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# Design and Analysis of Algorithms (DAA)

## Assignment - 1

1. Asymptotic notations are mathematical notations used to describe the time or space complexity of algorithms.

1. Big O notation: This notation represents the upper bound  $(O)$  of the algorithm's growth rate.
2. Omega notation: This notation represents the lower bound  $(\Omega)$  of algorithm's growth rate.
3. Theta notation: This notation represents the tight bound  $(\Theta)$  of the algorithm's growth rate.

ex:- Bubble sort has complexity of  $O(n)$ , in worst case it may be  $O(n^2)$

2. The value of  $i$  is doubled  $i = i * 2$  if  $i$  starts from  $i$  ~~to~~ it is doubled every time until  $i$  is in range of 1 to  $n$ .

Loop will terminate when  $2^K \geq n$   
So  $K \geq \log_2(n)$

So the complexity will be  $O(\log n)$

3.

$$f(n) = 3f(n-1)$$
$$f(n-1) = 3f(n-2)$$
$$f(n-2) = 3f(n-3)$$

Every call multiplies size by 3, so it forms a GP with common ratio of 3 & it takes  $n$  steps



So time complexity will be  $O(3^n)$

4.  $T(n) = 2T(n-1) \quad n > 0$

Every recursive call reduces the problem size by 1 in case of  $n=0$  function will return a constant value of 2

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

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

$$T(n-2) = 2T(n-3)$$

$$T(1) = 2T(0)$$

Recursive calls forms AP with steps equal to 1 until  $n > 0$

So time complexity -  $O(n)$

```
5. int i=1, s=1;
   while (s <= n) {
       i++;           -> n times
       s = s+i;       -> n times
       printf("#");  -> n times
   }
```

Time complexity will be  $O(n)$  since every line is implemented  $n$  times since loop runs from 1 to  $n$ .

```
6. void function (int n) {
    int i, count = 0;
    for (i=1; i <= n; i++) {
        count++;
    }
}
```

This loop will iterate while  $i^2 \leq n$   
initially  $i=1$  will continue while  $i^2 \leq n$

So the time complexity will be  $O(\sqrt{n})$   
since  $i$  will run until  $i \leq \sqrt{n}$ .

7. void function (int n) {  
    int i, j, k, count = 0;  
    for (i = 1; i <= n; i++) {  $\rightarrow n/2$   
        for (j = 1; j <= n; j = j \* 2)  $\rightarrow \log_2 n$   
            for (k = 1; k <= n; k = k \* 2) {  $\rightarrow \log_2 n$   
                count++;  
            }  
        }  
    }

In outer loop will be  $O(n/2)$  since it iterates  $n/2$  times similarly middle & inner loop the value is doubled every time so it will be  $\log_2 n$ .

$$\text{So } n/2 * \log_2 n * \log_2 n$$

$$= O(n \log_2 n^2)$$

8. function (int n) {  $\rightarrow T(n)$   
    if (n == 1) return;  
    for (i = 1 to n) {  $\rightarrow n$   
        for (j = 1 to n) {  $\rightarrow n^2$   
            print("\*");  
        }  
    }

function (n-3)  $\rightarrow T(n-3)$



So time complexity will be

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

Since we don't consider constants

$$\text{So } O(n^2)$$

9. 

```
void function(int n) {  
    for (i=1 to n) {  
        for (j=1; j<=n; j=j*i) {  
            print (" * ")  
        }  
    }  
}
```

here i will increment with 1, 2, 3, 4  
As j will increment with 1, 3, 6, 10

$$\begin{aligned} i=1 & \quad n \\ i=2 & \quad n/2 \\ i=3 & \quad n/3 \\ \text{+ } n/2 + n/3 + \dots \end{aligned}$$

so time complexity would be  $O(n \log n)$

10.  $n^k$  ( $k \geq 1$ )  $C^n$  ( $C > 1$ )  
 $C^n$  grows faster than  $n^k$ .

11. Complexity of adding a node in heap would be  $O(1)$   
but in worst case scenario the swapping will  
be done  $H$  times so  $O(H)$ , for a complete  
binary tree  $O(\log n)$  so overall complexity  
for insertion in minheap would be  $O(\log N)$