

# Generic Incremental Computation for Regular Datatypes

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Generic **Incremental Computation** for Regular Datatypes

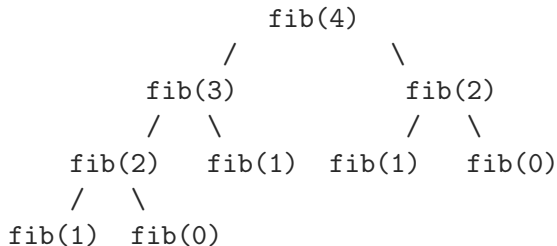
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Incremental computation tries to improve performance, when the same function is called again, by only computing the output of the changed input compared to the previous computations

# Title Explanation – Example Incremental Computation

```
fib :: Int -> Int
fib 0 = 0
fib 1 = 1
fib n = fib (n - 1) + fib (n - 2)
```

## Call Hierarchy



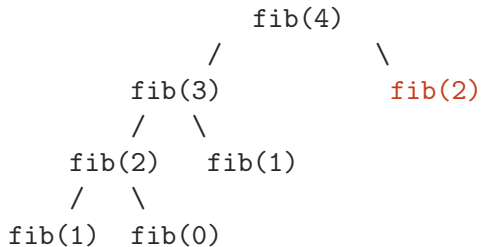
A technique for implementing incremental computations is:

### *Memoization*

stores the result of a computation and returns the cached result when the same input occurs again.

# Title Explanation – Example Incremental Computation

## New Call Hierarchy



Function Call	Result
<code>fib(2)</code>	1
<code>fib(3)</code>	2
<code>fib(4)</code>	3

## Cached Results

## Generic Incremental Computation for Regular Datatypes

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Generic refers to *datatype-generic programming*, which is a form of abstraction that allows defining functions that can operate on a large class of datatypes.

## Title Explanation – Generic Example

```
data List a = Nil | Cons a (List a) -- Haskell Notation [] / x : []
```

```
length :: List a -> Int
```

```
length Nil = 0
```

```
length (Cons _ t) = 1 + length t
```

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

```
length :: Tree a -> Int
```

```
length Leaf _ = 1
```

```
length (Node l _ r) = 1 + length l + length r
```

## Title Explanation – Generic Example

```
gLength :: (Generic f) => f a -> Int
gLength = ...
```

A *single* length function can be written, that can operate on lists, trees, and many other datatypes

```
> gLength (Cons 1 (Cons 2 (Cons 3 Nil))) -- List Int
3
```

```
> gLength (Node Leaf 1 Leaf) -- Tree Int
3
```



## Generic Incremental Computation for **Regular Datatypes**

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Regular datatypes are recursive datatypes, which can only recurse into themselves, such as lists, binary trees, etc.

# Title Explanation – Regular Datatypes Example

## Regular Datatypes

```
data List a = Nil | Cons a (List a)
```

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

## Not Regular Datatypes

```
data Tree a = Empty | Node a (Forest a)
```

```
data Forest a = Nil | Cons (Tree a) (Forest a)
```

## **Generic Incremental Computation for Regular Datatypes**

- Improve performance by only recomputing changed input
- Using generic programming to define functionality for a large class of datatypes
- The algorithm works for the regular datatypes

## Goal

Implement an incremental algorithm which performs better than the non-incremental algorithm for large regular datatypes

## Solution – Example with Memoization

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

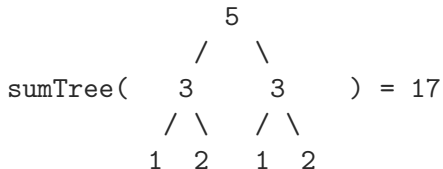
```
sumTree :: Tree Int -> Int
```

```
sumTree (Leaf x)      = x
```

```
sumTree (Node l x r) = x + sumTree l + sumTree r
```

```
exampleTree = Node (Node (Leaf 1) 3 (Leaf 2)) 5 (Node (Leaf 1) 3 (Leaf 2))
```

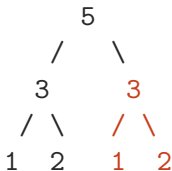
### Visual representation

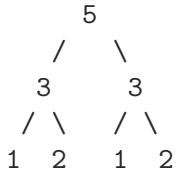



# Solution – Example with Memoization

## Memoized version of the sumTree

Example Tree



Tree	Result
	17
	6

Cached Results

## Solution – Equality

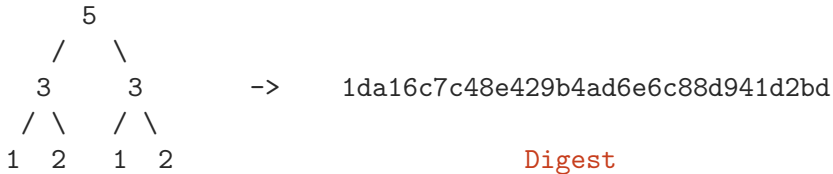
For every function call, the given tree needs to compare if there is an already cached input. This is accomplished by traversing through both trees.

```
instance Eq a => Eq (Tree a) where
  Leaf x1      == Leaf x2      = x1 == x2
  Node l1 x1 r1 == Node l2 x2 r2 = x1 == x2 && l1 == l2 && r1 == r2
  _            == _            = False
```

This is inefficient, and to improve the performance, we are going to use a *hash function*.

