

Recurrences

Definition

- A *recurrence* is an equation or inequality that describes a function in terms of its value on smaller inputs.
- Example from MERGE-SORT

$$T(n) = \begin{cases} \Theta(1) & \text{if } n=1 \\ 2T(n/2) + \Theta(n) & \text{if } n>1 \end{cases}$$

Technicalities

- Normally, independent variables only assume integral values
- Example from MERGE-SORT revisited

$$T(n) = \begin{cases} \Theta(1) & \text{if } n=1 \\ T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + \Theta(n) & \text{if } n>1 \end{cases}$$

- For simplicity, ignore floors and ceilings – often insignificant

Technicalities

- Boundary conditions (small n) are also glossed over

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

10 Value of $T(n)$ assumed to be small constant for small n

Recursion-Tree Method

- Straightforward technique of coming up with a good guess
- Can help the Substitution Method
- ***Recursion tree***: visual representation of recursive call hierarchy where each node represents the cost of a single subproblem

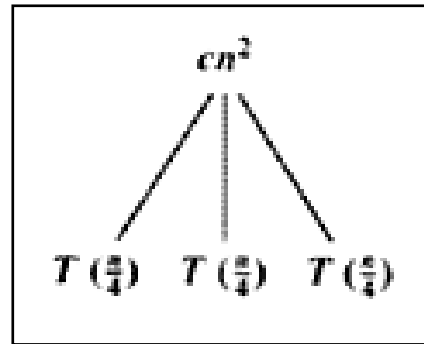
Recursion-Tree Method

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$

$$T(n)$$

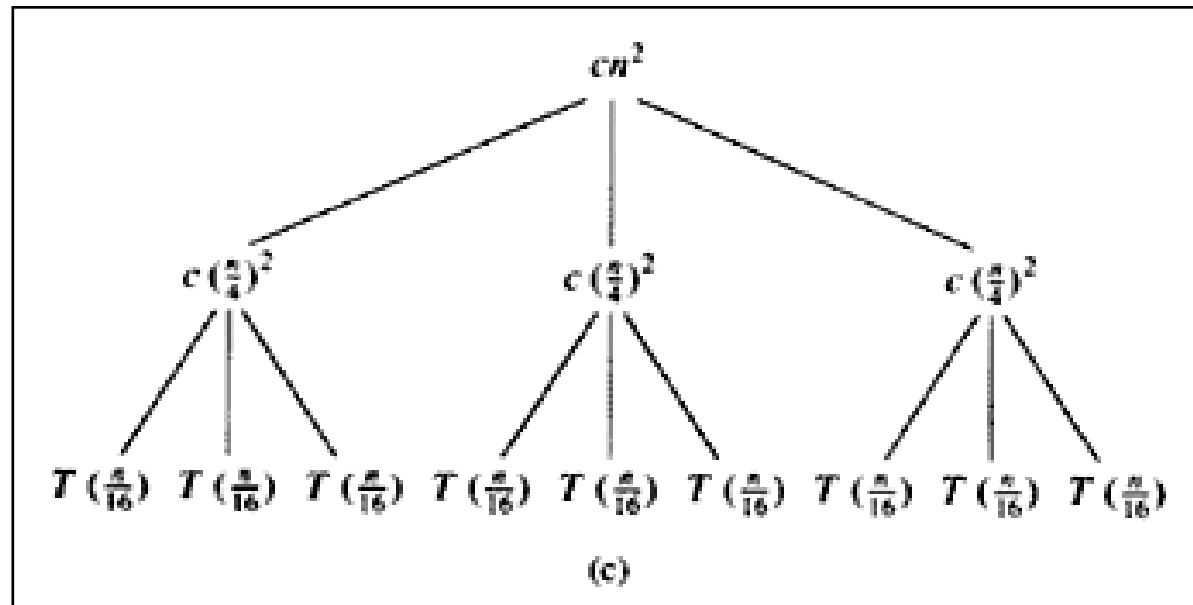
Recursion-Tree Method

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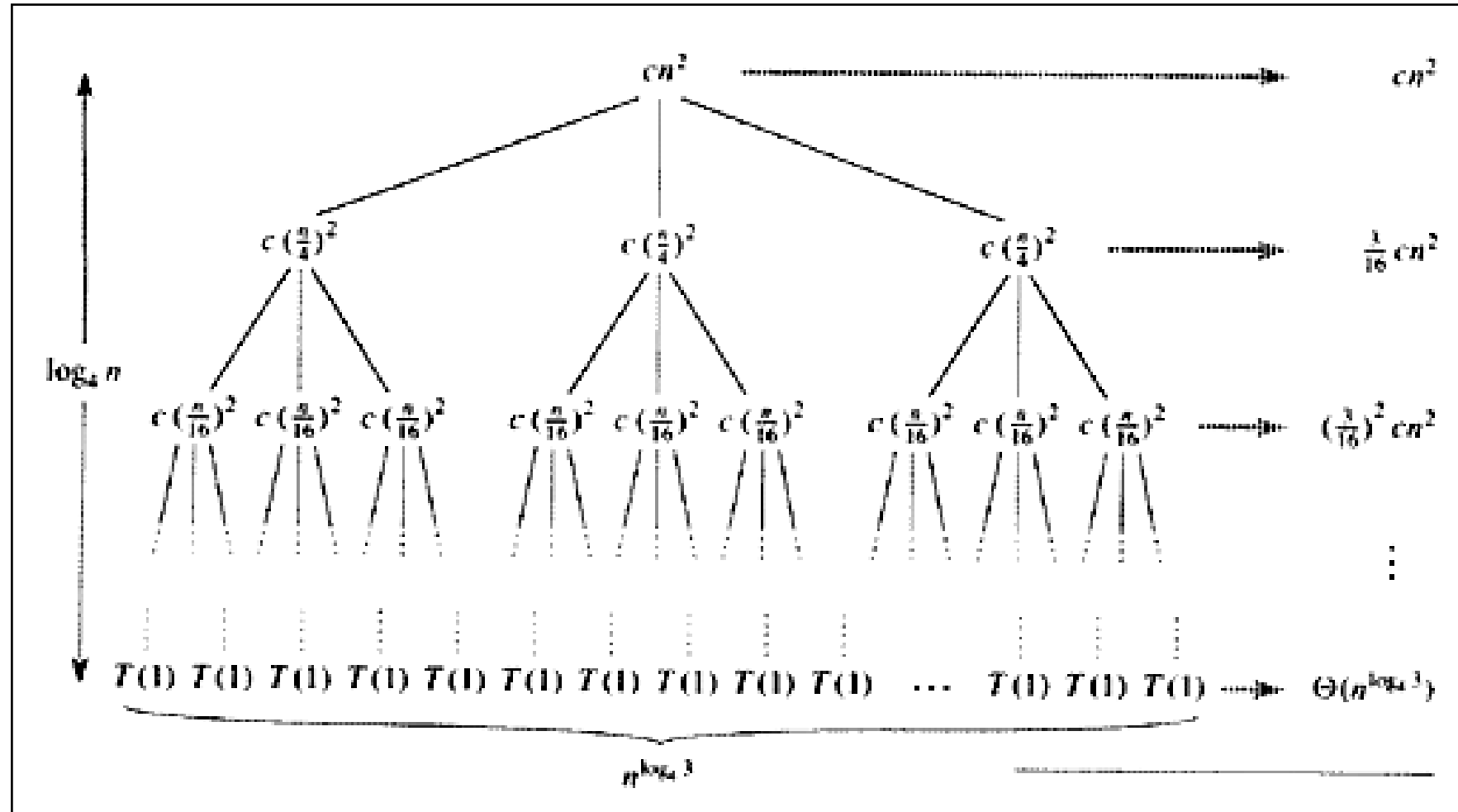


Recursion-Tree Method

$$T(n) = 3T(\lfloor n/4 \rfloor) + \Theta(n^2)$$



Recursion-Tree Method



Recursion-Tree Method

□ Gathering all the costs together:

$$T(n) = \sum_{i=0}^{\log_4 n - 1} (3/16)^i cn^2 + \Theta(n^{\log_4 3})$$

$$T(n) \leq \sum_{i=0}^{\infty} (3/16)^i cn^2 + o(n)$$

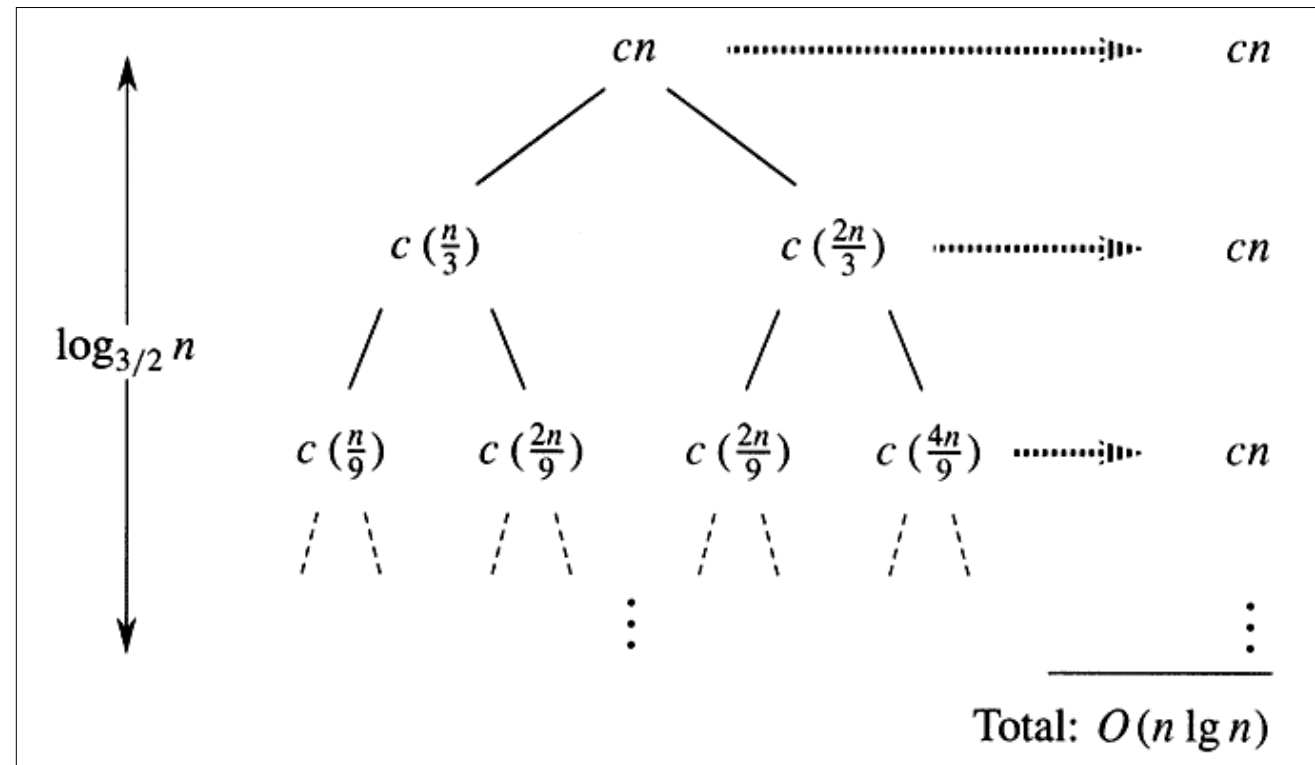
$$T(n) \leq (1/(1-3/16))cn^2 + o(n)$$

$$T(n) \leq (16/13)cn^2 + o(n)$$

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

Recursion-Tree Method

$$T(n) = T(n/3) + T(2n/3) + O(n)$$



Recursion-Tree Method

- An overestimate of the total cost:

$$T(n) = \sum_{i=0}^{\log_{3/2} n - 1} cn + \Theta(n^{\log_{3/2} 2})$$

- Counter-indications:

$$T(n) = O(n \lg n) + \omega(n \lg n)$$

- Notwithstanding this, use as “guess”:

$$T(n) = O(n \lg n)$$

Substitution method

The most general method:

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

Example: $T(n) = 4T(n/2) + 100n$

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

Example of substitution

$$\begin{aligned}T(n) &= 4T(n/2) + 100n \\&\leq 4c(n/2)^3 + 100n \\&= (c/2)n^3 + 100n \\&= cn^3 - ((c/2)n^3 - 100n) \quad \leftarrow \textit{desired} - \textit{residual} \\&\leq cn^3 \quad \leftarrow \textit{desired}\end{aligned}$$

whenever $(c/2)n^3 - 100n \geq 0$, for
example, if $c \geq 200$ 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.

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 $k < n$:

$$\begin{aligned} T(n) &= 4T(n/2) + 100n \\ &\leq cn^2 + 100n \\ &\leq cn^2 \end{aligned}$$

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 $k < n$.

$$\begin{aligned} T(n) &= 4T(n/2) + 100n \\ &\leq 4(c_1(n/2)^2 - c_2(n/2)) + 100n \\ &= c_1 n^2 - 2c_2 n + 100n \\ &= c_1 n^2 - c_2 n - (c_2 n - 100n) \\ &\leq c_1 n^2 - c_2 n \quad \text{if } c_2 > 100. \end{aligned}$$

Pick c_1 big enough to handle the initial conditions.

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

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

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

Examples

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

CASE 1: $f(n) = O(n^{2-\varepsilon})$ for $\varepsilon = 1$.
 $\therefore T(n) = \Theta(n^2).$

Ex. $T(n) = 8T(n/2) + n^2$

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

Three common cases

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

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

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

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

Examples

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

Ex. $T(n) = T(n/2) + c$ (Binary search)

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

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

$$af(n/b) \leq cf(n) \text{ for some constant } c < 1.$$

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

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

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

Ex. $T(n) = 9T(n/3) + n^3$

Examples

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 $n^\varepsilon = \omega(\lg n)$.

Reference:

<https://courses.csail.mit.edu/6.046/spring04/lectures/l2.ppt>