AN ALGEBRAIC MODEL OF COMPUTATION FOR SYSTEMS WITH DYNAMIC STRUCTURE

Eloi Pereira\(^1\) and Raja Sengupta

Systems Engineering, CEE Dept., UC Berkeley - Berkeley, CA, USA

ABSTRACT

We introduce the Structure Model - a formalism for modelling and controlling networked systems with dynamic structure. The pervasiveness of networked computing devices is raising attention on models of computation which entail the network structure as a first-class concept. Modelling how the network evolves and how entities interact is now paramount for a correct understanding of the overall system and for the design and synthesis of effective controllers. We address this problem by introducing a formalism for modelling and controlling systems with dynamic structures. The Structure Model (SM) consists of a set of entities, and a set of unary relations and a binary relation over the set of entities. The unary relations are used to model state properties of the entities while the binary relation is used to model interaction between entities such as communication. The SM can be manipulated using a set of algebraic operators. The dynamics of the structure are modelled using a transition system. The SM is designed to compose with an underlying model that describes the entities behaviour. We present how the structure can be abstracted from an underlying model and how the changes at the structure level can be propagated back to the system. This provides means controlling the system from a structure perspective.

The structure model aims at being agnostic to the underlying model of computation. In this work we compose the structure model with Nancy Lynch’s Synchronous Network Model and present a case study using a version of the “agree and pursue” communication and control law.

KEYWORDS

Models of Computation, Dynamic Networks, Control.

1. INTRODUCTION

With the advent of the internet and mobile devices computation has been becoming ubiquitous. Consider a smartphone that provides the user with information about the closest open restaurant, or a team of aerial robots that collaborate while searching for a target. In both examples, computing devices interact with their environment and with other computing devices. Thus, interaction has becoming a major aspect in system’s design. Indeed, some researchers have been claiming that we are facing a paradigm shift on computation: from calculation to interaction [GSW06, Mil06].

A major aspect of interaction is communication. It is paramount to understand how entities share information, what are the constraints introduced by communication infrastructures, and how the topology of such infrastructures varies over time. In this paper we introduce an algebraic model of computation for modelling and controlling discrete-time structures. Fig. 1 depicts our approach.

We assume that the behavior of the system is modelled by a given concurrent Model of Computation (MoC). The structure, denoted by \( \mathcal{S} \), is extracted by an abstraction map \( \alpha \) from the underlying MoC. A structure controller takes \( \mathcal{S} \) and produces a set of controllable events \( \mathcal{E}_c \) that steers the system in order to meet a given structure specification. The system can operate in adversarial environment. This is modelled by introducing a set of uncontrollable events \( \Sigma_u \) that can change \( \mathcal{S} \) arbitrarily. Changes at the structure level are then mapped back to the underlying MoC by a control map \( \beta \).

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\(^1\) Author funded by the Fundação para a Ciência e Tecnologia, SFRH/BD/43596/2008, and by the Portuguese MoD PITVANT project.
The literature is rich on concurrent models of computation, e.g. Kahn Process Network (KPN) [KM77], Milner’s Calculus of Communicating Systems (CCS) [Mil83], Communicating Sequential Processes (CSP) by Hoare [Hoa78], and the actor model by Hewitt and Agha [Agh86]. While the communication topology on KPN, CCS and CSP does not vary over time, the actor model can model dynamic topologies by changing the set of addresses that each actor knows at each instance of time. Influenced by the idea of dynamic structures Milner, Parrow and Walker introduced π-calculus [MPW89,Mil99]. π-calculus allows channel names to be communicated through other channels. This provides means for describing concurrent computations whose structure may change over time. Later, Milner introduced the Bigraph model which also entails changing of structure as a first-class concept using two graph structures that describes location and connectivity of computation.

The main contribution of the work presented in this paper is the introduction of the Structure Model - a formalism for extracting structural information of a system represented by a given model of computation, and for controlling the system from a structural perspective.

