Memoisation, Hylomorphism and Nexuses





"Memo" Functions

"It would be useful if computers could learn from experience and thus automatically improve the efficiency of their own programs during execution." Donald Michie

Created recursive functions for:

- Factorial
- Highest Common Factor
- Member
- Reverse

"Memo" Functions

We define the rule part of factorial as follows: if n < 0 or if not (n. isinteger) then undef else if n = 0 then 1 else n * fact (n - 1) close end

To endow fact with the "memo" facility, using Popplestone's routines, we merely write newmemo (fact, 100, nonop =) \rightarrow fact;

...rote has an upper fixed limit of 100 entries...the symbol nonop warns the machine not to try to operate the "=" function at this stage...

Ruby Factorial

```
def fact(n)
  n < 1 ? 1 : n * fact(n - 1)
end
require 'method_decorators/decorators/memoize'
+Memoize
def ffact(n)
  n < 1 ? 1 : n * ffact(n - 1)
end
```

Coffeescript Factorial

```
fact = (n) \rightarrow
  if n < 1 then 1 else n * fact(n - 1)
memos = | |
ffact = (n) \rightarrow
  memos[n] ?=
    if n < 1
    else
       n * ffact(n - 1)
```



Coffeescript Factorial

```
ffact:
  do ->
    memos = \{\}
    (n) ->
      memos[n] ?=
        if n < 1
        else
           n * ffact(n - 1)
```

Haskell Factorial

```
fac :: <u>Int</u> -> <u>Int</u>
fac 0 = 0
fac 1 = 1
fac n = n * fac (n - 1)

facs :: [<u>Int</u>]
facs = <u>scanl</u> (*) 1 [1...]
ffac n = facs !! n
```



Fibonacci Example

Named after a guy called Fibonacci, well actually Leonardo of Paris.

$$F_n = F_{n-1} + F_{n-2}$$

Letters and Numbers

Based on the popular television series "Countdown" or French series "La Conte est Bou".

Given a list of numbers: 2, 5, 8, 10, 11, 17, 24, 50

Target: 53280

Use: +, *, /, -

Answer: II + 5 * 8 - I7 * 24 * I0 * 2

```
data Op = Add | Sub | Mul | Div
data Expr = Num Int | App Op Expr Expr
type Value = Int
subseqs [x] = [[x]]
subseqs (x:xs) = xss ++ [x] : map(x:) xss
        where xss = subseqs xs
value :: Expr -> Value
value (Num x) = x
value (App op e1 e2) = apply op (value e1)
  (value e2)
```

```
legal :: Op -> Value -> Value -> <u>Bool</u>
legal Add v1 v2 = (v1 <= v2)
legal Sub v1 v2 = (v2 < v1)
legal Mul v1 v2 = (1 < v1) && (v1 <= v2)
legal Div v1 v2 = (1 < v2) && (v1 \text{ mod} \text{ v2} == 0)
```



```
unmerges :: [a] -> [([a], [a])]
unmerges [x, y] = [([x], [y])]
unmerges (x:xs) = [([x], xs)] ++
    concatMap (add x) (unmerges xs)
    where
    add x (ys, zs) = [(x : ys, zs), (ys, x : zs)]
```

Adding Memoisation

```
data Trie a = Node a [(Int, Trie a)]
type Memo = Trie [(Expr, Value)]

memoise :: [[Int]] -> Memo
memoise = foldl insert empty
insert memo xs = store xs (mkExprs memo xs) memo
```

Adding Memoisation

Skeleton Tree

```
data Tree = Tip Int | Bin Tree Tree
type Memo = Trie [Tree]

memoise :: [[Int]] -> Memo
memoise = foldl insert empty
insert memo xs = store xs (mkTrees memo xs)
```

data Trie a = Node a [(Int, Trie a)]

Tuesday, 25 September 12

memo



Three Examples



Results

PROBLEM	STANDARD	TRIE	SKELETON TREE
1,3,7,10,11,12, 14,50 → 12831	0.286079s	13.618973s	13.869919s
2,3,7,10,12,19, $24,50 \rightarrow 53280$	1.998651s	28.124236s	28.863705s
2,5,8,10,11,17, 24,50 → 53280	1.425602s	7.130803s	7.416698s



What's Going On?

