## Problem Set 4 Solution

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### Question 1

a. Statement:  $\forall f, g : \mathbb{N} \to \mathbb{R}^+, b \in \mathbb{R}^+, (g(n) \in \Theta(f(n))) \land (n_0 \in \mathbb{N}, n \geq n_0 \Rightarrow f(n) \geq b \land g(n) \geq b) \land (b > 1) \Rightarrow \log_b(g(n)) \in \Theta(\log_b(f(n)))$ 

Statement Expanded:  $\forall f, g : \mathbb{N} \to \mathbb{R}^+, b \in \mathbb{R}^+, \left(\exists c_1, c_2, n_0 \in \mathbb{R}^+, \forall n \in \mathbb{N}, n \geq n_0 \Rightarrow c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)\right) \land \left(\exists n_1 \in \mathbb{N}, n \geq n_1 \Rightarrow f(n) \geq b \land g(n) \geq b\right) \land \left(b > 1\right) \Rightarrow \left(\exists d_1, d_2, n_2 \in \mathbb{R}^+, \forall n \in \mathbb{N}, n \geq n_2 \Rightarrow d_1 \cdot \log_b(g(n)) \leq \log_b(f(n)) \leq d_2 \cdot \log_b(g(n))\right)$ 

Proof. Let  $f, g : \mathbb{N} \to \mathbb{R}^+$ , and  $b \in \mathbb{R}^+$ . Assume  $c_1 = 1$ ,  $c_2 = b$ , and  $n_0 = 1$ , and  $n \in \mathbb{N}$  such that  $n \geq n_0$  and  $c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$ . Assume f(n) and g(n) are eventually  $\geq b$ . Assume b > 1. Let  $d_1 = 1$ ,  $d_2 = 2$ , and  $n_2 = n_0$ . Assume  $n \geq n_2$ .

We need to show  $d_1 \cdot \log_b g(n) \le \log_b f(n) \le d_2 \cdot \log_b g(n)$ .

We will do so in two parts. One for  $(d_1 \cdot \log_b g(n) \le \log_b f(n))$  and the other for  $(\log_b f(n) \le d_2 \cdot \log_b g(n))$ .

Part 1  $(d_1 \cdot \log_b g(n) \le \log_b f(n))$ :

The assumption tell us

$$c_1 \cdot g(n) \le f(n) \tag{1}$$

Then, it follows from the fact  $\forall x, y \in \mathbb{R}^+, x \geq y \Leftrightarrow \log x \geq \log y$ 

$$\log(c_1 \cdot g(n)) \le \log(f(n)) \tag{2}$$

Then, using the fact b > 1, we can calculate

$$\frac{\log(c_1 \cdot g(n))}{\log b} \le \frac{\log(f(n))}{\log b} \tag{3}$$

$$\frac{\log(c_1) + \log(g(n))}{\log b} \le \frac{\log(f(n))}{\log b} \tag{4}$$

Then,

$$\frac{\log(g(n))}{\log b} \le \frac{\log(f(n))}{\log b} \tag{5}$$

by the fact  $c_1 = 1$  and  $\log c_1 = 0$ .

Then, since  $\frac{\log f(x)}{\log b} = \log_b f(x)$ ,

$$\log_b(g(n)) \le \log_b(f(n)) \tag{6}$$

Then, because we know  $d_1 = 1$ , we can conclude

$$\log_b(g(n)) \le d_1 \cdot \log_b(f(n)) \tag{7}$$

Part 2 ( $\log_b f(n) \le d_2 \cdot \log_b g(n)$ ):

The assumption tells us

$$f(n) \le c_2 \cdot g(n) \tag{8}$$

Then, it follows from the fact  $\forall x, y \in \mathbb{R}^+, x \geq y \Leftrightarrow \log x \geq \log y$ 

$$\log(f(n)) \le \log(c_2 \cdot g(n)) \tag{9}$$

Then, using the fact b > 1, we can calculate

$$\frac{\log(f(n))}{\log b} \le \frac{\log(c_2 \cdot g(n))}{\log b} \tag{10}$$

$$\frac{\log(f(n))}{\log b} \le \frac{\log(c_2) + \log(g(n))}{\log b} \tag{11}$$

Then, since  $c_2 = b$ ,

$$\frac{\log(f(n))}{\log b} \le \frac{\log(b) + \log(g(n))}{\log b} \tag{12}$$

Then, using the fact g(n) is eventually  $\geq b$ , we can write

$$\frac{\log(f(n))}{\log b} \le \frac{\log(g(n)) + \log(g(n))}{\log b} \tag{13}$$

