Abstract

We present a set of simulation processors and describe their temporal collaboration in support of the execution of models specified by an advanced Discrete Event Systems Specification (DEVS) formalism. Our formalism improves the existing parallel DEVS formalism by adopting the universal coupling mechanism. The mechanism enables coupled models to create, permute, and delete their couplings in a proactive or reactive manner. Each coupled model is a decomposable, constructive, multi-scale, discrete-event system that contains at least one component. Proposed simulation processors will allow exploration of spatiotemporal spaces generated from specified models. A simulation protocol is realized as a sequence of well-defined temporal interactions between those processors. It is expected that current modeling and simulation (M&S)-driven systems biology will be empowered by the new DEVS formalism and proposed simulation processors.

1 Introduction

Many complex biological phenomena result from dynamic spatiotemporal interactions between components and subsystems. To cope with the dynamic nature of the interactions, several computational methods have been applied. Process algebra describes an interaction between systems as a set of communication link and its dynamic permutation [1][2][3]. It focuses mainly on relationships between systems rather than on their internal temporal dynamics. Meanwhile, discrete event systems and discrete-event driven M&S focus on the temporal dynamics of systems [4][5][6]. They describe an interaction by message (or event) exchange or propagation links. The message is consumed or produced by a system as the result of a reactive response to incoming messages or its proactive dynamic state transitions. Specifically, agent-based M&S describes system level biological phenomena as resulting from interactions between agents, where some agents represent biological entities and others the environment within which these entities reside [7]. An agent is commonly postulated to be an intelligent and autonomous (discrete event) system. Membrane computing, also known as P-system, provides similar capabilities for dealing with dynamic interactions covered by process algebra from a system-oriented perspective [8]. However, it does not explicitly discuss temporal aspects of the relation. It is desirable to merge the advantages of above approaches to efficiently describe and explore complex dynamic interactions of the type that occur between and inside biological systems.

We have developed an advanced DEVS modeling formalism to describe adaptive spatiotemporal interactions of biological systems and the phenomena they produce [9]. The formalism has been applied to a set of biology problems including DNA formation and synthesis. It demonstrated the power and effectiveness of the new formalism in describing dynamic spatiotemporal interactions and the temporal dynamics that occur between and within biological systems.

In this paper, we present a set of simulation processors and describe their temporal collaboration to execute models described by the new modeling formalism. Proposed simulation processors extend an existing parallel DEVS abstract simulator and coordinators by integrating the universal interaction mechanism [9][10]. The universal interaction specification and revised DEVS model specifications will be briefly described before simulation processors are presented.

2 Universal coupling

A coupling is a directional influence from one entity to another. The influencer and influencee are commonly described as a pair: a model and its property (e.g., communication I/O port). It is classified into three categories: static, dynamic, and adaptive.
A static coupling consists of one influencer \((m_i, p_i)\) and one influencee \((m_j, p_j)\). It represents a causal relation from the influencer to the influencee. The coupling is specified by a 4-tuple \(C_{i,j}^d : (m_i, p_i, m_j, p_j)\) where \(m_i, m_j \in M, p_i, p_j \in P\). \(M\) is a set of models, and \(P\) is a set of properties.

A dynamic coupling is an extension of the static coupling that enables its dynamic permutation. Dynamic permutation includes the change of influencer or influencee, construction of a new static or dynamic coupling, and destruction of an existing dynamic coupling. It is specified by a 9-tuple \(C_{i,j}^d : (m_i, p_i, m_j, p_j, m_k, p_k, \xi, \rho, l)\) where \(m_i, m_j, m_k \in M, p_i, p_j, p_k \in P, \xi\) is a coupling constraint, \(\rho : C^d \rightarrow C^d\) is a coupling permutation function, and \(l \in Z^+\) is a lifespan variable. It can be alternatively specified by an endomorphic 5-tuple \(C_{i,j}^d : (C_{i,j}^s, m_k, p_k, \xi, \rho, l)\) where \(C_{i,j}^s\) is a static coupling, \(m_k, p_k, \xi, \rho\), and \(l\) are as same as above. \((m_k, p_k)\) is the candidate that will be promoted to a new influencer or influencee after dynamic permutation. \(\xi\) specifies when the permutation occurs. Satisfying \(\xi\) triggers \(\rho\). \(\rho\) provides details on how the coupling is changed. The dynamic coupling is valid only when \(l > 0\). \(l\) can be increased or decreased after execution of \(\rho\). An important property of the coupling is its capability to construct, permute, and evolve coupling networks from an existing coupling or coupling network. By tracking changes in network topology, we can trace dynamic coupling permutation between models during their lifetime.

