

Boolean Networks

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Brief History

“There is an urgent need for theories about the ways in which integrated genetic control systems might function.”

- Stuart Kauffman (1969)

“The mere description of a situation as it is seen at a given state of research has often become heavy and tedious, requiring long sentences which are easily ambiguous or misleading. I have felt for some years an increasing necessity for a formalization of the concepts in the field.”

- René Thomas (1973)

Boolean Abstraction

In the simplicity of the formalism, the modelled real system is necessarily abstracted.

Boolean network modelling constitutes a “top-down” abstraction, which allows us to generate hypotheses about the concrete system.

What kind of systems can be modelled by BNs

1. Made up of interacting entities.
2. Each entity is characterised by a variable quantity.
3. The events (and their mechanisms) are not directly observable, only their complete consequences.

“The base” model of complex interacting systems.

Definition - Syntax

$$\mathbb{B} = \{0, 1\}$$

A Boolean network of dimension $n \in \mathbb{N}$ is a function $f : \mathbb{B}^n \rightarrow \mathbb{B}^n$.

$x : \{1, \dots, n\} \rightarrow \mathbb{B}$ is a configuration.

equivalently $x \in \mathbb{B}^n$

local function : $f_i : \mathbb{B}^n \rightarrow \mathbb{B}$

Influence Graph

Directed graph with signed edges.

$$G = (V, E)$$

$$E \subseteq V \times \{+, -\} \times V$$

In-neighbours of a node are known as **activators** or **inhibitors**.

i is an activator of $j \Leftrightarrow (i, +, j) \in E$

i is an inhibitor of $j \Leftrightarrow (i, -, j) \in E$

An influence graph is compatible with a Boolean network, if the edges "explain" the network.

difference

$\exists x, y \in \mathbb{B}^n$ with $\Delta(x, y) = \{i\} \wedge x_i = 0$ such that

$$\begin{cases} f_j(x) = 0 \wedge f_j(y) = 1 & \Rightarrow (i, +, j) \in E \\ f_j(x) = 1 \wedge f_j(y) = 0 & \Rightarrow (i, -, j) \in E \end{cases}$$

$$f_1(x) = \neg x_3$$

$$f_2(x) = x_2 \wedge (x_1 \vee x_3)$$

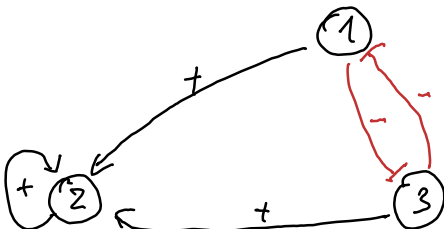
$$f_3(x) = \neg x_1$$

$$x = (0, 0, 0)$$

$$y = (0, 0, 1)$$

$$j=1$$

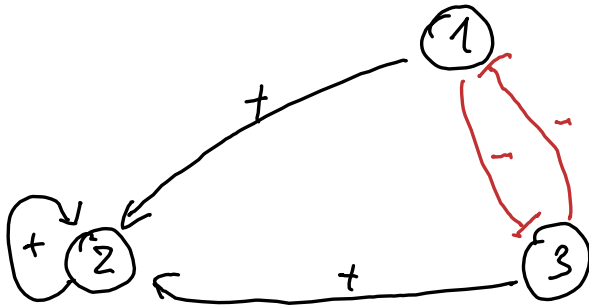
$$i=3$$



$$f_1(x) = \neg x_3$$

$$f_2(x) = x_2 \wedge (x_1 \vee x_3)$$

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



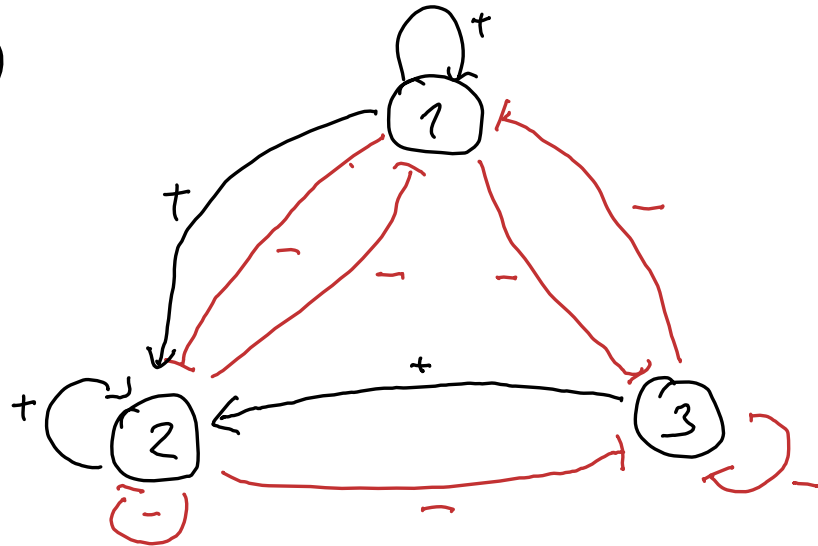
$$g_1(x) = x_1 \wedge \neg x_2 \vee \neg x_3$$

