Program Adverbs and Tlön Embeddings

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Free monads (and their variants) have become a popular general-purpose tool for representing the semantics of effectful programs in proof assistants. These data structures support the compositional definition of semantics parameterized by uninterpreted events, while admitting a rich equational theory of equivalence. But monads are not the only way to structure effectful computation, why should we limit ourselves?

In this paper, inspired by applicative functors, selective functors, and other structures, we define a collection of data structures and theories, which we call program adverbs, that capture a variety of computational patterns. Program adverbs are themselves composable, allowing them to be used to specify the semantics of languages with multiple computation patterns. We use program adverbs as the basis for a new class of semantic embeddings called Tlön embeddings. Compared with embeddings based on free monads, Tlön embeddings allow more flexibility in computational modeling of effects, while retaining more information about the program’s syntactic structure.

CCS Concepts: • Software and its engineering → Semantics; Syntax; Software verification; • Theory of computation → Program verification.

Additional Key Words and Phrases: formal verification, mechanized reasoning, embedding

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1 INTRODUCTION

Suppose that you want to formally verify a program written in your favorite language—be it Verilog, Haskell, or C—you first step would be to translate that program and a description of its semantics to a formal reasoning system, such as Coq [Coq development team 2022]. This step is known as semantic embedding [Boulton et al. 1992].

There are multiple approaches to semantic embeddings. The two most well-known were proposed by Boulton et al. [1992]: shallow embeddings, which represent terms of the embedded language using equivalent terms of the embedding language, and deep embeddings, which represent terms using abstract syntax trees (ASTs) and represent their semantics via some interpretation function.

Shallow embeddings are convenient because they are simple, but they have their limitations. It is impossible to use them to state and reason about properties related to syntax, because they do not retain the syntactic structure of the original program. Furthermore, shallow embeddings fix a single semantics, so they are less robust to changes in program interpretations. Such edits require changing the translation process, in addition to the semantic domain (i.e., the type used for representing the semantics of the embedded language).

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On the other hand, deep embeddings are more modular thanks to an extra layer (i.e., the AST) that defines the syntax of the embedded language. When we need to change the semantics, we only need to change the interpretation that maps the AST into some semantic domain—the translation to the formal reasoning system remains unchanged. Furthermore, the AST makes it possible to state and prove properties related to the original program’s syntactic structure. The downside is that interpreting and reasoning about properties based on this AST takes more effort than with shallow embeddings.

The pros and cons make it hard to choose between shallow and deep embeddings. Fortunately, we don’t need to commit to a single option. We can use mixed embeddings, a style of embedding that includes characteristics of each. In this style, parts of a language are embedded “shallowly” while other parts are embedded “deeply”. However, in any mixed embedding, we must ask: where should we draw the line to separate the shallowly embedded part from the deeply embedded part?

Recent efforts have focused on mixed embeddings based on freer monads [Kiselyov and Ishii 2015] or their variants [Capretta 2005; Dylus et al. 2019; McBride 2015; Piróg and Gibbons 2014; Swamy et al. 2020; Xia et al. 2020]. The style has been shown useful for representing and reasoning about effectful computation in various applications [Chlipala 2021; Christiansen et al. 2019; Foster et al. 2021; Letan et al. 2021; Nigron and Dagand 2021; Zakowski et al. 2021; Zhang et al. 2021]. Beyond these applications, this style points out a useful guideline for answering the question above. That is: modeling the pure parts of the computation “shallowly” and the effectful parts “deeply”.

Our work builds on this idea of separating pure and effectful parts in a mixed embedding, but inspects the following question: Why freer monads? We find that this is because freer monads model one general computation pattern that is common in many languages. However, the finding also implies that there are other computation patterns not captured by freer monads.

Following this observation, we propose a new class of mixed embeddings called Tlön embeddings. Tlön embeddings model programs using structures called program adverbs, which are reifications of familiar type classes (e.g., Applicative, Selective, Monad) paired with equational theories. Like freer monads, these free structures can be used to combine shallowly embedded pure computation with deeply embedded computational effects. However, program adverbs provide choices in the semantics through the selection of the structure and equational theory. For example, the “statically” adverb, based on applicative functors and their free theory, models computation where control flow and data flow in the semantics are fixed. Or, by modifying the equational theory of the free applicative structure to include commutativity, we can describe computation that is “statically and in parallel”.

We make the following contributions:

- We compare the trade-offs of different styles of semantic embeddings in the context of formal reasoning and propose Tlön embeddings (Section 2).
- We define program adverbs and show how to define their syntactic parts and their semantic parts (Section 3).
- We refactor program adverbs to support composition and extension. We motivate why we want to compose program adverbs and define a composition algebra (Section 4).
- We implement composable program adverbs using the Coq proof assistant. A major challenge for implementing them in Coq is supporting extensible inductive data types [Wadler 1998]. We show one way of addressing this challenge by adapting the Meta Theory à la Carte (MTC) approach [Delaware et al. 2013] (Section 4).

The name Tlön embedding is a reference to the short story Tlön, Uqbar, Orbis Tertius by Jorge Luis Borges. In the short story, Tlön is an imaginary world, where its parent language does not have any nouns, but only “impersonal verbs, modified by monosyllabic suffixes (or prefixes) with an adverbial value” [Borges 1940].
We identify five basic program adverbs from commonly used type classes in Haskell and we prove that these program adverbs are sound (Section 3). We also identify two add-on program adverbs that are used in combination with basic program adverbs (Section 4).

- We demonstrate the usefulness of program adverbs via three distinct language examples including a simple circuit language (Section 2), Haxl [Marlow et al. 2014], and a networked server adapted from Koh et al. [2019] (Section 5).

Additionally, we discuss the choice of adverb data types we use and alternative approaches to implement composable program adverbs in Section 6 and the related work in Section 7. We provide the Coq formalization of all the key concepts, theorems, and examples shown in this paper in our supplementary artifact [Li and Weirich 2022].

2 SEMANTIC EMBEDDINGS

In this section, we first demonstrate different forms of semantic embeddings using a simple circuit language called $B$ and compare how each form of embedding can be used to reason about programs written in this language. To distinguish the embedded language and the embedding language, we use mathematical notation to describe $B$ and use Coq code to describe its embeddings.

The syntax of $B$ appears in Fig. 1. Semantically, we want the Boolean operators to have their usual semantics. However, $B$ can read from the variables that represent references to external devices and we don’t want to fix those values in the semantics. Furthermore, we don’t know if the values are immutable: they might change over time, or they might change after each read, etc.

The four embeddings that we consider in this section are defined in the right column of Fig. 2. We use $\llbracket \cdot \rrbracket_S$, $\llbracket \cdot \rrbracket_D$, $\llbracket \cdot \rrbracket_M$, and $\llbracket \cdot \rrbracket_A$ to represent the translation from a term of $B$ to shallow, deep, and two mixed embeddings, respectively. These translations refer to the definitions in the left column as well as to the standard classes and notations for functors, monads, etc., shown in Fig. 3.

To compare embeddings, we will use each to consider the following questions regarding the semantics of $B$:

1. Is $x$ equivalent to $x \land x$?
2. Is $x$ equivalent to $x \land \text{true}$?
3. Is $t \land u$ equivalent to $u \land t$?
4. Is the number of variable accesses always less than or equal to $2$ to the power of the circuit’s depth?

Because we are modeling a circuit language that uses unknown external devices, we don’t want to be able to prove or disprove property (1). This property may hold or not hold depending on the situation. If the external devices are immutable, this property will be true. Otherwise, we may be able to falsify it. In contrast, we would like our semantic embedding to give us tools to verify properties (2) and (3) because these properties should hold regardless of the properties of our external device. The former holds because on both sides of the equivalence relation we have only accessed the variable $x$ once. The latter is due to circuits run $\land$ in parallel—whatever result appears in $t \land u$ can also appear in $u \land t$ and vice versa, regardless of what effects could be involved in $t$ or $u$. The last property (4) relates a dynamic property of the semantics (the number of variable accesses) to a syntactic property of the circuit (the size of the circuit itself).

Fig. 1. The syntax of $B$. 
Shallow Embedding

Definition Reader (A : Type) : Type :=
  (var -> bool) -> A.

Definition ret {A} (a : A) : Reader A :=
  fun _ => a.

Definition bind {A B} (m : Reader A)
  (k : A -> Reader B) : Reader B :=
  fun v => k (m v) v.

Definition ask (k : var) : Reader bool :=
  fun m => m k.

Deep Embedding

Inductive term :=
  | Var (v : var)
  | Lit (b : bool)
  | Neg (t : term)
  | And (t : term) (u : term)
  | Or (t : term) (u : term).

Freer Monad Embedding

Inductive FreerMonad (E : Type -> Type) R :=
  | Ret (r : R)
  | Bind {X} (m : E X)
    (k : X -> FreerMonad E R).
  | Fixpoint bind {E A B} (m : FreerMonad E A)
    (k : A -> FreerMonad E B) : FreerMonad E B :=
    match m with
    | Ret r => k r
    | Bind m' k' =>
      Bind m' (fun a => bind (k' a) k) end.

Variant DataEff : Type -> Type :=
  | GetData (v : var)

Reified Applicative Embedding

Inductive ReifiedApp (E : Type -> Type) R :=
  | EmbedA (e : E R)
  | Pure (r : R)
  | LiftA2 {X Y} (f : X -> Y -> R)
    (a : ReifiedApp E X) (b : ReifiedApp E Y).

Fig. 2. Semantic embeddings of \( B \) in Coq. We use the infix operator \(<\$>\) to represent a functor’s \( \text{fmap} \) method and a notation similar to Haskell’s \( \text{do} \) notation to represent monadic binds. The functions \( \neg\text{g}, \text{and}, \) and \( \text{orb} \) are Coq’s functions defined on the bool type.
Fig. 3. Coq type classes for functors, applicative functors [McBride and Paterson 2008], selective functors [Mokhov et al. 2019], and monads [Moggi 1991; Wadler 1992], as well as default definitions of fmap.

