Hierarchical Hybrid Symbolic Robot Motion Planning and Control†
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ABSTRACT

This paper addresses the motion planning problem using hybrid symbolic techniques. The proposed approach develops a unified hierarchical hybrid control framework using a bisimulation-based abstraction technique over the partitioned motion space that can be applied to autonomous aerial robots (3-D symbolic motion planning) or ground vehicles (2-D symbolic motion planning). The bisimulation relation between the abstracted model and the original continuous system guarantees that their behaviors are the same. This allows to design a discrete supervisor for the abstracted model, and then, the designed supervisor can be applied to the original system while the closed-loop behavior does not change. To apply the discrete supervisor to the original continuous system, an interface layer is developed, which on the one hand translates discrete commands of the supervisor to a continuous form applicable to the continuous plant and on the other hand, abstracts the continuous signals of the continuous low layer to discrete symbols understandable by the supervisor. The proposed algorithm is verified through implementation of a hybrid symbolic algorithm for the formation control of unmanned aerial vehicles.

Key Words: Symbolic Control, Hybrid Control, Abstraction, Robot Motion Planning

I. INTRODUCTION

With advances in technologies it is becoming possible to develop fully autonomous vehicles that can accomplish complicated tasks. Such sophisticated capabilities require the control structure of an autonomous vehicle to have various types of sensors to recognize itself and the environment, analyze high amount of sensor readings, process the collected data and accordingly, make decision to compute the control signals. Here, for motion planning of the robots, the question is that how to process and manage excessive amount of data, and then, how to integrate the decision making unit with continuous low level control structure of the system. In fact, a typical robot has an inherent hybrid nature whose event-triggered discrete logic of the decision making unit and the corresponding discrete commands influence the continuous dynamics of the robot. This logic may require the system to satisfy several goals with a particular order, which is beyond the traditional methods for robot motion control based on optimal control techniques. To comprehensively analyze and design a control system for motion planning and control, a proper solution is to utilize the hybrid modeling and control theory and consider the discrete and continuous dynamics of the system, simultaneously and within a unified framework [1].

To develop a hybrid symbolic motion planning and control mechanism, the challenge is to design and develop a computationally effective hybrid approach for the robot motion control so that the closed-loop system satisfies the discrete logic of the decision making unit.
In [2], a complicated search and rescue and in [3], the motion control of a team of robots are addressed using symbolic control methods and abstraction techniques. These schemes reduce the system to a finite state transition system [4], [5], [6] for which one can design a proper discrete supervisor [7] to achieve certain properties expressed in high-level human-like languages such as linear or branching temporal logics [8], [2], [9], [10]. Most efforts in the literature have been devoted to partitioning the motion space to obtain an abstracted model [11, 12, 13, 14, 15, 16]. Here, the key is to realize the relationship between the abstract model and the original continuous system. Understanding the relationship between the abstract model and the original system allows to design a supervisor for the abstract model that is essentially a simpler model, and then, convert it back to the original continuous model. The immediate question is that whether it can be guaranteed that the control objectives, achieved in the abstract domain, can be achieved by the original system as well. To address this problem, this paper aims at developing a bisimulation-based abstraction technique by which, the equivalent behaviors of the abstract model and the original plant allows the designer to synthesize the discrete supervisor for the abstract model and then, apply it to the original plant. The preliminary result of this work was presented in [17]. Compared with [17], here, the bisimulation relation between the abstract model and the original system is rigorously proved, and the results are extended to the polar partitioned spaces. Furthermore, actual implementation results are provided to verify the algorithm.

The main contribution of this paper is to develop a unified hierarchical hybrid framework for symbolic motion planning and control of robots based on a bisimulation-based abstraction technique over a rectangular or polar partitioned space. Starting from the low level continuous dynamics of the system, and using rectangular or polar partitioning techniques, the motion dynamics of the robots can be abstracted to a finite state machine over the partitioned motion space, for which we can design a discrete supervisor to achieve the desired specification. We prove that the bisimulation relation between the abstracted model and the original continuous model holds for a plant with multi-affine dynamics over the rectangular or polar partitioned space. This bisimulation relation guarantees the same behavior of the plant and its abstract model and therefore, the discrete supervisor designed for the abstracted model can be applied to the original continuous plant so that the closed-loop system’s behavior does not change. To implement the idea, a hierarchical hybrid control structure is proposed whose lowest layer is a plant with continuous dynamics, and its top layer is a discrete supervisor which controls the system to satisfy the given specification. To connect the discrete supervisor to the continuous plant, an interface layer is introduced by which the discrete commands of the supervisor can be converted to a continuous form applicable to the plant. Furthermore, when the system trajectory crosses the partitioning curves, the interface layer generates detection events which inform the supervisor about the current state of the system. then, based on observing these detection events, the supervisor can issue new commands. The proposed algorithm is implemented on unmanned aerial helicopters and the flight test results are provided to evaluate the algorithm.

