

ECE 235: Growth of Functions, Data structures, Thu 9/27/01 (Ch. 3,10)

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This lecture:

- big O , big Ω , and big Θ , how they grow
- fundamental data structures

1 Asymptotic Notations

- purpose: express run time as $T(n)$, n = input size
- assumption: $n \in \mathbb{N} = \{0, 1, 2, \dots\}$ natural numbers

1.1 big O = asymptotic upper bound: (page 44)

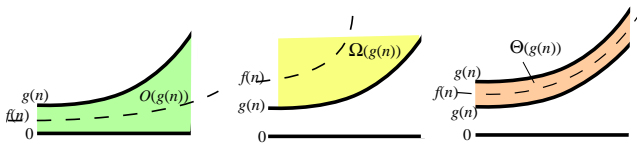
- $f(n) = O(g(n))$ is a set of functions that are upper-bounded by $0 \leq f(n) \leq c \cdot g(n)$ for some $n \geq n_0$
- $\Rightarrow f(n)$ grows no faster than $g(n)$
- example: $2n^2 = O(n^3)$, for $c = 1, n_0 \geq 2$
- funny equality, more like a membership: $2n^2 \in O(n^3)$
- n^2 and n^3 are actually functions, not values.
 \Rightarrow sloppy but convenient

1.2 big Ω = asymptotic lower bound (page 45)

- swap the place between $f(n)$ and $c \cdot g(n)$:
 $0 \leq c \cdot g(n) \leq f(n)$ for some $n \geq n_0$
- example: $\sqrt{n} = \Omega(\lg n)$, for $c = 1, n_0 \geq 16$

1.3 big Θ = asymptotic tight bounds

- need two constants c_1 and c_2 to sandwich $f(n)$.
- example: $\frac{1}{2}n^2 - 2n = \Theta(n^2)$ for $c_1 = \frac{1}{4}, c_2 = \frac{1}{2}, n_0 = 8$
- Theorem: (O and Ω) $\Leftrightarrow \Theta$



Example: insertion sort

$O(n^2)$ worst case, $\Omega(n)$ best case

$$\begin{aligned} T(n) = & c_1 \cdot n && // \text{for loop} \\ & + c_2 \cdot (n-1) && // \text{key} \leftarrow A \\ & + \sum_{j=2}^n \text{Insert}(j) && // \text{loop to insert } j \\ & + c_8 \cdot (n-1) && // A[i+1] \leftarrow \text{key} \end{aligned}$$

- Best case of $\text{Insert}(j)$ is $\Theta(1)$, Worst is $\Theta(j)$
- Issue: $\Theta(n^2) + \Theta(n)$ is still $\Theta(n^2)$

Example: MergeSort(A, p, r)

where p, r are the lower and upper bound indexes to array A .

if $p < r$ then

$$q := \lfloor (p+r)/2 \rfloor$$

MergeSort(A, p, q), MergeSort($A, q+1, r$)

Merge(A, p, q, r)

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ 2 \cdot T(\frac{n}{2}) + \Theta(n) & \text{if } n > 1 \end{cases}$$

$T(n)$ is expressed in terms of T itself! This is called *recurrence*. We want a closed-form solution (no T on the right-hand-side of the equation).

To solve recurrence:

- rewrite as $T(n) = \begin{cases} c & \text{if } n = 1 \\ 2 \cdot T(\frac{n}{2}) + c \cdot n & \text{if } n > 1 \end{cases}$

$$T(n) = 2T(\frac{n}{2}) + cn \quad \text{Expand}$$

$$= 2(2T(\frac{n}{4}) + c \cdot \frac{n}{2}) + cn$$

- $= 4T(\frac{n}{4}) + 2c \cdot \frac{n}{2} + cn = 4T(\frac{n}{4}) + 2cn$

$$= 8T(\frac{n}{8}) + 3cn = \dots$$

$$= (2^i)T(\frac{n}{2^i}) + i \cdot cn$$

- We can divide n by 2 at most $\lg n$ times before $n = 1$. So,
 $T(n) = (\lg n) \cdot cn + cn = \Theta(n \lg n)$.

