

# PERSPECTIVES ON LOGIC-BASED APPROACHES FOR REASONING ABOUT ACTIONS AND CHANGE

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# MOTIVATIONS

- Explore alternative approaches to encode action languages (e.g.,  $\mathcal{B}$ ) in different logic programming paradigms
  - ▶ ASP
  - ▶ CLP(FD)
- Comparative experimental performance testing
- Learn lessons from CLP(FD) encoding and raise constraints to the action language— $\mathcal{B}^{MV}$

# OUTLINE

1 ACTION DESCRIPTION LANGUAGES

2 ASP ENCODING

3 CLP ENCODING

4 MULTIVALUED LANGUAGES

5 EXPERIMENTAL RESULTS

6 DISCUSSION

7 ACKNOWLEDGMENTS

# ACTION DESCRIPTION LANGUAGES

Research on planning requires the resolution of two key problems [Lifschitz]:

- Declarative and Elaboration Tolerant **Languages** to describe planning domains
- Efficient and Scalable Reasoning **Algorithms**

Action Description Languages are formal models to represent knowledge on actions and change (e.g.,  $\mathcal{A}$  and  $\mathcal{B}$  [Gelfond and Lifschitz])

Specifications are given through declarative assertions that permit

- to describe actions and their effects on states
- to express queries on the underlying transition system

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# ACTION DESCRIPTION LANGUAGES

## PLANNING

A *planning domain*  $\mathcal{D}$  can be described through an **action domain description**, which defines the notions of

**FLUENTS** i.e., *variables* describing the state of the world, and whose value can change

**STATES** i.e., possible configurations of the domain of interest: an assignment of values to the fluents

**ACTIONS** that affect the state of the world, and thus cause the transition from a state to another

A **Planning Problem**  $P = \langle \mathcal{D}, I, O \rangle$  includes

- Description (complete or partial) of the **Initial** state
- Description of the **Final** state

# THE LANGUAGE $\mathcal{B}$

## ACTION DESCRIPTION

Let  $a$  be an action and  $\ell$  be a Boolean literal. We have:

- Executability conditions:

`executable(a, [list-of-preconditions])`

asserting that the given preconditions have to be satisfied in the current state for the action  $a$  to be executable

- Dynamic causal laws:

`causes(a,  $\ell$ , [list-of-preconditions])`

describes the effect (the fluent literal  $\ell$ ) of the execution of action  $a$  in a state satisfying the given preconditions

- Static causal laws:

`caused([list-of-preconditions],  $\ell$ )`

describes the fact that the fluent literal  $\ell$  is true in a state satisfying the given preconditions

# THE LANGUAGE $\mathcal{B}$

## GOAL

- *Initial state*

$\text{initially}(\ell)$

asserts that  $\ell$  holds in the initial state.

- *Goal*

$\text{goal}(\ell)$

asserts that  $\ell$  is required to hold in the final state.

# ASP ENCODING OF $\mathcal{B}$

Ideas from [Gelfond and Lifschitz 92]

We designed a Prolog program that translates an action description  $\mathcal{D}$ , with initial and final constraints  $\mathcal{O}$  and a plan length  $N$ , in an ASP program  $\Pi_{\mathcal{D}}(N, \mathcal{O})$ .

time(0..N)

fluent( $f$ ).

action( $a$ ).

literal( $F$ ):  $\neg$ fluent( $F$ ).

complement( $F, \neg F$ ).

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complement( $\neg F, F$ ).

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# ASP ENCODING OF $\mathcal{B}$

The predicate **holds(FluentLiteral,Time)** is defined using the axioms:

- executable( $a, [\ell_1^1, \dots, \ell_{r_1}^1], \dots, \text{executable}(a, [\ell_1^h, \dots, \ell_{r_h}^h]):$

**possible**( $a, T$ ):  $\neg \text{time}(T), \text{holds}(\ell_1^1, T), \dots, \text{holds}(\ell_{r_1}^1, T).$

...

**possible**( $a, T$ ):  $\neg \text{time}(T), \text{holds}(\ell_1^h, T), \dots, \text{holds}(\ell_{r_h}^h, T).$

- **Static causal laws**: caused( $[\ell_1, \dots, \ell_r], \ell$ ):

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- **Dynamic causal laws**: causes( $a, \ell, [\ell_1, \dots, \ell_r]$ ):

$\text{holds}(\ell, T + 1) : - \text{time}(T), \text{occ}(a, T),$   
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# ASP ENCODING OF $\mathcal{B}$

- State consistency:

: –time( $T$ ), fluent( $F$ ), holds( $F$ ,  $T$ ), holds(neg( $F$ ),  $T$ ).

