Abstract Refinement Types

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Abstract. We present *abstract refinement types* which enable quantification over the refinements of data- and function-types. Our key insight is that we can avail of quantification while preserving SMT-based decidability, simply by encoding refinement parameters as *uninterpreted* propositions within the refinement logic. We illustrate how this mechanism yields a variety of sophisticated means for reasoning about programs, including: *parametric* refinements for reasoning with type classes, *index-dependent* refinements for reasoning about key-value maps, *recursive* refinements for reasoning about recursive data types, and *inductive* refinements for reasoning about higher-order traversal routines. We have implemented our approach in a refinement type checker for Haskell, and present experiments using our tool to verify correctness invariants of various programs.

1 Introduction

Refinement types offer an automatic means of verifying semantic properties of programs, by decorating types with predicates from logics efficiently decidable by modern SMT solvers. For example, the refinement type {v: Int | v > 0} denotes the basic type Int refined with a logical predicate over the "value variable" v. This type corresponds to the set of Int values v which additionally satisfy the logical predicate, *i.e.*, the set of positive integers. The (dependent) function type x: {v:Int | v > 0} -> {v:Int | v < x} describes functions that take a positive argument x and return an integer less than x. Refinement type checking reduces to *subtyping* queries of the form $\Gamma \vdash {\tau: \nu \mid p} \preceq {\tau: \nu \mid q}$, where p and q are refinement predicates. These subtyping queries reduce to logical *validity* queries of the form $\llbracket \Gamma \rrbracket \land p \Rightarrow q$, which can be automatically discharged using SMT solvers [6].

Several groups have shown how refinement types can be used to verify properties ranging from partial correctness concerns like array bounds checking [27, 23] and data structure invariants [16] to the correctness of security protocols [2], web applications [14] and implementations of cryptographic protocols [10].

Unfortunately, the automatic verification offered by refinements has come at a price. To ensure decidable checking with SMT solvers, the refinements are quantifier-free predicates drawn from a decidable logic. This significantly limits expressiveness by precluding specifications that enable abstraction over the refinements (*i.e.*, invariants). For example, consider the following higher-order for-loop where set $i \times v$ returns the vector v updated at index i with the value x.

We would like to verify that initUpto returns a vector whose *first* n elements are equal to x. In a first-order setting, we could achieve the above with a loop invariant that asserted that at the ith iteration, the first i elements of the vector were already initalized to x. However, in our higher-order setting we require a means of *abstracting* over possible invariants, each of which can *depend on* the iteration index i. Higher-order logics like Coq and Agda permit such quantification over invariants. Alas, validity in such logics is well outside the realm of decidability, and hence their use precludes automatic verification.

In this paper, we present *abstract refinement types* which enable abstraction (quantification) over the refinements of data- and function-types. Our key insight is that we can preserve SMT-based decidable type checking by encoding abstract refinements as *uninterpreted* propositions in the refinement logic. This yields several contributions:

- First, we illustrate how abstract refinements yield a variety of sophisticated means for reasoning about high-level program constructs (§2), including: *parametric* refinements for type classes, *index-dependent* refinements for key-value maps, *recursive* refinements for data structures, and *inductive* refinements for higher-order traversal routines.
- Second, we demonstrate that type checking remains decidable (§3) by showing a fully automatic procedure that uses SMT solvers, or to be precise, decision procedures based on congruence closure [19] to discharge logical subsumption queries over abstract refinements.
- Third, we show that the crucial problem of *inferring* appropriate instantiations for the (abstract) refinement parameters boils down to inferring (non-abstract) refinement types (§3), which we have previously automated via the abstract interpretation framework of Liquid Types [23].
- Finally, we have implemented abstract refinements in HSOLVE, a new Liquid Typebased verifier for Haskell. We present experiments using HSOLVE to concisely specify and verify a variety of correctness properties of several programs ranging from microbenchmarks to some widely used libraries (§4).

2 Overview

We start with a high level overview of abstract refinements, by illustrating how they can be used to uniformly specify and automatically verify various kinds of invariants.

2.1 Parametric Invariants

Parametric Invariants via Type Polymorphism. Suppose we had a generic comparison (\leq =) :: a \rightarrow a \rightarrow Bool as in OCAML. We could use it to write:

```
max :: a -> a -> a
max x y = if x <= y then y else x
maximum :: [a] -> a
maximum (x:xs) = foldr max x xs
```

In essence, the type given for maximum states that *for any* a, if a list of a values is passed into maximum, then the returned result is also an a value. Hence, for example, if a list of *prime* numbers is passed in, the result is prime, and if a list of *even* numbers is passed in, the result is prime, and if a list of *even* numbers is passed in, the result is even. Thus, we can use refinement types [23] to verify

```
type Even = {v:Int | v % 2 = 0 }
maxEvens :: [Int] -> Even
maxEvens xs = maximum (0 : xs')
where xs' = [ x | x <- xs, x `mod` 2 = 0]</pre>
```

Here the % represents the modulus operator in the refinement logic [6] and we type the primitive mod :: x:Int -> y:Int -> {v: Int | v = x % y}. Verification proceeds as follows. Given that xs :: [Int], the system has to verify that maximum (0 : xs'):: Even. To this end, the type parameter of maximum is instantiated with the *refined* type Even, yielding the instance:

maximum :: [Even] -> Even

Then, maximum's argument should be proved to have type [Even]. So, the type parameter of (:) is instantiated with Even, yielding the instance:

(:) :: Even -> [Even] -> [Even]

Finally, the system infers that 0 :: Even and xs' :: [Even], *i.e.*, the arguments of (:) have the expected types, thereby verifying the program. The refinement type instantiations can be inferred, from an appropriate set of logical qualifiers, using the abstract interpretation framework of Liquid Types [23]. Here, once $v \ge 2 = 0$ is added to the set of qualifiers, either manually or (as done by our implementation) by automatically scraping predicates from refinements appearing in specification signatures, the refinement type instantiations, and hence verification, proceed automatically. Thus, parametric polymorphism offers an easy means of encoding second-order invariants, *i.e.*, of quantifying over or parametrizing the invariants of inputs and outputs of functions.

