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Automatic reduction of stochastic rules-based models in a nutshell

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Overview

- 1. Context and motivations
- 2. Information flow
- 3. Symmetric sites
- 4. Generic framework
- 5. Conclusion

Signalling Pathways



Eikuch, 2007

Rule-based approach

We use site graph rewrite systems



- 1. The description level matches with both
 - the observation level
 - and the intervention level

of the biologist.

We can tune the model easily.

- 2. Model description is very compact.
- 3. Quantitative semantics can be defined.

Complexity walls



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A model with ubiquitination



Statistical independence

We check numerically that:



Reduced model





 ${}^{*}\mathsf{P} \xrightarrow{k_{3}} \emptyset$ + side effect: remove one P

$$\xrightarrow{k_4} \emptyset$$

+ side effect: remove one P

P*

Comparison between the two models



Coupled semi-reactions



$$A \stackrel{k_{A+}}{\underset{k_{A-}}{\longrightarrow}} A^{\star}, AB \stackrel{k_{A+}}{\underset{k_{A-}}{\longrightarrow}} A^{\star}B, AB^{\star} \stackrel{k_{A+}}{\underset{k_{A-}}{\longrightarrow}} A^{\star}B^{\star}$$



$$B \stackrel{k_{B+}}{\underset{k_{B-}}{\overset{}}} B^{\star}, AB \stackrel{k_{B+}}{\underset{k_{B-}}{\overset{}}} AB^{\star}, A^{\star}B \stackrel{k_{B+}}{\underset{k_{B-}}{\overset{}}} A^{\star}B^{\star}$$



$$A + B \xleftarrow[k_{AB}]{k_{A.B}} AB, \quad A^{\star} + B \xleftarrow[k_{AB}]{k_{A.B}} A^{\star}B,$$
$$A + B^{\star} \xleftarrow[k_{AB}]{k_{A.B}} AB^{\star}, \quad A^{\star} + B^{\star} \xleftarrow[k_{A^{\star}B^{\star}}]{k_{A.B}} A^{\star}B^{\star}$$

Reduced model



$$A \stackrel{k_{A+}}{\underset{k_{A-}}{\longleftarrow}} A^{\star}, \quad AB^{\diamond} \stackrel{k_{A+}}{\underset{k_{A-}}{\longleftarrow}} A^{\star}B^{\diamond},$$



$$\mathsf{B} \stackrel{k_{\mathsf{B}+}}{\underbrace{}_{k_{\mathsf{B}-}}} \mathsf{B}^{\star}, \quad \mathsf{A}^{\diamond}\mathsf{B} \stackrel{k_{\mathsf{B}+}}{\underbrace{}_{k_{\mathsf{B}-}}} \mathsf{A}^{\diamond}\mathsf{B}^{\star},$$



$$A + B \xrightarrow{k_{AB}} AB^{\diamond} + A^{\diamond}B,$$

$$A^{\star} + B \xrightarrow{k_{AB}} A^{\star}B^{\diamond} + A^{\diamond}B,$$

$$A^{\star} + B \xrightarrow{k_{AB}} A^{\star}B^{\diamond} + A^{\diamond}B,$$

$$A + B^{\star} \xrightarrow{k_{AB}} AB^{\diamond} + A^{\diamond}B^{\star},$$

$$A + B^{\star} \xrightarrow{k_{AB}} AB^{\diamond} + A^{\diamond}B^{\star},$$

$$A^{\star} + B^{\star} \xrightarrow{k_{A*B^{\star}}} AB^{\diamond} + A^{\diamond}B^{\star},$$

$$A^{\star} + B^{\star} \xrightarrow{k_{A*B^{\star}}} A^{\star}B^{\diamond} + A^{\diamond}B^{\star},$$

Comparison between the two models



Although the reduction is correct in the ODE semantics.

Degree of correlation (in the unreduced model)



Distant control









 $A + A^{\star} \xrightarrow{k_+} A_{\star} + A^{\star}$ $A^{\star} + A^{\star} \xrightarrow{k_{+}} A^{\star}_{\star} + A^{\star}$ $\mathsf{A} + \mathsf{A}_{\star}^{\star} \xrightarrow{k_{+}} \mathsf{A}_{\star} + \mathsf{A}_{\star}^{\star}$ $\mathsf{A}^{\star} + \mathsf{A}^{\star}_{\star} \xrightarrow{\mathsf{k}_{+}} \mathsf{A}^{\star}_{\star} + \mathsf{A}^{\star}_{\star}$

 $\begin{array}{cccc} \mathsf{A}^{\star}_{\star} & \xrightarrow{k_{-}} & \mathsf{A}^{\star} \\ \mathsf{A}_{\star} & \xrightarrow{k_{-}} & \mathsf{A} \end{array}$

Jérôme Feret

Reduced model







 k_{-}

$$\mathsf{A} + \mathsf{A}^{\star} \xrightarrow{\mathsf{k}_{+}} \mathsf{A}_{\star} + \mathsf{A}^{\star}$$



Jérôme Feret

Comparison between the two models



Degree of correlation (in the unreduced model)



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A model with symmetries







 ${}^{\star}\mathsf{P}^{\star} \xrightarrow{k_2} \emptyset$

Degree of correlation (in the unreduced model)



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Equivalent chemical species

We check numerically that:

 $\mathsf{E}_{\mathsf{t}}(\mathsf{n}_{\mathsf{P}^{\star}}) = \mathsf{E}_{\mathsf{t}}(\mathsf{n}_{\mathsf{\star}_{\mathsf{P}}}).$



and two instances of P at time t = 0.

