

# Recursive Analysis

Recursion : A function calling itself is called "Recursion"

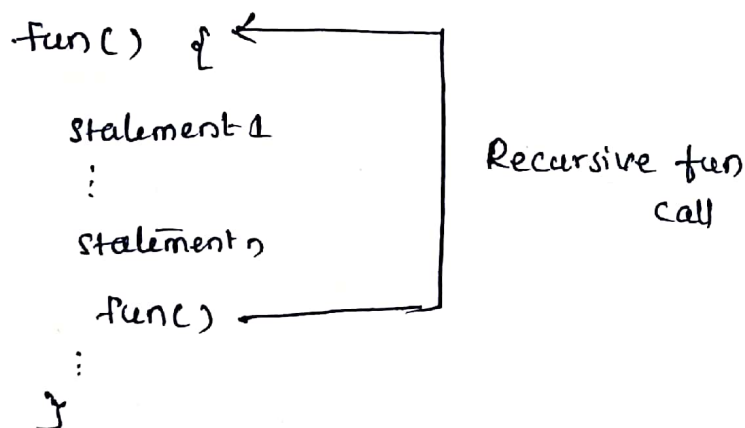
↳ Recursion is preferable when, In a given problem every subproblem is same as given problem.

For Recursive analysis space complexity is more compare to normal programs. for every recursive call again it will create stack.

In order to analyse Time complexity of Recursive function is we have various methods they are :

1. Back substitution Method
2. Recursive Tree Method
3. Master Theorem

Eg:



Steps:

1. In order to Analyse any Recursive program we need to write Recurrence relation.

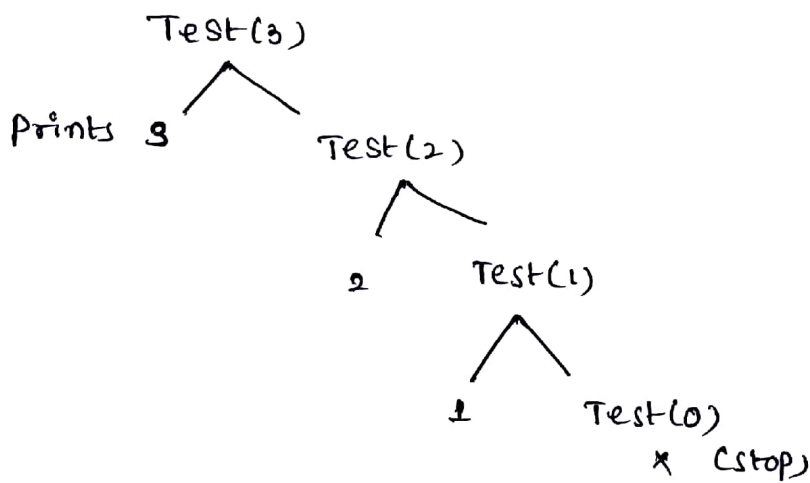
2. By solving that recurrence relation only we will get Time complexity.

Eg1:

```
void Test (int n)
{
    if (n > 0) → O(1)
    {
        printf("%d", n); → O(1)
        Test(n-1); → T(n-1)
    }
}
```

Tree method:

Let  $n=3$



The amount of work done here → printing value + calling function

As I passed '3' → 3 times printing + 1

→ passed '5' → 5 times prints

for each call it is printing 1 unit of time so.

n. of calls  $3+1=4$

n. of prints = 3

In general

n. of calls ' $n+1$ '

for any 'n'

n. of prints 'n'

Time function  $f(n) = O(n)$

$\therefore \text{Time complexity} = O(n)$

Eg1: void Test(n) { ———  $T(n)$   
     if (n > 0) ———  $O(1)$   
         { printf("%d", n); ——— 1  
             Test(n-1); ———  $T(n-1)$   
         }

Step 1: Write down recurrence relation, for if and printf statement it will take constant amount of time and again it will call Test(n-1) so the relation is

$$T(n) = T(n-1) + \underbrace{1}_{\text{as if and printf takes } O(1)}$$

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

Step 2: Solve Recurrence relation by Back substitution method.

