
CSE 301

Combinatorial Optimization

Lecture 2

Recurrence

Today's Topic

- Recurrence
 - Substitution Method
 - Recursive Method
 - Master Method
 - Akra-bazzi method

Recurrence Relations

- A *recurrence relation* is the recursive part of a *recursive definition* of either a number sequence or integer function.

Recursively Defined Sequences

- Fibonacci sequence:
 - $\{f_n\} = 0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, \dots$
 - Recursive definition for $\{f_n\}$:
 - INITIALIZE: $f_0 = 0, f_1 = 1$
 - RECURSE: $f_n = f_{n-1} + f_{n-2}$ for $n > 1$.
 - The recurrence relation is the recursive part
 - $f_n = f_{n-1} + f_{n-2}$. Thus a recurrence relation for a sequence consists of an equation that expresses each term in terms of lower terms.

Substitution method

The most general method:

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

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

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

Example of substitution

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

$$\leq 2(c(n/2) \lg(n/2)) + n$$

$$\leq cn \lg(n/2) + n$$

$$= cn \lg n - cn \lg 2 + n$$

$$= cn \lg n - (cn - n) \quad \leftarrow \textit{desired} - \textit{residual}$$

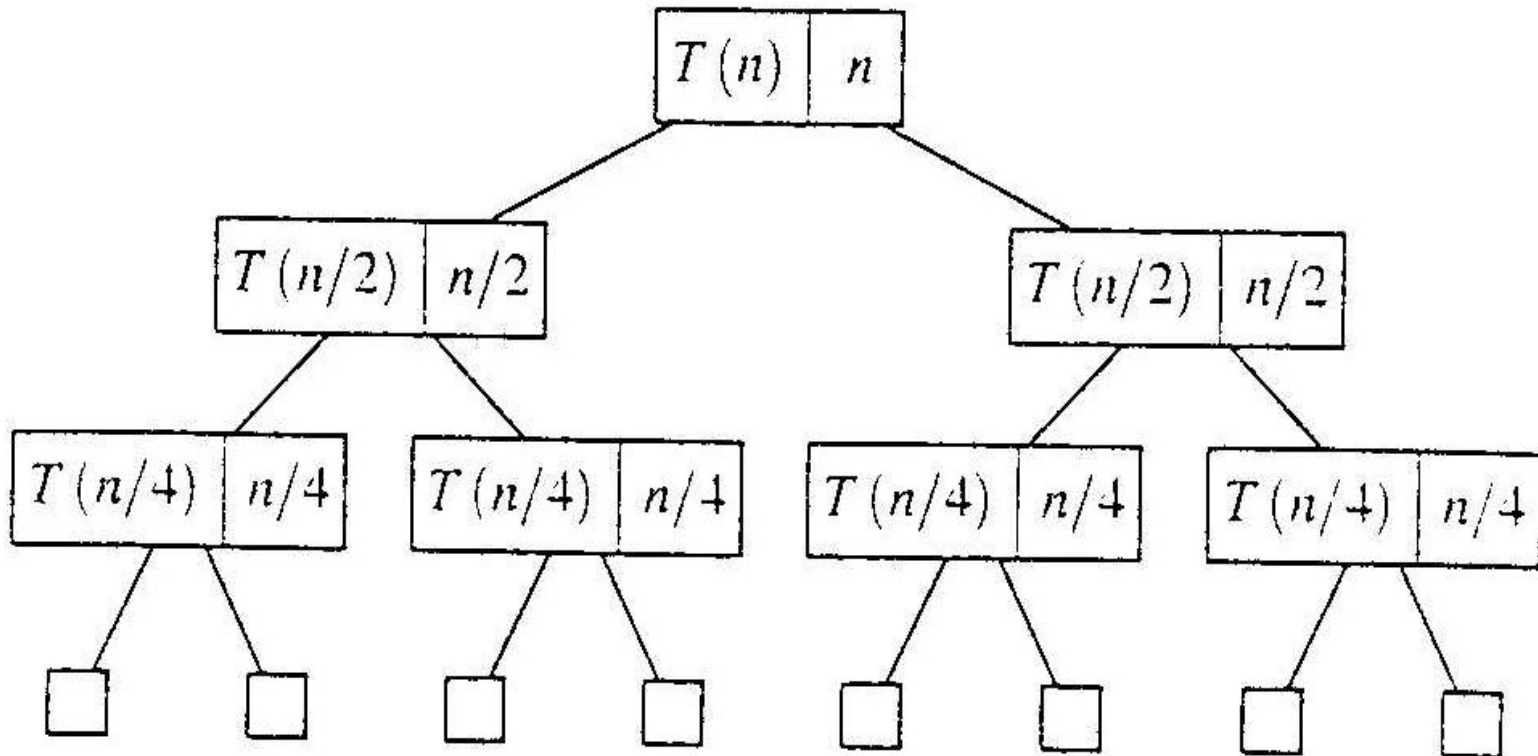
$$\leq cn \lg n \quad \leftarrow \textit{desired}$$

whenever $cn - n \geq 0$, for example, if $c \geq 2$ and $n \geq 1$.