### 2.1 A Shallow Embedding

To use a shallow embedding to represent the semantics of $B$, we need a way to represent the effects of reading from external devices—the most common way of doing this is using monads (Fig. 3). But which one? A simple option is the reader monad [Jones 1995; Wadler 1992]. We show core definitions of a specialized reader monad at the top left of Fig. 2. The translation from $B$ to Reader bool is given in the same figure. Following the terminology used by Svenningsson and Axelsson [2012], we call Reader bool the semantic domain of our shallow embedding. Of course, the reader monad is just one possible semantic domain, other candidates include Dijkstra monads [Swamy et al. 2013], predicate transformer semantics [Swierstra and Baanen 2019], etc.

Using the reader monad, we can prove that property (1) is true, using ($\equiv_S$), the pointwise equality of functions. More specifically, we can prove the following Coq theorem:

$$\forall x, \text{ask} x \equiv_S x_1 \leftarrow \text{ask} x; x_2 \leftarrow \text{ask} x; \text{ret (andb x1 x2)}$$

We “ask” twice on the right hand side of the equivalence to model accessing variable $x$ twice during program runtime. However, $x_1$ equals to $x_2$ in our case since nothing has changed the global store. After proving that, the theorem can be proved via a case analysis on $x_1$.

However, note that our proof relies on “nothing has changed the global store,” but we don’t know if this is true, as we don’t know anything about the characteristics of the external device. Indeed, property (1) should not be true if we have a device where its values change over time: the value of $x$ might change between two variable access. This is a problem with our choice of semantic domain. By choosing the reader monad, we introduce more assumptions over the semantics of $B$, which results in proving a property that is not supposed to be true in the original language $B$.

Although this is not a problem with the approach of shallow embedding—we can choose a different monad than the reader monad, the style does force us to choose a concrete semantic domain.
domain early. In practice, we sometimes need to change the semantic domain, either because we
made a wrong assumption or because the language evolves. With shallow embeddings, we would
need to change the entire translation process to change this domain.

Unlike property (1), property (2) is true even though we don’t know anything about the external
device. This is because on both sides of the equivalence relation we have only accessed the variable
\(x\) once. The proof follows from the theories of Coq’s \texttt{bool} type and the \texttt{Reader} monad. However,
even though this property should be true regardless of the external device, our mechanical proof
still relies on the assumption that the external device is immutable—this is again because the
property is stated in terms of the reader monad. If we change the shallow embedding to use a
different semantic domain, we would need to prove this property again.

Property (3) is true and we can prove it to be true using our shallow embedding, but that is
just a lucky hit. Even though we know nothing about the external device, there is an equivalence
between \(t \land u\) and \(u \land t\) because the two operands \(t\) and \(u\) run in parallel in a circuit. A proof based
on our shallow embedding would, on the other hand, be based on the wrong assumption that the
external device is immutable.

We cannot state property (4) with our shallow embedding. Our shallow embedding does not
retain the syntactic structure of the original program so we cannot define a function that calculates
the depth of the circuit.

### 2.2 A Deep Embedding

In a deep embedding, we first define an abstract syntax tree (AST) for \(\mathcal{B}\). For example, we can use
the \texttt{term} data type shown in Fig. 2. Our translation from \(\mathcal{B}\) to the \texttt{term} is shown in the same figure.
Note that the \texttt{term} data type does not encode any semantic meaning.

Without an interpretation, we cannot prove any of the first three properties. This is actually
ideal for answering question (1) since we know nothing about the external device so we should
not be able to prove it (nor should we be able to prove it wrong!). However, by leaving the entire
syntax tree uninterpreted we are now unable to prove property (2) or (3), either.

A way out of this quandary is to define a coarser equivalence relation for ASTs and use that
relation in the statement of properties (2) and (3). For example, we can interpret each \texttt{term}
using the reader monad (as in the shallow embedding) and use the pointwise equality for that type. The
proofs are essentially the same as the above.

One advantage of the deep embedding in this case is that, if we would like to change our definition
of equivalence, we can do so by choosing a different interpretation without changing the translation
process. In other words, deep embeddings achieve better modularity by introducing an intermediate
layer. The price, however, is that it takes effort to build that extra intermediate layer. This extra
effort seems small here, but can become tedious with some languages, e.g., those with features like
“let” that introduce variable bindings [Aydemir et al. 2005].

However, we still face a similar problem with the shallow embedding: If we would like to change
the interpretation in our definition of equivalence, we need to prove our properties again. This
suggests that another intermediate layer between deep and shallow embeddings might be helpful,
as we will see in the next subsection.

The primary benefit we have by using the deep embedding is that we can now state and prove
property (4). This is because the deep embedding gives us a representation of the program’s original
syntactic structure. This allows us to define the following function that counts the depth of a circuit:

```ml
Fixpoint depth (t : term) : nat :=
match t with
| Var _ => 0
```

Left identity: \[ \text{ret } a \gg= h = h \ a \]

Right identity: \[ m \gg= \text{ret } = m \]

Associativity: \[ (m \gg= g) \gg= h = m \gg= (\text{fun } x \Rightarrow g \ x \gg= h) \]

Fig. 4. The monad laws. The \( \gg= \) symbol is the infix operator for \text{bind}.

\[
\begin{align*}
| \text{Lit } _ & \Rightarrow 0 \\
| \text{Neg} \ t & \Rightarrow \text{depth } t + 1 \\
| \text{And} \ t \ u & \Rightarrow \text{max} (\text{depth } t) (\text{depth } u) + 1 \\
| \text{Or} \ t \ u & \Rightarrow \text{max} (\text{depth } t) (\text{depth } u) + 1 \\
\end{align*}
\]

Since we assume a straightforward semantics for \( B \), the number of variable access at runtime equals to the number of variables appeared in a term, so we can directly prove property (4) by an induction over the term data type.

### 2.3 A Mixed Embedding Based on Freer Monads

A semantic embedding can be partially shallow and partially deep. We use the term *mixed embeddings* to describe embeddings with this property. One style of mixed embeddings that is popular today is based on freer monads [Chlipala 2021; Dylus et al. 2019; Letan et al. 2021; McBride 2015; Nigron and Dagand 2021; Swamy et al. 2020; Xia et al. 2020]. In this type of mixed embeddings, the pure parts of the program are embedded shallowly, while effects are embedded deeply (and abstractly) using algebraic data types “connected” by freer monads.

The core definitions of freer monads are in the left column of Fig. 2. The \text{FreerMonad} data type is parameterized by an abstract effect \( E \) of \( \text{Type} \rightarrow \text{Type} \) and a return type \( R \) of \( \text{Type} \). Conceptually, it collects all the deeply embedded effects \( E \) in a right-associative monadic structure.

For any effect type, \text{FreerMonad} \( E \) is a monad as demonstrated by the \text{Ret} constructor and \text{bind} function. The \text{bind} function pattern matches its first argument \( m \) and, in the case of \text{Bind}, passes its second arguments \( k \) to the continuation of \( m \). This “smart constructor” ensures that binds always associate to the right.

To embed \( B \), we model reading data from external devices using the effect type \text{DataEff}. This datatype includes only one (abstract) effect, called \text{GetData}. This constructor represents a data retrieval with the variable \( v : \text{var} \) that returns an unknown \text{bool}. Similar to how the term data type says nothing about the semantics of \( B \), the effect data type \text{DataEff} says nothing about the semantics of a data read. As a result, we say that the effects are embedded deeply in this style.

The embedding function appears on the right side of Fig. 2. The translation strategy is almost the same as embedding \( B \) using the reader monad. The only exception is in the variable case (the effectful part): here the \text{Bind} constructor marks the occurrence of the \text{GetData} effect.

In this mixed embedding, the pure parts of a \( B \) program have been translated to a shallow semantic domain, but the effectful parts remain abstract. It turns out that this separation is useful for both questions (1) and (2).

For question (1), we cannot answer it. This is desirable since we don’t know if it’s true without knowing more about the external device.

We can prove that property (2) is true even though the read effect is not interpreted—this is because the property follows from the monad laws (Fig. 4). However, we cannot prove property (3) because the commutativity law is not one of the monad laws.
Ideally, we would also like to state and prove property (4). However, the dynamic nature of freer monads forbids us from statically inspecting the syntactic structure of the program. Interpreting the embedding does not help us, either, since that would not preserve the original syntactic structure.

Our success with questions (1) and (2) suggests that we have found an useful intermediate layer between shallow and deep embeddings, but our failure in stating or proving properties (3) and (4) indicates that we haven’t yet found the most suitable representation for this circuit language.

2.4 Another Mixed Embedding Based on Reified Applicative Functors

The last embedding shown in the figure uses a type that reifies the interface of applicative functors (Fig. 3). As in freer monads, this datatype is parameterized by deeply embedded abstract effects. These effects, of type E R, are recorded by the EmbedA data constructor.

However, instead of constructors for ret and bind, this datatype includes constructors for pure and liftA2, the two operations that define applicative functors.3 The Pure constructor shallowly “embeds” a pure computation into the domain, and LiftA2 “connects” two computations that potentially contain effect invocations. These constructors provide a trivial implementation of the Applicative type class for this datatype.

The translation of B to this datatype uses a deep embedding of variable reads, using the EmbedA data constructor with the DataEff type from the previous embedding. Because, as in freer monads, this effect is modeled abstractly, we cannot prove or disprove (1).

The translation function uses the applicative interface in the datatype to translate the constants, unary and binary operators. These components are modeled shallowly (i.e., as Boolean constants and operators), but the program’s syntactic structure is retained by the translation. However, because of the retainment, we need an additional equivalence relation to equate semantically equivalent terms that are not syntactically equal. To prove (2), we include the right identity law of applicative functors in the equivalence (denoted by ≈):

\[
\forall y, \ (\text{fun } _\_ x => x) a y = f a y \\
\text{liftA2 } f (\text{pure } a) b \equiv b
\]

This law is sufficient to show that (2) holds.

To model the parallelism of circuits, we could include the commutativity law in the equivalence:

\[
\text{liftA2 } f a b \equiv \text{liftA2 } (\text{flip } f) b a
\]

This is sufficient to show (3). Note that this is not one of the applicative functor laws. We defer showing the soundness of including this rule in the equivalence to Section 3.4.

