

# Learning any memory-less discrete semantics for dynamical systems represented by logic programs

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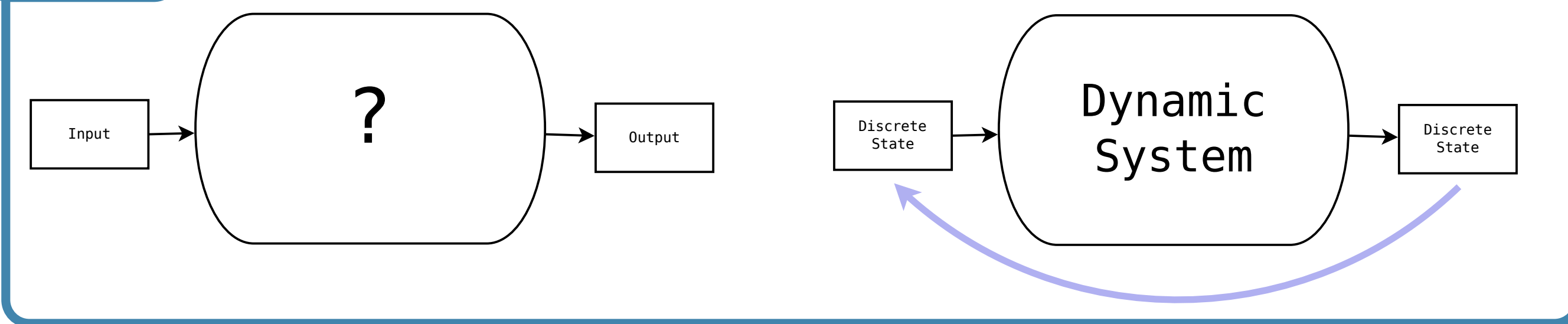
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## Motivations: Learning Dynamics

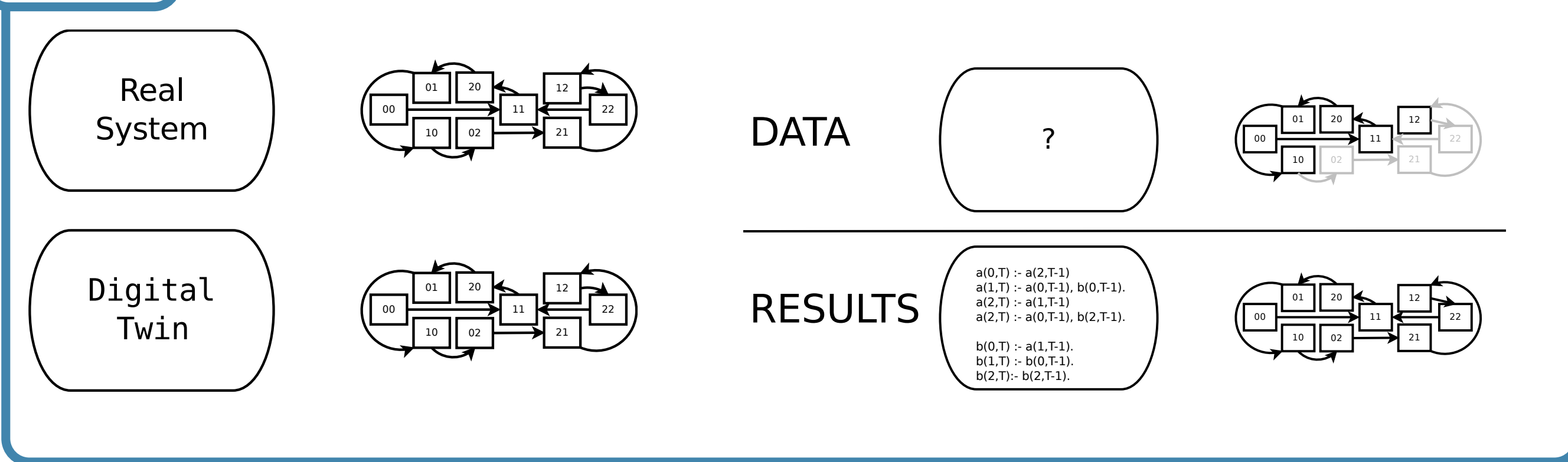
- Given a set of input/output states of a black-box system, learn its internal mechanics.
- Discrete system:** input/output are vectors of same size which contain discrete values.
- Dynamic system:** input/output are states of the system and output is the next input.

