## **Refinement Reflection:**

# Parallel Legacy Languages as Theorem Provers

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We introduce *refinement reflection*, a method to extend *legacy* languages—with highly tuned libraries, compilers, and runtimes—into theorem provers. The key idea is to reflect the code implementing a user-defined function into the function's (output) refinement type. As a consequence, at *uses* of the function, the function definition is unfolded into the refinement logic in a precise and predictable manner. We have implemented our approach in Liquid Haskell thereby retrofitting theorem proving into Haskell. We show how to use reflection to verify that many widely used instances of the Monoid, Applicative, Functor, and Monad typeclasses actually satisfy key algebraic laws required to make the code using the typeclasses safe. Finally, transforming a mature language—with highly tuned parallel runtime—into a theorem prover enables us to build the first *deterministic parallelism library* that verifies assumptions about associativity and ordering—that are crucial for determinism but simply assumed by existing systems.

### 1 INTRODUCTION

We introduce *refinement reflection*, a method to extend *legacy* programming languages—with highly tuned libraries, compilers, and run-times—into theorem provers, by letting programmers specify and verify arbitrary properties of their code simply by writing programs in the legacy language.

Previously, SMT-based refinement types (Constable and Smith 1987; Rushby et al. 1998) have been used to retrofit so-called "shallow" verification e.g. array bounds checking into ML (Bengtson et al. 2008; Rondon et al. 2008; Xi and Pfenning 1998), C (Condit et al. 2007; Rondon et al. 2010), Haskell (Vazou et al. 2014), TypeScript (Vekris et al. 2016), and Racket (Kent et al. 2016). To keep checking decidable, the specifications are restricted to quantifier free formulas over decidable theories. However, to verify "deep" specifications as in theorem proving languages like Agda, Coq, Dafny, F\* and Idris we require mechanisms that *represent* and *manipulate* the exact descriptions of user-defined functions. So far, this has entailed either building new languages on specialized type theories or encoding functions with SMT axioms which makes verification undecidable and hence, unpredictable.

- **1. Refinement Reflection** Our first contribution is the notion of refinement reflection: the code implementing a user-defined function can be *reflected* into the function's (output) refinement type, thus converting the function's (refinement) type signature into a deep specification of the functions behavior. This simple idea has a profound consequence: at *uses* of the function, the standard rule for (dependent) function application yields a precise, predictable and most importantly, programmer controllable means of *instantiating* the deep specification that is not tethered to brittle SMT heuristics. Specifically, we show how to use ideas for *defunctionalization* from the theorem proving literature which encode functions and lambdas using uninterpreted symbols, to encode terms from an expressive higher order language as decidable refinements, letting us use SMT-based congruence closure for decidable and predictable verification (§ 3).
- **2.** A Library of Proof Combinators Our second contribution is a *library of combinators* that lets programmers *compose proofs* from basic refinements and function definitions. We show how to represent proofs simply as unit-values refined by the proposition that they prove. We show how to build up sophisticated proofs using a small library of combinators that permits reasoning in an algebraic or equational style. Furthermore, since proofs are literally just programs, our proof combinators let us use standard language mechanisms like branches (to

encode case splits), recursion (to encode induction), and functions (to encode auxiliary lemmas) to write proofs that look very much like transcriptions of their pencil-and-paper analogues (§ 2).

- **3. Verified Typeclass Laws** Our third contribution is an implementation of refinement reflection in Liquid Haskell (Vazou et al. 2014), thereby converting the legacy language Haskell into a theorem prover. We demonstrate the benefits of this conversion by proving typeclass laws. Haskell's typeclass machinery has led to a suite of expressive abstractions and optimizations which, for correctness, crucially require typeclass *instances* to obey key algebraic laws. We show how reflection can be used to verify that widely used instances of the Monoid, Applicative, Functor, and Monad typeclasses satisfy the respective laws, making the use of these typeclasses safe (§ 6).
- **4. Verified Deterministic Parallelism** Finally, to showcase the benefits of retrofitting theorem proving onto legacy languages, we perform a case study in *deterministic parallelism*. Existing deterministic languages place unchecked obligations on the user to guarantee, *e.g.* the associativity of a fold. Violations can compromise type soundness and correctness. Closing this gap requires only modest proof effort—touching only a small subset of the application. But for this solution to be possible requires a *practical*, *parallel* programming language that supports deep verification. Before Liquid Haskell there was no such parallel language. We show how Liquid Haskell lets us verify the unchecked obligations from benchmarks taken from three existing parallel programming systems, and thus, paves the way towards high-performance with correctness guarantees (§ 7).

#### 2 OVERVIEW

We begin with an overview of how refinement reflection allows us to write proofs of and by Haskell functions.

### 2.1 Refinement Types

First, we recall some preliminaries about specification and verification with refinement types.

**Refinement types** are the source program's (here Haskell's) types decorated with logical predicates drawn from an SMT decidable logic (Constable and Smith 1987; Rushby et al. 1998). For example, we define the Nat type as Int refined by the predicate  $0 \le v$  which belongs to the quantifier free logic of linear arithmetic and uninterpreted functions (QF-UFLIA (Barrett et al. 2010)).

```
type Nat = \{ v: Int \mid 0 \le v \}
```

Here, v names the value described by the type so the above denotes "set of Int values v that are not less than 0". **Specification** & **Verification** We can use refinements to define and type the textbook Fibonacci function as:

```
fib :: Nat \rightarrow Nat
fib 0 = 0
fib 1 = 1
fib n = fib (n-1) + fib (n-2)
```

Here, the input type's refinement specifies a *pre-condition* that the parameters must be Nat, which is needed to ensure termination, and the output types's refinement specifies a *post-condition* that the result is also a Nat. Refinement type checking lets us specify and (automatically) verify the shallow property that if fib is invoked with a non-negative Int, then it terminates and yields a non-negative Int.

**Propositions** We can use refinements to define a data type representing propositions simply as an alias for unit, a data type that carries no useful runtime information:

```
type Prop = ()
```

which can be refined with propositions about the code. For example, the following type states that 2 + 2 equals 4

```
type Plus_2_2_eq_4 = { v: Prop | 2 + 2 = 4 }
```

For clarity, we abbreviate the above type by omitting the irrelevant basic type Prop and variable v:

```
type Plus_2_2_eq_4 = { 2 + 2 = 4 }
```

Function types encode universally quantified propositions:

```
type Plus_com = x:Int \rightarrow y:Int \rightarrow \{ x + y = y + x \}
```

The parameters x and y refer to input values: any inhabitant of the above is a proof that **Int** addition commutes. **Proofs** We *prove* the above theorems by writing Haskell programs. To ease this task, Liquid Haskell provides primitives to construct proof terms by "casting" expressions to **Prop**.

```
data QED = QED (**) :: a \rightarrow QED \rightarrow Prop- ** - = ()
```

To resemble mathematical proofs, we make this casting post-fix. Thus, we write e \*\* QED to cast e to a value of Prop. For example, we can prove the above propositions by defining trivial = () and then writing:

Via standard refinement type checking, the above code yields the respective verification conditions (VCs),

$$2 + 2 = 4$$
  $\forall x, y . x + y = y + x$ 

which are easily proved valid by the SMT solver, allowing us to prove the respective propositions.

A Note on Bottom: Readers familiar with Haskell's semantics may be concerned that "bottom", which inhabits all types, makes our proofs suspect. Fortunately, as described in Vazou et al. (2014), Liquid Haskell ensures that all terms with non-trivial refinements provably terminate and evaluate to (non-bottom) values, which makes our proofs sound.

### 2.2 Refinement Reflection

Suppose we wish to prove properties about the fib function, e.g. that fib 2 equals 1.

```
type fib2_eq_1 = { fib 2 = 1 }
```

Standard refinement type checking runs into two problems. First, for decidability and soundness, arbitrary user-defined functions do not belong to the refinement logic, *i.e.* we cannot *refer* to fib in a refinement. Second, the only specification that a refinement type checker has about fib is its shallow type  $Nat \rightarrow Nat$  which is too weak to verify fib2\_eq\_1. To address both problems, we **reflect** fib into the logic which sets the three steps of refinement reflection in motion.

**Step 1: Definition** The annotation tells Liquid Haskell to declare an *uninterpreted function* fib :: Int  $\rightarrow$  Int in the refinement logic. By uninterpreted, we mean that the logical fib is *not* connected to the program function fib; in the logic, fib only satisfies the *congruence axiom*  $\forall n, m. \ n = m \Rightarrow \text{fib } n = \text{fib } m$ . On its own, the uninterpreted function is not terribly useful, as it does not let us prove fib2\_eq\_1 which requires reasoning about the *definition* of fib.

**Step 2: Reflection** In the next key step, Liquid Haskell reflects the definition into the refinement type of fib by automatically strengthening the user defined type for fib to:

```
fib :: n:Nat \rightarrow \{ v:Nat \mid fibP \lor n \}
```

where  $\mathtt{fibP}$  is an alias for a refinement  $automatically\ derived$  from the function's definition:

```
fibP v n = v = if n = 0 then 0 else (if n = 1 then 1 else fib (n-1) + fib (n-2))
```

**Step 3: Application** With the reflected refinement type, each application of fib in the code automatically unfolds the fib definition *once* in the logic. We prove fib2\_eq\_1 by:

```
pf_fib2 :: { fib 2 = 1 }
pf_fib2 = let { t0 = fib 0; t1 = fib 1; t2 = fib 2 } in ()
```

We write **f** to denote places where the unfolding of f's definition is important. Via refinement typing, the above proof yields the following VC that is discharged by the SMT solver, even though **fib** is uninterpreted:

```
((fibP(fib\ 0)\ 0)\ \land (fibP(fib\ 1)\ 1)\ \land (fibP(fib\ 2)\ 2)) \Rightarrow (fib\ 2 = 1)
```

Note that the verification of pf\_fib2 relies merely on the fact that fib was applied to (*i.e.* unfolded at) 0, 1 and 2. The SMT solver automatically *combines* the facts, once they are in the antecedent. The following is also verified:

```
pf_fib2' :: { fib 2 = 1 }
pf_fib2' = [ fib 0, fib 1, fib 2 ] ** QED
```

Thus, unlike classical dependent typing, refinement reflection *does not* perform any type-level computation.

