Model-checking logical models of large regulatory networks

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1 Introduction

2 Reduction/NuSMV enconding

3 Application



4 Conclusions and Prospects

General motivation: study large biological networks



(Saez-Rodriguez et al., PLoS Comput. Biol. 2007)

- Signalling pathways, regulatory modules
- Lack of quantitative data
- ON/OFF mechanisms, thresholds
- \Rightarrow Discrete modelling

Discrete modelling: logical formalism (Thomas and d'Ari, Biological Feedback 1989)

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Logical regulatory graph (LRG) $\mathcal{R} = (\mathcal{G}, K)$

- $\mathcal{G} = \{g_i\}_{i=0,...,n}$ is a set of regulatory components
- $Max : \mathcal{G} \to \mathbb{N}^*$ associates a maximum level M_i to each component g_i
- $S = \prod_{g_i \in \mathcal{G}} D_i$: is the state space, where $D_i = \{0, \dots, Max(g_i)\}$
- $\forall g_i : K_i : S \rightarrow D_i$ is the regulatory function specifying the behaviour of g_i

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State transition graph (STG)

The dynamic behaviour of an LRG, is represented by an STG where:

- \blacksquare nodes are states in ${\mathcal S}$
- and arcs $(v, w) \in S^2$ denote transitions between states

Toy example (Boolean)



- $K_5(v) = 1$ if $v_1 = 1$
 - $\mathsf{K}_0(\mathsf{v}) = 1 \qquad \textit{if } \mathsf{v}_1 = 1 \lor \mathsf{v}_2 = 1$ $K_1(v) = 1$ if $v_0 = 1 \lor v_1 = 1 \lor v_2 = 1$ $K_2(v) = 1$ if $v_3 = 1$ $K_3(v) = input$ fixed or unconstrained $K_4(v) = 1$ if $v_0 = 1 \lor v_5 = 1$

Toy example (Boolean)



$$egin{aligned} & {\cal K}_0(v) = 1 \ & {\cal K}_1(v) = 1 \ & {\cal K}_2(v) = 1 \ & {\cal K}_3(v) = inpu \ & {\cal K}_4(v) = 1 \ & {\cal K}_5(v) = 1 \end{aligned}$$

 $\begin{array}{l} \mbox{if } v_1 = 1 \lor v_2 = 1 \\ \mbox{if } v_0 = 1 \lor v_1 = 1 \lor v_2 = 1 \\ \mbox{if } v_3 = 1 \\ \mbox{if } v_3 = 1 \\ \mbox{if } v_0 = 1 \lor v_5 = 1 \\ \mbox{if } v_1 = 1 \end{array}$

Interesting properties

- What are the attractors of the system? (stable states, complex attractors)
- Are these attractors reachable from initial conditions?
- Are these attractors maintained under input variations?
- ...

 $1st\ objective:\ automate\ model\ verification$

Confront model predictions with biological observations

1st objective: automate model verification

Confront model predictions with biological observations

Approach: use of formal verification techniques

Formal verification based on temporal logic and model checking provides a powerful technology to query models of interaction networks.

(Chabrier-Rivier et al., Theor Comput Sci 2004) (Batt et al., Bioinformatics 2005)

(Monteiro et al., Bioinformatics 2008)

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Model checking

Fully automated exhaustive exploration of the state space of the model.

- Transform models into a Kripke structure K = (S, AP, L, TR), where:
 - S are the states

(Kripke, Acta Phil. Fennica 1963)

- $TR \subseteq S \times S$ the transition relation between states
- $L: \overline{S} \to 2^{AP}$ a state labeling function, with a set of atomic propositions true in that state (values of variables, signs of derivatives, ...)
- Specify dynamical properties as statements in temporal logic that are interpreted on state transition graph.

(Emerson and Clarke, ICALP 1980)

2nd objective: ease dynamical analysis

Deal with the combinatorial explosion! \rightarrow define methods to **safely reduce** the state space

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Complete state transition graph has $2^6 = 64$ states

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Reduction?

Remove components: reduce complexity, control the dynamical impact

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Reduction?

Remove components: reduce complexity, control the dynamical impact

$Output \ reduction$

- No computation
- No impact
- Retrieve values

Extend to pseudo-outputs

- No impact
- Harder retrieval
- ⇒ Rewire the model: pseudo-outputs become outputs





Reduction of (pseudo-)outputs components

Implementation in GINsim

- Lossless reduction of (pseudo-)outputs
- Preservation of attractors and their reachability

Reduction of (pseudo-)outputs components

Implementation in GINsim

- Lossless reduction of (pseudo-)outputs
- Preservation of attractors and their reachability



- Generation of the STG without (pseudo-)output components
- Computation of all (pseudo-)output values on demand

Reduction of (pseudo-)outputs components

Implementation in NuSMV export

Effective representation for logical models:

- Symbolic model representation
- Combine different updating policies
- (Pseudo-)Outputs are:
 - Not part of the state description
 - \rightarrow reduction of the state-space
 - Defined as macros
 - \rightarrow computation of all (pseudo-)output values on demand

```
MDDULE main
VAR
properVar1 : { 0, 1 };
...
properVar1 : { 0, 1 };
ASSION
next(properVar1) :=
case
logicalRule1 : 1;
...
TRUE : 0;
esac;
...
```

```
DEFINE
outputVar :=
case
logicalRule1 : 1;
...
logicalRulej : 2;
TRUE : 0;
esac;
...
```