- Frame problem:

holds( $L$ ,  $T + 1$ ) : – time( $T$ ), literal( $L$ ), holds( $L$ ,  $T$ ),  
complement( $L$ ,  $L_1$ ), **not** holds( $L_1$ ,  $T + 1$ ).

- Initial state and goal:

holds( $L$ , 0): –initially( $L$ ). : –goal( $L$ ), **not** holds( $L$ , N).

- One action per time:

1{occ( $A$ ,  $T$ ) : action( $A$ )}1: –time( $T$ ),  $T < N$ .  
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In absence of static laws a simplified mapping has been implemented  
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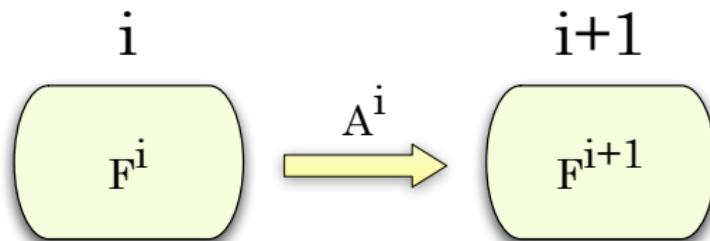


# MODELING $\mathcal{B}$ IN CLP(FD)

- Action descriptions are mapped to finite domain constraints
- Constrained variables are introduced for fluents and action occurrences
- Executability conditions and causal laws are rendered by imposing constraints
- Solutions of the constraints identify plans and trajectories

# MODELLING $\mathcal{B}$ IN CLP(FD)

## MAIN IDEA



For all states  $s_0, \dots, s_N$

- Every fluent  $f$  is represented as a Boolean variable  $F^i$ .
- Every action  $a$  is represented as a Boolean variable  $A^i$  ( $i < N$ ).

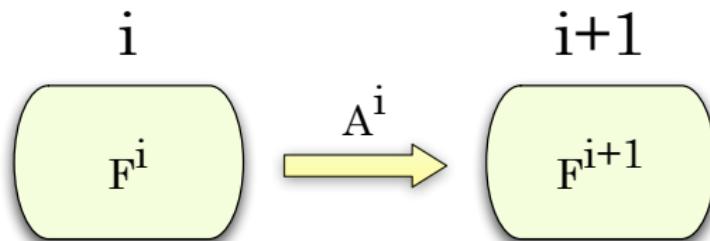
$$F^{i+1} = 1 \Leftrightarrow$$

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# MODELLING $\mathcal{B}$ IN CLP(FD)

## SOME CONSTRAINTS

$$F^{i+1} = 1 \equiv \text{PosFired}_f^i = 1 \vee (\text{NegFired}_f^i = 0 \wedge F^i = 1) \quad (1)$$

$$\text{PosFired}_f^i = 0 \vee \text{NegFired}_f^i = 0 \quad (2)$$

$$\text{PosFired}_f^i = 1 \equiv \text{PosDyn}_f^i = 1 \vee \text{PosStat}_f^{i+1} = 1 \quad (3)$$

$$\text{PosDyn}_f^i = 1 \equiv \bigvee_{j=1}^m (\hat{\alpha}_j^i \wedge A_{t_j}^{i+1} = 1) \quad (4)$$

$$\text{PosStat}_f^i = 1 \equiv \bigvee_{j=1}^h \hat{\delta}_j^i \quad (5)$$

$$\text{NegFired}_f^i = 1 \equiv \text{NegDyn}_f^i = 1 \vee \text{NegStat}_f^{i+1} = 1 \quad (6)$$

$$\text{NegDyn}_f^i = 1 \equiv \bigvee_{j=1}^p (\hat{\beta}_j^i \wedge A_{z_j}^{i+1} = 1) \quad (7)$$

$$\text{NegStat}_f^i = 1 \equiv \bigvee_{j=1}^k \hat{\gamma}_j^i \quad (8)$$

---

$$A_j^{i+1} = 1 \rightarrow \bigvee_{j=1}^q \hat{\eta}_{r_j}^i \quad (9)$$

$$\sum_{a_j \in \mathcal{A}} A_j^i = 1 \quad (10)$$

# THE LANGUAGE $\mathcal{B}^{MV}$

**FLUENTS:** introduced through *domain declarations*:

$\text{fluent}(f, v_1, v_2)$ ,  $\text{fluent}(f, \text{Set})$

**ANNOTATED FLUENTS:** modeling backward references:

$f^{-a}$  with  $a \in \mathbb{N}$

**FLUENT EXPRESSIONS:**

$\text{FE} ::= n \mid \text{AF} \mid \text{abs}(\text{FE}) \mid \text{FE}_1 \oplus \text{FE}_2 \mid \text{rei}(\text{FC})$

