

1) a) Is the following statement scientifically sound? "The running-time of algorithm A is at least $O(n^2)$ "?

→ A scientific sentence must contain conclusive and correct judgments.

Complexity of the algorithm that $O(n^2)$ is upper bound. So, The running-time of Algorithm A is the most quadratic at $O(n^2)$. (Bigger polynomial). Big-O notation is a upper bound that can not be least. Therefore, Not scientifically-sound.

b) Are the following true? (c is constant)

i) $2^{n+1} = O(2^n)$? $\Rightarrow 2^n \cdot 2 \leq c \cdot 2^n$ $c > 0$ $\Rightarrow 2 \leq c$ $\boxed{c=2}$
 $n \geq n_0$ $\Rightarrow 1 \leq 1$ \checkmark \Rightarrow for all n is true.

ii) $2^{2n} = O(2^n)$? $\Rightarrow 2^{2n} \leq c \cdot 2^n$ $c > 0$ $\Rightarrow \frac{2^{2n}}{2^n} \leq c$ $\Rightarrow 2^n \leq c$ $\rightarrow 2^n$ is not bounded by constant number
 $n \geq n_0$ \Rightarrow false

c) Let $f(n)$ and $g(n)$ be asymptotically nonnegative functions. Is the equation $\max(f(n), g(n)) = \Theta(f(n) + g(n))$ true? (c₁ and c₂ constant)

Theta has omega and big-O \Rightarrow ① $\rightarrow \max(f(n), g(n)) \geq \frac{1}{2}(f(n) + g(n))$
 \Rightarrow ② $\rightarrow \max(f(n), g(n)) \leq c_2(f(n) + g(n))$

$f, g \in N \Rightarrow$ ② i) $\max(f(n), g(n)) \geq c_1(f(n) + g(n))$ \checkmark
 ii) $\max(f(n), g(n)) \leq c_2(f(n) + g(n))$ \checkmark

i) Proof
 $g(n)$ or $f(n)$ must equal or smaller than $\max(f(n), g(n))$
 $\max(f(n), g(n)) \geq g(n)$ and $\max(f(n), g(n)) \geq f(n)$

$$\begin{aligned} 2 \max(f(n), g(n)) &\geq f(n) + g(n) \\ \max(f(n), g(n)) &\geq \frac{1}{2}(f(n) + g(n)) \end{aligned} \Rightarrow \boxed{c_1 = \frac{1}{2}}$$

ii) $\max(f(n), g(n)) \leq c_2(f(n) + g(n)) \rightarrow c_2$ can be 1 or bigger \checkmark
 $\boxed{c_2 \geq 1}$

2 ideas are proved
 So, this equation is true.

①

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2) In each of the following situations, indicate whether $f \in O(g)$ or $f \in \Omega(g)$ or both (in which case $f \in \Theta(g)$).

$$\textcircled{a} \frac{f(n)}{n^{1.01}}, \frac{g(n)}{n \cdot \log^2 n} \Rightarrow \lim_{n \rightarrow \infty} \frac{n^{1.01}}{n \cdot \log^2 n} = \frac{n^{0.01}}{\log^2 n} = \frac{\infty}{\infty}$$

L'Hospital are used $\rightarrow \lim_{n \rightarrow \infty} \frac{0.01 \cdot n^{-0.99}}{2 \cdot \log n \cdot \frac{1}{n}} = \lim_{n \rightarrow \infty} \frac{0.01 \cdot \ln 2 \cdot n}{2 \cdot \log n} = \frac{0.01 \cdot \ln 2}{2}$

Again L'Hospital $\rightarrow \lim_{n \rightarrow \infty} \frac{(0.01)^2 \ln 2 \cdot n^{-0.99}}{2 \cdot \frac{1}{n \ln 2}} = \frac{(0.01)^2 (\ln 2)^2 \cdot n^{0.01}}{2}$

$$= \infty \Rightarrow f \in \mathcal{L}(g)$$

Notation Rules:

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \text{Result}$$

① result = 0 $\Rightarrow f(n) \in O(g(n))$

② $|| = \infty \Rightarrow f(n) \in \Omega(g(n))$

③ " = constant $\Rightarrow f(n) \in O(g(n))$

(b) $\frac{f(n)}{n!}, \frac{g(n)}{2^n} \Rightarrow \lim_{n \rightarrow \infty} \frac{n!}{2^n} = \frac{\infty}{\infty} \Rightarrow \lim_{n \rightarrow \infty} \frac{\sqrt{2\pi n} \cdot \frac{n^n}{e^n}}{2^n} \Rightarrow \lim_{n \rightarrow \infty} \frac{(2\pi)^{1/2} \cdot n^{1/2}}{(2^n) \cdot e^n} = \infty$

Stirling Formula

Because

grown rate of exponential " n^n " is faster than grown rate of exponential " 2^n "

$$n! \approx \sqrt{2\pi n} \cdot \left(\frac{n}{e}\right)^n$$

$$f \in \Omega(g)$$

$$\textcircled{E} \frac{f(n), g(n)}{\sqrt{n}} \Rightarrow \lim_{n \rightarrow \infty} \frac{\sqrt{n}}{(\log n)^3} = \frac{\infty}{\infty} \Rightarrow \text{L'Hospital} \lim_{n \rightarrow \infty} \frac{\frac{1}{2} \cdot n^{-0.5}}{3(\log n)^2 \cdot \frac{1}{n \ln 2}} = \lim_{n \rightarrow \infty} \frac{n^{0.5} \cdot \ln 2}{6 \cdot (\log n)^2} = \frac{\infty}{\infty}$$

② L'Hospital Again

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$$\lim_{n \rightarrow \infty} \frac{\frac{1}{2} \cdot n^{-0.5} \cdot \ln 2}{12 \cdot \log \frac{1}{n \cdot \ln 2}} = \lim_{n \rightarrow \infty} \frac{n^{0.5} \cdot (\ln 2)^2}{24 \cdot \log n} \xrightarrow{\text{L'Hospital Again}} \lim_{n \rightarrow \infty} \frac{\frac{1}{2} \cdot n^{-0.5} \cdot (\ln 2)^2}{24 \cdot \frac{1}{n \cdot \ln 2}} \Rightarrow \lim_{n \rightarrow \infty} \frac{n^{0.5} \cdot (\ln 2)^3}{48} = \infty \Rightarrow f \in \Omega(g)$$

$$\textcircled{d} \frac{f(n)}{g(n)} = \frac{n \cdot 2^n}{3^n} \Rightarrow \lim_{n \rightarrow \infty} \frac{n \cdot 2^n}{3^n} = \frac{\infty}{\infty} \Rightarrow \lim_{n \rightarrow \infty} \frac{n \cdot 2^n}{3^n} = \lim_{n \rightarrow \infty} \frac{\ln(n \cdot 2^n)}{\ln 3^n} = \lim_{n \rightarrow \infty} \frac{\ln n + \ln 2^n}{\ln 3^n} = \lim_{n \rightarrow \infty} \frac{\ln n + n \ln 2}{n \ln 3} = \lim_{n \rightarrow \infty} \frac{\ln n}{n \ln 3} + \lim_{n \rightarrow \infty} \frac{n \ln 2}{n \ln 3} = 0 + \frac{\ln 2}{\ln 3} = \frac{\ln 2}{\ln 3}$$

$$\boxed{\log 2 - \log 3 < 0} \Rightarrow \lim_{n \rightarrow \infty} \frac{n \cdot 2^n}{3^n} = 0 \quad f \in O(g)$$

we proved with \log algorithm

②

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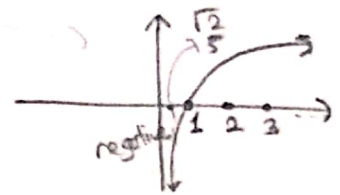
e) $\frac{f(n)}{g(n)} = \frac{\sum_{i=1}^n i^k}{n^{k+1}} \Rightarrow \lim_{n \rightarrow \infty} \frac{1^k + 2^k + \dots + (n-2)^k + (n-1)^k + n^k}{n^{k+1}} = \frac{1}{k+1}$
 it means sum up from 1^k to n^k constant

