

# Discrete Probabilistic Programming Languages<sup>1</sup>

Steven Holtzen

s.holtzen@northeastern.edu

October 2, 2023

<sup>1</sup> CS7470 Fall 2023: Foundations of Probabilistic Programming.

## 1 DISC: A simple discrete PPL

- Syntax:

```
1 e ::=
2   | x ← e; e
3   | observe e; e
4   | flip q           // q is a rational value
5   | if e then e else e
6   | return e
7   | true | false
8   | e ∧ e | e ∨ e | ¬ e |
9   | ( e )
10 p ::= e
```

- DISC looks very similar to a standard functional programming language, but has two some interesting new keywords: `flip`, `observe`, and `return`
- `flip  $\theta$`  allocates a new random quantity that is `true` with probability  $\theta$  and `false` with probability  $1 - \theta$
- `return e` turns a non-probabilistic quantity into a probabilistic one, i.e. `return true` is a *probabilistic quantity* that is `true` with probability 1 and `false` with probability 0
- Example program and its interpretation:

```
1 x ← flip 0.5;
2 y ← flip 0.5;
3 return x ∧ y
```

This program outputs the probability distribution  $[\text{true} \mapsto 0.25, \text{false} \mapsto 0.75]$ .

- `observe` is a powerful keyword that lets us *condition* the program. For instance, suppose I want to model the following scenario: “flip two coins and observe that at least one of them is heads. What is the probability that the first coin was heads?”

We can encode this scenario as a DISC program:

```
1 x ← flip 0.5;
2 y ← flip 0.5;
3 observe x ∨ y;
4 return x
```

This program outputs the probability distribution:

$$[\text{true} \mapsto (0.25 + 0.25)/0.75, \text{false} \mapsto 0.25/0.75]$$

- **Type system:** terms can either be pure Booleans of type  $\mathbb{B}$  or distributions on Booleans of type  $\text{Dist}(\mathbb{B})$ . So, we have the following type definition:

$$\tau ::= \mathbb{B} \mid \text{Dist}(\mathbb{B}). \quad (1)$$

- We define a typing judgment  $\Gamma \vdash e : \tau$  that associates each term with a type. The typing context  $\Gamma$  is a map from identifiers to types.

$$\begin{array}{c} \Gamma \vdash \text{true} : \mathbb{B} \quad \Gamma \vdash \text{false} : \mathbb{B} \quad \Gamma \vdash \text{flip } \theta : \text{Dist}(\mathbb{B}) \\[10pt] \frac{\Gamma \vdash e : \mathbb{B}}{\Gamma \vdash \text{return } e : \text{Dist}(\mathbb{B})} \quad \frac{\Gamma \vdash e_1 : \mathbb{B} \quad \Gamma \vdash e_2 : \tau}{\Gamma \vdash \text{observe } e_1; e_2 : \tau} \\[10pt] \frac{\Gamma \vdash e_1 : \text{Dist}(\mathbb{B}) \quad \Gamma \cup [x \mapsto \mathbb{B}] \vdash e_2 : \tau}{\Gamma \vdash x \leftarrow e_1; e_2 : \tau} \\[10pt] \frac{\Gamma \vdash e_1 : \mathbb{B} \quad \Gamma \vdash e_2 : \tau \quad \Gamma \vdash e_3 : \tau}{\Gamma \vdash \text{if } e_1 \text{ then } e_2 \text{ else } e_3 : \tau} \quad \frac{\Gamma \vdash e_1 : \mathbb{B} \quad \Gamma \vdash e_2 : \mathbb{B}}{\Gamma \vdash e_1 \wedge e_2 : \mathbb{B}} \end{array}$$

