

Lecture 04 - Introduction to Algorithms Recursion – Part 2

Design and Analysis of Algorithms – IT1205

Year 02 Semester 02

Contents for today

- Finding a solution to a recurrence
 - Repeated Substitution method.
 - Recursion tree.
 - Master Theorem.

The Recursion-Tree Method

- In a **Recursion Tree**, each node represents the cost of a single sub problem somewhere in the set of recursive function invocations.

Steps in Recursion Tree Method

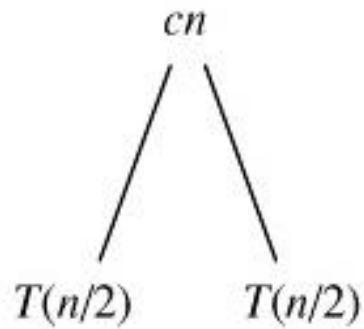
- Draw a recursive tree for given recurrence relation
- Calculate the cost at each level and count the total no of levels in the recursion tree.
- Count the total number of nodes in the last level and calculate the cost of the last level
- Sum up the cost of all the levels in the recursive tree

Examples

Let's try the Recursion Tree method

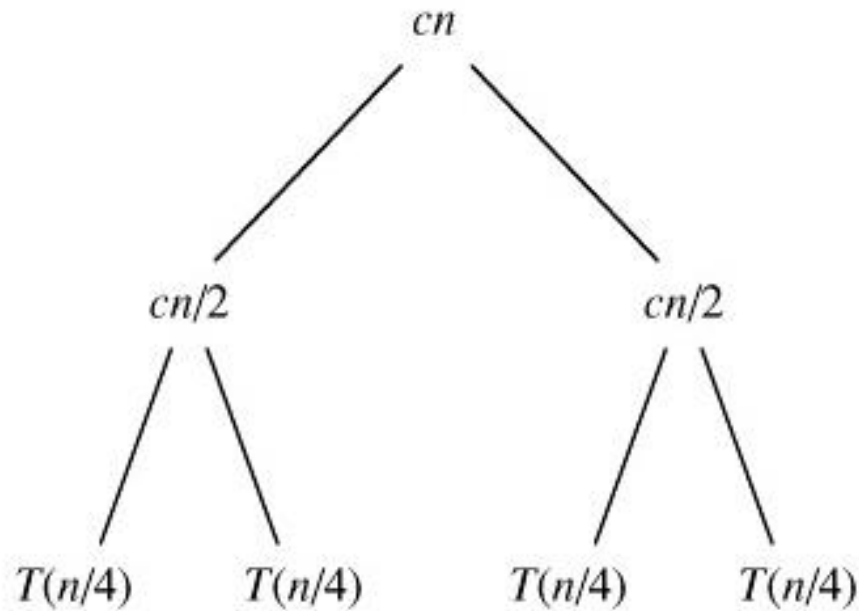
Recursion tree for $T(n) = 2T(n/2) + cn$

$T(n)$



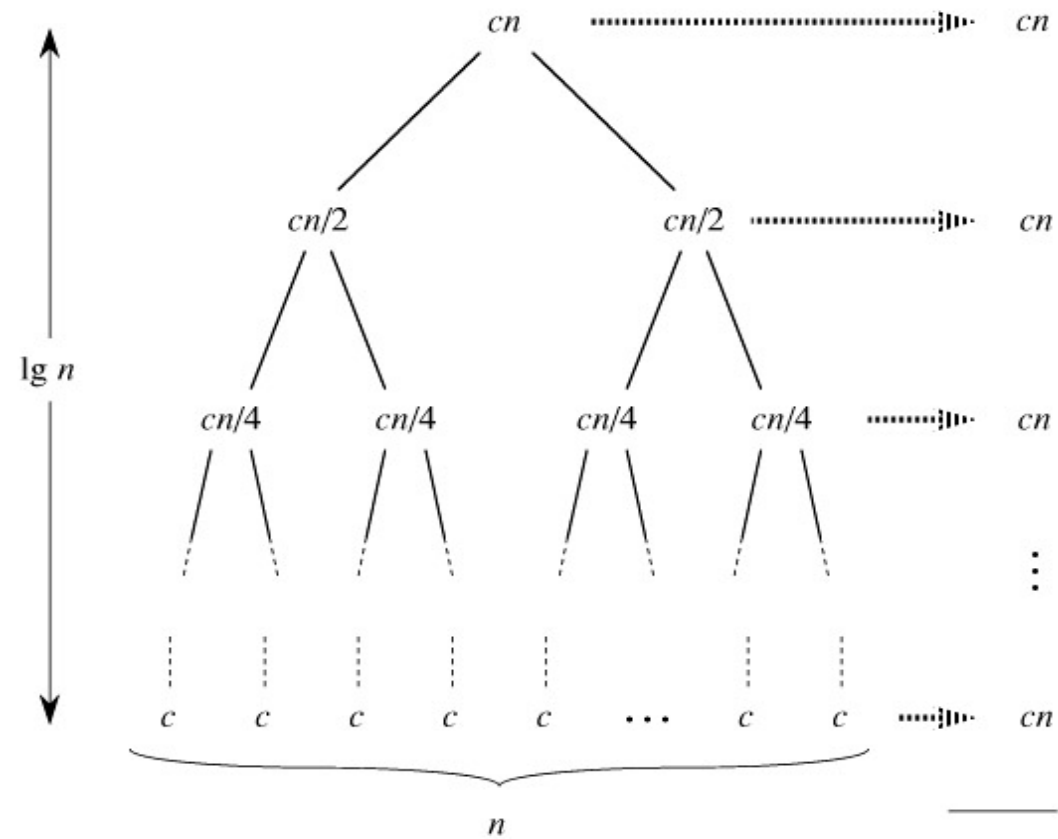
(a)

