# sybila 🗖

### Parameter Synthesis and Robustness Analysis of Rule-Based Models

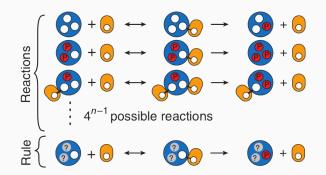
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- **systems biology** understand the (biological) system of interest by building a model
- modelling dominated by *explicit* model formats such as ODEs, reaction-based systems, Petri nets, ...
- hard to organise, develop, and edit (e.g. extend by new information)
- scalability of the description

# **Rule-based modelling**

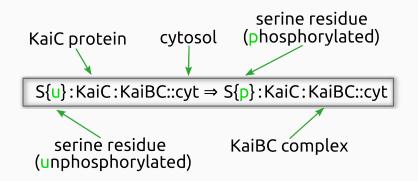
- compact extension of reaction-based systems
- suitable for the description of biological systems since the context/details are often not known



• there exist several representatives (Kappa, BNGL, Chromar, etc.)

# **BioChemical Space Language (BCSL)**

- a high-level rule-based language for biochemical interactions
- focus on human-readability while preserving formal aspects
- part of an annotation format for Comprehensive Modelling Platform



# **BCSL** – examples of rules

### • State change

- 1.  $S{u}::cell \Rightarrow S{p}::cell$
- 2.  $E.S\{u\}$ ::cell  $\Rightarrow E.S\{p\}$ ::cell
- 3.  $R(active{off})::cyt \Rightarrow R(active{on})::cyt$

# • Complex formation

- 4. E::cell +  $S{u}$ ::cell  $\Rightarrow$  E.S ${u}$ ::cell
- 5.  $\alpha.\alpha::$ out +  $\beta.\beta::$ out  $\Rightarrow \alpha.\alpha.\beta.\beta::$ out

### • Transport

- 6. Prot::cell  $\Rightarrow$  Prot::out
- 7.  $\Rightarrow$  mRNA::nuc

- Zooming inside the structure act{off}:A():A().B()::cell ⇒ act{on}:A():A().B()::cell
- Define an alias for complex

AB = A() . B() AC = A() . C()AD = A() . D()

• Usage of a variable

 $act{off}:A():?::cell \Rightarrow act{on}:A():?::cell ; ? = {AB, AC, AD}$ 

- extension by quantitative dynamical aspects (in this paper)
- associating rules with rational rate functions
- quantifies the rule rate how likely it is

 $S{u}::cell \Rightarrow S{p}::cell @ k1 \times [S{u}::cell]$ 

```
#! rules
T(P{i})::cyt ⇒ T(P{m})::cyt @ a1×[T(P{i})::cyt]
T(P{m})::cyt ⇒ 2 T(P{i})::cyt @ a2×[T(P{m})::cyt]
T(P{i})::cyt ⇒ @ d1×[T(P{i})::cyt]
T(P{m})::cyt ⇒ @ d2×[T(P{m})::cyt]
```

```
#! inits
2 T(P{m})::cyt
1 T(P{i})::cyt
#! definitions
```

```
a2 = 0.5
```

```
d1 = 0.3
```

- a rule can be applied to a state by match-replace pair of actions
  - 1. appropriate agents are *matched* according to LHS of rule and rate function is evaluated
  - 2. matched agents are modified according to the RHS
  - 3. original agents are *replaced* by the modified ones
- given an initial state of the model, by transitive application of rules we obtain a labelled transition system
  - by normalisation of transition labels, the semantics of BCSL model is a (parametrised) Discrete Time Markov Chain

• probabilistic semantics  $\Rightarrow$  behavioural properties expressed in **Probabilistic CTL** (PCTL) P => 0.3 [F S{u}::cell > 0]

$$P = ? \qquad [F S{u}::cell > 0]$$

- model checking and local robustness in non-parametric case
- parameter synthesis and global robustness in parametric case
  - for PCTL analysis used external tool Storm model checker method based on state elimination (finite state automaton → regular expression)

