Julia's efficient algorithm for subtyping unions and covariant tuples (Pearl)

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- Abstract

The Julia programming language supports multiple dispatch and provides a rich type annotation language to specify method applicability. When multiple methods are applicable for a given call, Julia relies on subtyping between method signatures to pick the correct method to invoke. Julia's subtyping algorithm is surprisingly complex, and determining whether it is correct remains an open question. In this paper, we focus on one piece of this problem: the interaction between union types and covariant tuples. Previous work normalized unions inside tuples to disjunctive normal form. However, this strategy has two drawbacks: complex type signatures induce space explosion, and interference between normalization and other features of Julia's type system. In this paper, we describe the algorithm that Julia uses to compute subtyping between tuples and unions—an algorithm that is immune to space explosion and plays well with other features of the language. We prove this algorithm correct and complete against a semantic-subtyping denotational model in Coq.

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1 Introduction

Union types, originally introduced by Barbanera and Dezani-Ciancaglini [2], are being adopted in mainstream languages. In some cases, such as Julia [5] or TypeScript [11], they are exposed at the source level. In others, such as Hack [8], they are only used internally as part of type inference. As a result, subtyping algorithms between union types are of increasing practical import. The standard subtyping algorithm for this combination of features has, for some time, been exponential in both time and space. An alternative algorithm, linear in space but still exponential in time, has been tribal knowledge in the subtyping community [15]. In this paper, we describe and prove correct an implementation of that algorithm.

We observed the algorithm in our prior work formalizing the Julia subtyping relation [17]. There, we described Julia's subtyping relation as it arose from its decision procedure but were unable to prove it correct. Indeed, we found bugs in the Julia implementation and identified unresolved correctness issues. Contemporary work addresses some correctness concerns [3] but leaves algorithmic correctness open.

Julia's subtyping algorithm [4] is used for method dispatch. While Julia is dynamically typed, method arguments can have type annotations. These annotations allow one method to be implemented by multiple functions. At run time, Julia searches for the most specific applicable function for a given invocation. Consider these declarations of multiplication:

```
*(x::Number, r::Range) = range(x*first(r),...)

*(x::Number, y::Number) = *(promote(x,y)...)

*(x::T, y::T) where T <: Union{Signed, Unsigned} = mul_int(x,y)
```

The first two methods implement, respectively, multiplication of a range by a number and generic numeric multiplication. The third method invokes native multiplication when both arguments are either signed or unsigned integers (but not a mix of the two). Julia uses subtyping to decide which of the methods to call at any specific site. The call 1*(1:4) dispatches to the first, 1*1.1 the second, and 1*1 the third.

Julia offers programmers a rich type language to express complex relationships in type signatures. The type language includes nominal primitive types, union types, existential types, covariant tuples, invariant parametric datatypes, and singletons. Intuitively, subtyping between types is based on semantic subtyping: the subtyping relation between types holds when the sets of values they denote are subsets of one another [5]. We write the set of values represented by a type t as [t]. Under semantic subtyping, the types t_1 and t_2 are subtypes iff $[t_1] \subseteq [t_2]$. From this, we derive a forall-exists intuition for subtyping: for every value denoted on the left-hand side, there must exist some value on the right-hand side to match it, thereby establishing the subset relation. This simple intuition is, however, complicated to check algorithmically.

In this paper, we focus on the interaction of two features: covariant tuples and union types. These two kinds of type are important to Julia's semantics. Julia does not record return types, so a function's signature consists solely of the tuple of its argument types. These tuples are covariant, as a function with more specific arguments is preferred to a more generic one. Union types are widely used as shorthand to avoid writing multiple functions with the same body. As a consequence, Julia library developers write many functions with union typed arguments, functions whose relative specificity must be decided using subtyping. To prove the correctness of the subtyping algorithm, we first examine typical approaches in the presence of union types. Based on Vouillon [16], the following is a typical deductive system for subtyping union types:

While this rule system might seem to make intuitive sense, it does not match the semantic intuition for subtyping. For instance, consider the following judgment:

```
\texttt{Tuple}\{\texttt{Union}\{t',t''\},t\} <: \texttt{Union}\{\texttt{Tuple}\{t',t\},\texttt{Tuple}\{t'',t\}\}
```