Parametric Invariants via Abstract Refinements. Instead, suppose that the comparison operator was monomorphic, and only worked for Int values. The resulting (monomorphic) signatures

```
max :: Int -> Int -> Int
maximum :: [Int] -> Int
```

preclude the verification of maxEvens (*i.e.*, typechecking against the signature shown earlier). This is because the new type of maximum merely states that *some* Int is returned as output, and not necessarily one that enjoys the properties of the values in the input list. This is a shame, since the property clearly still holds. We could type

max :: forall t <: Int. t -> t -> t

but this route would introduce the complications that surround bounded quantification which could render checking undecidable [22].

To solve this problem, we introduce *abstract refinements* which let us quantify or parameterize a type over its constituent refinements. For example, we can type max as

max :: forall <p::Int->Bool>. Int -> Int -> Int

where Int is an abbreviation for the refinement type {v:Int | p(v) }. Intuitively, an abstract refinement p is encoded in the refinement logic as an *uninterpreted function symbol*, which satisfies the *congruence* axiom [19]

$$\forall \overline{X}, \overline{Y} : (\overline{X} = \overline{Y}) \Rightarrow P(\overline{X}) = P(\overline{Y})$$

Thus, it is trivial to verify, with an SMT solver, that max enjoys the above type: the input types ensure that both p(x) and p(y) hold and hence the returned value in either branch satisfies the refinement {v:Int | p(v)}, thereby ensuring the output type. By the same reasoning, we can generalize the type of maximum to

maximum :: forall Bool>. [Int] -> Int

Consequently, we can recover the verification of maxEvens. Now, instead of instantiating a *type* parameter, we simply instantiate the *refinement* parameter of maximum with the concrete refinement { $\langle v \rangle \rangle \langle 2 \rangle = 0$ }, after which type checking proceeds as usual [23]. Later, we show how to retain automatic verification by inferring refinement parameter instantiations via liquid typing (§ 3.4).

Parametric Invariants and Type Classes. The example above regularly arises in practice, due to type classes. In Haskell, the functions above are typed

(<=) :: (Ord a) => a -> a -> Bool
max :: (Ord a) => a -> a -> a
maximum :: (Ord a) => [a] -> a

We might be tempted to ignore the typeclass constraint, and treat maximum as [a] ->

a. This would be quite unsound, as typeclass predicates preclude universal quantification over refinement types. Consider the function sum :: (Num a) => [a] -> awhich adds the elements of a list. The Num class constraint implies that numeric operations occur in the function, so if we pass sum a list of odd numbers, we are *not* guaranteed to get back an odd number.

Thus, how do we soundly verify the desired type of maxEvens without instantiating class predicated type parameters with arbitrary refinement types? First, via the same analysis as the monomorphic Int case, we establish that

max:: forall <p::a->Bool>. (Ord a) => a -> a -> a a a -> a -> a a

Next, at the call-site for maximum in maxEvens we instantiate the type variable a with Int, and the abstract refinement p with $\{v \rightarrow v & 2 = 0\}$ after which, the verification proceeds as described earlier (for the Int case). Thus, abstract refinements allow us to quantify over invariants without relying on parametric polymorphism, even in the presence of type classes.

2.2 Index-Dependent Invariants

Next, we illustrate how abstract invariants allow us to specify and verify indexdependent invariants of key-value maps. To this end, we develop a small library of *extensible vectors* encoded, for purposes of illustration, as functions from Int to some generic range a. Formally, we specify vectors as

```
data Vec a <dom :: Int -> Bool, rng :: Int -> a -> Bool>
    = V (i:Int<dom> -> a <rng i>)
```

Here, we are parameterizing the definition of the type Vec with *two* abstract refinements, dom and rng, which respectively describe the *domain* and *range* of the vector. That is, dom describes the set of *valid* indices, and r specifies an invariant relating each Int index with the value stored at that index.

Creating Vectors. We can use the following basic functions to create vectors:

```
empty :: forall <p::Int->a->Bool>.Vec<{\_ -> False}, p> a
empty = V (\_ -> error "Empty Vec")
create :: x:a -> Vec <{\_ -> True}, {\_ v -> v = x}> a
create x = V (\_ -> x)
```

The signature for empty states that its domain is empty (*i.e.*, is the set of indices satisfying the predicate **False**), and that the range satisfies any invariant. The signature for create, instead, defines a *constant* vector that maps every index to the constant x. Accessing Vectors. We can write the following get function for reading the contents of a vector at a given index:

The signature states that for any domain d and range r, if the index i is a valid index, *i.e.*, is of type, Int<d> then the returned value is an a that additionally satisfies the range refinement at the index i. The type for set, which *updates* the vector at a given index, is even more interesting, as it allows us to *extend* the domain of the vector:

The signature for set requires that (a) the input vector is defined everywhere at d *except* the index i, and (b) the value supplied must be of type a < r i>, *i.e.*, satisfy the range relation at the index i at which the vector is being updated. The signature ensures that the output vector is defined at d and each value satisfies the index-dependent range refinement r. Note that it is legal to call get with a vector that is *also* defined at the index i since, by contravariance, such a vector is a subtype of that required by (a).

Initializing Vectors. Next, we can write the following function, init, that "loops" over a vector, to set each index to a value given by some function.

```
initialize :: forall <r :: Int -> a -> Bool>.
        (z: Int -> a<r z>)
        -> i: {v: Int | v >= 0}
        -> n: Int
        -> Vec <{\v -> 0 <= v && v < i}, r> a
        -> Vec <{\v -> 0 <= v && v < n}, r> a
initialize f i n a
        | i >= n = a
        | otherwise = initialize f (i+1) n (set i (f i) a)
```

The signature requires that (a) the higher-order function f produces values that satisfy the range refinement r, and (b) the vector is initialized from 0 to i. The function ensures that the output vector is initialized from 0 through n. We can thus verify that

```
idVec :: Vec \{ v \rightarrow 0 \le v \& v \le n \}, \{ v \rightarrow v = i \} > Int idVec n = initialize (<math>i \rightarrow i) 0 n empty
```

i.e., idVec returns a vector of size n where each key is mapped to itself. Thus, abstract refinement types allow us to verify low-level idioms such as the incremental initialization of vectors, which have previously required special analyses [12, 15, 5].