\nwarrow *residual*

Evaluate recursive equation using 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 e.g.

- 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 rowSum for each row is n
- Therefore, the total cost, $T(n) = n \lg(n)$

Example of recursion tree

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

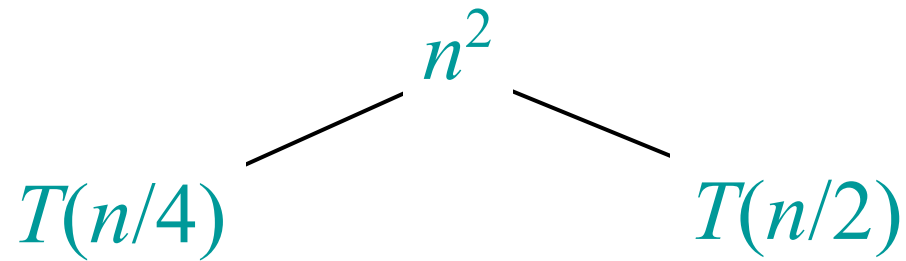
Example of recursion tree

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

$$T(n)$$

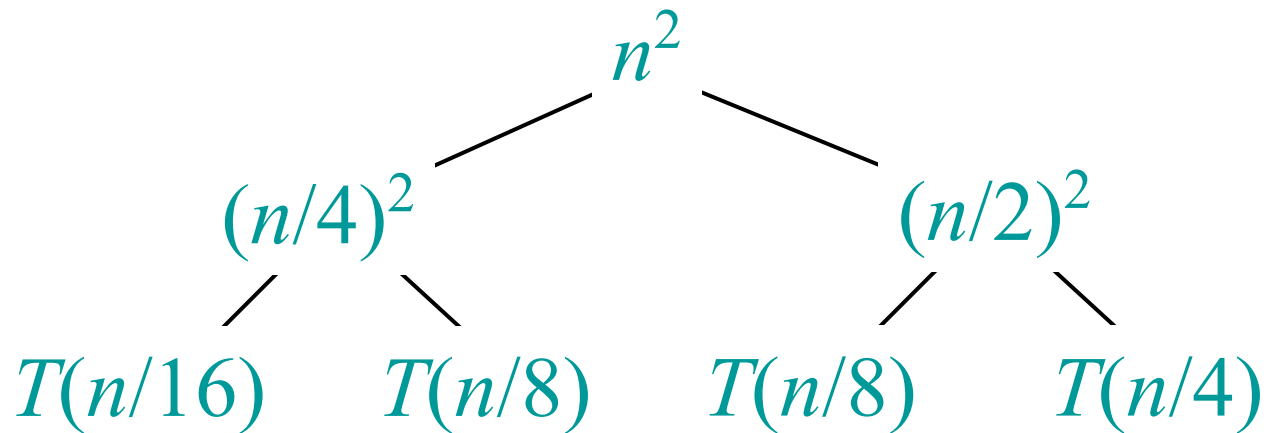
Example of recursion tree

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



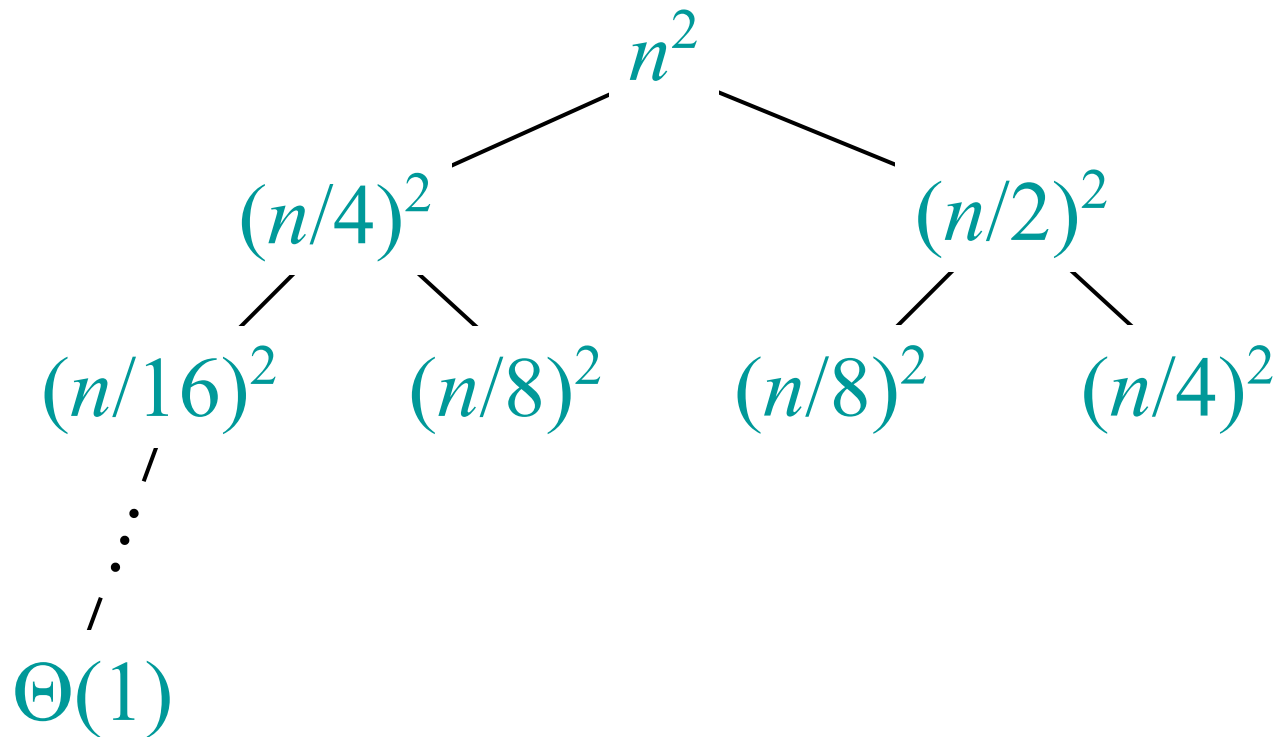
Example of recursion tree

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



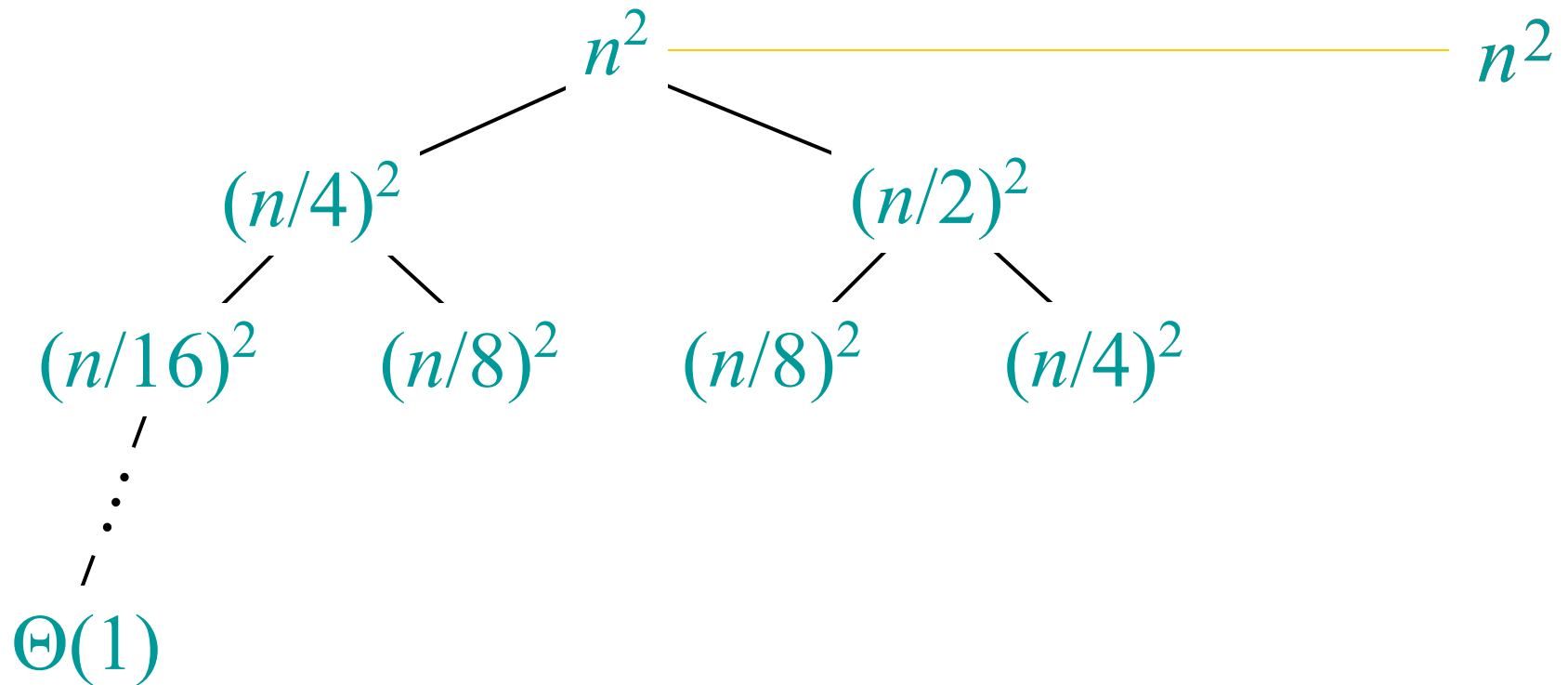
