

Stochastic Concurrent Constraint Programming and Differential Equations

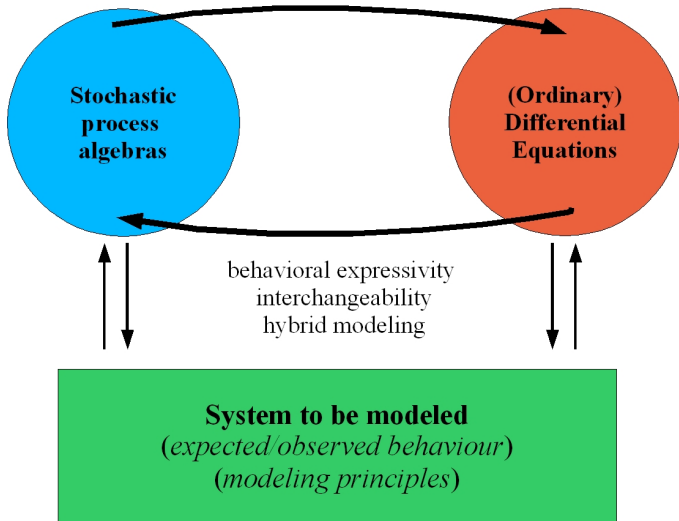
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The picture



Outline

- 1 Preliminaries
- 2 From ODE to SPA
- 3 From SPA to ODE

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Concurrent Constraint Programming

Constraint Store

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

Agents

Agents can perform two basic operations on this store (**asynchronously**):

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

Syntax of CCP

$$\text{Program} = \text{Decl}.A$$

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

$$A = \mathbf{0} \mid \text{tell}(c).A \mid \text{ask}(c_1).A_1 + \text{ask}(c_2).A_2 \mid A_1 \parallel A_2 \mid \exists_x A \mid p(x)$$

Syntax of sCCP

Syntax of Stochastic CCP

$$\begin{aligned}
 \text{Program} &= D.A \\
 D &= \varepsilon \mid D.D \mid p : -A \\
 \pi &= \text{tell}_\lambda(c) \mid \text{ask}_\lambda(c) \quad M = \pi.G \mid M + M \\
 G &= \mathbf{0} \mid \text{tell}_\infty(c).G \mid p \mid M \quad A = \mathbf{0} \mid M \\
 N &= A \mid A \parallel N
 \end{aligned}$$

L. Bortolussi, *Stochastic Concurrent Constraint Programming*, QAPL, 2006

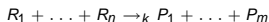
Stochastic Rates

Rates are functions from the constraint store \mathcal{C} to positive reals:

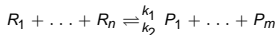
$$\lambda : \mathcal{C} \longrightarrow \mathbb{R}^+.$$

Rates can be thought as **speed** or **duration** of communications.

Biochemical Arrows to sCCP processes



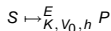
$\text{reaction}(k, [R_1, \dots, R_n], [P_1, \dots, P_m]) : -$
 $\text{ask}_{r_{MA}}(k, R_1, \dots, R_n) (\bigwedge_{i=1}^n (R_i > 0)) \cdot$
 $(\parallel_{j=1}^m \text{tell}_{\infty}(R_j = R_j - 1) \parallel_{j=1}^m \text{tell}_{\infty}(P_j = P_j + 1)).$
 $\text{reaction}(k, [R_1, \dots, R_n], [P_1, \dots, P_m])$



$\text{reaction}(k_1, [R_1, \dots, R_n], [P_1, \dots, P_m]) \parallel$
 $\text{reaction}(k_2, [P_1, \dots, P_m], [R_1, \dots, R_n])$



$\text{mm_reaction}(K, V_0, S, P) : -$
 $\text{ask}_{r_{MM}}(K, V_0, S) (S > 0) \cdot$
 $(\text{tell}_{\infty}(S = S - 1) \parallel \text{tell}_{\infty}(P = P + 1)).$
 $\text{mm_reaction}(K, V_0, S, P)$