Both $f(n)$ and $g(n)$ are n term and both degree are k . There is proportion between $g(n)$ and $f(n)$. $f \in \Theta(g)$

f) $\frac{f(n)}{g(n)} = \frac{2^n}{2^{n+1}} \Rightarrow \lim_{n \rightarrow \infty} \frac{2^n}{2^{n+1}} = \lim_{n \rightarrow \infty} \frac{2^n}{2^n \cdot 2} = \lim_{n \rightarrow \infty} \frac{1}{2} = \boxed{\frac{1}{2}} \Rightarrow f \in \Theta(g)$

g) $\frac{f(n)}{g(n)} = \frac{n^{1/2}}{5^{\log_2 n}} \Rightarrow \lim_{n \rightarrow \infty} \frac{n^{1/2}}{5^{\log_2 n}} \Rightarrow \lim_{n \rightarrow \infty} \frac{n^{\log_2 \sqrt{2}}}{n^{\log_2 5}} = \lim_{n \rightarrow \infty} n^{\log_2 \sqrt{2} - \log_2 5}$

$= \lim_{n \rightarrow \infty} n^{\log_2 \sqrt{2} - \log_2 5} = \lim_{n \rightarrow \infty} n^{\log_2 \frac{\sqrt{2}}{5}} \Rightarrow \lim_{n \rightarrow \infty} \frac{1}{n^{1.8}} = 0 \Rightarrow f \in O(g)$



b) $\frac{f(n)}{g(n)} = \frac{\log_2 n}{\log_3 n} \Rightarrow \lim_{n \rightarrow \infty} \frac{\log_2 n}{\log_3 n} = \frac{\infty}{\infty} \Rightarrow \text{L'Hospital} \lim_{n \rightarrow \infty} \frac{1}{\frac{2n \cdot \ln 2}{1}} = \frac{3}{2} \Rightarrow f \in \Theta(g)$

3) List the following functions according to their order of growth and prove your assertions
 $\log n, \sqrt{n+10}, n+10, 10^n, 100^n, n^2 \log n, 32^{\log n}, n^6$

Firstly, Growth rates must be found by $\frac{f(n+1)}{f(n)}$ over $f(n)$ for each functions.

① $\frac{f(n+1)}{f(n)} = \frac{\log(n+1)}{\log n} \Rightarrow n=2 \Rightarrow \frac{\log 3}{\log 2} \approx 1.58$
 $\Rightarrow n=3 \Rightarrow \frac{\log 4}{\log 3} \approx 1.26$

② $\frac{f(n+1)}{f(n)} = \frac{\sqrt{n+11}}{\sqrt{n+10}} \Rightarrow n=2 \Rightarrow \frac{\sqrt{13}}{\sqrt{12}} \approx 1.04$
 $\Rightarrow n=3 \Rightarrow \frac{\sqrt{14}}{\sqrt{13}} \approx 1.03$

③ $\frac{f(n+1)}{f(n)} = \frac{n+11}{n+10} \Rightarrow n=2 \Rightarrow \frac{13}{12} \approx 1.08$
 $\Rightarrow n=3 \Rightarrow \frac{14}{13} \approx 1.08$

$$\textcircled{5} \frac{f(n+1)}{f(n)} = \frac{100^{n+1}}{100^n} = 100 \text{ constant growth rate}$$

$$\textcircled{6} \frac{f(n+1)}{f(n)} = \frac{(n+1)^2 \cdot \log n + 1}{n^2 \cdot \log n} \Rightarrow n=2 \Rightarrow \frac{9 \cdot \log 3}{4 \cdot \log 2} = 3.56$$

$$\Rightarrow n=3 \Rightarrow \frac{16 \cdot \log 4}{9 \cdot \log 3} = 2.24$$

$$\textcircled{7} 32^{\log_2 n} \xrightarrow{\text{we can swap } n \text{ and } 32} n^{\log_2 32} \rightarrow n^5 \Rightarrow \frac{f(n+1)}{f(n)} = \frac{(n+1)^5}{n^5} \Rightarrow n=2 \Rightarrow \frac{3^5}{2^5} = 7.59$$

$$\Rightarrow n=3 \Rightarrow \frac{4^5}{3^5} = 4.21$$

$$\textcircled{8} \frac{f(n+1)}{f(n)} = \frac{(n+1)^6}{n^6} = n=2 \Rightarrow \frac{3^6}{2^6} = 11.39$$

$$n=3 \Rightarrow \frac{4^6}{3^6} = 5.61$$

Let's compare the closest growth then find the list.

