



Digital
Forensics
Investigation
Research Laboratory

Probabilistic / evidential reasoning about algorithms



<http://dfire.ucd.ie/>

ProbLog

- A probabilistic extension of Prolog
- Created by the Declarative Languages and Artificial Intelligence research group at KU Leuven.
- According to the authors: "...ProbLog is a tool that allows you to intuitively build programs that do not only encode **complex interactions** between a large sets of **heterogenous components** but also the inherent **uncertainties** that are present in real-life situations."
- In this workshop we will use version 2 of ProbLog.



Useful Links on ProbLog 2



- Interactive Tutorial
 - <https://dtai.cs.kuleuven.be/problog/tutorial.html>
- Online ProbLog Programming Environment (for your own code)
 - <https://dtai.cs.kuleuven.be/problog/editor.html>



Recap: main syntactic constructs of Prolog

- Facts

`it_is_late.` (asserts that proposition it_is_late=True)

- Rules

`is_alien :- is_clingon.` (is_alien=true if is_clingon=True)

- Conjunction (and) is expressed using “,” comma

`is_superman :- looks_human, has_superpowers.`

- Disjunction (or) is expressed using ";" semicolon

`is_twilight :- before_sunrise ; after_sunset.`

- Negation (not) is expressed using "\+"

`is_unhappy :- \+ is_happy.`



Probabilistic extensions of ProbLog



- “Probabilistic” facts:

`0.5::coin_heads. % coin_heads=True with probability 0.5`

- Probabilistic rules (a.k.a. “intuitionistic” facts):

`0.7::breathalyzer_alert :- drunk. % if drunk=True, then breathalyzer_alert=True with probability 0.7`

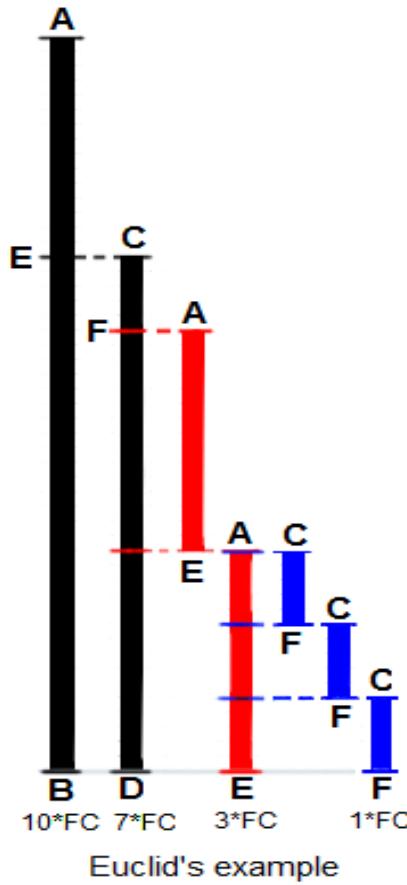
- Probabilistic disjunctions (specify probability distribution for a set of mutually exclusive events rather than a single true/false fact):