$\text{hill_reaction}(K, V_0, h, S, P) : -$
 $\text{ask}_{r_{Hill}}(K, V_0, h, S) (S > 0) \cdot$
 $(\text{tell}_{\infty}(S = S - h) \parallel \text{tell}_{\infty}(P = P + h)).$
 $\text{Hill_reaction}(K, V_0, h, S, P)$

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

Enzymatic reaction



Mass Action Kinetics

```
enz_reaction(k1, k-1, k2, S, E, ES, P) :-
  reaction(k1, [S, E], [ES]) ||
  reaction(k-1, [ES], [E, S]) ||
  reaction(k2, [ES], [E, P])
```

Mass Action Equations

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

Michaelis-Menten Equations

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

Michaelis-Menten Kinetics

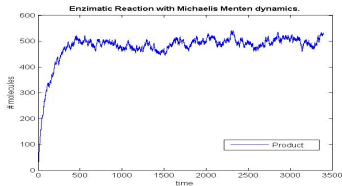
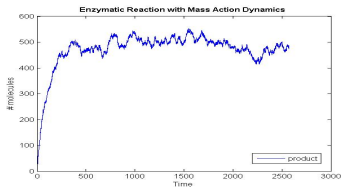
$$\text{mm_reaction} \left(\frac{k_2 + k_{-1}}{k_1}, k_2 \cdot E, S, P \right)$$

Enzymatic reaction



Mass Action Kinetics

```
enz_reaction(k1, k-1, k2, S, E, ES, P) :-
  reaction(k1, [S, E], [ES]) ||
  reaction(k-1, [ES], [E, S]) ||
  reaction(k2, [ES], [E, P])
```

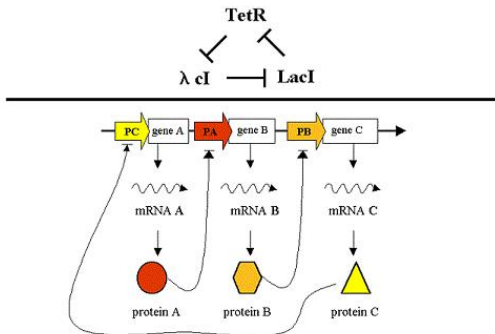


Michaelis-Menten Kinetics

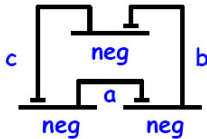
$$\text{mm_reaction} \left(\frac{k_2 + k_{-1}}{k_1}, k_2 \cdot E, S, P \right)$$

A paradigmatic example — the Repressilator

The Repressilator:
a cyclic, three-repressor, transcriptional network

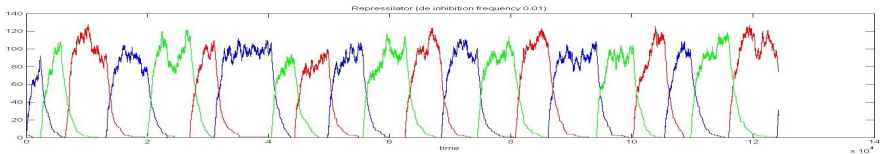


Repressilator in sCCP



$$\text{degr}(X) \text{ :- tell}_{[\lambda_D \cdot X]}(X = X - 1).\text{degr}(X)$$

$$\begin{aligned} \text{neg}(X, Y) \text{ :- } & (\text{tell}_{[\lambda_P]}(X = X + 1) \\ & + \text{ask}_{[\lambda_I \cdot Y]}(Y \geq 1).\text{ask}_{[\lambda_U]}(\text{true})).\text{neg}(X, Y) \\ &).\text{neg}(X, Y) \end{aligned}$$

$$\text{neg}(A, C) \parallel \text{neg}(B, A) \parallel \text{neg}(C, B) \parallel \text{degr}(A) \parallel \text{degr}(B) \parallel \text{degr}(C)$$


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From ODE to sCCP

What?

We want to associate a set an sCCP program to a set of ODEs.

Why?

- To unveil the logical patterns of interactions hidden in the ODEs.
- To study **expressivity** in terms of **representable behaviors** of sCCP.

How?

