Development of a Belief Merging Framework for dlvhex

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May 31, 2010

Outline

- 1 Motivation
- 2 Reasons for Incompatibility
- 3 Task Definition
- 4 Architecture of the Belief Merging Framework
- 5 Using the Framework Hands-on
- 6 Application Scenario
- 7 Summary

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- Many different merging techniques

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- fusion of business databases

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Examples

- relational databases
- object-orientated databases
- RDF ontologies
- logic programs

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Example

<u>name</u> is the primary key; a constraint forces the height to be unique for each person.

(a) name height
Marge 1,78m
Homer 1,82m
Bart 1,67m

(b)	name	height
	Marge	1,78m
	Homer	1,82m
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Example

Merging of address tables, one with and one without abbreviations

Logic Programs as Belief Bases

Given

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common signature vector of mapping functions merging operators merging plan

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- answer sets stay semantically equivalent!

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$$\circ_{i}^{n,m}: \underbrace{2^{\mathcal{A}} \times \dots \times 2^{\mathcal{A}}}_{n \text{ times}} \times \underbrace{\mathcal{D}_{1} \times \dots \times \mathcal{D}_{m}}_{additional \ parameters} \rightarrow 2^{\mathcal{A}}$$

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Example for a merging operator

The union operator $\circ_{U}^{2,0}$ is defined as follows:

$$\circ^{2,0}_{\cup}:2^{\mathcal{A}}\times2^{\mathcal{A}}\rightarrow2^{\mathcal{A}}$$

$$\circ_{\cup}^{2,0}(SAS_1,SAS_2) = \{AS_1 \cup AS_2 | AS_1 \in SAS_1, AS_2 \in SAS_2, AS_1 \cup AS_2 \not\models \bot\}$$

 $(\circ_{\cup}^2$ is binary, no additional parameters)

Merging Plans

A merging plan is hierarchical and defines

- the order
- of operators
- to be applied on which belief bases

Merging Plans

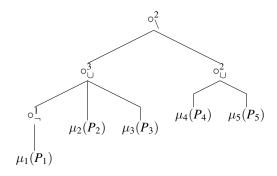
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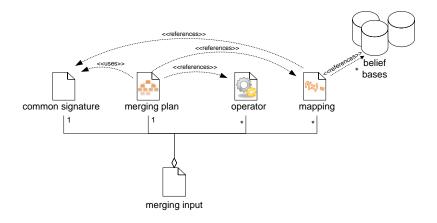
The result

the set of answer sets delivered by the topmost operator

Example merging plan

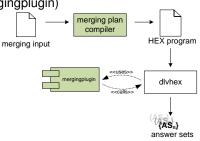


Merging Input



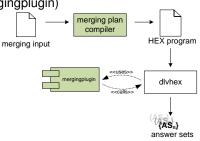
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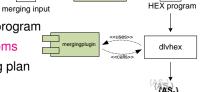
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merging plan compiler

translates merging plan into HEX program



answer sets = result of the merging plan



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mergingplugin <uses>> dlvhex

HEX program

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defines external atoms for:

- calling of nested HEX programs
- calling of merging operators

Rapid prototyping

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- 1 Define your merging task "merging.mp"
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Typical call

Command-line:

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$ mpcompiler merging.mp | dlvhex --filter=a,b,c --
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- Command-line:
 - \$ mpcompiler merging.mp | dlvhex --filter=a,b,c --
- Alternatively:
 - \$ dlvhex --merging --filter=a,b,c merging.mp