Using semantic subtyping, the judgment should hold. The set of values denoted by a union $[\![Union\{t_1,t_2\}]\!]$ is just the union of the set of values denoted by each of its members $[\![t_1]\!] \cup [\![t_2]\!]$. A tuple $[\![t_1,t_2]\!]$'s denotation is the set of tuples of the respective values $[\![t_1]\!] \cup [\![t_1]\!] \wedge v_1 \in [\![t_1]\!] \wedge v_2 \in [\![t_2]\!]$. Therefore, the left-hand side denotes the values $[\![t_1]\!] \cup [\![t_1]\!] \wedge v_2 \in [\![t_2]\!]$, while the right-hand side denotes $[\![t_1]\!] \cup [\![t_1]\!] \cap [\![t_1]$

Most prior work addresses this problem by normalization [2, 14, 1], rewriting all types into their disjunctive normal form, as unions of union-free types, *before* building the derivation.

Now all choices are made at the top level, avoiding the structural entanglements that cause difficulties. The correctness of this rewriting step comes from the semantic denotational model, and the resulting subtyping algorithm can be proved both correct and complete. Other proposals, such as Vouillon [16] and Dunfield [7], do not handle distributivity. Normalization is used by Frisch et al.'s [9], by Pearce's flow-typing algorithm [13], and by Muehlboeck and Tate in their general framework for union and intersection types [12]. Few alternatives have been proposed, with one example being Damm's reduction of subtyping to regular tree expression inclusion [6].

However, a normalization-based algorithm has two major drawbacks: it is not space efficient, and other features of Julia render it incorrect. The first drawback is caused because normalization can create exponentially large types. Real-world Julia code [17] has types like the following whose normal form has 32,768 constituent union-free types:

The second drawback arises because of type-invariant constructors. For example, Array{Int} is an array of integers, and is not a subtype of Array{Any}. In conjunction with type variables, this makes normalization ineffective. Consider Array{Union{t', t''}}, the set of arrays whose elements are either t' or t''. It wrong to rewrite it as Union{Array{t'}, Array{t''}}, as this denotes the set of arrays whose elements are either all t' or t''. A weaker disjunctive normal form, only lifting union types inside each invariant constructor, is a partial solution. However, this reveals a deeper problem caused by existential types. Consider the judgment:

```
Array{Union{Tuple}{t}, Tuple}{t'}} <: \exists T.Array{Tuple}{T}}
```

It holds if the existential variable T is instantiated with $\mathtt{Union}\{t,t'\}$. If types are in invariant-constructor weak normal form, an algorithm would strip off the array type constructors and proceed. However, since type constructors are invariant, the algorithm must test that both $\mathtt{Union}\{\mathtt{Tuple}\{t\},\mathtt{Tuple}\{t'\}\}$ <: $\mathtt{Tuple}\{T\}$ and $\mathtt{Tuple}\{T\}$ <: $\mathtt{Union}\{\mathtt{Tuple}\{t\},\mathtt{Tuple}\{t'\}\}$ hold. The first of these can be concluded without issue, producing the constraint $\mathtt{Union}\{t,t'\}$ <: T. However, this constraint on T is retained for checking the reverse direction, which is where problems arise. When checks the reverse direction, the algorithm has to prove that $\mathtt{Tuple}\{T\}$ <: $\mathtt{Union}\{\mathtt{Tuple}\{t'\}\}$, and in turn either T <: t or T <: t'. All of these are unprovable under the assumption that $\mathtt{Union}\{t,t'\}$ <: T. The key to deriving a successful judgment for this relation is to rewrite the right-to-left check into $\mathtt{Tuple}\{T\}$ <: $\mathtt{Tuple}\{\mathtt{Union}\{t,t'\}\}$, which is provable. This anti-normalization rewriting must be performed on sub-judgments of the derivation; to the best of our knowledge it is not part of any subtyping algorithm based on ahead-of-time disjunctive normalization.

Julia's subtyping algorithm avoids these problems, but it is difficult to determine how: the complete subtyping algorithm is implemented in close to two thousand lines of highly optimized C code. In this paper, we describe and prove correct only one part of that algorithm: the technique used to avoid space explosion while dealing with union types and covariant tuples. This is done by defining an iteration strategy over type terms, keeping a string of bits as its state. The space requirement of the algorithm is bounded by the number of unions in the type terms being checked.