We aim at making SM applicable to a wide range of different MoCs that describes underlying behaviour of the system. In this paper we use as a case study the Synchronous Network Model (SNM) by Nancy Lynch [Lyn96]. We choose the SNM because it is a simple model but yet able to model a wide range of interesting concurrent systems. The remainder of the paper goes as follows. In Section 2 we introduce the formalism for modelling discrete-time structures. Section 3 goes on to explain how a structure can be controlled in a centralized way. In Section 4 we research how to embed state of entities into an SM. In Section 5 we introduce a procedure to abstract the structure from a Synchronous Network Model (SNM) and how to control this SNM from the structure level. Section 6 introduces a case study based on the “agree-and-pursue control and communication law”. In Section 7 we state the major conclusions and future work.

2. FORMALISM FOR MODELLING DYNAMIC STRUCTURES

In this section we start by formalizing the SM and then introduce a transition system to model the discrete-time behaviour. Our formalism is based on relational and algebraic structures², motivated by the work of Courcelle [Cou91] and Staruch [Sta05]. The former uses a logical approach to formalize graphs, while the later models multi-agent systems using partial algebraic structures.

**Definition 1** (Structure of SM). An SM is a relational structure of the form \( S = (\text{Ent}, \sigma, I_\sigma) \) where the domain Ent represents a set of entities of the system, \( \sigma \) is the signature of the structure and \( I_\sigma \) is an interpretation function which interprets symbols of the signature on the domain of entities. We denote the set of all possible entities as \( \mathcal{E} \).

**Definition 2** (Signature of SM). The signature of \( S \) is a tuple \( \sigma = (\tau_1, \tau_2, \ldots, \tau_n, \text{ed}) \) where \( \tau_i \) are unary relation symbols. Each \( \tau_i \) has an interpretation \( I_\sigma(\tau_i) \subseteq \text{Ent} \) where \( e \in I_\sigma(\tau_i) \) iff \( e \) has the property \( \tau_i \). ed is a binary relation symbol with an interpretation \( I_\sigma(\text{ed}) \subseteq \text{Ent} \times \text{Ent} \) such that \( (e_1, e_2) \in I_\sigma(\text{ed}) \) iff entities \( e_1 \) and \( e_2 \) are related up to some criterion (e.g. communication relation).

An SM can be seen as a coloured graph where the set of vertices is the domain of \( \tau_i \), the colors are defined by the unary relations \( \tau_i \), and the edges are defined by the binary relation \( \text{ed} \). Note that the signature could be easily enriched with more relation symbols that could represent different structure relations: communication topology, synchronization relations, hierarchical dependencies, etc. For the sake of a clear presentation we choose to keep only one binary relation in the models presented in this paper.

We proceed by formalizing how an SM can evolve over time. Let \( S_\delta \) be the set of SMs obtained with all possible interpretations \( I_\sigma \) and domains \( \text{Ent} \subseteq \mathcal{E} \). Let \( \Sigma \) be a set of events that may lead the current structure to change another structure.

**Definition 3** (Structure Transition System). A Structure Transition System (STS) is a many-sorted algebra denoted as \( T = (S_\Sigma, \Sigma, \delta, S_0) \) where \( \delta : S_\Sigma \times \Sigma \rightarrow S_\delta \) is a 2-arity function called structure transition function, and \( S_0 : \{ \} \rightarrow S_\delta \) is a 0-arity function (constant) called initial structure.

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² Note that the term `"structure"` here refers to the mathematical logic definition: a set of functions and relation symbols interpreted in a given domain.
An STS formalizes how an SM evolves from the initial structure \( S_0 \) to another structure in the presence of an event. An execution trace of an STS is represented as a sequence \( S_0 \xrightarrow{e_0} S_1 \xrightarrow{e_1} S_2 \xrightarrow{e_2} \ldots \) where \( e_i \in \Sigma \).

### 3. CONTROL DESIGN

In this section we formalize how a STSs can be controlled. Our approach is inspired by the work of Ramadge and Wonham on supervisory control of Discrete Event Systems [RW87, RW89].