The rest of this paper is organized as follows. After explaining the preliminaries and notations in Section II, the symbolic motion planning and control problem is described in Section III. Then, in Section IV, the partitioning of the motion space will be described. Several controllers will be introduced to drive the system trajectory over the partitioning elements. In Section V, the partitioned system will be bisimilarly abstracted to a finite state machine and the bisimulation relation will be proven. For the resulting finite state machine one can design a discrete supervisor as explained in Section VI. Section VII describes how to implement the whole control structure through an illustrative example. Finally, the paper is concluded in Section VIII.

II. Preliminaries

To address the motion planning problem, we partition the motion space. In the literature, there are several methods that can be used for partitioning the space such as using natural invariants of the plants [18], rectangulation [12] or triangulation [11] of the motion space, or polar and spherical partitioning [16], [15], of the space. To elaborate the idea, without loss of generality, here we will use rectangular and polar partitioning of the motion space which are more convenient to work with.

Consider that the motion space is a $[0, x_N] \times [0, y_N]$ rectangle, which is partitioned by the curves $\{x = x_i | 0 \leq x_i \leq x_N$ such that for $i < j : x_i < x_j, i, j = 1, ..., N_a, x_1 = 0, x_{N_a} = x_N\}$, and $\{y = y_i | 0 \leq y_i \leq y_N$ such that for $i < j : y_i < y_j, i, j = 1, ..., N_b, y_1 = 0, y_{N_b} = y_N\}$. In this partitioned space, the region $R_{i,j} = \{(x, y) | x_i \leq x \leq x_{i+1}, y_j \leq y \leq y_{j+1}\}$ is a
rectangular partitioning element which is surrounded by the curves \( x = x_i, x = x_{i+1}, y = y_j, \) and \( y = y_{j+1} \). The interior of the region \( R_{i,j} \) is denoted by \( R_{i,j} \). Each region has four vertices \( v_m \), with \( m = (a, b) \), where \( m_a \) and \( m_b \) are the binary indices which refer to the partitioning curves that have generated the vertex \( v_m \). Hence, we have \( v_0 = v(00) = [x_i, y_j]^T \), \( v_1 = v(01) = [x_{i+1}, y_j]^T \), \( v_2 = v(10) = [x_i, y_{j+1}]^T \), and \( v_3 = v(11) = [x_{i+1}, y_{j+1}]^T \) as the vertices of the region \( R_{i,j} \) as shown in Fig. 1. The set \( V(*) \) stands for the vertices that belong to *, and \( E(v_m) \) is the set of edges that touch the vertex \( v_m \). Furthermore, the element \( R_{i,j} \) has four edges \( \{E_x^+, E_x^-, E_y^+, E_y^-\} \) and correspondingly, four outer normal vectors \( \{n_x^+, n_x^-, n_y^+, n_y^-\} = \{[1, 0]^T, [-1, 0]^T, [0, 1]^T, [0, -1]^T\} \).

Similarly, for polar partitioning of the motion space, consider the motion space as a circle with the radius of \( R_m \). With the aid of the partitioning curves \( \{r_i = \frac{R_m}{N_a}(i - 1), i = 1, \ldots, N_a\} \) and \( \{\theta_j = \frac{2\pi}{N_b}(j - 1), j = 1, \ldots, N_b\} \), this circle can be partitioned into \((N_a - 1)(N_b - 1)\) partitioning elements. An element \( R_{i,j} = \{p = (r, \theta) | r_i \leq r \leq r_{i+1}, \theta_j \leq \theta \leq \theta_{j+1}\} \) has four vertices, \( v_0, v_1, v_2, v_3 \), four edges, \( E_x^+, E_x^-, E_y^+, E_y^- \), and correspondingly, four outer normal vectors \( n_x^+, n_x^-, n_y^+, n_y^- \) (Fig. 2).

For these partitioned spaces, \( \exists(\ast) = \ast \) relates the label \( \ast \) to the set \( \ast \). This partitioned space can be captured by the equivalence relation \( Q = \{(x_1, x_2) \exists \ast s.t. x_1, x_2 \in \exists(\ast)\} \), where \( \ast \) is one of the above-mentioned partitioning elements. Correspondingly, \( \pi_Q(x) = \ast s.t. x \in \ast \) and \( \exists(\ast) = \ast \), where \( \pi_Q(x) \) is a projection map.

