

Datatype-Generic Programming Meets Elaborator Reflection

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Datatype-generic programming is natural and useful in dependently typed languages such as Agda. However, datatype-generic libraries in Agda are not reused as much as they should be, because traditionally they work only on datatypes decoded from a library's own version of datatype descriptions; this means that different generic libraries cannot be used together, and they do not work on native datatypes, which are preferred by the practical Agda programmer for better language support and access to other libraries. Based on elaborator reflection, we present a framework in Agda featuring a set of general metaprograms for instantiating datatype-generic programs as, and for, a useful range of native datatypes and functions —including universe-polymorphic ones— in programmer-friendly and customisable forms. We expect that datatype-generic libraries built with our framework will be more attractive to the practical Agda programmer. As the elaborator reflection features used by our framework become more widespread, our design can be ported to other languages too.

CCS Concepts: • Software and its engineering — Functional languages; Data types and structures.

Additional Key Words and Phrases: datatype-generic programming, dependently typed programming, inductive families, universe polymorphism, elaborator reflection, metaprogramming

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1 INTRODUCTION

Parametrised by datatype structure, datatype-generic programs [Gibbons 2007] are ideal library components since they can be instantiated for a usually wide range of datatypes, including user-defined ones as long as their structures are recognisable by the datatype-generic programs. Particularly in dependently typed programming [Stump 2016; Brady 2017; Kokke et al. 2020], datatype-genericity has long been known to be naturally achievable [Benke et al. 2003; Altenkirch and McBride 2003], and is even more useful for organising indexed datatypes with intrinsic constraints and their operations. However, there is hardly any datatype-genericity in, for example, the Agda standard library, which instead contains duplicated code for similar datatypes and functions. The existing dependently typed datatype-generic libraries [McBride 2011, 2014; Dagand and McBride 2014; Diehl and Sheard 2016; Ko and Gibbons 2017; Allais et al. 2021] —mostly in Agda, which will be our default language— are not reused as much as they should be either. What is going wrong?

One major problem, we argue, is the lack of interoperability. The prevalent approach to datatype-generic programming in Agda (recapped in Section 2) is to construct a family of datatype 'descriptions' and then decode the descriptions to actual datatypes via some least fixed-point operator μ . Generic programs take descriptions as parameters and work only on datatypes decoded from descriptions. Although this approach is theoretically rooted in the idea of universe à la Tarski [Martin-Löf

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97 98 1975, 1984] and serves as a simulation of a more recent theory of datatypes [Chapman et al. 2010] (discussed in Section 7), it is not what we want: Generic libraries usually use their own version of datatype descriptions and are incompatible with each other, so only one library can be chosen at a time, which is unreasonable. Moreover, decoded datatypes are essentially segregated from native datatypes, and there is no point for the Agda programmer to abandon most of the language support and libraries developed for native datatypes in exchange for one generic library.

So what do we want from datatype-generic libraries? We want to write our own native datatypes and then instantiate generic programs for them. And in a dependently typed setting, we should be able to instantiate theorems (and, in general, constructions) about native datatypes and functions too. For a standard example, from the List datatype, we want to derive not only its fold operator

```
foldr : \{A : \mathsf{Set}\ \ell\} \{B : \mathsf{Set}\ \ell'\} \to (A \to B \to B) \to B \to \mathsf{List}\ A \to B
foldr f e = e
foldr f e (a :: as) = f a (foldr f e as)
```

but also theorems about foldr, such as the following 'fold fusion' theorem (which allows us to optimise the composition of a foldr and a function *h* as a single foldr):

```
foldr-fusion : \{A : \operatorname{Set} \ell\} \{B : \operatorname{Set} \ell'\} \{C : \operatorname{Set} \ell''\}
                   (h: B \to C) \{e: B\} \{f: A \to B \to B\} \{e': C\} \{f': A \to C \to C\} \to C
                   (he: he \equiv e') (hf: \forall a b c \rightarrow h b \equiv c \rightarrow h (f a b) \equiv f' a c)
                   (as : List A) \rightarrow h (foldr f e as) \equiv foldr f' e' as
foldr-fusion h he hf \square
                                     = he
foldr-fusion h he hf (a :: as) = hf a_{-} (foldr-fusion h he hf as)
```

Note that both foldr and foldr-fusion are fully universe-polymorphic (otherwise they would be inconvenient to use). Also important (especially in a dependently typed setting) is the ability to derive new datatypes — the standard example is the derivation of vectors from list length,

```
data Vec (A : Set \ell) : \mathbb{N} \to Set \ell where
                                                                                    length : \{A : \operatorname{Set} \ell\} \to \operatorname{List} A \to \mathbb{N}
                                                       Vec A zero
                                                                                    length [] = zero
   :: A \to \forall \{n\} \to \operatorname{Vec} A \ n \to \operatorname{Vec} A \ (\operatorname{suc} n)
                                                                                    length (a :: as) = suc (length as)
```

and subsequently we want to derive constructions about vectors too. What we want is conceptually simple but immediately useful in practice: automated generation of native entities that had to written manually —including all the entities shown above— from datatype-generic programs.

Luckily, datatype-generic programming has a long history of development in Haskell [Löh 2004; Magalhães 2012], where we can find much inspiration for our development in Agda. Generic programs in Haskell have always been instantiated for native datatypes, so the interoperability problem in Agda does not exist there. However, generic program instantiation in Haskell traditionally proceeds by inserting conversions back and forth between native and generic representations, causing a serious efficiency problem. The conversions are even more problematic in Agda because their presence would make it unnecessarily complicated to reason about instantiated functions. The Haskell community addressed the conversion problem using compiler optimisation [Magalhães 2013] and, more recently, staging [Pickering et al. 2020]. Unfortunately, compiler optimisation does not work for us because instantiated functions are reasoned about even before they are compiled, and they need to be as clean as hand-written code right after instantiation; this need could be met by staging, which is not available in current dependently typed languages though.

Luckily again, in Agda there is a mechanism that can take the place of staging for generic program instantiation: elaborator reflection (inspired by Idris [Christiansen and Brady 2016]), through which the Agda metaprogrammer has access to operations for elaborating the surface language to the core,

in addition to the usual metaprogramming features such as quoting and unquoting. In fact, elabor-99 ator reflection is powerful enough for the metaprogrammer to develop general facilities for practical 100 101 datatype-generic programming. Like in Haskell, we can quote programmer-defined datatypes as 102 103 104 106 107 108 110 111 112 113 114 115

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146 147 descriptions for processing by generic programs; conversely, newly computed descriptions can be unquoted as programmer-friendly datatypes rather than decoded using a fixed-point operator, and functions can be defined by unquoting too. Besides these standard tasks, elaborator reflection achieves more and serves our purpose nicely: To express dependency in types, descriptions are usually higher-order and can be difficult to manipulate, but our metaprogram that manufactures datatypes from descriptions is surprisingly natural thanks to the 'local variable creation' technique [Nanevski and Pfenning 2005; Schürmann et al. 2005], which is easily implemented using a few elaborator reflection primitives that interact with the context during type-checking. Moreover, we can specialise generic programs to more efficient forms straightforwardly by using (open-term) normalisation —also an elaborator reflection primitive— to perform some of the computation early during elaboration. (By contrast, to achieve a similar effect with staging, it would be necessary to add annotations to generic programs, making generic libraries harder to develop.) Based on elaborator reflection, we have developed a framework in Agda where datatype-generic

programs can be instantiated as, and for, a useful range of native datatypes and functions in programmer-friendly and customisable forms. Central to the framework is a set of general metaprograms performing transformations between native and generic entities, for example deriving descriptions from datatypes and manufacturing datatypes from descriptions. These metaprograms are general in the sense that they are decoupled from generic libraries, and can be independently maintained and widely reused; the decoupling also allows generic libraries to be developed or adapted from old ones in largely the same, traditional way, without having to deal with native entities themselves. To interface with the metaprograms, generic libraries should (either directly or indirectly) target the datatype descriptions and function representations provided by our framework, which are expressive: we support inductive families [Dybjer 1994] and both fold and inductive functions, all of which can be parametrised and universe-polymorphic.

We expect that datatype-generic libraries built with our framework will be more attractive to the practical Agda programmer. As the elaborator reflection features used by our framework become more widespread, our design will be portable to other languages too (in particular dependently typed ones) — for example, by including an open-term normalisation operation, a metaprogramming system immediately gains the ability to optimise definitions. Moreover, Agda's currently unique design of universe polymorphism —where universe levels are made explicit and first-class—plays an important role in our universe-polymorphic datatype descriptions, and our work serves as a practical justification for further investigation into such design [Kovács 2022].

For the rest of the paper: After recapping standard datatype-generic programming (Section 2) and refining and replacing some definitions for our framework (Section 3), we present some of our metaprograms that showcase the power of elaborator reflection (Section 4). To simplify the presentation, up to this point we assume Set: Set and introduce only a slimmed-down version of our framework. Then, leaving Set: Set behind, we sketch how the full framework supports universe polymorphism (Section 5), and give a demo of the framework using some existing generic constructions (Section 6). Finally we conclude with some discussions (Section 7).

A RECAP OF DATATYPE-GENERIC PROGRAMMING

We start from a recap of standard datatype-generic programming (in a dependently typed setting). The core idea of datatype-genericity is to encode datatype definitions as 'descriptions', which can take a variety of forms but should be some kind of first-class data on which computation can be performed. Programs that manipulate datatype descriptions in some way are dubbed '(datatype-)

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```
data ConDs (I : Set) : Set where
                                                                                                                              \llbracket \_ \rrbracket_{\mathsf{Cs}} \, : \, \mathsf{ConDs} \, I \, \to (I \, \to \, \mathsf{Set}) \, \to (I \, \to \, \mathsf{Set})
                                                                                                                              [\![[]] \quad ]\!]_{\mathsf{Cs}} X i = \bot
                                                                       ConDs I
     \underline{} :: \underline{} : ConD I \rightarrow ConDs I \rightarrow ConDs I
                                                                                                                              [\![D::Ds]\!]_{\mathsf{Cs}} \ X \ i \ = \ [\![D]\!]_{\mathsf{C}} \ X \ i \ \uplus \ [\![Ds]\!]_{\mathsf{Cs}} \ X \ i
                                                                                                                              \llbracket \_ \rrbracket_{\mathcal{C}} : \mathsf{ConD} \ I \to (I \to \mathsf{Set}) \to (I \to \mathsf{Set})
data ConD (I : Set) : Set where
                                                                              \rightarrow ConD I
                                                                                                                              [\![\iota\,j] \quad ]\!]_{\mathsf{C}} \ X \ i = i \equiv j
     \sigma: (A: Set) \rightarrow (A \rightarrow ConD I) \rightarrow ConD I
                                                                                                                              \llbracket \boldsymbol{\sigma} A D \rrbracket_{\mathsf{C}} X i = \boldsymbol{\Sigma} [a:A] \llbracket D a \rrbracket_{\mathsf{C}} X i
                                                         \mathsf{ConD}\:I\:\:\to\:\mathsf{ConD}\:I
                                                                                                                              [\![\boldsymbol{\rho}\ D\ E\ ]\!]_{\mathsf{C}}\ X\ i\ =\ [\![\boldsymbol{D}]\!]_{\mathsf{R}}\ X\times [\![\boldsymbol{E}]\!]_{\mathsf{C}}\ X\ i
     \rho : \text{RecD } I \rightarrow
data RecD (I: Set): Set where
                                                                                                                              \llbracket \_ \rrbracket_{\mathsf{R}} : \mathsf{RecD}\, I \to (I \to \mathsf{Set}) \to \mathsf{Set}
                                                                                                                              [\![\iota\,i\,\,]\!]_{\mathsf{R}}\ X\ =\ X\,i
                                                                               \rightarrow RecD I
    \pi: (A: \mathsf{Set}) \to (A \to \mathsf{RecD}\ I) \to \mathsf{RecD}\ I
                                                                                                                              [\![\boldsymbol{\pi}\,A\,D\,]\!]_{\mathsf{R}}\ X\ =\ (a:A)\ \rightarrow [\![D\,a]\!]_{\mathsf{R}}\ X
```

Fig. 1. A basic version of datatype descriptions and their base functor semantics

generic programs', and can perform constructions tailored for individual datatypes by analysing input descriptions or produce new datatypes by computing output descriptions. The power of generic programs depends crucially on the range of datatypes encoded by the chosen descriptions. In Section 2.1 we fix on a class of descriptions covering inductive families [Dybjer 1994] in the form of sums of products. Then we review (*F*-)algebras in Section 2.2, the kind of generic program that our presentation will focus on (but not the only kind we use).

2.1 Datatype Descriptions

There have been quite a few variants of datatype descriptions [Altenkirch et al. 2007; Chapman et al. 2010; McBride 2011; Dagand and McBride 2014; Ko and Gibbons 2017]; here we use a three-layered version that closely follows the structure of an Agda datatype definition (comparable to de Vries and Löh's [2014] encoding). As a small running example, consider this accessibility datatype:

```
data Acc_{<}: \mathbb{N} \to Set where acc: (n: \mathbb{N}) \ (as: (m: \mathbb{N}) \ (lt: m < n) \to Acc_{<} \ m) \to Acc_{<} \ n
```

The first layer is the list of constructors, which for $Acc_{<}$ consists of only acc; the type of acc has two fields n and as, which constitute the second layer; the type of the field as is described in the third layer as it ends with the recursive occurrence $Acc_{<}m$, in front of which there are function arguments m and lt. Corresponding to the three layers, we use three datatypes of descriptions ConDs, ConD, and RecD in Figure 1—all parametrised by an index type I— to encode datatype definitions. Generic programs can then perform constructions depending on the number of constructors, the types of fields, the indices of recursive occurrences, etc. For example, $Acc_{<}$ is described by

```
 \begin{array}{l} \mathsf{Acc}_{<}\mathsf{D} : \mathsf{ConDs}\,\mathbb{N} \\ \mathsf{Acc}_{<}\mathsf{D} = (\sigma\,\mathbb{N}\,(\lambda\,n\,\rightarrow\,\rho\,(\pi\,\mathbb{N}\,(\lambda\,m\,\rightarrow\,\pi\,(m\,<\,n)\,(\lambda\,lt\,\rightarrow\,\iota\,m)))\,(\iota\,n))) :: [] \end{array}
```

Inhabitants of ConDs I are just lists of constructor (type) descriptions of type ConD I. Inhabitants of ConD I are also list-like: the elements can either be the type of a non-recursive field, marked by σ , or describe a recursive occurrence, marked by ρ , and the 'lists' end with ι . Different from ordinary lists, in the case of σ A D a new variable of type A is brought into the context of D (for example, in the type of acc, the field n appears in the type of as); this is done by making D a function from A, using the host language's function space to extend the context. The ι at the end of a ConD I should specify the index targeted by the constructor (for example, the final n in the

¹The computation power of the host language's function space has been better utilised in the datatype-generic programming literature (for example by McBride [2011, Section 2.1]), but we will refrain from abusing the function space in the descriptions we write for tasks beyond context extension, keeping our descriptions in correspondence with native datatypes. In general, if there are abuses, they will be detected at the meta-level (Section 4.2).

```
\begin{split} &\operatorname{fmap}_{\operatorname{Cs}}: (D:\operatorname{ConDs}I) \to (\forall \ \{i\} \to X\ i \to Y\ i) \to \forall \ \{i\} \to [\![D]\!]_{\operatorname{Cs}}\ X\ i \to [\![D]\!]_{\operatorname{Cs}}\ Y\ i \\ &\operatorname{fmap}_{\operatorname{Cs}}\ (D:Ds)\ f\ (\operatorname{inl}\ xs) = \operatorname{inl}\ (\operatorname{fmap}_{\operatorname{C}}\ D\ f\ xs) \\ &\operatorname{fmap}_{\operatorname{Cs}}\ (D:\operatorname{Ds})\ f\ (\operatorname{inr}\ xs) = \operatorname{inr}\ (\operatorname{fmap}_{\operatorname{Cs}}\ Ds\ f\ xs) \\ &\operatorname{fmap}_{\operatorname{C}}\ : (D:\operatorname{ConD}\ I) \to (\forall \ \{i\} \to X\ i \to Y\ i) \to \forall \ \{i\} \to [\![D]\!]_{\operatorname{C}}\ X\ i \to [\![D]\!]_{\operatorname{C}}\ Y\ i \\ &\operatorname{fmap}_{\operatorname{C}}\ (\mathfrak{o}\ A\ D)\ f\ (a,\ xs) = a, \operatorname{fmap}_{\operatorname{C}}\ (D\ a)\ f\ xs \\ &\operatorname{fmap}_{\operatorname{C}}\ (\mathfrak{o}\ D\ E)\ f\ (xs,\ xs') = \operatorname{fmap}_{\operatorname{R}}\ D\ f\ xs, \operatorname{fmap}_{\operatorname{C}}\ E\ f\ xs' \\ &\operatorname{fmap}_{\operatorname{R}}\ : (D:\operatorname{RecD}\ I) \to (\forall \ \{i\} \to X\ i \to Y\ i) \to [\![D]\!]_{\operatorname{R}}\ X \to [\![D]\!]_{\operatorname{R}}\ Y \\ &\operatorname{fmap}_{\operatorname{R}}\ (\mathfrak{o}\ A\ D)\ f\ xs = f\ x \\ &\operatorname{fmap}_{\operatorname{R}}\ (\pi\ A\ D)\ f\ xs = \lambda\ a \to \operatorname{fmap}_{\operatorname{R}}\ (D\ a)\ f\ (xs\ a) \end{split}
```

Fig. 2. Datatype-generic functorial maps of base functors

type of **acc**). Inhabitants of RecD *I* use the same structure to describe dependent function types ending with a recursive occurrence.

A couple of syntax declarations will make descriptions slightly easier to write and read:

```
syntax \pi A (\lambda a \rightarrow D) = \pi [a:A] D; syntax \sigma A (\lambda a \rightarrow D) = \sigma [a:A] D
```

For example, Acc_D can be rewritten as $(\sigma[n:\mathbb{N}] \rho(\pi[m:\mathbb{N}] \pi[lt:m < n] \iota m)(\iota n)) :: [].$

A description D: ConDs I is converted to a datatype μ $D: I \to$ Set for day-to-day programming by taking the least fixed point of the base functor $[\![D]\!]_{Cs}: (I \to Set) \to (I \to Set)$:

```
data \mu (D: ConDs I) : I \rightarrow Set where con : \forall \{i\} \rightarrow [\![D]\!]_{Cs} (\mu D) i \rightarrow \mu D i
```

For example, we can redefine $Acc_{<}$ as μ $Acc_{<}D: \mathbb{N} \to Set$, whose inhabitants are now constructed by the generic constructor **con**. Specified by the definition of the base functor $[\![D]\!]_{Cs}$ in Figure 1,² the argument of **con** encodes the choice of a constructor and the arguments of the chosen constructor in a sum-of-products structure; for example, in Agda it is customary to use a pattern synonym [Pickering et al. 2016] to define **acc** in terms of **con**,

```
pattern acc n as = con (inl (n, as, refl))
```

where the arguments n and as of acc are collected in a tuple (product structure), tagged by inl (left injection into a sum type), and finally wrapped up with con as an inhabitant of μ Acc_D n. In general, when there are multiple constructors, the injection parts will look like inl ..., inr (inl ...), inr (inr (inl ...)), etc, specifying the constructor choice in Peano style. The equality proof refl at the end of the tuple needs a bit more explanation: in the type of con, the index i is universally quantified, which seems to suggest that we could construct inhabitants of μ D i for any i, but the equality proof forces i to be n, the index targeted by acc.

2.2 Algebras as Generic Programs

Now we can write programs on Acc_<, for example its fold operator:

 $^{^2\}bot$ is the empty type with no constructors. $A \uplus B$ is the sum of the types A and B with constructors $\mathbf{inl}: A \to A \uplus B$ and $\mathbf{inr}: B \to A \uplus B$. $\Sigma[a:A]$ B is a dependent pair type, where $\Sigma[a:A]$ binds the variable a, which can appear in B; the pair constructor $_$ associates to the right. Free variables in types (such as I in the types of $[\![_]\!]_{Cs}$, $[\![_]\!]_{C}$, and $[\![_]\!]_{R}$) are implicitly universally quantified.

 However, the point of using described datatypes such as μ Acc_<D is that we do not have to write foldAcc_< ourselves but can simply derive it as an instantiation of a generic grogram. The class of generic programs we will focus on in this paper is '(*F*-)algebras' [Bird and de Moor 1997] (where the functor *F* is always some base functor $\llbracket D \rrbracket_{Cs}$ in this paper), whose type is defined by

```
\begin{array}{ll} \mathsf{Alg} : \mathsf{ConDs} \ I \to (I \to \mathsf{Set}) \to \mathsf{Set} \\ \mathsf{Alg} \ D \ X \ = \ \forall \ \{i\} \to [\![D]\!]_{\mathsf{Cs}} \ X \ i \to X \ i \end{array}
```

Algebras are useful because they are the interesting part of a fold function: By a 'fold function' we mean a function defined recursively on an argument of some datatype by (i) pattern-matching the argument with all possible constructors, (ii) applying the function recursively to all the recursive fields, and (iii) somehow computing the final result from the recursively computed sub-results and the non-recursive fields. For example, foldAcc $_{<}$ is a fold function, and so are a lot of common functions such as list length. The first two steps are the same for all fold functions on the same datatype, whereas the third step is customisable and represented by an algebra, whose argument of type $[\![D]\!]_{Cs} X i$ represents exactly the input of step (iii). We can define a generic fold operator that expresses the computation pattern of fold functions and can be specialised with an algebra,

```
{-# TERMINATING #-} fold : (D : \text{ConDs } I) \rightarrow \text{Alg } D X \rightarrow \forall \{i\} \rightarrow \mu D i \rightarrow X i fold D f (\text{con } ds) = f (\text{fmap}_{Cs} D (\text{fold } D f) ds)
```

where fmap_{Cs} is the functorial map for $[D]_{Cs}$ (defined in Figure 2),³ used here to apply fold Df to the recursive fields in ds.⁴ Libraries may provide generic programs in the form of algebras parametrised by descriptions, and the user gets a fold function for their datatype by applying fold to an algebra specialised to the description of the datatype. For example, by specialising a generic program in Section 6.1, we get an algebra (with some parameters of its own)

```
foldAcc<sub><</sub>Alg : \{P : \mathbb{N} \to \text{Set}\}\ (p : \forall n \to (\forall m \to m < n \to P m) \to P n) \to \text{Alg Acc}_< D P \text{ foldAcc}_< Alg p (inl <math>(n, ps, refl)) = p n ps
```

which we then use to specialise fold to get foldAcc<:

Being able to treat folds generically means that we can write generic programs whose types have the form $\forall \{i\} \to \mu D i \to X i$, but this is not enough when, for example, we want to prove generic theorems by induction on $d: \mu D i$, in which case the types take the more complex form $\forall \{i\} (d: \mu D i) \to P d$ (where $P: \forall \{i\} \to \mu D i \to Set$). Therefore we have another set of definitions for generic induction, corresponding to the scheme of elimination rules of inductive families [Dybjer 1994, Section 3.3]. The technical details of generic induction are omitted from the presentation, however, since the treatment is largely standard (closely following, for example, McBride [2011]), and our metaprograms (Section 4) work for fold and induction in the same way.

 $^{^3}$ For most of the generic programs in this paper we will provide only a sketch, because they are not too different from those in the literature. But as a more detailed example, the functorial map (Figure 2) is a typical generic program: The functorial map should apply a given function f to all the recursive fields in a sum-of-products structure while leaving everything else intact, and it does so by analysing the input description layer by layer — fmap_{Cs} keeps the choices of inl or inr, fmap_C keeps the σ -fields and ι -equalities, and finally fmap_R applies f to the recursive fields (of type X i for some i) pointwise. 4 Agda's termination checker cannot verify that this generic fold operator is terminating, hence the unsafe TERMINATING pragma. This is not a problem for us because instead of using this fold operator, we will only manufacture fold functions on specific datatypes such as foldAcc_<, which do pass the termination check.

3 PARAMETERS AND CONNECTIONS

 To make our framework expressive enough for practical use, we need a couple more features. The first feature is datatype parameters. For example, it was probably tempting to generalise $Acc_{<}$ (Section 2.1) to a version parametrised by a type A and a binary relation R on A:

```
data Acc \{A : Set\} (R : A \rightarrow A \rightarrow Set) : A \rightarrow Set where acc : (x : A) \rightarrow ((y : A) \rightarrow R y x \rightarrow Acc R y) \rightarrow Acc R x
```

The second feature is a data structure —which we call datatype *connections*— that abstracts and replaces the generic μ operator (although similar structures have long been present in the Haskell tradition of datatype-generic programming). Here is a motivating example: From a description D, we will use a metaprogram to manufacture a native datatype N (in place of μ D). Subsequently we may need to compute from D a new description that refers to N and its constructors. For example, in Section 6.3 we will define a datatype-generic predicate All P stating that a given predicate P holds for all the elements in a container-like structure; for lists, All specialises to

