

Reactive High-level Behavior Synthesis for an ATLAS Humanoid Robot

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Abstract—In this paper we present an *end-to-end* approach for the automatic generation of software that implements reactive high-level robot behaviors. Our approach starts with an informal description of the system, as well as a task’s goals and initial conditions. First, we abstract the problem and automatically construct a formal task specification in a fragment of Linear Temporal Logic (LTL). One highlight of our formalism is that it accounts for the possible failures of the various system components. We then synthesize a *reactive* mission plan that is guaranteed to satisfy the formal specification. Finally, we automatically generate a state machine that instantiates the synthesized plan in software.

Specifically, this paper focuses on Team ViGIR’s software and ATLAS humanoid robot. The proposed approach has been implemented and open-sourced as a collection of Robot Operating System (ROS) packages. We demonstrate the efficacy of our approach via experiments with ATLAS in the lab.

Body of my TODO example

LIST OF TODOS, FIXES, OPEN ISSUES

Title of my TODO example (hyperlink)	1
Consider mutex for grounding conflicts in this paper?	1
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Replace 2014 JFR with new JFR paper	2
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Generate “software implementation” or FlexBE SM?	3
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Fig. 1: Team ViGIR’s ATLAS humanoid robot on the first day of the DRC Finals. (Photo credit: DARPA)

I. INTRODUCTION

...

Example 1: Consider the high-level task: “Walk to the valve and turn it” (Fig. 1). This would be an intuitive way to express the task from a non-expert user’s point-of-view. However, ... (not formal, preconditions, templates, reaction to failures, etc.)

Contributions (brain dump):

- Partial to full specification
 - Most intuitive from the users point-of-view
 - Limited message size over bad comms (send partial specification → compile and synthesize onboard)
- Multi-paradigm specification (objectives and initial conditions from user, topology/modes, preconditions, task)

Also consider mutex for grounding conflicts?

- Generalization of activation-completion paradigm [1] Formalize how we deal with failures (and other outcomes). Graceful degradation. Human operators also included in environment.
- Integration with FlexBE and ROS
- Experimental validation on ATLAS

Literature Survey:

- Vasu’s “fast-slow” paradigm [1]
- Alternatives to LTL and GR(1) Synthesis
 - Classic AI planners, STRIPS-type planners
 - Optimization-based LTL synthesis (Eric Wolff)
 - **co-safe LTL** (Lydia Kavraki, Calin Belta, etc.) We could have used it as a different formalism and generated FlexBE state machines that way.

Check whether co-safe LTL encodes reactivity w.r.t. adversarial environment (worst case)

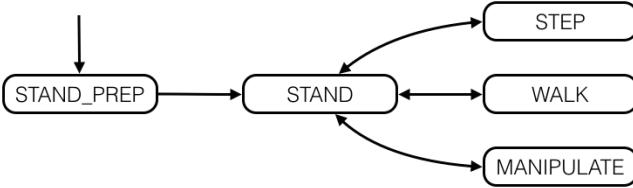


Fig. 2: Excerpt from the BDI control mode interface. Some changes between modes (depicted as arrows) are unidirectional while others are bidirectional. After Team ViGIR’s initial checkout, ATLAS is in the STAND_PREP control mode.

II. PRELIMINARIES

A. ATLAS Humanoid Robot

ATLAS (Fig. 1) is an anthropomorphic robot developed by Boston Dynamics, Inc. (BDI). Team ViGIR chose to leverage the basic capabilities provided by the Boston Dynamics Application Programming Interface (API). In the context of this work, we are especially interested in BDI’s “control mode” interface (Fig. 2). The active control mode dictates which joints are controlled by the low-level BDI controllers and which joints we can command. For example, in STAND and MANIPULATE BDI’s software handles balancing.

ATLAS is equipped with a number of sensors, most notably a Carnegie Robotics Multisense SL¹ mounted as the head. For the DRC Finals, ATLAS was equipped with two Robotiq 3-finger hands,² providing manipulation capabilities.

B. Team ViGIR’s Approach to the DRC Finals

Team ViGIR based its software on the Robot Operating System (ROS) [2], [3]. In this section, we highlight some elements of the software’s design that are relevant to high-level control and behavior synthesis. For a complete overview of Team ViGIR’s approach, we refer the interested reader to [4]. From this point on, we will refer to ATLAS running Team ViGIR’s software as the *system*, \mathcal{S} .