**Reflection vs. Axiomatization** An alternative *axiomatic* approach, used by Dafny (Leino 2010) and  $F^*$  (Swamy et al. 2016), is to encode fib using a universally quantified SMT formula (or axiom):  $\forall n$ . fibP (fib n) n. Axiomatization offers greater automation than reflection. Unlike Liquid Haskell, Dafny will verify the following by *automatically instantiating* the above axiom at 2, 1 and 0:

```
axPf_fib2 :: { fib 2 = 1 }
axPf_fib2 = trivial ** QED
```

The automation offered by axioms is a bit of a devil's bargain, as axioms render checking of the VCs undecidable. In practice, automatic axiom instantation can easily lead to infinite "matching loops". For example, the existence of a term fib n in a VC can trigger the above axiom, which may then produce the terms fib (n-1) and fib (n-2), which may then recursively give rise to further instantiations ad infinitum. To prevent matching loops an expert must carefully craft "triggers" and provide a "fuel" parameter (Amin et al. 2014) that can be used to restrict the numbers of the SMT unfoldings, which ensure termination, but can cause the axiom to not be instantiated at the right places. In short, per the authors of Dafny, the undecidability of the VC checking and its attendant heuristics makes verification unpredictable (Leino and Pit-Claudel 2016).

#### 2.3 Structuring Proofs

In contrast to the axiomatic approach, with refinement reflection the VCs are deliberately designed to always fall in an SMT-decidable logic, as function symbols are uninterpreted. It is up to the programmer to unfold the definitions at the appropriate places, which we have found, with careful design of proof combinators, to be quite a natural and pleasant experience. To this end, we have developed a library of proof combinators that permits reasoning about equalities and linear arithmetic, inspired by Agda (Mu et al. 2009).

**"Equation" Combinators** We equip Liquid Haskell with a family of equation combinators  $\odot$ . for each logical operator  $\odot$  in  $\{=, \neq, \leq, <, \geq, >\}$ , the operators in the theory QF-UFLIA. The refinement type of  $\odot$ . *requires* that  $x \odot y$  holds and then *ensures* that the returned value is equal to x. For example, we define = as:

```
(=.) :: x:a \to y:\{ a \mid x = y\} \to \{ v:a \mid v = x \}
 x = 0
```

and use it to write the following "equational" proof:

```
eqPf_fib2 :: { fib 2 = 1 }
eqPf_fib2 = fib 2 =. fib 1 + fib 0 =. 1 ** OED
```

"Because" Combinators Often, we need to compose "lemmata" into larger theorems. For example, to prove fib 3 = 2 we may wish to reuse eqPf\_fib2 as a lemma. To this end, Liquid Haskell has a "because" combinator:

```
(:) :: (Prop \rightarrow a) \rightarrow Prop \rightarrow a
f :: y = f y
```

The operator is simply an alias for function application that lets us write  $x \odot y : p$  (instead of  $(\odot x) y p$ ) where  $(\odot x)$  is extended to accept an *optional* third proof argument via Haskell's typeclass mechanisms. We use the because combinator to prove that fib 3 = 2 with a Haskell function:

```
eqPf_fib3 :: { fib 3 = 2 }
eqPf_fib3 = fib 3 =. fib 2 + fib 1 =. 2 : eqPf_fib2 ** QED
```

**Arithmetic and Ordering** SMT based refinements let us go well beyond just equational reasoning. Next, lets see how we can use arithmetic and ordering to prove that fib is (locally) increasing, *i.e.* for all n, fib  $n \le$ fib (n + 1).

```
fibUp :: n:Nat \rightarrow { fib n \leq fib (n+1) }

fibUp n | n == 0

= fib 0 <. fib 1 ** QED

| n == 1

= fib 1 \leq. fib 1 + fib 0 \leq. fib 2 ** QED

| otherwise

= fib n

=. fib (n-1) + fib (n-2)

\leq. fib n + fib (n-2) \therefore fibUp (n-1)

\leq. fib n + fib (n-1) \therefore fibUp (n-2)

\leq. fib (n+1) ** QED
```

Case Splitting and Induction The proof fibUp works by induction on n. In the *base* cases 0 and 1, we simply assert the relevant inequalities. These are verified as the reflected refinement unfolds the definition of fib at those inputs. The derived VCs are (automatically) proved as the SMT solver concludes 0 < 1 and  $1 + 0 \le 1$  respectively. In the *inductive* case, fib n is unfolded to fib (n-1) + fib (n-2), which, because of the induction hypothesis (applied by invoking fibUp at n-1 and n-2) and the SMT solver's arithmetic reasoning, completes the proof.

**Reflection and Refinement Types** The above examples illustrate how refinement types critically allow us to restrict the domain of both reflected functions and proof terms in order to enable sound theorem proving.

- **Totality** First, by refining Int to Nat we restrict the domain of Haskell's partial function fib which lets us use fib as a total function in the logic. Specifically, refinement type checking ensures soundness by verifying that fib is never called on negative numbers.
- *Induction* Second, the refinement type Nat in the argument of fibUp constrains fibUp's domain, turning fibUp into a total proof that encodes induction on Natural numbers, while reasoning about Haskell's runtime integers.

Thus, refinement types let us encode proofs that only hold on refined types, like Nat while using Haskell's Integers, instead of forcing to switch to a user or library defined Data.Natural Haskell type. As a less elegant alternative, usually followed in dependently typed languages, fib could be defined as a function returning a Maybe – *i.e.* returning Nothing on negative inputs. Then the code and proofs would be littered with cases analyses required to extract fib's result. Another alternative is Dafny's approach that uses code level assumptions and assertions to restrict function's domain and range. While these assumptions and assertions can be semi-automatically

discharged using the SMT solver, they impose programmer overheads as they cannot be inferred unlike refinement types that are inferred using the abstract interpretation framework of liquid typing (Rondon et al. 2008).

### 2.4 Case Study: Deterministic Parallelism

One benefit of an in-language prover is that it lowers the barrier to *small* verification efforts that touch only a fraction of the program, and yet ensure critical invariants that Haskell's type system cannot. Here we consider parallel programming, which is commonly considered error prone and entails proof obligations on the user that typically go unchecked.

The situation is especially precarious with parallel programming frameworks that claim to be *deterministic* and thus usable within purely functional programs. These include Deterministic Parallel Java (DPJ (Bocchino et al. 2009)), Concurrent Revisions for .NET (Burckhardt et al. 2010) and Haskell's LVish (Kuper et al. 2014), Accelerate (McDonell et al. 2013), and REPA (Keller et al. 2010). Accelerate's parallel fold function, for instance, claims to be deterministic—and its purely functional type means the Haskell optimizer will *assume* its referential transparency—but its determinism depends on an associativity guarantee which must be assured *by the programmer* rather than the type system. Thus, simply folding the minus function, fold (-) 0 arr, is sufficient to violate determinism and Haskell's pure semantics.

Likewise, DPJ goes to pains to develop a new type system for parallel programming, but then provides a "commutes" annotation for methods updating shared state, compromising the *guarantee* and going back to trusting the user. LVish has the same Achilles heel. Consider set insertion:

```
insert :: Ord a \Rightarrow a \rightarrow Set s a \rightarrow Par s ()
```

which returns an (effectful) Par computation that can be run within a pure function to produce a pure result. At first glance it would seem that trusting the implementation of the concurrent set is sufficient to assure a deterministic outcome. However, the interface has an Ord constraint, which means this polymorphic function works with user-defined data types and thus, orderings. What if the user fails to implement a total order? Then, even a correct implementation of, e.g. a concurrent skiplist (Fomitchev and Ruppert 2004), can reveal different insertion orders due to concurrency, thus wrecking the determinism guarantee.

In summary, parallel programs naturally need to communicate, but the mechanisms of that communication—such as folds or inserts into a shared structure—typically carry additional proof obligations. This in turn makes parallelism a liability. Fortunately, we can remove the risk with verification.

Verified Typeclasses Our solution involves simply changing the Ord constraint above to VerifiedOrd.

```
insert :: VerifiedOrd a \Rightarrow a \rightarrow Set s a \rightarrow Par s ()
```

This constraint changes the interface but not the implementation of insert. The additional methods of the verified type class don't add operational capabilities, but rather impose additional proof obligations:

```
class Ord a \Rightarrow VerifiedOrd a where antisym :: x:a \rightarrow y:a \rightarrow { x \leq y && y \leq x \Rightarrow x = y } trans :: x:a \rightarrow y:a \rightarrow z:a \rightarrow { x \leq y && y \leq z \Rightarrow x \leq z } total :: x:a \rightarrow y:a \rightarrow { x \leq y | | y \leq x }
```

Verified Monoids Similarly, we can refine the Monoid typeclass with refinements expressing the Monoid laws.

```
class Monoid a \Rightarrow VerifiedMonoid a where lident :: x:a \rightarrow \{ mempty <> x = x \} rident :: x:a \rightarrow \{ x <> mempty = x \} assoc :: x:a \rightarrow y:a \rightarrow z:a \rightarrow \{ x <> (y <> z) = (x <> y) <> z \}
```

The VerifiedMonoid typeclass constraint requires the binary operation to be associative, thus can be safely used to fold on an unknown number of processors.