(b)



(c)

Recursion tree for $T(n) = 2T(n/2) + cn$



(d)

Total: $cn \lg n + cn$

The Master Method

- The Master method applies to recurrences of the form

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

where $a \geq 1$ and $b > 1$, and $f(n)$ is an asymptotically positive function.

The recurrence describes the running time of an algorithm that divides a problem of size n into a sub problems, each of size n/b , where a and b are positive constants. The a sub problems are solved recursively, each in time $T(n/b)$. The cost of dividing the problem and combining the results of the sub problems is described by the function $f(n)$.

The master theorem

- The master method depends on the following theorem.
- Let $a \geq 1$ and $b > 1$ be constants, let $f(n)$ be a function, and let $T(n)$ be defined on the nonnegative integers by the recurrence

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

where we interpret n/b to mean either $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. Then $T(n)$ can be bounded asymptotically as follows.

The Master Theorem

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

Case 1: $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$.
($f(n)$ is polynomially smaller than $n^{\log_b a}$.)
Solution: $T(n) = \Theta(n^{\log_b a})$.

Case 2: $f(n) = \Theta(n^{\log_b a} \lg^k n)$, where $k \geq 0$.

($f(n)$ is within a polylog factor of $n^{\log_b a}$, but not smaller.)

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

(Intuitively: cost is $n^{\log_b a} \lg^k n$ at each level, and there are $\Theta(\lg n)$ levels.)

Simple case: $k = 0 \Rightarrow f(n) = \Theta(n^{\log_b a}) \Rightarrow T(n) = \Theta(n^{\log_b a} \lg n)$.

Case 3: $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$ and $f(n)$ satisfies the regularity condition $af(n/b) \leq cf(n)$ for some constant $c < 1$ and all sufficiently large n .

($f(n)$ is polynomially greater than $n^{\log_b a}$.)

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

(Intuitively: cost is dominated by root.)

The master theorem

$$T(n) = \begin{cases} \Theta\left(n^{\log_b a}\right) & f(n) = O\left(n^{\log_b a - \varepsilon}\right) \rightarrow f(n) < n^{\log_b a} \\ \Theta\left(n^{\log_b a} \lg n\right) & f(n) = \Theta\left(n^{\log_b a}\right) \rightarrow f(n) = n^{\log_b a} \\ \Theta(f(n)) & \begin{aligned} &f(n) = \Omega\left(n^{\log_b a + \varepsilon}\right) \rightarrow f(n) > n^{\log_b a} \\ &\text{if } af(n/b) \leq cf(n) \text{ for } c < 1 \text{ and large } n \end{aligned} \end{cases}$$

Master Theorem

Give tight asymptotic bound for

$$T(n) = 9T(n/3) + n$$

Solution:

$a=9$, $b=3$, and $f(n) = n$.

$$n^{\log_b a} = n^{\log_3 9} = n^2$$

$f(n) = O(n^{\log_3 9 - \varepsilon})$ for $\varepsilon = 1$ or $f(n) < n^{\log_3 9} \rightarrow \text{case 1}$

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

Master Theorem

Give tight asymptotic bound for

$$T(n) = T(2n/3) + 1$$

Solution:

$a=1$, $b=3/2$, and $f(n) = 1$.

$$n^{\log_b a} = n^{\log_{3/2} 1} = n^0 = 1$$

$f(n) = \Theta(n^{\log_b a})$ or $f(n) = n^{\log_b a} \rightarrow \text{case 2}$

$$\therefore T(n) = \Theta(\log n)$$

Master Theorem

- Give tight asymptotic bound for

$$T(n) = 3T(n/4) + n \log n$$

Solution:

$a=3$, $b=4$, and $f(n) = n \log n$

$$n^{\log_b a} = n^{\log_4 3} = O(n^{0.793})$$

$f(n) = \Omega(n^{\log_4 3 + \varepsilon})$, for $\varepsilon \approx 0.2$ or $f(n) > n^{\log_4 3} \rightarrow$ case 3

$$\text{Note : } n \lg n \geq c \cdot n^{\log_4 3} \cdot n^{0.2}$$

$$\therefore T(n) = \Theta(n \log n)$$