If the new JFR paper is accepted by February, replace this old reference, [4], in the final submission.

As mentioned, Team ViGIR uses BDI’s control mode interface. This means that some system capabilities are preconditioned on a certain control mode being active. For example, in order to execute an arm trajectory, the system must be in MANIPULATE. In addition, the system’s operation has to respect the constraints on the possible control mode changes (Fig. 2).

In terms of manipulation, Team ViGIR employs the concept of object templates [5]. In brief, the system presents the human operator with perception data, e.g., a point cloud. Then, the operator detects objects of interest and overlays an object template on top of them. These templates contain metadata, such as relative robot poses from which the object is reachable, relative pre-grasp and grasp end-effector poses, as well as finger configurations corresponding to different

¹<http://carnegierobotics.com/multisense-sl>

²<http://robotiq.com/products/industrial-robot-hand>

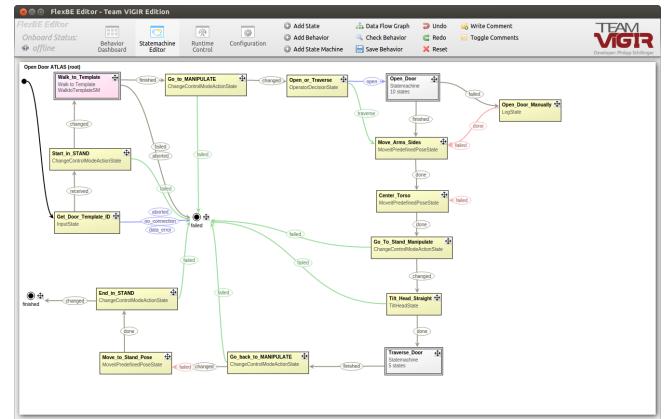


Fig. 3: A manually designed high-level behavior for carrying out the DRC Finals’ “Door” task. The initial state is indicated by the black arrow originating from the top left. The behavior has two outcomes, “finished” (bottom left) and “failed” (center). Yellow states are parametrized state implementations, gray states are state machines, and purple states are other high-level behaviors embedded in this one.

grasps. In addition, object templates provide manipulation affordances. For instance, the “door” template provides affordances such as “turn (handle) clockwise” and “push”.

Skim Albert’s paper and improve this description of OT.

High-level Control: Team ViGIR’s approach to high-level control is especially relevant to this work. Its corner stone is the Flexible Behavior Engine³ (FlexBE) [6], [7], which is a major extension of the SMACH high-level executive [8].

Based on the FlexBE framework, developers create “state implementations”, $s \in S$. These are small, atomic blocks of code that each interface with some primitive, lower-level, system capability. Furthermore, each state implementation defines a number of outcomes $Out(s)$, e.g. $\{\text{done}, \text{failed}, \text{aborted}\}$. The state implementations can be composed to form hierarchical state machines (SM), which encode the logic of execution as well as the flow of data. Specifically, state machines consist of *parametrized instantiations*, $s_p \in S_p$, of the state implementations S . For example, if a state implementation corresponds to changing control modes, its parametrized counterparts correspond to changing to specific control modes. State machines also have outcomes themselves. The top-level state machine will be referred to as a “behavior”.⁴ The composition of new state machines is historically done manually by an expert user.

Finally, composition of behaviors and supervision of their execution takes place in FlexBE’s graphical user interface⁵ (GUI). Figure 3 depicts an example of a high-level behavior designed manually in the FlexBE GUI’s editor.

³https://github.com/team-vigir/flexbe_behavior_engine

⁴https://github.com/team-vigir/vigir_behaviors

⁵https://github.com/team-vigir/flexbe_chrome_app

C. Linear Temporal Logic and Reactive LTL Synthesis

Linear Temporal Logic (LTL) is a formal language that combines Boolean (\neg , \wedge , \vee) and temporal (next \bigcirc , until \mathcal{U}) operators. Additional temporal operators, always \square , eventually \diamond , can be derived from those. LTL formulas are constructed from Boolean atomic propositions $\pi \in AP$. In the context of our work, the set of atomic propositions, AP , consists of propositions controlled by the system, \mathcal{Y} , and propositions controlled by the dynamic, and possibly adversarial, environment, \mathcal{X} . That is, $AP = \mathcal{X} \cup \mathcal{Y}$.