**Verified instances for primitive types** VerifiedOrd instances for primitive types like Int and Double are trivial to write; they just appeal to the SMT solver's built-in theories. For example, the following proves totality (of ordering) for Int.

```
totInt :: x:Int \rightarrow y:Int \rightarrow \{x \le y \mid | y \le x\}
totInt _ _ = trivial ** QED
```

**Verified instances for algebraic datatypes** We use structural induction to prove the class laws for user defined algebraic datatypes. For example, we can inductively define Peano numerals and then write a function to compare two Peano numbers:

```
data Peano = Z | S Peano

reflect leq :: Peano → Peano → Bool
leq Z _ = True
leq (S _) Z = False
leq (S n) (S m) = leq n m
```

In § 3 we will describe exactly how the reflection mechanism (illustrated via fibP) is extended to account for ADTs like Peano. Liquid Haskell automatically checks that leq is total (Vazou et al. 2014), which lets us safely **reflect** it into the logic. Next, we prove that leq is total on Peano numbers

```
totalPeano :: n:Peano \rightarrow m:Peano \rightarrow {leq n m || leq m n} / [toInt n + toInt m] totalPeano Z m = leq Z m ** QED totalPeano n Z = leq Z n ** QED totalPeano (S n) (S m) = leq (S n) (S m) || leq (S m) (S n) = . leq n m || leq m n \therefore totalPeano m n ** OED
```

The proof goes by induction, splitting cases on whether the number is zero or non-zero. Consequently, we pattern match on the parameters n and m and furnish separate proofs for each case. In the "zero" cases, we simply unfold the definition of leq. In the "successor" case, after unfolding we (literally) apply the induction hypothesis by using the because operator. The termination hint [toInt n + toInt m], where toInt maps Peano numbers to integers, is used to verify well-formedness of the totalPeano proof term. Liquid Haskell's termination and totality checker use the hint to verify that we are in fact doing induction properly (§ 3). We can write the rest of the VerifiedOrd proof methods in a fashion similar to totalPeano and use them to create verified instances:

Proving all the four VerifiedOrd laws is a burden on the programmer. Since Peano is isomorphic to Nats, next we present how to reduce the Peano proofs into the SMT automated integer proofs.

**Isomorphisms** In order to reuse proofs for a custom datatype, we provide a way to translate verified instances between isomorphic types (Barthe and Pons 2001). We design a typeclass Iso to witnesses type isomorphism.

```
class Iso a b where

to :: a \rightarrow b

from :: b \rightarrow a

toFrom :: x:a \rightarrow \{to (from x) = x\}

fromTo :: x:a \rightarrow \{from (to x) = x\}
```

For two isomorphic types a and b we compare instances of b using a's comparison method.

```
instance (Ord a, Iso a b) \Rightarrow Ord b where x \le y = \text{from } x \le \text{from } y
```

Then, we prove that VerifiedOrd laws are closed under isomorphisms. For example, we prove totality of comparison on b using the VerifiedOrd totality on a as:

Now, we can simply use Haskell's instances, to obtain a VerifiedOrd instance for Peano by creating an Iso Nat instance for Peano.

**Proof Composition via Products** Finally, we present a mechanism to automatically reduce proofs on product types to proofs of the product components. For example, lexicographic ordering preserves the ordering laws. First, we use class instances to define lexicographic ordering.

```
instance (VerifiedOrd a, VerifiedOrd b) \Rightarrow Ord (a, b) where (x1, y1) \leq (x2, y2) = if x1 == x2 then y1 \leq y2 else x1 \leq x2
```

Then, we prove that lexicographic ordering preserves the ordering laws, for example, totality:

```
instance (VerifiedOrd a, VerifiedOrd b) \Rightarrow VerifiedOrd (a, b) where total :: p:(a, b) \rightarrow q:(a, b) \rightarrow {p \leq q || q \leq p} total p@(x1, y1) q@(x2, y2) = p \leq q || q \leq p =. if x1 == x2 then (y1 \leq y2 || y2 \leq y1) else True :: total x1 x2 =. True :: total y1 y2 ** OED
```

Haskell's typeclass machinery will now let us obtain a VerifiedOrd instance for (Peano, Peano) from the VerifiedOrd instance for Peano. In short, we can decompose an algebraic datatype into an isomorphic type using sums and products to generate verified instances for arbitrary Haskell datatypes. This could be combined with GHC's support for generics (Magalhães et al. 2010) to fully automate the derivation of verified instances for user datatypes. In §7, we use these ideas to develop safe interfaces to LVish modules and to verify patterns from DPJ.

## 3 REFINEMENT REFLECTION: $\lambda^R$

We formalize refinement reflection in two steps. First, we develop a core calculus  $\lambda^R$  with an *undecidable* type system based on denotational semantics. We show how the soundness of the type system allows us to *prove* theorems using  $\lambda^R$ . Next, in § 4 we define a language  $\lambda^S$  that soundly approximates  $\lambda^R$  while enabling decidable SMT-based type checking.

#### 3.1 Syntax

Figure 1 summarizes the syntax of  $\lambda^R$ , which is essentially the calculus  $\lambda^U$  (Vazou et al. 2014) with explicit recursion and a special reflect binding to denote terms that are reflected into the refinement logic. The elements of  $\lambda^R$  are layered into primitive constants, values, expressions, binders and programs.

```
Operators \odot ::= = | <
 Constants
                 c ::= \land | \neg | \odot | +, \neg, \dots
                                                                      Predicates
                                                                                           := r \oplus_2 r \mid \oplus_1 r
                            True | False | 0, \pm 1, \ldots
                        n \mid b \mid x \mid D \mid x \overline{r}
                     := c \mid \lambda x.e \mid D \overline{w}
      Values w
                                                                                                   if r then r else r
Expressions
                 e ::= w | x | e e
                                                                         Integers
                                                                                          ::= 0, -1, 1, \dots
                             case x = e of \{D \overline{x} \rightarrow e\}
                                                                       Booleans
                                                                                        b ::= True \mid False
    Binders b ::= e \mid \text{let rec } x : \tau = b \text{ in } b
                                                                                                  = | < | \ \ | + | - | \ \ \ . \ .
                                                                    Binary Ops
                                                                                           ::=
                            b \mid \text{reflect } x : \tau = e \text{ in } p
   Program
                     ::=
                                                                     Unary Ops
                                                                                            ::= ¬ | ...
Basic Types
                B ::=
                            Int | Bool | T
                                                                       Sort Args
                                                                                           ::= Int | Bool | U | Fun s_a s_a
 Ref. Types
                 \tau ::= \{v : B^{[\downarrow]} \mid e\} \mid x : \tau \to \tau
                                                                             Sorts
                                                                                           := s_a \mid s_a \rightarrow s
```

Fig. 1. **Syntax of**  $\lambda^R$ 

Fig. 2. **Syntax of**  $\lambda^S$ 

**Constants** The primitive constants of  $\lambda^R$  include primitive relations  $\odot$ , here, the set  $\{=,<\}$ . Moreover, they include the primitive booleans True, False, integers -1,  $\emptyset$ , 1, etc., and logical operators  $\wedge$ ,  $\vee$ ,  $\neg$ , etc..

**Data Constructors** Data constructors are special constants. For example, the data type [*Int*], which represents finite lists of integers, has two data constructors: [] (nil) and : (cons).

**Values & Expressions** The values of  $\lambda^R$  include constants,  $\lambda$ -abstractions  $\lambda x.e$ , and fully applied data constructors D that wrap values. The expressions of  $\lambda^R$  include values, variables x, applications e e, and case expressions.

**Binders & Programs** A *binder b* is a series of possibly recursive let definitions, followed by an expression. A *program p* is a series of reflect definitions, each of which names a function that is reflected into the refinement logic, followed by a binder. The stratification of programs via binders is required so that arbitrary recursive definitions are allowed in the program but cannot be inserted into the logic via refinements or reflection. (We *can* allow non-recursive let binders in expressions *e*, but omit them for simplicity.)

## 3.2 Operational Semantics

We define  $\hookrightarrow$  to be the small step, call-by-name  $\beta$ -reduction semantics for  $\lambda^R$ . We evaluate reflected terms as recursive let bindings, with extra termination-check constraints imposed by the type system:

```
reflect x : \tau = e \text{ in } p \hookrightarrow \text{let rec } x : \tau = e \text{ in } p
```

We define  $\hookrightarrow^*$  to be the reflexive, transitive closure of  $\hookrightarrow$ . Moreover, we define  $\approx_{\beta}$  to be the reflexive, symmetric, and transitive closure of  $\hookrightarrow$ .

**Constants** Application of a constant requires the argument be reduced to a value; in a single step, the expression is reduced to the output of the primitive constant operation, *i.e.*  $c \ v \hookrightarrow \delta(c, v)$ . For example, consider =, the primitive equality operator on integers. We have  $\delta(=, n) = n$  where  $\delta(=, n)$  equals True iff m is the same as n.

**Equality** We assume that the equality operator is defined *for all* values, and, for functions, is defined as extensional equality. That is, for all f and f',  $(f = f') \hookrightarrow \text{True}$  iff  $\forall v. f \ v \approx_{\beta} f' \ v$ . We assume source *terms* only contain implementable equalities over non-function types; while function extensional equality only appears in *refinements*.

