

# *Introduction to Algorithms*

**6.046J/18.401J/SMA5503**

## *Lecture 2*

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# Solving recurrences

- The analysis of merge sort from *Lecture 1* required us to solve a recurrence.
- Recurrences are like solving integrals, differential equations, etc.
  - Learn a few tricks.
- *Lecture 3*: Applications of recurrences.

# Substitution method

*The most general method:*

- 1. *Guess*** the form of the solution.
- 2. *Verify*** by induction.
- 3. *Solve*** for constants.

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

- [Assume that  $T(1) = \Theta(1)$ .]
- Guess  $O(n^3)$  . (Prove  $O$  and  $\Omega$  separately.)
- Assume that  $T(k) \leq ck^3$  for  $k < n$  .
- Prove  $T(n) \leq cn^3$  by induction.

# Example of substitution

$$\begin{aligned}T(n) &= 4T(n/2) + n \\ &\leq 4c(n/2)^3 + n \\ &= (c/2)n^3 + n \\ &= cn^3 - ((c/2)n^3 - n) \leftarrow \textit{desired} - \textit{residual} \\ &\leq cn^3 \leftarrow \textit{desired}\end{aligned}$$

whenever  $(c/2)n^3 - n \geq 0$ , for example,  
if  $c \geq 2$  and  $n \geq 1$ .

*residual*

# Example (continued)

- We must also handle the initial conditions, that is, ground the induction with base cases.
- **Base:**  $T(n) = \Theta(1)$  for all  $n < n_0$ , where  $n_0$  is a suitable constant.
- For  $1 \leq n < n_0$ , we have “ $\Theta(1)$ ”  $\leq cn^3$ , if we pick  $c$  big enough.

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***This bound is not tight!***

# A tighter upper bound?

We shall prove that  $T(n) = O(n^2)$ .

Assume that  $T(k) \leq ck^2$  for  $k < n$ :

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

$$\leq 4cn^2 + n$$

$$= \cancel{cn^2} \quad \text{Wrong!} \quad \text{We must prove the I.H.}$$

$$= cn^2 - (-n) \quad [ \text{desired} - \text{residual} ]$$

$$\leq cn^2$$

for *no* choice of  $c > 0$ . Lose!

# A tighter upper bound!

**IDEA:** Strengthen the inductive hypothesis.

- ***Subtract*** a low-order term.

*Inductive hypothesis:*  $T(k) \leq c_1 k^2 - c_2 k$  for  $k < n$ .

$$\begin{aligned} T(n) &= 4T(n/2) + n \\ &\leq 4(c_1(n/2)^2 - c_2(n/2)) + n \\ &= c_1 n^2 - 2c_2 n + n \\ &= c_1 n^2 - c_2 n - (c_2 n - n) \\ &\leq c_1 n^2 - c_2 n \quad \text{if } c_2 > 1. \end{aligned}$$

Pick  $c_1$  big enough to handle the initial conditions.

# 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, just like any method that uses ellipses (...).
- The recursion-tree method promotes intuition, however.