In order to synthesize *reactive* mission plans in a computationally tractable manner, we use the GR(1) fragment of LTL [9]. GR(1) formulas φ have an assume-guarantee structure between the dynamic environment (e) and the system (s):

$$\begin{aligned}\varphi &= (\varphi_e \Rightarrow \varphi_s), \\ \varphi_e &= \varphi_e^i \wedge \varphi_e^t \wedge \varphi_e^g, \\ \varphi_s &= \varphi_s^i \wedge \varphi_s^t \wedge \varphi_s^g,\end{aligned}\quad (1)$$

where the superscript i denotes initial conditions, t safety assumptions/requirements, and g liveness assumptions/requirements (i.e., goals) for e and s , respectively.

GR(1) synthesis involves setting up a two-player game between e and s [9]. If a GR(1) specification φ is realizable for s , we can extract a finite-state automaton; specifically, a Mealy machine. This automaton encodes a strategy for s that guarantees φ_s for any evolution of e that satisfies φ_e .

If there's any space left, define the FSA/Mealy machine mathematically. Possibly connect to FlexBE SMs later.

III. PROBLEM STATEMENT

Problem 1 (Discrete Abstraction): Given ATLAS' control mode transition constraints and the available actions \mathcal{A} , define a discrete abstraction \mathcal{D} of the robot-plus-software system, S , that captures the execution and outcomes of the primitive capabilities (control mode transitions and actions). In addition, maintain a mapping, \mathcal{D}_M , between the primitive capabilities of the system S and the corresponding elements of the discrete abstraction.

Problem 2 (Formal Task Specification): Given a task in terms of goals \mathcal{G} , the task's initial conditions \mathcal{I} , and the discrete abstraction \mathcal{D} of the system S that is to carry out the task,⁶ automatically generate a specification \mathcal{T}_S that encodes, in a formal language, the task being carried out by S .

To illustrate these concepts, consider again the scenario in Example 1. We could say that the task's goals \mathcal{G} are `{turn_valve}` and its initial conditions \mathcal{I} are, e.g., `{stand}`. The system S is ATLAS running Team ViGIR's software, as described in Section II-B.

Problem 3 (Behavior Synthesis): Given a formal task specification, \mathcal{T}_S , and the mapping \mathcal{D}_M , automatically generate a software implementation of a discrete, high-level, control strategy that is verifiably guaranteed to satisfy \mathcal{T}_S .

Generate "software implementation" or FlexBE SM?

⁶In general, a discrete abstraction of the robot's workspace would also be an input to this problem, but we are not modeling it explicitly here.

IV. DISCRETE ABSTRACTION

A. Control Modes & Actions

We model ATLAS' control mode interface (c.f. Fig. 2) as a transition system $(\mathcal{M}, \rightarrow)$, where \mathcal{M} is the set of states, each corresponding to one control mode, $m \in \mathcal{M}$, and \rightarrow is a set of valid control mode transitions (subset of $\mathcal{M} \times \mathcal{M}$). In addition, we define $Adj(m) = \{m' \in \mathcal{M} \mid (m, m') \in \rightarrow\}$ and also allow self-transitions, i.e., $m \in Adj(m)$, $\forall m \in \mathcal{M}$.

Furthermore, ATLAS can perform actions. Each action, $a \in \mathcal{A}$, corresponds to an atomic capability of the system, e.g., generation of a footprint plan or closing the robot's fingers. Actions may also have one or more preconditions. Action preconditions can be control modes or other actions, i.e., $Prec(a) \in 2^{(\mathcal{A} \cup \mathcal{M})}$, $a \notin Prec(a)$, $\forall a \in \mathcal{A}$.

Are actions, preconditions, and the control mode TS part of the discrete abstraction or should they be defined in the Preliminaries instead? What is \mathcal{D} mathematically?

B. Atomic Propositions

Similar to question about \mathcal{D}_M . Define atomic propositions in terms of FlexBE states or primitive capabilities?

