



Design and Analysis of Algorithms

Recurrence

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Topics

- **Induction**
- **Substitution Method**
- **Recursion-Tree Method**
- **Master Method**



Induction

Induction used to prove that a statement $T(n)$ holds for all integers n :

- Base case: prove $T(0)$
- Assumption: assume that $T(n-1)$ is true
- Induction step: prove that $T(n-1)$ implies $T(n)$ for all $n > 0$

Strong induction: when we assume $T(k)$ is true for ***all*** $k \leq n - 1$ and use this in proving $T(n)$



Integer Multiplication

Let X and Y be n bit integers. $X = \boxed{A|B}$ and $Y = \boxed{C|D}$ where A , B , C , and D are $n/2$ bit integers.

Simple Method:
$$XY = (A2^{\frac{n}{2}} + B)(C2^{\frac{n}{2}} + D)$$
$$= AC2^n + (AD + BC)2^{\frac{n}{2}} + BD$$

Running Time Recurrence:
$$T(n) = 4T\left(\frac{n}{2}\right) + bn$$

How do we solve it?



Induction

The most general strategy:

Guess: the form of the solution.

Verify: by induction.

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

Base case $T(1) = \Theta(1)$.

Guess $O(n^3)$.

Assume that $T(k) \leq ck^3$ for $k < n$.

Prove $T(n) \leq cn^3$ by induction.



Induction

$$\begin{aligned}T(n) &= 4T\left(\frac{n}{2}\right) + bn \\&\leq 4c\left(\frac{n}{2}\right)^3 + bn \\&= \left(\frac{c}{2}\right)n^3 + bn \\&= cn^3 - \left(\left(\frac{c}{2}\right)n^3 - bn\right) \\&\leq cn^3\end{aligned}$$

$$T(k) \leq ck^3 \text{ for } k < n$$

For example, if $c \geq 2b$, then $\left(\frac{c}{2}\right)n^3 - bn \geq 0$.

This bound is not tight!



Induction

We also try that $T(n) = O(n^2)$.

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

$$\begin{aligned} T(n) &= 4T\left(\frac{n}{2}\right) + bn \\ &\leq 4c\left(\frac{n}{2}\right)^2 + bn \\ &= cn^2 + bn \\ &\leq cn^2 \text{ X} \end{aligned}$$



A Tighter Upper Bound

Strengthen the inductive hypothesis.

Subtract a low-order term.

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

$$\begin{aligned} T(n) &= 4T\left(\frac{n}{2}\right) + bn \\ &\leq 4\left(c_1 \left(\frac{n}{2}\right)^2 - c_2 \left(\frac{n}{2}\right)\right) + bn \\ &= c_1 n^2 - 2c_2 n + bn \\ &= c_1 n^2 - c_2 n - (c_2 n - bn) \\ &\leq c_1 n^2 - c_2 n \end{aligned}$$

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

For example, if $c_2 \geq b$, then $c_2 n - bn \geq 0$.



Example of Substitution

Use algebraic manipulation to make an unknown recurrence similar to what you have seen before.

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

Set $m = \log n$ and we have $T(2^m) = 2T(2^{m/2}) + m$

Set $S(m) = T(2^m)$ and we have $S(m) = 2S(m/2) + m$

$\rightarrow S(m) = O(m \log m)$

As a result, we have $T(n) = O(\log n \log \log n)$



A Useful Recurrence Relation

- $T(n)$ = max number of compares to Merge-Sort a list of size $\leq n$
- $T(n)$ is monotone nondecreasing.

Merge-Sort recurrence

$$T(n) \leq \begin{cases} 0, & \text{if } n = 1 \\ T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + n, & \text{otherwise} \end{cases}$$

Solution. $T(n)$ is $O(n \log n)$

Assorted proofs. We describe several ways to solve this recurrence. Initially we assume n is a power of 2 and replace “ \leq ” with “ $=$ ” in the recurrence.



Proof by Induction

If $T(n)$ satisfies the following recurrence, then $T(n)$ is $O(n \log n)$.

$$T(n) = \begin{cases} 0, & \text{if } n = 1 \\ 2T(n/2) + n, & \text{otherwise} \end{cases}$$

assuming n is a power of 2

- **Base case:** when $n = 1$, $T(1) = 0 = n \log n$.
- **Inductive hypothesis:** assume $T(n) = n \log n$.
- **Goal:** show that $T(2n) = 2n \log(2n)$

$$\begin{aligned} T(2n) &= 2T(n) + 2n \\ &= 2n \log n + 2n \\ &= 2n(\log(2n) - 1) + 2n \\ &= 2n \log(2n) \end{aligned}$$



