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Design & Analysis of Algorithms

1. Asymptotic Notations

Asymptotic Notations are the mathematical notations used to describe the running time of an algorithm when the input tends towards a particular value or a limiting value.

There are mainly three asymptotic notations :-

- i) Big-O notations
- ii) Omega (Ω) notations
- iii) Theta (Θ) notations.

I. Big-O notations

It represents the upper bound of the running time of an algorithm.

$O(g(n)) = \{ f(n) : \text{there exist positive constants } c \text{ &} n_0 \text{ such that } 0 \leq f(n) \leq c g(n) \text{ for all } n \geq n_0 \}$

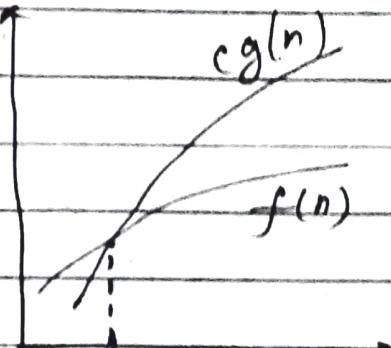
where $\epsilon > 0$ and $n \geq n_0$.

- This notation is known as the upper bound of the algorithm, or a worst case of an algorithm.

Ex. $f(n) = 3\log n + 100$
 $g(n) = \log n$

$$3\log n + 100 \leq c \times \log(n)$$

$$c = 150 \text{ & } n \geq 2 \text{ (undefined at } n=1)$$

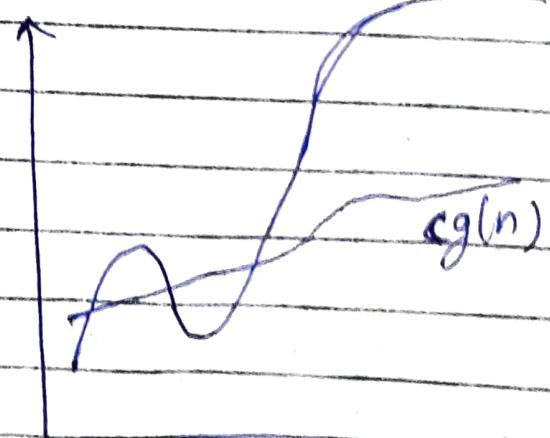


II. Omega (Ω) Notation

- Omega notation represents the lower bound of the running time of an algorithm.

$$\Omega(g(n)) = \{f(n) : \text{there exist positive constant } c \text{ & } n_0 \text{ such that } 0 \leq c g(n) \leq f(n)\}$$

for all $n, n \geq n_0$



- This notation is known as the lower bound of the algorithm, or a best case of an algorithm.

$n = \text{no. of input}$

Example: $f(n) = 3n + 2$

$c \cdot g(n) \leq f(n)$

$c = \text{constant } g(n) = n$

$$c \cdot n \leq 3n + 2$$

$$cn - 3n \leq 2$$

$$n(c-3) \leq 2 \Rightarrow n \leq \frac{2}{c-3}$$

if we assume $c=4$, then $n_0 = 2$

$$c=4, n_0 = 2$$

III Theta (Θ) notation

- Theta notation encloses the function from above & below since it represents the upper & the lower bound of the running time of an algorithm.

$\Theta(g(n)) = \{f(n) : \text{there exist positive constant } c_1, c_2 \text{ & } n_0 \text{ such that}$

$$0 \leq c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$$

for all $n > n_0$.

- This is known as tight bounds of an algorithm or average case of algorithm.