We adopt a paradigm that generalizes the one in [1]. We abstract the discrete actions, $a \in \mathcal{A}$, that ATLAS can perform using one system proposition, π_a , per action and one environment proposition, π_a^o , per possible outcome of that action, $o \in Out(a)$. Similarly,⁷ for each control mode, $m \in \mathcal{M}$, we have a system proposition π_m and a number of outcome propositions π_m^o . For both actions and control mode transitions, the outcomes that are of most interest in the context of this paper are completion (c) and failure (f) of the action. That is, $Out(a) = Out(m) = \{c, f\}$. Therefore, the set of atomic propositions AP is given by Eq. (2):

$$\mathcal{Y} = \bigcup_{a \in \mathcal{A}} \pi_a \bigcup_{m \in \mathcal{M}} \pi_m, \quad (2a)$$

$$\mathcal{X}' = \mathcal{X} \bigcup_{a \in \mathcal{A}} \bigcup_{o \in Out(a)} \pi_a^o \bigcup_{m \in \mathcal{M}} \bigcup_{o \in Out(m)} \pi_m^o, \quad (2b)$$

where \mathcal{X} are environment propositions other than outcome propositions, e.g., ones that abstract sensors, as per [10].

C. Proposition Mapping

The mapping $\mathcal{D}_M : AP \rightarrow S_p$ from Problem 1 relates parametrized FlexBE state implementations with atomic propositions. Specifically, each activation proposition π_p is mapped to the execution of a parametrized state implementation s_p . In addition, each outcome proposition π_p^o is mapped to an outcome $o \in Out(s)$ of the state implementation. In practice, an outcome proposition can correspond to multiple outcomes of the state implementation. For example, we may want to treat the outcomes `failed` and `aborted` of s_p as failure, thus mapping the proposition π_p^f to both of them.

⁷The distinction between action and control mode propositions is purely for the sake of clarity of notation. There is nothing special about either.

Define \mathcal{D}_M in terms of FlexBE states as above or more generally in terms of primitive system capabilities?

V. FORMAL TASK SPECIFICATION

A. Multi-Paradigm Specification

Specifying a robot task in a formal language can be time consuming and error prone. It also requires an expert user. To alleviate these issues, we employ a multi-paradigm specification approach. We first observe that there are portions of the task specification \mathcal{T}_S that are going to be system-specific and portions that are going to be task-specific, such as the task's goals. Intuitively, a user should only have to specify the goals without worrying about the internals of the robot and the software it is running. For ATLAS, we can automatically infer which control modes and actions are pertinent to a task. Finally, the initial conditions are either specified by the user or detected at runtime.

Referring to Problem 2, we get the goals, \mathcal{G} , and initial conditions, \mathcal{I} , from the user. The discrete abstraction, \mathcal{D} , is ATLAS-specific and we assume that it has been created a priori from expert developers, according to Section IV. We can now automatically compile the task specification \mathcal{T}_S in (the GR(1) fragment [9] of) Linear Temporal Logic. Since LTL is compositional, we can generate individual LTL formulas and then conjunct them to get the full LTL specification.

Demonstrate how, given partial specification, we can bring in only those actions \mathcal{A} that are necessary.

B. Specification of Actions and Control Mode Constraints

1) *Generic Formulas*: We say that an activation proposition π_p , $p \in \{a, m\}$, is True when the corresponding primitive functionality is being activated and False when it is not being activated⁸. Therefore, the system safety requirement (3) dictates that all activation propositions $\pi_p \in \mathcal{Y}$ should turn False once an outcome has been returned.

$$\bigwedge_{o \in Out(p)} \square (\pi_p \wedge \bigcirc \pi_p^o \Rightarrow \bigcirc \neg \pi_p) \quad (3)$$

The environment safety assumption (4) dictates that the outcomes, π_p^o , of the activation of any primitive functionality p are mutually exclusive. For example, an action cannot both succeed and fail. Formula (4) also allows for no outcome being True.

Formula (4) requires the \bigcirc operators to synthesize properly (slugs), but intuitively, they shouldn't be there.

$$\bigwedge_{o \in Out(p)} \square (\bigcirc \pi_p^o \Rightarrow \bigwedge_{o' \neq o} \bigcirc \neg \pi_p^{o'}) \quad (4)$$

The environment safety assumption (5) constraints the value of outcomes. Specifically, it dictates that, if an outcome is False and the corresponding primitive functionality is not

⁸Note that this is in contrast to [1], where π_p being False stands for the primitive functionality p being *deactivated*, e.g., turning a camera off.

being activated, then that outcome should remain False. It is a generalization of formula (4) in [1].

$$\bigwedge_{o \in Out(p)} \square (\neg \pi_p^o \wedge \neg \pi_p \Rightarrow \bigcirc \neg \pi_p^o) \quad (5)$$

2) *Action-specific Formulas*: The following formulas encode the connection between the activation and the possible outcomes of the robot's actions, $a \in \mathcal{A}$.