```
data ListAll \{A: \mathsf{Set}\}\ (P: A \to \mathsf{Set}): \mathsf{List}\ A \to \mathsf{Set}\ \mathbf{where}
[] : \mathsf{ListAll}\ P\ []
\underline{::}: \forall \{a\} \to P\ a \to \forall \{as\} \to \mathsf{ListAll}\ P\ as \to \mathsf{ListAll}\ P\ (a:: as)
```

whose description can be computed from the description of List. If μ were in use, then N would simply be μ D, whose constructor would also be known to be con; without μ , however, N and its constructors have to be provided as additional input to the generic construction of All to allow the latter to specialise to the description of ListAll. In general, datatype connections capture the extensional behaviour of datatypes generically such that what we can do with connections are more or less the same as what we can do with those datatypes manufactured with μ . Thus, by replacing μ with datatype connections, we can easily adapt generic programs that assume the presence of μ to work on native datatypes instead. The extensionality also allows us to customise the forms of native datatypes flexibly as long as they still behave the same.

After treating datatype parameters and connections respectively in Sections 3.1 and 3.2, we will also define parametrised algebras and their connections with fold functions in Section 3.3. In this section it may appear that, to use our framework, the programmer needs to write the additional components such as descriptions and connections, which are usually tedious, but keep in mind that we will be able to generate them automatically by metaprograms (Section 4).

3.1 Parametrised Datatype Descriptions and Telescopes

It is conceptually straightforward to encode a parametrised datatype, since parameters are just variables in the context which can be referred to by the index type and constructor types, and we know an easy way to extend the context — just use the host language's function space. So, traditionally, a parametrised datatype could be described by a parameter type P: Set, a parametrised index type $I:P\to$ Set, and a parametrised list of constructor descriptions $(p:P)\to$ ConDs (Ip). For example, we could describe Acc with $P=\Sigma[A:\text{Set}]$ $(A\to A\to \text{Set}), I=\lambda$ $(A,_)\to A$, and a parametrised description that looks like Acc $_<$ D. This encoding works well in principle, but there is one refinement that can make the encoding work better in practice.

We will eventually need to convert a description to a datatype, and it would be unsatisfactory if the parameter and index types in the datatype were not in the conventional curried form. To make this currying easier, we introduce telescopes [de Bruijn 1991] to represent lists of parameter or index types, as shown in Figure 3.⁵ Again we use the host language's function space to bring variables of

 $^{^{5}}$ T is the unit type with one constructor **tt**.

```
mutual
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                                data Tel: Set where
                                                                                                                     \llbracket \_ \rrbracket_\mathsf{T} : \mathsf{Tel} \to \mathsf{Set}
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                                        : Tel
                                                                                                                     [ [ ] ]_T = \top
346
                                    \underline{\quad}::\underline{\quad}:(A:\mathsf{Set})\;(T:A\longrightarrow\mathsf{Tel})\to\mathsf{Tel}
                                                                                                                     [\![A::T]\!]_T = \Sigma[a:A][\![Ta]\!]_T
347
                                    _+ : (T : \mathsf{Tel}) (U : [\![T]\!]_\mathsf{T} \to \mathsf{Tel}) \to \mathsf{Tel}
                                                                                                                    [T + U]_T = \Sigma[t : [T]_T] [U t]_T
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                                         Fig. 3. (Tree-shaped) telescopes and their semantics as nested \Sigma-types
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```

```
\begin{array}{lll} \operatorname{Curried_T}: (T:\operatorname{Tel}) \to ([\![T]\!]_\mathsf{T} \to \operatorname{Set}) \to \operatorname{Set} \\ \operatorname{Curried_T}[\!] & X = X \operatorname{tt} \\ \operatorname{Curried_T}(A::T) \ X = (a:A) \to & \operatorname{Curried_T}(Ta) \ (\lambda \ t \to X \ (a \ , t)) \\ \operatorname{Curried_T}(T + U) \ X = \operatorname{Curried_T}(\lambda \ t \to \operatorname{Curried_T}(U \ t) \ (\lambda \ u \to X \ (t \ , u))) \end{array}
```

Fig. 4. Curried function types from telescopes

Fig. 5. Parametrised datatype descriptions and lifted base functors

the types in the front of a telescope into the context of the rest of the telescope. Besides the usual cons constructor '::', we also include a constructor '::' for appending telescopes (which requires indexed induction-recursion [Dybjer and Setzer 2006] to define), making our telescopes tree-shaped. This allows us to combine tuples $t: [T]_T$ and $u: [Ut]_T$ directly into $(t, u): [T + U]_T$, from which we can still easily retrieve t and u because we did not insist on flattening t + t (to use only '::') and re-associating the possibly deeply nested tuple t0 to the right. We will see how this structure is useful to generic libraries when we reach Section 6.1. A couple of syntax declarations will make telescopes slightly easier to write and read:

```
syntax := A (\lambda x \rightarrow T) = [x : A] T; syntax = T (\lambda t \rightarrow U) = [t : T] U
```

For example, the parameters of Acc can be represented as [A:Set] $[R:(A \to A \to Set)]$ [] instead of Set $::(\lambda A \to (A \to A \to Set) ::(\lambda R \to []))$. From a telescope T it is straightforward to compute a curried function type Curried $_T$ T X (Figure 4) which has arguments with the types in T, and ends with a given type $X:[\![T]\!]_T$ \to Set that can refer to all the arguments (collectively represented as a tuple of type $[\![T]\!]_T$). It is also straightforward to convert between this curried function type and its uncurried counterpart with the functions $\operatorname{curry}_T:((t:[\![T]\!]_T) \to X$ $t) \to \operatorname{Curried}_T$ T and uncurry $_T$ in the opposite direction (whose definitions are omitted).

With telescopes, we can now define the type PDataD of parametrised datatype descriptions in Figure 5, refining the parameter and index sets P and I into telescopes. For example, the Acc datatype can now be described by

```
AccD : PDataD

AccD = record

{Param = [A : Set] [R : (A \rightarrow A \rightarrow Set)] []

; Index = \lambda (A, R, \_) \rightarrow [\_ : A] []

; applyP = \lambda (A, R, \_) \rightarrow (\sigma[x : A] \rho (\pi[y : A] \pi[\_ : R y x] \iota (y, tt)) (\iota (x, tt))) :: []}}
```

Fig. 6. Datatype connections

When accessing the fields in the PDataD structures, the postfix projection syntax works better, as shown in the definitions of base functor lifted pointwise to the PDataD layer in Figure 5 (which we will use later).

3.2 Datatype Connections

 When μ was present, generic programs only needed to take a description D: PDataD as input, and the corresponding native datatype would simply be μ D. Without μ , a corresponding native datatype N needs to be passed as an additional argument, and the first issue is the type of N: the native datatype is usually in a curried form, but it is easier for generic programs to handle an uncurried form, which can be computed by PDataT D as defined in Figure 6. Regardless of how many parameters and indices there actually are, this uncurried form always represents parameters and indices as two arguments ps and s, presenting a uniform view to generic programs. The conversion from a curried form to the uncurried form is purely cosmetic and can be done with a 'wrapper' function, for example,

```
AccT : PDataT AccD
AccT (A, R, \_) (as, \_) = Acc R as
```

Note that AccT allows the form of the native datatype to be customised: we can change the order and visibility of the arguments (for example, the visibility of A is set to implicit in Acc) as long as we change AccT accordingly. Also, corresponding to the **con** constructor of μ , we need a function toN to construct inhabitants of N, and moreover, we need to perform pattern matching, which can be simulated by an inverse fromN of toN. These are packed into the record type PDataC of 'datatype connections' in Figure 6, replacing μ 's functionalities. (Strictly speaking, the inverse property fromN-toN here is only propositional whereas for **con** it is definitional, but this does not pose a problem for our examples in Section 6.) An inhabitant of PDataC D N performs invertible conversion between the branches of the sum structure in D with the constructors of N, and the conversion is highly mechanical — for example,

```
\begin{array}{lll} \mathsf{AccC} : \mathsf{PDataC} \; \mathsf{AccD} \; \mathsf{AccT} \\ \mathsf{AccC} &= \; \mathbf{record} \; \{\mathsf{toN} &= \lambda \; \{ (\mathsf{inl} \; (x \, , \mathit{as} \, , \mathsf{refl})) \; \rightarrow \; \mathsf{acc} \; x \; \mathit{as} \quad \} \\ ; \mathsf{fromN} &= \lambda \; \{ (\mathsf{acc} \; x \; \mathit{as}) & \rightarrow \; \mathsf{inl} \; (x \, , \mathit{as} \, , \mathsf{refl}) \} \\ ; \mathsf{fromN-toN} &= \lambda \; \{ (\mathsf{inl} \; (x \, , \mathit{as} \, , \mathsf{refl})) \; \rightarrow \; \mathsf{refl} \quad \} \\ ; \mathsf{toN-fromN} &= \lambda \; \{ (\mathsf{acc} \; x \; \mathit{as}) & \rightarrow \; \mathsf{refl} \quad \} \} \end{array}
```

Note that the order and visibility of constructor arguments can be customised as well.

The introduction of PDataC supports a symmetric architecture where generic and native entities may grow separately but can be kept in sync (reminiscent of 'delta-based bidirectional transformations' [Abou-Saleh et al. 2018, Section 3.3]): we may compute a new description from an old one and then manufacture a native datatype from the new description, or write a native datatype and then derive its description; in either case, a connection is established between the generic and native entities at the end. This architecture generalises the standard one involving μ , where D has a connection only with μ D, whereas in our architecture, connections can be established between

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```
record PFoldP: Set where field
                                                                PFoldT : PFoldP \rightarrow Set
                                                                PFoldT F = \forall ps \{is\} \rightarrow
   {Desc} : PDataD
                                                                   let open PFoldP F in Native (param ps) is \rightarrow Carrier ps is
   { Native } : PDataT Desc
          : PDataC Desc Native
                                                                record PFoldC (F : PFoldP) (f : PFoldT F) : Set where field
  Param : Tel
                                                                   equation:
  param : [Param]_T \rightarrow [Desc.Param]_T
                                                                      \forall \{ps \ is\} \rightarrow let \ open \ PFoldP \ F \ in
  Carrier : \forall ps \rightarrow [\![ Desc . Index (param ps) ]\!]_T \rightarrow Set
                                                                      (ns : [Desc]_{PD} (Native (param ps)) is) \rightarrow
  applyP : \forall ps \rightarrow Alg (Desc.applyP (param ps))
                                                                     f ps (Con.toN ns) \equiv applyP ps (fmap_{PD} Desc (f ps) ns)
                             (Carrier ps)
```

Fig. 7. Parametrised fold programs (algebras) and connections

any pair of description and datatype as long as they correspond. In particular, the forms of native datatypes and constructors (curried versus uncurried forms, order and visibility of arguments, etc) are not tightly coupled with descriptions (especially datatype-generically computed ones, which usually have prescribed forms) and can be customised by the programmer, which is vital in practice.

3.3 Parametrised Fold Programs and Fold Connections

foldAcc : $\{A : Set\} \{R : A \rightarrow A \rightarrow Set\} \{P : A \rightarrow Set\}$

Like descriptions, algebras can also be parametrised — in fact, foldAcc_Alg is Section 2 is already an algebra with two parameters P and p. Analogous to PDataD (Figure 5), we use the type PFoldP (for 'parametrised fold programs') defined in Figure 7 to store an algebra with its parameters. There are some additional fields that require explanation: PFoldP is designed to contain sufficient information for manufacturing a corresponding native fold function. The fold function needs a type, which refers to the native datatype on which the fold function operates, so PFoldP includes a field Con: PDataC connecting the datatype description Desc on which the algebra operates to a Native datatype, enabling us to compute the type of the fold function using PFoldT in Figure 7.6 For example, the fold operator of Acc

```
(p: \forall x \to (\forall y \to R \ y \ x \to P \ y) \to P \ x) \to \forall \{x\} \to \operatorname{Acc} R \ x \to P \ x foldAcc p (acc x as) = p \ x \ (\lambda y \ lt \to \operatorname{foldAcc} p \ (as y \ lt)) is encoded as the fold program foldAccP: PFoldP foldAccP = \operatorname{record} {Con = AccC; Param = [A: \operatorname{Set}] \ [R: (A \to A \to \operatorname{Set})] \ [P: (A \to \operatorname{Set})] \ [p: (\forall x \to (\forall y \to R \ y \ x \to P \ y) \to P \ x)] \ []; param = \lambda \ (A, R, P, p, \_) \to A, R, \operatorname{tt}; Carrier = \lambda \ (A, R, P, p, \_) \ (x, \_) \to P \ x
```

Following the same architecture, we are also going to connect algebras with native fold functions. In the definition of PFoldT, we also see that the field param is used to compute the parameters for the native datatype argument from the parameters of the fold function. So, given F: PFoldP, it can be connected to some f: PFoldT F, but what should the connection be? Since f is supposed to replace an instantiation of the generic fold operator, what we need to know about f is that it

; applyP = $\lambda \{(A, R, P, p, _) (inl(x, ps, refl)) \rightarrow p \times ps\}\}$

⁶Agda's **open** statement can be used to bring the fields of an inhabitant of a record type into scope — for example, the name Native in the definition of PFoldT stands for *F* .Native because of **open** PFoldP *F*. Moreover, an **open** statement can be used in a **let**-expression to limit its effect to the body of the **let**-expression.

 satisfies a suitably instantiated version of the defining equation of fold. This equation constitutes the only field of the record type PFoldC in Figure 7. Proofs of the equation are usually by definition — for example, foldAccP and foldAcc are connected by

