

Diffing Mutually Recursive Types

A code tour

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1 Our Universe

The universe we are using is a *Sums-of-Products* over type variables and constant types.

```
data Atom (n : ℕ) : Set where
  I : Fin n      → Atom n
  K : Fin ks#    → Atom n
```

Constructor `I` refers to the n -th type variable whereas `K` refers to a constant type. Value `ks#` is passed as a module parameter. We denote products by π and sums by σ , but they are just lists.

```
 $\pi$  : ℕ → Set
 $\pi$  = List ∘ Atom
```

```
 $\sigma\pi$  : ℕ → Set
 $\sigma\pi$  = List ∘  $\pi$ 
```

Interpreting these codes is very simple. Here, `Parms` is a valuation for the type variables.

```
Parms : ℕ → Set1
Parm s n = Fin n → Set
```

```
[_]a : {n : ℕ} → Atom n → Parm s n → Set
[I x]a    A = A x
[K x]a    A = lookup x ks
```

```
[_]p : {n : ℕ} →  $\pi$  n → Parm s n → Set
[[]]p    A = Unit
[a :: as]p A = [a]a A × [as]p A
```

```
[_] : {n : ℕ} →  $\sigma\pi$  n → Parm s n → Set
[[]]    A =  $\perp$ 
[p :: ps] A = [p]p A  $\uplus$  [ps] A
```

Note that here, `Parms n` really is isomorphic to n types that serve as the parameters to the functor `[F]`. When we introduce a fixpoint combinator, these parameters are used to tie the recursion knot, just like a simple fixpoint: $\mu F \equiv F (\mu F)$. In fact, a mutually recursive family can be easily encoded in this setting. All we need is n types that refer to n type-variables each!

```
Fam : ℕ → Set
Fam n = Vec ( $\sigma\pi$  n) n
```

```
data Fix {n : ℕ} (F : Fam n) : Fin n → Set where
  (<_>) : ∀ {k} → [lookup k F] (Fix F) → Fix F k
```

This universe is enough to model Context-Free grammars, and hence, provides the basic bare bones for diffing elements of an arbitrary programming language. In the

future, it could be interesting to see what kind of diffing functionality indexed functors could provide, as these could have scoping rules and other advanced features built into them.

1.1 SoP peculiarities

One slightly cumbersome problem we have to circumvent is that the codes for type variables and constant types are modeled by `Atom`; whereas the codes for types are modelled by `σπ`. This requires more discipline to organize our code, since we have to separate, on the type level, functions that handle one or the other. Nevertheless, we may wish to see `Atom`s as a trivial *Sum-of-Product*.

```
α : {n : ℕ} → Atom n → σπ n
α a = (a :: []) :: []
```

Instead of having binary injections into coproducts, like we would on a *regular-like* universe, we have *n*-ary injections, or, *constructors*. We encapsulate the idea of constructors of a `σπ` into a type and write a *view* type that allows us to look at an inhabitant of a sum of products as a *constructor* and *data*.

First, we define constructors:

```
cons# : {n : ℕ} → σπ n → ℕ
cons# = length

Constr : {n : ℕ} (ty : σπ n) → Set
Constr ty = Fin (cons# ty)
```

Now, a constructor of type *C* expects some arguments to be able to make an element of type *C*. This is a product, we call it the `typeOf` the constructor.

```
typeOf : {n : ℕ} (ty : σπ n) → Constr ty → π n
typeOf [] ()
typeOf (x :: ty) fz = x
typeOf (x :: ty) (fs c) = typeOf ty c
```

Injecting is fairly simple.

```
inject : {n : ℕ} {A : Parm n} {ty : σπ n}
→ (i : Constr ty) → [ typeOf ty i ]p A
→ [ ty ] A
inject {ty = []} () cs
inject {ty = x :: ty} fz cs = i1 cs
inject {ty = x :: ty} (fs i) cs = i2 (inject i cs)
```

We finish off with a *view* of `[ty]A` as a constructor and some data. This greatly simplify the algorithms later on.

```
data SOP {n : ℕ} {A : Parm n} {ty : σπ n} : [ ty ] A → Set where
strip : (i : Constr ty) (ls : [ typeOf ty i ]p A)
→ SOP (inject i ls)
```

1.2 Agda Details

Here we clarify some Agda specific details that are agnostic to the big picture. This can be safely skipped on a first iteration.

As we mentioned above, our codes represent functors on *n* variables. Obviously, to program with them, we need to apply these to something. The denotation receives a function `Fin n → Set`, denoted `Parm n`, which can be seen as a valuation for each type variable.

In the following sections, we will be dealing with values of `[ty]A` for some class of valuations *A*, though. We need to have decidable equality for *A* *k. well-behaved parameter*:

```

record WBParms {n : ℕ}(A : Params n) : Set where
  constructor wb-parms
  field
    parm-size : ∀{k} → A k → ℕ
    parm-cmp   : ∀{k}(x y : A k) → Dec (x ≡ y)

```

The following sections discuss functionality that does not depend on *parameters to codes*. Hence, we will be passing them as Agda module parameters. We also set up a number of synonyms to already fix the aforementioned parameter. The first diffing technique we discuss is the trivial diff. It's module is declared as follows:

```

module RegDiff.Diff.Trivial.Base
  {ks# : ℕ}(ks : Vec Set ks#)(keqs : Vec1 Eq ks)
  {parms# : ℕ}(A : Params parms#)(WBA : WBParms A)
  where

```

We stick to this nomenclature throughout the code. The first line handles constant types: $ks\#$ is how many constant types we have, ks is the vector of such types and $keqs$ is an indexed vector with a proof of decidable equality over such types. The second line handles type parameters: $parms\#$ is how many type-variables our codes will have, A is the valuation we are using. The last parameter is a family of decidable equality for A .

Below are the synonyms we use for the rest of the code:

```

U : Set
U = σπ parms#

Atom : Set
Atom = Atom' parms#

Π : Set
Π = π parms#

sized : {p : Fin parms#} → A p → ℕ
sized = parm-size WBA

_≡?-A_ : {p : Fin parms#}(x y : A p) → Dec (x ≡ y)
_≡?-A_ = parm-cmp WBA

```

```

[[_]]a : Atom → Set
[[ a ]]a = interpa a A

[[_]]p : Π → Set
[[ p ]]p = interpp p A

[[_]] : U → Set
[[ u ]] = interps u A

UUSet : Set1
UUSet = U → U → Set

AASet : Set1
AASet = Atom → Atom → Set

ΠΠSet : Set1
ΠΠSet = Π → Π → Set

contr : ∀{a b}{A : Set a}{B : Set b}
  → (A → A → B) → A → B
contr p x = p x x

UU→AA : UUSet → AASet
UU→AA P a a' = P (α a) (α a')

→α : {a : Atom} → [[ a ]]a → [[ α a ]]
→α k = il (k , unit)

```

2 Computing and Representing Patches

Intuitively, a *Patch* is some description of a transformation. Setting the stage, let A and B be types, $x : A$ and $y : B$ elements of such types. A *patch* between x and y must specify a few parts:

- i) An `apply` : $A \rightarrow \text{Maybe } B$ function,
- ii) such that `apply` $x \equiv \text{just } y$.

