



Synthesis of Boolean Networks from Biological Dynamical Constraints using Answer-Set Programming

Stéphanie Chevalier

LRI, CNRS, U. Paris-Sud
U. Paris-Saclay, France
stephanie.chevalier@lri.fr

Christine Froidevaux

LRI, CNRS, U. Paris-Sud
U. Paris-Saclay, France
christine.froidevaux@lri.fr

Loïc Paulevé

LaBRI, CNRS, U. Bordeaux
Bordeaux INP, France
loic.pauleve@labri.fr

Andrei Zinovyev

Institut Curie, INSERM
U. PSL, Mines ParisTech, France
andrei.zinovyev@curie.fr

Automatically design models
(Boolean networks)
from knowledge on a system
(structure & behaviors)

Issue: **logical synthesis of a function**
from constraints on its properties

Motivation:
biological applications

Boolean network (BN): a discrete dynamical system

A BN of dimension n

is a function $f = (f_1, \dots, f_n)$ with

$$\forall i \in \{1, \dots, n\}, f_i : \{0, 1\}^n \rightarrow \{0, 1\}$$

A configuration is a vector $x \in \{0, 1\}^n$

example: $f_1(x) := \neg x_2$

$$f_2(x) := \neg x_1$$

$$f_3(x) := \neg x_1 \wedge x_2$$

Boolean network (BN): a discrete dynamical system

A BN of dimension n

is a function $f = (f_1, \dots, f_n)$ with

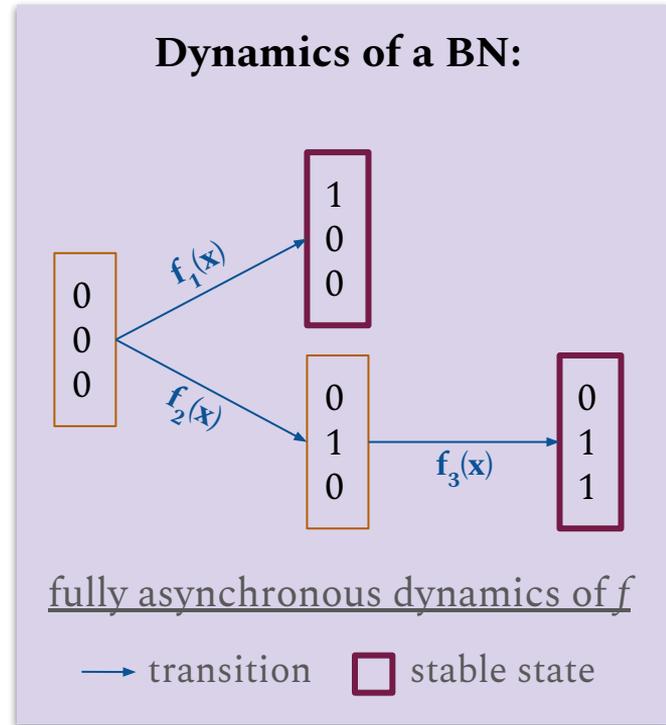
$$\forall i \in \{1, \dots, n\}, f_i : \{0, 1\}^n \rightarrow \{0, 1\}$$

A **configuration** is a vector $x \in \{0, 1\}^n$

example: $f_1(x) := \neg x_2$

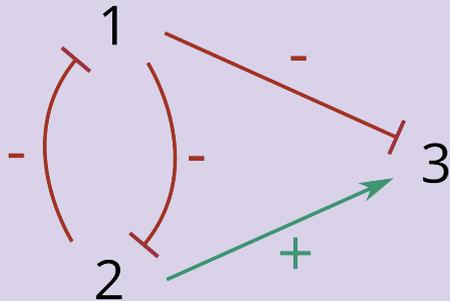
$$f_2(x) := \neg x_1$$

$$f_3(x) := \neg x_1 \wedge x_2$$

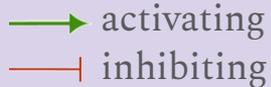


Boolean network (BN): a discrete dynamical system

Domain of BNs:

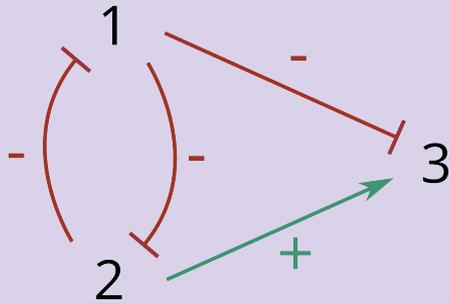


Influence graph (IG): specifies the domain of compatible BNs

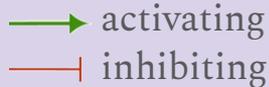


Boolean network (BN): a discrete dynamical system

Domain of BNs:



Influence graph (IG): specifies the domain of compatible BNs



BN compatible with an IG:
uses at most influences in the IG

example:

2 BNs compatible with the same IG

BN f :

$$f_1(x) = \neg x_2$$

$$f_2(x) = \neg x_1$$

$$f_3(x) = \neg x_1 \wedge x_2$$

BN g :

$$g_1(x) = 1$$

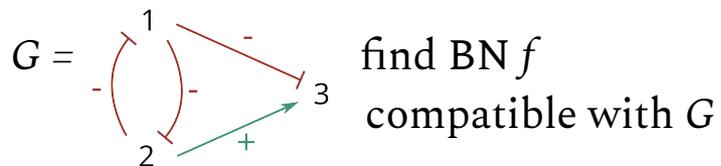
$$g_2(x) = \neg x_1$$

$$g_3(x) = \neg x_1 \vee x_2$$

dynamics(f) \neq dynamics(g)

Synthesis problem: find $f: \{0, 1\}^n \rightarrow \{0, 1\}^n$

Given a domain



Combinatorial problem:

indegree	# monotonic Boolean functions
0	2
2	6
4	168
6	7,828,354
8	56,130,437,228,687,557,907,788

Given dynamics

of configurations compatible with
partial observations of the system

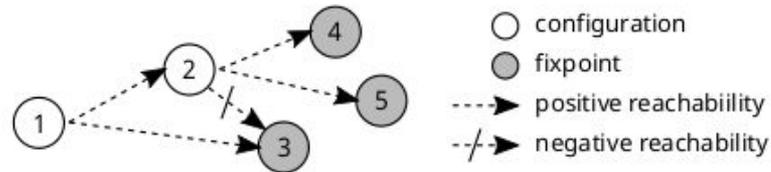
obs. $n^\circ 2$

$$\begin{matrix} x_1 = 1 \\ x_2 = 0 \end{matrix}$$

\rightarrow

compatible conf.

$$\begin{matrix} x_1 = 1 \\ x_2 = 0 \\ x_3 = 0 \end{matrix} \quad \begin{matrix} x_1 = 1 \\ x_2 = 0 \\ x_3 = 1 \end{matrix}$$



Answer-Set Programming (ASP)

brief overview
of ASP syntax

A Logic Program in ASP is a set of logical rules of the form:

$$a_0 \leftarrow a_1, \dots, a_n, \text{ not } a_{n+1}, \dots, \text{ not } a_{n+k}.$$

with integrity constraints as:

$$\leftarrow a_1, \dots, a_n, \text{ not } a_{n+1}, \dots, \text{ not } a_{n+k}.$$

Suitable for solving combinatorial satisfaction problem

Computes **stable models**: minimal sets of a_i satisfying the rules



[Gelfond and Lifschitz, 1988]

Contribution: ASP encoding

Boolean function: expressed in propositional logic under Disjunctive Normal Form

encoded as $\text{clause}(N,C,L,S)$ specifying that:

atom L with sign S (-1, 1) is included in the C^{th} clause of f_N

Canonicity of f : 2 solutions = 2 non-equivalent BNs enforced by a total ordering between the clauses

Dynamical constraints (*suitable for addressed bio. processes*):
positive & negative reachability, several stable properties

example: $f_a(x) = x_c \vee (\neg x_a \wedge x_b)$

is encoded: $\text{clause}(a,1,c,1).$
 $\text{clause}(a,2,a,-1).$
 $\text{clause}(a,2,b,1).$

Complexity

with:

- n #nodes
- d #variables
- k the **fixed upper bound** on #DNF clauses per local function (*the max. being* $\binom{d}{\lfloor d/2 \rfloor}$)

Linear:

- **BN encoding without canonicity** ($O(ndk)$ predicates and rules)
- **Pos. reachability and stable properties** ($O(nk)$ predicates and $O(ndk)$ rules)

Quadratic:

- **BN encoding with canonicity** ($O(nd^2k^2)$ predicates and $O(ndk^2)$ rules)
- **Neg. reachability** ($O(n^2k)$ predicates and $O(n^2dk)$ rules)

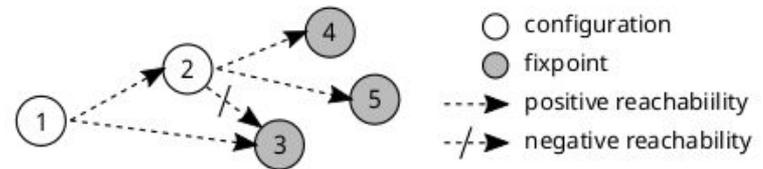
Synthesis with scales and types of knowledge not addressed before

Scalability results

Synthesis on random networks

Domains: random scale-free directed graphs, with different biases on the in-degree of nodes

Dynamics: a generic dynamical property of a two stages differentiation process



Scalability results

Exhaustive synthesis with all constraints:

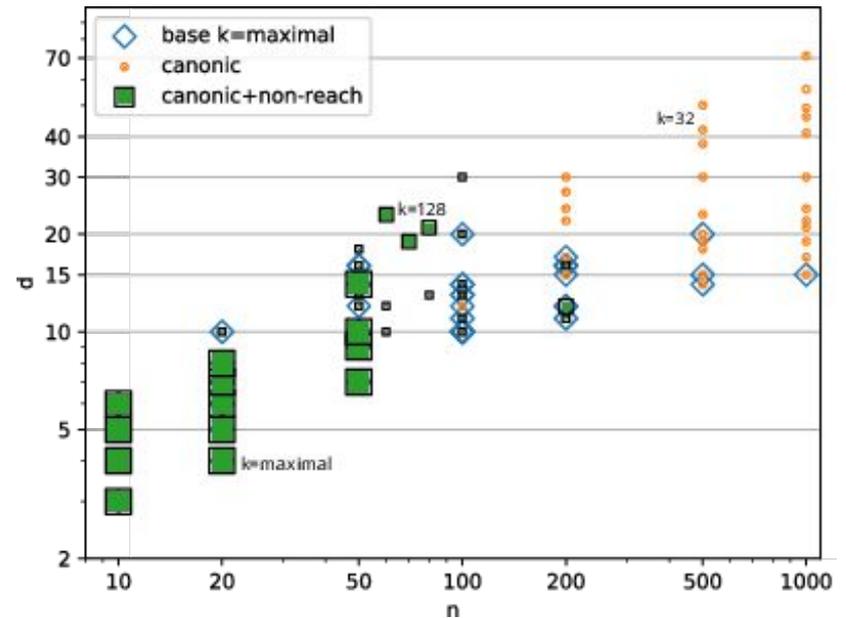
→ up to 50 nodes

Synthesis with a bounded number of clauses:

→ up to 200 nodes

Synthesis with approximate neg. reachability:

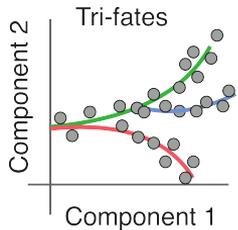
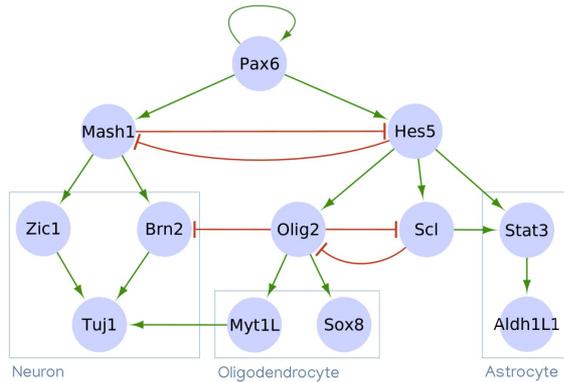
→ up to 1000 nodes



*Successfully solved instances
within 2h of CPU time (2.5Ghz)*

Biological application

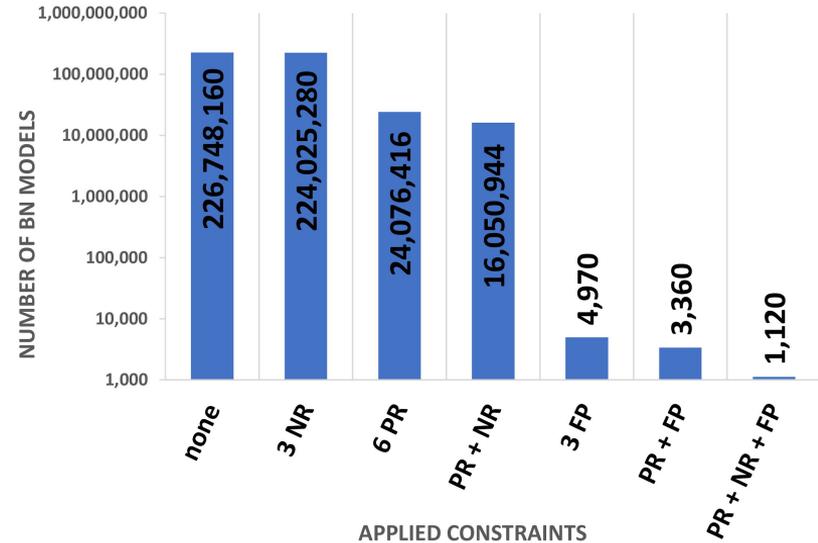
central nervous system development



6 pos. reach (PR)
3 neg. reach (NR)
3 fixpoint (FP)

Impact of the constraints:

NUMBER OF BNs COMPATIBLE WITH CNS DATA
W.R.T. VARIOUS PROPERTIES



Conclusion

Contribution:

Boolean network synthesis method in ASP

Features w.r.t. the state of the art:

- new constraints (negative reachability)
- mix reachability and stable properties
- scalability

Future work:

- Encoding of 2QBF constraints to check universal properties
- Application on blood single-cell differentiation data