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

$$\frac{\log(f(n))}{\log b} \le \frac{2 \cdot \log(g(n))}{\log b}$$
(13)

Then, since  $\frac{\log f(x)}{\log b} = \log_b f(x)$ ,

$$\log_b(f(n)) \le 2 \cdot \log_b(g(n)) \tag{15}$$

Then, because we know  $d_2 = 2$ , we can conclude

$$\log_b(f(n)) \le d_2 \cdot \log_b(g(n)) \tag{16}$$

Notes:

- $\forall x, y \in \mathbb{R}^+, x > y \Leftrightarrow \log x > \log y$
- $\exists c_1, c_2, n_0 \in \mathbb{R}^+, \forall n \in \mathbb{N}, n \ge n_0 \Rightarrow c_1 \cdot g(n) \le f(n) \le c_2 \cdot g(n)$
- Definition of Eventually:  $\exists n_0 \in \mathbb{N}, n \geq n_0 \Rightarrow P$ , where  $P : \mathbb{N} \to \{\text{True}, \text{False}\}$

### b. Proof. Let $k \in \mathbb{N}$ .

First, we will analyze the cost of loop 2 over iteration of loop 1.

The code tells us loop 2 starts at  $j_k = 1$  with  $j_k$  increasing by a factor of 3 per iteration until  $j_k \ge 1$ .

Using these facts, we can calculate that the terminating condition occurs when

$$3^k \ge i \tag{1}$$

$$k \ge \log_3 i \tag{2}$$

Because we know the number of iterations is the smallest value of k satisfying the above inequality, we can conclude loop 2 has

$$\lceil \log_3 i \rceil \tag{3}$$

iterations.

Next, we need to determine the total number of iterations of loop 2 over all iterations of loop 1.

The code tells us loop 1 starts at i = 1 and ends at i = n with each i increasing by 1 per iteration.

Using these facts, we can conclude loop 2 has total of

$$\lceil \log_3 1 \rceil + \lceil \log_3 2 \rceil + \dots + \lceil \log_3 n \rceil = \sum_{i=1}^n \lceil \log_3 i \rceil$$
 (4)

iterations. 
$$\Box$$

c. After scratching head and looking at solution many times, I realized that there are many things I do not yet understand, and it's the best to write what I have and learn from the solution. Here is my best attempt:).

Proof. Let  $n \in \mathbb{N}$ .

The previous answer tells us the exact cost of the algorithm is

$$\sum_{i=1}^{n} \lceil \log_3 i \rceil \tag{1}$$

Then, it follows by changing the variable i to  $i' = \log_3 i$  we can write

$$\sum_{i'=0}^{\lceil \log_3 n \rceil} i' \tag{2}$$

Then, because we know  $\sum_{i=0}^{n} i = \frac{n(n+1)}{2}$ , we can conclude

$$\sum_{i'=0}^{\lceil \log_3 n \rceil} i' = \frac{(\lceil \log_3 n \rceil)(\lceil \log_3 n \rceil + 1)}{2} \tag{3}$$

$$=\frac{\lceil \log_3 n \rceil^2 + \lceil \log_3 n \rceil}{2} \tag{4}$$

Then, we can conclude the runtime of the algorithm is  $\Theta(\log_3^2 n)$ .

### **Correct Solution:**

We need to determne  $\Theta$  of the algorithm.

We will prove that the  $\Theta$  of the algorithm is  $\Theta(n \log n)$ .

The answer to previous question tells us the total exact cost of the algorithm is

$$\sum_{i=1}^{n} \lceil \log_3 i \rceil \tag{5}$$

Then, by using fact  $1 \ \forall x \in \mathbb{R}, x \leq \lceil x \rceil \leq x + 1$ , we can calculate

$$\sum_{i=1}^{n} \log_3 i \le \sum_{i=1}^{n} \lceil \log_3 i \rceil \le \sum_{i=1}^{n} \left( \log_3 i + 1 \right) \tag{6}$$

$$\sum_{i=1}^{n} \log_3 i \le \sum_{i=1}^{n} \lceil \log_3 i \rceil \le \left(\sum_{i=1}^{n} \log_3 i + \sum_{i=1}^{n} 1\right) \tag{7}$$

$$\sum_{i=1}^{n} \log_3 i \le \sum_{i=1}^{n} \lceil \log_3 i \rceil \le \sum_{i=1}^{n} \log_3 i + n \tag{8}$$

Then,

$$\log_3\left(\prod_{i=1}^n i\right) \le \sum_{i=1}^n \lceil \log_3 i \rceil \le \log_3\left(\prod_{i=1}^n i\right) + n \tag{9}$$

$$\log_3(n!) \le \sum_{i=1}^n \lceil \log_3 i \rceil \le \log_3(n!) + n \tag{10}$$

by the fact  $\forall a, b \in \mathbb{R}^+$ ,  $\log(a) + \log(b) = \log(ab)$ .