Well, `apply` is clearly a partial function, and hence, can be seen as an arrow in the category **Par** of sets and partial functions (that is, the Kleisli category of `Maybe`!). We will denote the application function of p by p^\flat .

It is intuitive that if we can apply a patch p to an element a , resulting in `just` a' , there should exist a patch \bar{p} that *undo* those changes! That is, when \bar{p} is applied to a' it must result in `just` a . We can not ask for pointwise invertibility of p^\flat . This is the same as asking p^\flat to be total¹.

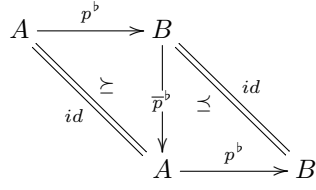
We can, however, get away if we require that $\bar{p}^\flat \bullet p^\flat$ be *less than* the identity function, in the sense that if a belongs in the domain of p^\flat , then $\bar{p}^\flat (p^\flat a) \equiv a$, where \bullet denotes Kleisli composition.

Take the canonical partial order on `Maybe` A , which puts `nothing` as the lower bound:

$$\begin{aligned} \text{nothing} &\preceq y \\ \text{just } x &\preceq \text{just } y \text{ iff } x \equiv y \end{aligned}$$

Then require the following diagrams, in **Par**, to commute up to \preceq :

¹To prove this, look at p^\flat as a functional relation: $\text{img } R \subseteq \text{id}$. Then require that \bar{p}^\flat be of the form R° . If R° also needs to be function we have that R needs to be entire. If R is entire and functional, it is a function.



We lift \preceq pointwise to functions, that is define $f \preceq g$ by $\forall x; f x \preceq g x$, then the diagrams above require $\bar{p}^b \bullet p^b \preceq id$ iff $p^b \bullet \bar{p}^b \preceq id$. Which is precisely what an Antitone Galois Connection is!

Hence, we want p^b and \bar{p}^b to form a (antitone) Galois Connection with respect to \preceq .

Needs discussion:

[Inspired by Tabareau's "Aspect-oriented programming: a language for 2-categories", I'm toying with (but not proposing for adoption, this is still very sketchy!) the following idea: we define a 2-category where 0-cells are types, 1-cells are antitone Galois connections (pairs of partial functions) (ie. diff and inverse diff) between types and 2-cells are residuals. There is a terminal object "unit": it is a 0-cell 1 such that for all type A, there exists a (trivial) 1-cell $skip : A \rightarrow 1$ and a unique 2-cell $1_{skip} : skip \Rightarrow skip$. The specification of "patch" becomes: for every 1-cell $x : 1 \rightarrow A$ and $y : 1 \rightarrow B$ there exists a 1-cell $p : A \rightarrow B$ and a 2-cell $apply : x \Rightarrow y$. Take this with a pinch of salt: it is more "wishful thinking" than "sound categorical reasoning".]

[I like the sketch, let me know where this goes! By the way, not sure we need to carry around the pair of partial functions. The inverse diff is uniquely determined by the diff!]

[re "inverse diff uniquely determined by the diff": I don't think that this is true. You could for example provide an inverse diff that fails all the time: it would satisfy our spec above. However, given a diff code, you can always compute *one* specific inverse. (Which, btw, should have some sort of universal property that we ought to characterize!)]

Now, let us discuss some code and build some intuition for what is what in the above schema. We will present different parts of the code, how do they relate to this relational view and give examples here and there!

2.1 Trivial Diff

The simplest possible way to describe a transformation is to say what is the source and what is the destination of such transformation. This can be accomplished by the Diagonal functor, Δ , just fine.

Now, take an element $(x, y) : \Delta ty tv$. The application is intuitive! We are expecting to transform x into y . Before formalizing this, we need two simple partial functions. Let $f|_S$ be the domain restriction of f to S :

$$(f|_S) a = \begin{cases} f a & , \text{if } a \in S \\ \perp & , \text{otherwise} \end{cases}$$

We write $f|_x$ instead of $f|_{\{x\}}$ whenever the type of x is clear. Let \underline{y} denote the *constant* y (total) function lifted to **Par**. We then define:

$$(x, y)^b = \underline{y}|_x$$

Needs discussion:

In the code, we actually define the application of Δ as

$$\begin{aligned}(x, x)^b &= id \\ (x, y)^b &= \underline{y}|_x\end{aligned}$$

This is because we want the patch between x and y to be the patch p that maximizes the domain of p^b maintaining $p^b x \equiv \text{just } y$.

In fact, let p_0, \dots, p_k be all the possible patches between x and y . We want to systematically choose the p_i such that: $p_j^b \preceq p_i^b$ for $j \leq k$.

2.1.1 Trivial Diff, in Agda

We will be using Δ *ty* *tv* for the three levels of our universe: atoms, products and sums. We distinguish between the different Δ 's with subscripts $_a$, $_p$ and $_s$ respectively. They only differ in type. The treatment they receive in the code is exactly the same! Below is how they are defined:

```
delta : ∀{a}{A : Set a}(P : A → Set)
  → A → A → Set
delta P a1 a2 = P a1 × P a2
```

Hence, we define $\Delta_x = \text{delta } \llbracket \cdot \rrbracket_x$, for $x \in \{a, p, s\}$.

2.2 Spines

We can make the trivial diff better by identifying whether or not x and y agree on something! In fact, we will aggressively look for copying opportunities. We start by checking if x and y are, in fact, equal. If they are, we say that the patch that transforms x into y is *copy*. If they are not equal, they might have the same *constructor*. If they do, we say that the constructor is copied and we put the data side by side (zip). Otherwise, there is nothing we can do on this phase and we just return $\Delta x y$.

Note that the *spine* forces x and y to be of the same type! In practice, we are only interested in diffing elements of the same language. It does not make sense to diff a C source file against a Haskell source file.

Nevertheless, we define an **S** structure to capture this longest common prefix of x and y ; which, for the *SoP* universe is very easy to state.

```
data S (P : UUSet) : U → Set where
  SX  : {ty : U} → P ty ty → S P ty
  Scp : {ty : U} → S P ty
  Scns : {ty : U} (i : Constr ty)
    → List! (contr P ∘ α) (typeOf ty i)
    → S P ty
```

Remember that $\text{contr } P x = P x x$ and $\alpha : \text{Atom } n \rightarrow \sigma\pi n$; Here, $\text{List! } P l$ is an indexed list where the elements have type $P l_i$, for every $l_i \in l$. We will treat this type like an ordinary list for the remainder of this document.