The environment safety assumptions (6) dictate that the value of an outcome should not change if the corresponding action has not been activated again. In other words, outcomes persist.

$$\bigwedge_{o \in Out(a)} \square (\pi_a^o \wedge \neg \pi_a \Rightarrow \bigcirc \pi_a^o) \quad (6)$$

The environment liveness assumption (7) is a fairness condition. It states that (always) eventually, the activation of an action will result in an outcome. The disjunct $\neg \pi_a$ is added in order to prevent the system from winning the game by never activating the action.

$$\square \diamond ((\pi_a \wedge \bigvee \bigcirc \pi_a^o) \vee \neg \pi_a) \quad (7)$$

The system safety requirement (8) demonstrates how a formula encoding the preconditions of an action, $Prec(a)$, looks like in the activation–outcomes paradigm.

$$\square \left(\bigvee_{p \in Prec(a)} \neg \pi_p^c \Rightarrow \neg \pi_a \right) \quad (8)$$

where the superscript $c \in Out(p)$ stands for “completion”.

3) *Control Mode Formulas*: For brevity of notation, let

$$\varphi_m = \pi_m \wedge \bigwedge_{m' \neq m} \neg \pi_{m'}$$

Activating φ_m , as opposed to π_m , takes into account the mutual exclusion between control modes $m \in \mathcal{M}$. Also let

$$\varphi_{\mathcal{M}}^{none} = \bigwedge_{m \in \mathcal{M}} \neg \pi_m,$$

where $\varphi_{\mathcal{M}}^{none}$ being True stands for not activating any control mode transitions, i.e., staying in the same control mode.

The system safety requirements (9) encode a topological transition relation, such as the BDI control mode transition system.

$$\bigwedge_{m \in \mathcal{M}} \square (\bigcirc \pi_m^c \Rightarrow \bigvee_{m' \in Adj(m)} \bigcirc \varphi_{m'} \vee \bigcirc \varphi_{\mathcal{M}}^{none}) \quad (9)$$

The environment safety assumptions (10) enforce mutual exclusion between the BDI control modes.

Formula (10) also requires the \bigcirc operators to synthesize properly (slugs), but intuitively, they shouldn't be there.

$$\bigwedge_{m \in \mathcal{M}} \square (\bigcirc \pi_m^c \Leftrightarrow \bigwedge_{m' \neq m} \bigcirc \neg \pi_{m'}^c) \quad (10)$$

The environment safety assumptions (11) govern how the active control mode can change in a single time step in response to the activation of a control mode transition.

$$\bigwedge_{m \in \mathcal{M}} \bigwedge_{m' \in \text{Adj}(m)} \square \left(\pi_m^c \wedge \varphi_{m'} \Rightarrow \left(\bigcirc \pi_m^c \bigvee_{o \in \text{Out}(m')} \bigcirc \pi_{m'}^o \right) \right) \quad (11)$$

The environment safety assumptions (12) dictate that the value of the outcomes of control mode transitions must not change if no transition is being activated, i.e., they must persist.

$$\bigwedge_{m \in \mathcal{M}} \bigwedge_{o \in \text{Out}(m)} \square \left(\pi_m^o \wedge \varphi_{\mathcal{M}}^{\text{none}} \Rightarrow \bigcirc \pi_m^o \right) \quad (12)$$

The environment liveness assumption (13) is the equivalent of the fairness condition (7) for control mode transitions. This single formula accounts for all control modes.

$$\square \diamond \left(\bigvee_{m \in \mathcal{M}} \left(\varphi_m \wedge \bigvee_{o \in \text{Out}(m)} \bigcirc \pi_m^o \right) \vee \varphi_{\mathcal{M}}^{\text{none}} \right) \quad (13)$$

C. Specification of Initial Conditions

For each action, a , and control mode, m , in the initial conditions, \mathcal{I} , the completion proposition should be True in the environment initial conditions (14). All other outcome propositions corresponding to those actions and control modes, as well as all outcome propositions corresponding to any other actions and control modes, should be False.

