Utrecht University

Computing Science MSc Thesis

Incremental Computation for Algebraic Datatypes in Haskell

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Write abstract

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Introduction

1.1 Motivation

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Write the research questions

Specific Implementation

```haskell
data Tree a = Leaf a
            | Node (Leaf a) a (Leaf a)

sumTree :: Tree Int -> Int
sumTree (Leaf x)     = x
sumTree (Node l x r) = x + (sumTree l) + (sumTree r)
```

Computing a value of a data structure can easily be defined in Haskell, but every time there is a small change in the `Tree`, the entire `Tree` needs to be recomputed. This is inefficient, because most of the computation has already been performed in the previous computation.

To prevent recomputation of already computed values, the technique memoization is introduced. Memoization is a technique where the results of computational intensive tasks are stored and when the same input occurs, the results are reused.

Write more about the equality and storage of data structure compared to hashes

Write about the bad side of using hashes

Implementing memoization for incrementally computing a value over a data structure, we need a way to efficiently represent the data structure. We do this by using hashes. Hashing is a process of transforming any string of characters into a fixed-length value, where the same input always generates the same output. Using hashes, we can efficiently store the data structure and check in constant time if two data structures are equal.

```haskell
sumTreeInc :: Tree Int -> (Int, Map Hash Int)
sumTreeInc l@(Leaf x)     = (x, insert (hash l) x empty)
sumTreeInc n@(Node l x r) = (y, insert (hash n) y (ml <> mr))
    where
        y = x + xl + xr
        (xl, ml) = sumTreeInc l
        (xr, mr) = sumTreeInc r
```

```haskell
sumTreeIncMap :: Map Hash Int -> Tree Int -> (Int, Map Hash Int)
sumTreeIncMap m l@(Leaf x) = case lookup (hash l) m of
    Just x  -> (x, m)
    Nothing -> (x, insert (hash l) x empty)
sumTreeIncMap m n@(Node l x r) = case lookup (hash n) m of
    Just x  -> (x, m)
    Nothing -> (y, insert (hash n) y (ml <> mr))
        where
            y = x + xl + xr
            (xl, ml) = sumTreeIncMap m l
            (xr, mr) = sumTreeIncMap m r
```

Generating a hash for every subset of the data structure for every execution is time-consuming and is unnecessary because most of the data structure stays the same. So, the Merkle Tree concept is used, which is inspired by the work of Miraldo and Swierstra[2]. The Merkle Tree integrates the hashes with the data structure.

2.1 Merkle Tree (`TreeH`)

First we introduce a new datatype `TreeH`, which contains a `Hash` for every constructor in `Tree`. Then to convert the `Tree` datatype into the `TreeH` datatype, the structure of the Tree is hashed and stored into the datatype using the `merkle` function.

```haskell
data TreeH a = LeafH Hash a
             | NodeH Hash (Leaf a) a (Leaf a)

merkle :: Tree Int -> TreeH Int
merkle (Leaf x) = LeafH h x
    where
        h = hash ["Leaf", x]
merkle (Node l x r) = NodeH h l' x r'
    where
        h = hash ["Node", x, getHash l', getHash r']
        l' = merkle l
        r' = merkle r
```

The precomputed hashes can then be used to easily create a `Map`, without computing the hashes every time the `sumTreeIncH` function is called.

```haskell
sumTreeIncH :: TreeH Int -> (Int, Map Hash Int)
```

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```
sumTreeIncH (LeafH h x)     = (x, insert h x empty)
sumTreeIncH (NodeH h l x r) = (y, insert h y (ml <> mr))
    where
        y = x + xl + xr
        (xl, ml) = sumTreeInc l
        (xr, mr) = sumTreeInc r
```

The problem with this implementation is, that when the `Tree` datatype is updated, the entire `Tree` needs to be converted into a `TreeH`, which is linear in time. This can be done more efficiently, by only updating the hashes which are impacted by the changes. Which means that only the hashes of the change and the parents need to be updated.

The first intuition to fixing this would be using a pointer to the value that needs to be changed. But because Haskell is a functional programming language, there are no pointers. Luckily, there is a data structure which can be used to efficiently update the data structure, namely the Zipper[1] data structure.

2.2 Zipper

The Zipper data structure works by keeping track of how the data structure is being traversed through. The Zipper keeps track by using a *context*. The context is the inverse of the direction which the data structure is traversed through. Meaning that when we traverse to the left side of the `Tree` the right side of the `Tree` gets stored in the context.