Note that **S** makes a functor (actually, a free monad!) on P , and hence, we can map over it:

```
S-map : {ty : U}
  {P Q : UUSet} (X : ∀{k v} → P k v → Q k v)
  → S P ty → S Q ty
S-map f (SX x)    = SX (f x)
S-map f Scp        = Scp
S-map f (Scns i xs) = Scns i (mapi f xs)
```

Computing a spine is easy, first we check whether or not x and y are equal. If they are, we are done. If not, we look at x and y as true sums of products and check if their constructors are equal, if they are, we zip the data together. If they are not, we zip x and y together and give up.

```

spine-cns : {ty : U} (x y :  $\llbracket ty \rrbracket$ ) → S  $\Delta_s$  ty
spine-cns x y with sop x | sop y
spine-cns _ _ | strip cx dx | strip cy dy
  with cx  $\stackrel{?}{=}$  Fin cy
... | no _ = SX (inject cx dx , inject cy dy)
spine-cns _ _ | strip _ dx | strip cy dy
  | yes refl = Scns cy (zipp dx dy)

spine : {ty : U} (x y :  $\llbracket ty \rrbracket$ ) → S  $\Delta_s$  ty
spine {ty} x y
  with dec-eq  $\stackrel{?}{=}$  A- ty x y
... | yes _ = Scp
... | no _ = spine-cns x y

```

Note that when both arguments to `spine-cns` have the same constructor (they already have the same type) we can safely zip their arguments! This speeds up the process, as this is *not* an alignment problem. Besides some more complicated types, the zip function is as usual:

```

zipp : {ty :  $\Pi$ }
  →  $\llbracket ty \rrbracket_p$  →  $\llbracket ty \rrbracket_p$  → List! (λ k →  $\Delta_s$  (α k) (α k)) ty
zipp {[]} _ _ = []
zipp {_ :: ty} (x , xs) (y , ys)
  = (il (x , unit) , il (y , unit)) :: zipp xs ys

```

The application functions specified by a spine $s = \text{spine } x \ y$, denoted s^b are defined, in **Par**, by:

$$\begin{aligned}
\text{Scp}^b &= A \xleftarrow{id} A \\
(\text{SX } p)^b &= A \xleftarrow{p^b} A \\
(\text{Scns } i \ [s1 , \dots , sN])^b &= \Pi_k \Pi_j A_{kj} \xleftarrow{\text{inj}_i \bullet (s_1^b \times \dots \times s_n^b) \bullet \text{match}_i} \Pi_k \Pi_j A_{kj}
\end{aligned}$$

where inj_i is the injection, with constructor i , into $\Pi_k T_k$. It corresponds to the function `injection`; whereas `matchi` is the inverse: pattern matching on constructor i .

Appendix A clarifies some aspects about products and coproducts on **Par**. Long story short, we have that $\pi_1 \bullet \langle f, g \rangle \preceq f$ and $\pi_2 \bullet \langle f, g \rangle \preceq g$! The proof is not very complicated once we know that `Maybe` is a commutative monad. Semantically speaking, $(s_1^b \times \dots \times s_n^b)$ is only defined on an input (x_1, \cdot, x_n) iff every s_i^b is defined on x_i , which meets one intuition.

Note that, in the $(\text{SX } p)$ case, we simply ask for the application function of p . The algorithm produces a $\text{S } \Delta_s$, so we have pairs on the leaves of the spine. In fact, either we have only one leave or we have *arity* C_i leaves, where C_i is the common constructor of x and y in `spine` $x \ y$.

For a running example, let's consider a datatype defined by:

```

2-3-TREE-F : σπ 1
2-3-TREE-F = []
  ⊕ (K kN) ⊗ I ⊗ I ⊗ []
  ⊕ (K kN) ⊗ I ⊗ I ⊗ I ⊗ []
  ⊕ []

```

We omit the `fz` for the `I` parts, as we only have one type variable. We also use

$_ \oplus _$ and $_ \otimes _$ as aliases for $_ :: _$ with different precedences. As expected, there are three constructors:

```
2-node' 3-node' nil' : Constr 2-3-TREE-F
nil'      = fz
2-node'   = fs fz
3-node'   = fs (fs fz)
```

We can then consider a few spines over $\llbracket 2-3-TREE-F \rrbracket_{\text{Unit}}$ to illustrate the algorithm:

```
spine nil' (3-node' 10 unit unit unit) = SX (nil' , 3-node' 10 unit unit unit)
spine (2-node' 10 unit unit) (2-node' 15 unit unit) = Scns 2-node' [ (10 , 15) , (unit , unit) , (unit , unit) ]
spine nil' nil' = Scp
```

In the case where the spine is **Scp** or **Scns** i there is nothing left to be done and we have the best possible diff. Note that on the **Scns** i case we do *not* allow for rearranging of the parameters of the constructor i .

In the case where the spine is **SX**, we can do a better job! We can record which constructor changed into which and try to reconcile the data from both the best we can. Going one step at a time, let's first change one constructor into the other.

It is important to note that if the output of **spine** is a **SX**, then the constructors are *different*.

2.3 Constructor Changes

Let's take an example where the spine can not copy anything:

```
s = spine (2-node' 10 unit unit) (3-node' 10 unit unit unit)
    = SX (2-node' 10 unit unit , 3-node' 10 unit unit unit)
```

Here, we wish to say that we changed a **2-node'** into a **3-node'**. But we are then left with a problem about what to do with the data inside the **2-node'** and **3-node'**; this is where the notion of alignment will be in the picture. For now, we abstract it away by the means of a parameter, just like we did with the **S**. This time, however, we need something that receives products as inputs.

```
data C (P : IIISet) : U → U → Set where
  CX : {ty tv : U}
      → (i : Constr ty)(j : Constr tv)
      → P (typeOf ty i) (typeOf tv j)
      → C P ty tv
```

Note that **C** also makes up a functor, and hence can be mapped over:

```
C-map : {ty tv : U}
        {P Q : IIISet} (X : ∀ {k v} → P k v → Q k v)
        → C P ty tv → C Q ty tv
C-map f (CX i j x) = CX i j (f x)
```

Computing an inhabitant of **C** is trivial:

```
change : {ty tv : U} → [ ty ] → [ tv ] → C Δp ty tv
change x y with sop x | sop y
change _ _ | strip cx dx | strip cy dy = CX cx cy (dx , dy)
```

Now that we can compute change of constructors, we can refine our s above. We can compute **S-map change** s and we will have:

```
c = S-map change s
    = SX (CX 2-node' 3-node' ((10 , unit , unit) , (10 , unit , unit , unit)))
```


The application induced by \mathbf{C} is trivial. We just need to pattern match, change the data of the constructor in whatever way we need, then inject into another type.

$$\begin{array}{ccc}
 V & \xleftarrow{(\mathbf{CX} \ i \ j \ p)^b} & T \\
 \uparrow \text{inj}_j & & \downarrow \text{match}_i \\
 \text{typeOf } V \ j & \xleftarrow{p^b} & \text{typeOf } T \ i
 \end{array}$$

Note that up until now, everything was deterministic! This is something we are bound to lose when talking about alignment.