$$T(n) = 1 + T(n-1) \rightarrow \textcircled{1}$$

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

$$T(n-2) = 1 + T(n-3)$$

⋮

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

consider  $\textcircled{1}$   $T(n) = 1 + T(n-1)$

substitute  $T(n-1)$  value from  $\textcircled{2}$

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

$$T(n) = 1 + \overbrace{[1 + T(n-2)]} = 2 + T(n-2)$$

$$T(n) = 2 + 1 + T(n-3) = 3 + T(n-3)$$

⋮

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

when it will stop whenever  $n-k=0 \Rightarrow n=k$

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

Substitute  $m=k$

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

$$T(n) = T(0) + n \Rightarrow 1+n$$

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

Eg 2:

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

recursive call

Step 1:

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

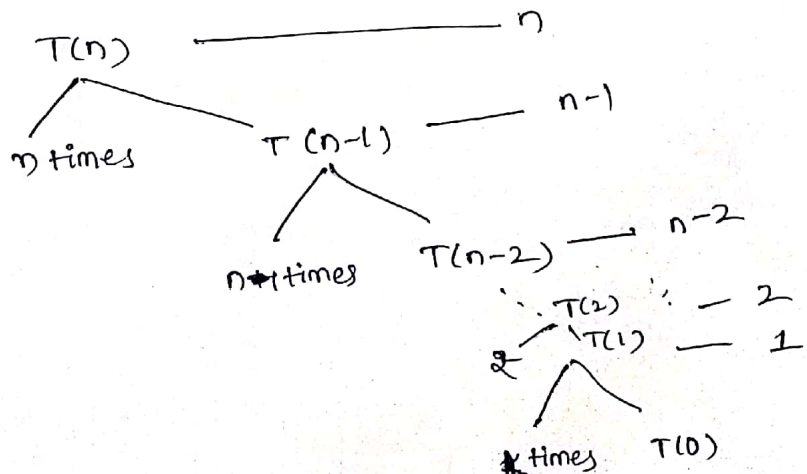
$$= T(n-1) + 2n + 2$$

writing it as Asymptotically  $O(n)$

So recurrence relation now,

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

Tree method:



Total time is:

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

sum of 1<sup>st</sup> natural numbers

$$T(n) = \frac{n(n+1)}{2} \approx O(n^2)$$

$$\boxed{\therefore T(n) = O(n^2)}$$

Backsubstitution Method:

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

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

$$T(n-2) = T(n-3) + 1$$

⋮

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

from 1

$$\begin{aligned} T(n) &= T(n-1) + 1 \\ &= [T(n-2) + 1] + 1 \\ &= T(n-2) + (n-1) + 1 \\ &= T(n-3) + (n-2) + (n-1) + 1 \\ &\vdots \\ \text{k times} \quad &= T(n-(k+1)) + n-k + \dots + T(n-3) + (n-2) + (n-1) + 1 \\ &= T(n-k) + (n-(k-1)) + n-(k-2) + \dots \\ &\quad \dots + (n-1) + 1 \end{aligned}$$

$n-k=0 \Rightarrow n=k$  substitute

$$\begin{aligned} T(n) &= T(0) + (n-n+1) + (n-n+2) + \dots + (n-1) + 1 \\ &= 1 + 1 + 2 + 3 + \dots + (n-1) + 1 \\ &= \frac{n(n+1)}{2} \approx O(n^2) \quad // \end{aligned}$$

Eg 3:

void Test(n)