Null-Terminated Strings. We can also use abstract refinements to verify code which manipulates C-style null-terminated strings, represented as Char vectors for ease of exposition. Formally, a null-terminated string of size n has the type

```
type NullTerm n
= Vec <{\v -> 0<=v<n}, {\i v -> i=n-1 => v='\0'}> Char
```

The above type describes a length-n vector of characters whose last element must be a null character, signalling the end of the string. We can use this type in the specification of a function, upperCase, which iterates through the characters of a string, uppercasing each one until it encounters the null terminator:

Note that the length parameter n is provided solely as a "witness" for the length of the string s, which allows us to use the length of s in the type of upperCase; n is not used in the computation. In order to establish that each call to get accesses string s within its bounds, our type system must establish that, at each call to the inner function ucs, i satisfies the type {v: Int | 0 <= v && v < n}. This invariant is established as follows. First, the invariant trivially holds on the first call to ucs, as n is positive and i is 0. Second, we assume that i satisfies the type {v: Int | 0 <= v && v < n}, and, further, we know from the types of s and get that c has the type {v: Char $| i = n - 1 => v = ' \setminus 0'$ }. Thus, if c is non-null, then i cannot be equal to n - 1. This allows us to strengthen our type for i in the else branch to {v: Int | 0 <= v && v < n - 1} and thus to conclude that the value i + 1 recursively passed as the i parameter to ucs satisfies the type {v: Int | 0 <= v && v < n - 1} and thus the safety of the upperCase function.

Memoization. Next, let us illustrate how the same expressive signatures allow us to verify memoizing functions. We can specify to the SMT solver the definition of the Fibonacci function via an uninterpreted function fib and an axiom:

```
measure fib :: Int -> Int
axiom: forall i. fib(i) = i<=1 ? 1 : fib(i-1) + fib(i-2)</pre>
```

Next, we define a type alias FibV for the vector whose values are either 0 (*i.e.*, undefined), or equal to the Fibonacci number of the corresponding index.

type FibV = Vec<{_->True}, {\i v-> v!=0 => v=fib(i)} > Int

Finally, we can use the above alias to verify fastFib, an implementation of the Fibonacci function, which uses a vector memoize intermediate results

Thus, abstract refinements allow us to define key-value maps with index-dependent refinements for the domain and range. Quantification over the domain and range refinements allows us to define generic access operations (*e.g.*, get, set, create, empty) whose types enable us establish a variety of precise invariants.

2.3 Recursive Invariants

Next, we turn our attention to recursively defined datatypes, and show how abstract refinements allow us to specify and verify high-level invariants that relate the elements of a recursive structure. Consider the following refined definition for lists:

```
data [a]  a -> Bool> where
[] :: [a]
(:) :: h:a -> [a] -> [a]
```

The definition states that a value of type [a] is either empty ([]) or constructed from a pair of a *head* h:: a and a *tail* of a list of a values *each* of which satisfies the refinement $(p \ h)$. Furthermore, the abstract refinement p holds recursively within the tail, ensuring that the relationship p holds between *all* pairs of list elements.

Thus, by plugging in appropriate concrete refinements, we can define the following aliases, which correspond to the informal notions implied by their names:

```
type IncrList a = [a] <{ \h v -> h <= v }>
type DecrList a = [a] <{ \h v -> h >= v }>
type UniqList a = [a] <{ \h v -> h != v }>
```

That is, IncrList a (resp. DecrList a) describes a list sorted in increasing (resp. decreasing) order, and UniqList a describes a list of *distinct* elements, *i.e.*, not containing any duplicates. We can use the above definitions to verify

[1, 2, 3, 4] :: IncrList Int
[4, 3, 2, 1] :: DecrList Int
[4, 1, 3, 2] :: UniqList Int

More interestingly, we can verify that the usual algorithms produce sorted lists:

```
insertSort :: (Ord a) => [a] -> IncrList a
insertSort [] = []
insertSort (x:xs) = insert x (insertSort xs)
insert :: (Ord a) => a -> IncrList a -> IncrList a
insert y [] = [y]
insert y (x:xs)
  | y <= x = y : x : xs
  | otherwise = x : insert y xs
```

Thus, abstract refinements allow us to *decouple* the definition of the list from the actual invariants that hold. This, in turn, allows us to conveniently reuse the same underlying (non-refined) type to implement various algorithms unlike, say, singleton-type based implementations which require up to three different types of lists (with three different "nil" and "cons" constructors [24]). This, makes abstract refinements convenient for verifying complex sorting implementations like that of Data.List.sort which, for efficiency, use lists with different properties (*e.g.*, increasing and decreasing).

Multiple Recursive Refinements. We can define recursive types with multiple parameters. For example, consider the following refined version of a type used to encode functional maps (Data.Map):

```
data Tree k v <l :: k->k->Bool, r :: k->k->Bool>
= Bin { key :: k
    , value :: v
    , left :: Tree <l, r> (k <l key>) v
    , right :: Tree <l, r> (k <r key>) v }
  | Tip
```

The abstract refinements 1 and r relate each key of the tree with *all* the keys in the *left* and *right* subtrees of key, as those keys are respectively of type k <1 key> and k <r key>. Thus, if we instantiate the refinements with the following predicates

```
type BST k v = Tree<{x y \rightarrow x > y}, {x y \rightarrow x < y} k v

type MinHeap k v = Tree<{x y \rightarrow x < y}, {x y \rightarrow x < y} k v

type MaxHeap k v = Tree<{x y \rightarrow x < y, {x y \rightarrow x < y} k v
```

then BST k v, MinHeap k v and MaxHeap k v denote exactly binary-searchordered, min-heap-ordered, and max-heap-ordered trees (with keys and values of types k and v). We demonstrate in (§ 4) how we use the above types to automatically verify ordering properties of complex, full-fledged libraries.