# Example of recursion tree

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

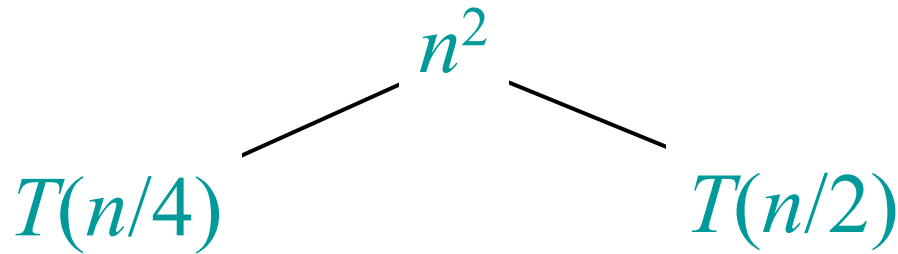
# Example of recursion tree

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

$$T(n)$$

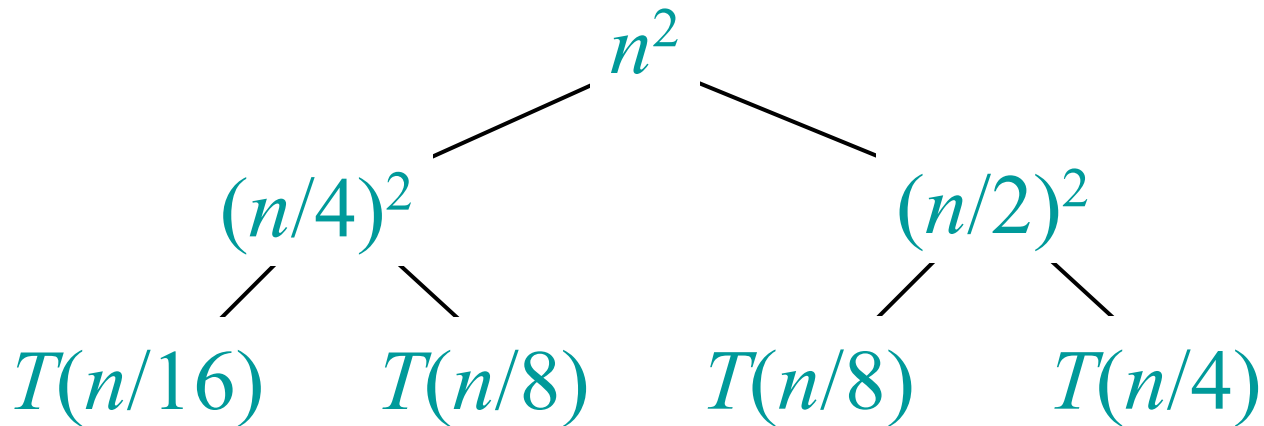
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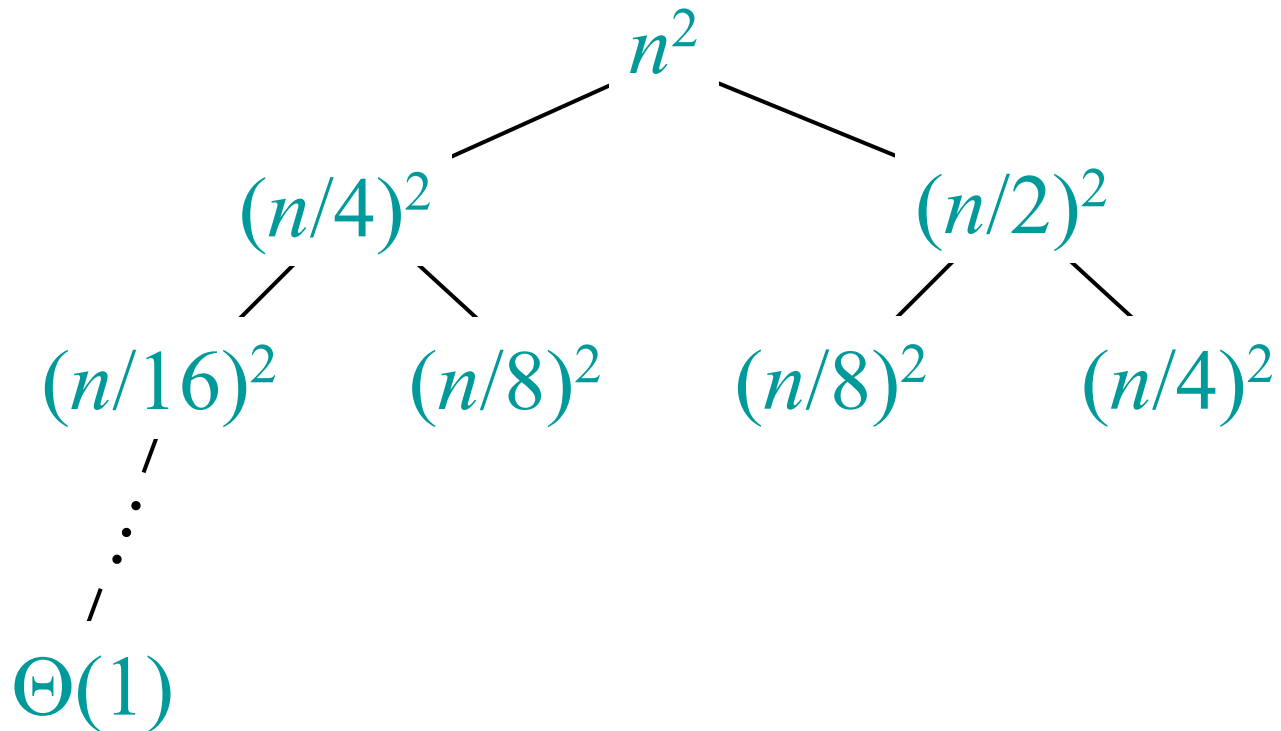
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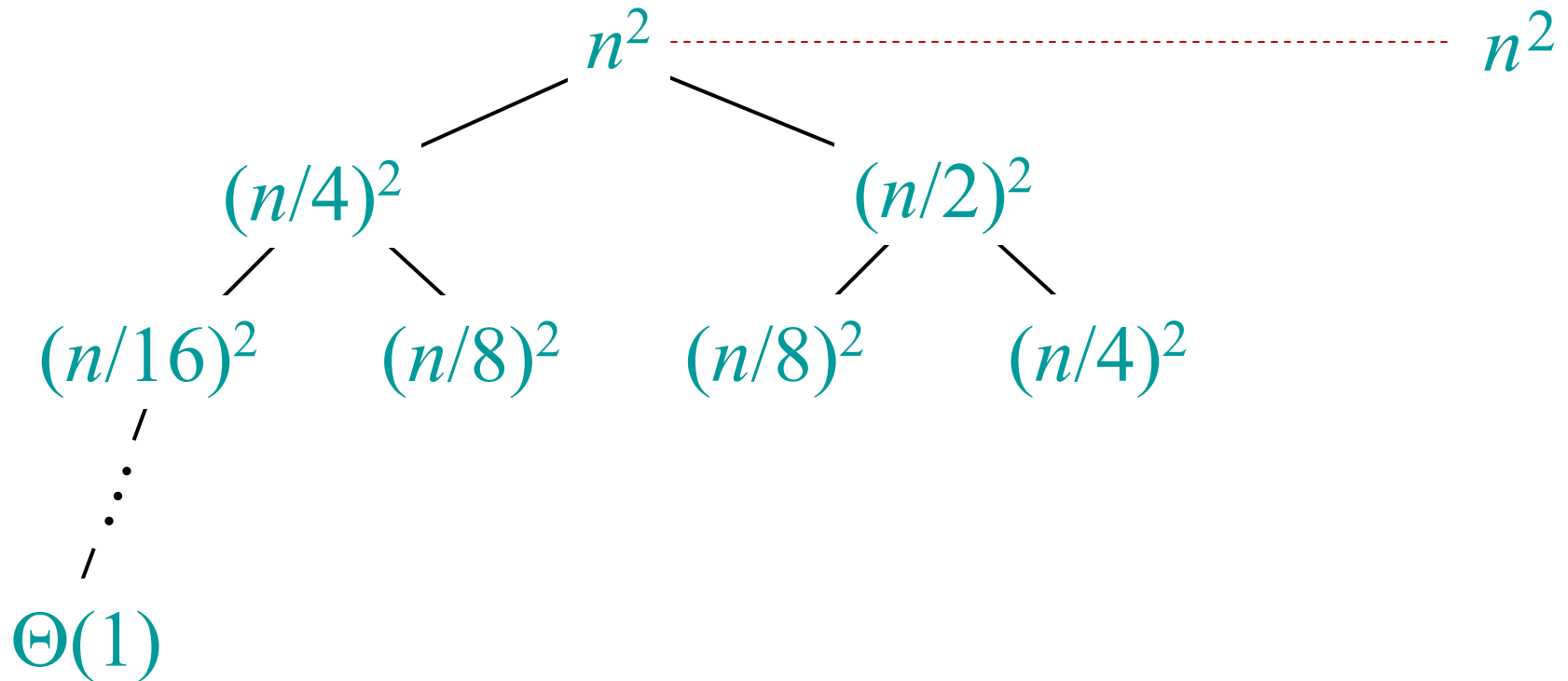