An adaptive coupling is an extension of the dynamic coupling that permits meta entities in the coupling specification. The entities are not precisely described in an initial specification. However, they are adaptively resolved and bound to concrete entities at later times. The coupling is mainly used to adaptively create a dynamic coupling in a proactive or reactive manner. It divides into two adaptive couplings - proactive and reactive. A proactive coupling is an extension of the dynamic coupling that contains a meta influence or influencer \((*, *)\), a binding constraint \(\psi\), and a binding function \(\varphi : C^d \rightarrow C^d, (*,*)\) is resolved and bound by \(\varphi\) when \(\psi\) is satisfied. \(\varphi\) creates a dynamic coupling. The coupling is represented by a 11-tuple \(C_{i,j}^{\varphi} : (m_i, p_i, *, *, m_k, p_k, \xi, \rho, l, \psi, \varphi)\) or \(C_{s,i}^{\varphi} : (m_s, p_s, m_j, p_j, \xi, m_k, p_k, \rho, l, \psi, \varphi)\), respectively, depending on the existence of the meta influence or influencee. It is alternatively simplified to an endomorphic 3-tuple - \(C_{i,j}^{\varphi} : (C_{s,i}^{\varphi}, \psi, \varphi)\) or \(C_{s,j}^{\varphi} : (C_{i,j}^{\varphi}, \psi, \varphi)\). A reactive coupling is an adaptive coupling that contains a meta candidate. Unlike the proactive coupling, the influencer and influencee are "clearly" presented but the candidate \((m_k, p_k)\) is not precisely defined in the coupling. The coupling is denoted as \(C_{i,j}^{\varphi_m} : (m_i, p_i, m_j, p_j, *, *, \xi, \rho, l, \psi, \varphi)\). The binding function \(\varphi : C^d \rightarrow C^d\) resolves the meta candidate when \(\psi\) is satisfied. \(C_{i,j}^{\varphi_m} : (C_{i,j}^{\varphi_m}, \psi, \varphi)\) is its endomorphic representation.

A universal coupling \(C_{i,j}^{u} : C_{i,j}^{s} \oplus C_{i,j}^{p} \oplus C_{s,j}^{u} \oplus C_{i,j}^{p} \) is the composition of all couplings introduced above. See [4] and [9] for more details on coupling mechanisms.

3 DEVS coupled model with universal coupling

Our new DEVS coupled model strengthens the existing coupling specification of Parallel DEVS (P-DEVS) by replacing it with the universal coupling specification [4][9][10]. For the congruent integration of the universal coupling specification with the DEVS formalism, we use causal relation forms instead of tuples. Specifically, \((m_i, p_i) \rightarrow (m_j, p_j)\) for \(C_{i,j}^d\), \((m_i, p_i) \leftarrow (m_j, p_j)\) for \(C_{i,j}^p\), \((m_s, p_s) \leftarrow (m_j, p_j)\) for \(C_{i,j}^p\), and \((m_s, p_s) \leftarrow (m_j, p_j)\) for \(C_{i,j}^p\). The DEVS coupled model with the universal coupling \(N = \{X, Y, D, \{M_d\}, \{I_d\}, \{Z_d, i\}\}\). \(X\) is a set of inputs. \(Y\) is a set of outputs. \(D\) is a set of components. For each \(d \in D\), \(M_d\) is a DEVS model, which is either atomic or coupled. For each \(d \in D \cup \{self\}, I_d\) is the set of influencees of \(d, self\) is a reserved reference of \(N\) itself. For each \(i \in I_d, Z_d, i\) is a function, \(d\)-to- \(i\) output translation with \(Z_d, i \in (Z_d^d \cup Z_d^d) \cup Z_d^d \cup Z_d^d\). \(Z_d^a\) is the static translation function with \((i) Z_d, i : X \rightarrow X, d = self\); \((ii) Z_d, i : Y_d \rightarrow Y, i = self\) and \((iii) Z_d, i : Y_d \rightarrow X, d \neq self \land i \neq self\). \(Z_d^a\) is the dynamic translation function with \((i) Z_d, i : X \leftarrow X, d = self\); \((ii) Z_d, i : Y_d \rightarrow Y, i = self\); \((iii) Z_d, i : Y_d \rightarrow X, d \neq self \land i \neq self\). \(Z_d, i\) is the reactive translation function with \((i) Z_d, i : X \rightarrow X, d = self\); \((ii) Z_d, i : Y \rightarrow Y, i = self\); \((iii) Z_d, i : Y \rightarrow X, d \neq self \land i \neq self\). Where \(* \in D\) is an unspecified component. \(Z_d^a\) is the reactive translation function with \((i) Z_d, i : X \rightarrow X, d = self\); \((ii) Z_d, i : Y \rightarrow Y, i = self\); \((iii) Z_d, i : Y \rightarrow X, d \neq self \land i \neq self\).
4 Simulation protocol