Hypothesis I:

Haskell Optimisations, GC

Hypothesis 2:

Cost of Memoisation > Recalculation



Problems

Initialising

Cache Miss

Extra Resources



Recalculate Intead?

2010

Improvement

Littles Law

Concurrency = Bandwidth * Latency

CPU Speed	3Mhz	3Ghz	×1000
Memory Bandwidth	~I0Mb/sec	~I0Gb/sec	×1000
Memory Latency	200ns	10ns	×20

1975



Hylomorphisms

A computation that consists of the duals fold and unfold.

hylo = fold f g . unfold p v h

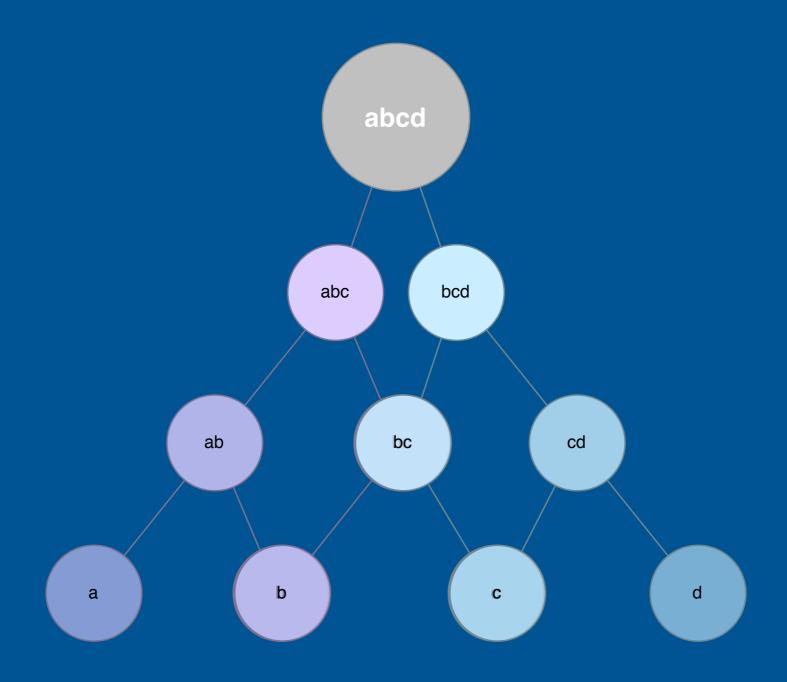
Catamorphism Lists

```
prod(x:xs) = x * prod xs
type ListCata x u = (u, x -> u -> u)
listCata :: ListCata x u -> [x] -> u
listCata (a, f) = cata where
       cata (x:xs) = f x (cata xs)
prod2 = listCata (1, (*))
rev = listCata ([], (\a b -> b ++ [a]))
```

Anamorphism Lists

```
count 0 = []
count n = n : count (n - 1)
type ListAna u x = u \rightarrow Either () (x, u)
listAna :: ListAna u x -> u -> [x]
listAna \ a = ana \ where
  ana u = case a u of
    Left _ -> []
    Right (x, xs) \rightarrow x : ana xs
count2 = listAna destructCount where
  destructCount 0 = Left ()
  destructCount n = Right (n, n - 1)
```

Nexus



Nexus

```
data Tree a = Leaf a | Node [Tree a]
```

```
unfold :: (b -> Bool) -> (b -> a) -> (b ->
[b]) -> b -> Tree a
unfold p v h x = if p x then Leaf (v x) else
Node (map (unfold p v h) (h x))
```

```
mkTree :: ([a] -> [[a]]) -> [a] -> Tree [a]
mkTree h = unfold single <u>id</u> h
```

Nexus

```
data LTree a = LLeaf a | LNode a [LTree a]
treeToNexus :: Tree [a] -> LTree [a]
treeToNexus = fill id recover
fill :: (a -> b) -> ([b] -> b) -> Tree a -> LTree b
fill f g = fold (lleaf f) (lnode g)
fold :: (a -> b) -> ([b] -> b) -> Tree a -> b
fold f g (Leaf x) = f x
fold f g (Node ts) = g (map (fold f g) ts)
```

Nexus and Countdown

```
minors :: [a] -> [[a]]
minors [x, y] = [[x], [y]]
minors (x : xs) = map (x :) (minors xs) ++ [xs]

unmerges :: [a] -> [([a], [a])]
unmerges x = unmerge $ halved $
   [treeToNexus $ mkTree minors x]
```