### Denotational semantics of *Disc*

- Associates each term with an *unnormalized probability distribution* (i.e., the total probability mass may be less than 1).
- Has the type  $\llbracket e \rrbracket : \text{Bool} \rightarrow [0, 1]$ , and has the following inductive definition:

$$\begin{aligned} \llbracket \text{flip } \theta \rrbracket(v) &= \begin{cases} \theta & \text{if } v = T \\ 1 - \theta & \text{if } v = F \end{cases} \\[10pt] \llbracket \text{return } e \rrbracket(v) &= \begin{cases} 1 & \text{if } v = \llbracket e \rrbracket \\ 0 & \text{otherwise} \end{cases} \\[10pt] \llbracket x \leftarrow e_1; e_2 \rrbracket(v) &= \sum_{v'} \llbracket e_1 \rrbracket(v') \times \llbracket e_2[x \mapsto v'] \rrbracket(v) \\[10pt] \llbracket \text{observe } e_1; e_2 \rrbracket(v) &= \sum_{\{v' \mid \llbracket e_1 \rrbracket(v') = T\}} \llbracket e_2 \rrbracket(v) \end{aligned}$$

- The semantics for non-probabilistic terms is standard. The semantic evaluation has type  $\llbracket e \rrbracket : \text{Bool}$  for closed terms.
- These semantics give an unnormalized distribution. The main semantic object of interest is the normalized distribution, which is given by the **normalized semantics**:

$$\llbracket e \rrbracket_D(T) = \frac{\llbracket e \rrbracket(T)}{\llbracket e \rrbracket(T) + \llbracket e \rrbracket(F)},$$

defined analogously for the false case.

## 2 Observation

## 3 Compiling DISC to PROP

- **Goal:** give a semantics-preserving compilation  $\rightsquigarrow$  that compiles DISC to PROP
- In order to handle observations, we will compile DISC programs into *two* PROP programs: one that computes the unnormalized probability of returning true, and one that computes the probability of evidence (i.e. normalizing constant)
- Inductive description has the shape  $e \rightsquigarrow (p_1, p_2)$ . We want to define this relation to satisfy the following **adequacy condition**:

$$\llbracket e \rrbracket_D(\text{true}) = \frac{\llbracket p_1 \rrbracket}{\llbracket p_2 \rrbracket}. \quad (2)$$

We will shorten this description to  $e \rightsquigarrow (\varphi, \varphi_A, w)$  and assume that the two formulae share a common  $w$ . The adequacy condition then becomes:

$$\llbracket e \rrbracket_D(\text{true}) = \frac{\llbracket (\varphi, w) \rrbracket}{\llbracket (\varphi_A, w) \rrbracket}. \quad (3)$$

- **Compilation relation:**

$$\begin{array}{l} \text{true} \rightsquigarrow (\text{true}, \text{true}, \emptyset) \qquad \text{false} \rightsquigarrow (\text{false}, \text{true}, \emptyset) \\[10pt] x \rightsquigarrow (x, \text{true}, \emptyset) \qquad \frac{\text{fresh } x}{\text{flip } \theta \rightsquigarrow (x, \text{true}, [x \mapsto \theta, \bar{x} \mapsto 1 - \theta])} \\[10pt] \frac{e_1 \rightsquigarrow (\varphi, \varphi_A, w) \quad e_2 \rightsquigarrow (\varphi', \varphi'_A, w')}{x \leftarrow e_1; e_2 \rightsquigarrow (\varphi'[\varphi/x], \varphi'_A[\varphi/x] \wedge \varphi_A, w_1 \cup w_2)} \\[10pt] \frac{e_1 \rightsquigarrow (\varphi, \varphi_A, w) \quad e_2 \rightsquigarrow (\varphi', \varphi'_A, w')}{\text{observe } e_1; e_2 \rightsquigarrow (\varphi', \varphi'_A \wedge \varphi_A, w_1 \cup w_2)} \end{array}$$

**Theorem 1** (Adequacy). *For well-typed term  $e$ , assume  $e \rightsquigarrow (\varphi, \varphi_A, w)$ .  
Then,  $\llbracket e \rrbracket_D(\text{true}) = \llbracket (\varphi, w) \rrbracket / \llbracket (\varphi_A, w) \rrbracket$ .*

*References*