Ornaments and Proof Transport applied to Numerical Representations

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Abstract

Todo list

	revisit this when further
	revisit this when further
	Indexed what, example fin
	Why circles are points with K. Why circles are not points with
	univalence
	Merge
	Express what this accomplishes, and why this is impressive com-
	pared to without univalence
	Generalizing the observation that lists look like unary naturals and
	Braun trees look like binary naturals
	And more
	Formalizing the "looks like relation"
	Adapt and split into background and actual work 16
	Is there a simple twist or other interesting example that we can run
	through instead, or would anything else be too abrupt without
	starting from this simple case?
	Put some minimal definitions here
	Sigma plus/minus
	Currently, without proof
	Adapt this to the non-proposal form
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1 Introduction

Program verification is an indispensible aspect of programming, whether you're coding up your own Asteroids or you're developing a linear algebra library, it would be a waste of time to hunt for bugs which could have been uncovered by random testing. When testing does not offer enough certainty or cannot handle the complexity of the input, we can instead use formal program verification: it would be embarassing if someone else suffers the consequences of a bug in your library, so you might prove your library or parts of it correct in a proof assistant like Coq or Agda. In a more extreme example, you might code directly into a proof assistant, specifying the behaviour of your program beforehand, having it checked while you're implementing it.

Yet, program verification, especially of the last kind, is a double-edged sword: while it becomes easier to write code without bugs, it becomes harder to write code in the first place. A proof assistant has to enforce total and terminating programs (at least by default), as incomplete or circular steps would undermine the correctness of a proof. Non-total operations are abundant in most languages, like getting the first element of a list; such operations would require the programmer to provide evidence that the operation can not fail at each usage. In this example the evidence can be encoded by modifying a list to remember its length, and generally we can create variations on datastructures for use in correct-by-construction programs.

This might prompt defining variations for each use case, and duplicating all operations on them, making little or no use of the fact that types like lists and vectors are strongly related. But this can be avoided, since a broad class of relations has been tamed by ornaments [McB14; KG16]. Informally, an ornament describes the pieces of information necessary to construct a new type from an existing type.

However, we do not have to stop at relating lists and vectors. Just like vectors can be described as lists with more information, lists can be described

as natural numbers with more information [McB14]. This can be generalized to other datastructures, such as binary numbers and trees. The idea of instead constructing datastructures from number systems has been studied as numerical representations [Oka98; HS22]. This provides a way to talk about datastructures using their underlying numbers, and allows one to mechanically calculate datastructures and some of their properties from these numbers, albeit manually.

By calculating a datastructure, one hopes to gain an isomorphism between the datatype represented as a lookup function, and the concrete version of the datatype. As the representation and the concrete type are equivalent, one can reason about properties of the concrete side by looking at the representation, which is often simpler. In the usual context, one would still have to manually convert proofs back and forth. More conveniently, we would like to apply representation independence; similarly to how equality of indiscernables ensures that exchanging equal terms cannot change the behaviour of a program, the same should hold for isomorphic types. While such results usually only exists in the meta-theory, or can only be applied on concrete types by manually weaving conversions through proofs, structured equivalences [Ang+20] can internalize this, at the cost of using Cubical Agda.

1.1 The Problem

The main question of this project is: can we describe finger trees [HP06] in the frameworks of numerical representations and ornamentation [KG16], simplifying the verification of their properties as flexible two-sided arrays? This question generates a number of interesting subproblems, such as that the number system corresponding to finger trees has many representations for the same number, which we expect to describe using quotients [VMA19] and reason about using representation independence [Ang+20].

revisit this when further

1.2 Contributions

In this paper, we

revisit this when fur-

- x adapt ornaments to nested types
- x allow ornaments to refer to sub-ornaments
- x define a small universe of typical number systems
- x give a generic derivation of numerical representations as ornaments from these number systems
- instantiate a Structure Identity Principle for these representations.

We follow this up by enumerating these, and more structures. We

x define hierarchies to enumerate terms by levels

- x track the cardinalities of each level
- include parametrized datatypes into this setup
- modify this to include nested types
- ? adapt this approach to index-first datatypes
- iterate the accessible indices per level

Along the way, we also

- x characterize identities of W-types
- x express heterogeneous variants of datastructures as ornaments.

2 Background

2.1 Agda

We formalize our work in Agda [Tea23], a functional programming language with dependent types. Using dependent types we can use Agda as a proof assistant, allowing us to state and prove theorems about our datastructures and programs. These proofs can then be run as algorithms, or in some cases be extracted to a Haskell program¹.

Syntactically Agda is remeniscent of Haskell. One difference is that Agda allows most characters and words in identifiers with only a small set of exceptions. For example, we can write

```
- \bigcirc - \bigcirc - : Bool \rightarrow A \rightarrow A \rightarrow A false \bigcirc t \bigcirc e = e true \bigcirc t \bigcirc e = t
```

Another is that datatypes are always either given as generalized algebraic datatypes (GADTs) or record types.

The type system of Agda is an extension of (intensional) Martin-Löf type theory (MLTT), a constructive type theory in which we can interpret intuitionistic logic: the Curry-Howard isomorphism states that certain formulas correspond to certain types, and proofs of a formula correspond to terms of the corresponding type. The atomic formula true can be represented as the empty record

```
record τ : Type where constructor tt
```

so that tt proves τ . False can be represented by a datatype with no constructors record ι : Type where

since there is (hopefully) no way to make get a term of 1 without inconsistent assumptions. The logical implication $A \implies B$ corresponds to the type of functions $A \to B$: a proof of A can be converted to a proof of B. Using implication, we can define the negation $\neg A$ of a formula A as the type $A \to \bot$. Disjunction (logical or) is described by a sum type A + B:

 $^{^1{\}rm Or}$ JavaScript, if you want.

```
data _+_ A B : Type where
inl : A → A + B
inr : B → A + B
```

if we have either A or B, we can prove A+B. Conjunction (logical and) is given as a product type:

```
record _x_ A B : Type where
  constructor _,_
  field
    fst : A
    snd : B
```

we need both A and B to prove $A \times B$. Using the correspondence, we reason in propositional logic by writing functional programs. As an example, consider the proof of the tautology

```
\rightarrow-x-undistr : ((A \rightarrow C) \times (B \rightarrow C)) \rightarrow (A + B) \rightarrow C
\rightarrow-x-undistr (a\rightarrow c , b\rightarrow c) (inl a) = a\rightarrowc a
\rightarrow-x-undistr (a\rightarrow c , b\rightarrow c) (inr b) = b\rightarrowc b
```

Compared to Haskell, Agda allows the type of a codomain of a function to vary with the applied value: given a function P from A into Type, a type family over A, we can form the dependent function type $(a:A) \to P$ a. Applying a function $f:(a:A) \to P$ a to a value a:A then will have type f a:P a. Similarly, the type of a field in a record type can depend on values of earlier fields, e.g.,

```
record ∃ A (P : A → Type) : Type where
  constructor _,_
  field
   fst : A
  snd : P fst
```

The presence of these dependent types enriches the interpretation of logic into programs. To interpret first-order logic we need to describe formulas containing variables, which are called predicates. Predicates correspond to functions into Type

```
P : A → Type
```

Using predicates, we can interpret quantifiers as the dependent types above. Universal quantification (for all) is a dependent function type

```
(a : A) \rightarrow P a
```

since for each a:A, we have a proof of P a. Likewise, existential quantification (exists) is the dependent pair type \exists , since this gives an a:A and a proof P a.

Predicates can also be expressed using indexed data types, in which the choice of constructor can influence the index. Equality of elements of a type A can then be interpreted as the type Indexed what, example fin

```
data Eq (a : A) : A → Type where
  refl : Eq a a
```

Closed terms of this type can only be constructed for definitionally equal elements, but crucially, variables of this type can contain equalities between different elements. As the second argument is an index, pattern matching on refl unifies the elements, such that properties like substitution follow

```
subst : Eq a b \rightarrow P a \rightarrow P b subst refl x = x
```

Unlike most languages, Agda rules out non-terminating functions by restricting their definitions to structural recursion. The termination checker (together with other restrictions which we will encounter in due time) prevents trivial proofs which would be tolerated in Haskell, like

```
undefined : ∀ {A : Type} → A undefined = undefined
```

This ensures that all our interpretations mentioned above remain consistent.

2.2 Cubical Agda

Intuitively, one expects that like how isomorphic groups share the same group-theoretical properties, isomorphic types also share the same type-theoretical properties. Meta-theoretically, this is known as representation independence, and is evident. Inside (ordinary) Agda this is not so practical, as this independence only holds when applied to concrete types, and is then only realized by manually substituting along the isomorphism. On the other hand, in Cubical Agda, the Structure Identity Principle internalizes a kind of representation independence [Ang+20].

Cubical Agda modifies the type theory of Agda to a kind of homotopy type theory, looking at equalities as paths between terms rather than the equivalence relation generated by reflexivity. In cubical type theories, the role played by pattern matching on refl or by axiom J, in MLTT and "Book HoTT" respectively, is instead acted out by directly manipulating cubes². In Cubical Agda, univalence

..

is not an axiom but a theorem.

2.3 The Structure Identity Principle

To give an understanding of the basics of Cubical Agda [VMA19] and the Structure Identity Principle (SIP), we walk through the steps to transport proofs about addition on Peano naturals to Leibniz naturals. We give an overview of some features of Cubical Agda, such as that paths give the primitive notion of equality, until the simplified statement of univalence. We do note that Cubical Agda has two downsides relating to termination checking and universe levels, which we encounter in later sections.

Starting by defining the unary Peano naturals and the binary Leibniz naturals, we prove that they are isomorphic by interpreting them into eachother. We explain that these interpretations are easily seen to be mutual inverses by proving lemmas stating that both interpretations "respect the constructors" of the types. Next, we demonstrate how this isomorphism can be promoted into

Why circles are points with K. Why circles are not points with univalence

 $^{^{2}}$ Under the analogy where a term is a point, an equality between points is a line, a line between lines is a square.

an equivalence or an equality, and remark that this is sufficient to transport intrinsic properties, such as having decidable equality, from one natural to the other.

Noting that transporting unary addition to binary addition is possible but not efficient, we define binary addition while ensuring that it corresponds to unary addition. We present a variant on refinement types as a syntax to recover definition from chains of equality reasoning, allowing one to rewrite definitions while preserving equalities.

We clarify that to transport proofs referring to addition from unary to binary naturals, we indeed require that these are meaningfully related. Then, we observe that in this instance, the pairs of "type and operation" are actually equated as magmas, and explain that this is an instance of the SIP.

Finally, we describe the use case of the SIP, how it generalizes our observation about magmas, and how it can calculate the minimal requirements to equate to implementations of an interface. This is demonstrated by transporting associativity from unary addition to binary addition, noting that this would save many lines of code provided there is much to be transported.

Let us quickly review some features of Cubical Agda [VMA19] that we will use in this section.

In Cubical Agda, the primitive notion of equality arises not (directly) from the indexed inductive definition we are used to, but rather from the presence of the interval type I. This type represents a set of two points i0 and i1, which are considered "identified" in the sense that they are connected by a path. To define a function out of this type, we also have to define the function on all the intermediate points, which is why we call such a function a "path". Terms of other types are then considered identified when there is a path between them.

Paths between types are incredibly useful, as they effectively let us directly transport properties between isomorphic structures. However, they do not come without downsides, such as that the negation of axiom K complicates both some termination checking and some universe levels.³

We will discuss how to deal with these issues in later sections, so let us not be distracted from what we *can* do with paths. For example, the different perspective gives intuitive interpretations to some proofs of equality, like

```
sym : x \equiv y \rightarrow y \equiv x
sym p i = p (\sim i)
```

where \sim is the interval reversal, swapping i0 and i1, so that sym simply reverses the given path.

Also, because we can now interpret paths in record and function types in a new way, we get a host of "extensionality" for free. For example, a path in $A \to B$ is indeed a function which takes each i in I to a function $A \to B$. Using this, function extensionality becomes tautological

```
funExt: (\forall x \rightarrow f x \equiv g x) \rightarrow f \equiv g
funExt p i x = p x i
```

Merge

³In particular, this prompts rather far-reaching (but not fundamental) changes to the code of previous work, such as to the machinery of ornaments [KG16] in Appendix A.

Finally, equivalences, the HoTT-compatible variant of bijections, have the univalence theorem

```
ua : \forall \{A B : Type \ell\} \rightarrow A \simeq B \rightarrow A \equiv B
```

stating that "equivalent types are identified", such that equivalences like $1 \to A \simeq A$ become paths $1 \to A \equiv A$, making it so that we can transport proofs along them. We will demonstrate this by a more practical example in the next section.

2.3.1 Unary numbers are binary numbers

Let us demonstrate an application of univalence by exploiting the equivalence of the "Peano" naturals and the "Leibniz" naturals. Recall that the Peano naturals are defined as

```
data N : Type where
  zero : N
  suc : N → N
```

This definition enjoys a simple induction principle and is well-covered in most libraries. However, the definition is also impractically slow, since most arithmetic operations defined on \mathbb{N} have time complexity in the order of the value of the result.

As an alternative we can use binary numbers, for which for example addition has logarithmic time complexity. Standard libraries tend to contain few proofs about binary number properties, but this does not have to be a problem: the N naturals and the binary numbers should be equivalent after all!

