

# Learning Disjunctive Logic Programs from Interpretation Transition

Yi Huang<sup>1,3</sup>   Yisong Wang<sup>1</sup>   Ying Zhang<sup>2</sup>  
and Mingyi Zhang<sup>2</sup>

<sup>1</sup>Guizhou University

<sup>2</sup>Guizhou Academy of sciences

<sup>3</sup>Chongqing University of Arts and Sciences

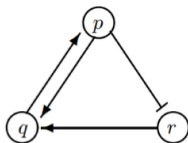
September 4, ILP2016

- 1 Motivation
  - Dynamic System
  - Synchronous Boolean Network As NLP
  - Asynchronous Boolean Network As DLP
- 2 Learning DLP from Interpretation Transition
  - The learning task
  - The LFDT learning algorithm
  - Two resolutions
  - An example
- 3 Concluding Remarks

Dynamics is involved in [wikipedia]:

- Mathematics and Computer Science:
  - A **dynamical system** is a system in which a function describes the time dependence of a point in a space.
  - At any given time a dynamical system has a state that can be represented by a point in an appropriate state space.
  - The *evolution rule* of the dynamical system is a function that describes what future states follow from the current state.

**Boolean networks is an important tool to model dynamical systems!**



$$p^{t+1} = q^t,$$

$$q^{t+1} = p^t \wedge r^t,$$

$$r^{t+1} = \neg p^t.$$

Figure: A Boolean network  $N$

- Inoue proposed Synchronous Boolean Network can be direct translated into a normal logic program(NLP).
  - discarding its time parameter  $t$ , and
  - replacing “=” by “ $\leftarrow$ ”.

- Inoue and Ribeiro and Sakama proposed a novel framework to learn NLP from transitions of interpretations.
  - naive/ground resolution
  - LF1T algorithm

**Input:** observations/examples  $E \subseteq 2^A \times 2^A$ , and a background program  $B$ ,

**Output:** A NLP  $P$  such that  $T_{P \cup B}(I) = J$  for each  $\langle I, J \rangle \in E$ .

$P \leftarrow \emptyset$ ;

for each  $A \in J$ ,

$$R_I^A := (A \leftarrow \bigwedge (I \cup \neg I)).$$

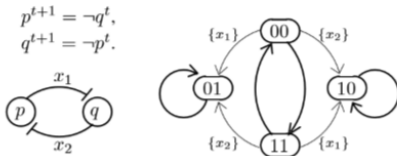
Adding  $R_A^I$  into  $P$ .

Simplifying  $P$  using  $B$  by ground resolution and subsumption.

# Questions:

A (asynchronous) Boolean network (ABN) differs from SBNs only in the state update:

Not all nodes are necessary updated at the same time!  
The next state of the current state is nondeterministic!



We consider that the next states are all minimal in an ABN!  
That can be captured by a disjunctive logic program!

- An observation/example is a state transition  $\langle I, J \rangle \in 2^A \times 2^A$ ,
- A set of observations/examples  $E$  is *coherent* if  $J$  and  $J'$  are incomparable for every two pairs  $\langle I, J \rangle$  and  $\langle I, J' \rangle$  of  $E$ .
- $E$  is *consistent* with a logic program  $B$  if, for each  $\langle I, J \rangle \in E$ ,  $I \models bd(r)$  implies  $J \models hd(r)$  for every  $r \in B$ .

### Learning from nondeterministic 1-step Transition (LFDT) task:

**Input:** coherent observations/examples  $E \subseteq 2^A \times 2^A$ , which is consistent with the background program  $B$ ,

**Output:** A DLP  $P$  such that  $J \in T_{P \cup B}^d(I)$  for each  $\langle I, J \rangle \in E$ .

---

## Algorithm 1: LFDT( $E, B$ )

---

**input** : A set  $E$  of state transitions and a background theory  $B$  such that  $E$  is coherent

**output**: A logic program  $P$

$P \leftarrow B$

**foreach**  $\langle I, J \rangle \in E$  **do**

$Q \leftarrow \{R_i^q \mid q \in J\}$

$E \leftarrow E \setminus \{\langle I, J \rangle\}$

**foreach**  $\langle I', J' \rangle \in E$  with  $I' = I$  **do**

$E \leftarrow E \setminus \{\langle I', J' \rangle\}$

**foreach**  $p \in J'$  and  $r \in Q$  **do**

$Q \leftarrow Q \cup \{hd(r) \cup \{p\} \leftarrow bd(r)\}$

**end**

**end**

**foreach**  $r \in Q$  **do** *AddRule*( $r, P$ )

**end**

$P \leftarrow P \setminus B$

**return**  $P$

---



---

## Algorithm 2: AddRule( $r, P$ )

---

**Input:** A rule  $r$  and a logic program  $P$   
**if**  $\exists r' \in P$  s.t.  $r' \prec r$  **then return;**  
**foreach**  $r' \in P$  **do** **if**  $r \prec r'$  **then**  $P \leftarrow P \setminus \{r'\};$   
 $P \leftarrow P \cup \{r\};$   
**while**  $r, r_1, \dots, r_k \in P$  are combined resolvable **do**  
 | **AddRule**( $cr(r, r_1, \dots, r_k), P$ );  
**end**  
**foreach**  $r' \in P$  **do**  
 | **if**  $r$  is disjunctively ground resolvable w.r.t.  $r'$  **then**  
 | | **AddRule**( $gr(r, r'), P$ );  
 | **else if**  $r'$  is disjunctively ground resolvable w.r.t.  $r$  **then**  
 | | **AddRule**( $gr(r', r), P$ );  
 | **end**  
**end**  
**return**  $P$ ;

---

$r$  subsumes  $r'$ , written  $r \prec r'$ , if  $hd(r)\theta \subseteq hd(r')$  and  $bd(r)\theta \subseteq bd(r')$  for some substitution  $\theta$ .

