

Solve the following recurrence relation.

a) $x(n) = x(n-1) + 5$ for $n \geq 1$ 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 $= 5$

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) $x(n) = 3x(n-1)$, $n \geq 1$ 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

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

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

- c) $x(n) = x(n/2) + n$ for $n \geq 1$ with $x(1) = 1$ (solve for $n=2^k$)
for $n=2^k$, we can write recurrence in terms of k .
- i) substitute $n=2^k$ in the recurrence.

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

- ii) write down the first few terms to identify the pattern

$$x(1) = 1$$

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

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

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

- iii) 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 $x(1) = 1$!

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

The geometric series with the term $a=2$ and the last term 2^k except for the additional $+1$ term.

The sum of a geometric series with ratio $r=2$ is given by

$$S = a \frac{r^n - 1}{r - 1}$$

Here $a=2$, $r=2$ and $n=k$:

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

Adding the $+1$ term

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

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

(3)

- d. $x(n) = x(n/3) + 1$ for $n \geq 1$ with $x(1) = 1$ (solve for $n = 3^k$)
 for $n = 3^k$, we can write the recurrence in terms of k
 * substitute $n = 3^k$ in the recurrence

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

* write down the first few terms to identify the pattern.

$$x(1) = 1$$

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

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

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

* Identify the general term:

we observe that:

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

summing up the series

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

$$x(3^k) = k + 1 \text{, the solution is } x(3^k) = k + 1$$

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

ii) iteration the recurrence

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

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

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

iii) Generalize the pattern

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

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

* assume $T(1)$ is a constant C .

$$T(n) = C + \log_3 n$$

The solution is

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

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

The recurrence can be solved using the master 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 $\log_b a$:

$$\log_b a = \log_3 2$$

using the properties of logarithms

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

now we compare $f(n) = cn$ with $n^{\log_3 2}$.

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

$$n = n^1$$

since $\log_3 2 < 1$ we are in the third case of the master's theorem.

$$f(n) = O(n^c)$$
 with $c > \log_3 2$

The solution is:

$$T(n) = O(f(n)) = O(cn) = O(n)$$

consider the following recurrence algorithm?

$\min(A[0 \dots n-2])$

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

else temp = $\min(A[0 \dots n-2])$

if temp <= $A(n-1)$ return temp

else return $A(n-1)$

what does this algorithm compute?

The given algorithm, $\min[A[0 \dots (n-1)]$ computes the minimum value in the array 'A'. It form index '0' for '(n-1)'. It does it by recurrence finding the minimum value in the sub array $A[0 \dots (n-2)]$ and then computing it with the last element $A[n-1]$ to determine the maximum value.

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

To determine the recurrence relation for the algorithm basic operation count, let's analyze the step involved in the algorithm. The basic operations are the comparisons 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 $\min[A[0 \dots (n-2)]$; Perform 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 .

* Base case

$$T(1)=1$$

* 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 $[n-1]$.

To solve this recurrence relation we can use iteration method.

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

The solution is,

$$T(n) = n$$

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

4. Analyze the Order of growth.

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

To analyze 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 fast as $g(n)$.

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

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

* Identify dominant terms.

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

* Establish the inequality.

→ We want to find constant c and n_0 such that:

$$2n^2 + 5 \geq cn \text{ for all } n \geq n_0$$

* Simplify the inequality:

* ignore the lower order term 5 for larger

$$2n^2 \geq cn$$

* Divide both sides by n .

$$2n \geq cn$$

* Solve for n

$$n \geq c/2$$

x choose constant

let $c=1$

$$n \geq \frac{7.1}{2} = 3.5$$

\therefore for $n \geq n_0$, the inequality holds:

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

We have shown that there will be $c=1$ and $n_0 \in \mathbb{N}$ 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 $7n$. 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 $7n$.