

# ADVANCED ANALYSIS OF ALGORITHMS

## CPS 5440

OMAR DIB

# SUBSTITUTION METHOD

# SUBSTITUTION METHOD: CONCEPT

- Substitution method for solving recurrences consists of two steps:
- Guess the form of the solution, e.g.,  $T(n) = O(g(n))$ , then
- Use mathematical induction to find constants ( $c$  and  $n_0$ ) in the form and show that the solution works
  - **Step 1 (Base step)** – Prove that the guess is true for the initial value
  - **Step 2 (Inductive step)** – Prove that if the guess is true for  $T(k) \leq c g(k)$ ,  $\forall k < n$ , then this implies that  $T(n) \leq c g(n)$ , for some  $c > 0$  and  $n \geq n_0$
- The inductive hypothesis is applied to smaller values, similar like recursive calls bring us closer to the base case
- The substitution method is a powerful way to establish lower or upper bounds on a recurrence
- It applies in cases when it is easy to guess the form of the solution

# SUBSTITUTION METHOD: MAKING A GOOD GUESS

- There is no general way to guess the correct solution to recurrences.
- Guessing a solution takes experience and, occasionally, creativity.
- There are some heuristics that can help us make a good guess (*e. g.*, Recursion Tree)
- If a recurrence is similar to a one, we have seen before, then guessing a similar solution is reasonable
- For example,  $T(n) = 2 T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + 17) + n$ , we make the guess that  $T(n) = O(n \lg n)$  like Merge Sort
- Another way to make a good guess is to prove the loose upper and lower bounds on the recurrence and then reduce the range of uncertainty. For example:
  - Start with and prove the initial lower bound of  $T(n) = \Omega(n)$  for the recurrence
  - Start with and prove the initial upper bound of  $T(n) = O(n^2)$  for the recurrence
  - Then gradually lower the upper bound and raise the lower bound until convergence to correct, asymptotically tight solution of  $T(n) = \Theta(n \lg n)$
- Sometimes the correct guess at an asymptotic bound on the solution of a recurrence does not work. This can be solved by revising the guess and subtracting a lower-order term in the guess.

# SUBSTITUTION METHOD: MERGE SORT

- $T(n) = \begin{cases} 1 & n = 1 \\ 2 T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n & n > 1 \end{cases}$
- Guess:  $T(n) = O(n \lg n)$ , or  $T(n) \leq c \cdot n \lg n$ , for some constant  $c$  and  $n_0 \leq n$
- Hypothesis:  $T(k) \leq c \cdot k \lg k, \forall k < n$ , we will use  $k = \frac{n}{2}$
- **Inductive Step:**

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

$$\leq 2 \cdot c \cdot \left(\left\lfloor \frac{n}{2} \right\rfloor\right) \lg \left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n$$

$$\leq c \cdot n \lg \left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n$$

$$= c \cdot n \lg n - c \cdot n \lg 2 + n$$

$$= c \cdot n \lg n - c \cdot n + n$$

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

# SUBSTITUTION METHOD: MERGE SORT

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- Guess:  $T(n) = O(n \lg n)$ , or  $T(n) \leq c \cdot n \lg n$ , for some constant  $c$  and  $n_0 \leq n$
- Hypothesis:  $T(k) \leq c \cdot k \lg k$ ,  $\forall k < n$
- From inductive step:  $T(n) \leq c \cdot n \lg n$  **when  $c \geq 1$**
- **Base step:**  $T(1) \leq c \cdot 1 \lg 1$  ?
  - Impossible as  $T(1) = 1 \not\leq c \cdot 1 \lg 1 = 0$ . (Problem!)
  - But we only want to show that  $T(n) \leq c \cdot n \lg n$  for sufficiently large values of  $n$ ; *i. e.*,  $\forall n \geq n_0$ .
  - Solution: Try  $n_0 > 1$
- **Base steps (check boundaries)**
  - We must check both  $T(2)$  and  $T(3)$  simultaneously because of the nature of the recursive equation
  - Check  $T(2)$  and  $T(3)$

# SUBSTITUTION METHOD : MERGE SORT

- $T(n) = \begin{cases} 1 & n = 1 \\ 2 T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n & n > 1 \end{cases}$
- Guess:  $T(n) = O(n \lg n)$ , or  $T(n) \leq c \cdot n \lg n$ , for some constant  $c$  and  $n_0 \leq n$
- Hypothesis:  $T(k) \leq c \cdot k \lg k, \forall k < n$
- From inductive step:  $T(n) \leq c \cdot n \lg n$  **when  $c \geq 1$**
- **Base step:**  $T(1) \leq c \cdot n \lg n \forall n \geq 1$  ( $n_0 > 1$ )
- Base step boundaries:
  - $T(1) = 1 \Rightarrow \begin{cases} T(2) = 4 \\ T(3) = 5 \end{cases}$
  - We want to satisfy simultaneously
  - $\begin{cases} 4 = T(2) \leq c \cdot 2 \lg 2 \\ 5 = T(3) \leq c \cdot 3 \lg 3 \end{cases} \Rightarrow \begin{cases} c \geq 2 \\ c \geq 1.052 \end{cases} \Rightarrow c \geq 2$
- We prove that  $T(n) \leq c \cdot n \lg n$ , with  $c = 2$ , and  $n_0 = 2$ , So  $T(n) = O(n \lg n)$

# SUBSTITUTION METHOD : MERGE SORT

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

*What happens if we make a wrong guess?*

■ Guess: what if we choose a **wrong guess**, e. g.,  $T(n) = O(n)$ , or  $T(n) \leq c \cdot n$ , for some constant  $c$  and  $n_0 \leq n$

■ Hypothesis:  $T(k) \leq c \cdot k, \forall k < n$ , **we will use  $k = \frac{n}{2}$**

■ Inductive Step:

$$\begin{aligned} T(n) &= 2 T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n \\ &\leq 2 \cdot c \cdot \left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n \\ &\leq c \cdot n + n \\ &= (c + 1) \cdot n \end{aligned}$$

**$\nless c \cdot n$**  *The above inequality does not hold because  $c + 1$  cannot be less than  $c$  (Contradiction).*

**Hence  $T(n) \neq O(n)$**



# SUBSTITUTION METHOD : MERGE SORT

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

Substitution method can also be used to guess a lower bound!

- Let's guess a **lower bound**
- We want to show  $T(n) \geq c \cdot n \lg n$ , for some constant  $c$  and  $n_0 \leq n$ . (Assume that  $n$  is a power of 2)
- Hypothesis:  $T(k) \geq c \cdot k \lg k$ ,  $\forall k < n$ , we will use  $k = \frac{n}{2}$
- Inductive Step: 
$$\begin{aligned} T(n) &= 2 T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n \\ &\geq 2 \cdot c \cdot \left(\left\lfloor \frac{n}{2} \right\rfloor\right) \lg \left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n \\ &\geq c \cdot n \lg \left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n \\ &= c \cdot n \lg n - c \cdot n \lg 2 + n \\ &= c \cdot n \lg n - c \cdot n + n \\ &= c \cdot n \lg n - n(c - 1) \\ &\geq c \cdot n \lg n \end{aligned}$$

True as long as  $(c - 1) \leq 0 \Rightarrow c \leq 1$

# SUBSTITUTION METHOD: MERGE SORT

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

Substitution method can also be used to guess a lower bound!

- Let's guess **lower bound**
- We want to show  $T(n) \geq c \cdot n \lg n$ , for some constant  $c$  and  $n_0 \leq n$ . (Assume that  $n$  is a power of 2)
- Hypothesis:  $T(k) \geq c \cdot k \lg k$ ,  $\forall k < n$ , **we will use  $k = \frac{n}{2}$**
- From inductive step:  $T(n) \geq c \cdot n \lg n$  **when  $c \leq 1$**
- **Base step:**  $T(1) \geq c \cdot 1 \lg 1$ ?
  - True as  $T(1) = 1 \geq c \cdot 1 \lg 1 = 0$ .
- **Check boundaries**
  - We also want to satisfy the boundary condition ( $T(2) = 4$ )
  - $T(2) \geq c \cdot 2 \lg 2 \Rightarrow 4 \geq 2 \cdot c$ ? (True as long as  $c < 2$ . By the requirement of the inductive step  $c \leq 1 \Rightarrow c = 1$ )

# SUBSTITUTION METHOD: MERGE SORT

■  $T(n) = \begin{cases} 1 & n = 1 \\ 2 T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + n & n > 1 \end{cases}$  To prove a bound, we must also prove that  $T(n)$  is strictly increasing!