This embedding also preserves enough of the syntax of the original program to prove (4). To do so, we must first calculate the depth of circuits and the number of variables under this encoding.

\[
\text{Fixpoint } \text{app_depth } \langle E A \rangle (t : \text{ReifiedApp } E A) : \text{nat} := \\
\text{match } t \text{ with} \\
| \text{EmbedA } _ => 0 \\
| \text{Pure } _ => 0 \\
| \text{LiftA2 } _ t u => 1 + \text{max } (\text{app_depth } t) (\text{app_depth } u) \\
\text{end.}
\]

We omit the function that counts the number of variables as it is similar to app_depth. Then we can formalize (4) in Coq as follows:

3Alternatively, Applicative can also be defined by pure and another operation of type \( F (A -> B) -> F A -> F B \), where \( F \) is an Applicative instance. These two definitions are equivalent, as we can derive the definition of \( \langle\rangle \) from liftA2 and vice versa.
Theorem  heightAndVar : forall (c : ReifiedApp DataEff bool),
    app_numVar c <= Nat.pow 2 (app_depth c).

The theorem is provable by an induction over c.

Furthermore, this embedding also allows us to reason about semantic properties that depend on syntactic structures of circuits. One example is a semantics with some cost model. In the semantics, we may not want our equivalence to equate, for example, \( x \land y \land z \land w \) and \( (x \land y) \land (z \land w) \) because they are not equivalent in their costs when parallelization is present. Indeed, we cannot show that they are equivalent with our embedding due to the absence of associativity in our equivalence.

2.5 Tlön embeddings

Just as the reader monad models one particular effect, freer monads model one particular computation pattern. Unfortunately, that particular computation pattern is not suitable for our \( \mathcal{B} \) example, because it does not model parallel computation (i.e., property (3)), nor does it capture the static data and control flows (i.e., property (4)). Instead we saw that the mixed embedding in the previous subsection, based on reified applicative functors, is a better approach.

Can we generalize the key idea even further? If we go beyond \( \mathcal{B} \), we might need to model other computation patterns. Are there other mixed embeddings that would be suitable for these tasks? How might we derive them?

To that end, we identify a novel set of mixed embeddings that we call Tlön embeddings. The goal of these embeddings is to provide flexibility in our models of effectful computation.\(^4\) We define Tlön embeddings by identifying a set of program adverbs that specify the embedding type and equational theory used in the embedding. For example, the embedding in Section 2.4 is based on an adverb composed of the ReifiedApp type and some rules of commutative applicative functors.

The flexibility that program adverbs provide can perhaps be understood by comparing them with effects: effects do certain actions, and program adverbs model how these actions are done—similar to the difference between verbs and adverbs. For example, the adverb we used in Section 2.4 is called "statically and in parallel", which states that there is a static dependency between different effect invocations and some of these effect invocations are executed in parallel.

In the next section, we define our set of program adverbs more precisely and discuss the reasoning principles that they provide for effectful computation.

3 PROGRAM ADVERBS

Program adverbs are the building blocks of Tlön embeddings. Mathematically, they are composed of two parts: a syntactic part, called the adverb data type, and a semantic part, called the adverb theory. More formally, we define program adverbs as follows:\(^5\)

**Definition 3.1 (Program Adverb).** A program adverb is a pair \( (D, \equiv_D) \). \( D \) is called the adverb data type and is parameterized by an effect \( E \) and a return type \( R \). The \( \equiv_D \) operation is called the adverb theory of \( D \). It is a binary operation that defines an equivalence relation on \( D(E, R) \) for any \( E \) and \( R \).

In the rest of the paper, we abbreviate \( \equiv_D \) as \( \equiv \) when \( D \) is clear from the context.

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\(^4\)Here, we define effects as communications with external environment that are performed by some explicit operations. For example, mutable states are effects which can be explicitly incurred by operations such as get and set. For the same reason, we also consider I/O (with operations like `read`, `print`, etc.) and exceptions (with operations like `throw`, etc.) as effects.

\(^5\)The Coq code of all definitions and theorems shown in this section can also be found in our supplementary artifact [Li and Weirich 2022].
In Coq terms, an adverb data type $D$ has the type $(\text{Type} \rightarrow \text{Type}) \rightarrow \text{Type} \rightarrow \text{Type}$. The first parameter of $\text{Type} \rightarrow \text{Type}$ is the effect $E$ and it's parameterized by its own return type; the second parameter is the return type $R$. The adverb theory $\simeq$ is a typed binary relation. More concretely:

**Class** Adverb (D : (Type -> Type) -> Type -> Type) :=
{ Equiv {E R} : relation (D E R) ;
  equiv {E R} : Equivalence (@Equiv E R) }.
**Notation** "a $\simeq$ b" := (Equiv a b).

where $D$ is the adverb data type, $\text{Equiv}$ is the adverb theory $\simeq$, and $\text{equiv}$ is a proof showing that $\text{Equiv}$ is an equivalence relation. The datatype relation is defined as:

**Definition** relation (A : Type) := A -> A -> Prop.

This definition is overly general, so we focus our attention only on program adverbs that are *sound* according to the definition that we will develop below. Furthermore, in this paper we will only consider adverbs defined by reifying classes of functors.

### 3.1 Adverb Data Types and Theories

The four key adverb data types, shown in Fig. 5, are derived from the four type classes shown in Fig. 3. We have already seen one before in the applicative embedding in Fig. 2. Other definitions

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**Congruence Rule**

\[
\begin{align*}
\text{Congruence} & : \quad a_1 \simeq a_2 \quad b_1 \simeq b_2 \\
\text{liftA2 f a_1 b_1} & \simeq \text{liftA2 f a_2 b_2}
\end{align*}
\]

**Applicative Functor Laws**

**Left Identity**

\[
\forall y, \text{(fun } x \Rightarrow x) \quad a \cdot y = f \cdot a \cdot y
\]

\[
\text{liftA2 f (pure a) b} \simeq b
\]

**Right Identity**

\[
\forall x, \text{(fun } x \Rightarrow x) \quad x \cdot b = f \cdot x \cdot b
\]

\[
\text{liftA2 f a (pure b)} \simeq a
\]

**Associativity**

\[
\forall x y z, f \cdot x \cdot y \cdot z = g \cdot y \cdot z \cdot x
\]

\[
\text{liftA2 id (liftA2 f a b) c} \equiv \text{liftA2 (flip id) a (liftA2 g b c)}
\]

\[
\forall x y z, p \cdot (q \cdot x \cdot y) \cdot z = f \cdot x \cdot (g \cdot y \cdot z)
\]

**Naturality**

\[
\text{liftA2 p (liftA2 q a b) } \equiv \text{liftA2 f a \cdot liftA2 g b}
\]

**Equivalence Properties**

**Reflexivity**

\[
a \simeq a
\]

**Symmetry**

\[
b \simeq a
\]

**Transitivity**

\[
a \simeq b \quad b \simeq c
\]

\[
a \simeq c
\]

Fig. 6. The equivalence relation for ReifiedApp. The infix operator \( \cdot \) denotes function compositions.

follow a similar pattern: the constructors of each data type include one for embedding effects (of type \( E \cdot R \)) and a constructor that reifies the interface of each method of the type class.

In addition to an adverb data type, every program adverb also comes with some theories, defined by an equivalence relation \( \simeq \). The purpose of the \( \simeq \) relation is to equate all computations that are semantically equivalent regardless of what effects are present.

For example, an adverb called Statically is composed of the ReifiedApp datatype with an equational theory based on three sorts of rules: (1) a congruence rule with respect to LiftA2, (2) the laws of applicative functors [McBride and Paterson 2008], and (3) the equivalence properties (i.e., reflexivity, symmetry, transitivity). We show the concrete rules in Fig. 6.

Why do we call this adverb Statically? The data dependency in the LiftA2 constructor of ReifiedApp shows that the data type imposes a “static” data flow and control flow on the computation: we will always need to run both parameters of type ReifiedApp \( E \cdot A \) and ReifiedApp \( E \cdot B \) to get the result of type ReifiedApp \( E \cdot C \), i.e., we cannot skip either computation. In addition, neither of the two parameters depends on the result of the other, which allows us to statically inspect either of them without running the other.

**Remark.** The adverb data types and their associated theories form free structures similar to those in Capriotti and Kaposi [2014]; Kiselyov and Ishii [2015]; Mokhov [2019]; Mokhov et al. [2019]. However, one distinction is that we intentionally do not normalize the adverb data types to preserve syntactic structures. To distinguish un-normalized free structures and normalized free structures, we use the term reified structures to describe the former and the term free structures to exclusively
Fixpoint interpA {E I : Type -> Type} `{Applicative I} {A : Type} (interpE : forall A, E A -> I A) (t : ReifiedApp E A) : I A :=
  match t with
  | EmbedA e => interpE _ e
  | Pure a => pure a
  | LiftA2 f a b => liftA2 f (interpA interpE a) (interpA interpE b)
  end.

Fig. 7. The interpretation from ReifiedApp to any instance of the Applicative type class.

describe the latter. We defer the detailed comparison and trade-offs between reified structures and free structures to Section 6.

3.2 Adverb Simulation

One important property of ReifiedApp is that it can be interpreted to any other instance of the Applicative class, as long as its embedded effects can be interpreted to that instance. We can show this via the abstract interpreter interpA shown in Fig. 7. The interpreter shows that given any effect E and any instance I of Applicative, as long as we can find an effect interpretation from E A to I A for any type A, we can interpret a ReifiedApp E A to an I A for any type A.

For example, we can interpret a ReifiedApp DataEff to the reader applicative functor (Fig. 2)\(^7\) by supplying the following function to the parameter interpE of interpA:

Definition interpDataEff {A : Type} (e : DataEff A) : Reader A :=
  match e with GetData v => ask v end.

Similarly, we can interpret ReifiedApp DataEff to other semantic domains that are applicative functors.

Why do we care if ReifiedApp can be interpreted into any instance of Applicative? This is because different instances of Applicative model different effects—if we have a data structure that can be interpreted to all instances, we can develop a theory of it that can be used for reasoning about properties that are true regardless of what effects are present.

