Integrating Answer Set Programming with Object-oriented Languages

Jakob Rath and Christoph Redl

jakob.rath@student.tuwien.ac.at, redl@kr.tuwien.ac.at



TECHNISCHE UNIVERSITÄT WIEN Vienna University of Technology



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Outline

1 Motivation

- 2 General Approach
- 3 Input and Output Specification Language
- 4 Implementation and Applications
- 5 Conclusion

Answer Set Programming

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 - graphical user interfaces
 - presentation of results
 - interfaces to data sources
 - etc.

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Limitations

- Typical end-user applications contain components which cannot be (easily) solved in ASP:
 - graphical user interfaces
 - presentation of results
 - interfaces to data sources
 - etc.
- Realizing such components is in the domain of traditional object-oriented (OOP) languages.

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- To this end, object-oriented code
 - adds input as facts,
 - 2 evaluates the ASP program, and
 - 3 interprets the answer sets.
- But: An implementation from scratch is similar for most applications ⇒ repetitive work.

Contribution

- The ASP program is extended with annotations which specify input/output.
- Input specifications define how objects are mapped to facts.
- Output specifications define how answer sets are mapped back to objects.
- Based on annotations, the integration with object-oriented code is automated.

Contribution

- The ASP program is extended with annotations which specify input/output.
- Input specifications define how objects are mapped to facts.
- Output specifications define how answer sets are mapped back to objects.
- Based on annotations, the integration with object-oriented code is automated.
- In contrast to existing approaches, ours is independent of a concrete OOP language.
- We provide a prototypical implementation PY-ASPIO for Python.

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Overview

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- Approach: the ASP program is annotated with input/output specifications. Annotations are added as special comments of form %! to the ASP code.
- We then provide an interpreter library for evaluating ("calling") such an annotated program:
 - it takes an annotated ASP program and a list of input parameters (objects) as input, and
 - returns a set of objects (corresponding the results of the ASP program).

Evaluation

More precisely, the interpreter library performs the following tasks:

- **1** Parameters v_1, \ldots, v_n are converted to facts according to input specification ι .
- 2 These facts along with the ASP program *P* are passed to the ASP solver.
- **3** The answer sets are mapped to objects O according to output specification ω .

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Language independence

- The specification language is largely independent of a concrete OOP language
 - \Rightarrow porting the interpreter library to other OOP languages is easily possible.
 - \Rightarrow the same annotated program can be used with multiple OOP languages.

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- Currently, we provide a prototypical implementation PY-ASPIO for Python.

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- Data is organized in classes, which consist of named attributes and methods.
- 2 The language must provide the classes str and int.
- 3 We presuppose the following collection types:
 - Set $\langle T \rangle$:

a collection of unique objects of type T.

Dictionary $\langle K, V \rangle$:

a mapping from objects of type K (the keys) to objects of type V (the values).

• $Tuple\langle T_1,\ldots,T_n\rangle$:

an ordered list of fixed length *n*, where the component at position *i* is of type T_i for $1 \le i \le n$.

• Sequence $\langle T \rangle$:

a finite ordered sequence containing objects of type T, where elements are addressable by an integer index.

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Input Specification

Example (3-Colorability)

Assume we want to use a 3-colorability program from our object-oriented code.

Let the graph in the object-oriented code be represented by sets of nodes and edges, where

- nodes are instances of class Node with the attribute label as a unique string identifiying the node, and
- edges are instances of class Edge, where the attributes first and second are the nodes at both ends of the edge.

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Mapping of the input graph to predicates vertex and edge and problem encoding:

```
%! INPUT (Set<Node> nodes, Set<Edge> edges) {
%! vertex(n.label) for n in nodes;
%! edge(e.first.label, e.second.label) for e in edges; }
color(X,r) v color(X,g) v color(X,b) :- vertex(X)
:- color(X,C), color(Y,C), edge(X,Y)
```

Input Specification Language

Language Definition

In general, an input specification ι is of the form **INPUT** $(t_1 v_1, \ldots, t_n v_n) \{s_1; s_2; \ldots s_k; \}$

where

 v_1, \ldots, v_n are input parameters of types t_1, \ldots, t_n ,

 s_1, \ldots, s_k are predicate specifications defined as follows.

