## Advances in Efficient Probabilistic Reasoning with Answer Set Semantics

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Probabilistic Stream Reasoning Approach

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Interest in quantitative reasoning on top of Stream Reasoning:

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Probabilistic Reasoning [Nickles and Mileo, 2014]

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Interest in quantitative reasoning on top of Stream Reasoning:

- Probabilistic Reasoning [Nickles and Mileo, 2014]
- wLARS [Eiter and Kiesel, 2020] for
  - Probabilistic Reasoning
  - Preferential Reasoning
  - Algebraic Model Counting

Probabilistic Stream Reasoning Approach

#### Probabilistic Reasoning via Knowledge Compilation

 "Compile" the logical theory into a *tractable circuit* representation like d-DNNF or SDD

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Probabilistic Stream Reasoning Approach

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- Logical connectives from the temporal domain
- Answer set semantics
  - $\hookrightarrow \mathsf{Our} \ \mathsf{recent} \ \mathsf{work}$

Probabilistic Stream Reasoning Approach

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 Goal: Perform cycle breaking and Clark completion in such a way that compilation and evaluation are fast

Probabilistic Stream Reasoning Approach

#### Treewidth I

#### Definition (Tree decomposition, Treewidth)

Let G be a graph. Then a tree decomposition is a pair  $(T, \chi)$ , where T is a tree and  $\chi$  is a labeling of V(T) by subsets of V(G)s.t.

- ▶ for all nodes  $v \in V(G)$  there is  $t \in V(T)$  s.t.  $v \in \chi(t)$ ;
- for every edge  $\{v_1, v_2\} \in V(E)$  there exists  $t \in V(T)$  s.t.  $v_1, v_2 \in \chi(t)$ ;

for all nodes v ∈ V(G) the set of nodes
 {t ∈ V(T) | v ∈ χ(t)} forms a (connected) subtree of T.
 The width of (T, χ) is max<sub>t∈V'</sub> |χ(t)| − 1. The treewidth of a
 graph is the minimal width of any of its tree decompositions.

Probabilistic Stream Reasoning Approach

# Treewidth II

CNF:

$$a \lor b$$
$$\neg b \lor c \lor d$$
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#### Treewidth and Knowledge Compilation

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- Tree decomposition-based variable selection performs well [Korhonen and Järvisalo, 2021]

Cycle-Breaking Approaches Our Cycle-Breaking

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  - ProbLog [Mantadelis and Janssens, 2010]: Resulting treewidth is O(k · 2<sup>|C|</sup>)

where |C| is the size of the largest strongly connected component (SCC) of the dependency graph of the program

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## Our Cycle-Breaking

- The largest SCC of the dependency graph may be large
- Observation: We can achieve a smaller increase when the cyclicity of the dependency graph is low
- Idea: Split the strongly connected components into subgraphs of low cyclicity

#### Component-Boosted Backdoor Size

## Definition (cbs(G))

Let G be a digraph. Then the component-boosted backdoor size of G, denoted cbs(G), is

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Intuitively, cbs(G) measures the *cyclicity* of G by decomposition into "easy to solve" subgraphs

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#### Main Result

#### Theorem

For every answer set program  $\Pi,$  there exists an equivalent program  $\Pi'$  such that

- 1. the answer sets are preserved bijectively
- 2.  $\Pi'$  is tight/acyclic
- 3. the treewidth of Π' is less or equal to k · cbs(DEP(Π)), where k is the treewidth of Π.

Benchmark Settings Results

#### Scenarios

- S1 Probabilistic reasoning: Computing probabilities for atoms of Problog programs
- S2 Counting (small number of solutions on average): Counting the number of different paths between stations in public transport networks
- S3 Counting (many solutions on average): Counting conflict-free extensions in abstract argumentation

### Solvers

- Problog, version 2.1.0.42, run with arguments "-k sdd"
- clingo, version 5.4.0, run with arguments "-q -n 0"
- Ip2sat+c2d: cycle breaking due to [Bomanson, 2017] followed by compilation using c2d [Darwiche, 2004]
- aspmc+c2d: our cycle breaking followed by compilation using c2d [Darwiche, 2004]

Benchmark Settings Results

#### Results S1



Benchmark Settings Results

#### Results S2



Benchmark Settings Results

#### Results S3

solver configuration	$\sum$	0-300	tw ranges 300-600	>600	unique	time[h]
aspmc+c2d	241	185	26	30	12	45.16
lp2sat+c2d	182	182	0	0	0	73.85
clingo	144	97	21	26	2	94.78

Conclusion & Outlook

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- Treewidth-awareness seems to be important for probabilistic reasoning
- Improved approach for probabilistic reasoning under ASP semantics
- Need to tackle the time domain!

   Gompile once and reuse: great!

   Already hard for one timepoint, how problematic for more?

Conclusion & Outlook



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arXiv preprint arXiv:1811.02944, 2018.



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Conclusion & Outlook

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