Let us make this formal. We define the Leibniz naturals as follows:

```
data Leibniz : Set where
  Ob : Leibniz
  _1b : Leibniz → Leibniz
  _2b : Leibniz → Leibniz
```

Here, the 0b constructor encodes 0, while the _1b and _2b constructors respectively add a 1 and a 2 bit, under the usual interpretation of binary numbers:

```
toN: Leibniz \rightarrow N

toN 0b = 0

toN (n 1b) = 1 N.+ 2 N. · toN n

toN (n 2b) = 2 N.+ 2 N. · toN n

[-] = toN
```

This defines one direction of the equivalence from \mathbb{N} to Leibniz, for the other direction, we can interpret a number in \mathbb{N} as a binary number by repeating the successor operation on binary numbers:

```
bsuc : Leibniz → Leibniz
bsuc 0b = 0b 1b
bsuc (n 1b) = n 2b
bsuc (n 2b) = (bsuc n) 1b
fromN : N → Leibniz
```

```
fromN 0 = 0b
fromN (suc n) = bsuc (fromN n)
```

To show that toN is an isomorphism, we have to show that it is the inverse of fromN. By induction on Leibniz and basic arithmetic on N we see that

```
toN-suc : \forall \ x \to [\![\!]\!] bsuc x \ [\![\!]\!] \equiv suc [\![\!]\!] x \to x \to x so toN respects successors. Similarly, by induction on x \to x we get fromN-1+2x \to x \to x (fromN x) 1b and
```

fromN-2+2·: $\forall x \rightarrow$ fromN (2 + double x) \equiv (fromN x) 2b so that fromN respects even and odd numbers. We can then prove that applying toN and fromN after each other is the identity by repeating these lemmas

```
N↔L : Iso N Leibniz
N↔L = iso fromN toN sec ret
where
sec : section fromN toN
ret : retract fromN toN
```

This isomorphism can be promoted to an equivalence

```
N≃L : N ≃ Leibniz
N≃L = isoToEquiv N↔L
```

which, finally, lets us identify N and Leibniz by univalence

```
N≡L : N ≡ Leibniz
N≡L = ua N≃L
```

The path $\mathbb{N}\equiv \mathbb{L}$ then allows us to transport properties from \mathbb{N} directly to $\mathsf{Leibniz}$; as an example, we have not yet shown that $\mathsf{Leibniz}$ is discrete, i.e., has decidable equality. Using substitution, we can quickly derive this⁴

```
discreteL : Discrete Leibniz
discreteL = subst Discrete N≡L discreteN
```

This can be generalized even further to transport proofs about operations from ${\tt N}$ to Leibniz.

2.3.2 Functions from specifications

As an example, we will define addition of binary numbers. We could transport _+_ as a binary operation

```
BinOp: Type → Type
BinOp A = A → A → A
from Nto Leibnizto get
_+'_: BinOp Leibniz
_+'_= subst BinOp N≡L N._+_
```

But this is inefficient, incurring an O(n+m) overhead when adding n and m. It is more efficient to define addition on Leibniz directly, making use of the binary nature of Leibniz, while agreeing with the addition on N. Such a definition can be derived from the specification "agrees with $_{-+-}$ ", so we implement a syntax

⁴Of course, this gives a rather inefficient equality test, but for the homotopical consequences this is not a problem.

for giving definitions by equational reasoning, inspired by the "use-as-definition" notation used by Hinze and Swierstra [HS22]: Using an implicit pair type

With this we can define addition on $\tt Leibniz$ and show it agrees with addition on $\tt N$ in one motion

Now we can easily extract the definition of ${\tt plus}$ and its correctness with respect to ${\tt _+_}$

```
plus : \forall x \ y \rightarrow \text{Leibniz}
plus x y = defined-by (plus-def x y)

plus-coherent : \forall x \ y \rightarrow \text{fromN} \ (x + y) \equiv \text{plus} \ (\text{fromN} \ x) \ (\text{fromN} \ y)
plus-coherent x y = cong fromN
   (cong<sub>2</sub> _+_ (sym (N\leftrightarrowL .leftInv x)) (sym (N\leftrightarrowL .leftInv _))) •
   by-definition (plus-def (fromN x) (fromN y))
```

We remark that Def is close in concept to refinement types⁵, but extracts the value from the proof, rather than requiring it before. ⁶

 $^{^5\}grave{\rm A}$ la Data. Refinement.

 $^{^6\}mathrm{Unfortunately}$, normalizing an application of a defined-by function also causes a lot of unnecessary wrapping and unwrapping, so Def is mostly only useful for presentation.

2.3.3 The Structure Identity Principle

We point out that N with N.+ and Leibniz with plus form magmas, that is, inhabitants of

```
Magma': Type<sub>1</sub>
Magma' = \Sigma[ X \in Type ] BinOp X
```

Using that a path in a dependent pair corresponds to a dependent pair of paths, we get a path from (N, N.+) to (Leibniz, plus). This observation is further generalized by the Structure Identity Principle (SIP) as a form of representation independence [Ang+20]. Given a structure, which in our case is just a binary operation

```
MagmaStr : Type → Type
MagmaStr = BinOp
```

this principle produces an appropriate definition "structured equivalence" ι . The ι is such that if structures X,Y are ι -equivalent, then they are identified. In the case of MagmaStr, the ι asks us to provide something with the same type as plus-coherent, so we have just shown that the plus magma on Leibniz

```
MagmaL : Magma
    fst MagmaL = Leibniz
    snd MagmaL = plus
and the _+_ magma on N and are identical
    MagmaN≃MagmaL : MagmaN ≡ MagmaL
    MagmaN≃MagmaL = equivFun (MagmaΣPath _ _) proof
    where
        proof : MagmaN ≃[ MagmaEquivStr ] MagmaL
        fst proof = N≃L
        snd proof = plus-coherent
As a consequence, properties of _+_ directly yield corresponding properties of plus. For example,
        plus-assoc : Associative _≡_ plus
```

Express what this accomplishes, and why this is impressive compared to without univalence

2.4 Numerical representations

 $(\lambda A \rightarrow Associative _=_ (snd A))$

plus-assoc = subst

MagmaN≃MagmaL N-assoc

2.5 Generic programming and ornaments

The deriving-mechanism in Haskell can take writing functions which consist primarily of boilerplate out of the hands of the programmer by deriving default implementations. Using reflection we can write similar macros and generic pro-

Generalizing the observation that lists look like unary naturals and Braun trees look like binary naturals.

grams inside the type-checking monad; with it one can quote types or values, inspect their definitions, and unquote terms them to inject them into the code as if they were written manually.

However, programming in this monad is generally not pleasant, as terms enjoy none of the safety we are used to from Agda, and type errors are only detected when applying macros as opposed to when writing them. That is not to say that effective generic programming is impossible in Agda, and quite the opposite is true [EC22]. We will take a closer look at constructions which we can use for datatype generic programming.

And more

To inspect a datatype and manipulate its values safely, we have to represent the datatype internally. This can be done by defining another datatype encoding how datatypes can be formed, henceforth description, together with a function which assigns meanings to this encoding, henceforth interpretation. We will start from an encoding which captures only a small set of types, and work towards an encoding of parametrized indexed types.

Each of the constructors of this description represents a type-former for the described universe. In this case, the universe only contains the finite types; the meaning of the type formers should be evident from the interpretation:

```
\mu: Desc \rightarrow Type

\mu 0 = I

\mu 1 = T

\mu (D \oplus E) = \mu D \oplus \mu E

\mu (D \otimes E) = \mu D \times \mu E
```

Booleans live in this universe as

```
BoolD : Desc
BoolD = 1 @ 1
```

but to encode a type like Nwe need a different setup. Consider the definition

```
data N : Type where
  zero : N
  suc : N → N
```

we can interpret this as the declaration $\mathbb{N} \simeq \mathsf{T} \uplus \mathbb{N}$, and formally, \mathbb{N} is indeed the least fixpoint of this equation. Intuitively, this tells us that \mathbb{N} can be formed by applying $\mathsf{T} \uplus$ a countable number of times to \bot . More generally, we see that recursive types are the fixpoints of polynomial functors

```
data Desc : Type where

1 ρ : Desc

_⊕_ _⊗_ : Desc → Desc → Desc
```

Now we will have to split the interpretation and the fixpoint, where the interpretation now translates a description to a polynomial functor

```
[D \otimes E] X = ([D] X) \times ([E] X)
```

mapping 1 to the constant 1 polynomial, ρ to the variable x polynomial, and θ and ∞ to the ordinary polynomial sum and product. (Conspicuously, 0 is missing, but you can see that 1 is not. After all, μ ρ has no values). Taking the fixpoint gives the actual type

```
data \mu (D : Desc) : Type where con : [D] (\muD) \rightarrow \muD so that we can describe N as ND : Desc ND = 1 \oplus \rho
```

To make describing complex types more practical we can merge ρ and \otimes , and add a variant σ of \otimes allowing arbitrary types in descriptions

```
data Desc : Type₁ where
1 : Desc
ρ : Desc → Desc
σ : (S : Type) → (S → Desc) → Desc
_⊕_ : Desc → Desc → Desc
```

In σ , we ask for a function $S \to Desc$ rather than just a Desc, modelling a Desc with a bound variable of S. We interpret $\sigma S D$ as $\Sigma[s \in S] \ [D S] X$.

In this universe we can describe types in which the fields be either X, the type itself, or another type S. For example, we can describe List using an external parameter

```
ListD: Type \rightarrow Desc
ListD A = 1 \oplus (\sigma A \lambda \rightarrow \rho 1)
```

We will soon see how we can internalize parameters, but since internalizing indices is easier, we will tackle indices first.

We should note that there are two strategies we can use to describe an indexed type. First, we can define a description of a type indexed by I to simply be a function $I \to Desc$, yielding a universe of index-first types. Second, we can pull the index completely into Desc, and let 1 declare the index at the leaf of a constructor, more closely resembling Agda's datatypes. Both have their advantages and disadvantages, mainly, index-first datatypes are more space efficient. We opt however for the second option, because as we will see later, this allows us to keep descriptions "relatively small" (i.e., something like foldable) and more flexible in their levels.

```
data Desc (I : Type) : Type₁ where

1 : I → Desc I

ρ : I → Desc I → Desc I

σ : (S : Type) → (S → Desc I) → Desc I

_⊕_ : Desc I → Desc I → Desc I
```

Now 1 i says that this branch constructs a term of X i, while ρ i asks for a recursive field X i. As Desc I describes a type indexed by I, which is a function I \rightarrow Type, we also have to interpret Desc I as an indexed functor

```
 [\![ \_ ]\!] : \mathsf{Desc} \ I \to (I \to \mathsf{Type}) \to (I \to \mathsf{Type}) \\ [\![ \ 1 \ j \ ]\!] \ \mathsf{X} \ i = i \equiv j \\ [\![ \ \rho \ j \ D \ ]\!] \ \mathsf{X} \ i = \mathsf{X} \ j \times [\![ \ D \ ]\!] \ \mathsf{X} \ i
```

```
[ \sigma S D ] X i = \Sigma [ s \in S ] [ D s ] X i
[ D \oplus E ] X i = [ D ] X i \uplus [ E ] X i
```

Applying an interpretation to an index i asks for the constructors at i. We see that by interpreting 1 j as an equality, we ensure that if we ask for i, then we also must get something at index i. In this universe we can also describe vectors

```
VecD : Type \rightarrow Desc N
VecD A = (1 zero) \oplus (\sigma N \lambda n \rightarrow \sigma A \lambda _ \rightarrow \rho n (1 (suc n)))
```

making use of the variable binding in σ to state that if we get a vector of length n, then we can construct a vector of length suc n.

The observant reader might have noticed that we claim $I\to Desc$ does not give small descriptions, but still allow for $S\to Desc.$ We can fix this issue at the same time we implement parameters, keeping a form of variable binding. We could implement types with a single parameter by interpreting to "endofunctors" Type $\to I\to T$ ype, adding another type-former accessing the parameter. This is however a bit less flexible than we want, does not distinguish between types with actually no parameters and a type applied to 1, nor allows us to refer to fields in the parameters.

We will first need some structure expressing the kinds of parameters that we can have. We could try using List Type, but this rules out types like Σ (A: Type) (B: A \rightarrow Type). Instead, we use a telescope, a list of types which does capture the dependency. To define a type depending on the preceding telescope, we define telescopes and their meaning by induction-recursion

so a telescope can either be empty, or be formed from a telescope and a type in the context of that telescope, where the context is defined as However, to deal with variables, we will also need to be able to describe variable telescopes. This means that while the parameter telescope in a description stays constant, the variable telescope grows independently when we add more σ 's. We can represent this by parametrizing telescopes over a type

```
data Tel (P: Type): Type₁
[_]tel: Tel P → P → Type
_⊢_: Tel P → Type a → Type a
Γ ⊢ A = Σ _ [ Γ ] tel → A

data Tel P where
∅ : Tel P
_⊳_: (Γ: Tel P) (S: Γ ⊢ Type) → Tel P
```

We define a shorthand $\Gamma \vdash A$ for type of S, representing a value of A in context Γ . By changing $\llbracket _ \rrbracket$ tel to depend on a value of P as

```
[ \lozenge ] \text{tel } p = T
[ \Gamma \triangleright S ] \text{tel } p = \Sigma [ x \in [ \Gamma ] \text{tel } p ] S (p , x)
```

```
a telescope of the form
        ExTel: Tel T → Type<sub>1</sub>
        ExTel Γ = Tel ([Γ]tel tt)
can access all values of \Gamma, and can be treated as an extension of \Gamma. To interpret
them, we define
        [_&_]tel : (\Gamma : Tel \tau) (V : ExTel \Gamma) \rightarrow Type
        [\Gamma \& V]tel = \Sigma ([\Gamma]tel tt) [V]tel
To make use of this we also split \(\Phi\) and Desc, making Desc a list of constructors,
in line with actual Agda datatypes
        data Con (Γ : Tel τ) (V : ExTel Γ) (I : Type) : Type<sub>1</sub>
        data Desc (Γ : Tel τ) (I : Type) : Type<sub>1</sub> where
           [] : Desc Γ I
           _::_ : Con \Gamma \oslash I \to Desc \Gamma I \to Desc \Gamma I
A constructor then starts off with the empty variable context, which grows as
fields are added
        data Con Γ V I where
           1: V \vdash I \rightarrow Con \Gamma V I
           \rho: V \vdash I \rightarrow Con \Gamma V I \rightarrow Con \Gamma V I
          \sigma: (S: V \vdash Type) \rightarrow Con \Gamma (V \triangleright S) I \rightarrow Con \Gamma V I
replacing I by V \vdash I in 1 and \rho allows the index of a constructor or argument
to depend on the preceding fields, of which the values are made accessible by
appending them to the context as V \triangleright S in \sigma. Finally, we interpret this as
        [\![\_]\!]C: \mathsf{Con} \; \Gamma \; \mathsf{V} \; \mathsf{I} \; \rightarrow \; ([\![\; \Gamma \; \& \; \mathsf{V} \;]\!] \; \mathsf{tel} \; \rightarrow \; \mathsf{I} \; \rightarrow \; \mathsf{Type}) \; \rightarrow \; ([\![\; \Gamma \; \& \; \mathsf{V} \;]\!] \; \mathsf{tel} \; \rightarrow \; \mathsf{I} \; \rightarrow \; \mathsf{Type})
        [1] j C \times pv i = i \equiv (j pv)
        [\rho j C] C X pv i = X pv (j pv) \times [C] C X pv i
        [\sigma S C] C X pv@(p, v) i = \Sigma[s \in S pv] [C] C (X \circ map_2 proj_1) (p, v, s) i
        [\_]D : Desc \Gamma I \rightarrow ([\Gamma]tel tt \rightarrow I \rightarrow Type) \rightarrow ([\Gamma]tel tt \rightarrow I \rightarrow Type)
        [ [ ]
                    D \times p i = \bot
        [ C :: Cs ] D X p i = [C] C (X \circ proj_1) (p, tt) i \uplus [Cs] D X p i
with the fixpoint
        data \mu (D : Desc \Gamma I) (p : [\Gamma]tel tt) (i : I) : Type where
           con: [D]D(\mu D)pi \rightarrow \mu Dpi
                                                                                                                       Formal-
                                                                                                                       izing the
```

Part I

Numerical representations and ornaments

Adapt and split into background and actual work

"looks like

relation".