$$\bullet \frac{n+10}{a} > \frac{\sqrt{n+10}}{\sqrt{a}} \Rightarrow a > \sqrt{a} \quad n > 0 \text{ all } n \quad \bullet 100^n > 10^n \Rightarrow 100 > 10$$

$$\bullet \text{ we simply } n^{\log_2 32} \rightarrow n^5 \text{ so, } n^{\log_2 32} > n^2 \cdot \log n \text{ for all } n \geq 2$$

$$\bullet \text{ Then, } n^{\log_2 35} \rightarrow n^5 \text{ so, } n^6 > n^{\log_2 35} \quad 6 > 5$$

$\rightarrow \log n$ is smallest polynomial function "

Not! There is no constant function

$$\log n < \sqrt{n+10} < n+10 < n^2 \cdot \log n < 32^{\log n} < n^6 < 10^n < 100^n$$

4) Analyze the complexity in time (big-Oh notation) of the following operations at a given binary search tree (BST) that has height n :

a) Find Min.

b) Searching a Node.

c) Delete a leaf node.

d) Merging with another BST that has height n .

(4)

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⑥ Find Min

// The algorithm that is finding minimum node of Binary Search Tree

procedure FindMin(root)

while root.left != null do

root = root.left

end while

return root

end

// The complexity of algorithm is $T(n)$. Because, Algorithm traverse root to leftmost node and that could be traversed n (node numbers) time that happens worst case for left trees.

So, complexity of the algorithm. $T(n) \in O(n)$

⑦ Searching a Node

// The algorithm that is searching desired node of Binary Search Tree.

procedure SearchNode(root, desiredNode)

① if root.data == desiredNode.data OR root == NULL then
return root
② else if root.data > desiredNode.data then
return (call SearchNode(root.left, desiredNode))
③ else
return (call SearchNode(root.right, desiredNode))

c) Delete a leaf node.

// The algorithm that is removing desired leaf node of binary search tree.

① procedure DeleteALeafNode (root, parentNode, deleteleaf)

② if root == NULL // if there is no what you want leaf node.

return NULL

end if

③ if root.data == deleteleaf.data AND root.left == NULL AND root.right then

④ if parent.leaf.data == deleteleaf.data then

delete parent.left

else

delete parent.right

endif

⑤ else if root.data < deleteleaf.data

call DeleteALeafNode (root.right, root, deleteleaf)

else

call DeleteALeafNode (root.left, root, deleteleaf)

endif

endif

end

① - function gets parameters which are root (for checking leaf node), parentNode (the node is deleted node of parent), and deleteleaf (a leaf node).

② Base condition that there is no desired leaf node and return Nothing.

③ if current node (root) is equal leaf node then, check current node is leaf node or not.

④ if current node (root) is desired leaf node. In order to delete desired node, we have to check whether left or right of parent. Thus, we will delete a desired leaf.

⑤ We are searching to delete a leaf through comparing with current node (root) and leaf node data.

each compare operation will take

$O(1)$ but worst case $T(n)$

⑥

So, complexity $\rightarrow T(n) \in O(n)$

because that can be following example

ex:

① ② ③

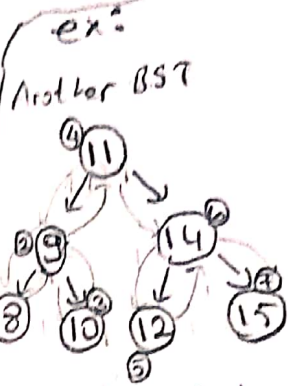
• Binary search tree could be like right example.

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④ Merging with another BST that has height n.