Example of recursion tree

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



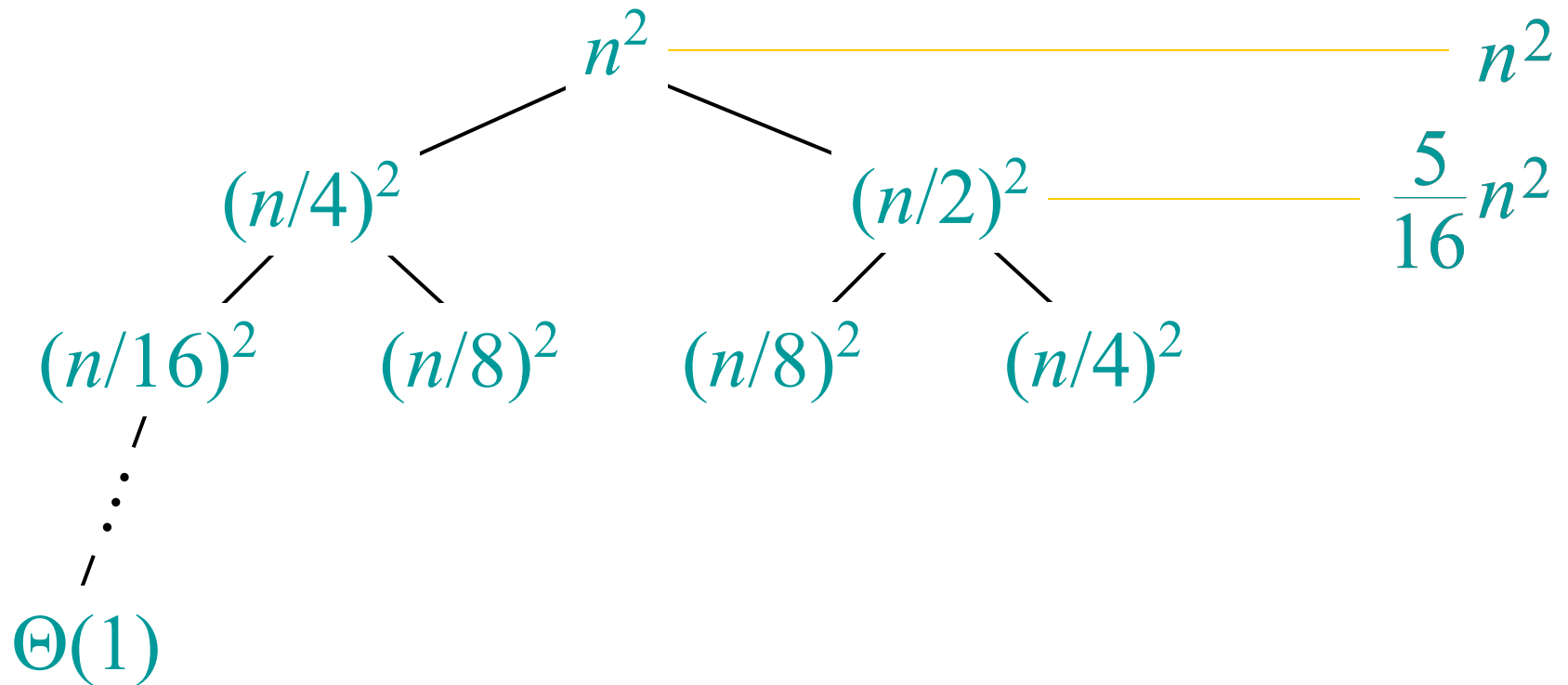
Example of recursion tree

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



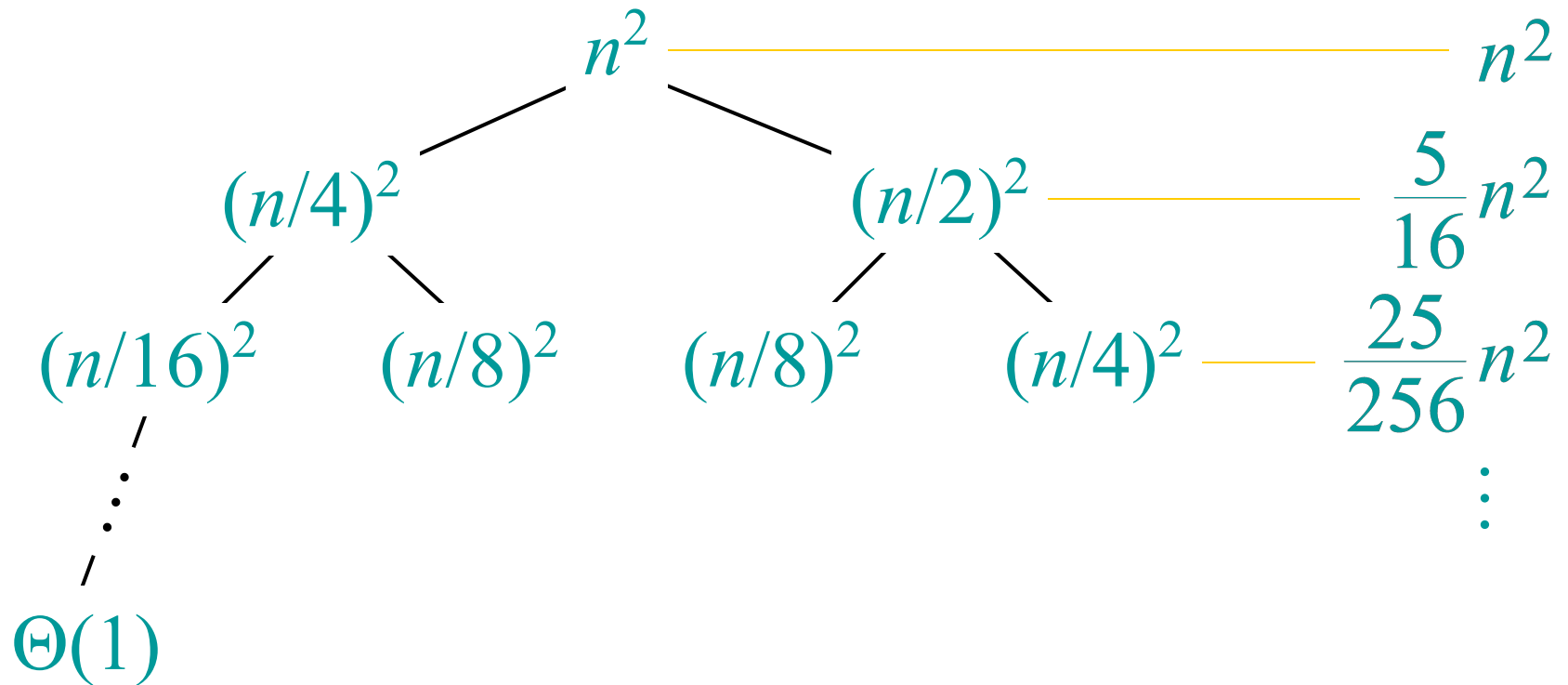
Example of recursion tree

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



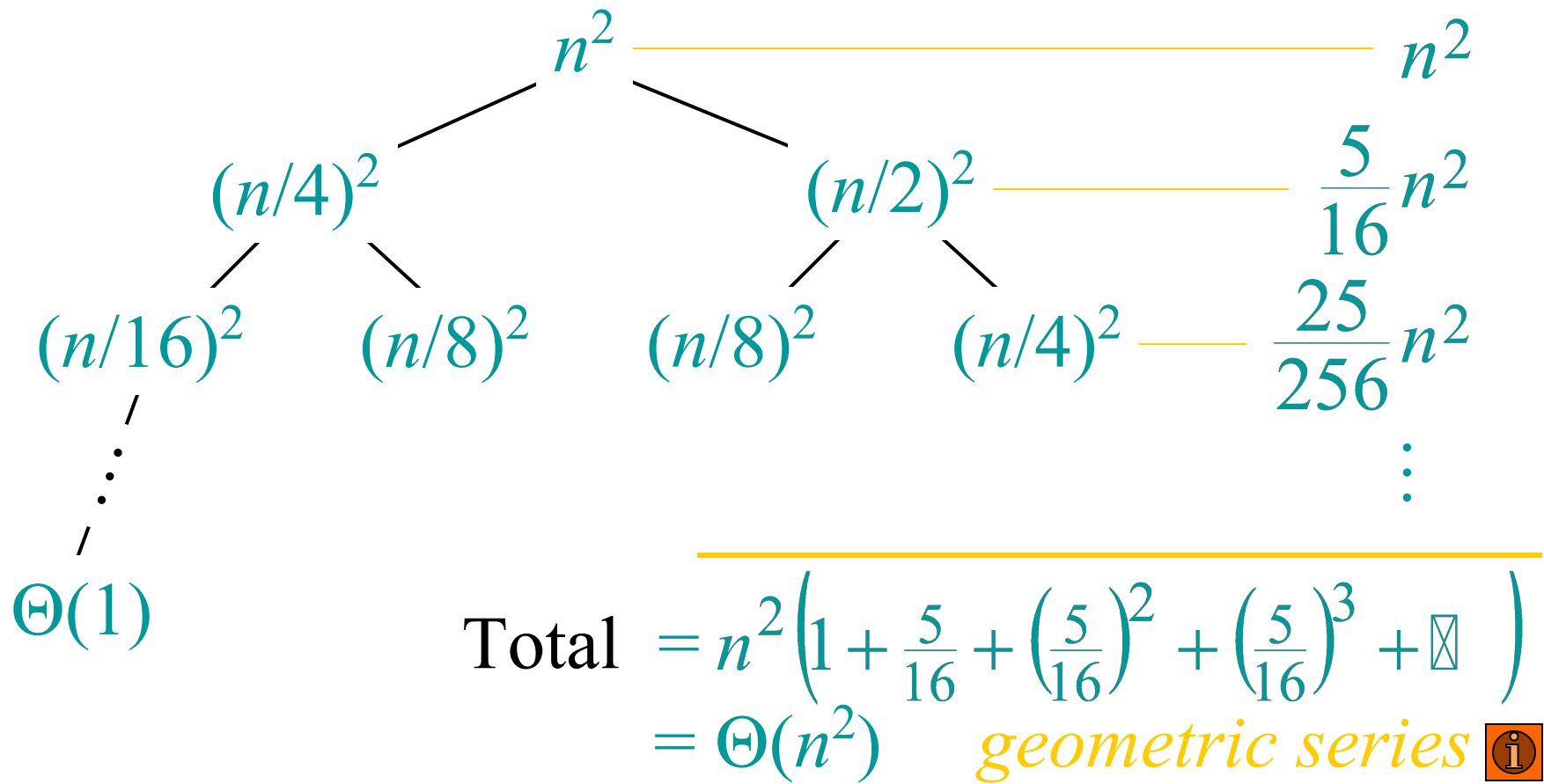
Example of recursion tree

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



Example of recursion tree

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



The divide-and-conquer design paradigm

1. *Divide* the problem (instance) into subproblems.
2. *Conquer* the subproblems by solving them recursively.
3. *Combine* subproblem solutions.

