Ghostbuster: A Tool for Simplifying and Converting GADTs

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Abstract

Generalized Algebraic Dataypes, or simply GADTs, can encode non-trivial properties in the types of the constructors. Once such properties are encoded in a datatype, however, all code manipulating that datatype must provide proof that it maintains these properties in order to typecheck. In this paper, we take a step towards gradualizing these obligations. We introduce a tool, Ghostbuster, that produces simplified versions of GADTs which elide selected type parameters, thereby weakening the guarantees of the simplified datatype in exchange for reducing the obligations necessary to manipulate it. Like ornaments, these simplified datatypes preserve the recursive structure of the original, but unlike ornaments we focus on information-preserving bidirectional transformations. Ghostbuster generates type-safe conversion functions between the original and simplified datatypes, which we prove are the identity function when composed. We evaluate a prototype tool for Haskell against thousands of GADTs found on the Hackage package database, generating simpler Haskell'98 datatypes and round-trip conversion functions between the two.

Categories and Subject Descriptors D.1.1 [*Programming Techniques*]: Applicative (Functional) Programming

Keywords GADTs, Datatypes, Haskell

1. Introduction

Languages in the Haskell, OCaml, Agda, and Idris traditions can encode complicated invariants in datatype definitions. This introduces safety at the cost of complexity. For example, consider the standard GADT (generalized algebraic datatype) formulation of length-indexed lists:

```
data Vec a n where
VNil :: Vec a Zero
VCons :: a \rightarrow Vec a n \rightarrow Vec a (Succ n)
```

Although this datatype provides additional static guarantees—for example, that we cannot take the head of an empty list—writing functions against this type necessarily involves additional work to manage the indexed length type n. In some situations however, such as when prototyping a new algorithm, the user may prefer to *delay*

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ICFP'16, September 18–24, 2016, Nara, Japan ACM. 978-1-4503-4219-3/16/09... http://dx.doi.org/10.1145/2951913.2951914 the effort required to fulfill these type obligations until they can verify that the new algorithm is beneficial. In that case, we can convert the length-indexed list into a regular list by *erasing* the length index n and operating over a simplified representation:

data Vec' a where Nil' :: Vec' a Cons' :: a \rightarrow Vec' a \rightarrow Vec' a

before converting back to the original datatype in order to test the changes within the larger code base.

However, this final step requires re-establishing the type-level invariants that were encoded in the original datatype, which may not be straightforward. Perhaps the user should stick to regular ADTs for this project? Unfortunately, that too may not be an option. In the 16,183,864 lines of public Haskell code we surveyed, we found 11,213 existing GADTs. A person tasked with working in an *existing* project is unlikely to be able to reimplement all of a project's datatypes and operations on them from scratch.

Inspired by the theory of *ornaments* [10, 15], we can think about moving between families of related datatypes that have the same recursive structure: rather than always working with a GADT, a user could choose to (initially) write code against a simpler datatype, while still having it seamlessly interoperate with code using the fancier one. A practical tool to do this could enable a *gradual* approach to discharging obligations of indexed datatypes. In this paper, we present such a tool. We require that it (1) define canonical simplified datatypes; and (2) create conversion functions between the original and simplified representations.

Is it possible to define such a simplification strategy by merely choosing which type indices to remove from a datatype? While such a method would be convenient for the user, it is far from obvious that there exists a class of datatypes for which such an erasure selection yields a canonical simplified datatype *and* guarantees that conversion functions can successfully round-trip all values of the GADT through the simplified representation and back.

In this work, we show how to do exactly that. Using our tool named *Ghostbuster*—the user simply places the following pragma above the definition of Vec:

{-# Ghostbuster: synthesize n #-}

and Ghostbuster will generate the definition Vec' as well as conversion functions between the two representations:

```
upVec :: Typeable n \Rightarrow Vec a n \rightarrow Vec' a downVec :: Typeable n \Rightarrow Vec' a \rightarrow Maybe (Vec a n)
```

Since downVec may fail at runtime if the actual size of the vector does not match the expected size as specified (in the type n and checked at runtime via Typeable) by the caller, we return the result wrapped in Maybe, but we could also choose to throw an error on failure or return a diagnostic message using Either. If we do not know the specific type index that must be *synthesized* during the conversion process, we can keep it *sealed* under an existential binding and still make use of it in contexts that operate over any instance of the sealed type:

```
data SealedVec a where
SealedVec :: Typeable n \Rightarrow Vec a n \rightarrow SealedVec a
downVecS :: Vec' a \rightarrow SealedVec a
withVecS :: SealedVec a \rightarrow (\forall n. \text{ Vec a } n \rightarrow b) \rightarrow b
```

Assuming we had such functionality, would that truly make our lives any easier, or have we just moved our type-checking responsibilities elsewhere? We will show that manipulating these simplified—or ghostbusted—datatypes is not at all burdensome, and can indeed make life simpler. As an example, consider implementing deserialization for our indexed list. With Haskell'98 datatypes such as Vec', a Read instance can be derived automatically, but an attempt to do so with the Vec GADT results in a cryptic error message mentioning symbols and type variables only present in the compiler-generated code. Disaster! On the other hand, by leveraging Ghostbuster we can achieve this almost trivially:¹

In this paper we scale up the above type-index erasure approach to handle a large number of datatypes automatically. We make the following contributions.

- We introduce the first practical solution to incrementalize the engineering costs associated with GADTs.
- We give an algorithm for deleting *any* type variable that meets a set of non-ambiguity criteria. Our ambiguity criteria establish a *gradual erasure guarantee*: if a multi-variable erasure is valid, then any subset of these variables also forms a valid erasure (Section 5).
- We formalize the algorithm in the context of a core language. We show that up-conversion functions are total and up-then-down is exactly the identity function on all values in the original GADT (Section 6).
- We show how the encoding of dynamically typed values that emerges from the algorithm can be asymptotically more efficient than a traditional type Dynamic (Section 2.2).
- Viewed in the context of the literature on deriving type class instances for datatypes, Ghostbuster increases the reach of deriving capabilities beyond previous functional language implementations, by lifting derivations on simpler types to fancier ones, as with Read above (Section 3).
- We describe the Ghostbuster tool, currently implemented as a source-to-source translator for Haskell, but directly generalizable to other languages. We evaluate the runtime performance of Ghostbuster conversions compared to the ad-hoc approach to constructing GADTs using a runtime eval, and apply it to existing datatypes in all 9026 packages on the Hackage Haskell package server (Section 8).

Although our approach does not handle all datatypes or Haskell features, it clearly delineates the class of valid erasures and lays the groundwork for future research.

2. Design Constraints

The central facility provided by Ghostbuster is a method to allow users to select a subset of type variables of a given GADT, from which we derive a new datatype that does not contain those type variables—they have been *erased* from the datatype. Furthermore, we generate an *up-conversion* function from the original datatype to the newly generated one, as well as an *down-conversion* function from the simplified type back to the original, re-establishing typelevel invariants as necessary.

Before jumping into the details of how this is implemented, we first highlight some of the different problems that can occur when attempting to erase a type variable from a GADT, which will give us some intuition on the design constraints and behavior of Ghostbuster. Section 3 explores a larger example in more detail.

2.1 Prerequisite: Testing Types at Runtime

Ghostbuster blurs the line between having a statically-typed and dynamically-checked program. With Ghostbuster, we can explicitly remove type-level information in one part of the program (upconversion), which we then re-establish at some later point (downconversion). To accomplish this, a central requirement for Ghostbuster is the ability to examine types at runtime and to take action based on those tests. Haskell has supported (open-world) type representations for years via the Typeable class:

```
class Typeable a where --GHC-7.10 (current)
typeRep :: proxy a \rightarrow TypeRep
```

However, this is insufficient for our purposes because examining a TypeRep value gives us no type-level information about the type that value represents. Instead, we require a *type-indexed* type representation, which makes the connection between the two visible to the type system:

```
class Typeable a where ----GHC-8.2
typeRep :: TypeRep a
```

This type-indexed representation is currently in development for GHC and is scheduled for release as part of GHC-8.2.² We assume the presence of this new design throughout the paper, although until this new Typeable design becomes available in GHC proper we generate these type-indexed TypeRep values ourselves.^{3,4}

We can then use the following functions to compare two types and gain type-level information when those types are equal:

```
eqT :: (Typeable a, Typeable b) \Rightarrow Maybe (a :~: b)
eqTT :: TypeRep a \rightarrow TypeRep b \rightarrow Maybe (a :~: b)
data a :~: b where
```

Refl :: a :~: a

2.2 Erasure Method: Checked versus Synthesized

The basic operation that we provide to users is the ability to erase type variables from a GADT. However, there are restrictions on which type variables are valid erasure candidates. Consider the standard list:

⁴ Additionally, the use of embedded TypeRep *values* rather than embedded Typeable class constraints simplifies our core language (Section 4).

