## 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  $\Omega(n^2)$ 

### Correct Solution:

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  $j_k = k$ , we can calculate that loop 2 stops when

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

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

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

iterations.

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

$$\left(\left\lceil \frac{n}{2}\right\rceil + 1\right) \cdot 1 = \left\lceil \frac{n}{2}\right\rceil + 1\tag{3}$$

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  $i_k = n - k$ , we can write loop 1 stops when

$$n - k \le \left\lceil \frac{n}{2} \right\rceil \tag{4}$$

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

$$k \ge n - \left\lceil \frac{n}{2} \right\rceil \tag{6}$$

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

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

iterations.

Since each iteration costs  $\lceil \frac{n}{2} \rceil + 2$  steps (where  $\lceil \frac{n}{2} \rceil + 1$  is the cost of loop 2 and +1 is the cost of line 14), we can conclude loop 1 has cost of

$$\left(\left\lceil \frac{n}{2}\right\rceil + 2\right)\left(n - \left\lceil \frac{n}{2}\right\rceil\right) \tag{8}$$

steps.

Finally, since the loop takes  $\lceil \frac{n}{2} \rceil + 1$  extra steps (where  $\lceil \frac{n}{2} \rceil$  is the cost of traveling from j = 0 until  $j = \lceil \frac{n}{2} \rceil$  and +1 is the cost of line 14) before coming to a full stop, the total running time is at least

$$\left(\left\lceil \frac{n}{2}\right\rceil + 2\right)\left(n - \left\lceil \frac{n}{2}\right\rceil\right) + \left\lceil \frac{n}{2}\right\rceil + 1\tag{9}$$

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

#### Notes:

- Noticed there is no room for errors. (most of mark deductions are from not being careful with the analysis).
- Realized I need to take time to verify and re-verify steps using examples at a very fine level (i.e at this step this happens ... at this step this happens) until conclusion.
- Noticed professor uses  $i_k = n k$  when going backward starting from n. And for the inequality,  $i_k \leq$  is used as opposed to the normal  $i_k \geq$ .

### c. Proof. Let $k, n \in \mathbb{N}$ .

We will prove the statement using proof by cases.

### Case 1: When all elements in nums are even

Let  $nums = [a_1, a_2, \dots, a_n]$  where  $a_1, \dots, a_n$  are even numbers.

We want to prove the best-case lower bound running time of this algorithm is  $\Omega(n)$ .

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

Given the iteration count k, the code tells us, the loop starts at  $j_k = 0$  and increases by 1 per iteration, and so we know  $j_k = k$ .

Because we know loop 2 runs until  $j_k \geq i$ , we can conclude loop 2 stops when

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

Since we are looking for the smallest value of k (because it represents the number of iterations), we can conclude loop 2 has i iterations.

Because we know each iteration of loop 2 costs a constant time (1 step), we can conclude loop 2 has cost of at least

$$k \cdot 1 = k \tag{2}$$

steps.

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

The code tells us loop 1 starts at i = n and ends at i = n due to the truthy condition of line 14.

Using these facts, we can conclude loop 1 has

$$\lceil n - n + 1 \rceil = 1 \tag{3}$$

iteration.

Because we know each iteration of loop 1 costs i+2 steps (where i is from the cost of loop 2, and +2 are from the cost of line 8 and line 16), we can conclude loop 1 has cost of at least

$$(i+2) \cdot 1 = i+2 \tag{4}$$

steps.

Finally, because we know i = n, the total running time is at least n + 2, which is  $\Omega(n)$ .

### Case 2: When one or more elements in nums are odd

Let  $nums = [1, a_2, a_3, \dots, a_{n-1}]$  where  $a_2, a_3, \dots, a_{n-1}$  are even numbers.

We will prove the algorithm has best-case lower bound running time of  $\Omega(n)$ .

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

The code tellus us loop 2 starts at j = 0 and ends at j = 0 due to the truthy condition of line 10.

Using these facts, we can calculate loop 2 has 1 iteration.

Because we know loop 2 takes constant time (1 step) per iteration, we can conclude loop 2 has cost of 1 step.