Chapter 4 will show how to solve recurrence systematically.

1.4 little- o , little- ω not asymptotically tight

$$o(g(n)) = \{f(n) : \forall c > 0 \exists n_0 > 0 : 0 \leq f(n) < c g(n) \forall n \geq n_0\}$$

$$\omega(g(n)) = \{f(n) : \forall c > 0 \exists n_0 > 0 : 0 \leq c g(n) < f(n) \forall n \geq n_0\}$$

Example

- $2n^2 = O(n^2)$ is a tight bound $\Rightarrow 2n^2 \neq o(n^2)$.
- $2n = O(n^2)$ is not tight $\Rightarrow 2n = o(n^2)$.
- does not make sense to have little- θ

2 Polynomials, log, factorial, fibonacci

2.1 Polynomial in n of degree d :

$$p(n) = \sum_{i=0}^d a_i n^i = \Theta(n^d)$$

Polynomially bounded: $f(n) = O(n^k)$ for constant k .

2.2 logarithm

- useful identity: $a^{\log_b c} = c^{\log_b a}$.
- polylogarithmically bounded if $f(n) = O(\lg^k n)$ for constant k
- grows slower than any polynomial: $\lg^b n = o(n^a)$ for any constant $a > 0$.
- “log-star”: $\lg^* n = \min\{i \geq 0 : \lg^{(i)} n \leq 1\}$
(log of log of log ... of log of n) grows very very slowly.
Example: $\lg^*(2^{65536}) = 5$.

2.3 Factorial:

- Loose bound: $n!$ is faster than 2^n but slower than n^n
- precisely, $n! = \Theta(n^{n+1/2} e^{-\lg n})$ (Stirling’s approximation)
- $\lg(n!) = \Theta(n \lg n)$

2.4 Fibonacci

grows exponentially

2.5 misc

- 2^{2^n} grows faster than $n!$
- $(n+1)!$ grows faster than $n!$
- $n!$ grows faster than $n \cdot 2^n$ faster than 2^n
- $2^{\lg n} = n$

3 Python: Functions

- function definition: **def** functionName (paramList) :
- variables have local scope by default
- control flow: **for**, **while**, **if**
- range(a, b) function: returns the list $[a, \dots, b-1]$
>>> range(1, 5)
[1, 2, 3, 4]

Algorithm from book (p.24)	Python code (file isort.py)
INSERTION-SORT(A)	def InsertionSort(A):
1 for $j \leftarrow 2$ to length[A]	for j in range(1, len(A)):
2 do $key \leftarrow A[j]$	$key = A[j]$
3 $i \leftarrow j - 1$	$i = j - 1$
4 while $i > 0$ and $A[i] > key$	while ($i \geq 0$) and ($A[i] > key$):
5 do $A[i+1] \leftarrow A[i]$	$A[i+1] = A[i]$
6 $i \leftarrow i - 1$	$i = i - 1$
7 $A[i+1] \leftarrow key$	$A[i+1] = key$

- Don’t forget the colon : for **def**, **for**, and **while** constructs
- Book $A[1 \dots n] \Rightarrow$ Python $A[0 \dots n-1]$
Book’s $2 \dots n \Rightarrow$ Python $1 \dots n-1$.
- line 1: $\text{range}(1, \text{len}(A)) \Rightarrow [1 \dots \text{len}(A)-1]$

```
% python
>>> from isort import *    # read file isort.py
>>> x = [2,7,3,8,1]        # create test case
>>> InsertionSort(x)       # call routine
>>> x                      # look at result
[1, 2, 3, 7, 8]
```

Works for other data types, too!

4 Data Structures Review

Concepts:

- Stacks: Last-in First-out
- Queue: First-in First-out
- Trees: Root, children

4.1 Array implementation

Stack S : need a stack pointer (top)

```
Push(x):  S[++top] ← x
Pop():    return S[top--]
```

Note: Actually stacks can grow downward also.
Note: To be safe, check stack underflow/overflow.