To make the relation between an adverb data type like ReifiedApp and a class of functors like Applicative more precise, we define the following adverb simulation relation:

Definition 3.2 (Adverb Simulation). Given an adverb data type \(D\), a class of functors \(C\), and a transformer \(T\) on all instances of \(C\), we say that there is an adverb simulation from \(D\) to \(C\) under \(T\), written \(D \models_T C\), if we can define a function that, for any effect type \(E\), instance \(F\) of type class \(C\), and interpreter \(f\) from \(E(A)\) to \(F(A)\) for any type \(A\), interprets a value of \(D(E, A)\) to \(T(F)(A)\) for any type \(A\).

We add some flexibility to this definition by making it parameterize over a transformer \(T\)—we do not need this extra flexibility for now, but we will see why it is useful in Section 3.4.

We also define an adverb interpretation as follows:

Definition 3.3 (Adverb Interpretation). Given an adverb data type \(D\), a class of functors \(C\), and a transformer \(T\) on all instances of \(C\), an interpreter \(I\) that shows \(D \models_T C\) is called an adverb interpretation, and we write \(I \in D \models_T C\).

\(^7\)Every monad is also an applicable functor, so the reader monad is also a reader applicative functor.
Our interpA in Fig. 7 is an adverb interpretation. More specifically, we say that

\[ \text{interpA} \in \text{ReifiedApp} \models \text{IdT Applicative} \]

where the IdT transformer is an identity Applicative transformer that “does nothing”. In the rest of the paper, when we have \( D \models \text{IdT} C \) for any \( D \) and \( C \), we abbreviate it as \( D \models C \).

### 3.3 Sound Adverb Theories

To know that our adverb theory is sound, i.e., it doesn’t equate computations that are not semantically equivalent, we define the following soundness property of adverb theories:

**Definition 3.4 (Soundness of Adverb Theories).** Given a program adverb \((D, \simeq)\) and an adverb interpretation \( I \in D \models T C \), we say that the adverb theory \( \simeq \) is sound with respect to \( I \) if there exists a lawful equivalence relation \( \equiv \) such that for all \( d_1, d_2 \in D \),

\[ d_1 \simeq d_2 \implies I(d_1) \equiv I(d_2). \]

Let us use \( \text{idT} \) for the transformer \( T \) for the moment. The equivalence relation \( \equiv \) on \( C \) is lawful if they respect the congruence laws and the class laws of \( C \). For Applicative, we use the common applicative functor laws regarding \( \equiv \). Based on the soundness of adverb theories, we can define the following soundness property of program adverbs with respect to their adverb interpretations:

**Definition 3.5 (Soundness of Program Adverbs).** Given a program adverb \((D, \simeq)\) and an adverb interpretation \( I \in D \models T C \), we say that the adverb is sound if the \( \simeq \) relation is sound with respect to \( I \).

We can now prove that the Statically adverb is sound:

**Theorem 3.6.** The Statically adverb \((\text{ReifiedApp}, \simeq)\) is sound with respect to the adverb interpretation \( \text{interpA} \in \text{ReifiedApp} \models \text{Applicative} \).

**Proof.** By induction over the \( \simeq \) relation. \( \square \)

### 3.4 “Statically and in Parallel”

Two adverbs can use the same data type yet differ in their theories. Let’s look at a variant of the Statically adverb called StaticallyInParallel. As its name suggests, it adds parallelization to a static computation pattern.

Recall that the two computations connected by liftA2 do not depend on each other. This suggests that an implementation of liftA2 can choose to run them in parallel. Indeed, that observation is one of the key ideas behind Haxl [Marlow et al. 2014].

Based on this idea, we also define the StaticallyInParallel adverb. The adverb data type of this adverb is the same as that of Statically. However, its theory differs from Statically in the following ways: (1) it adds the commutativity rule:

\[ \text{liftA2 } f \ a \ b \ \simeq \ \text{liftA2 } (\text{flip } f) \ b \ a \]

and (2) it does not include the associativity and naturality rules (Fig. 6).

The addition of commutativity rule states that the order that effects are invoked does not matter. Note that compared with other rules, the commutativity rule is not satisfied by every applicative functor. This might suggest that we should not add it to the theory, as it might be a theory that only holds for certain effects. Nevertheless, we can prove the soundness of the adverb theory with respect to the following adverb simulation:

\[ \text{ReifiedApp} \models \text{PowerSet Applicative} \]
**Definition** \( \text{PowerSet} \ (I : \text{Type} \rightarrow \text{Type}) \ (A : \text{Type}) := I \ A \rightarrow \text{Prop} \).

**Definition** \( \text{embedPowerSet} \ {A : \text{Type}} \ (a : I \ A) : \text{PowerSet} \ I \ A := \text{fun} \ r \Rightarrow r \equiv a. \)

**Definition** \( \text{purePowerSet} \ {A : \text{Type}} \ (a : A) : \text{PowerSet} \ I \ A := \text{fun} \ r \Rightarrow r \equiv \text{pure} \ a. \)

**Definition** \( \text{liftA2PowerSet} \ {A B C} \ (f : A \rightarrow B \rightarrow C) \ (a : \text{PowerSet} \ I \ A) \ (b : \text{PowerSet} \ I \ B) : \text{PowerSet} \ I \ C := \)
\[
\text{fun} \ r \Rightarrow \exists \ a', \ a' /\ \exists \ b', \ b' /\ \left( \text{liftA2} \ f \ a' \ b' \equiv r \lor \text{liftA2} \ (\text{flip} \ f) \ b' \ a' \equiv r \right).
\]

**Definition** \( \text{EqPowerSet} \ {A} : \text{relation} \ (\text{PowerSet} \ I \ A) := \text{fun} \ p \ q \Rightarrow \forall \ a, \ p \ a \leftrightarrow q \ a. \)

Fig. 8. The core definitions of a powerset applicative functor transformer.

The PowerSet transformer is a transformer on applicative functors and its core definitions are shown in Fig. 8. The key of PowerSet is the liftA2PowerSet operation. When executed, it creates two nondeterministic branches (indicated by the disjunction \( \lor \)) on one branch, it computes \( a' : I \ A \) before \( b' : I \ B \), and vice versa on the other branch. Intuitively, this is to model the nondeterministic execution order in a parallel evaluation. Many of these operations depend on \( \equiv \), which is the lawful \( \equiv \) relation on \( I \).

**Lemma 3.7.** If \( \equiv \) is a lawful equivalence relation on Applicative, EqPowerSet is an equivalence relation on PowerSet that satisfies congruence, left identity, right identity, and commutativity laws.

**Proof.** By definition. \( \square \)

Note that EqPowerSet does not satisfy the associativity or naturality laws. Consider that we have liftA2PowerSet id (liftA2PowerSet f a b) c, for some \( f, a, b, \) and \( c \). One of the possible evaluations in this powerset is liftA2 id (liftA2 (flip f) b a) c, which does not belong to the powerset of liftA2PowerSet (flip id) a (liftA2PowerSet g b c), for some \( g \) that is equivalent to \( \text{flip} \ f \). The case for naturality is similar. For this reason, we do not include these two rules in \( \equiv \). We do not know if there exists an alternative nontrivial transformer with an equivalence relation that satisfies all the applicative laws in addition to commutativity.

Nevertheless, we can show the following theorem with the help of Lemma 3.7:

**Theorem 3.8.** The adverb is sound: \( \text{ReifiedApp} \models \text{PowerSet Applicative} \).

**Proof.** We can construct an interpPowerSet \( \in \text{ReifiedApp} \models \text{PowerSet Applicative} \) by modifying interpA (Fig. 7) so that it uses embedPowerSet on the EmbedA case, purePowerSet on the Pure case, and liftA2PowerSet on the LiftA2 case. With the help of Lemma 3.7, we can show that for all \( d_1, d_2 \in \text{ReifiedApp} \),
\[
d_1 \equiv d_2 \implies \text{interpPowerSet}(d_1) \equiv \text{interpPowerSet}(d_2)
\]
where \( \equiv \) is EqPowerSet. \( \square \)

Intuitively, we can define StaticallyInParallel as an adverb because, even though with an effect running computations in different order might return different results, a language can be implemented in a parallel way such that the difference in evaluation orders is no longer observable.
The lack of associativity and naturality rules in the theory of StaticallyInParallel might initially sound limiting, but, as we have shown in the end of Section 2.4, it turns out to be desirable for applications like circuits.

### 3.5 Other Basic Adverbs

Besides Statically and StaticallyInParallel, we also identify three other basic adverbs, namely Streamingly, Conditionally, and Dynamically, defined using the adverb data types in Fig. 5.

**Streamingly.** This program adverb simulates Functor under IdT. The most simple form of stream processing computes the data directly as it is received. This is captured by the fmap interface (Fig. 3).

**Dynamically.** This adverb simulates Monad (Fig. 3). A monad is the most expressive and dynamic among all four classes of functors thanks to its core operation bind. Any kind of computation can happen in the second operand and we can’t know it without knowing a value of type A, which we can only get by running the first operand. This program adverb is commonly used in representing many programming language for its expressiveness, but it also allows for the least amount of static reasoning.

Unlike Statically, this variant does not have an InParallel variant. This might be surprising because there are many commutative monads. However, those monads are commutative because their specific effects are commutative. We cannot define a general powerset monad transformer that can make any monad satisfy the commutativity law.

**Conditionally.** We use this adverb to model conditional execution. The definition of its adverb data type is shown in Fig. 5. It reifies the Selective type class (Fig. 3). The signature operation of Selective is the selectBy operation. Loosely, “applying” a function of type A → ((B → C) + C) to a computation of type F A gets you either F (B → C) or F C. In the first case, you will need to run the computation of type F B. You don’t need to run the computation of type F B in the second case, but you can still choose to run it.

Because we can encode conditional execution with this adverb, it is more expressive than Statically. However, the extra expressiveness also makes static analysis less accurate. Since we cannot know statically if the computation F B in selectBy is executed, we can only get an under-approximation (assuming that F B is not executed) and an over-approximation (assuming that F B is executed) of the effects that would happen, but not an exact set.

