  then return  $\sigma$ 

```

---

A strategy in  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  is a partial function  $\sigma : \mathcal{S} \rightarrow 2^{\mathcal{A}}$  such that:

1.  $\sigma$  is defined on every initial state of  $\mathbb{A}_{\mathcal{G}}^{\Phi}$
2. if  $\sigma$  is defined on  $s \in \mathcal{S}$ , then  $\sigma(s)$  is valid in  $s$
3. if  $\sigma$  is defined on  $s \in \mathcal{S}$ ,  $\alpha \in \sigma(s)$ , and  $s \xrightarrow{\alpha} s'$  for some  $s' \in \mathcal{S}$ , then  $\sigma$  is defined on  $s'$

We also write  $s \xrightarrow{\sigma} s'$  if there is  $\alpha \in \sigma(s)$  such that  $s \xrightarrow{\alpha} s'$ . A strategy  $\sigma$  induces a set of plays  $\text{plays}(\sigma)$ , which are those paths in  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  consistent with  $\sigma$ . Formally,  $p = \langle s_0, \dots, s_n \rangle \in \text{plays}(\sigma)$  if

1.  $s_0$  is an initial state of  $\mathbb{A}_{\mathcal{G}}^{\Phi}$
2. for each  $i$ ,  $s_i \xrightarrow{\sigma} s_{i+1}$
3.  $\sigma(s_n) \subseteq \mathcal{A}_E$  and  $s_n$  is a final state of  $\mathbb{A}_{\mathcal{G}}^{\Phi}$

A play is winning if it ends in an accepting state. A strategy  $\sigma$  is winning if every play  $p \in \text{plays}(\sigma)$  is winning. We call a strategy  $\sigma$  terminating if for every infinite sequence of states  $s_0, s_1, \dots$  with  $s_k \xrightarrow{\sigma} s_{k+1}$  for every  $k$ , it holds that for every  $i$ , there is a  $j \geq i$  such that  $\sigma(s_j) \subseteq \mathcal{A}_E$  and  $s_j$  is final.

**Proposition 8.** *There is a terminating and winning strategy  $\sigma$  in  $\mathbb{A}_{\mathcal{G}}^{\Phi}$  if and only if there exists a terminating policy  $\pi$  for  $\mathcal{G}$  that satisfies  $\Phi$ .*

Hence, we need to determine a terminating and winning strategy in  $\mathbb{A}_{\mathcal{G}}^{\Phi}$ . In principle, this can be done with backward search starting in a set of good states and then checking whether the agent can force every play to end in a good state. However, not every final and accepting state is necessarily good, as the environment may force a play from this state that ends in a non-accepting state. On the other hand, every winning play must end in an accepting state, so if a strategy exists, there must be an enforceable set of final and accepting states. Hence, we can guess which final and accepting states are enforceable and then check if there is indeed a strategy that can force every play to end in those states.

This approach is formalized in Algorithm 1. It starts with a hypothesis  $H \subseteq \mathcal{S}_F \cap \mathcal{S}_A$  of good states  $G$  and tracks the states  $R$  that can reach  $G$ . It then iteratively checks the predecessors of all states in  $G$  whether the agent can force the play to end in  $G$ . This is the case if all environment successors  $\text{Succ}_E(s)$  are in  $G$  or if there is a control successor

$\text{Succ}_C(s)$  in  $G$ . If a state is found that can be forced to end in  $G$ , it is added to  $G$  and  $R$  and  $\sigma$  are updated accordingly. Finally, if all states of  $H$  and all initial states  $\mathcal{S}_0$  can in fact reach  $G$ , then  $\sigma$  is a winning and terminating strategy:

**Theorem 9.** *Algorithm 1 terminates and returns a winning and terminating strategy if one exists.*

## 5 Experiments

We implemented the method (Claßen and Hofmann 2025) in the Prolog-based Golog interpreter `vergo` (Claßen 2018), that, different from other implementations, uses full FOL as base logic, where an embedded theorem prover (Schulz, Cruanes, and Vukmirovic 2019) is used for reasoning tasks such as deciding entailment and consistency. The system contains optimizations for handling FO expressions, in particular an FO variant of binary decision diagrams. We evaluated the method on two domains, a dishwasher robot and a warehouse robot, that we will be described in detail below. All experiments were conducted on an Intel® Core™ i5-7300U @2.60GHz with 8GB of RAM, running Debian 12 with WSL2 under Windows 10, using SWI-Prolog 9.0.4 and version 3.2 of the E theorem prover.