¹Admittedly, this instance would be improved if the constructors of our simplified datatype used the exact same names as the original, but we append an apostrophe to constructor and type names as a convention to clearly distinguish the generated, simplified datatypes.

²https://ghc.haskell.org/trac/ghc/wiki/Typeable

³ For simplicity our local TypeRep and Typeable definitions do make a closed-world assumption, but since we require only the interface we have shown here, and as will appear in GHC-8.2 for an *open* world of types, there is no loss of generality.

If we remove the type parameter a and attempt to *synthesize* it when converting back to the original datatype, we will find that it is not possible to write this down-conversion function. In contrast to our initial Vec example (Section 1), if we remove the information about the type of the list elements, we cannot later infer that information based solely on the recursive structure of the list.

For this reason, we allow a second, weaker form of type index erasure. Given the declaration

{-# Ghostbuster: check a #-}

Ghostbuster will generate the following simplified representation of List together with its conversion functions:

```
data List' where

Nil' :: List'

Cons' :: \forall a. TypeRep a \rightarrow a \rightarrow List' \rightarrow List'

upList :: Typeable a \Rightarrow List a \rightarrow List'

downList :: Typeable a \Rightarrow List' \rightarrow Maybe (List a)
```

In contrast to Vec', where the erased type was synthesized during down-conversion, when erasing type variables in checked mode we must *embed* a representation of the type directly into the constructor Cons', otherwise this information will be lost. We refer to the type parameter a as *newly existential*, as it was not existentially quantified in the original datatype fed to Ghostbuster. It is *only* newly existential type variables that require an explicit type representation to be embedded within the simplified datatype. This is important, as we surely do not want the user to have to create and manipulate TypeRep values for all erased parameters.

During down-conversion, we check that each element of the list does indeed have the same type the user expects:

Compare this to the definition of down-conversion for our original Vec datatype, which erased its type-indexed length parameter in synthesized mode:

This highlights the key difference between erasures in checked versus synthesized mode. In order to perform down-conversion on List' we must examine the type of each element and compare it to the type that we expect; thus, we can not create a SealedList which hides the type of the elements, since we would not know what type to compare against in order to perform the conversion. In contrast, down-conversion for Vec' does *not* need to know a priori what the type n should be; only if we wish to open the SealedVec do we need to check (via Data.Typeable.gcast) that the type that was synthesized is indeed the type we anticipate.

Connection to dynamic typing We note that this embedded type representation essentially makes each list element a value of type Dynamic. Why then do we use explicit, *unbundled* type representations when Dynamic has existed in Haskell for years? For the List type above, we would perform the same O(n) number of runtime

type checks with either approach, but consider the following list-oflists datatype:

```
data LL a where
NilL :: LL a
ConsL :: [a] \rightarrow LL a \rightarrow LL a
```

These two competing approaches would each yield the following simplified types for ConsL:

Thus, during down-conversion the former would require a runtime type check on every element of the *inner* list, whereas our unbundled representation requires only a single check for each element of the *outer* list—an improvement in asymptotic efficiency. This is one reason that we design Ghostbuster to inject explicit type representations using TypeRep.

Finally, this observation suggests an appealing connection to gradual typing—when Ghostbusted, data structures that were refined by type indexing become regular, parametrically polymorphic data structures, which in turn become dynamic datatypes once all type parameters are erased.

2.3 Unrecoverable Information

Consider the following definition of a strange binary tree:

Here, only the rightmost leaf of the tree is usable, since every leftward branch is of some unknown, unusable type y. According to our policy of embedding an explicit type representation for any newly-existential types (Section 2.2) we will add a TypeRep to the Leaf constructor to record the erased type x. However, what type representation do we select for y? Since this type is already unknowable in the original structure we cannot possibly construct its type representation, so such erasures are not supported.

2.4 A Policy for Allowed Erasures

As we saw in Section 2.2, the defining characteristic of which mode a type variable can be erased in is determined by whether the erased information can be recovered from what other information remains. As a more complex example (which we explore further in Section 3) consider the application case for an expression language:

```
{-# Ghostbuster: check env, synthesize ans #-}
data Exp env ans where
App :: Exp e (a \rightarrow b) \rightarrow Exp e a \rightarrow Exp e b
```

Why does the type variable a, which is existentially quantified, not cause a problem? It is because a is a *pre-existing* existential type (not made existential by a Ghostbuster erasure). The type a can be synthesized by recursively processing fields of the constructor, unlike the Bad example above. Thus, we will not need to embed a type representation so long as we can similarly rediscover in the simplified datatype the erased type information at runtime. This is an information-flow criterion that has to do with how the types of the fields in the data constructor constrain each other.

Checked mode: right to left In the App constructor, because the env type variable is erased in checked mode, its type representation forms an *input* to the downExp down-conversion function. This means that since we know the type e of the result Exp e b (on the right), we must be able to determine the e in the fields to the left, namely in Exp e a and Exp e ($a \rightarrow b$). Operationally, this makes sense if we think how the downExp function must call itself on each of the fields of the constructor, passing the (same) representation for the type e to each recursive call.

Synthesized mode: left to right Conversely, the type ans forms part of the *output* of the down-conversion process, since this type is synthesized by downExp, and we only check after the conversion that the generated type is the type that we anticipate. This means that the recursive calls on the fields of the constructor will generate the types $(a \rightarrow b)$ and a from the left, which in turn are used to determine the output type b on the right.

Fortunately, whether or not type variables a and b can be determined by examining the other types in the constructor is a purely *local* check that can be determined in isolation on a perconstructor/per-datatype basis.⁵ The same local reasoning holds for the requirements on checked types as well as synthesized. We formalize these information flow checks in Section 5.

Erased types that escape Ghostbuster performs one final check before declaring an erasure is valid: datatypes undergoing erasure can only be used directly in the fields of a constructor, not as arguments to other type constructors. For example, what should the behavior be if we attempt to erase the type variable a in the following:

```
data T a where
MkT :: [T a] \rightarrow T a
```

We might expect a sufficiently clever implementation to notice that it can utilize the Functor instance to apply up- and down-conversion to each element of the list. But what if the type constructor does not have a Functor instance, or is only exported abstractly, thereby prohibiting further analysis? This appears to be a tricky design space, so our current implementation simply rules out these erasures altogether.

3. Life with Ghostbuster

In this section, we describe several scenarios in which Ghostbuster can make life easier, taking as a running example the simple expression language which we define below.

3.1 A Type-safe Expression Language

Implementing type-safe abstract syntax trees (ASTs) is perhaps the most common application of GADTs. Consider the following language representation:⁶

```
data Exp env ans where

Con :: Int \rightarrow Exp e Int

Add :: Exp e Int \rightarrow Exp e Int \rightarrow Exp e Int

Var :: Idx e a \rightarrow Exp e a

Abs :: Typ a \rightarrow Exp (e, a) b \rightarrow Exp e (a \rightarrow b)

App :: Exp e (a \rightarrow b) \rightarrow Exp e a \rightarrow Exp e b
```

Each constructor of the GADT corresponds to a term in our language, and the types of the constructors encode both the type that that term evaluates to (ans) as well as the type and scope of variables in the environment (env). This language representation enables the developer to implement an interpreter or compiler which will statically rule out any ill-typed programs and evaluations. For example, it is impossible to express a program in this language which attempts to Add two functions.

