

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

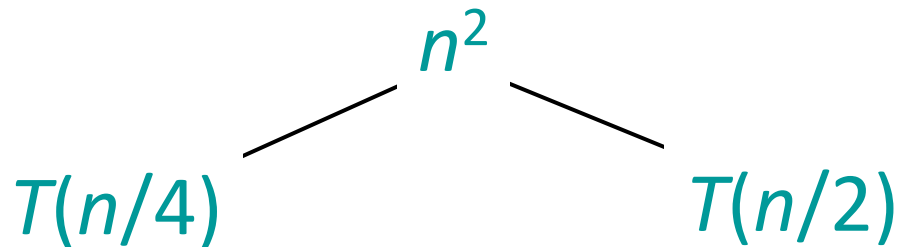
Example of recursion tree

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

$$T(n)$$

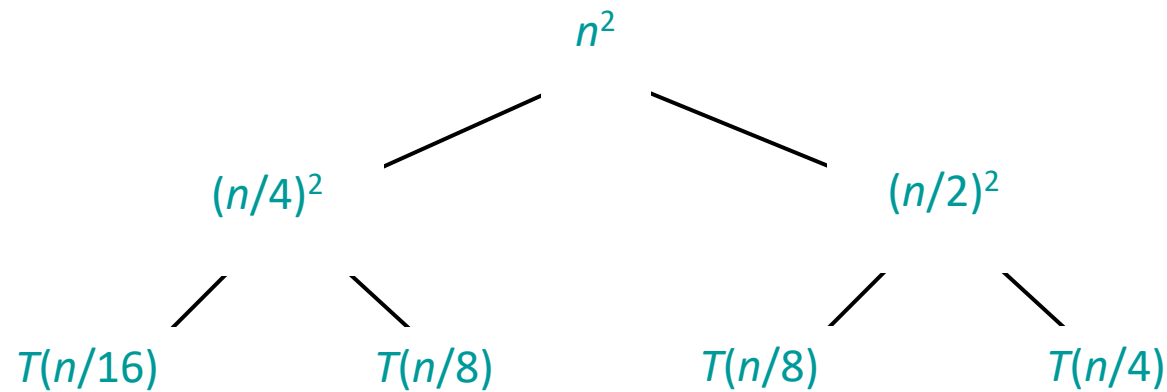
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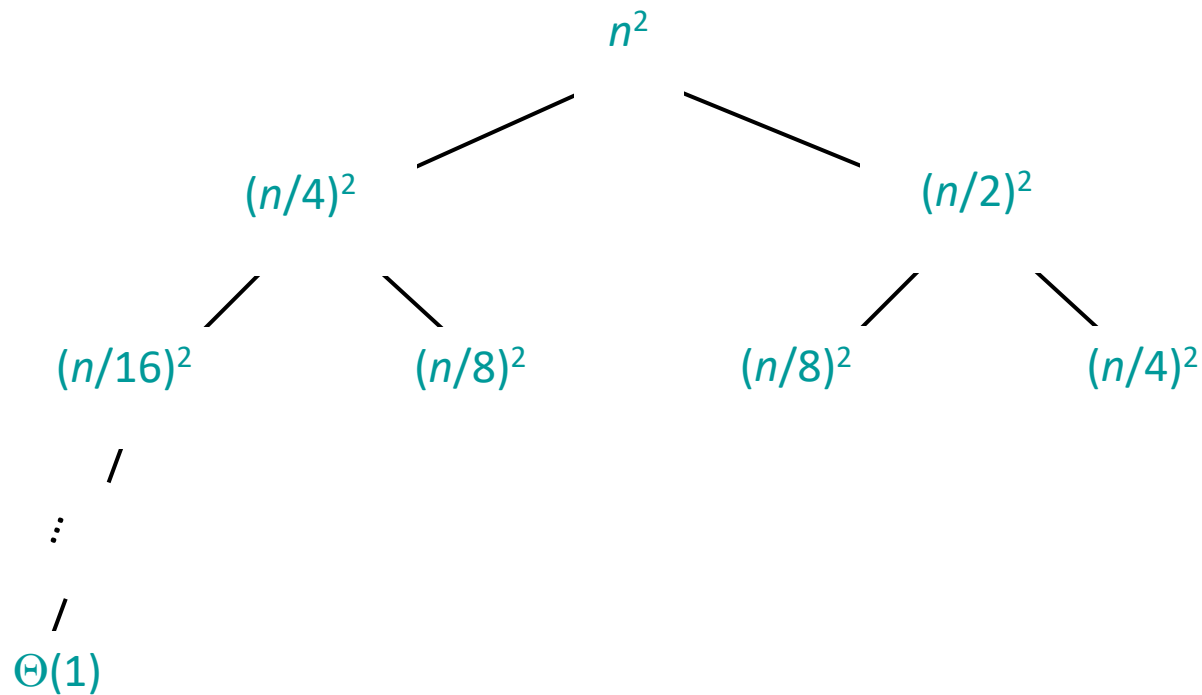
Example of recursion tree