Let \( \Sigma = \Sigma_c \cup \Sigma_u \) where \( \Sigma_c \) is the set of controllable events while \( \Sigma_u \) is the set of uncontrollable events. The events \( \Sigma_c \) can be disabled by a controller at any time while the set \( \Sigma_u \) denotes the set of events which the controller has no effect.

A control input for \( T \), denoted by \( \gamma \), is a subset of \( \Sigma \) such that \( \Sigma_u \subseteq \gamma \), i.e., all uncontrollable events are included in \( \gamma \). The set of control inputs, denoted by \( \Gamma \), is a subset of \( \mathcal{P}(\Sigma) \).

Assume that an SM at a given instance of time entails all the information needed for the controller to decide the next control input \( \gamma \). Thus we can define a supervisory controller for the structure of the form \( \text{ctr}: L \rightarrow \Gamma \) where \( w \in L \) is a sequence of \( S_0, S_1, S_2, \ldots \) with \( S_i \in S \). \( w \) represents the sequence of SMs from the initial step until the current step. We assume an interleaving semantics for the interactions between the plant and the controller, i.e., the controller performs an action between two plant executions.

On Discrete Event Systems, the domain of a model is a set of primitive states, i.e., there aren’t any signature on a given state. The domain of an STS is a set of SMs which each element encapsulates information such as properties of entities (unary relations \( \tau_i \)) or relations between entities (binary relation \( ed \)).

We introduce an algebra that aims at making the manipulation of SMs explicit. This algebra is used for the specification of the structure transition function of STSs. This approach is inspired by Algebraic Graph Theory [BW04]. In this section we introduce three different operators. At this stage of the work it is not clear yet if these operators are sufficient for modelling the desired structure-level controllers.

**Definition 4** (Algebra on SMs). The algebra over the domain \( S \) is an algebraic structure denoted as \( \mathcal{A} = (S, u_S, \setminus_S, +_S) \) where \( u_S: S \times S \rightarrow S \) is an 2-arity function called **union**, \( \setminus_S: S \times S \rightarrow S \) is a 2-arity function called **minus**, and \( +_S: S \times S \rightarrow S \) is a 2-arity function called **join**.

**Example 1.** Fig. 2 presents two structures \( S_1 \) and \( S_2 \) and the union of both structures. \( \text{Ent}_1 = \{0, 1, 2\}, \text{Ent}_2 = \{3, 4\} \), and \( \text{Ent}_{1, u,2} = \{0, 1, 2, 3, 4\} \). The signature includes one unary relation \( \text{BLACK} \) where \( \text{BLACK} = \{0\} \) for \( S_1 \), \( \text{BLACK} = \{3\} \) for \( S_2 \), and \( \text{BLACK} = \{0, 3\} \) for \( S_1 \cup S_2 \).

The minus function is defined as \( S_1 \setminus S_2 = (\text{Ent}_1 \setminus \text{Ent}_2, \sigma, I_0) \) i.e. the set of entities is the set of entities of \( S_1 \) except the entities that are also in \( S_2 \). The interpretation of \( \sigma \) is defined as follows:

- for each \( \tau_i \in \sigma \), \( L_0(\tau_i) = L_0(\tau_i) \setminus \text{Ent}_2 \)
- for each entity, \( L(\tau) = L(\tau) \setminus \text{Ent}_2 \)

Let \( \text{connect}: \mathcal{P}(\mathcal{E}) \times \mathcal{P}(\mathcal{E}) \rightarrow \mathcal{P}(\mathcal{E}) \) be a function that, given a set of entities \( \text{Ent} \) and a set of edges \( ed \), returns the subset of edges that contains any of the entities in \( \text{Ent} \). The set of edges of the \( S_1 \setminus S_2 \) is then \( I_0(ed) = I_0(ed) \setminus \text{connect}(\text{Ent}_1, I_0(ed)) \)

The minus function can be interpreted as an operator that takes a structure and removes a part from it.