2.4 Aligning Everything

On the literature for version control system, the *alignment* problem is the problem of mapping two strings l_1 and l_2 in \mathcal{L} into $\mathcal{L} \cup \{-\}$, for $\{-\} \not\subseteq \mathcal{L}$ such that the resulting strings l'_1 and l'_2 are of the same length such that for all i , it must not be happen that $l'_1[i] = - = l'_2[i]$. For example, Take strings $l_1 = \text{"CGTCG"}$ and $l_2 = \text{"GATAGT"}$, then, the following is an (optimal) alignment:

$$\begin{array}{ccccccc}
 \mathbf{C} & \mathbf{G} & - & \mathbf{T} & \mathbf{C} & \mathbf{G} & - \\
 - & \mathbf{G} & \mathbf{A} & \mathbf{T} & \mathbf{A} & \mathbf{G} & \mathbf{T}
 \end{array}$$

Let $\mathcal{DNA} = \{A, T, C, G\}$. Finding the table above is the same as finding a partial map:

$$f : \mathcal{DNA}^5 \rightarrow \mathcal{DNA}^6$$

such that $f(C, G, T, C, G) = (G, A, T, A, G, T)$. There are many ways of defining such a map. We would like, however, that our definition have a maximal domain, that is, we impose the least possible amount of restrictions. In this case, we can actually define f with some pattern matching as:

$$\begin{array}{ll}
 f \ (C, x, y, C, z) & = (x, A, y, A, z, T) \\
 f \ - & = \text{undefined}
 \end{array}$$

And it is easy to verify that, in fact, $f(C, G, T, C, G) = (G, A, T, A, G, T)$. Moreover, this is the *maximal* such f that still (provably) assigns the correct destination to the correct source.

On our running example, the leaf of c has type Δ_p (*typeOf 2-node'*) (*typeOf 3-node'*), and it's value is $((10, \text{unit}, \text{unit}), (10, \text{unit}, \text{unit}, \text{unit}))$. Note that we are now dealing with products of different arity. This step will let us say how to *align* one with the other!

On our example, as long as we align the 10 with the 10, the rest does not matter. One optimal alignment could be:

$$\begin{array}{cccc}
 10 & - & \text{unit} & \text{unit} \\
 10 & \text{unit} & \text{unit} & \text{unit}
 \end{array}$$

2.4.1 Back to Agda

We will look at alignments from the “finding a map between products” perspective. Here is where our design space starts to grow, and so, we should start making some distinctions:

- We want to allow sharing. This means that there can be more than one variable in the defining pattern of our f .
- We do *not* allow permutations, as the search space would be too big. This means that the variables appear in the right-hand side of f in the same order as they appear in the left-hand-side.
- We do *not* allow contractions nor weakenings. That is, every variable on the left-hand-side of f must appear *exactly* once on the right-hand-side.

Needs discussion:

As Pierre points out:

The positive type/negative type distinction I keep referring to and which drives my intuition is used to structure proof search in linear logic. There, the product would be called an *asynchronous* connective while the sum would be called a *synchronous* connective. Quoting “Focusing and Polarization in Intuitionistic Logic”, “the search for a focused proof can capitalize on this classification by applying [...] all invertible rules [related to an asynchronous connective] in any order (without the need for backtracking) and by applying a chain of non-invertible rules [related to a synchronous connective] that focus on a given formula and its positive subformulas.”. So my gut tells me that your diff computation is structured in two (repeating) phases: one that generates the spine & change, yielding several **independent** alignment problems which could be solved concurrently.

Needs discussion:

Which is, in fact true! The spine & change is deterministic and the alignment problems we have to solve are independent. I believe we could exploit some parallelism. It will be far from trivial though.

The following datatype describe such maps:

```
data AI (P : AASet) : II → II → Set where
  A0   : AI P [] []
  Ap1  : ∀{a ty tv} → [ a ]_a → AI P ty tv → AI P (a :: ty) tv
  Ap1° : ∀{a ty tv} → [ a ]_a → AI P ty tv → AI P ty (a :: tv)
  AX   : ∀{a a' ty tv} → P a a' → AI P ty tv → AI P (a :: ty) (a' :: tv)
```

Note that the indexes of `AI`, although represented as lists are, in fact, products. Well, turns out that lists and products are not so different after all. Let us represent the f we devised on the \mathcal{DNA} example using `AI`. Recall $f(C, x, y, C, z) = (x, A, y, A, z, T)$.

$$f \equiv \text{Ap1 } C \text{ (AX Scp (Ap1° A (AX Scp (AX (C, A) (AX Scp (Ap1° T A0))))))}$$

If we rename `Ap1` to `del`; `Ap1°` to `ins` and `AX` to `mod` we see some familiar structure arising! Aligning products is the same as computing the diff between heterogeneous lists! In fact, the `align` function is defined as:

```
align* : {ty tv : II} → [ ty ]_p → [ tv ]_p → List (AI Δ_a ty tv)
align* [] [] m n = return A0
align* [] {v :: tv} m (n , nn)
  = Ap1° n <$> align* m nn
align* {y :: ty} [] (m , mm) n
  = Ap1 m <$> align* mm n
align* {y :: ty} {v :: tv} (m , mm) (n , nn)
  = AX (m , n) <$> align* mm nn
++ Ap1 m <$> filter (not ∘ is-ap1°) (align* mm (n , nn))
++ Ap1° n <$> filter (not ∘ is-ap1) (align* (m , mm) nn)
where
  is-ap1 : {ty tv : II} → AI Δ_a ty tv → Bool
  is-ap1 (Ap1 _ _) = true
  is-ap1 _ = false

  is-ap1° : {ty tv : II} → AI Δ_a ty tv → Bool
  is-ap1° (Ap1° _ _) = true
  is-ap1° _ = false
```

We are now doing things in the *List* monad. This is needed because there are many possible alignments between two products. For the moment, we refrain from choosing and compute all of them.

On another note, some of these alignments are simply dumb! We do not want to have both $\text{Ap1 } x (\text{Ap1}^\circ y a)$ and $\text{AX } (x, y) a$. They are the same alignment. The *filters* are in charge of pruning out those branches from the search space.

Needs discussion:

We need a better way to optimize this. Preferably one that we can also use to optimize the mutually recursive variant.

