

## Recurrence Relations:-

In mathematics, a recurrence r/n is an equation that expresses the  $n^{\text{th}}$  term of a sequence as a function of the  $k$  preceding terms, for some fixed  $k$  (independent from  $n$ ), which is called the order of relation.

- \* An equation which represents a based on some rule.

- \* It helps in finding the subsequent term (next term) dependent upon the preceding term (previous term).

- \* If the previous term ~~is~~ <sup>is a given</sup> series, then we can easily determine the next term.

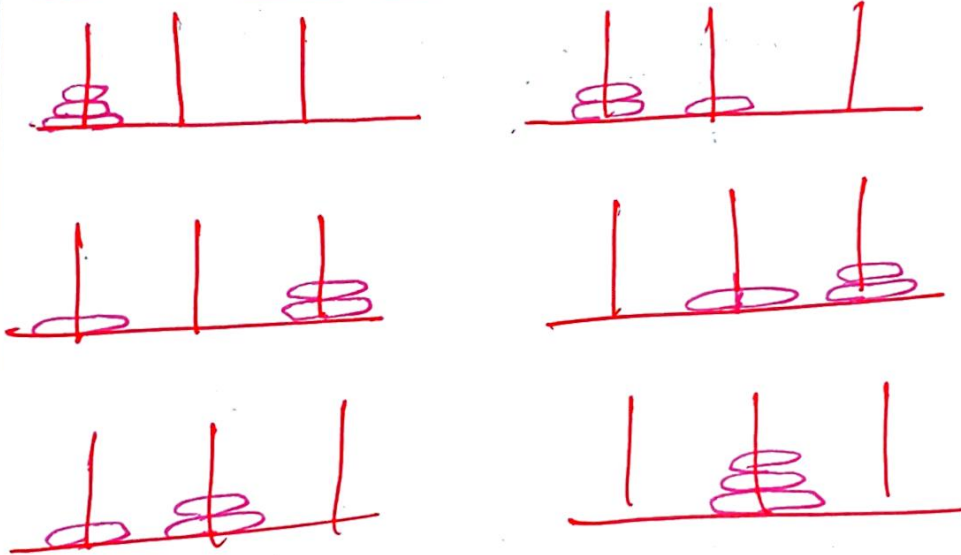
- \* Recurrence relations are used to reduce complicated problems to an iterative process based on simpler versions of the problem.

Ex: Tower of Hanoi jigsaw.

A recurrence or recurrence relation defines an infinite sequence by describing how to calculate the  $n^{\text{th}}$  element of the sequence given the values of smaller elements, as in:

$$T(n) = T(n/2) + n, \quad T(0) = T(1) = 1$$

Tower of Hanoi



## # RECURRENCE RELATION :-

\* Recurrence relations often arise in calculating the time and space complexity of algorithms.

\* Recursive algorithm is one which makes a recursive call to itself with a smaller i/p's.

## # Recurrences & running time :-

An equation that describe a function in terms of its values or smaller i/p's.

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

## To solve the recurrence :

- i) Find an explicit formula of the expression.
- ii) Bound the recurrence by an exp. that involves  $n$

Ex:-

$$S(n) = \begin{cases} 0 & n=0 \\ c+S(n-1) & n>0 \end{cases}$$



$$S(n) = \begin{cases} 0 & n=0 \\ n+S(n-1) & n>0 \end{cases}$$

$$T(n) = \begin{cases} c & n=1 \\ 2T(n/2) + c & n>1 \end{cases}$$

$$T(n) = \begin{cases} c & n=1 \\ \alpha T(n/2) + cn & n>1 \end{cases}$$

### Examples of Recurrences:-

- $T(n) = T(n-1) + n$   $\Theta(n)^2$   
 $\Rightarrow$  Recursive algorithm that loops through the i/p to eliminate one item.
- $T(n) = T(n/2) + c$   $\Theta(\log n)$   
 $\Rightarrow$  Recursive algorithm that halves the i/p in one step.
- $T(n) = T(n/2) + n$   
 $\Rightarrow$  Recursive algorithm that halves the i/p but must examine every item in the i/p.
- $T(n) = 2T(n/2) + 1$   
 $\Rightarrow$  Recursive algorithm that splits the i/p into 2 halves and does a constant amount of other work.

