

Programming in Haskell

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Lecture II

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Measuring efficiency

- Computation is reduction
 - Application of definitions as rewriting rules
- Count the number of reduction steps
 - Running time is $T(n)$ for input size n

Example: Complexity of ++

- $[] ++ y = y$
 $(x:xs) ++ y = x:(xs++y)$
- $[1,2,3] ++ [4,5,6] \Rightarrow$
 $1:([2,3] ++ [4,5,6]) \Rightarrow$
 $1:(2:([3] ++ [4,5,6])) \Rightarrow$
 $1:(2:(3:([] ++ [4,5,6]))) \Rightarrow$
 $1:(2:(3:([4,5,6])))$
- $l_1 ++ l_2$: use the second rule length l_1 times, first rule once, always

Example: elem

- `elem :: Int -> [Int] -> Bool`
`elem i [] = False`
`elem i (x:xs)`
 - | `(i==x) = True`
 - | `otherwise = elem i xs`
- `elem 3 [4,7,8,9] ⇒ elem 3 [7,8,9] ⇒`
`elem 3 [8,9] ⇒ elem 3 [9] ⇒ elem 3 [] ⇒ False`
- `elem 3 [3,7,8,9] ⇒ True`
- Complexity depends on input size and value

Variation across inputs

- Worst case complexity
 - Maximum running time over all inputs of size n
 - Pessimistic: may be rare
- Average case
 - More realistic, but difficult/impossible to compute

Asymptotic complexity

- Interested in $T(n)$ in terms of orders of magnitude
- $f(n) = O(g(n))$ if there is a constant k such that $f(n) \leq kg(n)$ for all $n > 0$
 - $an^2 + bn + c = O(n^2)$ for all a, b, c
(take $k = a + b + c$ if $a, b, c > 0$)
- Ignore constant factors, lower order terms
 - $O(n)$, $O(n \log n)$, $O(n^k)$, $O(2^n)$, ...

Asymptotic complexity ...

- Complexity of ++ is $O(n)$, where n is the length of the first list
- Complexity of elem is $O(n)$
 - Worst case!

Complexity of reverse

- `myreverse :: [a] -> [a]`
`myreverse [] = []`
`myreverse (x:xs) = (myreverse xs) ++ [x]`
- Analyze directly (like `++`), or write a recurrence for $T(n)$
 - $T(0) = 1$
 $T(n) = T(n-1) + n$
- Solve by expanding the recurrence

Complexity of reverse ...

$$T(n) = T(n-1) + n$$

$$= (T(n-2) + n-1) + n$$

$$= (T(n-3) + n-2) + n-1 + n$$

...

$$= T(0) + 1 + 2 + \dots + n$$

$$= 1 + 1 + 2 + \dots + n = 1 + n(n+1)/2$$

$$= O(n^2)$$

$$T(0) = 1$$

$$T(n) = T(n-1) + n$$

Speeding up reverse

- Can we do better?
- Imagine we are reversing a heavy stack of books
- Transfer to a new stack, top to bottom
- New stack is in reverse order!

Speeding up reverse ...

- `transfer :: [a] -> [a] -> [a]`
`transfer [] l = l`
`transfer (x:xs) l = transfer xs (x:l)`
- Input size for `transfer l1 l2` is length `l1`
- Recurrence
 - $T(0) = 1$
 $T(n) = T(n-1) + 1$
- Expanding: $T(n) = 1 + 1 + \dots + 1 = O(n)$

Speeding up reverse ...

- `fastreverse :: [a] -> [a]`
`fastreverse l = transfer l []`
- Complexity is $O(n)$
- Need to understand the computational model to achieve efficiency

Summary

- Measure complexity in Haskell in terms of reduction steps
- Account for input size and values
 - Usually worst-case complexity
- Asymptotic complexity
 - Ignore constants, lower order terms
 - $T(n) = O(f(n))$

Sorting

- Goal is to arrange a list in ascending order
- How would we sort a hand of cards?
 - A single card is sorted
 - Put second card before/after first
 - “Insert” third, fourth,... card in correct place
- Insertion sort

Insertion sort : insert

- Insert an element in a sorted list
- `insert :: Int -> [Int] -> [Int]`
`insert x [] = [x]`
`insert x (y:ys)`
 - | `(x <= y) = x:y:ys`
 - | `otherwise y:(insert x ys)`
- Clearly $T(n) = O(n)$

Insertion sort : *isort*

- `isort :: [Int] -> [Int]`
`isort [] = []`
`isort (x:xs) = insert x (isort xs)`
- Alternatively
- `isort = foldr insert []`
- Recurrence
 - $T(0) = 1$
 $T(n) = T(n-1) + O(n)$
- Complexity: $T(n) = O(n^2)$

A better strategy?