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

The code tells us that loop 1 starts at i = n, and i increases by 1 until  $i_k \le -1$ , where k represents the iteration count of loop 1.

Because we know  $i_k = n - k$ , we can conclude the loop stops when

$$n - k \le -1 \tag{5}$$

$$k \ge n + 1 \tag{6}$$

Since we are looking for the smallest value of k (because it represents the number of iterations), we can conclude loop 1 has

$$n+1 \tag{7}$$

Since each iteration of loop 1 takes 2 steps (where 1 is the cost of loop 2 and the other 1 is the cost of line 8), we can conclude that loop 1 has cost of at least

$$2 \cdot (n+1) \tag{8}$$

steps.

Finally, adding the cost of line 8, we can conclude the algorithm has running time of at least 2(n+1)+1 steps, which is  $\Omega(n)$ .

### Attempt 2:

Let  $k, n \in \mathbb{N}$ .

We will prove this statement using proof by cases.

#### Case 1: When all elements in nums are even

Let  $nums = [a_1, a_2, \dots, a_n]$  where  $a_1, \dots, a_n$  are even numbers.

We want to prove the best-case lower bound running time of this algorithm is  $\Omega(n)$ .

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

Given the iteration count k, the code tells us, the loop starts at  $j_k = 0$  and increases by 1 per iteration, and so we know  $j_k = k$ .

Because we know loop 2 runs until  $j_k \geq i$ , we can conclude loop 2 stops when

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

Since we are looking for the smallest value of k (because it represents the number of iterations), we can conclude loop 2 has i iterations.

Because we know each iteration of loop 2 costs a constant time (1 step), we can conclude loop 2 has cost of at least

$$k \cdot 1 = k \tag{2}$$

steps.

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

The code tells us loop 1 starts at i = n and ends at i = n due to the truthy condition of line 14.

Using these facts, we can conclude loop 1 has

$$\lceil n - n + 1 \rceil = 1 \tag{3}$$

iteration.

Because we know each iteration of loop 1 costs i + 2 steps (where i is from the cost of loop 2, and +2 are from the cost of line 8 and line 16), we can conclude loop 1 has cost of at least

$$(i+2) \cdot 1 = i+2 \tag{4}$$

steps.

Finally, because we know i=n, the total running time is at least n+2, which is  $\Omega(n).$ 

### Case 2: When one or more elements in nums are odd

In this case, let m be the index of first odd number in nums.

We need to prove this algorithm has best-case lower bound running time of  $\Omega(n)$ .

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

Given loop 2 iteration count k, the code tells us loop 2 starts at j = 0, and j increases by 1 until  $j_k \ge m + 1$ .

Since we know  $j_k = k$ , using these facts, we can calculate loop 2 terminates when

$$k \ge m + 1 \tag{5}$$

Since we are looking for the smallest value of k (because it represents the number of iterations), we can conclude loop 2 has

$$m+1$$
 (6)

iterations.

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

Given loop 1 iteration count k,, The code tells us that loop 1 starts at i = n, and i decreases by 1 until  $i_k \leq m - 1$ .

Since we know  $i_k = n - k$ , using these facts, we can calculate loop 1 stops when

$$n - k \le m - 1 \tag{7}$$

$$k \ge n - m + 1 \tag{8}$$

Since we are looking for the smallest value of k (because it represents the number of iterations), we can conclude loop 1 has

$$n - m + 1 \tag{9}$$

iterations.

Because we know that for the first n-k iterations, each iteration of loop 1 costs m+2 steps (where m+1 is the cost of loop 2 and +1 is the cost of line 8), and last iteration of loop 1 costs another m+2 (where m is the cost of loop 2 and +2 are the cost of line 8 and 15), we can conclude loop 1 has cost of

$$(n-m+1)(m+2)$$
 (10)

steps.

Next, adding the cost of line 6, we can conclude the algorithm has total cost of at least

$$(n-m+1)(m+2)+1 (11)$$

steps.

Finally, we need to show this algorithm has runtime of  $\Omega(n)$ .