Then,

$$\frac{\ln n!}{\ln 3} \le \sum_{i=1}^{n} \lceil \log_3 i \rceil \le \frac{\ln(n!)}{\ln 3} + n \tag{11}$$

by changing the base to e using the formula  $\log_3 n! = \frac{\log_e n!}{\log_e 3} = \frac{\ln n!}{\ln 3}$ .

Now, the fact 2 tells us  $n! \in \Theta(e^{n \ln n - n + \frac{1}{2} \ln n})$ .

Because we know from fact 3 that  $n \ln n - n + \frac{1}{2} \ln n$  is eventually  $\geq 1$ , we can conclude  $e^{n \ln n - n + \frac{1}{2} \ln n}$  is eventually  $\geq e$ .

Since n! is also eventually  $\geq e$ , by using solution to problem 1.a with g(n) = n! and  $f(n) = e^{n \ln n - n + \frac{1}{2} \ln}$  and b = e, we can write

$$\ln(n!) \in \Theta(\ln(e^{n\ln n - n + \frac{1}{2}\ln n})) \tag{12}$$

$$\ln(n!) \in \Theta(n \ln n - n + \frac{1}{2} \ln n) \tag{13}$$

Then,

$$\ln(n!) \in \Theta(n \ln n) \tag{14}$$

by the fact  $n \ln n - n + \frac{1}{2} \ln n \in \Theta(n \ln n)$ .

So, since the algorithm runs at least  $\frac{\ln n!}{\ln 3}$ , we can conclude it has asymptotic lower bound of  $\Omega(n \ln n)$ , and since the algorithm runs at most  $\frac{\ln n!}{\ln 3} + n$ , we can conclude it has upper bound running time of  $\mathcal{O}(n \ln n)$ .

Since the value of  $\Omega$  and  $\mathcal{O}$  are the same, we can conclude the algorithm has running time of  $\Theta(n \ln n)$  or  $\Theta(n \log n)$ .

### Notes:

- In a main flow of proof, when there is a huge interruption like showing  $\ln(n!) \in \Theta(n \ln n)$ , how can a sentence be started to tell the audience we are working on another major idea?
- When an interruption in proof has been occurred for another major part of a proof, how can a sentence be started to combine parts together?
- How can a sentence be written to say condition  $x_1$ ,  $x_2$ , and  $x_3$  are satisfied, so a statement y can be used to an equation or an idea?

### Question 2

a. We need to evaluate tight asymptotic upper bound.

We will prove that the tight asymptotic upper bound of the algorithm is  $\mathcal{O}(n^2)$ .

First, we need to analyze the number of iterations of loop 2 per iteration of loop 1.

The code tells us loop 2 starts at j = 0 and ends at most j = i - 1 with j increasing by 1 per iteration.

Then, using these facts, we can conclude loop 2 has at most

$$\left\lceil \frac{i-1-0+1}{1} \right\rceil = i \tag{1}$$

iterations.

Next, we need to determine the total number of iterations of loop 2 over all iterations of loop 1.

The code tells us that loop 1 starts at i = n and ends at most i = 0 with i decreasing by 1 per iteration.

Because we know each iteration of loop 1 takes i iterations by loop 2, using these facts, we can conclude the total number of iterations of loop 2 is at most

$$n + (n-1) + (n-2) + \dots + 0 = \sum_{i=1}^{n}$$
 (2)

$$=\frac{n(n+1)}{2}\tag{3}$$

iterations, or  $\mathcal{O}(n^2)$ .

#### **Correct Solution:**

We need to evaluate tight asymptotic upper bound.

We will prove that the tight asymptotic upper bound of the algorithm is  $\mathcal{O}(n^2)$ .

First, we need to analyze the cost of loop 2.

The code tells us loop 2 starts at j = 0 and ends at most j = i - 1 with j increasing by 1 per iteration.