Handling variable references is an especially tricky aspect for this style of encoding. We use typed de Bruijn indices (Idx) to project a type t out of a type level environment env, which ensures that bound variables are used at the correct type [3].



Figure 1. In this scenario, we wish to add a prototype transformation into a compiler that uses sophisticated types, but against a simpler representation. For example, we may want to verify that an optimization does indeed improve performance, before tackling the type-preservation requirements of the GADT representation.

Finally, our tiny language has a simple closed world of types Typ, containing Int and (\rightarrow) .

Using GADTs to encode invariants of our language (above) into the type system of the host language it is written in (Haskell) amounts to the static verification of these invariants every time we run the Haskell type checker. Furthermore, researchers have shown that this representation does indeed scale to realistically sized compilers: Accelerate [8, 16, 17] is an embedded language in Haskell for array programming which includes optimizations and code generation all written against a GADT AST, maintaining these well-typed invariants through the entire compiler pipeline.

Where then does the approach run into trouble? The problem is that manipulating this representation requires the developer to discharge a (potentially non-trivial) proof to the type system that all of these invariants are maintained. As such, the programmer's time may be spent searching for a type-preserving formulation of their algorithm, rather than working on the algorithm itself. While ultimately such effort is justified in that it rules out entire classes of bugs from the compiler, we question whether or not this effort should be required *up front*, and wonder if, without this extra initial burden, other optimizations or language features might have been implemented by the Accelerate authors and external contributors over the life of the project so far.

In the next section, we discuss how Ghostbuster can be used to realize the situation shown in Figure 1, where we wish to implement a *prototype* transformation over our expression language, without needing to discharge all of the typing obligations up front. Of course, other alternatives exist:

Competing approach #1: hand-written conversions Rather than using a tool such as Ghostbuster, a user could just as well build the same conversion functions to and from a less strictly typed AST representation themselves. Indeed, Ghostbuster itself is written entirely in Haskell, and no modifications to the GHC compiler were required to support it. However, this introduces a maintenance burden. Moreover, these conversion functions are tricky to implement, and since the Haskell type checker cannot stop us from writing ill-typed conversions to or from our untyped representation, these errors will only be caught when the runtime type tests fail.

Competing approach #2: runtime eval Another approach is to avoid the fine-grained runtime type checks necessary for down-conversion entirely, by generating the GADT term we require as a *string*, and using GHC embedded as a library in our program to typecheck (eval) the string at runtime. Implementing a pretty-printer is arguably less complex than the method we advocate in this work, but there are several significant disadvantages to this approach which we will demonstrate in Section 8.

⁵ This is more local than other (tangentially related) features such as the "." notation in Idris [5] and Agda, which signifies a type is runtime-irrelevant and should be erased during compilation. Irrelevance requires a whole-program check to verify whether the annotation can be fulfilled.

⁶https://github.com/shayan-najd/MiniFeldspar

```
class Syntactic f where
  \texttt{varIn} \quad :: \; \texttt{Idx} \; \texttt{env} \; \texttt{t} \; \rightarrow \; \texttt{f} \; \texttt{env} \; \texttt{t}
  expOut :: f env t \rightarrow f env t
  weaken :: f env t \rightarrow f (env, s) t
instance Syntactic Idx
instance Syntactic Exp
shift :: Syntactic f
        \Rightarrow (\dot{\forall} t'. Idx env t' \rightarrow f env' t')
        \rightarrow Idx (env, s) t
        \rightarrow f (env', s) t
shift _ ZeroIdx
                           = varIn ZeroIdx
shift v (SuccIdx ix) = weaken (v ix)
rebuild :: Syntactic f
           \Rightarrow (\forall t'. Idx env t' \rightarrow f env' t')
          \rightarrow Exp env t
           \rightarrow Exp env' t
rebuild v exp =
  case exp of
     Var ix \rightarrow expOut (v ix)
     Abs t e \rightarrow Abs t (rebuild (shift v) e)
     . . .
substitute :: Exp (env, s) t \rightarrow Exp env s \rightarrow Exp env t
substitute old new = rebuild (subTop new) old
  where
     subTop :: Exp env s \rightarrow Idx (env, s) t \rightarrow Exp env t
     subTop = ...
```

Listing 1. Substitution algorithm for richly-typed terms

3.2 Example #1: Substitution

Consider the task of inlining a term into all use sites of a free variable. For our richly-typed expression language, where the types of terms track both the type of the result as well as the type and scope of free variables, this requires a type-preserving but environment changing value-level substitution algorithm. Luckily, the simultaneous substitution method of McBride [14] provides exactly that, where renaming and substitution are instances of a single traversal, propagating operations on variables structurally through terms. Listing 1 outlines the method.

Although the simultaneous substitution algorithm is very elegant, we suspect that significant creativity was required to come up with it. Compare this to the implementation shown in Listing 2, which is just a simple structural recursion on terms. In particular, this is implemented against the simplified representation generated by Ghostbuster using the erasure pragma:^{7,8}

{-# Ghostbuster: check env, synthesize ans #-}

which yields the following expression datatype:

data Exp'		where				
Con'	::	Int \rightarrow Exp'				
Add'	::	$Exp' \rightarrow Exp' \rightarrow Exp'$				
Mul'	::	$Exp' \rightarrow Exp' \rightarrow Exp'$				
Var'	::	$Idx' \rightarrow Exp'$				
Abs'	::	$Typ' \rightarrow Exp' \rightarrow Exp'$				
App'	::	$Exp' \rightarrow Exp' \rightarrow Exp'$				

and conversion functions:

⁸ We simultaneously request erased versions of Idx and Typ using the same settings, but elide those for brevity.

```
shift :: Idx' \rightarrow Exp' \rightarrow Exp'
shift j exp =
  case exp of
     Var' ix | ix < j \rightarrow Var' ix
               otherwise \rightarrow Var' (SuccIdx' ix)
     Abs' t e \rightarrow Abs' t (shift (SuccIdx' j) e)
substitute :: Exp' \rightarrow Exp' \rightarrow Exp'
substitute = go ZeroIdx'
  where
     go j old new =
       case old of
          Var' ix | ix == j
                                                      \rightarrow new
                     | ix > j, SuccIdx' i \leftarrow ix \rightarrow Var' i
                    | ix < j
                                                      \rightarrow old
          Abs' t e \rightarrow Abs' t (go (SuccIdx' j) e
                                       (shift ZeroIdx' new))
```

Listing 2. Substitution algorithm implemented against the simplified datatype generated by Ghostbuster

```
upExp :: (Typeable env, Typeable t)

⇒ Exp env t → Exp'

downExp :: (Typeable env, Typeable t)

⇒ Exp' → Maybe (Exp env t)
```

Referring to the implementation of Listing 2, note that although the Var' and Abs' cases constitute environment changing operations, we do *not* need to manipulate any embedded TypeRep env values; needing to do so would seriously compromise usability, and Ghostbuster is instead able to recover this information automatically (see Sections 2.2 and 2.4).

3.3 Example #2: Template Haskell and Typeclass Deriving

One great feature of Haskell is its ability to automatically derive certain standard typeclass instances such as Show and Read for Haskell'98 datatypes. Unfortunately, attempting to do the same for GADTs results only in disappointment and cryptic error messages from compiler-generated code. However, as we saw in Section 1, we can regain this capability by using Ghostbuster and leveraging derived instances for the simplified datatypes instead.

Similarly, some libraries include Template Haskell [22] routines that can be used to automatically generate instances for the typeclasses of that library. Although these run into problems when applied to GADTs, once more we can use Ghostbuster to circumvent this limitation. As an example, we can easily generate JSON (de)serialization instances for the aeson package⁹ applied to our richly-typed terms:

\$(deriveJSON defaultOptions ''Exp')

```
instance (...) ⇒ ToJSON (Exp env t) where
toJSON = toJSON . upExp
instance (...) ⇒ FromJSON (Exp env t) where
parseJSON v = do
 v' ← parseJSON v :: Parser Exp'
return $ fromMaybe (error "...") (downExp v')
```

These examples demonstrate that Ghostbuster enables a *synergy* with existing Haskell libraries and deriving mechanisms, providing a convenient method to lift these operations to GADTs.