- $P \bowtie n \ [\varphi]$  with  $\bowtie \in \{\leq, <, >, \geq\}$  and  $n \in [0, 1]$  $\rightarrow$  boolean answer
- $P = ? [\varphi]$

 $\rightarrow$  probability of satisfaction of path formula  $\varphi$ 

•  $P \bowtie n \ [\varphi]$  with  $\bowtie \in \{\leq, <, >, \geq\}$  and  $n \in [0, 1]$ 

 $\rightarrow$  partitioning of given parameter space (allowed intervals for unknown parameters) to SAT, UNSAT, and UNKNOWN regions (with defined precision)

•  $P = ? [\varphi]$ 

 $\rightarrow$  satisfiability function of parameters

(its evaluation for particular parametrisation – point in parameter space – provides the probability that the model satisfies path formula  $\varphi$ )

The global robustness of a system s is defined as

$$R^{s}_{a,P} = \int_{P} \psi(p) D^{s}_{a}(p) dp$$

- *a* is the property of the system under scrutiny,
- P is the set of all perturbations,
- $\psi(p)$  is the probability of the perturbation p,
- *local robustness*  $D_a^s(p)$  is a measure stating how much the property *a* is preserved in perturbation *p*.

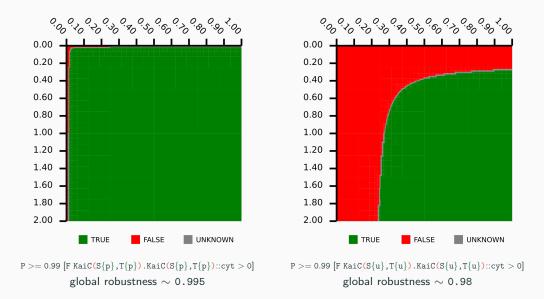
For a BCSL model  $\mathcal{M}$ , the global robustness can be computed as

$${\sf R}^{\mathcal{M}}_{\phi, oldsymbol{P}} = \int_{oldsymbol{P}} rac{1}{|oldsymbol{P}|} { t f({ t p})} { t d}{ t p}$$

- $\phi$  is the property (with the outermost operator P = ?),
- **P** is the parameter space,
- uniform probability  $\psi(\mathbf{p}) = \frac{1}{|\mathbf{P}|}$  of parametrisation p,
- the *local robustness* can be computed as evaluation of satisfiability function of parameters f(p) (obtained from parameter synthesis)

- an adopted (and simplified) version of Miyoshi et al. 2007 model describing circadian rhythms in cyanobacteria
- typical oscillatory behaviour of the model performed by repeated phosphorylation and de-phosphorylation of a protein complex KaiC
- two experiments to validate oscillatory behaviour of key product
  - 1. reach a *phosphorylated* form from an *unphosphorylated* one with **high probability**
  - 2. vice versa

#### Case studies – circadian clock



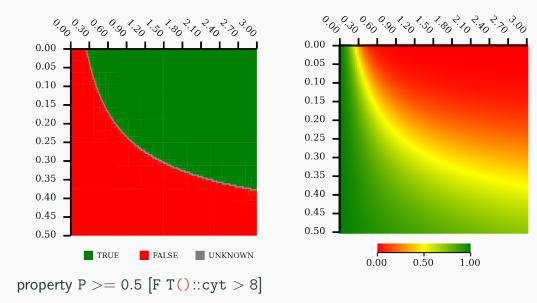
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#### Case studies – tumour growth



- BioChemical Space Language with methods for exact analysis of probabilistic behaviour
- future steps are to focus on scalability issues avoid explicit transition system enumeration using symbolic approaches and static analysis
- tool eBCSgen (submitted to CMSB'20)
  - available at https://biodivine-vm.fi.muni.cz/galaxy
  - all case-study data available in shared dataset https://biodivine-vm.fi.muni.cz/galaxy/u/xtrojak/h/ebcsgen-nfm-2020