

replicate $(Int \rightarrow a \rightarrow [a])$, iterate $((a \rightarrow a) \rightarrow a \rightarrow [a])$, splitAt $(Int \rightarrow [a] \rightarrow ([a], [a]))$

1 Evaluation

Definition 1.1 (Counting) To evaluate $T(e)$:

- **Application** $T(f\ e_1 \dots e_n) = T(f)\ e_1 \dots e_n + T(e_1) + \dots + T(e_n)$
- **Variable** $T(x) = 0$
- **Primitives** For primitive function f , $T(f)\ x_1 \dots x_n = 0$
- **Conditional** $T(\text{if } p \text{ then } e_1 \text{ else } e_2) = T(p) + \text{if } p \text{ then } T(e_1) \text{ else } T(e_2)$

2 Asymptotics

Definition 2.1 (L-function) is a real, positive, monotonic, continuous function where as $n \rightarrow \infty$, $f(n) \rightarrow$ one of $0, \infty$ or some value k .

Definition 2.2 (Du Bois-Reymond and Bachman-Landau Notation) For L-functions f and g ,

- $f \prec g \Leftrightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0 \Leftrightarrow f \in o(g(n))$ where $o(g(n)) = \{f | \forall \delta > 0. \exists n_0 > 0. \forall n > n_0. f(n) < \delta g(n)\}$
- $f \preceq g \Leftrightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} < \infty \Leftrightarrow f \in O(g(n))$ where $O(g(n)) = \{f | \exists \delta > 0. \exists n_0 > 0. \forall n > n_0. f(n) \leq \delta g(n)\}$
- $f \asymp g \Leftrightarrow 0 < \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} < \infty \Leftrightarrow f \in \Theta(g(n))$ where $\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$
- $f \succeq g \Leftrightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} > 0 \Leftrightarrow f \in \Omega(g(n))$ where $\Omega(g(n)) = \{f | \exists \delta > 0. \exists n_0 > 0. \forall n > n_0. f(n) \geq \delta g(n)\}$
- $f \succ g \Leftrightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty \Leftrightarrow f \in \omega(g(n))$ where $\omega(g(n)) = \{f | \forall \delta > 0. \exists n_0 > 0. \forall n > n_0. f(n) > \delta g(n)\}$

Theorem 2.3 (Master Theorem) $T(n) = aT(n/b) + f(n)$. Where $E = \frac{\log a}{\log b}$, if

- $n^{E+\epsilon} = O(f(n))$ for some $\epsilon > 0$, $T(n) = \Theta(f(n))$ — first level
- $f(n) = \Theta(n^E)$, $T(n) = \Theta(f(n) \log n)$ — every level
- $f(n) = O(n^{E-\epsilon})$ for some $\epsilon > 0$, $T(n) = \Theta(n^E)$ — base level

3 Data Structures

foldl $(\diamond) \epsilon$ is extensionally equivalent to foldr $(\diamond) \epsilon$ when $x \diamond (y \diamond z) = (x \diamond y) \diamond z$, $\epsilon \diamond y = y$ and $x \diamond \epsilon = x$

Definition 3.1 (Boom Hierachy)

1. **Associativity** $(x \cup y) \cup z = x \cup (y \cup z)$ Trees
2. **Identity** $\emptyset \cup x = x = x \cup \emptyset$ Trees, Lists
3. **Commutativity** $x \cup y = y \cup x$ Trees, Lists, Bags
4. **Idempotence** $x \cup x = x$ Trees, Lists, Bags, Sets

Definition 3.2 (Member) Using only $\log(x)+1$ comparisons

```
member x HTip = False
member x (HNode _ lte p gt) | x <= p = go p x lte
                             | otherwise = member x gt

go y x HTip = y <= x
go y x (HNode _ lte p gt) | x <= p = go p x lte
                             | otherwise = go y x gt
```

Definition 3.3 (Difference Lists) $(DList\ fxs) ++ (DList\ fys) = DList\ (fxs.fys)$; $cons\ x\ (DList\ fxs) = DList\ ((x:).fxs)$; $empty = DList\ id$; $toList\ (DList\ fxs) = fxs []$; $fromList\ xs = DList\ (xs++)$

Definition 3.4 (Deque) data Deque $a = Deque\ [a]\ [a]$; $toList\ (Deque\ xs\ sy) = xs ++ reverse\ sy$; $fromList\ xs = Deque\ ys\ (reverse\ zs)$ where $(ys, zs) = splitAt\ (length\ xs\ 'div'\ 2)\ xs$.