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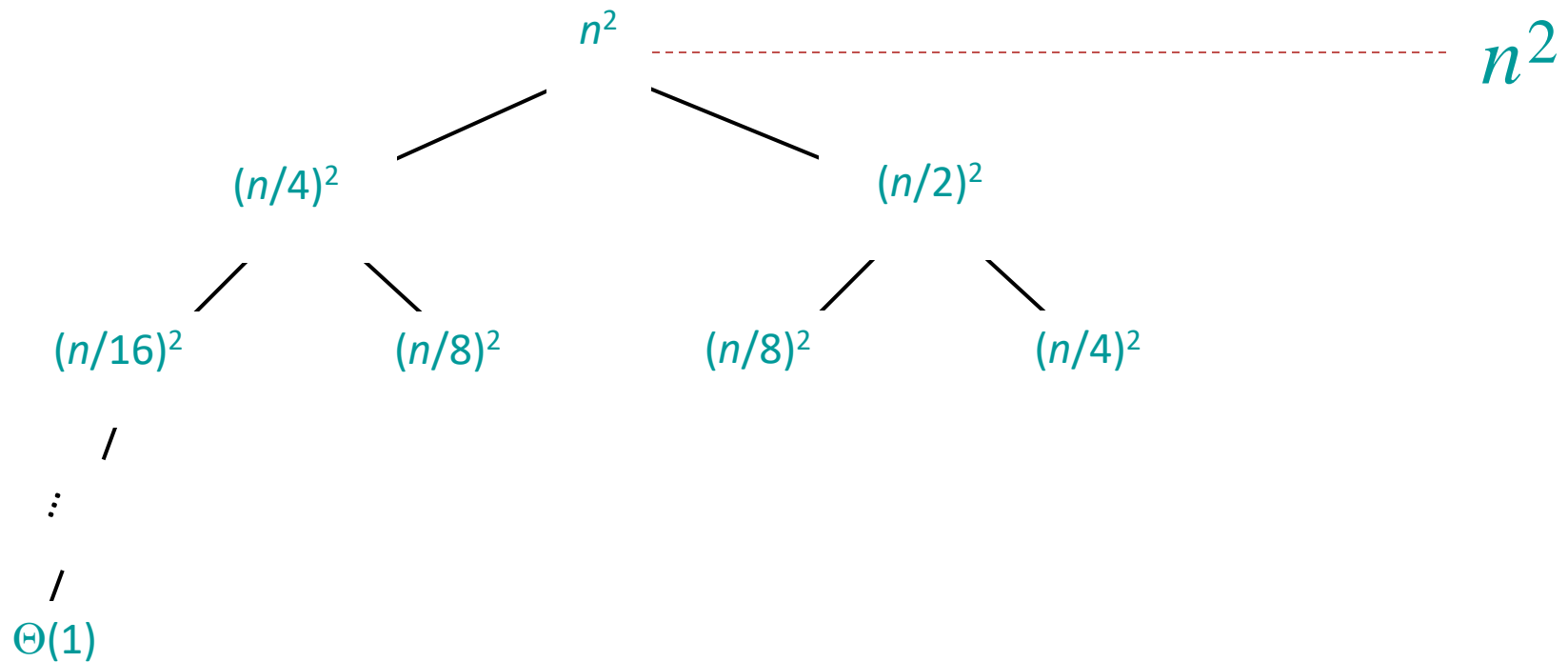
Example of recursion tree

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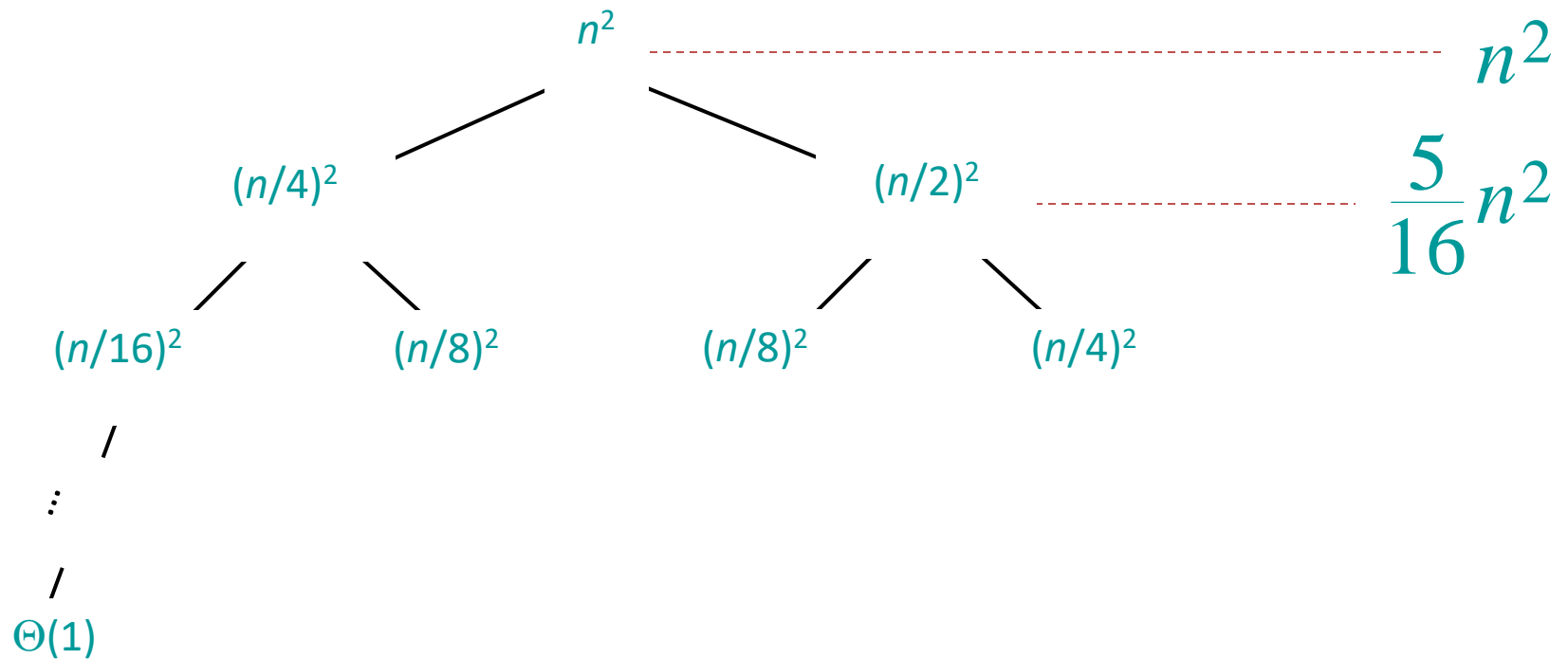
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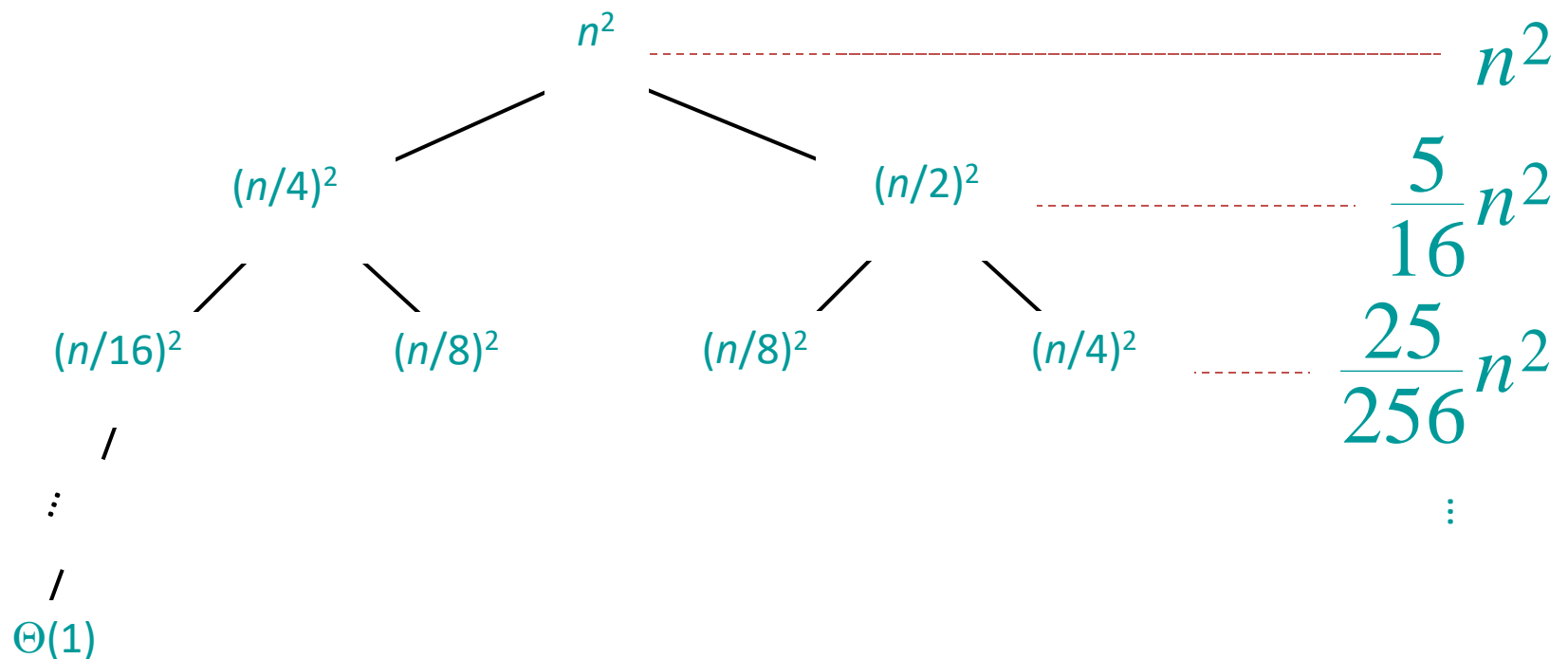
Example of recursion tree