**Example 2.** Fig. 2(b) presents two structures \( S_1 \) and \( S_2 \) and \( S_1 \setminus S_2 \). \( \text{Ent}_1 = \{0, 1, 2\}, \text{Ent}_2 = \{1\} \), and \( \text{Ent}_{1, u,2} = \{0, 2\} \). The signature includes the same unary relation \( \text{BLACK} \) where \( \text{BLACK} = \{0\} \) for \( S_1 \), \( \text{BLACK} = \{1\} \) for \( S_2 \), and \( \text{BLACK} = \{0\} \) for \( S_1 \setminus S_2 \).

The join function is defined as \( S_1 \cup S_2 = (\text{Ent}_1 \setminus \text{Ent}_2, \sigma, I_0) \) i.e. the set of entities is the union of the domains of \( S_1 \) and \( S_2 \). The interpretation of \( \sigma \) is defined as follows:

- for each \( \tau_i \in \sigma \), \( L(\tau_i) = L(\tau_i) \cup I_0(\tau_i) \)

\(^3\mathcal{P}(A)\) denotes the power set of the set \( A \), i.e. the set of all subsets of \( A \) including the empty set.

\(^4\) We add \( S \) subscripted in each command to differentiate this commands from the usual set operators \( \setminus \) and \( \cup \) which we use extensively through the paper. For the sake of simplicity of notation we drop the subscript \( \sigma \) in \( S_\sigma \).
Let cross: \( \mathcal{P}(\mathcal{E}) \times \mathcal{P}(\mathcal{E}) \rightarrow \mathcal{P}(\mathcal{E} \times \mathcal{E}) \) be a function that, given two sets of entities \( \text{Ent}_1 \) and \( \text{Ent}_2 \), it returns all of the edges from \( \text{Ent}_1 \) to \( \text{Ent}_2 \). Then, \( I_{\text{ed}}(\text{ed}) = I_{\text{ed}}^1(\text{ed}) \cup I_{\text{ed}}^2(\text{ed}) \cup \text{cross}(\text{Ent}_1, \text{Ent}_2) \).

The join operator is another operator for merging two structures. More than doing the union between two structures, the join operator also creates edges between all entities on the first structure to all entities on the second structure. This is useful when one wants to add a new entity to the system that should be connected to all the old entities.

**Example 3.** Fig. 2(c) presents two structures \( S_1 \) and \( S_2 \) and \( S_1+S_2 \). \( \text{Ent}_1 = \{0, 1, 2\} \), \( \text{Ent}_2 = \{1\} \), and \( \text{Ent}_{S_1+S_2} = \{0, 1, 2, 3, 4\} \). The signature includes the same unary relation \( \text{BLACK} \) where \( \text{BLACK} = \{0\} \) for \( S_1 \), \( \text{BLACK} = \{3\} \) for \( S_2 \) and \( \text{BLACK} = \{0, 3\} \) for \( S_1+S_2 \).

We define a \( \mathcal{A} \)-expression as an expression that syntactically agrees with the signature of \( \mathcal{A} \). For the sake of clarity we introduce a grammar for \( \mathcal{A} \)-expressions:

\[
\mathcal{A}\text{Exp} ::= S \mid \mathcal{A}\text{Exp} \lor \mathcal{A}\text{Exp} \mid \mathcal{A}\text{Exp} \land \mathcal{A}\text{Exp} \mid \mathcal{A}\text{Exp} + \mathcal{A}\text{Exp}
\]

where \( S \in \Sigma \). We denote the set of all \( \mathcal{A} \)-expressions as \( \mathcal{A}\text{Exp} \).

Each event in \( \Sigma \) of the transition system \( T \) maps to a \( \mathcal{A} \)-expression that takes the current structure and transforms it to the next structure.

### 4. MODELLING STATE OF ENTITIES

In this section we present two ways of embedding the state of the entities into the structure model: using unary relations and embedding the state-space directly into the SM.

