

Algorithm Engineering

Lab 1 – Lecture

In this lab, you will learn-

1. **design** algorithms
 - a. come up with solutions
 - b. generate test inputs
 - c. debug
2. **analyze** algorithms
 - a. time complexity
 - b. space complexity
 - c. compare with other approaches/algorithms
3. **optimize** algorithms
 - a. make it faster
 - b. make it use less space

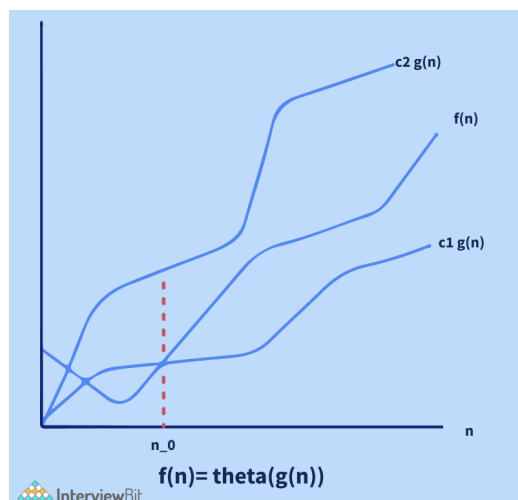
Asymptotic Notations

Asymptotic analysis is a technique that is used for determining the efficiency of an algorithm that does not rely on machine-specific constants and avoids the algorithm from comparing itself to the time-consuming approach. For asymptotic analysis, asymptotic notation is a mathematical technique that is used to indicate the temporal complexity of algorithms.

The following are the three most common asymptotic notations.

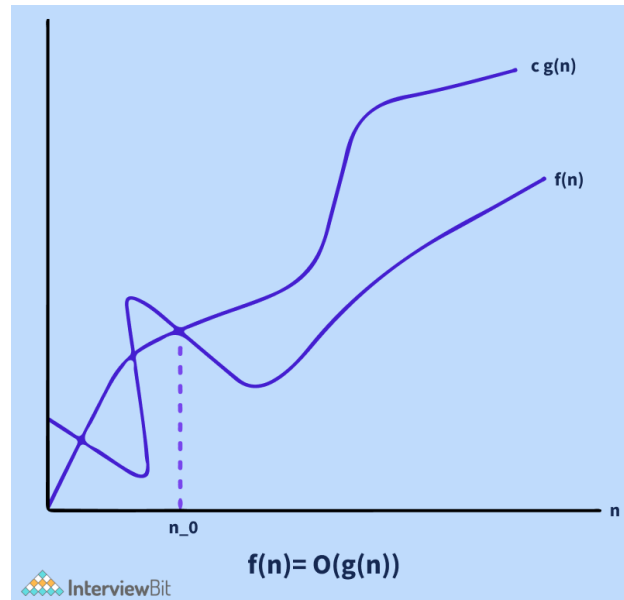
- **Big Theta Notation: (θ Notation)**

The exact asymptotic behavior is defined using the theta (θ) Notation. It binds functions from above and below to define behavior. Dropping low order terms and ignoring leading constants is a convenient approach to get Theta notation for an expression.



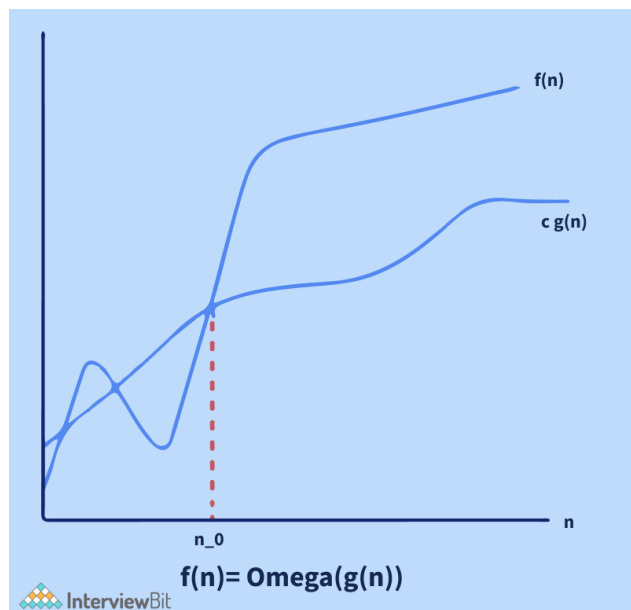
- **Big O Notation:**

The Big O notation defines an upper bound for an algorithm by bounding a function from above. Consider the situation of insertion sort: in the best-case scenario, it takes linear time, and in the worst case, it takes quadratic time. Insertion sort has a time complexity $O(n^2)$. It is useful when we just have an upper constraint on an algorithm's time complexity.



- **Big Omega (Ω) Notation:**

The Ω Notation provides an asymptotic lower bound on a function, just like Big O notation does. It is useful when we have a lower bound on an algorithm's time complexity.



