

Learning Dynamics with Synchronous, Asynchronous and General Semantics

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Outline

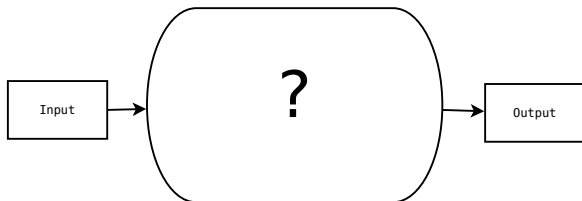
- 1 Motivations
- 2 Formalization
- 3 Learning Process
- 4 Semantics
- 5 Evaluation

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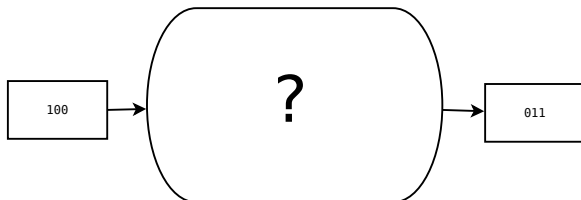
Research area

Idea: given a set of **input/output** states of a **black-box** system, learn its **internal mechanics**.



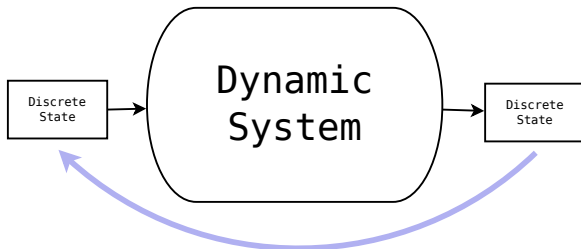
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Discrete system: input/output are vectors of **same size** which contain **discrete values**.



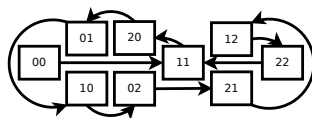
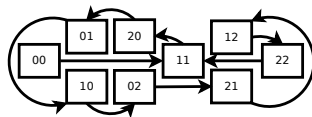
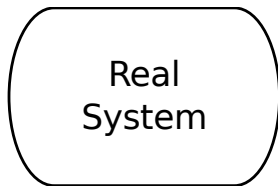
Research area

Dynamic system: input/output are state of the system and output becomes the next input.



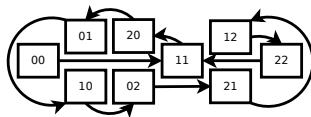
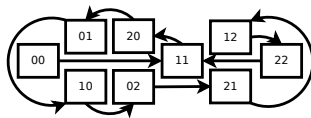
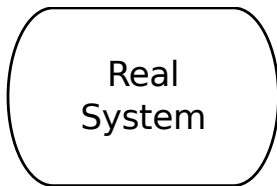
Research area

Goal: produce an **artificial system** with the **same behavior** as the one observed, i.e. a **digital twin**.



Research area

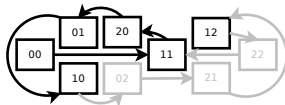
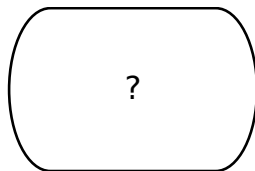
Representation: propositional **logic programs** with annotated atoms encoding **multi-valued variables**.



Research area

Method: learn the dynamics of systems from the observations of some of its state transitions.

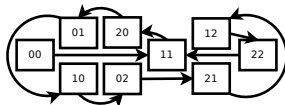
DATA



RESULTS

$a(0,T) :- a(2,T-1)$
 $a(1,T) :- a(0,T-1), b(0,T-1).$
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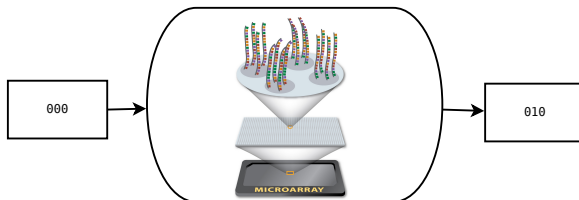
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Motivation

Data: time series of **genes expression** levels in a organic cell.

Goal: model genes interactions to **understand** their influences.