Analysis of Merg-Sort Recurrence

If $T(n)$ satisfies the following recurrence, then $T(n) \leq n \lceil \log n \rceil$.

$$T(n) \leq \begin{cases} 0, & \text{if } n = 1 \\ T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + n, & \text{otherwise} \end{cases}$$

- Base case: $n=1$, $T(1) = 0$.
- Define: $n_1 = \lceil n/2 \rceil$ and $n_2 = \lfloor n/2 \rfloor$.
- Induction step: assume true for $1, 2, \dots, n-1$.



$$\begin{aligned} T(n) &\leq T(n_1) + T(n_2) + n \\ &\leq n_1 \lceil \log_2 n_1 \rceil + n_2 \lceil \log_2 n_2 \rceil + n \\ &\leq n_1 \lceil \log_2 n_2 \rceil + n_2 \lceil \log_2 n_2 \rceil + n \\ &= n \lceil \log_2 n_2 \rceil + n \\ &\leq n (\lceil \log_2 n \rceil - 1) + n \\ &= n \lceil \log_2 n \rceil \end{aligned}$$

$$\begin{aligned} n_2 &= \lfloor n/2 \rfloor \\ &\leq \lceil 2^{\lceil \log_2 n \rceil} / 2 \rceil \\ &= 2^{\lceil \log_2 n \rceil} / 2 \end{aligned}$$

$$\log_2 n_2 \leq \lceil \log_2 n \rceil - 1$$

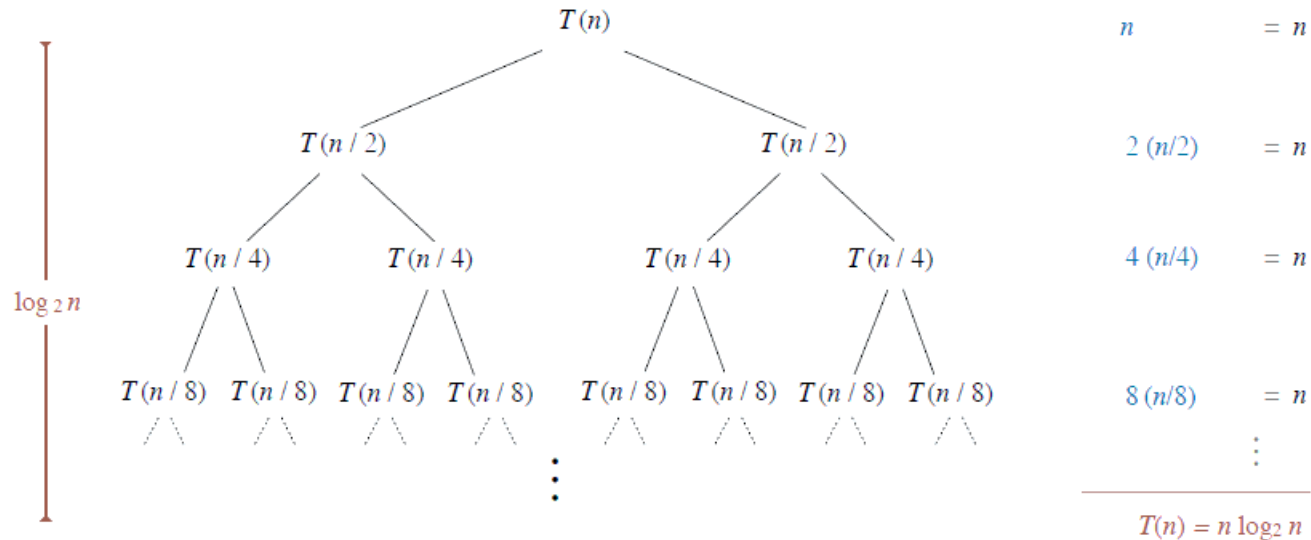


Recursion Tree

If $T(n)$ satisfies the following recurrence, then $T(n)$ is $O(n \log n)$.

$$T(n) = \begin{cases} 0, & \text{if } n = 1 \\ 2T(n/2) + n, & \text{otherwise} \end{cases}$$

assuming n is a power of 2





Example of Recursion Tree

Solve $T(n) = 3T(n/4) + n^2$:



Example of Recursion Tree

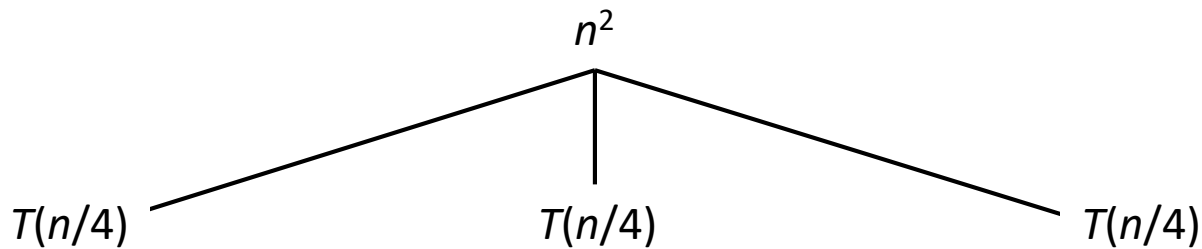
Solve $T(n) = 3T(n/4) + n^2$:

$$T(n)$$



Example of Recursion Tree

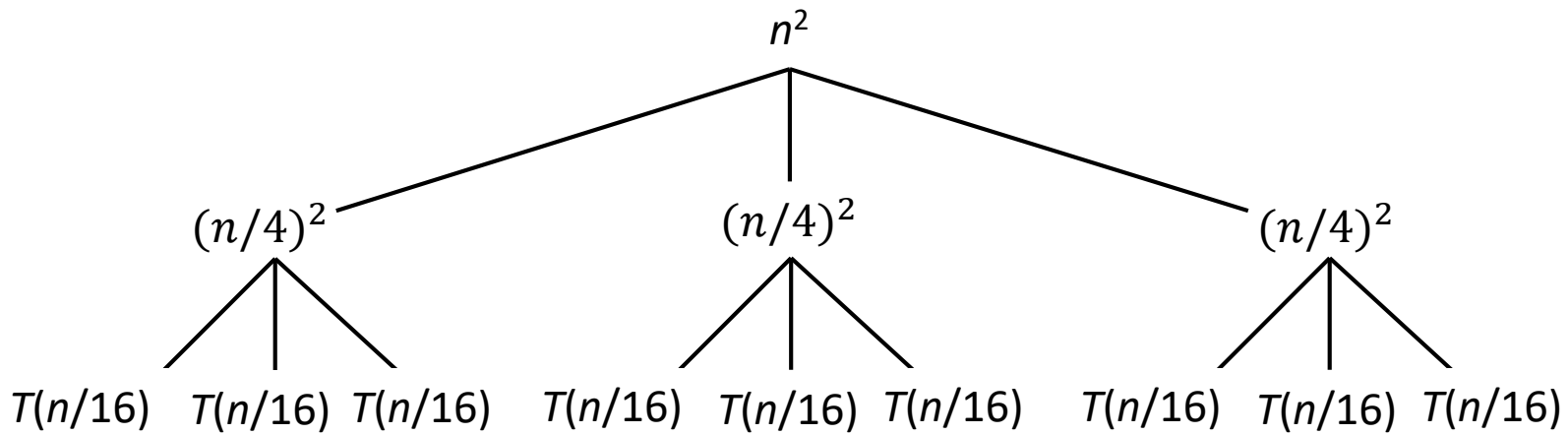
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Example of Recursion Tree