We use a minimal type language with union, tuples, and primitive types to avoid being drawn into the vast complexity of Julia's type language. This tiny language is expressive

enough to highlight the decision strategy and illustrate the structure of the algorithm. Empirical evidence from Julia's implementation suggests that this technique extends to invariant constructors and existential types [17], among others. We expect that the algorithm 142 we describe can be leveraged in other modern language designs.

Our mechanized proof is available at: benchung.github.io/subtype-artifact.

A space-efficient subtyping algorithm

Formally, our core type language consists of binary unions, binary tuples, and primitive types ranged over by $p_1 \dots p_n$, as shown below:

```
149
                  Prim of int
                                | Tuple of typ * typ | Union of typ * typ
    type typ =
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```

We define subtyping for primitives as the identity, so $p_i \ll p_i$.

2.1 Normalization

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To explain the operation of the space-efficient algorithm, we first describe how normalization can be used as part of subtyping. Normalization rewrites types to move all internal unions to the top level. The resultant term consists of a union of union-free terms. Consider the following relation:

```
{\tt Union} \{ {\tt Tuple} \{p_1, p_2\}, {\tt Tuple} \{p_2, p_3\} \} \quad <: \quad {\tt Tuple} \{ {\tt Union} \{p_2, p_1\}, {\tt Union} \{p_3, p_2\} \}.
```

The term on the left is in normal form, but the right term needs to be rewritten as follows: 159

```
Union{Tuple{p_2, p_3}, Union{Tuple{p_2, p_2}, Union{Tuple{p_1, p_3}, Tuple{p_1, p_2}}}}
```

The top level unions can then be viewed as sets of union-free-types equivalent to each side,

```
\ell_1 = \{ \text{Tuple}\{p_1, p_2\}, \text{Tuple}\{p_2, p_3\} \}
162
       and
163
             \ell_2 = \{ \text{Tuple}\{p_2, p_3\}, \text{Tuple}\{p_2, p_2\}, \text{Tuple}\{p_1, p_3\}, \text{Tuple}\{p_1, p_2\} \}.
```

Determining whether $\ell_1 <: \ell_2$ is equivalent to checking that for each tuple component t_1 in ℓ_1 , there should be an element t_2 in ℓ_2 such that $t_1 <: t_2$. Checking this final relation is straightforward, as neither t_1 nor t_2 may contain unions. Intuitively, this mirrors the rules ([ALLEXIST], [EXISTL/R], [TUPLE]).

A possible implementation of normalization-based subtyping can be written compactly, as shown in the code below. The subtype function takes two types and returns true if they are related by subtyping. It delegates its work to allexist to check that all normalized terms in its first argument have a supertype, and to exist to check that there is at least one supertype in the second argument. The norm function takes a type term and returns a list of union-free terms.

```
176
    let subtype(a:typ)(b:typ) = allexist (norm a) (norm b)
177
178
    let allexist(a:list typ)(b:list typ) =
179
      foldl (fun acc a' => acc && exist a' b) true a
180
181
```

However, as previously described, this expansion is space-inefficient. Julia's algorithm is more complicated, but avoids having to pre-compute the set of normalized types as norm does.

2.2 Iteration with choice strings

Given a type term such as the following,

 $\texttt{Tuple}\{\texttt{Union}\{\texttt{Union}\{p_2,p_3\},p_1\},\texttt{Union}\{p_3,p_2\}\}$

we want an algorithm that checks the following tuples,

 $Tuple\{p_2, p_3\}, Tuple\{p_2, p_2\}, Tuple\{p_1, p_3\}, Tuple\{p_1, p_2\}, Tuple\{p_3, p_3\}, Tuple\{p_3, p_2\}$

without having to compute and store all of them ahead-of-time. This algorithm should be able to generate each tuple on-demand while still being guaranteed to explore every tuple of the original type's normal form.

To illustrate the process that the algorithm uses to generate each tuple, consider the type term being subtyped. An alternative representation for the term is a tree, where each occurrence of a union node is a *choice point*. The following tree thus has three choice points, each represented as a ? symbol:

At each choice point we can go either left or right; making such a decision at each point leads to visiting one particular tuple.

Each tuple is uniquely determined by the original type term t and a choice string c. In the above example, the result of iteration through the normalized, union-free, type terms is defined by the strings LLL, LLR, LRL, LRR, RL, RR. The length of each string is bounded by the number of unions in a term.

The iteration sequence in the above example is thus $LL\underline{L} \to L\underline{L}R \to LR\underline{L} \to \underline{L}RR \to R\underline{L} \to RR$, where the underlined choice is next one to be toggled in that step. Stepping from a choice string c to the next string consists of splitting c in three, c' L c'', where c' can be empty and c'' is a possibly empty sequence of Rs. The next string is $c' R c_{pad}$, that is to say

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it retains the prefix c', toggles the L to an R, and is padded by a sequence of Ls. The leftover tail c'' is discarded. If there is no L in c, iteration terminates.

One step of iteration is performed by calling the next function with a type term and a choice string (encoded as a list of choices); next either returns the next string in the sequence or None. Internally, it calls step to toggle the last L and shorten the string (constructing c' R). Then it calls on pad to add the trailing sequence of Ls (constructing $c' R c_{pad}$).

