

Looking Back: A Probabilistic Inverse Perspective on Test Generation

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Inverse Programs

Definition (Invertible)

A function `f :: A -> B` is called invertible, if there exists another function `unf :: B -> A`, that satisfies the equations

$$\begin{aligned} \text{unf } (\text{f } a) &== a \\ \text{f } (\text{unf } b) &== b \end{aligned}$$

Examples

```
dec :: Int -> Int
```

```
dec    suc_n = suc_n - 1
```

```
undec    n =      n + 1
```

```
fib, unfib :: (Int, Int) -> (Int, Int)
```

```
fib    (a, b) = (a + b, a)
```

```
unfib (ab, a) = (a, ab - a)
```

```
fib_pair :: Int -> (Int, Int)
```

```
fib_pair 0 = (1, 1)
```

```
fib_pair n = fib (fib_pair (dec n))
```

```
unfib_pair :: (Int, Int) -> Int
```

```
unfib_pair (1, 1) = 0
```

```
unfib_pair p      = undec (unfib_pair (unfib p))
```

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- ▶ **Explicit inverse program semantics (Jeopardy)**

Jeopardy (Syntax)

$x \in \mathbf{Name}$	(Well-formed variable names).
$c \in \mathbf{Name}$	(Well-formed constructor names).
$\tau \in \mathbf{Name}$	(Well-formed datatype names).
$f \in \mathbf{Name}$	(Well-formed function names).
$p ::= x \mid [c \ p_i]$	(Patterns).
$v ::= [c \ v_i]$	(Values).
$\Delta ::= f \ (p : \tau_p) : \tau_t \ = \ t \ . \ \Delta$ data $\tau \ = \ [c \ \tau_i]_j \ . \ \Delta$ main $g \ .$	(Function definition). (Data type definition). (Main function declaration).
$g ::= f \mid (\mathbf{invert} \ g)$	(Function).
$t ::= p$ $g \ p$ case $t : \tau$ of $p_i \rightarrow t_i$	(Patterns in terms). (Function application). (Case statement).

Jeopardy (Semantics)

$$\begin{array}{ll} \downarrow \text{Variable} : \frac{}{\Delta \Gamma \vdash x \downarrow v} (\Gamma(x) = v) & \downarrow \text{Constructor} : \frac{\Delta \Gamma \vdash p_i \downarrow v_i}{\Delta \Gamma \vdash [c \ p_i] \downarrow [c \ v_i]} \\ \downarrow \text{Cases} : \frac{\Delta \Gamma \vdash t \downarrow v \quad \Delta(\Gamma \circ \text{unify}(v, p_i)) \vdash t_i \downarrow v_i}{\Delta \Gamma \vdash \text{case } t : \tau \text{ of } p_i \rightarrow t_i \downarrow v_i} (\psi) & \\ \downarrow \text{Application} : \frac{\Delta \Gamma \vdash p \downarrow v' \quad \Delta(\text{unify}(v', p')) \vdash t' \downarrow v}{\Delta[f \ (p' : \cdot) : \cdot = t'] \Gamma \vdash f \ p \downarrow v} & \\ \downarrow \text{Inversion} : \frac{\Delta \Gamma \vdash g \ p \uparrow v}{\Delta \Gamma \vdash (\text{invert } g) \ p \downarrow v} & \end{array}$$

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Jeopardy (Environment Inference)

$$\boxed{\Delta[t \downarrow v] \rightsquigarrow \Gamma} \quad (\text{for linear terms } t)$$

$$\begin{array}{lcl} \rightarrow \text{Variable} : \frac{}{\Delta[x \downarrow v] \rightsquigarrow \{x \mapsto v\}} & \rightarrow \text{Constructor} : \frac{\Delta[p_i \downarrow v_i] \rightsquigarrow \Gamma_i}{\Delta[(c p_i) \downarrow (c v_i)] \rightsquigarrow \circ \Gamma_i} \\ \rightarrow \text{Cases} : \frac{\Delta[t_i \downarrow v_i] \rightsquigarrow \Gamma_i \quad \Delta \Gamma_i \vdash p_i \downarrow v \quad \Delta[t \downarrow v] \rightsquigarrow \Gamma}{\Delta[\text{case } t : \tau \text{ of } p_i \rightarrow t_i \downarrow v_i] \rightsquigarrow \Gamma_i \circ \Gamma} \quad (\psi) \\ \rightarrow \text{Application} : \frac{\Delta[t' \downarrow v] \rightsquigarrow \Gamma' \quad \Delta \Gamma' \vdash p' \downarrow v' \quad \Delta[p \downarrow v'] \rightsquigarrow \Gamma}{\Delta[f(p' : \cdot) : \cdot = t'][f p \downarrow v] \rightsquigarrow \Gamma} \\ \rightarrow \text{Inversion} : \frac{\Delta[g p \uparrow v] \rightsquigarrow \Gamma}{\Delta[(\text{invert } g) p \downarrow v] \rightsquigarrow \Gamma} \end{array}$$

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Explicit Program Inversion

Problems

- ▶ Reversibility (local invertibility) Vs (global) Invertibility

```
swap :: (a, b) -> (b, a)
swap p = (snd p, fst p)
```

- ▶ Non-injective functions

```
sorted :: Ord a => [a] -> Bool
sorted [] = True
sorted (a : as) = all (a<=) as && sorted as
```

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Approach (WIP)

- ▶ Probabilistic program inversion

```
((invert f) : (b : B) -> Generator (A b))
```

- ▶ Stochastic types

```
((invert f) : B -> Aδ)
```

Probabilistic program inversion

Computable distribution

- ▶ Consider the term $(\text{even } x)$ to be *open*.

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Possible Solutions

- ▶ Size types
- ```
m : fin 5 = (invert even)(true)
```

# Probabilistic program inversion

## Computable distribution

- ▶ Consider the term  $(\text{even } x)$  to be *open*.
- ▶ Assume a distribution  $\mathcal{D}$  over  $x$
- ▶  $\mathcal{D}$  cannot be the uniform distribution.

## Possible Solutions

- ▶ Size types  
`m : fin 5 = (invert even)(true)`
- ▶ Stochastic Choice

# Stochastic Types (WIP)

Example syntax

$$t := x \mid \dots \mid \text{let } x \text{ from } \tau_\delta \text{ in } t$$
$$\delta := \text{uniform} \mid \dots$$
$$\tau := \dots$$

Intended rule for let

$$\delta\text{-let} : \frac{\Gamma[x \mapsto \tau_{x\delta}] \vdash t : \tau_\delta}{\Gamma \vdash \text{let } x \text{ from } \tau_{x\delta} \text{ in } t : \tau_\delta}$$

Thank you {~\_^}.