For simulation time management, we use \( t_L, t, t_N, e, \sigma \) to represent the time the last event occurs, the current simulation time, the time the next event will occur, the elapsed time since \( t_L \), and the time left until \( t_N \), respectively. Specifically, \( e = t - t_L \), \( \sigma = t_N - t = t_N - (t_L + e) \), and \( t_N = t_L + \tau(s) = t + \sigma \) where \( t_L \leq t \leq t_N \).

![Figure 1: DEVS simulation time](image)

There exist three different simulation processors - root coordinator, coordinator, and simulator. A root coordinator controls the top-most coordinator. Each coordinator controls its components by exchanging simulation control messages with both its parent coordinator and child components. Each simulator executes its associated atomic model whenever it receives a simulation control message from its parent coordinator. Algorithm 1 presents an atomic simulator that refines the original P-DEVS abstract simulator.

**Algorithm 1 Simulator for an atomic model**

1. \( t_L := e := 0; t_N := \tau(s); x^b = \phi \)
2. when \((%, t)\) message arrives from parent
3. \( \text{send } (\%, \tau(s) - (t - t_L)) \) to parent
4. end when
5. when \((0, t)\) message arrives from parent
6. \( \text{send } (y, ((t = t_N) \lor \lambda(s) = \phi), t) \) to parent
7. end when
8. when \((x, x_d^b, t)\) message arrives from parent
9. \( x^b := x^b \oplus x_d^b \)
10. \( \text{send } (done, t_N) \) to parent
11. end when
12. when \((s, t)\) message arrives from parent
13. if \( t < t_L \lor t > t_N \) then \( \text{timing synchronization error} \) else
14. end if
15. e := \( t - t_L \)
16. if \( t_L \leq t \leq t_N \) then
17. \( s := (x^b \neq \phi) \lor \delta_{ext}(s, e, x^b) : s \)
18. else if \( t = t_N \) then
19. \( s := (x^b = \phi) \lor \delta_{int}(s) : \delta_{conf}(s, e, x^b) \)
20. else \( \text{raise error} \)
21. end if
22. end if
23. \( t_L := t; t_N := t_L + \tau(s); x^b := \phi \)
24. end when
25. A coordinator advances the simulation time by

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the amount of \( \sigma_{min} \), which is \( \min \{ \sigma_d | d \in D \} \) where \( \sigma_d = \tau(s_d) - e_d \), \( s_d, e_d, \tau(s_d) \), and \( \sigma_d \) are \( s, e, \tau(s) \), and \( \sigma(s) \) of \( M_d \in M \), respectively. \( IMM \) is the set of imminent components that produce outputs, \( \{ d \sigma_d = \tau(s_d) \land \lambda_d^b(s_d) \neq \phi \} \) at time \( t_L + \sigma_{min} \). \( \lambda_d^b(s_d) \) is the bag of output messages produced by \( M_d \). \( INF \) is the set of components influenced by \( d \in IMM \), \( \{ i | i \in d_d, d \in IMM \land x_i^b \neq \phi \} \) where \( x_i^b \in \{ Z_d_i(\lambda^b_d(s_d)) | d \in IMM \land i_d \} \). It is equivalent to \( \{ d \sigma_d = \tau(s_d) \land \lambda_d^b(s_d) \neq \phi \} \) in the algorithm we present here because \( \lambda_d^b(s_d) \) produced by each \( d \in IMM \) is translated to \( \{ x_i^b \} \in D \) through \( Z_d_i(\lambda^b_d(s_d)) \) based on \( I_d \) before \( INF \) is computed. Dynamic or adaptive permutation is managed when the coordinator receives the message \((!t)\) from its parent coordinator. A collection of Algorithm 2 and 3 represents a coordinator that refines and extends the P-DEVS coordinator with the integration of the universal coupling specification.