```
type choice = L | R

type choice = L | R

let rec next(a:typ)(1:choice list) =
    match step 1 with
    | None -> None
    | Some(1') -> Some(fst (pad a 1'))
```

The step function delegates the job of flipping the last occurrence of L to toggle. For ease of programming, it reverses the string so that toggle can be a simple recursion without an accumulator. If the given string has no L, then toggle returns empty and step returns None.

```
238
    let step(1:choice list) =
239
       match rev (toggle (rev 1)) with
240
        [] -> None
241
       | hd::tl -> Some(hd::tl)
242
243
    let rec toggle = function
244
        [] -> []
245
        L::tl -> R::tl
246
        R::tl -> toggle tl
<del>24</del>7
```

The pad function takes a type term and a choice string to be padded. It returns a pair, whose first element is the padded string and second element is the string left over from the current type. Each union encountered by pad in its traversal of the type consumes a character from the input string. Unions explored after the exhaustion of the original choice string are treated as if there was an L remaining in the choice string. The first component of the returned value is the original choice string extended with an L for every union encountered after exhaustion of the original.

```
257
    let rec pad t l =
258
       match t,1 with
259
       | (Prim i,1) -> ([],1)
260
       | (Tuple(t,t'),1) ->
          let (h,tl) = pad t l in
          let (h',tl') = pad t' tl in (h @ h',tl')
         (Union(t,_),L::r) ->
          let (h,t1) = pad t r in (L::h,t1)
265
         (Union(_,t),R::r) ->
266
          let (h,t1) = pad t r in (R::h,t1)
267
       | (Union(t,_),[]) -> (L::(fst(pad t [])),[])
268
269
```

To obtain the initial choice string, the string composed solely of Ls, it suffices to call pad with the type term under consideration and an empty list. The first element of the returned tuple is the initial choice string. For convenience, we define the function initial for this.

```
273
274
275
276
let initial(t:typ) = fst (pad t [])
```

2.3 Subtyping with iteration

 Julia's subtyping algorithm visits union-free type terms using choice strings to iterate over types. The subtype function takes two type terms, a and b, and returns true if they are related by subtyping. It does so by iterating over all union-free type terms t_a in a, and checking that for each of them, there exists a union-free type term t_b in b such that $t_a <: t_b$.

```
let subtype(a:typ)(b:typ) = allexist a b (initial a)
```

The allexist function takes two type terms, a and b, and a choice string f, and returns true if a is a subtype of b for the iteration sequence starting at f. This is achieved by recursively testing that for each union-free type term in a (induced by a and the current value of f), there exists a union-free super-type in b.

Similarly, the exist function takes two type terms, a and b, and choice strings, f and e. It returns true if there exists in b, a union-free super-type of the type specified by f in a. This is done by recursively iterating through e. The determination if two terms are related is delegated to the sub function.

Finally, the sub function takes two type terms and choice strings and returns a value of type res. A res can be either NotSub, indicating that the types are not subtypes, or IsSub(_,_) when they are subtypes. If the two types are primitives, then they are only subtypes if they are equal. If the types are tuples, they are subtypes if each of their respective elements are subtypes. Note that the return type of sub, when successful, holds the unused choice strings for both type arguments. When encountering a union, sub follows the choice strings to decide which branch to take. Consider, for instance, the case when the first type term is Union(t1,t2) and the second is type t. If the first element of the choice string is an L, then t1 and t are checked, otherwise sub checks t2 and t.

ECOOP 2019

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```
325
    let rec sub t1 t2 f e =
      match t1,t2,f,e with
327
        (Prim i, Prim j,f,e) -> if i==j then IsSub(f,e) else NotSub
328
        (Tuple(a1,a2), Tuple(b1,b2),f,e)
329
          (match sub a1 b1 f e with
330
            IsSub(f', e') \rightarrow sub a2 b2 f'
331
           | NotSub -> NotSub)
332
        (Union(a,_),b,L::f,e) \rightarrow sub a b f e
333
        (Union(_,a),b,R::f,e) -> sub a b f e
334
        (a,Union(b,_),f,L::e) \rightarrow sub a b f e
335
        (a,Union(_,b),f,R::e) -> sub a b f e
336
```