```
foldAccC : PFoldC foldAccP foldAccT foldAccC = record {equation = \lambda {(inl (x, as, refl)) \rightarrow refl}} where foldAccT (A, R, P, p, _) = foldAcc p is a wrapper function.
```

Although we do not present the details of generic induction, the definitions are largely the same as those for folds, including IndP, IndT, IndC, etc. When we get to examples that require induction in Section 6, it should suffice to think of those generic programs as a more complex kind of parametrised algebras.

4 ESTABLISHING CONNECTIONS USING ELABORATOR REFLECTION

As explained at the end of Section 3.2, our framework can be thought of as keeping native and generic entities in sync through connections. This syncing can be tedious: whenever we write a native datatype N, we need to produce its description D, a wrapper T around N, and a connection between D and T; conversely, whenever we compute a new datatype description, we also need to produce the corresponding native datatype, wrapper, and connection; and the same goes for instantiating generic programs as native functions. Fortunately, such tasks can be automated by a set of metaprograms supplied by our framework.

Our metaprograms are based on Agda's elaborator reflection (Section 4.1), which provides a few important features that make some of the tasks much easier to accomplish than with traditional metaprogramming. We will focus on two particularly noteworthy examples (Sections 4.2 and 4.3): The first is the translation from our datatype descriptions —which use a higher-order representation of binders— to the elaborator reflection API's first-order representations —which use de Bruijn indices—for generating a native datatype. Surprisingly, we are able to avoid the tedious and error-prone manipulation of de Bruijn indices using the 'local variable creation' technique [Nanevski and Pfenning 2005; Schürmann et al. 2005], which is easily supported by elaborator reflection. The second is the instantiation of fold programs as native fold functions. The elaborator reflection API exposes open-term normalisation as a primitive, which we can directly use for non-trivial computation such as expanding generic definitions of function types and partially evaluating function bodies, without having to implement heavyweight term transformations in the metaprogram.

4.1 Elaborator Reflection in Agda

The elaborator reflection API is based on a set of datatypes reflecting Agda's core language. Among these datatypes, the one we will see most frequently is Term, the datatype of first-order representations of core expressions. The quotation of an expression e can be obtained as **quoteTerm** e: Term; this syntax makes it easy to produce concrete examples of Term — for example, we may evaluate **quoteTerm** $((A : Set) \rightarrow Vec \ A \ zero)$ and get⁷

```
pi (agda-sort (lit 0)) (def (quote Vec) (var 0 [] :: con (quote zero) [] :: [])) : Term
```

where the structure of the type expression is turned into the composition of several of the Term constructors, namely

- $pi : Term \rightarrow Term \rightarrow Term$, which represents a dependent function type,
- agda-sort : Sort \rightarrow Term, where Sort is the datatype representing sorts such as Set and Set ℓ ,

 $^{^{7}}$ In fact there is some additional information embedded in Terms such as argument visibility (being implicit or not) and binder names, which we suppress in our presentation for brevity.

- def: Name → List Term → Term and con: Name → List Term → Term, representing the application of a top-level definition or constructor (referred to by a quoted name of the form quote n: Name in this example) to a list of arguments,
- $var : \mathbb{N} \to List Term \to Term$, which is similar to **def** and **con** except that the first argument is a variable in the form of a de Bruijn index.

The central component of elaborator reflection is the elaborator monad TC (short for 'type-checking monad' [Agda Team 2022a]), which stores states needed for elaboration such as the context of the call site, the scope of names with its definition, and the set of metavariables. Metaprograms take the form of TC computations, and have access to a set of primitive operations used during elaboration — for example, the primitive unify: Term \rightarrow Term \rightarrow TC \top unifies two given terms and solves some of the metavariables (thereby changing the elaborator state).

At elaboration time, we can run a metaprogram and splice an expression into the source file. A more convenient way to do so is to use macros, a special kind of metaprograms of type $A_1 \to \cdots \to A_n \to \mathsf{Term} \to \mathsf{TC} \top \mathsf{declared}$ with the keyword **macro**. During elaboration, the call site of a macro M becomes a metavariable x, which is represented as **meta** x []: Term and supplied as the last argument of M for manipulation inside M. A minimal example is

```
macro give = unify
```

Elaborating give e will splice the given expression e: Term in place of the call, if Agda does nothing to its argument e. In fact, Agda quotes macro arguments of types Term and Name upon invocation, so elaborating give e amounts to running unify (**quoteTerm** e) (**meta** x []). Afterwards, the call site becomes e (and is elaborated again). In general, we can compute whatever expression we need inside a macro and then place it at the call site by unifying it with x.

Another way to use a metaprogram is to compute a top-level function by

```
unquoteDecl f = \cdots
```

where f is the function being defined, and the right-hand side calls the metaprogram. This metaprogram usually needs to take the name f as an argument, so that —analogously to how we write functions by hand—it can declare the type of the function using declareDef: Name \rightarrow Term \rightarrow TC \top and give the definition of the function using defineFun: Name \rightarrow List Clause \rightarrow TC \top (where Clause is the datatype of reflected function clauses).

We have extended the **unquoteDecl** mechanism to allow metaprograms to define datatypes as well. The extended syntax is

```
unquoteDecl data d constructor c_1 \ldots c_n = \cdots
```

which introduces a datatype d and its constructors c_1, \ldots, c_n into the scope. The definitions of d and c_1, \ldots, c_n are supplied by the metaprogram using the new primitives declareData and defineData, whose details are omitted from the presentation.

Below we look at some actual metaprograms for generating the components in our framework.

4.2 Translating Higher-Order Representations with Local Variable Creation

Our first task is to translate PDataD, a fully typed higher-order representation, into the reflected language to declare native datatypes. The reflected language is, by contrast, a uni-typed first-order representation using de Bruijn indices and *not* hygienic, posing a challenge. Rather than presenting the full detail, it suffices to see how telescopes (Figure 3) are handled to get the essence of the translation. For example, the tree-shaped telescope $\|[(A,_):[_:Set][]]\|[_:(A\to A)][]$ of type Tel should be translated to this (flattened) list of reflected types

 where the type Telescope is a part of the elaborator reflection API and stands for List Term (and the variables var 0 [] and var 1 [] in pi both refer to the quotation 'Set = agda-sort (lit 0) of Set).

One obvious approach is to analyse the quotation of the input Tel. Such a macro needs to analyse abstract syntax trees *modulo judgemental equality* — it has to check which case is being analysed by reducing, say, a reflected expression \mathbf{def} (\mathbf{quote} f) xs for a definition f to one of the three cases \mathbf{con} (\mathbf{quote} Tel.[]) [], \mathbf{con} (\mathbf{quote} Tel._::_) xs, and \mathbf{con} (\mathbf{quote} Tel._++_) xs. This approach is error-prone, and moreover, Agda fails to check if the macro terminates or not.

Instead, let us analyse the input Tel by pattern matching. The case for the empty telescope [] is simple to define but the other two cases A :: T and U + V seem impossible:

```
fromTel : Tel \rightarrow TC Telescope
fromTel [] = return []
fromTel (A :: T) = ... fromTel (T?) ...
fromTel (U + V) = ... fromTel (V?) ...
```

Note that T is a function from A, and V a function from $[\![U]\!]_T$, for some $arbitrary\ A$ and U; how do we give their arguments? We solve this problem by creating a $local\ variable$. The TC monad stores the context during elaboration, which can be extended by a variable of a given type to run a TC computation locally, using the primitive extendContext: Type \to TC $A \to$ TC A. The first argument of extendContext is a reflected type, which should be the quotation of the actual value of A, but this value is not known until elaboration and thus cannot be obtained by $\mathbf{quoteTerm}$; to obtain the quotation, we use the primitive quoteTC: $A \to$ TC Term. Then, in the second argument of extendContext, the local variable can be actually brought into scope by another primitive unquoteTC: Term \to TC A which unquotes a given reflected expression within the TC monad. The above construction amounts to a TC computation

```
exCxtT : (B : Set) \rightarrow (\mathsf{Type} \rightarrow B \rightarrow \mathsf{TC} \ A) \rightarrow \mathsf{TC} \ A
exCxtT Bf = \mathbf{do} \ `B \leftarrow \mathsf{quoteTC} \ B
extendContext `B \ (\mathsf{unquoteTC} \ (\mathsf{var} \ 0 \ []) \gg \lambda \ x \rightarrow f \ `B \ x)
```

which creates a local variable x of type B for use in a TC computation f. This function can be generalised to extend the context with a telescope:

```
\begin{array}{ll} \operatorname{exCxtTel}: (T:\operatorname{Tel}) \ (f: \llbracket T \rrbracket_{\mathsf{T}} \to \operatorname{TC} A) \to \operatorname{TC} A \\ \operatorname{exCxtTel} \ [] \qquad \qquad f = f \ \operatorname{tt} \\ \operatorname{exCxtTel} \ (A::T) \ f = \operatorname{exCxtT} \ A \ (\lambda \_ x \to \operatorname{exCxtTel} \ (Tx) \ \lambda \ t \to f \ (x,t)) \\ \operatorname{exCxtTel} \ (T+U) \ f = \operatorname{exCxtTel} \ T \ (\lambda \quad t \to \operatorname{exCxtTel} \ (Ut) \ \lambda \ u \to f \ (t,u)) \end{array}
```

As for $[\![U]\!]_T$, if we merely created a local variable $u:[\![U]\!]_T$, then each reference to a component of u would be formed by projections fst and snd. For example, instead of (1) we would have

```
'Set :: pi (def (quote fst) ((var 0 []) :: [])) (def (quote fst) ((var 1 []) :: [])) :: []
```

To eliminate projections, we use exCxtTel to create a list of local variables for each type in U: Tel as a tuple. The last two cases of fromTel can then be defined by

```
\begin{array}{lll} \text{fromTel } (A :: T) &=& \textbf{do} & \text{fromTel } (U +\!\!\!\!\! + V) &=& \textbf{do} \; \Gamma \leftarrow \text{fromTel } U \\ &=& \text{exCxtT} \; A \; \lambda \; `A \; x \; \rightarrow \; \textbf{do} & \text{exCxtTel } U \; \lambda \; u \; \rightarrow \; \textbf{do} \\ &\Gamma \leftarrow \text{fromTel } (T \; x) & \Delta \leftarrow \text{fromTel } (V \; u) \\ &=& \text{return } (`A :: \Gamma) & \text{return } (\Gamma +\!\!\!\!\! + \Delta) \end{array}
```

As long as local variables are not pattern-matched, the computation can proceed. Indeed, we are exploiting the fact that our representations are used as if they are higher-order abstract syntax!

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685 686 The approach scales well for our metaprogram that defines a native data type from a description. Conversely, we also have a macro that expands to a description of a given native data type. This direction is syntactical and unsurprising though — for example, telescopes are handled simply by

```
to Tel : Telescope \rightarrow Term
to Tel = foldr (\lambda 'A 'T \rightarrow 'A ':: 'T) '[]
```

where '[] = con (quote Tel.[]) [] and 'A ':: ' $T = \text{con (quote Tel.}_{::}$) ('A :: lam ' T :: []) (using lam : Term \rightarrow Term to construct a reflected λ -expression).

4.3 Instantiating Fold Programs with Normalisation

In this paper we are only interested in the direction of manufacturing native fold functions from PFoldP, leaving the opposite direction as future work. We will first work out a concrete example manually (Section 4.3.1) while developing some generic definitions used by the metaprogram definePFold (Section 4.3.2). The full detail of definePFold is not shown since it requires a deeper understanding of the reflection API, but the metaprogram is essentially just a careful formalisation of the manual process.

4.3.1 Instantiation by Hand. As a concrete example, let us manufacture from the fold program foldAccP in Section 3.3 a variant of the fold operator for Acc (whose arguments are all made explicit to avoid the complication of handling argument visibility):

```
foldAcc': (A : Set) (R : A \rightarrow A \rightarrow Set) (P : A \rightarrow Set)

(p : \forall x \rightarrow (\forall y \rightarrow R \ y \ x \rightarrow P \ y) \rightarrow P \ x) \rightarrow \forall x \rightarrow Acc \ R \ x \rightarrow P \ x

foldAcc' A \ R \ P \ p \ x \ (acc \ x \ as) = p \ x \ (\lambda \ y \ lt \rightarrow foldAcc' \ A \ R \ P \ p \ y \ (as \ y \ lt))
```

First we need a curried type for the fold function, which can be computed by a variant of PFoldT that uses Curried_T (Figure 4):