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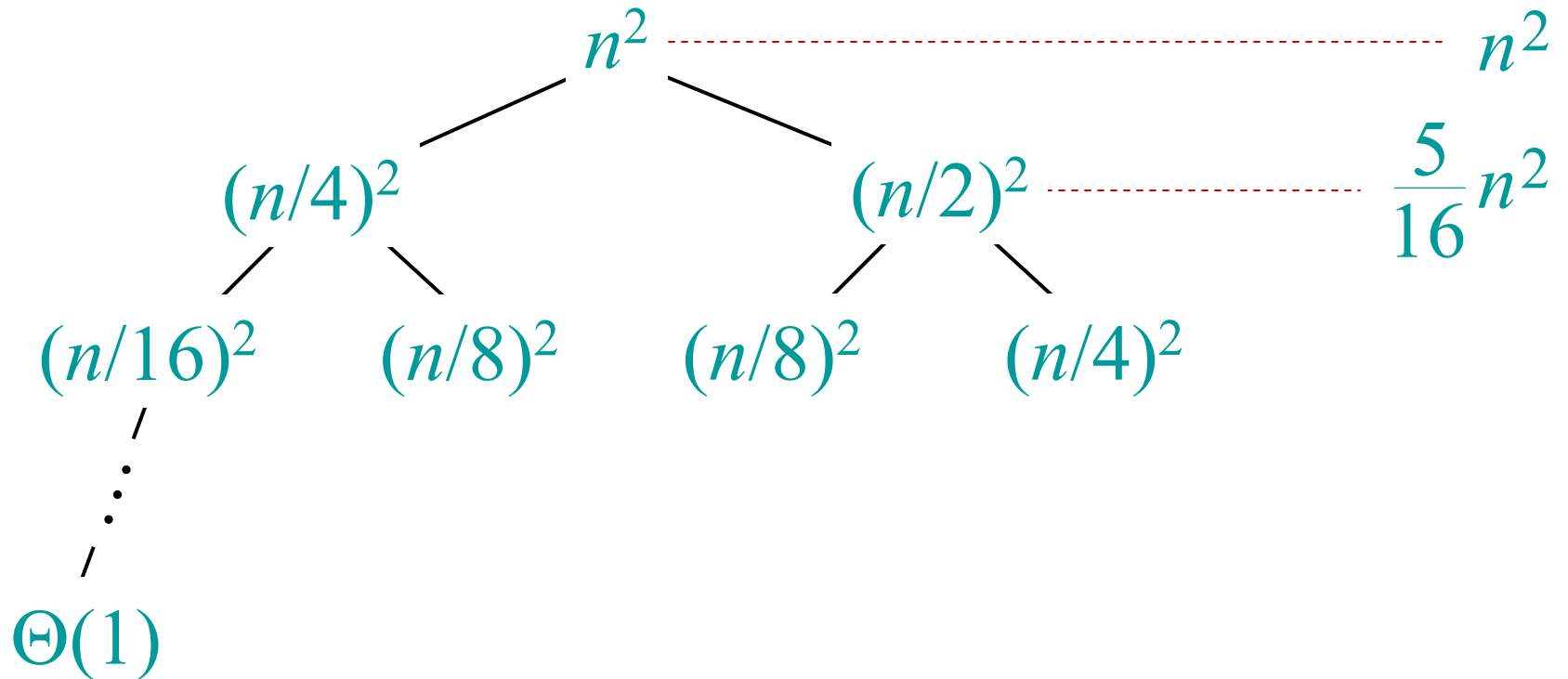
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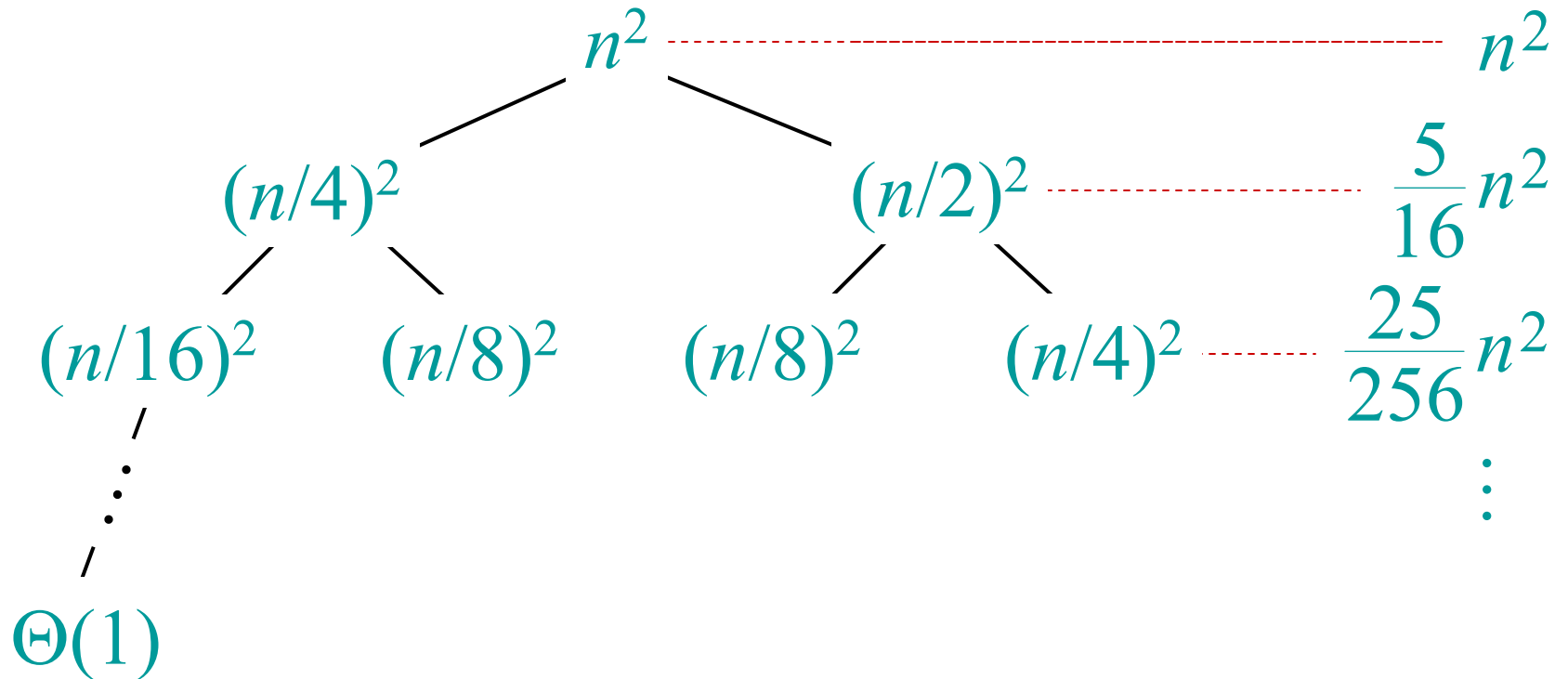
# Example of recursion tree