{ if (n > 0) —————  $O(1)$

{ for (i = 1; i < n; i++)  
    pf("%d", i); } —  $O(\log n)$

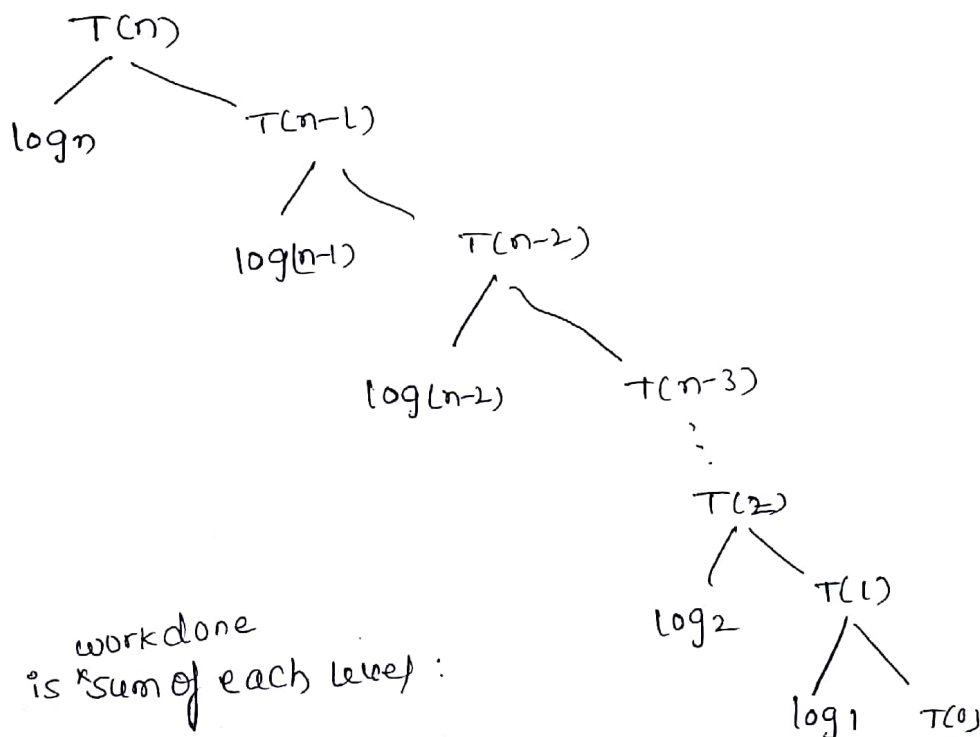
Test(n-1) —————  $T(n-1)$

}

write Recurrence relation :  $T(n) = T(n-1) + \log n + O(1)$

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

Tree method:



∴ Time taken is <sup>work done</sup> sum of each level :

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

$$\log [n * (n-1) * (n-2) * (n-3) * \dots * 1]$$

$$= \log [n!]$$

$$[\because n! \approx n^n]$$

$$= \log n^n$$

There is no bound so  
upper bound  $n! \approx n^n$

$$\boxed{T(n) \Rightarrow O(n \log n)}$$



Back substitution method :

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

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

$$= \underline{T(n-2)} + \log(n-1) + \log n$$

$$= [T(n-3) + \log(n-2)] + \log(n-1) + \log n$$

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

After  
'k' times

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

$$\log(n-1) + \log n \rightarrow \textcircled{1}$$

we know

$$n-k=0$$

$$n=k \rightarrow \text{substitute in } \textcircled{1}$$

$$T(n) = T(0) + \underbrace{\log 1 + \log 2 + \log 3 + \dots + \log n}_{\log n!}$$

$$= 1 + \log n!$$

$$= 1 + \log n^n$$

$$= 1 + n \log n$$

$$\boxed{T(n) = O(n \log n)}$$

## Conclusions:

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

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

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

↓

fun is multiplied by 'n' times

Instead by  $T(n-1)$  we can take  $T(n-2)$

Then also same thing.

$$T(n) = T(\underline{n-2}) + 1 \rightarrow n/2 \simeq O(n)$$

$$= T(\underline{n-100}) + 1 \rightarrow O(n)$$

↓

constant

∴ In general we can write from above observations

If any recurrence relation in the form of

$$T(n) = T(n - \underline{k}) + f(n)$$

↓

constant

↓

Asymptotic fun

Then time complexity is n times of  $f(n)$



Ex 4:

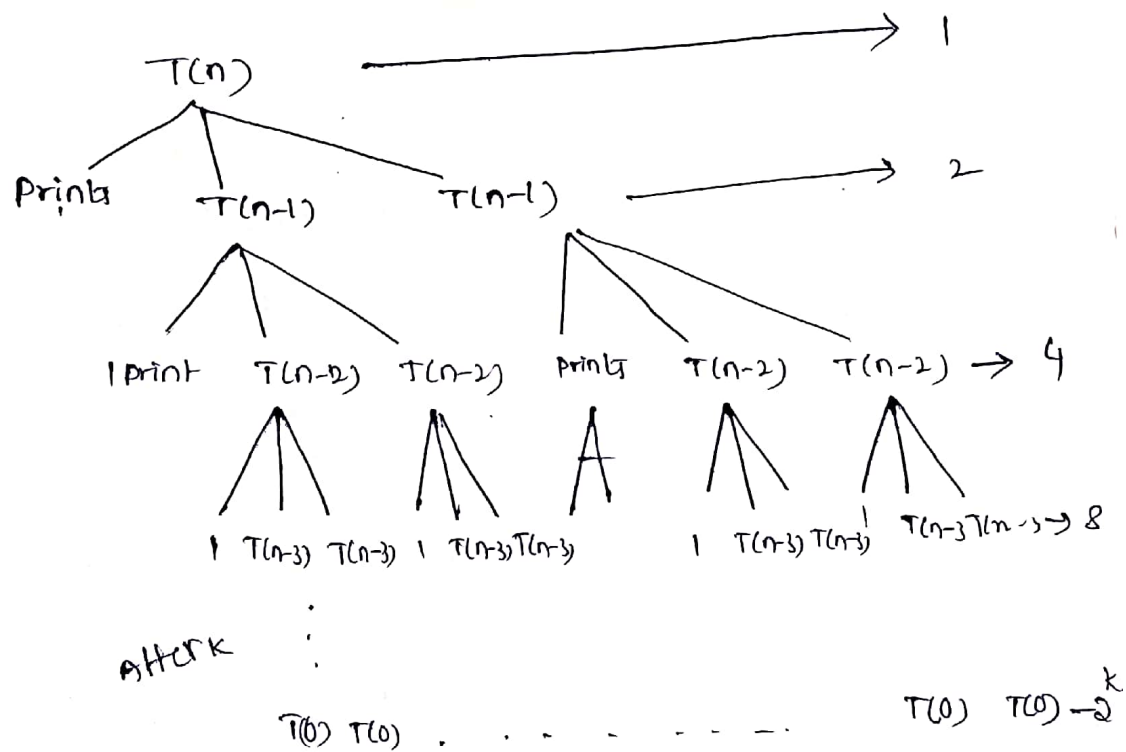
```

void Test(n) {  $\rightarrow T(n)$ 
    if (n > 0)  $\rightarrow 1$ 
    {
        pf("%d", n);  $\rightarrow 1$ 
        Test(n-1);  $\rightarrow T(n-1)$ 
        Test(n-1);  $\rightarrow T(n-1)$ 
    }
}

```

$$T(n) = \begin{cases} 2T(n-1) + 1 & n > 0 \\ 1 & n = 0 \end{cases}$$

Using Recursion Tree method :



## Back substitution Method:

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

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

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

$$= 2^2 [2T(n-3) + 1] + 2 + 1$$

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

⋮

After k times

$$T(n) = 2^k T(n-k) + 2^{k-1} + 2^{k-2} + \dots + 2^3 + 2^2 + 2 + 1$$

Assume  $n-k=0$

$$n=k$$

Substitute 'k' value as 'n'

$$\begin{aligned} T(n) &= 2^k T(0) + 2^{k-1} + 2^{k-2} + \dots + 2^3 + 2^2 + 2 + 1 \\ &= 2^n \times 1 + 2^n - 1 \end{aligned}$$

$$\Rightarrow 2^n + 2^n - 1 \Rightarrow 2^{n+1} - 1$$

$$T(n) = 2^{n+1} - 1$$

$$\boxed{T(n) \approx O(2^n)}$$

Ex5:  $\text{Test}(n) \{ \longrightarrow T(n)$

if ( $n > 1$ )

