



Recurrence Relations

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Need of Recursive Relations

- The following shows the recursive and iterative versions of the factorial function:

Recursive version	Iterative version
<pre>int factorial (int n) { if (n == 0) return 1; else return n * factorial (n-1); }</pre>	<pre>int factorial (int n) { int i, product=1; for (i=n; i>1; --i) product=product * i; return product; }</pre>

Recursive Call

```
factorial(n)
{
    if n == 0 ✓ 1
        return 1
    else
        return n * factorial(n-1)
}
```

↑ 1 ↑ 1

$$T(n) = T(n-1) + 3 \quad \text{if } n > 0$$

$$T(0) = 1$$

$$\begin{aligned} T(n) &= T(n-1) + 3 \\ &= T(n-2) + 6 \\ &= T(n-3) + 9 \\ &= T(n-k) + 3k \end{aligned}$$

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

$$\begin{aligned} \Rightarrow T(n) &= T(0) + 3n \\ &= 3n \end{aligned}$$

Recurrence Relations (1/2)

- A recurrence relation is an equation which is defined in terms **of itself** with smaller value.
- Why are recurrences good things?
 - Many natural functions are easily expressed as recurrences:
 - $a_n = a_{n-1} + 1, a_1 = 1 \rightarrow a_n = n$ (polynomial)
 - $a_n = 2a_{n-1}, a_1 = 1 \rightarrow a_n = 2^n$ (exponential)
 - $a_n = na_{n-1}, a_1 = 1 \rightarrow a_n = n!$ (weird function)
- It is often easy to find a recurrence as the solution of a counting problem

Recurrence Relations (2/2)

- In both, we have general and boundary conditions, with the general condition breaking the **problem into smaller and smaller pieces**.
- The initial or boundary condition terminate the recursion.

Recurrence Equations

- A recurrence equation defines a function, say $T(n)$. The function is defined recursively, that is, the function $T(\cdot)$ appear in its definition. (recall recursive function call). The recurrence equation should have a base case.

For example:

$$T(n) = \begin{cases} T(n-1)+T(n-2), & \text{if } n>1 \\ 1, & \text{if } n=1 \text{ or } n=0. \end{cases}$$

base case

for convenient, we sometime write the recurrence equation as:

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

$$T(0) = T(1) = 1.$$

Recurrence Examples

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

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

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

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

Methods for Solving Recurrences

- Iteration method (Backward Substitution Method
- Substitution method
- Recursion tree method
- Master method

Simplifications:

- There are two simplifications we apply that won't affect asymptotic analysis
 - Ignore floors and ceilings (justification in text)
 - Assume base cases are constant, i.e., $T(n) = \Theta(1)$ for n small enough

Iteration Method (Backward Substitution)

- Expand the recurrence
- Work some algebra to express as a summation
- Evaluate the summation

Iteration Method

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

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

$$= c + c + T(n/4)$$

$$= c + c + c + T(n/8)$$

Assume $n = 2^k$ $k = \log_2 n$

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

$$= \underbrace{c + c + \dots + c}_{\log_2 n \text{ times}} + T(1)$$

$$= \Theta(\lg n)$$

$$T(n/2) = c + T(n/4)$$

$$T(n/4) = c + T(n/8)$$

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

- $s(n) =$
 - $c + s(n-1)$
 - $c + c + s(n-2)$
 - $2c + s(n-2)$
 - $2c + c + s(n-3)$
 - $3c + s(n-3)$
 - ...
 - $kc + s(n-k) = ck + s(n-k)$
- What if $k = n$?
 - $s(n) = cn + s(0) = cn$

Iteration Method

Example: $T(n) = 4T(n/2) + n$

$$T(n) = 4T(n/2) + n \quad /**T(n/2)=4T(n/4)+n/2$$

$$= 4(4T(n/4)+n/2) + n \quad /**simplify**/$$

$$= 16T(n/4) + 2n + n \quad /**T(n/4)=4T(n/8)+n/4$$

$$= 16(4T(n/8)+n/4) + 2n + n \quad /**simplify**/$$

$$= 64(T(n/8) + 4n + 2n + n$$

$$= 4^{\log n} T(1) + \dots + 4n + 2n + n \quad /** \#levels = \log n **/$$

$$= c4^{\log n} + n \sum_{k=0}^{\log n - 1} 2^k \quad /** convert to summation**/$$

$$= cn^{\log 4} + n\left(\frac{2^{\log n} - 1}{2 - 1}\right) \quad /** a^{\log b} = b^{\log a} **/$$

Solving Recurrences: Iteration

(convert to summation) (cont.)

$$\begin{aligned} &= cn^2 + n(n^{\log 2} - 1) && /** \ 2^{\log n} = n^{\log 2} \ */ \\ &= cn^2 + n(n - 1) \\ &= cn^2 + n^2 - n \\ &= \Theta(n^2) \end{aligned}$$

$$1. T(n) = T(n-1) + n \quad n > 1$$

$$T(n) = 1 \quad n = 1$$

Solution:

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

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

$$\text{Substituting (2) in (1)} \quad T(n) = T(n-2) + n-1 + n$$

$$\dots(3)$$

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

$$\dots(4)$$

Substituting (4) in (3)

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

$$\dots(5)$$

General equation

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

$$\dots(6)$$

$$T(1) = 1$$

$$n-k=1$$

$$k = n-1$$

$$\dots(7)$$

Substituting (7) in (6)

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

$$= 1 + 2 + 3 + \dots + n$$

$$= n(n+1)/2 \quad T(n) = O(n^2)$$

$$2. T(n) = T(n-1) + b \quad n > 1$$

$$T(n) = 1 \quad n = 1$$

Solution:

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

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

Substituting (2) in (1)

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

$$T(n) = T(n-2) + 2b \quad \dots\dots(3)$$

$$T(n-2) = T(n-3) + b \quad \dots\dots(4)$$

Substituting (4) in (3)

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

General equation $T(n)$

$$= T(n-k) + k.b \quad T(1)$$

$$= 1$$

$$n-k=1 \quad k=n-1$$

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

$$= 1 + bn - b \quad T(n) =$$

$$O(n)$$

$$3. T(n) = 2T(n-1) + b \quad n > 1$$

$$T(n) = 1 \quad n = 1$$

Solution:

$$T(1) = 1$$

$$T(2) = 2T(1) + b$$

$$= 2 + b$$

$$= 2^1 + b$$

$$T(3) = 2T(2) + b$$

$$= 2(2 + b) + b$$

$$= 4 + 2b + b$$

$$= 4 + 3b$$

$$= 2^2 + (2^2 - 1)b \quad T(4) =$$

$$2T(3) + b$$

$$= 2(4 + 3b) + b$$

$$= 8 + 7b$$

$$= 2^3 + (2^3 - 1)b$$

General equation

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

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

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

$$= 2^n(b + 1)/2 - b \quad \text{Let } c$$

$$= (b + 1)/2 \quad T(n) = c 2^n$$

$$- b$$

$$= O(2^n)$$

$$4. T(n) = T(n/2) + b \quad n > 1$$

$$T(1) = 1 \quad n = 1$$

Solution:

$$T(n) = T(n/2) + b \quad \dots(1)$$

$$T(n/2) = T(n/4) + b \quad \dots(2)$$

Substituting (2) in (1)

$$T(n) = T(n/4) + b + b$$

$$= T(n/4) + 2b \quad \dots(3)$$

$$T(n/4) = T(n/8) + b \quad \dots(4)$$

Substituting (4) in (3)

$$T(n) = T(n/8) + 3b$$

$$= T(n/2^3) + 3b$$

General equation

$$T(n) = T(n/2^k) + kb \quad \dots\dots(5)$$

$$T(1) = 1$$

$$n/2^k = 1$$

$$2^k = n$$

$$K = \log n \quad \dots\dots(6)$$

Substituting (6) in (5)