with  $\oplus \in \{+, -, *, /, \text{mod}\}$

**FLUENT CONSTRAINTS:**

$\text{FC} ::= \text{FE}_1 \text{ op } \text{FE}_2$

with  $\text{op} \in \{\text{eq}, \text{neq}, \text{geq}, \text{leq}, \text{lt}, \text{gt}\}$

# THE LANGUAGE $\mathcal{B}^{MV}$

## ACTION DESCRIPTION SPECIFICATION

- Dynamic causal laws

causes (a, C<sub>1</sub>, C)

- Static causal laws

caused (C, C<sub>1</sub>)

- Executability conditions

executable (a, C)

where a is an action, C<sub>1</sub> is a fluent constraint, and C a conjunction of fluent constraints.

# THE LANGUAGE $\mathcal{B}^{MV}$

## MAIN ADVANTAGES

- It allows the compact representation of numerical domains.
- An encoding in  $\mathcal{B}$  is still possible, but the number of fluents explodes
- Consider

$$\text{causes}(a, f = f^{-1} + 1, [])$$

with  $\text{dom}(f) = [1..100]$ .

- ▶ In  $\mathcal{B}$ :

$$\text{causes}(a, f_2, [f_1]). \quad \dots \quad \text{causes}(a, f_{100}, [f_{99}])$$

- ▶ Alternative encodings (e.g., bit-based) incurs analogous problems

- The CLP(FD) mapping requires minor changes.

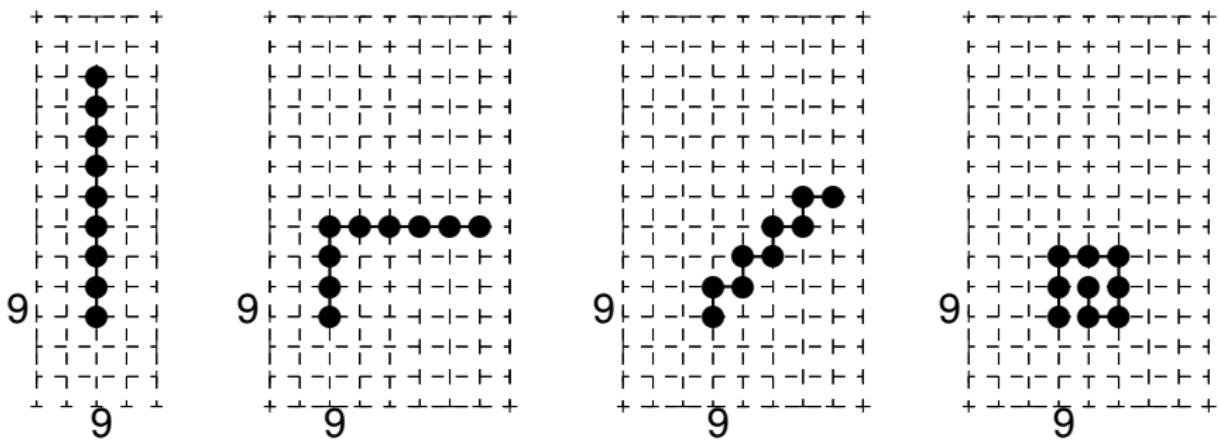
# EXPERIMENTAL COMPARISON

## BENCHMARKS USED

- *Hydraulic planning* (by Michael Gelfond et al., ASP 2009)
- *Peg Solitaire* (ASP 2009)
- *Sam Lloyd's 15 puzzle* (ASP 2009)
- *Towers of Hanoi* (ASP 2009)
- The *trucks* domain from the (IPC5)
- A generalized version of the classical *3-barrels* problem
- The *Gas Diffusion problem*
- The *reverse folding problem*.
- The *Tangram* puzzle.

