



Algorithms and Complexity Calculation Part 3

Course on Data Structure and Algorithms Using Python

Recursive Algorithm

```
def fun(int n):
```

```
    if n == 1:
```

```
        return 1
```

```
    Else:
```

```
        return n * fun(n-1)
```

5 * fun(4)

4 * 3 \Rightarrow 2
3 \Rightarrow 3 } \Rightarrow 1
4 * 10 \Rightarrow 1

$$T(1) = 1$$

$$T(n) = \text{Calling} + \text{Cost}$$

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

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

5 * 4

4 * f(6)

7 *

```
def fun(int n):
```

```
    if n == 1 :
```

```
        return n
```

```
    Else :
```

```
        return  $\frac{n}{2} + f(n-1)$ 
```

$$T(1) = 1$$

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

```
def fun(int n):
```

```
    if n == 1 :
```

```
        return n
```

```
    Else :
```

```
        return  $n + f(\frac{n}{2})$ 
```

$$T(1) = 1$$

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

$$T(1) = 1$$

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

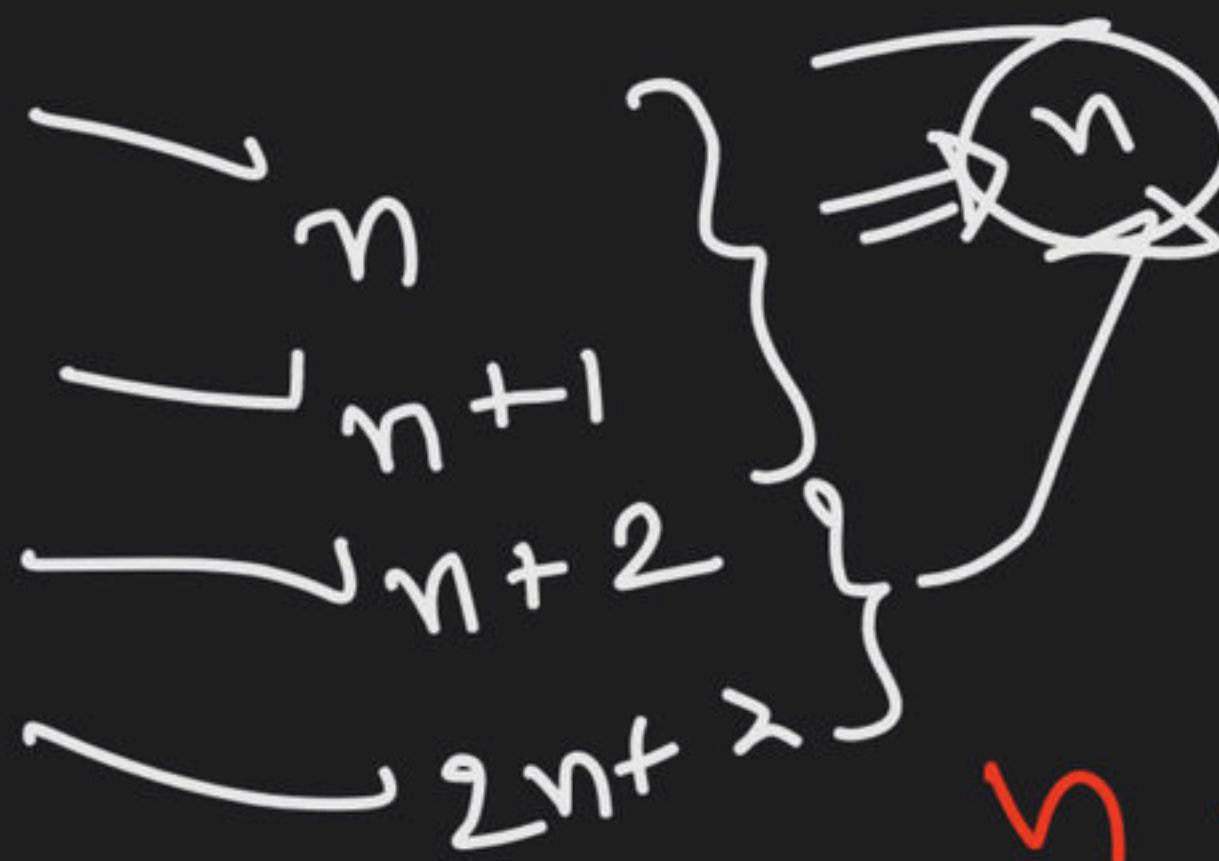
Cost

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

```
def fun(int n):
```

```
    if n == 1:
        return n
```

```
    else:
        for i in range(1, n+1):
            print i
        return n * f(n-1)
```



$$n, n+1, \frac{n}{2}, 2n$$

```
def fun( int n ):
```

```
    if n == 1 :
```

```
        return n
```

```
    Else
```

```
        return fun(n-1) + fun(n-1)
```