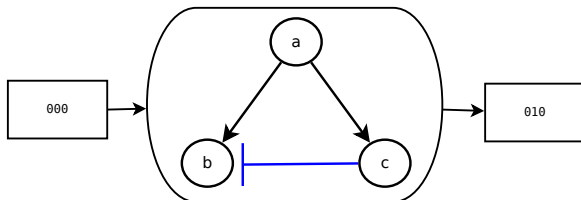
Example (Possible Applications)

- **Bioinformatics:** Construct gene regulatory networks.
- **Robotics:** Learn action models from robot observations.

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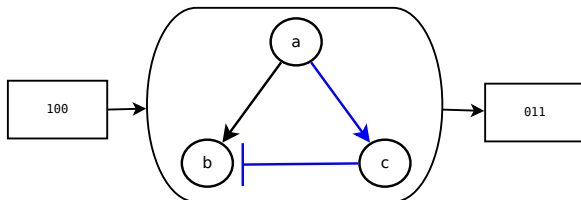
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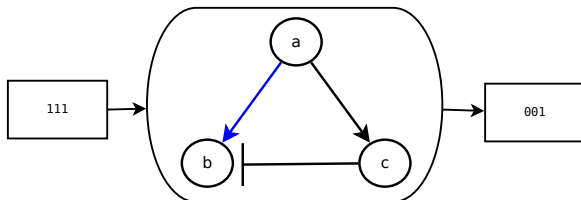
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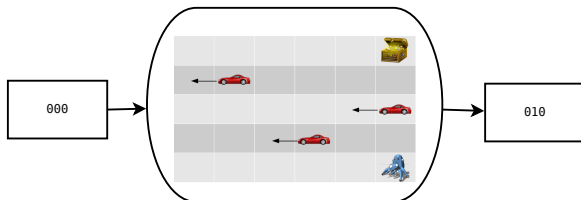
Example (Possible Applications)

- **Bioinformatics:** Construct gene regulatory networks.
- **Robotics:** Learn action models from robot observations.

Motivation

Data: observations of **environment evolution** according to a robot actions.

Goal: produce a **predictive** model of the environment for action **planning**.



Example (Possible Applications)

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- **Robotics:** **Learn action models from robot observations.**

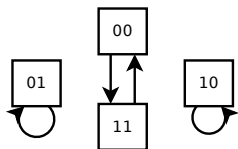
Semantics

Boolean network transitions differ according to the update semantics used.

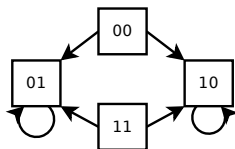


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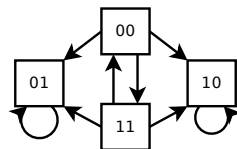
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Synchronous



Asynchronous



General

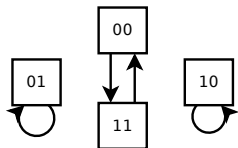
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- **Asynchronous**: only one variable is updated
- **General**: any number of variables can be updated

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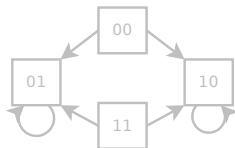
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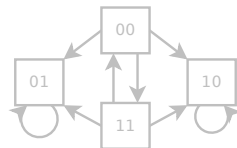
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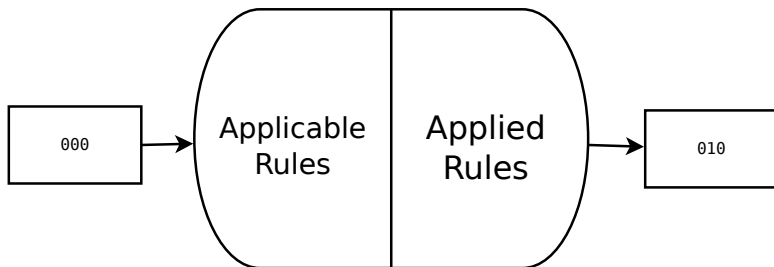


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What is a semantics?

For those three **semantics** atleast its about computing the next state by **selecting** among **applicable** local rules the ones that will be **applied**.

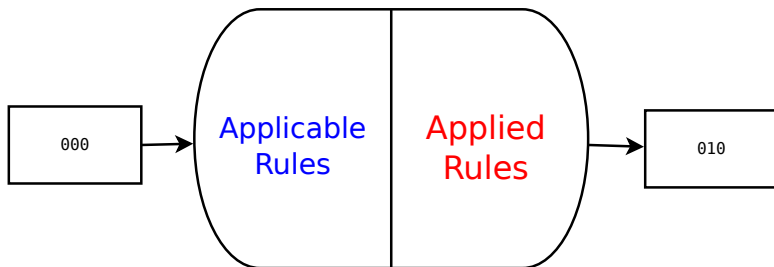


Semantics: what is an applicable rule and what is a valid set of applied rule.