Even though we derive this adverb by reifying Selective, we do not wish to model the adverb’s theory using the laws of selective functors. This is because the laws of selective functors do not distinguish them from applicative functors. Indeed, every applicative functor is also a selective functor (by running the second argument even when not required) and vice versa, so adhering to the “default” laws do not allow us to prove more properties. Therefore, we add one simple rule to the selective functor laws:

\[
\text{select} \ (\text{inr} \ <$> \ a) \ b \ \equiv \ a
\]

The function select has the type F (A + B) → F (A → B) → F B, where F is an instance of Selective. It is equivalent to

\[
\text{selectBy} \ (\text{fun} \ x \ => \ \text{match} \ x \ \text{with} \\
| \ \text{inl} \ x \ => \ \text{inl} \ (\text{fun} \ y \ => \ y \ x) \\
| \ \text{inr} \ x \ => \ \text{inr} \ x \\
\text{end}).
\]

This rule forces select to ignore the second argument when it does not need run. However, we can no longer show that Conditionally adverb simulates Selective by adding this laws, because ≡
is no longer an under-approximation of \( \equiv \). Instead, we show the following adverb simulation:

\[
\text{ReifiedSelective} \models \text{Monad}
\]

In this way, Conditionally serves as a compromise between Statically and Dynamically. Its adverb data type is more similar to Statically and allows for some static analysis, while its theories are more similar to Dynamically.

4 COMPOSABLE PROGRAM ADVERBS

From a monad instance, we can derive an applicative functor instance. From an applicative functor instance, we can derive a functor instance. We can derive a selective instance from an applicative functor and vice versa.\(^8\) This subsumption hierarchy among classes of functors means that we can choose the most expressive abstract interface of a data type, and that choice automatically includes the less expressive interfaces.

However, although we can derive a “default” applicative functor from a monad, we don’t always want to do that—e.g., we may want to define a different behavior for \( \text{liftA2} \) than the one derived from \( \text{bind} \). Indeed, Haxl is one such example, where \( \text{bind} \) is defined as a sequential operation and \( \text{liftA2} \) is parallel so that certain tasks with no data dependencies can be automatically parallelized [Marlow et al. 2014]. In the program adverbs terminology, the semantics of their language is composed of a “statically and in parallel” adverb and a “dynamically” adverb.

In addition, some languages may have a part that corresponds to the “statically” adverb and some extensions that correspond to “dynamically”. If we only use the “dynamically” adverb to reason about programs written in this language, we lose the ability to state properties for the “statically” subset.

We need a way to compose multiple program adverbs. Therefore, in this section, we refactor program adverbs to composable program adverbs.

4.1 Uniform Treatment of Effects and Program Adverbs

Effects are commonly considered secondary to monads. This treatment of effects carries over to the approaches based on freer monads and our previous implementation of program adverbs, where the effects are a parameter of adverb data types.

This approach works well when we use one fixed program adverb, but needs an update when multiple adverbs are involved. This is because, in both scenarios we mentioned earlier, our intention is not to combine program adverbs that each contain their own set of effects—we would like the composed program adverbs to share the same set of effects. One solution is requiring that we can only join program adverbs when they share the same set of effects, but that would require extra machinery.

In our work, we choose to give a uniform treatment to effects and program adverbs. Figure 9 shows our algebra for effects and program adverbs. The algebra includes an \( \oplus \) operator which is a disjoint union of effects and adverb data types. We define an equivalence relation \( \approx \) on effects and adverb data types as follows: for all \( A, B \) that are effects and adverb data types, \( A \approx B \) if there exists a bijection between \( A \) and \( B \). Similarly, we define an \( \uplus \) operator for the disjoint union of adverb theories. We define an equivalence relation \( \iff \) on adverb theories as follows: for any adverb data type \( D \) and adverb theories \( P, Q \), which are adverb theories of \( D \), \( P \iff Q \) if \( a P b \iff a Q b \) for all \( a, b \in D \), where \( \iff \) is the logical symbol for “if and only if”. Properties of this algebra are also shown in Fig. 9.

\(^8\)This is one special thing about selective functors: every selective functor is an applicative functor and the reverse is also true. However, separating these two classes is still useful because the automatically derived instances might not be what we want, as discussed in Mokhov et al. [2019].
effects and adverb data types  \( A, B, C \)  
\[ := \text{Effect } E \mid \text{AdverbDataType } D \mid A \oplus B \]

adverb theories  \( P, Q, R \)  
\[ := \text{AdverbTheory } \equiv_D \mid P \uplus Q \]

Properties of \( \oplus \)

**Commutativity**  
\( A \oplus B \approx B \oplus A \)

**Associativity**  
\[ (A \oplus B) \oplus C \approx A \oplus (B \oplus C) \]

Properties of \( \uplus \)

**Commutativity**  
\( P \uplus Q \iff Q \uplus P \)

**Associativity**  
\[ (P \uplus Q) \uplus R \iff P \uplus (Q \uplus R) \]

**Idempotence**  
\( P \uplus P \iff P \)

Fig. 9. The algebra for effects and composable program adverbs.

4.2 The Coq Implementation

All the adverb data types we have seen (Fig. 5) are recursive. When we compose these program adverbs, we cannot simply put these inductive types into a sum type—we need to adapt each adverb so that it recurses on the new composed adverb rather than itself. In other words, we need *extensible inductive types*. However, extensible inductive types are not directly supported by most formal reasoning systems including Coq. In fact, how to support extensible inductive types is part of an open problem known as the *expression problem* [Wadler 1998].

In this paper, we address the problem and implement composable adverbs in Coq using a technique presented in *Meta Theory à la Carte* (MTC) [Delaware et al. 2013]. The key idea of MTC is using Church encodings of data types [d. S. Oliveira 2009; Wadler 1990] instead of Coq’s native inductive types. We apply and extend this idea to define the two least fixpoint operators \( \text{Fix1} \) and \( \text{FixRel} \) that work on adverb data types and adverb theories, respectively. We show the definitions of these operators in Fig. 10a.

We define the disjoint union \( \oplus \) by first refactoring the types of adverb data types and effects. We make both adverb data types and effects have the type \((\text{Set} \to \text{Set}) \to \text{Set} \to \text{Set}\) where the first parameter is a *recursive parameter* and the second parameter is a return type. We can then define \( \oplus \) simply as a sum type on \((\text{Set} \to \text{Set}) \to \text{Set} \to \text{Set}\), as shown in Fig. 10b. Similarly, we define \( \uplus \) as a sum type on \((\forall A : \text{Set}, \text{relation } (F A)) \to \forall A : \text{Set}, \text{relation } (F A)\) for any \( F : \text{Set} \to \text{Set}\).

Figure 11a shows the definitions of composable adverb data types. Compared with the adverb data types in Fig. 5, a composable adverb data type replaces the effect parameter (which is named \( E \)) with a recursive parameter (which is named \( K \)) so that it “recurses” on \( K \) instead of itself.

We also factor out the Pure constructor, a common part shared by multiple basic adverb data types, as a separate composable adverb data type called ReifiedPure. In this way, we avoid introducing multiple Pure constructors, e.g., by combining Statically and Conditionally. Furthermore, we remove the Embed constructors in composable adverb data types. Thanks to the uniform treatment of effects and program adverbs, we can now embed effects simply by including them in \( K \), so we have no need for those constructors.

As an example, we can define an “inductive type” \( T : \text{Set} \to \text{Set} \) that is composed of ReifiedPure, ReifiedApp, and some effect \( E : (\text{Set} \to \text{Set}) \to \text{Set} \to \text{Set} \) as follows:

**Definition**  
\( T := \text{Fix1} (\text{ReifiedPure} \oplus \text{ReifiedApp} \oplus E) \).
\textbf{Definition} \texttt{Alg1} \ ((F : \text{(Set -> Set) -> Set -> Set}) \ (E : \text{Set -> Set}) : \text{Type} :=
\forall \{A : \text{Set}\}, F \ E \ A \to E \ A.

\textbf{Definition} \texttt{Fix1} \ ((F : \text{(Set -> Set) -> Set -> Set}) \ (A : \text{Set}) :=
\forall (E : \text{Set -> Set}), \text{Alg1} \ F \ E \to E \ A.

\textbf{Definition} \texttt{AlgRel} \ ((F : \text{Set -> Set}) \ (R : \text{(\forall (A : \text{Set}), relation (F A)) -> \forall (A : \text{Set}), relation (F A)})
\forall (K : \text{\forall (A : \text{Set}), relation (F A)}),
\text{fun} A (a : F A) \Rightarrow R \ K \_ a b \to K \_ a b.

\textbf{Definition} \texttt{FixRel} \ ((F : \text{Set -> Set}) \ (R : \text{(\forall (A : \text{Set}), relation (F A)) -> \forall (A : \text{Set}), relation (F A)})
\forall (A : \text{Set}), \text{relation (F A)} :=
\text{fun} A (a : F A) \Rightarrow \forall (K : \text{\forall (A : \text{Set}), relation (F A)}),\ (\forall (A : \text{Set}) (a : F A), \text{AlgRel} \ R \ K \_ a b) \to K \_ a b.

(a) The algebra and the least fixpoint operators for effects and adverb data types (\texttt{Alg1}, \texttt{Fix1}), and for adverb theories (\texttt{AlgRel}, \texttt{FixRel}).

\textbf{Variant} \texttt{Sum1} \ ((F \ G : \text{Set -> Set) -> Set -> Set}) \ (K R) :=
\texttt{Inl1} (a : F K R) | \texttt{Inr1} (a : G K R).

\textbf{Variant} \texttt{SumRel} \ ((F : \text{Set -> Set}) \ (P Q : \text{(\forall (A : \text{Set}), relation (F A)) -> \forall (A : \text{Set}), relation (F A)})
\forall (K : \text{\forall (A : \text{Set}), relation (F A)}),
| \texttt{InlRel} \ (A : \text{Set}) (a : F A) : P K \_ a b \to \text{SumRel} P Q K \_ a b
| \texttt{InrRel} \ (A : \text{Set}) (a : F A) : Q K \_ a b \to \text{SumRel} P Q K \_ a b.