### Incremental Construction

In our implementation, the construction of the abstract game arena follows closely Definition 13. However, the construction is done in an incremental fashion, where only the relevant and reachable parts are actually materialized. This is achieved by keeping the types as general as possible, and only including additional formulas once they are needed. More specifically, the method works by iterating the following steps, until no more changes occur:

**Initialize:** Create initial states  $(\tau, \emptyset, A, \delta)$ , where types  $\tau$  are constructed only from formulas in  $\mathcal{D}_0$  and literals  $L(P)$  of propositional assignments over  $A$ .

**Split:** If there is a state  $(\tau, E, A, \rho)$  that does not entail a truth value for some required condition  $\psi$  (the transition condition for an action  $\alpha$ , the termination condition  $\varphi(\rho)$ , the condition  $\kappa$  of an effect, or a literal  $l \in L(P)$  of a propositional assignment over  $A$ ), then create two copies of all states and transitions, where one includes  $\psi$  and the other includes  $\neg\psi$  into  $\tau$ , discarding states with inconsistent  $\tau$ .

**Expand:** If a state  $s = (\tau, E, A, \rho)$  admits an action  $\alpha$ , create the successor state  $s'$  and the transition  $s \xrightarrow{\alpha} s'$ .

We represent  $\tau$  directly by the regressed versions of formulas to avoid having to regress them repeatedly. The construction also stops in states where  $A = \emptyset$ , since the corresponding traces can never satisfy the input property.

### Dishwasher Robot

The first domain is inspired by the dishwasher robot example used in (Claßen et al. 2014), but adds additional fluents. A robot can move between a number of rooms and the kitchen, load (an arbitrary number of) dirty dishes onto itself, and unload dishes it carries into the dishwasher. The environment

Algorithm 2: The program for the dishwasher robot

```

loop
  while  $\exists x. \text{OnRobot}(x)$  do  $\pi x : \{d_1, d_2\}. \text{unload}(x)$ 
   $\pi y : \{r_1, r_2\}. \text{goto}(y)$ ;
  while  $\exists x. \text{DirtyDish}(x, y)$  do  $\pi x : \{d_1, d_2\}. \text{load}(x, y)$ 
   $\text{goto}(\text{kitchen})$ 
||
loop  $\pi x : \{d_1, d_2\}, y : \{r_1, r_2\}. \text{addDish}(x, y)$ 

```

R	D	Nodes (TS)	Edges (TS)	Nodes (St)	Edges (St)	Time [s]
1	1	22	25	16	19	2.6
1	2	150	203	128	176	155.4
1	3	–	–	–	–	–
2	1	109	168	87	110	69.0
2	2	–	–	–	–	–
3	1	483	857	413	543	1885.7
3	2	–	–	–	–	–

Table 1: Evaluation Results for the Dish Robot Domain

has actions that represent used dishes being placed in arbitrary rooms. Every dish can only be used once in this fashion. The basic action theory, program, and temporal specification are specified below.