- We will **prove that  $T(n)$  is strictly increasing**
- Assuming for all  $k \leq n$  it holds  $T(k) > T(k - 1)$ , we want to show that  $T(n + 1) > T(n)$
- For the base case, note that  $T(1) = 1 < 4 = T(2)$
- $$\begin{aligned} T(n + 1) &= 2 T\left(\left\lfloor \frac{n+1}{2} \right\rfloor\right) + n + 1 \\ &= 2 T\left(\frac{n}{2}\right) + n + 1 // \text{Note: } T\left(\left\lfloor \frac{n+1}{2} \right\rfloor\right) = T\left(\frac{n}{2}\right) \text{ (assuming } n \text{ is a power of 2)} \\ &= \left[2 T\left(\frac{n}{2}\right) + n\right] + 1 \\ &= T(n) + 1 \\ &> T(n) \end{aligned}$$
- We prove that  $T(n) \geq c \cdot n \lg n$ , with  $c = 1$ , and  $n_0 = 2$ , So  $T(n) = \Omega(n \lg n)$

# SUBSTITUTION METHOD: QUICK SORT

- $T(n) = \begin{cases} 1 & n = 1 \\ T(n-1) + n & n > 1 \end{cases}$
- Guess:  $T(n) = O(n^2)$ , or  $T(n) \leq c \cdot n^2$ , for some constant  $c$  and  $n_0 \leq n$
- Hypothesis:  $T(k) \leq c \cdot k^2, \forall k < n$ , **we will use  $k = n - 1$**
- Inductive Step:  $T(n) = T(n-1) + n$ 
$$\begin{aligned} &\leq c \cdot (n-1)^2 + n \\ &= c \cdot (n^2 - 2 \cdot n + 1) + n \\ &= c \cdot n^2 - 2 \cdot c \cdot n + c + n \\ &\leq c \cdot n^2 \end{aligned}$$
**True if:  $-2 \cdot c \cdot n + c + n \leq 0 \Rightarrow c \geq 1$**
- Base Step:  $T(1) = 1 \leq c \cdot (1)^2 \Rightarrow T(1) \leq c$
- We prove that  $T(n) \leq c \cdot n^2$ , with  $c = 1$ , and  $n_0 = 1$ , So  $T(n) = O(n^2)$

# SUBSTITUTION METHOD: LOOSE BOUND

- $T(n) = \begin{cases} 1 & n = 1 \\ 2 T\left(\frac{n}{2}\right) + 1 & n > 1 \end{cases}$
- Guess:  $T(n) = O(n)$ , or  $T(n) \leq c \cdot n$ , for some constant  $c$  and  $n_0 \leq n$
- Hypothesis:  $T(k) \leq c \cdot k, \forall k < n$ , we will use  $k = \frac{n}{2}$
- Inductive Step: 
$$\begin{aligned} T(n) &= 2 T\left(\frac{n}{2}\right) + 1 \\ &\leq 2 \cdot c \cdot \left(\frac{n}{2}\right) + 1 \\ &= c \cdot n + 1 \\ &\not\leq c \cdot n \end{aligned}$$
- Which does not imply that  $T(n) \leq c \cdot n$ , for any  $c$ . We need to show the exact form.
- To overcome this hurdle:
  - Revise our guess: say  $T(n) = O(n^2)$ . However, our original guess was correct!
  - Sometimes it is easier to prove something stronger!

# SUBSTITUTION METHOD: LOOSE BOUND

- $T(n) = \begin{cases} 1 & n = 1 \\ 2 T\left(\frac{n}{2}\right) + 1 & n > 1 \end{cases}$
- Solution: Try tighter bound
- Guess:  $T(n) = O(n)$ . Let's try  $T(n) \leq c \cdot n - b$ , where  $b$  is another constant
- Hypothesis:  $T(k) \leq c \cdot k - b, \forall k < n$ , we will use  $k = \frac{n}{2}$
- Inductive Step:  $T(n) = 2 T\left(\frac{n}{2}\right) + 1$ 
$$\leq 2 \cdot \left[ c \cdot \left(\frac{n}{2}\right) - b \right] + 1$$
$$= c \cdot n - 2 \cdot b + 1$$
$$\leq c \cdot n - b \quad \text{True if: } -b + 1 \leq 0 \Rightarrow b \geq 1$$
- Base Step:  $T(1) = 1 \leq c \cdot 1 - b \Rightarrow 1 \leq c - b \Rightarrow b \leq c$

We prove that  $T(n) \leq c \cdot n - b$ , with  $c = 2, b = 1$  and  $n_0 = 1$ , So  $T(n) = O(n)$

# SUBSTITUTION METHOD: CHANGING VARIABLES

- Consider the recurrence  $T(n) = 2 T(\lfloor \sqrt{n} \rfloor) + \lg n$
- We can simplify the recurrence with a change of variables
- Rename  $m = \lg n$ . We have:

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

- Define  $S(m) = T(2^m)$ . We get:

$$S(m) = 2 S\left(\frac{m}{2}\right) + m$$

- Hence, the solution is  $O(m \lg m)$ , or with substitution  $O(\lg n \cdot \lg(\lg n))$

# SUBSTITUTION METHOD: PRACTICE QUESTIONS

- Apply the substitution method to the given recurrences and show that the given guesses are the solution to these recurrences. Assume  $T(1) = 1$  as the base case.
- $T(n) = T(n - 1) + T(n - 2) + 1$  (Guess:  $O(2^n)$ )
- $T(n) = 2.T(n - 1) + 1$  (Guess:  $O(2^n)$ )
- $T(n) = 3.T\left(\frac{n}{4}\right) + n^2$  (Guess:  $O(n^2)$ )
- $T(n) = 3.T\left(\frac{n}{4}\right) + n$  (Guess:  $O(n)$ )



# RECURSION TREE METHOD

# RECURSION TREE METHOD: CONCEPT

- Making a good guess is sometimes difficult with the substitution method
- Recursion tree method can be used to devise a good guess
- Recursion trees show successive expansions of recurrences using trees
- RT model the costs (time) of a recursive execution of an algorithm that is composed of two parts:
  - Cost of non-recursive part
  - Cost of recursive call on smaller input size
- A tree node represents the cost of a sub-problem (recursive function invocation)
- To determine the total cost of the recursion tree, evaluate:
  - Cost of individual node at depth " $i$ "
  - Sum up the cost of all nodes at depth " $i$ "
  - Sum up all per-level costs of the recursion tree

# RECURSION TREE METHOD: EXAMPLE I

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

- Solve the following recurrence using the Recurrence Tree Method

- Assumption: We assume that  $n$  is an exact power of 2

- Some Useful Properties:

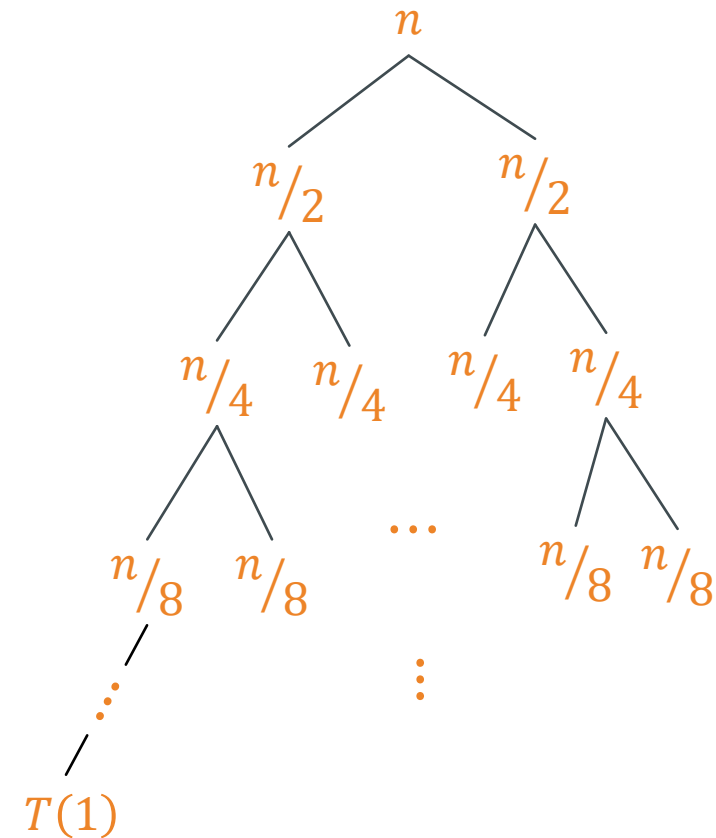
- $x^{\log_y n} \Rightarrow n^{\log_y x}$

- $x^0 + x^1 + x^2 + \dots + x^n = \frac{x^{n+1}-1}{x-1}$  *for  $x \neq 1$*

- $x^0 + x^1 + x^2 + \dots = \frac{1}{1-x}$  *for  $|x| < 1$*

# RECURSION TREE METHOD: EXAMPLE I

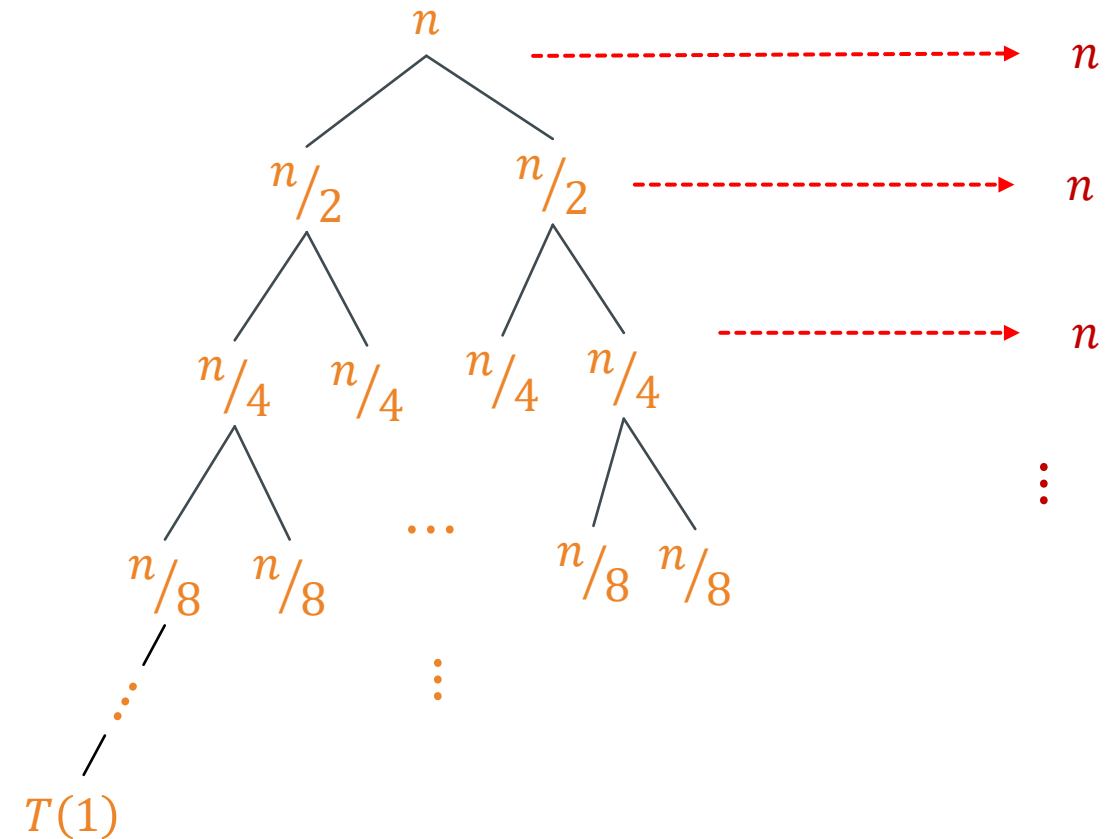
- $T(n) = 2.T\left(\frac{n}{2}\right) + n$
- $T\left(\frac{n}{2}\right) = 2.T\left(\frac{n}{2^2}\right) + \frac{n}{2}$
- $T\left(\frac{n}{2^2}\right) = 2.T\left(\frac{n}{2^3}\right) + \frac{n}{2^2}$
- $T\left(\frac{n}{2^{k-1}}\right) = 2.T\left(\frac{n}{2^k}\right) + \frac{n}{2^{k-1}}$
- $T\left(\frac{n}{2^k}\right) = T(1)$



- Total cost = [Cost of Leaf Nodes] + [Cost of Internal Nodes]
- Total cost = [cost of leaf node  $\times$  total leaf nodes] + [sum of costs at each level of internal nodes]
- Total cost =  $L_c + I_c$

# RECURSION TREE METHOD: EXAMPLE I

- $T\left(\frac{n}{2^k}\right) = T(1) \Rightarrow n = 2^k \Rightarrow k = \lg n$
- $L_c = 2^k \Rightarrow 2^{\lg n} \Rightarrow n^{\lg 2} \Rightarrow n$
- $I_c = k \cdot n = n \lg n$
- Total cost =  $L_c + I_c \Rightarrow n + n \lg n$
- Hence,  $T(n) \in O(n \lg n)$



## RECURSION TREE METHOD: EXAMPLE 2

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

- Solve the following recurrence using the Recursion Tree Method

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

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

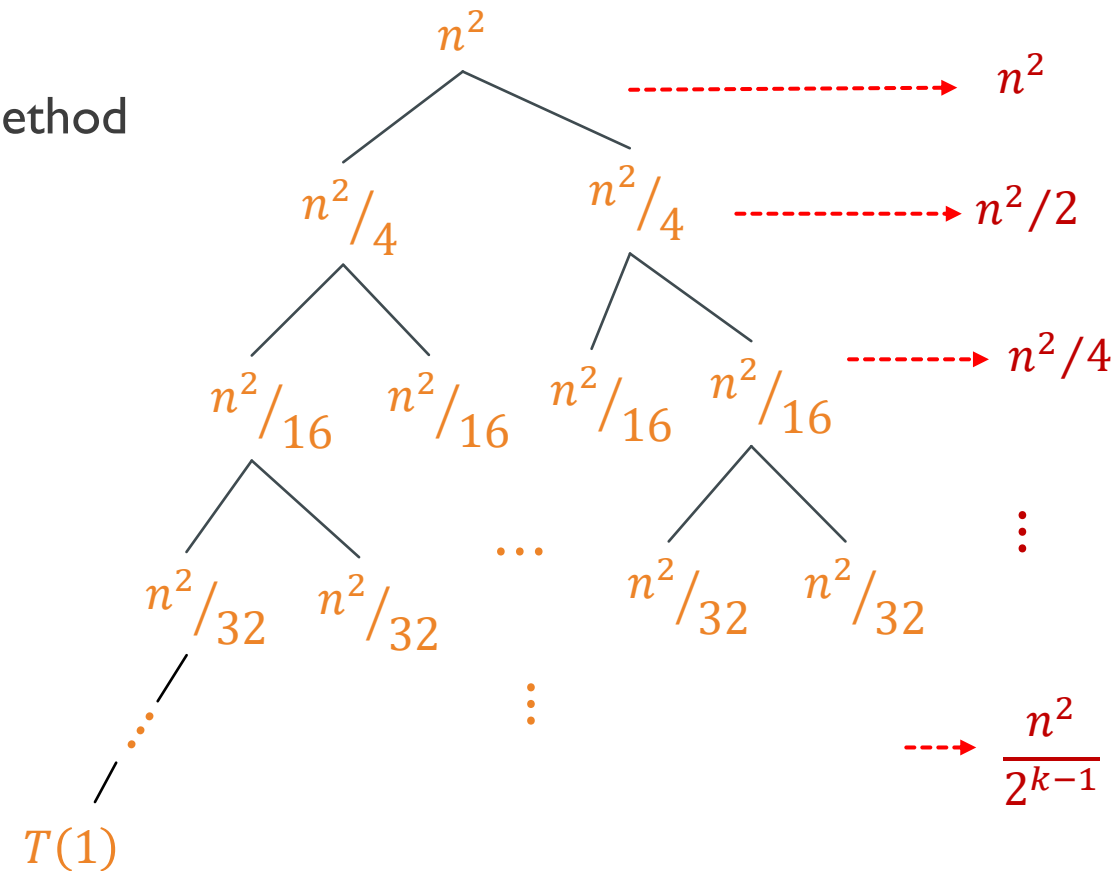
$$T\left(\frac{n}{2^2}\right) = 2 \cdot T\left(\frac{n}{2^3}\right) + \frac{n^2}{4^2}$$