3 Types from Specifications: Ornamentation and Calculation

Suppose that we started writing and verifying some code using a vector-based implementation of the two-sided flexible array interface, but later decide to reimplement more efficiently using trees. It would be a shame to lay aside our vector lemmas, and rebuild the correctness proofs for trees from scratch. Instead, we note that both vectors and trees can be represented by their lookup function. In fact, we can ask for more, and rather than defining an array-like type and then showing that it is represented by a lookup function, we can go the other way around and define types by insisting that they are equivalent to such a function. This approach, in particular the case in which one calculates a container with the same shape as a numeral system, was dubbed numerical representations by Okasaki [Oka98], and has some formalized examples due to Hinze and Swierstra [HS22] and Ko and Gibbons [KG16]. Numerical representations are our starting point for defining more complex datastructures based on simpler ones, so we demonstrate such a calculation.

3.1 From numbers to containers

We can compute the type of vectors starting from \mathbb{N} .

Is there a simple twist or other interesting example that we can run through instead, or would anything else be too abrupt without starting from this simple case?

⁷ For simplicity, we define them as a type computing function via the "use-as-definition" notation from before. We expect vectors to be represented by

```
Lookup: Type \rightarrow \mathbb{N} \rightarrow \text{Type}
Lookup A n = Fin n \rightarrow A
```

where we use the finite type Fin as an index into vector. Using this representation as a specification, we can compute both Fin and a type of vectors. The finite type can be computed from the evident definition

```
Fin-def: \forall n \rightarrow Def (\Sigma[ m \in \mathbb{N} ] m < n)

Fin-def zero =

(\Sigma[ m \in \mathbb{N} ] m < 0)

\equiv \langle 1-strict (\lambda ()) \rangle

1 use-as-def

Fin-def (suc n) =

(\Sigma[ m \in \mathbb{N} ] m < suc <math>n)

\equiv \langle ua (\leftarrow-split n) \rangle

\vdash \forall (\Sigma[ m \in \mathbb{N} ] m < n)

\equiv \langle cong (\top \forall) (by-definition (Fin-def n)) \rangle

\vdash \forall defined-by (Fin-def n) use-as-def
```

⁷This is adapted (and fairly abridged) from Calculating Datastructures [HS22]

```
Fin : N → Type
      Fin n = defined-by (Fin-def n)
using
      \leftarrow-split : \forall n → (\Sigma[ m ∈ N ] m < suc n) \simeq (T \uplus (\Sigma[ m ∈ N ] m < n))
Likewise, vectors can be computed by applying a sequence of type isomorphisms
      Vec-def : \forall A n \rightarrow Def (Lookup A n)
      Vec-def A zero =
            (\bot \rightarrow A)
        ≡⟨ isContr→≡Unit isContr⊥→A ⟩
            τ ■ use-as-def
      Vec-def A (suc n) =
            ((T \uplus Fin n) \rightarrow A)
        ≡( ua Π⊎≃ )
            (T \rightarrow A) \times (Fin n \rightarrow A)
         ≡( cong<sub>2</sub> _×_
              (UnitToTypePath A)
              (by-definition (Vec-def A n)) >
            A × (defined-by (Vec-def A n)) ■ use-as-def
      Vec : ∀ A n → Type
      Vec A n = defined-by (Vec-def A n)
   SIP doesn't mesh very well with indexed stuff, does HSIP help?
   We can implement the following interface using Vec
      record Array (V: Type → N → Type): Type<sub>1</sub> where
        field
           lookup : \forall \{A n\} \rightarrow V A n \rightarrow Fin n \rightarrow A
           tail: \forall \{A n\} \rightarrow V A (suc n) \rightarrow V A n
and show that this satisfies some usual laws like
      record ArrayLaws {C} (Arr : Array C) : Type1 where
        field
           lookup∘tail : ∀ {A n} (xs : C A (suc n)) (i : Fin n)
                         → Arr .lookup (Arr .tail xs) i = Arr .lookup xs (inr i)
Since we defined Vec such that it agrees with Lookup, we can relate their imple-
mentations as well.
   The implementation of arrays as functions is straightforward
      FunArray: Array Lookup
      FunArray .lookup f = f
      FunArray .tail f = f ∘ inr
and clearly satisfies our interface
      FunLaw: ArrayLaws FunArray
      FunLaw .lookupotail _ _ = refl
We can implement arrays based on Vec as well<sup>8</sup>
```

⁸Note that, like any other type computing representation, we pay the price by not being

```
VectorArray : Array Vec
VectorArray .lookup {n = n} = f n
  where
  f : ∀ {A} n → Vec A n → Fin n → A
  f (suc n) (x , xs) (inl _) = x
  f (suc n) (x , xs) (inr i) = f n xs i
VectorArray .tail (x , xs) = xs
```

Now, rather than rederiving the laws for vectors, the equality allows us to transport them from Lookup to $\mathsf{Vec.}^9$

As you can see, taking "use-as-definition" too literally prevents Agda from solving a lot of metavariables.

This computation can of course be generalized to any arity zeroless numeral system; unfortunately beyond this set of base types, this "straightforward" computation from numeral system to container loses its efficacy. In a sense, the n-ary natural numbers are exactly the base types for which the required steps are convenient type equivalences like $(A + B) \to C = (A \to C) \times (B \to C)$?

3.2 Numerical representations as ornaments

Reflecting on this derivation for \mathbb{N} , we could perform the same computation for Leibniz to get Braun trees. However, we note that these computations proceed with roughly the same pattern: each constructor of the numeral system gets assigned a value, and is amended with a field holding a number of elements and subnodes using this value as a "weight". This kind of "modifying constructors" is formalized by ornamentation [KG16], which lets us formulate what it means for two types to have a "similar" recursive structure. This is achieved by interpreting (indexed inductive) datatypes from descriptions, between which an ornament is seen as a certificate of similarity, describing which fields or indices need to be introduced or dropped to go from one description to the other. *Ornamental descriptions*, which act as one-sided ornaments, let us describe new datatypes by recording the modifications to an existing description.

Put some minimal definitions here.

Looking back at Vec, ornaments let us show that express that Vec can be formed by introducing indices and adding a fields holding an elements to \mathbb{N} . However, deriving List from \mathbb{N} generalizes to Leibniz with less notational overhead, so we tackle that case first. We use the following description of \mathbb{N}

```
NatD : Desc \tau \ell-zero NatD _ = \sigma Bool \lambda
```

able to pattern match directly on our type.

⁹Except that due to the simplicity of this case, the laws are trivial for Vec as well.

```
{ false → y []
; true → y [ tt ] }
```

Here, σ adds a field to the description, upon which the rest of the description can vary, and γ lists the recursive fields and their indices (which can only be tt). We can now write down the ornament which adds fields to the suc constructor

```
NatD-ListO : Type \rightarrow OrnDesc \tau ! NatD NatD-ListO A (ok _) = \sigma Bool \lambda { false \rightarrow \gamma _ ; true \rightarrow \Delta A (\lambda _ \rightarrow \gamma (ok _ , _)) }
```

Here, the σ and γ are forced to match those of NatD, but the Δ adds a new field. Using the least fixpoint and description extraction, we can then define List from this ornamental description. Note that we cannot hope to give an unindexed ornament from Leibniz

```
LeibnizD : Desc τ ℓ-zero

LeibnizD _ = σ (Fin 3) λ

{ zero → γ []

; (suc zero) → γ [ tt ]

; (suc (suc zero)) → γ [ tt ] }
```

into trees, since trees have a very different recursive structure! Thus, we must keep track at what level we are in the tree so that we can ask for adequately many elements:

```
power : \mathbb{N} \to (\mathbb{A} \to \mathbb{A}) \to \mathbb{A} \to \mathbb{A}

power \mathbb{N}.\mathsf{zero}\ f = \lambda \ x \to x

power (\mathbb{N}.\mathsf{suc}\ n)\ f = f \circ \mathsf{power}\ n\ f

Two : Type \to Type

Two X = X \times X

LeibnizD-TreeO : Type \to OrnDesc \mathbb{N} ! LeibnizD

LeibnizD-TreeO \mathbb{A}\ (\mathsf{ok}\ n) = \sigma\ (\mathsf{Fin}\ 3)\ \lambda

\{\ \mathsf{zero} \ \to \ \mathsf{v}\ \_

; (\mathsf{suc}\ \mathsf{zero}) \to \Delta\ (\mathsf{power}\ n\ \mathsf{Two}\ \mathbb{A})\ \lambda\ \_ \to \ \mathsf{v}\ (\mathsf{ok}\ (\mathsf{suc}\ n)\ ,\ \_)

; (\mathsf{suc}\ (\mathsf{suc}\ \mathsf{zero})) \to \Delta\ (\mathsf{power}\ (\mathsf{suc}\ n)\ \mathsf{Two}\ \mathbb{A})\ \lambda\ \_ \to \ \mathsf{v}\ (\mathsf{ok}\ (\mathsf{suc}\ n)\ ,\ \_)\ \}
```

We use the **power** combinator to ensure that the digit at position n, which has weight 2^n in the interpretation of a binary number, also holds its value times 2^n elements. This makes sure that the number of elements in the tree shaped after a given binary number also is the value of that binary number.

3.3 Heterogeneization

The situation in which one wants to collect a variety of types is not uncommon, and is typically handled by tuples. However, if e.g., you are making a game in Haskell, you might feel the need to maintain a list of "Drawables", which may be of different types. Such a list would have to be a kind of "heterogeneous list". In Haskell, this can be resolved by using an existentially quantified list, which, informally speaking, can contain any type implementing a given constraint, but

can only be inspected as if it contains the intersection of all types implementing this constraint.

This ports directly to Agda, but becomes cumbersome quickly, and impractical if we want to be able to inspect the elements. The alternative is to split our heterogeneous list into two parts; one tracking the types, and one tracking the values. In practice, this means that we implement a heterogeneous list as a list of values indexed over a list of types. This approach and mainly its specialization to binary trees is investigated by Swierstra [Swi20].

We will demonstrate that we can express this "lift a type over itself" operation as an ornament. For this, we make a small adjustment to RDesc to track a type parameter separately from the fields. Using this we define an ornament-computing function, which given a description computes an ornamental description on top of it:

```
Heto': (D E: RDesc \tau \ell-zero) (x: \dot{F} (\lambda \rightarrow D) (\mu (\lambda \rightarrow E) Type) Type tt) \rightarrow ROrnDesc (\mu (\lambda \rightarrow E) Type tt) ! D

Heto' (\gamma is) E x = \gamma (map-\gamma is x)

where

map-\gamma: (is: List \tau) \rightarrow \dot{P} is (\mu (\lambda \rightarrow E) Type) \rightarrow \dot{P} is (Inv!)

map-\gamma [] _ = _

map-\gamma (_:: is) (x , xs) = ok x , map-\gamma is xs

Heto' (\sigma S D) E (s , x) = \nabla s (Heto' (D s) E x)

Heto' (\dot{P} D) E (A , x) = \Delta [ _ \in A ] \dot{P} (Heto' D E x)

Heto : (D: RDesc \tau \ell-zero) \rightarrow OrnDesc (\mu (\lambda \rightarrow D) Type tt) ! \lambda \rightarrow D

Heto D (ok (con x)) = Heto' D D x
```

This ornament relates the original unindexed type to a type indexed over it; we see that this ornament largely keeps all fields and structure identical, only performing the necessary bookkeeping in the index, and adding extra fields before parameters.