#### 4.1 Model State Using Unary Relations

Let \( Q \) be the state-space of entities. For the sake of simplicity we assume that the entities have homogeneous state-space. Let \( \tau_q \subseteq \text{Ent} \) be a unary relation such that \( e \in \tau_q \iff \text{the state of entity } e \text{ is } q \) where \( q \in Q \). This solution allows modelling the state of entities into an SM without the need of extra notation.

**Definition 5 (State Consistency).** Let \( S = (\text{Ent}, \sigma = \{\tau_{q_1}, \ldots, \tau_{q_n}, \text{ed}\}, I) \) be an SM where \( Q = \{q_1,\ldots,q_n\} \) denotes the state-space of entities. \( S \) is state-consistent iff \( (\forall e \in \text{Ent}. (\exists q \in Q. (e \in \tau_q))) \). State consistency ensures that each entity \( e \in \text{Ent} \) is at one and only one state \( q \in Q \). If one wants to model non-determinism where a given entity can be at different states at a given time, this should be introduced at the level of the SM (modelling different SMs) and not at the level of the state.

**Example 4.** Consider an SM with the domain \( \text{Ent} = \{v_1, v_2, \ldots, v_n, t_1, t_2, \ldots, t_m\} \) where \( v_i \) denotes a vehicle and \( t_j \) denotes a task. Vehicles can be assigned to tasks. Each vehicle is an entity of the system and is modelled as DES. All vehicles share the same DES definition where the state space is defined as \( Q = \{\text{IDLE, ASSIGNED, DONE, FAIL, TODO}\} \). The state IDLE means that the vehicle is operative but not assigned to any task, the state ASSIGNED means that the vehicle is assigned to a task, the state DONE means that the vehicle finished a task, and the state FAIL means that the vehicle is inoperative. A task can be on state TODO which means that it was instantiated but not yet assigned, ASSIGNED, or DONE. The set of unary relations is then \( T = \{t_{\text{IDLE}}, t_{\text{ASSIGNED}}, t_{\text{DONE}}, t_{\text{FAIL}}, t_{\text{TODO}}\} \). Fig. 3 presents an example of one vehicle that is assigned to two tasks.

![Figure 3. Dynamic state and structure](image_url)
4.2 Embed the State-Space into an SM

Another approach is to redefine the SM definition in order to include the state-space of the entities. Let the function \( \text{state}: \text{Ent} \rightarrow Q \) be defined as \( \text{state}(e) = q \Leftrightarrow e \in \tau_q \) where \( e \in \text{Ent} \) and \( \tau_q \)'s are defined as per the previous approach. We can now extend the definition of a SM from a relational structure to a many-sorted algebra where \( \text{state} \) is functional symbol. We call this algebraic model the Structure-State Model (SSM).

**Definition 6 (Structure-State Model).** A Structure-State Model is a many-sorted algebra denoted as \( S = (\text{Ent}, Q, \{\sigma, \text{state}\}, \{I_{\text{in}}, I_{\text{state}}\}) \) which results from the overloading of a structure model \( S = (\text{Ent}, \sigma, I_{\text{en}}) \). \( Q \) is the state space of entities, state is a function symbol that is interpreted by \( I_{\text{state}} \) to a function of type \( \text{Ent} \rightarrow Q \) which, given an entity, provides the state of such entity.

For the remainder of the paper we are going to use the approach with the unary relations. The use of SSM remains future work.

5. COMPOSITION WITH AN UNDERLYING MOC

We aim at developing a model that can be used together with different MoCs that models the underlying dynamics of entities. In this section we compose the SM with Nancy Lynch’s Synchronous Network Model. The composition consists of a map that extracts the SM from an SNM and a map that changes the SNM up to future work.