$$T\left(\frac{n}{2^k}\right) = T(1) \Rightarrow n = 2^k \Rightarrow k = \lg n$$

$$L_c = 2^k \Rightarrow 2^{\lg n} \Rightarrow n^{\lg 2} \Rightarrow n$$

$$I_c = n^2 \cdot \left[ \left(\frac{1}{2}\right)^0 + \left(\frac{1}{2}\right)^1 + \left(\frac{1}{2}\right)^2 + \dots + \left(\frac{1}{2}\right)^{k-1} \right] = n^2 \cdot \left[ \frac{1}{1 - \frac{1}{2}} \right] \Rightarrow 2 \cdot n^2$$

$$\text{Total cost} = L_c + I_c = n + 2 \cdot n^2, \text{ Hence } T(n) \in O(n^2)$$



# RECURSION TREE METHOD: EXAMPLE 3

$$T(n) = \begin{cases} 1 & n = 1 \\ 3T\left(\frac{n}{4}\right) + n^2 & n > 1 \end{cases}$$

Solve the following recurrence using the Recursion Tree Method

Assumption:  $n$  is an exact power of 4

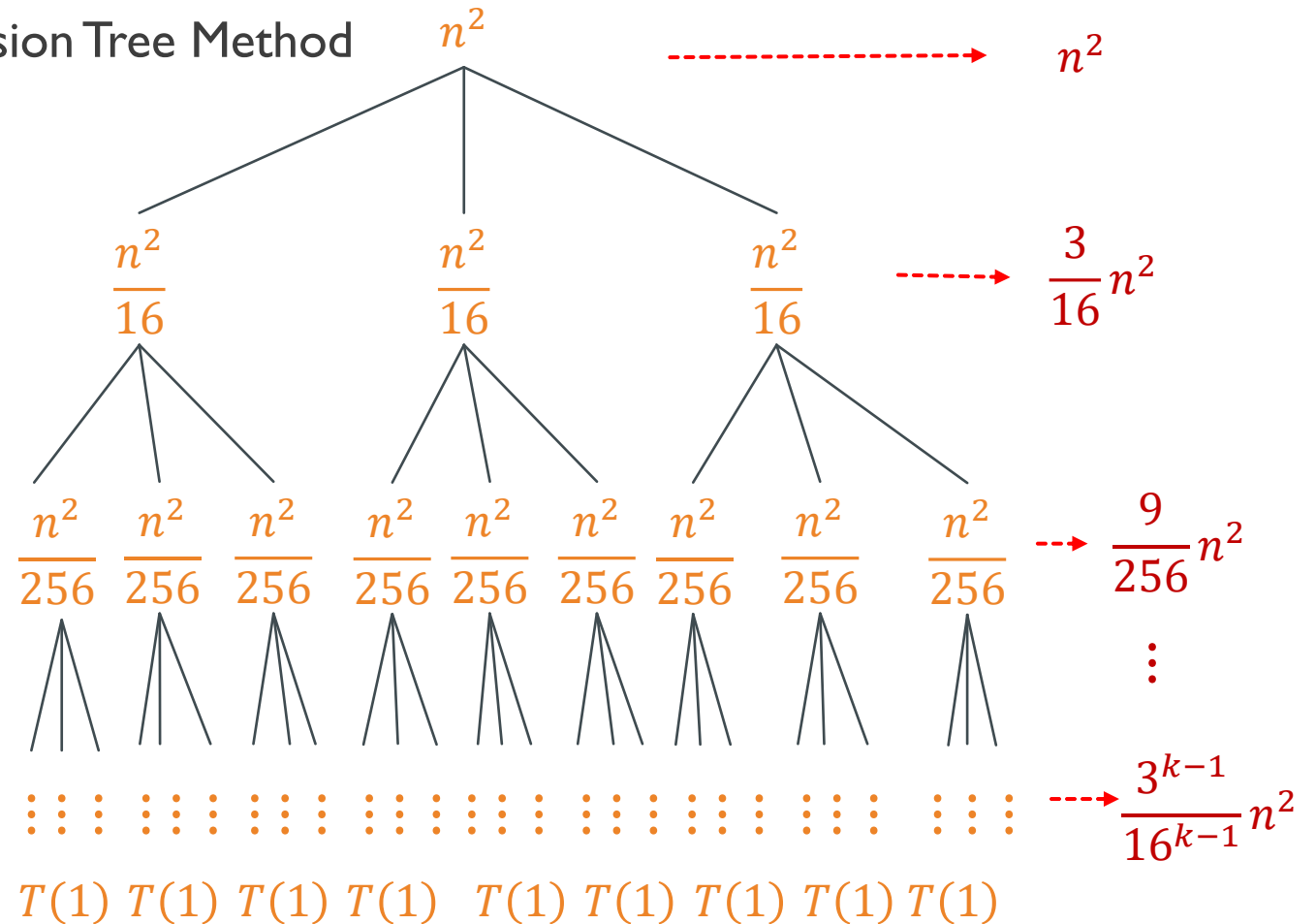
$$T(n) = 3 \cdot T\left(\frac{n}{4}\right) + n^2$$

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

$$T\left(\frac{n}{4^2}\right) = 3 \cdot T\left(\frac{n}{4^3}\right) + \frac{n^2}{16^2}$$