2.4 Further optimization

This implementation represents choice strings as linked lists, but this design requires allocation and reversals when stepping. However, the implementation can be made more efficient by using a mutable bit vector instead of a linked list. Additionally, the maximum length of the bit vector is bounded by the number of unions in the type, so it need only be allocated once. Julia's implementation uses this efficient representation.

3 Correctness and completeness of subtyping

To prove the correctness of Julia's subtyping, we take the following general approach. We start by giving a denotational semantics for types from which we derive a definition of semantic subtyping. Then we easily prove that a normalization-based subtyping algorithm is correct and complete. This provides the general framework for which we prove two iterator-based algorithms correct. The first iterator-based algorithm explicitly includes the structure of the type in its state to guide iteration; the second is identical to that of the prior section.

The order in which choice strings iterate through a type term is determined by both the choice string and the type term being iterated over. Rather than directly working with choice strings as iterators over types, we start with a simpler structure, namely that of iterators over the trees induced by type terms. We prove correct and complete a subtyping algorithm that uses these simpler iterators. Finally, we establish a correspondence between tree iterators and choice string iterators. This concludes our proof of correctness and completeness, and details can be found in the Coq mechanization.

The denotational semantics we use for types is as follows:

We define subtyping as follows: if $[t] \subseteq [t']$, then t <: t'. This leads to the definition of subtyping in our restricted language.

```
▶ Definition 1. The subtyping relation t_1 <: t_2 \text{ holds iff } \forall t_1' \in \llbracket t_1 \rrbracket, \exists t_2' \in \llbracket t_2 \rrbracket, t_1' = t_2'.
```

The use of equality for relating types is a simplification afforded by the structure of primitives.

3.1 Subtyping with normalization

The correctness and completeness of the normalization-based subtyping algorithm requires proving that the norm function returns all union-free type terms.

Lemma 2 (NF Equivalence). $t' \in [t]$ iff $t' \in norm t$.

Theorem 3 states that the subtype relation of Section 2.1 abides by Definition 1 because it uses norm to compute the set of union-free type terms for both argument types, and directly checks subtyping.

Theorem 3 (NF Subtyping). For all a and b, subtype a b iff a <: b.

Therefore, normalization-based subtyping is correct against our definition.

3.2 Subtyping with tree iterators

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Reasoning about iterators that use choice strings, as described in Section 2.2, is tricky as it requires simultaneously reasoning about the structure of the type term and the validity of the choice string that represents the iterator's state. Instead, we propose to use an intermediate data structure, called a tree iterator, to guarantee consistency of iterator state with type structure

A tree iterator is a representation of the iteration state embedded in a type term. Thus a tree iterator yields a union-free tuple and can either step to a successor state or a final state. Recalling the graphical notation of Section 2.2, we can represent the state of iteration as a combination of type term and a choice or, equivalently, as a tree iterator.

Choice string: Tree iterator:

This structure-dependent construction makes tree iterators less efficient than choice strings. A tree iterator must have a node for each structural element of the type being iterated over, and is thus less space efficient than the simple choices-only strings. However, it is easier to prove subtyping correct for tree iterators first.

Tree iterators depend on the type term they iterate over. The possible states are IPrim at primitives, ITuple at tuples, and for unions either ILeft or IRight.

```
395
396    Inductive iter: Typ -> Set :=
397    | IPrim : forall i, iter (Prim i)
398    | ITuple : forall t1 t2, iter t1 -> iter t2 -> iter (Tuple t1 t2)
399    | ILeft : forall t1 t2, iter t1 -> iter (Union t1 t2)
400    | IRight : forall t1 t2, iter t2 -> iter (Union t1 t2).
```

The next function for tree iterators steps in depth-first, right-to-left order. There are four cases to consider:

- A primitive has no successor.
- A tuple steps its second child; if that has no successor step, then it steps its first child and resets the second child.
- An ILeft tries to step its child. If it has no successor, then the ILeft becomes an IRight with a newly initialized child corresponding to the right child of the union.

