



Data Structures and Data Management

CS 240



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Preface

Disclaimer Much of the information on this set of notes is transcribed directly/indirectly from the lectures of CS 240 during Spring 2019 as well as other related resources. I do not make any warranties about the completeness, reliability and accuracy of this set of notes. Use at your own risk.

This set of notes is quite incomplete. I recompiled the old tex code using the new template. I will most likely make a new set of notes sooner or later. Stay tuned!

For any questions, send me an email via <https://notes.sibeliusp.com/contact>.

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Sibelius Peng

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May 7

Problem count positive integers in an array.

An Instance [-5, 10, -5, 20]

The Solution 2

Size of the input length of the array

```
Count(A) // A is an array of length n
res = 0
for i = 0 ... n-1
  if A[i] > 0
    res++
return res
```

1.1 Order Notation

Example $f(n) = 2n^2 + 3n + 11$ $g(n) = n^2$

Proof For $n \geq 1$,

$$\begin{aligned} 2n^2 &\leq 2n^2 \\ 3n &\leq 3n^2 \implies f(n) \leq 16n^2 \\ 11 &\leq 11n^2 \end{aligned}$$

Taking $c = 16, n_0 = 1$ this proves that $f(n) \in O(n^2)$

May 9

$$f(n) = 75n + 500, g(n) = 5n^2?$$

Proof

1. For $n \geq 20, 100n \leq 5n^2$

2. For $n \geq 20, 500 \leq 25n$

So if $n \geq 20, f(n) = 500 + 75n \leq 25n + 75n \leq 5n^2 = g(n)$. Since also $f(n) \geq 0$ for all n , taking $n_0 = 20$ and $c = 1$, this proves $f(n) \in O(g(n))$

Another Proof for $n \geq 1, 75n \leq 75n^2, 500 \leq 500n^2$

$$f(n) \leq 575n^2 = 115g(n)$$

So taking $n_0 = 1, c = 115$, this proves $f(n) \in O(g(n))$

Prove that $f(n) = 2n^2 + 3n + 11 \in \Omega(n^2)$ from first principles.

Proof

$$2n^2 \geq 2n^2$$

$$3n \geq 0$$

$$11 \geq 0$$

$f(n) \geq 2n^2 = 2g(n)$. Taking $n_0 = 1, c = 2$, this completes the proof. ($n_0 = 1, c = 1$ work as well)

Prove that $\frac{1}{2}n^2 - 5n \in \Omega(n^2)$ from first principles.

Proof For $n \geq 20, n^2 \geq 20n$, then $-5n \geq \frac{-1}{4}n^2$

$$\text{add } \frac{1}{2}n^2, \quad \frac{1}{2}n^2 - 5n \geq \frac{1}{2}n^2 - \frac{1}{4}n^2 = \frac{1}{4}g(n)$$

$$f(n) \geq \frac{1}{4}g(n)$$

So taking $n_0 = 20, c = \frac{1}{4}$, this completes the proof.

Prove that $\log_b(n) \in \Theta(\log n)$ for all $b > 1$ from first principles.

Proof

$$f(n) = \frac{\log n}{\log b} = \frac{g(n)}{\log b}$$

$$\frac{g(n)}{\log b} \leq f(n) \leq \frac{g(n)}{\log b}$$

Taking $n_0 = 1, c_1 = c_2 = \frac{1}{\log b}$, this completes the proof.

Example $f(n) = 2000n^2, g(n) = n^n$.

Given $c > 0$, we have to find n_0 , (depend on c), such that for $n \geq n_0, |f(n)| < |cg(n)| \iff 2000n^2 < cn^n$ (*)

(*) is equivalent to $2000 < cn^{n-2}$

1. For $n \geq 3, n-2 \geq 1$, so $n^1 \leq n^{n-2}$

2. For $n \geq 3$ and $n \geq \frac{2000}{c} + 1$

$$\frac{2000}{c} < \frac{2000}{c} + 1 \leq n \leq n^{n-2}$$

So taking $n_0 = \max(3, \frac{2000}{c} + 1)$, this proves $f(n) \in o(g(n))$

Example Let $f(n)$ be a polynomial of degree $d \geq 0$,

$$f(n) = c_d n^d + c_{d-1} n^{d-1} + \dots + c_1 n + c_0$$

for some $c_d > 0$, prove $f(n) \in \Theta(n^d)$

Proof Then

$$\frac{f(n)}{g(n)} = \frac{c_d n^d + c_{d-1} n^{d-1} + \dots + c_1 n + c_0}{n^d} = c_d + c_{d-1} \frac{1}{n} + \dots + \frac{c_0}{n^d}$$