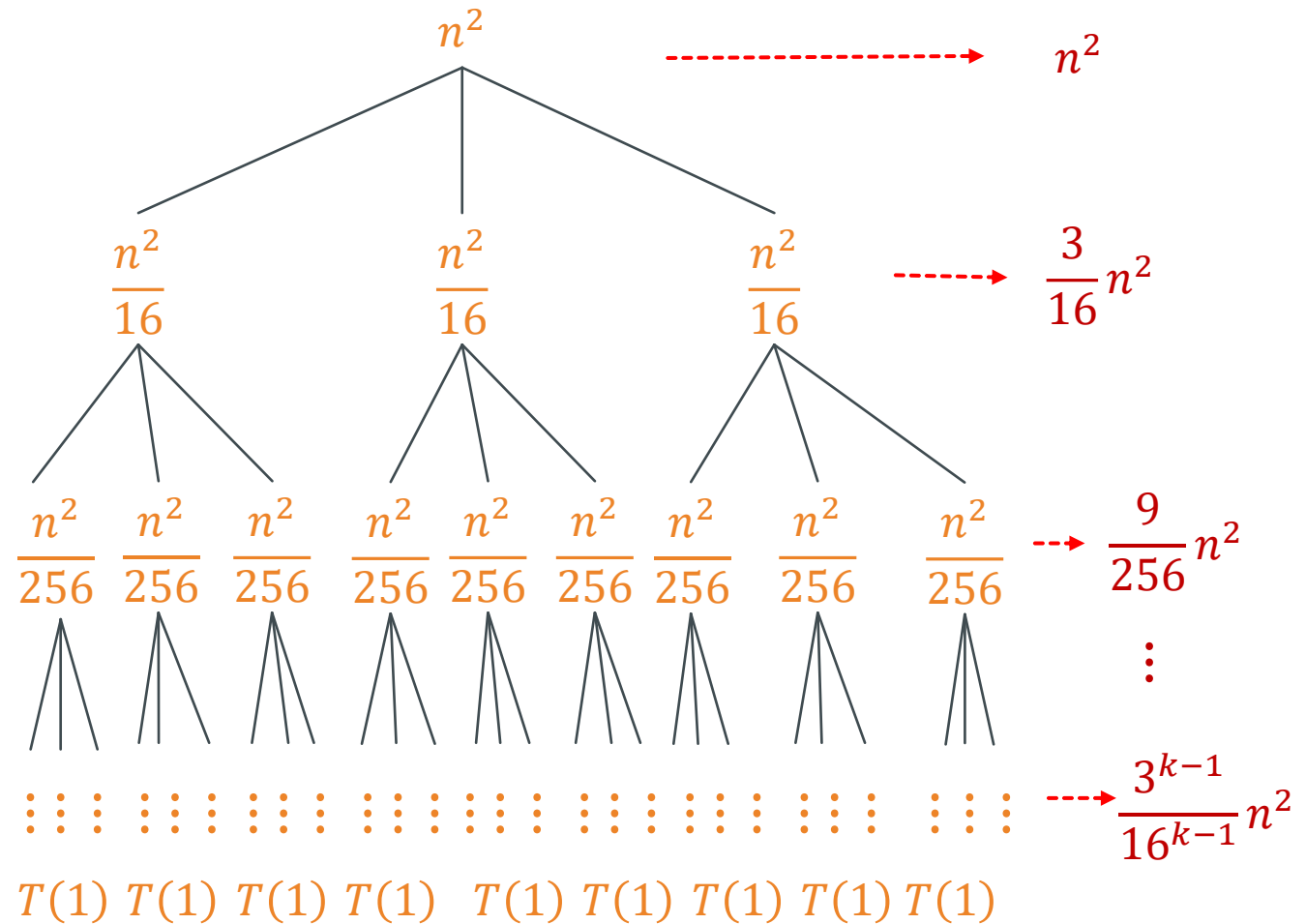
$$T\left(\frac{n}{4^k}\right) = T(1) \Rightarrow n = 4^k \Rightarrow k = \log_4 n$$

$$L_c = 3^k \Rightarrow 3^{\log_4 n} \Rightarrow n^{\log_4 3} \Rightarrow n$$



# RECURSION TREE METHOD: EXAMPLE 3

- $T(n) = 3 \cdot T\left(\frac{n}{4}\right) + n^2$
- $T\left(\frac{n}{4}\right) = 3 \cdot T\left(\frac{n}{4^2}\right) + \frac{n^2}{4^2}$
- $T\left(\frac{n}{4^2}\right) = 3 \cdot T\left(\frac{n}{4^3}\right) + \frac{n^2}{16^2}$
- $T\left(\frac{n}{4^k}\right) = T(1) \Rightarrow n = 4^k \Rightarrow k = \log_4 n$
- $L_c = 3^k \Rightarrow 3^{\log_4 n} \Rightarrow n^{\log_4 3}$
- $I_c =$   
 $= n^2 \cdot \left[ \left(\frac{3}{16}\right)^0 + \left(\frac{3}{16}\right)^1 + \left(\frac{3}{16}\right)^2 + \dots + \left(\frac{3}{16}\right)^{k-1} \right]$   
 $= n^2 \cdot \left[ \frac{1}{1 - \frac{3}{16}} \right] \Rightarrow \frac{16}{13} \cdot n^2$
- Total cost =  $L_c + I_c = n^{\log_4 3} + \frac{16}{13} \cdot n^2$ ,  
Hence  $T(n) \in O(n^2)$





# RECURSION TREE METHOD: EXAMPLE 4

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

Solve the following recurrence using the Recursion Tree Method

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

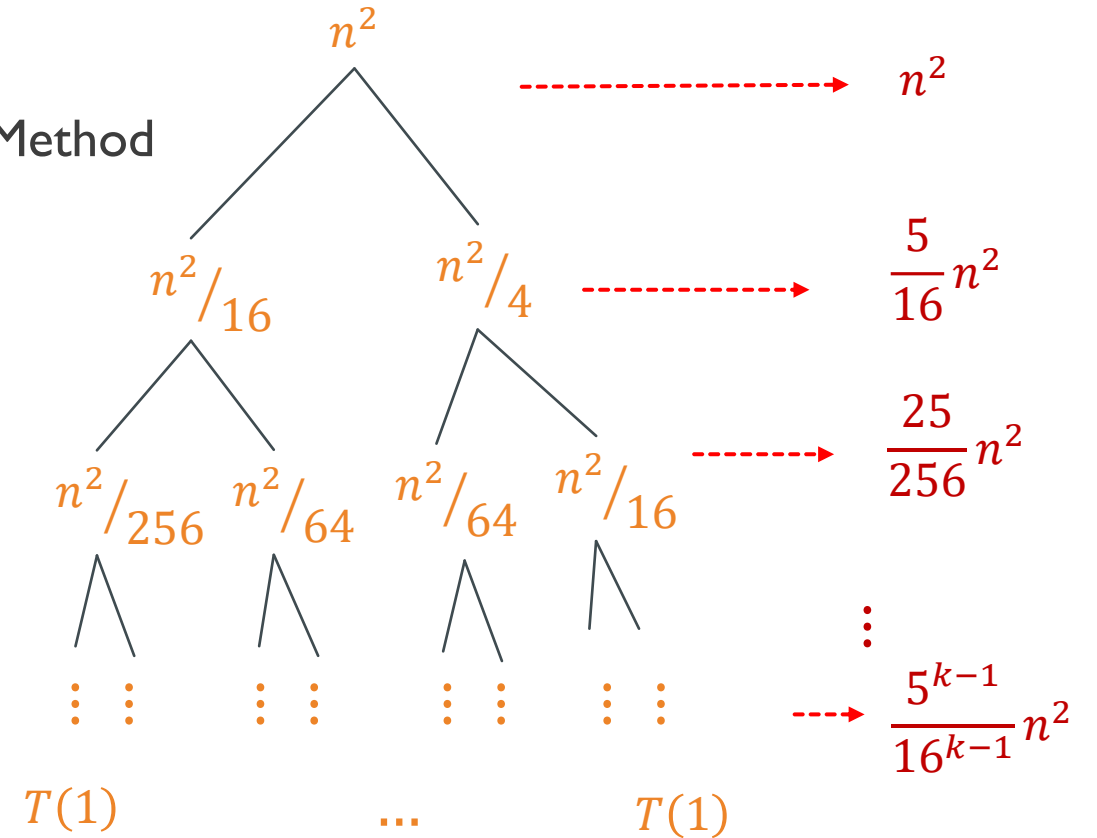
$$\left\{ \begin{array}{l} T\left(\frac{n}{4}\right) = T\left(\frac{n}{16}\right) + T\left(\frac{n}{8}\right) + \frac{n^2}{16} \\ T\left(\frac{n}{2}\right) = T\left(\frac{n}{8}\right) + T\left(\frac{n}{4}\right) + \frac{n^2}{4} \end{array} \right.$$

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

$$\left\{ \begin{array}{l} T\left(\frac{n}{16}\right) = T\left(\frac{n}{64}\right) + T\left(\frac{n}{32}\right) + \frac{n^2}{256} \\ T\left(\frac{n}{8}\right) = T\left(\frac{n}{32}\right) + T\left(\frac{n}{16}\right) + \frac{n^2}{64} \end{array} \right.$$

$$T\left(\frac{n}{8}\right) = T\left(\frac{n}{32}\right) + T\left(\frac{n}{16}\right) + \frac{n^2}{64}$$

$$\left\{ \begin{array}{l} T\left(\frac{n}{4}\right) = T\left(\frac{n}{16}\right) + T\left(\frac{n}{8}\right) + \frac{n^2}{16} \end{array} \right.$$



