

① solve the following recurrence relation.

a) $x(n) = x(n-1) + 5$ for $n > 1$ with $x(1) = 0$

1) Write down the first two terms to identify the pattern.

$$x(1) = 0$$

$$x(2) = x(1) + 5 = 5$$

$$x(3) = x(2) + 5 = 10$$

$$x(4) = x(3) + 5 = 15$$

2) Identify the pattern (or) the general term

→ The first term $x(1) = 0$

The common difference $d = 5$

The general formula for the n^{th} term of an AP is

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

Substituting the given values

$$x(n) = 0 + (n-1) \cdot 5 = 5(n-1)$$

∴ The solution is $x(n) = 5(n-1)$

b) $x(n) = 3x(n-1)$ for $n > 1$ with $x(1) = 4$

1) Write down the first two terms to identify the pattern

$$x(1) = 4$$

$$x(2) = 3x(1) = 3 \cdot 4 = 12$$

$$x(3) = 3x(2) = 36$$

$$x(4) = 3x(3) = 108$$

2) Identify the general term

→ The first term $x(1) = 4$

→ The common ratio $r = 3$

The general formula for the n^{th} term of a gp is

$$x(n) = x(1) \cdot r^{n-1}$$

Substituting the given values

$$x(n) = 4 \cdot 3^{n-1}$$

\therefore The solution is $x(n) = 4 \cdot 3^{n-1}$

c) $x(n) = x(n/2) + n$ for $n > 1$ with $x(1) = 1$ (solve for $n = 2^k$)

for $n = 2^k$, we can write recurrence in terms of k .

1) Substitute $n = 2^k$ in the recurrence

$$x(2^k) = x(2^{k-1}) + 2^k$$

2) Write down the first few terms to identify the pattern

$$x(1) = 1$$

$$x(2) = x(2^1) = x(1) + 2 = 1 + 2 = 3$$

$$x(4) = x(2^2) = x(2) + 4 = 3 + 4 = 7$$

$$x(8) = x(2^3) = x(4) + 8 = 7 + 8 = 15$$

3) Identify the general term by finding the pattern
we observe that:-

$$x(2^k) = x(2^{k-1}) + 2^k$$

we sum the series:-

$$x(2^k) = 2^k + 2^{k-1} + 2^{k-2} + \dots$$

Since $x(1) = 1$:

$$x(2^k) = 2^k + 2^{k-1} + 2^{k-2} + \dots$$

\therefore The geometric series with the term $a = 2$ and the last term 2^k except for the additional term.

The sum of a geometric series s with ratio $r = 2$

is given by $s = a \frac{r^n - 1}{r - 1}$

where, $a = 2$, $r = 2$ and $n = k$.

$$s = \frac{2^k - 1}{2 - 1} = 2(2^k - 1) = 2^{k+1} - 2$$

Adding the +1 term

$$x(2^k) = 2^{k+1} - 2 + 1 = 2^{k+1} - 1$$

∴ Solution is

$$x(2^k) = 2^{k+1} - 1$$

d) $x(n) = x(n/3) + 1$ for $n > 1$ with $x(1) = 1$ (solve for $n = 3^k$)
For $n = 3^k$, we can write the recurrence in terms of k .

1) substitute $n = 3^k$ in recurrence

$$x(3^k) = x(3^{k-1}) + 1$$

2) Write down the first few terms to identify the pattern

$$x(1) = 1$$

$$x(3) = x(3^1) = x(1) + 1 = 1 + 1 = 2$$

$$x(9) = x(3^2) = x(3) + 1 = 2 + 1 = 3$$

$$x(27) = x(3^3) = x(9) + 1 = 3 + 1 = 4$$

3) Identify the general term:-

$$\text{we observe that } x(3^k) = x(3^{k-1}) + 1$$

Summing up the series

$$x(3^k) = 1 + 1 + 1 + \dots + 1$$

$$x(3^k) = k + 1$$

∴ The solution is $x(3^k) = k + 1$

2) Evaluate the following recurrences complexity

i) $T(n) = T(n/2) + 1$, where $n = 2^k$ for all $k \geq 0$

The recurrence relation can be solved using iteration method.

i) Substitute $n = 2^k$ in the recurrence.

2) Iterate the recurrence

$$\text{for } k=0: T(2^0) = T(1) = T(1)$$

$$k=1: T(2^1) = T(2) = T(1) + 1$$

$$k=2: T(2^2) = T(4) = T(n) + 1 = (T(1) + 1) + 1 = T(1) + 2$$

$$k=3: T(2^3) = T(8) = T(n) + 1 = (T(1) + 2) + 1 = T(1) + 3$$

3) Generalize the pattern

$$T(2^k) = T(1) + k$$

$$\text{Since } n = 2^k, k = \log_2 n$$

$$T(n) = T(2^k) = T(1) + \log_2 n$$

4) Assume $T(1)$ is a constant c

$$T(n) = c + \log_2 n$$

\therefore The solution is $T(n) = O(\log n)$

ii) $T(n) = T(n/3) + T(2n/3) + cn$ where c is constant and n is input

The recurrence can be solved using the masters theorem for divide-and-conquer recurrence of the form

$$T(n) = aT(n/b) + f(n)$$

where, $a=2$, $b=3$ and $f(n) = cn$.