### Problem



- Goal:** produce an artificial system with the same behavior, i.e., a digital twin.
- Representation:** propositional logic programs encoding multi-valued discrete variables.
- Method:** learn the dynamics of a system from its state transitions.

### Goal



## Formalization: MVL and DMVLP

**Definition 1 (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 MVL (denoted  $\mathcal{A}$ ) are of the form  $v^{val}$  where  $v \in \mathcal{V}$  and  $val \in \llbracket 0; \text{dom}(v) \rrbracket$ .

**Definition 2 (Multi-valued logic program).** A DMVLP is a set of MVL rules:

$$\underbrace{v_0^{val_0}}_{\text{head}} \leftarrow \underbrace{v_1^{val_1} \wedge v_2^{val_2} \wedge v_3^{val_3} \wedge \dots \wedge v_m^{val_m}}_{\text{body}}$$

**Definition 3 (Dynamic MVL).** Let  $\mathcal{T} \subseteq \mathcal{V}$  and  $\mathcal{F} \subseteq \mathcal{V}$  such that  $\mathcal{F} = \mathcal{V} \setminus \mathcal{T}$ . A DMVLP  $P$  is a MVL such that  $\forall R \in P, \text{var}(\text{head}(R)) \subseteq \mathcal{T}$  and  $\forall v^{val} \in \text{body}(R), v \in \mathcal{F}$ .

**Definition 4 (Discrete state).** A discrete state  $s$  on  $\mathcal{T}$  (resp.  $\mathcal{F}$ ) of a DMVLP is a function from  $\mathcal{T}$  (resp.  $\mathcal{F}$ ) to  $\mathbb{N}$ .  $\mathcal{S}^{\mathcal{T}}$  (resp.  $\mathcal{S}^{\mathcal{F}}$ ) denote the set of all discrete states of  $\mathcal{T}$  (resp.  $\mathcal{F}$ ).

**Definition 5 (Transition).** A transition is a couple of states  $(s, s') \in \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$ .

**Definition 6 (Semantics).** A dynamical semantics is a function  $\text{DMVLP} \rightarrow (\mathcal{S}^{\mathcal{F}} \rightarrow \wp(\mathcal{S}^{\mathcal{T}}) \setminus \{\emptyset\})$  where DMVLP is the set of DMVLPs and  $\wp$  is the power set symbol.

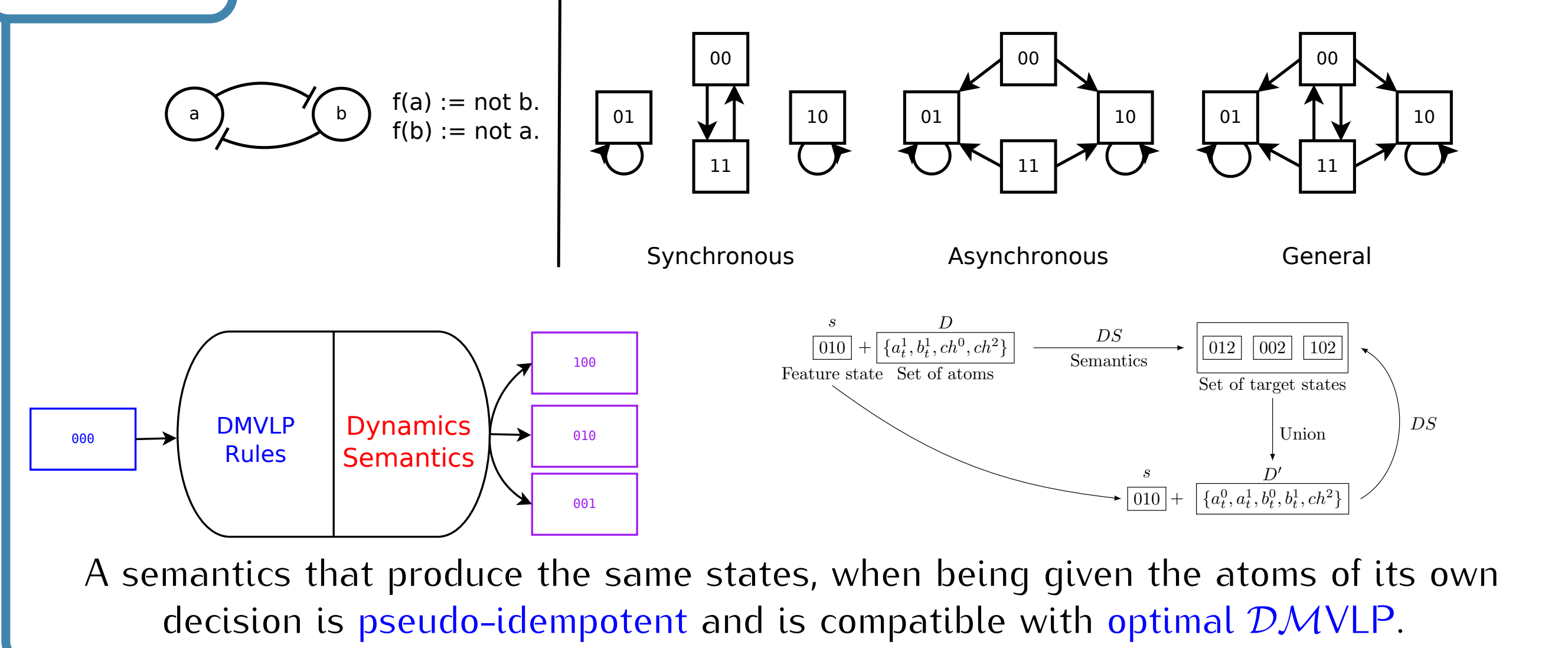
- $R_1$  dominates  $R_2$ , written  $R_1 \geq R_2$ , if  $\text{head}(R_1) = \text{head}(R_2)$  and  $\text{body}(R_1) \subseteq \text{body}(R_2)$ .
- $R$  matches  $s \in \mathcal{S}^{\mathcal{F}}$ , written  $R \sqcap s$ , if  $\text{body}(R) \subseteq s$ .
- $R$  realizes the transition  $(s, s') \in \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$ , if  $R \sqcap s$  and  $\text{head}(R) \in s'$ .
- $R$  conflicts with  $T \subseteq \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$  if  $\exists (s, s') \in T, (R \sqcap s \wedge \forall (s, s'') \in T, \text{head}(R) \notin s'')$ .

