

Revisiting and extending probabilistic Boolean networks

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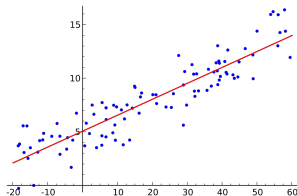
Supervised by : Elisabeth REMY & Claudine CHAOUIYA



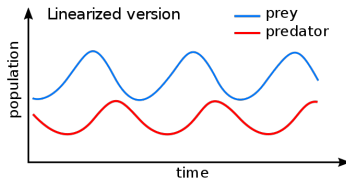
Mathematical models

Applied to many fields of science

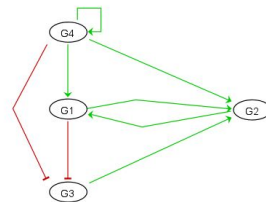
- Elaborated from ground data
- Used to reproduce, understand, and predict behaviors



Linear regression.



Differential equations.

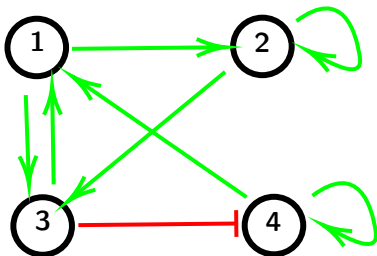


Boolean networks.

Gene Regulatory Networks

Interaction between genes

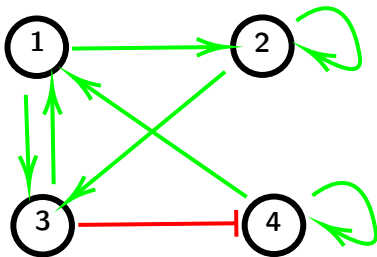
- Positive interactions
- Negative interactions



Boolean Networks

Transition functions

- One boolean function associated to each gene
- Computes the next state depending on the current one



$$f_1(x) = \begin{cases} 1 & \text{if } (x_3 = 1) \vee (x_4 = 1) \\ 0 & \text{otherwise} \end{cases}$$

$$f_2(x) = \begin{cases} 1 & \text{if } (x_1 = 1) \vee (x_2 = 1) \\ 0 & \text{otherwise} \end{cases}$$

$$f_3(x) = \begin{cases} 1 & \text{if } (x_1 = 1) \wedge (x_2 = 1) \\ 0 & \text{otherwise} \end{cases}$$

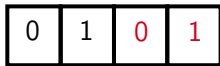
$$f_4(x) = \begin{cases} 1 & \text{if } (x_3 = 0) \wedge (x_4 = 1) \\ 0 & \text{otherwise} \end{cases}$$

Boolean Networks

Transition functions

- One boolean function associated to each gene
- Computes the next state depending on the current one

$$x = (x_1 \quad x_2 \quad x_3 \quad x_4)$$



$$f_1(x) = x_3 \vee x_4$$



$$f_1(x) = \begin{cases} 1 & \text{if } (x_3 = 1) \vee (x_4 = 1) \\ 0 & \text{otherwise} \end{cases}$$

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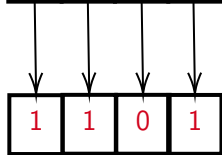
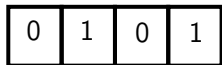
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Boolean Networks

Transition functions

- One associated to each gene
- Computes the next state depending on the current one

$$x = (x_1 \quad x_2 \quad x_3 \quad x_4)$$



$$f(x) = (f_1(x) \quad f_2(x) \quad f_3(x) \quad f_4(x))$$

$$f_1(x) = \begin{cases} 1 & \text{if } (x_3 = 1) \vee (x_4 = 1) \\ 0 & \text{otherwise} \end{cases}$$

$$f_2(x) = \begin{cases} 1 & \text{if } (x_1 = 1) \vee (x_2 = 1) \\ 0 & \text{otherwise} \end{cases}$$

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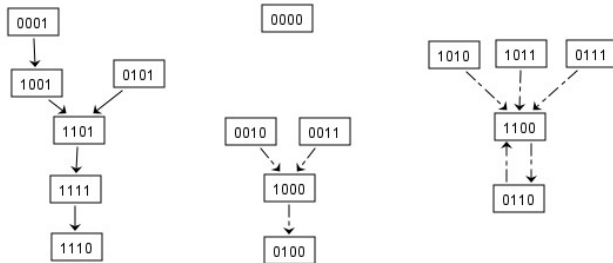
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Boolean Networks

State transitions

0000 → 0000
 0001 → 1001
 0010 → 1000
 0011 → 1000
 0100 → 0100
 0101 → 1101
 0110 → 1100
 0111 → 1100
 1000 → 0100
 1001 → 1101
 1010 → 1100
 1011 → 1100
 1100 → 0110
 1101 → 1111
 1110 → 1110
 1111 → 1110

State Transition Graph (STG)

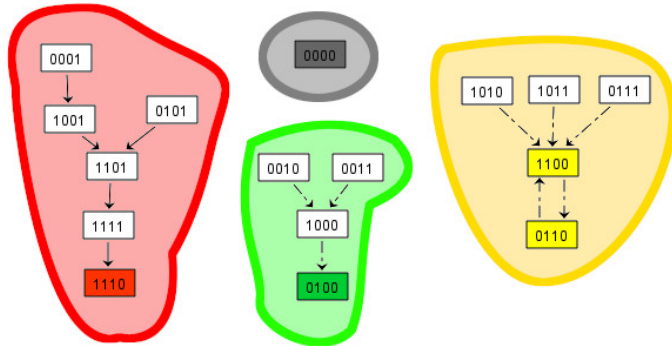


Boolean Networks

State transitions

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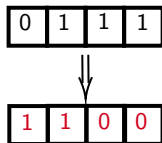
State Transition Graph (STG) with attractors