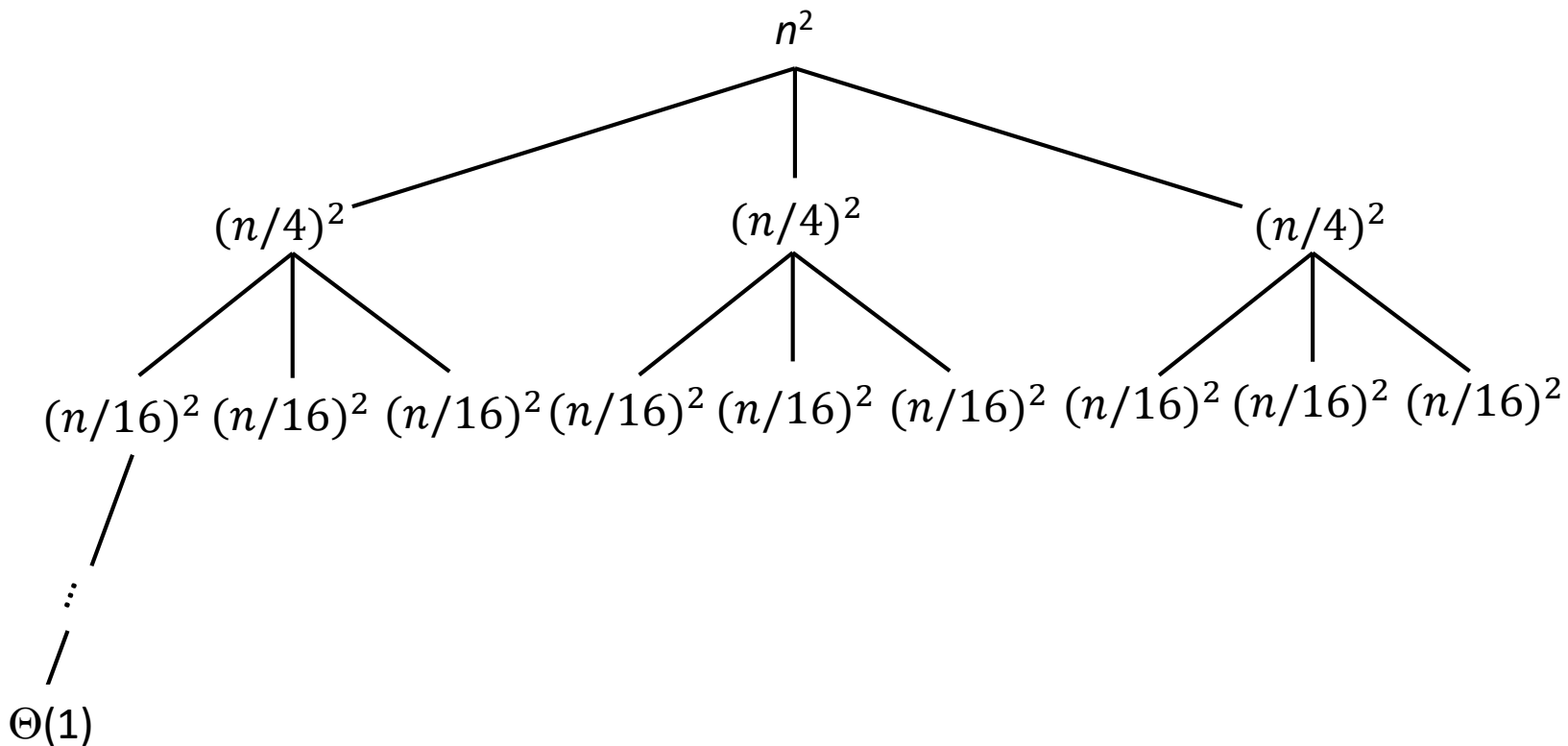
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Example of Recursion Tree

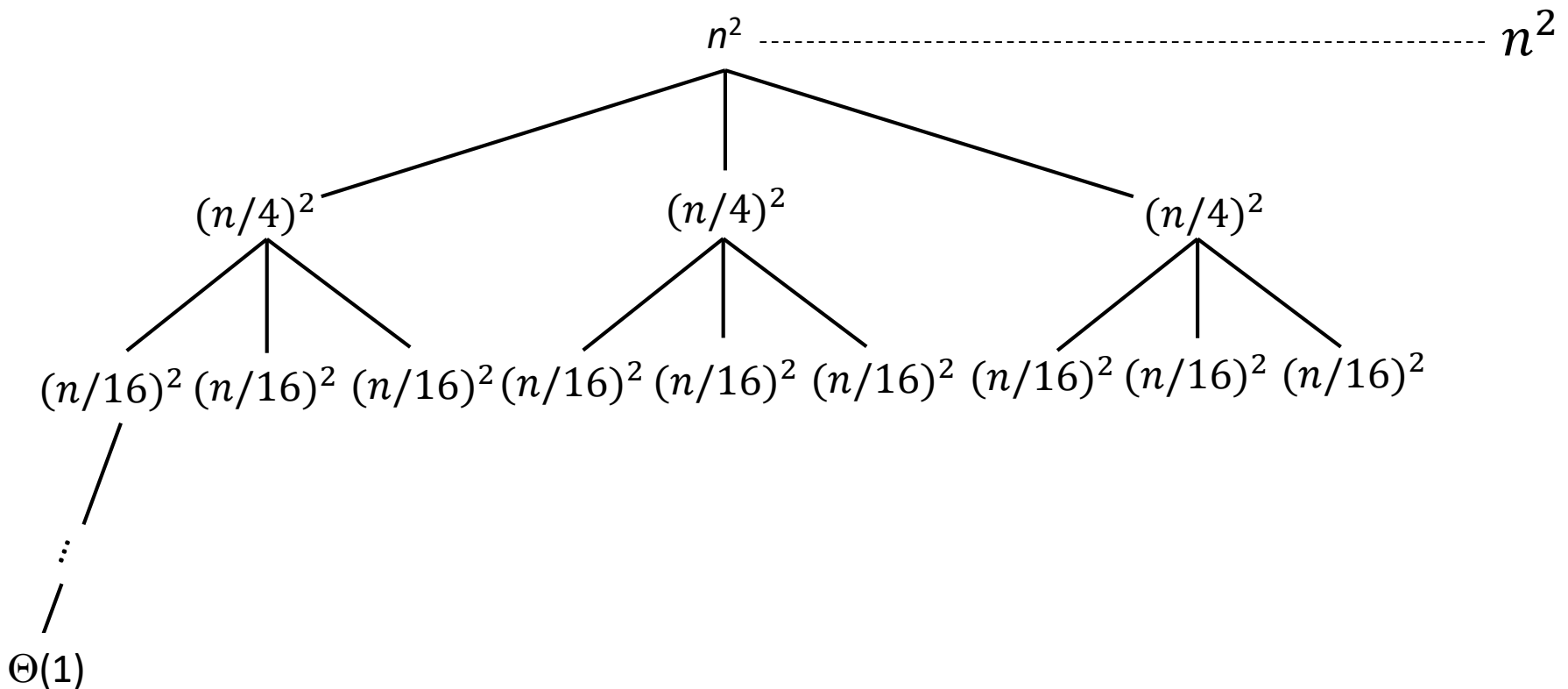
Solve $T(n) = 3T(n/4) + n^2$:





Example of Recursion Tree

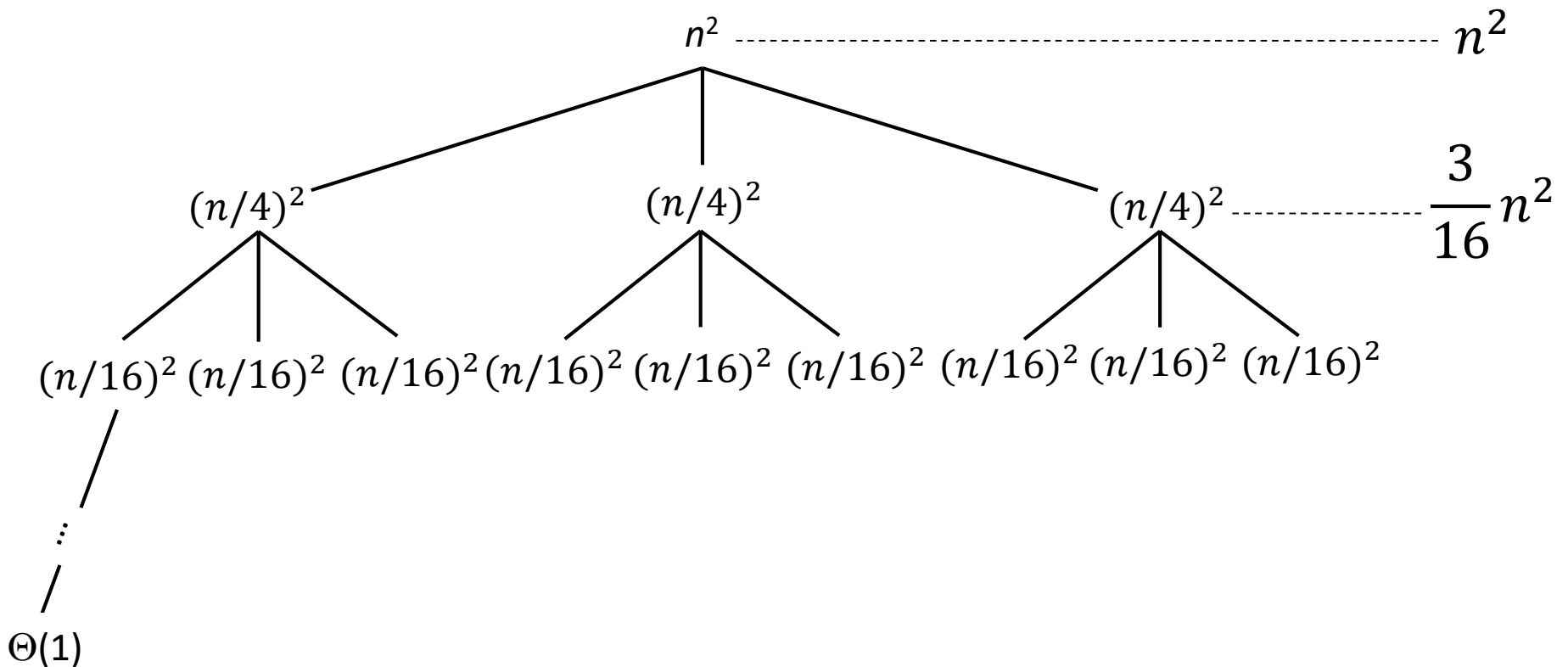
Solve $T(n) = 3T(n/4) + n^2$:





Example of Recursion Tree

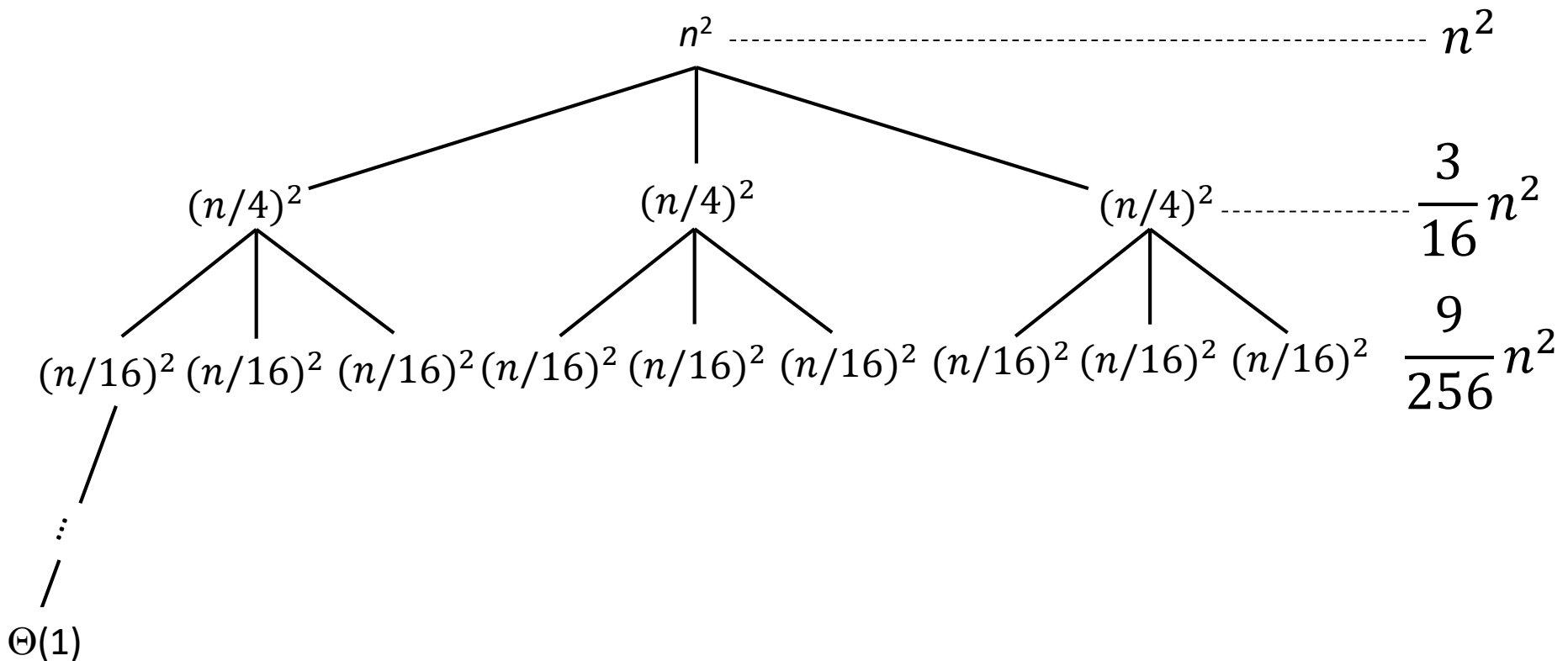
Solve $T(n) = 3T(n/4) + n^2$:





Example of Recursion Tree

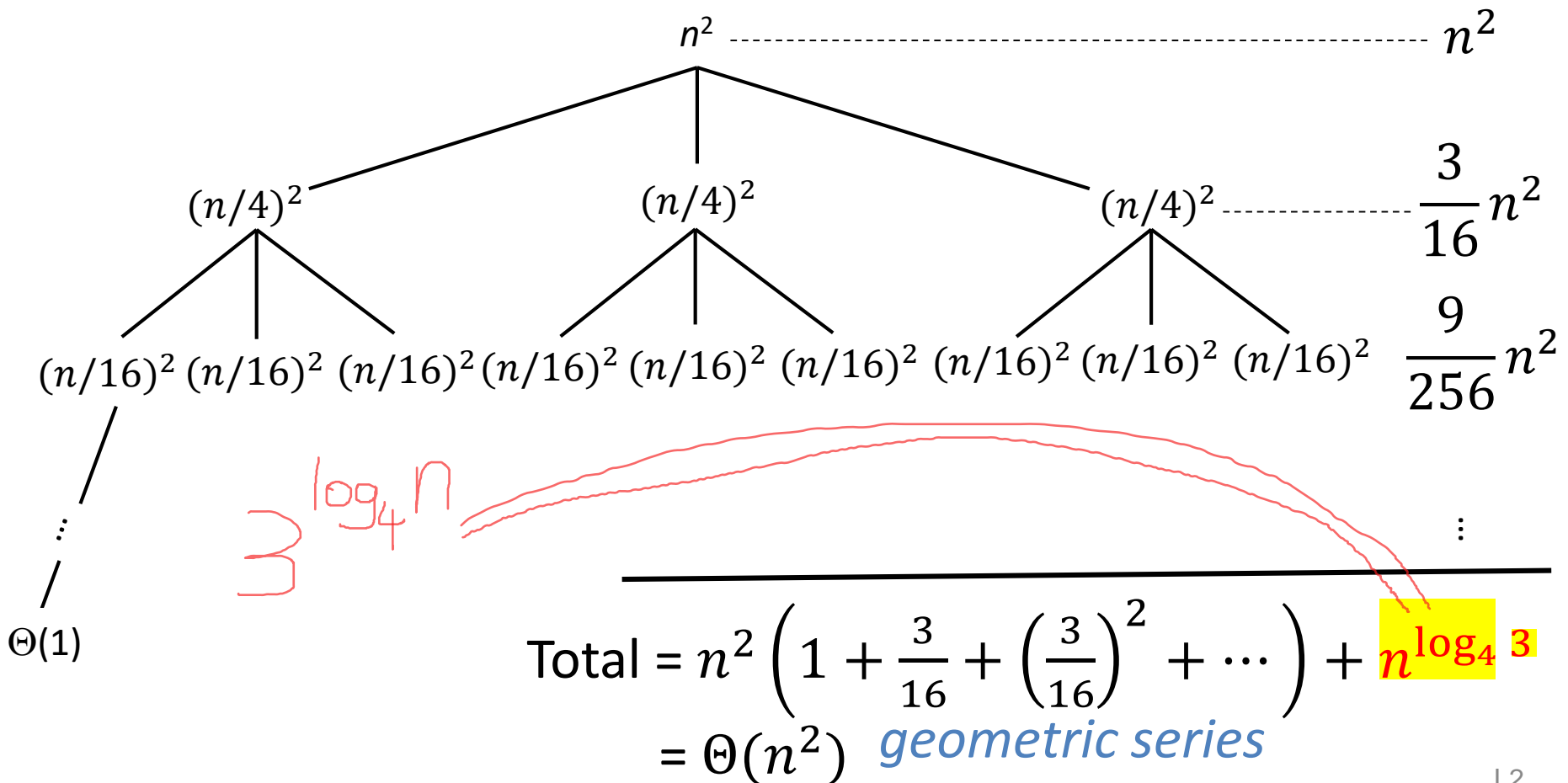
Solve $T(n) = 3T(n/4) + n^2$:





Example of Recursion Tree

Solve $T(n) = 3T(n/4) + n^2$:



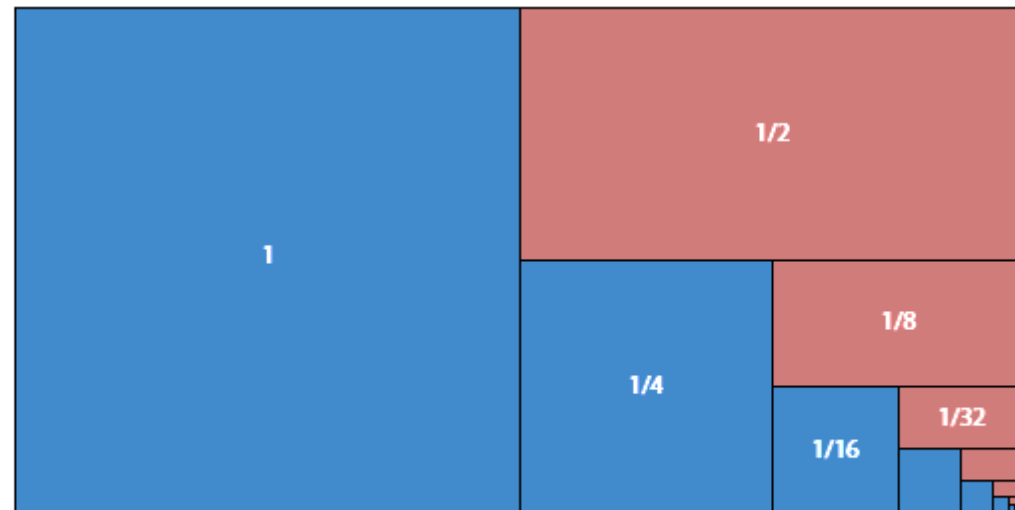


Geometric Series

Fact 1. For $r \neq 1$, $1 + r + r^2 + r^3 + \dots + r^{k-1} = \frac{1 - r^k}{1 - r}$

