

Question 1:

1. The claim is True.

From the Given we receive that $f(n) = \Theta(g(n)) \implies \exists c_1, c_2, n_0 > 0$ such that $\forall n > n_0$ we get that $0 \leq c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$

From the given we also know that $g(n) = \Theta(h(n)) \implies \exists d_1, d_2, n_1: \forall n > n_1$ such that $0 \leq d_1 \cdot h(n) \leq g(n) \leq d_2 \cdot h(n)$

We want to prove the following: $\exists e_1, e_2, n_2 > 0: \forall n > n_2: 0 \leq e_1 h(n) \leq f(n) \leq e_2 h(n)$

We will define n_2 such that $n_2 = \max\{n_0, n_1\} : \forall n > n_2$ the following is true:

$$f(n) \leq c_2 \cdot g(n) \leq c_2 \cdot d_2 h(n)$$

$$f(n) \geq c_1 \cdot g(n) \geq c_1 \cdot d_1 h(n)$$

We will choose $e_1 = \min\{c_1, d_1, c_1 \cdot d_1\}$, $e_2 = \max\{c_2, d_2, c_2 \cdot d_2\}$, (c_1, d_1, c_2, d_2 are all positive constants therefore e_1, e_2 are also bigger than 0) Therefore we will get that $0 \leq e_1 h(n) \leq f(n) \leq e_2 h(n) \implies$ by definition $f(n) = \Theta(h(n))$ and thus the claim is correct. ■

2. The claim is False.

We will show such with a counter example: $f(n) = n$, $g(n) = \frac{1}{n}$

$$f(n) + g(n) = n + \frac{1}{n} = \mathcal{O}(n)$$

$$f(n) \cdot g(n) = n * \frac{1}{n} = 1 = \mathcal{O}(1)$$

Therefore, $\exists n_0, c_1 > 0 : \forall n > n_0$ such that

$$0 \leq n + \frac{1}{n} = f(n) + g(n) \leq c_1 f(n) \cdot g(n) = c_1$$

but this is a contradiction because $\forall n_0 \exists n > n_0 : c_1 < n + \frac{1}{n}$ but we get that $n + \frac{1}{n} \leq c_1$

Thus, the Claim is incorrect ■

3. The claim is False.

We will show such with a counter example: $f(n) = \frac{1}{n}$, $g(n) = n^2$

let there be some $c > 0$ such that for $n_0 = \min\{1, c\} : \forall n > n_0$ the following happens:

$$c \cdot f(n) = c \cdot \frac{1}{n} \leq c \cdot \frac{1}{n_0} \leq c \cdot 1 = c \leq n^2$$

Therefore, $g(n) = \omega(f(n))$ but we will show that $f(n) + g(n) \neq \mathcal{O}(f(n))$ and from a theorem we will also get that $f(n) + g(n) \neq \Theta(f(n))$

$$\forall n_1 \exists n > n_1 : f(n) + g(n) > c \cdot f(n)$$

$$\frac{1}{n} + n^2 > c \cdot \frac{1}{n} / \cdot n \implies 1 + n^3 > c \implies n^3 > c - 1 \implies n > \sqrt[3]{c-1}$$

We will choose $n = \max\{n_1, \sqrt[3]{c-1}\}$ such that we get $f(n) + g(n) \neq \mathcal{O}(f(n))$

Therefore, according to the theorem mentioned before, $f(n) + g(n) \neq \Theta(f(n))$ ■

4. The claim is True.

It is given that $\forall c > 0 \exists n_0 > 0 : \forall n > n_0$ such that: $0 \leq g(n) < c \cdot f(n)$

We would like to show that there

$$\exists n_1, c_1, c_2 > 0 : \forall n > n_1 : 0 \leq f(n) + g(n) \leq f(n) + c \cdot f(n) = (c + 1) \cdot f(n)$$

Let $c_2 = c + 1$ such that we get by definition that $f(n) + g(n) = \mathcal{O}(f(n))$

Furthermore, Lets choose $c_1 = 1$ such that we get $0 \leq 1 \cdot f(n) = f(n) \leq f(n) + g(n)$ and we know that $f(n), g(n) \geq 0$ from the given.

Therefore, $f(n) + g(n) = \Omega(f(n))$

Now, according to a theorem since we found that $f(n) + g(n) = \mathcal{O}(f(n))$ and $f(n) + g(n) = \Omega(f(n)) \implies f(n) + g(n) = \Theta(f(n))$ ■

5. The claim is True.

Let $m = \log(n)$ Therefore, n^ϵ using log theorems is the same as $2^{\epsilon \cdot m} = (2^\epsilon)^m$

From the given rule $\forall a \in \mathbb{R}_{>1}$ and $d \in \mathbb{R}$ the following is true: $n^d = o(a^n)$

$$*(\log(n))^b \rightarrow m^b$$

Using this rule, we get $m^b = o((2^\epsilon)^m) \implies f(n) = o(n^\epsilon) = o(g(n))$ Thus proving the desired. ■