5.1 Synchronous Network Model

An SNM is defined as a set of processes \( \text{Proc} \) and a set of messages \( M \). Each \( p \in \text{Proc} \) is defined as a tuple \( p = (\text{states}_p, \text{start}_p, \text{msgs}_p, \text{in}_p, \text{out}_p) \) where \( \text{states}_p \) is a (possibly infinite) set of states, \( \text{start}_p \subseteq \text{states} \) is the set of initial states, \( \text{in}_p \subseteq \text{Proc} \) denotes the incoming neighbours of \( p, \text{out}_p \subseteq \text{Proc} \) denotes the outgoing neighbours of \( p, \text{msgs}_p: \text{states}_p \times \text{out}_p \rightarrow M \cup \{\text{null}\} \) is a message-generation function, and \( \text{trans}_p: \text{states}_p \times (M \cup \{\text{null}\})^{\text{in}_{\text{port}}} \rightarrow \text{states}_p \) is a state-transition function.

For every round, each process \( p \) in the network sends a message to each of its outgoing neighbors (specified by \( \text{msgs} \) function) and receives a message from each of its incoming neighbours, computing its new state (specified by \( \text{trans} \) function). Note that a SNM forms a directed graph \( G = (V, E) \) where \( V = \text{Proc} \) and the set of edges \( E \) is formed according to the incoming and outgoing neighbours of each process. In order to instantiate new SNM processes during execution it is convenient to have a set of “templates” abstracted from any incoming and outgoing neighbours. We define an SNM template has an SNM process where incoming neighbours and outgoing neighbours are abstracted by place-holders called ports. The sets \( \text{in}_{\text{port}} \) and \( \text{out}_{\text{port}} \) are respectively the place-holders for incoming neighbours and outgoing neighbours. The functions \( \text{msgs} \) and \( \text{trans} \) are defined using the sets of ports instead of incoming and outgoing neighbours. We denote the set of process templates as \( P_{\text{imp}}. \) An \( n \)-process SNM can be instantiated using a partial function \( \text{inst}_n: \mathcal{P}(P_{\text{imp}})^n \times \mathcal{P}(\text{Ports}_{\text{in}} \times \text{Ports}_{\text{out}}) \rightarrow \mathcal{N} \) where \( \text{Ports}_{\text{in}} \) and \( \text{Ports}_{\text{out}} \) are respectively the set of all input ports and the set of all output ports for all the \( n \) process templates. \( \text{inst}_n \) takes \( n \) process templates and a set of edges representing connections between input ports and output ports. After generating the processes, \( \text{inst}_n \) binds the ports according to the set of edges. Note that all ports must be bonded in order to \( \text{inst}_n \) to be defined. Otherwise it cannot produce a sound SNM.

5.2 Synchronous Network Model

We define a map \( \epsilon: \mathcal{N} \times V_{\text{states}} \rightarrow S \) that extracts the SM from an SNM. \( \mathcal{N} \) is the set of all SNMs, and \( V_{\text{states}} \) is a set of vectors where each \( [p_1], [p_2], \ldots \) \( \in V_{\text{states}} \) represents a snapshot of the state of all SNM processes.

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5 Given a tuple \( s = (x_1, x_2, \ldots) \), we use the notation \( x_{i_{|s}} \) to access to the value of \( x_i \).
5.3 Map from Structure Model to Synchronous Network Model

In order to be able to control the SNM from the structure level we need to introduce the map \( \beta : S \times S \times \mathcal{N} \times V_{\text{states}} \to \mathcal{N} \) which takes the current structure, the new structure, and the old SNM and returns the new SNM. Note that adding and removing processes to and from an SNM might also require the redefinition of in\(_p\), out\(_p\), \text{msgs}_p\), and \text{trans}_p\) for some processes in \(\text{Proc}\).

6. CASE STUDY

6.1 The Agree-and-Pursue Law (APL)

Our case study is an adaptation of the “agree-and-pursue control and communication law” (APL) used by Martínez et al. [MBCF07] to present an SNM-based model called Synchronous Robotic Networks. The example consists of a set of locally connected agents that can move on a circle. Each agent can travel clockwise or counter-clockwise. The APL can be informally defined as follows: each agent transmits to its neighbours a message with its position, direction and the unique identifier (UID) of their “leader” denoted as “prior”; each agent compares its “prior” with the “prior” of all its incoming messages. If there exists a message with greater “prior” then the agent adopts it, assumes its direction, and increases its position proportionally to the difference between its current position and the position stated in the elected message. Otherwise the agent increases its position according to its previous pace. We also assume that agents can be added and deleted during execution. This feature is not part of the original APL. We introduce it in order to enrich the behavior of the structure.