\textbf{Notation} "\textit{F} \oplus \textit{G}" := (\texttt{Sum1} \ F \ G).

\textbf{Notation} "\textit{F} \uplus \textit{G}" := (\texttt{SumRel} \ F \ G).

(b) The Coq definitions for the \(\oplus\) and \(\uplus\) operators.

Fig. 10. Key definitions for implementing composable program adverbs in Coq.

The \(\text{T}\) data type here is equivalent to the non-composable \texttt{ReifiedApp} shown in Fig. 5.

Adverb interpretation can be defined as an algebra of type \texttt{Alg1} \(F E\) (Fig. 10a) where \(F\) is the adverb data type and \(E\) is the instance we are interpreting to. To apply this “interpretation algebra” to the composed “inductive type”, we fold it over \texttt{Fix1} as follows:

\textbf{Definition} \texttt{foldFix1} \ (E A) (alg : \texttt{Alg1} \ F E) (f : \texttt{Fix1} \ F A) : E A := f \_ alg.

We define all composable adverb data types using \texttt{Set} rather than \texttt{Type} because we use the impredicative sets extension in Coq, following MTC. The consequence of this decision is that (1) certain types cannot inhabit \texttt{Set}, and (2) the extension is inconsistent with certain set of axioms such as the axiom of unique choice together with the law of excluded middle.\footnote{\texttt{https://github.com/coq/coq/wiki/Impredicative-Set}} We also develop other mechanisms like the injection type classes, the induction principles following MTC. We omit more detail here due to the space constraint. The interested readers can find them in MTC or our supplementary artifact [Li and Weirich 2022].

Besides MTC, there are other solutions that address the expression problem in theorem provers like Coq. We discuss those alternative solutions in Section 6.

\footnote{\texttt{https://github.com/coq/coq/wiki/Impredicative-Set}}

**Variant** ReifiedPure (K : \text{Set} \rightarrow \text{Set}) (R : \text{Set}) : \text{Set} :=
| Pure (r : R).

**Variant** ReifiedFunctor (K : \text{Set} \rightarrow \text{Set}) (R : \text{Set}) : \text{Set} :=
| FMap (X : \text{Set}) (g : X \rightarrow R) (f : K X).

**Variant** ReifiedApp (K : \text{Set} \rightarrow \text{Set}) (R : \text{Set}) : \text{Set} :=
| LiftA2 (X Y : \text{Set}) (f : X \rightarrow Y \rightarrow R) (g : K X) (a : K Y).

**Variant** ReifiedSelective (K : \text{Set} \rightarrow \text{Set}) (R : \text{Set}) : \text{Set} :=
| SelectBy {X Y : \text{Set}} (f : X \rightarrow ((Y \rightarrow R) + R)) (a : K X) (b : K Y).

**Variant** ReifiedMonad (K : \text{Set} \rightarrow \text{Set}) (R : \text{Set}) : \text{Set} :=
| Bind {X : \text{Set}} (m : K X) (g : X \rightarrow K R).

(a) The composable adverb data types.

**Class** AppKleenePlus (F : \text{Type} \rightarrow \text{Type}) `{\text{Applicative F}} :=
\{ \text{kplus} \{A\} : F A \rightarrow F A \}.

**Class** FunctorPlus (F : \text{Type} \rightarrow \text{Type}) `{\text{Functor F}} :=
\{ \text{plus} \{A\} : F A \rightarrow F A \rightarrow F A \}.

(* The adverb data type for Repeatedly. *)

**Variant** ReifiedKleenePlus (K : \text{Set} \rightarrow \text{Set}) (R : \text{Set}) : \text{Set} :=
| KPlus : K R \rightarrow \text{ReifiedKleenePlus} K R.

(* The adverb data type for Nondeterministically. *)

**Variant** ReifiedPlus (K : \text{Set} \rightarrow \text{Set}) (R : \text{Set}) : \text{Set} :=
| Plus : K R \rightarrow K R \rightarrow \text{ReifiedPlus} K R.

(b) The adverb data types of Nondeterministically and Repeatedly.

Fig. 11. The composable adverb data types and add-on adverb data types.

### 4.3 Add-on Adverbs

Another benefit of making program adverbs composable is that we can now define two add-on adverbs, namely Repeatedly and Nondeterministically, which are not suitable as standalone adverbs. These two adverbs reify two classes of functors, namely AppKleenePlus and FunctorPlus, that we define ourselves. We show these classes of functors and their reifications in Fig. 11b.

AppKleenePlus is a subclass of Applicative and represents the “Kleene plus”.\(^{10}\) It is a “Kleene plus” rather than a “Kleene star” because no empty element is defined. FunctorPlus is similar to the commonly-used Alternative and MonadPlus type classes in Haskell, but contains no empty element and only requires itself to be a subclass of Functor. We define these type classes’ reifications as add-on adverbs so that these adverbs can be composed with classes of functors at different expressive levels: e.g., Repeatedly can be composed with Statically as well as Dynamically.

We show the adverb theories of Repeatedly and Nondeterministically in Fig. 12. Both of these two add-on adverbs are somewhat nondeterministic, so one change we make to their adverb theories is adding refinement relations (⊆) in addition to equivalence relations (≡).

We show that these two adverbs are sound with respect to the following adverb simulations:

```
ReifiedKleenePlus \models_{\text{PowerSet}} \text{AppKleenePlus}
ReifiedPlus \models_{\text{PowerSet}} \text{FunctorPlus}
```

\(^{10}\)https://en.wikipedia.org/wiki/Kleene_star#Kleene_plus
\[
\begin{align*}
\text{REPEAT} & : \forall n, \text{repeat } a n \subseteq kplus a \\
\text{KPLUS} & : \quad a \subseteq kplus b \\
\text{KPLUS} & : \quad kplus a \subseteq kplus b \\
\text{COMMUTATIVITY} & : \quad \text{plus } a \ b \equiv \text{plus } b \ a \\
\text{ASSOCIATIVITY} & : \quad \text{plus } a \ (\text{plus } b \ c) \equiv \text{plus } (\text{plus } a \ b) \ c \\
\text{PLUS} & : \quad a \subseteq c \quad b \subseteq c \\
\text{LEFT PLUS} & : \quad a \subseteq \text{plus } a \ b \\
\text{RIGHT PLUS} & : \quad b \subseteq \text{plus } a \ b
\end{align*}
\]

Fig. 12. The adverb theories for Repeatedly and Nondeterministically. The function repeat \( a \ n \) repeats \( a \) for \( n \) times. Functions \( kplus \) and \( plus \) are smart constructors of \( KPlus \) and \( Plus \), respectively.

(* FunctorPlus transformer. *)

Definition fmapPowerSet \( \{A \to B\} \) (\( a \to \text{PowerSet } I \ A \) \) : PowerSet \( I \ B := \)
\[
\text{fun } r \Rightarrow \text{exists } a', \ a' /\ \text{fmap } f \ a' \equiv r.
\]

Definition plusPowerSet \( \{A \to \text{PowerSet } I \ A \} \) : PowerSet \( I \ A := \)
\[
\text{fun } r \Rightarrow a \ r \ \lor \ b \ r.
\]

(* AppKleenePlus transformer. *)

Definition liftA2PowerSet \( \{A \to B \to C\} \) (\( a \to \text{PowerSet } I \ A \) \) : PowerSet \( I \ A := \)
\[
\text{match } n \text{ with } \begin{array}{c}
| 0 \Rightarrow a \\
| S n \Rightarrow \text{liftA2PowerSet } (\text{fun } x \Rightarrow x) \ a \ (\text{repeatPowerSet } a \ n) \end{array}
\]

Definition kplusPowerSet \( \{A \to \text{PowerSet } I \ A \} \) : PowerSet \( I \ A := \)
\[
\text{fun } r \Rightarrow \text{exists } n, \ \text{repeatPowerSet } a \ n \ r.
\]

Fig. 13. The FunctorPlus transformer instance and the AppKleenePlus transformer instance of the PowerSet data type. \( \equiv \) is the lawful equivalence relation on original functor/applicative functor \( I \).

The definition of PowerSet data type is the same as that in Fig. 8, but we are using its AppKleenePlus transformer and FunctorPlus transformer instances here. The core definitions of these transformers are shown in Fig. 13.

5 EXAMPLES

In this section, we use two different examples to demonstrate the usefulness and different aspects of program adverbs and Tlön embeddings.
5.1 Haxl

In our first example, we show that we can use composable adverbs to capture two different computation patterns in the same library. We also demonstrate interpreting composable adverbs to a shallow embedding in a modular way.

We illustrate these aspects via an example based on the core ideas of Haxl. Haxl is a Haskell library developed and maintained by Meta (formerly known as Facebook) that automatically parallelizes certain operations to achieve better performance [Marlow et al. 2014]. As an example, suppose that we want to fetch data from a database and we have a `Fetch : Type -> Type` data type that encapsulates the fetching effect. The key insight of the Haxl library is to distinguish the operations of `Fetch`'s `Monad` instance and those of its `Applicative` instance. When we use `>>=` to bind two `Fetch`, those data fetches are sequential; but when we use `liftA2` to bind them, those data fetches are batched and will be sent to the database together. To achieve this, it is important that the definition of `liftA2` is not equivalent to the “default” definition derived from `>>=`.

This design of Haxl poses a challenge to mixed embeddings based on freer monads or any other basic adverbs discussed in Section 3, because we need to distinguish when `Applicative` operations are used and when `Monad` operations are used. This is exactly where composable adverbs are useful.

In this example, we assume that we already have a translation from Haxl’s `Applicative` and `Monad` operations to those operations in Coq. In our embedding, we use the following `T` datatype to encode the Tlön embedding of a data fetching program:

\[
T := \text{Fix1} (\text{ReifiedPure} \oplus \text{ReifiedApp} \oplus \text{ReifiedMonad} \oplus \text{DataEff}).
\]

We use `ReifiedApp` to model batched operations and the theory of `StaticallyInParallel` to model their parallel nature. We use `ReifiedMonad` to model sequential operations.

