Modeling Biological Systems in Stochastic Concurrent Constraint Programming

Luca Bortolussi¹ Alberto Policriti¹

¹Department of Mathematics and Computer Science University of Udine, Italy.

Workshop on Constraint Based Methods for Bioinformatics, Nantes, 25th September 2006

Modeling Biological Systems with Stochastic Process Algebras



Pros

- Simple Language
- Compositionality

Cons

- Hard to encode general information
- Lacking computational extensibility

Constraints... why not?

Outline



- Concurrent Constraint Programming
- Continuous Time Markov Chains
- Stochastic CCP



- Modeling Biochemical Reactions
- Modeling Gene Regulatory Networks

Outline



- Concurrent Constraint Programming
- Continuous Time Markov Chains
- Stochastic CCP

2 Bio-Modeling

- Modeling Biochemical Reactions
- Modeling Gene Regulatory Networks

Concurrent Constraint Programming

Constraint Store

- In this process algebra, the main object are constraints, which are formulae over an interpreted first order language (i.e. X = 10, Y > X 3).
- Constraints can be added to a "pot", called the constraint store, but can never be removed.

Agents

Agents can perform two basic operations on this store:

- Add a constraint (tell ask)
- Ask if a certain relation is entailed by the current configuration (ask instruction)

Syntax of CCP

Program = Decl.A

$$D = \varepsilon \mid Decl.Decl \mid p(x) : -A$$

$$A = \mathbf{0}$$

tell(c).A
ask(c₁).A₁ + ask(c₂).A₂
 $A_1 \parallel A_2 \mid \exists_x A \mid \rho(x)$

Continuous Time Markov Chains

A **Continuous Time Markov Chain** (CTMC) is a direct graph with edges labeled by a real number, called the rate of the transition (representing the speed or the frequency at which the transition occurs).



- In each state, we select the next state according to a *probability distribution* obtained normalizing rates (from *S* to *S*₁ with prob. $\frac{r_1}{r_1+r_2}$).
- The time spent in a state is given by an exponentially distributed random variable, with rate given by the sum of outgoing transitions from the actual node $(r_1 + r_2)$.

Syntax of sCCP

Syntax of Stochastic CCP

Program = D.A $D = \varepsilon \mid D.D \mid p(\mathbf{x}) : -A$ $\pi = \operatorname{tell}_{\lambda}(c) \mid \operatorname{ask}_{\lambda}(c)$ $M = \pi.A \mid \pi.A.p(\mathbf{y}) \mid M + M$ $A = \mathbf{0} \mid \operatorname{tell}_{\infty}(c).A \mid \exists_{x}A \mid M \mid (A \parallel A)$

Stochastic Rates

Each basic instruction (tell, ask, procedure call) has a rate attached to it. Rates are functions from the constraint store C to positive reals: $\lambda : C \longrightarrow \mathbb{R}^+$.

sCCP soup

Operational Semantics

- There are *two transition relations*, one instantaneous (finite and confluent) and one stochastic.
- Traces are sequences of events with variable time delays among them.

Implementation

- We have an interpreter written in Prolog, using the *CLP engine of SICStus* to manage the constraint store.
- Efficiency issues.

Stream Variables

- Quantities varying over time can be represented in sCCP as unbounded lists.
- Hereafter: special meaning of X = X + 1.

Outline



- Concurrent Constraint Programming
- Continuous Time Markov Chains
- Stochastic CCP



Bio-Modeling

- Modeling Biochemical Reactions
- Modeling Gene Regulatory Networks

General Principles

- Measurable Entities \leftrightarrow Stream Variables
 - Logical Entities ↔ Processes (Control Variables)
 - Interactions \leftrightarrow Processes