As an example, we adapt the list description

```
ListD: Desc τ ℓ-zero

ListD = σ Bool λ

{ false → γ []

; true → ṗ (γ [ tt ]) }

List': Type ℓ → Type ℓ

List' A = μ ListD A tt
```

which is easily heterogeneized to an <code>HList</code>. In fact, <code>HetO</code> seems to act functorially; if we lift <code>Maybe</code> like

```
MaybeD: Desc τ ℓ-zero
MaybeD = σ Bool (λ
{ false → y []
; true → ṗ (y []) })

Maybe: Type ℓ → Type ℓ
Maybe A = μ MaybeD A tt
```

```
HMaybeD = L HetO (MaybeD tt) J
HMaybe = μ HMaybeD τ
then we can lift functions like head as
head: List' A → Maybe A
head (con (false , _)) = con (false , _)
head (con (true , a , _)) = con (true , a , _)

hhead: (As: List' Type) → HList As → HMaybe (head As)
hhead (con (false , _)) (con _) = con _
hhead (con (true , A , _)) (con (a , _)) = con (a , _ , _)
```

4 Finger trees

We know that some datastructures can be presented as non-redundant numerical representations, for example lists by unary numbers, random access lists by binary numbers [HS22], and, skew binary heaps by skew binary numbers [KG16]. So far, some of these examples do support amortized constant time cons, but they have at best logarithmic time snoc. This is reflected by their number systems, for which either the natural successor operation is logarithmic time, or is constant time, but can only act at the front. Instead, we will look at more redundant number systems, and refine these step-by-step to produce structures similar to finger trees. This gives us datastructures with fast access to both ends, and some of their properties for free.

4.1 Binary finger trees

If a datastructure has a numerical representation, we see that the operations on the datastructure must be coherent with the number system. Hence, if we want to have constant time cons and snoc, we must first have constant time suc anc cus. By starting from a symmetric number system, we can ensure good performance for both.

Note that such a system is necessarily redundant: if suc and cus both are amortized constant time, there must be cases where neither recurses (otherwise, there is a value and a sequence of sucs and cuss which cannot be amortized constant). On the other hand, both must clearly yield different values!

Symmetric unary numbers could be represented by a pair of Peano naturals, but would lead to a linear time lookup. By using a binary backbone for the numbers, we can get good suc and lookup

```
data Digit : Set where
   1 2 : Digit

data Bin : Set where
   0 1 : Bin
   _⟨___ : Digit → Bin → Digit → Bin
```

However, this shape is still not ideal. We can see that for values like

```
bad-1 : Bin
bad-1 = 2 \langle 2 \langle 2 \langle 1 \rangle 1 \rangle 1 \rangle 1
applying suc would give
bad-2 : Bin
bad-2 = 1 \langle 1 \langle 1 \langle 1 \langle 1 \langle 1 \rangle 1 \rangle 1 \langle 1 \langle 1 \langle 1 \rangle 1 \ra
```

Applying pred would take us back, so composing the two always takes logarithmic time [Cla20]. To avoid this, we can give the numbers bigger digits (the system merely goes from redundant to slightly more redundant)

```
data Digit: Set where
          123: Digit
       data Bin : Set where
          0 1 : Bin
          _{\langle - \rangle}: Digit \rightarrow Bin \rightarrow Digit \rightarrow Bin
Now applying suc to the pathological case
       good-1: Bin
       good-1 = 3 \langle 3 \langle 3 \langle 1 \rangle 1 \rangle 1 \rangle 1
produces
       good-2: Bin
       good-2 = 2 \langle 2 \langle 2 \langle 1 \langle 0 \rangle 1 \rangle 1 \rangle 1 \rangle 1
instead, for which both suc and pred are constant time ^{10}. We interpret this
number system as
       [_{-}]D : Digit \rightarrow \mathbb{N}
       [ 1 ]D = 1
       [ 2 ] D = 2
       [ 3 ]D = 3
       [\_]B : Bin \rightarrow \mathbb{N}
       [ 0 ]B = 0
       [ 1 ]B = 1
        [l(m)r]B = [l]D + 2 * [m]B + [r]D
```

To extract the datastructure, we must find a suitable index type for these numbers. Since the numbers are redundant, we can also get trees of different shapes with the same size, each having a different and incompatible index type. However, the trees of a fixed shape are represented by functions, and the isomorphisms will still hold.

The computation of the index type from the interpretation of the numbers is straightforward. We first compute the indices for digits, which yields the indices for the numbers

 $^{^{10}\}mathrm{More}$ formally, we can use recursive slowdown [Oka98; KT95] to show that any sequence of operations amortizes to constant time.

```
data IxB : Bin → Set where
          1-1: IxB 1
          \langle \rangle - l : IxD d \rightarrow IxB (d \langle n \rangle e)
          \langle \rangle-m1 : IxB n \rightarrow IxB (d \langle n \rangle e)
          \langle \rangle - m2 : IxB n \rightarrow IxB (d \langle n \rangle e)
          \langle \rangle-r : IxD e \rightarrow IxB (d \langle n \rangle e)
This represents simple finger trees as
       Array : Set → Bin → Set
       Array A b = IxB b \rightarrow A
To define the basic array operations like cons on these functions as datastruc-
tures, we again construct a Fin-like view for the indices. For this we produce
values corresponding to zero
       izero : \forall \{n\} \rightarrow IxB (succ n)
and induce the successor on the indices using
       isucc: IxB n \rightarrow IxB (succ n)
The view is similarly defined by
       data IxV: IxB (succ n) → Set where
          as-izero : IxV {n} izero
          as-isucc : (i : IxB n) → IxV (isucc i)
       iview : \{n : Bin\} \rightarrow (i : IxB (succ n)) \rightarrow IxV i
letting us define
       head : Array A (succ n) → A
       head \{n = 0\} f = f 1-1
       head \{n = 1\} f = f (\langle \rangle-l 1-1)
       head \{n = 1 \langle m \rangle r\} f = f(\langle \rangle - 1 2 - 1)
       head \{n = 2 \langle m \rangle r\} f = f(\langle \rangle - 1 \cdot 3 - 1)
       head \{n = 3 \langle m \rangle r\} f = f(\langle \rangle - 1 2 - 1)
       cons : A \rightarrow Array A n \rightarrow Array A (succ n)
       cons {n = n} x f i with iview i
        ... | as-izero = x
        ... | as-isucc i = f i
We can again trieify this to get a concrete datastructure<sup>11</sup>
       data Finger (A : Set) : Digit → Set where
          1 : A → Finger A 1
          2: A \rightarrow A \rightarrow Finger A 2
          3: A \rightarrow A \rightarrow A \rightarrow Finger A 3
       data Array' (A : Set) : Bin → Set where
          \mathbb{O} : Array' A \mathbb{O}
          1 : A → Array' A 1
          -\langle - \rangle_-: Finger A d \rightarrow Array' A n \rightarrow Array' A n \rightarrow Finger A e \rightarrow Array' A (d \langle n \rangle e)
```

 $^{^{11}\}mathrm{I'll}$ probably not do this manually, because it is theoretically analogous to the other trees, but hellish in practice

Consequently, the concrete version will now obey all the relations the representable arrays obey as well. For example, for representable arrays we can easily see

```
(x : A) (xs : Array A n) \rightarrow head (cons x xs) \equiv x hence, the concrete arrays obey this as well.
On the other hand, as \forall n \rightarrow succ (cuss n) \equiv cuss (succ n)
```

does not generally hold for symmetric binary, cons will not interchange with snoc for finger trees either¹²; it seems that binary finger trees are not a very nice array type. Likewise, indexing into the finger trees is impractical, as changing shapes would require inefficient re-indexing.

4.2 Restoring efficient lookup

Can we restore lookup? We can probably do something similar to the original finger trees, and maintain the sizes internally (hopelessly breaking the isomorphism¹³). Then we could state that a fingertree of a given size is just a finger tree of a shape paired with a proof that this shape has the right size.

5 Nested ornaments

To capture finger trees as an ornament over a number system, we will need to describe ornaments over nested datatypes. In this section we will work out descriptions and ornaments suitable for nested datatypes.

We construct descriptions for nested datatypes by extending the encoding of parametric and indexed datatypes with three features: information bundles, parameter transformation, and description composition. Also, to make sharing constructors easier, we introduce variable transformations. Transforming variables before they are passed to child descriptions allows both aggressively hiding variables and introducing values as if by let-constructs.

We base the encoding of off existing encodings [Sij16; EC22], and are shaped as sums of products descriptions, enforce indices at leaf nodes, and have explicit split parameter and variable telescopes. Unlike some encodings, we do not allow higher-order inductive arguments.

We use type-in-type and with-K to simplify the presentation, noting that these can be eliminated respectively by moving to Type ω and by implementing interpretations as datatypes.

5.1 The descriptions

We use telescopes identical to those in Subsection 2.5:

```
data Tel (P : Type) : Type [-]tel : (\Gamma : Tel P) \rightarrow P \rightarrow Type
```

¹²For starters, the types are different

 $^{^{13}{}m Or}$ would it stay intact, since the shape determines the size anyway?

```
_⊢_: (Γ: Tel P) → Type → Type

Γ ⊢ A = \Sigma _ [Γ]tel → A

data Tel P where

∅: Tel P

_▷_: (Γ: Tel P) (S: Γ⊢ Type) → Tel P

[∅]tel p = τ

[Γ▷ S]tel p = \Sigma ([Γ]tel p) (S∘ (p,_))

ExTel: Tel \tau → Type

ExTel Γ = Tel ([Γ]tel tt)
```

Recall that a Tel represents a sequence of types, which can depend on the external type P. This lets us represent a telescope succeeding another using ExTel. A term of the interpretation [-]tel is then a sequence of terms of all the types in the telescope.

We define a couple more shorthands to handle extension telescopes and telescope transformations

```
_□'_: (Γ : Tel P) (S : Type) → Tel P

Γ □' S = Γ □ const S

_&_⊢_: (Γ : Tel τ) → ExTel Γ → Type → Type

Γ & V ⊢ A = V ⊢ A

[_&_]tel : (Γ : Tel τ) (V : ExTel Γ) → Type

[ Γ & V ]tel = Σ ([ Γ ]tel tt) [ V ]tel

Cxf : (Γ Δ : Tel τ) → Type

Cxf Γ Δ = [ Γ ]tel tt → [ Δ ]tel tt

Vxf : (Γ : Tel τ) (V W : ExTel Γ) → Type

Vxf Γ V W = ∀ {p} → [ V ]tel p → [ W ]tel p

Vxf0 : (f : Cxf Γ Δ) (V : ExTel Γ) (W : ExTel Δ) → Type

Vxf0 f V W = ∀ {p} → [ V ]tel p → [ W ]tel (f p)
```

As we will see in Subsection 5.3, some generics require descriptions augmented with more information. For example, a number system needs to describe both a datatype and its interpretation into naturals. This can be incorporated into a description by allowing description formers to query specific pieces of information. We will control where and when which pieces get queried by parametrizing descriptions over information bundles

```
record Info : Type where
  field
    1i : Type
    ρi : Type
    σi : (S : Γ & V ⊢ Type) → Type
    δi : Tel T → Type → Type
```

Here a bundle declares for example that 1i is the type of information has to be provided at a 1 former. Remark that in σi the type of the field is passed along to the information. We can recover the conventional descriptions by providing the plain bundle

```
Plain: Info
Plain.1i = T
Plain.pi = T
Plain.oi = T
Plain.8i = T
```

To allow reusing more specific descriptions in a less specific one, e.g., a number system in a plain datatype, we define the "down-casting" of information as follows

We can now define the descriptions, which should represent a mapping between parametrized indexed functors

```
PIType : Tel T \rightarrow Type \rightarrow Type
PIType \Gamma J = [\Gamma]tel tt \rightarrow J \rightarrow Type
```

Recall that a description is simply a list of constructor descriptions

```
data DescI If Γ J where
[] : DescI If Γ J
_::_: ConI If Γ J Ø → DescI If Γ J → DescI If Γ J
```

We define constructor descriptions

```
data ConI (If : Info) (\Gamma : Tel \tau) (J : Type) (V : ExTel \Gamma) : Type where as follows.
```

Leaves are formed by

```
1: {if: If.1i} (j: \Gamma \& V \vdash J) \rightarrow ConI If \Gamma J V
```

where if is the queried information. The function j determines the index of the leaf given the parameters and variables.

A recursive field is formed by

```
\begin{array}{l} \rho: \; \{\text{if}: \; \text{If} \; .\rho \mathbf{i}\} \\ \qquad (\texttt{j}: \; \Gamma \; \& \; \mathsf{V} \vdash \mathsf{J}) \; (\texttt{g}: \; \mathsf{Cxf} \; \Gamma \; \Gamma) \; (\texttt{C}: \; \mathsf{ConI} \; \mathsf{If} \; \Gamma \; \mathsf{J} \; \mathsf{V}) \\ \rightarrow \; \mathsf{ConI} \; \mathsf{If} \; \Gamma \; \mathsf{J} \; \mathsf{V} \end{array}
```

where j now determines the index of the recursive field. The parameter transform g is new, and determines the parameters of the recursive field; this is exactly what allows us to describe a limited form of nested datatypes. The remainder of the fields are described by C. Note that a recursive field is intentionally not brought into scope: making use of it requires induction-recursion anyway!

A non-recursive field is formed similarly to a recursive field

```
\sigma: (S: V \vdash \mathsf{Type}) \{ \mathsf{if}: \mathsf{If} . \sigma \mathsf{i} S \} 
 (h: \mathsf{Vxf} \Gamma (V \triangleright S) \ \mathsf{W}) \ (C: \mathsf{ConI} \ \mathsf{If} \Gamma \ \mathsf{J} \ \mathsf{W})
```

```
→ ConI If Γ J V
```

The type of the field is given by S, which may depend on the values of the preceding fields. Since this field is brought into scope, as opposed to a recursive field, we should continue the description in an extended context. However, we allow the remainder of the description to select a different context W, provided we can convert V extended with S into W. This makes it possible to hide fields on which the subsequent fields do not depend.