Invariant: $isEmpty\ xs \Rightarrow isEmpty\ sy$ or $isSingle\ sy$ and $isEmpty\ sy \Rightarrow isEmpty\ xs$ or $isSingle\ xs$

Definition 3.5 (Random Access List) *toList (RList ts) = (concat . map toList) ts*

```
cons x xs = RList (consTree (Leaf x) xs) where
  consTree t (RList []) = [t]
  consTree t (RList (Tip:ts)) = t:ts
  consTree t (RList (t':ts))
    = Tip:consTree (fork t t') (RList ts)

  RList (t:ts) !! k
    | isEmpty t = RList k
    | k < length t = t !! k
    | otherwise = RList ts !! (k-m)
```

Definition 3.6 (AVL Tree) *data HTree a = HTip | HNode Height (HTree a) a (HTree a)*

```
insert x HTip = hnode HTip x HTip          balance lt y rt
insert x t@(HNode _ lt y rt)
  | x == y = t
  | x < y = balance (insert x lt) y rt
  | otherwise = balancer lt y (insert x rt)

  | height lt - height rt <= 1 = hnode lt y rt
  | otherwise = case lt of HNode _ llt x rlt
    | height llt >= height rlt -> rotr (hnode lt y rt)
    | otherwise -> rotr (hnode (rotr llt) y rt)

rotr (HNode _ (HNode _ p x q) y r) = hnode p x (hnode q y r)
rotr (HNode _ p x (HNode _ q y r)) = hnode (hnode p x q) y r

rotr (HNode _ (HNode _ p x q) y r) = hnode p x (hnode q y r)
rotr (HNode _ p x (HNode _ q y r)) = hnode (hnode p x q) y r
```

fromOrdList xs = fst (go (length xs) xs) where

```
go 0 xs = (HTip, xs)
go n xs = let m = n `div` 2
  (lhs, x:xs') = go m xs
  (rhs, xs'') = go (n-m-1) xs'
  in (HNode (1+max (height lhs, height rhs)) lhs x rhs, xs'')
```

hTreeToList rt = go rt [] where

```
go HTip ks = ks
go (HNode _ lt p gt) ks = go lt (p:go gt ks)
```

insert x DTip = (True, DNode Zero DTip x DTip)

insert x (DNode bl lt y rt) = case compare x y of

```
EQ -> (False, DNode bl lt y rt)
LT -> case insert x lt of
  (False, lt') -> (False, DNode bl lt' y rt)
  (True, lt') -> case bl of
    MinusOne -> (False, DNode Zero lt' y rt)
    Zero -> (True, DNode PlusOne lt' y rt)
    PlusOne -> rotr lt' y rt
```

rotr (DNode PlusOne a y b) x c = (False, DNode Zero a y (DNode Zero b x c))

rotr (DNode Zero a y b) x c = (True, DNode MinusOne a y (DNode PlusOne b x c))

rotr (DNode MinusOne a y (DNode bl b z c)) x d =

```
(False, DNode Zero (DNode (balr bl) a y b) z (DNode (ball bl) c x d))
```

balr PlusOne = Zero; balr Zero = Zero; balr MinusOne = PlusOne

ball PlusOne = MinusOne; ball Zero = Zero; ball MinusOne = Zero

Definition 3.7 (Red Black Tree) *data RBTre a = E | N Colour (RBTre a) a (RBTre a). Every red node must have a black parent and every path from root to leaf must have same number of black nodes.*

insert x t = blacken (go t) where

```
go E = N R E x E
go t@(N c lt y rt)
  | x < y = balance c (go lt) y rt
  | x == y = t
  | otherwise = balance c lt y (go rt)

  blacken (N R lt x rt) = N B lt x rt
  blacken t = t
```

balance B (N R (N R a x b) y c) z d = N R (N B a x b) y (N B c z d)

balance B (N R a x (N R b y c)) z d = N R (N B a x b) y (N B c z d)

balance B a x (N R (N R b y c) z d) = N R (N B a x b) y (N B c z d)

balance B a x (N R b y (N R c z d)) = N R (N B a x b) y (N B c z d)

balance c lt x rt = N c lt x rt

Definition 3.8 (Treap) *data Treap a = Empty | Node (Treap a) a Int (Treap a)*

insert x p Empty = Node Empty x p Empty

insert x p (Node a y q b)

```
| x < y = lnode (insert x p a) y q b
```

```
| x == y = Node a y q b
```

```
| x > y = rnode a y q (insert x p b)
```

```
lnode Empty y q c = Node Empty y q c
```

```
lnode l@(Node a x p b) y q c
```

```
| q <= p = Node l y q c
```

```
| otherwise = Node a x p (Node b y q c)
```

Definition 3.9 (Tries) *data Trie a = (Bool, [a, Trie a])*

```
insert [] (e, ts) = (True, ts)
insert (x:xs) (e,ts) = e, ins ts where
  ins [] = [(x, insert xs (False, []))]
  ins ((y,ys):ts) | x == y = (y, insert xs yt):ts
                  | otherwise = (y,yt): ins ts
```

4 Divide & Conquer

1. Divide a problem into subproblems.
2. Divide and conquer subproblems into subsolutions.
3. Conquer subsolutions into a solution.

Algorithm 4.1 (Merge Sort) *msort [] = []; msort [x] = [x]; msort xs = merge (msort us) (msort vs) where (us,vs) = splitAt ((length xs) 'div' 2) xs; merge (x:xs) (y:ys) = if x ≤ y then x : merge xs (y:ys) else y : merge (x:xs) ys; merge xs [] = xs; merge [] ys = ys;*

Worst Case: $n \log n$. $T_{msort}(n) = T_{length}(n) + T_{splitAt}(\frac{n}{2}) + T_{merge}(\frac{n}{2}) + 2 \times T_{msort} \frac{n}{2}$

Algorithm 4.2 (Quicksort) *Worst Case: n^2 . qsort [] = []; qsort [x] = [x]; qsort (x:xs) = qsort us ++ [x] ++ qsort vs where (us,vs) = partition (≤ x) xs; partition p xs = (filter p xs, filter (not p) xs)*

5 Dynamic Programming

1. Write an inefficient recursive algorithm that solves the problem.
2. Improve inefficiency by storing intermediate shared results.

tabulate :: $I \times i \Rightarrow (i,i) \rightarrow (i \rightarrow a) \rightarrow \text{Array } i \ a$
 tabulate (u,v) f = array (u,v) [(i, f i) | i ← range(u,v)]
 Array starts from index 0

Algorithm 5.1 (Fibonacci Numbers) *fib' n = table ! n where*
table = tabulate (0,n) memo :: Array Int Integer
memo 0 = 0; memo 1 = 1; memo n = table ! (n-1) + table ! (n-2)

Algorithm 5.2 (Edit Distance) *dist xs ys = table ! (m,n) where*
table = tabulate ((0,0),(m,n)) (uncurry memo)
memo i 0 = i; memo 0 j = j;
memo i j = minimum [table ! (i,j-1) + 1, table ! (i-1,j) + 1,
table ! (i-1,j-1) + if (axs!(m-i))==(ays!(n-j)) then 0 else 1]
m = length xs; n = length ys; axs = fromList xs; ays = fromList ys

Algorithm 5.3 (Bitonic Travelling Salesman) *bitonic d n = table ! n where*
table = tabulate (0,n) memo
*memo 0 = Path 0 [(0,0)]; memo 1 = Path (2 * d 0 1) [(0,1),(0,1)]*
memo n = minimum [table ! k - d' (k-1) k + d' (k-1) n + sum[d' i (i+1) | i <- [k..n-1]]
| k <- [1..n-1]]
where d' i j = Path (d i j) [(min i j, max i j)]

6 Amortized Analysis

Definition 6.1 (Amortization) $C_{op_i}(xs_i) \leq A_{op_i}(xs_i) + S(xs_i) - S(xs_{i+1})$

- A cost function $C_{op_i}(xs_i)$ for each operation op_i on data xs_i .
- An amortized cost function $A_{op_i}(xs_i)$ for each operation op_i on data xs_i .
- A size function $S(xs)$ that calculates the size of data xs which increases by a small amount normally and decreases by a large amount on a huge operation.
 where $S(xs)$ increases by a small amount usually but decreases by a large amount on a costly operation
 If $S(xs_0) = 0$ then $\sum_{i=0}^{n-1} C_{op_i}(xs_i) \leq \sum_{i=0}^{n-1} A_{op_i}(xs_i)$

7 Randomized Algorithm

Monte Carlo algorithms have predictable run time but unpredictable results. Las Vegas algorithms vice versa.

Algorithm 7.1 (Randomised Pi) montePi = loop (mkStdGen 42) 1000 0 where
loop seed 0 m = 4 * fromIntegral m / fromIntegral 1000
loop seed n m = let (x,seed') = randomR (0,1) seed
 (y,seed'') = randomR (0,1) seed'
 in loop seed'' (n-1) (if inside (x,y) then m+1 else m)

Algorithm 7.2 (Randomised Treap) data RTreap a = RTreap StdGen (Treap a)

insert' x (RTreap seed t) = RTreap seed' (insert x p t) where (p,seed') = random seed
rquicksort xs = toList' (fromList' xs)

8 Mutable Data Structures

To get STArray from list, axs ← newListArray (0, length xs - 1) xs

Algorithm 8.1 (Checklist) minfree xs = length (takeWhile id (checklist xs))
checklist xs = runST \$ do
 ays <- newArray (0,length xs - 1) False :: ST s (STArray s Int Bool)
 sequence [writeArray ays x True | x <- xs, x < length xs]
 getElem ays

Algorithm 8.2 (Quicksort) qsort xs = runST \$ do
 axs <- newListArray (0,(length xs - 1)) xs
 aqsort axs 0 (length xs - 1)
 getElem axs
aqsort axs i j | i >= j = return ()
 | otherwise = do
 k <- apartition axs i j
 aqsort axs i (k-1)
 aqsort (k+1) j
apartition axs p q = do
 x <- readArray axs p
 let loop i j | i > j = do swap axs p j
 return j
 | otherwise = do
 u <- readArray axs i
 if u < x then loop (i+1) j
 else do swap axs i j
 loop i (j-1)
 in loop (p+1) q

Algorithm 8.3 (Quickselect) qselect k xs = runST \$ do
 axs <- newListArray (0,(length xs-1)) xs
 mx <- aqselect k axs 0 (length xs-1)
 return mx
aqselect k axs i j | i > j = return Nothing
 | otherwise = do
 p <- apartition axs i j
 if | k < p -> aqselect k axs i (p-1)
 | k > p -> aqselect k (p+1) j
 | otherwise -> do x <- readArray axs p
 return x