Using the total cost of algorithm, we can calculate

$$(n-m+1)(m+2)+1 = (n-m)(m+2)+(m+2)+1$$
(12)

$$> (n-m)(m+2) + (m+2)$$
 (13)

$$= (n-m)m + 2(n-m) + (m+2)$$
 (14)

$$> (n-m)m + (n-m) + m$$
 (15)

$$= (n-m)m + n \tag{16}$$

Because we know  $n-m \ge 0$  and  $m \ge 0$ , we can conclude that

$$(n-m+1)(m+2)+1 > n (17)$$

and the algorithm has best case lower bound running time of  $\Omega(n)$ .

### Notes:

- The solution in problem 2.b adds constant time operations into total cost where as the solution to this problem doesn't... Is there a rule behind when and when not they can be included?
- Noticed professor reduces the exact cost to n by separating it from the rest of the terms

$$(n-m+1)(m+2) > (n-m)m+n$$

• Realized the best-case lower bound running time doesn't use input family like worst-case lower bound running time

## Question 3

a. Proof. Let  $n \in \mathbb{N}$  and lst be a list with all negative numbers.

Then, the code tells us line 9-12 will run for all elements in the list.

Because we know i increases by a factor of 2 per iteration, we can conclude that at  $k^{th}$  iteration, i has value of  $i_k = 2^k$ .

Because we know loop terminates when  $i_k \geq n$ , we can conclude this is true when

$$2^k \ge n \tag{1}$$

$$k \ge \log n \tag{2}$$

Since we are looking for the smallest value of k (since it represents the number of iterations), we can conclude loop has

$$\lceil \log n \rceil \tag{3}$$

iterations.

Since each iteration of while loop takes a constant time (1 step), we can conclude the loop has cost of

$$\lceil \log n \rceil$$
 (4)

steps.

Finally, since lines 2 to 4 have cost of 1 each, by adding to the costs together, we can conclude the algorithm has total running time of  $\lceil \log n \rceil + 3$ , which is  $\Theta(\log n)$ .

### **Correct Solution:**

Let  $n, k \in \mathbb{N}$  and lst be a list with all negative numbers.

In this case, the loop follows this pattern

- iteration 1 else condition executes and j increases by a factor of 2
- iteration 2 if condition executes and i increases by a factor of 2, and moves to where j is
- iteration 3 else condition executes again and j increases by a factor of 2
- iteration 4 if condition executes again and i increases by a factor of 2 and moves to where j is.
- and this pattern repeats until the end of while loop.

Now, we need to determine the total number of iterations in while loop.

Because we know i increases by a factor of 2 per execution of **if** lst[i] >= 0: condition, we can conclude that at  $k^{th}$  execution of **if** lst[i] >= 0: condition, i has value of  $i_k = 2^k$ .

Because we know loop terminates when  $i_k \geq n$ , we can conclude this is true when

$$2^k > n \tag{1}$$

$$k \ge \log n \tag{2}$$

Since we are looking for the smallest value of k (since it represents the number of executions caused by the **if** lst[i] >= 0: condition), we can conclude loop has

$$\lceil \log n \rceil$$
 (3)

executions due to the if lst[i] >= 0: condition.

Because we know every execution of **if** lst[i] >= 0: condition in an iteration, is followed by the execution of **else**: condition in previous iteration, we can conclude while loop has total of

$$2 \cdot \lceil \log n \rceil \tag{4}$$

executions, or iterations.

Since each iteration of while loop takes a constant time (1 step), we can conclude the while loop has cost of

$$2 \cdot \lceil \log n \rceil \tag{5}$$

steps.

Finally, adding cost of 1 for the constant time operations on line 2-4, we can conclude the algorithm has total running time of  $2 \cdot \lceil \log n \rceil + 1$  steps, which is  $\Theta(\log n)$ .

#### Notes:

- Noticed professor bundles up time of constant operations (i.e. line 2-4) to 1, and same for the ones within while loop.
- Noticed professor introduces k in body as  $k^{th}$  execution of the if/else branch, and he doesn't introduce the variable in header.
- Noticed professor uses the word 'execution' to focus on the number of iterations caused by the if condition.
- Noticed professor lays out the pattern in while loop before moving onto proof.
- b. Proof. Let  $n \in \mathbb{N}$ , and lst be a list of integers where lst[0] to  $lst[\frac{n}{2}]$  have value 0, and  $lst[\frac{n}{2}+1]$  to lst[n-1] have value -1.