$$g_2(x) = x_3 \wedge (x_1 \neq x_2)$$

$$g_3(x) = \neg x_1 \vee \neg x_2 \vee \neg x_3$$

\neq exclusive or (XOR)

$$x \neq y \approx (x \wedge \neg y) \vee (\neg x \wedge y)$$



Discrete Regulatory Networks

Extension to a finite discrete domain

$$M = \{0, \dots, k_1\} \times \dots \times \{0, \dots, k_n\}, \quad k_1, \dots, k_n \in \mathbb{N}.$$

$$f: M \rightarrow M$$

$$x \in M$$

A special case are "Thomas networks", which retain the Boolean function.

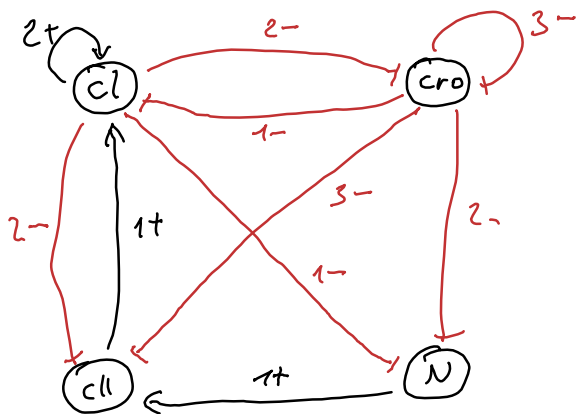
D. Thieffry and R. Thomas. [Dynamical behaviour of biological regulatory networks—ii. immunity control in bacteriophage lambda.](#) *

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$$f: \mathbb{B}^n \rightarrow \mathbb{B}^n$$

$x \in M$

edges have a threshold
step-wise updates



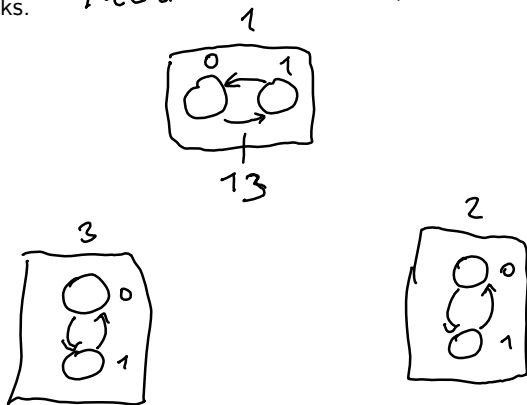
$$M = \{0, 1, 2\} \times \{0, 1, 2, 3\}_x \\ \times \mathbb{B} \times \mathbb{B}$$

Automata Networks

Networks of communicating automata.

Cellular automata. von Neumann, 1966

Artificial neural networks. McCulloch & Pitts, 1943



Dynamics

Updating function modifies configurations according to f .

$$\phi_i : \mathbb{B}^n \rightarrow \mathbb{B}^n$$

$$\phi_i : x \mapsto (x_1, \dots, x_{i-1}, f_i(x), x_{i+1}, \dots, x_n)$$

$$W \in \{1, \dots, n\}$$

$$\phi_W : x \mapsto y \quad \text{such that}$$

$$\forall i \in \{1, \dots, n\} \quad y_i = \begin{cases} f_i(x) & \text{if } i \in W \\ x_i & \text{otherwise} \end{cases}$$

Transitions

An **elementary** transition exists for each application of an updating function.

$$x \xrightarrow{w}_f y \quad \Leftrightarrow \quad \phi_w(x) = y$$

A **non-elementary** transition exists for each predefined sequence of updating function applications.

$$x \xrightarrow{w_1 \dots w_k}_f y \quad \Leftrightarrow \quad \phi_{w_k} \circ \dots \circ \phi_{w_1}(x) = y$$

