

# Coloured Hybrid Petri Nets for Systems Biology

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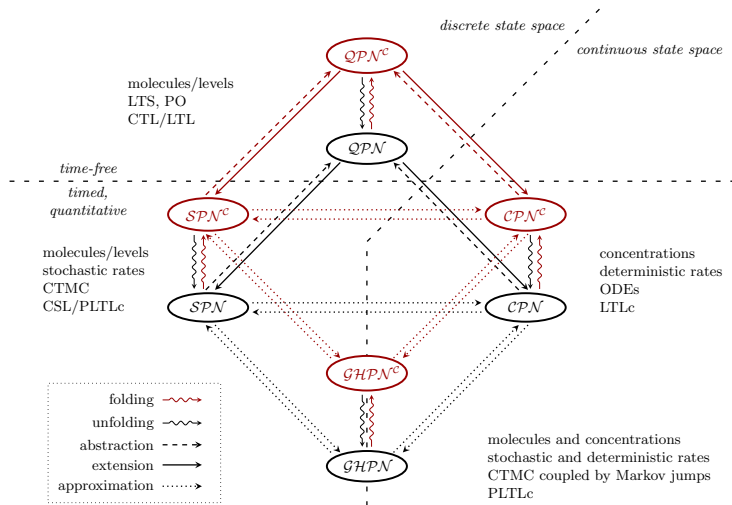
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# Agenda

- 1 Introduction
- 2  $GHPN^c$
- 3 Simulation of  $GHPN^c$
- 4 Case Study
- 5 Implementation
- 6 Conclusion

# Petri Nets in Snoopy<sup>1</sup>



<sup>1</sup>Heiner et al. 2012

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Coloured Hybrid Petri Nets for Systems Biology

# Motivations for $GHPN^c$

- The rapid change of the size of biological models
- Certain biological phenomena necessitates the existence of discrete and continuous variables as well as continuous and stochastic processes in one and the same model
- Coloured Petri nets can easily be used to model biological systems with repetition of components

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- Can easily be used to model scalable biological systems
- Combines both CPN and GSPN into one class
- Different transition types → different reaction types can be modelled using GHPN
- Stiff biochemical networks can be easily modelled and simulated using GHPN
- The final model can be simulated using either static or dynamic partitioning
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# Elements

**Places**

**Transitions**

**Arcs**

# Elements

## Places



Discrete



Continuous

## Transitions

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## Places



Discrete



Continuous

## Transitions



Stochastic



Continuous



Immediate



<1>

Deterministic



[.SimStart,1,.SimEnd]

Scheduled

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Standard



Read



Inhibitor



Equal



Reset



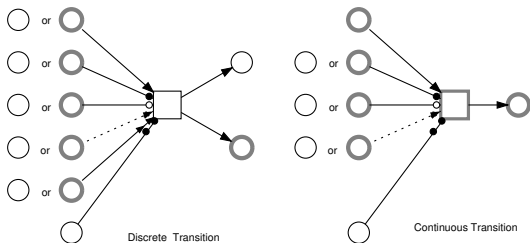
Modifier

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M. Herajy and M. Heiner, NAHS (2012)



# Connectivity



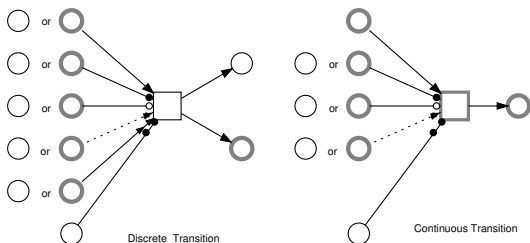
The restrictions are:

- discrete places cannot be connected with continuous transitions using standard arcs,
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- and continuous transitions cannot use reset arcs.

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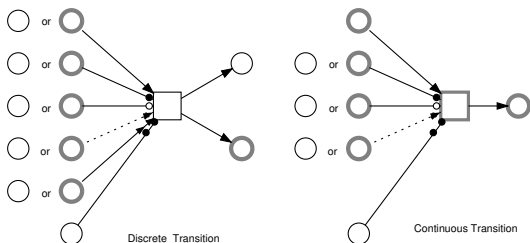
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# Simulation of $GHPN^c$

# Unfolding

- Coloured hybrid Petri nets can be unfolded into uncoloured ones where many analysis and simulation techniques can be applied.
- The conversion between uncoloured and coloured Petri nets changes the style of representation, but does not change the actual net structure of the underlying biological reaction network.

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# Unfolding (Cont.)

- If the colour set of each variable in a transition guard has a finite integer domain, a constraint satisfaction approach is used to obtain all valid transition instances.
- Otherwise, a general unfolding algorithm is adopted, in which some optimization techniques like partial binding partial test and pattern matching are used

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# Simulation of $\mathcal{GHPN}^c$

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- The partitioning of the net transitions into stochastic and continuous ones.
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# Simulation of $GHPN^C$ (Cont.)