$$T(1) = 1$$

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

or

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

β_2 def fun(int n):

```
if n == 1 :
```

```
    return n
```

```
Else
```

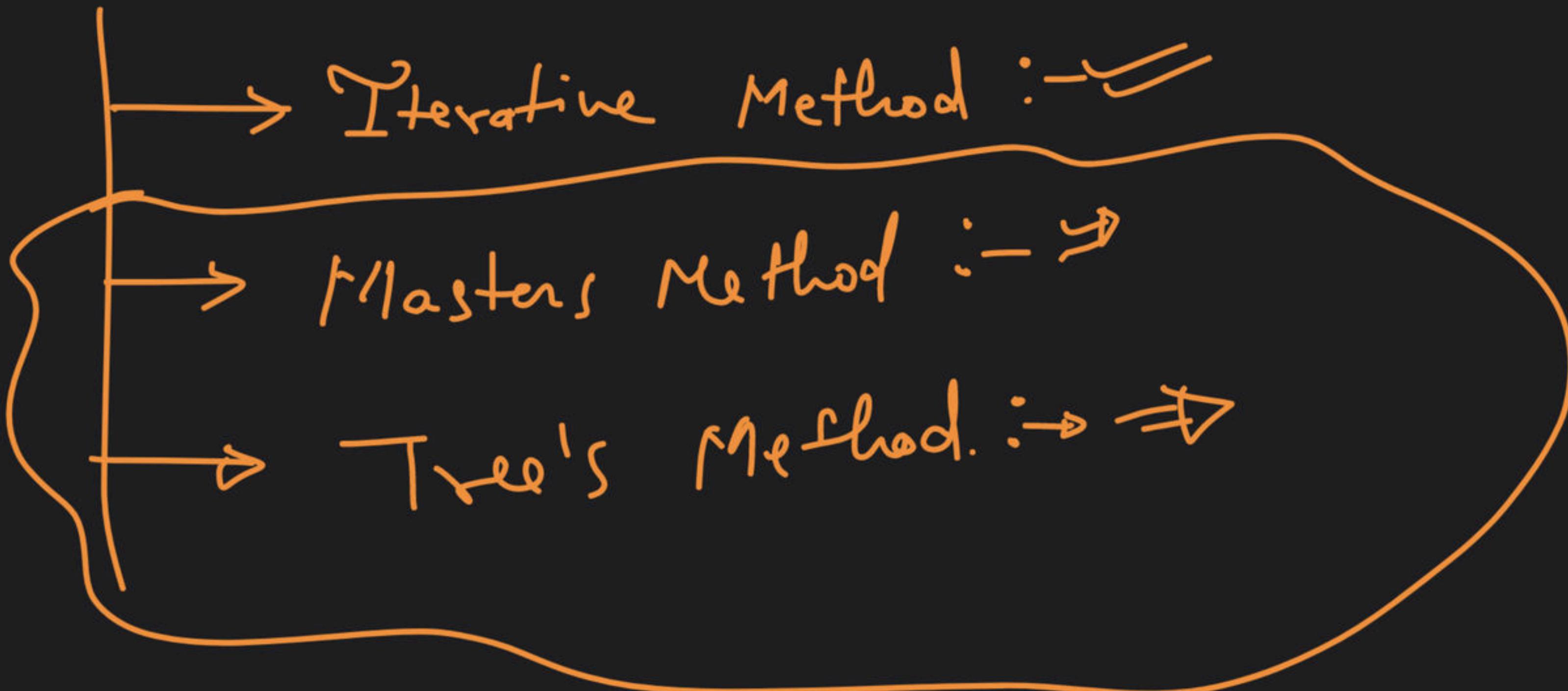
```
    return 2*fun(n-1)
```

$$T(1) = 1$$

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



To Solve Recursive Relation !



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

$$= \underbrace{T(n-2)}_{\vdots} + 1 + 1$$

$$= T(n-3) + 1 + \underbrace{1 + 1}_{\vdots}$$

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

↓

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

~~$$= 1 + n - 1$$~~

$\Rightarrow \underline{\underline{\Theta(n)}}$

$$T(0) = 1$$

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

$$n-k = 1$$

$$k = n-1$$

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

$$T(1) = 1$$

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

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

⋮

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

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

$$\begin{aligned} &= 1 + 2 + \dots + (n-1) + n \\ &= \frac{n(n+1)}{2} \Rightarrow \Theta(n^2) \end{aligned}$$

$$\begin{array}{c} n-k=1 \\ \hline k=n-1 \end{array}$$

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

$$T(1) = 1$$

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

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

$$= 2^2 \left[T(n-2) + 2 + 1 \right] + 2T(n-2) + 2 + 1$$

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

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

⋮

$$2^K T(n-K) + 2^{K-1} + \dots + 2^2 + 2 + 1$$

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

$$\underline{n-K=1}$$

$$\underline{K=n-1}$$

Time and Space Complexity of an Algorithm

- **Time Complexity:** The time complexity of an algorithm quantifies the amount of time taken by an algorithm to run as a function of the length of the input.
- The **space complexity** of an algorithm quantifies the amount of space taken by an algorithm to run as a function of the length of the input.
 - *Auxiliary Space*
 - *Input Space*