A **trajectory** is a sequence of consecutive transitions.

$$x_0 \rightarrow x_1 \rightarrow \dots \rightarrow x_n \approx x_0 \rightarrow^* x_k$$

An **updating mode** specifies a subset of elementary and non-elementary transitions.

$$(\mathcal{B}^n, \rightarrow_{(f, \mu)}) \quad \mu\text{-updating mode}$$

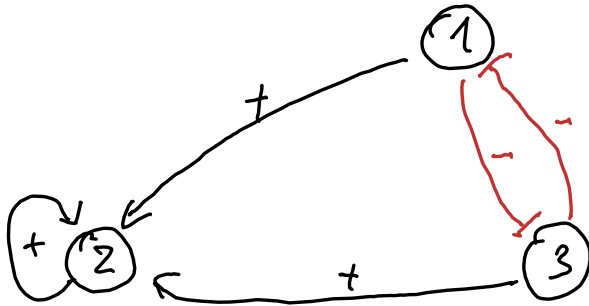
\hookrightarrow subset of elementary and non-elementary transitions

$$f_1(x) = \neg x_3$$

$$f_2(x) = x_2 \wedge (x_1 \vee x_3)$$

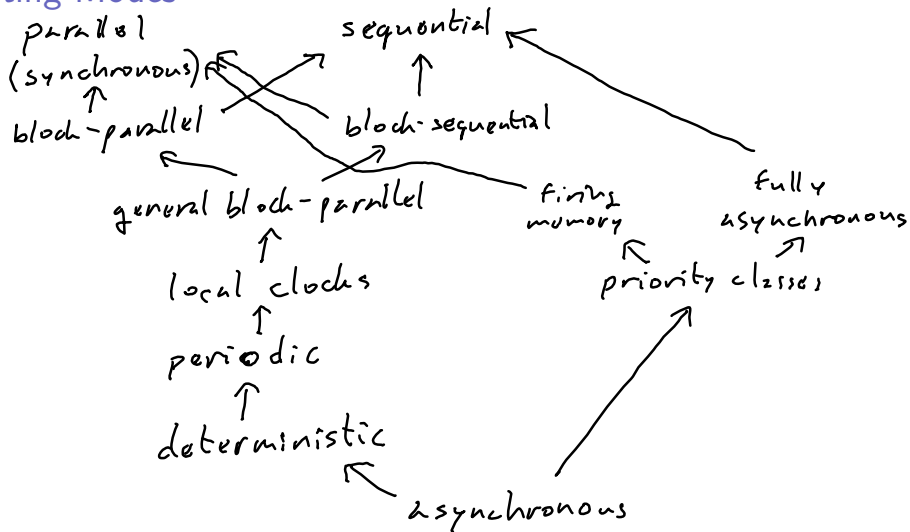
$$f_3(x) = \neg x_1$$

$$x \in \mathbb{B}^3$$



x	$f_1(x)$	$f_2(x)$	$f_3(x)$	ϕ_1	$\phi_{\{1,3\}}$	$\phi_{\{1,2,3\}}$
$(1, 0, 0)$	1	0	1	$(1, 0, 0)$	$(0, 0, 1)$	$(1, 0, 1)$
$(0, 0, 1)$	0	0	1	$(0, 0, 1)$	$(0, 0, 1)$	$(0, 0, 1)$
$(0, 1, 0)$	1	0	1	$(1, 1, 0)$	$(0, 0, 1)$	$(1, 0, 1)$

Updating Modes



Synchronous

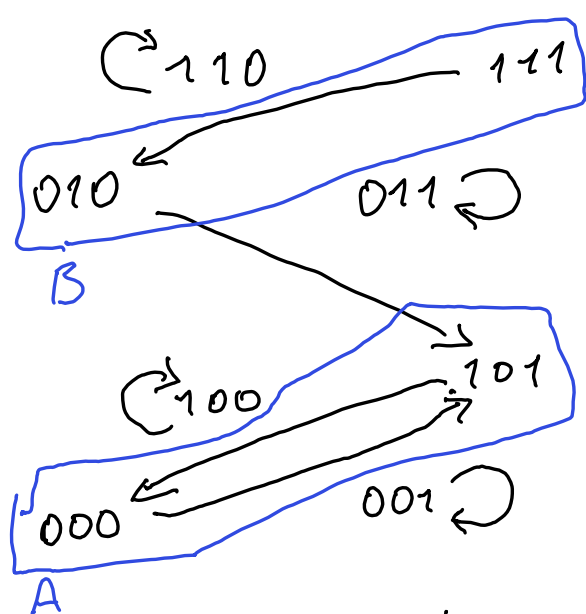
μ_s only transitions $x \xrightarrow{w}_f y$