In these partitioned spaces, let’s define \( V_r \) as the set of all vertices of the partitioning elements, \( P \) as the perimeter of the motion space in which the vertices are excluded, and \( W \) as the exterior of the motion space. Also consider the detection element \( d([i, j], [i', j']) = R_{i,j} \cap R_{i',j'} - V_r \), which is defined for two adjacent regions \( R_{i,j} \) and \( R_{i',j'} \) (the order is not important). With this procedure, the whole space will be partitioned into \( V_r \cup R_{i,j} \cup d([i, j], [i', j']) \cup P \cup W \). Correspondingly, consider \( V_r, R_{i,j}, d([i, j], [i', j']), P, \) and \( W \) as the labels for these partitioning elements.

### III. Problem Description

Consider a robot with the dynamics \( \dot{X}(t) = f(X(t), u(t)) \) where \( X \) is the robot position and \( u \) is the control input. For the motion control of this robot, the motion space can be partitioned into several disjoint regions which are separated by hypersurfaces. Our objective here is to construct a hybrid controller to drive the robot through the partitioned space to satisfy a given specification. Let \( R_1, R_2, \ldots, R_n \) be the elements of the partitioned space, and correspondingly \( \bar{R}_1, \bar{R}_2, \ldots, \bar{R}_n \) as the finite set of symbols that label these elements, where \( \exists(R_i) = R_i \). The motion planning objective may require the robot to visit particular regions with a specific order while avoiding some other regions which can be specified either by a sequence of events in the form of an automaton or by a LTL formula [8]. A LTL formula over the set of propositions \( P = \{R_1, R_2, \ldots, R_n\} \) can be constructed using the combination of traditional logical operators including negation (\( \neg \)), disjunction (\( \lor \)), conjunction (\( \land \)), and the temporal operators including next (\( O \)), until (\( U \)).
Lemma 1 (Adopted from [16] and [21]) Given a multi-affine function \( g(X) \) defined over a partitioned element \( R_{i,j} \), the function \( g \) can be uniquely described based on the values of \( g \) at vertices of \( R_{i,j} \) as follows:

\[
\forall X = (a, b) \in \tilde{R}_{i,j} : g(X) = \sum_{m=0}^{3} \lambda_m g(v_m),
\]

where \( v_m, m = 0, \ldots, 3, \) are the vertices of the element \( R_{i,j} \). The coefficients \( \lambda_m \), can be obtained uniquely as follows:

\[
\lambda_m = \lambda_a^{m_a} (1 - \lambda_a)^{1-m_a} \lambda_b^{m_b} (1 - \lambda_b)^{1-m_b},
\]

where \( m_a, m_b \), are the corresponding binary digits of the index \( m \).

For rectangular partitioning, \( \lambda_a \) and \( \lambda_b \) can be found as follows:

\[
\lambda_a = \frac{a - x_i}{x_i+1 - x_i}, \quad \lambda_b = \frac{b - y_j}{y_j+1 - y_j}
\]

and for polar partitioning,

\[
\lambda_a = \frac{a - r_i}{r_i+1 - r_i}, \quad \lambda_b = \frac{b - \theta_j}{\theta_j+1 - \theta_j}
\]

In this theorem, it can be verified that \( \lambda_m \geq 0 \), and \( \sum_m \lambda_m = 1 \). Also, since the above theorem holds true for all points in \( \tilde{R}_{i,j} \), the theorem can be also applied to the points on the edges.

Now, using these properties, for a system with multi-affine dynamics it is possible to construct multi-affine controllers to either keep the system’s trajectory inside the region (invariant region) or to push it out from the desired edge (exit edge) as it is described in the following two lemmas.

Lemma 2 [16] (Constructing an invariant region) For a continuous multi-affine vector field \( \dot{X} = h(X, u(X)) = g(X) \), the region \( R_{i,j} \) is an invariant region if there exists a controller \( u \), such that for each vertex \( v_m, m = 0, 1, 2, 3, \) with incident edges \( E_{q}^{\pm} \in E(v_m) \), and corresponding outer normals \( n_q^s \) we have \( U_m(Inv) = \{ u| n_{q}^{s^T} . g(v_m) < 0, \text{ for all } E_{q}^{\pm} \in E(v_m) \} \neq \emptyset \).