#### Initial situation:

$\text{Dish}(x) \equiv (x = d_1 \vee x = d_2)$ ,  $\text{Room}(x) \equiv (x = r_1 \vee x = r_2)$   
 $\forall x. \text{At}(x) \equiv x = \text{kitchen}$   
 $\forall x. \text{New}(x) \equiv \text{Dish}(x) \wedge \forall y. \neg \text{DirtyDish}(x, y) \wedge \neg \text{OnRobot}(x)$   
 $\text{OnRobot}(x) \supset \text{Dish}(x) \wedge \neg \exists y. \text{DirtyDish}(x, y)$   
 $\text{DirtyDish}(x, y) \supset \text{Dish}(x) \wedge \text{Room}(y) \wedge \neg \text{OnRobot}(x)$

#### Precondition axioms:

$\square \text{Poss}(\text{load}(x, y)) \equiv \text{DirtyDish}(x, y) \wedge \text{At}(y)$   
 $\square \text{Poss}(\text{unload}(x)) \equiv \text{OnRobot}(x) \wedge \text{At}(\text{kitchen})$   
 $\square \text{Poss}(\text{addDish}(x, y)) \equiv \text{New}(x) \wedge \text{Room}(y)$   
 $\square \text{Poss}(\text{goto}(x)) \equiv \text{Room}(x) \vee x = \text{kitchen}$

#### Successor state axioms:

$\square[a] \text{DirtyDish}(x, y) \equiv a = \text{addDish}(x, y) \vee \text{DirtyDish}(x, y) \wedge a \neq \text{load}(x, y)$   
 $\square[a] \text{OnRobot}(x) \equiv \exists y. a = \text{load}(x, y) \vee \text{OnRobot}(x) \wedge a \neq \text{unload}(x)$   
 $\square[a] \text{New}(x) \equiv \text{New}(x) \wedge \neg \exists y. a = \text{addDish}(x, y)$   
 $\square[a] \text{At}(x) \equiv a = \text{goto}(x) \vee \text{At}(x) \wedge \neg \exists y. a = \text{goto}(y)$

**Program:** The program is shown in Algorithm 2. It is to be understood as being *precondition extended*, i.e., an underlined action  $\underline{A}(\vec{\sigma})$  stands for  $\pi?$ ;  $A(\vec{\sigma})$ , where  $\pi$  is the right-hand side of the precondition axiom for  $A$ , instantiated by  $\vec{\sigma}$ . For better readability,  $\delta^*$  is written as **loop**  $\delta$ .

**Specification:**  $\mathcal{F} \mathcal{G} \neg \exists x, y. \text{DirtyDish}(x, y)$

**Results:** Table 1 summarizes the experimental results on the dishwasher domain. Here,  $R$  and  $D$  denote the number of rooms and dishes, respectively,  $\text{Nodes (TS)}$  and  $\text{Edges (TS)}$  refer to the number of nodes and edges of the resulting transition system, while  $\text{Nodes (St)}$  and  $\text{Edges (St)}$  denote the

corresponding metrics of the discovered strategy. *Time* indicates the duration in seconds for completing the algorithm, with a timeout set at 120 minutes. As expected, the size of transition system, and the time needed to construct it, grows with additional rooms or dishes. Interestingly, the number of dishes has a bigger impact than the number of rooms. Intuitively, this is because the program contains more choices for dishes than for rooms, which are furthermore nested inside inner loops. Accordingly, adding one more dish results in a more significant blow-up than adding a room.

## Warehouse Robot

The second domain is a warehouse robot, adapted from an example in (Claßen and Zarriß 2017). Here, the robot can move boxes from one shelf of a warehouse to another. The boxes may contain an unknown number of objects, and it is unknown whether and which objects are fragile. Accidentally (i.e., due to the environment’s choice), the robot may drop a box, breaking all fragile objects in it, unless the box contains bubble wrap. The robot has the option to put bubble wrap into a box.

### Initial situation:

$$\begin{aligned}
&\forall x. Shelf(x) \equiv (x = s_1 \vee x = s_2) \\
&\forall x. Box(x) \equiv (x = b_1 \vee x = b_2) \\
&\forall x. \exists y. In(x, y) \supset \neg Shelf(x) \wedge \neg Box(x) \\
&\exists x. Wrap(x) \\
&\forall x. \neg Broken(x) \wedge \neg Holding(x) \\
&RAt(s_1) \wedge \forall x. Box(x) \supset At(x, s_1) \\
&\forall x, y. (In(x, y) \wedge Box(y)) \supset At(x, s_1) \\
&\forall y. y \neq s_1 \supset \neg RAt(y) \wedge \forall x. \neg At(x, y) \\
&\forall x, y. In(x, y) \supset \neg Wrap(x)
\end{aligned}$$

### Precondition axioms:

$$\begin{aligned}
&\square Poss(take(x, y)) \equiv At(x, y) \wedge RAt(y) \\
&\square Poss(move(x, y)) \equiv RAt(x) \wedge Shelf(y) \wedge (x \neq y) \\
&\square Poss(put(x, y)) \equiv Holding(x) \wedge RAt(y) \\
&\square Poss(addWrap(x)) \equiv \exists y. RAt(y) \wedge At(x, y) \\
&\square Poss(drop(x)) \equiv Holding(x)
\end{aligned}$$

### Successor state axioms:

$$\begin{aligned}
&\square[a]RAt(y) \equiv \exists x. a = move(x, y) \\