Types of Algorithm

- Sequential Algorithm
- Iterative Algorithm
- Recursive Algorithm

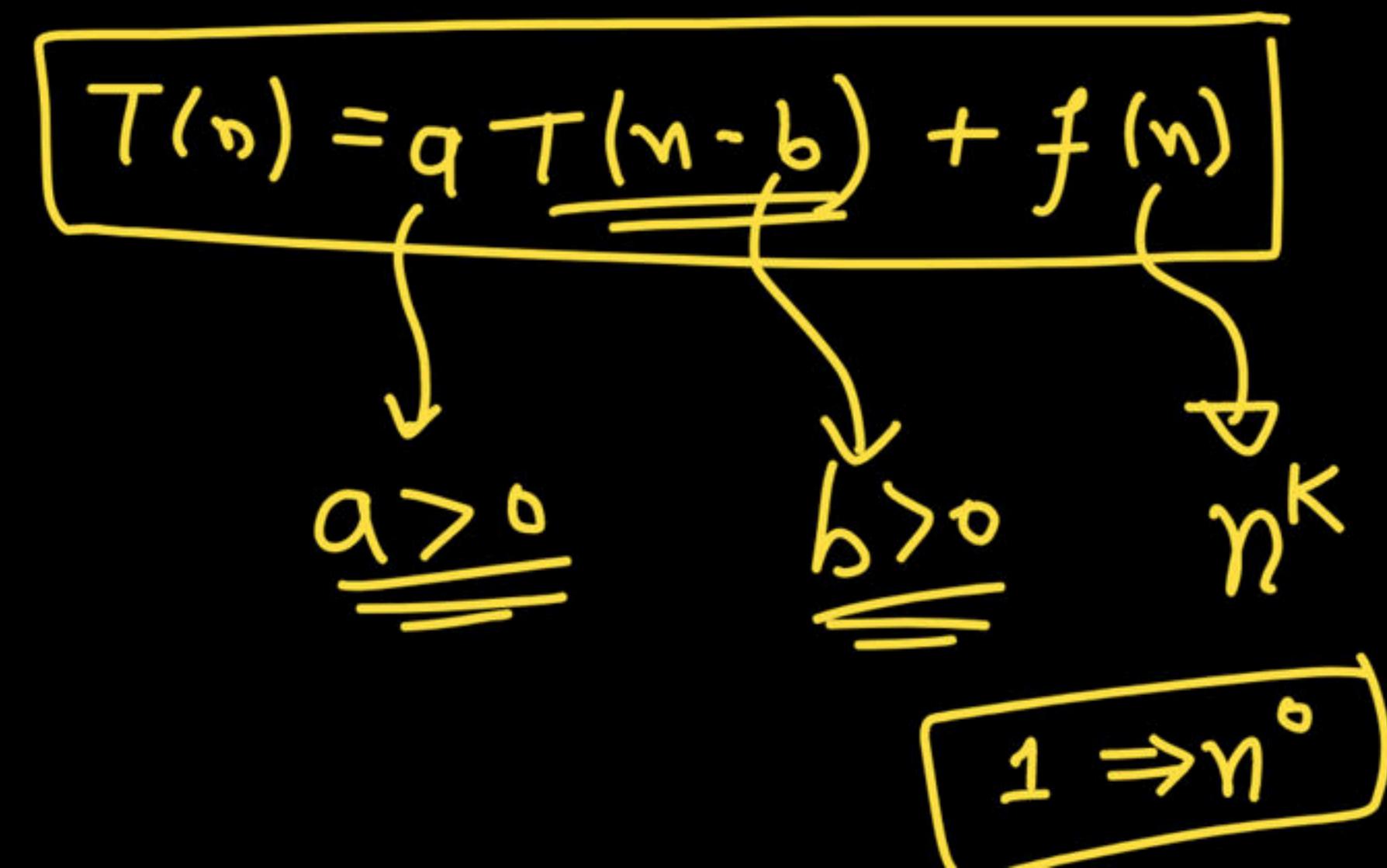
Master Theorem For Subtract Recurrences

- Let $T(n)$ be a function defined on positive n as shown below:

$$T(n) \leq \begin{cases} c, & \text{if } n \leq 1, \\ aT(n-b) + f(n), & n > 1, \end{cases}$$

for some constants c , $a > 0$, $b > 0$, $k \geq 0$ and function $f(n)$. If $f(n)$ is $O(n^k)$, then

- If $a < 1$ then $T(n) = O(n^k)$
- If $a = 1$ then $T(n) = O(n^{k+1})$
- If $a > 1$ then $T(n) = O(n^k a^{n/b})$





$$T(n) = \underbrace{a}_{a=1} T(n-1) + \underbrace{b}_{b=1}$$

$f(n) = n^0$

Case II

$$a = 1 \Rightarrow \Theta(n^{0+1}) \Rightarrow \underline{\Theta(n)}$$

$$T(n) = 2T(n-1) + \boxed{1 \Rightarrow n^0}$$

$a > 1 \Rightarrow \Theta(n^{k \cdot a^n/b}) \Rightarrow \Theta(1 \cdot 2^{n/1}) \Rightarrow \underline{\Theta(2^n)}$

$$(1) T(n) = T(n-2) + 1 \rightarrow a=1 \text{ then } \Theta(n^{k+1}) \Rightarrow \Theta(n^{0+1}) \Rightarrow \Theta(n)$$

$$(2) T(n) = 0.5T(n-1) + n \rightarrow a < 1 \text{ then } \Theta(n)$$

$$(3) T(n) = 2T(n-2) + 1 \rightarrow a \geq 1 \text{ then } \Theta(1 \cdot 2^{\frac{n}{2}}) \Rightarrow \Theta((\sqrt{2})^n)$$

$$(4) T(n) = 2T(n-1) + n \Rightarrow a > 1 \text{ then } (n \cdot 2^n)$$

$$(5) T(n) = 2T(n-2) + n \Rightarrow a > 1 \Rightarrow \text{then } (n \cdot 2^{\frac{n}{2}})$$

$$(6) T(n) = 3T(n-1) + n^2 \Rightarrow a > 1 \Rightarrow \text{then } (n^2 \cdot 3^n)$$

$$(7) T(n) = 4T(n-2) + n^2 \Rightarrow a > 1 \Rightarrow \text{then } (n^2 \cdot 4^{\frac{n}{2}})$$