# RECURSION TREE METHOD: EXAMPLE 4

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

$$T\left(\frac{n}{2^k}\right) = T(1) \Rightarrow n = 2^k \Rightarrow k = \lg n$$

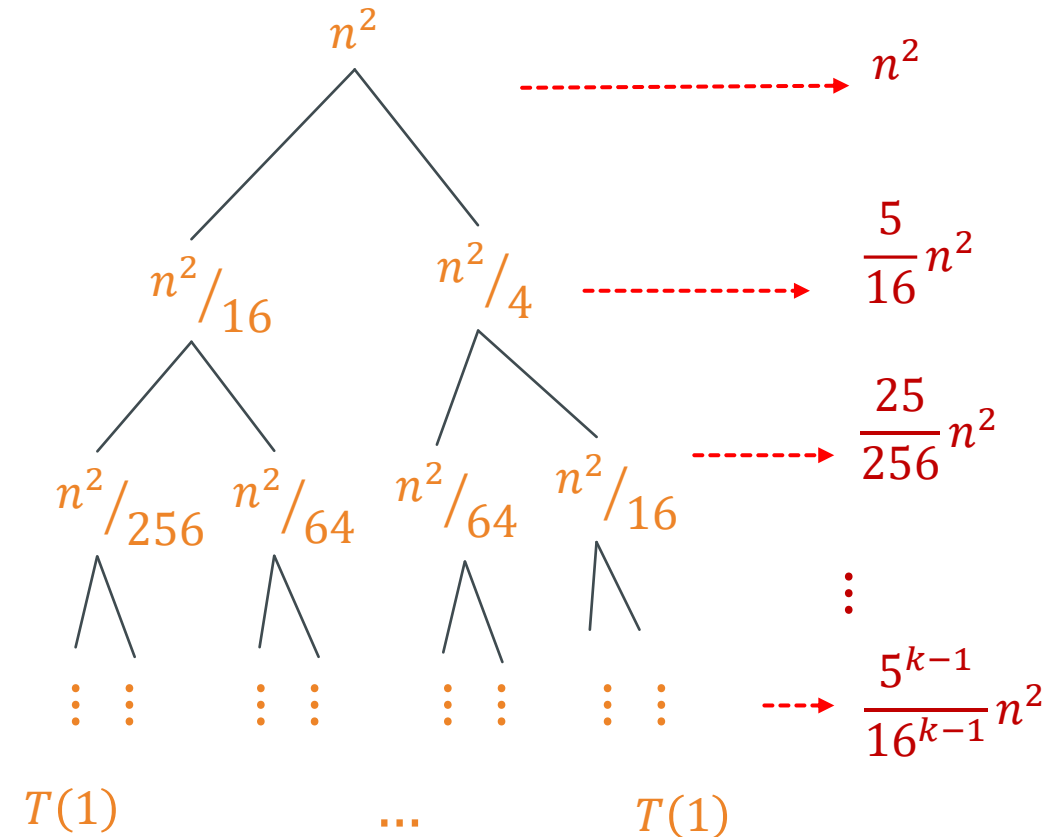
$$L_c = 2^k \Rightarrow 2^{\lg n} \Rightarrow n^{\lg 2} \Rightarrow n$$

$$I_c =$$

$$= n^2 \cdot \left[ \left(\frac{5}{16}\right)^0 + \left(\frac{5}{16}\right)^1 + \left(\frac{5}{16}\right)^2 + \dots + \left(\frac{5}{16}\right)^{k-1} \right]$$

$$= n^2 \cdot \left[ \frac{1}{1 - \frac{5}{16}} \right] \Rightarrow \frac{16}{11} \cdot n^2$$

$$\text{Total cost} = L_c + I_c = n + \frac{16}{11} \cdot n^2, \text{ Hence } T(n) \in O(n^2)$$



# RECURSION TREE METHOD: EXAMPLE 5

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

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

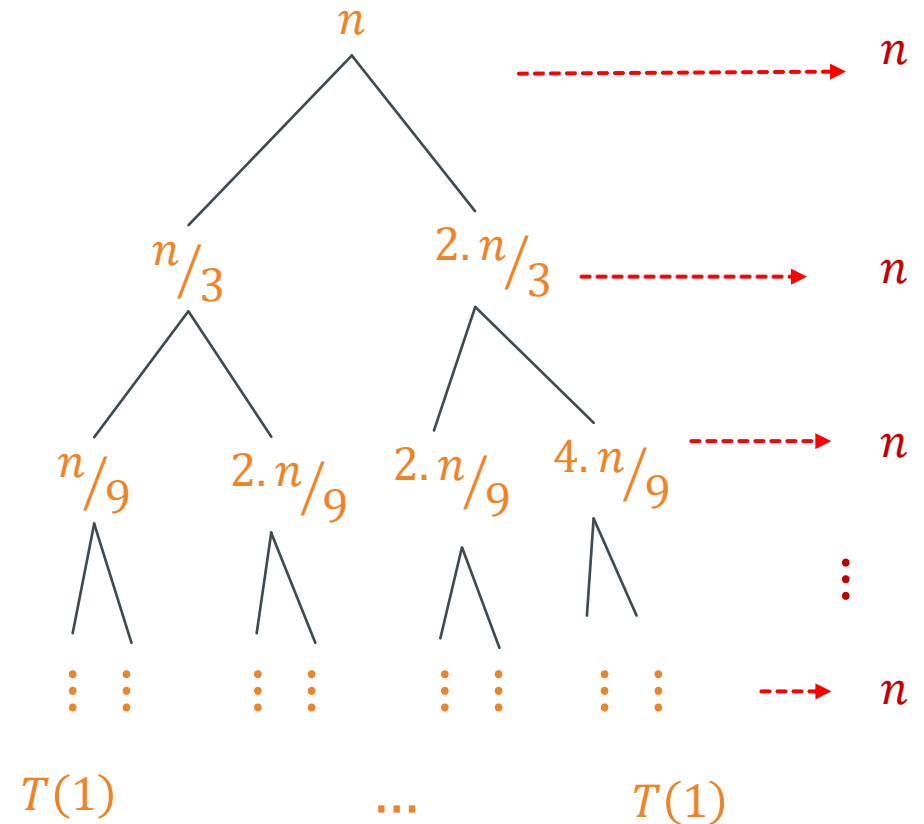
$$\blacksquare T\left(\frac{n}{3}\right) = T\left(\frac{n}{9}\right) + T\left(\frac{2n}{9}\right) + \frac{n}{3}$$

$$\blacksquare T\left(\frac{2n}{3}\right) = T\left(\frac{2n}{9}\right) + T\left(\frac{4n}{9}\right) + \frac{2n}{3}$$

$$\blacksquare T\left(\frac{n}{9}\right) = T\left(\frac{n}{27}\right) + T\left(\frac{2n}{27}\right) + \frac{n}{9}$$