⁷ The environment type env needs to be provided by the client (checked mode) because otherwise it is ambiguous. For example, the constant term Con 42 can be typed in any environment.

⁹https://hackage.haskell.org/package/aeson



Figure 2. The architecture of the Ghostbuster tool, which processes data definitions in several passes, resulting in pretty-printed Haskell source on disk. Note that only the ingestion and code generation phases are Haskell specific: the ambiguity check through lowering phases are implemented in terms of our core language.

4. Core Language Definition

Before covering host-language-specific aspects of our Ghostbuster implementation, we first formalize a core language to facilitate the precise description of the transformations performed by the Ghostbuster tool. This core language also serves as the intermediate representation of the Ghostbuster implementation. Although we implement our prototype in Haskell, it is easily extended to generate code for any language that supports GADTs.

The core language definition is given in Figure 3. The input to Ghostbuster is a set of datatype definitions, $dd_1 \dots dd_n$. The term language is used only as an *output language* for generating upand down-conversion functions. As such, we are not interested in the problem of type inference for GADTs, rather we assume type annotations that allow us to use the permissive, natural type system for GADTs [21], which supports decidable checking [9, 24] (but not inference). Our implementation runs a checker for this type system, and, to support checking, case and typecase forms are labelled with their return types as well, though we will elide these in the code through the rest of the paper.

4.1 Syntax

The syntax of terms and types in Figure 3 resembles Haskell syntax with extensions for type representation handling and extra conventions related to type arguments ($\overline{k} \ \overline{c} \ \overline{s}$) to indicate the erasure level (respectively, type variables which are kept unchanged in the output, and those which are erased in checked and synthesized mode, as discussed in Section 2.2). Without loss of generality, we assume that type constructor arguments are *sorted* into these kept, checked, and synthesized categories. This simplifies the discussion of which type arguments occur in which *contexts*, based on position. The implemented Ghostbuster tool does not have this restriction and the status of type arguments are specified in pragmas, as we saw earlier.

A program consists of a number of datatype declarations followed by mutually-recursive value definitions (vd) and a "main" term e. The generated up- and down-conversions will form a series of vds. Terms in our language consist of the lambda calculus, a nonrecursive let with explicit type signatures, simple case expressions and ways of creating, casing on, and querying equality of runtime type representations, which we call (typerep, typecase, and \simeq_{τ}). The \simeq_{τ} operator must work over arbitrary monotype representations, comparing them for equality at runtime. typecase also performs runtime tests on type representations, and enables *deconstructing* type representations into their component parts—for example, splitting a function type into a input type and output type.

We deviate from the standard presentation of GADTs. Typically, the return type of each constructor is normalized to the form $T \bar{a}$,

Programs and da	atatyp	be dec	clarations	5	
prog ::	=	dd_1 .	$\ldots dd_n; u$	$d_1 \dots$	$vd_m; e$
dd ::	=	data	$T \ \overline{k} \ \overline{c} \ \overline{s}$	where	
		v	$\forall \overline{k} \overline{a} \overline{a}$	\overline{b}	
		к	v n, c, s,	0.	
vd :::	=	$x :: \sigma$	r; x = e	$\cdots \rightarrow $	$I_p \rightarrow I \ I_k \ I_c \ I_s$
Data constru	ctors	K			
Type constru	ictors	<i>T</i>	S		
Type variable	es	,,	kcs		
Monotypes	00	τ	, 10, 0, 0		$a \mid \tau \to \tau \mid T \overline{\tau}$
monotypes		'			Two Ben τ
Type Schem	A C	~			$\pi \mid \forall \overline{a} \neq \sigma$
Type Selleni	05	0		–	7 Va.7
Term variabl	65	ri	1 7		
Constraints	05	C	,, ~ Д		$\epsilon \tau \sim \tau C \wedge C$
Substitution		ф,	D		$\alpha \mid \phi \mid a := \sigma$
Substitutions	`	φ			$\emptyset \mid \psi, \{u := i\}$
Terms	e :	:=	$K \mid x \mid \lambda$	$\lambda x :: \tau.$	$e \mid e \mid$
			let x ::	$\sigma = e$	in e
		i	$case[\tau]$	e of [p]	$i \rightarrow e_i]_{i \in I}$
		i	typerep	T	
		i	typecas	$e[\tau] e$	of
		i	(typer	$e\vec{p} \vec{T}$) a	$x_1 \dots x_n \to e \mid \to e$
		i	if $e \simeq_{\tau}$	e then	e else e
Patterns	p :	:=	$K x_1 \dots$	x_n	
Type names	\mathbb{T} :	:=	$T \mid Arre$	wTu	Existential

Figure 3. The core language manipulated by Ghostbuster

with any constraints on the output type pushed into a per-dataconstructor constraint store (C):

$$K_i :: \forall \overline{a}, b.C \Rightarrow \tau_1 \to \cdots \to \tau_p \to T \overline{a}$$

We avoid this normalization. Because we lack type class constraints in the language (and equality constraints over existentially-bound variables can easily be normalized away), we simply omit perdata-constructor constraints. This means that when scrutinizing a GADT with case, we must synthesize constraints equating the scrutinee's type $T \overline{\tau}$ with $T \overline{\tau_k} \overline{\tau_c} \overline{\tau_s}$ in each K_i clause and then add this into a constraint store C, which we will use during typechecking (Figure 5). The advantage is that avoiding per-constructor constraints greatly simplifies our definition of the allowable space of input datatypes for Ghostbuster (Section 5). The absence of perconstructor type-class constraints from our core language is also why we require type-indexed TypeRep values (rather than equivalent Typeable constraints) to observe the type of newly existential type variables (Sections 2.1 and 2.2).

4.2 Type System

The typing rules for our language are syntax-directed and are given in Figures 4 to 7. Many of the typing rules are standard, but a few in particular Pat, TypeRep, TypeCase, and IfTyEq—are unique to our language and separated into Figure 4. Here the TypeRep and TypeCase rules only cover the type constructor T cases, the elided rules for the built-in type representations $\mathbb{T} = Existential$ and $\mathbb{T} = ArrowTy$ are nearly identical.

We lack a full kind system, but we do track the arity of constructors, with $T: \overline{\star}^n \in \Gamma$ as a shorthand for the *T* being an arity-*n* type constructor. We require that all type constructors be fully applied except when referenced by name through the (typerep *T*) form.

5. Preconditions and Ambiguity Checking

Before Ghostbuster can generate up- and down-conversion functions, it first performs a sanity check that the datatypes, together with requested parameter erasures, meet all preconditions necessary

$$\begin{array}{c} C, \Gamma \vdash_{e} \text{typerep } T: \overline{\text{TypeRep } a^{n}} \rightarrow \text{TypeRep } (T \ \overline{a}^{n}) & C, \Gamma \vdash_{e} e: \text{TypeRep } a_{0} \\ \hline C \land (a_{0} \sim T \ \overline{a}^{n}), \Gamma \cup \{x_{1}: \text{TypeRep } a_{1}, \ldots, x_{n}: \text{TypeRep } a_{n}\} \vdash_{e} e': \tau & C, \Gamma \vdash_{e} e'': \tau \\ \hline C, \Gamma \vdash_{e} \text{typecase}[\tau] \ e \ \text{of} \ ((\text{typerep } T) \ x_{1} \ldots x_{n}) \rightarrow e' \mid_{-} \rightarrow e'': \tau \\ \hline C, \Gamma \vdash_{e} \text{typeRep } a^{n} \rightarrow \text{TypeRep } (T \ \overline{a}^{n}) \\ \hline T : \overline{\star}^{n} \in \Gamma \\ \hline C, \Gamma \vdash_{e} \text{typerep } T: \overline{\text{TypeRep } a^{n}} \rightarrow \text{TypeRep } (T \ \overline{a}^{n}) \\ \end{array}$$

$$\begin{array}{c} C, \Gamma \vdash_{e} e_{1}: \text{TypeRep } \tau_{1} & C, \Gamma \vdash_{e} e_{2}: \text{TypeRep } \tau_{2} \\ \hline C \land (\tau_{1} \sim \tau_{2}), \Gamma \vdash_{e} e': \tau & C, \Gamma \vdash_{e} e'': \tau \\ \hline C, \Gamma \vdash_{e} \text{ if } e_{1} \simeq_{\tau} e_{2} \text{ then } e' \text{ else } e'': \tau \\ \end{array}$$

$$\begin{array}{c} \text{If TypeRep } \tau_{1} & T_{1} \vdash_{e} t_{1} \\ \hline T_{1} \\ \hline T_{1} \vdash_{e} t_{1} \\ \hline T_{1} \vdash_{e} t_{1} \\ \hline T_{1} \\ \hline T_{1} \vdash_{e} t_{1} \\ \hline T_$$