Almost analogously, we make composition of descriptions internal by a variant of σ , taking a description and acting like the σ of its fixpoint, only with more ceremony

```
δ: {if: If.δi Δ K} {iff: InfoF If' If}
    (j: Γ & V ⊢ K) (g: Γ & V ⊢ [[ Δ ]] tel tt) (R: DescI If' Δ K)
    (h: Vxf Γ (V ⊳ liftM2 (μ R) g j) W) (C: ConI If Γ J W)
    → ConI If Γ J V
```

Similar to ρ , the functions j and g control indices and parameters, only now of the applied description. As we allow the subdescription R to have a different kind of information bundle If', we must ask that we can down-cast it into If via iff. This constructor allows us to form descriptions by composing other descriptions, which lets us avoid multiplying the number of constructors of composite datatypes as we will see later.

Descriptions and constructor descriptions can then be interpreted to appropriate kind of functor, constructor descriptions also taking variables

We see that a leaf becomes a constraint between expected index and the actual index. A recursive field passes down a transformation of the current parameters and the expected index computed from the variables, before interpreting the remainder of the description. Likewise, a non-recursive field adds a field with type depending on variables, but also adds this field to the variables, which

are then transformed and passed on to the remainder. The composite field is analogous, only adding a field from a description rather than a type. Finally, the list of constructor descriptions are interpreted as alternatives.

The fixpoint can then be taken over the interpretation of a description

```
con: \forall {i} \rightarrow [ D ] (\mu D) p i \rightarrow \mu D p i giving the datatype represented by the description.

Let's look at some examples, we define more notation

\neg \bot: (X Y: A \rightarrow Type) \rightarrow Type

X \rightrightarrows Y = \forall a \rightarrow X a \rightarrow Y a

\neg \bot: (X Y: A \rightarrow B \rightarrow Type) \rightarrow Type

X \equiv Y = \forall a b \rightarrow X a b \rightarrow Y a b

liftM2: (A \rightarrow B \rightarrow C) \rightarrow (X \rightarrow A) \rightarrow (X \rightarrow B) \rightarrow X \rightarrow C

liftM2 f g h x = f (g x) (h x)

!: {A: Type} \rightarrow A \rightarrow T
```

We can then give a generic fold for the represented datatypes which descends the description, mapping itself over all recursive fields before applying the folding function.

Remark 5.1. The situation of fold is very common when dealing with different kinds of recursive interpretations: functions from the fixpoint are generally defined from functions out of the interpretation, generalizing over the inner description while pattern matching on the outer description.

Note that the fold requires a rather general function, limiting its usefulness: because of the parameter transformations, we cannot instantiate the fold to a single parameter. Defining, e.g., the vector sum, would require us to inspect the description, and ask that a vector of naturals can be converted into a vector of naturals, which is trivial in this case.

As an example, we can encode the naturals as a type parametrized by \varnothing and indexed by \intercal

Sigma plus/minus

```
NatD : Desc ∅ T
NatD = 1 _
:: ρ0 _ (1 _)
:: []
```

data µ D p where

Lists can be encoded similarly, but this time using the telescope

```
ListTel: Tel τ
ListTel = ∅ ⊳ const Type
```

declaring that lists have a single type parameter. Compared to the naturals, the description now also asks for a field in the second case

Since the type parameter is at the top of the parameter telescope, the type of the field is given as par top.

Vectors are described using the same structure, but have indices in N.

In the first case, the index is fixed at 0. The second case declares that to construct a vector of length n+1, here computed by $\mathsf{suc} \, \circ \, \mathsf{top}$, the recursive field must have length n, top . Note that unlike index-first types, we cannot know the expected index from inside the description, so much like native indexed types, we must add a field choosing an index.

Recall the type of finger trees. Using parameter transformations and composition, we can give a description of full-fledges finger trees! First, we describe the digits

```
DigitD : Desc (∅ ⊳ const Type) т
       DigitD = \sigma- (par top) (1 _)
                :: \sigma- (par top) (\sigma- (par top) (1_))
                :: \sigma- (par top) (\sigma- (par top) (\sigma- (par top) (1 _)))
                :: []
and define the nodes<sup>14</sup>
       data Node (A: Type): Type where
         two : A \rightarrow A
                               → Node A
         three: A \rightarrow A \rightarrow A \rightarrow Node A
We encode finger trees as
       FingerD : Desc (∅ ⊳ const Type) τ
       FingerD = 1_{-}
                 ∷ σ- (par top) (1 _)
                 :: \delta - (par((tt, -) \circ top)) DigitD
                 (\rho_{-}(\lambda \{ (-, A) \rightarrow (-, Node A) \})
                 (\delta - (par((tt, -) \circ top)) DigitD(1 -)))
```

In the third case, we have digits which are passed the parameters on both sides in composite fields, and a recursive field in the middle. The recursive field has a parameter transformation, turning the type parameter A into a Node A in the recursive child.

Variable transforms are both less essential and less troublesome than I first thought. We can move variable transforms into a new Desc-former, and it probably simplifies the definition of ornaments a lot.

This setup does not allow nesting over recursive fields, which is necessary

¹⁴We could give the nodes as a description, but in this case we only use them in the recursive fields, so we would take the fixpoint without looking at their description anyway.

for structures like RoseTrees. This is actually kind of essential for enumeration. Nesting over a recursive field is problematic: we can incorporate it by adding "this" implicitly to a δ , but then the R needs to be strictly positive in its last argument, meaning we need to split **Desc** into a strictly positive part and normal part. The strictly positive part should then only allow strictly positive parameter transforms in recursive and non-recursive fields, requiring an embedding of transforms.

If we levitate, then informed descriptions become ornaments over desc. This gives us the best of both worlds (modulo reflecting the description into a datatype): in plain descriptions, information does not even exist, and in informed descriptions, it is explicit. For levitation, we likely need induction-recursion and mutual recursion.

Making Desc coinductive would do a couple of things. First, recursion and composition become identical. Second, nesting of both types becomes easier to describe, but potentially impossible to prove strictly positive.

We intentionally dodge having index telescopes (or having the index type depend on the parameters and values). Does this really change anything?

5.2 The ornaments

Now that we have descriptions, we can start relating them. We will be constructing ornaments as a binary relation on descriptions

data Orn {If} {If'} (f: Cxf $\Delta \Gamma$) (e: K \rightarrow J): DescI If Γ J \rightarrow DescI If' Δ K \rightarrow Type which, informally, will encode a proof that "E is a more informative variant of D". This means that we can always convert the parameters and indices of E into those of D, as witnessed by f and e. Furthermore, we also have to be able to convert¹⁵ values of E to D

```
ornForget 0 p = fold (ornAlg 0) p _
```

Thus, the relation should be precise enough pairs of E and D for which we could not define ornForget.

We will walk through the constructor ornaments

```
data ConOrn {If} {If'} {c : Cxf \Delta \Gamma} (f : VxfO c W V) (e : K \rightarrow J)
: ConI If \Gamma J V \rightarrow ConI If' \Delta K W \rightarrow Type where
```

again, an ornament between datatypes is just a list of ornaments between their constructors

```
data Orn f e where
[] : Orn f e [] []
```

 $^{^{15}\}mathrm{I}$ would love to require this conversion to be epi, but we can sigma a field of the empty type.

```
_::_ : ∀ {D E D' E'}

→ ConOrn {c = f} id e D' E'

→ Orn f e D E

→ Orn f e (D' :: D) (E' :: E)
```

Note that all ornaments completely ignore information bundles! They cannot affect the existence of ornForget after all.

Copying parts from one description to another, up to parameter and index refinement, corresponds to reflexivity. Preservation of leaves follows the rule

```
1 : ∀ {k j}

→ (∀ p → e (k p) ≡ j (over f p))

→ ∀ {if if'}

→ ConOrn f e (1 {if = if} j) (1 {if = if'} k)
```

where the only condition is that the indices, and the parameter and index conversions, fit in a commuting square.

Remark 5.2. Rather than having the user provide two indices and show that the square commutes, we can ask for a "lift" k



and derive the indices as i=ek, j=kf. However, this is more restrictive, unless f is a split epi, as only then pairs i,j and arrows k are in bijection. In addition, this makes ornaments harder to work with, because we have to hit the indices definitionally, whereas asking for the square to commute gives us some leeway (i.e., the lift would require the user to transport the ornament).

Preserving a recursive field¹⁶ similarly requires a square of indices and conversions to commute, and additionally requires the recursive parameters to commute with the conversion

```
\begin{split} \rho : & \forall \; \{k \; j \; g \; h \; D \; E\} \\ & \rightarrow ConOrn \; f \; e \; D \; E \\ & \rightarrow (\forall \; p \rightarrow g \; (c \; p) \equiv c \; (h \; p)) \\ & \rightarrow (\forall \; p \rightarrow e \; (k \; p) \equiv j \; (over \; f \; p)) \\ & \rightarrow \forall \; \{if \; if'\} \\ & \rightarrow ConOrn \; f \; e \; (\rho \; \{if = if\} \; j \; g \; D) \; (\rho \; \{if = if'\} \; k \; h \; E) \end{split}
```

Preservation of non-recursive fields and preservation of description fields is analogous

 $^{^{16}\}mathrm{Kind}$ of breaking the "ornaments relate types with similar recursive structure" interpretation.

differing only in that non-recursive fields appears transformed on the right hand, while description fields have their conversions modified instead. For this rule, we need that the variable transformations fit into a commuting square with the parameter conversions. The condition on indices for descriptions, which is a commuting triangle, is encoded in the return type¹⁷.

Ornaments would not be very interesting if they only related identical structures. The left-hand side can also have more fields than the right-hand side, in which case ornForget will simply drop the fields which have no counterpart on the right-hand side. As a consequence, the description extending rules have fewer conditions than the description preserving rules. Extension with a recursive field has no conditions

```
\Delta \rho: \forall {D : ConI If \Gamma J V} {E} {k} {h} 

\rightarrow ConOrn f e D E

\rightarrow \forall {if}

\rightarrow ConOrn f e D (\rho {if = if} k h E)
```

and extending by a non-recursive field or a description field again only requires the variable transform to interact well with the parameter conversion

```
Δσ: ∀ {W'} {S} {D: ConI If Γ J V} {E: ConI If ΄ Δ K W'}

→ (f': VxfO c _ _) → {h: Vxf Δ _ _}

→ ConOrn f' e D E

→ (∀ {p'} (p: [ W ▷ S ]tel p') → f (p.proj₁) ≡ f' (h p))

→ ∀ {if'}

→ ConOrn f e D (σ S {if = if'} h E)

Δδ: ∀ {W'} {R: DescI If" Θ L} {D: ConI If Γ J V} {E: ConI If' Δ K W'}

{f': VxfO c _ _} {m} {k} {h: Vxf Δ _ _}

→ ConOrn f' e D E

→ (∀ {p'} (p: [ W ▷ liftM2 (μ R) m k ]tel p') → f (p.proj₁) ≡ f' (h p))

→ ∀ {if' iff'}

→ ConOrn f e D (δ {if = if'} {iff = iff'} k m R h E)
```

In the other direction, the left-hand side can also omit a field which appears on the right-hand side, provided we can produce a default value

 $^{^{17}}$ Should this become a problem like with ρ , modifying the rule to require a triangle is trivial.

```
∇σ: ∀ {S} {V'} {D: ConI If Γ J V'} {E: ConI If ΄Δ K W} {g: Vxf Γ _ _}

→ (s: V ⊨ S)

→ ConOrn ((g∘ ⊨-▷ s)∘ f) e D E

→ ∀ {if}

→ ConOrn f e (σ S {if = if} g D) E

∇δ: ∀ {R: DescI If "Θ L} {V'} {D: ConI If Γ J V'} {E: ConI If ΄Δ K W}

{m} {k} {g: Vxf Γ _ _}

→ (s: V ⊨ liftM2 (μ R) m k)

→ ConOrn ((g∘ ⊨-▷ s)∘ f) e D E

→ ∀ {if iff}

→ ConOrn f e (δ {if = if} {iff = iff} k m R g D) E
```

These rules let us describe a decent set of useful relations between datatypes.

But we would intuitively also expect a conversion to exist when two constructors have description fields which are not equal, but are only related by an ornament. Such a composition of ornaments takes two ornaments, one between the field, and one between the outer descriptions. We first require two commuting squares, one relating the parameters of the fields to the inner and outer parameter conversions, and one relating the indices of the fields to the inner index conversion and the outer parameter conversion. Then, the last square has a rather complicated equation, which merely states that the variable transforms for the remainder respect the outer parameter conversion. The composition rule then reads

```
•δ : ∀ {ΘΛΜLW'V'} {D : ConI If Γ J V'} {E : ConI If ' Δ K W'}
             \{R : DescI \ If'' \ O \ L\} \ \{R' : DescI \ If''' \ A \ M\} \ \{c' : Cxf \ A \ O\} \ \{e' : M \rightarrow L\}
             \{f'': Vxf0 c W' V'\} \{f\Theta: V \vdash [\Theta] tel tt\} \{f\Lambda: W \vdash [\Lambda] tel tt\}
             \{l: V \vdash L\} \{m: W \vdash M\} \{g: Vxf_(V \triangleright_) V'\} \{h: Vxf_(W \triangleright_) W'\}
       → ConOrn f'' e D E
       → (0 : Orn c' e' R R')
       \rightarrow (p<sub>1</sub>: \forall q w \rightarrow c' (f\Lambda (q , w)) \equiv f\Theta (c q , f w))
       \rightarrow (p<sub>2</sub>: \forall q w \rightarrow e' (m (q , w)) \equiv l (c q , f w))
       \rightarrow ( \forall {p'} (p : [ W \triangleright liftM2 (\mu R') fA m ]tel p')
             → f'' (h p)
                \equiv g (Vxf0-\triangleright-map f (liftM2 (\mu R) f\Theta l) (liftM2 (\mu R') f\Lambda m)
                     (\lambda \neq x \rightarrow subst2 (\mu R) (p_1 - ) (p_2 - ) (ornForget O (f \land (q, w)) x)) p))
       \rightarrow \forall \{if if'\}
       → ∀ {iff iff'}
       \rightarrow ConOrn f e (\delta {if = if} {iff = iff} l f\Theta R g D)
                           (\delta \{if = if'\} \{iff = iff'\} m f \land R' h E)
We will construct ornForget as a fold. Using
   pre_2 : (A \rightarrow B \rightarrow C) \rightarrow (X \rightarrow A) \rightarrow (Y \rightarrow B) \rightarrow X \rightarrow Y \rightarrow C
   pre_2 fabxy = f(ax)(by)
   erase : ∀ {D : DescI If Γ J} {E : DescI If 'Δ K} {f} {e} {X : PIType Γ J}
            \rightarrow Orn f e D E \rightarrow \forall p k \rightarrow [ E ] (pre<sub>2</sub> X f e) p k \rightarrow [ D ] X (f p) (e k)
```

we can define the algebra which forgets the added structure of the outer layer

```
\begin{array}{l} \mbox{ornAlg}: \ \forall \ \{D: DescI \ \mbox{if} \ \Gamma \ J\} \ \{E: DescI \ \mbox{if}' \ \Delta \ K\} \ \{f\} \ \{e\} \\ \rightarrow \mbox{Orn} \ f \ e \ D \ E \\ \rightarrow \ [ \ E \ ] \ (\lambda \ p \ k \rightarrow \mu \ D \ (f \ p) \ (e \ k)) \ \equiv \ \lambda \ p \ k \rightarrow \mu \ D \ (f \ p) \ (e \ k) \\ \mbox{ornAlg} \ O \ p \ k \ x = con \ (erase \ O \ p \ k \ x) \end{array}
```

Folding over this algebra gives the wanted function

```
ornForget 0 p = fold (ornAlg 0) p _
```

Now we can show that the descriptions we gave in Subsection 5.1 are related. The ornament between naturals and lists is

We use ! to convert parameters, naturals have no parameters, so we can map every parameter of lists to the empty sequence. The index conversion is id, since neither type has an index. All structure is preserved; we just have to note that lists have an added field using $\Delta \sigma$, and all commutativity squares are trivial, since naturals have neither parameters nor indices.