# Recurrence Relations:-

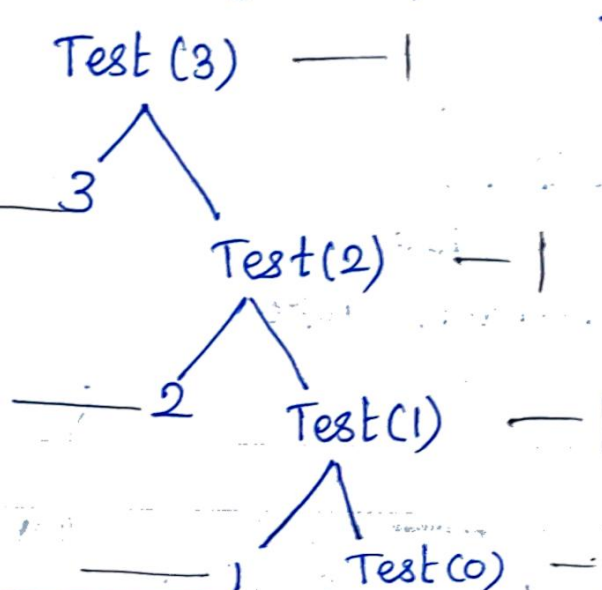
$$T(n) = T(n-1) + n$$

Ex:1

Tracing Tree / Recursive tree.

$$f(n) = n+1$$

$$T(n) = O(n)$$



$T(n)$  Void test (int n)

```
{
    if (n > 0)
    {
        printf("%d", n);
        Test(n-1);
    }
}
```

Ex:2 - decreasing

$T(n)$  - Void Test (int n)

```
{
    if (n > 0)
    {
        for (i = 0; i < n; i++)
        {
            printf("%d", n);
        }
        Test(n-1);
    }
}
```

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

↑ recurrence eqn.  $T(n-1) + n$

any constant  $c/a/k$

Ex:1

$T(n)$  Void test (int n)

```
{
    if (n > 0)
    {
        printf("%d", n);
        Test(n-1);
    }
}
```

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

→  $C/k/a$  any

$$T(n) = \begin{cases} 1 & n=0 \\ T(n-1) + 1 & n>0 \end{cases}$$

1. Ret 1a → 1b

### Ex 83 Factorial.

```
int factorial (unsigned int n)
{
    if (n == 0)
        return 1;
    return n * factorial (n-1);
}
```

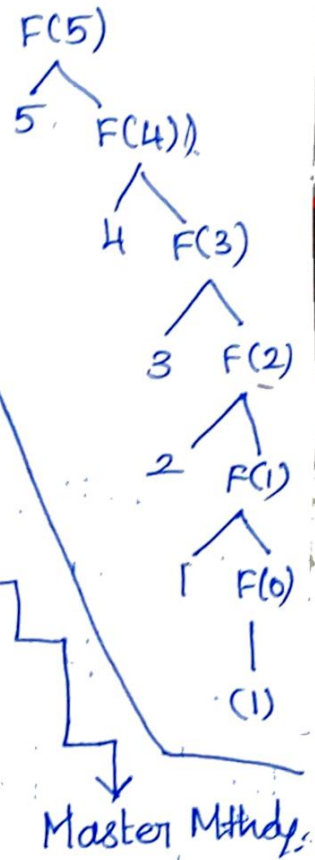
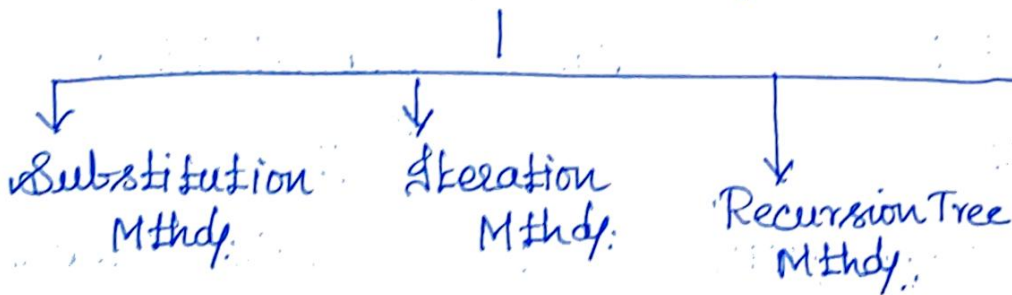
$$\begin{aligned} T(n) &= 1 + 1 + T(n-1) \\ &= n - 1 + 3 \\ &= n - 2 \\ \boxed{T(n) = O(n)} \end{aligned}$$

Recurrence relation is

$$T(n) = \begin{cases} 1 & n=0 \\ T(n-1) + 1 & n > 0 \end{cases}$$

### # Solutions of Recurrence r/n's:-

Four Methods for solving  
recurrence r/n's.





$$T(n) = O(n^2)$$

# Solys. to Recurrence Relations:- definition

↳ Solys. to Recursive Tree Method & Substitution Mthd.

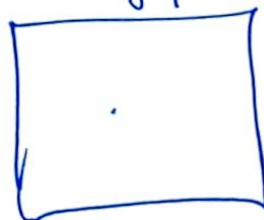
Ex: 1.



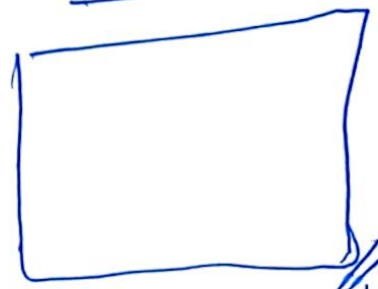
Ex: Recursive Tree/Trav



egy.