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409 An IRight also tries to step its child, but is final if its child has no successor.

```
410
411
    Fixpoint next(t:Typ)(i:iter t): option(iter t) := match i with
412
       | IPrim _ => None
413
       | ITuple t1 t2 i1 i2 =>
414
         match (next t2 i2) with
415
         | Some i' => Some(ITuple t1 t2 i1 i')
416
         | None =>
417
           match (next t1 i1) with
418
             Some i' => Some(ITuple t1 t2 i' (start t2))
419
           | None => None
420
           end
421
         end
422
423
       | ILeft t1 t2 i1 =>
424
         match (next t1 i1) with
425
         | Some(i') => Some(ILeft t1 t2 i')
426
         | None => Some(IRight t1 t2 (start t2))
         end
427
       | IRight t1 t2 i2 =>
         match (next t2 i2) with
         | Some(i') => Some(IRight t1 t2 i')
430
         | None => None
431
         end
432
      end.
433
434
```

An induction principle for tree iterators is needed to reason about all iterator states for a given type. First, we show that iterators eventually reach a final state. This is done with a function inum, which assigns natural numbers to each state. It simply counts the number of remaining steps in the iterator. To count the total number of union-free types denoted by a type, we use the tnum helper function.

```
441
    Fixpoint tnum(t:Typ):nat :=
442
      match t with
443
      | Prim i => 1
444
       Tuple t1 t2 => tnum t1 * tnum t2
445
      | Union t1 t2 \Rightarrow tnum t1 + tnum t2
446
      end.
447
448
    Fixpoint inum(t:Typ)(ti:iter t):nat :=
449
      match ti with
450
      | IPrim i => 0
451
        ITuple t1 t2 i1 i2 => inum t1 i1 * tnum t2 + inum t2 i2
452
        IUnionL t1 t2 i1 => inum t1 i1 + tnum t2
        IUnionR t1 t2 i2 => inum t2 i2
      end.
455
```

This function then lets us define the key theorem needed for the induction principle. At each step, the value of inum decreases by 1, and since it cannot be negative, the iterator must therefore reach a final state.

▶ **Lemma 4** (Monotonicity). If next t it = it' then inum t it = 1 + inum t it'.

It is now possible to define an induction principle over next. By monotonicity, next eventually reaches a final state. For any property of interest, if we prove that it holds for the final state and for the induction step, we can prove it holds for every state for that type.

▶ Theorem 5 (Tree Iterator Induction). Let P be any property of tree iterators for some type t. Suppose P holds for the final state, and whenever P holds for a successor state it then it holds for its precursor it' where next t it' = it. Then P holds for every iterator state over t.

Now, we can prove correctness of the subtyping algorithm with tree iterators. We implement subtyping with respect to choice strings in the Coq implementation in a two-stage process. First, we compute the union-free types induced by the iterators over their original types using here. Second, we decide subtyping between the two union-free types in ufsub. The function here walks the given iterator, producing a union-free type mirroring its state. To decide subtyping between the resulting union-free types, ufsub checks equality between Prims and recurses on the elements of Tuples, while returning false for all other types. Since here will never produce a union type, the case of ufsub for them is irrelevant, and is false by default.

```
Definition sub (a b:Typ) (ai:iter a) (bi:iter b) :=
   ufsub (here a ai) (here b bi).
```

This version of sub differs from the algorithmic implementation to ensure that recursion is well founded. The previous version of sub was, in the case of unions, decreasing on alternating arguments when unions were found on either of the sides. In contrast, the proof's version of sub applies the choice string to each side first using here, a strictly decreasing function that recurs structurally on the given type. This computes the union-free type induced by the iterator applied to the current type. The algorithm then checks subtyping between the resultant union-free types, which is entirely structural. These implementations are equivalent, as they both apply the given choice strings at the same places while computing subtyping; however, the proof version separates choice string application while the implementation intertwines it with the actual subtyping decision.