Then $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$ exists, and is equal to

$$c_d + 0 + \dots + 0 = c_d > 0$$

By the limit test, $f(n) \in \Theta(g(n))$

Example Prove that $n(2 + \sin n\pi/2)$ is $\Theta(n)$. Note that $\lim_{n \rightarrow \infty} (2 + \sin n\pi/2)$ does not exist.

Proof for $n \geq 1, -1 \leq \sin n\pi/2 \leq 1 \dots n \leq f(n) \leq 3n$. So taking $n_0 = 1, c_1 = 1, c_2 = 3$, this completes the proof.

On the other hand,

$$\frac{f(n)}{g(n)} = 2 + \sin n\pi/2 \quad \text{has no limit at } n = \infty$$

the limit test does not apply

May 14

Example 3 $f(n) = \log(n) = \frac{\ln n}{\ln 2} \rightarrow f'(n) = \frac{1}{\ln 2 \cdot n}$ $g(n) = n \rightarrow g'(n) = 1$

So

$$\lim_{n \rightarrow \infty} \frac{f'}{g'} = 0 \implies \lim_{n \rightarrow \infty} \frac{f}{g} = 0 \implies f(n) \in o(g(n))$$

$$\begin{aligned} f(n) &= \log n \rightarrow f'(n) = \frac{1}{\ln n} \cdot \frac{1}{n} \\ g(n) &= n^a \end{aligned}$$

$$\implies \frac{f'}{g'} = \frac{1}{\ln 2} \frac{1}{a} \frac{1}{n^a}$$

As before, limit $f'/g' = 0 \implies$ limit $f/g = 0$. Therefore $f(n) \in o(g(n))$

$$f(n) = (\log n)^c, \quad g(n) = n^d$$

$$\frac{f}{g} = \left(\frac{\log n}{n^{d/c}} \right)^c$$

Taking $a = \frac{d}{c}$, we saw that $\lim_{n \rightarrow \infty} \frac{\log n}{n^{d/c}} = 0$, so $\lim f/g = 0$. So $f(n) \in o(g(n))$

3.1 Algorithm Analysis

Test1(n)

1. sum \leftarrow 0
2. for i \leftarrow 1 to n do
3. for j \leftarrow i to n do
4. sum \leftarrow sum + (i-j)²
5. return sum

Let $T_1(n)$ be the runtime of Test1(n). Then $T_1(n) \in \Theta(S_1(n))$ where $S_1(n)$ is the number of time we enter Step4.

$$S_1(n) = \sum_{i=1}^n \sum_{j=1}^n 1$$

1. $\sum_{j=1}^n 1 = n - i + 1$

2. So

$$S_n = \sum_{i=1}^n (n - i + 1) = \sum_{i=1}^n n - \sum_{i=1}^n i + \sum_{i=1}^n 1 = n^2 - \frac{n(n+1)}{2} + 2 = \frac{1}{2}n^2 + \frac{1}{2}n \in \Theta(n^2)$$

So $T_1(n) \in \Theta(n^2)$

3.2 two strategies

Test2(A, n)

1. max \leftarrow 0
2. for i \leftarrow 1 to n do
3. for j \leftarrow i to n do
4. sum \leftarrow 0
5. for k \leftarrow i to j do
6. sum \leftarrow sum + A[k]
7. max \leftarrow max (max, sum)
8. return max

Insertion sort: sorting A in a descending order.

Worst case A sorted in increasing order.

Then for all i , A[i] goes to order o in i steps \rightarrow worst case runtime $\Theta(\sum_{i=1}^n i) = \Theta(n^2)$.

Best Case A sorted in decreasing order.