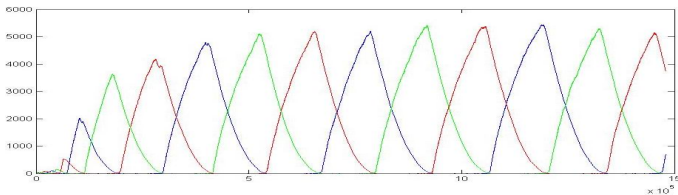
Using the functional form of rates of sCCP.

Example: encoding S-System's Repressilator

$$\begin{aligned}\dot{X}_1 &= \alpha_1 X_3^{-1} - \beta_1 X_1^{0.5}, \\ \dot{X}_2 &= \alpha_2 X_1^{-1} - \beta_2 X_2^{0.5}, \\ \dot{X}_3 &= \alpha_3 X_2^{-1} - \beta_3 X_3^{0.5}, \\ \alpha_i &= 0.2, \quad \beta_i = 1.\end{aligned}$$

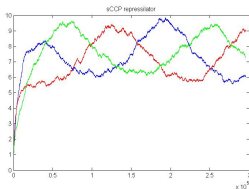
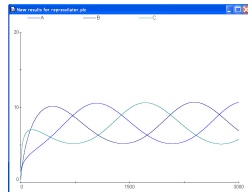
```
subs(X1) ::= (tell(X1 = X1 +  $\sigma$ )[ $\alpha_1 X_3^{-1}$ ]
+ tell(X1 = X1 -  $\sigma$ )[ $\beta_1 X_1^{0.5}$ ]).subs(X1)
subs(X2) ::= (tell(X2 = X2 +  $\sigma$ )[ $\alpha_2 X_1^{-1}$ ]
+ tell(X2 = X2 -  $\sigma$ )[ $\beta_2 X_2^{0.5}$ ]).subs(X2)
subs(X3) ::= (tell(X3 = X3 +  $\sigma$ )[ $\alpha_3 X_2^{-1}$ ]
+ tell(X3 = X3 -  $\sigma$ )[ $\beta_3 X_3^{0.5}$ ]).subs(X3)
```

$\sigma = 1$

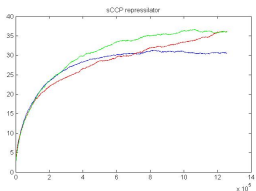
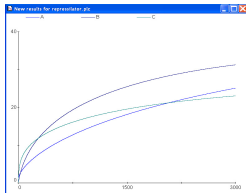


Repressilator gone wild

S-System's model of repressilator suffers from an **high sensitivity from parameters**, differently from the usual PA models. sCCP model with variable rates has the same “wild” behaviour!

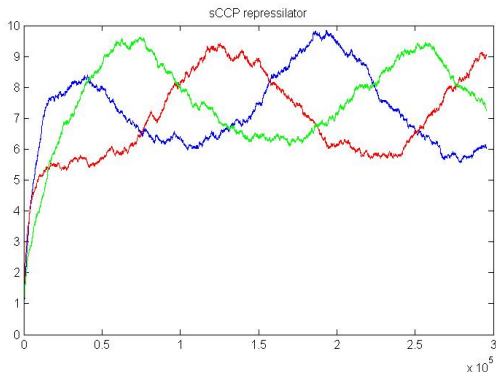


$$\beta_i = 0.01$$



$$\beta_i = 0.001$$

There's a trick...



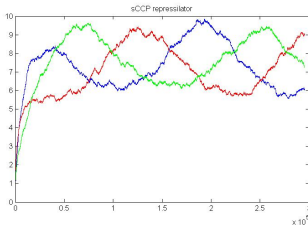
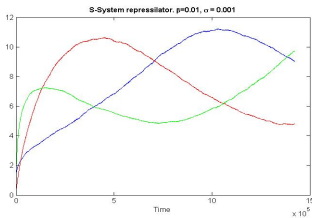
The magnitude
of fluctuations
is small.