Reduction of inputs components

$\mathcal{K}_0(v) = 1$	if $v_1=1 \lor v_2=1$
$\mathcal{K}_1(v) = 1$	if $v_0 = 1 \lor v_1 = 1 \lor v_2 = 1$
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$\mathcal{K}_3(v) = input$	fixed or unconstrained



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STG with input reduction



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Kripke transition system
IVAR G3 : { 0, 1 };
VAR G0 : { 0, 1 }; G1 : { 0, 1 }; G2 : { 0, 1 };
(Müller-Olm et al., SAS 1999)

STG with input reduction



Reduction of inputs components: stable patterns

$Types \ of \ stable \ states$

- Strong stable state
- Weak stable state

Types of stable core ensembles

- Strong stable core ensemble
- Weak stable core ensemble



$$egin{array}{lll} \mathcal{K}_0(\mathbf{v}) = 1 & ext{if} \ \mathcal{K}_1(\mathbf{v}) = 1 & ext{if} \ \mathcal{K}_2(\mathbf{v}) = 1 & ext{if} \ \mathcal{K}_3(\mathbf{v}) = ext{input} & ext{fi} \end{array}$$

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Question: What's the impact of different switches of input conditions on the reachability of the biological attractors? and the system's behaviour?

Pseudo-inputs components not subject to reduction

Reduction of pseudo-inputs can cause reachability problems!



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Lost transitions



NuSMV export

Approach: Use a Kripke Transition System, representing information both:

- on states (core + pseudo-input components)
- on transitions (input components)

Advantages:

- Implicit representation of the model
- Reduction of (pseudo-)outputs by defining them as macros
- Projection of input components over transitions

In GINsim, input components remain (constant) part of state characterization

Considered temporal logics

- Computation Tree Logic (CTL)
 Verifying properties with all unconstrained inputs
- Action Restricted CTL (AR-CTL) (Pecheur and Raimondi, MoChArt 2006) Verifying properties with some (or all) fixed inputs

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Example of reachability properties

With unconstrained inputs (CTL):

```
INIT s000;
EF(s110);
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With fixed inputs (AR-CTL):

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TCR activation model

T-cell activation through TCR is a key part of the specific immune response



(Saez-Rodriguez, PLoS Comp. Biol. 2007)

LRG with 91 components

3 inputs, 14 fixed-inputs 35 core components

28 pseudo-outputs, 14 outputs

STG before reductions

 $2^3=8$ disconnected STGs $2^{88}=3\times 10^{26}$ states for each Size: 2.5×10^{27} states

STG after reductions

1 single compacted STG! Size: $2^{35} = 3.4 \times 10^{10}$ states

TCR activation model: structure of the dynamics

Approach

Impose combinations of fixed inputs (AR-CTL), to test reachability properties:

- From the initial state towards the attractors
- Between all the attractors

TCR activation model: structure of the dynamics

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- cd28: creates a separation on the state space
- tcrlig=0: system evolves towards a stable state
- tcrlig=1: system evolves towards a complex attractor
- cd4: augments the size of the complex attractor



⁽Sánchez et al, Intl J. Dev. Biol 2008)



(Sánchez et al, Intl J. Dev. Biol 2008)

Externa	l inputs													
Wg	Hh	Wg	Fz	Dsh	Slp	Nkd	En	Hh	Ci	Ciact	Cirep	Pka	Ptc	Letter code
0	0	0	0	0	0	1	0	0	1	0	1	2	1	T (trivial)
0	1	0	0	0	0	1	0	0	1	1	0	0	0	C (CiCiact)
0	1	2	1	1	1	2	0	0	1	2	0	0	0	W (Wg)
1	0	0	1	1	0	0	1	1	0	0	0	0	0	E (En)
1	0	0	1	1	1	2	0	0	1	1	0	2	2	N (Nkd)
1	1	2	1	1	1	2	0	0	1	2	0	0	0	w
1	1	0	1	1	0	0	1	1	0	0	0	0	0	E

Stabl	Stable patterns direct reachability: 4 fixed inputs																
	TT	TC	ΤN	CT	CC	CE	CN	EC	EE	EW	NT	NC	NN	NW	WE	WN	WW
TT																	
TC																	
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СТ																	
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Legen	nd:			∃inp	out cor	nbinat	ions, E	one	direct	path co	onnecti	ng two	states	5			
				∄ inp	out cor	nbinat	ions, =	one o	direct	path co	onnecti	ng two	states	5			

Stabl	Stable patterns direct reachability: 4 varying inputs																
	TT	TC	ΤN	СТ	CC	CE	CN	EC	EE	EW	NT	NC	NN	NW	WE	WN	WW
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Stabl	Stable patterns direct reachability: 4 varying inputs																
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				With	ı varyi	ng inpi	uts,∄	one pa	th cor	nectin	g two s	states					

- Reduction of the state space without loss of information
- Identification of WE and EW patterns as strong stable states

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In the NuSMV export

- Symbolic representation
 Profit from NuSMV internal OMDD representation
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Study of the structure of the system's dynamics

- Counterexample may contain information about necessary environmental conditions to ensure specific reachability properties
- Impact of input components on attractor switches:
 - Definition of strong/weak stable states
 - Definition of strong/weak stable core ensembles

Study of the structure of the system's dynamics

Automated uncovering of necessary conditions for attractor reachability

$Complex \ attractors$

- Efficient methods for complex attractor identification (without performing simulation)
- Complex attractor characterization in terms of strong/weak patterns

Thank you!

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Questions?!