Generating Random Logic Programs Using Constraint Programming

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CP 2020





Probabilistic Logic Programs (PROBLOG)

"Smokers" (Domingos et al. 2008; Fierens et al. 2015)

```
0.2::stress(P):-person(P).
0.3::influences(P_1, P_2):-friend(P_1, P_2).
0.1::cancer_spont(P):-person(P).
0.3::cancer_smoke(P):-person(P).
    smokes(X):-stress(X).
    smokes(X):-smokes(Y), influences(Y, X).
    cancer(P):-cancer_spont(P).
    cancer(P) : -smokes(P), cancer_smoke(P).
    person(michelle).
    person(timothy).
    friend(timothy, michelle).
```

Applications





Moldovan et al. 2012

```
is_malignant(Case):-
    biopsyProcedure(Case, usCore),
    changes_Sizeinc(Case, missing),
    feature_shape(Case).

is_malignant(Case):-
    assoFinding(Case, asymmetry),
    breastDensity(Case, scatteredFDensities),
    vacuumAssisted(Case, yes).

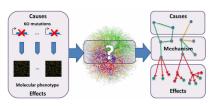
is_malignant(Case):-
    needleGauge(Case,9),
    offset(Case,14),
    vacuumAssisted(Case,yes).
```

Côrte-Real, Dutra, and Rocha 2017

Q1: In a group of 10 people, 60 percent have brown eyes. Two people are to be selected at random from the group. What is the probability that neither person selected will have brown eyes?

Q2: Mike has a bag of marbles with 4 white, 8 blue, and 6 red marbles. He pulls out one marble from the bag and it is red. What is the probability that the second marble he pulls out of the bag is white?

Dries et al. 2017



De Maeyer et al. 2013

Inference Algorithms and Knowledge Compilation Maps

NNF negation normal form

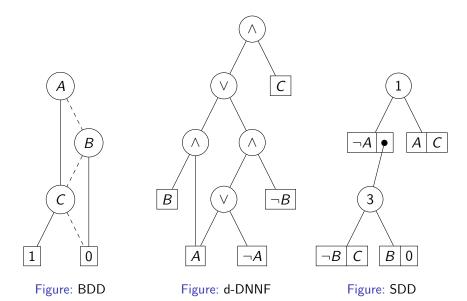
d-DNNF deterministic decomposable negation normal form

BDD binary decision diagrams

SDD sentential decision diagrams

k-Best only use the *k* most probable proofs

Example Diagrams for $C \wedge (A \vee \neg B)$



Anytime Inference in Probabilistic Logic Programs with T_p -Compilation

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Inference and learning in probabilistic logic programs using weighted Boolean formulas

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k-Optimal: a novel approximate inference algorithm for ProbLog

Joris Renkens · Guy Van den Broeck · Siegfried Nijssen

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ProbLog Technology for Inference in a Probabilistic First Order logic

Maurice Bruynooghe and Theofrastos Manta-lie and Angelika Kimmig and Bernd Gutmann and Joost Vennekens and Gerda Janssens and Luc De Raedt

```
0.2::stress(P):-person(P).
0.3::influences(P_1, P_2):-friend(P_1, P_2).
0.1::cancer_spont(P):-person(P).
0.3::cancer_smoke(P):-person(P).
    smokes(X) : - stress(X).
    smokes(X):-smokes(Y), influences(Y, X).
    cancer(P):-cancer_spont(P).
    cancer(P): - smokes(P), cancer_smoke(P).
    person(michelle).
    person(timothy).
    friend(timothy, michelle).
```

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0.2:stress(P):-person(P).
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                                                    predicates.
                                                     arities
0.3::cancer_smoke(P):-person(P).
    smokes(X):-stress(X).
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0.2::stress(P):-person(P).
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0.1::cancer\_spont(P):-person(P).
                                                    predicates.
                                                     arities
0.3::cancer\_smoke(P):-person(P).
                                                    variables
    smokes(X):-stress(X).
    smokes(X):-smokes(Y), influences(Y, X).
    cancer(P):-cancer_spont(P).
    cancer(P) := smokes(P), cancer_smoke(P).
    person(michelle).
    person(timothy).
    friend(timothy, michelle).
```

