

# Assignment-9

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1) Solve the following recurrence relations.

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

Solution:

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

$$\Rightarrow x(1) = 0$$

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

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

$$\Rightarrow x(4) = x(3) + 5 = 10 + 5 = 15$$

e) identify the pattern. The general term

$$\Rightarrow \text{The first term } x(1) = 0$$

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

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

$\therefore$  The solution is

$$x(n) = 5(n-1)$$

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

Solution:

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

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

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

$$\Rightarrow x(3) = 3x(2) = 3 \cdot 12 = 36$$

$$\Rightarrow x(4) = 3x(3) = 3 \cdot 36 = 108$$

e) identify the general term:

$$\Rightarrow \text{The first term } x(1) = 4$$

$$\Rightarrow \text{The common ratio } r = 3$$

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  $n \geq 1$  with  $x(1) = 1$  (solve for  $n = 2^k$ )

Solution:

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

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

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

d) write down the first few terms to identify the pattern:

$$x(1) = 1$$

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

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

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

e) identify the general term by finding the pattern:

we observe that:

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

Let's 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 first term  $a=2$  and the last term  $2^k$  except for the additional 1 term.

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



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

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

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

Adding the +1 term

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

$\therefore$  solution is

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

d)  $x(n) = x(n/3) + 1$  for  $n$  not 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 the recurrence

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

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

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

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

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

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

3) identify the general form:

we observe that:

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

Summing up the series:

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

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

$\therefore$  the solution is:

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

Evaluate the following recurrence complexity.

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

Solution

The recurrence relation can be solved using iteration method

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

2) Iterate the recurrence:

for  $k=0$ :  $T(2^0) = T(1) = T(1)$   
 for  $k=1$ :  $T(2^1) = T(2) = T(1) + 1$   
 for  $k=2$ :  $T(2^2) = T(4) = T(2) + 1 = (T(1) + 1) + 1 = T(1) + 2$   
 for  $k=3$ :  $T(2^3) = T(8) = T(4) + 1 = (T(1) + 2) + 1 = T(1) + 3$

3) Generalize the pattern:

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

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 size.

Solution:

This recurrence can be solved using the master theorem for divide-and-conquer recurrence of the form:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

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

The master let, determine the value of  $\log_b a$ :

$$\log_b a = \log_3 2$$

using the properties of logarithms

$$\log_3 2 = \frac{\log_2 2}{\log_2 3}$$

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

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

$$n = n^1$$

since  $1 > \log_3 2$ , we are in the third case of the master 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?

$\text{mini}(A[0 \dots n-1])$

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

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

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

Else

return  $A[n-1]$

a) what does this algorithm compute?

Solution:

The given algorithm, ' $\text{mini}(A[0 \dots n-1])$ ', computes the minimum value in the array 'A' from index '0' to '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 minimum value.

b) setup a recurrence relation for the algorithm's basic operation count and solve it.



## Solution:

To determine the recurrence relation for the algorithm's basic operation count, let's analyze the steps 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  $AEU$

2) Recursive case: when  $n>1$ , the algorithm:

→ makes a recursive call to  $\text{mini}(A[0 \dots n-2])$ ;

→ performs a comparison between  $\text{temp}$  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  $\text{mini}(A[0 \dots n-2])$ ; and the  $+1$  accounts for the comparison between  $\text{temp}$  and  $A[n-1]$

To solve this recurrence relation, we can use iteration method.

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

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

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

...

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

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

$$= n$$

∴ The solution is

$T(n) = n$   
this means the algorithm performs  $n$  basic operations for an input array of size  $n$ .

2) Analyze the order of growth.

1)  $F(n) = 2n^2 + 5$  and  $g(n) = 7n$ . Use the  $\Omega(g(n))$  notation.

Solution:

To analyze the order of growth and use the  $\Omega$  notation, we need to compare the given functions  $F(n)$  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 of a function. Specifically,  $F(n) = \Omega(g(n))$  means that for sufficiently large  $n$ ,  $F(n)$  grows at least as fast as  $g(n)$ .  
formally,  $F(n) = \Omega(g(n))$  if there exist positive constants  $C$  and  $n_0$  such that for all  $n \geq n_0$ :

$$F(n) \geq C \cdot g(n)$$

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  $7n$

2) establish the inequality:



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

3) simplify the inequality!

→ ignore the lower order terms for larger

⇒ Divide both sides by  $n$ .

→ solve for  $n$ :

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

4) Choose constants

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

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

∴ 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  $c=1$  and  $n_0=4$  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  $\Omega$  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  $7n$ .