We can also relate lists and vectors

```
ListD-VecD : Orn id ! ListD VecD
ListD-VecD = 1 (const refl)

:: \sigma id

( \Delta \sigma _ ( \rho (1 (const refl)) (\lambda p \rightarrow refl) (const refl))

\lambda p \rightarrow refl)

(const refl)

:: []
```

Now the parameter conversion is the identity, since both have a single type parameter. The index conversion is !, since lists have no indices. Again, most structure is preserved, we only note that vectors have an added field carrying the length.

Instantiating ornForget to these ornaments, we now get the functions length and toList for free!

Having a function of the same type as **ornForget** is not sufficient to deduce an ornament. An obstacle is that the usual empty type (no constructors) and the non-wellfounded empty type (only a recursive field) don't have an ornament. Also, while the leaf-preservation case spells itself out, the substitutions obviously don't give us a way to recover the equalities.

5.2.1 Ornamental descriptions

Can these be simpler? Right now, these just construct the ornament and description on the fly, rather than actually asking for less.

Does "Ornamental algebras, algebraic ornaments" still hold for our ornaments?

5.3 Numerical descriptions, and the Trie ornament

We demonstrate the capabilities of our encoding by applying it to the calculation of datastructures. We first define the kind of information constituting a type of "natural numbers" the semantics of the type is given by the conversion to Ndefined by generalizing over the inner information bundle and folding using This restricts the numbers to those that are computed by linear functions, which certainly do not include all interesting number systems, but do include almost all systems that have associated containers 18 . Note that an arbitrary number system of this kind is not necessarily isomorphic to \mathbb{N} , as can be incomplete or redundant.

The description of Nand Leibnizis straightforward The essence of the calculation of arrays is that given a number system, we can calculate a datastructure which still has the same shape, and has the correct number of elements. We can generalize the calculation to all number systems while proving that the shape is preserved by presenting the datastructure by an ornamental description.

We could directly compute indexed array, using the index for the proof of representability, and from it the correctness of numbers of elements. However, we give the unindexed array first: we can get the indexed variant for free[McB14]! We claim that also is foldable in its argument, and folding with λ x y \rightarrow suc y coincides with ornForget. The ornamental description is computed by generalizing over the inner information bundle, and by cases on the description Overall, most structure is directly preserved, only inserting fields containing vectors of the appropriate size.

Currently, without proof

Index types are a simple ornament over number types: paths. This is not quite like [DS16].

Is Ix $x \rightarrow A$ initial for the algebra of the algebraic ornament induced by TrieO? (This is [HS22]).

 $^{^{18}\}mbox{Notably, polynomials still calculate to data$ structures, interpreting multiplication as precomposition.

While evidently Ix x = Fin (toN x) for arbitrary number systems, does the suspected iso Ix x -> A = Trie A x yield Traversable, for free?

5.4 Comparison

We compare our implementation to a selection of previous work, considering the following features

g	Haskell	[JG07]	[Cha+10]	0] [McB14]	[KG16]	
Fixpoint	yes*	yes	no	yes?	yes	
Index		_	first**	equality	first	
Poly	yes	1	externa	l external	external	
Levels		_	no	no	no	
Sums	list	_	large	large	large	
IndArg	any	any	$\cdots \rightarrow X$	i X i	X^{-i}	
Compose	yes	yes	no	no	no	
Extension	_	_	no		_	
Ignore		_	_		_	
Set	_	_	_	_	_	
	[Sij16]	[eff:	20]	[EC22]	Shallow	Deep (old)
Fixpoint	[Sij16] yes	[eff:		[EC22] no	Shallow	Deep (old) yes
Fixpoint Index			es			
_	yes equality telescope	y€	es	no	yes	
Index	yes equality	ye equa	es llity rnal	no equality	$_{ m yes}$ equality	
Index Poly	yes equality telescope	ye equa exte	es dity rnal lative	no equality telescope	yes equality telescope	
Index Poly Levels	yes equality telescope no***	ye equa exte cumul lar	es ality rnal lative ge	no equality telescope Typeω	yes equality telescope Type-in-Type	
Index Poly Levels Sums	yes equality telescope no*** list	ye equa exte cumul lar	es Ility rnal lative ge X v i ···	no equality telescope Type \Omega list	yes equality telescope Type-in-Type list	yes
Index Poly Levels Sums IndArg	yes equality telescope no*** list X pv i	$y\epsilon$ equa exter cumul lar $\cdots \rightarrow$	es sality rnal lative $X \ v \ i \cdots$?2	no equality telescope Type ω list $\cdots \rightarrow X \ pv \ i$	$\begin{array}{c} \text{yes} \\ \text{equality} \\ \text{telescope} \\ \text{Type-in-Type} \\ \text{list} \\ X(fpv)i \end{array}$	yes
Index Poly Levels Sums IndArg Compose	yes equality telescope no*** list X pv i	equa exte cumul lar > yes	es sality rnal lative $x = x + y = 0$ $x = x $	no equality telescope Type ω list $\cdots \rightarrow X \ pv \ i$ no	yes equality telescope Type-in-Type list $X(fpv)i$ yes	yes

- IndArg: the allowed shapes of inductive arguments. Note that none other than Haskell, higher-order functors, and potentially ?1, allow full nested types!
- Compose: can a description refer to another description?
- Extension: do inductive arguments and end nodes, and sums and products coincide through a top-level extension?
- Ignore: can subsequent constructor descriptions ignore values of previous ones? (Either this, or thinnings, are essential to make composites work)
- Set: are sets internalized in this description?
- * These descriptions are "coinductive" in that they can contain themselves, so the "fixpoint" is more like a deep interpretation.

- ** This has no fixpoint, and the generalization over the index is external.
- *** But you could bump the parameter telescope to Type ω and lose nothing.
- *4 A variant keeps track of the highest level in the index.
- ?1 Deeply encoding all involved functors would remove the need for positivity annotations for full nested types like in other implementations.
- ?2 The "simplicity" of this implementation, where data and constructor descriptions coincide, automatically allows composite descriptions.

We take away some interesting points from this:

- Levels are important, because index-first descriptions are incompatible with "data-cumulativity" when not emulating it using equalities! (This results in datatypes being forced to have fields of a fixed level).
- Coinductive descriptions can generate inductive types!
- Type ω descriptions can generate types of any level!
- Large sums do not reflect Agda (a datatype instantiated from a derived description looks nothing like the original type)! On the other hand, they make lists unnecessary, and simplify the definition of ornaments as well.
- We can group/collapse multiple signatures into one using tags, this might be nice for defining generic functions in a more collected way.
- Everything becomes completely unreadable without opacity.

5.5 Descriptions

At the very least, descriptions will need sums, products, and recursive positions as well. While we could use coinductive descriptions, bringing normal and recursive fields to the same level, we avoid this as it also makes ornaments a bit more wild¹⁹. We represent indexed types by parametrizing over a type I. Since we are aiming for nested types, external polymorphism²⁰ does not suffice: we need to let descriptions control their contexts.

We describe parameters by defining descriptions relative to a context. Here, a context is a telescope of types, where each type can depend on all preceding types:

• • •

Much like the work Escot and Cockx [EC22] we shove everything into Typew, but we do not (yet) allow parameters to depend on previous values, or indices on parameters²¹.

¹⁹For better or worse, an ornament could refer to a different ornament for a recursive field.

 $^{^{20}}$ E.g., for each type A a description of lists of A à la [KG16]

²¹I do not know yet what that would mean for ornaments.

We use equalities to enforce indices, simply because index-first types are not honest about being finite, and consequently mess up our levels. For an index type and a context a description represents a list of constructors:

. . .

These represent lists of alternative constructors, which each represent a list of fields:

. . .

We separate mere fields from "known" fields, which are given by descriptions rather than arbitrary types. Note that we do not split off fields to another description, as subsequent fields should be able to depend on previous fields

. . . .

We parametrize over the levels, because unlike practical generic, we stay at one level.

Q: what happens when you precompose a datatype with a function? E.g. (List . f) A = List (f A)

Q: practgen is cool, compact, and probably necessary to have all datatypes. Note that in comparison, most other implementations (like Sijsling) do not allow functions as inductive arguments. Reasonably so.

Q: I should probably update my agda and make use of the new opaque features to make things readable when refining

Part II

Enumeration

6 Enumeration

Property based testing frameworks often rely on random generation of values, consider for example the Arbitrary class of Quickcheck [CH00]. How these values are best generated depends on the property being tested; if we are testing an implementation of <code>insertSorted</code>, we should probably generate sorted lists [Res19]! Some frameworks like Quickcheck do provide deriving mechanisms for Arbitrary instances, but this relinquishes most control over the distribution. This leaves manually re-implementing Arbitrary as necessary as the only option for a user who wants to test properties with more sophisticated preconditions.

A more controllable alternative to random generation is the complete enumeration of all values. Provided that such an enumeration supports efficient (and fair) indexing, one can adjust a random distribution of values by controlling the sampling from enumerations. There is rich theory of enumeration, and this problem has also been tackled numerous times in the context of functional programming. Some approaches focus on the efficient indexing of enumerations

[DJW12], others focus on generating indexed types as a means of enumerating values with invariants [RS22].

We will describe a framework generalizing these approaches, which will support:

- 1. unique and complete enumeration
- 2. indexing by (exact) recursive depth
- 3. fast skipping through the enumeration
- 4. indexed, nested, and mutually recursive types

We will follow an approach similar to the list-to-list approach [RS22], but rather than expressing enumerations as a step-function, computing the next generation of values from a list of predecessors, we will keep track of the entire depth indexed hierarchies.

6.1 Basic strategy

We define a hierarchy of elements as

```
Hierarchy: Type \rightarrow Type
Hierarchy A = N \rightarrow List A
```

When applied to a number n, a hierarchy should then return the list of elements of exactly depth n. To iteratively approximate hierarchies, we define a hierarchy-builder type

```
Builder : (A B : Type) → Type
Builder A B = Hierarchy A → Hierarchy B
```

Hierarchy-builders should be able to take a partially defined hierarchy, and return a hierarchy which is defined at one higher level.

We implement some basic hierarchy building blocks, such as the one-element builder

```
pure : B → Builder A B
pure x _ zero = [ x ]
pure x _ (suc n) = []
```

which represents nullary constructors, and the shift builder

```
rec B zero = []
rec B (suc n) = B n
```

which represents recursive fields.

To interpret sum types, we use an interleaving operation. Consider that for the disjoint sum, the elements at level n must be formed from elements which are also at level n, regardless whether they are from the left summand or the right.

```
_{(| )_{-}}: Builder A B → Builder A C → Builder A (B \uplus C)
(B<sub>1</sub> (| ) B<sub>2</sub>) V n = interleave (mapL inl (B<sub>1</sub> V n)) (mapL inr (B<sub>2</sub> V n))
```

For product types, the elements at level n are those which contain at least one component at level n, so we have to sum all possible combinations of products

```
pair : Builder A B → Builder A C → Builder A (B × C)
      pair B_1 B_2 V n =
             (downFrom (suc n) >>= \lambda i \rightarrow (prod (B_1 V n) (B_2 V i)))
         ++ (downFrom n >>= \lambda i \rightarrow (prod (B_1 V i) (B_2 V n)))
We claim that this is sufficient to enumerate the following simple universe of
types
      data Desc: Set where
         one : Desc
         var : Desc
         _{-\otimes_{-}}: (D E : Desc) \rightarrow Desc
         _⊕_ : (D E : Desc) → Desc
       [_] : Desc → Set → Set
       [ one ] X = T
       var | X = X
       [\![ D \otimes E ]\!] X = [\![ D ]\!] X \times [\![ E ]\!] X
       \llbracket \ \mathsf{D} \oplus \mathsf{E} \ \rrbracket \ \mathsf{X} = \llbracket \ \mathsf{D} \ \rrbracket \ \mathsf{X} \uplus \llbracket \ \mathsf{E} \ \rrbracket \ \mathsf{X}
      data \mu (D : Desc) : Set where
         con : [D](\mu D) \rightarrow \mu D
In the same vein as other generic constructions, we can define a generic builder
by cases over the interpretentation
      builder : \forall \{D\} D' \rightarrow Builder (\mu D) ([\![ D' \!]\!] (\mu D))
      builder one = pure tt
      builder var = rec
      builder (D ⊗ E) = pair (builder D) (builder E)
      builder (D ⊕ E) = builder D ⟨|⟩ builder E
By applying constructors, we can wrap this up into an endomorphism at a
fixpoint
       gbuilder : ∀ D → Builder (µ D) (µ D)
      gbuilder D V = mapH con (builder D V)
Finally, we observe that applying this builder n+1 times to the empty hierarchy
is sufficient to approximate the hierarchy up to level n
       iterate : \mathbb{N} \rightarrow (A \rightarrow A) \rightarrow A \rightarrow A
       iterate zero f x = x
      iterate (suc n) f x = f (iterate n f x)
      build: Builder A A → Hierarchy A
      build B n = iterate (suc n) B (const []) n
      hierarchy : \forall D \rightarrow Hierarchy (\mu D)
      hierarchy D = build (gbuilder D)
which gives us the generic hierarchy
    We can for example apply this to generate binary trees of given depths
      TreeD : Desc
      TreeD = one ⊕ (var ⊗ var)
```

```
TreeH = hierarchy TreeD
which returns the following trees of level 2
  node (node leaf leaf) (node leaf leaf)
  :: node (node leaf leaf) leaf
     :: node leaf (node leaf leaf)
     :: []
  However, it would be even cooler if
```

- 1. An enumeration could tell us how many elements there are of some depth
- 2. An enumeration was a map from constructor to subsequent enumerations
- 3. The possible indices get computed as we go down.