2.4 Inductive Invariants

Finally, we explain how abstract refinements allow us to formalize some kinds of structural induction within the type system.

Measures. First, let us formalize a notion of *length* for lists within the refinement logic. To do so, we define a special len measure by structural induction

```
measure len :: [a] -> Int
len [] = 0
len (x:xs) = 1 + len(xs)
```

We use the measures to automatically strengthen the types of the data constructors[16]:

```
data [a] where
[] :: forall a.{v:[a] | len(v) = 0}
(:) :: forall a.a -> xs:[a] -> {v:[a]|len(v)=1+len(xs)}
```

Note that the symbol len is encoded as an *uninterpreted* function in the refinement logic, and is, except for the congruence axiom, opaque to the SMT solver. The measures are guaranteed, by construction, to terminate, and so we can soundly use them as uninterpreted functions in the refinement logic. Notice also, that we can define *multiple* measures for a type; in this case we simply conjoin the refinements from each measure when refining each data constructor.

With these strengthened constructor types, we can verify, for example, that append produces a list whose length is the sum of the input lists' lengths:

append :: l:[a] -> m:[a] -> {v:[a] |len(v)=len(l)+len(m)}
append [] zs = zs
append (y:ys) zs = y : append ys zs

However, consider an alternate definition of append that uses foldr

append ys zs = foldr (:) zs ys

where foldr :: $(a \rightarrow b \rightarrow b) \rightarrow b \rightarrow [a] \rightarrow b$. It is unclear how to give foldr a (first-order) refinement type that captures the rather complex fact that the fold-function is "applied" all over the list argument, or, that it is a catamorphism. Hence, hitherto, it has not been possible to verify the second definition of append.

Typing Folds. Abstract refinements allow us to solve this problem with a very expressive type for foldr whilst remaining firmly within the boundaries of SMT-based decidability. We write a slightly modified fold:

The trick is simply to quantify over the relationship p that foldr establishes between the input list xs and the output b value. This is formalized by the type signature, which encodes an induction principle for lists: the base value b must (1) satisfy the relation with the empty list, and the function op must take (2) a value that satisfies the relationship with the tail xs (we have added the xs as an extra "ghost" parameter to op), (3) a head value x, and return (4) a new folded value that satisfies the relationship with x:xs. If all the above are met, then the value returned by foldr satisfies the relation with the input list ys. This scheme is not novel in itself [3] — what is new is the encoding, via uninterpreted predicate symbols, in an SMT-decidable refinement type system.

Using Folds. Finally, we can use the expressive type for the above foldr to verify various inductive properties of client functions:

```
length :: zs:[a] -> {v: Int | v = len(zs)}
length = foldr (\_ n -> n + 1) 0
append :: l:[a] -> m:[a] -> {v:[a] | len(v)=len(l)+len(m)}
append ys zs = foldr (\_ -> (:)) zs ys
```

The verification proceeds by just (automatically) instantiating the refinement parameter p of foldr with the concrete refinements, via Liquid typing:

 $\{ xs v \rightarrow v = len(xs) \} -- for length \\ \{ xs v \rightarrow len(v) = len(xs) + len(zs) \} -- for append$

Fig. 1. Syntax of Expressions, Refinements, Types and Schemas

3 Syntax and Semantics

Next, we present a core calculus λ_P that formalizes the notion of abstract refinements. We start with the syntax (§ 3.1), present the typing rules (§ 3.2), show soundness via a reduction to contract calculi [17, 1] (§ 3.3), and inference via Liquid types (§ 3.4).

3.1 Syntax

Figure 1 summarizes the syntax of our core calculus λ_P which is a polymorphic λ calculus extended with abstract refinements. We write b, $\{v : b \mid e\}$ and $b\langle p \rangle$ to abbreviate $\{v : b\langle true \rangle \mid true\}$, $\{v : b\langle true \rangle \mid e\}$, and $\{v : b\langle p \rangle \mid true\}$ respectively. We say a type or schema is *non-refined* if all the refinements in it are *true*. We write \overline{z} to abbreviate a sequence $z_1 \dots z_n$.

Expressions. λ_P expressions include the standard variables x, primitive constants c, λ -abstraction $\lambda x : \tau.e$, application e, type abstraction $A\alpha.e$, and type application e $[\tau]$. The parameter τ in the type application is a *refinement type*, as described shortly. The two new additions to λ_P are the refinement abstraction $A\pi : \tau.e$, which introduces a refinement variable π (together with its type τ), which can appear in refinements inside e, and the corresponding refinement application e [e].

Refinements. A *concrete refinement* e is a boolean valued expression e drawn from a strict subset of the language of expressions which includes only terms that (a) neither diverge nor crash, and (b) can be embedded into an SMT decidable refinement logic including the theory of linear arithmetic and uninterpreted functions. An *abstract refinement* p is a conjunction of refinement variable applications of the form $\pi \bar{e}$.

Types and Schemas. The basic types of λ_P include the base types int and bool and type variables α . An *abstract refinement type* τ is either a basic type refined with an abstract and concrete refinements, $\{v : b\langle p \rangle \mid e\}$, or a dependent function type where the parameter x can appear in the refinements of the output type. We include refinements for functions, as refined type variables can be replaced by function types. However, type-checking ensures these refinements are trivially true. Finally, types can be quantified over refinement variables and type variables to yield abstract refinement schemas.