The three semantics differ on the selection but share the same definition of what is an applicable rule.

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Learning algorithm intuition: classification problem

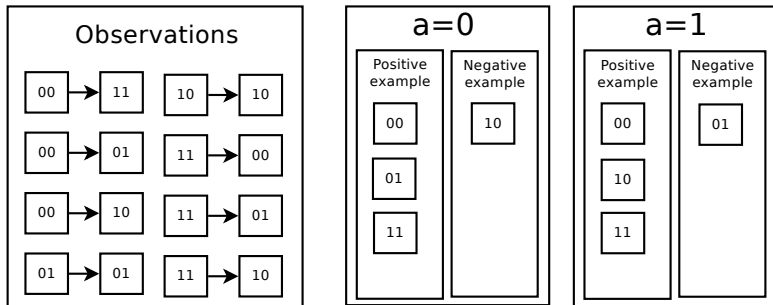
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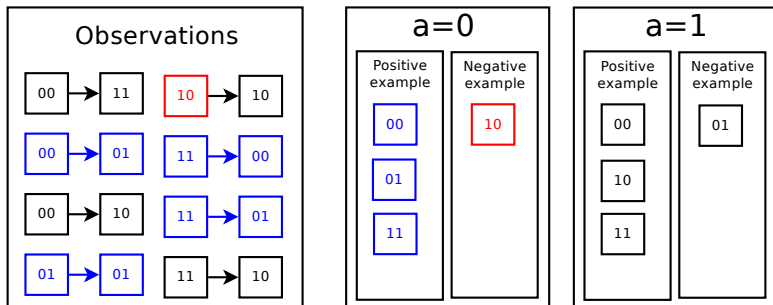
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Equivalent to a **classification problem**: for each variable value, what is a **typical state** where the variable **can** takes this value in the next state ?

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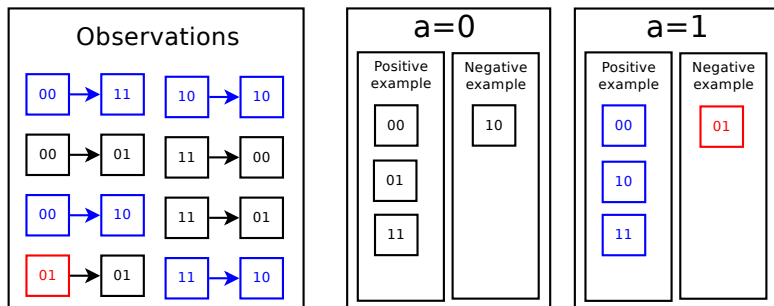
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Multi-valued Logic (\mathcal{MVL})

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Definition (Atoms)

Let $\mathcal{V} = \{v_1, \dots, v_n\}$ be a finite set of $n \in \mathbb{N}$ variables, and $\text{dom} : \mathcal{V} \rightarrow \mathbb{N}$. The atoms of \mathcal{MVL} are of the form v^{val} where $v \in \mathcal{V}$ and $val \in \llbracket 0; \text{dom}(v) \rrbracket$. The set of such atoms is denoted by $\mathcal{A}_{\text{dom}}^{\mathcal{V}}$ or simply \mathcal{A} .

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A \mathcal{MVL} rule R is defined by:

$$R = v_0^{val_0} \leftarrow v_1^{val_1} \wedge \dots \wedge v_m^{val_m} \quad (1)$$

where $\forall i \in \llbracket 0; m \rrbracket, v_i^{val_i} \in \mathcal{A}$ are atoms in \mathcal{MVL} .

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Left-hand side is called the head of R and is denoted $h(R) := v_0^{val_0}$.

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$\text{var}(h(R)) := v_0$ denotes the variable that occurs in $h(R)$.

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A *multi-valued logic program* (\mathcal{MVLP}) is a set of \mathcal{MVL} rules.

Rules Properties

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Definition (Rule Domination)

Let R_1, R_2 be two \mathcal{MVL} rules. R_1 dominates R_2 , written $R_2 \leq R_1$ if $h(R_1) = h(R_2)$ and $b(R_1) \subseteq b(R_2)$.

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Proposition (Double domination is equality)

If $R_1 \leq R_2$ and $R_2 \leq R_1$ then $R_1 = R_2$.

Rules with the most general bodies dominate the other rules.
These are the rules we want since they cover the most general cases.

