CSCI 570 - Fall 2022 - HW 2

Due: Sep 7, 2022 at 11.59 PM PST

1. What is the worst-case runtime performance of the procedure below?

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\begin{array}{l} c=0\\ i=n\\ \textbf{while } i>1 \ \textbf{do}\\ \textbf{for } j=1 \ \textbf{to } i \ \textbf{do}\\ c=c+1\\ \textbf{end for}\\ i=\text{floor}(i/2)\\ \textbf{end while}\\ \textbf{return } c \end{array}
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Provide a brief explanation for your answer.

Solution:

There are i operations in the for loop and the while loop terminates when i becomes 1. The total time is

$$n + |n/2| + |n/4| + \dots \le (1 + 1/2 + 1/4 + \dots) \cdot n \le 2n = O(n).$$

Rubric (4 pts):

- 2 pts: if bound correctly found as O(n)
- 2 pts: Provides a correct explanation of the runtime
- 2 pts total if bound given is $O(n \log n)$ (i.e., not a tight upper bound)
- 2. Arrange these functions under the O notation using only = (equivalent) or \subset (strict subset of):
 - (a) $2^{\log n}$
 - (b) 2^{3n}
 - (c) $n^{n \log n}$
 - (d) $\log n$
 - (e) $n \log (n^2)$
 - (f) n^{n^2}

(g) $\log(\log(n^n))$

E.g. for the function $n, n+1, n^2$, the answer should be

$$O(n+1) = O(n) \subset O(n^2)$$
.

Provide brief explanations for your arrangement.

Solution:

First separate functions into logarithmic, polynomial, and exponential. Note that

$$2^{\log n} = n$$
, $n^{n \log n} = 2^{n(\log n)^2}$, $n^{n^2} = 2^{n^2 \log n}$,

we have:

(a) Logarithmic: $\log n$, $\log(\log(n^n))$

(b) Polynomial: $2^{\log n}$, $n \log(n^2)$

(c) Exponential: 2^{3n} , $n^{n \log n}$, n^{n^2}

• Since

$$\log n \le 1 \cdot \log(n \log n) = \log(\log(n^n)),$$

so $\log n = O(\log(\log(n^n)))$. On the other hand,

$$\log(\log(n^n)) = \log(n\log n) \le \log(n^2) = 2 \cdot \log n,$$

so $\log(\log(n^n)) = O(\log n)$. Thus

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

• Since every logarithmic grows slower than every polynomial,

$$O(\log(\log(n^n))) \subset O(2^{\log n}).$$

• $2^{\log n} = O(n) \subset O(n \log n) = O(2 \cdot n \log n) = O(n \log n^2)$. Thus

$$O(2^{\log n}) \subset O(n \log(n^2)).$$

• Since every exponential grows faster than every polynomial,

$$O(n\log(n^2)) \subset O(2^{3n}).$$

• Since

$$O(3n) \subset O(n(\log n)^2) \subset O(n^2 \log n),$$

so

$$O(2^{3n}) \subset O(2^{n(\log n)^2}) = O(n^{n\log n}) \subset O(2^{n^2\log n}) = O(n^{n^2}).$$

Therefore,

$$O(\log n) = O(\log(\log(n^n))) \subset O(2^{\log n}) \subset O(n\log(n^2)) \subset O(2^{3n}) \subset O(n^{n\log n}) \subset O(n^{n^2})$$

Rubric (10 pts):

- 1 point for correctly placing each of the 7 functions. '=' instead of ⊂ or vice versa is treated as incorrect placement of one of the functions.
- 3 points for brief justifications.
- 3. Given functions f_1, f_2, g_1, g_2 such that $f_1(n) = O(g_1(n))$ and $f_2(n) = O(g_2(n))$. For each of the following statements, decide whether it is true or false and briefly explain why.
 - (a) $f_1(n) \cdot f_2(n) = O(g_1(n) \cdot g_2(n))$
 - (b) $f_1(n) + f_2(n) = O(\max(g_1(n), g_2(n)))$
 - (c) $f_1(n)^2 = O(g_1(n)^2)$
 - (d) $\log_2 f_1(n) = O(\log_2 g_1(n))$

Solution:

By definition, there exist $c_1, c_2 > 0$ such that

$$f_1(n) \le c_1 \cdot g_1(n) \text{ and } f_2(n) \le c_2 \cdot g_2(n)$$

for n sufficiently large.

