# Petri Nets for Systems and Synthetic Biology

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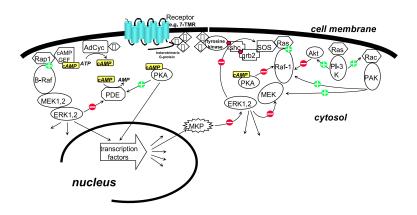
joint work with David Gilbert, Robin Donaldson Bioinformatics Research Centre, University of Glasgow

Workshop on Computational Models for Cell Processes Turku, May 27, 2008



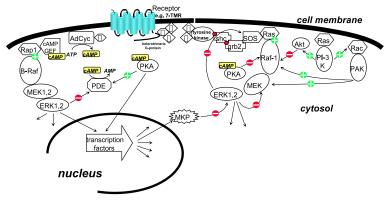
## Biochemical Networks

• ... are networks of (bio-) chemical reactions



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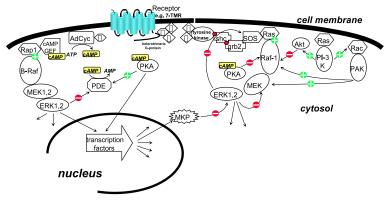
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How to model this?

## Biochemical Networks

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How to model this?

How to analyse this?



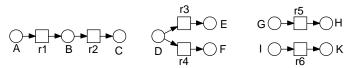
# Biochemical Networks, Three Basic Properties

• bipartite - species & reactions :  $r: 2H_2 + O_2 \rightarrow 2H_2O$ 

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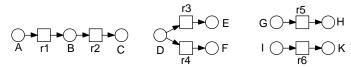
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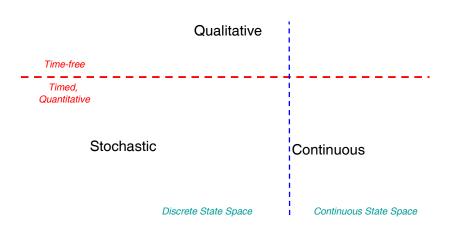
behaviour - stochastic

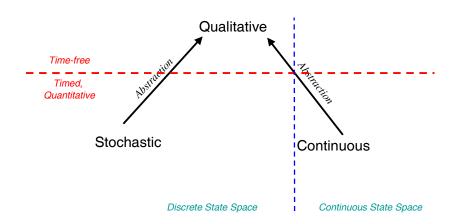


Qualitative

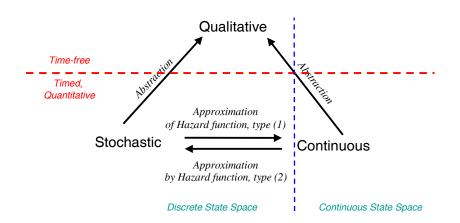
Stochastic

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#### Definition:

A place/transition Petri net is a quadruple

 $\mathcal{PN} = (P, T, f, m_0)$ , where

• P, T - finite, non empty, disjoint sets (places, transitions)

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Interleaving Semantics: reachability graph / CTL, LTL

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- $v: T \to H$  (stochastic firing rate functions) with
  - $H := \bigcup_{t \in T} \left\{ h_t \mid h_t : \mathbb{N}_0^{|\bullet t|} \to \mathbb{R}^+ \right\}$
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Semantics: Continuous Time Markov Chain / CSL, PLTLc



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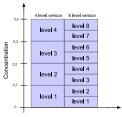
Semantics: ODEs / LTLc



## Discrete Petri nets

## Interpretation of tokens:

- tokens = molecules, moles
- tokens = concentration levels



## Specialised stochastic firing rate function, two examples :

• molecules semantics

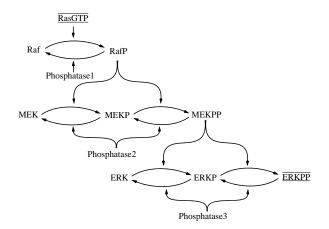
$$h_t := \frac{c_t}{c_t} \cdot \prod_{p \in {}^{\bullet}t} \binom{m(p)}{f(p,t)} \tag{1}$$

• concentration levels semantics

$$h_t := \frac{\mathbf{k_t}}{N} \cdot N \cdot \prod_{p \in {}^{\bullet}t} \left(\frac{m(p)}{N}\right) \tag{2}$$

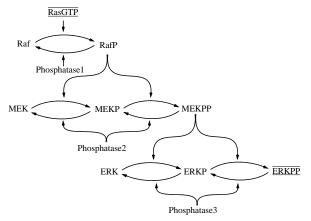
## Running Case Study

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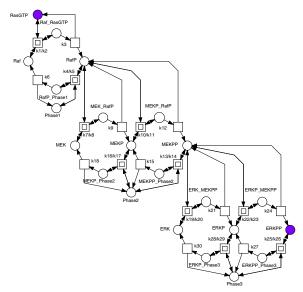
modelled in [Levchenko et al. 2000] like this ...

## Running Case Study - Origin

#### [Levchenko et al. 2000], Supplemental Material: ODEs

```
dRaf/dt
                              k_2 * Raf RasGTP + k_6 * RafP Phase1 - k_1 * Raf * RasGTP
         dRasGTP/dt
                               k2 * Raf RasGTP + k3 * Raf RasGTP - k1 * Raf * RasGTP
   dRaf RasGTP/dt
                               k<sub>1</sub> * Raf * RasGTP - k<sub>2</sub> * Raf RasGTP - k<sub>3</sub> * Raf RasGTP
            dRafP/dt
                               k3 * Raf RasGTP + k12 * MEKP RafP + k9 * MEK RafP+
                               k<sub>5</sub> * RafP Phase1 + k<sub>8</sub> * MEK RafP + k<sub>11</sub> * MEKP RafP -
                               k_7 * RafP * MEK - k_{10} * MEKP * RafP - k_4 * Phase1 * RafP
                               k4 * Phase1 * RafP - k5 * RafP Phase1 - k6 * RafP Phase1
   dRafP Phase1/dt
                               \mathbf{k_7} * \mathbf{RafP} * \mathbf{MEK} - \mathbf{k_8} * \mathbf{MEK} \quad \mathbf{RafP} - \mathbf{k_9} * \mathbf{MEK} \quad \mathbf{RafP}
     dMEK RafP/dt
   dMEKP RafP/dt
                               k_{10} * MEKP * RafP - k_{11} * MEKP RafP - k_{12} * MEKP RafP
 dMEKP Phase2/dt
                               k<sub>16</sub> * Phase2 * MEKP - k<sub>18</sub> * MEKP Phase2 - k<sub>17</sub> * MEKP Phase2
dMEKPP Phase2/dt
                               k_{13} * MEKPP * Phase2 - k_{15} * MEKPP Phase2 - k_{14} * MEKPP Phase
            dERK/dt
                              k_{20} * ERK MEKPP + k_{30} * ERKP Phase3 - k_{19} * MEKPP * ERK
  derk mekpp/dt
                               k_{19} * MEKPP * ERK - k_{20} * ERK MEKPP - k_{21} * ERK MEKPP
dERKP MEKPP/dt
                               k<sub>22</sub> * MEKPP * ERKP - k<sub>24</sub> * ERKP MEKPP - k<sub>23</sub> * ERKP MEKPP
               etcetera
```

# Running Case Study



• initial marking construction P-invariants

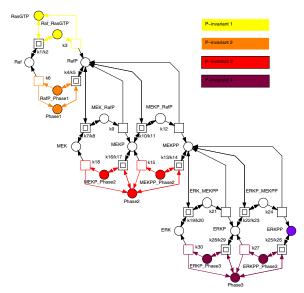
- initial marking construction
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  - P-invariants : token preserving modules (mass conservation)
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  - boundedness: every place gets finite token number only
  - liveness: every transition may happen forever
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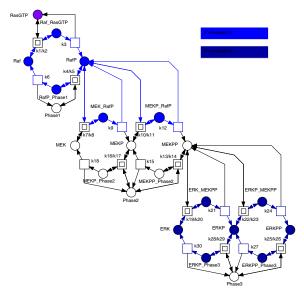
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- general behavioural properties
  - boundedness: every place gets finite token number only
  - liveness: every transition may happen forever
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- special behavioural properties
  - CTL / LTL model checking