## 3.3 Types

 $\lambda^R$  types include basic types, which are *refined* with predicates, and dependent function types. *Basic types B* comprise integers, booleans, and a family of data-types T (representing lists, trees *etc.*). For example, the data type [Int] represents lists of integers. We refine basic types with predicates (boolean-valued expressions e) to obtain *basic refinement types*  $\{v: B \mid e\}$ . We use  $\downarrow$  to mark provably terminating computations and use refinements to ensure that if  $e:\{v: B^{\downarrow} \mid e'\}$ , then e terminates. As discussed by Vazou et al. (2014) termination labels can be checked using refinement types and are used to ensure that refinements cannot diverge as required for soundness of type checking under lazy evaluation. Finally, we have dependent *function types*  $x: \tau_x \to \tau$  where the input x has the type x and the output x may refer to the input binder x. We write x to abbreviate x to abbreviate x if x does not appear in x.

**Denotations** Each type  $\tau$  *denotes* a set of expressions  $[\![\tau]\!]$ , that is defined via the operational semantics (Knowles and Flanagan 2010). Let  $\lfloor \tau \rfloor$  be the type we get if we erase all refinements from  $\tau$  and  $e: \lfloor \tau \rfloor$  be the standard typing relation for the typed lambda calculus. Then, we define the denotation of types as:

```
\begin{split} & \llbracket \{x:B \mid r\} \rrbracket \ \doteq \ \{e \mid e:B, \text{ if } e \hookrightarrow^{\bigstar} w \text{ then } r[x \mapsto w] \hookrightarrow^{\bigstar} \text{ True} \} \\ & \llbracket \{x:B^{\parallel} \mid r\} \rrbracket \ \doteq \ \llbracket \{x:B \mid r\} \rrbracket \cap \{e \mid e \hookrightarrow^{\bigstar} w \} \\ & \llbracket x:\tau_x \to \tau \rrbracket \ \doteq \ \{e \mid e:\lfloor\tau_x \to \tau\rfloor, \forall e_x \in \llbracket\tau_x\rrbracket. \ (e e_x) \in \llbracket\tau[x \mapsto e_x] \rrbracket \} \end{split}
```

**Constants** For each constant c we define its type Ty(c) such that  $c \in [Ty(c)]$ . For example,

```
\begin{array}{lll} \mathsf{Ty}(3) & \doteq & \{v: \mathsf{Int}^{\Downarrow} \mid v = 3\} \\ \mathsf{Ty}(+) & \doteq & \mathsf{x}: \mathsf{Int}^{\Downarrow} \to \mathsf{y}: \mathsf{Int}^{\Downarrow} \to \{v: \mathsf{Int}^{\Downarrow} \mid v = x + y\} \\ \mathsf{Ty}(\leq) & \doteq & \mathsf{x}: \mathsf{Int}^{\Downarrow} \to \mathsf{y}: \mathsf{Int}^{\Downarrow} \to \{v: \mathsf{Bool}^{\Downarrow} \mid v \Leftrightarrow x \leq y\} \end{array}
```

### 3.4 Refinement Reflection

The key idea in our work is to *strengthen* the output type of functions with a refinement that *reflects* the definition of the function in the logic. We do this by treating each reflect-binder (reflect  $f : \tau = e \text{ in } p$ ) as a let rec-binder (let rec  $f : \text{Reflect}(\tau, e) = e \text{ in } p$ ) during type checking (rule T-Refl in Figure 3).

**Reflection** We write Reflect( $\tau$ , e) for the reflection of the term e into the type  $\tau$ , defined by strengthening  $\tau$  as:

```
 \begin{array}{lll} \mathsf{Reflect}(\{v:B^{\Downarrow}\mid r\},e) & \doteq & \{v:B^{\Downarrow}\mid r \wedge v=e\} \\ \mathsf{Reflect}(x:\tau_{x} \rightarrow \tau, \lambda x.e) & \doteq & x:\tau_{x} \rightarrow \mathsf{Reflect}(\tau,e) \\ \end{array}
```

As an example, recall from § 2 that the reflect fib strengthens the type of fib with the refinement fibP.

**Termination Checking** We defined Reflect(·, ·) to be a *partial* function that only reflects provably terminating expressions, *i.e.* expressions whose result type is marked with  $\Downarrow$ . If a non-provably terminating function is reflected in an  $\lambda^R$  expression then type checking will fail (with a reflection type error in the implementation). This restriction is crucial for soundness, as diverging expressions can lead to inconsistencies. For example, reflecting the diverging  $f \times f = 1 + f \times f = 1$ .

**Automatic Reflection** Reflection of  $\lambda^R$  expressions into the refinements happens automatically by the type system, not manually by the user. The user simply annotates a function f as reflect f. Then, the rule T-Refl in Figure 3 is used to type check the reflected function by strengthening the f's result via Reflect( $\cdot$ ,  $\cdot$ ). Finally, the rule T-Let is used to check that the automatically strengthened type of f satisfies f's implementation.

**Consequences for Verification** Reflection has a major consequence for verification: instead of being tethered to quantifier instantiation heuristics or having to program "triggers" as in Dafny (Leino 2010) or F\* (Swamy et al.

Fig. 3. Typing of  $\lambda^R$  and algorithmic subtyping of  $\lambda^S$ 

2016), the programmer can predictably "unfold" the definition of the function during a proof simply by "calling" the function, which, as discussed in § 6, we have found to be a very natural way of structuring proofs.

#### 3.5 Typing Rules

Next, we present the type-checking rules of  $\lambda^R$ .

**Environments and Closing Substitutions** A *type environment*  $\Gamma$  is a sequence of type bindings  $x_1 : \tau_1, \ldots, x_n : \tau_n$ . An environment denotes a set of *closing substitutions*  $\theta$  which are sequences of expression bindings:  $x_1 \mapsto e_1, \ldots, x_n \mapsto e_n$  such that:

$$[\![\Gamma]\!] \ \doteq \ \{\theta \mid \forall x : \tau \in \Gamma.\theta(x) \in [\![\theta \cdot \tau]\!]\}$$

**Typing** A judgment  $\Gamma \vdash p : \tau$  states that the program p has the type  $\tau$  in the environment  $\Gamma$ . That is, when the free variables in p are bound to expressions described by  $\Gamma$ , the program p will evaluate to a value described by  $\tau$ . **Rules** All but two of the rules are standard (Knowles and Flanagan 2010; Vazou et al. 2014). First, rule T-Refl is used to strengthen the type of each reflected binder with its definition, as described previously in § 3.4. Second, rule T-Exact strengthens the expression with a singleton type equating the value and the expression (*i.e.* reflecting the expression in the type). This is a generalization of the "selfification" rules from (Knowles and Flanagan 2010; Ou et al. 2004) and is required to equate the reflected functions with their definitions. For

example, the application fib 1 is typed as  $\{v : \text{Int}^{\Downarrow} \mid \text{fibP } v \mid 1 \land v = \text{fib } 1\}$  where the first conjunct comes from the (reflection-strengthened) output refinement of fib § 2 and the second comes from rule T-EXACT.

Well-formedness A judgment  $\Gamma \vdash \tau$  states that the refinement type  $\tau$  is well-formed in the environment  $\Gamma$ . Following Vazou et al. (2014),  $\tau$  is well-formed if all the refinements in  $\tau$  are Bool-typed, provably terminating expressions in  $\Gamma$ .

**Subtyping** A judgment  $\Gamma \vdash \tau_1 \leq \tau_2$  states that the type  $\tau_1$  is a subtype of  $\tau_2$  in the environment  $\Gamma$ . Informally,  $\tau_1$  is a subtype of  $\tau_2$  if, when the free variables of  $\tau_1$  and  $\tau_2$  are bound to expressions described by  $\Gamma$ , the denotation of  $\tau_1$  is *contained in* the denotation of  $\tau_2$ . Subtyping of basic types reduces to denotational containment checking, shown in rule  $\leq$ -BASE- $\lambda^R$ . That is,  $\tau_1$  is a subtype of  $\tau_2$  under  $\Gamma$  if for any closing substitution  $\theta$  in the denotation of  $\Gamma$ ,  $[\![\theta \cdot \tau_1]\!]$  is contained in  $[\![\theta \cdot \tau_2]\!]$ .

**Soundness** Following  $\lambda^U$  (Vazou et al. 2014), in Supplementary-Material (2017) we prove that evaluation preserves typing and typing implies denotational inclusion.

THEOREM 3.1. [Soundness of  $\lambda^R$ ]

- **Denotations** If  $\Gamma \vdash p : \tau$  then  $\forall \theta \in [\![\Gamma]\!] . \theta \cdot p \in [\![\theta \cdot \tau]\!].$
- **Preservation** If  $\emptyset \vdash p : \tau$  and  $p \hookrightarrow^* w$  then  $\emptyset \vdash w : \tau$ .

Theorem 3.1 lets us interpret well typed programs as proofs of propositions. For example, in § 2 we verified that the term fibUp proves

$$n: \mathsf{Nat} \to \{\mathsf{fib}\ n \leq \mathsf{fib}\ (n+1)\}$$

Via soundness of  $\lambda^R$ , we get that for each valid input n, the result refinement is valid.

$$\forall n.0 \leq n \hookrightarrow^{\star} \text{True} \Rightarrow \text{fib } n \leq \text{fib } (n+1) \hookrightarrow^{\star} \text{True}$$

## 4 ALGORITHMIC CHECKING: $\lambda^S$

Next, we describe  $\lambda^S$ , a conservative, first order approximation of  $\lambda^R$  where higher order features are approximated with uninterpreted functions and the undecidable type subsumption rule  $\leq$ -BASE- $\lambda^R$  is replaced with a decidable one  $\leq$ -BASE- $\lambda^S$ , yielding an SMT-based algorithmic type system that is both sound and decidable.