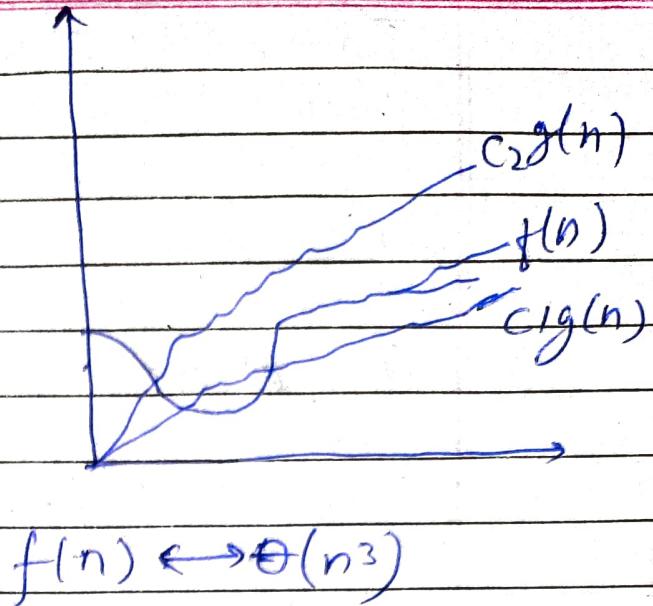
- Example

$$f(n) = 5n^3 + 16n^2 + 3n + 8$$

$$= 5n^3 \leq 5n^3 + 16n^2 + 3n + 8 \leq (5+16+3+8)n^3$$

$$= 5n^3 \leq f(n) \leq 32n^3$$

$$C_1 = 5, C_2 = 32, n_0 = 1$$



$$f(n) \leftrightarrow \Theta(n^3)$$

Ques 2 :- $i = 1, 2, 4, 8, 16, \dots, \text{kth term} \dots n$

$$G_n = a \cdot 2^{n-1}$$

$$G_n = 1(2)^{k-1}$$

$$n = 2^{k-1}$$

$$\log_2 n = (k-1) \log_2 2$$

$$k = \log_2 n + 1$$

$$O(n) = \log n$$

Ans 3 $T(n) = 3T(n-1)$

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

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

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

$$T(n) = 3 \times 3 \times 3T(n-3)$$

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

$$T(n-3) = 3T(n-4)$$

$$T(n) = 3^3 \times 3T(n-4)$$

$$T(n) = 3^4 \times T(n-4)$$

⋮

⋮

⋮

General form

$$T(n) = 3^i T(n-i) - 1 \quad T(0) = 1$$

$$T(n-i) = T(0)$$

$$n-i = 0 \quad [n=i]$$

Put $n=i$ in Eq, I

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

$$T(n) = 3^n T(0)$$

$(T(0) = 1 \{ \text{Given} \})$

$$T(n) = 3^n$$

$$T(n) = O(3^n)$$

$$\text{Ans-4} \quad T(n) = \{ 2T(n-1) - 1 \quad \text{if } n \geq 0 ; \text{ otherwise } 1 \}$$

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

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

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

$$T(n) = 2^2 (2T(n-3) - 1) - 2 - 1$$

⋮

⋮

$$T(n) = 2^i (T(n-i) - (2^{i-1} + 2^{i-2} + 2^{i-3} + \dots + 1))$$

$$T(n) = T(n-i) + T(i)$$

$$n-i=0$$

$$n=i$$

④ Time complexity = $O(1)$

Ans 5 & 11 :- No. of step (k)

	i	j
0	0	1
1	1	2
2	3	3
3	6	4
4	10	5
5	15	6
6	21	7
7	1	1
8	1	1
9	1	1
K step	n	1

$$T(n) = O(k)$$

$$i = 0, 1, 3, 6, 10, \dots, n.$$

$$S_n = 1 + 3 + 6 + 10 + 15 + \dots + n$$

$$S_n = 1 + 3 + 6 + 10 + \dots + (n-1) + n$$

\rightarrow \leftarrow

$$0 = 1 + 2 + 3 + 4 + 5 + \dots + n$$

$$n = 1 + 2 + 3 + 4 + \dots + K \text{ step}$$

$$n = \frac{k}{2} [2(1) + (k-1)1]$$

$$2n = k[2 + k-1]$$

$$2n = k^2 + k \Rightarrow 2n = (k+1/2)^2 - (1/2)^2$$

$$2n + (1/2)^2 = (k+1/2)^2$$

$$k+1/2 = \sqrt{2n + (1/2)^2}$$

$$k = \sqrt{2n + (1/2)^2} - 1/2$$

$$T(n) = T(k)$$

$$T(n) = T(\sqrt{2n + (1/2)^2} - 1/2)$$

$$T(n) = O(\sqrt{n})$$

Ans. 6 void function (int n) {
 int i, count = 0;