$n^2 \cdot 4^{\frac{n}{2}}$

Master's Method for Divide and Conquer

$$\boxed{T(n) = aT\left(\frac{n}{b}\right) + f(n) \cdot (\log n)^k}$$

$a > 0, b > 1, k \geq 0, f(n) = \text{Any Polynomial function}$

\hookrightarrow all are +ve constant

(I) $n^{\log_b a} > f(n) \Rightarrow \Theta(n^{\log_b a})$

(II) $n^{\log_b a} < f(n) \Rightarrow \Theta(f(n) \cdot (\log n)^k)$

(III) $n^{\log_b a} = f(n) \Rightarrow \Theta(f(n) \cdot (\log n)^{k+1})$

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

$$a=1 \quad b=2 \quad f(n) = \underline{\underline{n}} \quad k=0$$

$$n^{\log_2 1} \Rightarrow n^0 \Rightarrow \begin{matrix} 1 \\ \vdots \end{matrix} < n$$

→ $\Theta(n)$

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

$$a=2 \quad b=2 \quad f(n) = \underline{\underline{1}}, \quad K=0$$

$$\log_b a \Rightarrow \frac{\log_b n}{\log_b b} \cdot n^{\log_2 2} \Rightarrow \frac{n}{\underline{\underline{1}}} > 1$$

$\Theta(n)$

$$n \equiv 2n$$

$$(1) \ T(n) = T(n/2) + 1 \xrightarrow{n^{\log_2 1}} n^0 \Rightarrow 1 \equiv 1 \Rightarrow \Theta(1 \cdot (\log n)^{0+1}) \Rightarrow \underline{\Theta(\log n)}$$

$$(2) \ T(n) = 2T(n/2) + 1 \xrightarrow{\Theta(n)}$$

$$(3) \ T(n) = 4T(n/2) + n \xrightarrow{n^{\log_2 4}} n^2 > n \Rightarrow \underline{\Theta(n^2)}.$$

$$(4) \ T(n) = T(n/2) + 1 \cdot \log n \xrightarrow{n^{\log_2 1}} n^0 \Rightarrow 1 \equiv 1 \Rightarrow \Theta(1 \cdot (\log n)^{0+1})$$

$$(5) \ T(n) = 2T(n/2) + n \log n \xrightarrow{\dots} n^{\log_2 2} \Rightarrow n \equiv n \Rightarrow \underline{\Theta(n(\log n)^2)}$$

- (1) $T(n) = 2T(n/4) + \cancel{\sqrt{n} \log n} \rightarrow n^{\log_4 2} \Rightarrow n^{\frac{1}{\log_2 4}} \Rightarrow n^{\frac{1}{2}} \Rightarrow \underline{\underline{\sqrt{n}}} \equiv \sqrt{n}$
- (2) $T(n) = 8T(n/2) + \underline{n^2} \rightarrow n^{\log_2 8} \Rightarrow n^3 > n^2 \Rightarrow \underline{\Theta(n^3)} \quad \Theta(\sqrt{n} (\log n)^2)$
- (3) $T(n) = 2T(n/8) + \cancel{n} \rightarrow n^{\log_8 2} \Rightarrow n^{\frac{1}{\log_2 8}} \Rightarrow n^{1/3} < n \Rightarrow \underline{\underline{\Theta(n)}}$
- (4) $T(n) = T(n/2) + \cancel{n^2} \rightarrow n^{\log_2 1} \Rightarrow n^0 < n^2 \Rightarrow \underline{\underline{\Theta(n^2)}}.$
- (5) $T(n) = 4T(n/2) + \cancel{n^2} \rightarrow n^{\log_2 4} \Rightarrow n^2 = n^2 \rightarrow \underline{\underline{\Theta(n^2 \log n)}}.$

