Homework 2 M1522.000900 Data Structure (2019 Fall)

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1 Q1

Growth rate has an order like $n! > a^n > n^k > logn > c$ from fastest. By this, growth rate above has an order in $n! > 3 \cdot 2^n > 6n^2 > 20n > log_2n > log_2log_2n > 20n$.

2 Q2

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(1)  f(n) = log n^2 = 2log n
      q(n) = log n + 5
      For n_0 = 10^6 and c = 1, f(n) \ge c \cdot g(n), \forall n > n_0
      \therefore f(n) = \Omega(g(n))
      And for n_0 = 0 and c = 2, f(n) \le c \cdot g(n), \forall n > n_0
      \therefore f(n) = O(q(n))
      f(\mathbf{n}) = \theta(\mathbf{g}(\mathbf{n}))
(2) f(n) = \sqrt{n} = n^{\frac{1}{2}}
      q(n) = log n^2
      For n_0 = 10^2 and c = 1, f(n) \ge c \cdot g(n), \forall n > n_0
      \therefore \mathbf{f}(\mathbf{n}) = \Omega(\mathbf{g}(\mathbf{n}))
(3) f(n) = n
      q(n) = log^2 n = (log n)^2
      At n > 10, log n < n^{\frac{1}{2}}, thus, (log n)^2 < (n^{\frac{1}{2}})^2 = n
      For n_0 = 10, and c = 1, f(n) \ge c \cdot g(n), \forall n > n_0
      \therefore \mathbf{f}(\mathbf{n}) = \Omega(\mathbf{g}(\mathbf{n}))
(4) f(n) = log n^2 = 2log n
      g(n) = log^2 n = (log n)^2
      At n > 10, log n > 1 and (log n)^2 > log n
      For n_0 = 10 and c = 2, f(n) < c \cdot q(n), \forall n > n_0
      \therefore \mathbf{f}(\mathbf{n}) = \mathbf{O}(\mathbf{g}(\mathbf{n}))
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3 Q3

- (1) (a) Before the 1st for loop, time complexity is $\theta(1)$. For the 1st for loop, all the comparison and calculation and assignment occur constant times at each repetitions. So, 1st for loop's time complexity is $\theta(N)$. Similarly, 2nd for loop's time complexity is $\theta(M)$.
 - \therefore Total time complexity is $\theta(N + M)$.
 - (b) Before the 1st for loop, memory allocating occurs for a and b. So, its space complexity is $\theta(1)$. For the 1st for loop, memory allocating occurs only for i

and a at each repetitions. So, its space complexity is $\theta(1)$. Similarly, 2nd for loop's space complexity is also $\theta(1)$.

- \therefore Total space complexity is $\theta(1)$.
- (2) Before the 1st for loop, time complexity is $\theta(1)$. For the outer for loop, i increases by 1, so it is repeated $c_1 \cdot n$ times. For the inner for loop, j increases in $2^1, 2^2, \dots, 2^{(lgn)}$, so it is repeated $c_2 \cdot lgn$ times. Therefore, the whole repetition occurs $c_1 \cdot c_2 \cdot n \cdot lgn$ times.
 - \therefore Total time complexity is $\theta(\mathbf{n} \cdot \mathbf{lgn})$.
- (3) An algorithm X is asymptotically more efficient than Y means X will always be a better choice for large inputs.
 ∴ (b)

4 Q4

- (1) $T(n) + 1 = 3 \cdot (T(n-1) + 1)$ $T(n) + 1 = 3^{n-1} \cdot (T(0) + 1)$ $\therefore T(n) = 3^{n-1} \cdot (T(0) + 1) - 1 = 3^{n-1} \cdot T(0) + 3^{n-1} - 1$ Thus, for c = T(0) + 1 and $n_0 = 0$, $T(n) \le c \cdot 3^{n-1}$, $\forall n > n_0$ $\therefore \mathbf{T}(\mathbf{n}) = \mathbf{O}(3^{\mathbf{n}})$
- (2) Let's assume that the new machine Y has similar algorithm with X. X takes t seconds for n inputs. As Y is 27 times faster than X, Y takes $\frac{1}{27}t$ for the same n inputs.

$$c \cdot 3^{n-1} = 27 \cdot c' \cdot 3^{n-1} = t$$

 $c' = \frac{1}{27}c$

 \therefore Time complexity of $Y, Y(n) = c \cdot 3^{n-4}$

By the result, **Y** can handle n + 3 inputs while X handles n inputs in the same t.

5 Q5

- (1) In the function powerN, comparison occurs 1 time in the 1st line. At the 2nd line, function call occurs. If time complexity of function in Figure 1. is T(n), then, T(n) = 1 + T(n-1)
 - \therefore If *n* is large enough, $\mathbf{T}(\mathbf{n}) = \theta(\mathbf{n})$.
- (2) If time complexity of function in Figure 2. is S(n), then,

$$S(n) = \begin{cases} 1 & \text{for } n = 0\\ 1 + S(\frac{n}{2}) & \text{for } n = 2^k\\ 1 + S(\frac{n-1}{2}) & \text{for } n = 2^k - 1 \end{cases}$$

 \therefore If *n* is large enough, $\mathbf{S}(\mathbf{n}) = \theta(\mathbf{lgn})$.