- Divide list in two equal parts
- Separately sort left and right half
- Combine the two sorted halves to get the full list sorted

Combining sorted lists

- Given two sorted lists l_1 and l_2 , combine into a sorted list l_3
 - Compare first element of l_1 and l_2
 - Move it into l_3
 - Repeat until all elements in l_1 and l_2 are over
- Merging l_1 and l_2

Merging two sorted lists

~~32~~ ~~74~~ ~~89~~

~~21~~ ~~55~~ ~~64~~

21

32

55

64

74

89

Merge Sort

- Sort $l[0]$ to $l[n/2-1]$
- Sort $l[n/2]$ to $l[n-1]$
- Merge sorted halves into l'
- How do we sort the halves?
 - Recursively, using the same strategy!

Merge Sort

13	22	32	43	57	63	78	91
---------------	---------------	---------------	---------------	---------------	---------------	---------------	---------------

22	32	43	78
---------------	---------------	---------------	---------------

63	57	63	91
---------------	---------------	---------------	---------------

32	43
---------------	---------------

22	78
---------------	---------------

57	63
---------------	---------------

91	91
---------------	---------------

43

32

22

78

63

57

91

13

Merge sort : merge

- `merge :: [Int] -> [Int] -> [Int]`
`merge [] ys = ys`
`merge xs [] = xs`

```
merge (x:xs) (y:ys)
  | x <= y    = x:(merge xs (y:ys))
  | otherwise = y:(merge (x:xs) ys)
```

- Each comparison adds one element to output
- $T(n) = O(n)$, where n is sum of lengths of input lists

Merge sort

- ```
mergesort :: [Int] -> [Int]
mergesort [] = []
mergesort [x] = [x]
mergesort l = merge (mergesort (front l))
 (mergesort (back l))

where
front l = take ((length l) `div` 2) l
back l = drop ((length l) `div` 2) l
```

# Analysis of Merge Sort

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- $T(n)$ : time taken by Merge Sort on input of size  $n$ 
  - Assume, for simplicity, that  $n = 2^k$
  - $T(n) = 2T(n/2) + cn$
  - Two subproblems of size  $n/2$
  - Splitting the list into front and back takes  $n$  steps
  - Merging solutions requires time  $O(n/2 + n/2) = O(n)$
- Solve the recurrence by unwinding

# Analysis of Merge Sort ...

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- $T(1) = 1$
- $T(n) = 2T(n/2) + cn$   
 $= 2 [ 2T(n/4) + cn/2 ] + cn = 2^2 T(n/2^2) + 2cn$   
 $= 2^2 [ 2T(n/2^3) + cn/2^2 ] + 2cn = 2^3 T(n/2^3) + 3cn$   
...  
 $= 2^j T(n/2^j) + cjn$
- When  $j = \log n$ ,  $n/2^j = 1$ , so  $T(n/2^j) = 1$
- $T(n) = 2^j T(n/2^j) + cjn = 2 \log n + 2(\log n) n = n + 2n \log n = O(n \log n)$

# Avoid merging

---

- Some elements in left half move right and vice versa
- Can we ensure that everything to the left is smaller than everything to the right?
- Suppose the median value in list is  $m$ 
  - Move all values  $\leq m$  to left half of list
  - Right half has values  $> m$
- Recursively sort left and right halves
- List is now sorted! No need to merge

# *Avoid merging ...*

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- How do we find the median?
  - Sort and pick up middle element
  - But our aim is to sort!
- Instead, pick up some value in list — pivot
  - Split list with respect to this pivot element

# Quicksort

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- Choose a **pivot** element
  - Typically the first value in the list
- Partition list into lower and upper parts with respect to pivot
- Move pivot between lower and upper partition
- Recursively sort the two partitions

# Quicksort

---

|    |    |    |    |    |    |    |    |
|----|----|----|----|----|----|----|----|
| 43 | 32 | 32 | 48 | 63 | 63 | 98 | 98 |
|----|----|----|----|----|----|----|----|

# Quicksort

---

- ```
quicksort :: [Int] -> [Int]
quicksort [] = []
quicksort (x:xs) = (quicksort lower) ++
                  [splitter] ++
                  (quicksort upper)

where
splitter = x
lower    = [ y | y <- xs, y <= x ]
upper    = [ y | y <- xs, y > x ]
```

Analysis of Quicksort

- Worst case
- Pivot is maximum or minimum
 - One partition is empty
 - Other is size $n-1$
 - $T(n) = T(n-1) + n = T(n-2) + (n-1) + n$
 $= \dots = 1 + 2 + \dots + n = O(n^2)$
- Already sorted array is worst case input!

Analysis of Quicksort

- But ...
- Average case is $O(n \log n)$
 - Sorting is a rare example where average case can be computed
- What does average case mean?

Quicksort: Average case

- Assume input is a permutation of $\{1, 2, \dots, n\}$
 - Actual values not important
 - Only relative order matters
 - Each input is equally likely (uniform probability)
- Calculate running time across all inputs
- Expected running time can be shown $O(n \log n)$

Summary

- Sorting is an important starting point for many functions on lists
- Insertion sort is a natural inductive sort whose complexity is $O(n^2)$
- Merge sort has complexity $O(n \log n)$
- Quicksort has worst-case complexity $O(n^2)$ but average-case complexity $O(n \log n)$