Lemma 3 [12] (Constructing an exit edge) For a continuous multi-affine vector field \( \dot{X} = h(X, u(X)) = g(X) \), the edge \( E_q^s \) with the outer normal \( n_q^s \) is an exit edge if there exists a controller \( u \), such that for each vertex \( v_m, m = 0, 1, 2, 3, \) we have \( U_m(Ex(F_q^s)) = \{ u \in \mathbb{R}^2 | n_{q}^{s^T} . g(v_m) > 0, \text{ for all } v_m \text{ and } n_{q}^{s^T} . g(v_m) < 0, \text{ for all } E_{q}^{\pm} \neq E_q^s, v_m \in V(E_{q}^{\pm}) \} \neq \emptyset \).
Next proposition shows that if we construct a controller based on Lemma 3, all of the points on an exit edge are reachable.

**Proposition 1** For a continuous multi-affine vector field \( X = h(X, u(X)) = g(X) \), in a region \( R_{i,j} \) with the exit edge \( E_q \) constructed by Lemma 3, all \( y \in E_q \setminus E \) are reachable from a point inside the region \( R_{i,j} \).

**Proof:** Respecting the condition of Lemma 3 for the points on the exit edge \( E_q \), we will have \( n_q(y) = 0 \), \( \forall y \in E_q \). This strictly positive inequality guarantees that the trajectories that leave the region do not return back any more. In addition, it shows that the points on the exit edge are not reachable from other points on the edge. Therefore, \( y \in E_q \) is not reachable form an adjacent region or from another point on \( E_q \). Then, considering \( n_q(y) = 0 \), by continuity of \( g \), it can be concluded that there is a point inside the region \( R_{i,j} \) on the neighborhood of \( y \) from which \( y \) is reachable.

With these controllers defined over the partitioned space, it is possible to drive the system’s trajectory to one of the adjacent regions or to keep it inside the current region. This system can be captured by a transition system \( T_Q = (X_Q, X_{Q_0}, U_Q, \rightarrow_Q, Y_Q, H_Q) \), where

- \( X_Q = V_r \cup R_{i,j} \cup d((i, j), [i', j']) \cup P \cup W \) is the set of system’s states, where \( 1 \leq i, i' \leq N_a - 1, 1 \leq j, j' \leq N_b - 1 \).
- \( X_{Q_0} \subseteq R_{i,j} \) is the set of initial states. Here, we assume that the system initially starts from the inside of the regions \( R_{i,j} \).
- \( U_Q = U_r \cup U_d \), where
  - \( U_r = U_{ex} \cup \{C_0\} \) is the set of labels where the label \( C_0 \) corresponds to the the controller that makes the region \( R_{i,j} \) an invariant region. For rectangular partitioning, the set \( U_{ex} \) consists of the labels \( C_{r}^+, C_{r}^-, C_{t}^+, C_{t}^- \) correspond to the controllers that make the edges \( E_r^+, E_r^-, E_t^+, E_t^- \) exit edges, respectively. For the polar partitioning, the set \( U_{ex} \) includes the labels \( C_{r}^+, C_{r}^-, C_{t}^+, C_{t}^- \) correspond to the controllers that make the edges \( E_r^+, E_r^-, E_t^+, E_t^- \) exit edges, respectively.
  - For these control labels, the sets of control actions that can be activated in this region are: \( r(C_r^+) = \{u(X)|u(X) = \sum_m \lambda_m u(v_m) \ m = 0, 1, 2, 3 v_m \in V(R_{i,j}) \} \) and \( r(C_r^-) = \{u(X)|u(X) = \sum_m \lambda_m u(v_m) \ m = 0, 1, 2, 3 v_m \in V(R_{i,j}) \} \) and \( r(C_t^+) = \{u(X)|u(X) = \sum_m \lambda_m u(v_m) \ m = 0, 1, 2, 3 v_m \in V(R_{i,j}) \} \) and \( r(C_t^-) = \{u(X)|u(X) = \sum_m \lambda_m u(v_m) \ m = 0, 1, 2, 3 v_m \in V(R_{i,j}) \} \), where \( \lambda_m \) can be obtained by (2).
important transitions have been captured, this transition continues signals. The block diagram of this control actuator, will translate these discrete commands to the control commands to the plant are continuous. specification. This command has a discrete nature and should be injected to the plant to satisfy the desired symbols, the supervisor decides which control signal generated, which inform the current situation of the bols, upon crossing partitioning hypersurfaces, plant sym-

(b) \( v \in \{d^-(i, j), [i', j']\} \subseteq U_d \); \( \pi_Q(X_1) \neq \pi_Q(X_2); \exists R_{i,j}, R_{i', j'}, d([i, j], [i', j']) \) \( i' \leq i \), and \( j' \leq j \) such that \( \pi_Q(X_1) = d([i, j], [i', j']) \) and \( \pi_Q(x_2) = R_{i', j'} \); \( \exists 0 < \varepsilon < \tau \) and \( \exists w \in \{C_x, C_y\} \) or \( w \in \{C^-, C^+_y\} \) such that \( \psi(t): [0, \tau) \rightarrow \mathbb{R}^2 \) is the solution of \( X = h(X, r(w)), \psi(\varepsilon) = X; \psi(\tau) = X_2, \pi_Q(\psi(t)) = R_{i,j}, \) for \( t \in (0, \varepsilon) \), and \( \pi_Q(\psi(t)) = W \) for \( t \in (\varepsilon, \tau) \).