$$\varphi_i^e = \bigwedge_{i \in \mathcal{I}} \left(\pi_i^c \bigwedge_{o \in \text{Out}(i) \setminus \{c\}} \neg \pi_i^o \right) \wedge \bigwedge_{j \notin \mathcal{I}} \bigwedge_{o \in \text{Out}(j)} \neg \pi_j^o \quad (14)$$

Activation propositions are False regardless of whether that action or control mode is in the initial conditions or not (15). The reason being that, intuitively, if we want something to be an initial condition, then we shouldn't have the resulting controller re-activate it at the beginning of execution.

$$\varphi_i^s = \bigwedge_{i \in \mathcal{I}} \neg \pi_i \wedge \bigwedge_{j \notin \mathcal{I}} \neg \pi_j \quad (15)$$

Do I move ICs to the end of the section and include memory props explicitly? Or leave them here, but state memory ICs along with “infinite-to-finite” formulas?

D. Specification of Task Goals

The system initial condition (16), safety requirements (17) and (18), and liveness requirement (19) are used to reason about the satisfaction of the system's goals, $g \in \mathcal{G}$, in a finite run (as opposed to infinite execution, which is what LTL is defined over). In this finite run paradigm, the synthesized state machine (SM) itself has outcomes, $o \in \text{Out}(SM)$. The propositions corresponding to the SM's outcomes, π_{SM}^o , are system, not environment, propositions. The system propositions, μ_g , serve as memory of having accomplished each goal (c.f. [11]).

$$\bigwedge_{g \in \mathcal{G}} \neg \mu_g \quad (16)$$

$$\bigwedge_{g \in \mathcal{G}} \square \left(\bigcirc \pi_g^c \vee \mu_g \Leftrightarrow \bigcirc \mu_g \right) \quad (17)$$

$$\square \left(\pi_{SM}^c \Leftrightarrow \bigwedge_{g \in \mathcal{G}} \mu_g \right) \quad (18a)$$

$$\square \left(\pi_{SM}^f \Leftrightarrow \bigvee_{\pi \in \mathcal{Y}} \pi^f \right) \quad (18b)$$

$$\bigwedge_{o \in \text{Out}(SM)} \square \left(\pi_{SM}^o \Rightarrow \bigcirc \pi_{SM}^o \right) \quad (18c)$$

$$\square \diamond \left(\bigvee_{o \in \text{Out}(SM)} \pi_{SM}^o \right) \quad (19)$$

The time complexity of synthesis is cubic in the number of liveness requirements. We save by only having one. Although that's probably dominated by the complexity being exponential in the number of propositions.

Formula (17) does not guarantee that the goals will be achieved in a specific order. However, that is often desirable. To this end, we can define the goals as an ordered set $\mathcal{G} = \{g_1, g_2, \dots, g_n\}$, where $g_i < g_j$ for $i < j$, and the relation $g_i < g_j$ means that goal g_i has to be achieved before g_j . With this definition, we can replace the safety requirement (17) with (20), whenever strict goal order is desired.

$$\bigwedge_{i=1}^n \square \left((\pi_{g_i} \wedge \bigcirc \pi_{g_i}^c) \wedge \mu_{g_{i-1}} \vee \mu_{g_i} \Leftrightarrow \bigcirc \mu_{g_i} \right), \quad (20)$$

where $\mu_{g_0} \triangleq \text{True}$. Formula (20) forces the system to carry out goal g_i after it has accomplished goal g_{i-1} . It can still activate the action corresponding to π_{g_i} earlier, as necessitated by other parts of the task, but that will not count towards achievement of g_i , as indicated by μ_{g_i} being True.

Finally, these auxiliary (memory and SM outcome) propositions have to be added to the system propositions:

$$\mathcal{Y}' = \mathcal{Y} \bigcup_{g \in \mathcal{G}} \mu_g \bigcup_{o \in \text{Out}(SM)} \pi_{SM}^o$$

VI. HIGH-LEVEL BEHAVIOR SYNTHESIS

We tackle Problem 3 in two sequential steps. First, we automatically generate a correct-by-construction automaton from the formal specification \mathcal{T}_S using GR(1) synthesis (see [9] and Section II-C). Specifically, we employ the synthesis algorithm in [12], which can handle a slightly larger fragment of LTL than GR(1). Namely, the one that includes \bigcirc (next) operators in liveness formulas, such as in formulas (7) and (13). This algorithm was first used in [1].