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

Extended Master's Method 1

• $T(n) = aT(n/b) + f(n) (\log n)^k$

• Here $f(n)$ is polynomial function

• a, b are positive constant $a > 0, b > 1, k = -1$

• Calculate $n^{\log_b a}$

Case 1: If $f(n) < n^{\log_b a}$ then $T(n) = \Theta(n^{\log_b a})$

Case 2: If $f(n) > n^{\log_b a}$ then $T(n) = \Theta(f(n))$

Case 3: If $f(n) = n^{\log_b a}$ then $T(n) = \Theta(f(n) \log \log n)$

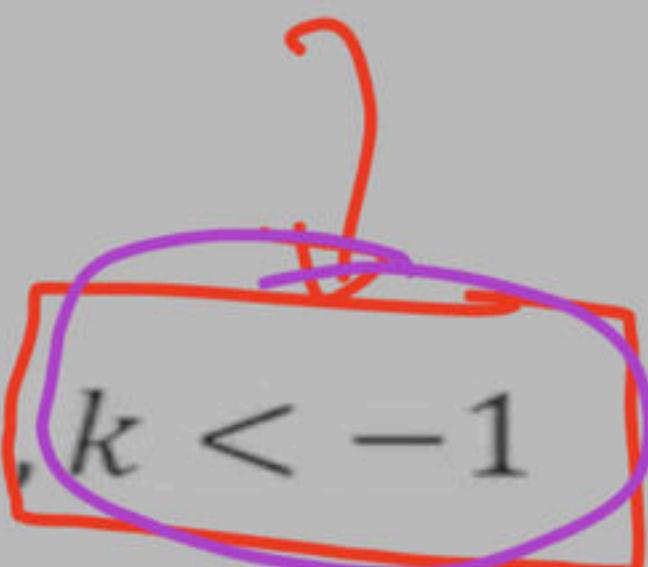
= =

Extended Master's Method 2

❖ $T(n) = aT(n/b) + f(n) (\log n)^k$

❖ Here $f(n)$ is polynomial function

❖ a, b, k are positive constant $a > 0, b > 1, k < -1$



❖ Calculate $n^{\log_b a}$

Case 1: If $f(n) < n^{\log_b a}$ then $T(n) = \Theta(n^{\log_b a})$

Case 2: If $f(n) > n^{\log_b a}$ then $T(n) = \Theta(f(n))$

Case 3: If $f(n) = n^{\log_b a}$ then $T(n) = \Theta(n^{\log_b a})$

(1) $T(n) = T(n/2) + \frac{1}{(\log n)^{-1}} \rightarrow n^{\log_2 1} \Rightarrow n^0 \Rightarrow 1 \quad \underline{\Theta(1 \cdot \log \log n)}$

(2) $T(n) = 2T(n/2) + \frac{1}{(\log n)^{-2}} \rightarrow n^{\log_2 2} \Rightarrow n > 1 \Rightarrow \Theta(n)$

(3) $T(n) = 4T(n/2) + \frac{n}{\log n} \rightarrow n^{\log_2 4} \Rightarrow n^2 > n \Rightarrow n^2$

(4) $T(n) = T(n/2) + \frac{1}{(\log n)^2} \rightarrow n^{\log_2 1} \Rightarrow n^0 \Rightarrow \boxed{1=1} \Rightarrow \underline{\Theta(1)}$

(5) $T(n) = 2T(n/2) + \frac{n}{\underline{\log n}} \rightarrow n^{\log_2 2} \Rightarrow \boxed{n=n} \Rightarrow \underline{\Theta(n \log \log n)}$

(6) $T(n) = 2T(n/4) + \frac{\sqrt{n}}{(\log n)^3} \rightarrow n^{\log_4 2} \Rightarrow \boxed{\sqrt{n} \approx \sqrt{n}} \Rightarrow \underline{\Theta(\sqrt{n})}$

(7) $T(n) = 8T(n/2) + \frac{n^2}{\log n} \rightarrow n^{\log_2 8} \Rightarrow n^3 > n^2 \Rightarrow \underline{\Theta(n^3)}$

- Recursive Relation
 - Substitution Method ✓
 - Trees Method ✓

Trees Method : \rightarrow

$$T(n) = aT(n/b) + f(n)(\log n)^k$$

\hookrightarrow Here we can apply
Master's -

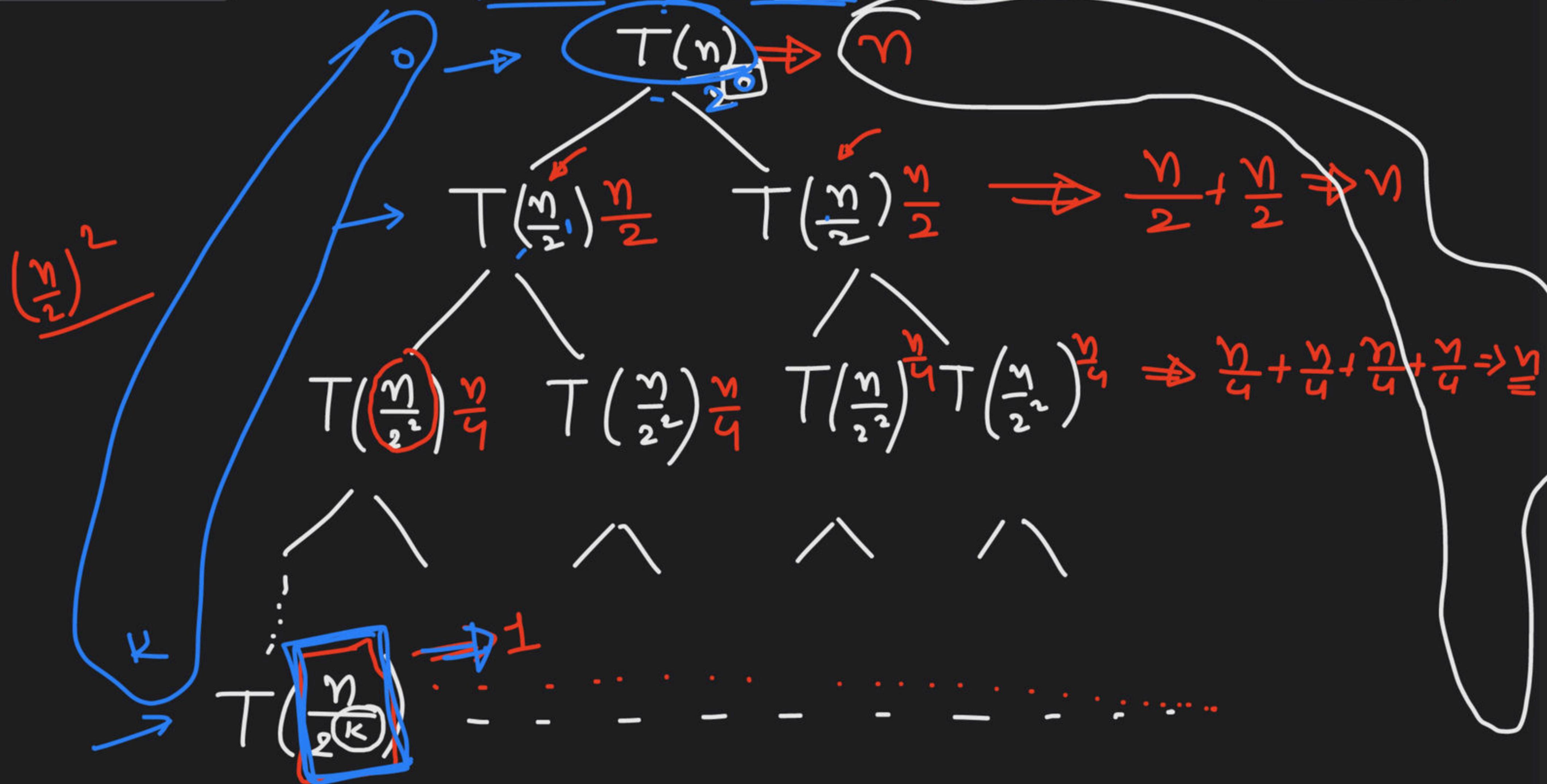
$$T(n) = \underline{\underline{T\left(\frac{n}{4}\right)}} + \underline{\underline{T\left(\frac{3n}{4}\right)}} + n$$