- Crossing the motion space’s boundary: \( v = P; \) \( \pi_Q(X_1) = P \) and \( \pi_Q(X_2) = W \); \( \exists R_{i,j} \) and \( \exists 0 < \varepsilon < \tau \) and \( \exists w \in U_e \) such that \( \psi(t): [0, \tau) \rightarrow \mathbb{R}^2 \) is the solution of \( X = h(X, r(w)), \psi(\varepsilon) = X; \psi(\tau) = X_2, \pi_Q(\psi(t)) = R_{i,j}, \) for \( t \in (0, \varepsilon) \), and \( \pi_Q(\psi(t)) = W \) for \( t \in (\varepsilon, \tau) \).

- \( Y_Q = X_Q \) is the output space.
- \( H_Q: X \rightarrow Y_Q \) is the output map. Here, we have chosen \( H_Q(X) = \pi_Q(X) \).

Remark 1 The transition system \( T_Q \) captures only important transitions form one region to another region under the exit commands, or from one region to itself under the invariant controller, \( C_Q \).

Analogous with [18], to model this partitioned system, we can define an interface layer which connects this partitioned system to a higher discrete supervision layer. The interface layer has two main blocks: Detector and Actuator. The detector converts continuous time signals to a sequence of symbols. Upon crossing partitioning hypersurfaces, plant symbols, \( d^+(i, j), [i', j'], d^-(i, j), [i', j'], \) and \( P \), will be generated, which inform the current situation of the plant to the supervisor. Based on the observed plant symbols, the supervisor decides which control signal should be injected to the plant to satisfy the desired specification. This command has a discrete nature and the control commands to the plant are continuous. The actuator, will translate these discrete commands to continuous signals. The block diagram of this control structure is shown in Fig. 3.

V. Abstraction over the partitioned space

In the partitioned system \( T_Q \), although all important transitions have been captured, this transition system still has infinite number of states which makes the control synthesis problem very difficult or even impossible. Abstraction [22] is a technique that reduces the number of states by aggregating similar states. Hence, using this strategy, and considering each partitioning element as one of the states in the abstracted model, the resulting model will be:

\[ T_\xi = (X_\xi, X_{\xi_0}, U_\xi, R_\xi, Y_\xi, H_\xi) \]

- \( X_\xi = \{\hat{R}_{i,j}|1 \leq i \leq N_a - 1, 1 \leq j \leq N_b - 1\} \cup \{d([i, j], [i', j'])|1 \leq i, i' \leq N_a - 1, 1 \leq j, j' \leq N_b - 1\} \cup \{P, W\} \). Note that since the system starts from a point inside the regions \( R_{i,j} \) and due to strictly negative inequalities in Lemmas 2 and 3, the system trajectory never crosses the vertices, and hence, the set \( V_\xi \) does not need to be considered in the abstracted system.
- \( X_{\xi_0} \subseteq \{\hat{R}_{i,j}|1 \leq i, i' \leq N_a - 1, 1 \leq j, j' \leq N_b - 1\} \).
- \( U_\xi = U_a \cup U_d \) is like what we have in \( T_Q \).
- \( (r_1, r_2, v) \in \rightarrow_\xi \), denoted by \( r_1 \rightarrow_\xi r_2 \), if \( \exists v \in U_\xi \) such that \( X_1 \rightarrow_Q X_2 \).
- \( Y_\xi = X_\xi \).
- \( H_\xi(r) = r \) is the output map.

