

① Solve the following recurrence relation.

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

(i) write down the first two terms to identify the pattern

$$x(1) = 0$$

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

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

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

(ii) Identify the pattern (or) the general term.

$\Rightarrow$  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  $x(1) = 4$

(i) write down the first two terms to identify the pattern

$$x(1) = 4$$

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

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

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

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  $n > 1$  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.

$$x(1) = 1$$

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

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

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

(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:

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

$$\text{Since } x(1) = 1$$

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

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

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

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

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

relation is

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

(a)  $x(n) = x(n/3) + 1$  for  $n > 1$  with  $x(1) = 1$  (value for  $n = 3^k$ ).

for  $n = 3^k$ , we can write the recurrence in term of  $k$ .

(1) substitute  $n = 3^k$  in the 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:

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

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

(v) Evaluate the following recurrences completely.

(1)  $T(n) = T(n/2) + 1$ , 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.

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

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

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

$$k=3 : T(2^3) = T(4) = T(2) + 1 = (T(1) + 2) + 1 = T(1) + 3.$$

(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  $c$ .

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

The solution is

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

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

The recurrence can be solved using the master's 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 algorithms

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

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

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

$$n = n^1$$

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

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

The solution is:

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

(3) Consider the following recursive algorithms.

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

$$\text{if } n=1 \text{ return } A[0]$$

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

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

else

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

(a) what does this algorithm compute?

The given algorithm,  $\min(A[0, \dots, n-1])$  computes the minimum value in the array  $A$  from index 0 for  $(n-1)$  in the sub array  $A[0, \dots, n-2]$  and then comparing it with the last element  $A[n-1]$  to determine the overall minimum value.

(b) Setup a recurrence relation for the algorithm basic operation count and solve it.

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  $\Omega$  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  $g(n)$ .

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

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

(1) identify Dominant terms:

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

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

→ establish the inequality.

→ we want to find constants  $c$  and  $n_0$  such that:

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

(2) simplify the inequality:-

→ ignore the lower term 5 for larger

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

→ divide both sides by  $n$ .

$$2n \geq 7c.$$

→ solve for  $n$ :

$$n \geq 7c/2.$$

(4) choose constants

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

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

∴ for  $n \geq 4$ , the inequality holds.

$$2n^2 + 5 \geq 7n \text{ for all } n \geq 4$$

we have shown that there exist constants  $(2)$  and  $n_0 = 4$  such that for all  $n \geq n_0$ .

$$2n^2 + 527n$$

Thus, we can conclude that:-

$$f(n) = 2n^2 + 527n = \Omega(n^2).$$

In  $\Omega$  notation, the dominant term  $2n^2$  in  $f(n)$  clearly grows faster than  $fn$ . Hence

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

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

Showing that  $f(n)$  grows at least as fast as  $fn$ .