```
data Cxt a = Top
           | L (Cxt a) (Tree a) a
           | R (Cxt a) (Tree a) a
```

Then combining the `Tree` and the `Cxt`, we can traverse through the `Tree` and keeping the *context* of how we got to the current point.

```
type Loc a = (Tree a, Cxt a)
```

To use the Loc, first we need to create a `Cxt`.

```
enter :: Tree a -> Loc a
enter t = (t, Top)
```

Using the `Loc`, we can define multiple function on how to traverse through the `Tree`.

```
left :: Loc a -> Loc a
left (Node l x r, c) = (l, L c r x)

right :: Loc a -> Loc a
right (Node l x r, c) = (r, R c l x)

up :: Loc a -> Loc a
up (t, L c r x) = (Node t x r, c)
up (t, R c l x) = (Node l x t, c)
```

When we get to the desired point in the `Tree` which needs to be updated, we call the `modify` function.

```
modify :: (Tree a -> Tree a) -> Loc a -> Loc a
modify f (t, c) = (f t, c)
```

Then to get the entire updated tree back, we only need to call the `up` function indefinitely until the `Loc` is at the top.

```
leave :: Loc a -> Loc a
leave l@(t, Top) = l
leave l = top (up l)
```

2.2.1 Zipper `TreeH`

To implement the Zipper for `TreeH`, the same setup can be used as in the previous section, but when modifying the `TreeH` we also need to update all the parent nodes. First the `Cxt` needs to be updated to additionally store the hashes. Then the direction functions need to be updated to also store the `Hash` in the `Cxt`.

```
data Cxt a = Top
           | L (Cxt a) (Tree a) Hash a
           | R (Cxt a) (Tree a) Hash a
```

When modifying the value in the `TreeH`, the parents nodes hashes also needs to be updated.

```
updateLoc :: (TreeH a -> TreeH a) -> Loc a -> Loc a
updateLoc f l = if top l' then l' else updateParents (up l')
    where
        l' = modify f l

        updateParents :: Loc a -> Loc a
        updateParents (Loc x Top) = Loc (updateHash x) Top
        updateParents (Loc x cs)  = updateParents $ up (Loc (updateHash x) cs)

update :: (TreeH a -> TreeH a) -> [Loc a -> Loc a] -> TreeH a -> TreeH a
update f dirs t = leave $ updateLoc f l'
    where
        l' = applyDirs dirs (enter t)
```

3

Prototype Implementation

3.1   Generic programming in Haskell

Introduce generic programming and how it works

3.1.1   Pattern Functors vs Sums of products

Describe the differences between defining generic data types

3.1.2   Mutually recursive datatypes

Describe what mutually recursive datatypes are and why do we need to know about it

3.1.3   Comparison Generic Libraries

3.2   Prototype language

```
data I r        = I r
data K a r      = K a
data (:+:) f g r = Inl (f r) | Inr (g r)
data (:*:) f g r = Pair (f r, g r)
```

The definition of the pattern functor only leads to shallow recursion. Meaning that pattern functor can only be used to observe a single layer of recursion. To apply a function over the complete data structure, deep recursion is used. To implement deep recursion, the fix point is introduced.

```
data Fix f = In { unFix :: f (Fix f) }
```

The fix point is then used to describe the recursion of the datatype on a type-level basis. Using pattern functors and fix point most of the Haskell datatypes can be represented. For example:

```
data Tree a = Leaf a
            | Node (Tree a) a (Tree a)


type TreeG a = Fix (TreeF a)
type TreeF a = K a                     -- Leaf
             :+: ((I :*: K a) :*: I)   -- Node
```

Because the generic representation of the Haskell datatypes can be represented using pattern functors, we can use Functors. Using the Functor class a `cata` function can be defined, which is a generic fold function.

```
cata :: Functor f => (f a -> a) -> Fix f -> a
cata alg t = alg (fmap (cata alg) (unFix t))

cataSum :: TreeG Int -> Int
cataSum = cata f
  where
    f (Inl (K x))                     = x
    f (Inr (Pair (Pair (I l, K x), I r))) = x + l + r
```

To store the intermediate results of `cata`, we want the structure of the data to be hashed. This way we can easily compare if the data structure has changed over time, without completely recomputing the resulting digests. To do this, first a fix point is introduced which additionally stores the digest.

```
type Merkle f = Fix (f :*: K Digest)
```

Then to convert the fix point to a fix point containing the structural digest, the `Merkelize` class is introduced.

```
class Hashable f where
    hash :: Hashable g => f (Fix g) -> (f :*: K Digest) (Fix (g :*: K Digest))


merkleG :: Hashable f
        => f (Fix (g :*: K Digest)) -> (f :*: K Digest) (Fix (g :*: K Digest))
merkleG f = f :*: K (hash f)


merkle :: Hashable f => Fix f -> Merkle f
merkle = In . merkleG . fmap merkle . from
```

Using the new fix point with its structural digest, a new `cata` function can be defined which can store its intermediate values in a `Map Digest` a.

```
cataMerkleState :: (Functor f, Traversable f, Container c, Show (c a), Show a)
                => (f a -> a) -> Merkle f -> State (c a) a
cataMerkleState alg (In (Pair (x, K h))) = do m <- get
```

```
  case lookup h m of
    Just a  -> return a
    Nothing -> do y <- mapM (cataMerkleState alg) x
                  let r = alg y
                  modify (insert h r) >> return r


cataMerkle :: (Traversable f, Container c, Show (c a), Show a)
           => (f a -> a) -> Merkle f -> (a, c a)
cataMerkle alg t = runState (cataMerkleState alg t) empty
```

Zipper

Describe the use of the Zipper and how the hashes are updated

3.3   HashMap vs Trie

Write a piece about the comparison of storing it in a HashMap or a Trie datastructure

Write about Hdiff and the use of Trie datastructure

4

Generic Implementation

4.0.1 Regular

> Write about why Regular is chosen

> Write about the implementation of Regular and what had to change compared to the proto-type language

```haskell
newtype K a r     = K { unK :: a}      -- Constant value
newtype I r       = I { unI :: r }     -- Recursive value
data U r          = U                  -- Empty Constructor
data (f :+: g) r  = L (f r) | R (g r)  -- Alternatives
data (f :*: g) r  = f r :*: g r        -- Combine
data C c f r      = C { unC :: f r }   -- Name of a constructor
data S l f r      = S { unS :: f r }   -- Name of a record selector

merkle :: (Regular a, Hashable (PF a), Functor (PF a))
       => a -> Merkle (PF a)
merkle = In . merkleG . fmap merkle . from

cataSum :: Merkle (PF (Tree Int)) -> (Int, M.Map Digest Int)
cataSum = cataMerkle
  (\case
    L (C (K x))                 -> x
    R (C (I l :*: K x :*: I r)) -> l + x + r
  )
```

4.1 Complexity

> Describe for every function used the complexity and what leads to the complete complexity

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4.2 Memory Strategies

Describe multiple memory strategies for keeping memory usage and execution time low

Write about paper selective memoization

4.3 Pattern Synonyms

Explain Pattern Synonyms

```haskell
{-# COMPLETE Leaf_, Node_ #-}

pattern Leaf_ :: a -> PF (Tree a) r
pattern Leaf_ x <- L (C (K x)) where
  Leaf_ x = L (C (K x))


pattern Node_ :: r -> a -> r -> PF (Tree a) r
pattern Node_ l x r <- R (C (I l :*: K x :*: I r)) where
  Node_ l x r = R (C (I l :*: K x :*: I r))

cataSum :: MerklePF (Tree Int) -> (Int, M.Map Digest Int)
cataSum = cataMerkle
  (\case
    Leaf_ x      -> x
    Node_ l x r -> l + x + r
  )
```

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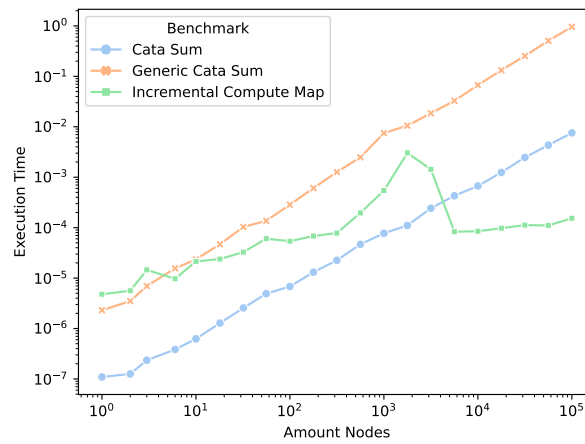# Experiments

5.1 Execution Time
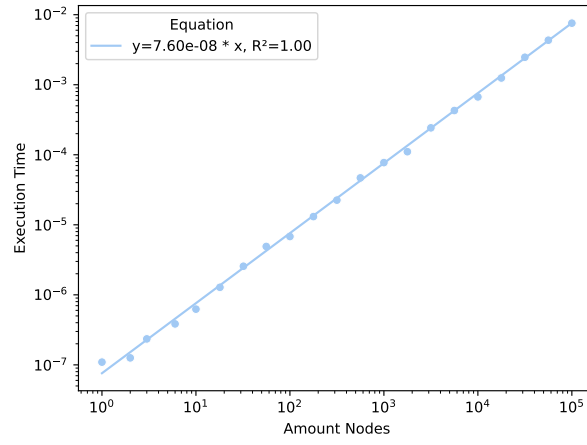

Figure 5.1: Overview execution time
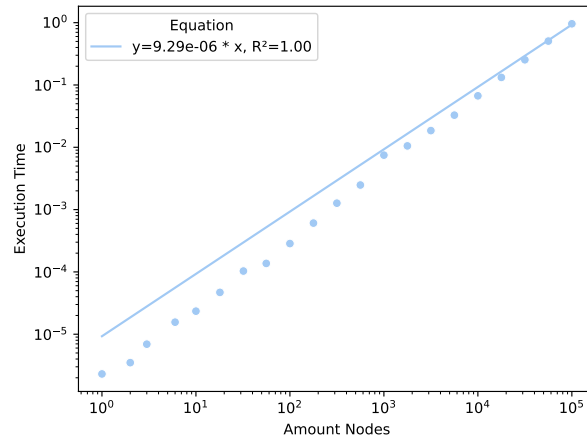
Figure 5.2: Execution time for Cata Sum

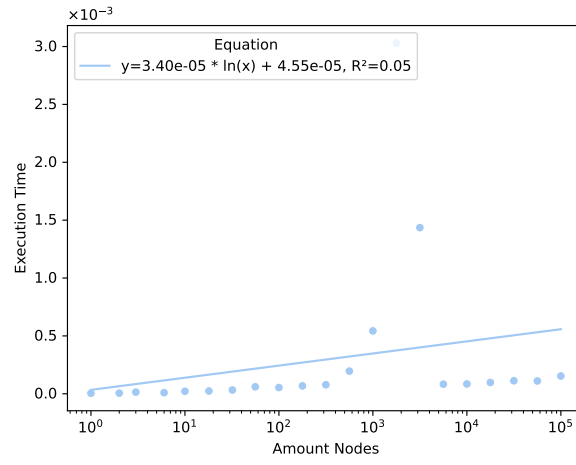Figure 5.3: Execution time for Generic Cata Sum

Figure 5.4: Execution time for Incremental Cata Sum