With this method, the partitioned system, \( T_Q \), which previously was modelled by the regulation layer and the interface layer, now is abstracted to a finite state transition system \( T_\xi \), for which we can design a discrete supervisor [7] to achieve the desired specification. Then, with the aid of the interface layer, the designed supervisor for the abstract model can be applied to the original continuous model. To guarantee that the discrete supervisor for the abstract model can also work for the original continuous model, it is necessary that the abstract model and the original continuous model represent the same behavior which requires them.
to be bisimilar. A bisimulation relation between two transition systems can be formally defined as follows:

**Definition 1** [22] Given $T_i=\langle Q_i, Q_i^0, U_i, \rightarrow_i, Y_i, H_i \rangle$, $(i=1,2)$, $R$ is a bisimulation relation between $T_1$ and $T_2$, denoted by $T_1 \approx_R T_2$, iff:

1. $\forall q_1 \in Q_1^0$ then $\exists q_2 \in Q_2^0$ so that $(q_1, q_2) \in R$. Also, $\forall q_2 \in Q_2^0$ then $\exists q_1 \in Q_1^0$ so that $(q_1, q_2) \in R$.
2. $\forall q_1 \rightarrow q_1'$, and $(q_1, q_2) \in R$ then $\exists q_2' \in Q_2$ such that $q_2 \rightarrow q_2'$ and $(q_1', q_2') \in R$, Also, $\forall q_2 \rightarrow q_2'$, and $(q_1, q_2) \in R$ then $\exists q_1' \in Q_1$ such that $q_1 \rightarrow q_1'$.

For multi-affine functions defined over a rectangular or polar partitioned space, and with the controllers which we defined to construct exit edges or to make a region invariant, the abstract model and the original partitioned system are bisimilar as proven in Theorem 1.

**Theorem 1** The original partitioned system, $T_Q$, and the abstract model, $T_{\xi}$, are bisimilar.

**Proof:**

consider the relation $R = \{(q_0, q_0) | q_0 \in X_Q, q_0 \in X_{\xi}, \text{ and } q_0 \in \mathcal{S}(q_0)\}$. We will show that this relation is a bisimulation relation between $T_Q$ and $T_{\xi}$. To prove this bisimulation relation we should verify both conditions of Definition 1.

To verify the first condition of the bisimulation relation in Definition 1, we know that for any $q_0 \in X_Q$, there exists a region $R_{i,j}$ such that $q_0 \in R_{i,j}$. For this region, there exists a label, $\tilde{R}_{i,j}$ such that $R_{i,j} = \mathcal{S}(\tilde{R}_{i,j})$ and $\tilde{R}_{i,j} \in X_{\xi}$. Hence, $(q_0, R_{i,j}) \in R$. Conversely, it can be similarly shown that for any $q_0 \in X_{\xi}$, there exists a region $q_0 \in X_Q$ such that $(q_0, q_0) \in R$.

To verify the second condition of the bisimulation relation, following from the definition of $T_{\xi}$, we know that for any $(q_0, q_0) \in R$ and $q_0 \rightarrow q_0' Q q_0'$, there exists a transition $q_0 \rightarrow_{\xi} q_0'$, where $q_0' \in \mathcal{S}(q_0')$ or equivalently $(q_0', q_0') \in R$. For the converse case, assume that $q_0 \rightarrow_{\xi} q_0'$. According to the definition of $R$, all $x \in \mathcal{S}(q_0')$ are related to $q_0$. Hence, to prove the second condition of the bisimulation relation, we should investigate it for all $x \in \mathcal{S}(q_0)$. Based on the control construction procedure, the labels $u$, $q_0$, and $q_0'$ can be the one of the following cases:

1. $u = C_0$ and $q_0 = q_0'$. In this case, since the controller $C_0$ makes the region an invariant region (Proposition 2), all of the trajectories starting from any $q_0 \in \mathcal{S}(q_0)$ will remain inside the region $\mathcal{S}(q_0)$. Therefore, for any $q_0 \in \mathcal{S}(q_0)$, there exists a $q_0' \in \mathcal{S}(q_0)$ such that $q_0 \rightarrow_{\xi} q_0'$ and $q_0 = q_0'$.

2. $u \in U_{ex}$, $q_0 \in \{R_{i,j}\} \subseteq \{R_{i,j}\}$, and $q_0' \in \{\tilde{d}([i,j],[i',j'])\} \subseteq \{U_{ex}\}$, where $q_0' \in \mathcal{S}(q_0')$, $q_0' \in \mathcal{S}(q_0')$, and $u$ is the (finite) set of events. Therefore, for any $q_0 \in \mathcal{S}(q_0)$, there exists a $q_0' \in \mathcal{S}(q_0')$ such that $q_0 \rightarrow_{\xi} q_0' q_0' q_0$.

3. $u \in \{d^+([i,j],[i',j'])\} \subseteq U_d$, $q_0 \in \{R_{i,j}\}$, and $q_0' \in \{\tilde{d}([i,j],[i',j'])\} \subseteq \{U_{ex}\}$, where $q_0' \in \mathcal{S}(q_0')$, $q_0' \in \mathcal{S}(q_0')$, and $u$ is the (finite) set of events.