In this case, the loop follows this pattern:

• When j=1, if branch performs  $\frac{n}{2}+1$  executions, stops at  $i=\frac{n}{2}+1$ 

- else branch performs, and value of j doubles, and i resets to 0
- When j=2, if branch performs  $\frac{n}{4}+1$  executions, stops at  $i=\frac{n}{2}+2$
- ullet else branch performs, and value of j doubles, and i resets to 0
- When j=4, if branch performs  $\frac{n}{8}+1$  executions, stops at  $i=\frac{n}{2}+4$
- $\bullet$  else branch performs, and value of j doubles, and i resets to 0
- When j=8, if branch performs  $\frac{n}{16}+1$  executions, stops at  $i=\frac{n}{2}+8$
- This pattern repeats until  $k^{th}$  execution of else branch of statements has value of j half the size of n.
- Loop terminates one iteration after i reaches the end of array.

Now, we will prove this statement in two parts: one for determining the number of executions of else branch of statements in while loop, and another for determining the runtime of whole algorithm.

### Part 1: Determining the number of executions of else branch:

We need to prove the else branch executes  $\Omega(\log n)$  times.

The pattern tells us while loop depends on j, and j increases by a factor of 2 per execution of else branch until  $j_{k+1} \geq n$ .

Because we know at  $k^th$  execution of else branch has j with value of  $j_k = 2^k$ , using these facts, we can calculate

$$2^{k+1} \ge n \tag{1}$$

$$k+1 \ge \log n \tag{2}$$

$$k \ge \log n - 1 \tag{3}$$

So we know the else branch executes at least  $\log n - 1$  times, which is  $\Omega(\log n)$ .

## Part 2: Determining running time of algorithm:

We need to prove this algorithm has running time of  $\Theta(n)$ .

First, we need to determine number of executions of if branch in while loop.

The pattern tells us at  $k^{th}$  execution of else branch of statements in while loop,  $\frac{n}{2^{k+1}} + 1$  many executions of if branch of statements are performed.

Since loop performs  $\log n - 1$  many executions of the else branch of statements, we can conclude

$$\sum_{k=1}^{\log n - 2} \left( \frac{n}{2^{k+1}} + 1 \right) \tag{4}$$

many executions of the if branch are performed.

Then, since we know  $\log n \in \mathbb{N}$  due to n being in factors of 2, using the fact  $\forall n \in \mathbb{N}, \forall r \in \mathbb{R}, \sum_{i=0}^{n-1} r^i = \frac{1-r^n}{1-r}$ , we can calculate that

$$\sum_{k=1}^{\log n-2} \left( \frac{n}{2^{k+1}} + 1 \right) = \sum_{k=1}^{\log n-2} \frac{n}{2^{k+1}} + \sum_{k=1}^{\log n-2} 1$$
 (5)

$$= \frac{n}{2} \cdot \sum_{k=1}^{\log n - 2} \frac{1}{2^k} + \sum_{k=1}^{\log n - 2} 1 \tag{6}$$

$$= \frac{n}{2} \cdot \sum_{k=1}^{\log n - 2} \frac{1}{2^k} + (\log n - 2) \tag{7}$$

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

$$= n \cdot \left(1 - \frac{2}{n}\right) + (\log n - 2) \tag{9}$$

$$= n - 2 + \log n - 2 \tag{10}$$

$$= n + \log n - 4 \tag{11}$$

Now, adding the cost of the number of executions of else statements and the extra iteration taken to verify loop's terminating condition, we can conclude while loop has total of

$$n + \log n - 4 + (\log n - 1) + 1 = n + 2\log n - 4 \tag{12}$$

executions or iterations.

Since each execution takes a constant time (1 step), we can conclude while loop has cost of

$$1 \cdot (n+2\log n - 4) = (n+2\log n - 4) \tag{13}$$

steps.

Finally, adding constant time operations on line 2 to 4 (1 step), the algorithm has running time of

$$n + 2\log n - 3\tag{14}$$

which is  $\Theta(n)$ .