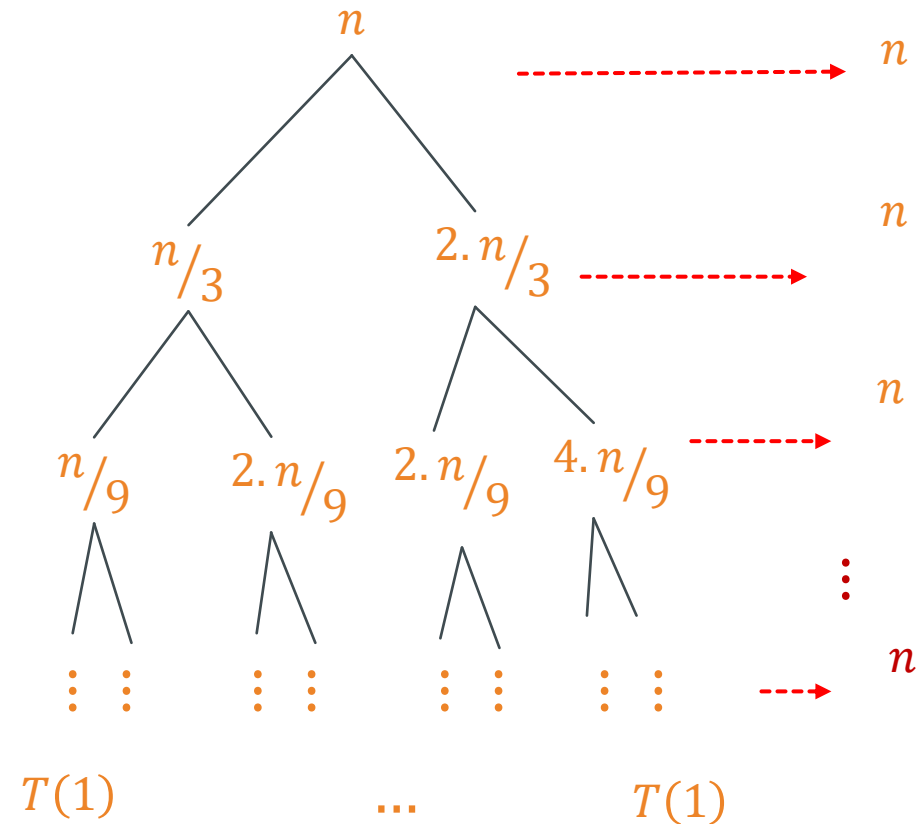
$$\blacksquare T\left(\frac{2n}{9}\right) = T\left(\frac{2n}{27}\right) + T\left(\frac{4n}{27}\right) + \frac{2n}{9}$$

$$\blacksquare T\left(\frac{4n}{9}\right) = T\left(\frac{4n}{27}\right) + T\left(\frac{8n}{27}\right) + \frac{4n}{9}$$



# RECURSION TREE METHOD: EXAMPLE 5

- $$T(n) = \begin{cases} 1 & n = 1 \\ T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + n & n > 1 \end{cases}$$
- $$T\left(\frac{2^k}{3^k}n\right) = T(1) \implies n = \frac{3^k}{2^k} \implies k = \log_{3/2} n$$
- $$L_c = 2^k \implies 2^{\log_{3/2} n} \implies n^{\log_{3/2} 2}$$
- $$I_c = n \cdot k = n \cdot \log_{3/2} n$$
- $$\text{Total cost} = L_c + I_c = n^{\log_{3/2} 2} + n \cdot \log_{3/2} n,$$
- $T(n) \in O(n \cdot \log n)$ ? Difficult to decide which side is bigger  $\implies O$  is ambiguous
- Solution:** Use both terms in the total cost as guesses in the substitution method (Try it out: You will find  $n \cdot \log n$  is a correct guess ...)



## RECURSION TREE METHOD: CAUTION NOTE

- Recursion trees are best used to generate good guesses
  - Verify guesses using the substitution method
- A small amount of “sloppiness” can be tolerated
  - Using an infinite decreasing geometric series as an upper bound
  - Assuming “ $n$ ” to be an exact power of 2, 3, or 4
  - Assuming the tree is complete. In reality, the tree may have fewer internal and leaf nodes
- By carefully drawing out a recursion tree and summing the costs, the recursion tree method can be used as a direct proof of a solution to any recurrence

# RECURSION TREE METHOD: PRACTICE QUESTIONS

- Solve the following recurrences using the recurrence tree method.

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

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

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

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

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

# MASTER THEOREM

# MASTER THEOREM

- The Master Method depends on the following theorem

- **Theorem:**

Let  $a \geq 1, b > 1$  be constants, let  $f(n)$  be a function, and let  $T(n)$  be defined on the non-negative integers by the recurrence:  $T(n) = a T\left(\frac{n}{b}\right) + f(n)$

Then  $T(n)$  can be bounded asymptotically as follows:

1. If  $f(n) = O(n^{\log_b a - \varepsilon})$  for some constant  $\varepsilon > 0$ , 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  $f(n) = \Omega(n^{\log_b a + \varepsilon})$  for some constant  $\varepsilon > 0$ ,  
**and**  $f(n)$  satisfies the regularity condition:  $a f\left(\frac{n}{b}\right) \leq c f(n)$  for some constant  $c < 1$ ,  
and all sufficiently large  $n$   
then  $T(n) = \Theta(f(n))$



# MASTER THEOREM

- Important to note that the three cases do not cover all the possibilities
  - Gap between cases 1 and 2  
when  $f(n)$  is smaller than  $n^{\log_b a}$  but not polynomially smaller
  - Gap between cases 2 and 3  
when  $f(n)$  is larger than  $n^{\log_b a}$  but not polynomially larger
- If  $f(n)$  falls into one of these gaps, or if the regularity condition in case 3 fails to hold, the master method cannot be used to solve the recurrence

# MASTER THEOREM: EXAMPLE I

- $T(n) = 2 T\left(\frac{n}{2}\right) + n$  // Merge Sort
- $a = 2$
- $b = 2$
- $f(n) = n$
- $n^{\log_b a} \Rightarrow n^{\log_2 2} \Rightarrow n$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) = n^{\log_b a}$  so case 2 is applied. [ $f(n) = \Theta(n^{\log_b a})$ ]
  - $\Rightarrow T(n) = \Theta(n^{\log_b a} \lg n)$   
 $= \Theta(n^{\log_2 2} \lg n)$   
 $= \Theta(n \lg n)$
- Hence:  $T(n) = \Theta(n \lg n)$

## MASTER THEOREM: EXAMPLE 2

- $T(n) = 2 T\left(\frac{n}{2}\right) + n^2$  //Quick Sort
- $a = 2$
- $b = 2$
- $f(n) = n^2$
- $n^{\log_b a} \Rightarrow n^{\log_2 2} \Rightarrow n$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) > n^{\log_b a}$  so case 3 is applied.  $[f(n) = \Omega(n^{\log_b a + \varepsilon})]$
  - $\Rightarrow T(n) = \Theta(f(n))$   
 $= \Theta(n^2)$
- Hence:  $T(n) = \Theta(n^2)$

*Verify Regularity Condition:*

✓ a.  $f\left(\frac{n}{b}\right) \leq c f(n)$

✓ 2.  $f\left(\frac{n}{2}\right) \leq c \cdot n^2$

✓ 2.  $\frac{n^2}{4} \leq c \cdot n^2$

✓  $\frac{1}{2} \leq c$

## MASTER THEOREM: EXAMPLE 3

- $T(n) = 9 T\left(\frac{n}{3}\right) + n$
- $a = 9$
- $b = 3$
- $f(n) = n$
- $n^{\log_b a} \Rightarrow n^{\log_3 9} \Rightarrow n^2$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) < n^{\log_b a}$  so case 1 is applied. [ $f(n) = O(n^{\log_b a - \varepsilon})$ ]
  - $\Rightarrow T(n) = \Theta(n^{\log_b a})$   
 $= \Theta(n^{\log_3 9})$   
 $= \Theta(n^2)$
- Hence:  $T(n) = \Theta(n^2)$

## MASTER THEOREM: EXAMPLE 4

- $T(n) = T\left(\frac{n}{2}\right) + 1$  // Binary Search
- $a = 1$
- $b = 2$
- $f(n) = n^0$
- $n^{\log_b a} \Rightarrow n^{\log_2 1} \Rightarrow n^0$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) = n^{\log_b a}$  so case 2 is applied. [ $f(n) = \Theta(n^{\log_b a})$ ]
  - $\Rightarrow T(n) = \Theta(n^{\log_b a} \lg n)$   
 $= \Theta(n^{\log_2 1} \lg n)$   
 $= \Theta(n^0 \lg n)$
- Hence:  $T(n) = \Theta(\lg n)$