// If we merge the BST with another BST that will be done in order traversal.

- ① procedure Merge Another BST (root, another BST root)
- ② nodes = call convert SortedListInorderTraverse (another BST root)
- ③ While node != nodes do
 - call insert (root, node)
- end while
- end



This tree is perfect tree. That was for case for converting sort list. Every node traversed in $2^{n+1}-1$ with in order traverse.

① → Merge Another BST function is get arguments which is first BST root and another BST root

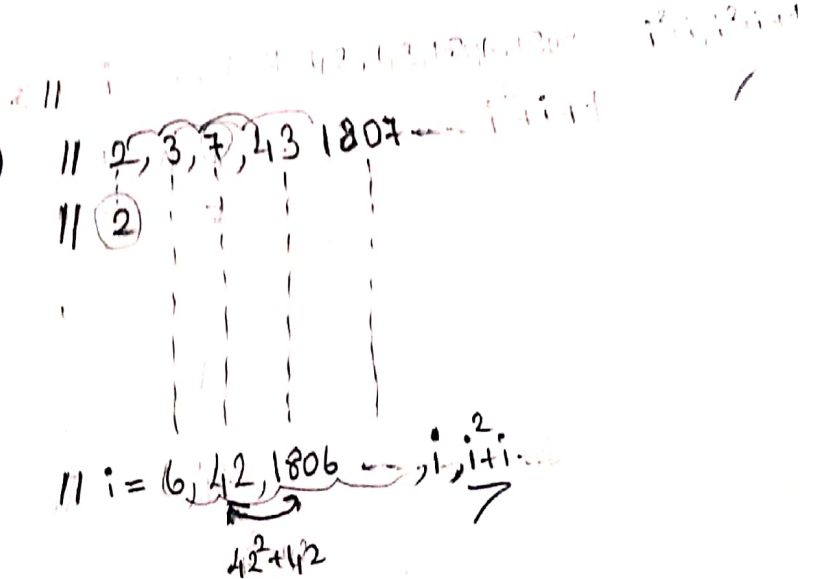
② → In convert... function traverse in Another BST and using the in order traversal with $2^{n+1}-1$ nodes. The function returns sorted list of nodes. (n height) → $T(2^n)$

③ Each nodes of BST are inserted one by one. When we insert a node, running is taking long time → Each inserting

$$[k] \text{ is node of first BST} \Rightarrow T(k) + T(k+1) + \dots + T(k+n) \Rightarrow O(k+n)$$

5) Find the time complexity (big-Oh notation) of the following program.

```
void function(int n)
{
    int count = 0;
    for (int i = 2; i <= n; i++)
    {
        if (i % 2 == 0)
        {
            count++;
        }
        else
        {
            i = (i-1) * i;
        }
    }
}
```



In the loop, $i \% 2 == 0$ is done just one time. Then, all the time, index is increasing

$$i^2 + i \rightarrow 1^2 + 1, 2^2 + 2, 3^2 + 3, \dots \rightarrow O(\log n)$$

⑦

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Induction Proof $\rightarrow 1, 3, 7, 15, 31, 63, 127, 255, 511, 1023, \dots i, i^2, i^3$

If we know i is a index, let's say t iteration of i .

$$i[t]^2 \leq i[t+1] = i[t]^2 + c \cdot i[t]$$

\rightarrow where c is a constant to be determined.

\rightarrow Then taking the base-2 logarithm

$$2 \log_2 i[t] \leq 2 \log_2 i[t] + \log_2 c$$

\rightarrow Taking algorithm again,

$$\log_2 2 \log_2 i[t] \leq \log_2 2 \log_2 i[t] + \log_2 \log_2 c$$

\rightarrow As, t represents the number of iterations of i for $n = i[t]$

$$\log_2 \log_2 n \leq \log_2 \log_2 n + \log_2 \log_2 c$$

\rightarrow we can take $c \geq 2$ let's take 4

$$\log_2 \log_2 n \leq \log_2 \log_2 n + \log_2 \log_2 4 \Rightarrow f(n) \in O(\log_2 \log_2 n)$$

(8)

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