In all of the above mentioned cases, the second condition of the bisimulation relation for the converse case holds true. Since both conditions of the bisimulation relation hold, $T_{\xi}$ and $T_Q$ are bisimilar.

**VI. Adopting the DES supervisory control to the abstracted model**

For the abstracted model with finite number of states we can design a discrete supervisor using Discrete Event Systems (DES) supervisory control theory initiated by Ramadge and Wonham [7]. Formally, the finite state machine model of the abstracted system can be represented by an automaton $G = (Q, \Sigma, \alpha, Q_0, Q_m)$, where $Q = Q_0$ is the set of states; $Q_0 = Q_0 \subseteq Q$ is the set of initial states; $\Sigma = U_{ex} \cup U_d$ is the (finite) set of events; $Q_m \subseteq Q$ is the set of final
(marked) states, and \( \alpha : Q \times \Sigma \to Q \) is the transition function which is a partial function and determines the possible transitions in the system caused by different events. Based on the transitions in \( T_\xi \), the function \( \alpha \) can be defined as follows:

\[
\alpha(R_{i,j},\sigma) = \begin{cases} 
R_{i,j} & \text{if } \sigma = C_0 \\
 d([i,j],[i+1,j]) & \text{if } \sigma = C_x^+, C_{x}^- \text{ and } i \neq N_a - 1 \\
 d([i,j],[i-1,j]) & \text{if } \sigma = C_x^-, C_{x}^+ \text{ and } i \neq 1 \\
 d([i,j],[j+1]) & \text{if } \sigma = C_y^+, C_{y}^- \text{ and } j \neq N_b - 1 \\
 d([i,j],[j-1]) & \text{if } \sigma = C_y^-, C_{y}^+ \text{ and } j \neq 1 \\
 d([i,j],[N_a-1,j]) & \text{if } \sigma = C_y^- \text{ and } j = 1 \\
 P & \text{if } \sigma = C_x^+ \text{ and } j = N_b - 1 \\
 i = 1; \sigma = C_x^+, j = N_b - 1, & \text{or} \\
 \sigma = C_x^-, j = 1 \\
 \alpha(d([i,j],[i',j'])),\sigma) = R_{i',j'} & \text{if } \sigma = d^+(([i,j],[i',j'])), i' \geq i, j' \geq j, \\
 \text{or } \sigma = d^-(([i,j],[i',j'])), i' \leq i, j' \leq j \\
 \alpha(\hat{P},\hat{P}) = \hat{W} 
\end{cases}
\]

In this automaton, the sequence of events generates a string, \( \varepsilon \) is an empty string, and \( \Sigma^* \) is the set of all possible strings over the set \( \Sigma \) including \( \varepsilon \). The function \( \alpha \) can be extended from acting on events to acting on the strings as \( \alpha_{ext} : Q \times \Sigma^* \to Q \) in which \( \alpha_{ext}(q,\varepsilon) = q \) and \( \alpha_{ext}(q,s\sigma) = \alpha(\alpha_{ext}(q,s),\sigma) \forall s \in \Sigma^* \) and \( \sigma \in \Sigma \). The language of the automaton is a sequence of strings that can be generated by \( G \) and can be defined as \( L(G) = \{ s \in \Sigma^* \mid \exists q_0 \in Q_0 \text{ s.t. } \alpha_{ext}(q_0,s) \text{ is defined} \} \). The marked language, denoted by \( L_m(G) \), consists of the strings that can be generated by the automaton \( G \) and end with the marked states, which formally can be defined as \( L_m(G) = \{ s \in \Sigma^* \mid \exists q_0 \in Q_0 \text{ s.t. } \alpha_{ext}(q_0,s) \text{ is defined and } \alpha_{ext}(q_0,s) \in Q_m \} \). The event set \( \Sigma \) consists of two types of events: the controllable event set \( \Sigma_c = U_a \) and the uncontrollable event set \( \Sigma uc = U_d \). The controllable events are those that can be disabled or enabled by an external supervisor; however, the uncontrollable events cannot be affected by the supervisor. Playing with the controllable events, the supervisor can modify the plant’s generable language so that \( \varepsilon \in L(S(G)) \) and \( \{ s \in L(S(G)) \land (s\sigma \in L(G)) \land (s \in L(S)) \} \leftrightarrow [s\sigma \in L(S(G))] \). Accordingly, the closed-loop marked language will be \( L_m(S(G)) = L(S(G)) \cup L_m(G) \). This supervisor can be used to achieve a controllable language specification. A language specification \( K \) is said to be controllable with respect to the language of the plant \( G \) and set of uncontrollable events \( E_{uc} \) if \( \forall s \in K, e \in E_{uc}, se \in L(G) \Rightarrow se \in K \). To realize this control strategy and to combine the plant discrete model and the supervisor, we can use parallel composition [23] which is a binary operation between two automata. Next theorem shows how the parallel composition can be used to modify the plant language to achieve a desirable specification given in terms of a controllable language.