Well-Formedness

Subtyping

$$\frac{\operatorname{SMT-Valid}(\llbracket\Gamma\rrbracket \land \llbracketp_1 v \rrbracket \land \llbrackete_1\rrbracket \Rightarrow \llbracketp_2 v \rrbracket \land \llbrackete_2\rrbracket)}{\Gamma \vdash \{v : b\langle p_1 \rangle \mid e_1\} \preceq \{v : b\langle p_2 \rangle \mid e_2\}} \preceq \operatorname{-BASE}$$

$$\frac{\Gamma \vdash \tau_2 \preceq \tau_1 \quad \Gamma, x_2 : \tau_2 \vdash \tau'_1[x_2/x_1] \preceq \tau'_2}{\Gamma \vdash \{v : (x_1 : \tau_1) \to \tau'_1 \mid e_1\} \preceq \{v : (x_2 : \tau_2) \to \tau'_2 \mid true\}} \preceq \operatorname{-Fun}$$

$$\frac{\Gamma, \pi : \tau \vdash \sigma_1 \preceq \sigma_2}{\Gamma \vdash \forall \pi : \tau.\sigma_1 \preceq \forall \pi : \tau.\sigma_2} \preceq \operatorname{-RVaR} \quad \frac{\Gamma \vdash \sigma_1 \preceq \sigma_2}{\Gamma \vdash \forall \alpha.\sigma_1 \preceq \forall \alpha.\sigma_2} \preceq \operatorname{-Poly}$$

Type Checking

$$\begin{array}{c|c} \hline \Gamma \vdash e: \sigma_{2} & \Gamma \vdash \sigma_{2} \preceq \sigma_{1} & \Gamma \vdash \sigma_{1} \\ \hline \Gamma \vdash e: \sigma_{1} & T-SUB & \hline \Gamma \vdash c: tc\left(c\right) & T-CONST \\ \hline \hline \mu \vdash e: \sigma_{1} & T-VAR-BASE & \hline \Gamma \vdash c: tc\left(c\right) & T-VAR \\ \hline \hline \Gamma \vdash x: \{v: b\langle p \rangle \mid e \land v = x\} & T-VAR-BASE & \hline \Gamma \vdash x: \tau & T-VAR \\ \hline \hline \Gamma \vdash x: \{v: b\langle p \rangle \mid e \land v = x\} & T-VAR-BASE & \hline \Gamma \vdash x: \tau & T-VAR \\ \hline \hline \Gamma \vdash \lambda x: \tau_{x} \vdash e: \tau & \Gamma \vdash \tau_{x} & T-FUN & \hline \Gamma \vdash e_{1}: (x: \tau_{x}) \rightarrow \tau & \Gamma \vdash e_{2}: \tau_{x} & T-APP \\ \hline \hline \hline \Gamma \vdash \lambda x: \tau_{x}.e: (x: \tau_{x}) \rightarrow \tau & T-GEN & \hline \Gamma \vdash e: \forall \alpha.\sigma & \Gamma \vdash \tau & T-INST \\ \hline \hline \hline \Gamma \vdash A\pi: \tau.e: \forall \pi: \tau.\sigma & T-PGEN & \hline \Gamma \vdash e[\lambda \overline{x}: \tau_{x}.e']: \sigma[\pi \vdash \lambda \overline{x}: \tau_{x}.e'] & T-PINST \end{array}$$

Fig. 2. Static Semantics: Well-formedness, Subtyping and Type Checking

3.2 Static Semantics

Next, we describe the static semantics of λ_P by describing the typing judgments and derivation rules. Most of the rules are standard [21, 23, 17, 2]; we discuss only those pertaining to abstract refinements.

 $\Gamma\vdash\sigma$

 $\Gamma \vdash \sigma_1 \preceq \sigma_2$

 $\varGamma \vdash e : \sigma$

Judgments. A type environment Γ is a sequence of type bindings $x : \sigma$. We use environments to define three kinds of typing judgments:

- Wellformedness judgments ($\Gamma \vdash \sigma$) state that a type schema σ is well-formed under environment Γ , that is, the refinements in σ are boolean expressions in the environment Γ .
- Subtyping judgments ($\Gamma \vdash \sigma_1 \preceq \sigma_2$) state that the type schema σ_1 is a subtype of the type schema σ_2 under environment Γ , that is, when the free variables of σ_1 and σ_2 are bound to values described by Γ , the set of values described by σ_1 is contained in the set of values described by σ_2 .
- Typing judgments ($\Gamma \vdash e : \sigma$) state that the expression e has the type schema σ under environment Γ , that is, when the free variables in e are bound to values described by Γ , the expression e will evaluate to a value described by σ .

Wellformedness Rules. The wellformedness rules check that the concrete and abstract refinements are indeed bool-valued expressions in the appropriate environment. The key rule is WF-BASE, which checks, as usual, that the (concrete) refinement e is boolean, and additionally, that the abstract refinement p applied to the value v is also boolean. This latter fact is established by WF-RAPP which checks that each refinement variable application $\pi \bar{e} v$ is also of type bool in the given environment.

Subtyping Rules. The subtyping rules stipulate when the set of values described by schema σ_1 is subsumed by the values described by σ_2 . The rules are standard except for \leq -VAR, which encodes the base types' abstract refinements p_1 and p_2 with conjunctions of *uninterpreted predicates* $[p_1 v]$ and $[p_2 v]$ in the refinement logic as follows:

$$\llbracket true \ v \rrbracket \doteq true$$
$$\llbracket (p \land \pi \ \overline{e}) \ v \rrbracket \doteq \llbracket p \ v \rrbracket \land \pi(\llbracket e_1 \rrbracket, \dots, \llbracket e_n \rrbracket, v)$$

where $\pi(\overline{e})$ is a term in the refinement logic corresponding to the application of the uninterpreted predicate symbol π to the arguments \overline{e} .

Type Checking Rules. The type checking rules are standard except for T-PGEN and T-PINST, which pertain to abstraction and instantiation of abstract refinements. The rule T-PGEN is the same as T-FUN: we simply check the body e in the environment extended with a binding for the refinement variable π . The rule T-PINST checks that the concrete refinement is of the appropriate (unrefined) type τ , and then replaces all (abstract) applications of π inside σ with the appropriate (concrete) refinement e' with the parameters \overline{x} replaced with arguments at that application. Formally, this is represented as $\sigma[\pi \triangleright \lambda \overline{x} : \overline{\tau}.e']$ which is σ with each base type transformed as

$$\{v: b\langle p \rangle \mid e\} [\pi \rhd z] \doteq \{v: b\langle p'' \rangle \mid e \land e''\}$$

where $(p'', e'') \doteq \mathsf{Apply}(p, \pi, z, true, true)$

Apply replaces each application of π in p with the corresponding conjunct in e'', as

In other words, the instantiation can be viewed as two symbolic reduction steps: first replacing the refinement variable with the concrete refinement, and then "beta-reducing" concrete refinement with the refinement variable's arguments. For example,

 $\{v: \operatorname{int} \langle \pi \ y \rangle \mid v > 10\} [\pi \rhd \lambda x_1 : \tau_1 . \lambda x_2 : \tau_2 . x_1 < x_2] \doteq \{v: \operatorname{int} \mid v > 10 \land y < v\}$

3.3 Soundness

As hinted by the discussion about refinement variable instantiation, we can intuitively think of abstract refinement variables as *ghost* program variables whose values are boolean-valued functions. Hence, abstract refinements are a special case of higher-order contracts, that can be statically verified using uninterpreted functions. (Since we focus on static checking, we don't care about the issue of blame.) We formalize this notion by translating λ_P programs into the contract calculus F_H of [1] and use this translation to define the dynamic semantics and establish soundness.

Translation. We translate λ_P schemes σ to F_H schemes $\langle \sigma \rangle$ as by translating abstract refinements into contracts, and refinement abstraction into function types:

$\langle true v \rangle \doteq true$	$\langle\!\langle \forall \pi : \tau . \sigma \rangle\!\rangle \ \doteq \ (\pi : \langle\!\langle \tau \rangle\!\rangle) \to \langle\!\langle \sigma \rangle\!\rangle$
$\langle\!\langle (p \land \pi \ \overline{e}) \ v \rangle\!\rangle \ \doteq \ \langle\!\langle p \ v \rangle\!\rangle \land \pi \ \overline{e} \ v$	$\langle\!\langle \forall \alpha.\sigma \rangle\!\rangle \doteq \forall \alpha. \langle\!\langle \sigma \rangle\!\rangle$
$\langle\!\langle \{v: b \langle p \rangle \mid e\} \rangle\!\rangle \doteq \{v: b \mid e \land \langle\!\langle p v \rangle\!\rangle\}$	$\langle\!\langle (x:\tau_1) \to \tau_2 \rangle\!\rangle \doteq (x:\langle\!\langle \tau_1 \rangle\!\rangle) \to \langle\!\langle \tau_2 \rangle\!\rangle$

Similarly, we translate λ_P terms e to F_H terms $\langle e \rangle$ by converting refinement abstraction and application to λ -abstraction and application

$\langle x \rangle \doteq x$	$\langle c \rangle \doteq c$
$\langle\!\langle \lambda x : \tau.e \rangle\!\rangle \ \doteq \ \lambda x : \langle\!\langle \tau \rangle\!\rangle.\langle\!\langle e \rangle\!\rangle$	$\langle e_1 \ e_2 \rangle \doteq \langle e_1 \rangle \langle e_2 \rangle$
$\langle\!\langle \Lambda \alpha. e \rangle\!\rangle \doteq \Lambda \alpha. \langle\!\langle e \rangle\!\rangle$	$\langle e[\tau] \rangle \doteq \langle e \rangle \langle \tau \rangle$
$\langle\!\langle A\pi:\tau.e\rangle\!\rangle \ \doteq \ \lambda\pi:\langle\!\langle \tau\rangle.\langle\!\langle e\rangle\!\rangle$	$\langle e_1 [e_2] \rangle \doteq \langle e_1 \rangle \langle e_2 \rangle$

Translation Properties. We can show by induction on the derivations that the type derivation rules of λ_P conservatively approximate those of F_H . Formally,

 $\begin{array}{l} - \mbox{ If } \Gamma \vdash \tau \mbox{ then } \langle\!\langle \Gamma \rangle\!\rangle \vdash_H \langle\!\langle \tau \rangle\!\rangle, \\ - \mbox{ If } \Gamma \vdash \tau_1 \preceq \tau_2 \mbox{ then } \langle\!\langle \Gamma \rangle\!\rangle \vdash_H \langle\!\langle \tau_1 \rangle\!\rangle <: \langle\!\langle \tau_2 \rangle\!\rangle, \\ - \mbox{ If } \Gamma \vdash e : \tau \mbox{ then } \langle\!\langle \Gamma \rangle\!\rangle \vdash_H \langle\!\langle e \rangle\!\rangle : \langle\!\langle \tau \rangle\!\rangle. \end{array}$

Soundness. Thus rather than re-prove preservation and progress for λ_P , we simply use the fact that the type derivations are conservative to derive the following preservation and progress corollaries from [1]:

- **Preservation:** If $\emptyset \vdash e : \tau$ and $\langle\!\langle e \rangle\!\rangle \longrightarrow e'$ then $\emptyset \vdash_H e' : \langle\!\langle \tau \rangle\!\rangle$
- **Progress:** If $\emptyset \vdash e : \tau$, then either $\langle e \rangle \longrightarrow e'$ or $\langle e \rangle$ is a value.

Note that, in a contract calculus like F_H , subsumption is encoded as a *upcast*. However, if subtyping relation can be statically guaranteed (as is done by our conservative SMT based subtyping) then the upcast is equivalent to the identity function and can be eliminated. Hence, F_H terms $\langle e \rangle$ translated from well-typed λ_P terms *e* have no casts.

3.4 Refinement Inference

Our design of abstract refinements makes it particularly easy to perform type inference via Liquid typing, which is crucial for making the system usable by eliminating the tedium of instantiating refinement parameters all over the code. (With value-dependent refinements, one cannot simply use, say, unification to determine the appropriate instantations, as is done for classical type systems.) We briefly recall how Liquid types work, and sketch how they are extended to infer refinement instantiations.