Sticking with our example, we can align the leaves of our c by computing the following expression, where C-mapM is simply the monadic variant of C-map .

```
a = C-mapM align* c
= SX (CX 2-node' 3-node' (AX (10, 10) (AX (unit, unit) ...)))
::SX (CX 2-node' 3-node' (Ap1 10 (AX (unit, 10) ...)))
::SX (CX 2-node' 3-node' (Ap1 10 (Ap1 unit ...)))
::SX (CX 2-node' 3-node' (Ap1° 10 (AX (10, unit) ...)))
...
```

Now we have a problem. Which of the patches above should we chose to be *the* patch? Recall that we mentioned that we wanted to find the alignment with *maximum domain*. Something interesting happens if we look at patches from their application function, but first, we define the application of A1 :

$$\begin{aligned} (\text{AX } a_1 a_2)^b &= B \times \Pi D \xleftarrow{a_1^b \times a_2^b} A \times \Pi C \\ (\text{Ap1 } x a)^b &= \Pi B \xleftarrow{\pi_2 \bullet (\mathbb{1}_x \times a^b)} X \times \Pi A \\ (\text{Ap1}^\circ x a)^b &= X \times \Pi B \xleftarrow{\langle \underline{x}, a^b \rangle} \Pi A \\ \text{A0}^b &= \mathbb{1} \xleftarrow{!} \mathbb{1} \end{aligned}$$

TODO

The next section needs to be completely restructured. I'm here!

3 Patches as Partial Functions

In order to better illustrate this concept, we need a simpler example first. Let's consider the following type with no type variables:

```
Type1 = K kN ⊗ []
       ⊕ K kN ⊗ K kN ⊗ []
       ⊕ []
```

It clearly has two constructors:

```
C1 C2 : Constr Type1
C1    = fz
C2    = fs fz
```

Now, let's take two inhabitants of Type1 .

```

x y : [ Type1 ]
x   = inject C2 (4 , 10 , unit)
y   = inject C1 (10 , unit)

```

There are two possible options for `diff x y`:

```

ds : Patch* Type1
ds = SX (CX C2 C1 (AX (4 , 10) (Ap1 10 A0))) - P1
  :: SX (CX C2 C1 (Ap1 4 (AX (10 , 10) A0))) - P2
  :: []

```

Consider the semantics for Δ as described in the discussion box at Section 2.1, that is,

$$\begin{aligned}
 (x, x)^b &= id \\
 (x, y)^b &= \underline{y}|_x
 \end{aligned}$$

Then it becomes clear that we want to select patch (P2) instead of (P1). In fact, there is a deeper underlying reason for that! Let's calculate the application function of (P1):

$$\begin{aligned}
 P1^b &= \text{inj}_{C_1} \bullet ((4, 10)^b \times (\pi_2 \bullet (!|_{10} \times !))) \bullet \text{match}_{C_2} \\
 &= \text{inj}_{C_1} \bullet \underbrace{10|_4 \times (\pi_2 \bullet (!|_{10} \times !))}_{\alpha_1} \bullet \text{match}_{C_2}
 \end{aligned}$$

And for (P2), we have:

$$\begin{aligned}
 P2^b &= \text{inj}_{C_1} \bullet \pi_2 \bullet (!|_4 \times ((10, 10)^b \times !)) \bullet \text{match}_{C_2} \\
 &= \text{inj}_{C_1} \bullet \underbrace{\pi_2 \bullet (!|_4 \times (id \times !))}_{\alpha_2} \bullet \text{match}_{C_2}
 \end{aligned}$$

Drawing them as a diagram, in **Par**, we have:

$$\begin{array}{ccc}
 \text{typeOf}_{\text{Type1}} C_2 \equiv \mathbb{N} \times \mathbb{N} & \xleftarrow{\text{inj}_{C_2}^\circ} & [\text{Type1}] \\
 \downarrow \alpha_2 \quad \downarrow \alpha_1 & & \downarrow \downarrow \\
 \text{typeOf}_{\text{Type1}} C_1 \equiv \mathbb{N} & \xrightarrow{\text{inj}_{C_1}} & [\text{Type1}]
 \end{array}$$

$P2^b \parallel P1^b$
 $\Downarrow \Downarrow$

Here, we have something curious going on... We have that $P1^b \preceq P2^b$. To see this is not very hard. First note that pre and post Kleisli composition is monotonous with respect to \preceq , hence we just need to prove $\alpha_1 \preceq \alpha_2$, that is:

$$10|_4 \times (\pi_2 \bullet (!|_{10} \times !)) \preceq \pi_2 \bullet (!|_4 \times (id \times !))$$

Well, α_1 has only one element in its domain, and it is $(4, 10, \text{unit})$. Hence, we just need to check that α_2 is defined for $(4, 10, \text{unit})$. this is trivial to do. In fact, we can see that α_2 is defined for elements of the form $(4, x, \text{unit})$, $\forall x \in \mathbb{N}$.

$$\begin{aligned}
 &\langle lalala, \underline{10} \rangle \subseteq \langle \underline{4}, id \rangle \\
 &\equiv \{ \text{split universal} \} \\
 &\pi_1 \cdot \langle lalala, \underline{10} \rangle \subseteq \underline{4} \wedge \pi_2 \cdot \langle lalala, \underline{10} \rangle \subseteq id
 \end{aligned}$$

The first proof obligation is easy to calculate with:

$$\begin{aligned}
& \pi_1 \cdot < lalala, \underline{10} > \subseteq \underline{4} \\
& \Leftarrow \{ \pi_1\text{-cancel} ; \subseteq\text{-trans} \} \\
& lalala \subseteq \underline{4} \\
& \Leftarrow \{ \text{Leibniz} \} \\
& lalala \cdot \underline{10} \subseteq \underline{4} \cdot \underline{10} \\
& \equiv \{ \underline{a}^\circ \cdot \underline{a} \equiv \top \} \\
& \underline{4} \cdot \top \subseteq \underline{4} \cdot \underline{10} \\
& \equiv \{ \underline{a} \cdot \underline{b} \equiv \underline{a} \} \\
& \underline{4} \cdot \top \subseteq \underline{4} \\
& \equiv \{ \underline{a} \cdot \top \equiv \underline{a} \} \\
& \underline{4} \subseteq \underline{4} \\
& \equiv \{ \subseteq\text{-refl} \} \\
& \text{True}
\end{aligned}$$

The second is easier to prove once we add variables!

$$\begin{aligned}
& \pi_2 \cdot < lalala, \underline{10} > \subseteq id \\
& \equiv \{ \text{Add variables} \} \\
& \forall x, y . x (\pi_2 \cdot < lalala, \underline{10} >) y \Rightarrow x = y \\
& \equiv \{ \text{PF expand composition} \} \\
& \forall x, y . \exists z . x (\pi_2) z \wedge z < lalala, \underline{10} > y \Rightarrow x = y \\
& \equiv \{ \text{Types force } z = (z_1, z_2) \} \\
& \forall x, y . \exists z_1, z_2 . x (\pi_2) (z_1, z_2) \wedge (z_1, z_2) < lalala, \underline{10} > y \Rightarrow x = y \\
& \equiv \{ \pi_2 \text{ def} \} \\
& \forall x, y . \exists z_1, z_2 . x = z_2 \wedge (z_1, z_2) < lalala, \underline{10} > y \Rightarrow x = y \\
& \equiv \{ \text{split def} \} \\
& \forall x, y . \exists z_1, z_2 . x = z_2 \wedge z_1 (lalala) y \wedge z_2 (\underline{10}) y \Rightarrow x = y \\
& \equiv \{ \text{points def} \} \\
& \forall x, y . \exists z_1, z_2 . x = z_2 \wedge z_1 = 4 \wedge y = 10 \wedge z_2 = 10 \Rightarrow x = y \\
& \equiv \{ \text{substitutions ; weakening} \} \\
& \forall x, y . \exists z_2 . x = 10 \wedge y = 10 \Rightarrow x = y \\
& \equiv \{ \text{trivial} \} \\
& \text{True}
\end{aligned}$$

Nevertheless, it is clear which patch we should choose! We should always choose the patch that gives rise to the biggest relation, as this is applicable to much more elements.