for (i=1; i*i <= n; i++)

count++

}

Ans. 6 \rightarrow Since, i is moving from 1 to \sqrt{n} with linear growth so,

$$T(n) = O(\sqrt{n})$$

Ans-7:- void function (int n)

```

int i, j, k, count = 0;
for (i = n/2; i <= n; i++)
    for (j = 1; j <= n; j = j*2)
        for (k = 1; k <= n; k = k*2)

```

Count ++;

}

⇒ $O(n \log n \log n)$
 ⇒ $O(n(\log n)^2)$

B. function (int n) → $T(n)$

```

    {
        if (n == 1) return;
        for (i = 1 to n) → n
            {
                for (j = 1 to n) → n
                    {

```

printf ("*"); } }

function (n-1); } → $T(n-1)$

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

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

$$T(n) = T(n-3) + n^2 + (n-1)^2 + (n-2)^2$$

⋮

General Term

$$T(n) = T(n-1) + n^2 + (n-1)^2 + (n-2)^2 + \dots + (n-i)^2$$

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

$$n = i+1 \Rightarrow [n-1 = i]$$

$$T(n) = T(n-(n-1)) + n^2 + (n-1)^2 + (n-2)^2 + \dots + (n-(n-1))^2$$

$$T(n) = T(1) + n^2 + (n-1)^2 + (n-2)^2 + \dots + 1^2$$

$$T(n) = 1 + 1^2 + 2^2 + 3^2 + \dots + n^2$$

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

$$\boxed{T(n) = O(n^3)}$$

9. Time complexity of

void function (int n)

for (i=1; i < n; i)

 for (j=1; j <= n; j++)

 print ("#")

3

⇒ O(n \sqrt{n})

10. If $c > 1$, then the exponential c^n for outgrows any term, so that ans is

$$n^k \text{ is } O(c^n)$$

$$12. T(n) = T(n-1) + T(n-2) + c$$

$$T(n-2) \leq T(n-1)$$

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

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

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

$$T(n) = 2^2 T(n-2) + 2c + c$$

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

$$T(n) = 2^2(2T(n-3) + c) + 2c + c$$

$$T(n) = 2^3 T(n-3) + 2^2 c + 2c + c$$

⋮

⋮

General Term

$$T(n) = 2^i T(n-i) + (2^0 + 2^1 + 2^2 + \dots + 2^{i-1})c$$

$$n-i=0$$

$$\boxed{n=i}$$

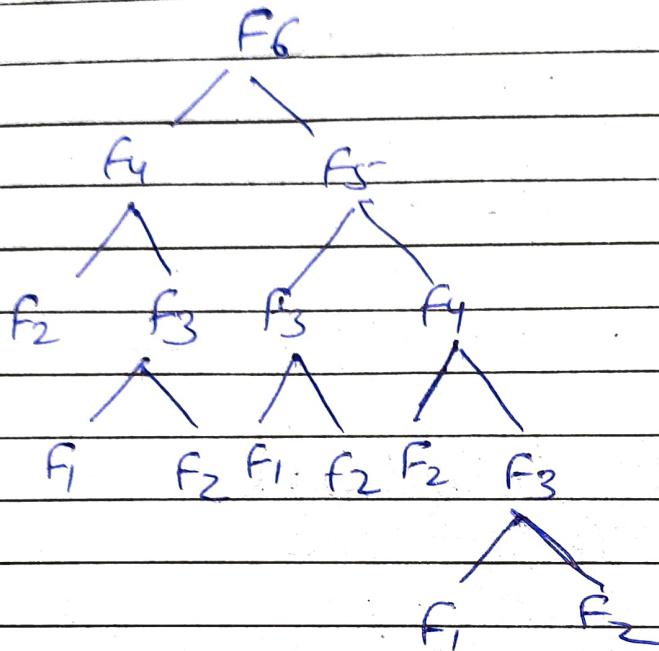
$$T(n) = 2^n T(0) + (2^0 + 2^1 + 2^2 + \dots + 2^{n-1})c$$

$$T(n) = 2^n(1) + \frac{2^0(2^{n-1} - 1)}{2-1} c$$

$$T(n) = 2^h(1+c) + c$$

$$T(n) = O(2^n)$$

fit(6)