→ we can't apply Masters.

$$T(n) = \underline{\underline{T\left(\frac{n}{12}\right)}} + \underline{\underline{T\left(\frac{n}{12}\right)}} + n$$

$$T(n) = 2\underline{\underline{T\left(\frac{n}{12}\right)}} + n$$

$$T(n) = 2T\left(\frac{n}{2}\right) + n \Rightarrow \Theta(n \log n)$$



$$\underbrace{n+n+\dots}_{\text{sum}} + \underbrace{\dots}_{\text{sum}} = \underbrace{(k+1)}_{\text{sum}} + \text{term}$$

ΔK
 $(K+1) \approx$

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

$$y = 2^x$$

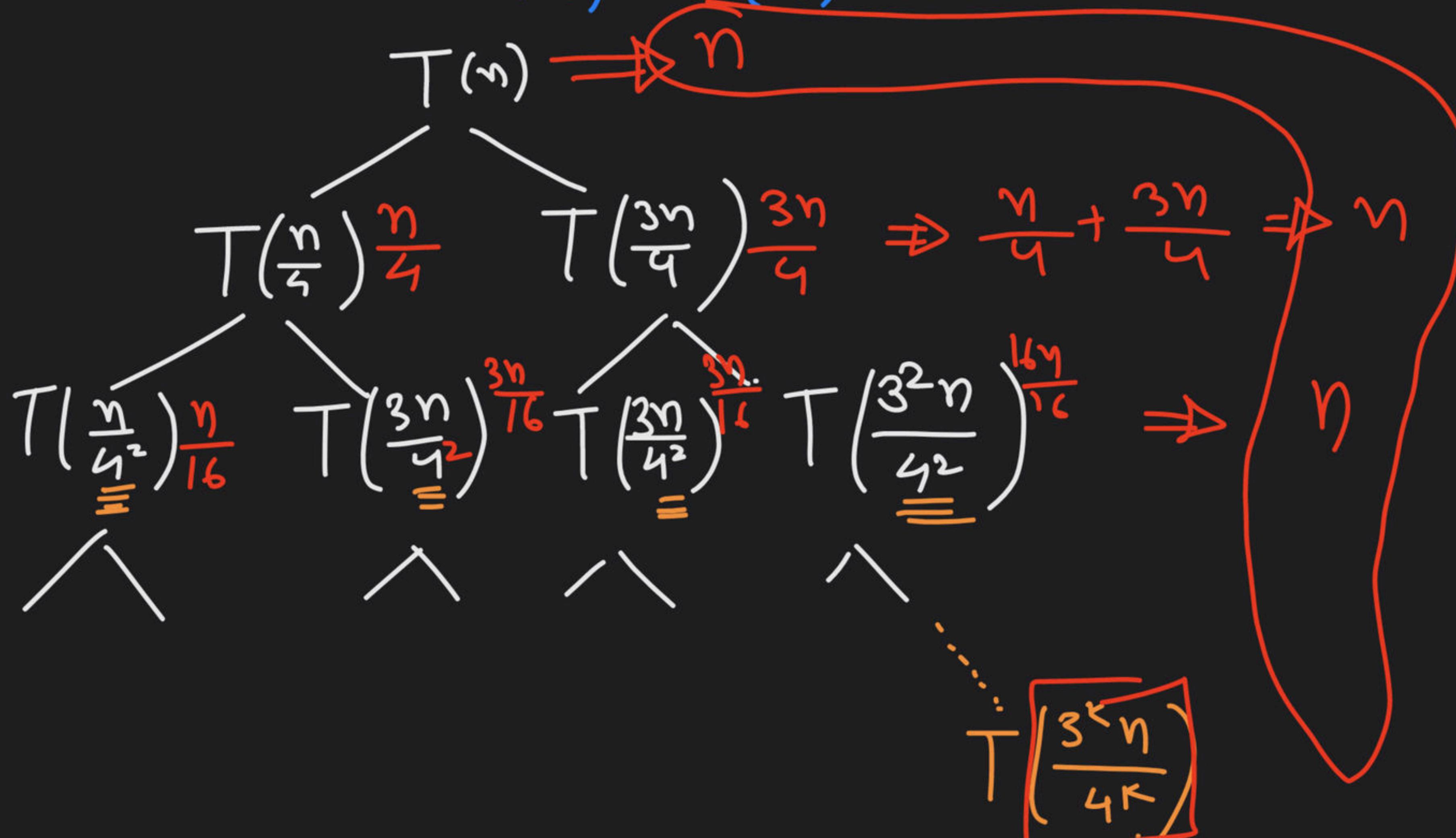
$$K = \log_2 N$$

$$(\log n + 1)^n$$

$$\underline{n \log n + n} =$$

$\rightarrow \Theta(n \log n)$

$$\underline{\underline{T(n)}} = T\left(\frac{n}{4}\right) + T\left(\frac{3n}{4}\right) + \underline{\underline{n}}$$