**Syntax: Terms & Sorts** Figure 2 summarizes the syntax of  $\lambda^S$ , the *sorted* (SMT-) decidable logic of quantifier-free equality, uninterpreted functions and linear arithmetic (QF-EUFLIA) (Barrett et al. 2010; Nelson 1981). The *terms* of  $\lambda^S$  include integers n, booleans b, variables x, data constructors D (encoded as constants), fully applied unary  $\oplus_1$  and binary  $\oplus_2$  operators, and application x  $\overline{r}$  of an uninterpreted function x. The *sorts* of  $\lambda^S$  include built-in integer Int and Bool for representing integers and booleans. The interpreted functions of  $\lambda^S$ , e.g. the logical constants = and <, have the function sort  $s \to s$ . Other functional values in  $\lambda^R$ , e.g. reflected  $\lambda^R$  functions and  $\lambda$ -expressions, are represented as first-order values with the uninterpreted sort Fun s s. The universal sort U represents all other values.

**Semantics: Satisfaction & Validity** An assignment  $\sigma$  maps variables to terms  $\sigma = \{x_1 \mapsto r_1, \dots, x_n \mapsto r_n\}$ . We write  $\sigma \models r$  if the assignment  $\sigma$  is a *model of r*, intuitively if  $\sigma \cdot r$  "is true" (Nelson 1981). A predicate r is satisfiable if there exists  $\sigma \models r$ . A predicate r is valid if for all assignments  $\sigma \models r$ .

## 4.1 Transforming $\lambda^R$ into $\lambda^S$

The judgment  $\Gamma \vdash e \leadsto r$  states that a  $\lambda^R$  term e is transformed, under an environment  $\Gamma$ , into a  $\lambda^S$  term r. Most of the transformation rules are identity and can be found in (Supplementary-Material 2017). Here we discuss the non-identity ones.

**Embedding Types** We embed  $\lambda^R$  types into  $\lambda^S$  sorts as:

$$\begin{array}{lll} (|\operatorname{Int}|) & \doteq & \operatorname{Int} \\ (|\operatorname{Bool}|) & \doteq & \operatorname{Bool} \\ (|T|) & \doteq & \operatorname{U} \end{array} \qquad \begin{array}{ll} (|\{\upsilon:B^{[\downarrow]}\mid e\}|) & \doteq & (|B|) \\ (|x:\tau_x\to\tau|) & \doteq & \operatorname{Fun} (|\tau_x|) (|\tau|) \end{array}$$

**Embedding Constants** Elements shared on both  $\lambda^R$  and  $\lambda^S$  translate to themselves. These elements include booleans, integers, variables, binary and unary operators. SMT solvers do not support currying, and so in  $\lambda^S$ , all function symbols must be fully applied. Thus, we assume that all applications to primitive constants and data constructors are fully applied, *e.g.* by converting source terms like (+ 1) to (\z \rightarrow z + 1).

**Embedding Functions** As  $\lambda^S$  is first-order, we embed  $\lambda$ -abstraction using the uninterpreted function lam.

$$\frac{\Gamma, x : \tau_x \vdash e \leadsto r \qquad \Gamma \vdash (\lambda x.e) : (x : \tau_x \to \tau)}{\Gamma \vdash \lambda x.e \leadsto \operatorname{lam}_{(\mid \tau_x \mid)}^{(\mid \tau_x \mid)} x r} \quad \text{T-Fun}$$

The term  $\lambda x.e$  of type  $\tau_x \to \tau$  is transformed to  $lam_s^{s_x} x r$  of sort Fun  $s_x$  s, where  $s_x$  and s are respectively  $(\tau_x)$  and  $(\tau)$ ,  $lam_s^{s_x}$  is a special uninterpreted function of sort  $s_x \to s \to Fun s_x$  s, and x of sort  $s_x$  and r of sort s are the embedding of the binder and body, respectively. As lam is an SMT-function, it *does not* create a binding for x. Instead, x is renamed to a *fresh* name pre-declared in the SMT logic.

**Embedding Applications** We embed applications via defunctionalization (Reynolds 1972) using the uninterpreted app:

$$\frac{\Gamma \vdash e' \leadsto r' \qquad \Gamma \vdash e \leadsto r \qquad \Gamma \vdash e : \tau_x \to \tau}{\Gamma \vdash e \; e' \leadsto \mathsf{app}_{\left(\mid \tau_x \mid\right)}^{\left(\mid \tau_x \mid\right)} \; r \; r'} \quad \mathsf{T-App}$$

The term e e', where e and e' have types  $\tau_x \to \tau$  and  $\tau_x$ , is transformed to  $\mathsf{app}_s^{s_x} r r' : s$  where s and  $s_x$  are  $(\tau)$  and  $(\tau_x)$ , the  $\mathsf{app}_s^{s_x}$  is a special uninterpreted function of sort Fun  $s_x$   $s \to s_x \to s$ , and r and r' are the respective translations of e and e'.

**Embedding Data Types** We translate each data constructor to a predefined  $\lambda^S$  constant  $s_D$  of sort (Ty(D)).

$$\Gamma \vdash D \leadsto \mathsf{s}_D$$

For each datatype, we assume the existence of reflected functions that *check* the top-level constructor and *select* their individual fields. For example, for lists, we assume the existence of measures:

isNil [] = True isCons 
$$(x:xs)$$
 = True sel1  $(x:xs)$  =  $x$  isNil  $(x:xs)$  = False isCons [] = False sel2  $(x:xs)$  =  $x$ 

Due to the simplicity of their syntax the above checkers and selectors can be automatically instantiated in the logic (*i.e.* without actual calls to the reflected functions at source level) using the measure mechanism of Vazou et al. (2015).

To generalize, let  $D_i$  be a data constructor such that

$$\mathsf{Ty}(D_i) \doteq \tau_{i,1} \to \cdots \to \tau_{i,n} \to \tau$$

Then the check function  $is_{D_i}$  has the sort Fun  $(\tau)$  Bool and the select function  $sel_{D_{i,j}}$  has the sort Fun  $(\tau)$   $(\tau_{i,j})$ .

**Embedding Case Expressions** We translate case-expressions of  $\lambda^R$  into nested if terms in  $\lambda^S$ , by using the check functions in the guards and the select functions for the binders of each case.

$$\Gamma \vdash e \leadsto r \qquad \Gamma \vdash e_i[\overline{y_i} \mapsto \overline{\operatorname{sel}_{D_i} x}][x \mapsto e] \leadsto r_i$$

$$\Gamma \vdash \operatorname{case} x = e \text{ of } \{D_i \ \overline{y_i} \to e_i\} \leadsto \operatorname{if app is}_{D_1} r \text{ then } r_1 \text{ else } \dots \text{ else } r_n$$

For example, the body of the list append function

```
[] ++ ys = ys
(x:xs) ++ ys = x : (xs ++ ys)
```

is reflected into the  $\lambda^S$  refinement: if isNil xs then ys else sell xs: (sel2 xs ++ ys). We favor selectors to the axiomatic translation of HALO (Vytiniotis et al. 2013) to avoid universally quantified formulas and the resulting unpredictability.

## 4.2 Correctness of Translation from $\lambda^R$ to $\lambda^S$

Informally, the translation relation  $\Gamma \vdash e \leadsto r$  is correct in the sense that if e is a terminating boolean expression then e reduces to True  $iff\ r$  is SMT-satisfiable by a model that respects  $\beta$ -equivalence.

*Definition 4.1 (β-Model).* A  $\beta$ -model  $\sigma^{\beta}$  is an extension of a model  $\sigma$  where lam and app satisfy the axioms of  $\beta$ -equivalence:

$$\forall x \ y \ e. \ \text{lam} \ x \ e = \ \text{lam} \ y \ (e[x \mapsto y])$$
  $\forall x \ e_x \ e. \ \text{app} \ (\text{lam} \ x \ e) \ e_x = \ e[x \mapsto e_x]$ 

**Semantics Preservation** We define the translation of a  $\lambda^R$  term into  $\lambda^S$  under the empty environment as (e) = r iff  $0 \vdash e \leadsto r$ . A lifted substitution  $\theta^{\perp}$  is a set of models  $\sigma$  where each "bottom" in the substitution  $\theta$  is mapped to an arbitrary logical value of the respective sort (Vazou et al. 2014). We connect the semantics of  $\lambda^R$  and  $\lambda^S$  via the following:

THEOREM 4.2. If  $\Gamma \vdash e \leadsto r$ , then for every  $\theta \in \llbracket \Gamma \rrbracket$  and every  $\sigma \in \theta^{\perp}$ , if  $\theta^{\perp} \cdot e \hookrightarrow^{\star} v$  then  $\sigma^{\beta} \models r = (v)$ .

Corollary 4.3. If  $\Gamma \vdash e : \mathsf{Bool}$ ,  $e \ reduces \ to \ a \ value \ and \ \Gamma \vdash e \leadsto r$ , then for every  $\theta \in \llbracket \Gamma \rrbracket$  and every  $\sigma \in \theta^{\perp}$ ,  $\theta^{\perp} \cdot e \hookrightarrow^{\star}$  True iff  $\sigma^{\beta} \models r$ .

## 4.3 Decidable Type Checking

We make the type checking from Figure 3 decidable by checking subtyping via an SMT solver.

**Verification Conditions** The implication or *verification condition* (VC) ( $\Gamma$ )  $\Rightarrow$  r is *valid* only if the set of values described by  $\Gamma$  is subsumed by the set of values described by r.  $\Gamma$  is embedded into logic by conjoining (the embeddings of) the refinements of provably terminating binders (Vazou et al. 2014):

$$(|\Gamma|) \doteq \bigwedge_{x \in \Gamma} (|\Gamma, x|) \quad \text{where we embed each binder as} \quad (|\Gamma, x|) \doteq \begin{cases} r & \text{if } \Gamma(x) = \{x : B^{\Downarrow} \mid e\}, \ \Gamma \vdash e \leadsto r \\ \text{True} & \text{otherwise}. \end{cases}$$

**Subtyping via Verification Conditions** We make subtyping, and hence, typing decidable, by replacing the denotational base subtyping rule  $\leq$ -BASE- $\lambda^R$  with the conservative, algorithmic version  $\leq$ -BASE- $\lambda^S$  of Figure 3 that uses an SMT solver to check the validity of the subtyping VC. We use Corollary 4.3 to prove soundness of subtyping.

LEMMA 4.4. If 
$$\Gamma \vdash_S \{v : B \mid e_1\} \leq \{v : B \mid e_2\}$$
 then  $\Gamma \vdash \{v : B \mid e_1\} \leq \{v : B \mid e_2\}$ .

**Soundness of**  $\lambda^S$  We write  $\Gamma \vdash_S e : \tau$  for the judgments that can be derived using the algorithmic subtyping rule  $\leq$ -BASE- $\lambda^S$  instead of the denotational rule  $\leq$ -BASE- $\lambda^R$ . Lemma 4.4 implies the soundness of  $\lambda^S$ .

```
Theorem 4.5 (Soundness of \lambda^S). If \Gamma \vdash_S e : \tau then \Gamma \vdash e : \tau.
```

Note that we *do not* require the  $\beta$ -equivalence axioms for soundness: if a VC is valid *without* the  $\beta$ -equivalence axioms, *i.e.* when lam and app are uninterpreted, then the VC is trivially valid *with* the axioms, and hence, by Theorem 4.5 the program is safe.

#### 5 REASONING ABOUT LAMBDAS

Encoding of  $\lambda$ -abstractions and applications via uninterpreted functions, while sound, is *imprecise* as it makes it hard to prove theorems that require  $\alpha$ - and  $\beta$ -equivalence or extensional equality. Using the universally quantified  $\alpha$ - and  $\beta$ - equivalence axioms would let the type checker accept more programs, but would render validity, and hence, type checking undecidable. Next, we identify a middle ground by describing an not provably complete, but sound and decidable approach to increase the precision of type checking by strengthening the VCs with instances of the  $\alpha$ - and  $\beta$ - equivalence axioms § 5.1 and by introducing a combinator for safely asserting extensional equality § 5.2. In the sequel, we omit app when it is clear from the context.

## 5.1 Equivalence

As soundness relies on satisfiability under a  $\sigma^{\beta}$  (see Definition 4.1), we can safely *instantiate* the axioms of  $\alpha$ - and  $\beta$ -equivalence on any set of terms of our choosing and still preserve soundness (Theorem 4.5). That is, instead of checking the validity of a VC  $p \Rightarrow q$ , we check the validity of a *strengthened VC*,  $a \Rightarrow p \Rightarrow q$ , where a is a (finite) conjunction of *equivalence instances* derived from p and q, as discussed below.

**Representation Invariant** The lambda binders, for each SMT sort, are drawn from a pool of names  $x_i$  where the index  $i = 1, 2, \ldots$  When representing  $\lambda$  terms we enforce a *normalization invariant* that for each lambda term  $\lim x_i e$ , the index i is greater than any lambda argument appearing in e.

 $\alpha$ -instances For each syntactic term  $\limsup x_i \ e$  and  $\lambda$ -binder  $x_j$  such that i < j appearing in the VC, we generate an  $\alpha$ -equivalence instance predicate (or  $\alpha$ -instance):

$$\lim x_i e = \lim x_i e[x_i \mapsto x_i]$$

The conjunction of  $\alpha$ -instances can be more precise than De Bruijn representation, as they let the SMT solver deduce more equalities via congruence. For example, this VC is needed to prove the applicative laws for Reader:

$$d = \operatorname{lam} x_1 (x x_1) \implies \operatorname{lam} x_2 ((\operatorname{lam} x_1 (x x_1)) x_2) = \operatorname{lam} x_1 (d x_1)$$

The  $\alpha$  instance  $\limsup x_1$  ( $d x_1$ ) =  $\limsup x_2$  ( $d x_2$ ) derived from the VC's hypothesis, combined with congruence immediately yields the VC's consequence.

 $\beta$ -instances For each syntactic term app (lam x e)  $e_x$ , with  $e_x$  not containing any  $\lambda$ -abstractions, appearing in the VC, we generate a  $\beta$ -equivalence instance predicate (or  $\beta$ -instance):

app (lam 
$$x_i e$$
)  $e_x = e[x_i \mapsto e_x]$ , s.t.  $e_x$  is  $\lambda$ -free

The  $\lambda$ -free restriction is a simple way to enforce that the reduced term  $e[x_i \mapsto e']$  enjoys the representation invariant. For example, consider the following VC needed to prove that the bind operator for lists satisfies the monadic associativity law.

$$(f x \gg g) = app (lam y (f y \gg g)) x$$

The right-hand side of the above VC generates a  $\beta$ -instance that corresponds directly to the equality, allowing the SMT solver to prove the (strengthened) VC.

**Normalization** The combination of  $\alpha$ - and  $\beta$ -instances is often required to discharge proof obligations. For example, when proving that the bind operator for the Reader monad is associative, we need to prove the VC:

$$lam x_2 (lam x_1 w) = lam x_3 (app (lam x_2 (lam x_1 w)) w)$$

The SMT solver proves the VC via the equalities corresponding to an  $\alpha$  and then  $\beta$ -instance:

$$\lim x_2 (\lim x_1 w) =_{\alpha} \lim x_3 (\lim x_1 w) =_{\beta} \lim x_3 (\operatorname{app} (\lim x_2 (\lim x_1 w)) w)$$

#### 5.2 Extensionality

Often, we need to prove that two functions are equal, given the definitions of reflected binders. Consider

```
reflect ic
id x = x
```

Liquid Haskell accepts the proof that id x = x for all x:

```
id_x_eq_x :: x:a \rightarrow \{id \ x = x\}
id_x_eq_x = \ x \rightarrow id \ x =. \ x ** QED
```

as "calling" id unfolds its definition, completing the proof. However, consider this  $\eta$ -expanded variant of the above proposition:

```
type Id_eq_id = \{(\x \rightarrow id \x) = (\y \rightarrow y)\}
```

Liquid Haskell *rejects* the proof:

```
fails :: Id_eq_id
fails = (\x \rightarrow id \x) = . (\y \rightarrow y) ** QED
```

The invocation of id unfolds the definition, but the resulting equality refinement {id x = x} is *trapped* under the  $\lambda$ -abstraction. That is, the equality is absent from the typing environment at the *top* level, where the left-hand side term is compared to  $y \to y$ . Note that the above equality requires the definition of id and hence is outside the scope of purely the  $\alpha$ - and  $\beta$ -instances.

**An Exensionality Operator** To allow function equality via extensionality, we provide the user with a (family of) *function comparison operator*(s) that transform an *explanation* p which is a proof that f x = g x for every argument x, into a proof that f = g.

```
= \forall :: f:(a \rightarrow b) \rightarrow g:(a \rightarrow b) \rightarrow exp:(x:a \rightarrow \{f \ x = g \ x\}) \rightarrow \{f = g\}
```

Of course, =\forall cannot be implemented; its type is *assumed*. We can use =\forall to prove Id\_eq\_id by providing a suitable explanation:

```
pf_id_id :: Id_eq_id pf_id_id = (\y \rightarrow y) = \forall (\x \rightarrow id x) \because expl ** QED where expl = (\x \rightarrow id x =. x ** QED)
```

The explanation is the second argument to  $\cdot$ : which has the following type that syntactically fires  $\beta$ -instances:

```
x:a \rightarrow \{(\x \rightarrow id x) x = ((\x \rightarrow x) x\}
```

## 6 EVALUATION

We have implemented refinement reflection in Liquid Haskell. In this section, we evaluate our approach by using Liquid Haskell to verify a variety of deep specifications of Haskell functions drawn from the literature and categorized in Figure 4, totalling about 4,000 lines of specifications and proofs. Verification is fast: from 1 to 5s per file (module), which is about twice as much time as GHC takes to compile the corresponding module. Overall there is about one line of type specification (theorems) for three lines of code (implementations and proof). Next, we detail each of the five classes of specifications, illustrate how they were verified using refinement reflection, and discuss the strengths and weaknesses of our approach. *All* of these proofs require refinement reflection, *i.e.* are beyond the scope of shallow refinement typing.

**Proof Strategies.** Our proofs use three building blocks, that are seamlessly connected via refinement typing:

• *Un/folding* definitions of a function f at arguments e1...en, which due to refinement reflection, happens whenever the term f e1 ... en appears in a proof. For exposition, we render the function whose un/folding is relevant as f;

CATEGORY			LOC		
I.	Arithmetic				
	Fibonacci	§ 2	48		
	Ackermann	(Tourlakis 2008)	280		
II. Algebraic Data Types					Monoid
	Fold Universal	(Mu et al. 2009)	105	Left Id.	
III.	Typeclasses	Fig 5		Right Id.	, •
	Monoid	Peano, Maybe, List	189	Assoc	$(x \diamond y) \diamond z \equiv x \diamond (y \diamond z)$
	Functor	Maybe, List, Id, Reader	296		Functor
	Applicative	Maybe, List, Id, Reader	578	Id.	fmap id $xs \equiv id xs$
	Monad	Maybe, List, Id, Reader	435	Distr.	$fmap (g \circ h) xs \equiv (fmap g \circ fmap h) xs$
IV.	IV. Functional Correctness				Applicative
	SAT Solver	(Casinghino et al. 2014)	133	Id.	pure id $\circledast$ $v \equiv v$
	Unification	(Sjöberg and Weirich 2015)	200	Comp.	$pure\;(\circ)\circledast u\circledast v\circledast w\equiv u\circledast (v\circledast w)$
	Chinication	(Sjoberg and Wenten 2013)		Hom.	pure $f \circledast \text{ pure } x \equiv \text{pure } (f x)$
V.	Deterministic Parallelism			Inter.	$u \circledast \text{pure } y \equiv \text{pure } (\$ y) \circledast u$
	Conc. Sets	§ 7.1	906		Monad
	<i>n</i> -body	§ 7.2	930	Left Id.	return $a \gg f \equiv f a$
	Par. Reducers	§ 7.3	55		$m \gg = \text{return} \equiv m$
TOTAL		4155	Assoc	$(m \gg f) \gg g \equiv m \gg (\lambda x \to f \ x \gg g)$	

Fig. 4. Summary of Case Studies

Fig. 5. Summary of Verified Typeclass Laws

- *Lemma Application* which is carried out by using the "because" combinator (:) to instantiate some fact at some inputs;
- *SMT Reasoning* in particular, *arithmetic*, *ordering* and *congruence closure* which kicks in automatically (and predictably!), allowing us to simplify proofs by not having to specify, *e.g.* which subterms to rewrite.