Then, since each iteration of loop 2 takes a constant step (1 step), using these facts, we can conclude the cost of loop 2 is at most

$$1 \cdot (i - 1 - 0 + 1) = i \tag{1}$$

steps.

Next, we need to determine cost of loop 1.

The code tells us that loop 1 starts at i = n and ends at most i = 0 with i decreasing by 1 per iteration.

Because we know each iteration of loop 1 takes i + 1 steps (where i is from loop 2 and 1 from line 8), using these facts, we can conclude the total cost of loop 1 is at most

$$(n+1) + n + (n-1) + (n-2) + \dots + 1 = \sum_{i=0}^{n} (i+1)$$
 (2)

$$= \sum_{i=0}^{n} i + \sum_{i=0}^{n} 1 \tag{3}$$

$$= \sum_{i=0}^{n} i + (n+1) \tag{4}$$

$$= \frac{n(n+1)}{2} + (n+1) \tag{5}$$

$$=\frac{(n+1)(n+2)}{2}$$
 (6)

steps.

Finally, adding the cost of line 6, we can conclude the algorithm has total cost of  $\frac{(n+1)(n+2)}{2} + 1$  steps, which is  $\mathcal{O}(n^2)$ .

### Notes:

- Noticed professor writes proof that gets to a point (i.e. ... where each iteration takes i+1 steps), and provides more detailed explanation in brackets (i.e. ... where each iteration takes i+1 steps (Adding the cost of loop 2 and 1 step for other constant time operations)).
- Noticed professor uses 'finally' when proof has reached the final step that leads to its
  conclusion.
- b. Let  $n, k \in \mathbb{N}$ , and  $list = [0, 0, \dots, 0, 1, 0, \dots, 0]$  where 1 is at  $\lceil \frac{n}{2} \rceil$  position.

We will prove that the tight asymptotic lower bound running time of this algorithm is  $\Omega(n^2)$ .

First, we need to evaluate the cost of loop 2.

The code tells us loop 2 starts at  $j_k = 0$ , and  $j_k$  will increase by 1 until  $j_k \ge \lceil \frac{n}{2} \rceil + 1$  (where +1 is because of loop 2 stopping at  $j_k = \lceil \frac{n}{2} \rceil$  by the if condition on line 10).

Using the fact  $j_k = k + 1$ , we can calculate that loop 2 stops when

$$k+1 \ge \left\lceil \frac{n}{2} \right\rceil + 1 \tag{1}$$

$$k \ge \left\lceil \frac{n}{2} \right\rceil \tag{2}$$

Since we are looking for the smallest value of k (because the smallest value of k translates to number of iterations), we can conclude the loop has

$$\left\lceil \frac{n}{2} \right\rceil \tag{3}$$

iterations.

Since each iteration takes a constant time (1 step), the cost of loop 2 is

$$\left\lceil \frac{n}{2} \right\rceil \cdot 1 = \left\lceil \frac{n}{2} \right\rceil \tag{4}$$

steps.

Next, we need to evaluate the cost of loop 1.

The code tell us loop 1 will start at  $i_k = n$ , and  $i_k$  will decrease by 1 per iteration until  $i_k \leq \lceil \frac{n}{2} \rceil$ .

Using the fact  $i_k = k - 1$ , we can write loop 1 stops when

$$k - 1 \le \left\lceil \frac{n}{2} \right\rceil \tag{5}$$

$$k \le \left\lceil \frac{n}{2} \right\rceil + 1 \tag{6}$$

Since we are looking for the largest value of k (because the largest value of k translates to number of iterations), we can conclude loop 1 has

$$\left\lceil \frac{n}{2} \right\rceil + 1 \tag{7}$$

iterations.

Since each costs  $\left\lceil \frac{n}{2} \right\rceil + 1$  steps, we can conclude loop 1 has cost of

$$\left(\left\lceil \frac{n}{2}\right\rceil + 1\right) \left(\left\lceil \frac{n}{2}\right\rceil + 1\right) = \left\lceil \frac{n}{2}\right\rceil^2 + 2 \cdot \left\lceil \frac{n}{2}\right\rceil + 1 \tag{8}$$

steps.

Finally, by adding the cost of line 6 (1 step), the total running time of this algorithm is

$$\left\lceil \frac{n}{2} \right\rceil^2 + 2 \cdot \left\lceil \frac{n}{2} \right\rceil + 2 \tag{9}$$

steps, which is  $\Omega(n^2)$ 

# Question 3

# Question 4