Example: merge sort

1. **Divide:** Trivial.
2. **Conquer:** Recursively sort 2 subarrays.
3. **Combine:** Linear-time merge.

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

subproblems

subproblem size

work dividing and combining

The master method

The master method applies to recurrences of the form

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

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^ε factor).

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

2. $f(n) = \Theta(n^{\log_b a})$.

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

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

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

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)

- The Master method is fairly powerful and results in a closed form solution for divide-and-conquer recurrences with a special form.
- Akra and Bazzi discovered a far more general solution to divide-and-conquer recurrences.

The Akra-Bazzi Method

$$T(x) = \begin{cases} \Theta(1) & \text{for } 1 \leq x \leq x_0 \\ \sum_{i=1}^k a_i T(b_i x) + g(x) & \text{for } x > x_0 \end{cases} \quad (1)$$

where¹

1. $x \geq 1$ is a real number,
2. x_0 is a constant such that $x_0 \geq 1/b_i$ and $x_0 \geq 1/(1 - b_i)$ for $1 \leq i \leq k$,
3. $a_i > 0$ is a constant for $1 \leq i \leq k$,
4. $b_i \in (0, 1)$ is a constant for $1 \leq i \leq k$,
5. $k \geq 1$ is a constant, and
6. $g(x)$ is a nonnegative function that satisfies the polynomial-growth condition specified below.

Definition. We say that $g(x)$ satisfies the *polynomial-growth condition* if there exist positive constants c_1, c_2 such that for all $x \geq 1$, for all $1 \leq i \leq k$, and for all $u \in [b_i x, x]$,

$$c_1 g(x) \leq g(u) \leq c_2 g(x).$$

The Akra-Bazzi Solution

Theorem 1 ([1]). *Given a recurrence of the form specified in Equation 1, let p be the unique real number for which $\sum_{i=1}^k a_i b_i^p = 1$. Then*

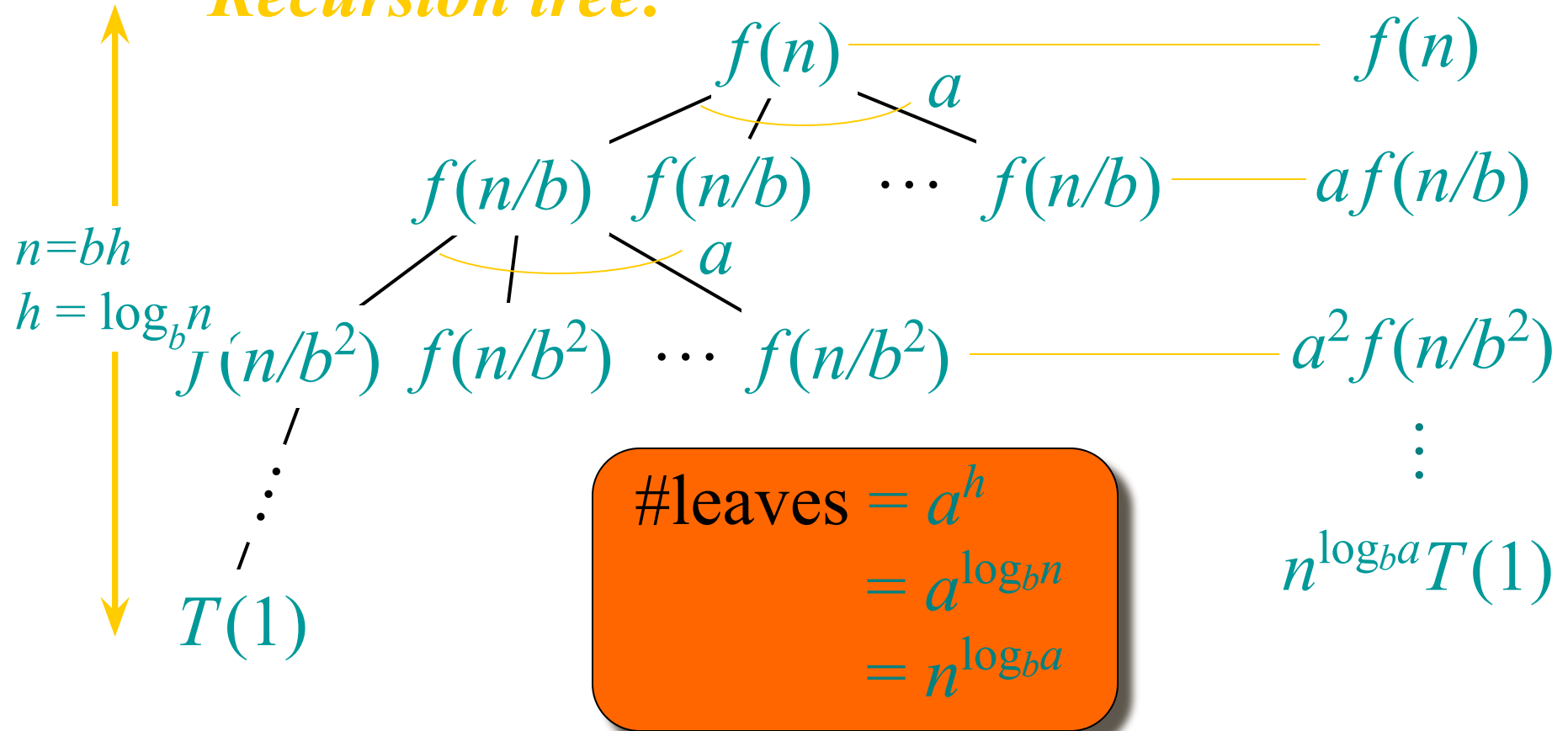
$$T(x) = \Theta \left(x^p \left(1 + \int_1^x \frac{g(u)}{u^{p+1}} du \right) \right).$$