Figure 5. Typing rules for the core language

Figure 6. Equality Theory for the Ghostbuster Type System



Figure 7. Environment and program typing rules

for the tool to generate well-typed conversion functions. Indeed, as we discussed in Section 2 not every erasure setting is valid. We therefore want to create sufficient preconditions such that *if* these preconditions are met, the Ghostbuster tool is guaranteed to generate a pair of well-typed functions (up, down), such that up-conversion followed by down-conversion is a total identity function. This section details these preconditions and ambiguity criteria.

5.1 Ambiguity Test

The goal of the ambiguity criteria are a concise specification of the class of programs handled by Ghostbuster. These nonambiguity prerequisites apply per-data-constructor, K_i , per-datatype that requests a type erasure (nonempty \overline{c} or \overline{s} variables). If all constructors of the datatypes undergoing erasure individually pass the ambiguity check, then the input program as a whole is valid. For a given data constructor:

$$K_i :: \forall \overline{k}, \overline{c}, \overline{s}, \overline{b}, \tau_1 \to \cdots \to \tau_p \to T \ \overline{\tau_k} \ \overline{\tau_c} \ \overline{\tau_s}$$

We refer to the types τ_i through τ_p as the *fields* of the constructor, and the $T \ \overline{\tau_k} \ \overline{\tau_c} \ \overline{\tau_s}$ expression as the right-hand side (RHS). Types which occur in a checked or synthesized *context* means that they occur within the arguments of some type constructor T in positions corresponding to its \overline{c} or \overline{s} type parameters. Likewise, *kept* (or *nonerased*) context \overline{k} refers to all types τ that are *not* in checked or synthesized context.

The ambiguity check is concerned with *information flow*. That is, whether the erased information can be recovered based on properties of the simpler datatype. If not, then these type variables would not be recoverable upon down-conversion and Ghostbuster rejects the program.

5.2 Type Variables Synthesized on the RHS

For each synthesized type $\tau' \in \overline{\tau_s}$ on the RHS, type variables occurring in that type, $a \in Fv[\![\tau']\!]$, must be computable based on:

- occurrences of a in any of the fields $\overline{\tau}^p$. That is, $\exists i \in [1, p] . a \in Fv_s[\![\tau_i]\!]$, using the $Fv_s[\![1]\!]$ function from Figure 8; or
- $a \in Fv[\overline{\tau_k}]$. That is, kept RHS types; or
- $a \in Fv[[\overline{\tau_c}]]$. That is, a occurs in the *checked* (input) type.

Note that the occurrences of a in fields can be in kept or in synthesized contexts, but *not* checked. For example, consider our Exp example (Section 3.1), where the a variable in the type of an expression Exp e a is determined by the synthesized a component

of its sub-expressions, bottoming out at leaf expressions such as constants and variables. In contrast, checked variables in the fields must be created by the down-conversion function as *inputs* to recursive down-conversion calls on the value's fields. Thus they cannot be a source of new information to determine synthesized outputs, and we use the $Fv_s[]$ rather than the Fv[] metafunction above. Conversely, notice that we do not worry about applying the above prerequisites to synthesized variables inside fields—these are the *outputs* of recursive down-conversion calls. Their computability is left to an inductive argument (bottoming out at "leaf" constructors such as Exp's Con).

5.3 Type Variables in Checked Context

All types in checked context in $\tau_1 \dots \tau_p$ are implicit arguments to the down-conversion function that will process that field. Thus for all τ_i in checked context, all $a \in Fv[[\tau_i]]$ must be computable based on information available *at that point*, which includes:

- kept or checked variables in the RHS, $a \in Fv[[\overline{\tau_c}]] \cup Fv[[\overline{\tau_k}]]$
- occurrences of a in non-erased context within any field
- occurrences of a in Fv [[τ_j]], for other fields τ_j that have already been processed before the field containing τ_i.

This last case—inter-field dependencies—can be found in the Abs case of our expression language (Section 3.1):

Abs :: Typ a
$$\rightarrow$$
 Exp (e, a) b \rightarrow Exp e (a \rightarrow b)

Recall that in our example, given $\text{Exp} \in a$, we erase e in checked mode and a in synthesized mode. Thus the type (e, a) is in checked context, so how is it determined? It cannot be resolved using $(a \rightarrow b)$ on the RHS, as this is a synthesized type (meaning it is an *output* of the down-conversion function); it must be determinable from the other fields of the constructor, in this case Typ a.

For a type in checked context, we must be able to determine which fields to examine in order to determine what the checked type should be. This requires that any inter-field dependencies do not form a cycle. As an example, we cannot erase the type t from Loop, because the types a and b in the fields of the constructor are in checked mode but depend on each other:

For simplicity our formal language assumes that fields are already topologically sorted so that dependencies are ordered left to right. That is, a field τ_{i+k} can depend on field τ_i . In the case of Abs, $a \in Fv_s[Typ a]$ and $\tau_1 = Typ a$ occurs before $\tau_2 = Exp$ (e,a) b, therefore Ghostbuster accepts the definition.

5.4 Gradual Erasure Guarantee

One interesting property of the class of valid inputs described by the above ambiguity check is that it is always valid to erase *fewer* type variables—to change an arbitrary subset of *erased* variables (either \overline{c} or \overline{s}) to *kept* (\overline{k}). That is:

Theorem 1 (Gradual erasure guarantee). For a given datatype with erasure settings \overline{k} , $\overline{c} = \overline{c_1} \overline{c_2}$ and $\overline{s} = \overline{s_1} \overline{s_2}$, then erasure settings $\overline{k}' = (\overline{k} \overline{c_2} \overline{s_2}), \overline{c}' = \overline{c_1}, \overline{s}' = \overline{s_1}$ will also be valid.

Proof. The requirements above are specified as a conjunction of constraints over *each* type variable in synthesized or checked position. Removing erased variables removes terms from this conjunction.

 Fv_s : extracting dependencies for synthesized type variables

$Fv_s\llbracket a\rrbracket$	=	$\{a\}$			
$Fv_s[\![au_1 \dots au_n]\!]$	=	$\bigcup_{i=1}^{n} Fv_s[\![\tau_i]\!]$			
$ \begin{aligned} F v_s \llbracket \tau_1 &\to \tau_2 \rrbracket \\ F v_s \llbracket T \overline{\tau_k} \ \overline{\tau_c} \ \overline{\tau_s} \rrbracket \end{aligned} $	=	$ \begin{aligned} \stackrel{i=1}{Fv_s} \llbracket \tau_1 \rrbracket & \cup Fv_s \llbracket \tau_2 \rrbracket \\ Fv_s \llbracket \overline{\tau_k} \rrbracket & \cup Fv_s \llbracket \overline{\tau_s} \rrbracket \end{aligned} $			
Fv_k : extracting free vars in non-erased context					
$Fv_k[\![a]\!]$	=	$\{a\}$			
$Fv_k[\![au_1 \dots au_n]\!]$	=	$\bigcup_{i=1}^{n} F v_k \llbracket \tau_i \rrbracket$			
$Fv_k[\![au_1 \to au_2]\!]$	=	$F_{v_k}^{i=1}[\tau_1] \cup F_{v_k}[\tau_2]$			
$Fv_k \llbracket T\overline{\tau_k} \ \overline{\tau_c} \ \overline{\tau_s} \rrbracket$	_	$Fv_{L}[\overline{\tau_{L}}]$			

Figure 8. Extracting free type variables in different contexts.