## MASTER THEOREM: EXAMPLE 5

- $T(n) = 4 T\left(\frac{n}{2}\right) + n^3$
- $a = 4$
- $b = 2$
- $f(n) = n^3$
- $n^{\log_b a} \Rightarrow n^{\log_2 4} \Rightarrow n^2$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) > n^{\log_b a}$  so case 3 is applied. [ $f(n) = \Omega(n^{\log_b a + \varepsilon})$ ]
  - $\Rightarrow T(n) = \Theta(f(n))$   
 $= \Theta(n^3)$
- Hence:  $T(n) = \Theta(n^3)$

*Verify Regularity Condition:*

✓ a.  $f\left(\frac{n}{b}\right) \leq c f(n)$

✓ 4.  $f\left(\frac{n}{2}\right) \leq c \cdot n^3$

✓ 4.  $\frac{n^3}{8} \leq c \cdot n^3$

✓  $\frac{1}{2} \leq c$

## MASTER THEOREM: EXAMPLE 6

- $T(n) = T\left(\frac{n}{2}\right) + n^2$
- $a = 1$
- $b = 2$
- $f(n) = n^2$
- $n^{\log_b a} \Rightarrow n^{\log_2 1} \Rightarrow n^0$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) > n^{\log_b a}$  so case 3 is applied. [ $f(n) = \Omega(n^{\log_b a + \varepsilon})$ ]
  - $\Rightarrow T(n) = \Theta(f(n))$   
 $= \Theta(n^2)$
- Hence:  $T(n) = \Theta(n^2)$

*Verify Regularity Condition:*

✓ a.  $f\left(\frac{n}{b}\right) \leq c f(n)$

✓ 1.  $f\left(\frac{n}{2}\right) \leq c \cdot n^2$

✓  $\frac{n^2}{4} \leq c \cdot n^2$

✓  $\frac{1}{4} \leq c$

## MASTER THEOREM: EXAMPLE 7

- $T(n) = 4 \cdot T\left(\frac{n}{2}\right) + n^2$
- $a = 4$
- $b = 2$
- $f(n) = n^2$
- $n^{\log_b a} \Rightarrow n^{\log_2 4} \Rightarrow n^2$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) = n^{\log_b a}$  so case 2 is applied. [ $f(n) = \Theta(n^{\log_b a})$ ]
  - $\Rightarrow T(n) = \Theta(n^{\log_b a} \lg n)$   
 $= \Theta(n^{\log_2 4} \lg n)$   
 $= \Theta(n^2 \lg n)$
- Hence:  $T(n) = \Theta(n^2 \lg n)$



## MASTER THEOREM: EXAMPLE 8

- $T(n) = 7.T\left(\frac{n}{3}\right) + n^2$
- $a = 7$
- $b = 3$
- $f(n) = n^2$
- $n^{\log_b a} \Rightarrow n^{\log_3 7} \Rightarrow n^{1.77}$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) > n^{\log_b a}$  so case 3 is applied. [ $f(n) = \Omega(n^{\log_b a + \varepsilon})$ ]
  - $\Rightarrow T(n) = \Theta(f(n))$   
 $= \Theta(n^2)$
- Hence:  $T(n) = \Theta(n^2)$

*Verify Regularity Condition:*

✓ a.  $f\left(\frac{n}{b}\right) \leq c f(n)$

✓  $7.f\left(\frac{n}{3}\right) \leq c \cdot n^2$

✓  $7 \cdot \frac{n^2}{9} \leq c \cdot n^2$

✓  $\frac{7}{9} \leq c$

## MASTER THEOREM: EXAMPLE 9

- $T(n) = 7 T\left(\frac{n}{2}\right) + n^2$
- $a = 7$
- $b = 2$
- $f(n) = n^2$
- $n^{\log_b a} \Rightarrow n^{\log_2 7} \Rightarrow n^{2.81}$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) < n^{\log_b a}$  so case 1 is applied.  $[f(n) = O(n^{\log_b a - \varepsilon})]$
  - $\Rightarrow T(n) = \Theta(n^{\log_b a})$   
 $= \Theta(n^{\log_2 7})$   
 $= \Theta(n^{2.81})$
- Hence:  $T(n) = \Theta(n^{2.81})$

## MASTER THEOREM: EXAMPLE 10

- $T(n) = 2 T\left(\frac{n}{2}\right) + \sqrt{n}$
- $a = 2$
- $b = 2$
- $f(n) = n^{(1/2)}$
- $n^{\log_b a} \Rightarrow n^{\log_2 2} \Rightarrow n^1$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) < n^{\log_b a}$  so case 1 is applied.  $[f(n) = O(n^{\log_b a - \varepsilon})]$
  - $\Rightarrow T(n) = \Theta(n^{\log_b a})$   
 $= \Theta(n^{\log_2 2})$   
 $= \Theta(n)$
- Hence:  $T(n) = \Theta(n)$

# MASTER THEOREM: EXAMPLE 11

- $T(n) = 3.T\left(\frac{n}{4}\right) + n \log n$
- $a = 3$
- $b = 4$
- $f(n) = n \log n$
- $n^{\log_b a} \Rightarrow n^{\log_4 3} \Rightarrow n^{0.79}$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow f(n) > n^{\log_b a}$  so case 3 is applied.  $[f(n) = \Omega(n^{\log_b a + \varepsilon})]$
  - $\Rightarrow T(n) = \Theta(f(n))$   
 $= \Theta(n \log n)$
- Hence:  $T(n) = \Theta(n \log n)$

*Verify Regularity Condition:*

- ✓  $a.f\left(\frac{n}{b}\right) \leq c.f(n)$
- ✓  $3.f\left(\frac{n}{4}\right) \leq c.n \lg n$
- ✓  $3.\frac{n}{4} \lg \frac{n}{4} \leq c.n \lg n$
- ✓  $\frac{3}{4} [(\lg n) - 2] \leq c.n \lg n$
- ✓  $\frac{3}{4} \leq c$

## MASTER THEOREM: EXAMPLE 12

- $T(n) = 4 \cdot T\left(\frac{n}{2}\right) + \frac{n^2}{\lg n}$
- $a = 4$
- $b = 2$
- $f(n) = \frac{n^2}{\lg n}$
- $n^{\log_b a} \Rightarrow n^{\log_2 4} \Rightarrow n^2$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - $\Rightarrow$  Non – polynomial difference between  $f(n)$  and  $n^{\log_b a}$
  - Master Method does not apply
    - The difference must be polynomially larger by a factor of  $n^\varepsilon$  where  $\varepsilon > 0$ .
    - In this case the difference is only larger by a factor of  $\frac{1}{\lg n}$

## MASTER THEOREM: EXAMPLE 13

- $T(n) = 2.T\left(\frac{n}{2}\right) + n \log n$
- $a = 2$
- $b = 2$
- $f(n) = n \log n$
- $n^{\log_b a} \Rightarrow n^{\log_2 2} \Rightarrow n^1$
- Compare  $f(n)$  and  $n^{\log_b a}$ :
  - *Seems like case 3 should apply*
  - **Master Method does not apply.** Non polynomial difference between  $f(n)$  and  $n^{\log_b a}$ 
    - The difference must be polynomially larger by a factor of  $n^\varepsilon$  where  $\varepsilon > 0$ .
    - In this case the difference is only larger by a factor of  $\log n$

# MASTER THEOREM : PRACTICE QUESTIONS

- Solve the following recurrences using the Master Method.

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

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

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

- $T(n) = 16.T\left(\frac{n}{4}\right) + n^2$

- $T(n) = 2.T\left(\frac{n}{4}\right) + n^{0.51}$

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

- $T(n) = T\left(\frac{2n}{5}\right) + n$

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

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

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

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

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

- $T(n) = 3.T\left(\frac{n}{3}\right) + 1$

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

- $T(n) = 6.T\left(\frac{n}{3}\right) + n^2 \lg n$

- $T(n) = 16.T\left(\frac{n}{4}\right) + n$

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

- $T(n) = 64.T\left(\frac{n}{8}\right) + n^2 \lg n$