Consider an APL model with \(N_a\) agents. For the sake of simplicity we assume that agents are homogeneous in terms of communication range. The space is discretized into \(N_s\) samples of a unit circle. The set of agents is modelled as a set of SNM processes \(\text{Proc}\) where each for each \(p \in \text{Proc}\), \(s \in \text{states}_p\) is modelled as a tuple \(s = (\text{pos}, \text{dir}, \text{prior})\) where \(\text{pos} \in \{0, 1, \ldots, N_s - 1\}\), \(\text{dir} \in \{\text{cw}, \text{ccw}\}\) and, \(\text{prior} \in \mathbb{N}_0\), and \(\text{start}_p = (\text{pos}_0, \text{dir}_0, \text{UID}(p))\). \text{pos}_0 and \text{dir}_0 are arbitrary initial position and direction and the function \text{UID} provides a unique identifier for the process.

The sets \(\text{in}_p\) and \(\text{out}_p\) include all processes except \(p\). The function \(\text{msgs}_p\) is defined as:

\[
\text{msgs}_p(s, p') = \begin{cases} 
  s, & \text{if dist}(s, \text{pos}, s', \text{pos}) \leq \lambda \\
  \text{null}, & \text{otherwise}
\end{cases}
\]

where \(s'\) is the state of \(p'\). The function \(\text{trans}_p\) implements the control algorithm informally stated above. Formally,

\[
\text{trans}_p(s, \{m_1, m_2, \ldots, m_{N_a}\}) = \begin{cases} 
  (s \mid \text{pos} \pm \delta_p, s \mid \text{dir}, s \mid \text{prior}) & \text{if } s \mid \text{prior} \geq M \mid \text{prior} \\
  (s \mid \text{pos} \pm \eta \text{dist}(s, M \mid \text{pos}), M \mid \text{dir}, M \mid \text{prior}) & \text{o.w.}
\end{cases}
\]

where \(M \in \{m_1, m_2, \ldots, m_{N_a}\}\) is the message received with the greatest prior. \(\eta \in \mathbb{N}\) is a fixed gain, and \(\delta_p\) is the previous step for process \(p\). We use the signal \(\pm\) in the update of the position is replaced by \(+\) if \(\text{dir}\) is \(\text{cw}\) and \(-\) if \(\text{dir}\) is \(\text{ccw}\). We denote the resulting SNM as \(A_{\text{SNM}}\).

Let \(S = (\text{Proc}, \{\text{CW}, ed\}, \{\text{CW}, f\})\) denote the SM for APL where the set of entities are the set of SNM processes of \(A_{\text{SNM}}\). \(\text{CW}\) is a unary relation that is true if the corresponding agent is rotating clockwise and false otherwise. \(\text{CW}\) is formally interpreted using a predicate \(f_{\text{CW}}(p) = \top\) iff \(p \mid \text{dir} = \text{cw}\) and \(f_{\text{CW}}(p) = \bot\) otherwise.
The binary relation ed models if two agents are within communication range. The interpretation \( f_{ed} \) is defined such that the following constraint is satisfied:
\[
\forall p,q \in \text{Proc.} (p,q) \in ed \Rightarrow \text{dist}(p|_{\text{pos}}, q|_{\text{pos}}) \leq \lambda.
\]

An incoming edge and an outgoing edge between two entities means that the corresponding agents are within communication range. Note that this information cannot be extracted directly from the edges of \( A_{\text{SNM}} \) since SNM doesn’t allow changing the edges during execution. This behavior is entailed in the messages exchanged by processes: if a null message is sent between two agents then they are out of range. In the structure model the behavior of the communication topology is a first-class citizen and can be accessed syntactically by looking at the set of edges.

