# Design and Implementation of Anglican Probabilistic Programming Language

David Tolpin Jan Willem van de Meent Hongseok Yang Frank Wood

September 1, 2016

https://bitbucket.org/probprog/anglican-white-paper https://bitbucket.org/probprog/anglican http://www.robots.ox.ac.uk/~fwood/anglican/index.html



#### Outline

#### Motivation

Design Outline

Inference Algorithms

Definitions and Runtime Library

#### Intuition

#### Probabilistic program:

- A program with random computations.
- Distributions are conditioned by 'observations'.
- Values of certain expressions are 'predicted' the output.

Can be written in any language (extended by sample and observe).

#### Example: Model Selection

```
(let [;; Guessing a distribution
1
          dist (sample (categorical
2
                           [[normal 1] [gamma 1]
3
                            [uniform-continuous 1]
4
                            [uniform-discrete 1]]))
5
          a (sample (gamma 1 1))
6
          b (sample (gamma 1 1))
7
          d (dist a b)]
8
      ;; Observing samples from the distribution
9
      (loop [data data]
10
        (when (seq data)
11
          (let [[x & data] data]
12
            (observe d x))
13
          (recur data)))
14
      ;; Predicting a, b and the distribution
15
      (predict :a a)
16
      (predict :b b)
17
      (predict :d d))
18
```

## More examples

▶ Intruder detection — given a log of **times** and **amounts** of payments in a bank account, how likely that the baccount was compromised?

## More examples

- Intruder detection given a log of times and amounts of payments in a bank account, how likely that the baccount was compromised?
- ➤ Counterfactual reasoning There are **two routes** from Jerusalem to Tel Aviv: 1 and 443. Based on traffic reports, I chose route 1 and was late. Would I arrive on time If I chose 443 instead?

## More examples

- Intruder detection given a log of times and amounts of payments in a bank account, how likely that the baccount was compromised?
- Counterfactual reasoning There are two routes from Jerusalem to Tel Aviv: 1 and 443. Based on traffic reports, I chose route 1 and was late. Would I arrive on time If I chose 443 instead?
- ► (Due to Stuart Russell) If you observe that a student GPA is exactly 4.0 in a model of transcripts of students from the USA (GPA's from 0.0 to 4.0) and India (GPA's from 0.0 to 10.0) what is the probability that the student is from India?

#### Inference Objective

► Suggest most probable explanation (MPE) - most likely assignment for all non-evidence variable given evidence.

#### Inference Objective

- Suggest most probable explanation (MPE) most likely assignment for all non-evidence variable given evidence.
- Approximately compute integral of the form

$$\Phi = \int_{-\infty}^{\infty} \varphi(x) p(x) dx$$

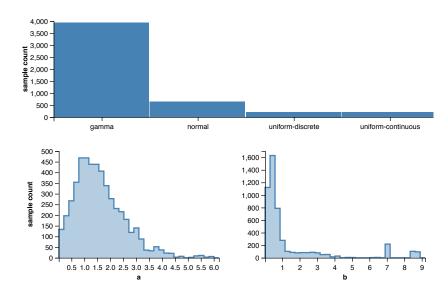
### Inference Objective

- Suggest most probable explanation (MPE) most likely assignment for all non-evidence variable given evidence.
- Approximately compute integral of the form

$$\Phi = \int_{-\infty}^{\infty} \varphi(x) p(x) dx$$

 Continuously and infinitely generate a sequence of samples drawn from the distribution of the output expression
 — so that someone else puts it in good use (vague but common). ✓

## Example: Inference Results



#### Importance Sampling

#### loop

Run program, computing weight based on observations. Output result and weight.

#### end loop

- ► Simple good.
- ▶ Slow convergence (unless one knows the answer) bad.

Can we do better?

## Lightweight Metropolis-Hastings (LMH)

Run program once, remembering random choices.

#### loop

Uniformly select one random choice.

Propose a new value for the choice.

Re-run the program.

Accept or reject with MH probability.

Output result.

#### end loop

#### Can we do better?

- Particle methods
- Variational inference
- **.**..

## Why functional?

We want a functional language because an inference algorithm controls the execution:

- ▶ A program is run many (often many hundreds of thousands) of times (with almost any algorithm).
- ▶ A program must be partially re-executed multiple times from different positions (particle methods).
- We want to reason about the distribution defined by the program.

▶ Runs on JVM — easy deployment and access to libraries.

- ▶ Runs on JVM easy deployment and access to libraries.
- ▶ A Lisp we (ab)use the macro facility.

- ▶ Runs on JVM easy deployment and access to libraries.
- ▶ A Lisp we (ab)use the macro facility.
- Church (https://en.wikipedia.org/wiki/Church\_ (programming\_language)) is derived from Scheme.

- ▶ Runs on JVM easy deployment and access to libraries.
- A Lisp we (ab)use the macro facility.
- Church (https://en.wikipedia.org/wiki/Church\_ (programming\_language)) is derived from Scheme.

#### Others use:

- Scheme (Church, Venture).
- Scala Figaro.
- Haskell Hakaru, Model-Bayes.
- **.**..

- ▶ Runs on JVM easy deployment and access to libraries.
- A Lisp we (ab)use the macro facility.
- Church (https://en.wikipedia.org/wiki/Church\_ (programming\_language)) is derived from Scheme.

#### Others use:

- Scheme (Church, Venture).
- Scala Figaro.
- Haskell Hakaru, Model-Bayes.
- **.**..

As well as Python, C#, and other languages.

#### Outline

Motivation

Design Outline

Inference Algorithms

Definitions and Runtime Library

# Language

## Macro-based compilation

#### Outline

Motivation

Design Outline

Inference Algorithms

Definitions and Runtime Library

## Managing stack size

#### Probabilistic forms

#### Memoization

#### Outline

Motivation

Design Outline

Inference Algorithms

Definitions and Runtime Library

#### Outline

Motivation

Design Outline

Inference Algorithms

Definitions and Runtime Library

Thank you! Questions?