such that $W = \{1, \dots, n\}$

$$f_1(x) = \neg x_3$$

$$f_2(x) = x_2 \wedge (x_1 \vee x_3)$$

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

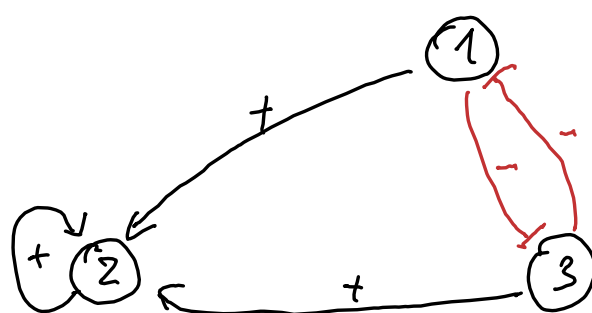


→ Deterministic

A → cyclic attractor (reachable bottom strongly connected component)

B → strong basin of attraction of A

"A is the only reachable attractor (weak = A is reachable, but there are other reachable attractors.)



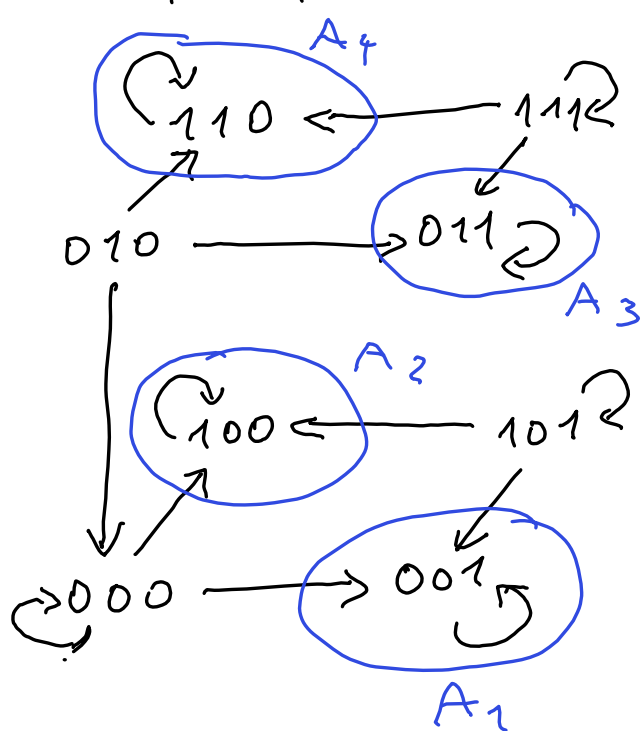
Fully asynchronous

μ_f - only transitions $x \xrightarrow{w}_f y$ such that $W = \{i\}$ for $i \in \{1, \dots, n\}$

$$f_1(x) = \neg x_3$$

$$f_2(x) = x_2 \wedge (x_1 \vee x_3)$$

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



→ Non-deterministic

A_1, \dots, A_4 → attractors (simple)

weak basins of attraction:

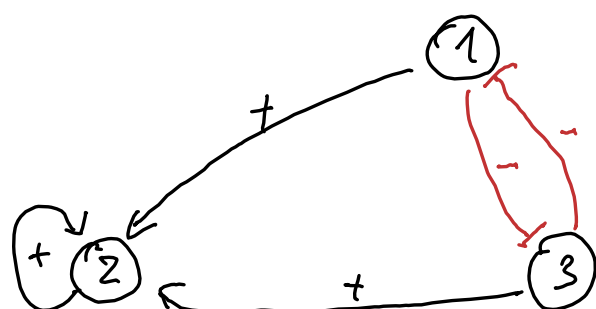
$A_1 \dots \{101, 000, 010\}$

$A_3 \dots \{111, 010\}$

$A_2 \dots \{101, 000, 010\}$

$A_4 \dots \{111, 010\}$

No strong basins!



Asynchronous

μ_a - all elementary transitions

$x \xrightarrow{w}_f y$, $W \subseteq \{1, \dots, n\}$

$$f_1(x) = \neg x_3$$

$$f_2(x) = x_2 \wedge (x_1 \vee x_3)$$

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