```
PFoldNT: PFoldP \rightarrow Set

PFoldNT F = let open PFoldP F in

Curried<sub>T</sub> Param \lambda ps \rightarrow Curried<sub>T</sub> (Desc .Index (param ps)) \lambda is \rightarrow

Native (param ps) is \rightarrow Carrier ps is
```

Simply normalising PFoldNT foldAccP gives the type of foldAcc'.

As for the definition of foldAcc', it should satisfy the equation of PFoldC foldAccP foldAccT' (Figure 7) —where foldAccT' (A, R, P, p, p) $\{x, y\}$ = foldAcc' ARPPX is a wrapper function—but the equation does not work directly as a definition because toN is not a constructor. We can, however, change toN on the left-hand side to fromN on the right-hand side to get a definition, which we write as

```
foldAcc' A R P p x a = fold-base foldAccP foldAcc' A R P p x a
```

where fold-base generically expresses the computation pattern of fold functions in the usual non-recursive form that abstracts the recursive call as an extra argument *rec*:

```
fold-base : (F : \mathsf{PFoldP}) \to \mathsf{FoldNT} F \to \mathsf{FoldNT} F

fold-base F \ rec =

let open \mathsf{PFoldP} F \ \mathbf{in} \ \mathsf{curry}_\mathsf{T} \ \lambda \ ps \to \mathsf{curry}_\mathsf{T} \ \lambda \ is \to

\mathsf{applyP} \ ps \circ \mathsf{fmap}_\mathsf{PD} \ \mathsf{Desc} \ (\lambda \ \{is\} \to \mathsf{uncurry}_\mathsf{T} \ (\mathsf{uncurry}_\mathsf{T} \ rec \ ps) \ is) \circ \mathsf{Con} \ .\mathsf{fromN}
```

definePFold F f = do

declareDef f 'type
-- step (ii)

;_

-- step (iii)

-- step (i)

```
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734 735 This definition of foldAcc', albeit one deemed non-terminating by Agda, implies the PFoldC.equation because of the inverse property PDataC.fromN-toN. To turn this into a valid definition, we patternmatch the variable *a* with all the possible constructors, although there is only one in this case:

Fig. 8. The metaprogram definePFold for generating a top-level fold function from a fold program

 $cls \leftarrow caseM$ (getDefinition = PFoldPToNativeName F) of λ { (data-type pars cs) \rightarrow for M cs (conClause pars)

 \rightarrow typeError [] }

$$foldAcc' A R P p . x (acc x as) = fold-base foldAccP foldAcc' A R P p x (acc x as)$$
 (2)

Now normalise the right-hand side:

```
foldAcc' A R P p .x (acc x as) = p x (\lambda y lt \rightarrow foldAcc' A R P p y (as y lt))
```

 $\mathsf{definePFold} : \mathsf{PFoldP} \, \to \, \mathsf{Name} \, \to \, \mathsf{TC} \, \top$

'type ← normalise \implies quoteTC (PFoldNT F)

defineFun $f \ll mapRHS$ normalise cls

In general, this final definition will pass the termination check if Con .fromN works normally by breaks its input into structurally smaller pieces, in which case the $fmap_{PD}$ Desc part in fold-base applies rec to those smaller pieces. And the connecting equation always holds definitionally:

```
foldAccC' : PFoldC foldAccP foldAccT' foldAccC' = record {equation = \lambda {(inl (x, as, refl)) \rightarrow refl}}
```

(The inverse property DataC.fromN-toN does not appear in the proof, but we need it at the meta-level to argue generically that the proof always works.)

4.3.2 Instantiation by a Metaprogram. To formalise the manual process in Section 4.3.1, we define a TC computation define PFold (Figure 8) to instantiate a given F: PFoldP as a native function f by (i) generating the instantiated type using PFoldNT, (ii) generating a clause for each constructor of the datatype specified in F, and (iii) normalising these clauses.

Step (i) is straightforward because the type of f can be obtained directly by normalising PFoldNT F. We use quoteTC to turn PFoldNT F into a Term, normalise it, and finally declare the normalised term as the type of f using declareDef.

For step (ii), we use the primitive getDefinition : Name \to TC Definition (where Definition is the datatype of reflected definitions of datatypes, record types, functions, etc) to get the list cs of constructor names of the datatype F. Native, and generate a clause for each constructor. First we need to explain how clauses are handled in a bit more detail: A clause takes the form $\Delta \vdash \overline{p} \hookrightarrow e$ where \overline{p} is a list of patterns and e a reflected expression; the types of the variables in \overline{p} are specified in the context Δ . It may appear that the context Δ needs to be fully specified beforehand, but actually it is not the case. This is because the reflected language plays the dual role of unchecked input and checked output of the elaborator [Cockx and Abel 2020]: the context of a checked clause is fully specified, whereas the context of an unchecked clause, which has the form $\overline{p} \hookrightarrow e$, can simply be filled with unsolved metavariables represented by **unknown**: Term.

Back to the clauses we should generate for f: Abstracted from (2), each clause has the form

$$\Delta \vdash \overline{p} \ \overline{x} \ (c_i \ \overline{a}) \hookrightarrow \text{fold-base} \ F \ f \ \overline{p} \ \overline{x} \ (c_i \ \overline{a}) \tag{3}$$

where \overline{p} the parameters, \overline{x} the indices, and \overline{a} the constructor arguments. In general, the context Δ needs to be given because it will be used by the elaborator to determine the types of the variables when the right-hand side of (3) is normalised. But in this case only the length of Δ needs to be specified, since all types will be determined upon synthesising the type of the right-hand side of (3), which is the first step of normalisation. Moreover, the patterns \overline{x} are forced since the values of indices is determined by pattern-matching with c_i \overline{a} . It follows that \overline{x} can be given as $\overline{\mathbf{unknown}}$ on both sides. Therefore, we only need to generate a simpler clause

$$\overline{p}$$
 .unknown $(c_i \overline{a}) \hookrightarrow \text{fold-base } F f \overline{p}$ unknown $(c_i \overline{a})$ (4)

for each constructor c_i .

For step (iii), we normalise the right-hand side of (4) within a context extended with variables from the left-hand side of (4) and obtain the list of clauses with normalised expressions on the right-hand side as the result of mapRHS normalise cls. Finally, we define f by the primitive defineFun, supplying the clauses we just constructed.

Following the same approach to generating functions, the remaining components of our framework, namely PDataT, PDataC, PFoldT, and PFoldC, can also be automatically generated from native datatypes and functions by metaprograms.

5 UNIVERSE POLYMORPHISM

We have assumed a single universe Set satisfying Set: Set in our presentation so far, but it is well known that this assumption leads to logical inconsistency, to avoid which we must use a hierarchy of universes $Set_0: Set_1: Set_2: \cdots$ instead [Martin-Löf 1975]. General library components are usually reusable across the hierarchy — for example, the most general form of Acc is

data Acc
$$\{\ell \ \ell' : \text{Level}\}\ \{A : \text{Set} \ \ell\}\ (R : A \to A \to \text{Set} \ \ell') : A \to \text{Set} \ (\ell \sqcup \ell') \text{ where }$$

 $acc : (x : A) \to ((y : A) \to R \ y \ x \to \text{Acc} \ R \ y) \to \text{Acc} \ R \ x$

where the universe levels ℓ and ℓ' in the types of A and R can be arbitrary. Our framework ought to support such *universe-polymorphic* datatypes given their prevalence in Agda.

Agda supports universe polymorphism through a currently unique system where (finite) universe levels are internalised and have a type Level: Set. Levels can be constructed with the primitives lzero: Level and lsuc: Level \rightarrow Level in the same way as natural numbers are constructed, but we cannot pattern-match levels with lzero and lsuc. There is also an operator $_\sqcup_$: Level \rightarrow Level \rightarrow Level that computes the maximum of two levels. The internalisation makes it convenient to encode universe-polymorphic datatypes - just put levels inside datatype descriptions. It is also possible to compute new datatypes where the universe levels are the results of non-trivial computation, which can be reasoned about. Besides showing that it is possible to deal with universe-polymorphic entities generically, our encoding in this section also serves as a practical justification of (Agda's) first-class universe levels, as we make essential use of the capabilities of computing and reasoning about levels internally.

Our framework makes a simplifying assumption that holds for common universe-polymorphic datatypes (for example Acc above): we assume that there is a list of level parameters separate from other ordinary parameters, and only these level parameters are involved in universe polymorphism. Under this assumption, to describe a possibly universe-polymorphic datatype, we start with a number $n: \mathbb{N}$ of level parameters, from which we can compute a type Level n of tuples of n levels (as defined by n n zero n n and n (suc n) n0, and then provide a function of type Level n0 n0 PDataD, which brings n1 level parameters into the scope of the definition of a PDataD. We create a new description layer DataD for level parametrisation, which is shown in Figure 9 along with the existing four layers adapted to accommodate levels, to be explained

```
record DataD : Set a, where field
                                                          data ConDs (I : Set \ \ell_I) : ConBs \rightarrow Set \ \omega where
   \#levels : \mathbb{N}
                                                             [] :
                                                                                                              ConDs I
                                                             :: : ConD I cb \rightarrow ConDs I cbs \rightarrow ConDs I (cb :: cbs)
   applyL : Level ^ #levels → PDataD
record PDataD : Set ω where field
                                                          data ConD (I : Set \ \ell_I) : ConB \rightarrow Set \ \omega where
   alevel {plevel} {ilevel} : Level
                                                                                                                \rightarrow ConD I []
                                                             \sigma: (A: Set \ \ell) \rightarrow (A \rightarrow ConD \ I \ cb) \rightarrow ConD \ I \ (inl \ \ell :: cb)
    {struct} : ConBs
                                                             \rho : \text{RecD } I \ rb \rightarrow
                                                                                              ConD I cb \rightarrow ConD I (inr rb :: cb)
   level-ineq : \maxMap \max-\pi struct \sqcup
                   maxMap max-σ struct
                                                          data RecD (I : Set \ \ell_I) : RecB \rightarrow Set \ \omega where

    □ alevel    □ ilevel

   Param : Tel plevel
                                                             \pi: (A: \mathsf{Set}\ \ell) \to (A \to \mathsf{RecD}\ I\ rb) \to \mathsf{RecD}\ I\ (\ell::rb)
   Index : [Param]_T \rightarrow Tel ilevel
   applyP : (ps : [Param]_T) \rightarrow ConDs [Index ps]_T struct
```

Fig. 9. Universe-polymorphic datatype descriptions (all five layers), where RecB = List Level, ConB = List (Level ⊎ RecB), and ConBs = List ConB (Modifications to Figures 1 and 5 are highlighted.)

```
mutual
```

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Fig. 10. Universe-polymorphic telescopes (Modifications to Figure 3 are highlighted.)

below in Sections 5.1 to 5.3. The rest of our framework are similarly adapted, so there are also definitions of DataT, DataC, FoldP, FoldT, and FoldC, which we omit from the presentation. Finally, the metaprograms in Section 4 are extended to treat level parameters, but due to the current limit of Agda's universe polymorphism, the treatment is different from that of ordinary parameters, and is briefly sketched in Section 5.4.

5.1 Telescopes

Before we adapt the descriptions, telescopes need to be adapted too, as shown in Figure 10. The core change is made to the type of the first argument of "::", from Set to Set ℓ . We also add an index of type Level to Tel such that $T: \text{Tel } \ell$ implies that the maximum level appearing in T is ℓ . This maximum level is important since it is the universe level of the type $[\![T]\!]_T$, which is a nested Σ -type inhabited by tuples whose components have the types in T. Since the elements of a telescope can now be Sets of arbitrary finite levels, the level of the type of the telescope itself has to be greater than all those levels, and is set as the first infinite level ω here.

5.2 Sum-of-Products Descriptions

Next we adapt the description datatypes ConDs, ConD, and RecD. A first instinct might be copying what has been done for Tel (as constructor descriptions can be viewed as a slightly more complex kind of telescopes), enriching the Set-arguments to Set ℓ and perhaps indexing the datatypes with the maximum level, but this is not enough: the range of definitions depending on Tel (such as $\llbracket _ \rrbracket_T$ and Curried_T) is limited and requires only the computation of the maximum level, so indexing suffices; on the other hand, generic libraries built on the description datatypes may construct whatever they want from descriptions, and the need for non-trivial level computation will naturally arise if those constructions are universe-polymorphic. For example, in Sections 6.2 and 6.3 we will compute new universe-polymorphic datatypes from old ones, and will need to specify the new levels in terms of the old ones (and even reason about them). For a concrete example we can look at now, consider how the type of a base functor $\llbracket D \rrbracket_{Cs}$ should be enriched: One place where we use

 the base functor is the type of an algebra $\{i:I\} \to [\![D]\!]_{Cs} X i \to X i$ where $X:I \to \operatorname{Set} \ell$ is the result type, which can have any level depending on what the algebra computes, so ℓ should be universally quantified in the type of $[\![D]\!]_{Cs}$. But then, what should the level of the type $[\![D]\!]_{Cs} X i$ be? This level —call it ℓ' — needs to be computed from ℓ and the structure of D, and the computation is non-trivial — for example, if D is $[\![D]\!]_{Cs} X i = \bot$, in which case ℓ' is simply lzero; if D is non-empty, then ℓ may or may not appear in ℓ' , depending on whether there is a constructor with a ρ -field or not.