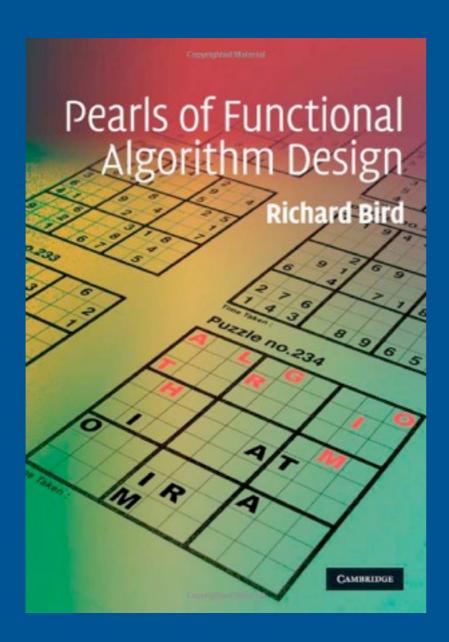


Results

PROBLEM	STANDARD	NEXUS
1,3,7,10,11,12, 14,50 → 12831	0.286079s	0.32372s
2,3,7,10,12,19, 24,50 → 53280	1.998651s	1.322064s
2,5,8,10,11,17, 24,50 → 53280	1.425602s	2.486383s



My Learning



Sorting morphisms

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Abstract. Sorting algorithms can be classified in many different ways. The way presented here is by expressing the algorithms as functional programs and to classify them by means of their recursion patterns. These patterns on their turn can be classified as the natural recursion patterns that destruct or construct a given data-type, the so called cata- and anamorphisms respectively. We show that the selection of the recursion pattern can be seen as the major design decision, in most cases leaving no more room for more decisions in the design of the sorting algorithm. It is also shown that the use of alternative data structures may lead to new sorting algorithms.

This presentation also serves as a gentle, light-weight, introduction into the various morphisms.

1 Introduction

In this paper we present several well known sorting algorithms, namely insertion sort, straight selection sort, hibble sort, quick sort, herp sort and merge sort (see e.g. [Knn73, Wn76]) in a non-standard way. We express the sorting algorithms as functional programs that obey a certain pattern of recursion. We show that for each of the sorting algorithms, the recursion patterns forms the major design decision, often leaving no more space for additional decisions to be taken. We make these recursion patterns explicit in the form of higher-order functions, much like the well-known map function on lists.

In order to reason about recursion patterns, we need to formalize that notion. Such a formalization is already available, based on a category theoretical modeling of recursive data types as can e.g. be found in [Fok92, Mei32]. In [BdM94] this theory is presented together with its application to many algorithms, including selection sort and quicksort. These algorithms can be understood however only after absorbing the underlying category theory. There is no need to present that theory here. The results that we need can be understood by anyone having some basic knowledge of functional programming, hence we repeat only the main results here. These usults show how to each recursive data type a number of morphisms is related, each capturing some pattern of recursion which involve the recursive structure of the data type. Of these morphisms, we use the so called catamorphism, anamorphism, hylomorphism and paramorphism on linear lists and binary trees. The value of this approach is not so much in obtaining a nice implementation of some algorithm, but in unraveling its structure.

This presentation gives the opportunity to introduce the various morphisms in a simple way, namely as patterns of recursion that are useful in functional programming, instead



Links

"'Memo' Function and Machine Learning" http://www.cs.utexas.edu/users/hunt/research/hash-cons/hash-cons-papers/michie-memo-nature-1968.pdf

"Extending Ruby with Ruby" and method_decorators gem https://github.com/newhavenrb/conferences/wiki/Extending-Ruby-with-Ruby
https://github.com/michaelfairley/method_decorators

"Memoized, the *practical* method decorator" https://github.com/raganwald/method-combinators



Links

"Functional Pearl Trouble Shared is Trouble Halved" http://www.cs.ox.ac.uk/ralf.hinze/publications/HW03.pdf

"Sorting Morphisms"
http://citeseerx.ist.psu.edu/viewdoc/summary?
doi:10.1.1.51.3315

Code and Examples: https://github.com/newmana/memoization