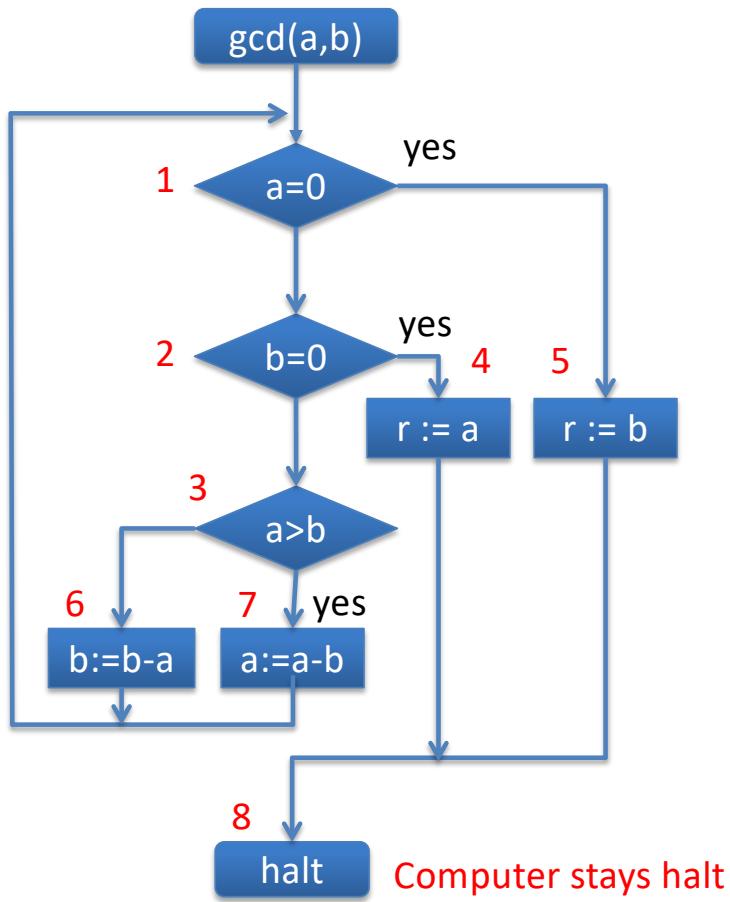
`1/6::dice(1); 1/6::dice(2); 1/6::dice(3); 1/6::dice(4); 1/6::dice(5); 1/6::dice(6). % only one of dice(x) events will occur!`



Reasoning about algorithms

- $\text{gcd}(a,b)$ – largest number that divides both a and b





gcd in ProbLog



```
gcd(A,B,Result) :- A <= 0, Result = B.
```

```
gcd(A,B,Result) :- B <= 0, Result = A.
```

```
gcd(A,B,Result) :- A > 0, B > 0, A>=B, NewA is A-B, gcd(NewA,B,Result).
```

```
gcd(A,B,Result) :- A > 0, B > 0, A<B, NewB is B-A, gcd(A,NewB,Result).
```

```
query(gcd(10,5,Result)).
```



Specifying probability distribution of initial states

```
% Probability distribution of A: 1<=A<=10  
0.1::a(1);  
0.1::a(2);  
0.1::a(3);  
0.1::a(4);  
0.1::a(5);  
0.1::a(6);  
0.1::a(7);  
0.1::a(8);  
0.1::a(9);  
0.1::a(10).
```

```
% Probability distribution of B: 1<=B<=10  
0.1::b(1);  
0.1::b(2);  
0.1::b(3);  
0.1::b(4);  
0.1::b(5);  
0.1::b(6);  
0.1::b(7);  
0.1::b(8);  
0.1::b(9);  
0.1::b(10).
```

```
model(A,B,Result) :- a(A),b(B),gcd(A,B,Result).  
query(model(A,B,Result)).
```



Resulting distribution



Query▼	Location	Probability
model(1,1,1)	32:7	0.01
model(1,2,1)	32:7	0.01
model(1,3,1)	32:7	0.01
model(1,4,1)	32:7	0.01
model(1,5,1)	32:7	0.01
model(1,6,1)	32:7	0.01
model(1,7,1)	32:7	0.01
model(1,8,1)	32:7	0.01

Querying probability of getting specific outputs

To get a distribution of specific algorithm outputs, we need to define a helper predicate (a ProbLog limitation):

```
result(Result) :- model(_,_,Result).
```

```
query(result(Result)).
```



Resulting distribution



Query▼	Location	Probability
result(1)	33:7	0.63
result(2)	33:7	0.19
result(3)	33:7	0.07
result(4)	33:7	0.03
result(5)	33:7	0.03
result(6)	33:7	0.01
result(7)	33:7	0.01
result(8)	33:7	0.01
result(9)	33:7	0.01
result(10)	33:7	0.01

Querying probability of input given evidence of output

To calculate a posterior distribution of algorithm inputs, we need to define an additional helper predicate inputs and specify evidence using built-in **evidence()** predicate:

```
result(Result) :- model(_,_,Result).
```

```
Inputs(A,B) :- model(A,B,_).
```

```
evidence(result(5)).
```

```
query(inputs(A,B)).
```



Resulting distribution



inputs(5,1)	36:7	0
inputs(5,2)	36:7	0
inputs(5,3)	36:7	0
inputs(5,4)	36:7	0
inputs(5,5)	36:7	0.33333333
inputs(5,6)	36:7	0
inputs(5,7)	36:7	0
inputs(5,8)	36:7	0
inputs(5,9)	36:7	0
inputs(5,10)	36:7	0.33333333
inputs(6,1)	36:7	0
inputs(6,2)	36:7	0

Exercise: try querying posterior probability of parameter A probability distribution



query(a(A)).