```
0.2::stress(P):-person(P).
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0.1::cancer\_spont(P):-person(P).

    predicates,

                                                      arities
0.3::cancer\_smoke(P):-person(P).
                                                    variables
    smokes(X):-stress(X).
                                                      constants
    smokes(X):-smokes(Y), influences(Y, X).
    cancer(P):-cancer_spont(P).
    cancer(P) := smokes(P), cancer_smoke(P).
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```

- predicates, arities
- variables
- constants
- probabilities

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- predicates, arities
- variables
- constants
- probabilities
- length

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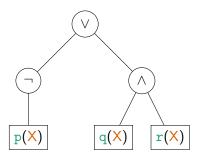
- predicates, arities
- variables
- constants
- probabilities
- length
- complexity

Clauses As Trees

$$\neg p(X) \lor (q(X) \land r(X))$$

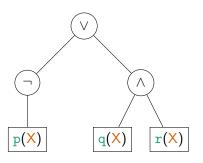
Clauses As Trees

$$\neg p(X) \lor (q(X) \land r(X))$$



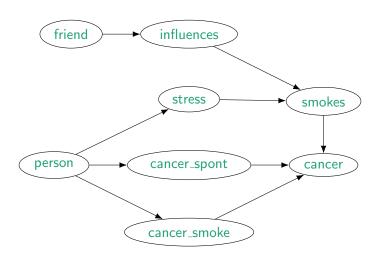
Clauses As Trees

$$\neg p(X) \lor (q(X) \land r(X))$$

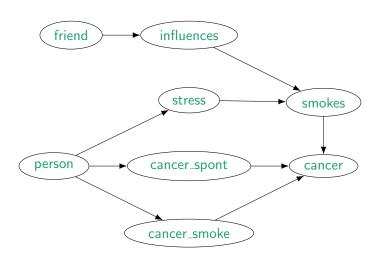


0	0	0	1	2	2
V	_	Λ	p(X)	q(X)	r(X)

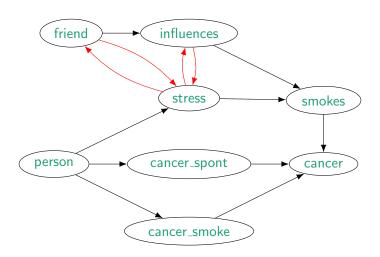
Predicate Dependency Graph



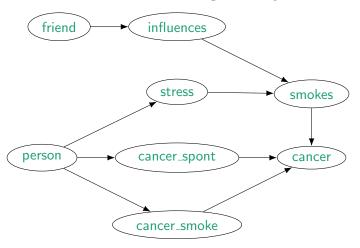
Independence: influences ⊥ stress



Independence: influences ⊥ stress

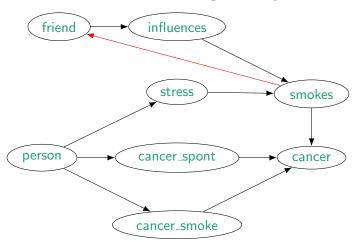


Stratification and Negative Cycles



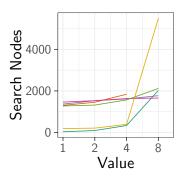
0.1::friend(X, Y):-\+smokes(Y).

Stratification and Negative Cycles



0.1::friend(X, Y):-\+smokes(Y).

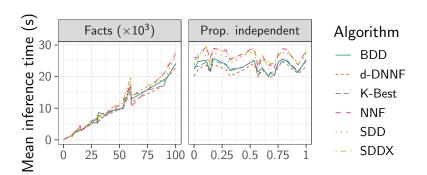
Scalability



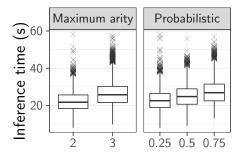
Variable

- The number of predicates
- Maximum arity
- The number of variables
- The number of constants
- The number of additional clauses
- The maximum number of nodes

Properties of Programs vs. Inference Algorithms



Properties of Programs vs. Inference Algorithms



Summary

- ...
- ...
- ...

The implementation of the model is available at

https://github.com/dilkas/random-logic-programs