@HKG, is it true that Vasu's paper was the first case of synthesis for this fragment? Had Ruediger used it before?

Second, we use the mapping $\mathcal{D}_M : AP \rightarrow S_P$ to instantiate the abstract automaton as a concrete software

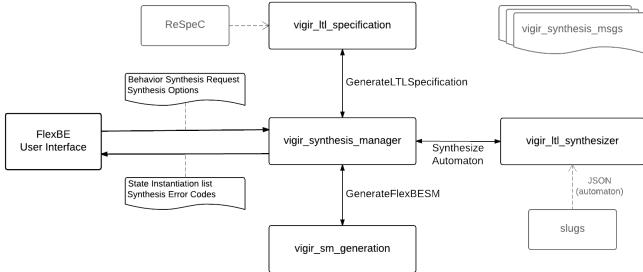


Fig. 4: Team ViGIR’s “Behavior Synthesis” ROS packages and the nominal workflow (clockwise, starting from the left).

- (i) Add ReSpeC to image (move msgs to the left)
- (ii) Either number all steps or none
- (iii) Rename 3 services.

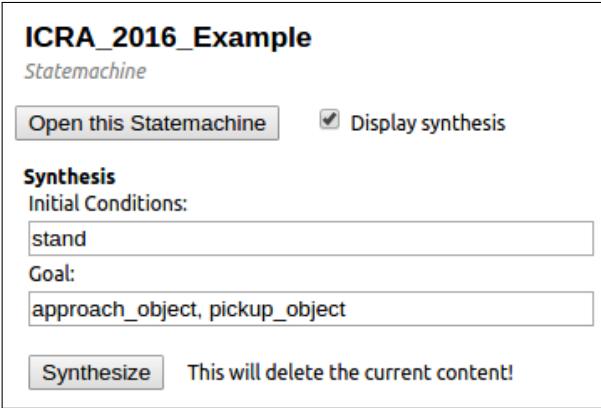


Fig. 5: Screenshot of the FlexBE Editor’s synthesis menu.

implementation, i.e., an executable state machine in the FlexBE framework introduced in Section II-B.

VII. ROS IMPLEMENTATION

We have implemented all aspects of our approach in `vigir_behavior_synthesis`,⁹ a collection of Robot Operating System (ROS) Python packages. Figure 4 depicts these packages as well as the nominal workflow.

The synthesis action server (`vigir_synthesis_manager`) receives a request from the user via FlexBE’s GUI. Given the user’s input (initial conditions and goals), the server first requests a full set of LTL formulas for ATLAS from the `GenerateLTLSpecification` service (`vigir_lt1_specification` package). The generation of the LTL formulas from Section V is delegated to our “Reactive Specification Construction kit” (ReSpeC),¹⁰ which is a Python framework with rudimentary ROS integration.

The `vigir_lt1_synthesizer` package acts as a wrapper for external synthesis tools (currently, [12] is supported). Given the generated LTL specification, the `SynthesizeAutomaton` service returns a finite-state automaton that is guaranteed to satisfy it, if one exists. Finally, the server requests a `StateInstantiation` message from

the `GenerateFlexBESM` service (`vigir_sm_generation` package). This message provides the FlexBE Editor with sufficient information to generate Python code, i.e., an executable state machine that instantiates the synthesized automaton. The corresponding action, services, and messages are defined in the `vigir_synthesis_msgs` package.

The following excerpt¹¹ is taken from the `StateInstantiation` list message, which is the end product of the `vigir_behavior_synthesis` workflow (Fig. 4). Specifically, this excerpt corresponds to the primitive functionality `object_template`, which appears in Fig. 6a.

```

state_path: /4_object_template
state_class: InputState
parameter_names: [request]
parameter_values: [InputState.SELECTED_OBJECT_ID]
outcomes: [no_connection, aborted, received, data_error]
transitions: [failed, failed, 6_manipulate, failed]
autonomy: [0, 0, 0, 0]
userdata_keys: [data]
userdata_remapping: [template_id]

```

VIII. EXPERIMENTAL VALIDATION

Provide data on how computationally costly/cheap behavior synthesis is. Time vs number of actions?