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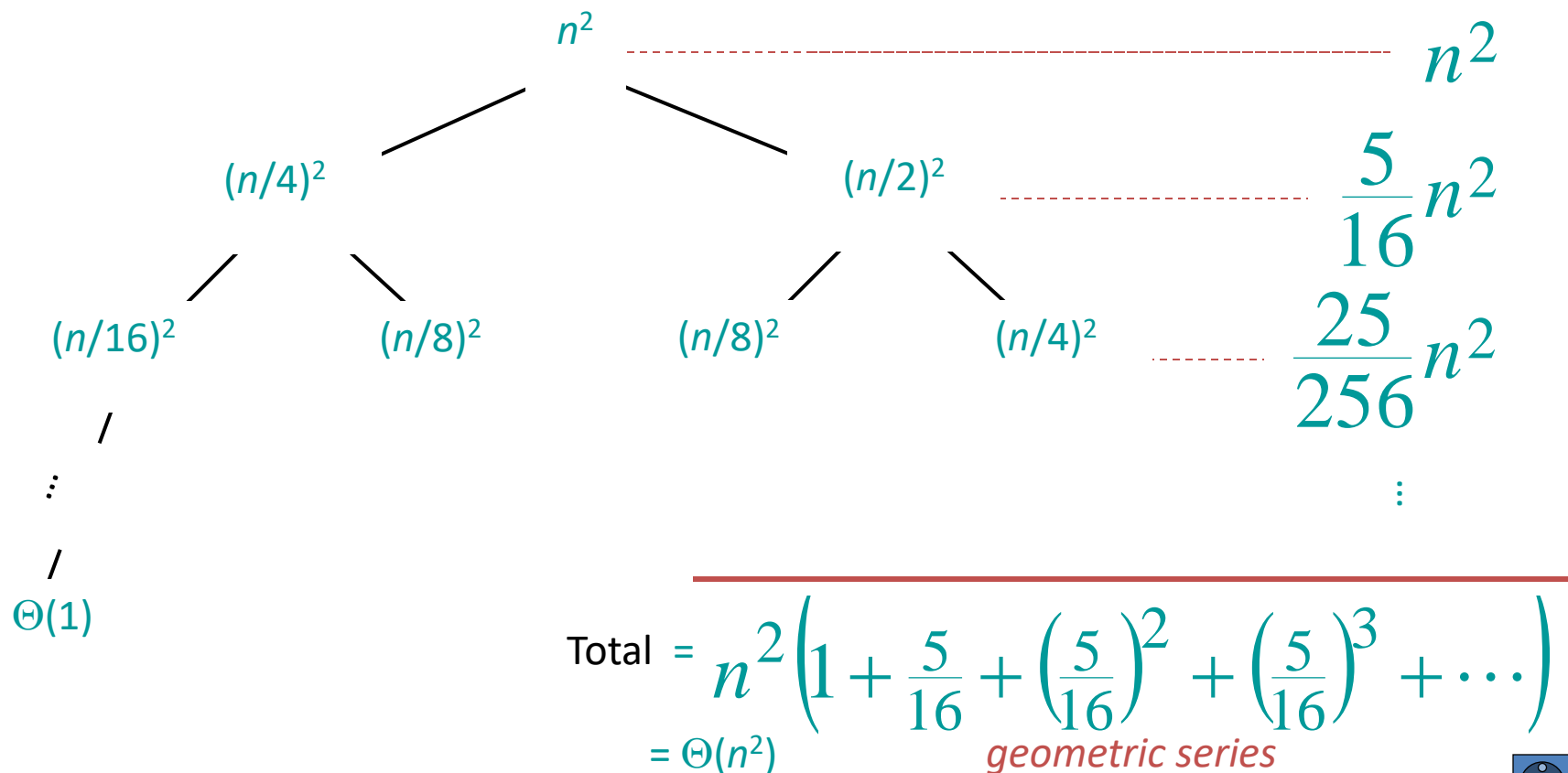
Example of recursion tree