### 6.1 Arithmetic Properties

The first category of theorems pertains to the textbook Fibonacci and Ackermann functions. In § 2 we proved that fib is increasing. We prove a higher order theorem that lifts increasing functions to monotonic ones:

```
fMono :: f: (Nat \rightarrow Int) \rightarrow fUp: (z:Nat \rightarrow \{f \ z \leq f \ (z+1)\}) \rightarrow x: Nat \rightarrow y: \{x < y\} \rightarrow \{f \ x \leq f \ y\}
```

By instantiating the function argument f of fMono with fib, we proved monotonicity of fib.

```
\label{eq:fibMono} \begin{array}{ll} \text{fibMono} :: \ n \colon \! Nat \ \to \ m \colon \! \{n < m\} \ \to \ \{fib \ n \le \ fib \ m\} \\ \text{fibMono} \ = \ fMono \ fib \ fibUp \end{array}
```

Using higher order functions like fMono, we mechanized the proofs of the Ackermann function properties from (Tourlakis 2008). We proved 12 equational and arithmetic properties using various proof techniques, including instantiation of higher order theorems (like fMono), proof by construction, and natural number and generalized induction.

## 6.2 Algebraic Data Properties

Fold Universality as encoded in Adga (Mu et al. 2009). The following encodes the universal property of foldr

The Liquid Haskell proof term of foldr\_univ is similar to the one in Agda. But, there are two major differences. First, specifications do not support quantifiers: universal quantification of x and xs in the step assumption is encoded as functional arguments. Second, unlike Agda, Liquid Haskell does not support optional arguments, thus to use foldr\_univ the proof arguments base and step that were implicit arguments in Agda need to be explicitly provided. For example, the following code calls foldr\_universal to prove foldr\_fusion by explicitly applying proofs for the base and the step arguments

```
foldr_fusion :: h:(b \rightarrow c)

\rightarrow f:(a \rightarrow b \rightarrow b)

\rightarrow g:(a \rightarrow c \rightarrow c)

\rightarrow e:b

\rightarrow z:[a]

\rightarrow x:a \rightarrow y:b

\rightarrow fuse: {h (f x y) == g x (h y)})

\rightarrow {(h.foldr f e) z == foldr g (h e) z}

foldr_fusion h f g e ys fuse = foldr_univ g (h . foldr f e) (h e) ys

(fuse_base h f e)

(fuse_step h f e g fuse)
```

Where, for example, fuse\_base is a function with type

```
fuse\_base :: h:(b \rightarrow c) \rightarrow f:(a \rightarrow b \rightarrow b) \rightarrow e:b \rightarrow \{(h . foldr f e) [] == h e\}
```

## 6.3 Typeclass Laws

We used Liquid Haskell to prove the Monoid, Functor, Applicative, and Monad Laws, summarized in Figure 5, for various user-defined instances summarized in Figure 4.

**Monoid Laws** A Monoid is a datatype equipped with an associative binary operator  $\diamond$  (mappend) and an *identity* element mempty. We prove that Peano (with a suitable add and Z), Maybe (with a suitable mappend and Nothing), and List (with append ++ and []) satisfy the monoid laws.

**Functor Laws** A type is a functor if it has a function fmap that satisfies the *identity* and *distribution* (or fusion) laws in Figure 5. For example, consider the proof of the map distribution law for the lists, also known as "mapfusion", which is the basis for important optimizations in GHC (Wiki 2008). We reflect the definition of map for lists and use it to specify fusion and verify it by an inductive proof:

```
map_fusion :: f:(b \rightarrow c) \rightarrow g:(a \rightarrow b) \rightarrow xs:[a] \rightarrow {map (f . g) xs = (map f . map g) xs}
```

**Monad Laws** The monad laws, which relate the properties of the two operators  $\gg$  and return (Figure 5), refer to  $\lambda$ -functions which are encoded in our logic as uninterpreted functions. For example, we can encode the associativity property as a refinement type alias:

```
type Assoclaw m f g = { m >>= f >>= g = m >>= (\x \rightarrow f x >>= g) }
```

and use it to prove that the list-bind is associative:

Where the bind\_append fusion lemma states that:

```
bind_append :: xs:[a] \rightarrow ys:[a] \rightarrow f:(a \rightarrow [b]) \rightarrow \{(xs++ys) >>= f = (xs >>= f)++(ys >>= f)\}
```

Notice that the last step requires  $\beta$ -equivalence on anonymous functions, which we get by explicitly inserting the redex in the logic, via the following lemma with trivial proof

```
\betaeq :: f:_ \rightarrow g:_ \rightarrow x:_ \rightarrow {f x >>= g = (\y \rightarrow f y >>= g) x} \betaeq _ _ = trivial
```

#### 6.4 Functional Correctness

Next, we proved correctness of two programs from the literature: a unification algorithm and a SAT solver.

**Unification** We verified the unification of first order terms, as presented in (Sjöberg and Weirich 2015). First, we define a predicate alias for when two terms s and t are equal under a substitution su:

```
eq_sub su s t = apply su s == apply su t
```

Now, we defined a Haskell function unify s t that can diverge, or return Nothing, or return a substitution su that makes the terms equal:

```
unify :: s:Term \rightarrow t:Term \rightarrow Maybe \{su \mid eq\_sub su s t\}
```

For specification and verification, we only needed to reflect apply and not unify; thus, we only had to verify that the former terminates, and not the latter. We proved correctness by invoking separate helper lemmata. For example, to prove the post-condition when unifying a variable TVar i with a term t in which i *does not* appear, we apply a lemma not\_in:

```
unify (TVar i) t2 | not (i \in freeVars t2) = Just (const [(i, t2)] \because not_in i t2)
```

i.e. if i is not free in t, the singleton substitution yields t:

```
not_in :: i:Int \rightarrow t:\{Term \mid not (i \in freeVars t)\} \rightarrow \{eq\_sub [(i,t)] (TVar i) t\}
```

This example highlights the benefits of partial verification on a legacy programming language: potential diverging code (*e.g.* the function unify) coexists and invokes proof terms.

**SAT Solver** As another example, we implemented and verified the simple SAT solver used to illustrate and evaluate the features of the dependently typed language Zombie (Casinghino et al. 2014). The solver takes as input a formula f and returns an assignment that *satisfies* f if one exists, as specified below.

```
solve :: f:Formula → Maybe {a:Assignment | sat a f}
```

The function sat a f returns True iff the assignment a satisfies the formula f. Verifying solve follows directly by reflecting sat into the refinement logic.

#### 7 VERIFIED DETERMINISTIC PARALLELISM

Finally, we evaluate our deterministic parallelism prototypes. Aside from the lines of proof code added, we evaluate the impact on runtime performance. Were we using a proof tool external to Haskell, this would not be necessary. But our proofs are Haskell programs—they are necessarily visible to the compiler. In particular, this means a proliferation of unit values and functions returning unit values. Also, typeclass instances are witnessed at runtime by "dictionary" data structures passed between functions. Layering proof methods on top of existing classes like Ord (from § 2.4) could potentially add indirection or change the code generated, depending on the details of the optimizer. In our experiments we find little or no effect on runtime performance. Benchmarks were run on a single-socket Intel® Xeon® CPU E5-2699 v3 with 18 physical cores and 64GiB RAM.

#### 7.1 LVish: Concurrent Sets

First, we use the verifiedInsert operation (from § 2.4) to observe the runtime slowdown imposed by the extra proof methods of VerifiedOrd. We benchmark concurrent sets storing 64-bit integers. Figure 6 compares the parallel speedups for a fixed number of parallel insert operations against parallel verifiedInsert operations, varying the number of concurrent threads. There is a slight observable difference between the two lines because the extra proof methods do exist at runtime. We repeat the experiment for two set implementations: a concurrent skiplist (SLSet) and a purely functional set inside an atomic reference (PureSet) as described in Kuper et al. (2014).

## 7.2 monad-par: *n*-body simulation

Next, we verify deterministic behavior of an *n*-body simulation program that leverages monad-par, a Haskell library which provides deterministic parallelism for pure code (Marlow et al. 2011).