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



# Example of recursion tree

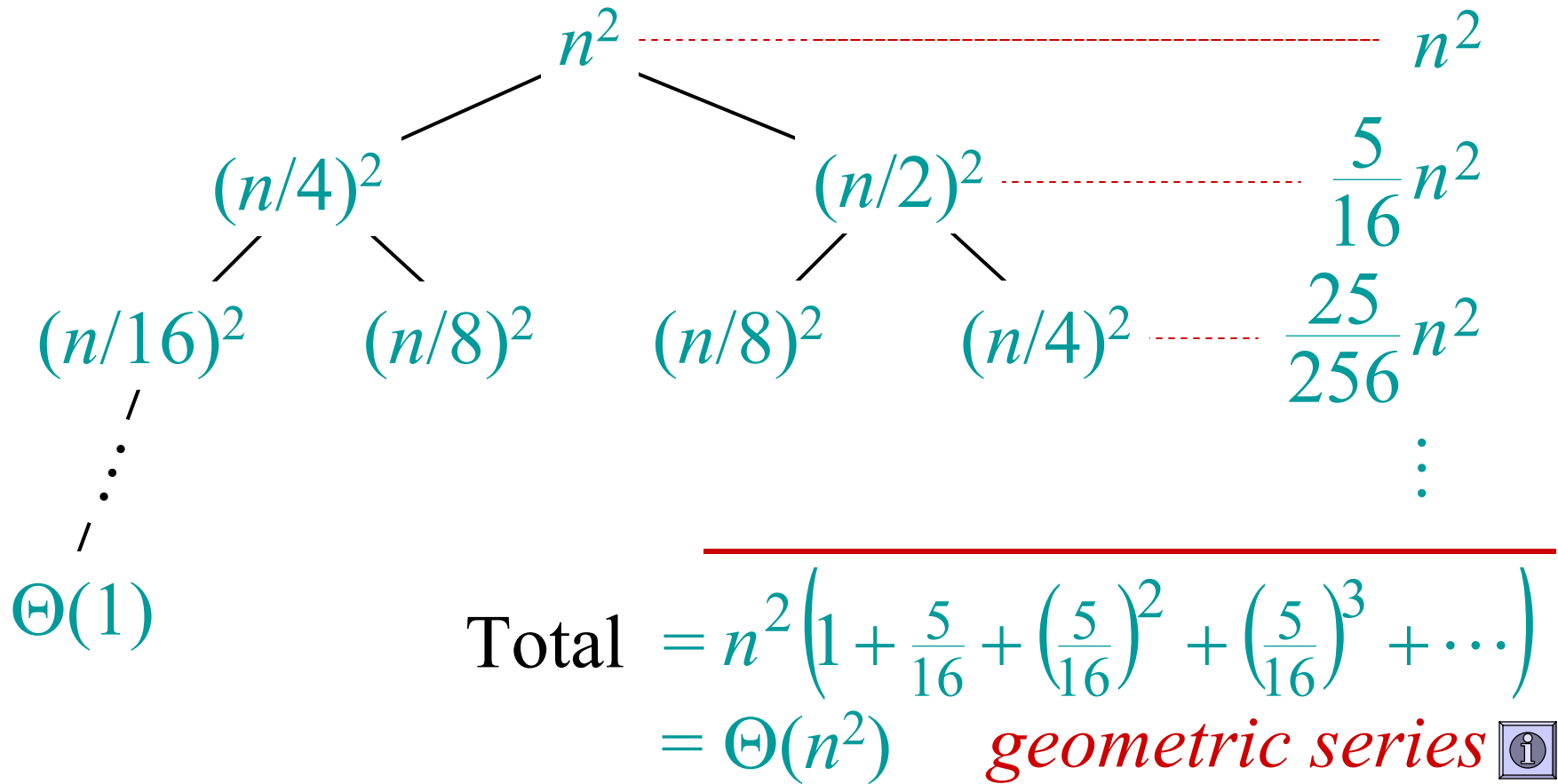
Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :





# Example of recursion tree

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



# The master method

The master method applies to recurrences of the form

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

where  $a \geq 1$ ,  $b > 1$ , and  $f$  is asymptotically positive.

# Three common cases

Compare  $f(n)$  with  $n^{\log_b a}$ :

1.  $f(n) = O(n^{\log_b a - \varepsilon})$  for some constant  $\varepsilon > 0$ .

- $f(n)$  grows polynomially slower than  $n^{\log_b a}$  (by an  $n^\varepsilon$  factor).

**Solution:**  $T(n) = \Theta(n^{\log_b a})$ .

2.  $f(n) = \Theta(n^{\log_b a} \lg^k n)$  for some constant  $k \geq 0$ .

- $f(n)$  and  $n^{\log_b a}$  grow at similar rates.

**Solution:**  $T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$ .

# Three common cases (cont.)

Compare  $f(n)$  with  $n^{\log_b a}$ :

3.  $f(n) = \Omega(n^{\log_b a + \varepsilon})$  for some constant  $\varepsilon > 0$ .

- $f(n)$  grows polynomially faster than  $n^{\log_b a}$  (by an  $n^\varepsilon$  factor),

*and*  $f(n)$  satisfies the **regularity condition** that  $a f(n/b) \leq c f(n)$  for some constant  $c < 1$ .

**Solution:**  $T(n) = \Theta(f(n))$ .

# Examples

**Ex.**  $T(n) = 4T(n/2) + n$

$$a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n.$$

**CASE 1:**  $f(n) = O(n^{2-\epsilon})$  for  $\epsilon = 1$ .

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

**Ex.**  $T(n) = 4T(n/2) + n^2$

$$a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^2.$$

**CASE 2:**  $f(n) = \Theta(n^2 \lg^0 n)$ , that is,  $k = 0$ .

$$\therefore T(n) = \Theta(n^2 \lg n).$$

# Examples

**Ex.**  $T(n) = 4T(n/2) + n^3$

$$a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^3.$$

**CASE 3:**  $f(n) = \Omega(n^{2 + \varepsilon})$  for  $\varepsilon = 1$

**and**  $4(cn/2)^3 \leq cn^3$  (reg. cond.) for  $c = 1/2$ .

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

**Ex.**  $T(n) = 4T(n/2) + n^2/\lg n$

$$a = 4, b = 2 \Rightarrow n^{\log_b a} = n^2; f(n) = n^2/\lg n.$$

Master method does not apply. In particular, for every constant  $\varepsilon > 0$ , we have  $n^\varepsilon = \omega(\lg n)$ .