## Solution – Using Hash function

A *hash function* is a process of transforming input into an arbitrary fixed-size value (i.e., digest), where the same input always generates the same output.



The comparison for equality is now **constant** time instead of *linear*.

**However**, the digest still needs to be computed for every comparison, while the structure stays the same. So, the digest needs to be stored somewhere.



## Solution – Storing the Digests

A *Merkle Tree* is a data structure which integrates the *digests*, which represents the internal structure, within the data structure.

```
data TreeH a = LeafH Digest a
             | NodeH Digest (TreeH a) a (TreeH a)
```

```
instance Eq (TreeH a) where
  LeafH d1 _      == LeafH d2 _      = d1 == d2
  NodeH d1 _ _ _  == NodeH d2 _ _ _  = d1 == d2
  _              == _                  = False
```

## Solution – Efficiently updating the Input

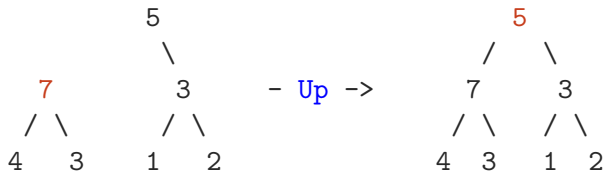
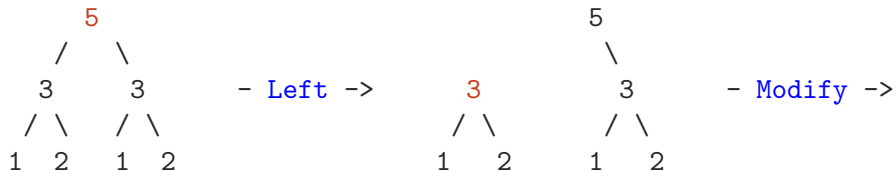
When a new tree is given to the function, the digests of the entire tree needs to be rehashed. This is **inefficient**, because, the affected digests are the changed parts of the tree and their parent nodes.

A more **efficient** method is, by passing a set of instructions on how to update the current input into the newly changed input. Using the set of instructions we can:

- traverse through the data structure
- modify the value of the current position with a given function
- and to return to a complete tree, we traverse back to the rootnode, while updating the digests

To implement the previously described functionality, we use a *Zipper*.

## Solution – Efficiently updating the Input



# Generic Programming – Pattern Functors

## Primitive Type Constructors

```
data U r          = U          -- Empty constructor
data I r          = I r        -- Recursive position
data K a r        = K a        -- Constant
data (f :+: g) r = L (f r) | R (g r) -- Sums (Choice)
data (f **: g) r = (f r) **: (g r) -- Products (Combine)
```

## Pattern functor for the Tree datatype

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

type instance PF (Tree a) = K a          -- Leaf
               :+: (I **: K a **: I)    -- Node
```

## Generic Implementation – Hashing Primitive Type Constructors

```
class Hashable f where  
  hash :: f (Merkle g) -> Digest
```

- The `f` represents the primitive type constructors (i.e., `U`, `I`, `K`, `:+:`, `:*:`)
- The `Merkle g` is the type of the recursive position (i.e., `I r`).
- `Merkle g` contains the `Digest` of its internal structure.
- The `hash` function only converts a single layer of the pattern functor.

## Generic Implementation – Hashing Primitive Type Constructors

```
instance Hashable U where
```

```
  hash _ = hash "U"
```

```
instance (Show a) => Hashable (K a) where
```

```
  hash (K x) = digestConcat [hash "K", hash x]
```

## Generic Implementation – Hashing Primitive Type Constructors

```
instance (Hashable f, Hashable g) => Hashable (f :+: g) where
  hash (L x) = digestConcat [hash "L", hash x]
  hash (R x) = digestConcat [hash "R", hash x]

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

## Generic Implementation – Hashing Primitive Type Constructors

```
class Hashable f where
  hash :: f (Merkle g) -> Digest

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



## Generic Implementation – Generic Merkle Tree

```
merkleG :: Hashable f => f (Merkle g) -> (f :: K Digest) (Merkle g)
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
```

```
> let x = merkle (Node (Leaf 1) 2 (Leaf 3))
    x :: (Merkle (PF (Tree Int)))
    x = In ((R (I (In (K 1)) :: K 2 :: I (In (K 3)))) -- PF (Tree Int)
             :: K 1da16c7c48e)                       -- Merkle
```

## Generic Implementation – Cata Merkle

`cata` means *catamorphism* which is a generalization of a *fold*. A fold combines the data structure into a single value (e.g., `sumTree` is a fold).

