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1 - Asymptotic notations are mathematical notations used to describe the limiting behaviour of function as the input size approaches infinity.

The three main asymptotic notations are:-

(i) Big O Notation \Rightarrow Describes the upper bound or worst-case scenario of an algorithm's runtime or space complexity.

example \Rightarrow An algorithm with a time complexity of $O(n^2)$ means that its runtime grows quadratically as the input size n increases.

(ii) Omega notation (Ω) \Rightarrow Describes the lower bound or best-case scenario of an algorithm's runtime or space complexity. It represents minimum growth rate of a function.

example \Rightarrow An algorithm with time complexity of ~~$O(n)$~~ $\Omega(n)$ means that the algorithm's runtime grows linearly at least as the input size (n) increases.

(iii) Theta notation (Θ) \Rightarrow Describes both the upper and lower bounds of an algorithm's runtime. Represents the tightest possible growth rate of the function.

example \Rightarrow An algorithm has time complexity of $\Theta(n)$ means that the algorithm's runtime grows linearly as the input size (n) increases and it is bounded both from above and below by a linear function.

$$2 \rightarrow \begin{array}{l} \text{for (int } i=0; i < n; i+=2) \\ \{ \\ \quad \text{---} \\ \} \end{array} \quad \begin{array}{l} t_k = a x^{k-1} \\ n = (2)^{k-1} \end{array}$$

Taking \log_2 both sides

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

$$k = \log_2 n + 1$$

Time complexity $\Rightarrow O(\log_2 n)$

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

$$T(0) = 1$$

Substituting $n=1 \Rightarrow T(1) = 3T(0) = 3$

substituting $n=2 \Rightarrow T(2) = 3T(1) = 3^2$

substituting $n=3 \Rightarrow T(3) = 3T(2) = 3 \cdot 3^2 = 3^3$

substituting $n=k \Rightarrow T(k) = 3^k$

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

Time complexity $\Rightarrow O(3^k)$

$$4 \Rightarrow T(n) = \begin{cases} 2T(n-1) - 1 & \text{if } n > 0, \text{ otherwise } 1 \end{cases}$$

$$T(0) = 1$$

substituting $n=1 \Rightarrow T(1) = 2T(0) - 1 = 1 - 1 = 0$

substituting $n=2 \Rightarrow T(2) = 2T(1) - 1 = -1$

substituting $n=3 \Rightarrow T(3) = 2T(2) - 1 = -3$

$n=4 \Rightarrow T(4) = 2T(3) - 1 =$

$$\begin{aligned} T(n) &= 2T(n-1) - 1 = 2(2T(n-2) - 1) - 1 = 2^2 T(n-2) - 2 - 1 \\ &= 2^3 T(n-3) - 2^2 - 2 - 1 \\ &= 2^n T(n-n) - (2^{n-1} + 2^{n-2} + \dots + 2^1) \\ &= 2^n - n - 1 \end{aligned}$$

Time complexity $\Rightarrow O(2^n)$

5 ⇒

int i=1, s=1;

while(s <= n)

{

i++;

s = s+i;

printf("#");

O(1)

}

i = 1 2 3 4 5

s = 1 3 6 10 15

} →

$$s = \frac{i(i+1)}{2}$$

$$t_k = \frac{k(k+1)}{2} \Rightarrow n = \frac{k(k+1)}{2} \Rightarrow k^2 + k - 2n = 0$$

$$k = \frac{-1 \pm \sqrt{1+8n}}{2} = \frac{-1 + \sqrt{1+8n}}{2}$$

Time complexity ⇒ $O(\sqrt{n})$

6 ⇒

void function(int n)

{

int i, count=0;

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

{

count++;

}

}

i = 1 2 3 4

i² = 1 4 9 16

$$t_k = k^2$$

$$n = k^2 \Rightarrow k = \sqrt{n}$$

Time complexity ⇒ $O(\sqrt{n})$

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

~~$O(n/2) \times O(\log_2 n) \times O(\log_2 n)$~~ Time complexity = $O\left(\frac{n}{2} \times \log_2 n \times \log_2 n\right) = O\left(\frac{n}{2} (\log_2 n)^2\right)$

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

8 → function (int n)
 {
 if (n == 1) return;
 for (int i = 1; i < n; i++) → $O(n)$
 {
 for (j = 1; j < n; j++) → $O(n)$
 {
 printf(" ");
 }
 }
 function(n-3); → $T(n-3)$
 }

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

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

9 → void function (int n)
 {
 for (i = 1; i < n; i++)
 {
 for (j = 1; j <= n; j += i)
 {
 printf("%d ", i);
 }
 }
 }

i 1 2 3 4 5 → n times

j 1 3 6 10 15 → $\left(n + \frac{n}{2} + \frac{n}{3} + \dots + 1\right)$ times

n n/2 n/3 n/4
 times

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

10 \Rightarrow n^k $k \geq 1$ c^n $c > 1$

c^n grows faster than n^k .

- 11 — The time complexity for `extractMin()`, given a min heap of n nodes is $O(\log n)$. This is because `extractMin()` removes the root node which is the minimum element and then calls `heapify()` to restore the heap property. `Heapify()` takes $O(\log n)$ time as it traverses the height of the heap which is $\log n$.

12 \Rightarrow

