

23-02-2024 | DAA

Assignment - 1

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Semester: - IV

① Asymptotic notations is used to describe the running time of an algorithm - how much time an algorithm takes with a given input, n . The types of notations are:-

a) Big O notation (O)

Denotes the $[g(n)]$ tight upper bound of $f(n)$

$$f(n) = O(g(n))$$

$$\text{iff } f(n) = (g(n))$$

$$\forall n = n_0$$

for some constant $c > 0$

Example: -

$$f(n) = 2n^2 + 3n + 1$$

is $O(n^2)$

b) Big Omega notation (Ω)

Denotes the $[g(n)]$ tight lower bound of $f(n)$.

$$f(n) = \Omega(g(n))$$

$$f(n) = g(n)$$

$$\forall n = n_0$$

for some constant $c > 0$.

Example: -

$$f(n) = n^2 \text{ is } \Omega(n)$$

c) Theta (Θ) Notation

$$f(n) = \Theta(g(n))$$

Theta gives both tight upper bound and tight lower bound.

$$f(n) = \Theta(g(n))$$

$$f(n) = O(g(n)) \text{ and } \Omega(g(n))$$

$$f(n) = \Theta(g(n))$$

$$\text{i.e. } c_1 g(n) \leq f(n) \leq c_2 g(n)$$

$$\forall n \geq \max(n_1, n_2) \text{ and}$$

for some constant $c_1 > 0$ and $c_2 > 0$

Example :-

$$f(n) = n^2 \text{ is } \Theta(n^2)$$

d) Small O notation (o)

Denotes the $[g(n)]$ upper bound of $f(n)$

$$f(n) = o(g(n))$$

$$\text{i.e. } f(n) < g(n)$$

$$\forall n > n_0$$

$$(\text{for all}) \quad c > 0$$

Example :

$$f(n) = n \text{ is } o(n^2)$$

e) Small Omega (ω) notation
Denotes the strict lower bound of a function's growth rate.

$$f(n) = \omega(g(n))$$

$g(n)$ is lower bound of $f(n)$

$$f(n) > g(n)$$

$$\text{iff } f(n) > g(n)$$

$$\forall n > n_0$$

$$\text{for all } \forall c > 0$$

Example:

~~$$f(n) = \omega(g(n))$$~~

$$f(n) = n^2 \text{ is } \omega(n)$$

② for ($i = 1$ to n)

$$\{ \quad i = i * 2; \quad \}$$

}

$$\Rightarrow \text{for } (i = 1; i \leq n; i = i * 2)$$

$$\underline{1, 2, 4, 8, 16, \dots}$$

G.P.

$$n = a r^{K-1}$$

$$n = 1 \times 2^{K-1}$$

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

$$2n = 2^K$$

$$K = \log_2(2n)$$

$$K = \log_2(2) + \log_2(n)$$

$$K = 1 + \log_2 n$$

$$\text{Time complexity} = O(\log_2 n)$$

$$\textcircled{3} T(n) = \begin{cases} 3(T(n-1)) & \text{if } n > 0, \\ 1 & \text{otherwise} \end{cases}$$

$$T(0) = 1$$

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

$$\text{for } T(1)$$

$$\begin{aligned} T(1) &= 3T(0) \\ &= 3 \times 1 \end{aligned}$$

$$\text{for } T(2)$$

$$\begin{aligned} T(2) &= 3T(2-1) \\ &= 3T(1) \\ &= 3T(0) \\ &= 3 \times 3 \times 1 \end{aligned}$$

$$\text{for } T(3)$$

$$\begin{aligned} T(3) &= 3T(3-1) \\ &= 3T(2) \\ &= 3 \times 3 \times 3 \times 1 \end{aligned}$$

$$\text{for } T(4)$$

$$\begin{aligned} T(4) &= 3T(4-1) \\ &= 3T(3) \\ &= 3 \times 3 \times 3 \times 3 \times 1 \end{aligned}$$

for $T(n)$

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

$$= 3 \times 3 \times 3 \times 3 \times \dots \times 3$$

$$= 3^n$$

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

$$(9) T(n) = \begin{cases} 2T(n-1) - 1 & \text{if } n > 0, \\ 1 & \text{otherwise} \end{cases}$$

$$T(0) = 1$$

for $T = 1$

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

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

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

$$= 2 - 1$$

$$= 1$$

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

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

$$= 2 \times 1 - 1$$

$$= 2 - 1$$

$$= 1$$

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

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

$$= 2 \times 1 - 1$$

$$= 2 - 1$$

$$= 1$$

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

$$= 2n - (2n-2) - \dots - 4 - 2 - 1$$

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


```

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