```
cataMerkleState :: (Functor f, Traversable f)
                => (f a -> a) -> Merkle f -> State (HashMap Digest a) a
cataMerkleState alg (In (x :: K d))
  = do m <- get
      case lookup d m of
        Just a  -> return a
        Nothing -> do y <- mapM (cataMerkleState alg) x
                        let r = alg y
                        modify (insert d r) >> return r
```

## Generic Implementation – Incremental Sum Tree

```
incSumTree :: Merkle (PF (Tree Int)) -> (Int, HashMap Digest Int)
incSumTree = cataMerkle
  (\case
    L (K x)          -> x
    R (I l :: K x :: I r) -> l + x + r
  )

> incSumTree (merkle (Node (Leaf 1) 2 (Leaf 3)))
(6, {"931090e5": 1, "7d1ef1c9": 3, "ba811ed5": 6})
```

## Generic Implementation – Pattern Synonyms

*Pattern synonyms* add an abstraction over patterns, which allows the user to move additional logic from guards and case expressions into patterns.

```
incSumTree :: Merkle (PF (Tree Int)) -> (Int, HashMap Digest Int)
incSumTree = cataMerkle
  (\case
    Leaf_ x      -> x
    Node_ l x r  -> l + x + r
  )
```

## Generic Implementation – Final Process

```
> let exampleTree = merkle (Node (Leaf 1) 2 (Leaf 3))

> let (y, m) = incSumTree exampleTree
      (6, {"931090e5": 1, "7d1ef1c9": 3, "ba811ed5": 6})

> let newTree = update [Left] (const (Leaf 4)) exampleTree

> incSumTreeMap m exampleTree
      (9, {"931090e5": 1, "7d1ef1c9": 3, "ba811ed5": 6, "61f159e6": 4
           , "16d55294": 9})
```

## Three functions:

- Cata Sum
  - ▶ Non-incremental algorithm (which computes sumTree)
- Generic Cata Sum
  - ▶ Incremental algorithm with an empty cache
- Incremental Cata Sum
  - ▶ Incremental algorithm with a filled cache

## Three scenarios:

- Worst case: updates the lowest left leaf with a new leaf
- Average case: updates a node in the middle of the data structure with a new leaf
- Best case: updates the left child of the root-node with a new leaf

# Experiments – Results - Execution Time

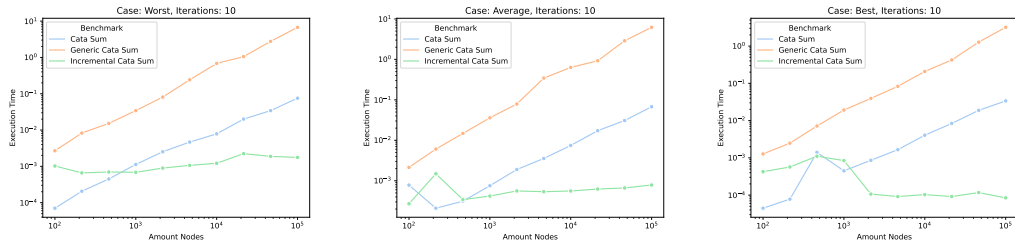


Figure: The execution time over 10 executions for the Worst, Average and Best case.

# Experiments – Results - Memory Usage

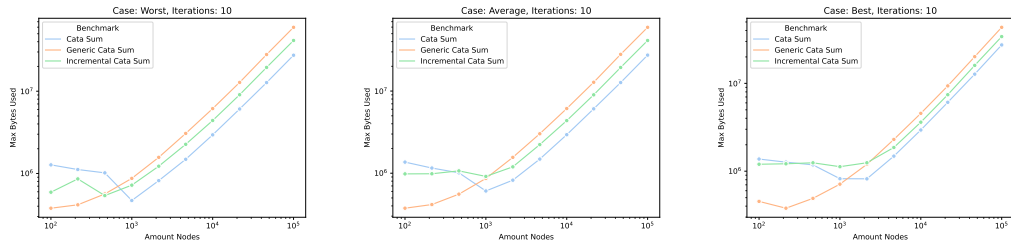


Figure: The max-bytes-used over 10 executions for the Worst, Average and Best case.



# Conclusion – Limitation & Future Work

## Limitations

- Only supports Regular Datatypes
- Pattern Synonyms needs to be handwritten

## Future Work

- Support Mutually Recursive Datatypes
- Generate Pattern Synonyms using TemplateHaskell

## Conclusion – Undiscussed Topics

- Cache management
- The explanation of fixed-point
- Implementation of the (Generic) Zipper

## Conclusion – Summary

- We have implemented an efficient incremental algorithm for regular datatypes
- The incremental algorithm is faster than the non-incremental version when the data structure contains more than  $10^3$  nodes
- We use pattern synonyms, so that the developer experience is almost at the same level as the non-incremental implementation
- However, the initial pass of the incremental algorithm is a lot slower than the non-incremental version. Therefore, the incremental algorithm needs to be performed a lot (preferably, with small changes), before being overall faster than the non-incremental version.