6. The claim is True.

We would like to prove the following: $\forall c > 0 \exists n_0 > 0 \forall n > n_0 : 0 \leq c \cdot n! < 2^{n \log n}$

Let there be some $c > 0$. We will choose $n_0 = c : \forall n > n_0$ the following occurs:

$$2^{n \log n} = 2^{\log n^n} > c \cdot n! > c \cdot n_0! \geq 0$$

(The third move is true because $n \cdot n \dots n > n \cdot (n - 1) \dots 1$ and with log rules, $2^{\log(x)} = x$)

Therefore, by definition we get that $f(n) = \omega(n!)$ ■

7. The claim is True.

For $k > 2$ where $k \in \mathbb{R}$, $f(n) = s^{\frac{n}{k} \cdot \log n} = \sqrt[k]{n^n}$ According to log rules.

We want to prove that $\forall c \exists n > n_0 : 0 < \sqrt[k]{n^n} < c \cdot n!$ which implies that $f(n) = o(n!)$

Let there be some arbitrary $c > 0$ that for some $n_0 = \max\{1, \frac{1}{c}\}$, $\forall n > n_0, n \in \mathbb{N}$ such that

$$\sqrt[k]{n^n} = n^{\frac{n}{k}} \leq n^{\frac{n}{2}}$$

now we will show that $n^{\frac{n}{2}} \leq n!$ which is the same as proving $n^n < (n!)^2$

$$(n!)^2 = (1 \cdot n)(2 \cdot (n - 1))(3 \cdot (n - 2)) \dots ((n - 2) \cdot 3)((n - 1) \cdot 2)(n \cdot 1) = n \cdot (2n - 2) \cdot (3n - 6) \dots (3n - 6)(2n - 2) \cdot n = \prod_{k=1}^n k \cdot (n - (k - 1))$$

we will show that $k(n - (k - 1)) > n$ for all $1 < k < n$ since when $k = 1, n$ we get that the expression equals n . We will open up the expression:

$$k(n+1-k) - n > 0 \implies k \cdot n + k - k^2 - n > 0 \implies n \cdot (k-1) - k \cdot (k-1) > 0 \\ \implies (k-1) \cdot (n-k) > 0 \implies \text{the expression is positive } \forall n > 2, n > k > 1$$

Thus we get

$$1 \cdot n = n, 2 \cdot (n-1) > n, 3 \cdot (n-2) > n, \dots, (n-2) \cdot 3, 2 \cdot (n-1) > n, n \cdot 1 = n \\ \implies \text{if we multiply each of these expressions we end up getting } (n!)^2 > n^n \text{ thus} \\ \text{proving what we wanted.}$$

Which is a total length of n separate multiplications where each multiplication is greater or equal to n .

Therefore, we get the desired and we showed that $(n!)^2 > n^n \implies n! > n^{\frac{n}{2}}$

Therefore, we get that $\sqrt[k]{n^n} = n^{\frac{n}{k}} \leq n^{\frac{n}{2}} < n! \leq n_0 \cdot n! < c \cdot n! \implies f(n) = o(n!) \blacksquare$

Question 2

1. The claim is True.

We know that there exists $c > 0$ that for every $n \in \mathbb{Z}_{>0}$ exists $I, |I| = n$ such that $T_A(I) \geq cg(n)$. We also know that $T_A(n) = \max_{I: |I|=n} T_A(I)$. For every instance $I, |I| = n$ in $\{I_n\}$ we get that $0 \leq cg(n) \leq T_A(I) \leq T_A(n)$.

Therefore, by the definition there exists (The same c, n_0 that we got from the given) $c, n_0 > 0$ that for every $n > n_0$, $0 \leq cg(n) \leq T_A(n)$. Thus, $T_A(n) = \Omega(g(n))$.

2. The claim is False.

Function A is a counter example:

```
n = A.length
for i to n do {
  i = 2 * i
}
```

For this algorithm, the time complexity is $\Theta(\log n)$, so by the definition there exists $c_1, c_2, n_0 > 0$ such that $\forall n > n_0 : 0 \leq c_1 \cdot \log n \leq T_A(n) \leq c_2 \cdot \log n$

so we can see that $T_A(I) \leq T_A(n) \leq c_2 \cdot \log n \leq c \cdot n^2 \implies T_A(I) \leq c \cdot n^2$
for every $n > n_0$ but $T_A(n) \neq \Omega(n)$ like we asked since the upper bound is $\log n$. \blacksquare

3. The claim is True.

If there exists $c, n_0 > 0$ that for all $n \geq n_0$ and all the instance $I, |I| = n$ there $T_A(I) < c \cdot g(n) \leq c \cdot g(n)$ so by the definition:
 $T_A(I) = O(g(n)) \implies T_A(n) = \max_{I: |I|=n} T_A(I) = O(g(n))$.