Liquid Types. The Liquid Types method infers refinements in three steps. First, we create refinement templates for the unknown, to-be-inferred refinement types. The shape of the template is determined by the underlying (non-refined) type it corresponds to, which can be determined from the language's underlying (non-refined) type system. The template is just the shape refined with fresh refinement variables κ denoting the unknown refinements at each type position. For example, from a type $(x : int) \rightarrow int$ we create the template $(x : \{v : int | \kappa_x\}) \rightarrow \{v : int | \kappa\}$. Second, we perform type checking using the templates (in place of the unknown types.) Each wellformedness check becomes a wellformedness constraint over the templates, and hence over the individual κ , constraining which variables can appear in κ . Each subsumption check becomes a subtyping constraint between the templates, which can be further simplified, via syntactic subtyping rules, to a logical implication query between the variables κ . Third, we solve the resulting system of logical implication constraints (which can be cyclic) via abstract interpretation — in particular, monomial predicate abstraction over a set of logical qualifiers [9, 23]. The solution is a map from κ to conjunctions of qualifiers, which, when plugged back into the templates, yields the inferred refinement types.

Inferring Refinement Instantiations. The key to making abstract refinements practical is a means of synthesizing the appropriate arguments e' for each refinement application e[e']. Note that for such applications, we can, from e, determine the non-refined type of e', which is of the form $\tau_1 \rightarrow \ldots \rightarrow \tau_n \rightarrow bool$. Thus, e' has the template $\lambda x_1 : \tau_1 \dots \lambda x_n : \tau_n . \kappa$ where κ is a fresh, unknown refinement variable that must be solved to a boolean valued expression over x_1, \dots, x_n . Thus, we generate a *well-formedness* constraint $x_1 : \tau_1, \dots, x_n : \tau_n \vdash \kappa$ and carry out typechecking with template, which, as before, yields implication constraints over κ , which can, as before, be solved via predicate abstraction. Finally, in each refinement template, we replace each κ with its solution e_{κ} to get the inferred refinement instantiations.

4 Evaluation

In this section, we empirically evaluate the expressiveness and usability of abstract refinement types by exploring the process of typechecking a set of challenging benchmark programs using a prototype type checker for Haskell. (We defer the task of extending the metatheory to a call-by-name calculus to future work.)

HSOLVE. We have implemented abstract refinement in HSOLVE, a refinement type checker for Haskell. HSOLVE verifies Haskell source one module (.hs file) at a time. It takes as input:

Program	LOC	Specs	Annot	Time (s)
Micro	32	19	4	2
Vector	56	56	2	14
ListSort	29	4	1	3
Data.List.sort	71	3	1	8
Data.Set.Splay	136	15	11	15
Data.Map.Base	1399	119	31	235
Total	1723	216	50	277

Table 1. (LOC) is the number of non-comment Haskell source code lines as reported by *sloc-count*, (**Specs**) is the number of lines of type specifications, (**Annot**) is the number of lines of other annotations, including refined datatype definitions, type aliases and measures, required for verification, (**Time**) is the time in seconds taken for verification.

- A *target* Haskell source file, with the desired refinement types specified as a special form of comment annotation,
- An (optional) set of type specifications for imported definitions; these can either be
 put directly in the source for the corresponding modules, if available, or in special
 .spec files otherwise. For imported functions for which no signature is given,
 HSOLVE conservatively uses the non-refined Haskell type.
- An (optional) set of logical qualifiers, which are predicate templates from which refinements are automatically synthesized [23]. Formally, a logical qualifier is a predicate whose variables range over the program variables, the special value variable ν , and *wildcards* \star , which HSOLVE instantiates with the names of program variables. Aside from the qualifiers given by the user, HSOLVE also uses qualifiers mined from the refinement type annotations present in the program.

After analyzing the program, HSOLVE returns as output:

- Either SAFE, indicating that all the specifications indeed verify, or UNSAFE, indicating there are refinement type errors, together with the positions in the source code where type checking fails (*e.g.*, functions that do not satisfy their signatures, or callsites where the inputs don't conform to the specifications).
- An HTML file containing the program source code annotated with inferred refinement types for all sub-expressions in the program. The inferred refinement type for each program expression is the strongest possible type over the given set of logical qualifiers. When a type error is reported, the programmer can use the inferred types to determine why their program does not typecheck: they can examine what properties HSOLVE *can* deduce about various program expressions and add more qualifiers or alter the program as necessary so that it typechecks.

Implementation. HSOLVE verifies the contents of a single file (module) at a time as follows. First, the Haskell source is fed into GHC, which desugars the program to GHC's "core" intermediate representation [26]. Second, the desugared program, the type signatures for the module functions (which are to be verified) and the type signatures for externally imported functions (which are assumed to hold) are sent to the constraint generator, which traverses the core bindings in a syntax-directed manner to

generate subtyping constraints. The resulting constraints are simplified via our subtyping rules (§ 3) into simple logical implication constraints. Finally, the implication constraints, together with the logical qualifiers provided by the user and harvested from the type signatures, are sent into an SMT- and abstract interpretation-based fixpoint computation procedure that determines if the constraints are satisfiable [13,9]. If so, the program is reported to be *safe*. Otherwise, each unsatisfiable constraint is mapped back to the corresponding program source location that generated it and a potential error is reported at that line in the program.

Benchmarks. We have evaluated HSOLVE over the following list of benchmarks which, in total, represent the different kinds of reasoning described in § 2. While we can prove, and previously have proved [16], many so-called "functional correctness" properties of these data structures using refinement types, in this work we focus on the key invariants which are captured by abstract refinements.