1 pf ("vd", n);  $\longrightarrow 1$

Test( $n/2$ );  $\longrightarrow T(n/2)$

}

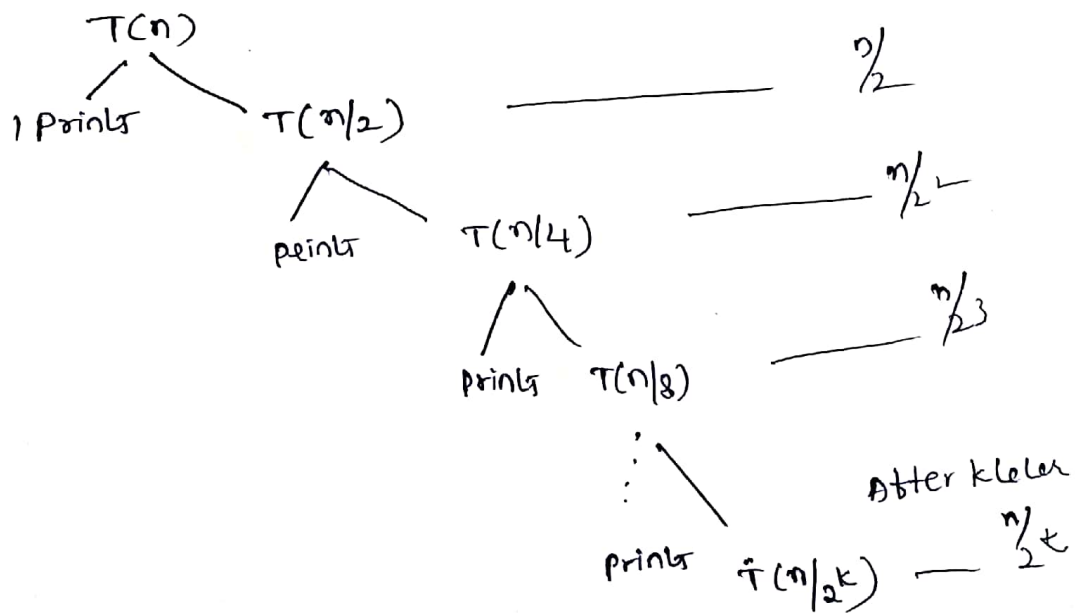
}

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

Recurrence relation:

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

Recursion Tree method:



## Back substitution Method:

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

$$T(n/2) = [T(n/2^2) + 1] + 1$$

$$= T(n/2^2) + 2 \rightarrow (2)$$

$$= [T(n/2^3) + 1] + 2$$

$$= T(n/2^3) + 3 \rightarrow (3)$$

After k :

$$T(n/2^k) + k$$

$$= T(1) + k$$

$$[ \because n/2^k = 1 ]$$

$$2^k = n$$

$$k = \log n$$

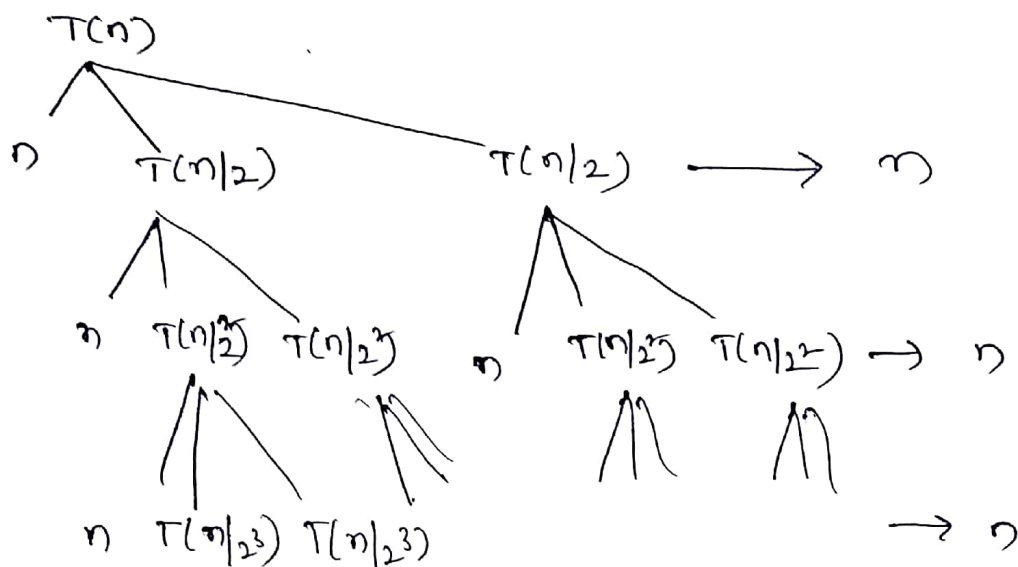
$$T(n) = 1 + \log n$$

$$\boxed{T(n) = O(\log n)}$$

Eg 6:

```
void Test (n) {  
    if (n > 1) {  
        for (i = 0; i < n; i++)  
            n ← { statements; }  
        T(n/2) ← Test(n/2)  
        T(n/2) ← Test(n/2)  
    }  
}
```

### Recursive Tree method:



After  
k level

$$\Gamma(m/2^k)$$

Total work done After  $k$  levels is  $\Rightarrow k \times n$

we know it will stop when  $n/a^k = 1$

$$n = 2^k$$

$$K = \log n$$

Substituto

$$T(n) = K * n$$

$$= \log n * n$$

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

Recurrence relation:

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

## Back substitution Method:

$$T(n) = 2T\left(\frac{n}{2}\right) + n \rightarrow (1)$$

$$\Rightarrow 2 \left[ 2T\left(\frac{n}{2^2}\right) + \frac{n}{2} \right] + n$$

$$\Rightarrow 2^2 T\left(\frac{n}{2^2}\right) + n + n$$

$$\Rightarrow 2^2 T\left(\frac{n}{2^2}\right) + 2n \rightarrow (2)$$

$$\Rightarrow 2^2 \left[ 2T\left(\frac{n}{2^3}\right) + \frac{n}{2^2} \right] + 2n$$

$$\Rightarrow 2^3 T\left(\frac{n}{2^3}\right) + 3n \rightarrow (3)$$

After  $k^{\text{th}}$  recursive calls

$$\Rightarrow 2^k T\left(\frac{n}{2^k}\right) + kn$$

$$\frac{n}{2^k} = 1 \Rightarrow k = \log n$$

$$\Rightarrow 2^{\log n} T(1) + \log n * n$$

$$\Rightarrow n + n \log n$$

$$[2^k = n = 2^{\log n}]$$

$$\boxed{\therefore T(n) = O(n \log n)}$$



Eg 7:

Test(n) {

if (n > 2)  $\longrightarrow$  1

  statements  $\longrightarrow$  1

  Test( $\sqrt{n}$ )  $\longrightarrow$  T( $\sqrt{n}$ )

}

y

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

Using Back substitution method:

$$T(n) = T(n^{1/2}) + 1 \rightarrow (1)$$

$$= [T(n^{1/2^2}) + 1] + 1$$

$$= T(n^{1/2^3}) + 2 \rightarrow (2)$$

$$= T(n^{1/2^k}) + 3 \rightarrow (3)$$

At k:

$$T(n^{1/2^k}) + k$$

$$n^{1/2^k}$$

Assume  $n = 2^m$

$$T(2^m) = T(2^{m/2^k}) + k$$

$$= \log_2 m$$

$$\Rightarrow \log_2 \log n$$

$$\boxed{T(n) = O(\log \log n)}$$

$$1 = \frac{m}{2^k}$$

$$2^k = m$$

$$k = \log_2 m$$

$$n = 2^m$$

$$m = \log n$$