Dynamical System Modeling

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Definition (Transitions)

We write \mathcal{S} to denote the set of all discrete states, and a couple of states $(s, s') \in \mathcal{S}^2$ is called a transition.

Dynamical System Modeling

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Definition (Rule-state matching)

Let $s \in \mathcal{S}$. The \mathcal{MVL} rule R matches s , written $R \sqcap s$, if $b(R) \subseteq s$.

When matching a state, a rule can be used to realize a transition.

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A rule R realizes the transition (s, s') , written $s \xrightarrow{R} s'$, if $R \sqcap s, h(R) \in s'$.

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Definition (Program realization)

A MVLP P realizes (s, s') , written $s \xrightarrow{P} s'$, if

$\forall v \in \mathcal{V}, \exists R \in P, \text{var}(h(R)) = v \wedge s \xrightarrow{R} s'$.

It realizes $T \subseteq \mathcal{S}^2$, written $\xrightarrow{P} T$, if $\forall (s, s') \in T, s \xrightarrow{P} s'$.

Desired Properties

In the following, for all sets of transitions $T \subseteq \mathcal{S}^2$, we denote:

$$\text{fst}(T) := \{s \in \mathcal{S} \mid \exists (s_1, s_2) \in T, s_1 = s\}.$$

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Definition (Conflicts)

A MVL rule R conflicts with a set of transitions $T \subseteq \mathcal{S}^2$ when
 $\exists s \in \text{fst}(T), (R \sqcap s \wedge \forall (s, s') \in T, h(R) \notin s').$

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Definition (Concurrent rules)

Two \mathcal{MVL} rules R and R' are concurrent when
 $R \sqcap R' \wedge \text{var}(h(R)) = \text{var}(h(R')) \wedge h(R) \neq h(R').$

Definition (Complete program)

A \mathcal{M} VLP P is complete if $\forall s \in \mathcal{S}, \forall v \in \mathcal{V}, \exists R \in P, R \sqcap s \wedge \text{var}(h(R)) = v$.

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Let $s \in \text{fst}(T)$, a program consistent with T will only realize the transitions $(s, s') \in T$.

Optimal \mathcal{MVLP}

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Definition (Suitable program)

Let $T \subseteq \mathcal{S}^2$. A \mathcal{MVLP} P is **suitable** for T when:

- P is **consistent** with T , Cover no negative example
- P **realizes** T , Cover all positive example
- P is **complete** Cover all state space
- $\forall R$ not conflicting with T , $\exists R' \in P$ s.t. $R \leq R'$. Cover all hypotheses

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- $\forall R$ not conflicting with T , $\exists R' \in P$ s.t. $R \leq R'$. Cover all hypotheses

Definition (Optimal program)

If in addition, $\forall R \in P$, all the rules R' belonging to a MVLP suitable for T are such that $R \leq R'$ implies $R' \leq R$ then P is called optimal.

An optimal program is the set of all rules that are not dominated by any consistent rules. Contains all minimal hypotheses

Optimal \mathcal{MVLP}

Optimal \mathcal{M} VLP

Proposition

Let $T \subseteq \mathcal{S}^2$. The \mathcal{M} VLP optimal for T is *unique* and denoted $P_{\mathcal{O}}(T)$.

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Proposition

Let $T \subseteq \mathcal{S}^2$. If P is a MVLP suitable for T , then

$$P_{\mathcal{O}}(T) = \{R \in P \mid \forall R' \in P, R \leq R' \implies R' \leq R\}$$

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Troll mode on: does it works if it is the empty set? :p

Proposition

$$P_{\mathcal{O}}(\emptyset) = \{v^{val} \leftarrow \emptyset \mid v^{val} \in \mathcal{A}\}.$$

Yeah ! And this property is the *starting point* of the learning algorithm.

Proposition

Let $T \subseteq \mathcal{S}^2$. If P is a MVLP suitable for T , then

$$P_{\mathcal{O}}(T) = \{R \in P \mid \forall R' \in P, R \leq R' \implies R' \leq R\}$$

We can obtain the optimal program from any suitable program by simply *removing the dominated rules*.

Outline

- 1 Motivations
- 2 Formalization
- 3 Learning Process**
- 4 Semantics
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Learning Process

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How to make a **minimal modifications** of a \mathcal{MVLP} in order to be suitable with a **new set of transitions**?