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



Example of recursion tree

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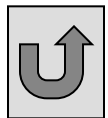


Appendix: geometric series

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

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

Return to last
slide viewed.



The master method

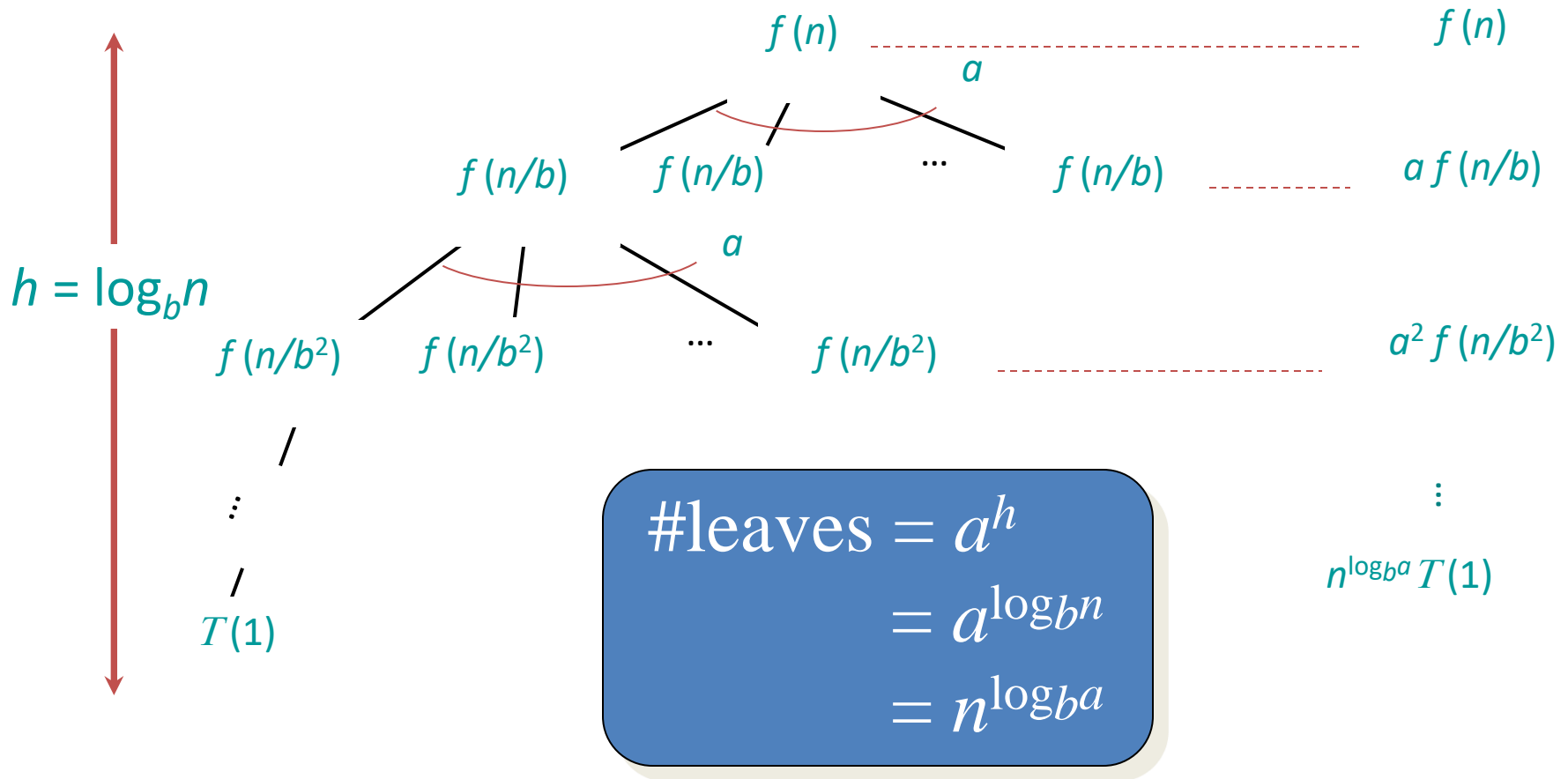
The master method applies to recurrences of the form

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

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

Idea of master theorem

Recursion tree:



Three common cases

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

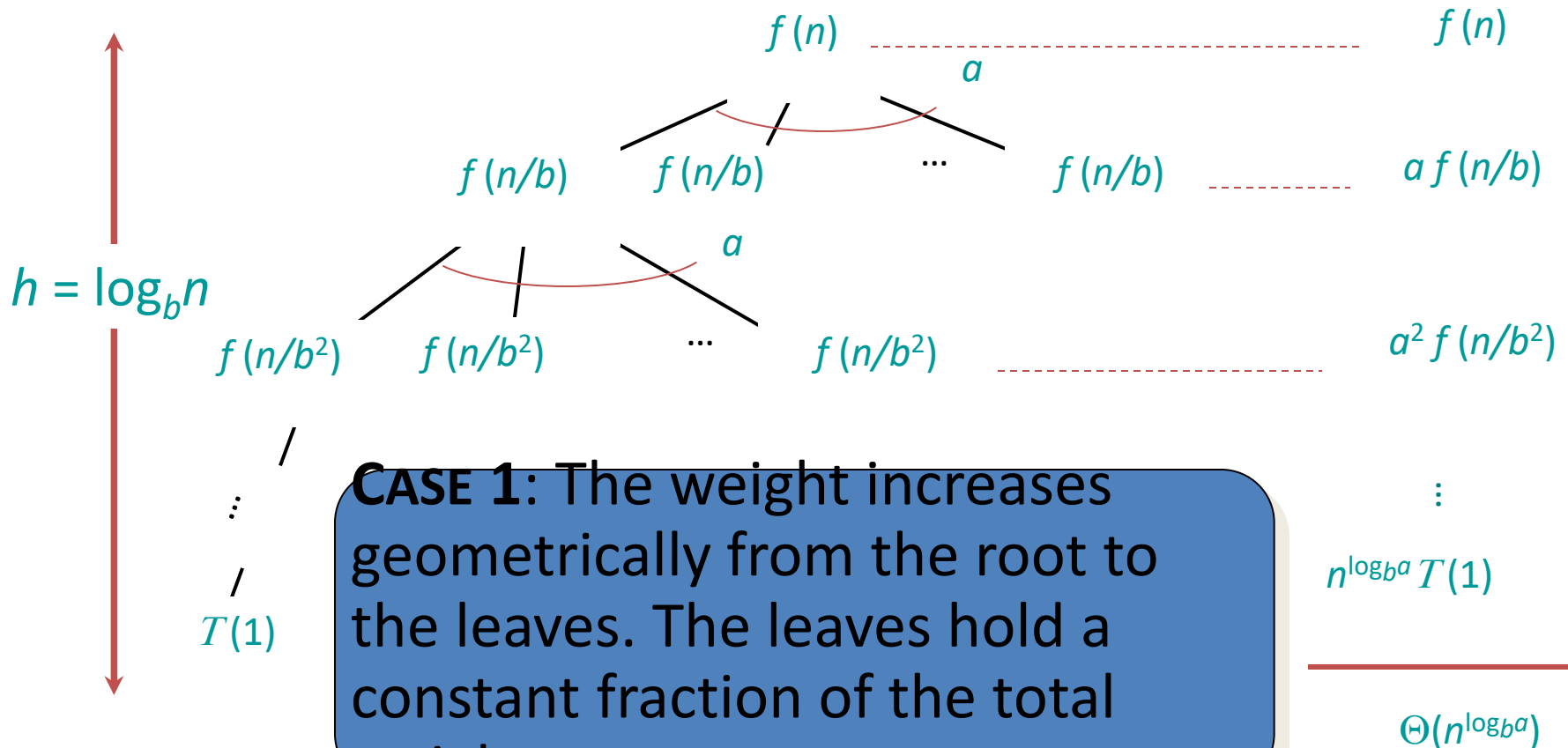
1. $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$.

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

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

Idea of master theorem

Recursion tree:



CASE 1: The weight increases geometrically from the root to the leaves. The leaves hold a constant fraction of the total weight.

Three common cases

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

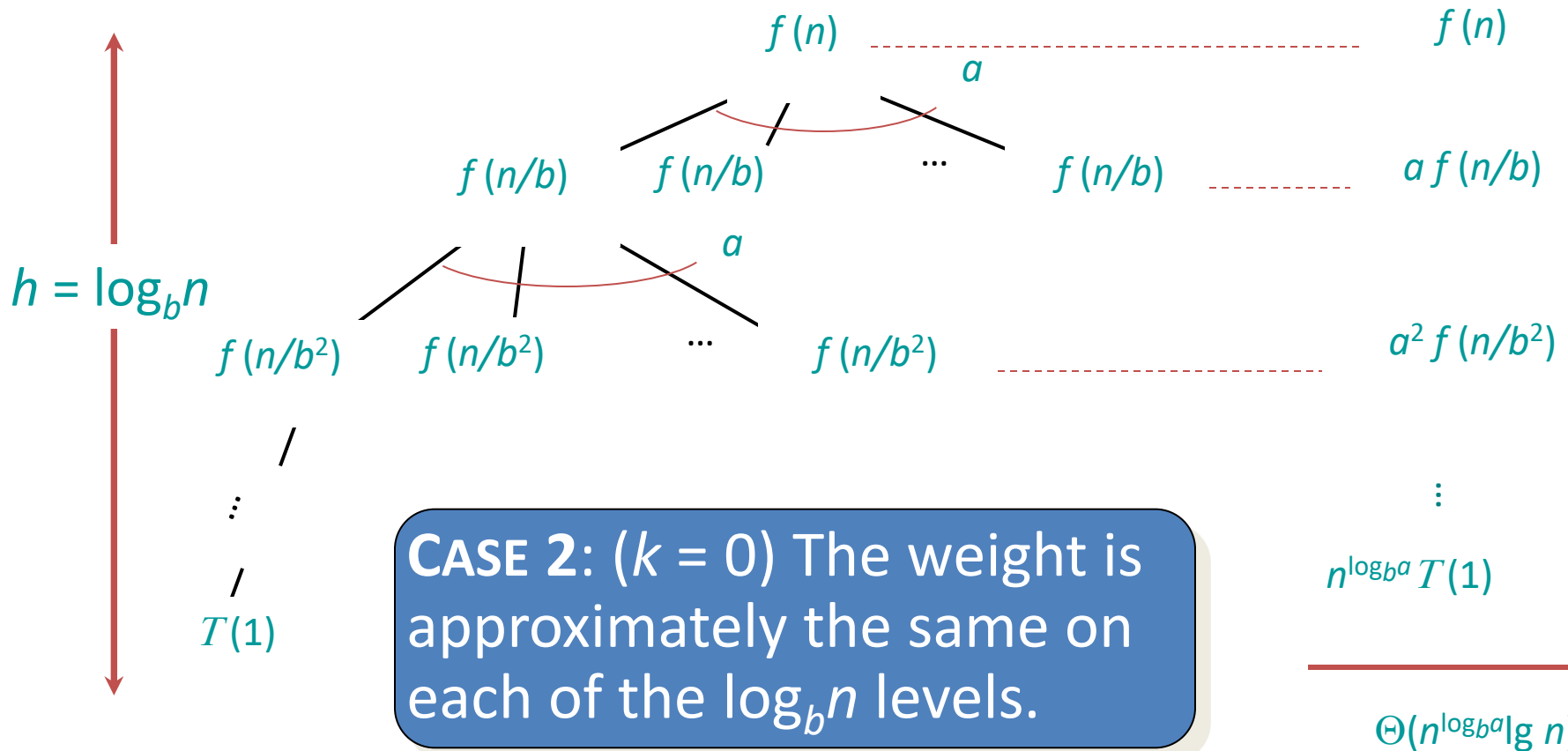
2. $f(n) = \Theta(n^{\log_b a} \lg^k n)$ for some constant $k \geq 0$.