# EXPERIMENTAL COMPARISON

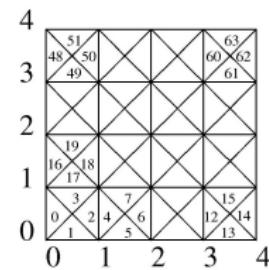
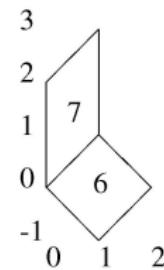
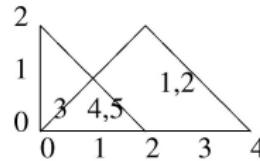
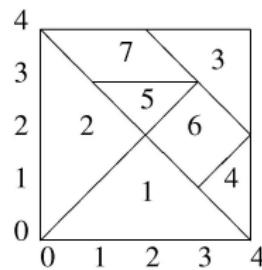
BENCHMARKS USED: REVERSE FOLDING PROBLEM



Four foldings with  $k=9$ : The initial (straight line) folding, the result of a clockwise pivot move on the 4th element, a zigzag folding, and a spiral folding.

# EXPERIMENTAL COMPARISON

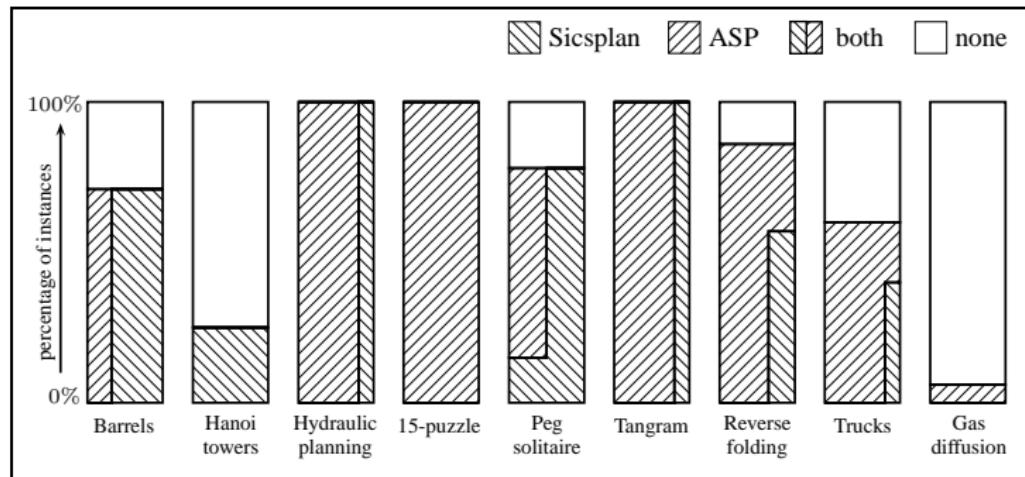
BENCHMARKS USED: TANGRAM



Tangram solution, “Base” position of the seven blocks, and space discretization using triangles.

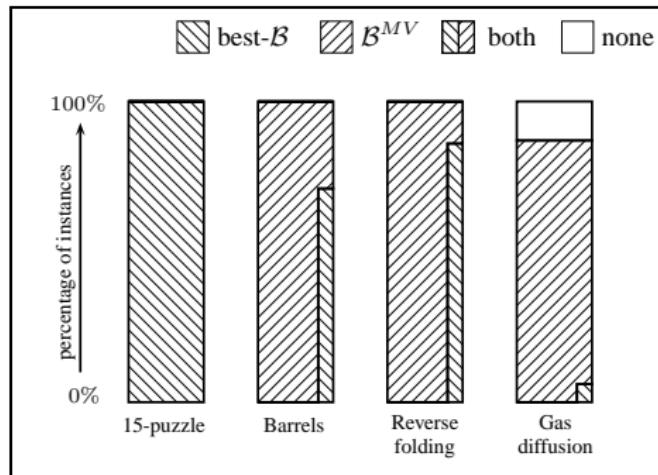
# EXPERIMENTAL COMPARISON

ASP (CLASP) vs CLP (SICSTUS)



# EXPERIMENTAL COMPARISON

$\mathcal{B}$  BEST vs  $\mathcal{B}^{MV}$ (CLP)



# DISCUSSION

## OTHER THINGS ALREADY DONE

- Other constraints (*cost* of each action, fluent, and about the global cost of a plan, capabilities of plan minimization, temporal fluent literals, integrity constraints, . . .) [TPLP 2010]
- Multiagent reasoning (with central coordination) [FUIN 2011]
- Nicer syntax

# DISCUSSION

## THINGS TO DO

- Looking into the future: reasoning/constraints referring to future steps of the trajectory
- Preferences (qualitative, quantitative) in action execution
- Distributed multiagent reasoning
- Heuristics

WEBSITE <http://www.dimini.uniud.it/dovier/CLPASP>

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# GRAZIE MICHAEL