- Static Partitioning
- Dynamic Partitioning

# Simulation of $\mathcal{GHPN}^c$ (Cont.)

- Static partitioning: partitioning is done off-line before the simulation starts.
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# Synchronization Mechanism

One option is to use (1)<sup>2</sup>:

$$g(\mathbf{x}) = \int_t^{t+\tau} a_0^s(\mathbf{x}) dt - \xi = 0, \quad (1)$$

where  $\xi$  is a random number exponentially distributed with a unit mean, and  $a_0^s(\mathbf{x})$  is the cumulative rate of all the stochastic transitions.

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<sup>2</sup>Gillespie 91, Markov Processes

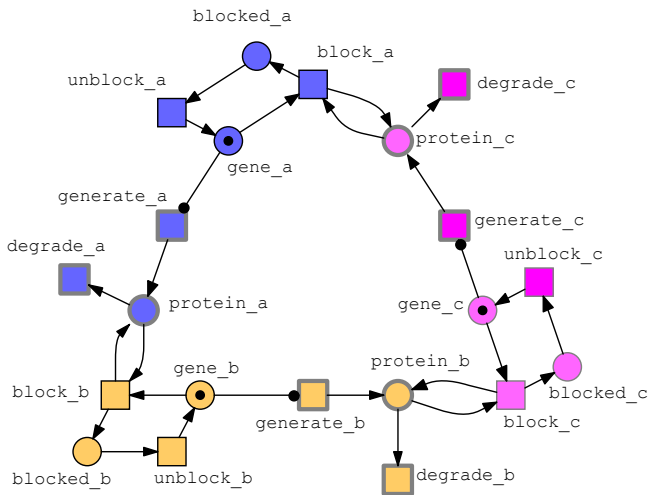
# Case Study

# Repressilator

- It is an example of a synthetic circuit.
- The repressilator system is a regulatory cycle of three genes (gene a, gene b and gene c).
- Each gene represses its successor, namely, gene a inhibits gene b, gene b inhibits gene c, and gene c inhibits gene a.
- This negative regulation is realized by the repressors, protein a, protein b and protein c.



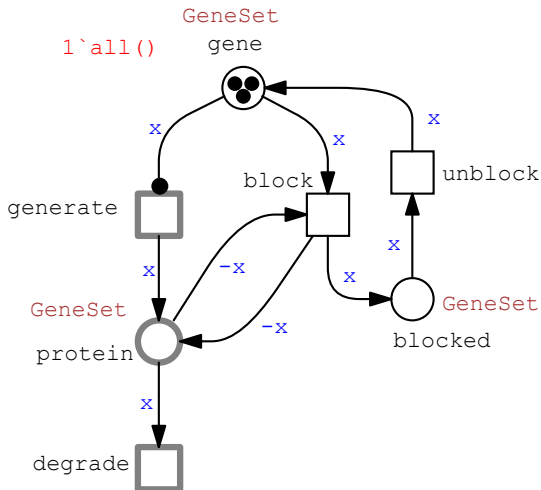
# GHPN Representation



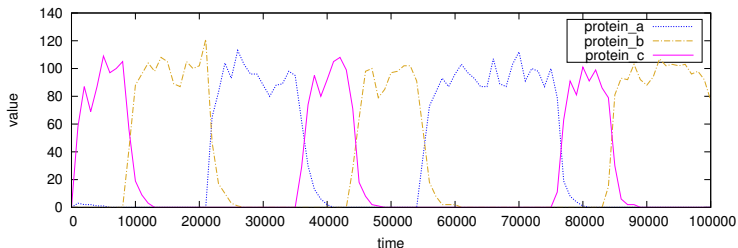
# $GHPN^c$ Net Partitioning

- The 1-bounded places as determined by P-invariant analysis and the related transitions as determined by T-invariant analysis are kept discrete.
- The unbounded places and related transitions are approximated by continuous places and transitions, respectively.

# GHPN<sup>C</sup> Representation



# $GHPN^c$ Simulation Result



# Implementation

All the features of  $\mathcal{GHPN}^c$  are implemented in:

- Snoopy – a unifying Petri net editing tool.
- $S^4$  – Snoopy Steering and Simulation Server

# Conclusion

- In this paper we have introduced a class of coloured Petri nets called Coloured Generalised Hybrid Petri Nets ( $GHPN^c$ ).
- $GHPN^c$  are particularly tailored to systems biologists needs to model and analyse multiscales models.
- $GHPN^c$  provide the interplay between stochastic and continuous regimes on the coloured level

# Future work

- Adding marking-dependent arc weights.
- Simulation of  $\mathcal{GHPN}^c$  on the coloured level
- Implementing more complex case studies.

# Acknowledgement

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# Thank You