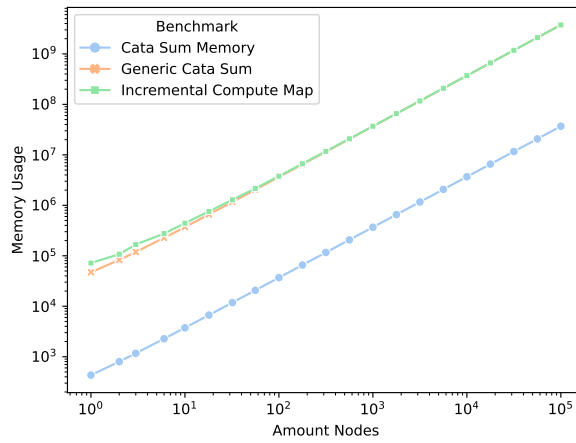
5.2 Memory Usage

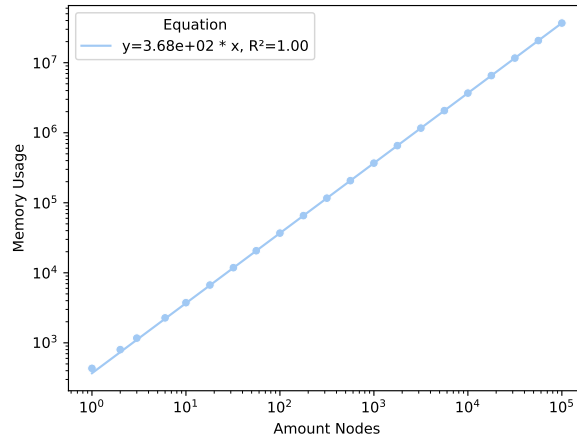

Figure 5.5: Overview memory usage
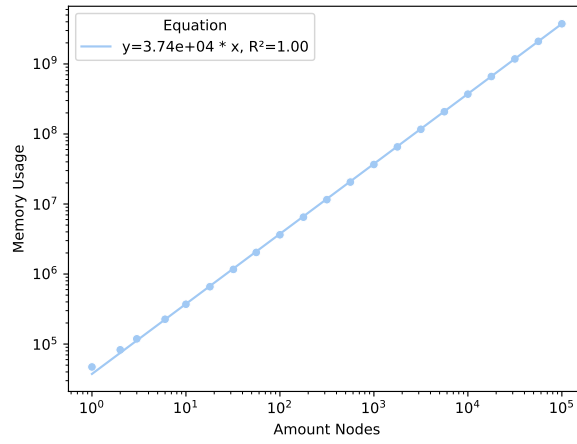
Figure 5.6: Memory usage for Cata Sum

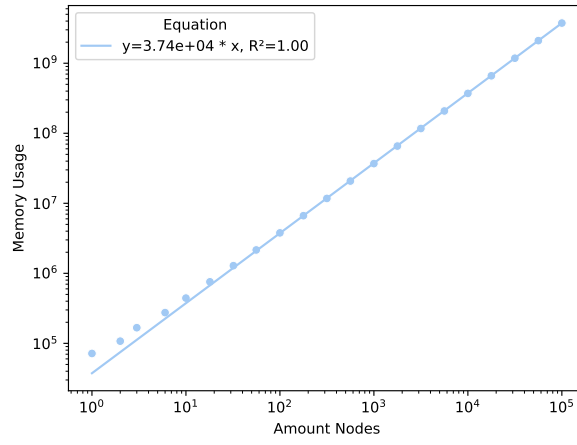Figure 5.7: Memory usage for Generic Cata Sum

Figure 5.8: Memory usage for Incremental Cata Sum
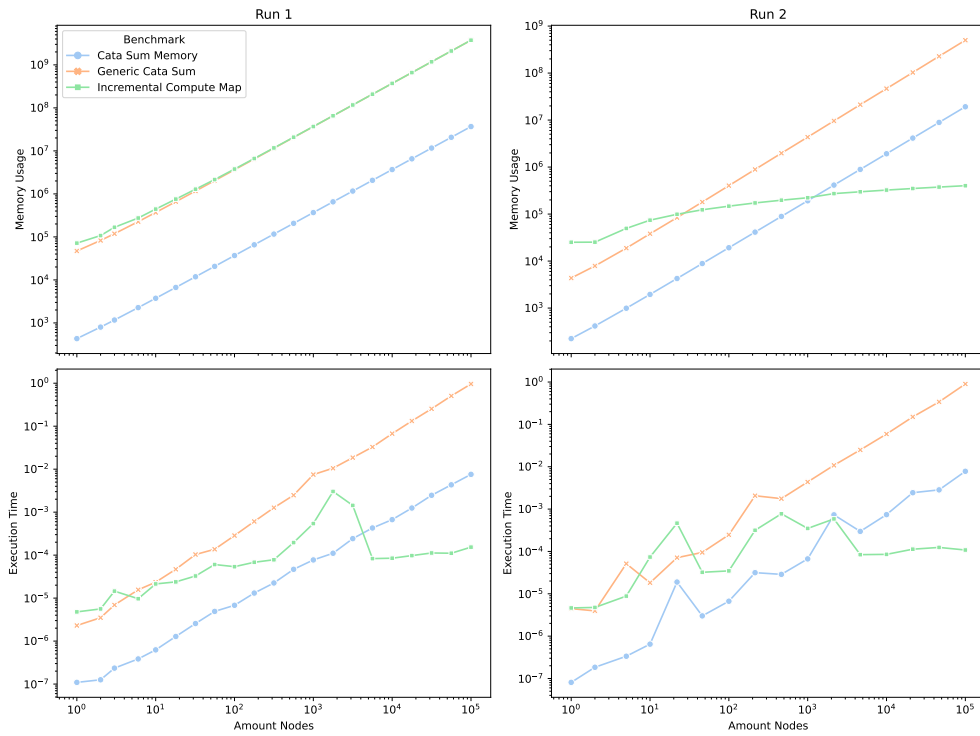
5.3 Comparison Memory Strategies

Figure 5.9: Comparison Memory Strategy

6
Conclusion and Future Work

6.1 Conclusion

A

Implementation Memo Cata

A.1 Definition Generic Datatypes