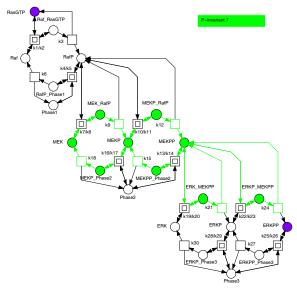
# Running Case Study - P-invariants



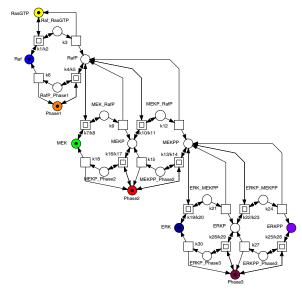
# Running Case Study - P-invariants



# Running Case Study - P-invariants



# Running Case Study - initial marking



# Running Case Study - general properties

#### • state space

levels	reachability graph	IDD data structure
	number of states	number of nodes
1	118	52
4	$2.4 \cdot 10^4$	115
8	$6.1\cdot 10^6$	269
80	$5.6 \cdot 10^{18}$	13,472
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- Covered by P-invariants (CPI) ⇒ bounded
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### Running Case Study - general properties

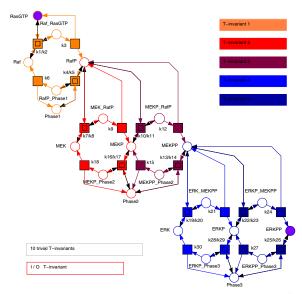
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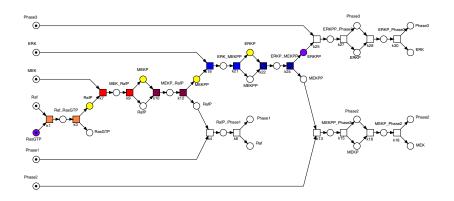
- Covered by P-invariants (CPI) ⇒ bounded
- Deadlock-Trap Property (DTP) holds ⇒ no dead states
- reachability graph
  - strongly connected ⇒ reversible
  - contains every transition (reaction) ⇒ live



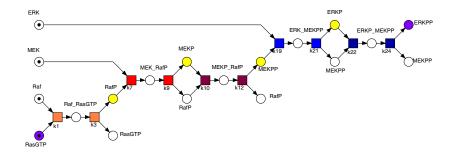
# Running Case Study - T-invariants



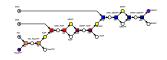
# Running Case Study - partial order run of I/O T-invariant



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# Qualitative Model Checking (CTL)



#### property Q1:

The signal sequence predicted by the partial order run of the I/O T-invariant is the only possible one; i.e., starting at the initial state, it is necessary to pass through RafP, MEKP, MEKPP and ERKP in order to reach ERKPP.

```
¬ [ E ( ¬ RafP U MEKP ) ∨
E ( ¬ MEKP U MEKPP ) ∨
E ( ¬ MEKPP U ERKP ) ∨
E ( ¬ ERKP U ERKPP ) ]
```

## Stochastic Model Checking - Preparation

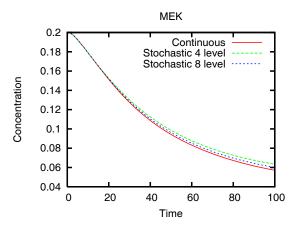
- isomorphy of reachability graph and CTMC, thus all qualitative properties still valid
- How many levels needed for quantitative evaluation?
  - state space(1 levels) = 118 (Boolean interpretation)
  - state space(4 levels) = 24,065
  - state space(8 levels) = 6,110,643
- equivalence check

$$C_{RafP}(t) = \frac{0.1}{s} \cdot \underbrace{\sum_{i=1}^{4s} (i \cdot P(L_{RafP}(t) = i))}_{expected \ value \ of \ L_{RafP}(t)}$$