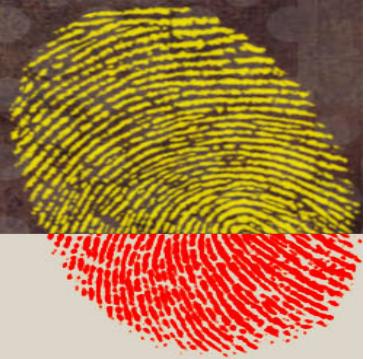


Resulting distribution

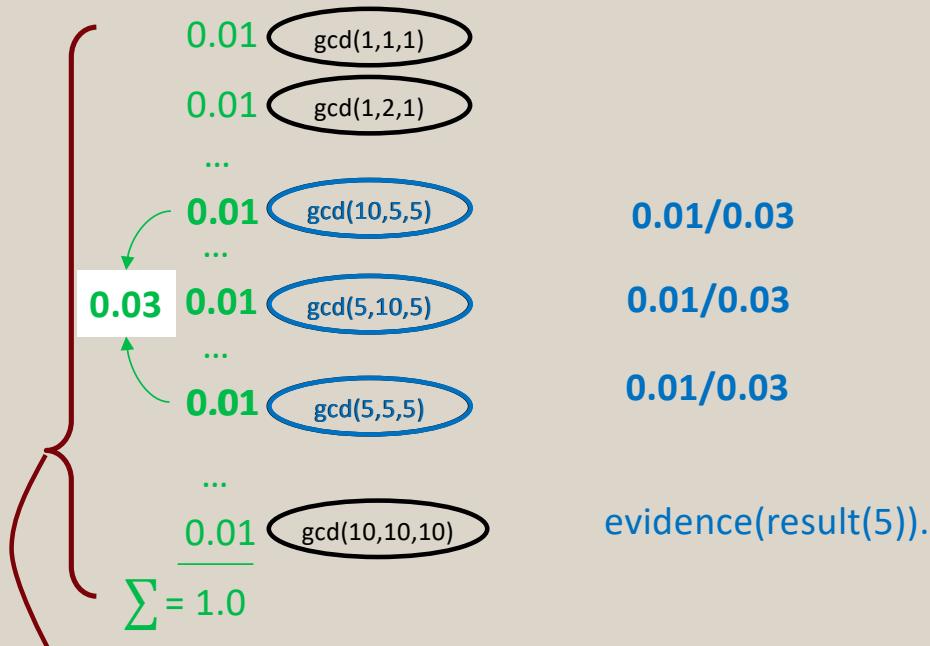


Query▼	Location	Probability
a(1)	36:7	0
a(2)	36:7	0
a(3)	36:7	0
a(4)	36:7	0
a(5)	36:7	0.666666667
a(6)	36:7	0
a(7)	36:7	0
a(8)	36:7	0
a(9)	36:7	0
a(10)	36:7	0.333333333

query(a(A))



Initial probabilities:



100 possible initial states determined by **model(A,B,Result) :- a(A),b(B),gcd(A,B,Result).**

What is the probability that parameter A was 5 at T=0? $(0.01+0.01)/0.03 = 0.666666(6)$



Further reading



- “Learn Prolog Now!”
 - <http://www.learnprolognow.org>
 - A very good and comprehensive Intro to programming in Prolog.
- ProbLog tutorials
 - <https://dtai.cs.kuleuven.be/problog/tutorial.html>



Optional self-practice exercise



- Try to modify our model so that instead of GCD it calculates a the sum of natural numbers from 1 to n for the given n.
- Contact me if you would like to get an answer! My email is on <http://dfire.ucd.ie/>