**Theorem 2** [23] Let \( G \) be the plant and \( K \subseteq \Sigma^* \) be a desired language. If \( \emptyset \neq K = K \subseteq L(G) \) and \( K \) is controllable, there exist a nonblocking supervisor \( S \) such that \( L(S(G)) = L(S||G) = K \). In this case, \( S \) could be any automaton that satisfies \( L_m(S) = L(S) = K \).

**VII. Illustrative example for symbolic motion planning and control for the formation control of unmanned helicopters**

Formation control is the jointly movement of a group of agents with a relatively fixed distances, and has been addressed by different methods (see e.g. [24], [25], [26], [27], [28]). Here, we use our proposed method of symbolic motion planning for formation control of a team of unmanned aerial vehicles. The team consists of two UAV helicopters, HeLion and SheLion which are developed by our research group at the National University of Singapore. The modelling and low level control structure of the NUS UAV helicopters are explained in [29], [30], [31]. For the regulation layer of these helicopters we have proposed a two-layer control structure in which the inner-loop controller stabilizes the system using \( H_\infty \) control design techniques, and their outer-loop is used to derive the system towards the desired location. As it has been discussed in [30], in this control structure, the inner-loop is fast enough to track the given references, so that the outer-loop dynamics can be approximately described as follows:

\[
\dot{x} = u, \quad x \in \mathbb{R}^2, \quad u \in U \subseteq \mathbb{R}^2,
\]

where \( x \) is the position of the UAV; \( u \) is the UAV velocity reference generated by the formation algorithm, and \( U \) is the velocity constraint set, which is a convex set.

Now, in a leader follower formation scenario, consider the follower velocity in the following form:

\[
V_{follower} = V_{leader} + V_{rel}.
\]

In this relative framework, consider a circle with the radius of \( R_m \) that is centered at the desired position...
of the follower, and is partitioned as discussed in Section II. A part of the discrete abstracted model of the follower motion dynamics is shown in Fig. 4.

For these helicopters, our aim is to design the formation controller to generate the relative velocity of the follower, $V_{rel}$, such that starting from any initial point inside the control horizon, it eventually reaches the desired relative distance with respect to the leader. Moreover, after reaching the formation, the follower UAV should remain at the desired position. Using the proposed polar partitioning approach, the formation can be achieved if the controller drives the system directly to the regions $R_{1,j}$, $1 \leq j \leq n_\theta - 1$. After reaching the formation by this scenario, the controller should maintain the state of the system at the final state to keep the formation by activating the command $C_0$. This specification can be achieved by the supervisor, $S_F$, as shown in Fig. 5.

To verify the algorithm and to monitor how the follower can reach the formation and maintain the achieved formation, a flight test has been conducted in which the leader tracks a line path, and the follower should reach and keep the formation. In this flight test, the control horizon $R_m$ is 50 meter, $N_a = 10$, and $N_b = 20$. The follower is initially located at a point which has a relative distance of $(dx, dy) = (-17.8 m, 11.4 m)$ with respect to the desired position and the distance between the desired position and the leader is $(dx, dy) = (-5 m, -15 m)$.

The position of the UAVs in x-y plane is shown in Fig. 6. The relative distance of the follower UAV from the desired position is shown in Fig. 7. As it can be seen the follower UAV has finally reached the first circle and then, it has been able to maintain the formation.

Fig. 4. An abstract model for the UAV motion dynamics over a partitioned space.

Fig. 5. Diagram of the polar partitioning approach.

Fig. 6. The position of the UAVs in the x-y plane.
VIII. Conclusion

In this paper, a hybrid framework was proposed for the symbolic motion planning and control of robots. The approach was based on rectangular and polar partitioning of the motion space and then, abstracting the original continuous system with infinite number of states to a finite state machine. To implement the idea, a multi-layer control structure was proposed in which the discrete supervisor was connected to the plant via an interface layer. The continuous plant and the interface layer together were shown to be bisimilar with the abstract model. This bisimilarity let us apply the discrete supervisor which was designed for the abstract model to the continuous plant while the closed-loop behavior does not change. The algorithm was successfully applied to the formation control of unmanned helicopters.

REFERENCES