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

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

Idea of master theorem

Recursion tree:



Three common cases (cont.)

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

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

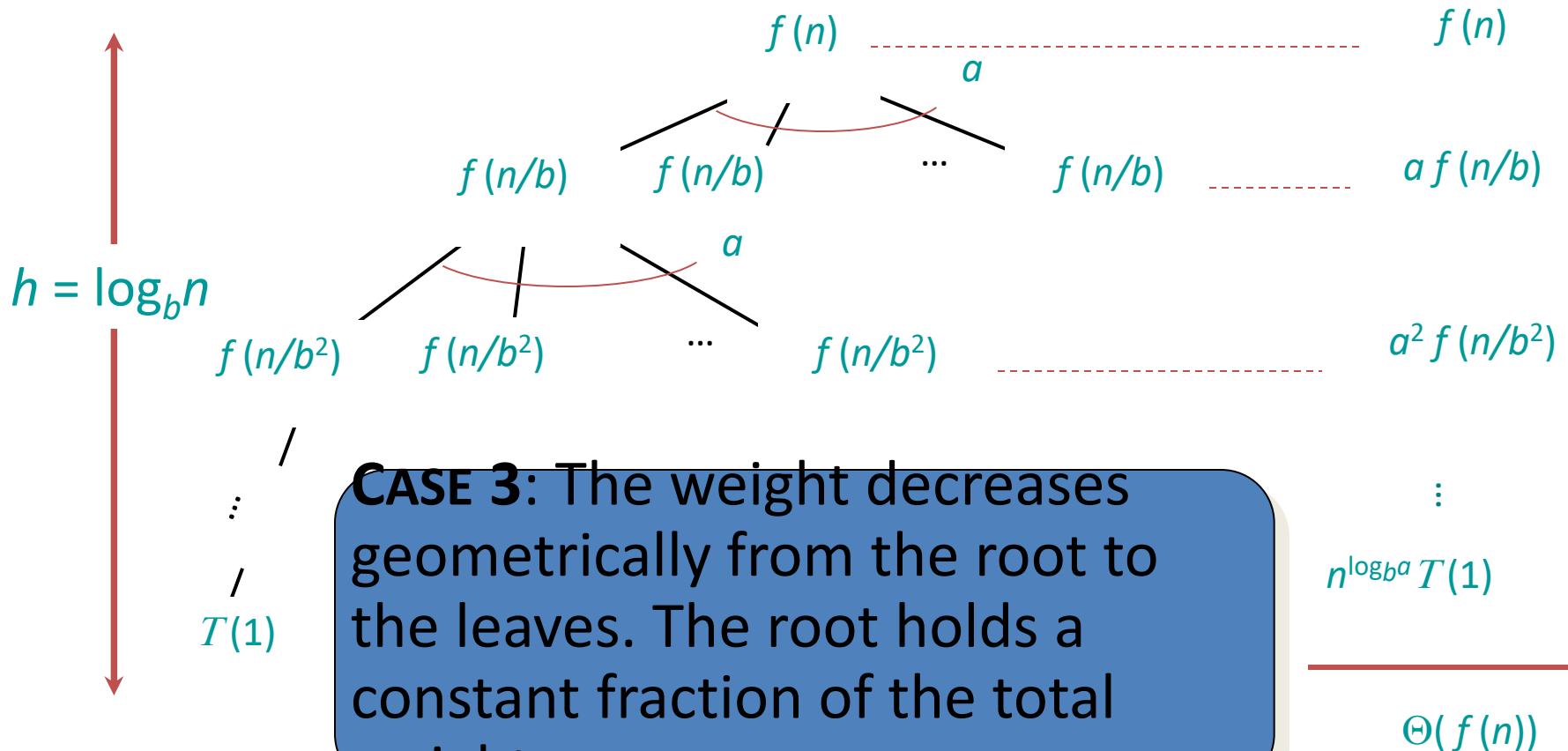
- $f(n)$ grows polynomially faster than $n^{\log_b a}$ (by an n^ε factor),

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

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

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.

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: $f(n) = \Omega(n^{2+\epsilon})$ for $\epsilon = 1$