### Notes:

• I analyzed the example [0, 0, -1, -1]. This is what I found.

- iteration 1: if brach of statement executes and i increases by a 1 (i = 1, j = 1)

- iteration 2: if brach of statement executes and i increases by a 1 (i = 2, j = 1)

- iteration 3: if brach of statement executes and i increases by a 1 (i = 3, j = 1)

- iteration 4: else branch of statement executes, causing lst[i] = abs(lst[i]), i = 0, and j to increase by twice of its size (i = 0, j = 2)

The following is how the list looks after update

$$[0,0,1,-1]$$

- iteration 5: if brach of statement executes and i increases by a 2 (i = 2, j = 2)

- **iteration 6:** if brach of statement executes and i increases by a 2 (i = 4, j = 2)

- iteration 7: Loop terminates,

## Here's what I found about j

- Loop terminates when  $k+1^{th}$  execution of else statement is greater than or equal to n.

### Here's what I found about i

– When j=1, loop performs  $\frac{n}{2}+1$  executions, stops at  $\frac{n}{2}+1$ 

– When j=2, loop performs  $\frac{n}{4}+1$  executions, stops at  $\frac{n}{2}+2$ 

- loop terminates 1 after

 Number of execution of if branch of statements depend on the number of execution of else branch of statements

num of exec. of if statements = 
$$\sum_{k=0}^{\text{num of exec. of else}} \left( \frac{n}{2^k + 1} + 1 \right)$$
 (15)

- I analyzed the example [0,0,0,0,-1,-1,-1,-1]. This is what I found.
  - **iteration 1:** if brach of statement executes and i increases by a 1 (i = 1, j = 1)
  - iteration 2: if brach of statement executes and i increases by a 1 (i = 2, j = 1)
  - iteration 3: if brach of statement executes and i increases by a 1 (i = 3, j = 1)
  - iteration 4: if brach of statement executes and i increases by a 1 (i = 4, j = 1)
  - iteration 5: else branch of statement executes, causing lst[i] = abs(lst[i]), i = 0, and j to increase by twice of its size (i = 0, j = 2)

The following is how the list looks after update

$$[0,0,0,0,1,-1,-1,-1]$$

- iteration 6: if brach of statement executes and i increases by a 2 (i = 2, j = 2)
- iteration 7: if brach of statement executes and i increases by a 2 (i = 4, j = 2)
- iteration 8: if brach of statement executes and i increases by a 2 (i = 6, j = 2)
- iteration 9: else branch of statement executes, causing lst[i] = abs(lst[i]), i = 0, and j to increase by twice of its size (i = 0, j = 4)

The following is how the list looks after update

$$[0,0,0,0,1,-1,1,-1]$$

- **iteration 10:** if brach of statement executes and i increases by a 4 (i = 4, j = 4)
- **iteration 11:** if brach of statement executes and i increases by a 4 (i = 8, j = 4)
- iteration 12: Loop terminates,

### Here's what I found about j

- Loop terminates when  $k+1^{th}$  execution of else statement is greater than or equal to n.

### Here's what I found about i

– When j=1, loop performs  $\frac{n}{2}+1$  iterations, stops at  $\frac{n}{2}+1$ 

- When j=2, loop performs  $\frac{n}{4}+1$  iterations, stops at  $\frac{n}{2}+2$
- When j=4, loop performs  $\frac{n}{8}+1$  iterations, stops at  $\frac{n}{2}+4$
- Loop terminates 1 after
- Number of execution of if branch of statements depend on the number of execution of else branch of statements

num of exec. of if statements = 
$$\sum_{k=0}^{\text{num of exec. of else}} \left( \frac{n}{2^k + 1} + 1 \right)$$
 (16)