&\quad \vee RAt(y) \wedge \neg \exists z. a = move(y, z) \\
&\square[a]At(x, y) \equiv \exists z[a = move(z, y) \wedge \\
&\quad \exists v(Holding(v) \wedge (v = x \vee In(x, v)))] \\
&\quad \vee At(x, y) \wedge \neg \exists z[a = move(y, z) \wedge \\
&\quad \exists v(Holding(v) \wedge (v = x \vee In(x, v)))] \\
&\square[a]Holding(x) \equiv \exists y. a = take(x, y) \\
&\quad \vee Holding(x) \wedge \neg \exists y. a = put(x, y) \\
&\square[a]Broken(x) \equiv \exists y. a = drop(y) \wedge In(x, y) \wedge Fragile(x) \\
&\quad \wedge \neg \exists z. In(z, y) \wedge Wrap(z) \vee Broken(x) \\
&\square[a]In(x, y) \equiv a = addWrap(y) \wedge Wrap(x) \vee In(x, y)
\end{aligned}$$

**Program:** The program for the warehouse robot is shown in Algorithm 3. The notation  $\delta^?$  stands for an optional execution of  $\delta$ , and is formally defined as  $\delta^? \doteq (\delta \mid nil)$ . Note that the choice for putting bubble wrap is up to the robot, but that of the box getting dropped is due to the environment.

## Algorithm 3: The program for the warehouse robot

```

loop
   $\pi_{l_0, l_1} : \{s_1, s_2\}. [move(l_0, l_1)^?;$ 
   $\pi_b : \{b_1, b_2\}. (wrap(b)^?; take(b, s_1); drop(b)^?;$ 
   $\pi_{l_2} : \{s_1, s_2\}. move(l_1, l_2); put(b, l_2)]$ 

```

B	Nodes (TS)	Edges (TS)	Nodes (St)	Edges (St)	Time [s]
1	162	162	46	58	11.8
2	7584	7830	2038	2647	13328.9
3	–	–	–	–	–

Table 2: Evaluation Results for the Warehouse Domain

**Specification:**  $\mathcal{F} \forall o. In(o, b_1) \supset \neg Broken(o) \wedge At(o, s_2)$

**Results:** Table 2 presents the results of the experiments on the warehouse robot domain, where B denotes the number of boxes, and the other columns are as before (with a timeout of 240 minutes). As can be seen, the method struggles more with this domain than the previous one, which is due to several reasons. For one, the successor state axioms for the warehouse robot actually exploit the expressivity of the class of acyclic theories more than do the ones for the dishwasher robot. Note that the dishwasher BAT actually falls into the class of *local-effect* theories (Liu and Lakemeyer 2009), a subset of acyclic theories where regression works much simpler (i.e., does not introduce additional quantifiers), and consequently results in less complex formulas. Moreover, the warehouse robot suffers from the same problem that causes the *Gripper* domain to be a challenge in classical planning: There is a number of objects, each of which has to be handled in the same way. For solving the task, the order in which objects are handled is hence irrelevant, yet the system considers all possible permutations, resulting in a blow-up. The problem is amplified by the fact that handling a single box in this domain is a slightly complex task in itself, containing a sequence of actions with several choice points. An interesting avenue for future work would be to improve our method to be able to detect and deal with symmetries of this kind.

## 6 Conclusion

In this paper, we have presented an approach to the realization of GOLOG programs with uncontrollable actions. We have formulated the realization problem as a synthesis problem, where parts of the program are under the environment’s control and the agent needs to determine a policy that realizes the program while satisfying the temporal specification. The presented approach synthesizes policies for LTL<sub>f</sub> specifications on GOLOG programs with first-order action theories that allow for an unbounded number of objects and non-local effects, an expressive and decidable fragment of the situation calculus. We have demonstrated the feasibility of the approach in two example domains. The synthesis method can also be understood as a (restricted) first-order variant of LTL<sub>f</sub> synthesis, where the user may provide a declarative specification of the agent’s capabilities along with a partial strategy. Future work could further investigate this relation.

## Acknowledgements

The research has been supported by the Alexander von Humboldt Foundation with funds from the Federal Ministry for Education and Research, Germany, by the European Research Council (ERC), Grant agreement No. 885107, and by the Excellence Strategy of the Federal Government and the Länder, Germany.

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