For the remaining erased type variables, their dependence check may have depended on formerly erased, now kept, variables. However, both the synthesized and checked dependency prerequisites include all variables in kept context. Thus, moving variables from erased to kept context never breaks any dependency. $\hfill\square$

6. Core Translation Algorithms

We now describe the core translation algorithms used in Ghostbuster using the language defined in Section 4. The resulting pipeline of translation passes is shown in Figure 2.

6.1 Simplified Datatype Generation

Creating simplified data definitions is straightforward. Fields τ_i are replaced with updated versions, τ'_i , that replace all type applications $T \tau_k \tau_c \tau_s$ with $T' \tau_k$:

$$\begin{split} \mathbf{K}_{i} : \forall \overline{k}, \overline{c}, \overline{s}, \overline{b}. \tau_{1} \to \cdots \to \tau_{p} \to T \ \overline{\tau_{k}} \ \overline{\tau_{c}} \ \overline{\tau_{s}} \\ \Rightarrow \\ : \forall \overline{k}, \overline{b}. \ \text{getTyReps}(\mathbf{K}_{i}) \to \tau_{1}' \to \cdots \to \tau_{p}' \to T \ \overline{\tau_{h}} \end{split}$$

Where getTyReps returns any newly existential variables for a constructor (Section 2.2):

getTyReps(K_i :
$$\forall \overline{k}, \overline{c}, \overline{s}, \overline{b}.\tau_1 \to \cdots \to \tau_p \to T \ \overline{\tau_k} \ \overline{\tau_c} \ \overline{\tau_s}) =$$

{TypeRep $a \mid a \in (Fv_k [\![\tau_1 \dots \tau_p]\!] - Fv[\![\overline{\tau_k}]\!]) - \overline{b}$ }

Recall here that \overline{b} are the preexisting existential type variables that do not occur in $\overline{\tau_k} \ \overline{\tau_c} \ \overline{\tau_s}$.

6.2 Up-conversion Generation

 K'_i

In order to generate the up-conversion function for a type T, we instantiate the following template:

$$\begin{array}{rll} {\rm up} T_i & :: & \overline{{\rm typeRep}\ c} \to \overline{{\rm typeRep}\ s} \to T_i \ k \ \overline{c} \ \overline{s} \to T_i' \ k \\ {\rm up} T_i \ {\rm c1_typerep}\ \dots {\rm sn_typerep} \ {\rm orig} = \\ \begin{array}{rll} {\rm case \ orig \ of} \\ {\rm K}_j \ x_1 \ \dots \ x_p \to \\ {\rm let}\ \phi & = \ {\rm unify}({\rm T}\ k \ {\rm c}\ {\rm s}, \ {\rm T}\ \tau_k \ \tau_c \ \tau_s) \\ {\rm KtyRep}_j & = \ {\rm map}\ (\lambda\tau \to {\rm bind}(\phi, \ [\tau], \ {\rm buildTyRep}(\tau))) \\ {\rm getTyReps}({\rm K}) \\ {\rm in} \\ {\rm K}_j' \ {\rm KtyRep}_j \\ {\rm dispatch}_{\uparrow}(\phi, \ x_1, \ \phi(\tau_1)) \dots {\rm dispatch}_{\uparrow}(\phi, \ x_p, \ \phi(\tau_p)) \end{array}$$

The Supplemental Material (Section B) includes the full, formal specification of up/down generation, but the procedure is straightforward: pattern match on each K_j and apply the K'_j constructor. The

complexity is in the type representation management of the *bind* and *dispatch*^{\uparrow} operations. Here we follow a naming convention where a type variable k is witnessed by a type representation bound to a term variable $k_typerep$. Ghostbuster performs a renaming of type variables in data definitions to ensure there is no collision between the variables used at the declaration head $T \ k \ c \ s$, and those used within each constructor K_j . For example, this already holds in Exp where we used env/ans interchangably with e/a.

In the let-binding of ϕ above, we unify the type of orig with the expected result type of K_j . This uses a unification function that is part of a type checking algorithm based on the type system of Figures 4 to 7. Because we use the $\overline{k} \ \overline{c} \ \overline{s}$ variables to refer to the type of the input, orig, this gives us a substitution binding these type variables. For example, in the Abs case of our expression language (Section 3.1):

Abs :: Typ r
$$\rightarrow$$
 Exp (e, r) s \rightarrow Exp e (r \rightarrow s)

unification yields:

$$\phi = \{env := e, ans := (r \to s)\}$$

It is the job of *bind* to navigate this substitution in order to create type representations for type variables mentioned in ϕ , such as r. Here, getting to r requires digging inside the type representation for ans using a typecase expression. Because the type representation added to K'_j will always be of the form TypeRep a (for type variable a), this is all the call to *bind* must do to create the type representations that decorate K'_j . Note that there may be multiple occurrences of $r \in \phi$, and thus multiple *paths* that bind might navigate; which path it chooses is immaterial.

Type representation construction in dispatch The *dispatch*^{\uparrow} function is charged with recursively processing each field f of κ_j . Based on the type of f this will take one of two actions:

- · Opaque object: return it unmodified.
- Ghostbusted type T: call upT.

In the latter case, it is necessary to build type representation arguments for the recursive calls. This requires not just accessing variables found in ϕ , but also building compound representations such as for the pair type (e, r) found in the Abs case of Exp.

Finally, when building type representations inside the $dispatch_{\uparrow}$ routine, there is one more scenario that must be handled: representations for pre-existing existential variables, such as the type variable a in App:

App :: Exp e (a
$$\rightarrow$$
 b) \rightarrow Exp e a \rightarrow Exp e b

In recursive calls to upExp, what representation should be passed in for a? We introduce an explicit ExistentialType in the output language of the generator which appears as an implicitly defined datatype such that (typerep Existential) is valid and has type \forall a. TypeRep a.

Theorem 2 (Reachability of type representations). All searches by bind for a path to v in ϕ succeed.

Proof. By contradiction. Assume that $v \notin \phi$. But then v must not be mentioned in the $T_i \ \overline{\tau_k} \ \overline{\tau_c} \ \overline{\tau_s}$ return type of K_j . This would mean that v is a preexisting existential variable, whereas only newly existential variables are returned by getTyReps.

6.3 Down-conversion Generation

Down-conversion is more challenging. In addition to the type representation binding tasks described above, it must also perform runtime type tests (\simeq_{τ}) to ensure that constraints hold for formerly

erased (now restored) type variables. The type signature of a downconverter takes type representation arguments only for checked type variables; synthesized types must be computed:

 $\operatorname{down} T_i :: \overline{\operatorname{TypeRep} c} \to T'_i \ \overline{k} \to \operatorname{Sealed} T'_i \ \overline{k} \ \overline{c}$

If the set of synthesized variables is empty, then we can elide the Sealed return type and return $T'_i \ \overline{k} \ \overline{c}$ directly. This is our strategy in the Ghostbuster implementation, because it reduces clutter that the user must deal with. However, it would also be valid to create sealed types which capture no runtime type representations, and we present that approach here to simplify the presentation.