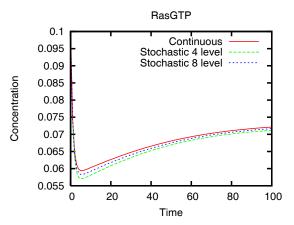
## Stochastic Model Checking - Preparation

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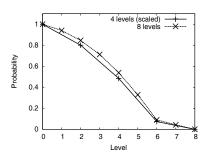


# Stochastic Model Checking (CSL)

#### property \$1:

What is the probability of the concentration of RafP increasing, when starting in a state where the level is already at L?

$$P_{=?} \; [\; (\; \mathsf{RafP} \; = \; \mathsf{L} \;) \; \mathsf{U}^{<=100} \; (\; \mathsf{RafP} > \mathsf{L} \;) \{\; \mathsf{RafP} = \mathsf{L} \;\} \;]$$

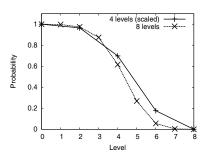


# Stochastic Model Checking (CSL)

#### property S2:

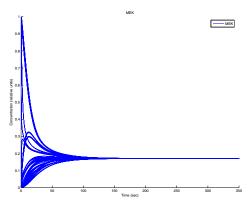
What is the probability that RafP is the first species to react?

$$\begin{array}{l} {\sf P}_{=?} \left[ \, \left( \, \left( \, {\sf MEKPP} \, = \, 0 \, \right) \, \wedge \, \left( \, {\sf ERKPP} \, = \, 0 \, \right) \, \right) \, {\sf U}^{<=100} \left( \, {\sf RafP} \, > \, L \, \right) \\ \left\{ \, \left( \, {\sf MEKPP} \, = \, 0 \, \right) \, \, \wedge \, \left( \, {\sf ERKPP} \, = \, 0 \, \right) \, \, \wedge \, \left( \, {\sf RafP} \, = \, 0 \, \right) \, \right\} \, \, \right] \end{array}$$



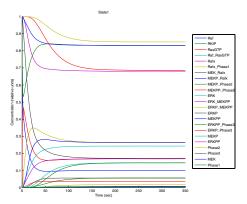
### Continuous Model Checking - Preparation

• steady state analysis, results for all 118 'good' states, e.g. for MEK:



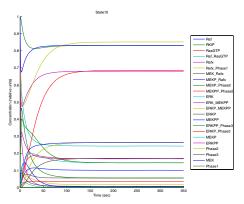
## Continuous Model Checking - Preparation

• steady state analysis for state 1:



## Continuous Model Checking - Preparation

• steady state analysis for state 10 :

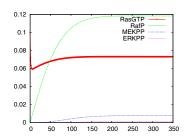


## Continuous Model Checking (LTLc)

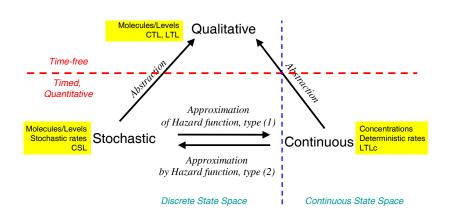
#### property C1:

The concentration of RafP rises to a significant level, while the concentrations of MEKPP and ERKPP remain close to zero; i.e. RafP is really the first species to react.

( (MEKPP 
$$<$$
 0.001)  $\land$  (ERKPP  $<$  0.0002) )  $\textbf{U}$  (RafP  $>$  0.06)



#### Framework



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  - MATLAB
  - BioNessie (Glasgow)
  - LTLc model checking: MC2(PLTLc) (Glasgow), BioCham



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- not bound to the Petri net perspective



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   stochastic and continuous behaviour may differ; why? when?
- sharing structure = sharing properties
   BUT, to which extend?
   relation: qualitative properties & continuous behaviour?



# Thanks!

- all data files and analysis results available at www-dssz.informatik.tu-cottbus.de/examples/levchenko
- M Heiner, D Gilbert, R Donaldson:
   Petri Nets for Systems and Synthetic Biology;
   SFM 2008, Springer LNCS 5016, pp. 215-264, 2008.
- laptop demonstration available