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

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

$$O(\log n)$$

The substitution method

1. Guess a solution
2. Use induction to prove that the solution works

Substitution method

- Guess a solution
 - $T(n) = O(g(n))$
 - Induction goal: **apply the definition of the asymptotic notation**
 - $T(n) \leq d g(n)$, for some $d > 0$ and $n \geq n_0$
 - Induction hypothesis: $T(k) \leq d g(k)$ for all $k < n$
- Prove the induction goal
 - Use the **induction hypothesis** to **find some values of the constants d and n_0** for which the **induction goal** holds

Example: Binary Search

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

- Guess: $T(n) = O(\lg n)$
 - Induction goal: $T(n) \leq d \lg n$, for some d and $n \geq n_0$
 - Induction hypothesis: $T(n/2) \leq d \lg(n/2)$
- Proof of induction goal:

$$T(n) = T(n/2) + c \leq d \lg(n/2) + c$$

$$= d \lg n - d + c \leq d \lg n$$

$$\text{if: } -d + c \leq 0, d \geq c$$

Example 2

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

- Guess: $T(n) = O(n^2)$
 - Induction goal: $T(n) \leq c n^2$, for some c and $n \geq n_0$
 - Induction hypothesis: $T(k-1) \leq c(k-1)^2$ for all $k < n$
- Proof of induction goal:

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

$$= cn^2 - (2cn - c - n) \leq cn^2$$

$$\text{if: } 2cn - c - n \geq 0 \Leftrightarrow c \geq n/(2n-1) \Leftrightarrow c \geq 1/(2 - 1/n)$$

- For $n \geq 1 \Rightarrow 2 - 1/n \geq 1 \Rightarrow$ any $c \geq 1$ will work

Example 3

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

- Guess: $T(n) = O(n \lg n)$
 - Induction goal: $T(n) \leq cn \lg n$, for some c and $n \geq n_0$
 - Induction hypothesis: $T(n/2) \leq cn/2 \lg(n/2)$
- Proof of induction goal:

$$\begin{aligned} T(n) &= 2T(n/2) + n \leq 2c (n/2) \lg(n/2) + n \\ &= cn \lg n - cn + n \leq cn \lg n \end{aligned}$$

$$\text{if: } -cn + n \leq 0 \Rightarrow c \geq 1$$

Changing variables

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

- Rename: $m = \lg n \Rightarrow n = 2^m$

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

- Rename: $S(m) = T(2^m)$

$$S(m) = 2S(m/2) + m \Rightarrow S(m) = O(m \lg m)$$

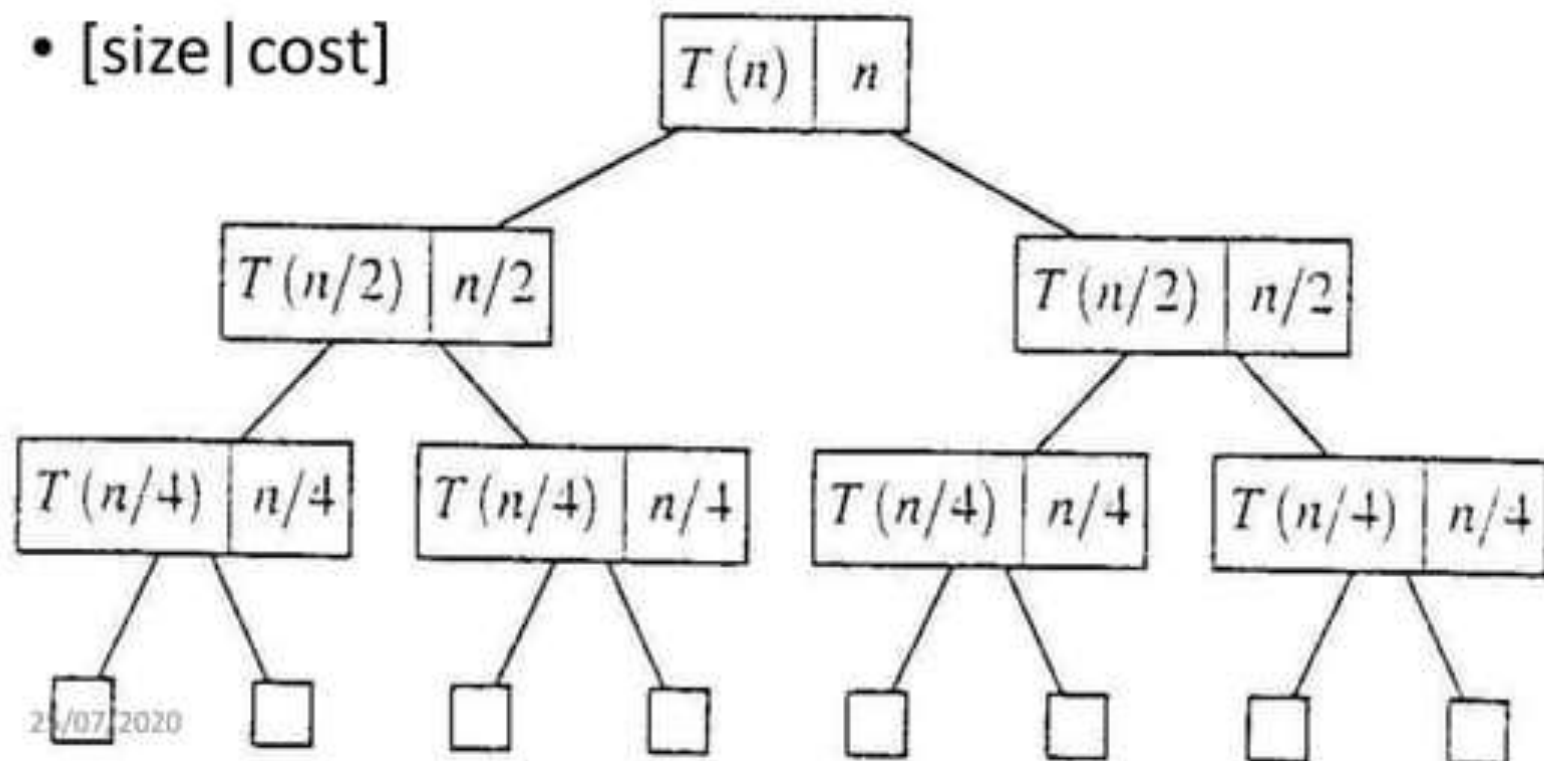
(demonstrated before)

$$T(n) = T(2^m) = S(m) = O(m \lg m) = O(\lg n \lg \lg n)$$

Idea: transform the recurrence to one that you have seen before

Recursion Tree

- Evaluate: $T(n) = T(n/2) + T(n/2) + n$
 - Work copy: $T(k) = T(k/2) + T(k/2) + k$
 - For $k=n/2$, $T(n/2) = T(n/4) + T(n/4) + (n/2)$
- [size | cost]



Recursion-tree method

- A recursion tree models the costs (time) of a recursive execution of an algorithm.
- The recursion tree method is **good for generating guesses** for the substitution method.
- The recursion-tree method can be **unreliable**.
- The recursion-tree method promotes intuition, however.

