

1. Asymptotic Notation :- It defines the time taken by an algorithm to run for a given input

ii Big O notation $O()$ \rightarrow It represents the upper bound or the maximum time that an algorithm can take to execute.

Eg $\rightarrow O(n) \rightarrow$ Accessing elements of 1-D array

$O(n^2) \rightarrow$ Bubble sort

$O(n \log n) \rightarrow$ quick sort

iii Omega Notation $\Omega()$ \rightarrow It represents the lower bound or the minimum time that an algorithm can take to execute.

Eg $\rightarrow \Omega(1) \rightarrow$ Searching an element in 1-D array
(If the element is at 1st position)

$\Omega(n) \rightarrow$ Bubble sort (if we are given sorted array)

iiii Theta Notation $\Theta()$ \rightarrow It represents the lower as well as upper bound or the average time that an algorithm can take to execute.

Eg $\rightarrow \Theta(n) \rightarrow$ Searching element

$\Theta(n^2) \rightarrow$ Bubble Sort

2. Time complexity of

for ($i=1$ to n) $\rightarrow n$

{

$i = i * 2;$

}

$i = 1, 2, 4, 8, \dots, n$

$n = 2^{k-1}$

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

$n = \frac{2^k}{2}$

$2^k = 2^k$

$\log_2 2^n = \log_2 2^k$

$\log_2 2 + \log_2 n = k \log_2 2$

$k = 1 + \log_2 n$

Time complexity = $O(\log_2 n)$

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

$T(n) = 3T(n-1), n > 0$

$= 1$

, otherwise

$T(0) = 1$

$T(n) = 3T(n-1) \dots (i)$

put $n = n-1$

$T(n-1) = 3T(n-2) \dots (ii)$

$T(n-1) = 3T(n-2) \dots (ii)$

put (ii) in (i)

$T(n) = 3 \cdot 3T(n-2) \dots (iii)$

put $n = n-2$

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

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

put in (iii)

$T(n) = 3 \cdot 3 \cdot 3T(n-3) \dots$

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

put $n-k = 1$

$k = n-1$

$T(n) = 3^{n-1} T(n-n+1)$

$= \frac{3^n}{3} \cdot T(1)$

$= \frac{1}{3} \cdot 3^n$

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

$$4. \quad T(n) = \begin{cases} 2T(n-1) - 1; & n > 0 \\ 1 & \text{otherwise} \end{cases}$$

$$T(0) = 1$$

$$T(n) = 2T(n-1) - 1 \quad \text{--- (i)}$$

put $n = n-1$ in (i)

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

$$T(n-1) = 2T(n-2) - 1 \quad \text{--- (ii)}$$

put in eq (ii)

$$T(n) = 2 \cdot 2T(n-2) - 1 - 1 \quad \text{--- (iii)}$$

put $n = n-2$

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

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

put in eq (iii)

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

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

put $n-k = 1$

$$k = n-1$$

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

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

$$= O(2^n)$$

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5. int i=1, s=1;
   while (s<=n)
   {
       i++;
       s=s+i;
       printf("%d #", i);
   }

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}

$i = 2, 3, 4, 5, 6, \dots, k$

$s = 1 + 2 + 3 + 4 + \dots + n$

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

$= 1 + (n-1)$

$T \rightarrow O(n)$

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++;

}

}

}

$k = 1, 2, 4, 8, \dots, n$

$n = 2^{k-1}$

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

$$n = \frac{2^k}{2}$$

$$2n = 2^k$$

$$\log_2 n = k \log_2 2$$

$$\boxed{k = \log_2 n}$$

Same for j loop

$$= \log_2 n$$

The first loop will execute $\frac{n}{2}$ times

$$T = \frac{n}{2} \times \log_2 n \times \log_2 n$$

$$= \frac{n}{2} \log_2^2 n$$

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

6. void function(int n) {

int i, count = 0

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

{

count++;

}

if n = 5

i = 1, 2

if n = 10

i = 1, 2, 3

if n = 20

i = 1, 2, 3, 4

So loop runs \sqrt{n} for all cases

i = 1, 2, 3, 4, ..., \sqrt{n}

$O(\sqrt{n})$

3. function (int n) {

if (n == 1):

return

for (i = 1 to n)

for (j = 1 to n)

printf("%d * %d"),

function(n-3);

}

4. void function (int n)

for (i = 1 to n) {

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

printf("%d * %d"),

}

}

}

i	j	
1	1, 2, 3, 4, ... n	n times
2	1, 3, 5, 7, ... n	n/2 times
3	1, 4, 7, ... n	n/3 times
⋮	⋮	⋮
n	1, 1+n, ...	n/n times

$$n \text{ times} \left| \frac{n}{1} + \frac{n}{2} + \frac{n}{3} + \dots + \frac{n}{n} \right.$$

$$= n \left(\frac{1}{1} + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \right) = O(\log n)$$

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