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Definition (Rule least specialization)

Let R be a $\mathcal{M}VL$ rule and $s \in \mathcal{S}$ such that $R \sqcap s$. The least specialization of R by s is:

$$L_{\text{spe}}(R, s) := \{h(R) \leftarrow b(R) \cup \{v^{val}\} \mid v^{val} \in \mathcal{A} \wedge v^{val} \notin s \wedge \forall val' \in \mathbb{N}, v^{val'} \notin b(R)\}.$$

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Thus, a \mathcal{MVLP} can be revised to only realizes given transitions from s .

Definition (Program least revision)

Let P be a \mathcal{MVLP} , $s \in \mathcal{S}$ and $T \subseteq \mathcal{S}^2$ such that $\text{fst}(T) = \{s\}$. Let $R_P := \{R \in P \mid R \text{ conflicts with } T\}$. The least revision of P by T is $L_{\text{rev}}(P, T) := (P \setminus R_P) \cup \bigcup_{R \in R_P} L_{\text{spe}}(R, s)$.

Learning Process

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Theorem

Let $s \in \mathcal{S}$ and $T, T' \subseteq \mathcal{S}^2$ such that $|\text{fst}(T')| = 1 \wedge \text{fst}(T) \cap \text{fst}(T') = \emptyset$. $L_{\text{rev}}(P_{\emptyset}(T), T')$ is a MVLP suitable for $T \cup T'$.

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Let $s \in \mathcal{S}$ and $T, T' \subseteq \mathcal{S}^2$ such that $|\text{fst}(T')| = 1 \wedge \text{fst}(T) \cap \text{fst}(T') = \emptyset$. $L_{\text{rev}}(P_{\mathcal{O}}(T), T')$ is a MVLP suitable for $T \cup T'$.

In association with previous results it gives a method to iteratively compute $P_{\mathcal{O}}(T)$ for any $T \subseteq \mathcal{S}^2$, starting from $P_{\mathcal{O}}(\emptyset)$.

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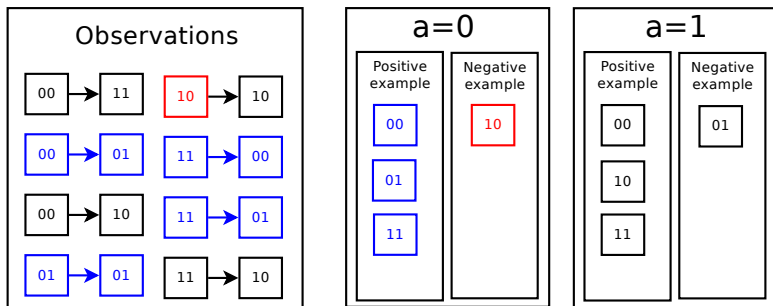
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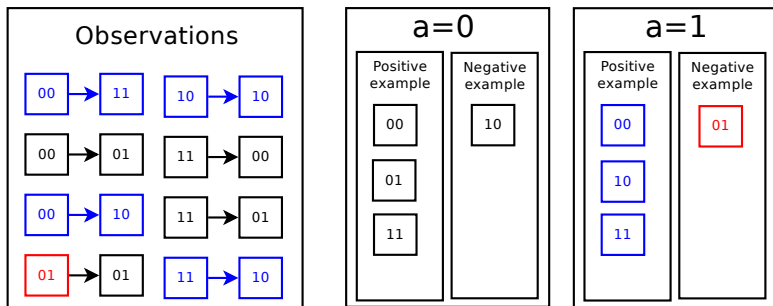
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OUTPUT: $P_{\mathcal{O}}(T) := P$.

Outline

- 1 Motivations
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Where is the semantics gone?

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The formalization of \mathcal{MVLP} is **independant of the semantics** that produced its transitions.

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Definition (Semantics)

Let $\mathcal{A}_{\text{dom}}^{\mathcal{V}}$ be a set of atoms and \mathcal{S} the corresponding set of states. A **semantics** (on $\mathcal{A}_{\text{dom}}^{\mathcal{V}}$) is a function that associates, to each complete \mathcal{MVLP} P , a set of transitions $T \subseteq \mathcal{S}^2$ so that: $\text{fst}(T) = \mathcal{S}$. Equivalently, a semantics can be seen as a function of $(\text{c-}\mathcal{MVLP} \rightarrow (\mathcal{S} \rightarrow \wp(\mathcal{S}) \setminus \emptyset))$ where $\text{c-}\mathcal{MVLP}$ is the set of complete \mathcal{MVLP} s and \wp is the power set operator.