```
data U r       = U
data I r       = I r
data K a r     = K a
data (:+:) f g r = L (f r) | R (g r)
data (:*:) f g r = (f r) :*: (g r)
data C c f r   = C (f r)


newtype Fix f = In { out :: f (Fix f) }
```

A.2 Implementation Hashable

```
class Hashable f where
  hash :: f (Fix (g :*: K Digest)) -> Digest


instance Hashable U where
  hash _ = digest "U"


instance (Show a) => Hashable (K a) where
  hash (K x) = digestConcat [digest "K", digest x]


instance Hashable I where
  hash (I x) = digestConcat [digest "I", getDigest x]
    where
      getDigest :: Fix (f :*: K Digest) -> Digest
      getDigest (In (_ :*: K h)) = h


instance (Hashable f, Hashable g) => Hashable (f :+: g) where
```

```haskell
  hash (L x) = digestConcat [digest "L", hash x]
  hash (R x) = digestConcat [digest "R", hash x]


instance (Hashable f, Hashable g) => Hashable (f :*: g) where
  hash (x :*: y) = digestConcat [digest "P", hash x, hash y]


instance (Hashable f) => Hashable (C c f) where
  hash (C x) = digestConcat [digest "C", hash x]
```

A.3   Implementation Merkle

```haskell
type Merkle f = Fix (f :*: K Digest)


merkleG :: Hashable f
        => f (Fix (g :*: K Digest))
        -> (f :*: K Digest) (Fix (g :*: K Digest))
merkleG f = f :*: K (hash f)


merkle :: (Regular a, Hashable (PF a), Functor (PF a))
       => a -> Merkle (PF a)
merkle = In . merkleG . fmap merkle . from
```

A.4   Implementation Cata Merkle

```haskell
cataMerkleState :: (Functor f, Traversable f)
                => (f a -> a) -> Fix (f :*: K Digest)
                -> State (M.Map Digest a) a
cataMerkleState alg (In (x :*: K h)) = do m <- get
  case M.lookup h m of
    Just a -> return a
    Nothing -> do y <- mapM (cataMerkleState alg) x
                  let r = alg y
                  modify (M.insert h r) >> return r


cataMerkle :: (Functor f, Traversable f)
           => (f a -> a) -> Fix (f :*: K Digest) -> (a, M.Map Digest a)
cataMerkle alg t = runState (cataMerkleState alg t) M.empty
```

A.5 Implementation Zipper Merkle

```
data Loc :: * -> * where
  Loc :: (Zipper a) => Merkle a
                    -> [Ctx (a :*: K Digest) (Merkle a)]
                    -> Loc (Merkle a)


modify :: (a -> a) -> Loc a -> Loc a
modify f (Loc x cs) = Loc (f x) cs


updateDigest :: Hashable a => Merkle a -> Merkle a
updateDigest (In (x :*: _)) = In (merkleG x)


updateParents :: Hashable a => Loc (Merkle a) -> Loc (Merkle a)
updateParents (Loc x []) = Loc (updateDigest x) []
updateParents (Loc x cs) = updateParents
                              $ expectJust "Exception: Cannot go up"
                              $ up (Loc (updateDigest x) cs)


updateLoc :: Hashable a => (Merkle a -> Merkle a)
                        -> Loc (Merkle a) -> Loc (Merkle a)
updateLoc f loc = if   top loc'
                  then loc'
                  else updateParents
                      $ expectJust "Exception: Cannot go up" (up loc')
  where
    loc' = modify f loc
```

B

Regular

B.1 Zipper

```
data instance Ctx (K a) r
data instance Ctx U r
data instance Ctx (f :+: g) r = CL (Ctx f r) | CR (Ctx g r)
data instance Ctx (f :*: g) r = C1 (Ctx f r) (g r) | C2 (f r) (Ctx g r)
data instance Ctx I r = CId
data instance Ctx (C c f) r = CC (Ctx f r)
data instance Ctx (S s f) r = CS (Ctx f r)
```

Bibliography

[1] Gérard Huet. "The zipper". In: *Journal of functional programming* 7.5 (1997), pp. 549–554.

[2] Victor Cacciari Miraldo and Wouter Swierstra. "An efficient algorithm for type-safe structural diffing". In: *Proceedings of the ACM on Programming Languages* 3.ICFP (2019), pp. 1–29.