Fact 2. For $r = 1$, $1 + r + r^2 + r^3 + \dots + r^{k-1} = k$

Fact 3. For $r < 1$, $1 + r + r^2 + r^3 + \dots = \frac{1}{1 - r}$



$$1 + 1/2 + 1/4 + 1/8 + \dots = 2$$



Master Method

Goal. Recipe for solving common divide-and-conquer recurrences:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

With $T(0) = 0$ and $T(1) = \Theta(1)$.

Terms.

- $a \geq 1$ is the (integer) number of subproblems.
- $b > 1$ is the (integer) factor by which the subproblem size decreases.
- $f(n)$ = work to divide and combine subproblems.

Recursion tree.

- Number of levels:
- Number of subproblems at level i :
- Size of subproblem at level i :
- Number of leaves:



Master Method

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$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

With $T(0) = 0$ and $T(1) = \Theta(1)$.

Terms.

- $a \geq 1$ is the (integer) number of subproblems.
- $b > 1$ is the (integer) factor by which the subproblem size decreases.
- $f(n)$ = work to divide and combine subproblems.

Recursion tree.

- Number of levels: $k = \log_b n$.
- Number of subproblems at level i : a^i .
- Size of subproblem at level i : n/b^i .
- Number of leaves: $n^{\log_b a}$.



Master Theorem

Master Theorem. Suppose that $T(n)$ is a function on the non-negative integers that satisfies the recurrence:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

with $T(0) = 0$ and $T(1) = \Theta(1)$, where n/b means either $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. Then,

Case 1. If $f(n) = O(n^k)$ for some constant $k < \log_b a$, then $T(n) = \Theta(n^{\log_b a})$.

Ex. $T(n) = 3T(n/2) + 5n$

$a = 3, b = 2, f(n) = 5n, k = 1, \log_b a = 1.58$

$T(n) = \Theta(n^{\log_2 3})$



Master Theorem

Master Theorem. Suppose that $T(n)$ is a function on the non-negative integers that satisfies the recurrence:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

with $T(0) = 0$ and $T(1) = \Theta(1)$, where n/b means either $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. Then,

Case 2. If $f(n) = \Theta(n^k \log^p n)$ for $p \geq 0$ and $k = \log_b a$, then $T(n) = \Theta(n^k \log^{p+1} n)$.

多一项式, 次数同 \Rightarrow 加 $\log n$

Ex. $T(n) = 2T(n/2) + 17n \log n$

$a = 2, b = 2, f(n) = 17n \log n, k = 1, p = 1, \log_b a = 1$

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



Master Theorem

Master Theorem. Suppose that $T(n)$ is a function on the non-negative integers that satisfies the recurrence:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

with $T(0) = 0$ and $T(1) = \Theta(1)$, where n/b means either $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. Then,

Case 3. If $f(n) = \Omega(n^k)$ for some constant $k > \log_b a$, and if $af(n/b) \leq cf(n)$ for some constant $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$.

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

$a = 3, b = 2, f(n) = n^2, k = 2, \log_b a = 1.58$

Regularity condition: $3(n/2)^2 \leq cn^2$ for $c = 3/4$

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



Master Theorem

Master Theorem. Suppose that $T(n)$ is a function on the non-negative integers that satisfies the recurrence:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

with $T(0) = 0$ and $T(1) = \Theta(1)$, where n/b means either $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$.

Case 1. If $f(n) = O(n^k)$ for some constant $k < \log_b a$, then $T(n) = \Theta(n^{\log_b a})$.

Case 2. If $f(n) = \Theta(n^k \log^p n)$ for $p \geq 0$ and $k = \log_b a$, then $T(n) = \Theta(n^k \log^{p+1} n)$.

Case 3. If $f(n) = \Omega(n^k)$ for some constant $k > \log_b a$, and if $af(n/b) \leq cf(n)$ for some constant $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$.



Master Theorem Need Not Apply

Gaps in master theorem

- Number of subproblems must be a constant.

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

- Number of subproblems must be ≥ 1 .

$$T(n) = \frac{1}{2}T(n/2) + n^2$$

- Non-polynomial separation between $f(n)$ and $\log n$.

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

- $f(n)$ is not positive.

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

- Regularity condition does not hold.

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