- I analyzed the example [0, 0, 0, 0, 0, 0, 0, 0, -1, -1, -1, -1, -1, -1, -1, -1]. This is what I found.
  - **iteration 1:** if brach of statement executes and i increases by a 1 (i = 1, j = 1)
  - **iteration 2:** if brach of statement executes and i increases by a 1 (i = 2, j = 1)
  - iteration 3: if brach of statement executes and i increases by a 1 (i = 3, j = 1)
  - iteration 4: if brach of statement executes and i increases by a 1 (i = 4, j = 1)
  - iteration 5: if brach of statement executes and i increases by a 1 (i = 5, j = 1)
  - **iteration 6:** if brach of statement executes and i increases by a 1 (i = 6, j = 1)
  - iteration 7: if brach of statement executes and i increases by a 1 (i = 7, j = 1)
  - **iteration 8:** if brach of statement executes and i increases by a 1 (i = 8, j = 1)
  - iteration 9: if brach of statement executes and i increases by a 1 (i = 9, j = 1)
  - iteration 10: else branch of statement executes, causing lst[i] = abs(lst[i]), i = 0, and j to increase by twice of its size (i = 0, j = 2)

The following is how the list looks after update

$$[0,0,0,0,0,0,0,1,-1,-1,-1,-1,-1,-1,-1]$$

- **iteration 11:** if brach of statement executes and i increases by a 2 (i = 2, j = 2)
- iteration 12: if brach of statement executes and i increases by a 2 (i = 4, j = 2)
- iteration 13: if brach of statement executes and i increases by a 2 (i = 6, j = 2)
- **iteration 14:** if brach of statement executes and i increases by a 2 (i = 8, j = 2)
- **iteration 15:** if brach of statement executes and *i* increases by a 2 (i = 10, j = 2)
- iteration 16: else branch of statement executes, causing lst[i] = abs(lst[i]), i = 0, and j to increase by twice of its size (i = 0, j = 4)

The following is how the list looks after update

$$[0, 0, 0, 0, 0, 0, 0, 0, 1, 1, -1, -1, -1, -1, -1, -1]$$

- iteration 17: if brach of statement executes and i increases by a 4 (i = 4, j = 4)
- **iteration 18:** if brach of statement executes and i increases by a 4 (i = 8, j = 4)
- **iteration 19:** if brach of statement executes and *i* increases by a 4 (i = 12, j = 4)
- iteration 20: else branch of statement executes, causing lst[i] = abs(lst[i]), i = 0, and j to increase by twice of its size (i = 0, j = 8)

The following is how the list looks after update

$$[0,0,0,0,0,0,0,0,1,1,-1,1,-1,-1,-1,-1]$$

- iteration 21: if brach of statement executes and i increases by a 8 (i = 8, j = 8)
- **iteration 22:** if brach of statement executes and *i* increases by a 8 (i = 16, j = 8)
- iteration 23: Loop terminates.

## Here's what I found about j

\* Loop terminates when  $k + 1^{th}$  execution of else statement is greater than or equal to n.

### Here's what I found about i

- \* When j=1, if branch performs  $\frac{n}{2}+1$  executions, stops at  $\frac{n}{2}+1$
- \* When j=2, if branch performs  $\frac{n}{4}+1$  executions, stops at  $\frac{n}{2}+2$
- \* When j=4, if branch performs  $\frac{n}{8}+1$  executions, stops at  $\frac{n}{2}+4$
- \* When j=8, if branch performs  $\frac{n}{16}+1$  executions, stops at  $\frac{n}{2}+8$
- \* Loop terminates 1 after
- \* Number of execution of if branch of statements depend on the number of execution of else branch of statements

num of exec. of if statements = 
$$\sum_{k=0}^{\text{num of exec. of else}} \left( \frac{n}{2^k + 1} + 1 \right)$$
 (17)

#### Notes:

- Realized the need to learn how to organize ideas for proof
- Realized the need to learn how to connect the dots or lay structure to proofs given sets of ideas
- Realized concepts involved are 1. finding examples 2. finding patterns in example
  3. generalizing patterns 4. write how am i going to solve problem 5. lay out big
  ideas 6. chunk out big ideas into smaller parts 7. solve the small parts

- Realized building a large proof without organizing ideas feels like jumping into solving pramp problems without pseudocode on how to solve it.
  - I wonder how to lay pseudocode or organize ideas for proofs...
- Realized I am keep losing details because my brain can't hold too much of information.
- Realized writing proof feels similar to writing algorithms

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# Question 4