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

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

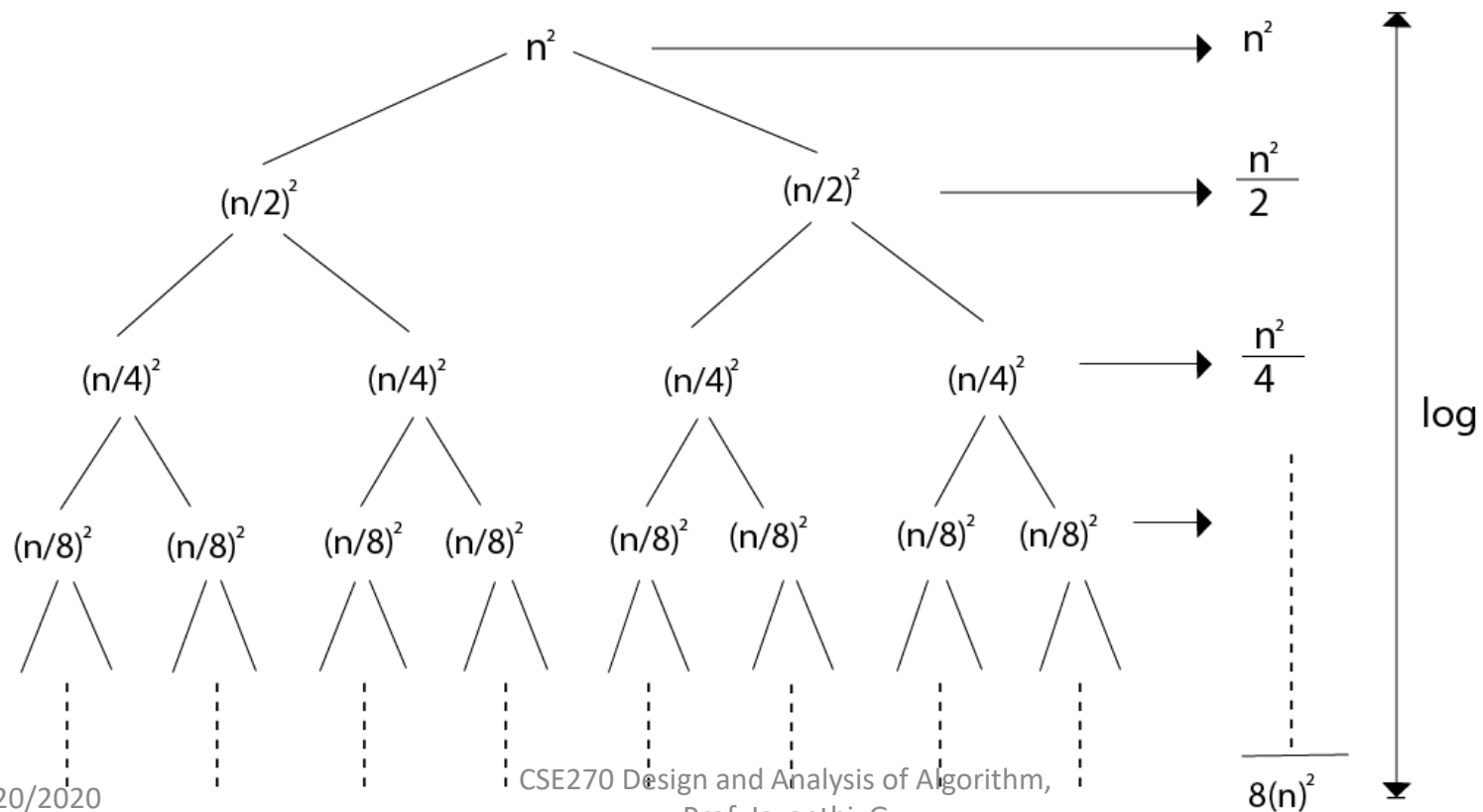
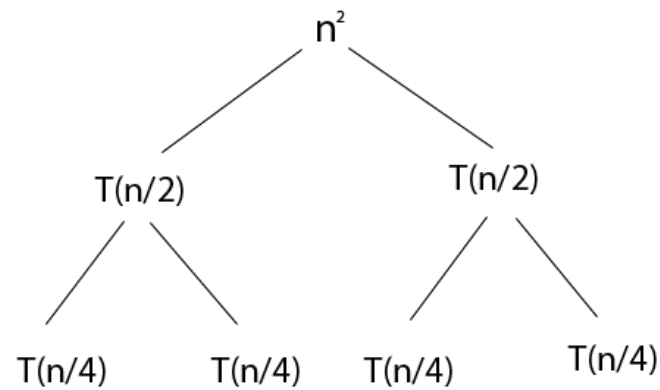
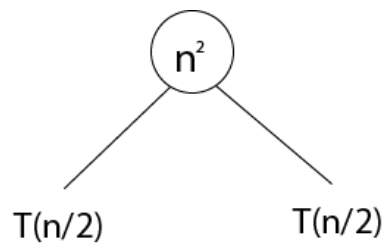
Ex. $T(n) = 4T(n/2) + n^2/\lg n$

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

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

Recursion tree

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



Cost of levels

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

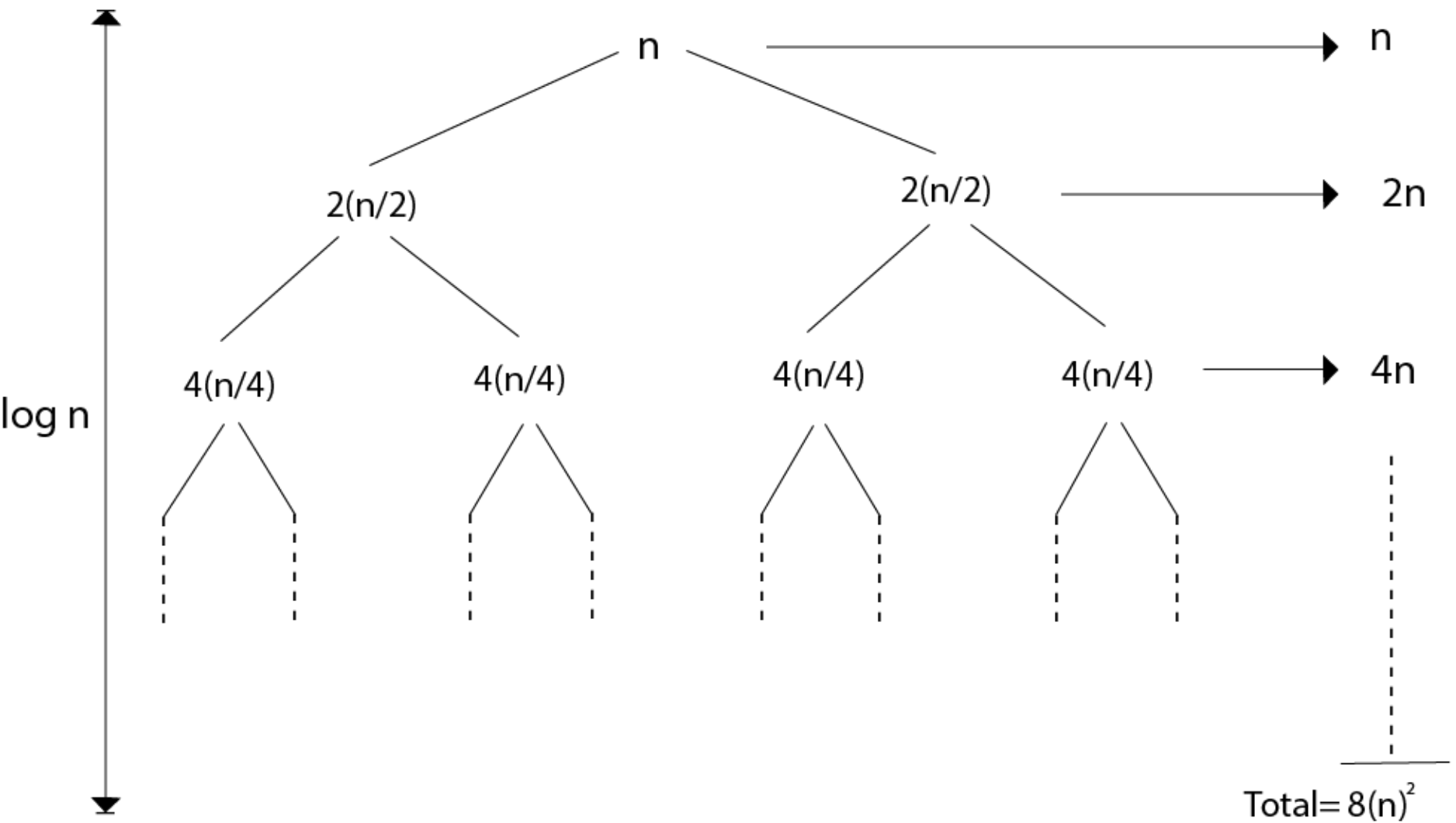
$$\leq n^2 \sum_{i=0}^{\infty} \left(\frac{1}{2^i} \right)$$