To invert the up function, down has the opposite relationship to the substitution ϕ . Rather than being *granted* the constraints ϕ by virtue of a GADT pattern match, it must test and witness those same constraints using (\simeq_{τ}). Here the initial substitution ϕ_0 is computed by unification just as in the up-conversion case above.

down
$$T_i$$
 c1_typerep ... cm_typerep lower =
case lower of
 K'_j ex_typerep ... $f_1... f_p \rightarrow$
let $\phi_0 = ...$ in
 $openConstraints(\phi_0, openFields(f_1...f_p))$
where
 $openConstraints(\emptyset, bod) =$ bod
 $openConstraints(a := b : \phi, bod) =$
if $a_typerep \simeq_{\tau} b_typerep$
then $openConstraints(\phi, bod)$
else $genRuntimeTypeError$
 $openConstraints(a := T \tau_1...\tau_n : \phi, bod) =$
typecase $a_typerep$ of
(typerep T) a_1 -typerep ... a_n -typerep \rightarrow
 $openConstraints(a_1 := \tau_1,...,a_n := \tau_n : \phi, bod)$

 \rightarrow genRuntimeTypeError

Again, a more formal and elaborated treatment can be found in the Supplemental Material (Section B). Above we see that *openConstraints* has two distinct behaviors. When equating two type variables, it can directly issue a runtime test. When equating an existing type variable (and corresponding _typerep term variable) to a compound type $T \overline{\tau}^n$, it must break down the compound type with a different kind of runtime test (typecase), which in turn brings more _typerep variables into scope. We elide the (\rightarrow) case, which is isomorphic to the type constructor one. Note that (\simeq_{τ}) works on any type of representation, but this algorithm follows the convention of only ever introducing variable references (e.g. a_typerep) to "simple" representations of the form TypeRep a.

Following *openConstraints*, *openFields* recursively processes the field arguments $f_1 \dots f_p$ from left to right:

$$\begin{array}{l} openFields(\texttt{f}::T \ \overline{\tau_k} \ \overline{\tau_c} \ \overline{\tau_s}: rst) = \\ \texttt{case} \ openRecursion(\phi_0,\texttt{f}) \ \texttt{of} \\ \texttt{Sealed}T_q \ \overline{\texttt{s'}}\texttt{-typerep} \ \texttt{f'} \rightarrow \\ openConstraints(unify(\overline{\texttt{s'}}\texttt{-typerep}, \overline{\tau_s}) \\ , openFields(rst)) \\ \end{array}$$

let f' = f in openFields(rst)

Here we show only the type constructor $(T \tau_k \tau_c \tau_s)$ case and the "opaque" case. We again omit the arrow case, which is identical to the type constructor one.

As before with $dispatch_{\uparrow}$, the openRecursion routine must construct type representations to make the recursive calls. Unsealing the result of a recursive call reveals more constraints that must be checked. For example, in the Add case of Exp, both recursions must synthesize a return type of Int and thus a type representation inside the Sealed type of (typerep Int). Likewise, in the

App case the function input and the argument types must match. *openConstraints* ensures these synthesized values are as expected before returning control to *openFields* to process the rest of the arguments.

Finally, in its terminating case, *openFields* now has all the necessary type representations in place that it can build the type representation for SealedT_i. Likewise, all the necessary constraints are present in the typing environment—from previous typecase and (\simeq_{τ}) operations—enabling a direct call to the more strongly typed K_j constructor.

 $openFields(\emptyset) =$ SealedT_i $\overline{buildTyRep(s_typerep)}$ (K_j f'₁ ··· f'_p)

The result of code generation is that Ghostbuster has augmented the *prog* with up- and down-conversion functions in the language of Figure 3, including the typecase and (\simeq_{τ}) constructs. What remains is to eliminate these constructs and emit the resulting program in the target language, which, in our prototype, is Haskell.

6.4 Validating Ghostbuster

We are now ready to state the main Ghostbuster theorem: upconversion followed by down-conversion is the identity after unsealing synthesized type variables.

Theorem 3. Round-trip Let prog be a program, and let $\mathbf{T} = \{(T_1, k_1, c_1, s_1), \ldots, (T_n, k_n, c_n, s_n)\}$ be the set of all datatypes in prog that have variable erasures. Let $\mathbf{D} = \{D_1, \ldots, D_n\}$ be a set of dictionaries such that $D_i = (D_i s, D_i c)$ contains all needed typeReps for the synthesized and checked types of T_i . We then have that if for each $(T_i, k_i, c_i, s_i) \in \mathbf{T}$ that T_i passes the ambiguity criteria, then Ghostbuster will generate a new program prog' with busted datatypes $\mathbf{T}' = \{(T'_1, k_1), \ldots, (T'_n, k_n)\}$, and functions $\operatorname{up} T_i$ and $\operatorname{down} T_i$ such that

$$\forall e \in \text{prog. prog} \vdash e :: T_i \ k_i \ c_i \ s_i \land (T_i, k_i, c_i, s_i) \in \mathbf{T} \\ \implies \text{prog}' \vdash (\text{up}T_i \ D_i \ e) :: T'_i \ k_i, \ where \ (T'_i, k_i) \in \mathbf{T}'$$
(1)

and

$$\forall e \in \text{prog. prog} \vdash e :: T_i \ k_i \ c_i \ s_i \land (T_i, k_i, c_i, s_i) \in \mathbf{T}$$

$$\implies \text{prog'} \vdash (\text{down} T_i \ D_i c \ (\text{up} T_i \ D_i \ e))$$
(2)
$$\equiv (\text{Sealed} T_i \ D_i s \ e :: \text{Sealed} T_i \ k_i \ c_i)$$

The full proof including supporting lemmas can be found in the Supplemental Material (Section C). We provide a brief proof-sketch here.

Proof Sketch. We first show by the definition of up-conversion that given any data constructor K of the correct type, that the constructor will be matched. Proceeding by induction on the type of the data constructor and case analysis on *bind* and *dispatch*₁ we then show that the map of *bind* over the types found in the constructor K succeeds in building the correct typeReps needed for the checked fields of K. After showing that every individual type-field is up-converted successfully and that this up-conversion preserves values, we are able to conclude that since we have managed to construct the correct type representations needed for the up-converted data constructor K', and since we can successfully up-convert each field of K, that the application of K' to the typeReps for the newly-existential types and the up-converted fields is well-typed and that the values that we wish to have preserved have been kept.

To show that down-conversion succeeds, we first show that given any data constructor K' of the correct type that the down-conversion function will match it. We then proceed by case analysis on the code-path executed on the right-hand-side of the case clause that matched the data constructor: we show that *openConstraints* succeeds in deriving suitable type representations for the call to openRecursion to succeed in constructing the correct downconverted datatypes for each of the busted recursive datatypes in the fields of K'. We then use this to show that openFields will succeed in down-converting the busted types that it encounters. We then use the fact that openFields has successfully down-converted the types it has encountered, coupled with the success of constructing suitable type representations to show that we are finally able to successfully construct the down-converted sealed type.

7. Implementing Ghostbuster for Haskell

The Ghostbuster prototype tool is a source-to-source translator, which currently supports Haskell but could be extended to other languages that incorporate GADTs. To build a practical tool implementing Ghostbuster, we need to import data definitions from, and generate code to, a target host language. Because our prototype targets Haskell, we extended our core language slightly to accommodate certain Haskell features of data definitions such as bang patterns. For the most part, code generation is a straightforward translation from our core-language into Haskell using the haskell-src-exts package,¹⁰ which we subsequently pretty-print to file. If erasure results in Haskell'98 datatypes, we add deriving clauses to the simplified datatypes for the standard typeclasses such as Show.

7.1 Current Limitations

Our current prototype comes with some limitations. Yet, as we will see in Section 8.2, a great many of the datatypes found in the wild are supported.

Runtime type representation As mentioned in Section 2.1, we require type-indexed TypeRep values, which are scheduled to appear in GHC-8.2. In the meantime, we use our own representation of runtime types synthesized on demand by the Ghostbuster tool and described in the Supplemental Material (Section A.3).

Advanced type system features There are some features we support indirectly by allowing them in the "opaque" regions of the datatype which Ghostbuster-generated code need not traverse, but we do not model explicitly in our core language. This currently includes type families [7, 20] and type classes [11, 19].

Erased datatypes as type parameters As we saw in Section 2.4, Ghostbuster does not allow datatypes undergoing erasure to be used as arguments to other type constructors, for example []. If available, we could lean on a Functor instance for that type, but in general there is not a single, clearly defined behaviour. Future work may allow a user to specify how Ghostbuster should traverse under type constructors to continue the erasure and conversion processes.