Merging plan language: merging.mp

```
[common signature]
   predicate: a/0:
   predicate: b/0:
   predicate: c/0:
   predicate: p/1:
   predicate: q/3;
[belief base]
   name:bb1;
   mapping: "some_rule."; % query external source here
   mapping: "q(X, Y, Z) := &rdf["..."](X, Y, Z).";
[belief base]
   name:bb2:
   source: "some_program.hex"; % or within this program
```

```
Merging plan language: merging.mp (ctn'd.)
    [merging plan]
       operator: setminus;
            operator: union;
                  operator: neg;
                     {bb1};
               {bb2};
               {bb3};
         operator: union;
            {bb4};
            {bb5};
          };
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Definition of implemented version

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$$d(A, P_i) = \min_{J \in AS(P_i)} \underline{d}(A, J)$$

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$$D^d(A,K) = D(d(A,P_1),\ldots,d(A,P_n))$$

$$\bullet \circ^n(K) = \arg\min_{G \in \mathcal{A}: consistent} D^d(G, K)$$

Fault Diagnosis

Finding an explanation for some observation

Definition

Propositional abduction problem (PAP): $\mathscr{P} = \langle V, H, M, T \rangle$

- V is a finite set of propositional variables
- \blacksquare $H \subseteq V$ is a set of hypothesis
- lacksquare $M \subseteq V$ is the set of manifestations
- T is a consistent theory

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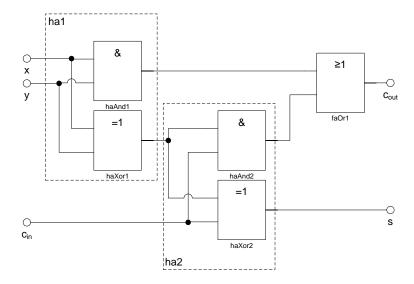
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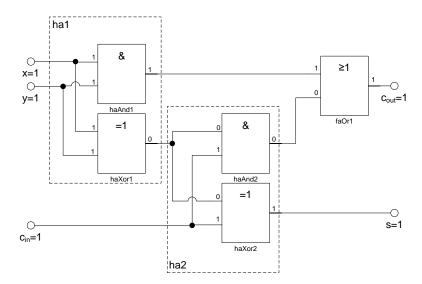
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Challenge: *multiple* experts with *different* explanations S_i **Task:** Finding a group decision S_G s.t.

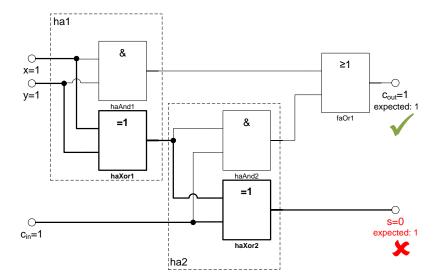
- \blacksquare S_G is a solution to \mathscr{P}
- \blacksquare S_G is as similar to $S_i \forall i$ as possible



Full Adder - Example interpretation



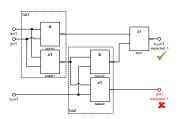
Full Adder - Malfunctioning



Implemented as logic program "fulladder.dl" (theory) with observations "fault.obs"

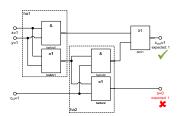
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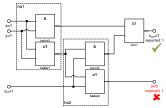
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- **3** ab(haAnd1).ab(haAnd2).ab(haXor2).ab(faOr1). **no** ab(haXor2)!



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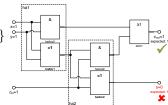
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(Minimal) Solutions

$$AS(P_{J_2}) = \{\{ab(haXor2)\}\}$$

$$AS(P_{J_3}) = \{\{ab(haXor1)\}\}$$



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I = individual explanation

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$$\neg a \in I \land \neg a \notin G$$

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```
[common signature]
predicate:ab/1;
[belief base]
name: juror1;
dlvargs: "-FRmin fulladder.dl abnormal1.hyp fault.obs";
[belief base] name: juror2; ...
[belief base] name: juror3; ...
[merging plan]
operator: dalal; aggregate: "sum";
penalize: "ignoring";
constraints: "fulladder.dl"; constraints: "fault.obs";
{juror1}; {juror2}; {juror3};
```

Full Adder - Group Decision

Individual explanations

Possible group explanations

$$\begin{array}{lll} \textbf{1} & AS(P_{J_1}) = & \textbf{1} & E_1 = \\ & \left\{ \{ab(haXor1)\}, \{ab(haXor2)\} \right\} & \left\{ ab(haXor1), ab(haXor2) \right\} \\ \textbf{2} & AS(P_{J_2}) = \left\{ \{ab(haXor2)\} \right\} & \textbf{2} & E_2 = \left\{ ab(haXor1) \right\} \\ \end{array}$$

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Possible group explanations

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2
$$AS(P_{J_2}) = \{\{ab(haXor2)\}\}$$
 2 $E_2 = \{ab(haXor1)\}$

$$AS(P_{J_3}) = \{\{ab(haXor1)\}\}\$$
 $E_3 = \{ab(haXor2)\}$

Distances to Individuals

Penalizing ignoring only

	$AS(P_{J_1})$	$AS(P_{J_2})$	$AS(P_{J_3})$	Sum
$\mathbf{E_1}$	0	0	0	0 ⇐
E_2	0	1	0	1
E_3	0	0	1	1

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Possible group explanations

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$$AS(P_{J_1}) =$$
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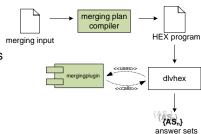
Distances to Individuals

Penalizing ignoring and unfounded group beliefs ($|I\Delta G|$)

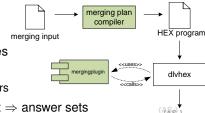
	$AS(P_{J_1})$	$AS(P_{J_2})$	$AS(P_{J_3})$	Sum
E_1	1	1	1	3
$\mathbf{E_2}$	0	2	0	2 ←
$\mathbf{E_3}$	0	0	2	2 ⇐

Summary

■ Task: Merging of several belief bases

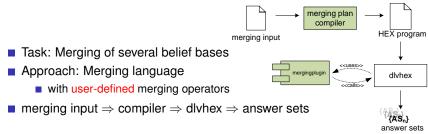


Summary



- Task: Merging of several belief bases
- Approach: Merging language
 - with user-defined merging operators
- lacktriangledown merging input \Rightarrow compiler \Rightarrow dlvhex \Rightarrow answer sets

Summary



Advantages

- Develop merging operators only once or select one of the preinstalled ones (like Dalal)
- No need for manual re-merging after each change of the setting
- Try out several operators and evaluate which behaves best
- No routine tasks (like information flow between sources)
- User can focus on development and optimization of merging procedures in narrower sense!