Examples.

- If $T(x) = 2T(x/4) + 3T(x/6) + \Theta(x \log x)$, then $p = 1$ and $T(x) = \Theta(x \log^2 x)$.
- If $T(x) = 2T(x/2) + \frac{8}{9}T(3x/4) + \Theta(x^2 / \log x)$, then $p = 2$ and $T(x) = \Theta(x^2 / \log \log x)$.
- If $T(x) = T(x/2) + \Theta(\log x)$, then $p = 0$ and $T(x) = \Theta(\log^2 x)$.
- If $T(x) = \frac{1}{2}T(x/2) + \Theta(1/x)$, then $p = -1$ and $T(x) = \Theta((\log x)/x)$.
- If $T(x) = 4T(x/2) + \Theta(x)$, then $p = 2$ and $T(x) = \Theta(x^2)$.

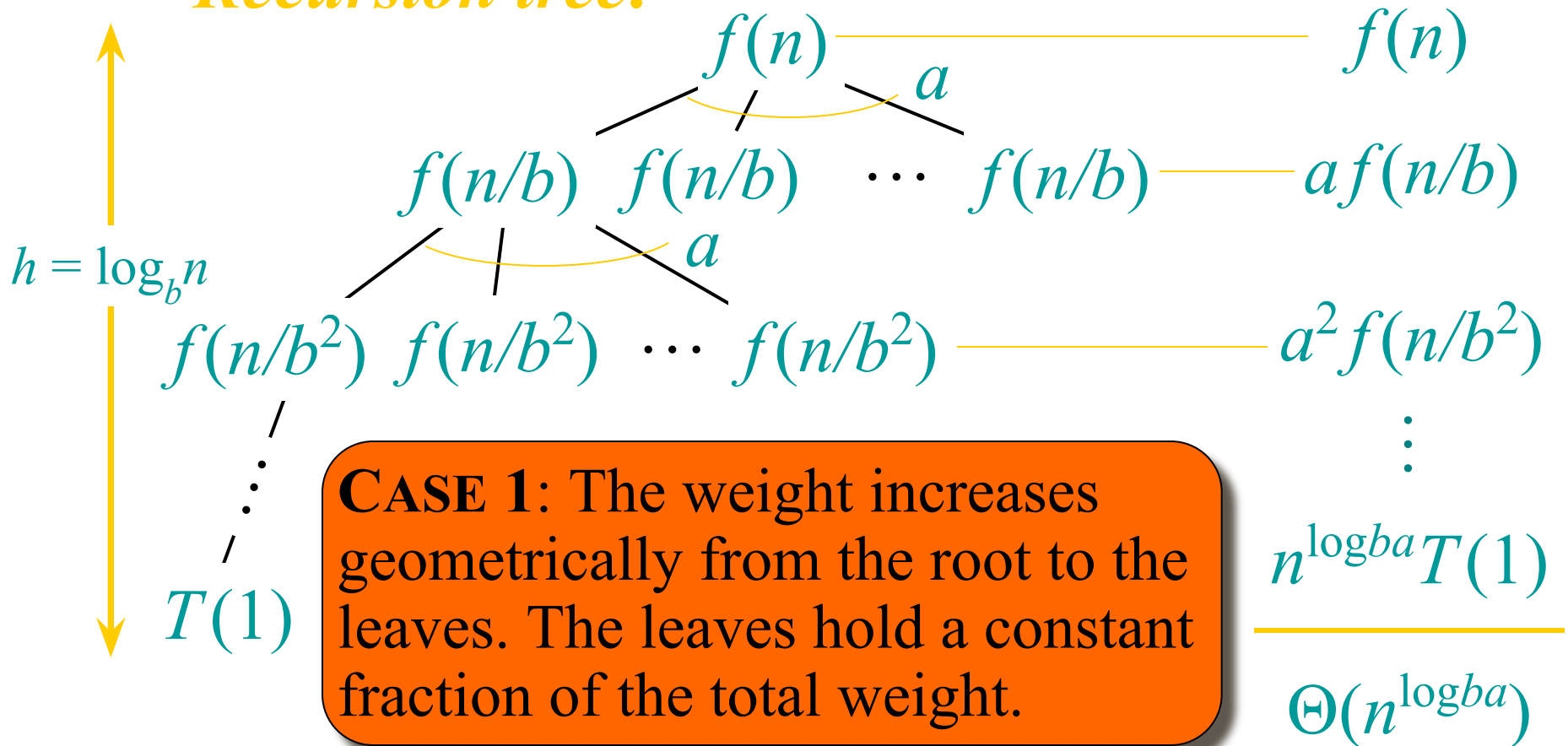
Idea of master theorem

Recursion tree:



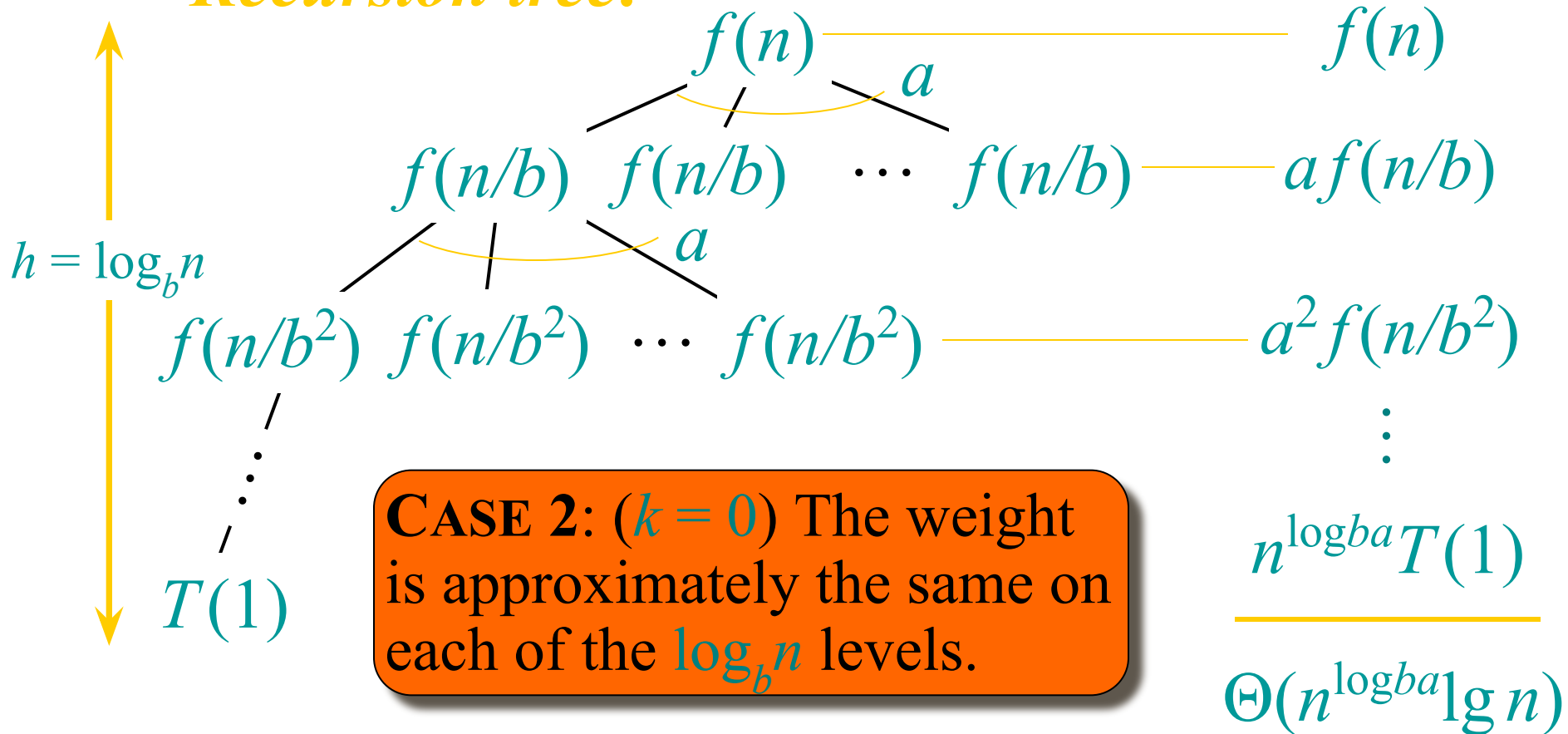
Idea of master theorem

Recursion tree:



Idea of master theorem

Recursion tree:



Idea of master theorem

Recursion tree:

