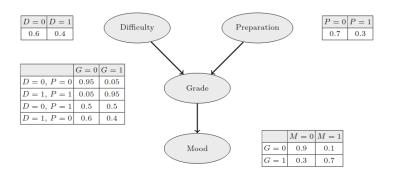
How long, O Bayesian network, will I sample thee? A program analysis perspective on expected sampling times

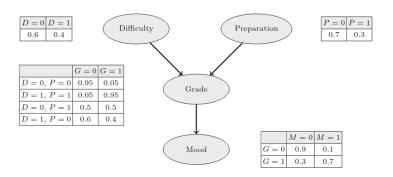
A paper for **ICALP 2018** by Kevin Batz Benjamin Lucien Kaminski Joost-Pieter Katoen Christoph Matheja

promoted by Márk Széles (me)

Bayesian networks



Bayesian networks



An example inference problem

What is the probability that a student prepares, but ends up in a bad mood?

$$P(M = 0|P = 1) = ???$$

How to solve inference problems?

Exact inference

- Gives an exact result
- Not Pleasant to compute (NP-hard)

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Sampling-based methods

- Draw random samples from distribution.
- Use the samples to compute an approximate result.
- This is often faster.

Rejection sampling

An example inference problem

What is the probability that a student prepares, but ends up in a bad mood?

$$P(M = 0|P = 1) = ???$$

Collect samples using a loop:

repeat {
$$(D,P,G,M):\approx \mu$$
 } until $(P=1)$

Result:

D	0	1	0	1
Р	1	1	1	1
G	1	1	1	0
М	1	1	1	0

$$P(M=0|P=1)=1/4$$

The problem

Compute how long it takes to get a sample via rejection sampling.

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The solution

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- Onvert the Bayesian network the an imperative, probabilistic program.
- ② The rejection sampling query can now be formulated as a loop.

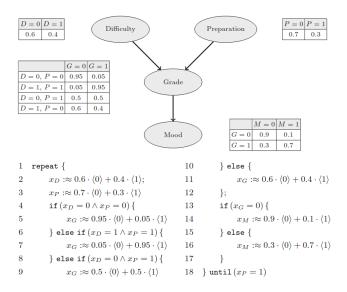
The problem

Compute how long it takes to get a sample via rejection sampling.

The solution

- Convert the Bayesian network the an imperative, probabilistic program.
- The rejection sampling query can now be formulated as a loop.
- Reason about the expected runtime of the loop using a weakest precondition-style calculus.

Rejection sampling queries as programs



A taste of the ert-calculus

ullet Let Σ be the set of possible program states. The expected runtime transformer has type

$$\operatorname{ert}[-]:\operatorname{\mathsf{Progs}} o (\Sigma o \mathbb{R}^\infty_{\geq 0}) o (\Sigma o \mathbb{R}^\infty_{\geq 0}).$$

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• Compare with weakest precondition:

$$wp[-]: \mathsf{Progs} \to (\Sigma \to 2) \to (\Sigma \to 2).$$

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• Compare with weakest precondition:

$$\mathsf{wp}[-]:\mathsf{Progs}\to (\Sigma\to 2)\to (\Sigma\to 2).$$

• There are rules similar to weakest precondition.

Why choose this paper?

- It is accessible.
- You can learn about probabilistic programming.
- You can learn about verification of probabilistic programs.
- The paper is a nice mix of theory and practice.
- There is a wide variety of possible second papers.