We cannot know statically how many database accesses would happen in a Haxl program, because a program can choose to do different things depending on the result of some data fetch. Therefore, we need to pick an effect interpretation for `DataEff` to reason about this property. In this example, we are assuming that the database does not change, so we interpret our Tlön embedding to a shallow embedding whose semantic domain is the update monad [Ahman and Uustalu 2013].

The key definitions of the update monad are shown in Fig. 14a. The update monad is essentially a combination of a reader monad and a writer monad. In our example, the “reader state” has type `var -> val` which represents an immutable key-value database we can read from. The “writer state” is a `nat`, which represents the accumulated number of database accesses. The `bind` operation propagates the key-value database and accumulates the cost.

Additionally, we define a `liftA2` operation, which only records the maximum number of database accesses in one of its branches. This is not the same as the `liftA2` operation that can be automatically derived from the monad instance of `Update`. Furthermore, this `liftA2` is commutative. Thanks to that, we can interpret `T` to the `Update` datatype without the help of a `PowerSet` transformer.

Figure 14b shows how we interpret composed adverbs in a modular way. First, we define a type class called `AdverbAlg` for interpretation algebras. We then define an interpretation from each individual composable adverb and effect in `T` to `Update`. Finally, the interpretation from `T` to `Update` can be automatically inferred by Coq thanks to the instance `AdverbAlgSum`. If we would like to add another effect or composable adverb to `T`, we only need to add one more instance of `AdverbAlg` and we do not need to modify any existing interpretation algebras.

Interested readers can find the full Coq implementation of the `Update` data type, the `AdverbAlg` type class and relevant instances, along with a few simple examples in our supplementary artifact [Li and Weirich 2022].

---

11 Tools like `hs-to-coq` [Breitner et al. 2021; Spector-Zabuksky et al. 2018] can be adapted to implement the translation.
Definition Update A := ((var -> val) -> A * nat).

Definition ret {A} (a : A) : Update A := fun map => (a, 0).

Definition bind {A B} (m : Update A) (k : A -> Update B) : Update B :=
  fun map => match m map with
    | (i, n) => match (k i map) with
      | (r, n') => (r, n + n')
    end
  end.

Definition liftA2 {A B C} (f : A -> B -> C) (a : Update A) (b : Update B) :
  Update C := fun map => match (a map, b map) with
    | ((a, n1), (b, n2)) => (f a b, max n1 n2)
  end.

Definition get (v : var) : Update val := fun map => (map v, 1).

(a) The Update datatype.

Class AdverbAlg (D : (Set -> Set) -> Set -> Set) (I : Set -> Set) :=
  { adverbAlg : Alg1 D I }.

Instance CostApp : AdverbAlg ReifiedApp Update :=
  { \| adverbAlg := fun d => match d with LiftA2 f a b => liftA2 f a b end \}.}

Instance CostMonad : AdverbAlg ReifiedMonad Update :=
  { \| adverbAlg := fun d => match d with Bind m k => bind m k end \}.

Instance CostPure : AdverbAlg ReifiedPure Update :=
  { \| adverbAlg := fun d => match d with Pure a => ret a end \}.

Instance CostData : AdverbAlg DataEff Update :=
  { \| adverbAlg := fun d => match d with GenData v => get v end \}.

Instance AdverbAlgSum D1 D2 I `{AdverbAlg D1 I} `{AdverbAlg D2 I} :
  AdverbAlg (D1 ⊕ D2) I name :=
  { \| adverbAlg := fun a => match a with
    | Inl1 a => adverbAlg a
    | Inr1 a => adverbAlg a
    end \}.

(b) Interpretation algebras that interpret composable adverbs and DataEff to Update. Thanks to Instance AdverbAlgSum and Coq’s type class inference, we can automatically get the interpretation from T to Update.

Fig. 14. The Update datatype and the interpretation from T to Update.

5.2 A Networked Server

A common technique used in formal verification is dividing the verification into multiple layers and establishing a refinement relation between every two layers [Gu et al. 2018; Koh et al. 2019; Lorch et al. 2020; Zakowski et al. 2021]. This approach offers better abstraction and modularity, as at each layer, we only need to consider certain subsets of properties.

In this example, we show the usefulness of program adverbs and Tlon embeddings in a layered approach. Specifically, we define an intermediate-level specification that omits implementation
Program Adverbs and Tlön Embeddings

1. newconn ::=<- accept ;;
2. IF (not (*newconn == 0)) THEN
3.   newconn_rec ::==
4.   connection *newconn READING ;;
5.   conns ::=++ newconn_rec
6. END ;;
7. FOR y IN conns DO
8.   IF (y->state == WRITING) THEN
9.     r ::=<- write y->id *s ;;
10.    y->state ::= CLOSED
11. END ;;
12. IF (y->state == READING) THEN
13.    r ::=<- read y->id ;;
14.    IF (*r == 0) THEN
15.      y->state ::= CLOSED
16.    ELSE
17.      s ::= *r ;;
18.      y->state ::= WRITING
19.    END
20. END
21. END.

(a) The implementation Impl in NetImp.

```
newconn ::=<- accept ;;
IF (not (*newconn == 0)) THEN
  newconn_rec ::==
  connection *newconn READING ;;
  conns ::=++ newconn_rec
END ;;
FOR y IN conns DO
  IF (y->state == WRITING) THEN
    r ::=<- write y->id *s ;;
    y->state ::= CLOSED
  END ;;
  IF (y->state == READING) THEN
    r ::=<- read y->id ;;
    IF (*r == 0) THEN
      y->state ::= CLOSED
    ELSE
      s ::= *r ;;
      y->state ::= WRITING
    END
  END
END
```

(b) The specification Spec in NetSpec.

```
(Or (newconn ::=<- accept ;;
  IF (not (*newconn == 0)) THEN
    newconn_rec ::==
    connection *newconn READING ;;
    conns ::=++ newconn_rec
  END)
  (OneOf (conns) y
    (Or (IF (y->state == WRITING) THEN
      r ::=<- write y->id *s ;;
      y->state ::= CLOSED
    END)
    (IF (y->state == READING) THEN
      r ::=<- read y->id ;;
      IF (*r == 0) THEN
        y->state ::= CLOSED
      ELSE
        s ::= *r ;;
        y->state ::= WRITING
      END
    END))))
```

Fig. 15. The implementation and the intermediate layer specification of our networked server.

details about execution order, etc. Since the specification is only more nondeterministic in its control flow, we would like our formal verification to show that an implementation refines the specification without interpreting effects to a shallow embedding. This is exactly where program adverbs and Tlön embeddings can help.

We demonstrate this vision above via a simple server adapted from that of Koh et al. [2019]. The server communicates with multiple clients via a network interface. A client initiates a communication with the server by sending a request that is a number. Whenever the server receives a request, it stores the number of that request and sends back a number in its store—a client does not necessarily receive what they sent before, because the server can interleave multiple sessions.

We show that a specific implementation of such a server refines an intermediate-level specification. We also show the refinements based on Tlön embeddings with the help of adverb theories. Unlike Koh et al. [2019], we do not show that the implementation further refines a higher-level specification based on observations over a network, as that is beyond the scope of this work.

The implementation. The server is implemented using a single-process event loop [Pai et al. 1999]. Instead of processing a request and sending back a response immediately, the server divides a session with a client into multiple steps. In each iteration of the event loop, the server advances the session of each request by one step, thus interleaving multiple sessions.

We show the main loop body of our adapted version of the networked server in Fig. 15a. For simplicity, we use a custom language called NetImp. NetImp supports datatypes like booleans, natural numbers, and a special record type called connection. It has network operations like accept, read, and write. All these operations return natural numbers, with 0 indicating failures. The language does not have a while loop but it has a FOR loop that iterates over a list. The loop

variable is implemented as a pointer that points to elements in the list iteratively. We also use C-like notations (i.e., * and ->) for operations on pointers.

The implementation Impl maintains a list of connections called conns. Each connection in conns is in one of the three possible states: READING, WRITING, or CLOSED. At the start of each loop, the server checks if there is a new connection waiting to be established by calling the non-blocking operation accept. If there is, the server creates a new connection with the READING state and adds it to conns. The server then goes over each connection in conns: if a connection is in the READING state, the server tries to read from the connection and updates an internal state s with the recently read value; if a connection is in the WRITING state, the server sends the current value of its internal state s to the connection; once a connection enters the CLOSED state, it remains that state forever and the server will not do anything with it—we design the server in this way for simplicity; a more realistic server should remove closed connections from conns.

The specification. We show our specification Spec in Fig. 15b. Spec is written in a language called NetSpec. NetSpec adds a few additional commands to NetImp: Some is an unary operation that models the "Kleene plus"; Or is a binary operation that models a nondeterministic choice wrapped inside a "Kleene plus"; OneOf is like Or, but it nondeterministically chooses from a list—line 8 means that we nondeterministically assign the variable y with one element from the list in conns.

Spec is more nondeterministic compared with Impl. At each iteration of the event loop, Impl always first tries to accept a connection. It then goes over the list of conns in a fixed order. Spec does not enforce order: an accept could happen immediately after another accept; we can access elements in conns in any order and some connection might get visited more often than others.

Tlön embeddings and the refinement proof. To show that Impl refines Spec, we embed both NetImp and NetSpec in Coq using program adverbs. We use the following datatype:

```
Definition T := Fix1 (ReifiedKleenePlus ⊕ ReifiedPlus ⊕ ReifiedPure ⊕ ReifiedMonad ⊕ NetworkEff ⊕ MemoryEff ⊕ FailEff).
```

We have already seen the first four adverbs. Effect NetworkEff models the effects incurred by network operations accept, read, and write. Effect MemoryEff models the effects incurred by assigning values to variables and retrieving values from them. Finally, effect FailEff models when the program crashes.

We use \( \llbracket \cdot \rrbracket_T \) to denote NetImp’s Tlön embedding in T and \( \llbracket \cdot \rrbracket_T^\text{NetSpec} \) to denote NetSpec’s Tlön embedding in T. For the sake of space, we omit the embeddings here.