Team ViGIR did not employ high-level behavior synthesis during the DRC Finals. However, we later carried out experimental demonstrations on ATLAS in the lab. Due to a hardware issue, ATLAS could not locomote. Thus, in addition to two experimental demonstrations, we present a simulation run carried out in Gazebo, using the same operator and onboard software. We summarize these demonstrations below. Please also refer to the accompanying video.

A. Behavior Development using Synthesis

In the first experimental demo, we show how a high-level behavior is specified and synthesized starting from scratch¹². Once the state machine has been instantiated (Fig. 6a), it is ready for execution (Fig. 6b).

B. Online Modifications using Synthesis

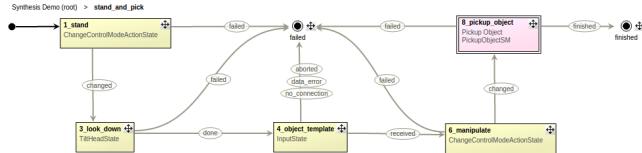
For the second experimental demonstration, consider a scenario where the operator has designed a state machine that addresses a high-level task (either manually or via synthesis). ATLAS is then deployed and starts carrying out this task. If, during execution, an *unexpected* situation arises, the operator can use FlexBE’s runtime modification capability (Fig. 7). In this case, behavior execution is “locked” at some state, i.e., this state is prevented from returning an outcome (Fig. 7a). Then, the operator specifies a new high-level behavior meant to address the unexpected situation. Once this new state machine is instantiated (Fig. 7b), it is connected to the previous one (Fig. 7a), and execution resumes.

¹¹We have omitted some details for the sake of brevity and clarity of presentation. For example, most list elements are strings, e.g., “`template_id`”.

¹²The LTL specification and the synthesized automaton are available at: <https://gist.github.com/spmaniato/c37fb12e874c73d986da>

⁹https://github.com/team-vigir/vigir_behavior_synthesis

¹⁰<https://github.com/team-vigir/ReSpeC>



(a) The state machine above was synthesized for the task with $\mathcal{I} = \{\text{stand_prep}\}$ and $\mathcal{G} = \{\text{look_down}, \text{pickup_object}\}$. The capability `object_template` requests an object template from the operator (Section II-B). It is a precondition of `pickup_object`.



(b) ATLAS finishing execution of state 8_pickup_object.

Fig. 6: Snapshots from the demo in Section VIII-A.

IX. CONCLUSIONS AND FUTURE WORK

...Once we have a synthesized SM, it can be treated as a primitive action with its outcomes, etc.

Hint at the question: “What does it mean to offer formal guarantees when the activation of primitives can result in failure?”

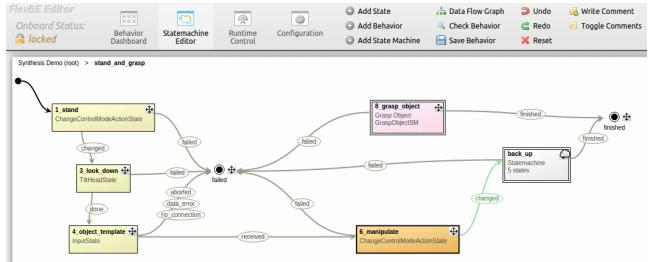
...Future work: Capability specification, integrate more robots, ...

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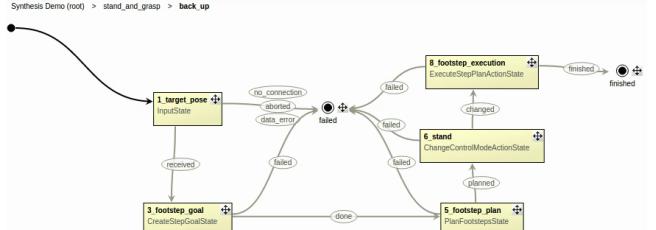
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(a) The operator “locks” the initial state machine at the state 6_manipulate (indicated by the orange color), which is allowed to be executed. Then, a new state machine, `back_up`, is synthesized with `manipulate` as the initial condition. The transition from 6_manipulate is then moved from 8_grasp_object to `back_up`.



(b) The new state machine, `back_up`, was synthesized for the task with $\mathcal{I} = \{\text{manipulate}\}$ and $\mathcal{G} = \{\text{footstep_execution}\}$.

Fig. 7: FlexBE Editor snapshots from the demo in Section VIII-B. In response to some unexpected event, the operator synthesized a state machine that has ATLAS back away (7b).

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