To allow level computation to be performed as freely as possible, we choose to index the description datatypes with as much useful information as possible. The index in the type of a description is a list which not only contains the levels of the fields but also encodes the description constructors used. Starting from the simplest RecD datatype, we index it with RecB = List Level, recording the levels of the π -fields. For ConD, the index type is ConB = List (Level \oplus RecB), whose element sum type is used to record whether a field is σ or ρ . Finally, ConDs is indexed with ConBs = List ConB, collecting information from all the constructors into one list. With some helper functions, which constitute a small domain-specific language for datatype level computation, we can now specify the output level of $[\![\]\!]_{Cs}$:

In prose, the output level is the maximum among the maximum level of the π -fields, the maximum level of the σ -fields, ℓ if the description has a ρ -field, and ℓ_I if the description has a constructor.

For our constructions, the approach works surprisingly well (even though the level expressions may look somewhat scary sometimes): we are able to write fully universe-polymorphic types while keeping almost all of the programs as they were — for example, the universe-polymorphic program of $\llbracket _ \rrbracket_{\text{Cs}}$ is exactly the same as the non-universe-polymorphic one in Figure 1. To see how the universe-polymorphic version of $\llbracket _ \rrbracket_{\text{Cs}}$ is type-checked, we need to show a couple of definitions:

```
\begin{array}{ll} \max \operatorname{Map}: (A \to \operatorname{Level}) \to \operatorname{List} A \to \operatorname{Level} & \operatorname{hasCon?}: \operatorname{Level} \to \operatorname{ConBs} \to \operatorname{Level} \\ \max \operatorname{Map} f \ [] &= \operatorname{Izero} & \operatorname{hasCon?} \ \ell = \max \operatorname{Map} \ (\lambda_- \to \ell) \\ \max \operatorname{Map} f \ (a :: as) &= f \ a \sqcup \max \operatorname{Map} f \ as \end{array}
```

It is easy to see that the output level in the $[\![[] \!]\!]_{Cs}$ case is lzero, which is indeed the level of \bot . In the $[\![D :: Ds]\!]_{Cs}$ case where $D: ConD \ I \ cb$ and $Ds: ConDs \ I \ cbs$, the output level expands to

```
\max \pi cb \sqcup \max \sigma cb \sqcup \operatorname{hasRec} \ell cb \sqcup \ell_I \sqcup
```

```
maxMap max-π cbs \sqcup maxMap max-σ cbs \sqcup maxMap (hasRec? ℓ) cbs \sqcup hasCon? ℓ_I cbs
```

where the first line is the level of $[\![D]\!]_{\mathbb{C}} X$ i and the second line is inductively the level of $[\![Ds]\!]_{\mathbb{C}s} X$ i, and indeed the level of the sum type is their maximum. It may appear that we skipped several steps applying the associativity and commutativity of ' \sqcup ', but in fact these properties (along with some others) are built into Agda's definitional equality on Level, so the definition of $[\![_]\!]_{\mathbb{C}s}$ type-checks without any manual proofs about levels.

5.3 Ordinary Parameters

The changes to PDataD should be mostly unsurprising except the new fields alevel and level-ineq, which make sure that a corresponding datatype definition would pass Agda's universe checker. Here we are using the simpler datatype level-checking rule employed when Agda's --without-K

option [Cockx et al. 2016] is turned on: the level of a datatype should at least be the maximum level of its index types, which is ilevel in our descriptions. If there are more components in the datatype level, they are specified in alevel, and the final datatype level is alevel $\ \ \ \$ ilevel. The datatype level is not uniquely determined by the content of the datatype —for example, we could define alternative versions of natural numbers at any level— but must be no less than the level of any π -or σ -field of the constructors; this is enforced by level-ineq, where the relation $\ell \sqsubseteq \ell'$ is defined by $\ell \sqcup \ell' \equiv \ell'$. With level-ineq, we could even define a universe-polymorphic version of the μ operator (Section 2), so even the traditional approach to datatype-genericity could be extended to incorporate universe polymorphism. In general, the ability to manipulate and reason about levels internally seems crucial to datatype-genericity, because computation of universe-polymorphic datatype descriptions—in particular the levels in the descriptions—can be arbitrarily complex, and it may no longer be feasible to infer levels or check level constraints automatically as can be done for specific datatypes in languages with typical ambiguity [Sozeau and Tabareau 2014].

5.4 Level Parameters

 Conceptually, level parameters could be treated in the same way as ordinary ones — in particular, Level $^{\land}$ n could be thought of as a kind of specialised telescope, so we could start with a construction similar to Curried_T (Figure 4), computing a curried function type with n level quantifications:

```
\begin{aligned} & \mathsf{Curried}_{\mathsf{L}} : (n : \mathbb{N}) \; \{ f : \mathsf{Level} \, ^{\wedge} \, n \, \to \, \mathsf{Level} \, \} \, \to ((\ell s : \mathsf{Level} \, ^{\wedge} \, n) \, \to \, \mathsf{Set} \; (f \; \ell s)) \, \to \, \mathsf{Set} \; ? \\ & \mathsf{Curried}_{\mathsf{L}} \; \mathsf{zero} \quad X \, = \, X \; \mathsf{tt} \\ & \mathsf{Curried}_{\mathsf{L}} \; (\mathsf{suc} \; n) \; X \, = \; (\ell : \mathsf{Level}) \, \to \, \mathsf{Curried}_{\mathsf{L}} \; n \; (\lambda \; \ell s \, \to \, X \; (\ell \; , \, \ell s)) \end{aligned}
```

However, the hole '?' is problematic since it should be a finite level when n is zero (meaning that there is no level quantification), or ω when n is non-zero, but currently Agda's universe polymorphism supports only finite levels. To produce curried types with level quantifications, we have to operate at the meta-level (evading the type-checker) and generate the level quantifications in their syntax trees in relevant metaprograms such as defineFold, of which definePFold (Figure 8) is a cut-down version. The need for the special treatment is one reason that we separate level parameters from ordinary ones instead of allowing the two kinds of parameters to mix freely.

6 EXAMPLES

As a demo of our framework, here we provide some samples of generic constructions that should have been made available to the Agda programmer. To be more precise, these constructions are not new (or not too novel compared to those in the literature), but they have not been in the main toolbox of the Agda programmer, who prefers to work with native datatypes and functions; our framework makes it possible to instantiate these constructions for native entities. We will omit the details except those related to the design of our framework, and briefly discuss mechanisms that could make these constructions more convenient to use.

6.1 Fold Operators

The generic program that instantiates to fold operators on native datatypes is given the type

```
fold-operator : (C : DataC D N) \rightarrow FoldP
```

As an example of instantiating the generic program, suppose that we have written the datatype Acc manually, and want to derive its fold operator. First we generate from Acc its description AccD, a wrapper AccT, and a datatype connection AccC between AccD and AccT by some macros:

```
AccD = genDataD Acc; AccT = genDataT AccD Acc; AccC = genDataC AccD AccT
```

 We can then use the connection AccC to instantiate fold-operator to the fold program

```
foldAccP = fold-operator AccC
```

from which the fold operator/function foldAcc can be manufactured by using the **unquoteDecl** syntax to introduce the name foldAcc into scope, and then invoking the metaprogram defineFold with the fold program and the newly introduced name:

unquoteDecl foldAcc = defineFold foldAccP foldAcc

Finally, we generate a fold connection between foldAccP and foldAcc by

```
foldAccC = genFoldC foldAccP foldAcc
```

which can be used to instantiate other generic programs (such as fold-fusion below) for foldAcc. The macros are manually invoked for now and might be invoked with wrong arguments, but usually the error messages are reasonable because the macros are typed — for example, checking genDataC AccD Acc will give rise to a standard error message saying that Acc does not have the type DataT AccD. (Perhaps the instantiation process could eventually be streamlined as, say, pressing a few buttons of an editor, which displays only valid options.)

It would not be too interesting if we could only manufacture functions but not prove some of their properties. For fold operators, one of the most useful theorems is the fusion theorem [Bird and de Moor 1997], of which foldr-fusion in Section 1 is an instance. The interface to the theorem is

```
fold-fusion : (C : DataC D N) (fC : FoldC (fold-operator C) f) \rightarrow IndP
```

where the fold connection fC is used to quantify over functions f that are fold operators of N. Suppose that we want to instantiate the theorem for the foldr operator on List in Section 1. Note that this version of foldr is different from the one manufactured by defineFold from the fold program foldListP = fold-operator ListC, where ListC is the datatype connection for List; in particular, the arguments of foldr are in a different order from that specified by foldListP. However, we can still connect foldr with foldListP by manually writing a wrapper to specify the argument order,

```
foldrT : FoldT foldListP
foldrT (\ell', \ell, tt) ((A, tt), B, e, f, tt) = foldr f e
```

and then generating a fold connection (using a variant of genFoldC that takes a wrapper)

```
foldrC = genFoldC' foldListP foldrT
```

Now we can obtain foldr-fusion in Section 1 by

unquoteDecl foldr-fusion = defineInd (fold-fusion ListC foldrC) foldr-fusion

and use foldr-fusion to prove, for example, the handy law that a map followed by a foldr (two list traversals) can be optimised as a foldr (a single list traversal),

```
 \begin{array}{c} \mathsf{foldr\text{-}map\text{-}fusion} : \{A: \mathsf{Set}\ \ell'\} \, \{B: \mathsf{Set}\ \ell''\} \, \\ (f: B \to C \to C)\ (e: C)\ (g: A \to B) \to \\ (as: \mathsf{List}\ A) \to \mathsf{foldr}\ f\ e\ (\mathsf{map}\ g\ as) \equiv \mathsf{foldr}\ (f\circ g)\ e\ as \\ \mathsf{foldr\text{-}map\text{-}fusion}\ f\ e\ g = \mathsf{foldr\text{-}fusion}\ (\mathsf{foldr}\ f\ e)\ \mathbf{refl}\ (\lambda\ \{\_\_\_\ \mathbf{refl}\ ) + \mathbf{refl}\}) \\ \end{array}
```

where map $g = \text{foldr } [] (\underline{::} \circ g)$. (The law could also be proved generically, but first we would need to give map a generic definition, which requires extra machinery not developed in this paper.)

In general, if the library user is not satisfied with the form of a manufactured function (argument order, visibility, etc), they can print the function definition, change it to a form they want, and connect the customised version back to the library in the same way as we treated foldr. This customisation can be tiresome if it has to be done frequently, however, and there should be ways to

get the manufactured forms right most of the time. We have implemented a cheap solution where argument name suggestions (for definition-printing) and visibility specifications are included in FoldP (and IndP) and processed by relevant components such as $Curried_T$, and the solution works well for our small selection of examples. More systematic solutions are probably needed for larger libraries though, for example name suggestion based on machine learning [Alon et al. 2019] and visibility calculation by analysing which arguments can be inferred by unification.

Let us switch to the perspective of the library programmer and look at fold-operator in a bit more detail. The ordinary parameters of fold-operator C are mainly a $[D]_D$ -algebra in a curried form, so the work of fold-operator is purely cosmetic: at type level, compute the types of a curried algebra, which are the curried types of the constructors of D where all the recursive fields are replaced with a given carrier, and at program level, uncurry a curried algebra. The level parameters of fold-operator C include those of D and one more for the carrier X appearing in the Param telescope shown below, which also contains the ordinary parameters of D and a curried algebra represented as a telescope computed (by FoldOpTel) from the list of constructor descriptions in D:

```
fold-operator \{D\} C .applyL (\ell, \ell s) .Param = \text{let } D_P = D .applyL \ell s in [ps: D_P] .Param [X: Curried_T (D_P]] .Index ps (\lambda_- \to Set \ell) [X: Curried_T (D_P]] .Index ps (X: Curried_T (D_P]] .Index ps (X: Curried_T (D_P]] .Index ps (X: Curried_T (D_P]] .Param (X: Curried_T (D_P]) .Index ps (X: Curried_T (D_P]) .Index ps (X: Curried_T (D_P]) .Index (X: Curried_T (D_P)) .Index (X
```

The type of X is in a curried form, which is then uncurried for FoldOpTel and other parts of the definition of fold-operator, for example the Carrier field:

```
fold-operator C .applyL (\ell, \ell s) .Carrier = \lambda (ps, X, calgs) \rightarrow uncurry_T X
```

This is a recurring pattern (which we first saw in Section 3.2): curried forms are exposed to the library user, whereas uncurried forms are processed by generic programs. The pattern is also facilitated by the telescope-appending constructor, which appears in Param above (disguised with the syntax $[\![\dots]\!]$): the parameters are instantiated in a curried form for the library user, but for generic programs they are separated into three groups ps, X, and calgs, making it convenient to refer to each group like in Carrier above.