Biochemical Arrows to sCCP processes

$$R_{1} + \ldots + R_{n} \Rightarrow_{k} P_{1} + \ldots + P_{m}$$

$$R_{1} + \ldots + R_{n} \rightleftharpoons_{k_{2}}^{k_{1}} P_{1} + \ldots + P_{m}$$

$$S \mapsto_{K, V_{0}}^{E} P$$

$$S \mapsto_{K, V_{0}, h}^{E} P$$

 $\begin{array}{l} \operatorname{reaction}(k, [R_1, \ldots, R_n], [P_1, \ldots, P_m]) : - \\ \operatorname{ask}_{f_MA}(k, R_1, \ldots, R_n) \left(\bigwedge_{i=1}^n (R_i > 0) \right) . \\ \left(\begin{array}{c} \|_{i=i}^n \operatorname{tell}_{\infty}(R_i = R_i - 1) \end{array} \right) \|_{j=1}^m \operatorname{tell}_{\infty}(P_j = P_j + 1) \right). \\ \operatorname{reaction}(k, [R_1, \ldots, R_n], [P_1, \ldots, P_m]) \end{array}$

reaction $(k_1, [R_1, ..., R_n], [P_1, ..., P_m]) \parallel$ reaction $(k_2, [P_1, ..., P_m], [R_1, ..., R_n])$

 $\begin{array}{l} & \operatorname{mm_reaction}(K,\,V_0,\,S,\,P):-\\ & \operatorname{ask}_{f_{MM}(K,\,V_0,\,S)}(S>0).\\ & (\operatorname{tell}_{\infty}(S=S-1)\parallel \textit{tell}_{\infty}(P=P+1))\,.\\ & \operatorname{mm_reaction}(K,\,V_0,\,S,\,P) \end{array}$

 $\begin{array}{l} \mbox{hill_reaction}(K,\,V_0,\,h,\,S,\,P): - \\ \mbox{ask}_{\textit{fHH}}(K,\,V_0,\,h,S)(S>0). \\ \mbox{(tell}_{\infty}(S=S-h) \parallel \textit{tell}_{\infty}(P=P+h)) \,. \\ \mbox{Hill_reaction}(K,\,V_0,\,h,\,S,\,P) \end{array}$

where
$$r_{MA}(k, X_1, \dots, X_n) = k \cdot X_1 \cdots X_n$$
; $r_{MM}(K, V_0, S) = \frac{V_0 S}{S + K}$; $r_{Hill}(k, V_0, h, S) = \frac{V_0 S^h}{S^h + K^h}$

A simple reaction: $H + CI \Longrightarrow HCI$

We have two reaction agents. The reagents and the products are stream variables of the constraint store (put down in the environment). *Independent on the number of molecules*.

reaction(100, [*H*, *CL*], [*HCL*]) || reaction(10, [*HCL*], [*H*, *CL*])





Another reaction: Na + Cl \Rightarrow Na⁺ + Cl⁻

$\operatorname{reaction}(100, [\textit{NA}, \textit{CL}], [\textit{NA}+, \textit{CL}-]) \parallel \operatorname{reaction}(10, [\textit{NA}+, \textit{CL}-], [\textit{NA}, \textit{CL}])$



Enzymatic reaction

$$S + E \rightleftharpoons_{k_{-1}}^{k_1} ES \rightarrow_{k_2} P + E$$

Mass Action Kinetics

 $\begin{array}{l} {\rm enz_reaction}(k_1, k_{-1}, k_2, S, E, ES, P) :- \\ {\rm reaction}(k_1, [S, E], [ES]) \parallel \\ {\rm reaction}(k_{-1}, [ES], [E, S]) \parallel \\ {\rm reaction}(k_2, [ES], [E, P]) \end{array}$

Mass Action Equations

$$\begin{array}{l} \frac{d[ES]}{dt} = k_1[S][E] - k_2[ES] - k_{-1}[ES] \\ \frac{d[E]}{dt} = -k_1[S][E] + k_2[ES] + k_{-1}[ES] \\ \frac{d[S]}{dt} = -k_1[S][E] \\ \frac{d[S]}{dt} = k_2[ES] \end{array}$$

Michaelis-Menten Equations

$$\frac{d[P]}{dt} = \frac{V_0 S}{S+K}$$
$$V_0 = k_2 [E_0]$$
$$K = \frac{k_2 + k_{-1}}{k_1}$$

Michaelis-Menten Kinetics

mm_reaction
$$\left(\frac{k_2 + k_{-1}}{k_1}, k_2 \cdot E, S, P\right)$$

Enzymatic reaction







Michaelis-Menten Kinetics $mm_{reaction}\left(\frac{k_2+k_{-1}}{k_1}, k_2 \cdot E, S, P\right)$