$$K = \frac{\log n}{\log 413}$$

$$k = \log_{4/3} 3$$

$$\frac{n(k+1)}{n(\log_{4/3} n + 1)} = \underline{\underline{n \log_{4/3} n + n}}$$
$$\underline{\underline{\Theta(n \log n)}}$$

$$\frac{3^k}{4^k} =$$

$$y = \left(\frac{y}{3}\right)\pi$$

$$\log n = \log(4/3)^k$$

$$\log n = \lfloor \log_4 n \rfloor$$

$$T(n) = 2T(\lfloor n/2 \rfloor) + \sqrt{n}$$

$$T(n) = 2T(n/2) + \sqrt{n}$$

$$n^{\log_2 2} \Rightarrow n > \sqrt{n}$$

$\Theta(n)$



$$T(n) = \underline{\underline{T}}(\underline{\underline{n}}_{l_2}) + n^{\checkmark} \Rightarrow$$

$$3|_{l_2} \leftarrow T(n) = \underline{\underline{T}}(\underline{\underline{n}}_{l_3}) + \underline{\underline{T}}\left(\frac{2n}{3}\right) + n^{\checkmark}$$

$$4|_{l_3} \leftarrow T(n) = \underline{\underline{T}}(\underline{\underline{n}}_{l_4}) + \underline{\underline{T}}\left(\frac{3n}{4}\right) + n^{\checkmark}$$

$$5|_{l_4} \leftarrow T(n) = \underline{\underline{T}}(\underline{\underline{n}}_{l_5}) + \underline{\underline{T}}\left(\frac{4n}{5}\right) + n^{\checkmark}$$

$\Theta(n \log n)$

====

$$T(n) = T\left(\frac{n}{3}\right) + T\left(\frac{n}{4}\right) + \boxed{n^2}$$

$$T(n) \Rightarrow n^2$$

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

$$T\left(\frac{n}{4}\right) \Rightarrow \left(\frac{n}{4}\right)^2$$

$$\frac{n^2}{9} + \frac{n^2}{16} \Rightarrow \frac{25n^2}{144}$$

$$T\left(\frac{n}{9}\right) \left(\frac{n}{9}\right)^2$$

$$T\left(\frac{n}{12}\right) \left(\frac{n}{12}\right)^2$$

$$T\left(\frac{n}{12}\right) \left(\frac{n}{12}\right)^2$$

$$T\left(\frac{n}{4^2}\right) \Rightarrow$$

$$\frac{n^2}{81} + \frac{n^2}{144} + \frac{n^2}{144} + \frac{n^2}{256} \Rightarrow \left(\frac{25}{144}\right)^2 n^2$$



$$n^2 + \frac{25n^2}{144} + \left(\frac{25}{144}\right)^2 n^2 + \dots$$

n^2 [$1 + \frac{25}{144} + \left(\frac{25}{144}\right)^2 + \dots + \left(\frac{25}{144}\right)^K$]

$\cancel{n^2}$ → ignore

G.P.

$\Rightarrow \frac{25}{144} < 1$

$\Theta(n^2)$

$$\therefore T(n) = T(\sqrt{n}) + 1$$

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

$$= T(n^{1/4}) + 1 + 1$$

$$= T(n^{1/8}) + 1 + 1 + 1$$

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

⋮

⋮

⋮

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

$$n^{\frac{1}{2^K}} = c$$

$$\log n^{\frac{1}{2^K}} = \log c$$

$$\frac{1}{2^K} \log n = \log c$$

$$\frac{1}{2^K} = \frac{\log c}{\log n}$$

$$= T(c) + K$$

$$\frac{1+K}{1+\log \log n}$$

$$\Rightarrow O(\log \log n)$$

$$2^K = \frac{\log n}{\log c}$$

$$K = \log_2 \log n$$

$$\Rightarrow \log \log n$$

$$2^K = \log_c n$$

$$T(c) = 1$$

$$T(n) = T(\sqrt{n}) + 1$$

$n=2^k$

$$T(2^k) = S(k)$$

$$S(k) = S(k/2) + 1$$

$\Rightarrow k \log_2 1 \Rightarrow k^0 \Rightarrow 1 \equiv 1 \Rightarrow \Theta(\underline{\log k})$

$$T(2^k) = \log k$$

$$T(n) \Rightarrow \log \log n$$

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$n = 2^k$

$T(2^k) = S(k)$

$K = \log n$

$T(n) = 2T(\sqrt{n}) + 1$

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

$S(k) = 2S(k/2) + 1$

$k \xrightarrow{\log_2 2} \frac{k}{2} \Rightarrow k \geq 1$

$\Theta(k)$

$T(2^k) = k$

$T(n) = \log n$

$\Rightarrow \Theta(\log n)$



$$n = 2^k$$

$$\log n = \log_2 k$$

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

How to Calculate Space Complexity of an Algorithm

Space complexity is the amount of memory used by the algorithm to execute and produce the result.