**Algorithm 2 Coordinator for a coupled model: Part I**

1. \( t_L := t_N := e := 0; sync := \phi \)
2. when \((!, t)\) message arrives from parent
3. \( \text{sync} := D \)
4. for each \( d \in D \) do \( \text{send } (\%, t) \) to child \( d \) end for
5. end when
6. when \((%, t)\) message arrives from child \( d \)
7. \( \sigma_d := t; \text{sync} := sync\setminus d \)
8. if \( \text{sync} = 0 \) then
9. \( t_N := t_L + min\{\sigma_d\} \)
10. \( \text{send } (done, t_N) \) to parent
11. end if
12. end when
13. when \((0, t)\) message arrives from parent
14. \( IMM := INF := \phi; \text{sync} := D \)
15. for each \( d \in D \) do
16. \( x^b_d := \phi \)
17. \( \text{send } (0, t) \) to child \( d \)
18. end for
19. end when
20. when \((y, y^b, t)\) message arrives from child \( d \)
21. \( \text{sync} := \text{sync}\setminus d \)
22. if \( y^b \neq \phi \) then
23. \( IMM := IMM \land d \)
24. for each \( i \in I_d\setminus self \) do \( x^b_i := x^b \oplus z_d,i(x^b) \) end for
25. if \( \text{self} \in I_d \) then \( y^b_{self} := y^b_{self} \oplus z_d,\text{self}(y^b) \) end if
26. end if
27. if \( \text{sync} = 0 \) then
28. \( \text{send } (y, y^b_{self}, t) \) to parent
29. \( y^b_{self} := \phi \)
30. end if
31. end when
32. when \((x, x^b, t)\) message is received from parent
33. for each \( i \in I_d\setminus self \land x^b_i \neq \phi \) do
34. \( x^b_i := x^b_i \oplus z_d,\text{self}(x) \)
35. end for
36. for each \( d \in D \land x^b_d \neq \phi \) do
37. \( INF := INF \land d \)
38. end for
39. \( \text{sync} := INF \)
40. for each \( d \in INF \) do \( \text{send } (x, x^b_d, t) \) to child \( d \) end for
41. end when
Algorithm 3 Coordinator for a coupled model: Part II
42: when (l, t) message is received from parent
43: for each z in Z do
44: if z ∈ (L^x, r ⊕ L^y, r ⊕ L^z, r) then
45: if z.ψ is true then execute z.ϕ end if
46: else if z ∈ (L^x, d ⊕ L^y, d) then
47: if z.ξ is true then execute z.ρ end if
48: end if
49: end for
50: end when
51: when (s, t) message is received from parent
52: e := t - t_L
53: if t_L ≤ t ≤ t_N then
54: sync := IMM ⊕ INF
55: for each d ∈ IMM ⊕ INF do
56: send (s, t) to child d end for
57: end for
58: else raise an error
59: end if
60: t_L := t
61: end when
62: when (done, t) message arrives from child d
63: sync := sync/d
64: if ∣sync⟩ = 0 then send (done, t) to parent end if
65: end when

The root coordinator initiates the whole simulation cycle by sending the (#, t_N) message to the top-most coordinator and terminates it if t_N is ∞.

Algorithm 4 Root coordinator
1: t_N := 0
2: while t_N ≠ ∞ do
3: send (#, t_N) to the top coordinator
4: wait for (done, t_N) from the top coordinator
5: send (@, t_N) to the top coordinator
6: wait for (y, t_N) from the top coordinator
7: send (z, x^b, t_N) to the top coordinator
8: wait for (done, t_N) from the top coordinator
9: send (!, t_N) to the top coordinator
10: wait for (done, t_N) from the top coordinator
11: send (*, t_N) to the top coordinator
12: wait for (done, t_N) from the top coordinator
13: end while

5 Conclusion

We present a set of three simulation processors to execute and control models specified using the advanced DEVS modeling formalism. The formalism defines coupling relations between components of a coupled model with the universal coupling specification. The formalism enables dynamic creation, permutation, and removal of couplings in a proactive or reactive manner. With this feature, the coupling networks between components in a coupled model can be evolved during simulation. The evolution can be traced by tracking changes in the coupling network topology. Proposed simulation processors enable exploration of spatiotemporal spaces generated by dynamic construction and permutation of interactions between biological system components. Our expectation is that these processors and the revised formalism will facilitate progress in how the hierarchical, multiscale mechanistic details of biological systems can be modeled, coupled, and simulated.

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