## Definition (disjunctive ground resolution)

Let  $r$  and  $r'$  be two rules such that

- (a)  $l \in bd(r)$  and  $\bar{l} \in bd(r')$ ,
- (b)  $bd(r') \setminus \{\bar{l}\} \subseteq bd(r) \setminus \{l\}$ , and
- (c)  $hd(r') \subseteq hd(r)$ .

The (*disjunctive*) *ground resolvent*  $gr(r, r')$  of  $r$  w.r.t.  $r'$  on  $l$  is:

$$hd(r) \leftarrow bd(r) \setminus \{l\}.$$

Given the following two rules  $r_1, r_2$ , to compute  $gr(r_1, r_2)$ :

$$r_1 : p \vee q \leftarrow p \wedge q \wedge \neg s,$$

$$r_2 : q \leftarrow q \wedge s.$$

$$gr(r_1, r_2) = p \vee q \leftarrow p \wedge q.$$

## Definition (combined resolution)

Let  $r_1, \dots, r_k$  and  $r$  be the following rules:

$$r_1 : hd(r_1) \leftarrow bd^+ \cup \neg(bd^- \cup \{b_1\}),$$

$$\vdots$$

$$r_k : hd(r_k) \leftarrow bd^+ \cup \neg(bd^- \cup \{b_k\}),$$

$$r : hd(r) \leftarrow bd^+ \cup \{b_i \mid 1 \leq i \leq k\} \cup \neg bd^{-'}, \text{ such that}$$

- (a)  $bd^- \cap \{b_i \mid 1 \leq i \leq k\} = \emptyset$ , and
- (b)  $hd(r_i) \subseteq hd(r)$  for every  $i$  ( $1 \leq i \leq k$ ).

The *combined resolvent*  $cr(r, r_1, \dots, r_k)$  of  $r, r_1, \dots, r_k$  is:

$$hd(r) \leftarrow bd^+ \cup \neg(bd^- \cup bd^{-'}). \quad (1)$$

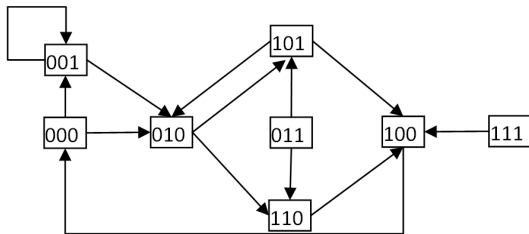


Table: Execution of LFDT with  $B = \emptyset$  and  $E$

step	$I \rightarrow J$	operation	rule	ID	$P$
1	$\epsilon \rightarrow r, \epsilon \rightarrow q$	$r_{q \vee r}^\epsilon$	$q \vee r \leftarrow \neg p \wedge \neg q \wedge \neg r$	1	1
2	$p \rightarrow \epsilon$	$r_\epsilon^p$			1
3	$q \rightarrow pr, q \rightarrow pq$	$r_p^q$	$p \leftarrow \neg p \wedge q \wedge \neg r$	2	1,2
		$r_{p \vee r}^q$	$p \vee r \leftarrow \neg p \wedge q \wedge \neg r$	3	1,2
		$r_{p \vee q}^q$	$p \vee q \leftarrow \neg p \wedge q \wedge \neg r$	4	1,2
		$r_{q \vee r}^q$	$q \vee r \leftarrow \neg p \wedge q \wedge \neg r$	5	1,2,5
		$gr(1,5)$	$q \vee r \leftarrow \neg p \wedge \neg r$	6	2,6

We get the result:

$$P = \left\{ \begin{array}{l} q \vee r \leftarrow \neg p, \\ p \vee q \leftarrow p \wedge \neg q \wedge r, \\ p \leftarrow q. \end{array} \right\} \quad (2)$$

A logic program  $P$  is *complete* for  $E$  w.r.t.  $B$  if

$\{J \mid \langle I, J \rangle \in E\} \subseteq T_{B \cup P}^d(I)$  for any  $I \in E^i$ , it is *sound* for  $E$  if  
 $T_{B \cup P}^d(I) \subseteq \{J \mid \langle I, J \rangle \in E\}$  for any  $I \in E^i$ .

## Theorem

The algorithm **LFDT** is sound and complete (with disjunctive ground resolution, combined resolution, and/or subsumption reduction).

What we did:

- the **LFDT** inductive learning framework for a restricted ABN
- that is both sound and complete.

Future work:

- Implementation **LFDT**,
- Experimentation,
- refinement the **LFDT** algorithm in a way of **LF1T**.