We used

$$\sigma = 0.01$$

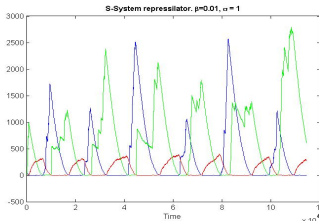
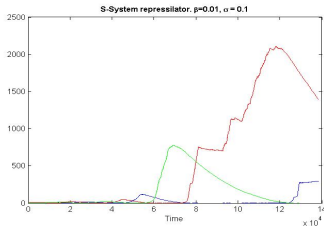
In this way we can reduce the perturbation effect of stochastic fluctuations.

Dependency on the step size σ



$\sigma = 0.001$

$\sigma = 0.01$



$\sigma = 0.1$

$\sigma = 1$

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From sCCP to ODE

What?

We want to associate a set of ODE to an sCCP program (written with a restricted syntax).

Why?

ODE can be numerically simulated faster than stochastic processes.

On the market...

There are (syntactic) methods to write set of ODEs for PEPA and stochastic π -calculus, looking at the speed of creation and destruction of terms. We did analogously for sCCP.

However, the ODE can show a behavior different from that of SPA models.

J. Hillston, Fluid Flow Approximation of PEPA models, *QEST*, 2005.

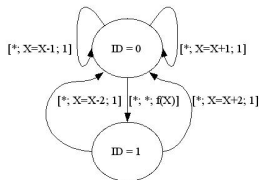
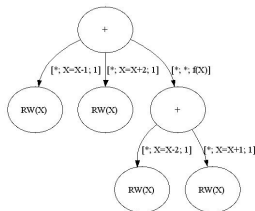
L. Cardelli, From Processes to ODEs by Chemistry, 2006.

L. Bortolussi, A. Policriti. Connecting Process Algebras and Differential Equations for systems biology, 2006.

From sCCP to ODE: example

RW_X :-
 $\text{tell}_1(X = X - 1).RW_X$
 $+ \text{tell}_1(X = X + 2).RW_X$
 $+ \text{ask}_{f(X)}(\text{true}).(\text{tell}_1(X = X - 2).RW_X$
 $\quad + \text{tell}_1(X = X + 1).RW_X)$

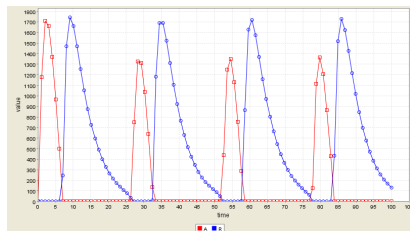
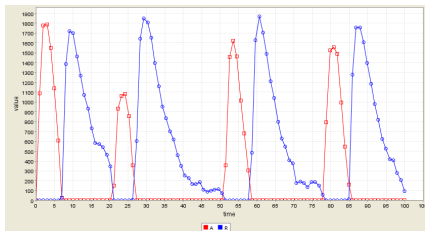
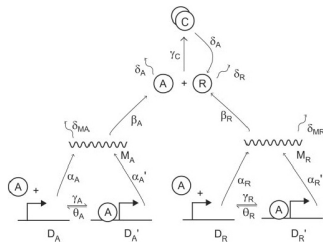
$$f(X) = \frac{1}{X^2+1}$$



X	-1	+2	0	-2	+1
P_0	0	0	-1	+1	+1
P_1	0	0	+1	-1	-1

$$\begin{cases} \dot{X} = P_0 - P_1 \\ \dot{P}_0 = -\frac{1}{X^2+1}P_0 + 2P_1 \\ \dot{P}_1 = \frac{1}{X^2+1}P_0 - 2P_1 \end{cases}$$

A “yes” case: the circadian clock



Preservation of chemical kinetics

Mass Action Kinetics

$\text{reaction}(k, [R_1, \dots, R_n], [P_1, \dots, P_m]) : -$
 $\text{ask}_{r_{MA}(k, R_1, \dots, R_n)} (\bigwedge_{i=1}^n (R_i > 0)) .$
 $(\parallel_{j=1}^m \text{tell}_{\infty} (R_j = R_j - 1) \parallel$
 $\parallel_{j=1}^m \text{tell}_{\infty} (P_j = P_j + 1)).$
 $\text{reaction}(k, [R_1, \dots, R_n], [P_1, \dots, P_m])$