Hence, our *cost* functions will count how many elements of the domain and range of the “application” relation of a patch are *fixed*. Note that the **S** and **C** parts of the algorithm are completely deterministic, hence they should *not* contribute to cost:

$$\begin{aligned}
& \text{S-cost} : \{ty : \mathbf{U}\} \{P : \mathbf{UUSet}\} (doP : \{k v : \mathbf{U}\} \rightarrow P \ k \ v \rightarrow \mathbb{N}) \\
& \quad \rightarrow \mathbf{S} \ P \ ty \rightarrow \mathbb{N} \\
& \text{S-cost } doP \ (\mathbf{SX} \ x) = doP \ x \\
& \text{S-cost } doP \ \mathbf{Scp} = 0 \\
& \text{S-cost } doP \ (\mathbf{Scns} \ i \ xs) = \text{foldr}_i (\lambda \ h \ r \rightarrow doP \ h + r) \ 0 \ xs \\
& \text{C-cost} : \{ty \ tv : \mathbf{U}\} \{P : \mathbf{IIISet}\} (doP : \{k v : \mathbf{II}\} \rightarrow P \ k \ v \rightarrow \mathbb{N}) \\
& \quad \rightarrow \mathbf{C} \ P \ ty \ tv \rightarrow \mathbb{N} \\
& \text{C-cost } doP \ (\mathbf{CX} \ i \ j \ x) = doP \ x
\end{aligned}$$

An **Alignment** might fix one element on the source, using **Ap1** or one element on the destination, using **Ap1°**.

```

Al-cost : {ty tv : Π} {P : AASet} (doP : {k v : Atom} → P k v → ℕ)
  → Al P ty tv → ℕ
Al-cost doP A0 = 0
Al-cost doP (Ap1 x a) = 1 + Al-cost doP a
Al-cost doP (Ap1° x a) = 1 + Al-cost doP a
Al-cost doP (AX x a) = doP x + Al-cost doP a

```

Last but not least, a **Δ** will either fix 2 elements: one in the source that becomes one in the destination; or none, when we just copy the source.

```

cost-delta-raw : ℕ
cost-delta-raw = 2

cost-delta : ∀ {α} {A : Set α} {ty tv : A} (P : A → Set)
  (eqA : (x y : A) → Dec (x ≡ y))
  (eqP : (k : A) (x y : P k) → Dec (x ≡ y))
  → delta P ty tv → ℕ
cost-delta {ty = ty} {tv = tv} P eqA eqP (pa1 , pa2)
  with eqA ty tv
... | no _ = cost-delta-raw
cost-delta {ty = ty} P eqA eqP (pa1 , pa2)
  | yes refl with eqP ty pa1 pa2
... | no _ = cost-delta-raw
... | yes _ = 0

```

According to these definitions, the cost of (P1) above is 3, where the cost of (P2) is 1.

3.1 Patches for Regular Types

Now that we have *spines*, *changes* and *alignments* figured out, we can define a patch as:

```

Patch : AASet → U → Set
Patch P = S (C (Al P))

```

Computing inhabitants of such type is done with:

```

diff1* : {ty : U} (x y : [[ ty ]]) → Patch* ty
diff1* x y = S-mapM (C-mapM (uncurry align*) ∘ uncurry change) (spine x y)

```

where **Patch*** is defined as **List** (**Patch** **Δ_a**).

3.2 Conjectures About the **cost** function

Here we conjecture a few lemmas about the interplay of the cost function and the “application” relation. Let *P*, *Q* and *R* be patches.

- i) If *P* has a lower cost than *Q*, then the domain and range of the “application” relation of *P* contains the “application” relation of *Q*.

$$\text{cost } P < \text{cost } Q \Rightarrow Q^b \subseteq P^b \bullet \top \bullet P^b$$

Needs discussion:

This is not as simple as

$$\text{cost } P < \text{cost } Q \Rightarrow Q^b \subseteq P^b$$

Take two *Deltas*, $px = (10, 50)$ and $py = (30, 30)$. Trivially, $\text{cost } py = 0$ and $\text{cost } px = 2$. Now, $px^b = \text{lalala}$ and $py^b = \text{id}$. It is not true that $\text{lalala} \subseteq \text{id}$!
If we state, however: Let P, Q in $\text{diff} * x y$; $\text{cost } P < \text{cost } Q \Rightarrow Q^b \subseteq P^b$
Seems more likely. As the above counter example would not work anymore.
 $\text{diff} * 10 \ 50 = (10, 50) :: []$.

- ii) If P and Q have equal cost, it means that there is at least one place where P and Q are doing *the same thing*, hence there is a patch that copies this *same thing* and costs strictly less.

$$\text{cost } P \equiv \text{cost } Q \Rightarrow \exists R \bullet \text{cost } R < \text{cost } P$$

4 Mutually Recursive Types

Now that we have a clear picture of regular types, extending this to recursive types is not very difficult.

First, recall that a mutually recursive family is defined as n codes that each reference n type variables:

```
Fam : ℕ → Set
Fam n = Vec (σπ n) n

data Fix {n : ℕ} (F : Fam n) : Fin n → Set where
  (⌊_⌋) : ∀ {k} → [ lookup k F ] (Fix F) → Fix F k
```

Another auxiliary definition we use here is the indexed coproduct, which let's us *extend* some indexed type.

$$\begin{aligned} _ +_u _ : \{n : \mathbb{N}\} \rightarrow (\sigma\pi n \rightarrow \sigma\pi n \rightarrow \text{Set}) \rightarrow (\sigma\pi n \rightarrow \sigma\pi n \rightarrow \text{Set}) \rightarrow (\sigma\pi n \rightarrow \sigma\pi n \rightarrow \text{Set}) \\ (P +_u Q) \text{ ty } tv = (P \text{ ty } tv) \uplus (Q \text{ ty } tv) \end{aligned}$$

Now, we already have the ingredients for detecting and representing common constructors or full copies, with **S**, constructor changes, with **C** and alignments with **Al**. We just need to handle type variables to tie the knot. Before we proceed with the nasty definitions, we still need two last synonyms:

```
Fami : Set
Fami = Fin fam#

T : Fami → σπ fam#
T k = lookup k fam
```

Here $T \ k$ represents the k -th type of the family, and Fam_i acts as the types in the family.