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

$$\mathbf{T(n) = \theta n^2}$$

Recursion Tree

$$T(n) = 4T_{\frac{n}{2}} + n$$



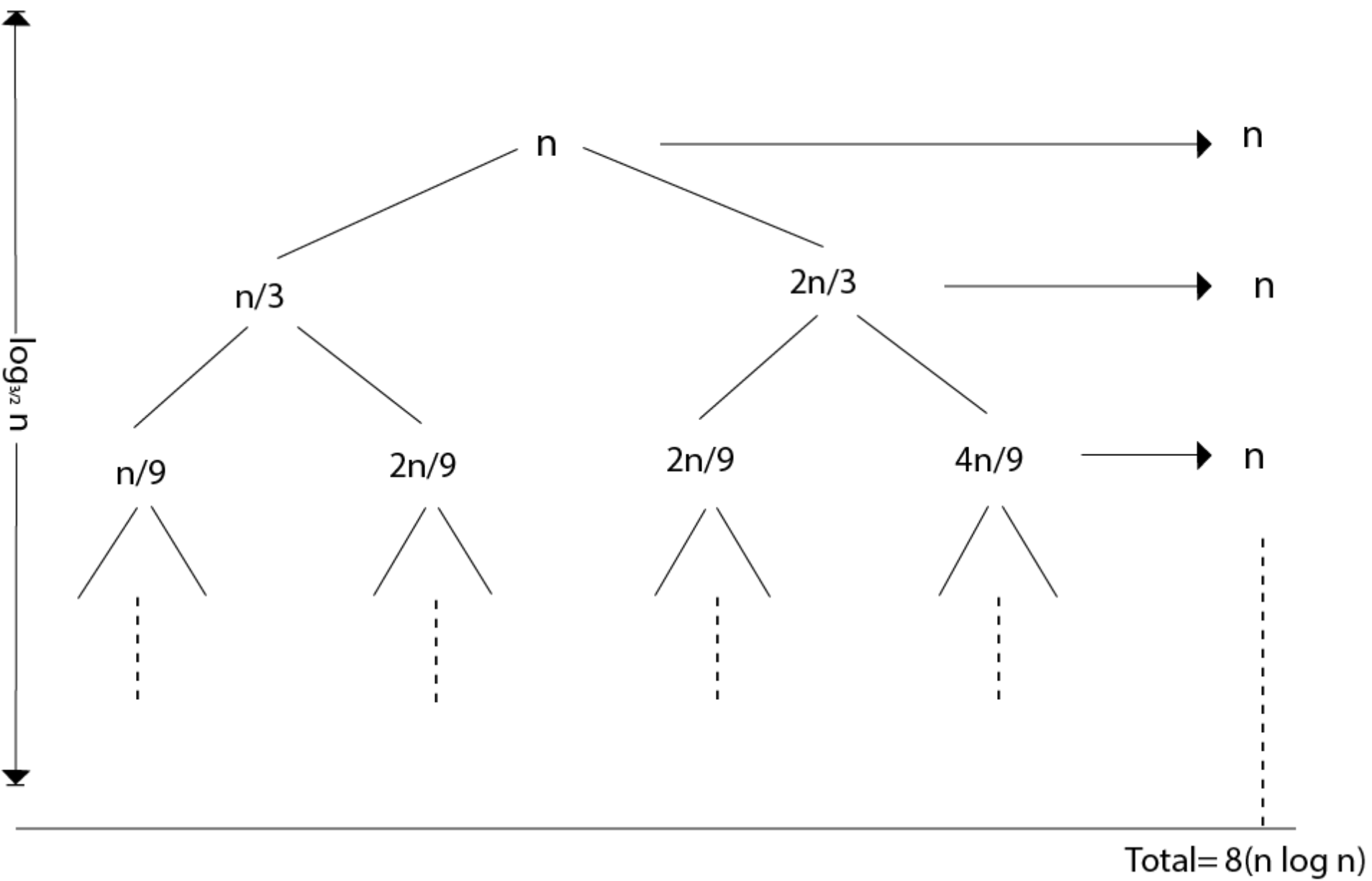
We have $n + 2n + 4n + \dots \log_2 n$ times

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

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

$$\mathbf{T(n) = \theta(n^2)}$$

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



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

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

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

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