**Definition 7 (Suitable program).** Let  $T \subseteq \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$ . A DMVLP  $P$  is suitable for  $T$  when:  $P$  is complete, consistent with  $T$ , realizes  $T$  and  $\forall R$  not conflicting with  $T, \exists R' \in P$  s.t.  $R \geq R'$ . If, in addition,  $\forall R \in P$ , all the rules  $R'$  belonging to a MVL suitable for  $T$  are such that  $R \geq R'$  implies  $R' \geq R$ , then  $P$  is unique, called optimal and denoted  $P_O(T)$ .

## Problem: Dynamical Semantics

Semantics decide the target states according to a DMVLP and a feature state.

### Dynamics



**Definition 8 (Pseudo-idempotent Semantics).** Let  $DS$  be a dynamical semantics.  $DS$  is said pseudo-idempotent if, for all  $P$  a DMVLP:  $DS(P_O(DS(P))) = DS(P)$ .

## Algorithm: GULA

**Definition 9 (Rule least specialization).** Let  $R$  be a MVL rule and  $s \in \mathcal{S}^{\mathcal{F}}$  such that  $R \sqcap s$ . The least specialization of  $R$  by  $s$  according to  $\mathcal{F}$  and  $\mathcal{A}$  is:

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

$\forall T \subseteq \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$ , we denote:  $\text{first}(T) := \{s \in \mathcal{S}^{\mathcal{F}} \mid \exists (s_1, s_2) \in T, s_1 = s\}$ .