- Micro, which includes several functions demonstrating parametric reasoning with base values, type classes, and higher-order loop invariants for traversals and folds, as described in § 2.1 and § 2.4;
- Vector, which includes the domain- and range-generic Vec functions and several "clients" that use the generic Vec to implement incremental initialization, nullterminated strings, and memoization, as described in § 2.2;
- ListSort, which includes various textbook sorting algorithms including insert-, merge- and quick-sort. We verify that the functions actually produce sorted lists, *i.e.*, are of type IncrList a, as described in § 2.3;
- Data.List.sort, which includes three non-standard, optimized list sorting algorithms, as found in the base package. These employ lists that are increasing and decreasing, as well as lists of (sorted) lists, but we can verify that they also finally produce values of type IncrList a;
- Data.Set.Splay, which is a purely functional, top-down splay set library from the llrbtree package. We verify that all the interface functions take and return binary search trees;
- Data.Map.Base, which is the widely-used implementation of functional maps from the containers package. We verify that all the interface functions preserve the crucial binary search ordering property and various related invariants.

Table 1 quantitatively summarizes the results of our evaluation. We now give a qualitative account of our experience using HSOLVE by discussing what the specifications and other annotations look like.

Specifications are usually simple. In our experience, abstract refinements greatly simplify writing specifications for the *majority* of interface or public functions. For example, for Data.Map.Base, we defined the refined version of the Tree ADT (actually called Map in the source, we reuse the type from § 2.3 for brevity), and then instantiated it with the concrete refinements for binary-search ordering with the alias BST k v as described in § 2.3. Most refined specifications were just the Haskell types with the Tree type constructor replaced with the alias BST. For example, the type of fromList is refined from (**Ord** k) => [(k, a)] -> Tree k a to (**Ord** k) => [(k, a)] -> BST k a. Furthermore, intra-module Liquid type inference permits the automatic synthesis of necessary stronger types for private functions.

Auxiliary Invariants are sometimes Difficult. However, there are often rather thorny *internal* functions with tricky invariants, whose specification can take a bit of work. For example, the function trim in Data.Map.Base has the following behavior (copied verbatim from the documentation): "trim blo bhi t trims away all subtrees that surely contain no values between the range blo to bhi. The returned tree is either empty or the key of the root is between blo and bhi." Furthermore blo (resp. bhi) are specified as option (*i.e.*, Maybe) values with Nothing denoting $-\infty$ (resp. $+\infty$). Fortunately, refinements suffice to encode such properties. First, we define measures

```
measure isJust :: Maybe a -> Bool
isJust (Just x) = true
isJust (Nothing) = false
measure fromJust :: Maybe a -> a
fromJustS (Just x) = x
measure isBin :: Tree k v -> Bool
isBin (Bin _ _ _) = true
isBin (Tip) = false
measure key :: Tree k v -> k
key (Bin k _ _) = k
```

which respectively embed the Maybe and Tree root value into the refinement logic, after which we can type the trim function as

where bound is simply a refinement alias

```
refinement bound(lo, v, hi)
= isBin(v) => isJust(lo) => fromJust(lo) < key(v)
&& isBin(v) => isJust(hi) => fromJust(hi) > key(v)
```

That is, the output refinement states that the root is appropriately lower- and upperbounded if the relevant terms are defined. Thus, refinement types allow one to formalize the crucial behavior as machine-checkable documentation.

Code Modifications. On a few occasions we also have to change the code slightly, typically to make explicit values on which various invariants depend. Often, this is for a trivial reason; a simple re-ordering of binders so that refinements for *later* binders can depend on earlier ones. Sometimes we need to introduce "ghost" values so we can write the specifications (*e.g.*, the foldr in § 2.4). Another example is illustrated by the use of list append in quickSort. Here, the append only produces a sorted list if the two input lists are sorted and such that each element in the first is less than each element

in the second. We address this with a special append parameterized on pivot

```
append :: pivot:a
    -> IncrList {v:a | v < pivot}
    -> IncrList {v:a | v > pivot}
    -> IncrList a
    append pivot [] ys = pivot : ys
    append pivot (x:xs) ys = x : append pivot xs ys
```

5 Related Work

The notion of type refinements was introduced by Freeman and Pfenning [11], with refinements limited to restrictions on the structure of algebraic datatypes, for which inference is decidable. Our present notion of refinement types has its roots in the *in*dexed types of Xi and Pfenning [27], wherein data types' ranges are restricted by indices, analogous to our refinement predicates, drawn from a decidable domain; in the example case explored by Xi and Pfenning, types were indexed by terms from Presburger arithmetic. Since then, several approaches to developing richer refinement type systems and accompanying methods for type checking have been developed. Knowles and Flanagan [17] allow refinement predicates to be arbitrary terms of the language being typechecked and present a technique for deciding some typing obligations statically and deferring others to runtime. Findler and Felleisen's [8] higher-order contracts, which extend Eiffel's [18] first-order contracts - ordinary program predicates acting as dynamic pre- and post-conditions — to the setting of higher-order programs, eschew any form of static checking, and can be seen as a dynamically-checked refinement type system. Bengtson et al. [2] present a refinement type system in which type refinements are drawn from a decidable logic, making static type checking tractable. Greenberg et al. [1] gives a rigorous treatment of the metatheoretic properties of such a refinement type system.

Refinement types have been applied to the verification of a variety of program properties [27, 7, 2, 10]. In the most closely related work to our own, Kawaguchi et al. [16] introduce *recursive* and *polymorphic* refinements for data structure properties. The present work unifies and generalizes these two somewhat ad-hoc notions into a single, strictly and significantly more expressive mechanism of abstract refinements.

A number of higher-order logics and corresponding verification tools have been developed for reasoning about programs. Example of systems of this type include NuPRL [4], Coq [3], F^* [25] and Agda [20] which support the development and verification of higher-order, pure functional programs. While these systems are highly expressive, their expressiveness comes at the cost of making logical validity checking undecidable. To help automate validity checking, both built-in and user-provided tactics are used to attempt to discharge proof obligations; however, the user is ultimately responsible for manually proving any obligations which the tactics are unable to discharge.

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