Each simulated particle is represented by a type Body that stores its position, velocity, and mass. The function accel computes the relative acceleration between two bodies:

```
accel :: Body \rightarrow Body \rightarrow Accel
```

where Accel represents the three-dimensional acceleration

```
data Accel = Accel Real Real Real
```

To compute the total acceleration of a body b we (1) compute the relative acceleration between b and each body of the system (Vec Body) and (2) we add each acceleration component. For efficiency, we use a parallel mapReduce for the above computation that first *maps* each vector body to get the acceleration relative to b (accel b) and then adds each Accel value by pointwise addition. mapReduce is only deterministic if the element is a VerifiedMonoid from § 2.4.

```
mapReduce :: VerifiedMonoid b \Rightarrow (a \rightarrow b) \rightarrow Vec a \rightarrow b
```

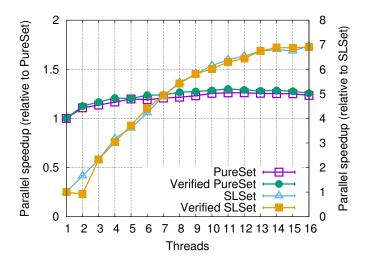


Fig. 6. Parallel speedup for doing 1 million parallel inserts over 10 iterations, verified and unverified, relative to the unverified version, for PureSet and SLSet.

To enforce the determinism of an n-body simulation, we need to provide a VerifiedMonoid instance for Accel. We can prove that (Real, +, 0.0) is a monoid. By product proof composition, we get a verified monoid instance for

```
type Accel' = (Real, (Real, Real))
which is isomorphic to Accel (i.e. Iso Accel' Accel).
```

Figure 7 shows the results of running two versions of the n-body simulation with 2,048 bodies over 5 iterations, with and without verification, using floating point doubles for Real<sup>1</sup>. Notably, the two programs have almost identical runtime performance. This demonstrates that even when verifying code that is run in a tight loop (like accel), we can expect that our programs will not be slowed down by an unacceptable amount.

## 7.3 DPJ: Parallel Reducers

The Deterministic Parallel Java (DPJ) project provides a deterministic-by-default semantics for the Java programming language (Bocchino et al. 2009). In DPJ, one can declare a method as commutative and thus *assert* that racing instances of that method result in a deterministic outcome. For example:

```
commutative void updateSum(int n) writes R { sum += n; }
```

But, DPJ provides no means to formally prove commutativity and thus determinism of parallel reduction. In Liquid Haskell, we specified commutativity as an extra proof method that extends the VerifiedMonoid class.

```
class VerifiedMonoid a \Rightarrow VerifiedCommutativeMonoid a where commutes :: x:a \rightarrow y:a \rightarrow { x <> y = y <> x }
```

Provably commutative appends can be used to deterministically update a reducer variable, since the result is the same regardless of the order of appends. We used LVish (Kuper et al. 2014) to encode a reducer variable with a value a and a region s as RVar s a.

<sup>&</sup>lt;sup>1</sup>Floating point numbers notoriously violate associativity, but we use this approximation because Haskell does net yet have an implementation of *superaccumulators* (Collange et al. 2014).

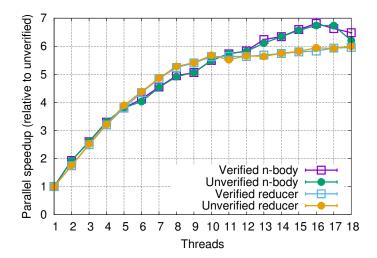


Fig. 7. Parallel speedup for doing a parallel *n*-body simulation and parallel array reduction. The speedup is relative to the unverified version of each respective class of program.

#### newtype RVar s a

We specify that safe (i.e. deterministic) parallel updates require provably commutative appending.

```
updateRVar :: VerifiedCommutativeMonoid a \Rightarrow a \rightarrow RVar s a \rightarrow Par s ()
```

Following the DPJ program, we used updateRVar's provably deterministic interface to compute, in parallel, the sum of an array with  $3x10^9$  elements by updating a single, global reduction variable using a varying number of threads. Each thread sums segments of an array, sequentially, and updates the variable with these partial sums. In Figure 7, we compare the verified and unverified versions of our implementation to observe no appreciable difference in performance.

## 8 RELATED WORK

**SMT-Based Verification** SMT-solvers have been extensively used to automate program verification via Floyd-Hoare logics (Nelson 1981). Our work is inspired by Dafny's Verified Calculations (Leino and Polikarpova 2016), a framework for proving theorems in Dafny (Leino 2010), but differs in (1) our use of reflection instead of axiomatization and (2) our use of refinements to compose proofs. Dafny, and the related F\* (Swamy et al. 2016) which like Liquid Haskell, uses types to compose proofs, offer more automation by translating recursive functions to SMT axioms. However, unlike reflection, this axiomatic approach renders typechecking and verification undecidable (in theory) and leads to unpredictability and divergence (in practice) (Leino and Pit-Claudel 2016).

**Dependent types** Our work is also inspired by dependently typed systems like Coq (Bertot and Castéran 2004) and Agda (Norell 2007). Reflection shows how deep specification and verification in the style of Coq and Agda can be *retrofitted* into existing languages via refinement typing. Furthermore, we can use SMT to significantly automate reasoning over important theories like arithmetic, equality and functions. It would be interesting to investigate how the tactics and sophisticated proof search of Coq *etc.* can be adapted to the refinement setting.

**Dependent Types in Haskell** Integration of dependent types into Haskell has been a long standing goal that dates back to Cayenne (Augustsson 1998), a Haskell-like, fully dependent type language with undecidable type

checking. In a recent line of work Eisenberg and Stolarek (2014) aim to allow fully dependent programming within Haskell, by making "type-level programming ... at least as expressive as term-level programming". Our approach differs in two significant ways. First, reflection allows SMT-aided verification, which drastically simplifies proofs over key theories like linear arithmetic and equality. Second, refinements are completely erased at run-time. That is, while both systems automatically lift Haskell code to either uninterpreted logical functions or type families, with refinements, the logical functions are not accessible at run-time and promotion cannot affect the semantics of the program. As an advantage (resp. disadvantage), refinements cannot degrade (resp. optimize) the performance of programs.

**Proving Equational Properties** Several authors have proposed tools for proving (equational) properties of (functional) programs. Systems Sousa and Dillig (2016) and Asada et al. (2015) extend classical safety verification algorithms, respectively based on Floyd-Hoare logic and Refinement Types, to the setting of relational or *k*-safety properties that are assertions over *k*-traces of a program. Thus, these methods can automatically prove that certain functions are associative, commutative *etc.*. but are restricted to first-order properties and are not programmer-extensible. Zeno (Sonnex et al. 2012) generates proofs by term rewriting and Halo (Vytiniotis et al. 2013) uses an axiomatic encoding to verify contracts. Both the above are automatic, but unpredictable and not programmer-extensible, hence, have been limited to far simpler properties than the ones checked here. HERMIT (Farmer et al. 2015) proves equalities by rewriting the GHC core language, guided by user specified scripts. In contrast, our proofs are simply Haskell programs, we can use SMT solvers to automate reasoning, and, most importantly, we can connect the validity of proofs with the semantics of the programs.

**Deterministic Parallelism** Deterministic parallelism has plenty of theory but relatively few practical implementations. Early discoveries were based on limited producer-consumer communication, such as single-assignment variables (Arvind et al. 1989; Tesler and Enea 1968), Kahn process networks (Kahn 1974), and synchronous dataflow (Lee and Messerschmitt 1987). Other models use synchronous updates to shared state, as in Esterel (Benveniste et al. 2003) or PRAM. Finally, work on type systems for permissions management (Naden et al. 2012; Westbrook et al. 2012), supports the development of *non-interfering* parallel programs that access disjoint subsets of the heap in parallel. Parallel functional programming is also non-interfering (Fluet et al. 2007; Marlow et al. 2009). Irrespective of which theory is used to support deterministic parallel programming, practical implementations such as Cilk (Blumofe et al. 1995) or Intel CnC (Budimlic et al. 2010) are limited by host languages with type systems insufficient to limit side effects, much less prove associativity. Conversely, dependently typed languages like Agda and Idris do not have parallel programming APIs and runtime systems.

#### 9 CONCLUSIONS AND FUTURE WORK

We have shown how refinement reflection—namely reflecting the definitions of functions in their output refinements—can be used to convert a legacy programming language, like Haskell, into a theorem prover. Reflection ensures that (refinement) type checking stays decidable and predictable via careful design of the logic and proof combinators. Reflection enables programmers working with the highly tuned libraries, compilers and run-times of legacy languages to specify and verify arbitrary properties of their code simply by writing programs in the legacy language. Our evaluation shows that refinement reflection lets us prove deep specifications of a variety of implementations, for example, the determinism of fast parallel programming libraries, and also identifies two important avenues for research.

First, while our proofs are often elegant and readable, they can sometimes be cumbersome. For example, in the proof of associativity of the monadic bind operator for the Reader monad three out of eight (extensional) equalities required explanations, some nested under multiple  $\lambda$ -abstractions. Thus, it would be valuable to explore how to extend the notions of tactics, proof search, and automation to the setting of legacy languages. Similarly, while our approach to  $\alpha$ - and  $\beta$ -equivalence is sound, we do not know if it is *complete*. We conjecture it is, due to

the fact that our refinement terms are from the simply typed lambda calculus (STLC). Thus, it would be interesting to use the normalization of STLC to develop a sound and complete system for SMT-based type-level computation and use it to automate proofs predictably.

Second, while Haskell's separation of pure and effectful code undoubtedly makes it easier to implement refinement reflection, we believe that the technique, like refinement typing in general, is orthogonal to purity. In particular, by carefully mediating between the interaction of pure and impure code using methods like permissions, uniqueness *etc.* refinement type systems have been developed for legacy languages like C (Rondon et al. 2010), JavaScript (Chugh et al. 2012; Vekris et al. 2016), and Racket (Kent et al. 2016) and so it would be interesting to see how to extend refinement reflection to other legacy languages.

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