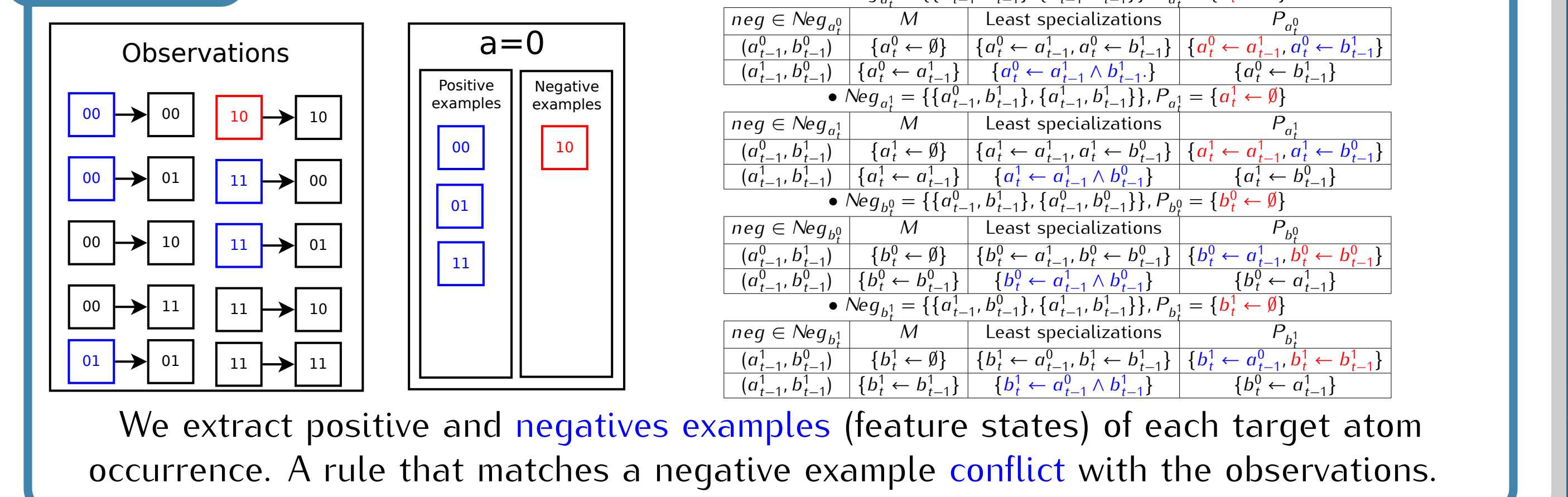
**Definition 10 (Program least revision).** Let  $P$  be a DMVLP,  $s \in \mathcal{S}^{\mathcal{F}}$  and  $T \subseteq \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$  such that  $\text{first}(T) = \{s\}$ . Let  $R_p := \{R \in P \mid R \text{ conflicts with } T\}$ . The least revision of  $P$  by  $T$  according to  $\mathcal{A}$  and  $\mathcal{F}$  is  $L_{\text{rev}}(P, T, \mathcal{A}, \mathcal{F}) := (P \setminus R_p) \cup \bigcup_{R \in R_p} L_{\text{spe}}(R, s, \mathcal{A}, \mathcal{F})$ .

Algorithmic properties:

- $P_O(\emptyset) = \{v^{val} \leftarrow \emptyset \mid v \in \mathcal{T} \wedge v^{val} \in \mathcal{A}\}$ .
- Let  $s \in \mathcal{S}^{\mathcal{F}}$  and  $T, T' \subseteq \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$  such that  $|\text{first}(T')| = 1 \wedge \text{first}(T) \cap \text{first}(T') = \emptyset$ .  $L_{\text{rev}}(P_O(T), T', \mathcal{A}, \mathcal{F})$  is a DMVLP suitable for  $T \cup T'$ .
- If  $P$  is a DMVLP suitable for  $T$ , then  $P_O(T) = \{R \in P \mid \forall R' \in P, R' \geq R \implies R' = R\}$ .

Idea: Starting from  $P = P_O(\emptyset)$  we group transitions by common feature state ( $T$ ) and iteratively revise  $P$  using  $L_{\text{rev}}(P, T', \mathcal{A}, \mathcal{F})$  and domination relation to obtain  $P_O(T)$ .

### GULA



We extract positive and negatives examples (feature states) of each target atom occurrence. A rule that matches a negative example conflict with the observations.

## Learning From Any Semantics Using Constraints

**Definition 11 (Constrained DMVLP).** Let  $P'$  be a DMVLP on  $\mathcal{A}, \mathcal{F}$  and  $T$  two sets of variables, and  $\varepsilon$  a special variable with  $\text{dom}(\varepsilon) = 1$  so that  $\varepsilon \notin T \cup \mathcal{F}$ . A CD MVLP  $P$  is a MVL such that  $P = P' \cup \{R \in \text{MVL} \mid \text{head}(R) = \varepsilon^1 \wedge \forall v^{val} \in \text{body}(R), v \in \mathcal{F} \cup T\}$ . A rule  $R$  such that  $\text{head}(R) = \varepsilon^1$  and  $\forall v^{val} \in \text{body}(R), v \in \mathcal{F} \cup T$  is called a MVL constraint.