The maximum depth is proportional to the N ; hence the space complexity of fibonaci recursive is $O(n)$.

$$13. \text{ i) } n \cdot \log n$$

```
void fun()  
int i,j;
```

for (i=1; i<=n; i++)

for (j=0; j<=n; j=j+2)
 minff("##");

printf ("\\n");

3

3

ii) n^3

```

void fun(int n)
{
    int i, j, k;
    for(i=0; i<=n; i++)
    {
        for(j=0; j<=n; j++)
        {
            for(k=0; k<=n; k++)
                printf("%#");
        }
    }
}

```

iii) $\log(\log n)$

```

void SieveofEratosthenes(int n)
{
    bool prime[n+1];
    memset(prime, true, sizeof(prime));
    for(int p=2; p*p <=n; p++)
    {
        if(prime[p] == true)
        {
            for(int i = p*p; i<=n; i+=p)
                prime[i] = false;
        }
    }
}

```

for (int p=2; p<=n; p++)
 if (prime[p])

cout << p << endl;

}

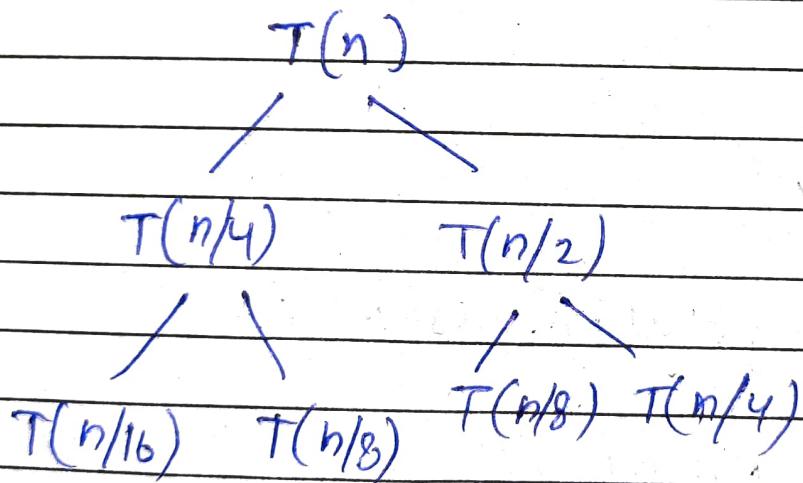
14.

$$T(n) = T(n/4) + T(n/2) + cn^2 \quad T(1) = c$$

$$\Rightarrow n = n/2$$

$$T(n/2) = T(n/8) + T(n/4) + c(n^2/4)$$

$$T(n) = T(n/4) + 2T(n/16) + c(n^2/16) + n^2/4 + n^2$$



$$T(n) = c \left[n^2 + \frac{5n^2}{16} + \frac{25n^2}{256} + \dots \right]$$

$$T(n) = \pi^2 c \left[1 + \frac{5}{16} + \frac{25}{256} + \dots \right]$$

$T(n) = O(n^2)$

15. for $i=1$, the inner loop is executed n times
 for $i=2$, the inner loop is executed $n/2$ times.

for $i=3$, the inner loop is executed $n/3$ times.
 for $i=4$, the inner loop is executed $n/4$ times

for $i=n$, the inner loop is executed n/n times.

$$\begin{aligned}\text{Total time} &= n + n/2 + n/3 + \dots + n/n \\ &= n(1 + 1/2 + 1/3 + \dots + 1/n) \\ &= n \log n\end{aligned}$$

$$T(n) = O(n \log n)$$

16. $O(\log(\log n))$

18. (a) $100, \log \log n, \log n, \sqrt{n}, n, n \log n, n^2, 2^n, 2^{2n}, 4^n, n!$