6.2 Algebraic Ornamentation

Ornaments [McBride 2011] are descriptions of relationships between two structurally similar datatype descriptions, one of which has more information than the other. For example, after computing the descriptions ListD and NatD of List and $\mathbb N$ using genDataD, the following ornament states that List —having an additional element field— is a more informative version of $\mathbb N$:

```
ListO: DataO ListD NatD

ListO = record {level = \lambda_{-} \rightarrow tt; applyL = \lambda (\ell, _) \rightarrow record

{param = \lambda_{-} \rightarrow tt; index = \lambda_{-} \rightarrow tt; applyP = \lambda_{-} \rightarrow \iota :: :: (\Delta[_] \rho \iota \iota) :: :: :[]}}
```

Do not worry about the details — the point here is that it is not difficult to write ornaments between concrete datatypes (and it will be even easier if there is a (semi-)automatic inference algorithm [Ringer et al. 2019] or an editing interface showing two datatypes side by side and allowing the user to mark their differences intuitively).

The first thing we can derive from an ornament is

```
\mathsf{forget} \,:\, \mathsf{DataC}\,\,D\,M \,\to\, \mathsf{DataC}\,\,E\,N \,\to\, \mathsf{DataO}\,\,D\,E \,\to\, \mathsf{FoldP}
```

which instantiates to a forgetful function from M to N — if we instantiate forget with ListO,

```
lenP = forget ListC NatC ListO
```

 (where ListC and NatC are connections for List and \mathbb{N} respectively) the function manufactured from lenP will be list length (Section 1), which discards the additional element field.

More can be derived from special kinds of ornaments, with a notable example being 'algebraic ornaments'. In our formulation, given a fold program F: FoldP we can compute a more informative version of the description F. Desc and an ornament between them:

```
AlgD : FoldP \rightarrow DataD; AlgO : (F : FoldP) \rightarrow DataO (AlgD F) (F .Desc)
```

The new datatype described by AlgD F has the parameters of F and an extra index that is the result of the fold corresponding to F. For example, the following datatype of 'algebraic lists' [Ko 2021] can be manufactured from the description AlgD (fold-operator ListC):

```
data AlgList \{A: \operatorname{Set} \ell\} \{B: \operatorname{Set} \ell'\} (e:B) (f:A \to B \to B): B \to \operatorname{Set} (\ell \sqcup \ell') where
[] : AlgList \ ef \ e
::_{::}: (a:A) \to \forall \{b\} \to \operatorname{AlgList} \ ef \ b \to \operatorname{AlgList} \ ef \ (f \ a \ b)
```

But it is usually easier to program with more specialised datatypes such as Vec (Section 1), which is a standard example of algebraic ornamentation. To manufacture Vec, we use the **unquoteDecl data** syntax to introduce the names of the datatype and its constructors into scope, and invoke the metaprogram defineByDataD with the description VecD = AlgD lenP and the newly introduced names (where the constructor names are collected in a list):

```
unquoteDecl data Vec constructor [] _::_ = defineByDataD VecD Vec ([] :: _::_ :: [])
```

Then we generate a datatype connection between VecD and (a wrapper around) Vec so that we can instantiate generic programs for Vec later:

```
VecC = genDataC VecD (genDataT VecD Vec)
```

From the algebraic ornament VecO = AlgO lenP between Vec and List, in addition to a forgetful function fromV instantiated from forget VecC ListC VecO, we can also derive its inverse toV and the inverse properties:

```
fromV : Vec A n \to \text{List } A; toV : (as : \text{List } A) \to \text{Vec } A (length as)
from-toV : (as : \text{List } A) \to \text{fromV } (\text{toV } as) \equiv as
to-fromV : (as : \text{Vec } A n) \to (\text{length } (\text{fromV } as) , \text{toV } (\text{fromV } as)) \equiv_{\sum \mathbb{N} \text{ (Vec } A)} (n, as)
```

Note that, besides a new datatype Vec, we have derived an isomorphism between List A and Σ \mathbb{N} (Vec A) from the ornament ListO, allowing us to promote a natural number n to a list if a vector of type Vec A n can be supplied (or the other way around). In general, every ornament gives rise to such a 'promotion isomorphism' [Ko and Gibbons 2013]. A more interesting and notable example is the conversion between extrinsically and intrinsically typed λ -terms [Kokke et al. 2020]:

```
\begin{array}{lll} \mathbf{data} \ \Lambda : \ \mathsf{Set} \ \mathbf{where} & \mathbf{data} \ \_ \vdash_{-} : \ \mathsf{List} \ \mathsf{Ty} \ \to \ \mathsf{Ty} \ \to \ \mathsf{Set} \ \mathbf{where} \\ \mathbf{var} \ : \ \mathbb{N} \ \to \ \Lambda & \mathbf{var} \ : \ \Gamma \ni \tau \ \to \ \Gamma \vdash \tau \\ \mathbf{app} \ : \ \Lambda \to \ \Lambda & \mathbf{app} \ : \ \Gamma \vdash \sigma \Rightarrow \tau \ \to \ \Gamma \vdash \sigma \ \to \ \Gamma \vdash \tau \\ \mathbf{lam} \ : \ \Lambda \to \ \Lambda & \mathbf{lam} \ : \ \sigma : \ \Gamma \vdash \tau \ \to \ \Gamma \vdash \sigma \Rightarrow \tau \\ \mathbf{data} \ \mathsf{Ty} \ : \ \mathsf{Set} \ \mathbf{where} & \mathbf{data} \ \_ \vdash_{-} : \ : \ \mathsf{List} \ \mathsf{Ty} \ \to \ \Lambda \ \to \ \mathsf{Ty} \ \to \ \mathsf{Set} \ \mathbf{where} \\ \mathbf{base} \ : \ \mathsf{Ty} & \mathbf{var} \ : \ (i : \ \Gamma \ni \tau) \ \to \ \Gamma \vdash \ \mathbf{var} \ (to \mathbb{N} \ i) : \tau \\ \mathbf{app} \ : \ \Gamma \vdash t : \sigma \Rightarrow \tau \ \to \ \Gamma \vdash u : \sigma \ \to \ \Gamma \vdash \ \mathbf{app} \ t \ u : \tau \\ \mathbf{lam} \ : \ \sigma : \Gamma \vdash t : \tau \ \to \ \Gamma \vdash \ \mathbf{lam} \ t : \sigma \Rightarrow \tau \\ \end{array}
```

(The list membership relation ' \ni ' will be defined in Section 6.3.) Write an ornament between the datatypes Λ and ' \vdash ' of untyped and intrinsically typed λ -terms, and we get the typing relation

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```
SC : DataD \rightarrow Set_{\omega}
                                                                                       record SC_P(D : PDataD) : Set_{\omega} where field
1079
                                                                                           {elevel} : Level
               SC D = \forall \{\ell s\} \rightarrow SC_P (D.applyL \ell s)
1080
                                                                                          EI : [D.Param]_T \rightarrow Set elevel
               SC_B : ConB \rightarrow Set
1081
                                                                                          pos: ListAll SC<sub>B</sub> (D.struct)
                SC_B = ListAll (\lambda \{ (inl \_) \rightarrow Bool \})
1082
                                                                                          coe : (ps : [D.Param]_T) \rightarrow SC_{Cs} (D.applyP ps) pos (El ps)
                                           ; (inr_) \rightarrow \top \})
1083
               \mathsf{SC}_\mathsf{Cs}: \{I: \mathsf{Set}\,\ell_I\} \to \mathsf{ConDs}\,I\,\mathit{cbs} \to \mathsf{ListAll}\,\mathsf{SC}_\mathsf{B}\,\mathit{cbs} \to \mathsf{Set}\,\ell \to \mathsf{Set}_\omega
1084
               SC<sub>Cs</sub> []
                                  X = T
1085
               SC_{Cs} (D :: Ds) (s :: ss) X = SC_{C} D s E \times SC_{Cs} Ds ss X
1086
               \mathsf{SC}_\mathsf{C} : \{I : \mathsf{Set}\,\ell_I\} \to \mathsf{ConD}\,I\,\mathit{cb} \to \mathsf{SC}_\mathsf{B}\,\mathit{cb} \to \mathsf{Set}\,\ell \to \mathsf{Set}_\mathsf{\omega}
1087
               SC_C(\iota i)_{-} X = T
1088
               SC_{C} (\sigma A D) (true :: s) X = ((\_, A) \equiv_{\Sigma[\ell : \mathsf{Level}]} \mathsf{Set}_{\ell} (\_, X)) \times ((a : A) \to \mathsf{SC}_{C} (D a) \mathsf{s} X)
               SC_C (\sigma A D) (false :: s) X =
                                                                                                                       (a:A) \rightarrow SC_C (D a) s X
               SC_C (\rho D E) (\_ :: s) X =
                                                                                                                                          SC_C E s X
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```

Fig. 11. The simple-container predicate on datatype descriptions

'_\-:_' and an isomorphism between $\Gamma \vdash \tau$ and $\Sigma[t : \Lambda] \Gamma \vdash t : \tau$ for free, allowing us to promote an untyped term t to an intrinsically typed one if a typing derivation for t can be supplied.

We have omitted the types of the generic programs related to algebraic ornamentation because they are somewhat verbose, making generic program invocation less cost-effective — for example, the generic programs proving the inverse properties need the connections for the original and the new datatypes, the fold used to compute the algebraic ornament, and the 'from' and 'to' functions. In general, we should seek to reduce the cost of invoking generic programs. We have tested a smaller-scale solution where generic programs use Agda's instance arguments [Devriese and Piessens 2011] to automatically look for the connections and other information they need, and the solution works — for example, to-fromV can be derived by supplying just the names Vec and List. However, instance searching currently brings serious performance overhead, and the solution still requires us to instantiate one definition at a time. Larger-scale solutions such as instantiating the definitions in a parametrised module all at once may be required in practice.

Finally, we should briefly mention how AlgD handles universe polymorphism. Given F: FoldP, the most important change from F. Desc to AlgD F is adding a suitably typed σ -field (for example, the field b in AlgList) in front of every ρ -field; this is mirrored in the computation of the struct field of AlgD F from that of F. Desc, primarily using the function

```
algConB : Level \rightarrow ConB \rightarrow ConB

algConB \ell [] = []

algConB \ell (inl \ell' :: cb) = inl \ell' :: algConB \ell cb

algConB \ell (inr rb :: cb) = inl (max-\ell rb \sqcup \ell) :: inr rb :: algConB \ell cb
```

(where max- ℓ rb is the maximum level in rb). Subsequently we need to prove level-ineq for AlgD F, which requires non-trivial reasoning and involves properties about algConB such as max- σ (algConB ℓ cb) \equiv max- π cb \sqcup max- σ cb \sqcup hasRec? ℓ cb. In this case the datatype computation is not difficult, but it still makes sense to reason that the computed levels will always pass the universe check, and Agda conveniently allows us to perform the reasoning internally.

6.3 Predicates on Simple Containers

There are not too many generic programs that work without assumptions on the datatypes they operate on; with dependent types, such assumptions can be formulated as predicates on datatype descriptions. As a simple example, below we characterise a datatype N as a 'simple container' type by marking some fields of its description as elements of some type X, and then derive predicates

All P and Any P on N lifted from a predicate P on X, stating that P holds for all or one of the elements in an inhabitant of N. For example, the ListAll datatype (Section 3) is an instance of All.

The definition of simple containers (in several layers) is shown in Figure 11. The top layer SC on DataD only quantifies over the level parameters, and the main definition is at the next layer SC_P on PDataD: First is the element type El, which can refer to the ordinary parameters. Then in pos we assign a Bool to every σ -field indicating whether it is an element or not. More precisely, the assignments are performed on the struct field of the description, and might not make sense since any σ -field could be marked with **true**, not just those of type El. However, the coe field of SC_P makes sure that the types of the fields marked with **true** are equal to El; subsequently, when a generic program sees such a field, it can use the equality to coerce the type of the field to El.