Substn



## Substitution Mthd

Fwd

Bkwd.

1 to n

n, n-1, n-2, ..., 3, 2, 1, 0

Adv:-

\* Easy to prove

\* prone to mistakes.



2/4 x:2

Approved by AICTE

$$T(n) = \begin{cases} 1 & n=0 \\ T(n-1) + n & n > 0 \end{cases}$$

Substitution Mthdy:

$$T(n) = T(n-1) + n \rightarrow (a)$$

Sub  $n=n-1 \Rightarrow \because T(n-1) = T(n-1-1) + n-1$

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

Sub (1) in (a)

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

$$= T(n-3) + n-2$$

$$T(n-2) = T(n-3) + n-2 \rightarrow (2)$$

$$n-1 \quad T(n-1) = [T(n-2) + n-1] + n$$

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

avoiding adding terms to prepare a seq.

$$n-2 \quad T(n-2) = T(n-3) + (n-2) + (n-1) + n \rightarrow (3)$$

o  
o  
o  
o  
o

Mthds followed for induction after, continue for k steps

$$T(n) = T(n-k) + T(n-k) + (n-(k-1)) + (n-k-2) + \dots + (n-1) + n$$

eg. Assume  $n-k$  has become 0  
 $\because n-k=0$   
 $n=k$

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

$$= T(0) + 1 + 2 + 3 + \dots + (n-1) + n$$

$$= 1 + \frac{n(n+1)}{2}$$

$$T(n) = O(n)^2$$

Ex:-

$$T(n) = \begin{cases} 1 & n=0 \\ T(n-1) + \log n & n>0 \end{cases}$$

Void Test (int n)

{ if (n>0)

{ for (i=1; i<n; i=i\*2)

{ printf("%d", i);

}

Test(n-1);

}

}

$\log n$

$T(n-1)$

$$\underline{T(n) = T(n-1) + \log n}$$

$$O(n \log n)$$

# Properties of Asymptotic Notations:-

## ① General properties

if  $f(n)$  is  $O(g(n))$  then  $a * f(n)$  is  $O(g(n))$

eg:  $f(n) = 2n^2 + 5$  is  $O(n^2)$

then  $= 7 * f(n) = 7(2n^2 + 5)$   
 $= 14n^2 + 35$  is  $O(n^2)$

Also true for  $\Omega$  &  $\Theta$ . i.e.  $\forall$  all three notations  
 $f(n) = \Omega(g(n)) \rightarrow a * f(n) \rightarrow \Omega(g(n))$

## ② Reflexive property:-

if  $f(n)$  is given then  $f(n)$  is  $O(f(n))$

eg:-  $f(n) = n^2 \Rightarrow O(n^2)$

## ③ Transitive property if $f(n)$ is $O(g(n))$ and $g(n)$ is $O(h(n))$

then  $f(n) = O(h(n))$

eg:-  $f(n) = n$   $\xrightarrow{\text{upper bound}} g(n) = n^2$   $\xrightarrow{\text{upper bound}} h(n) = n^3$

$n$  is  $O(n^2)$  &  $n^2$  is  $O(n^3)$

then  $n$  is  $O(n^3)$ .

$\forall$  three notations  $O, \Omega$  &  $\Theta$ .



(4) Symmetric property: (Lower & only)

If  $f(n)$  is  $O(g(n))$  then  $g(n)$  is  $O(f(n))$

eg:-  $f(n) = n^2 \rightarrow g(n) = n^2$

$$f(n) = O(n^2)$$

$$g(n) = O(n^2)$$

When both the ~~thry~~ are same, that they are symmetric.

(5) Transpose symmetric:- (O's) only

If  $f(n) = O(g(n))$  then  $g(n)$  is  $\Omega(f(n))$

eg:-  $f(n) = n$   $g(n) = n^2$  LB.

then  $n$  is  $O(n^2)$  and

$$n^2 \text{ is } \Omega(n)$$

If one ~~thry~~ forms an upper bound for other ~~thry~~ then the other ~~thry~~ will form a lower bound for the other ~~thry~~.

If  $f(n) = O(g(n))$  and  $g(n) = \Omega(f(n))$

$$g(n) \leq f(n) \leq g(n)$$

$$\boxed{f(n) = O(g(n))} //$$

When same ~~thry~~ is, both upper & lower bound

=

## properties of Asymptotic Notations:- #2.

✓ IF  $f(n) = O(g(n))$

✓ and  $d(n) = O(en)$

then  $f(n) + d(n) = O(\max(g(n), en))$

eg:  $f(n) = n = O(n)$

$d(n) = n^2 = O(n^2)$

$f(n) + d(n) = n + n^2 = O(n^2)$

IF  $f(n) = O(g(n))$

$d(n) = O(en)$

then  $f(n) * d(n) = O(g(n) * en)$

$n \quad n^2 = n^3$

—x—