Then for all i , we exit the while loop immediately \rightarrow best case runtime $\Theta(\sum_{i=1}^n 1) = \Theta(n)$

4

May 16

$$T(n) = 2T\left(\frac{n}{2}\right) + cn, \quad n > 1 \quad (*)$$
$$T(1) = c$$

$$\begin{aligned} n = 2^k \rightarrow T(2^k) &= 2T(2^{k-1}) + c2^k = 2(2T(2^{k-2}) + c2^{k-1}) + c2^k && \text{by } (*) \\ &= 2^2T(2^{k-2}) + 2c2^k \\ &= 2^2(2T(2^{k-3}) + c2^{k-2}) + 2c2^k && \text{by } (*) \\ &= 2^3T(2^{k-3}) + 3c2^k \\ &= 2^4T(2^{k-4}) + 4c2^k \\ &= \dots = 2^kT(2^{k-k}) + kc2^k \\ &= 2^kT(1) + kc2^k = c2^k(k+1) \end{aligned}$$

Since $n = 2^k$, $\log n = k$

$$T(2^k) = c2^k(k+1)$$

$$T(n) = cn(\log n + 1)$$

May 21

Insert(A, k)

- if A is full, double its size
- copy k into A

$$\text{cost of insert, if length}(A)=n \begin{cases} 1 & \text{copy if A not full} \\ 1+n & \text{new key + doubling, otherwise} \end{cases}$$

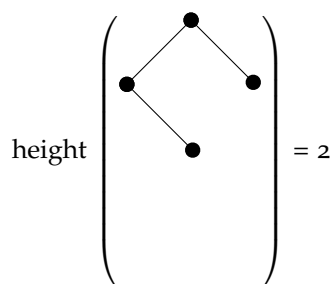
Suppose we start with $\text{length}(A)=1$. Total cost of n inserts (n a power of 2) is

$$\underbrace{1+1+\dots+1}_n \text{ (new key)} + \underbrace{1+2+4+8+\dots+n}_{\text{doubling}} = n + 2n - 1 = 3n - 1$$

5.1 Binary heaps

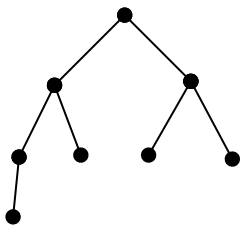
height of binary tree is length of the longest path from the root to a node.

$\text{height}(\bullet)=0$

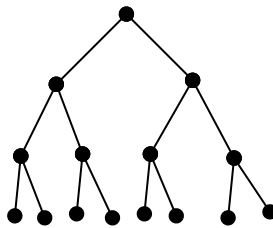


$\text{height}(\emptyset) = -1$

number of nodes in a heap of height 3



at least 8 nodes



at most 15 nodes

true for any binary tree

$$8 \leq n \leq 15 \text{ if } h = 3$$

$$2^h \leq n \leq 2^{h+1} - 1 \text{ any } h \quad \text{true for any binary trees}$$

$$h \leq \log n \text{ and } h \geq \log n + 1$$

Number of nodes in a heap of height h is

- at least

$$1 + 2 + 4 + \dots + 2^{h-1} + 1 = 2^h$$

- is at most $1 + \dots + 2^h = 2^{h+1} - 1$

May 23

recursive_heapify(T, n)

1. if $n = 1$, return
2. recursive_heapify (left child of T, # elements in left child)
3. recursive_heapify (right child, # in right)
4. fix down the root

Jun 18

7.1 Proof for slide 2 mod 6

Lower bound for search in a dictionary of size n , with keys k_1, \dots, k_n , values v_1, \dots, v_n . We count, comparisons between input key k and k_i 's. (comparisons can be $<$, $>$ or $=$).

The decision tree associated to a given search algorithm in size n has $n + 1$ leaves. $\begin{cases} (v_1, \dots, v_n) \\ \text{"not found"} \end{cases}$

$$n + 1 = \# \text{ leaves} \leq \# \text{ nodes} \leq 2^{h+1} - 1$$

$$\implies h \geq \log(n + 1) - 1$$

(and the height h is the most case # comparisons for this algorithm)

Suppose $A[0]$ and $A[n - 1]$ are fixed $A[1] \dots A[n - 2]$ chosen uniformly at random in $\{A[0] \dots A[n - 1]\}$

Can prove to interpolation search in an array of length n with probability $\geq 1/4$, we do a recursive call in length $\leq \sqrt{n}$

$$\implies T^{avg}(n) \leq c + \frac{1}{4}T^{avg}(\sqrt{n}) + \frac{3}{4}T^{avg}(n)$$