Versions of exist and allexist that use tree iterators are given next. They are similar to the string iterator functions of Section 2.2. exist tests if the subtyping relation holds in the context of the current iterator states for both sides. If not, it recurs on the next state. Similarly, allexist uses its iterator for a in conjunction with exist to ensure that the current left-hand iterator state has a matching right-hand state. We prove termination of both using Lemma 4.

```
Definition subtype(a b:Typ) = allexist a b (initial a)
Program Fixpoint allexist (a b:typ)(ia:iter a) {measure(inum ia)} =
```

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```
exists a b ia (initial b) &&
503
           (match next a ia with
504
              Some(ia') => allexist a b ia'
505
              None => true).
506
507
    Program Fixpoint exist(a b:typ)(ia:iter a)(ib:iter b)
508
                                                {measure(inum ib)} =
509
       subtype a b ia ib
                             - 1 1
510
           (match next b ib with
511
              Some(ib') => exist a b ia ib'
512
            | None => false).
513
514
```

The denotation of a tree iterator state $\mathcal{R}(i)$ is the set of states that can be reached using next from i. Let a(i) indicate the union-free type produced from the type a at i, and $|i|_a$ is the set $\{a(i') | i' \in \mathcal{R}(i)\}$, the union-free types that result from states in the type a reachable by i. This lets us prove that the set of types corresponding to states reachable from the initial state of an iterator is equal to the set of states denoted by the type itself.

▶ **Lemma 6** (Initial equivalence). $|initial \ a|_a = [a]$.

Next, Theorem 5 allows us to show that exists of a, b, with i_a and i_b tries to find an iterator state i'_b starting from i_b such that $b(i'_b) = a(i_a)$. The desired property trivially holds when $|i_b|_b = \emptyset$, and if the iterator can step then either the current union-free type is satisfying or we defer to the induction hypothesis.

▶ Theorem 7. exist a b i_a i_b holds iff $\exists t \in |i_b|_b, a(i_a) = t$.

We can then appeal to both Theorem 7 and Lemma 6 to show that exist $a \ b \ i_a$ (initial b)
finds a satisfying union-free type on the right-hand side if it exists in $[\![b]\!]$. Using this, we can
then use Theorem 5 in an analogous way to exist to show that allexist is correct up to the
current iterator state.

- Theorem 8. allexist a b i_a holds iff $\forall a' \in |i_a|_a, \exists b' \in [\![b]\!], a' = b'$.
- 531 Finally, we can appeal to Theorem 8 and Lemma 6 again to show correctness of the algorithm.
 - Theorem 9. subtype $a\ b\ holds\ iff\ \forall a'\in [a], \exists b'\in [b], a'=b'.$

3.3 Subtyping with choice strings

We prove the subtyping algorithm using choice strings correct and complete. We start by showing that iterators over choice strings simulate tree iterators. This lets us prove that the choice string based subtyping algorithm is correct by showing that the iterators at each step are equivalent. To relate tree iterators to choice string iterators, we use the itp function, which traverses a tree iterator state and linearizes it, producing a choice string using depth-first search.