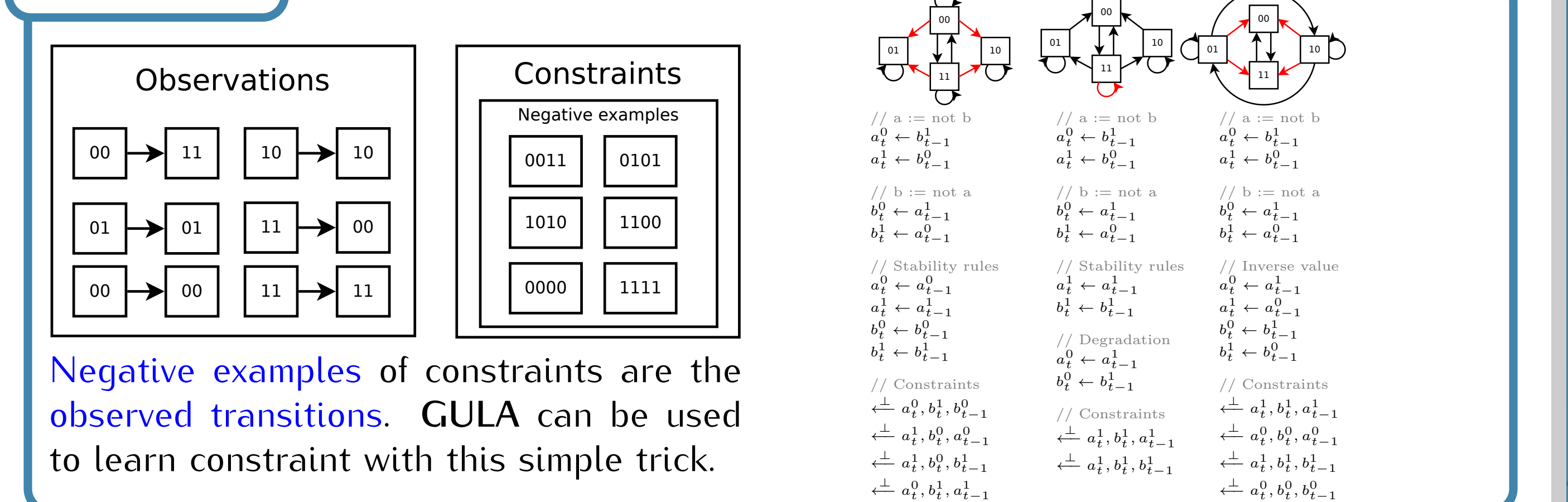
**Definition 12 (Constraint-transition matching).** Let  $(s, s') \in \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$ . The constraint  $C$  matches  $(s, s')$ , written  $C \sqcap (s, s')$ , iff  $\text{body}(C) \subseteq s \cup s'$ .

**Definition 13 (Suitable and optimal constraints).** Let  $T \subseteq \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$ . A set of MVL constraints  $SC$  is suitable for  $T$  when:  $SC$  is consistent with  $T$ , complete with  $T$  and for all constraints  $C$  not conflicting with  $T$ , there exists  $C' \in SC$  such that  $C \geq C'$ . In addition, for all  $C \in SC$ , all the constraint rules  $C'$  belonging to a set of constraints suitable for  $T$  are such that  $C' \geq C$  implies  $C \geq C'$ , then  $SC$  is called optimal, is unique and denoted  $C_O(T)$ .

**Definition 14 (Synchronous constrained Semantics).** The synchronous constrained semantics  $\mathcal{I}_{\text{Syn-c}}$  is defined by:

$$\mathcal{I}_{\text{Syn-c}} : P \mapsto \{(s, s') \in \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}} \mid s' \subseteq \{\text{head}(R) \in \mathcal{A} \mid R \in P, R \sqcap s\} \wedge \nexists C \in P, \text{head}(C) = \varepsilon^1 \wedge C \sqcap (s, s')\}$$

### Synchronizer



Negative examples of constraints are the observed transitions. GULA can be used to learn constraint with this simple trick.

Let  $T \subseteq \mathcal{S}^{\mathcal{F}} \times \mathcal{S}^{\mathcal{T}}$ , it holds that  $T = \mathcal{I}_{\text{Syn-c}}(P_O(T) \cup C_O(T))$ , i.e., any semantics is captured.

## Contributions

- Previous works:** Synchronous deterministic transitions only [1-3].
- Novelty:** Learn from any memory-less discrete dynamical semantics.
- Application:** The selection of a dynamical semantics, which has a major impact when trying to model a system, can now be done a posteriori. The rules can explain local interactions and constraint are hints of the behavior of the original semantics.
- Weakness:** The current complete method is too costly/sensitive to deal with real systems.
- Outlook:** Development of heuristic approaches (WDMVLP, PRIDE, ...) to tackle real data and tools (see other poster) to extract knowledge from the learned models.
- The source code is available as open source on Github. See QR-code