

Specifying Credal Sets With Probabilistic Answer Set Programming

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Center for AI (C4AI)



- Main funding by FAPESP and IBM.
- About 90 faculty associated, 100 students.



Probabilistic Answer Set Programming

- Logic Programming + Constraint Satisfaction
+ Uncertain Reasoning
= Probabilistic Answer Set Programming
- Example application: Argumentation under uncertainty (e.g. driving public discourse on climate change).

Probabilistic Answer Set Programming

0.3::a.

c :- not d.

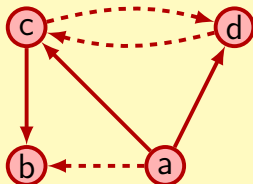
d :- not c.

c :- a.

d :- a.

b :- a.

b :- not a, c.



Probabilistic Answer Set Programming

0.3::a.

c :- not d.

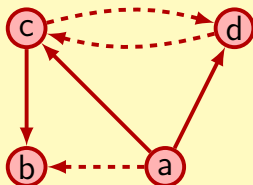
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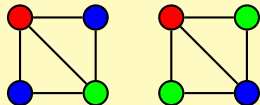
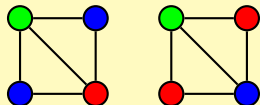
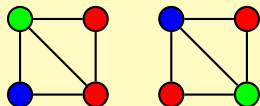


- Nondisjunctive acyclic \Rightarrow Bayesian networks.
- Nonstratified or disjunctive \Rightarrow belief functions.
- **This work:** How to extend to more general imprecise probability models?

Example: Coloring a random graph

```
node(1). node(2). node(3). node(4).  
0.2::edge(1,2). 0.3::edge(2,3).  
0.4::edge(3,4). 0.5::edge(1,4).  
0.6::edge(1,3).
```

```
edge(X,Y) :- edge(Y,X).  
conflict(X,Y) :- not conflict(X,Y),  
    edge(X,Y), color(X,C), color(Y,C).  
color(X,red); color(X,blue); color(X,green) :- node(X).
```



This Work: Extended PASP

Interval-valued PASP

```
[0.1,0.3]::red(X); [0.2,0.4]:: green(X); [0.4,0.6]::blue(X) :- node(X).
```

Parametrized PASP

```
W::win(X); D::draw(X); L::loose(X) :- match(X),  
                                     W > D, W > L, L <= 0.3.
```


Expressivity I

Theorem (Standard PASP capture belief functions)

*Every **infinitely monotone lower probability over a finite domain** can be specified by a probabilistic answer set program with precise probabilities in size proportional to the number of focal sets of its m -function characterization.*

Theorem (Interval-Valued PASP capture finite credal sets)

*Every **finitely-generated credal set over a finite domain** can be represented by an acyclic and positive probabilistic logic program with a single vacuous interval-valued annotated disjunction and a set of precise annotated disjunction.*

Expressivity II

Theorem (Simplifying Interval-Valued PASP)

Any interval-valued probabilistic answer set program with interval-valued annotated disjunctions can be converted into an equivalent program containing only interval-valued probabilistic facts (Bernoulli vars) and non-probabilistic rules. If the original program is acyclic (resp., nondisjunctive), the resulting program is also acyclic (resp., nondisjunctive).

Theorem (PASP semantics through credal networks)

*The semantics of an **acyclic parametrized probabilistic answer set program** is given by a credal network; if only probabilistic facts and nonprobabilistic rules appear, the network structure is the dependency graph of the program.*

Complexity

Theorem

*Deciding whether $\underline{Pr}(\text{atom}) \geq \gamma$ is NP^{PP} -**complete** in both interval-valued and parametrized probabilistic answer set programs.*

Inference

Compute $\underline{Pr}(a|b)$ by GBR (solve for μ using binary search):

$$\min_{Pr} Pr(a, b) - \mu Pr(b) = 0 \Leftrightarrow \min_{Pr} Pr(a, b) + \mu Pr(\neg b) = \mu$$

augment program with

query :- a,b.

mu::query :- not a, b.

To conclude...

- Probabilistic Answer Set Programming captures belief functions.
- **This work:** Extended language to capture any finite credal set.
- Interval-valued PASP implemented in dPASP.
- **Challenge:** Probabilistic inference is *too* costly.
 - Needed: approximate inference algorithms!
- In the works: connection with neural networks.