

Problem 1 (10 pts)

1. True False

$$IF, f(n) \in \Theta(g(n)), \text{ then } f(n) + g(n) \in \Theta(g(n))$$

True

Because if $f(n)$ is an element of $\Theta(g(n))$, it must contain the time of $f(n) + g(n)$, because adding the two together would only result in something like $2n$, which when removing the constant from the equation, is simply n .

2. True False

$$n^p \in \Omega(\log n)$$

for any constant $p > 0$.

True

Because in cases for Sigma, we consider the best case bound for the problem. The best case for such a problem as described above would be contained within $\Sigma(\log n)$ because there could be scenarios where the p is a lower number and does not take a large amount of time.

3. True False

$$n^2 \in \Theta(n \log n)$$

False

Because n^2 is not going to be equal to $n \log n$, as the second term in $n \log n$, will necessarily be less than n . This means that the two are not equal, so n^2 is not an element of $\Theta(n \log n)$. In order for it to be an element of the theta argument, it must be equal or approaching equal.

4. True False

$$n! \in \Omega(2^n)$$

True

Because performing n factorial operations takes less time than performing 2 to the n operations. So $n!$ is within the set of times covered by 2^n .

5. True False

$$n^3 + n^2 \log n \in \Theta(n^3)$$

True

Because n to the 3rd is our most impactful term, and since it is so much larger than the other terms present within the function, we can ignore them. Thus, we can consider the term as an element of $\Theta(n^3)$, because the two terms are effectively equal.

6. True False

The worst-case time complexity of Euclid's algorithm is $O(\log n)$ where n is the larger of the two input integers.

True

Because Euclid's algorithm will never go through all the terms in order to find it, but may go through a section of the terms equal to $\log n$, in the worst cases. It can also take only constant time.

7. True False

For any two functions f and g , we always have

$$f \in \Omega(g) // \text{ or } // g \in \Omega(f)$$

False

Because you could have something like \sin and \cos waves, meaning that the outcomes could be variable.

8. True False

Asymptotically, an $O(\log n)$ algorithm is faster than an $O(n)$ algorithm

True

Because asymptotically is for values at upper range, $\log n$ will always be faster than n , because $\log n$ is less than n . In other cases, where n is very small, this could be false.

9. True False

The time complexity of dequeue in a queue containing n elements implemented using an array is $\Theta(1)$.

True

Because we can implement a circular array using modulo

10. True False

A $\Theta(n)$ algorithm always runs faster than a $\Theta(n^2)$ algorithm.

False

Because there may be some cases where the n^2 algorithm is very small, and the n algorithm is larger, meaning that the n^2 algorithm would run faster. Were it most cases, this would be true, but because it is always, this is False.

Problem 2 (10 pts)

Find the time complexity of the following sorting algorithm in terms of the number of comparisons. Show your work and express the final result in Θ notation.

Algorithm 1 ExchangeSort($A[0, \dots, n-1]$)

```

1: for  $i = 0, \dots, n-2$  do
2:   for  $j = i+1, \dots, n-1$  do
3:     if  $A[i] > A[j]$  then
4:       swap( $A[i], A[j]$ )
5:     end if
6:   end for
7: end for

```

$$\sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 = \sum_{i=0}^{n-2} n - 1 - (i+1) + 1 = \sum_{i=0}^{n-2} (n-i-1) = \frac{n(n-1)}{2} \in \theta(n^2)$$

Method

Each for loop is a summation. Because it is a nested for loop we make a nested summation. The top of the summation is the top term, base case the bottom term. Set it equal to 1, because it's a Theta comparison. Top term minus bottom term the inner summation. Add +1 to the end because we count up from 0. Resolve it and get the next term. Top term in outer summation multiplies with the inner expression, divide it by 2 because it's two summations. Set $i = 0$. Final term, show that final term is an element of Θn^2 because n^2 is the most significant term in our final expression.

Problem 3 (10 pts)

Buggy Cody has recently switched majors to computer science. He just learned recursion and was excited to use it whenever possible. Below is a pseudo code he wrote to find if a string (S_1) is contained in another string (S_2) using recursion. After turning it into a Java program, he found that it produced correct results, but ran extremely slow for long strings S_2 . Let the length of S_1 and S_2 be m and n , respectively, i.e., $S_1.length() = m$ and $S_2.length() = n$. The syntaxes $S(1 : \text{end} - 1)$ and $S(2 : \text{end})$ represent the strings with the last character and first character removed from S , respectively. strcmp(S_1, S_2) returns true if the string S_1 is identical to string S_2 , and false otherwise. Please help Mr. Cody identify his algorithm's time complexity in terms of the number of strcmp() operations.

Algorithm 2 isSubstring(S_1, S_2)

```

1: if  $S_1.length() > S_2.length()$  then
2:   return false
3: else
4:   if  $S_1.length() == S_2.length()$  then
5:     return strcmp( $S_1, S_2$ )
6:   else
7:     return isSubstring( $S_1, S_2(2 : \text{end})$ ) OR isSubstring( $S_1, S_2(1 : \text{end} - 1)$ )
8:   end if
9: end if

```

Let $N = n-m$

$T(N) = 2T(N-1)$

$T(0) = 1$

$$T(N) = 2T(N-1) = 2^2T(N-2) = 2^3T(N-3) \dots = 2^N T(0) = 2^N$$

Method

Big N represents the way we go through the strings, which is n times, minus the length of m , our second string. $T(N)$ represents overall time complexity. We can traverse our time complexity terms over, and over again, finding $n-1, 2, 3, \dots$ all the way till $n-n$ which is 0. That cancels, and leaves us with 2^N , which represents how many comparisons we've made overall.

Bonus rewrite

Alg isSubstring(S_1, S_2)

if $S_1.len() > S_2.len()$

return false

else

if strcmp($S_1, S_2(1:len())$) == true

return true

else

return isSubstring($S_1, S_2(2:end)$)