- **Auxiliary Space**
- **Input Space**

Space Complexity

- How to calculate Space complexity of
 - Sequential Algorithm
 - Iterative Algorithm
 - Recursive Algorithm

➤ **Sequential
Algorithm**

➤ **Iterative Algorithm**

```
def initialize_array(A, n):
    for i in range(n):
        A[i] = 0

# Example usage:
n = 10 # Replace with the desired array size
A = [0] * n # Initialize an array of size n with zeros
initialize_array(A, n)
print(A) # You can print the array to verify the initialization
```

```
# Example array A  
A = [1, 2, 3, 4, 5] # Replace with your array
```

```
n = len(A)  
B = [0] * n # Initialize array B with zeros  
C = [0] * n # Initialize array C with zeros
```

```
for i in range(n):  
    B[i] = A[i]  
    C[i] = A[i]
```

```
# Printing the results  
print("Array B:", B)  
print("Array C:", C)
```

```
# Example array A
A = [1, 2, 3, 4, 5] # Replace with your array
n = len(A)

# Create an empty 2D array C with dimensions n x n
C = [[0 for _ in range(n)] for _ in range(n)]

# Copy the elements from A to C
for i in range(n):
    for j in range(n):
        C[i][j] = A[i]

# Printing the resulting 2D array C
for row in C:
    print(row)
```

➤ **Recursive
Algorithm**

```
def f(n):  
    if n / 2 <= 1:  
        return n  
    f(n / 2)  
    f(n / 2)  
  
# Example usage  
result = f(8) # Replace 8 with the desired value  
print(result)
```

Algo 1:

```
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def fun(n):
    if n <= 1:
        return n
    else:
        return fun(n - 1) + fun(n - 1)

# Example usage:
result = fun(5) # Replace 5 with the desired value
print(result)
```

Algo 2:

```
def fun(n):
    if n <= 1:
        return n
    else:
        return 2 * fun(n - 1)

# Example usage:
result = fun(5) # Replace 5 with the desired value
print(result)
```

```
int fun( int n)
{
if(n<=1)
return n;
else
{
fun(n-1);
for(i=1;i<=n; i=i+10);
return i;
}
```

```
def fun(n):
    if n <= 1:
        return n
    else:
        fun(n - 1)
        i = 1
        while i <= n:
            i += 10
        return i

# Example usage:
result = fun(5) # Replace 5 with the desired value
print(result)
```

```
int fun( int n)
{
if(n<=1)
return n;
else
{
fun(n/2);
for(i=1;i<=n; i=i+10);
return i;
}
```

```
def fun(n):
    if n <= 1:
        return n
    else:
        fun(n // 2) # Note: Use integer division (//) to match C++ behavior
        i = 1
        while i <= n:
            i += 10
        return i

# Example usage:
result = fun(20) # Replace 20 with the desired value
print(result)
```

What is Space Complexity?

- (a) $\Theta(n)$
- (b) $\Theta(\log n)$
- (c) $\Theta(n \log n)$
- (d) None

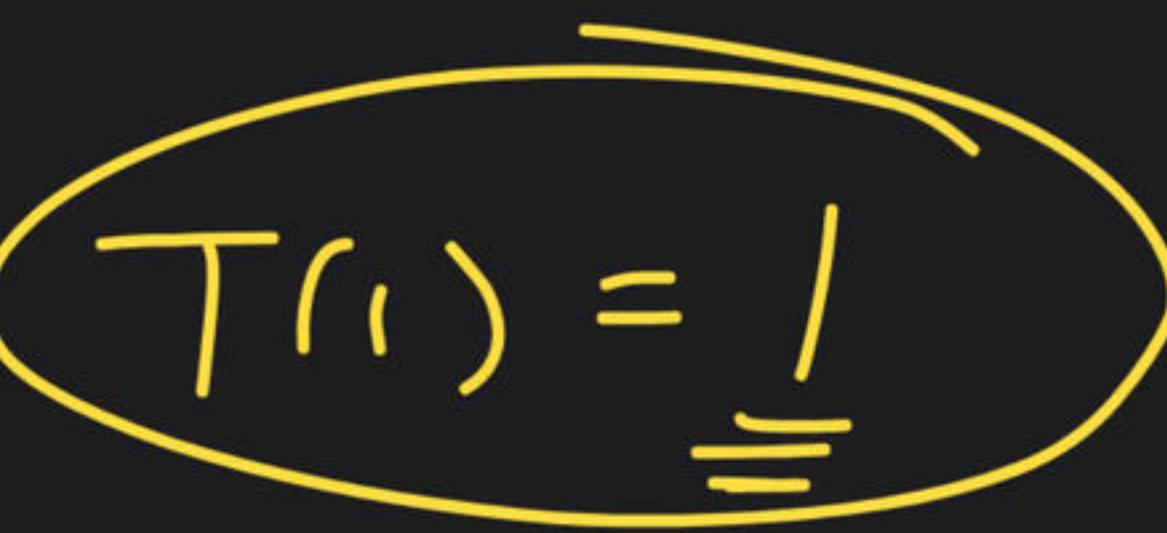
$$= 2 + 2 + 2^2 + \dots - - - - - 2^{n-1}$$

$$= \frac{1 \cdot (2^n - 1)}{(2 - 1)} \Rightarrow 2^n - 1$$

$$\Rightarrow \underline{\Theta(2^n)}$$

If $n = 1 :$





A hand-drawn diagram illustrating a single electron shell. It features a large oval containing the text "Tr(1) = 1" and three short horizontal lines below it, likely representing the radial states for one electron. The oval has a curved arrow at its top right corner.