Recursion Tree

- To evaluate the total cost of the recursion tree
Sum all the non-recursive costs of all nodes
= Sum (rowSum(cost of all nodes at the same depth))
- Determine the maximum depth of the recursion tree:
For our example, at tree depth d
the size parameter is $n/(2^d)$
the size parameter converging to base case, i.e. case 1
such that, $n/(2^d) = 1$,
 $d = \lg(n)$
The row Sum for each row is n
- Therefore, the total cost, $T(n) = n \lg(n)$

Example of recursion tree

Solve $T(n) = T(n/4) + T(n/2) + n^2$:

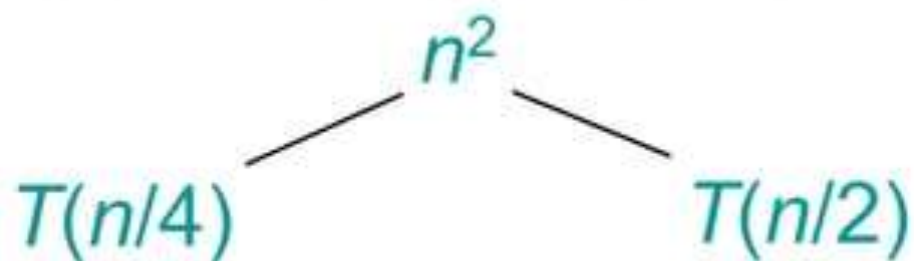
Example of recursion tree

Solve $T(n) = T(n/4) + T(n/2) + n^2$:

$T(n)$

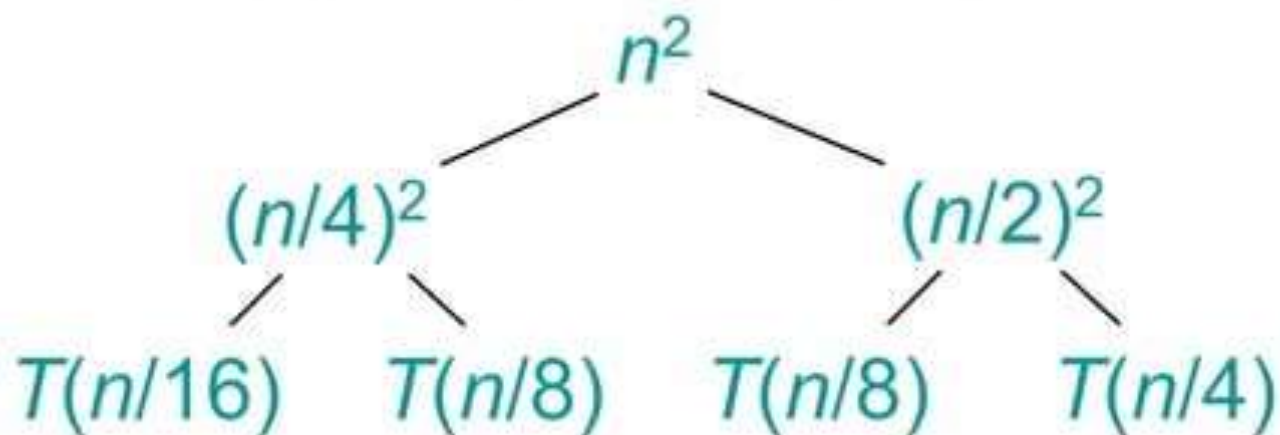
Example of recursion tree

Solve $T(n) = T(n/4) + T(n/2) + n^2$:



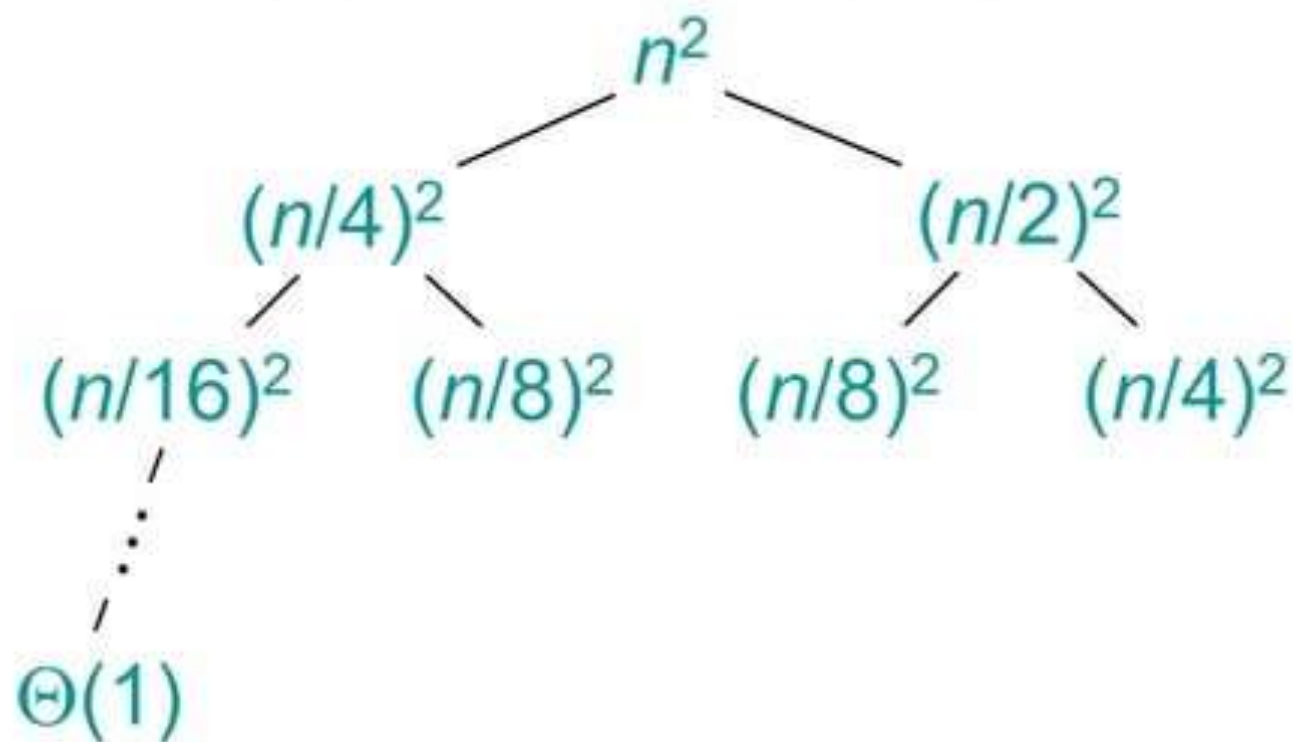
Example of recursion tree

Solve $T(n) = T(n/4) + T(n/2) + n^2$:



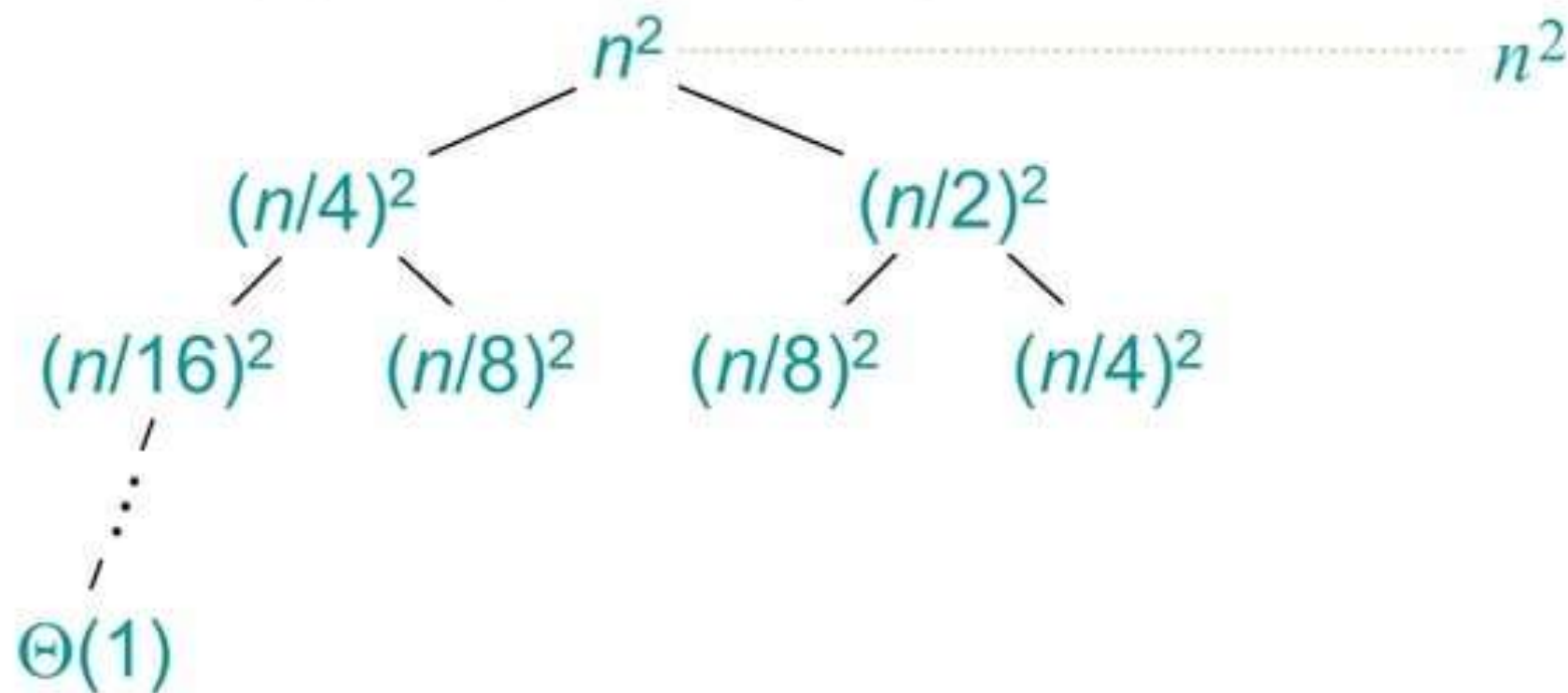
Example of recursion tree

Solve $T(n) = T(n/4) + T(n/2) + n^2$:



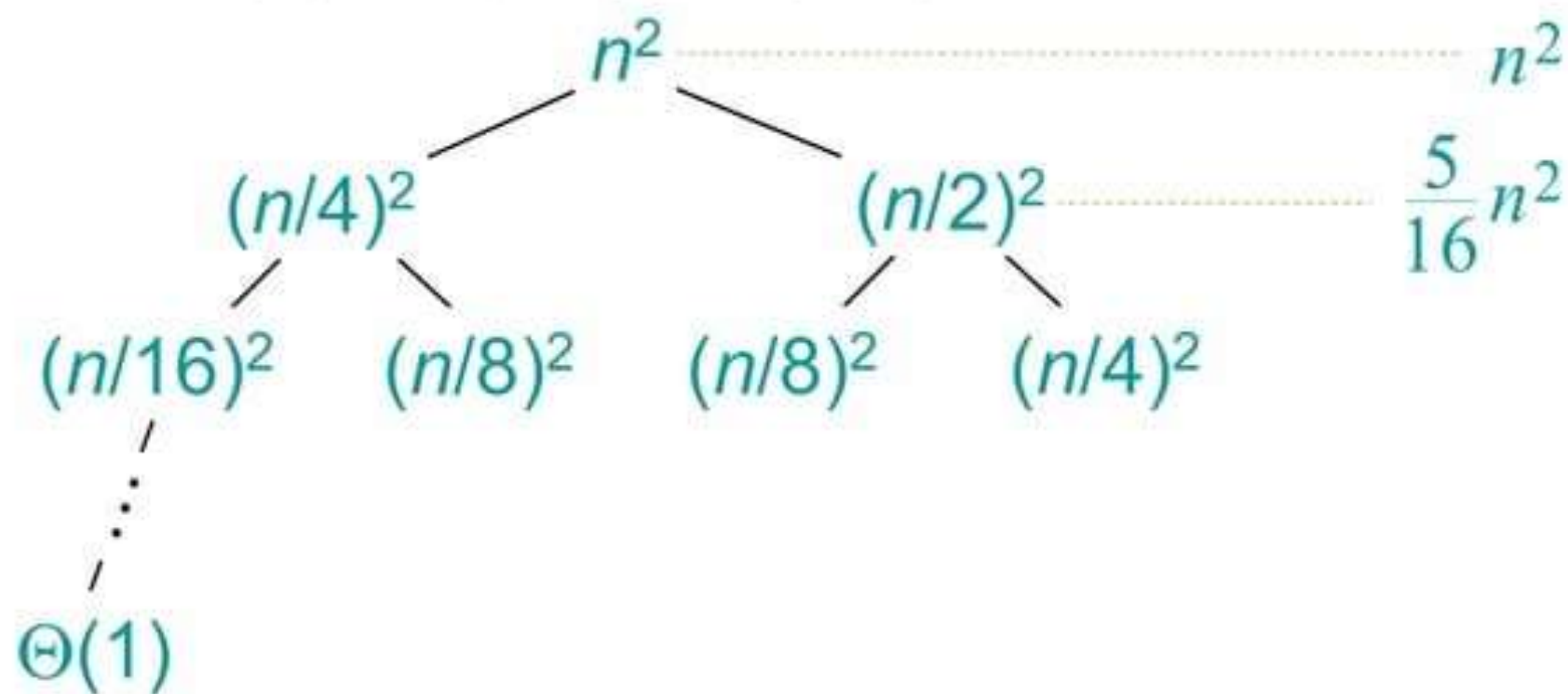
Example of recursion tree

Solve $T(n) = T(n/4) + T(n/2) + n^2$:



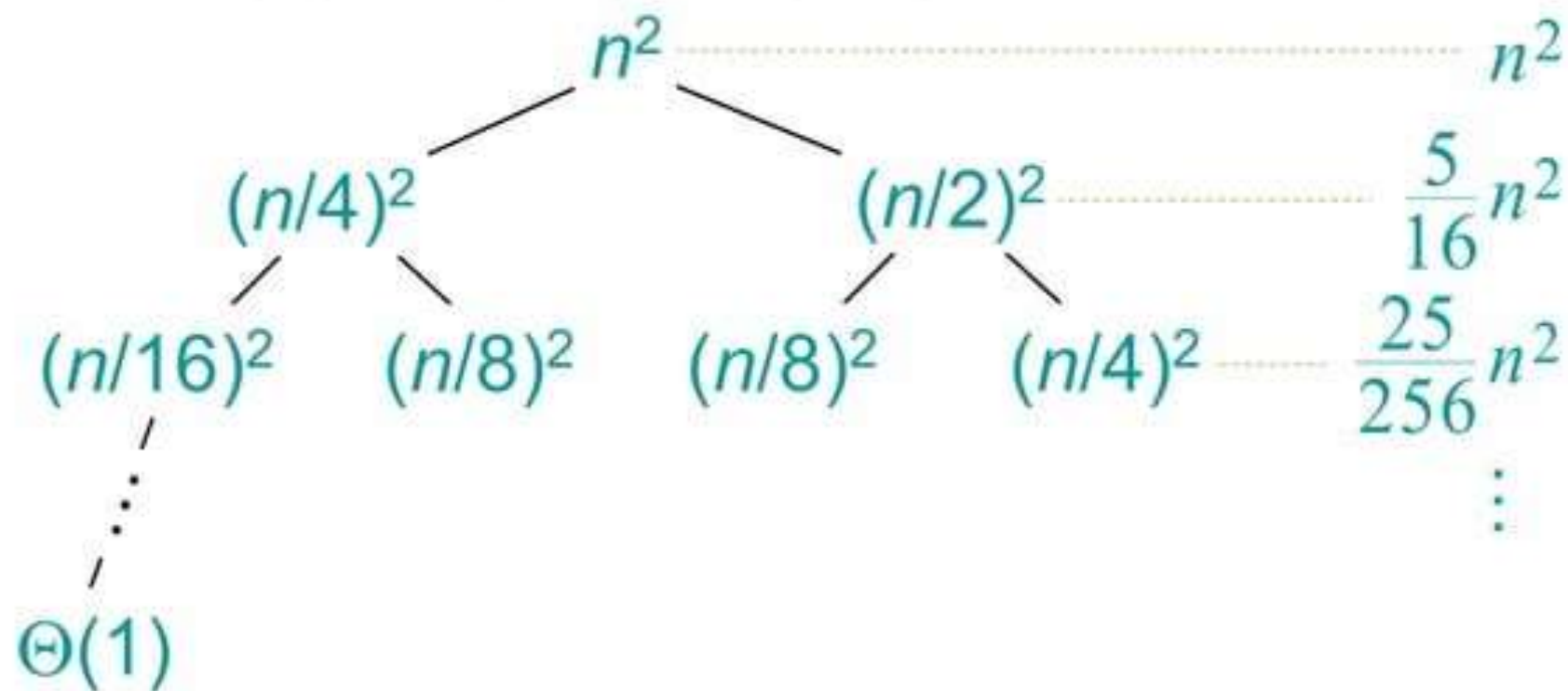
Example of recursion tree

Solve $T(n) = T(n/4) + T(n/2) + n^2$:



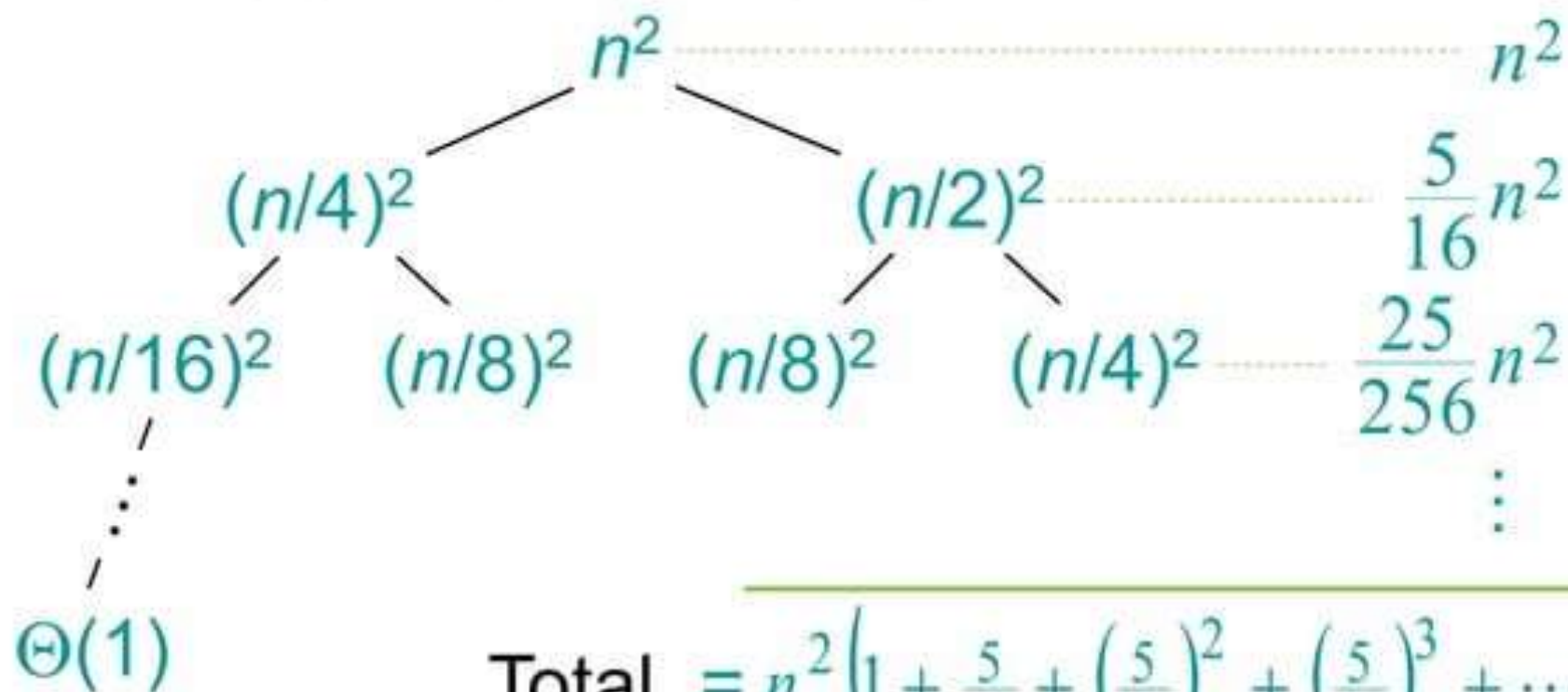
Example of recursion tree

Solve $T(n) = T(n/4) + T(n/2) + n^2$:



Example of recursion tree

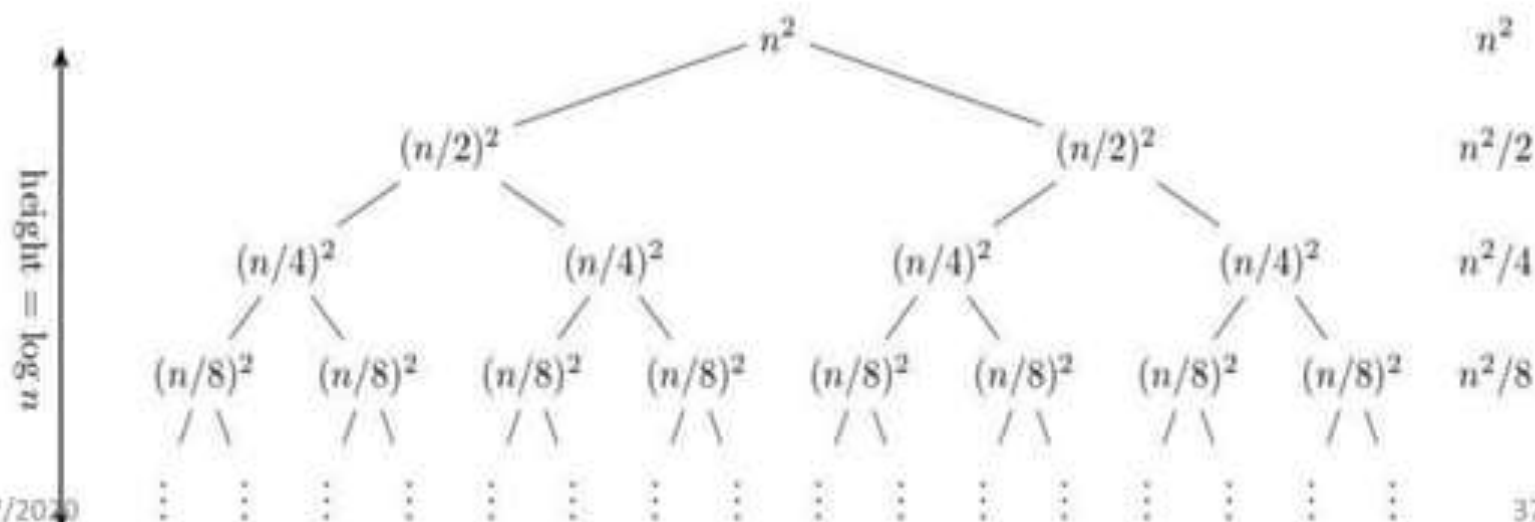
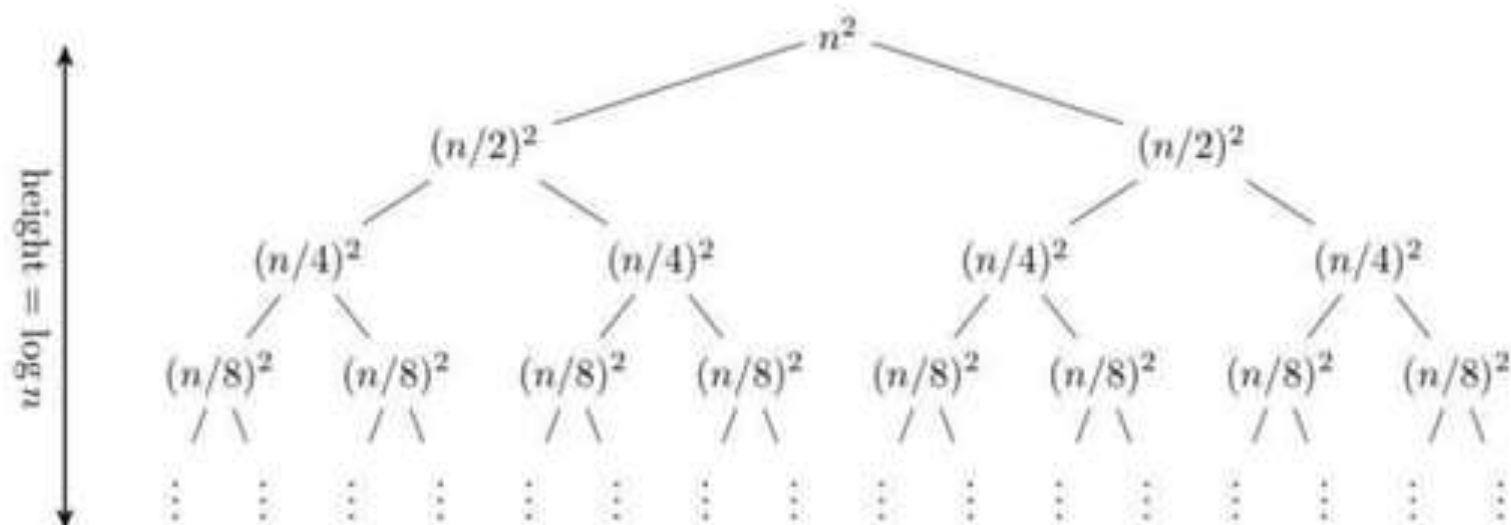
Solve $T(n) = T(n/4) + T(n/2) + n^2$:



$$\begin{aligned}\text{Total} &= n^2 \left(1 + \frac{5}{16} + \left(\frac{5}{16}\right)^2 + \left(\frac{5}{16}\right)^3 + \dots \right) \\ &= \Theta(n^2) \quad \text{geometric series}\end{aligned}$$

• **Example**

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



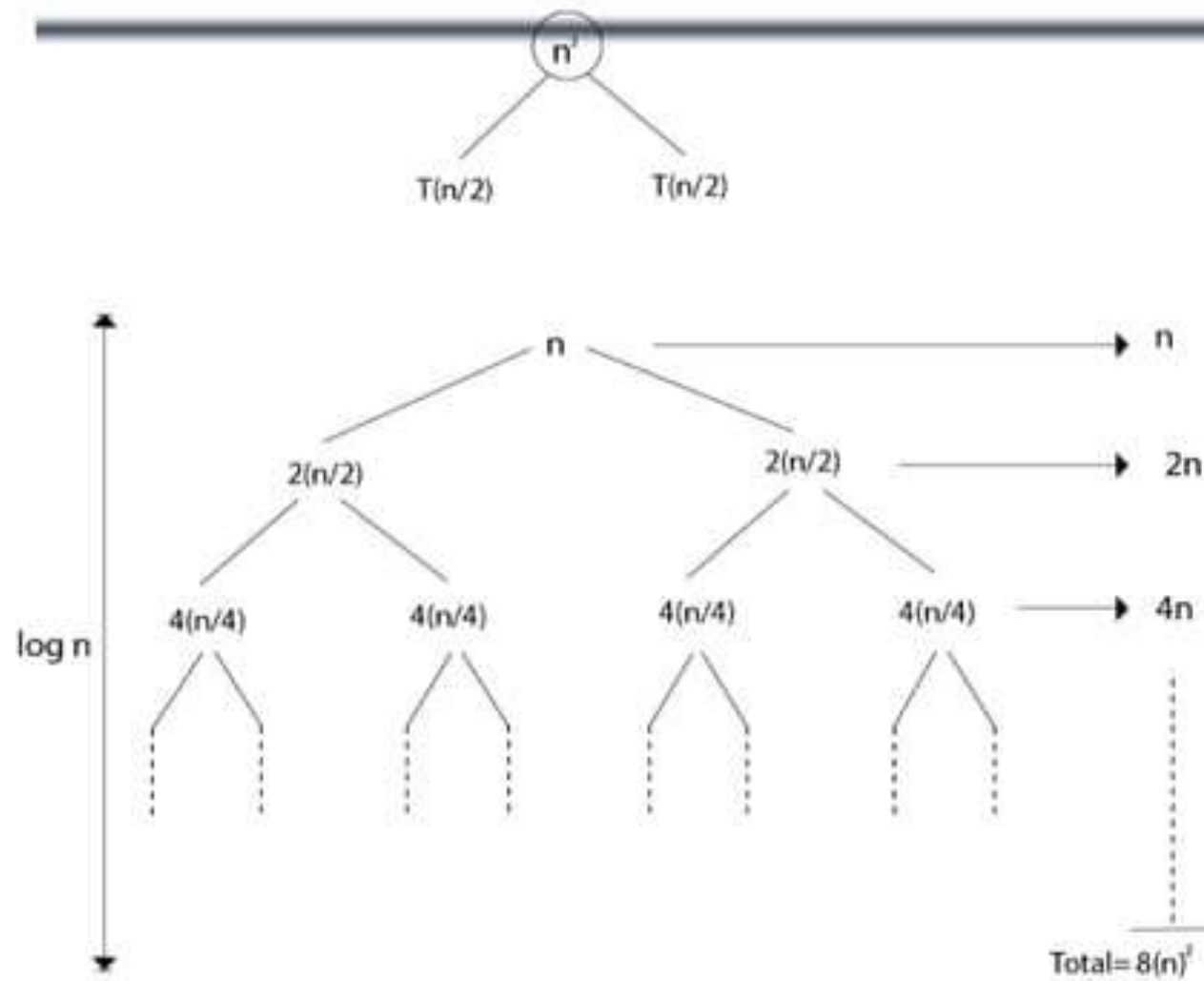
$$T(n) = n^2 + \frac{n^2}{2} + \frac{n^2}{4} + \dots \log n \text{ times.}$$

$$\leq n^2 \sum_{i=0}^{\infty} \left(\frac{1}{2^i}\right)$$

$$\leq n^2 \left(\frac{1}{1-\frac{1}{2}}\right) \leq 2n^2$$

$$T(n) = \Theta(n^2)$$

Example : $T(n) = 4T(n/2) + n$



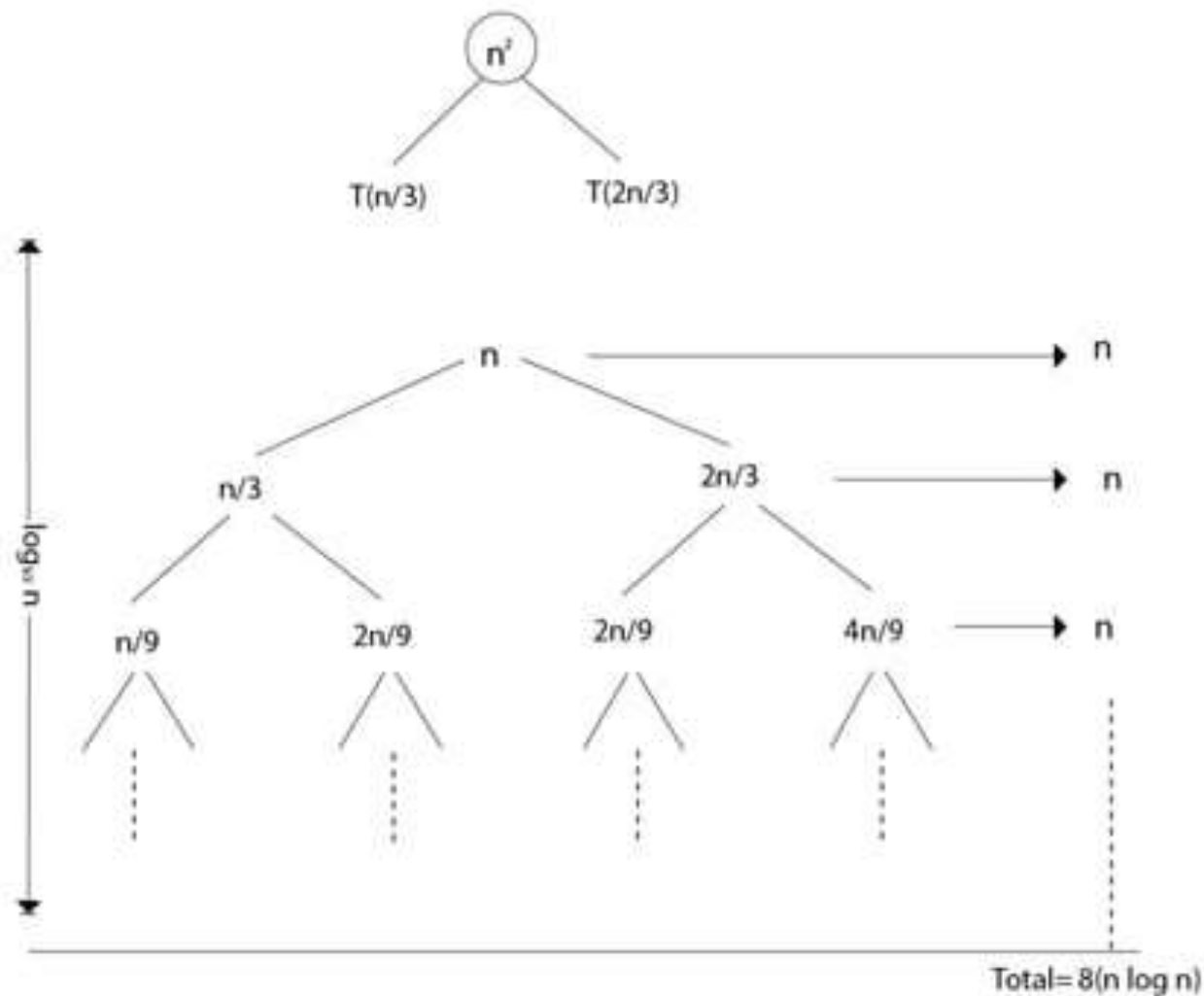
We have $n + 2n + 4n + \dots \log_2 n$ times