Finding Time Complexity of Programs with Loops

$O(n)$

```
for i in range(1,N):  
    print(i)  
for i in range(1,N):  
    print(i)
```

$O(n^2)$

```
for i in range(1,N):  
    print(i)  
    for j in range(1,N):  
        print(j)
```

$O(n)$ [as the inner loop runs finite times; the complexity is $O(2n)$ to be exact]

```
for i in range(1,N):  
    print(i)  
    for j in range(1,N):  
        if j>=2:  
            break  
        print(j)
```

Have to use summation to find out the exact complexity for the following snippet

```
for i in range(1,N):  
    print(i)  
    for j in range(1,i):  
        print(j)
```

```
i = 1, j = 1  
i = 2, j = 1, 2  
i = 3, j = 1, 2, 3  
...  
i = N, j = 1, 2, ... N
```

So, the total number of operations = $1 + 2 + 3 + \dots + N = O(N*(N+1)/2) = O(N^2)$

Finding Time Complexity of Programs with Recursion

Code Snippet:

```
int f(int n){
    if(n<=1)
        return 1;
    return f(n-1) + f(n-1);
}
```

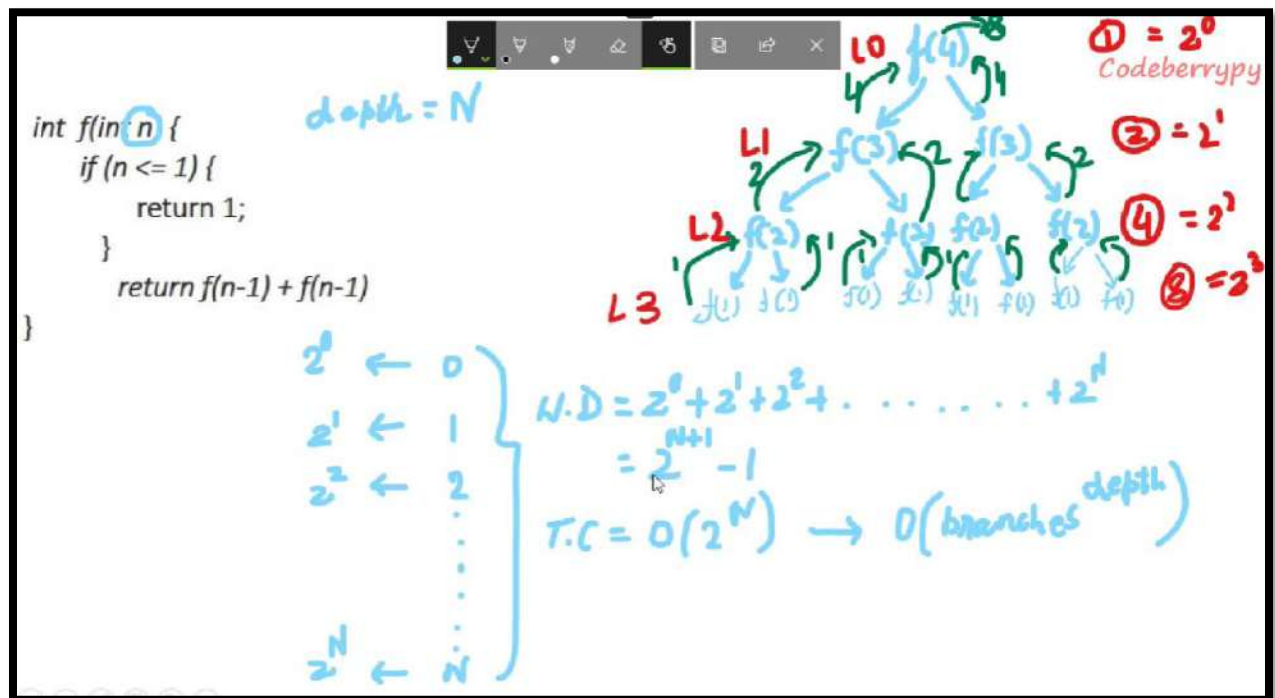
Complexity Formulation:

$$T(n) = 1 + T(n-1) + T(n-1) = 1 + 2T(n-1)$$

For this particular problem, the time complexity of a subproblem with size n is the 1 plus twice the time complexity of subproblem of size $n-1$.

Solution:

Draw the recursion. Count the number of operations in each node. Add them up.



Answer: $O(2^N)$

Reference:

<https://youtu.be/NyV0d5QadWM>

<https://youtu.be/ncpTxqK35PI>

Tasks

Solve the attached tasks in **Lab1_Tasks.pdf** and write the solution in a report using Latex/Overleaf. Upload one pdf containing the report. Rename the submitted pdf as **LabNo_LabGroup_FullStudentID.pdf** (Lab1_1A_160041010.pdf)

Use the following template for generating report; copy the following overleaf project and edit it to prepare the report.

<https://www.overleaf.com/read/kdpyrcpytkfx>