The first is essential for sampling. The second would give the user total control over the shapes of their generated values. And the third is particularly crucial when the set of possible indices is small.

6.2 Cardinalities

```
Simplifying our earlier approach a bit, we can tinker
      Hierarchy: Type → Type
      Hierarchy A = \mathbb{N} \rightarrow \mathbb{N} \times \text{List } A
to track the sizes. For example, our interleaving operation becomes
      _(|)_: Hierarchy A → Hierarchy B → Hierarchy (A ⊎ B)
      (V_1 \langle | \rangle V_2) n with V_1 n | V_2 n
      ... | c_1, xs | c_2, ys = c_1 + c_2, interleave (mapL inl xs) (mapL inr ys)
We can write down a generic hierarchy
      {-# TERMINATING #-}
      ghierarchy : \forall D \{E\} \rightarrow Hierarchy ([D](\mu E))
      ghierarchy one = pure tt
      ghierarchy var zero = 0 , []
      ghierarchy var (suc n) = mapH con (ghierarchy _) n
      ghierarchy (D ⊗ E) = ghierarchy D ⊕ ghierarchy E
      ghierarchy (D ⊕ E) = ghierarchy D ⟨|⟩ ghierarchy E
      -- note that the termination checker also does not like this case,
      -- so inline it if you want to get rid of the pragma
Then we can count
      numTrees : \mathbb{N} \to \mathbb{N}
      numTrees n = fst (TreeH n)
```

and see that there are 210065930571 trees of level 6, wow! It still takes a bit of time to walk across all branches and products in the description, because there is no memoization at all, but it's a lot better than counting the trees after generating them. Also indexing will be slow, even knowing this information, because we're working with plain lists. Things would probably already get a lot better if we worked with trees that know the sizes of their children.

6.3 Indexed types

Ideally, we get a meaningful list or enumeration of indices at the end: the nonempty ones. However, we do not (yet) require the index type to be enumerable.

The index-first presentation of indexed datatypes, while efficient and succinct, does not seem suitable for this. After all, the descriptions for such a presentation live in the function space from the index to the base descriptions. We would rather want to start "recklessly applying" constructors and seeing what kinds of indices that leaves us with.

This example explains why it's also pretty hopeless for Sijsling's descriptions: We would need a notion of "forward indexed type" in which the indices in the arguments must be strictly less crazy than those in the resulting type.

Anyway, we restrict our attention to indexed types that work, that is, we can decide whether an index fits. In the previous example, the constructor would instead compute whether n is n' + 2, and return n' if it is. This completely breaks any attempt at counting the enumeration.

In comparison, the index-first presentation tells us nothing about which indices are reachable, but probably does better with counting. I suppose you could combine them at the cost of a lot, and first run the forward idea on only the indices, and then see how much each index has, or something.

Part III Temporary

Part IV Related work

7 Related work

7.1 The Structure Identity Principle

If we write a program, and replace an expression by an equal one, then we can prove that the behaviour of the program can not change. Likewise, if we replace one implementation of an interface with another, in such a way that the correspondence respects all operations in the interface, then the implementations should be equal when viewed through the interface. Observations like these are instances of "representation independence", but even in languages with an internal notation of type equality, the applicability is usually exclusive to the metatheory.

In our case, moving from Agda's "usual type theory" to Cubical Agda, a cubical homotopy type theory, *univalence* [VMA19] lets us internalize a kind of rep-

Adapt this to the non-proposal form

resentation independence known as the Structure Identity Principle [Ang+20], and even generalize it from equivalences to quasi-equivalence relations. We will also be able to apply univalence to get a true "equational reasoning" for types when we are looking at numerical representations.

Still, representation independence in non-homotopical settings may be internalized in some cases [Kap23], and remains of interest in the context of generic constructions that conflict with cubical.

7.2 Numerical Representations

Rather than equating implementations after the fact, we can also "compute" datastructures by imposing equations. In the case of container types, one may observe similarities to number systems [Oka98] and call such containers numerical representations. One can then use these representations to prototype new datastructures that automatically inherit properties and equalities from their underlying number systems [HS22].

From another perspective, numerical representations run by using representability as a kind of "strictification" of types, suggesting that we may be able to generalize the approach of numerical representations, using that any (non-indexed) infinitary inductive-recursive type supports a lookup operation [DS16].

7.3 Ornamentation

While we can derive datastructures from number systems by going through their index types [HS22], we may also interpret numerical representations more literally as interpret to rewrite a number system to a container type. We can record this transformation internally using ornaments, which can then be used to derive an indexed version of the container [McB14], or can be modified further to naturally integrate other constraints, e.g., ordering, into the resulting structure [KG16]. Furthermore, we can also use the forgetful functions induced by ornaments to generate specifications for functions defined on the ornamented types [DM14].

7.4 Generic constructions

Being able to define a datatype and reflect its structure in the same language opens doors to many more interesting constructions [EC22]; a lot of "recipes" we recognize, such as defining the eliminators for a given datatype, can be formalized and automated using reflection and macros. We expect that other type transformations can also be interpreted as ornaments, like the extraction of heterogeneous binary trees from level-polymorphic binary trees [Swi20].

References

- [Ang+20] Carlo Angiuli et al. Internalizing Representation Independence with Univalence. 2020. arXiv: 2009.05547 [cs.PL].
- [CH00] Koen Claessen and John Hughes. "QuickCheck: A Lightweight Tool for Random Testing of Haskell Programs". In: SIGPLAN Not. 35.9 (Sept. 2000), pp. 268–279. ISSN: 0362-1340. DOI: 10.1145/357766. 351266. URL: https://doi.org/10.1145/357766.351266.
- [Cha+10] James Chapman et al. "The Gentle Art of Levitation". In: Proceedings of the 15th ACM SIGPLAN International Conference on Functional Programming. ICFP '10. Baltimore, Maryland, USA: Association for Computing Machinery, 2010, pp. 3–14. ISBN: 9781605587943. DOI: 10.1145/1863543.1863547. URL: https://doi.org/10.1145/1863543.1863547.
- [Cla20] Koen Claessen. "Finger Trees Explained Anew, and Slightly Simplified (Functional Pearl)". In: Proceedings of the 13th ACM SIG-PLAN International Symposium on Haskell. Haskell 2020. Virtual Event, USA: Association for Computing Machinery, 2020, pp. 31–38. ISBN: 9781450380508. DOI: 10.1145/3406088.3409026. URL: https://doi.org/10.1145/3406088.3409026.
- [DJW12] Jonas Duregård, Patrik Jansson, and Meng Wang. "Feat: Functional Enumeration of Algebraic Types". In: SIGPLAN Not. 47.12 (Sept. 2012), pp. 61–72. ISSN: 0362-1340. DOI: 10.1145/2430532.2364515. URL: https://doi.org/10.1145/2430532.2364515.
- [DM14] Pierre-Évariste Dagand and Conor McBride. "Transporting functions across ornaments". In: Journal of Functional Programming 24.2-3 (Apr. 2014), pp. 316–383. DOI: 10.1017/s0956796814000069. URL: https://doi.org/10.1017%2Fs0956796814000069.
- [DS16] Larry Diehl and Tim Sheard. "Generic Lookup and Update for Infinitary Inductive-Recursive Types". In: Proceedings of the 1st International Workshop on Type-Driven Development. TyDe 2016. Nara, Japan: Association for Computing Machinery, 2016, pp. 1–12. ISBN: 9781450344357. DOI: 10.1145/2976022.2976031. URL: https://doi.org/10.1145/2976022.2976031.
- [EC22] Lucas Escot and Jesper Cockx. "Practical Generic Programming over a Universe of Native Datatypes". In: Proc. ACM Program. Lang. 6.ICFP (Aug. 2022). DOI: 10.1145/3547644. URL: https://doi. org/10.1145/3547644.
- [eff20] effectfully. Generic. 2020. URL: https://github.com/effectfully/ Generic.
- [HP06] Ralf Hinze and Ross Paterson. "Finger trees: a simple general-purpose data structure". In: *Journal of Functional Programming* 16.2 (2006), pp. 197–217. DOI: 10.1017/S0956796805005769.

- [HS22] Ralf Hinze and Wouter Swierstra. "Calculating Datastructures". In: *Mathematics of Program Construction*. Ed. by Ekaterina Komendantskaya. Cham: Springer International Publishing, 2022, pp. 62–101. ISBN: 978-3-031-16912-0.
- [JG07] Patricia Johann and Neil Ghani. "Initial Algebra Semantics Is Enough!"
 In: Typed Lambda Calculi and Applications. Ed. by Simona Ronchi
 Della Rocca. Berlin, Heidelberg: Springer Berlin Heidelberg, 2007,
 pp. 207–222. ISBN: 978-3-540-73228-0.
- [Kap23] Kevin Kappelmann. Transport via Partial Galois Connections and Equivalences. 2023. arXiv: 2303.05244 [cs.PL].
- [KG16] Hsiang-Shang Ko and Jeremy Gibbons. "Programming with ornaments". In: Journal of Functional Programming 27 (2016), e2. DOI: 10.1017/S0956796816000307.
- [KT95] Haim Kaplan and Robert E. Tarjan. "Persistent Lists with Catenation via Recursive Slow-Down". In: Proceedings of the Twenty-Seventh Annual ACM Symposium on Theory of Computing. STOC '95. Las Vegas, Nevada, USA: Association for Computing Machinery, 1995, pp. 93–102. ISBN: 0897917189. DOI: 10.1145/225058.225090. URL: https://doi.org/10.1145/225058.225090.
- [McB14] Conor McBride. "Ornamental Algebras, Algebraic Ornaments". In: 2014.
- [Oka98] Chris Okasaki. Purely Functional Data Structures. USA: Cambridge University Press, 1998. ISBN: 0521631246.
- [Res19] Cas van der Rest. "Generating Constrained Test Data using Datatype Generic Programming". In: *Master's thesis* (2019).
- [RS22] Cas van der Rest and Wouter Swierstra. "A Completely Unique Account of Enumeration". In: *Proc. ACM Program. Lang.* 6.ICFP (Aug. 2022). DOI: 10.1145/3547636. URL: https://doi.org/10.1145/3547636.
- [Sij16] Yorick Sijsling. "Generic programming with ornaments and dependent types". In: *Master's thesis* (2016).
- [Swi20] WOUTER Swierstra. "Heterogeneous binary random-access lists". In: Journal of Functional Programming 30 (2020), e10. DOI: 10. 1017/S0956796820000064.
- [Tea23] Agda Development Team. Agda. 2023. URL: https://agda.readthedocs.io/en/v2.6.3/.
- [VMA19] Andrea Vezzosi, Anders Mörtberg, and Andreas Abel. "Cubical Agda: A Dependently Typed Programming Language with Univalence and Higher Inductive Types". In: *Proc. ACM Program. Lang.* 3.ICFP (July 2019). DOI: 10.1145/3341691. URL: https://doi.org/10.1145/3341691.

Part V

Appendix

A More equivalences for less effort

Noting that constructing equivalences directly or from isomorphisms as in Subsection 2.3 can quickly become challenging when one of the sides is complicated, we work out a different approach making use of the initial semantics of W-types instead. We claim that the functions in the isomorphism of Subsection 2.3 were partially forced, but this fact was not made use of.

First, we explain that if we assume that one of the two sides of the equivalence is a fixpoint or initial algebra of a polynomial functor (that is, the μ of a <code>Desc'</code>), this simplifies giving an equivalence to showing that the other side is also initial.

We describe how we altered the original ornaments [KG16] to ensure that μ remains initial for its base functor in Cubical Agda, explaining why this fails otherwise, and how defining base functors as datatypes avoids this issue.

In a subsection focusing on the categorical point of view, we show how we can describe initial algebras (and truncate the appropriate parts) in such a way that the construction both applies to general types (rather than only sets), and still produces an equivalence at the end. We explain how this definition, like the usual definition, makes sure that a pair of initial objects always induces a pair of conversion functions, which automatically become inverses. Finally, we explain that we can escape our earlier truncation by appealing to the fact that "being an equivalence" is a proposition.

Next, we describe some theory, using which other types can be shown to be initial for a given algebra. This is compared to the construction in Subsection 2.3, observing that intuitively, initiality follows because the interpretation of the zero constructor is forced by the square defining algebra maps, and the other values are forced by repeatedly applying similar squares. This is clarified as an instance of recursion over a polynomial functor.

To characterize when this recursion is allowed, we define accessibility with respect to polynomial functors as a mutually recursive datatype as follows. This datatype is constructed using the fibers of the algebra map, defining accessibility of elements of these fibers by cases over the description of the algebra. Then we remark that this construction is an atypical instance of well-founded recursion, and define a type as well-founded for an algebra when all its elements are accessible.

We interpret well-foundedness as an upper bound on the size of a type, leading us to claim that injectivity of the algebra map gives a lower bound, which is sufficient to induce the isomorphism. We sketch the proof of the theorem, relating part of this construction to similar concepts in the formalization of well-founded recursion in the Standard Library. In particular, we prove an irrelevance and an unfolding lemma, which lets us show that the map into any other algebra induced by recursion is indeed an algebra map. By showing that

it is also unique, we conclude initiality, and get the isomorphism as a corrolary.

The theorem is applied and demonstrated to the example of binary naturals. We remark that the construction of well-foundedness looks similar to view-patterns. After this, we conclude that this example takes more lines that the direct deriviation in Subsection 2.3, but we argue that most of this code can likely be automated.

Merge

Using Subsection 2.3 we can relate functionally equivalent structures, and using Section 3 we can relate structurally similar structures. However, both have downsides; the former requires us to construct isomorphisms, and the latter wraps all components behind a layer of constructors. In this section will alleviate these problems through generics and by alternative descriptions of equivalences.

In later sections we will construct many more equivalences between more complicated types than before, so we will dive right into the latter. Reflecting upon Subsection 2.3, we see that when one establishes an equivalence, most of the time is spent working out a series of lemmas that prove the conversion functions are to be mutual inverses. We note that the functions themselves were, in fact, forced for a large part.

First, we remark that μ is internalization of the representation of simple ^22 datatypes as W-types. Thus, we will assume that one of the sides of the equivalence is always represented as an initial algebra of a polynomial functor, and hence the μ of a Desc'.

A.1 Well-founded monic algebras are initial

Unfortunately, the machinery developed by Ko and Gibbons [KG16] relies on axiom K for a small but crucial part. To be precise, in a cubical setting, the type μ as given stops being initial for its base functor! In this section, we will be working with a simplified and repaired version. Namely, we simplify <code>Desc'</code> to

```
\begin{array}{c} \text{data Desc}' : \text{Set}_1 \text{ where} \\ \text{$\nu$ : $(n:N) \to \text{Desc}'$} \\ \text{$\sigma$ : $(S:Set)$ $(D:S \to \text{Desc}') \to \text{Desc}'$} \end{array} To complete the definition of $\mu$ \begin{array}{c} \text{data $\mu$ $(D:Desc')$ : $Set}_1$ \text{ where} \\ \text{con: Base $(\mu$ D)$ $D \to \mu$ D} \end{array}
```

we will need to implement Base. We remark that in the original setup, the recursion of mapFold is a structural descent in [D'] (μ D). Because [] is a type computing function and not a datatype, this descent becomes invalid²³, and mapFold fails the termination check. We resolve this by defining Base as a datatype

```
data Base (X : Set₁) : Desc' → Set₁ where
  in-y : ∀ {n} → Vec X n → Base X (y n)
```

 $^{^{22}{\}rm Of}$ course, indexed data types are indexed W-types, mutually recursive data types are represented yet differently...

 $^{^{23}}$ Refer to the without K page.

```
in-\sigma: \forall \{S D\} \rightarrow \Sigma[s \in S] (Base X (D s)) \rightarrow Base X (\sigma S D)
```

such that this descent is allowed by the termination checker without axiom K.²⁴

Recall that the Base functors of descriptions are special polynomial functors, and the fixpoint of a base functor is its initial algebra. We are looking for sufficient conditions on X to get the equivalence $e:X\cong \mu F$. Note that when $X\cong \mu F$, then there necessarily is an initial algebra $FX\to X$. Conversely, if the algebra (X,f) is isomorphic to $(\mu F,\operatorname{con})$, then $X\cong \mu F$ would follow immediately, so it is equivalent to ask for the algebras to be isomorphic instead.

A.1.1 Datatypes as initial algebras

To characterize when such algebras are isomorphic, we reiterate some basic category theory, simultaneously rephrasing it in Agda terms.²⁵

Let C be a category, and let a,b,c be objects of C, so that in particular we have identity arrows $1_a:a\to a$ and for arrows $g:b\to c,f:a\to b$ composite arrows $gf:a\to c$ subject to associativity. In our case, C is the category of types, with ordinary functions as arrows.

Recall that an endofunctor, which is simply a functor F from C to itself, assigns objects to objects and sends arrows to arrows

```
F<sub>0</sub>: Type \ell \to \text{Type } \ell
fmap: (A \to B) \to F_0 A \to F_0 B
These assignments are subject to the identity and composition laws
f-id: (x : FA)
\to \text{mapF} \text{ id } x \equiv x

f-comp: (g : B \to C) (f : A \to B) (x : FA)
\to \text{mapF} (g \circ f) x \equiv \text{mapF} g \text{ (mapF } f x)
An F-algebra is just a pair of an object a and an arrow Fa \to a
record Algebra (F : \text{Type } \ell \to \text{Type } \ell): Type (\ell\text{-suc }\ell) where field

Carrier: Type \ell
forget: F Carrier \to \text{Carrier}
```

Algebras themselves again form a category C^F . The arrows of C^F are the arrows f of C such that the following square commutes

$$\begin{array}{ccc}
Fa & \xrightarrow{Ff} Fb \\
U_a \downarrow & & \downarrow U_b \\
a & \xrightarrow{f} & b
\end{array}$$

So we define

²⁴This has, again by the absence of axiom K, the consequence of pushing the universe levels up by one. However, this is not too troublesome, as equivalences can go between two levels, and indeed types are equivalent to their lifts.

²⁵We are not reusing a pre-existing category theory library for the simple reasons that it is not that much work to write out the machinery explicitly, and that such libraries tend to phrase initial objects in the correct way, which is too restrictive for us.

Note that we take the propositional truncation of the square, such that algebra maps with the same underlying morphism become propositionally equal

```
Alg\rightarrow-Path : {F : RawFunctor \ell} {A B : Algebra (F .F<sub>0</sub>)}

\rightarrow (g f : Alg\rightarrow F A B) \rightarrow g .mor \equiv f .mor \rightarrow g \equiv f
```

The identity and composition in \mathbb{C}^F arise directly from those of the underlying arrows in \mathbb{C} .

Recall that an object \emptyset is initial when for each other object a, there is a unique arrow $!: \emptyset \to a$. By reversing the proofs of initiality of μ and the main result of this section, we obtain a slight variation upon the usual definition. Namely, unicity is often expressed as contractability of a type

```
is
Contr A = \Sigma[ x \in A ] (\forall y \to x \equiv y)
Instead, we again use a truncation
weak
Contr A = \Sigma[ x \in A ] (\forall y \to || x \equiv y ||_1)
```

but note that this also, crucially, slightly stronger than connectedness. We define initiality for arbitrary relations

```
record Initial (C : Type \ell) (R : C \rightarrow C \rightarrow Type \ell') (Z : C) : Type (\ell-max (\ell-suc \ell) \ell') where field initiality : \forall X \rightarrow weakContr (R Z X)
```

such that it closely resembles the definition of least element. Then, A is an initial algebra when

```
InitAlg RawF A = Initial (Algebra (RawF .F₀)) (Alg→ RawF) A
```

By basic category theory (using the usual definition of initial objects), two initial objects a and b are always isomorphic; namely, initiality guarantees that there are arrows $f:a\to b$ and $g:b\to a$, which by initiality must compose to the identities again.

Similarly, we get that

Because being an equivalence is a property, we can eliminate from the truncations to get the wanted result.

A.1.2 Accessibility

As a consequence, we get that X is isomorphic to μD when X is an initial algebra for the base functor of D; μD is initial by its fold, and by induction on

 μD using the squares of algebra maps.

Remark A.1. The fixpoint μD is not in general a strict initial object in the category of algebras. For a strict initial object, having a map $a \to \emptyset$ implies $a \cong \emptyset$. This is not the case here: strict initial objects satisfy $a \times \emptyset \cong \emptyset$, but for the $X \mapsto 1 + X$ -algebras $\mathbb N$ and $2^{\mathbb N}$ clearly $2^{\mathbb N} \times \mathbb N \cong \mathbb N$ does not hold. On the other hand, the "obvious" sufficient condition to let C^F have strict initial objects is that F is a left adjoint, but then the carrier of the initial algebra is simply \bot .

Looking back at Subsection 2.3, we see that Leibniz is an initial $F: X \mapsto 1+X$ algebra because for any other algebra, the image of 0b is fixed, and by bsuc all other values are determined by chasing around the square. Thus, we are looking for a similar structure on $f: FX \to X$ that supports recursion.

We will need something stronger than $FX \cong X$, as in general a functor can have many fixpoints. For this, we define what it means for an element x to be accessible by f. This definition uses a mutually recursive datatype as follows: We state that an element x of X is accessible when there is an accessible y in its fiber over f

```
data Acc D f x where
  acc : (y : fiber f x) → Acc' D f D (fst y) → Acc D f x
```

Accessibility of an element x of Base A E is defined by cases on E; if E is y n and x is a Vec A n, then x is accessible if all its elements are; if x is σ S E', then x is accessible if snd x is

```
data Acc' D f where

acc-\gamma: All (Acc D f) x \to Acc' D f (\gamma n) (in-\gamma x)

acc-\sigma: Acc' D f (E s) x \to Acc' D f (\sigma S E) (in-\sigma (s , x))
```

Consequently, X is well-founded for an algebra when all its elements are accessible

```
Wf D f = \forall x \rightarrow Acc D f x
```

We can see well-foundedness as an upper bound on the size of X, if it were larger than μD , some of its elements would get out of reach of an algebra. Now having $FX\cong X$ also gives us a lower bound, but note that having a well-founded injection $f:FX\to X$ is already sufficient, as accessibility gives a section of f, making it an iso. This leads us to claim

Claim A.1. If there is a mono $f: FX \to X$ and X is well-founded for f, then X is an initial F-algebra.

Proof sketch of Claim A.1. Suppose X is well-founded for the mono $f: FX \to X$. To show that (X, f) is initial, let us take another algebra (Y, g), and show that there is a unique arrow $(X, f) \to (Y, g)$.

This section is about as digestable as a brick.

By $\mathsf{Acc}\text{-}\mathsf{recursion}$ and because all x are accessible, we can define a plain map into Y

```
Wf-rec : (D : Desc') (X : Algebra (\dot{F} D)) \rightarrow Wf D (X .forget) \rightarrow (\dot{F} D A \rightarrow A) \rightarrow X .Carrier \rightarrow A
```

This construction is an instance of the concept of "well-founded recursion"²⁶, so we use a similar strategy. In particular, we prove an irrelevance lemma

```
Wf-rec-irrelevant: \forall x' y' x a b \rightarrow rec x' x a \equiv rec y' x b which implies the unfolding lemma unfold-Wf-rec: \forall x' \rightarrow rec (cx x') (cx x') (wf (cx x')) \equiv f (Base-map (\lambda y \rightarrow rec y y (wf y)) x')
```

The unfolding lemma ensures that the map we defined by Wf-rec is a map of algebras. The proof that this map is unique proceeds analogously to that in the proof that μD is initial, but here we instead use Acc-recursion

```
Wf+inj→Init : (D : Desc') (X : Algebra (F D)) → Wf D (X .forget)

→ injective (X .forget) → InitAlg (RawF D) X
```

Thus, we conclude that X is initial. The main result is then a corollary of initiality of X and the isomorphism of initial objects

```
Wf+inj\equiv \mu: (D: Desc') (X: Algebra (\dot{F} D)) \rightarrow Wf D (X.forget) \rightarrow injective (X.forget) \rightarrow X.Carrier \equiv \mu D
```

A.1.3 Example

Let us redo the proof in Subsection 2.3, now using this result. Recall the description of naturals NatD. To show that Leibniz is isomorphic to Nat, we will need a NatD-algebra and a proof of its well-foundedness. We define the algebra

```
bsuc' : Base Leibniz<sub>1</sub> NatD → Leibniz<sub>1</sub>
bsuc' zero = 0b<sub>1</sub>
bsuc' (succ n) = bsuc<sub>1</sub> n

L-Alg : Algebra (F NatD)
L-Alg .Carrier = Leibniz<sub>1</sub>
L-Alg .forget = bsuc'
```

For well-foundedness, we use something similar to view-patterns (the main difference being that we look through the entire structure, instead of a single layer)

```
data Peano-View : Leibniz₁ → Type₁ where
   as-zero : Peano-View 0b₁
   as-suc : (n : Leibniz₁) (v : Peano-View n) → Peano-View (bsuc₁ n)

view-1b : ∀ {n} → Peano-View n → Peano-View (n 1b₁)
view-2b : ∀ {n} → Peano-View n → Peano-View (n 2b₁)
view : (n : Leibniz₁) → Peano-View n
```

where the mutually recursive proof of view is "almost trivial". Well-foundedness follows immediately

```
view→Acc : ∀ {n} → Peano-View n → Acc NatD bsuc' n
Wf-bsuc : Wf NatD bsuc'
Wf-bsuc n = view→Acc (view n)
```

²⁶This is formalized in the standard-library with many other examples.

Injectivity of bsuc_1 happens to be harder to prove from retractions than directly, so we prove it directly, from which the wanted statement follows

```
L≃\muN : Leibniz<sub>1</sub> ≃ \mu NatD
L≃\muN = Wf+inj≃\mu NatD L-Alg Wf-bsuc \lambda x y p \rightarrow inj-bsuc x y p
```

In this case, we needed more lines of code to prove the same statement, however, the process of writing became more forced, and might be more amenable to automation.