#### ProbLog

A Probabilistic Prolog and its Application in Link Discovery

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Dutch-Belgian Database Day 2006

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## Motivating Example: Biological Networks



#### Scheme of biological database

- Huge amounts of molecular biological data available
- Probabilities of direct links obtained by various prediction techniques
- Probability of some items being in a specific kind of possibly complex relationship?

```
0.3779:edge('EntrezProtein_4885045', 'HGNC_620').
0.4928:edge('HGNC_620', 'PubMed_12653567').
0.6054:edge('EntrezProtein_4885045', 'HGNC_12850').
0.9022:edge('PubMed_2322535', 'HGNC_983').
0.8750:edge('HomoloGene_20065', 'HGNC_983').
...
```

```
1.0:path(X,Y):-edge(X,Y).
1.0:path(X,Y):-edge(X,Z),path(Z,Y).
```

What is the probability of path('HGNC\_620', 'HGNC\_983')?

Of course, more complex encodings and queries possible...

ProbLog = Prolog + Probability Labels

- Each clause has a label (value between 0 and 1) indicating the probability that the clause is in the logic program.
- Each clause independent of all other clauses
- No other assumptions or restrictions

Inference in ProbLog

■ Given a ProbLog program *T* and a query *q*, what is the probability that there is some proof of *q* in *T*?

#### Semantics

ProbLog program  $T = \{p_1 : c_1, \dots, p_n : c_n\}$ , logical part  $L_T = \{c_1, \dots, c_n\}$ 

Sampling logic programs  $L \subseteq L_T$ :

$$P(L|T) = \prod_{c_i \in L} p_i \prod_{c_i \in L_T \setminus L} (1 - p_i)$$

Given *T* and a query *q*:

$$P(q|T) = \sum_{L \subseteq L_T} P(q|L) \cdot P(L|T)$$

where P(q|L) = 1 if there is a proof of q in L, else P(q|L) = 0

Use monotone DNF formula d describing all proofs of q in  $L_T$ 

- Boolean variable  $b_i$  for each  $p_i : c_i \in T$
- Concrete proof: conjunction of clause variables
- Some proof: disjunction of formulae for all proofs

$$d = \bigvee_{b \in proofs(q)} \bigwedge_{b_i \in clauses(b)} b_i$$

- Formula constructed using standard SLD-tree
- Calculating probability of monotone DNF formula is NP-hard.

# Construction of DNF Formula Example



 $d = (p_1 \land e_4) \lor (p_2 \land e_3 \land p_1 \land e_2)$ 

## DNF Formula as Binary Decision Diagram (BDD)



 $egin{array}{ccc} (p_1 \wedge e_4) & arphi \ (p_2 \wedge e_3 \wedge p_1 \wedge e_2) \end{array}$ 

- Efficient graphical representation for Boolean functions [Bryant,1986]
- Nodes labeled with Boolean variables
- Solid edge: variable assigned 1
- Dashed edge: variable assigned 0
- Probability of formula starting at node *n*:

$$prob(n) = p_n \cdot prob(n = 1) + (1 - p_n) \cdot prob(n = 0)$$

 Experiments show: up to 100.000 conjuncts feasible, depending on formula

## Approximate Inference

- Incremental, levelwise search for proofs
- After each level, construct two DNF formulae:
  - *low*: all proofs found up to current level
  - up: all derivations up to current level that have not yet failed

• low 
$$\models d \models up$$

Bounds on probability obtained using BDDs for *low* and *up*:

$$P(low) \leq P(q|T) \leq P(up)$$

- Stop if  $|P(up) P(low)| \le \delta$  for some small  $\delta$
- Similar to [Poole,1992]

#### Approximate Inference Example revisited



$$egin{aligned} low &= (p_1 \wedge e_4) \ up &= (p_1 \wedge e_4) \lor (p_2 \wedge e_3) \lor (p_2 \wedge e_4) \end{aligned}$$

$$d = (p_1 \wedge e_4) \vee (p_2 \wedge e_3 \wedge p_1 \wedge e_2)$$

#### Experiments

- Real biological graph G around four random Alzheimer genes (5220 nodes, 11530 edges)
- Example query: connection between two of the genes
- 10 sequences of subgraphs  $G_1 \subset G_2 \subset \ldots$  of sizes 200, 400, ... edges obtained by randomly subsampling edges from *G*
- Each G<sub>i</sub> consists of exactly one connected component and contains both genes used in the query.
- Approximate inference using interval width  $\delta = 0.01$
- First level of search contains proofs with up to 4 clauses, this bound is incremented by one clause on each level.

## Convergence of Probability Interval Width



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- Results for 10 graphs with 1400 edges each
- 6 to 13 levels taking 15 seconds to 4 minutes

## Results of Scalability Experiments

- Good scalability as often only small fraction of proofs needed for approximation
- Example query solved for graphs with up to 1400 to 4600 edges, depending on the random sample
- Runtimes from some seconds up to four hours for larger graphs
- Runtimes influenced by many factors, difficult to predict based on size of graph



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## Conclusions

- ProbLog: a simple Probabilistic Prolog assuming independence between clauses
- Probability calculation using
  - Monotone DNF formulae
  - Binary decision diagrams
- Effective and efficient approximation algorithm
- Experimental evaluation on real-world problem of mining large biological networks

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