Queue Q : need 2 pointers ($head$, $tail$)

```
Enqueue(x): Q[tail] ← x
            tail ← (tail mod len(Q)) + 1
Dequeue():  temp ← Q[head]
            head ← (head mod len(Q)) + 1
            return temp
```

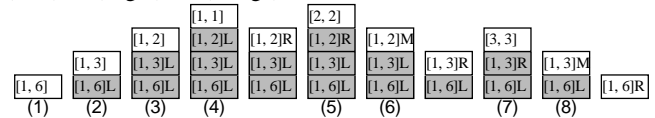
Note: To be safe, check queue underflow/overflow.

4.2 Stack and Calls/Recursion

- Calling routine \Rightarrow Push local, param, return address
- Return \Rightarrow Pop and continue

Example: MergeSort (cf. MergeSort Fig. from last lecture)

L (Left), R (Right), M (Merge)



4.3 Linked List x

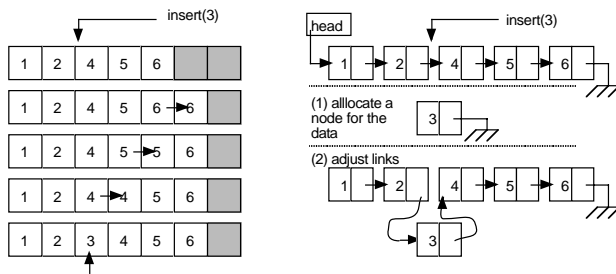
- Singly linked (next) or Doubly linked (next, prev)
- need a key
- need a NIL value

How to implement linked lists?

- Pre-allocate arrays $prev[1:n]$, $next[1:n]$, $key[1:n]$, and x is an index $\in \{0 \dots n\}$, where 0 is NIL.
- Records (structures in C, classes in Java): $x.prev$, $x.next$, $x.key$, and NULL for NIL.

Rearrange by adjusting links, not moving data (key)

- Recall: INSERTIONSORT with array need to copy object to insert.
- Linked list representation: once we know where to insert, adjust link in $O(1)$ time.
- Array version: best case $O(1)$, worst case $O(n)$, average case $O(\frac{n}{2}) = O(n)$.

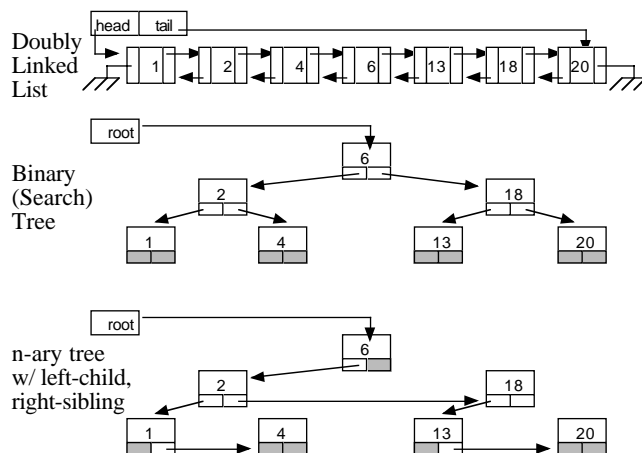


Search for the item

- e.g., where to insert?
- list is *unsorted*, worst/average case $O(n)$ time to find it! (compare keys one at a time).
- list is *sorted*:
 - Array: binary search in $O(\lg n)$ times
 - Singly linked list: $O(n)$ avg/worst case, need to do linear search
 - Doubly linked list: does not help! still $O(n)$ – can't jump to the middle of the list, because link is to the head/tail.

4.4 Binary search tree

- pointer to root; each node has pointer to left-child, right-child
- optional pointer to parent
- left \Rightarrow smaller, right \Rightarrow bigger key value
- $O(\lg n)$ search, if BALANCED
- Can still be $O(n)$ if not balanced!!
 \Rightarrow Red-black tree (Ch.13) is a way to keep BST balanced
- a *Heap* (Ch.6) implements balanced BST using array; no pointers



4.5 rooted trees w/ unbounded branches

Could have arbitrary number of children!

- left-child: head of linked list to children (down one level)
- right-sibling: link to next on same level
- optional link back to parent