Note that we are not modelling all the state-space of the SNM processes at the structure level but only the state information that we find important for the sake of this example. This state-space abstraction is a common technique to prevent state-explosion of combinatorial algorithms. In the structure model, state-space abstraction and refinement can be done in a very natural way by adding and removing unary relations on the signature of the structure.

Fig. 4 depicts an example of three execution steps of an APL with seven agents with the corresponding SMs. The black disks denote agents rotating counter-clockwise while the unfilled disks represent agents rotating clockwise. The circumference surrounding each agent represents its communication range. At the bottom of Fig. 4 we present the corresponding STS. This STS is derived from the \( A_{\text{SNM}} \) using the map \( \alpha \) at each step. The vectors \( v, v′, \) and \( v″ \) are the state of the processes at each execution step. Note that APL has a bottom-up emerging dynamic structure: the communication topology changes due to the communication constraints, and the entities state changes due to their direction.

First we need to define the map \( \beta \). We use the general definition provided in section 5. The only application-specific step is the definition of \( \text{insta} \) for the instantiation of new processes. The instantiation needs to be sound up to the physical constraints of the system. Thus, the initial position of processes generated by \( \text{insta} \) cannot violate the communication constraints.

Consider that we want to have two ”teams” of APL agents. Agents of each team only share information with their team-mates, and thus do not influence the behavior of agents outside the team. Let us design a structure-level controller that controls the network topology such that the system meets this new specification. The controller observes the current structure and if two agents from different teams share an edge (i.e. they are within communication range) then the controller disables communication between the agents by removing the edge.

Consider a new operator denoted as
\[
\forall ed, S \times (\text{Ent} \times \text{Ent}) \rightarrow S \quad \text{such that}
\]
\[
S' = S \forall ed (e_0, e_1) \text{ where } S' \text{ has the same structure as } S \text{ except } ed_{S'} = ed_S \setminus (e_0, e_1),
\]
i.e. the operator removes the edge from the original structure maintaining the remaining structure the same.

The following pseudo-code implements the team controller.

The first step consists of initializing the “agree and pursue law” as a set of SNM processes, their corresponding initial states, and the definition of the two teams. Inside the infinite loop the first step consists of obtaining the current structure using the map \( \alpha \). The structure controller is represented by the “forall” loop which checks for edges shared by agents at different teams and removes them. The next step consists of propagating the new structure to the system using the map \( \beta \). In this case \( \beta \) can disable communication channels even when the agents are within communication range. If the corresponding edges get removed at the structure model \( \beta \) disables the capacity for both entities to communicate. \( A_{\text{SNM}} \) is then executed and the states of the processes are calculated.

This simple example shows the convenience of accessing the network topology as a first-class concept. Remember that the structure of an SNM is not explicit but rather implicit in the messages sent by each
process. Thus, the structure model introduces a layer of abstraction that makes the structure explicit and easier to manipulate.

7. CONCLUSIONS AND FUTURE WORK

In this paper we introduce the Structure Model (SM) - an algebraic model of computation for modelling and controlling systems with dynamic structure. The SM is a mathematical structure defined by a set of entities, a set of unary relations and a binary relation. The unary relations are used to model the state of the entities while the binary relation models interactions between entities (e.g. communication network topology). An SM can be manipulated using a set of operators. The model for time evolution of SMs is formalized as a transition system. The structure model aims at being agnostic to the model of computation that describes the underlying system’s behaviour. The composition of an underlying model with the SM allows extracting structural information from the system up to the structure level and propagate changes from the structure level back to the system. This allows the specification of structure-level controllers. We introduce a composition with the Synchronous Network Model and present a simple case study based on the ”agree and pursue” communication and control law.

The structure model is at an early stage of development. We foresee work to be done on the definition of the set of operators for manipulating structures. We are looking for the least amount of operators that are expressive enough to model useful controllers. This work will lead us to the design and implementation of a Domain-Specific Language (DSL) where structure-level controllers can be easily specified. We are also exploring the use of Bigraphs for extending the structure model with location of computation.

REFERENCES