```

after first iteration:-

$$s = s + 1$$

after second iteration:-

$$s = s + 1 + 2$$

it goes on for x iterations.

$$1 + 2 + \dots + x \leq n$$

$$(x * (x + 1)) / 2 \leq n$$

$$O(x^2) \leq n$$

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

```

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++;
}

```

x	i	j	k
1	1	1	1
2	1	1+1=2	1+1+1+1=4
3	1+1	1+1=2	1+1+1+1=4
4	1+1	1+1+1=3	1+1+1+1+1+1+1+1=8
5	1+1	1+1+1=3	1+1+1+1+1+1+1+1=8
⋮	⋮	⋮	⋮
8	1+1+1	1+1+1+1=4	1+1+1+1+1+1+1+1=16
⋮	⋮	⋮	⋮
x		$\log x$	

no. of k increases with power of 2^n , so we can say that it will iterate $\log n$ times and will k while i iterates n times.

$$O(n) = (n * \log_2 n)$$

```

(6) int i = 1, count = 0
    for (i = 1; i * i <= n; i++)
    {
        count++;
    }

```


for $i = 1$

iteration = 1

$i = 2$

iteration = 2

$i = 3$

iteration = 3

$i = 4$

iteration = 4

$i = 5$

iteration = 5

$i = 6$

iteration = 6

$i = n$

iteration = 1

we can say that time complexity for the function will be $O(1)$

8) function (int n)

{
if ($n == 1$)

return;

for ($i = 1$ to n)

{

for ($y = 1$ to n)

{

print (" $>$ " + i)

}

}

for $n = 1 \rightarrow O(1)$

for $n \geq 2 \rightarrow$

i	j	$i * j$	
1	1	1	
2	2	4	
3	3	9	
4	4	16	
⋮	⋮	⋮	
n	n	n^2	

Since ~~i~~ i runs n time for j running $i * j$ we can say Time complexity = n^2
T.C. = $O(n^2) \quad \forall n \geq 2$

for $n = 1$, T.C. = $O(1)$

Q function(int n)

{
 for (i = 1 to n)

{
 for (j = 1; j <= n; j = j + 1)
 {
 print(" * ");
 }
}

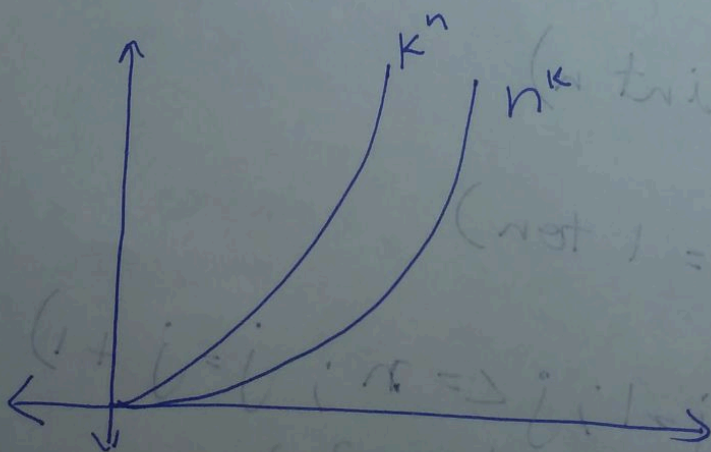
i	j	i	j
1	n	1	1
2	n/2	2	2
3	n/3	3	3
4	n/4	4	4
5	n/5	5	5
6	⋮	6	6
⋮	⋮	⋮	⋮
⋮	⋮	⋮	⋮
n	n/n = 1	⋮	⋮
n	log n	n	n

hence, we can say that:-

$$T.C. \text{ of } j = O(\log n)$$

$$T.C. \text{ of } i = O(n)$$

$$\text{nested} = O(n \log n)$$



For value of $n, k, c > 1$ all the value of

$$O(k^n) > O(n^k)$$

this is because (n) exponential time complexity is always greater than integer exponential.

for $n, k, c = 1$, $O(k^n) = O(n^k)$

and for;

$n, k, c < 1$, the condition is false & the program worst iterate once.