(b) $1, \log(\log(n)), \sqrt{\log n}, \log n, \log(2^n), \log(n!), 2 \log(n), n, 2n, 4n, n \log(n), n^2, 2(2^n), n!$

(c) $96, \log_8 n, \log_2 n, \log(n!), 5n, n \log_6 n, n \log_2 n, 8n^2, 7n^3, 8^{2n}, n!$

Q. LINEAR SEARCH (A, key)

Comp $\leftarrow 0$, $j \leftarrow 0$

for $i = 1$ to $A.length$

Comp \leftarrow Comp + 1

if $A[i] == \text{key}$

print "Element founded"

$j = 1$

if $j == 0$

print "Element not found"

print Comp

Q. INSERTION SORT (A)

for $j = 2$ to $A.length$

Key $\leftarrow A[j]$

$i \leftarrow j - 1$

while $i > 0$ and $A[i] > \text{key}$

$A[i+1] \leftarrow A[i]$

$i = i - 1$

$A[j+1] \leftarrow \text{key}$

Recursive Method of insertion sort

INSERTION-SORT (A, n)

if $n \leq 1$

return

INSERTION-SORT ($A, n-1$)

key = $A[n-1]$;

$j = n-2$;

while $j \geq 0$ & $A[j] > \text{key}$

$A[j+1] = A[j]$

$j = j-1$

$A[j+1] = \text{key}$

Insertion sort considers one input element per iteration & produces a partial solution without considering future elements that why it is called online sorting

Other sorting algorithm that have been discussed in lectures are :-

- Bubble sorting
- Selection sorting
- Quick sort
- Merge sort
- Heap sort
- Counting sort

Q1.

	Best case	Average case	Worst case
Bubble Sort	$\Omega(N)$	$\Theta(N^2)$	$O(N^2)$
Selection Sort	$\Omega(N^2)$	$\Theta(N^2)$	$O(N^2)$
Insertion Sort	$\Omega(N)$	$\Theta(N^2)$	$O(N^2)$
Merge Sort	$\Omega(N \log N)$	$\Theta(N \log N)$	$O(N \log N)$
Heap Sort	$\Omega(N \log N)$	$\Theta(N \log N)$	$O(N \log N)$
Quick Sort	$\Omega(N \log N)$	$\Theta(N \log N)$	$O(N^2)$
Counting Sort	$\Omega(N+K)$	$\Theta(N+K)$	$O(N+K)$

Q2.

	In Place	Stable	Online
Bubble Sort	Yes	Yes	Yes
Insertion Sort	Yes	Yes	Yes
Selection Sort	Yes	No	Yes
Merge Sort	No	Yes	Yes
Quick Sort	Yes	No	Yes
Heap Sort	Yes	No	Yes
Count Sort	No	Yes	Yes

Q3.

Linear Search

LINEAR SEARCH (A , key)

$found \leftarrow 0$

for $i=1$ to N

if $A[i] == key$

$found \leftarrow 1$

print "Element found"

break

If found = 20

print "Element Not found"

Time complexity - $O(n)$

Space complexity - $O(1)$

Binary Search (Iterative)

BINARY-SEARCH (A, beg, end, key)

while beg \leq end

 mid = beg + (end - beg) / 2

 if mid = key

 return mid

 if A[mid] < key

 beg = mid + 1

 if A[mid] > key

 end = mid - 1

return -1

Time complexity - $O(\log_2 n)$

Space complexity - $O(1)$

Binary Search (Recursion)

BINARY-SEARCH (A, beg, end, key)

If end \geq beg

 mid = (beg + end) / 2

if $A[mid] == item$

return $mid + 1$

else if $A[mid] < item$

return BINARY-SEARCH ($A, mid + 1, end, key$)

else,

return BINARY-SEARCH ($A, beg, mid - 1, end$)

return -1

Time complexity - $O(\log n)$

Space complexity - $O(1)$

$$24. T(n) = T(n/2) + c$$

————— X —————