Reduced model



Exponential reduction!!!

Comparison between the two models



and two instances of P at time t = 0.

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Weighted Labelled Transition Systems

A weighted-labelled transition system \mathcal{W} is given by:

- Q, a countable set of states;
- *L*, a set of labels;
- $w: \mathcal{Q} \times \mathcal{L} \times \mathcal{Q} \rightarrow \mathbb{R}^+_0$, a weight function;
- $\pi_0: \mathcal{Q} \to [0, 1]$, an initial probability distribution.

We also assume that:

- the system is finitely branching, i.e.:
 - the set $\{q \in \mathcal{Q} \mid \pi_0(q) > 0\}$ is finite
 - and, for any $q \in Q$, the set $\{l, q' \in \mathcal{L} \times Q \mid w(q, l, q') > 0\}$ is finite.
- the system is deterministic:

if $w(q, \lambda, q_1) > 0$ and $w(q, \lambda, q_2) > 0$, then: $q_1 = q_2$.

Trace distribution

A cylinder set of traces is defined as:

$$\tau \stackrel{\Delta}{=} q_0 \stackrel{\lambda_1, I_1}{\rightarrow} q_1 \dots q_{k-1} \stackrel{\lambda_k, I_k}{\rightarrow} q_k$$

where:

- $(q_i)_{0 \leq i \leq k} \in \mathcal{Q}^{k+1}$ and $(\lambda_i)_{1 \leq i \leq k} \in \mathcal{L}^k$,
- $(I_i)_{1 \le i \le k}$ is a family of open intervals in \mathbb{R}^+_0 .

The probability of a cylinder set of traces is defined as follows:

$$\begin{aligned} \mathcal{P}r(\tau) &\stackrel{\Delta}{=} \pi_0(q_0) \prod_{i=1}^k \frac{w(q_{i-1}, l_i, q_i)}{\alpha(q_{i-1})} \left(e^{-\alpha(q_{i-1}) \cdot \text{inf}(I_i)} - e^{-\alpha(q_{i-1}) \cdot \text{sup}(I_i)} \right), \end{aligned}$$
where $\alpha(q) \stackrel{\Delta}{=} \sum_{\lambda, q'} w(q, \lambda, q').$

Abstraction between WLTS



Soundness

Given:

- two WLTS $\mathcal{S} \stackrel{\Delta}{=} (\mathcal{Q}, \mathcal{L}, \rightarrow, w, \mathcal{I}, \pi_0)$ and $\mathcal{S}^{\sharp} \stackrel{\Delta}{=} (\mathcal{Q}^{\sharp}, \mathcal{L}^{\sharp}, \rightsquigarrow, w^{\sharp}, \mathcal{I}^{\sharp}, \pi_0^{\sharp})$,
- two abstraction functions $\beta^{\mathcal{Q}}: \mathcal{Q} \to \mathcal{Q}^{\sharp}$ and $\beta^{\mathcal{L}}: \mathcal{L} \to \mathcal{L}^{\sharp}$,

 S^{\sharp} is a sound abstraction of S, if and only if, for any cylinder set τ of traces of S, we have:

$$\mathcal{P}r(\beta^{\mathbb{T}}(\tau)) = \sum_{\tau'} (\mathcal{P}r(\tau') \mid \beta^{\mathbb{T}}(\tau) = \beta^{\mathbb{T}}(\tau')),$$

where,

$$\beta^{\mathbb{T}}(q_{0} \stackrel{\lambda_{1}, I_{1}}{\to} q_{1} \dots q_{k-1} \stackrel{\lambda_{k}, I_{k}}{\to} q_{k}) \\ \stackrel{\Delta}{=} \beta^{\mathcal{Q}}(q_{0}) \stackrel{\beta^{\mathcal{L}}(\lambda_{1}), I_{1}}{\to} \beta^{\mathcal{Q}}(q_{1}) \dots \beta^{\mathcal{Q}}(q_{k-1}) \stackrel{\beta^{\mathcal{L}}(\lambda_{k}), I_{k}}{\to} \beta^{\mathcal{Q}}(q_{k}).$$

Backward bisimulation

Let $\sim_{\mathcal{Q}}$ be an equivalence relation over \mathcal{Q} and $\sim_{\mathcal{L}}$ be an equivalence relation over \mathcal{L} .

We say that $(\sim_{\mathcal{Q}}, \sim_{\mathcal{L}})$ is a backward bisimulation, if and only if, there exists $\gamma : \mathcal{Q} \to \mathbb{R}^+$, such that: for any $q'_1, q'_2 \in \mathcal{Q}$ which satisfies $q'_1 \sim_{\mathcal{Q}} q'_2$:

• $a(q'_1) = a(q'_2);$





Bisimulation algebra

Backward bisimulations can be:

• composed:



• factored:

• combined with a symmetric product (c.f. lub or pushout):



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Conclusion

- A framework for reducing stochastic rule-based models.
 - We use:
 - * the sites the state of which are uncorrelated;
 - * the sites having the same capabilities of interactions.
 - Algebraic operators combine these abstractions.
- We use backward bisimulations in order to prove statistical invariants, we use them to reduce the dimension of the continuous-time Markov chains.

Future works

- Forward bisimulations are very convenient to abstract ODE semantics.
 ⇒investigate the use of hybrid bisimulation.
- Propose approximated simulation algorithms to approximate different scale rate reactions.
 - hybrid systems,
 - tau-leaping,
 - **-** . . .