$r_{MA}(k, X_1, \dots, X_n) = k \cdot X_1 \cdots X_n$

$$\dot{R}_1 = -kR_1 \cdots R_n$$

$$\vdots$$

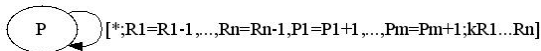
$$\dot{R}_n = -kR_1 \cdots R_n$$

$$\dot{P}_1 = kR_1 \cdots R_n$$

$$\vdots$$

$$\dot{P}_m = kR_1 \cdots R_n$$

$$\dot{P} = 0$$



Preservation of chemical kinetics

Michaelis-Menten Kinetics

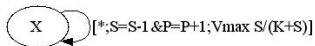
```
mm_reaction(K, V0, S, P) : -
  askrMM(K, V0, S)(S > 0).
  (tell∞(S = S - 1) || tell∞(P = P + 1)).
  mm_reaction(K, V0, S, P)
```

$$\dot{P} = \frac{v_{\max} S}{K+S}$$

$$\dot{X} = 0$$

$$\dot{S} = -\frac{v_{\max} S}{K+S}$$

$$r_{MM}(K, V_0, S) = \frac{V_0 S}{S+K}$$



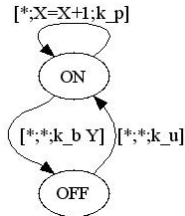
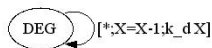
Back on repressilator

$\text{degr}(X) :- \text{tell}_{[k_d \cdot X]}(X = X - 1).\text{degr}(X)$

$\text{neg}(X, Y) :- (\text{tell}_{[k_p]}(X = X + 1)$
 $\quad + \text{ask}_{[k_b \cdot Y]}(Y \geq 1).\text{ask}_{[k_u]}(\text{true})$
 $\quad).\text{neg}(X, Y)$

$\text{neg}(A, C) \parallel \text{neg}(B, A) \parallel \text{neg}(C, B) \parallel$

$\text{degr}(A) \parallel \text{degr}(B) \parallel \text{degr}(C)$



$$\dot{A} = k_p Y_1 - k_d X_1$$

$$\dot{B} = k_p Y_2 - k_d X_2$$

$$\dot{C} = k_p Y_3 - k_d X_3$$

$$\dot{Y}_1 = k_u Z_1 - k_b Y_1 C$$

$$\dot{Y}_2 = k_u Z_2 - k_b Y_2 A$$

$$\dot{Y}_3 = k_u Z_3 - k_b Y_3 B$$

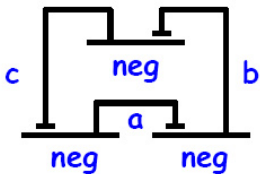
$$\dot{Z}_1 = k_b Y_1 C - k_u Z_1$$

$$\dot{Z}_2 = k_b Y_2 A - k_u Z_2$$

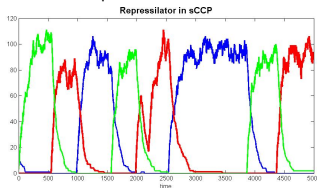
$$\dot{Z}_3 = k_b Y_3 B - k_u Z_3$$

The strange beast of repressilator

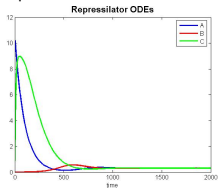
Repressilator with gene gates



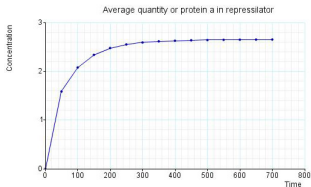
Repressilator in sCCP



Repressilator: ODE from sCCP



Repressilator: average of sCCP model



Conclusions

- We defined transformations from sCCP to ODE and back, preserving rate semantics.
- Dynamic behavior is not always conserved, the quest for behaviorally invariant translations (for a wider class of models) is still open.
- The concept of “behavioral equivalence” needs to be treated more formally.