Each predicate specification s_i for $1 \le i \le k$ is of form $p(x_1, \dots, x_m)$ for w_1 in $y_1 \dots$ for w_ℓ in y_ℓ (1)

where

 $p \dots$ predicate symbol, x_1, \dots, x_m are objects of any type, w_1, \dots, w_ℓ are (iteration) variables, y_1, \dots, y_ℓ are collections.

Here, the constructs for w_i in y_i are iteration clauses which are used to let w_i iterate over the contents of a collection object y_i .

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%! INPUT (Set<Node> nodes, Set<Edge> edges) {
%! vertex(n.label) for n in nodes;
%! edge(e.first.label, e.second.label) for e in edges; }
%! OUTPUT {
%! colorednodes = set { query: color(X,C);
%! content: ColoredNode(X,C); }; }
color(X,r) v color(X,g) v color(X,b) :- vertex(X)
:- color(X,C), color(Y,C), edge(X,Y)
```

The output is a set containing instances of ColoredNode, which are created by calling the constructor of the class with arguments X and C for each atom color(X,C) in the current answer set.

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Example (3-Colorability (cont'd))

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This program can be called with two parameters of types Set<Node> resp. Set<Edge> and its output is a set of type Set<ColoredNode>.

Output Specification Language

Language Definition

Building blocks are (possibly nested) expressions which transform atoms, sets of atoms, and/or results of subexpressions to objects.

Support types:

- Basic Expressions are integer and string constants *e*.
- Collection Expressions are of one of the following forms:
 - **set** { query: q; content: e; }
 - sequence { query: q; index: i; content: e; }
 - dictionary { query: q; key: k; content: e; }
- Composite Expressions are instances of custom classes of the object-oriented language.
- An output specification ω is then of the form

OUTPUT
$$\{w_1 = e_1; \dots w_k = e_k; \}$$

where

 w_1, \ldots, w_k are pairwise distinct attributes and e_1, \ldots, e_k are expressions.

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Implementation

PY-ASPIO

- We implemented an interpreter for our specification language for Python.
- <u>ASP</u> Interface to <u>Object</u>-oriented programs for Python.
- Available at https://github.com/hexhex/py-aspio.
- dlvhex is used as ASP solver (switching to other solvers is easily possible).

Implementations for other object-oriented languages are left for future work.



Example: Using the Library from Python

Suppose the ASP program from above is stored in file coloring.dl.

Example (3-Colorability (cont'd))

```
from collections import namedtuple
import aspio
# Define classes and create sample data
Node = namedtuple('Node', ['label'])
ColoredNode = namedtuple('ColoredNode', ['label', 'color'])
Edge = namedtuple('Edge', ['first', 'second'])
a, b, c = Node('a'), Node('b'), Node('c')
nodes = {a, b, c}
edges = {Edge(a, b), Edge(a, c), Edge(b, c)}
# Register class names with aspio
aspio.register.dict(globals())
```

Load ASP program and input/output specifications from file prog = aspio.Program(filename='coloring.dl')

Iterate over all answer sets
for result in prog.solve(nodes, edges):
 print(result.colored_nodes)

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Upcoming research application

- ASP extensions support epistemic negation (quantification over answer sets).
- The evaluation algorithm for such programs is based on evaluating sets of programs and reasoning about their answer sets.
- An implementation of epistemic ASP based on PY-ASPIO is work in progress.

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Existing approaches

- Focus on a particular language, cf. e.g. [?], [?], [?].
- Tweety [?] and PyASP (https://pypi.python.org/pypi/pyasp) provide only a generic atom-based interface, but no customizable mapping.

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Future work

- Language extensions.
- Implementation for other object-oriented languages.

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