Boolean Networks : update mode

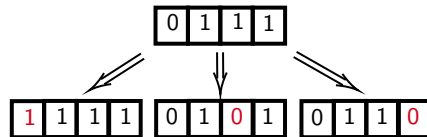
Synchronous

Updates all the genes simultaneously



Asynchronous

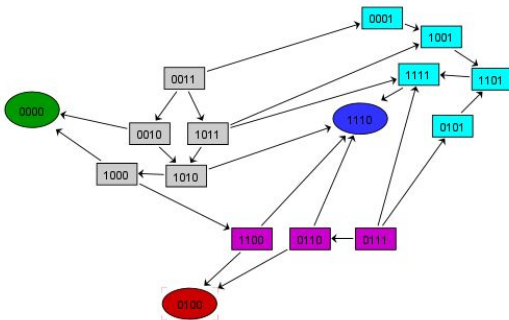
Updates one randomly-chosen gene at a time



→ Also modifies the attractors.

Boolean Networks : update mode

Example : STG for asynchronous update



- Attracted by the blue attractor
- Attracted by the blue and red attractors
- Attracted by all three attractors

Software tools

Analyze a Boolean network with GINSIM

- Takes a set of transition functions as an input
- Computes the state transition graph
- Identifies attractor states and their basins of attraction

Complexity

- With n genes, the STG contains 2^n states. . .

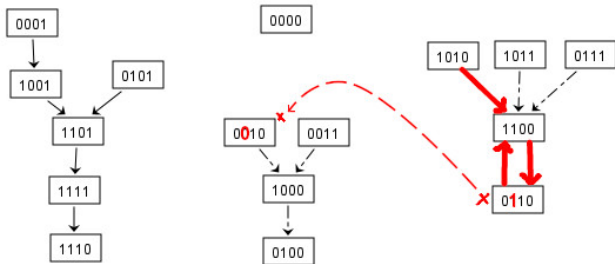
Reduction model

- Methods to approximate the attractors without combinatorial explosion

Extensions I

Perturbations

- Each iteration, a random gene is flipped with probability $p \ll 1$
- Emanates from biological noise

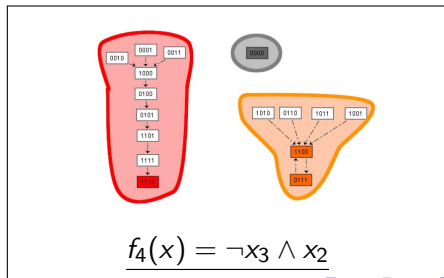
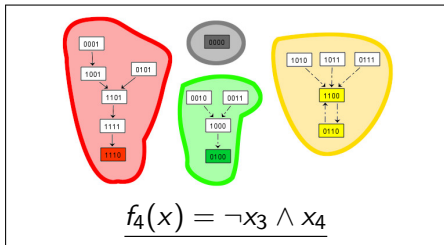


In this example, the gene n°2 is **flipped** from a '1' to a '0', and the current state moves to another attractor.

Extensions II

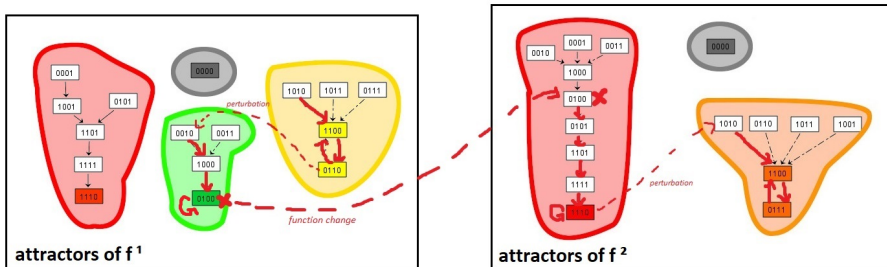
Probabilistic Boolean Networks (PBNs)

- Each iteration, the transition functions are re-drawn from a set with probability $q \ll 1$
- It can alter the 'attractor landscape' of the state space.



Extensions

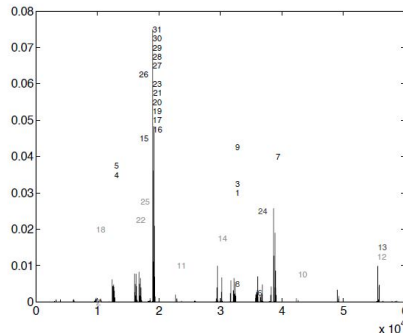
Example of a PBN evolution



Software tools

Sample the stationary law

- Simulating the model a large number of times
- Identify the recurrent states (=attractors)



Example with $n = 10$ [Kim et al., 2003].

Future extensions - My internship

Asynchronous PBNs

- Extend the formalism
- Study dynamical properties
- Adapt the algorithms

Thank you for your attention.

Selected sources :

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- W. Abou-Jaoudé, P. Traynard, P.T. Monteiro, J. Saez-Rodriguez, T. Helikar, D. Thieffry, and C. Chaouiya. Logical Modeling and Dynamical Analysis of Cellular Networks. Frontiers in Genetics, 7(94), 2016. doi :10.3389/fgene.2016.00094.
- Marcel Brun, Edward R. Dougherty, and Ilya Shmulevich. Steady-state probabilities for attractors in probabilistic boolean networks. Signal Processing, 85(10) :1993–2013, 2005. doi :10.1016/j.sigpro.2005.02.016.
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