$$= n (1 + 2 + 4 + \dots \log_2 n \text{ times})$$

$$= n \frac{(2^{\log_2 n} - 1)}{(2 - 1)} = \frac{n(n - 1)}{1} = n^2 - n = \theta(n^2)$$

$$T(n) = \theta(n^2)$$

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



$$n \rightarrow \frac{2}{3}n \rightarrow \left(\frac{2}{3}\right)^2 n \rightarrow \dots 1$$

Since $\left(\frac{2}{3}\right)^i n = 1$ when $i = \log_{\frac{3}{2}} n$.

Thus the height of the tree is $\log_{\frac{3}{2}} n$.

$$T(n) = n + n + n + \dots + \log_{\frac{3}{2}} n \text{ times.} = \theta(n \log n)$$

Let k th steps, it moves up to $n/3^k$

It moves until 1

The total sum up to k th step $= kn$

$$\text{If } \frac{n}{3^k} = 1$$

$$n = 3^k$$

$$k = \log n$$

Total time taken $= n \log n$

The Master Method

- Based on the **Master theorem**.
- **“Cookbook”** approach for solving recurrences of the form

$$T(n) = aT(n/b) + f(n)$$

- $a \geq 1, b > 1$ are constants.
 - $f(n)$ is asymptotically positive.
 - n/b may not be an integer, but we ignore floors and ceilings.
- Requires memorization of three cases.

The Master Theorem

Theorem 4.1

Let $a \geq 1$ and $b > 1$ be constants, let $f(n)$ be a function, and Let $T(n)$ be defined on nonnegative integers by the recurrence $T(n) = aT(n/b) + f(n)$, where we can replace n/b by $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. $T(n)$ can be bounded asymptotically in three cases:

1. If $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
2. If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \lg n)$.
3. If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$, and if, for some constant $c < 1$ and all sufficiently large n , we have $a \cdot f(n/b) \leq c f(n)$, then $T(n) = \Theta(f(n))$.

Master Method – Examples

- $T(n) = 16T(n/4) + n$

- $a = 16, b = 4, n^{\log_b a} = n^{\log_4 16} = n^2.$
- $f(n) = n = O(n^{\log_b a - \varepsilon}) = O(n^{2-\varepsilon})$, where $\varepsilon = 1 \Rightarrow$ **Case 1.**
- Hence, $T(n) = \Theta(n^{\log_b a}) = \Theta(n^2).$

- $T(n) = T(3n/7) + 1$

- $a = 1, b = 7/3$, and $n^{\log_b a} = n^{\log_{7/3} 1} = n^0 = 1$
- $f(n) = 1 = \Theta(n^{\log_b a}) \Rightarrow$ **Case 2.**
- Therefore, $T(n) = \Theta(n^{\log_b a} \lg n) = \Theta(\lg n)$

Master Method – Examples

- $T(n) = 3T(n/4) + n \lg n$

- $a = 3, b=4$, thus $n^{\log_b a} = n^{\log_4 3} = O(n^{0.793})$
- $f(n) = n \lg n = \Omega(n^{\log_4 3 + \varepsilon})$ where $\varepsilon \approx 0.2 \Rightarrow$ **Case 3.**
- Therefore, $T(n) = \Theta(f(n)) = \Theta(n \lg n)$.

- $T(n) = 2T(n/2) + n \lg n$

- $a = 2, b=2, f(n) = n \lg n$, and $n^{\log_b a} = n^{\log_2 2} = n$
- $f(n)$ is asymptotically larger than $n^{\log_b a}$, but **not polynomially larger**. The ratio $\lg n$ is asymptotically less than n^ε for any positive ε . Thus, the Master Theorem **doesn't** apply here.

Master Method – Examples

- $T(n) = 9T(n/3) + n$
 - $a=9, b=3, f(n) = n$
 - $n^{\log_b a} = n^{\log_3 9} = \Theta(n^2)$
 - Since $f(n) = n, f(n) < n^{\log_b a}$
 - **Case 1 applies:**

$$T(n) = \begin{cases} \Theta(n^{\log_b a}) & f(n) = O(n^{\log_b a - \epsilon}) \\ \Theta(n^{\log_b a} \log n) & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)) & f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ AND } af(n/b) < cf(n) \text{ for large } n \end{cases} \begin{matrix} \epsilon > 0 \\ c < 1 \end{matrix}$$

$$T(n) = \Theta(n^{\log_b a}) \text{ when } f(n) = O(n^{\log_b a - \epsilon})$$

- Thus the solution is $T(n) = \Theta(n^2)$

Problems on The Master Method

- $T(n) = 4T(n/2) + n^2$
- $a = 4, b = 2, f(n) = n^2$
- $n^{\log_b a} = n^2$
- Since, $f(n) = n^2$
- Thus, $f(n) = n^{\log_b a}$

$$T(n) = \begin{cases} \Theta(n^{\log_b a}) & f(n) = O(n^{\log_b a - \epsilon}) \\ \Theta(n^{\log_b a} \log n) & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)) & f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ AND } af(n/b) < cf(n) \text{ for large } n \end{cases} \begin{matrix} \epsilon > 0 \\ c < 1 \end{matrix}$$

- **Case 2 applies:**
 $f(n) = \Theta(n^2 \log n)$

- Thus the solution is $T(n) = \Theta(n^2 \log n)$.