We start defining a patch for fixed points by allowing normal patches to perform changes on the first layer.

```
data Patchμ : U → U → Set where
  skel : {ty : U} → Patch (UU → AA Patchμ) ty
  → Patchμ ty ty
```

However, a patch for a fixed point might not only follow the precise order of operations (**S**, then **C**, then **Al**) that regular types enjoyed. For instance, imagine we are transforming the following lists:

$$[5, 8, 13, 21] \rightsquigarrow [8, 13, 21]$$

Let lists be seen (as usual) as the initial algebras of $L_A X = 1 + A \times X$; then, both lists are inhabitants of μL_N , but, more precisely, the source is an inhabitant of $L_N(L_N(L_N(L_N \mathbb{1})))$ whereas the target is an inhabitant of $L_N(L_N(L_N \mathbb{1}))$.

Here, we are already beginning with different types, so a spine (which is homogeneous) might not be the best start! In fact, the best start is to say that the first 5 is deleted, then the spine can kick in and say that everything else is copied!

Deleting and inserting can be seen as alignments between a type variable and the type we wish to delete or insert. For instance, deleting 5, which is actually $5 :: _$, consists in deleting the $_ :: _$ constructor and aligning the $\mathbb{N} \times \mathbb{I} k$ with a type variable $\mathbb{I} k$. Insertions are analogous.

```

ins  : {ty : U}{k : Fami}(i : Constr ty)
      → AI (UU→AA Patchμ) (I k :: []) (typeOf ty i)
      → Patchμ (T k) ty
del  : {ty : U}{k : Fami}(i : Constr ty)
      → AI (UU→AA Patchμ) (typeOf ty i) (I k :: [])
      → Patchμ ty (T k)

```

Note, however, that we use `UU→AA Patchμ` to populate the leaves of `Patch` and `AI`. That's because their parameter is of type `Atom → Atom → Set`, but `Patchμ` has type `U → U → Set`. Nevertheless, when these leaves are just two type variables, we want to keep using `Patchμ` to record their differences. When these leaves are constant types, we give up and set their values with a delta.

```

fix  : {k k' : Fami}
      → Patchμ (T k) (T k')
      → Patchμ (α (I k)) (α (I k'))
set  : {ty tv : U}
      → Δs ty tv
      → Patchμ ty tv

```

Associating costs are trivial. We just piggy back on the previous cost definitions. I am still very puzzled for how this works.

```

{-# TERMINATING #-}
Patchμ-cost : {ty tv : U} → Patchμ ty tv → ℕ
Patchμ-cost (skel x)
  = Patch-cost Patchμ-cost x
Patchμ-cost (ins i x)
  = AI-cost Patchμ-cost x
Patchμ-cost (del i x)
  = AI-cost Patchμ-cost x
Patchμ-cost (fix p)
  = Patchμ-cost p
Patchμ-cost (set x)
  = cost-Δs x

```

Nevertheless, the heart of the algorithm is the same as any patching algorithm out there. Things can be modified, inserted or deleted:

```

{-# TERMINATING #-}
diffμ* : {k k' : Fami} → Fix fam k → Fix fam k' → List (Patchμ (T k) (T k'))
diffμ* {k} {k'} ⟨ x ⟩ ⟨ y ⟩
  = diffμ*-mod {T k} {T k'} x y
++ diffμ*-ins {lookup k' fam} {k} ⟨ x ⟩ y
++ diffμ*-del {lookup k fam} {k'} x ⟨ y ⟩

```

Modifying must happen at the same type only, otherwise we force an insertion or deletion. If we are, in fact, on the same type, we can get the patch for that layer and continue diffing atoms.


```

diff $\mu^*$ -mod : {ty tv : U} → [ ty ] → [ tv ] → List (Patch $\mu$  ty tv)
diff $\mu^*$ -mod {ty} {tv} x y with  $\sigma\pi$ -eq ty tv
...| no _ = []
diff $\mu^*$ -mod x y
| yes refl
= skel <$> (diff1* x y »= Patch-mapM (uncurry diff $\mu^*$ -atoms))

```

Atoms are easy to diff. If reach two type variables, we tie the knot and keep patching.
If we reach two constants, we set one into the other. Any other situation is forbidden.

```

diff $\mu^*$ -atoms : {ty tv : Atom} → [ ty ]a → [ tv ]a → List (UU→AA Patch $\mu$  ty tv)
diff $\mu^*$ -atoms {l ty} {l tv} x y = fix <$> diff $\mu^*$  x y
diff $\mu^*$ -atoms {K ty} {K tv} x y = return (set (→ $\alpha$  {K ty} x , → $\alpha$  {K tv} y))
diff $\mu^*$ -atoms {K ty} {l tv} x y = []
diff $\mu^*$ -atoms {l ty} {K tv} x y = []

```

Insertions and deletions are very simple:

```

diff $\mu^*$ -ins : {ty : U}{k : Fami} → Fix fam k → [ ty ] → List (Patch $\mu$  (T k) ty)
diff $\mu^*$ -ins x y with sop y
diff $\mu^*$ -ins x _ | strip cy dy
= ins cy <$> align $\mu'$  (x , unit) dy

diff $\mu^*$ -del : {ty : U}{k : Fami} → [ ty ] → Fix fam k → List (Patch $\mu$  ty (T k))
diff $\mu^*$ -del x y with sop x
diff $\mu^*$ -del _ y | strip cx dx
= del cx <$> align $\mu'$  dx (y , unit)

```

We just need to pay attention not to insert or delete some constructor without arguments; This is done on the auxiliar alignment function:

```

align $\mu'$  : {ty tv :  $\Pi$ } → [ ty ]p → [ tv ]p
→ List (Al (UU→AA Patch $\mu$ ) ty tv)
align $\mu'$  {[]} {[]} _ _ = []
align $\mu'$  {[]} {[]} _ _ = []
align $\mu'$  {_ :: _} {_ :: _} x y = align $\mu$  x y

```

Which ignores empty products. If we are not aligning empty products, we can just piggyback on the previous alignment function we had:

```

align $\mu$  : {ty tv :  $\Pi$ } → [ ty ]p → [ tv ]p
→ List (Al (UU→AA Patch $\mu$ ) ty tv)
align $\mu$  x y = align* x y »= Al-mapM (uncurry diff $\mu^*$ -atoms)

```

Now we just need to choose the patch with the least cost for the *deterministic* version.

```

diff $\mu$  : {k k' : Fami} → Fix fam k → Fix fam k' → Patch $\mu$  (T k) (T k')
diff $\mu$  {k} x y with diff $\mu^*$  x y
...| s :: ss = s < $\mu$ > ss
...| [] = set (unmu x , unmu y)

```

Needs discussion:

If we define a family of two types, say $K\ N::K\ Bool::[]$. Now take $x = \text{inject } fz\ 10$ and $y = \text{inject } (fs\ fz)\ true$.

Here, $\text{diffmu}^* x\ y = []$. It is easy to see why. We can't modify because we have two different types. We can't insert or delete because *both* inhabitants were constructed with constructors that have no recursive arguments.