Let's determine the value of $\log_b a$:

$$\log_b a = \log_3 2$$

(3)

using the properties of logarithms

$$\log_3 2 = \frac{\log 2}{\log 3}$$

Now, we compare $f(n) = cn$ with $n^{\log_3 2}$:

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

$$n = n'$$

Since $\log_3 2$ we are in third case of master's theorem

$$f(n) = O(n^c) \text{ with } c > \log_b a$$

\therefore The solution is $T(n) = O(f(n)) = O(cn) = O(n)$

(3) Consider the following recurrence algorithm

$\min[A[0, \dots, n-2]]$

if $n=1$ return $A[0]$

Else $\text{temp} = \min(A[0, \dots, n-2])$

if $\text{temp} \leq A[n-1]$ return temp

else

return $A[n-1]$

a) what does this algorithm compute?

The given algorithm, $\min[A[0, \dots, n-1]]$ computes the min value in the array 'A' from index '0' for 'n-1'. It does this by recursively finding the minimum value in the sub array $A[0, \dots, n-2]$ and then comparing it with the last element $A[n-1]$ to determine the overall max value.

b) Set up a recurrence relation for the algorithm basic operation count and solve it.

To determine the recurrence relation for the algorithm's basic operation count, let's analyse the steps involved in the algorithm. The basic operations are the comparison and function calls.

Recurrence relation Setup.

1) Base case when $n=1$, the algorithm performs a single operation to return $A[v]$.

2) Recursive case. when $n>1$, the algorithm makes a recursive call to $\min(A[0, \dots, n-2])$: performs a comparison b/w temp and $A[n-1]$.

Let $t(n)$ represent the no. of basic operation the algorithm performs for an array of size 'n'.

1) Base case :-

$$T(1) = 1$$

2) Recursive case :-

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

Here $T(n-1)$ accounts for the operations performed by the recursive call to $\min(A[0, \dots, n-2])$ and the $+1$ accounts for the comparison b/w temp and $A[n-1]$.

To solve this recurrence relation we can use iteration method:

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

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

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

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

$$= n$$

\therefore The Solution is $T(n) = n$

This means the algorithm performs 'n' basic operations for an input array of size n.

(4)

④ Analyze the order of growth

i) $f(n) = 2n^2 + 5$ and $g(n) = 7n$ use the $\Omega(g(n))$ notation

To analyse the order of growth and use the Ω notation, we need to compare the given function $f(n)$ and $g(n)$.

given functions:-

$$f(n) = 2n^2 + 5$$

$$g(n) = 7n$$

order of growth using $\Omega(g(n))$ notation.

The notation $\Omega(g(n))$ describes a lower bound on the growth rate that for sufficiently large n , $f(n)$ grows at least as $g(n)$

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

Let's analyse $f(n) = 2n^2 + 5$ with respect to $g(n) = 7n$

1) Identify Dominant terms:

→ The dominant terms in $f(n)$ is $2n^2$ since it grows faster than constant terms as 'n' increases.

→ The dominant term in $g(n)$ is $7n$.

2) Establish the inequality:-

→ we want to find constants 'c' and n_0 such that

$$2n^2 + 5 \geq c \cdot 7n \text{ for all } n \geq n_0$$

3) Simplify the inequality:-

→ Ignore the lower order term 5 for larger

$$2n^2 \geq 7cn$$

→ Divide both sides by n

$$2n \geq 7c$$

→ solve form: -

$$n \geq 7\frac{1}{2}$$

4) choose constants: -

$$\text{let } c=1$$

$$n \geq \frac{7 \cdot 1}{2} = 3.5$$

∴ for $n \geq n$, the inequality holds:

$$2n^2 + 5 \geq 7n \text{ for all } n \geq n$$

we have shown that there exist constant $c=1$ and $n_0=n$ such that for all $n \geq n_0$:

$$2n^2 + 5 \geq 7n$$

Thus, we can conclude that: -

$$f(n) = 2n^2 + 5 = \Omega(7n)$$

in Ω notation, the dominant term $2n^2$ in $f(n)$ clearly grows faster than $+n$. Hence

$$f(n) = \Omega(n^2)$$

However, for the specific comparison asked $f(n) = \Omega(7n)$ is also correct.

showing that $f(n)$ grows at least as fast as $7n$.