8. Evaluation

8.1 Runtime Performance

This section analyzes the performance of the conversion routines generated by Ghostbuster. Benchmarks were conducted on a machine with two 12-core Xeon E5-2670 CPUs (64-bit, 2.3GHz, 32GB RAM) running GNU/Linux (Ubuntu 14.04 LTS), using GHC version 7.10.1 at -02 optimization level. Each data point is generated via linear regression using the criterion package.¹¹

Figure 9 compares the performance of the Ghostbuster generated conversion routines for our simple expression language (Section 3.1). We generated large random programs that included all of the important cases of up- and down- conversion (Abs, App, etc.), and report the time to convert programs containing that number of terms.

Ghostbuster achieves comparable performance to a manually written up-conversion routine. The hand-written written downconversion routine, however, which uses embedded Typeable class constraints is significantly slower than the Ghostbuster generated version with embedded TypeRep values. Profiling reveals that our generated TypeRep encodings were more efficient than dictionary passing with Data.Typeable. However, this may be an artifact of the closed-world simplification we used to generate our TypeRep values, so this performance advantage may disappear once we transition to open-world, type-indexed Typeable arriving in GHC-8.2.

Even so, the size of the Ghostbuster generated up- and downconversion functions are comparable to the Data.Typeable based implementation:

Contender	SLOC	Tokens	Binary size
Ghostbuster	198	1426	1MB
Data.Typeable	122	1011	1MB
Hint	78	451	45MB

For the down-conversion process, we also compare against using GHC's interpreter as a library via the Hint package.¹² Due to the difficulty of writing the down-conversion process manually, it is appealing to be able to re-use the GHC Haskell type-checker itself in order to generate expressions in the original GADT. In this method, a code generator converts expressions in the simplified type into an equivalent Haskell expression using constructors of the original GADT, which is then passed to Hint as a string and interpreted, with the value returned to the running program. Unfortunately: (1) as shown in Figure 9, this approach is significantly slower than the alternatives; (2) the conversion must live in the IO monad; (3) generating strings of Haskell code is error-prone; and (4) embedding the entire Haskell compiler and runtime system into the program increases the size of the executable significantly.

Nevertheless, before Ghostbuster, this runtime interpretation approach was the only reasonable way for a language implemented in Haskell with sophisticated AST representations to read programs from disk. One DSL that takes this approach is Hakaru.¹³

8.2 Package Survey

We conclude our experimental evaluation by testing our prototype implementation against all 9026 packages currently available on hackage.haskell.org, the central open source package archive of the Haskell community. We seek to gather some insight into how many GADTs exist "in the wild" which might benefit from the automated up- and down-conversions explored in this work.

In this survey, we extract all of the ADT and GADT datatype declarations of a package, and group these data declarations into connected components. We elide any connected components where none of the data declarations are parameterised by a type variable, or do not contain at least one GADT. For each connected component, we then vary which type variables are kept, checked, or synthesized, and attempt to run ghostbuster on each configuration. For connected components containing many data types and/or type variables this can yield a huge search space, so we explore at most ten thousand erasure variants for each connected component. A summary of the results are shown in Table 1.

As discussed in Section 5, our current design has some restrictions on what datatypes and erasure settings it will accept. However, out of the variants explored, ghostbuster was successfully able to erase at least one type variable in 2,582,572 cases. Moreover, out

 $^{^{10}\,\}tt{http://hackage.haskell.org/package/haskell-src-exts}$

¹¹ http://hackage.haskell.org/package/criterion

¹²http://hackage.haskell.org/package/hint

¹³ https://hackage.haskell.org/package/hakaru



Figure 9. Time to convert a program in our richly-typed expression language (Section 3.1) with the given number of terms (i.e. nodes in the AST), from original GADT to simplified ADT (left) and vice-versa (right). Note the log-log scale.

Metric	
Total # packages	9026
Total # source files	94,611
Total # SLOC	16,183,864
Total # datatypes using ADT syntax	9261
Total # datatypes using GADT syntax	18,004
Total # connected components	15,409
ADTs with type variable(s)	1341
GADTs with type variable(s)	11,213
GADTs with type indexed variable(s)	8773
Actual search space	$185,\!056,\!322,\!576,\!712$
Explored search space	$9,\!589,\!356$
Ghostbuster succeeded	$2,\!582,\!572$
GADTs turned into ADTs	5525
Ambiguity check failure	5,374,628
Unimplemented feature in Ghostbuster	1,632,156

Table 1. Summary of package survey

of the 8773 "real" GADTs surveyed¹⁴, we were able to successfully ghostbust 5525 (63%) of these down to regular ADTs.

9. Related Work

Ornaments [10, 12, 15], from the world of dependent type theory, provides an interesting theoretical substrate for moving between inductive data structures that share the same recursive structure, where one type is refined, or ornamented, by adding and removing information. Unlike ornaments, we focus on *bidirectional* conversions from a richer to simpler type. Recent progress has been made in bringing ornaments from a theoretical topic to a practical language [28]. This prototype is semi-automated and leaves holes in the generated code for the user to fill in, rather than being an entirely *in language* and *fully-automatic* abstraction like Ghostbuster.

The eqT of Haskell's Typeable class and the $(typecase/\simeq_{\tau})$ and TypeRep of our core language, are both similar to typecase and Dynamic from Abadi et al. [1, 2]. However, while typecase (from dynamic) allows querying the type of expressions, it does not inject type-level evidence about the scrutinee into the local constraints the way that GADT pattern matching (and our typecase) do.

Another closely related work is on *staged inference* [23], which formulates dynamic typing as staged checking of a single unified

type system. While the mechanism is different, functions over Ghostbusted types defer type-checking obligations until downconversion. Likewise, Haskell's *deferred type errors* [27] are related, but are a coarse-grained setting at the module level and hence not practical for writing code against GADTs while deferring typechecking obligations.

The Yoneda lemma applied to Haskell provides a method of encoding GADTs as regular ADTs.¹⁵ However, this encoding does not offer the benefits of Ghostbuster simplified types because: (1) the encodings include function types, which preclude Show/Read deriving, and (2) the encoding cannot actually enforce its guarantees in Haskell due to laziness (lack of an initial object).

F# type providers [26] are related to Ghostbuster in that both automatically generate datatype definitions against which developers are expected to write code. Type providers do not include GADTs, but deal with type schemas that are too large (e.g. all of Wikipedia) or externally maintained (e.g. in a database) and must be populated dynamically, whereas Ghostbuster deals with maintaining simplified types for existing GADTs.

Checking whether input-output tags are consistent in a logic program is often approximated in practice based on a dependency graph of the variables. For example, the Mercury programming language [25] has tags: input, output, deterministic. Our ambiguity checking process is similar.

Ou et al. [18] define a language that provides interoperability between simply-typed and dependently-typed regions of code. Both regions are encoded in a common internal language (also dependentlytyped), with runtime checks when transitioning between regions. Similarly, the Trellys project [6] includes a two-level language design where each definition is labelled logical or programmatic. Because of the shared syntax, one can migrate code from programmatic to logical when ready to prove non-termination.

It is folklore in dependently typed programming communities (Idris, Agda, etc.) that if you need to write a parser for a compiler, you would parse to a raw, untyped term and write a type-checking function (i.e. down-conversion) manually. To our knowledge there are not currently any tools that automate this process. However, most fully dependent languages make these type checkers easier to write than they are in Haskell.

¹⁴ Some types were written in GADT syntax that didn't need to be.

¹⁵ The Yoneda lemma in Haskell is currently best explained in blog posts: http://www.haskellforall.com/2012/06/gadts.html and http://bartoszmilewski.com/2013/10/08/ lenses-stores-and-yoneda/.

10. Conclusion

We've shown how Ghostbuster enables the automatic maintenance of simplified datatypes that are easier to prototype code against. This resulted in some performance advantages in addition to software engineering benefits. Because of these advantages, we believe that in the coming years gradualization of type checking obligations for advanced type systems will become an active area of work and widely-used language implementations may better support gradualization of type-checking obligations directly.

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