# General method (Akra-Bazzi)

$$T(n) = \sum_{i=1}^k a_i T(n/b_i) + f(n)$$

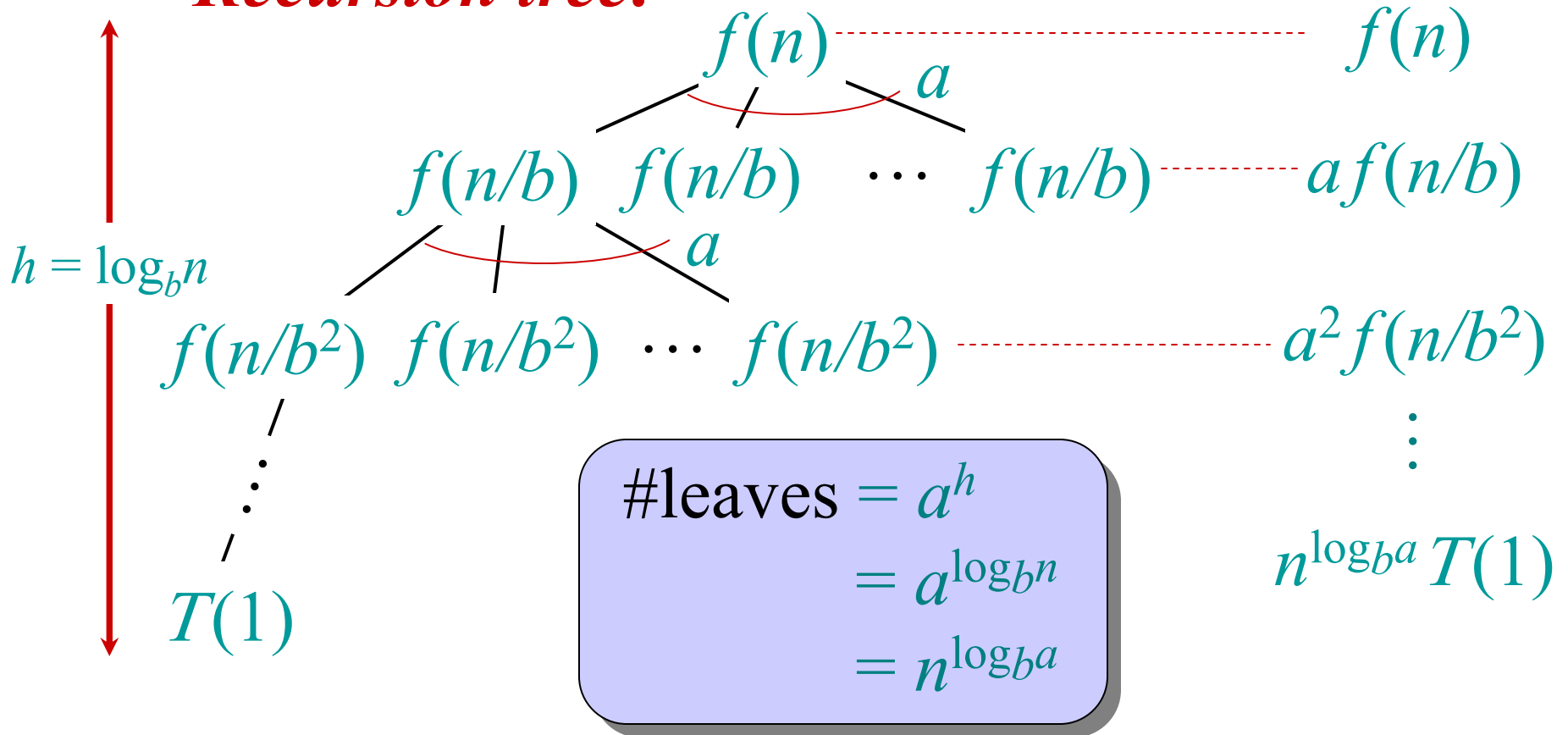
Let  $p$  be the unique solution to

$$\sum_{i=1}^k (a_i/b_i^p) = 1.$$

Then, the answers are the same as for the master method, but with  $n^p$  instead of  $n^{\log_b a}$ .  
(*Akra and Bazzi also prove an even more general result.*)