Definition (Synchronous semantics)

The synchronous semantics \mathcal{T}_{syn} is defined by:

$$\mathcal{T}_{syn} : P \mapsto \{(s, s') \in \mathcal{S}^2 \mid s' \subseteq \{h(R) \in \mathcal{A} \mid R \in P, R \sqcap s\}\}$$

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Semantic free modeling

Finally, we can state that the definitions and method developed in the previous section are independent of those three semantics.

Theorem (Semantics-free correctness)

Let P be a MVLP such that P is complete.

- $\mathcal{T}_{syn}(P) = \mathcal{T}_{syn}(P_{\mathcal{O}}(\mathcal{T}_{syn}(P)))$,
- $\mathcal{T}_{asyn}(P) = \mathcal{T}_{asyn}(P_{\mathcal{O}}(\mathcal{T}_{asyn}(P)))$,
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Whatever the semantic which produced T , given the optimal MVLP of T we can reproduce exactly T with the same semantic.

GULA: General Usage LFIT Algorithm

And **GULA** can learn such an optimal \mathcal{MVLP} from T .

Theorem (**GULA** Termination, soundness, completeness, optimality)

*Let $T \subseteq \mathcal{S}^2$. The call **GULA**(\mathcal{A}, T) terminates and **GULA**(\mathcal{A}, T) = $P_{\mathcal{O}}(T)$.*

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Victory! In theory, but how does it scale in practice?

Outline

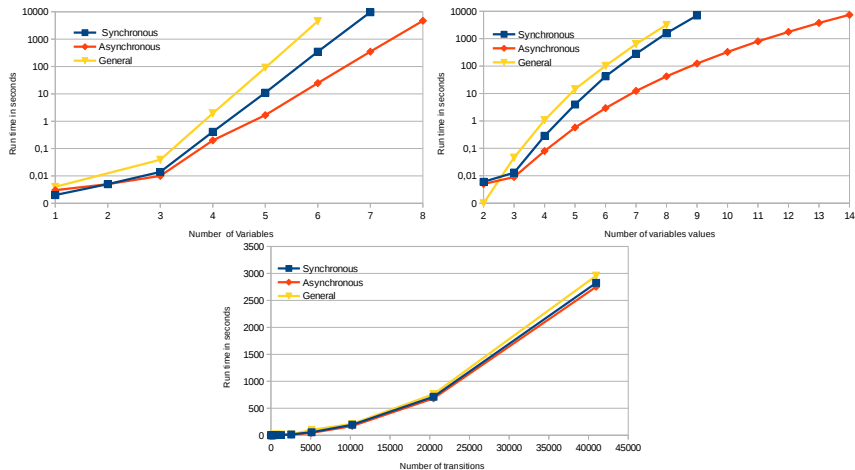
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Evaluation

Theorem (**GULA** Complexity)

Let $T \subseteq \mathcal{S}^2$ be a set of transitions, $n := |\mathcal{V}|$ be the number of variables of the system and $d := \max(\text{dom}(\mathcal{V}))$ be the maximal number of values of its variables. The worst-case time complexity of **GULA** when learning from T belongs to $\mathcal{O}(|T|^2 + 2n^3d^{2n+1} + 2n^2d^n)$ and its worst-case memory use belongs to $\mathcal{O}(d^{2n} + 2d^n + nd^{n+2})$.

Evaluation



Evaluation of **GULA**'s scalability w.r.t. number of variables (top left), number of variables values (top right) and number of input transitions (bottom).

Evaluation

Semantics	Mammalian (10)	Fission (10)	Budding (12)	Arabidopsis (15)
Synchronous	1.84s / 1, 024	1.55s / 1, 024	34.48s / 4, 096	2, 066s / 32, 768
Asynchronous	19.88s / 4, 273	19.18s / 4, 217	523s / 19, 876	T.O. / 213, 127
General	928s / 34, 487	1, 220s / 29, 753	T.O. / 261, 366	T.O. / > 500, 000

Run time of **GULA** (run time in seconds / number of transitions as input) for Boolean network benchmarks up to 15 nodes for the three semantics.

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- Interface with MetaGol for learning semantics too

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- Synchronous deterministic
- Markov(k) systems
- Synchronous non-deterministic (no minimality)
- continuous valued systems

New contribution

- Synchronous non-deterministic
- Asynchronous
- generalized

Ongoing

- Improve implementation + approximation
- Apply to learn construction network
- Interface with MetaGol for learning semantics too
- One algorithm to learn them all