Master Method – Examples

- **Ex.** $T(n) = 4T(n/2) + n^3$
 - $a = 4, b = 2, f(n) = n^3$ $n^{\log_b a}$
 - $= n^2; f(n) = n^3.$

Since, $f(n) = n^3$ Thus, $f(n) > n^{\log_b a}$

$$T(n) = \begin{cases} \Theta(n^{\log_b a}) & f(n) = O(n^{\log_b a - \epsilon}) \\ \Theta(n^{\log_b a} \log n) & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)) & f(n) = \Omega(n^{\log_b a + \epsilon}) \text{ AND } af(n/b) < cf(n) \text{ for large } n \end{cases} \left. \begin{array}{l} \epsilon > 0 \\ c < 1 \end{array} \right\}$$

- **Case 3 applies:**

$$f(n) = \Omega(n^3)$$

- **and** $4(n/2)^3 \leq cn^3$ (regularity condition) for $c = 1/2$.

$$T(n) = \Theta(n^3).$$

Master Method – Examples

1. $T(n) = 4T(n/2) + n$ $n > 1$

$T(n) = 1$ $n = 1$

From the above recurrence relation we obtain

$a = 4, b = 2, c = 1, d = 1, f(n) = n$

$\log_2 a = \log_2 4 = \log_2 2^2 = 2 \log_2 2 = 2$

$n^{\log_2 a} = n^2$

Master Method – Examples

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

$n = O(n^2)$ It will fall in Case 1. So that

$$T(n) = \theta(n^2)$$

$$2. \quad T(n) = 4T(n/2) + n^2 \quad n > 1 \quad T(n) = 1 \quad n = 1$$

From the above recurrence relation we obtain $a = 4, b = 2, c = 1, d = 1, f(n) = n^2$

$$p^{\log_b a} = n^2$$

$$f(n) = \theta(n^2) \quad n^2 = \theta(n^2)$$

It will fall in case 2.

$$T(n) = \theta(n^2 \log n)$$

$$3. \quad T(n) = 4T(n/2) + n^3 \quad n > 1$$

$$T(n) = 1 \quad n = 1$$

From the above recurrence relation we obtain

$$a = 4, b = 2, c = 1, d = 1, f(n) = n^3$$

$$p^{\log_b a} = n^3$$

$$f(n) = \Omega(n^{\log_b a + \epsilon})$$

Recurrence Relation: — ① Substitution ② Recursion Tree ③ Master Method

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

Ex: C program.

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

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

Substitution method:-

Assume about recurrence relation.

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

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

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

Substitute

$$T(n) = [T(n-3)] + 3$$

Continue for k times

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

Assume $n-k=0$

$$\therefore n=k$$

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

$$= T(0) + n$$

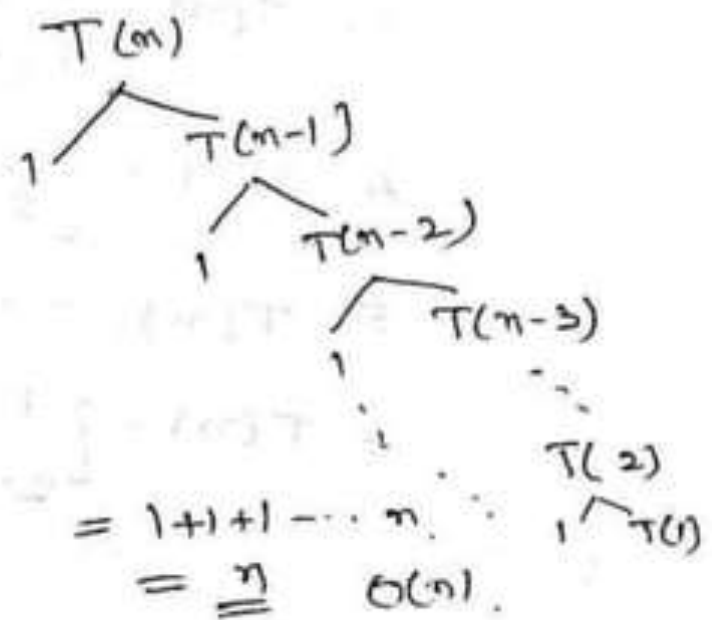
$$T(n) = 1 + n$$

$$= O(n)$$

25/07/2020

Substitution method.

Recursion Tree Method



Examples

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

Substitution method

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

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

$$T(\frac{n}{4}) = T(\frac{n}{2^3}) + c$$

Substitute:

$$T(n) = T(\frac{n}{2^k}) + kc$$

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

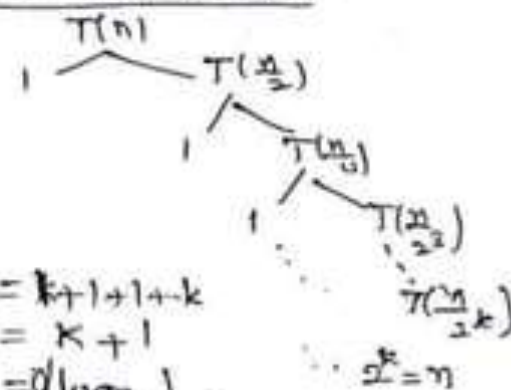
$$= 1 + kc$$

$$= 1 + c \log_2 n \quad \therefore O(\log_2 n)$$

```

Ex: T(n) - Void test(int n)
        {
            if (n > 1)
                print(" " + n)
            T(n/2)
        }
    
```

Recursion Tree method



Examples

Problems: on Iterative Substitution
Recursion Tree method.

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

$$2) T(n) = \begin{cases} 1 & n=1 \\ 2T(\frac{n}{2})+4n & n>1 \end{cases}$$

$$3) T(n) = \begin{cases} 2 & n=1 \\ 2T(\frac{n}{2})+7 & n>1 \end{cases}$$

$$4) T(n) = \begin{cases} 2 & n=1 \\ 2T(\frac{n}{2})+8 & n>1 \end{cases}$$

$$5) T(n) = 5T(\frac{n}{5}) + \frac{n}{\log n}$$

$$6) T(n) = \begin{cases} 1 & n \leq 4 \\ 2T(\sqrt{n}) + \log n & n > 4 \end{cases}$$

Solution for $T(n) = \begin{cases} 1 & n \leq 4 \\ 2T(\sqrt{n}) + \log n \end{cases}$

Substitution method.

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

$$T(n^{\frac{1}{2}}) = 2T(n^{\frac{1}{4}}) + \log \sqrt{n}$$

$$T(n^{\frac{1}{4}}) = 2T(n^{\frac{1}{8}}) + \frac{1}{2} \log n$$

Substitute -

$$T(n) = 2^3 T(n^{\frac{1}{8}}) + \frac{1}{2} \log n$$

$$= 2^3 T(n^{\frac{1}{8}})$$

Master Method – Examples

Master Method

$$T(n) = a T\left(\frac{n}{b}\right) + f(n), \quad \text{where } a \geq 1, b > 1$$

$f(n)$ should be +ve.

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

II. If $f(n) = n^{\log_b a}$ then $T(n) = \Theta(n^{\log_b a} \cdot \log n)$

III. If $f(n) > n^{\log_b a}$ then $T(n) = \Theta(f(n))$

Example: $T(n) = 4T\left(\frac{n}{2}\right) + n$.

given: $a=4$ $b=2$ $f(n)=n$ $\therefore n^{\log_b a} = n^{\log_2 4} = n^2$

then: $n < n^2 \Rightarrow$ case I then $T(n) = \Theta(n^2)$.

Example: $T(n) = 2T\left(\frac{n}{2}\right) + n$. Ans: $T(n) = \Theta(n \log n)$.

(i) $T(n) = 8T\left(\frac{n}{4}\right) + n^2$ — case III

(ii) $T(n) = 4T\left(\frac{n}{2}\right) + n^3$ — case III

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

{ 3rd: $T(n) = 5T\left(\frac{n}{2}\right) + n^3$

$T(n) = 3T\left(\frac{n}{2}\right) + n^3 \quad a=3 \quad b=2$ }

(v) $T(n) = 16T\left(\frac{n}{8}\right) + n^2$

Some Common Recurrence Relation

Recurrence Relation	Complexity	Problem
$T(n) = T(n/2) + c$	$O(\log n)$	Binary Search
$T(n) = 2T(n-1) + c$	$O(2^n)$	Tower of Hanoi
$T(n) = T(n-1) + c$	$O(n)$	Linear Search
$T(n) = 2T(n/2) + n$	$O(n \log n)$	Merge Sort
$T(n) = T(n-1) + n$	$O(n^2)$	Selection Sort, Insertion Sort
$T(n) = T(n-1) + T(n-2) + c$	$O(2^n)$	Fibonacci Series