In a real scenario, though, this will never happen. Mainly because we are only interested in diffing things of the same type. Sure, the types might change during the algorithm, but they start the same; hence at least a modify is always possible.

On another note; The aforementioned type family is not really a type family, but a single type: $N + B$.

Obviously, we can also define the “application” relation for fixed points, and that is done in `RegDiff/Diff/Multirec/Domain`. I believe it is more worthwhile to look at some example patches and their “application” relation instead of the general case, though. Let's begin with the lists we just discussed:

```
s0 : Patchμ LIST-F LIST-F
s0 = diffμ (5 > 8 > 13 > 21 > #) (8 > 13 > 21 > #)

s0-norm : Patchμ LIST-F LIST-F
s0-norm = del cons' (Ap1 5 (AX (fix (skel Scp)) A0))
```

Here, `LIST-F` is defined as `u1⊕N⊗l`, the usual list functor. The “application” relation for the above patch is (isomorphic to):

$$\begin{array}{ccccc}
 1 + N \times [N] & \xrightarrow{\quad s0 \quad} & 1 + N \times [N] \\
 \downarrow \iota_2^\circ & & \uparrow \text{in}^\circ \\
 N \times [N] & \xrightarrow{\quad \bar{5}^\circ \times id \quad} N \times [N] \xrightarrow{\quad \pi_2 \quad} & [N]
 \end{array}$$

4.1 Examples

Here we add some more examples of patches over fixpoints. These can be seen in the respective `Lab.agda` modules. Here are a few examples of list patches:

```

l0 l1 : list
l0 = (3 > 50 > 4 > #)
l1 = (1 > 50 > 4 > 20 > #)

s1 : Patch $\mu$  LIST-F LIST-F
s1 = diff $\mu$  l0 l1

s1-normalized : Patch $\mu$  LIST-F LIST-F
s1-normalized
= skel
  (Scns cons'
    (CX fz fz (AX (set ( $\rightarrow\alpha$  3 ,  $\rightarrow\alpha$  1)) A0) ::
      (CX fz fz
        (AX
          (fix
            (skel
              (Scns cons'
                (CX fz fz (AX (set ( $\rightarrow\alpha$  50 ,  $\rightarrow\alpha$  50)) A0) ::
                  (CX fz fz
                    (AX
                      (fix
                        (skel
                          (Scns cons'
                            (CX fz fz (AX (set ( $\rightarrow\alpha$  4 ,  $\rightarrow\alpha$  4)) A0) ::
                              (CX fz fz
                                (AX (fix (ins cons' (Ap1 $^\circ$  20 (AX (fix (skel Scp)) A0)))) A0
                                :: []))))))
                              A0)
                              :: []))))))
                              A0)
                              :: []))))
  A0)
  :: []))

```

Previously we had *2-3-Trees* as an example. Here are some patches over them:

```

import RegDiff.Diff.Multirec.Base konstants keqs
as DIFF
open DIFF.Internal (2-3-TREE-F :: []) public

k0 k1 k2 : 2-3-Tree
k0 = Leaf
k1 = 2-Node 1 Leaf Leaf
k2 = 3-Node 5 Leaf Leaf Leaf
k3 = 3-Node 3 k1 k2 k2

t1 t2 : 2-3-Tree
t1 = 2-Node 4 k1 k2
t2 = 3-Node 5 k1 Leaf k2

```

The patches we calculate are:

```

r1 r2 : Patch $\mu$  2-3-TREE-F 2-3-TREE-F
r1 = diff $\mu$  t1 t2
r2 = diff $\mu$  k1 k3

```

Which are normalized to the following patches. Note that it is the $A\otimes$ in **r1** that lets us copy **k1** and **k2** from the **2-node** to the **3-node**.

```

r1-normalized : Patch $\mu$  2-3-TREE-F 2-3-TREE-F
r1-normalized
= skel
  (SX
    (CX 2-node' 3-node'
      (AX (set (i1 (4 , unit) , i1 (5 , unit)))
        (AX (fix (skel Scp))
          (Ap1 $^\circ$  < i1 unit > (AX (fix (skel Scp)) A0)))))))

r2-normalized : Patch $\mu$  2-3-TREE-F 2-3-TREE-F
r2-normalized
= ins 3-node'
  (Ap1 $^\circ$  3
    (AX (fix (skel Scp))
      (Ap1 $^\circ$ 
        < i2 (i2 (i1 (5 , < i1 unit > , < i1 unit > , < i1 unit > , unit)))
        >
        (Ap1 $^\circ$ 
          < i2 (i2 (i1 (5 , < i1 unit > , < i1 unit > , < i1 unit > , unit)))
          >
          A0))))))

```

5 Comparing to the Rose Tree approach

Needs discussion:

- By using a list of edit operations, they lose the alignments.
- They allow for sharing of data even when the same constructor is used.

A Products in the Kleisli Category

Given a monad $M : \mathbf{C} \rightarrow \mathbf{C}$, let $\mathbf{Kl}(M)$ be the Kleisli Category of M , we denote by $\square^b : \mathbf{C} \rightarrow \mathbf{Kl}(M)$ the *identity on objects* inclusion functor into the Kleisli of M ; it's action on arrows is defined as $f^b = \eta \cdot f$. We will denote composition in \mathbf{C} by \cdot whereas composition in $\mathbf{Kl}(M)$ will be denoted \bullet .

Coproducts carry trivially as \cdot^b is a left adjoint² and, hence, preserve colimits. Products are not so straight forward.

We can define a notion of *almost products* if M is commutative, that is, there exists a left and right strength such that $\tau_l \bullet \tau_r \equiv \tau_r \bullet \tau_l$. We denote $\tau_l \bullet \tau_r$ by δ .

If $A \times B$ is a product in \mathbf{C} and M is commutative, $A \times B^b$ will be the *almost product* in $\mathbf{Kl}(M)$ where $\langle f, g \rangle^b = \delta \cdot \langle f, g \rangle$. Although we have that it is the unique arrow into the product, it does not satisfy the β -elimination laws, that is, $(\pi_i \cdot \langle f_1, f_2 \rangle)^b \neq f_i^b$. This is because side-effects cannot be undone.

Nevertheless, for $M = \text{Maybe}$, we have that $(\pi_i \cdot \langle f_1, f_2 \rangle)^b \preceq f_i^b$, and this suffices for pretty much all that we need.

We define the product of two arrows the usual way:

$$(f \times g)^b = \langle f \cdot \pi_1, g \cdot \pi_2 \rangle^b$$

²We can define a functor $U : \mathbf{Kl}(M) \rightarrow \mathbf{C}$ as $U A = M A$ and $U f = \mu \cdot M f$. We have that $\square^b \dashv U$. In fact, this is the initial adjunction that constructs the monad M , $U \cdot \square^b = M$! In fact, $U(A^b) = M A$ and $U(f^b) = M f$. There is a final such adjunction giving rise to the Eilenberg-Moore construction.