```
540
541
   Fixpoint itp{t:Typ}(it:iter t):choice list :=
542
      match it with
543
      | IPrim _ => nil
544
      | ITuple t1 t2 it1 it2 => (itp t1 it1)++(itp t2 it2)
545
      546
      | IRight _ t2 it1 => Right::(itp t2 it1)
547
      end.
548
549
```

- Next, we define an induction principle over choice strings by way of linearized tree iterators.

 The next function in Section 2.2 works by finding the last L in the choice string, turning it into an R, and replacing the rest with Ls until the type is valid. If we use itp to translate
- both the initial and final states for a valid next step of a tree iterator, we see the same
- structure.
- ▶ Lemma 10 (Linearized Iteration). For some type t and tree iterators it it', if next t it = it', there exists some prefix c', an initial suffix c'' made up of Rs, and a final suffix c''' consisting of Ls such that itp t it = c' Left c'' and itp t it' = c' Right c'''.
- We can then prove that stepping a tree iterator state is equivalent to stepping the linearized versions of the state using the choice string next function.
- ▶ Lemma 11 (Step Equivalence). If it and it' are tree iterator states and next it = it', then next($itp\ it$) = ($itp\ it'$).
- The initial state of a tree iterator linearizes to the initial state of a choice string iterator.
- ▶ **Lemma 12** (Initial Equivalence). itp(initial t) = pad t [].
- The functions exist and allexist for choice string based iterators are identical to those for tree iterators (though using choice string iterators internally), and sub is as described in Section 2.2. The correctness proofs for the choice string subtype decision functions use the tree iterator induction principle (Theorem 5), and are thus in terms of tree iterators. By Lemma 11, however, each step that the tree iterator takes will be mirrored precisely by itp into choice strings. Similarly, the initial states are identical by Lemma 12. As a result, the sequence of states checked by each of the iterators is equivalent with itp.
- **Lemma 13.** exist a b $(itp i_a)$ $(itp i_b)$ holds iff $\exists t \in |i_b|_b, a(ia) = t$.
- With the correctness of exist following from the tree iterator definition, we can apply the same proof methodology to show that allexist is correct. In order to do so, we instantiate Lemma 13 with Lemma 6 and Lemma 12 to show that if exist a b (itp ia) (pad t []) then $\exists t \in [b], a(ia) = t$, allowing us to check each of the exists cases while establishing the forall-exists relationship.
- **Lemma 14.** allexist a b $(itp i_a)$ holds iff $\forall a' \in [i_a|_a, \exists b' \in [b], a' = b'$.
- We can then instantiate Lemma 14 with Lemma 12 and Lemma 6 to show that allexist for choice strings ensures that the forall-exists relation holds.
- **Theorem 15.** allexist a b (pad t []) holds iff $\forall a' \in [a], \exists b' \in [b], a' = b'$.
- Finally, we can prove that subtyping is correct using the choice string algorithm.
- **Theorem 16.** subtype a b holds iff $\forall a' \in [a], \exists b' \in [b], a' = b'$.
- Thus, we can correctly decide subtyping with distributive unions and tuples using the choice string based implementation of iterators.

4 Complexity

The worst-case time complexity of Julia's subtyping algorithm and normalization-based approaches is determined by the number of terms that could exist in the normalized type. In

6:14 Subtyping union types and covariant tuples

the worst case, there are 2^n union-free tuples in the fully normalized version of a type that has n unions. Each of those tuples must always be explored. As a result, both algorithms have worst-case $O(2^n)$ time complexity. The approaches differ, however, in space complexity. The normalization approach computes and stores each of the exponentially many alternatives, so it also has $O(2^n)$ space complexity. However, Julia need only store the choice made at each union, thereby offering O(n) space complexity.

Julia's algorithm improves best-case time performance. Normalization always experiences worst-case time and space behavior as it has to precompute the entire normalized type. Julia's iteration-based algorithm can discover the relation between types early. In practice, many queries are of the form $uft <: union(t_1...t_n)$, where uft is an already union-free tuple. As a result, all that Julia needs to do is find one matching tuple in $t_1...t_n$, which can be done sequentially without needing explicit enumeration.

5 Future work

We plan to handle additional features of Julia. Our next steps will be subtyping for primitive types, existential type variables, and invariant constructors. Adding subtyping to primitive types would be the simplest change. The challenge is how to retain completeness, as a primitive subtype heirarchy and semantic subtyping have undesirable interactions. For example, if the primitive subtype hierarchy contains only the relations $p_2 <: p_1$ and $p_3 <: p_1$, then is p_1 a subtype of Union $\{p_2, p_3\}$? In a semantic subtyping system, they are, but this requires changes both to the denotational framework and the search space of the iterators. Existential type variables create substantial new complexities in the state of the algorithm. No longer is the state solely restricted to that of the iterators being attempted; now, the state includes variable bounds that are accumulated as the algorithm compares types to type variables. As a result, correctness becomes a much more complex contextually linked property to prove. Finally, invariant type constructors induce contravariant subtyping, which when combined with existential variables may create cycles within the subtyping relation.

6 Conclusion

It is likely that subtyping with unions and tuples is always going to be exponential time, as subtyping of regular expression types have been proven to be EXPTIME-complete [10]. However, it need not take exponential space to decide subtyping: we have described and proven correct a subtyping algorithm for covariant tuples and unions that uses iterators instead of normalization. This algorithm uses linear space and allows common patterns, such as testing if a tuple of primitives is a subtype of a tuple of unions, to be handled as a special case of the subtyping algorithm. Finally, based on Julia's experience with the algorithm, we think that it can generalize to rich type languages; Julia supports bounded polymorphism and invariant constructors enabled in part by its use of this algorithm.

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632

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