We would like to show that \( \llbracket \text{Impl} \rrbracket_T \subseteq \llbracket \text{Spec} \rrbracket_T^\text{NetSpec} \). Recall that \( \subseteq \) is the refinement relation on program adverbs (Section 4.3). The theorem states that the Tlön embedding of our implementation Impl in T refines the Tlön embedding of our specification Spec in T.

To show that, we first observe that Impl and Spec share some common program fragments, e.g., lines 1–6 of Impl are the same as lines 2–7 of Spec. Indeed, there are three such common

---

**Fig. 16.** Program L1 written in NetImp, and programs L2 and L3 written in NetSpec.
fragments and we name them $A$ (lines 1–6 of $\text{Impl}$), $B$ (lines 8–11 of $\text{Impl}$), and $C$ (lines 12–20 of $\text{Impl}$), respectively. We then define three programs $L_1$, $L_2$, and $L_3$ shown in Fig. 16. These programs represent some intermediate layers between $\text{Impl}$ and $\text{Spec}$. We prove the following theorem:

**Theorem 5.1.** $[\text{Impl}]^T_T \subseteq [L_1]^T_T \subseteq [L_2]^T_T \subseteq [L_3]^T_T \subseteq [\text{Spec}]^T_T$.

**Proof.** We show $[\text{Impl}]^T_T \subseteq [L_1]^T_T$ by associativity of $\text{Dynamically}$. Both $[L_1]^T_T \subseteq [L_2]^T_T$ and $[L_2]^T_T \subseteq [L_3]^T_T$ can be proven by an induction over $\text{conns}$ and with the help of theories of $\text{Dynamically}$, $\text{Repeatedly}$, and $\text{Nondeterministically}$. Finally, we prove $[L_3]^T_T \subseteq [\text{Spec}]^T_T$ by the theories of $\text{Dynamically}$, $\text{Repeatedly}$, and $\text{Nondeterministically}$. □

Interested readers can find the full Coq implementation of $\text{NetImp}$, $\text{NetSpec}$, the Tlön embeddings of these two languages, the implementation $\text{Impl}$, the specification $\text{Spec}$, as well as the full proof of Theorem 5.1 in our supplementary artifact [Li and Weirich 2022].

6 DISCUSSION

The expression problem. The composable program adverbs require extensible inductive types. We implement this feature in Coq by using the Church encodings of datatypes, following the precedent work of MTC [Delaware et al. 2013]. There are several consequences of using Church encodings instead of Coq’s original inductive datatypes.

First, we cannot make use of Coq’s language mechanisms, libraries, and plugins that make use of Coq’s inductive types (e.g., Coq’s built-in induction principle generator, the Equations plugin [Sozeau and Mangin 2019], the QuickChick plugin [Lampropoulos et al. 2018; Paraskevopoulou et al. 2022], etc.). Furthermore, the extra implementation overheads incurred by Church encodings (e.g., proving an algebra is a functor, proving the induction principle using dependent types, etc.) can be huge. However, this situation can be helped by developing tools or plugins for supporting Church encodings.

The other consequence is that, following the practice of MTC, we use Coq’s impredicative set extension. This causes two problems: (1) Certain types cannot inhabit $\texttt{Set}$, and (2) our Coq development is inconsistent with certain set of axioms such as the axiom of unique choice together with the law of excluded middle, as we have discussed in Section 4.2.

There are alternative methods for addressing the expression problem. One option is the meta-programming approach proposed by Forster and Stark [2020]. In this approach, we can define each composable adverb separately in a meta language and use a language plugin to generate a combined definition in Coq. This approach does not fully address the expression problem as extending the combined definition requires recompilation—but the amount of code that needs to be recompiled is much smaller and the generated code uses Coq’s built-in inductive types. Another option that has recently been explored by Kravchuk-Kirilyuk et al. [2021] is adding family polymorphism [Ernst 2001] to theorem provers. These works are promising. Unfortunately, they either lack mature tool support or is still in development at the moment. We would like to explore these approaches in the future and composable program adverbs might provide a good application to these approaches.

Reified vs. free structures. Even though the reified structures used in adverb data types are free structures, they are different from those free structures present in Capriotti and Kaposi [2014]; Kiselyov and Ishii [2015]; Mokhov [2019]; Mokhov et al. [2019]. The biggest difference between reified structures and these free structures are the parameters they recurse on: all the reified structures recurse on both their computational parameters, while each free structure only recurses on one of them.\footnote{With the exception of reified/free functors, since each of them has only one computational parameters to be recursed on.} For example, comparing $\text{FreerMonad}$ in Fig. 2 and $\text{ReifiedMonad}$ in
Fig. 5: FreerMonad only recurses on the parameter $k$ of Bind, while ReifiedMonad recurses on both parameters $m$ and $k$. This means that a free structure does not just reify a class of functors, it also converts the reification to a left- or right-associative normal form.

One advantage of the normal forms in free structure definitions is that the type class laws can be automatically derived from definitional equality (with the help of the axiom of functional extensionality). However, this conversion would eliminate some differences in the syntax. Taking ReifiedApp as an example, normalizing it would result in a “list” rather than a “binary tree”, making analyzing the depth of the tree impossible. Preserving the original tree structure of StaticallyInParallel also plays a crucial role in our examples shown in Section 2.4 and 5.1.

7 RELATED WORK

Semantic embeddings. There are various works that study different semantic embeddings. Boulton et al. [1992] are the pioneers who coined terms such as semantic embeddings, shallow embeddings, and deep embeddings. It is known that there are many styles of embeddings between shallow and deep embeddings, but there is not an agreed term on describing them. In this paper, we use the term mixed embeddings, which is borrowed from Chlipala [2021], where it is used to describe an embedding based on freer monads. Another term deeper shallow embeddings is proposed by Prinz et al. [2022], which shows a way of deepening any shallow embedding.

Freer Monads and Variants. Freer monads [Kiselyov and Ishii 2015] and their variants are studied by many researchers in formal verification to reason about programs with effects. Earlier work includes the study of the delay monad [Capretta 2005] and resumption monads [Piróg and Gibbons 2014]. More recent work includes Letan et al. [2021], where the authors use free monads to develop a modular verification framework based on effects and effect handlers called FreeSpec. Christiansen et al. [2019] develop a framework based on free monads and containers [Abbott et al. 2003] for reasoning about Haskell programs with effects. Swierstra and Baanen [2019] interpret free monads into a predicate transformer semantics that is similar to Dijkstra monads; Nigron and Dagand [2021] interprets free monads using separation logic.

On the coinductive side, Xia et al. [2020] develop a coinductive variant of freer monads called interaction trees that can be used to reason about general recursions and nonterminating programs. Koh et al. [2019] encode interaction trees in VST [Appel et al. 2014] to reason about networked servers. Mansky et al. [2020] use interaction trees as a lingua franca to interface and compose higher-order separation logic in VST and a first-order verified operating system called CertiKOS [Gu et al. 2019]. Zakowski et al. [2020] propose a technique called generalized parameterized coinduction for developing equational theory for reasoning about interaction trees. Zakowski et al. [2021] use interaction trees to define a modular, compositional, and executable semantics for LLVM. Yoon et al. [2022] further extend the modularity of interaction trees by extending them with layered monadic interpreters. Silver and Zdancewic [2021] connect interaction trees with Dijkstra monads [Maillard et al. 2017] for writing termination sensitive specifications based on uninterpreted effects. Lesani et al. [2022] use interaction trees to verify transactional objects. Foster et al. [2021] apply interaction trees to Isabelle/HOL to produce a verification and simulation framework for state-rich process languages, which is used by Ye et al. to give an operational semantics to RoboChart, a timed and probabilistic domain-specific language for robotics [Ye et al. 2022].

Among many variants of freer monads, one particular structure closely resembles program adverbs. That is the action trees defined in Swamy et al. [2020]. The action trees have four constructors, Act, Ret, Par, and Bind, whose types correspond to effects, ReifiedPure, ReifiedApp, and ReifiedMonad in composable program adverbs, respectively, another evidence that program adverbs are general models. In contrast to our work, compositionality and extensibility of “adverbs” are not
the main issue action trees try to address, so action trees are not built in a composable way. On the other hand, action trees are embedded with separation logic assertions, which are not the focus of program adverbs or Tlön embeddings.

Other Free Structures. Other free structures are also explored by various works. Capriotti and Kaposi [2014] propose two variants of freer applicative functors, which correspond to the left- and right-associative variants, respectively. Xia [2019] explores defining freer applicative functors in Coq, and points out that the right associative variant is harder to define in Coq. Milewski [2018] discusses deriving free monoidal functors. Mokhov [2019] defines the free selective functors.

Programming Abstractions. We are not the first to observe that monads are too dynamic for certain applications. For example, Swierstra and Duponcheel [1996] identify that a parser that has some static features cannot be defined as a monad. Inspired by their observation, Hughes [2000] proposes a new abstract interface called arrows. The relationship among arrows, applicative functors, monads are studied by Lindley et al. [2011]. Willis et al. [2020] observe that monads generate dynamic structures that are hard to optimize. They further show that, by using applicative and selective functors instead, it is possible to implement staged parser combinators that generate efficient parsers. Mokhov et al. [2020] observe that the datatype of tasks in a build system can be parameterized by a class constraint to describe various kinds of build tasks.

8 CONCLUSION

In this paper, we compare different styles of semantic embeddings and how they impact formal reasoning about programs with effects. We find that, if used properly, mixed embeddings can combine benefits of both shallow and deep embeddings, and be effective in (1) preserving syntactic structures of original programs, (2) showing general properties that can be proved without assumptions over external environment, and (3) reasoning about properties in specialized semantic domains.

We propose program adverbs and Tlön embeddings, a class of structures and a style of mixed embeddings based on these structures, that enable us to reap these benefits. Like free monads, program adverbs embed pure computations shallowly and effects deeply (and abstractly, but can later be interpreted). However, various program adverbs correspond to alternative computation patterns, and can be composed to model programs with multiple characteristics.

Based on program adverbs, Tlön embeddings cover a wide range of programs and allow us to reason about syntactic properties, semantic properties, and general semantic properties with no assumption over external environment within the same embedding.

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REFERENCES


Draft.


Email correspondence.


Blog post.