# Idea of master theorem

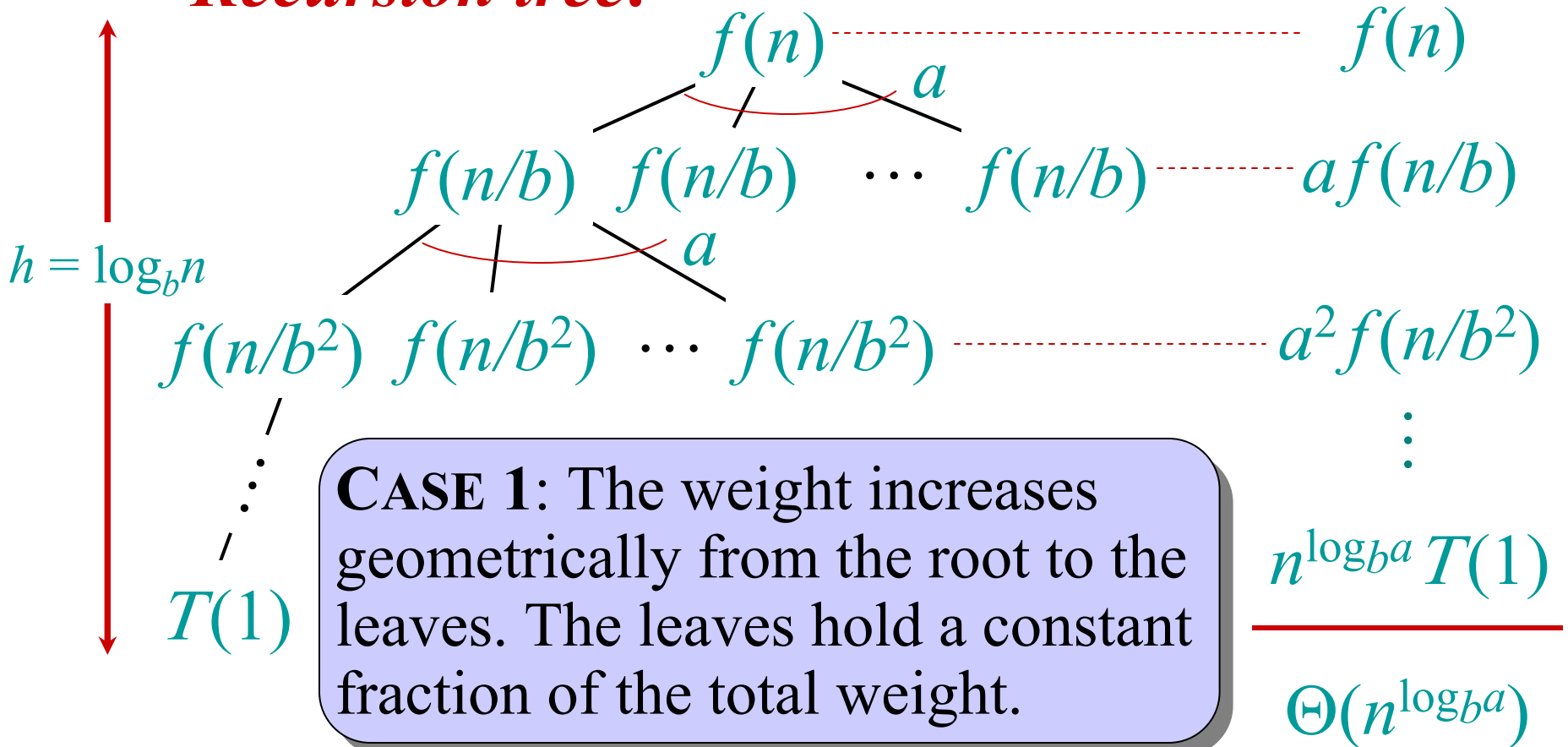
*Recursion tree:*





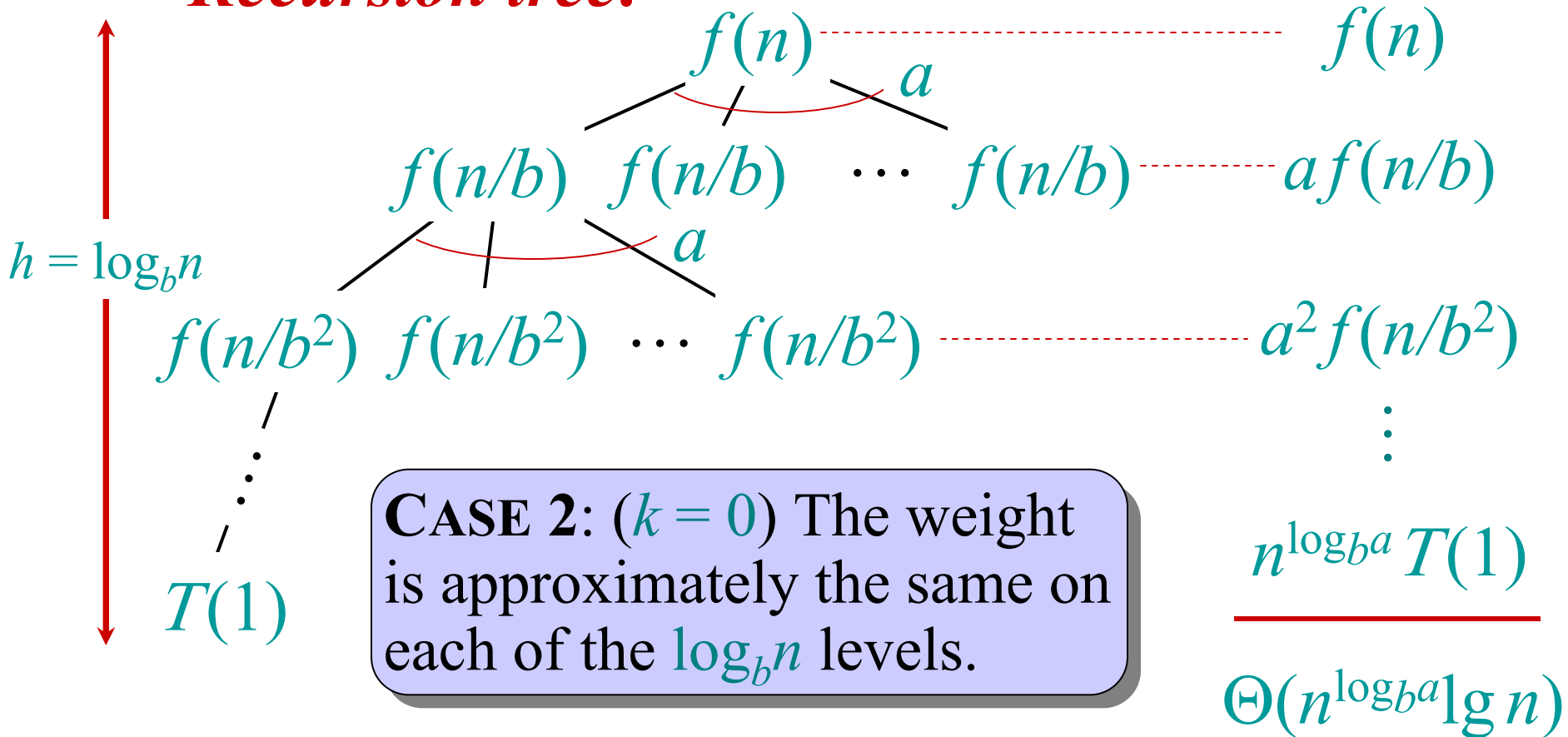
# Idea of master theorem

*Recursion tree:*



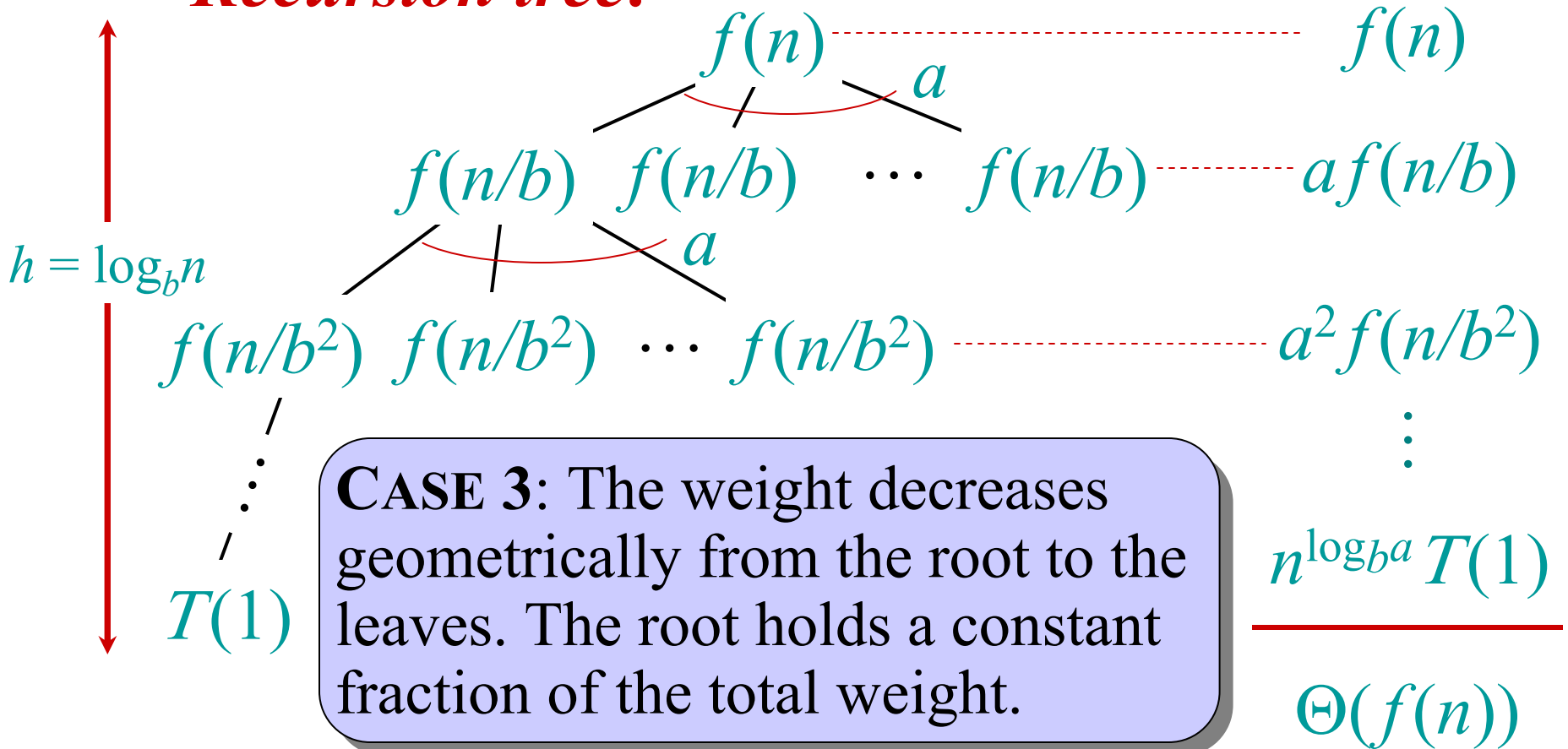
# Idea of master theorem

*Recursion tree:*



# Idea of master theorem

*Recursion tree:*



**CASE 3:** The weight decreases geometrically from the root to the leaves. The root holds a constant fraction of the total weight.

# Conclusion

- Next time: applying the master method.
- For proof of master theorem, see CLRS.

# Appendix: geometric series

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

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

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