The All predicate is simpler since it is just the datatype created along with a promotion isomorphism (Section 6.2). For example, to derive ListAll, we mark the element field of List in an SC ListD structure, from which we can compute a more informative ListWP datatype that requires every element a to be supplemented with a proof of P a:

```
\begin{array}{lll} \mathbf{data} \ \mathsf{ListWP} \ \{A: \mathsf{Set} \ \ell\} \ (P: A \to \mathsf{Set} \ \ell') : \mathsf{Set} \ (\ell \sqcup \ell') \ \mathbf{where} \\ & [] & : & \mathsf{ListWP} \ P \\ & \langle \_, \_ \rangle ::\_ : (a: A) \to P \ a \to \mathsf{ListWP} \ P \to \mathsf{ListWP} \ P \end{array}
```

Then the ornament between ListWP and List gives rise to ListAll and an isomorphism between ListWP P and Σ (List A) (ListAll P), allowing us to convert between a list of pairs of an element and a proof (ListWP P) and a pair of a list of elements (List A) and a list of proofs (ListAll P).

The Any predicate is more interesting since its structure is rather different from that of the original datatype, although in the case of List, the Any structure happens to degenerate quite a bit:

```
data ListAny \{A: \operatorname{Set} \ell\} (P: A \to \operatorname{Set} \ell'): \operatorname{List} A \to \operatorname{Set} (\ell \sqcup \ell') where here : \forall \{a \ as\} \to P \ a \to \operatorname{ListAny} P \ (a :: as) there : \forall \{a \ as\} \to \operatorname{ListAny} P \ as \to \operatorname{ListAny} P \ (a :: as)
```

The list membership relation $xs \ni x$ used in Section 6.2, defined by ListAny ($\lambda y \to x \equiv y$) xs, is a special case. In general, a proof of Any P is a path pointing to an element satisfying P, and we can write a generic lookup function that follows a path to retrieve the element it points to (resembling Diehl and Sheard's [2016] construction) — for ListAny, this lookup function specialises to

```
lookupListAny : \{A: \text{Set } \ell\} \{P: A \to \text{Set } \ell'\} \{as: \text{List } A\} \to \text{ListAny } P \ as \to \Sigma \ A \ P lookupListAny (here p) = _ , p lookupListAny (there i) = lookupListAny i
```

A path can be regarded as an (enriched) natural number that instructs the lookup function to stop (here/zero) or go further (there/suc) — that is, there is an ornament between Any and \mathbb{N} , allowing us to derive a forgetful function to \mathbb{N} that computes the length of a path. Moreover, a path should specify which element it points to if stopping, or which sub-tree to go into if going further, so the numbers of here and there constructors are exactly the numbers of element positions and recursive fields respectively. For example, the Any predicate for the datatype of balanced 2-3 trees below (taken from McBride [2014]) would have three here constructors and five there constructors:

```
data B23T : Height \rightarrow Value \rightarrow Value \rightarrow Set where-- both Height and Value are \mathbb{N}node_0: \{l \leq r\}\}\rightarrow B23T zero lrnode_2: (x: Value) \rightarrow B23T h l x \rightarrow B23T h x r\rightarrow B23T (suc h) lrnode_3: (x y: Value) \rightarrow B23T h l x \rightarrow B23T h x y \rightarrow B23T h x y \rightarrow B23T (suc h) l r
```

It is nice not having to write B23TAny and lookupB23TAny by hand.

7 DISCUSSION

 Interoperability between generic and native entities. Our framework makes generic constructions parametrised by our version of datatype descriptions interoperable with native datatypes and functions. Generic libraries using their own descriptions could achieve the same by duplicating our framework, but there are alternatives: In Haskell, Magalhães and Löh [2014] provide automatic conversions between the datatype representations of several libraries —targeting different families of datatypes— and a representative representation, through which native datatypes are connected to all the libraries at once; our descriptions (possibly with extensions to support mutual recursion, induction-recursion, etc) can serve as a representative representation. Moreover, in a dependently typed setting it is possible to attain better uniformity and reusability by defining families of datatypes as predicates on a single representation like in Section 6.3, in contrast to a simply typed setting where different families of datatypes need to be modelled by separate representations.

Optimisation of datatype-generic programs. A traditional way to instantiate a generic program is to compose the program with conversions between a native datatype and its generic description. If the generic program was defined on datatypes decoded by the μ operator and then instantiated by the recursive conversion between the native and μ -decoded datatypes, the conversion overhead would be roughly the same as that of unoptimised Haskell generic programs, and had to be eliminated. Rather than optimising the composition of several recursive functions, the Haskell community has employed a shallow encoding and studied relevant optimisation/specialisation (see, for example, de Vries and Löh [2014, Section 5.1]); this encoding is the basis of our design. Below we discuss previous attempts at optimising instantiated programs using the shallow encoding.

Recent work using staging [Yallop 2017; Pickering et al. 2020] eliminates performance overheads by generating native function definitions that are almost identical to hand-written ones. There is, however, no implementation of staging in existing dependently typed languages, so we cannot compare them properly on the same ground. But it shares a similar purpose with our framework of generating function definitions containing neither generic representations nor conversions, making them comparable regardless of the specific languages they work in.

We compare staging with our framework from the view of partial evaluation [Jones et al. 1993]. A partial evaluator takes a general program and known parts of its input, and generates a program that takes the remaining input; the resulting program is extensionally equal to —and usually more optimised than— the general program partially applied to the known input. Our metaprogram define(P)Fold (Section 4.3) is a partial evaluator that specialises a generic program (general program) to a given description (known input). Indeed, it has been observed that we can perform partial evaluation in functional languages by normalisation [Filinski 1999], which define(P)Fold does.

Similar to a partial evaluator, a staged generic program is a more specialised program-generator — the generic/general program to be specialised has been fixed. However, staging requires manually inserting staging annotations; this puts burdens on the programmer and mixes a part of the instantiation process with generic definitions. Our approach separates how we instantiate generic programs (by metaprograms) from how we define them (as algebras). As a result, our generic programs are annotation-free, making them easier to write and read. On the other hand, staging has its own benefit: Pickering et al. [2020, Section 4.1] provide principles followed by the programmer to avoid generic representations from appearing in specialised programs; without staging annotations, it seems difficult to formulate similar principles in our setting. But we can explore alternative approaches — for example, Alimarine and Smetsers [2004] give theorems guaranteeing that generic representations can be removed from instantiated programs with suitable types by normalisation.

There have also been attempts using compiler optimisation [de Vries 2004; Magalhães 2013], which are less relevant to our work as explained in Section 1. However, as generic programs become

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more complex, we may need more advanced code generation techniques that can be borrowed from compiler optimisation, possibly through extensions to elaborator reflection.

Performance overhead of the framework. While programs instantiated using our framework are optimised, there is still the overhead of type-checking the additional definitions required by our framework and running the metaprograms to generate code. Minimising this overhead is particularly important for an interactive development environment (which Agda is famous for). We performed a quick experiment comparing two versions of all our examples: the first version used our metaprograms to instantiate the generic programs, whereas the second version contained printed (hand-written) definitions only, without anything from our framework. The result was disappointing: on a typical laptop, it took about 66 seconds to check the first version, and about 5 seconds to check the second version. We think that the experiment is more an indication of how much the current implementation of Agda's type-checker and elaborator reflection (and, to a lesser extent, our metaprograms) can still be optimised, since the overhead inherently associated with our core idea --instantiating generic programs with normalisation and unquotation-- should not be so high (and our metaprograms are just straightforward implementations of the idea). But however high the overhead is, it can always be alleviated by separate compilation—collecting generic entities in separate files, which are checked only once and do not interfere with subsequent interactive development sessions— or copying and pasting printed definitions into the programmer's files. These solutions require the programmer's effort, but as generic libraries become more powerful and the user interface gets improved, the cost will be easily outweighed by the benefit of not having to write the definitions that can generated.

Analysis of higher-order encodings. The local variable creation technique has been used extensively to analyse higher-order encodings, but it has to be used with caution. Agda's elaborator checks if a local variable escapes its context in the result of extendContext during the runtime of elaboration.⁸ This is similar to the nu constructor of the typed tactic language Mtac [Ziliani et al. 2015], which checks the escape of local variables during tactic execution. Indeed, Chen [2019] implemented Mtac's nu constructor by elaborator reflection in the same way. By contrast, the v-operator for local name creation in a modal calculus [Nanevski and Pfenning 2005] and a two-level functional language [Schürmann et al. 2005] statically ensures that local names cannot escape their scope.

Our higher-order encodings are not actually higher-order abstract syntax [Harper et al. 1993], since exotic terms such as Bool :: λ {false \rightarrow [A : Set] []; true \rightarrow [P : (\mathbb{N} \rightarrow Set)] []} are not excluded by the typing discipline. As argued by McBride [2011], exotic terms (which we do not use) could be exploited for richer expressiveness, but then it is harder to analyse and deviates too much from Agda's current datatype declarations, which our framework targets.

Dependently typed datatype-generic libraries. Our framework makes it possible and worthwhile to develop datatype-generic libraries for wider and practical use in Agda. The change to traditional Agda generic programs required by our framework is a mild generalisation from the operators μ and fold to datatype and fold connections capturing the behaviour of the operators, so it is easy to adapt existing libraries to our framework as well as develop new ones. There are still many opportunities to explore for dependently typed datatype-generic libraries: Even for recursion schemes [Yang and Wu 2022], a standard datatype-generic example, we can start supplying theorems about them like in Section 6.1. Datatypes derived from others, such as the lifted predicates All and Any in Section 6.3, are also common and should be treated generically (as opposed to manually duplicating an instance for each datatype as in the standard library [Agda Team 2022b]). Domain-specific organisation of

⁸In previous versions of Agda, the existing de Bruijn indices in an extended context were not weakened [Agda Issue 2019] and the check was not implemented in full until the recent version 2.6.2 of Agda [Agda Issue 2020].

 datatypes with intrinsic invariants is another important goal, for which ornaments [McBride 2011] still have much potential (although the community has focussed mostly on lifting ornaments to programs and proofs [Dagand and McBride 2014; McDonell et al. 2016; Williams and Rémy 2018; Ringer et al. 2019]). For example, Section 6.2 mentions that the relationship between intrinsically and extrinsically typed λ -terms can be captured as an ornament, whose properties and derived constructions should be formulated generically for reuse in developments of typed embedded languages — a direction already proved fruitful by Allais et al. [2021].

Code generation versus first-class datatypes. Similar to staged approaches [Yallop 2017; Pickering et al. 2020], our framework instantiates generic programs by generating code separately for each native instance. A potential problem is code duplication, on which we take a conservative position: while we cannot solve the problem, which is inherent in languages with datatype declarations, we do alleviate it a little by removing the overhead of manual instantiation and maintaining explicit connections between generic and instantiated entities, which generic libraries can exploit. A possible solution to the problem was proposed by Chapman et al. [2010] in the form of a more radically redesigned type theory where datatype declarations are replaced with first-class descriptions, and the μ operator becomes the exclusive built-in mechanism for manufacturing datatypes. Generic programs in this theory are directly computable and do not require instantiation. However, there have been no subsequent developments of the theory, in particular a practical implementation. While waiting for better languages to emerge, it is also important to enable the development and practical use of datatype-generic libraries as soon as possible, which our framework does.

Foundations. While the use of universe polymorphic-datatypes is convenient in practice, its theory is, surprisingly, an unexplored territory, and the notion of universe polymorphism is less studied than it should be. To the best of our efforts, we did not find any type theory backing Agda's current design of first-class universe levels and universe polymorphism. Even worse, without universe cumulativity, subject reduction does not hold for expressions in Set $_{\omega}$ [Agda Issue 2022]. Kovács [2022] initiated a theoretical framework for first-class universe levels which is able to model features like bounded universe polymorphism, but its metatheory is bare-bones and lacks, for example, an elaboration algorithm needed for implementation. A type theory of first-class datatypes [Chapman et al. 2010] requires internal universe-polymorphic encodings, but most of internal encodings [Dybjer and Setzer 2006; Nordvall Forsberg 2013; Kaposi and Kovács 2020] do not encode universe polymorphism.

Due to the nature of uni-typed encodings, our experience with Agda's elaborator reflection was painful, especially in contrast to datatype-generic programming. Elaborator reflection is a useful paradigm but it has not been specified in theory, and to use it the understanding of the inner workings of elaborator is much needed. More importantly, the correctness of a macro can only be verified externally at best, but it is internally guaranteed by typed metaprogramming like staging. Christiansen and Brady [2016] argued that programs in the reflected elaborator are shorter and simpler than typed metaprograms, but they also admitted the additional cost is for the mandatory correctness. Hopefully, the best of two worlds could be combined. Say, suppose that we have a type TTerm A of A-typed reflected expressions. Normalisation (by evaluation) always works on well-typed expressions, so the type of normalise could be TTerm $A \to TTerm A$; type checking transforms a possibly ill-formed expression to a typed expression if successful, so the type of checkType could be Term $A \to TTerm A$. Typed reflected expressions also benefit efficiency, since they need not be elaborated again.

We hope that our framework can serve as inspiration and a call for a foundation for universes and metaprogramming not only for theoretical interests but also for practical needs.

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