4. The claim is False.

A counter example:

```
n = A.length
If n % 2 = 0{
  for i = 0 to n^3 do{
    print("hi")
```

```

}}
else{
print("by")
}

```

we can see that for every $n_0 > 0$ there will be an instance where n is odd, where $n \geq n_0$ such that $T_A(I) \leq c \cdot n^2$ since the runtime is $\Theta(1)$ shown by the last line in the function. But there is an instance I that $\forall n \geq n_0 : T_A(I) = \Theta(n^3)$ which can be seen in the for loop which runs from i to n^3 . Therefore $T_A(n) = \max_{I:|I|=n} T_A(I) \neq O(n^2)$.

Question 3:

Line 1 takes $\Theta(1)$ time.

Every iteration of the inner loop in lines 7-9 takes $\Theta(1)$ time.

In one iteration of the inner loop in lines 4-9, the most inner loop it runs at a worst case of $\log(n)$ iteration ($2^k = n \rightarrow k = \log(n)$). Therefore, in every iteration of the most inner loop (line 7) we get $(\log(n) \cdot \Theta(1)) = \Theta(\log(n))$ runtime.

In one iteration of the for outer loop in lines 2-9, the first inner loop run at the worst case $\log(n)$ iteration ($2^j = n \rightarrow j = \log(n)$). Therefore, in every iteration of the for loop we get $(\log(n) \cdot \Theta(\log(n))) = \Theta((\log(n))^2)$ time.

The for loop runs $\lfloor \frac{n}{2} \rfloor$ to $n = \Theta(n)$ time, so the time complexity of the algorithm overall is the multiplication of each inner loop which results in: $(\Theta(n) \cdot \Theta((\log(n))^2)) = \Theta(n((\log(n))^2))$. ■

Question 4:

Firstly we will analyze the function called *FindPow*(n)

we start on lines 1,2 by instantiating k, p to $k = 1, p = 0$ which takes $\Theta(1)$ time each.

given n as an input we have a *while* loop on line 3 that runs from $k = 1$ to n where the condition for when $k = n$ occurs according to line 4 when $2^k = n \implies k = \log n$

p is equal to the amount of iterations that the *while* loop runs which we found to be $\log n$ thus *FindPow*(n) will return the value of $p - 1$ which is $\log n$.

calculating asymptotic bound of $Alg_4(n)$

on line 1 we call the *FindPow* function and give it the value n . Therefore, $p := \log n$ and so far the function will run $\Theta(\log n)$ which is equal to the asymptotic runtime of *FindPow*(n).

all the lines that are relevant to *binary* variable run $\Theta(1)$ time and therefore since we found that the minimal runtime so far is $\log n$ thus it won't affect the asymptotic runtime of the function.

on line 3, $n = n - 2^p = n - 2^{\log n} = n - n = 0 \implies n = 0$

on line 4 we have a *loop* that runs from $p - 1$ to 0 meaning that it runs $\log n$ times.
 we have a variable called *NewP* on line 5 that is defined to be $FindPow(n)$ when $n = 0$ and n doesn't get changed since the condition on line 6 is never true until $i = 0$ which only occurs on the last iteration of the loop. Therefore, until then n remains 0. In the last iteration when it's True then the function will run lines 7, 8 which are anyways have $\Theta(1)$ runtime. In all the other iterations as the first condition is false we will enter the *else* condition on line 9, 10 which will run a total of $\Theta(1)$ time.

In addition line 11 which just returns the binary variable will run $\Theta(1)$ time.

In total we get that the functions runs:

$$\Theta(\log(n)) + 2 \cdot \Theta(1) + \Theta(\log(n)) \cdot \Theta(1) \implies \text{runtime} = \Theta(\log(n)) \implies \Omega(\log n) \text{ and } \mathcal{O}(\log n) \blacksquare$$

Question 5:

We will analyze the runtime of $Alg_5(n)$.

Lines 1 run $\Theta(1)$ and instantiates the following variables with the value of 0: x, y, k

Line 2 is a loop that runs from 1 to $2n$ thus runs at $\Theta(n)$.

The first inner loop on line 3 runs a total of i^2 throughout each iteration of i

The second inner loop starts at a value of $t = k := i^2$

Lines 7 through 11 are negligible as they run $\Theta(1)$ each. Therefore, as the runtime will be greater than $\Theta(1)$ we can ignore their runtime relative to the total function runtime.

The second inner loop (*while*) in each i iteration will run $2 \cdot i^2$ time since line 12 tells us that $t- = 0.5$ in each of the inner loops iteration.

We have two equations here that their runtime is in parallel.

Therefore the total runtime is equal to the following:

$$\sum_{i=1}^{2n} i^2 + \sum_{i=1}^{2n} (2 \cdot i^2) = \frac{1}{3} \cdot (n \cdot (8 \cdot n^2 + 6 \cdot n + 7)) + \frac{2}{3} \cdot (n \cdot (8 \cdot n^2 + 6 \cdot n + 7)) = \Theta(n^3) + \Theta(n^3) = \Theta(n^3)$$

Total runtime = $\Theta(n^3)$ ■