MAP-Kinase cascade



 $\begin{array}{l} \text{enz_reaction}(k_a, k_d, k_r, KKK, E1, KKKE1, KKKS) \parallel \text{enz_reaction}(k_a, k_d, k_r, KKKS, E2, KKKSE2, KKK) \parallel \\ \text{enz_reaction}(k_a, k_d, k_r, KK, KKKS, KKKKKS, KKP) \parallel \text{enz_reaction}(k_a, k_d, k_r, KKP, KKP1, KKPKKP1, KK) \parallel \\ \text{enz_reaction}(k_a, k_d, k_r, KKP, KKKS, KKPKKS, KKPP) \parallel \text{enz_reaction}(k_a, k_d, k_r, KP, KP1, KPKP1, K) \parallel \\ \text{enz_reaction}(k_a, k_d, k_r, K, KKP, KKKP, KKP), KP) \parallel \text{enz_reaction}(k_a, k_d, k_r, KKP1, KKP1, KKP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KKPP, KKPP, KP) \parallel \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KKPP, KKPP, KPP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KKPP, KPP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KKPP, KPP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KKPP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KKPP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP, KP) \\ \text{enz_reaction}(k_a, k_d, k_r, KP) \\ \text{$





The gene machine



The instruction set

► b

 $\begin{array}{l} \text{null_gate}(k_{p}, X) : -\\ \text{tell}_{k_{p}}(X = X + 1).\text{null_gate}(k_{p}, X) \end{array}$



$$\begin{array}{l} \text{pos_gate}(k_{p}, k_{\theta}, k_{f}, X, Y) : -\\ \text{tell}_{k_{p}}(X = X + 1).\text{pos_gate}(k_{p}, k_{\theta}, k_{f}, X, Y)\\ + \text{ask}_{r(k_{\theta}, Y)}(true).\text{tell}_{k_{\theta}}(X = X + 1).\text{pos_gate}(k_{p}, k_{\theta}, k_{f}, X, Y) \end{array}$$



 $\begin{array}{l} & \operatorname{neg_gate}(k_{\rho},k_{i},k_{d},X,Y):-\\ & \operatorname{tell}_{k_{\rho}}(X=X+1).\operatorname{neg_gate}(k_{\rho},k_{i},k_{d},X,Y)\\ & +\operatorname{ask}_{r(k_{j},Y)}(true).\operatorname{ask}_{k_{d}}(true).\operatorname{neg_gate}(k_{\rho},k_{i},k_{d},X,Y) \end{array}$

where $r(k, Y) = k \cdot Y$.

L. Cardelli, A. Phillips, 2005.

Bio-Modeling

Repressilator







Bio-Modeling

Circadian Clock





Circadian Clock

 $\begin{array}{l} \mathsf{pos_gate}(\alpha_A, \alpha'_A, \gamma_A, \theta_A, M_A, A) \parallel \mathsf{pos_gate}(\alpha_R, \alpha'_R, \gamma_R, \theta_R, M_R, A) \parallel \\ \mathsf{reaction}(\beta_A, [M_A], [A]) \parallel \mathsf{reaction}(\delta_{MA}, [M_A], []) \parallel \\ \mathsf{reaction}(\beta_R, [M_R], [R]) \parallel \mathsf{reaction}(\delta_{MR}, [M_R], []) \parallel \\ \mathsf{reaction}(\gamma_C, [A, R], [AR]) \parallel \mathsf{reaction}(\delta_A, [AR], [R]) \parallel \\ \mathsf{reaction}(\delta_A, [A], []) \parallel \mathsf{reaction}(\delta_R, [R], []) \end{array}$



Conclusions

- We have introduced a stochastic version of CCP, with functional rates.
- We showed that sCCP may be used for modeling biological systems, defining *libraries* for biochemical reactions and gene regulatory networks.
- We showed that non-constant rates allow to use more complex chemical kinetics than mass action one.

Bio-Modeling



THANKS FOR THE ATTENTION!

QUESTIONS?