BST 234 - Lab 2

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Run time complexity

$$f \in O(g) : \exists c > 0, \exists n_0 \text{ s.t. } \forall n > n_0 : f(n) \le c * g(n)$$

Intuitively this means f(n) is O(g(n)) if f(n) grows at most as fast as some constant times g(n) for large n.

This is read as ''f(x) is big-Oh of g(x)" or ''g asymptotically dominates f."

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Properties

- f= O(f)
- O(O(f)) = O(f)
- kO(f) = O(f)
- O(f + k) = O(f)
- $\bullet \ O(f) + O(g) = O(max(f , g))$
- O(f) * O(g) = O(f * g)

Example 1

Show that:

$$n^2$$
 is $notO(n)$

Solution: Suppose \exists constants c and n_o for which:

$$n^2 \le cn, \forall n > n_o$$

by *dividing* both sides of the inequality *by n*, then $n \le c$ must hold $\forall n > n_0$.

A contradiction!



Example 2

Show that:

$$2^{n+1}=O(2^n)$$

Solution:

$$2^{n+1} \le 2^n * 2, \forall n \ge 0$$

 $\therefore 2^{n+1} = O(2^n)$

Example 3

Show that:

$$2^{2n}$$
 is not $O(2^n)$

Solution: If the above was true, there would exist n_0 and c such that $n > n_0$:

$$2^n * 2^n = 2^{2n} \le c * 2^n$$

, so $2^n \le c$ for $n > n_0$ which is clearly impossible since c is a constant.



Run - Time Complexity - Alternate Definition

Assume $g(n) \neq 0$ near ∞ .

$$f \in O(g) \iff \limsup_{n \to \infty} \frac{f(n)}{g(n)} < \infty$$

$$\limsup_{n\to\infty} h(n) = \lim_{n\to\infty} \left(\sup_{m>n} h(m) \right)$$

(i.e. the limit of the least upper bound)

f in O(g)? some examples:

$$\lim_{n \to \infty} \frac{n}{2n} = \frac{1}{2}, \text{ n is } O(2n)$$

$$\lim_{n \to \infty} \frac{2n}{n} = 2, \text{ 2n is } O(n)$$

$$\lim_{n \to \infty} \frac{n}{\sqrt{n}} = \infty, \text{ n is not } O(\sqrt{n})$$

Prove:

$$\log(\log(n)) = O(\log^2 n)$$

Hint: L'Hospital's rule

Prove:

$$log(log(n)) = O(log^2 n)$$

$$\lim_{n \to \infty} \frac{log(log(n))}{log^2 n} =$$

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$$log(log(n)) = O(log^2n)$$

$$\lim_{n\to\infty}\frac{\log(\log(n))}{\log^2 n}=\lim_{n\to\infty}\frac{\frac{1}{n\log(n)}}{\frac{2\log(n)}{n}}=$$

Prove:

$$log(log(n)) = O(log^2n)$$

$$\lim_{n \to \infty} \frac{\log(\log(n))}{\log^2 n} = \lim_{n \to \infty} \frac{\frac{1}{n\log(n)}}{\frac{2\log(n)}{n}} = \frac{1}{2\log^2 n} = 0$$

$$\therefore \log(\log(n)) = O(\log^2 n)$$



Prove:

$$\lim_{n\to\infty}\frac{2^{n+1}-1}{2^n} \ \text{is} \ \textit{O}(1)$$

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Solution:

$$\lim_{n \to \infty} \frac{2^{n+1} - 1}{2^n} = \lim_{n \to \infty} \frac{(n+1)2^n}{n2^{n-1}} = \lim_{n \to \infty} \frac{2n+2}{n} = 2 = O(1)$$

Order from lowest to highest:

- $I O(log_{10} n)$
- II O(n!)
- III $O(2^n)$
- IV O(1)
- $V O(log_2 n)$
- VI O(nlog n)
- VII $O(n^2)$
- VIII O(n)

O(1)

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- **O**(1)
- \bigcirc O(log_{10} n)=O(log_2 n)

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- **O**(1)
- $\bigcirc O(log_{10} n) = O(log_2 n)$
- O(n)

Order from lowest to highest:

$$I O(log_{10} n)$$

III
$$O(2^n)$$

VII
$$O(n^2)$$

$$\bigcirc O(log_{10} n) = O(log_2 n)$$

Order from lowest to highest:

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- II O(n!)
- III $O(2^n)$
- IV O(1)
- $V O(log_2 n)$
- VI O(nlog n)
- VII $O(n^2)$
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- **0** O(1)
- $\bigcirc O(log_{10} n) = O(log_2 n)$
- O(n)
- O(nlog n)
- $O(n^2)$

Order from lowest to highest:

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Order from lowest to highest:

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- $O(n^2)$
- O(n!)

The "=" sign in Big-Oh Notation

$$n = O(n^4)$$

$$n^2 = O(n^4)$$

$$n^3 = O(n^4)$$

$$\implies n = n^2 = n^3$$

The "=" sign should be interpreted as a \leq .

The "=" sign in Big-Oh Notation

Is there an analog for <?

Little-Oh Notation

Assume $g(n) \neq 0$ near ∞ .

$$f \in O(g) \iff \limsup_{n \to \infty} \frac{f(n)}{g(n)} = 0$$

$$\limsup_{n\to\infty} h(n) = \lim_{n\to\infty} \left(\sup_{m\geq n} h(m) \right)$$

(i.e. the limit of the least upper bound)

What's the Difference?

True for Big-Oh but not Little-Oh:

- $x^3 \in O(x^3)$
- $x^3 \in O(x^3 + x)$
- $x^3 \in O(10x^3)$

True for Little-Oh:

- $x^2 \in O(x^3)$
- $x^3 \in O(x^4)$
- $x^3 \in O(x!)$

Prove: $n^k \in O(b^n)$ and $b^n \notin O(n^k) \forall b > 1, n > 1, k \ge 0$

Hint: use the following theorem:

Let f(n), g(n) be two non-negative functions. Then:

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}=0\implies f(n)\in O(g(n))\ \ \text{and}\ \ g(n)\notin O(f(n))$$

Use L'Hospitals!



Prove: $n^k \in O(b^n)$ and $b^n \notin O(n^k) \forall b > 1, n > 1, k \ge 0$

Hint: use the following theorem:

Let f(n), g(n) be two non-negative functions. Then:

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Use L'Hospitals!

$$\lim_{n\to\infty}\frac{n^k}{b^n}=\lim_{n\to\infty}\frac{kn^{k-1}}{\log(b)b^n}=\cdots\lim_{n\to\infty}\frac{k!}{b^n(\log(b))^k}$$

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Insertion sort

- Input an unsorted array A of length n
- Iterate from i=2:n
- Remove element i
- Compare with sorted elements<i</p>
- Insert element i in sorted position

Exercise 4 - Implement the insertion sort in Python

```
def insertionSort(alist):
    for index in range(1,len(alist)):
        currentvalue = alist[index]
        position = index
        while position>0 and alist[position-1]>currentvalue:
            alist[position]=alist[position-1]
            position = position-1
            alist[position]=currentvalue
        return(alist)
```

What is the worst-case runtime complexity of the insertion sort?Prove it. Best case?

Hint:
$$\sum_{n=1}^{\infty} i = \frac{n^2 + n}{2}$$



What is the worst-case runtime complexity of the insertion sort? Prove it. Best case?

Solution: $O(n^2)$. Note that we execute the outer loop n times and in the worst case scenario, we must make i comparisons each iteration. Combine this with the hint on the previous slide. Best case is O(n).