

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 ++

- $[] \text{ ++ } y = y$
 $(x:xs) \text{ ++ } y = x:(xs \text{ ++ } y)$
- $[1,2,3] \text{ ++ } [4,5,6] \Rightarrow \Rightarrow$
 $1:([2,3] \text{ ++ } [4,5,6]) \Rightarrow \Rightarrow$
 $1:(2:([3] \text{ ++ } [4,5,6])) \Rightarrow \Rightarrow$
 $1:(2:(3:([] \text{ ++ } [4,5,6]))) \Rightarrow \Rightarrow$
 $1:(2:(3:([4,5,6])))$
- $l_1 \text{ ++ } l_2$: 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] \Rightarrow elem 3 [7,8,9] \Rightarrow`
`elem 3 [8,9] \Rightarrow elem 3 [9] \Rightarrow elem 3 [] \Rightarrow False`
- `elem 3 [3,7,8,9] \Rightarrow 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 ...

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

$$T(0) = 1$$

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

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

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

...

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

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

$$= O(n^2)$$

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
 - $T(o) = I$
 $T(n) = T(n-1) + I$
- Expanding: $T(n) = I + I + \dots + I = O(n)$

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*

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- `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?*

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- 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*

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- Given two sorted lists **l1** and **l2**, combine into a sorted list **l3**
  - Compare first element of **l1** and **l2**
  - Move it into **l3**
  - Repeat until all elements in **l1** and **l2** are over
- Merging **l1** and **l2**

# *Merging two sorted lists*

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~~32~~      ~~74~~      ~~89~~

~~21~~      ~~55~~      ~~64~~

21

32

55

64

74

89

# Merge Sort

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- 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

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

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

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

|    |    |
|----|----|
| 32 | 43 |
|----|----|

|    |    |
|----|----|
| 22 | 78 |
|----|----|

|    |    |
|----|----|
| 63 | 63 |
|----|----|

|    |    |
|----|----|
| 91 | 91 |
|----|----|

|    |
|----|
| 43 |
|----|

|    |
|----|
| 32 |
|----|

|    |
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| 22 |
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| 78 |
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| 63 |
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| 57 |
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|    |
|----|
| 91 |
|----|

|    |
|----|
| 13 |
|----|

# Merge sort : merge

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- `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

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- ```
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

- $T(n)$: 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 ...

- $T(1) = 1$
- $T(n) = 2T(n/2) + cn$
 $= 2 [2T(n/4) + cn/2] + cn = 2^2 T(n/2^2) + 2cn$
 $= 2^2 [2T(n/2^3) + cn/2^2] + 2cn = 2^3 T(n/2^3) + 3cn$
...
 $= 2^j T(n/2^j) + cjn$
- 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...

- 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

- 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

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- 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*

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- 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

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- 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

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- 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  $O(n \log n)$