(a) True.

$$f_1(n) \cdot f_2(n) \le c_1 \cdot g_1(n) \cdot c_2 \cdot g_2(n) = (c_1c_2) \cdot (g_1(n) \cdot g_2(n)).$$

(b) True.

$$f_1(n) + f_2(n) \le c_1 \cdot g_1(n) + c_2 \cdot g_2(n)$$

$$\le (c_1 + c_2)(g_1(n) + g_2(n))$$

$$\le 2 \cdot (c_1 + c_2) \max(g_1(n), g_2(n)).$$

(c) True.

$$f_1(n)^2 \le (c_1 \cdot g_1(n))^2 = c_1^2 \cdot g_1(n)^2.$$

(d) False. Consider $f_1(n) = 2$ and $g_1(n) = 1$. Then

$$\log_2 f_1(n) = 1 \neq O(\log_2 g_1(n)) = O(0).$$

Rubric (3 pts for each subproblem):

- 1 pts: Correct T/F claim
- 2 pts: Providing a correct explanation or counterexample

4. Given an undirected graph G with n nodes and m edges, design an O(m+n) algorithm to detect whether G contains a cycle. Your algorithm should output a cycle if G contains one.

Solution:

Without loss of generality assume that G is connected. Otherwise, we can compute the connected components in O(m+n) time and deploy the below algorithm on each component.

Starting from an arbitrary vertex s, run BFS to obtain a BFS tree T, which takes O(m+n) time. If G=T, then G is a tree and has no cycles. Otherwise, G has a cycle and there exists an edge $e=(u,v)\in G\setminus T$. Let w be the least common ancestor of u and v. There exist a unique path T_1 in T from u to w and a unique path T_2 in T from w to v. Both T_1 and T_2 can be found in O(m) time. Output the cycle e by concatenating P_1 and P_2 .

Rubric (12 pts):

- No penalty for not mentioning disconnected case.
- 6 pts: for detecting whether G contains a cycle
- 4 pts: for finding (the edges in) a cycle if G contains one
- 2 pts: describing that the runtime is O(m+n) in each step (and thus total)
- 5. Solve Kleinberg and Tardos, Chapter 3, Exercise 6.

Solution:

Proof by Contradiction: assume there is an edge e=(x,y) in G that does not belong to T. Since T is a DFS tree, one of x or y is the ancestor of the other. On the other hand, since T is a BFS tree, x and y differs by at most 1 layer. Now since one of x and y is the ancestor of the other, x and y should differ by exactly 1 layer. Therefore, the edge e=(x,y) should be in the BFS tree T. This contradicts the assumption. Therefore, G cannot contain any edges that do not belong to T.

Rubric (8 pts):

- 4 pts: Correctly utilizing the fact that T is a DFS.
- 4 pts: Correctly utilizing using the fact that T is a BFS.

Ungraded problems

6. Solve Kleinberg and Tardos, Chapter 2, Exercise 6.

Solution:

- (a) The outer loop runs for exactly n iterations, the inner loop runs for at most n iterations, and the number of operations needed for adding up array entries A[i] through A[j] is j-i+1=O(n). Therefore, the running time is in $n^2 \cdot O(n) = O(n^3)$.
- (b) Consider those iterations that require at least n/2 operations to add up array entries A[i] through A[j]. When $i \leq n/4$ and $j \geq 3n/4$, the number of operations needed is at least n/2. So there are at least $(n/4)^2$ pairs of (i,j) such that adding up A[i] through A[j] requires at least n/2 operation. Therefore, the running time is at least $\Omega((n/4)^2 \cdot n/2) = \Omega(n^3/32) = \Omega(n^3)$.
- (c) Consider the following algorithm:

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\begin{array}{l} \mathbf{for} \ i = 1, 2, \dots, n-1 \ \mathbf{do} \\ B[i, i+1] \leftarrow A[i] + A[i+1] \\ \mathbf{end} \ \mathbf{for} \\ \mathbf{for} \ j = 2, 3, \dots, n-1 \ \mathbf{do} \\ \mathbf{for} \ i = 1, 2, \dots, n-j \ \mathbf{do} \\ B[i, i+j] \leftarrow B[i, i+j-1] + A[i+j] \\ \mathbf{end} \ \mathbf{for} \\ \mathbf{end} \ \mathbf{for} \\ \mathbf{end} \ \mathbf{for} \end{array}
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It first computes B[i,i+1] for all i by summing A[i] with A[j]. This for loop requires O(n) operations. For each j, it computes all B[i,i+j] by summing B[i,i+j-1] with A[i+j]. This works since the value B[i,i+j-1] were already computed in the previous iteration. The double for loop requires $O(n) \cdot O(n) = O(n^2)$ time. Therefore, the algorithm runs in $O(n^2)$.