

Introduction to Algorithms

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❖ Chapter.1 Foundations

1. The Role of Algorithms in Computing

- 1) Algorithms
- 2) Algorithms as a technology

2. Getting Started

- 1) Insertion sort
- 2) Analyzing algorithms
- 3) Designing algorithms

3. Growth of Functions

- 1) Asymptotic notation
- 2) Standard notations and common functions

4. Divide-and-Conquer

- 1) The maximum-subarray problem
- 2) Strassen's algorithm for matrix multiplication
- 3) The substitution method for solving recurrences
- 4) The recursion-tree method for solving recurrences
- 5) The master method for solving recurrences
- 6) Proof of the master theorem

5. Probabilistic Analysis and Randomized Algorithms

- 1) The hiring problem
- 2) Indicator random variables
- 3) Randomized algorithms
- 4) Probabilistic analysis and further uses of indicator random variables

❖ The Introduction for solving recurrences

• Introduction

➤ Now that we have seen how recurrences characterize the running times of divide-and-conquer algorithms

✓ Merge sort, Maximum-subarray problem, Strassen's algorithm

✓ Ex)
$$\begin{aligned} T(n) &= \Theta(1) + 2T(n/2) + \Theta(n) + \Theta(1) \\ &= 2T(n/2) + \Theta(n) . \end{aligned}$$

➤ From this chapter, learn how to solve recurrences

✓ Substitution method

✓ Recursion-tree method

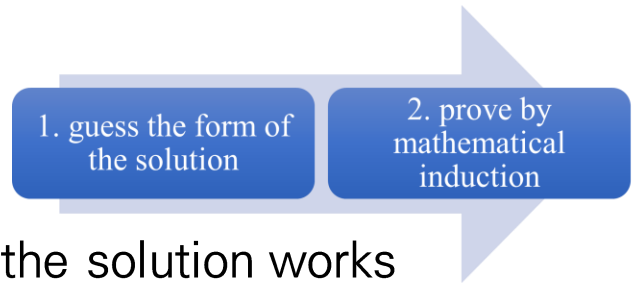
✓ Master method ★

✓ Ex)
$$\begin{aligned} T(n) &= \Theta(1) + 2T(n/2) + \Theta(n) + \Theta(1) \\ &= 2T(n/2) + \Theta(n) . \end{aligned} \quad \longrightarrow \quad T(n) = \theta(n \log n)$$

❖ The substitution method for solving recurrences

- Concept

- When we solve the recurrence function, substitute the guessed solution for the function and check if the guessed solution is true or not
- The substitution method comprises two steps
 - ✓ Guess the form of the solution
 - ✓ Use mathematical induction to find the constants and show that the solution works



- Property

- This method is powerful, but we must be able to guess the form of the answer
- We can use this method to establish either upper or lower bounds on a recurrence

❖ The substitution method for solving recurrences

• Example

➤ Determine the upper bound of the recurrence $T(n) = 2T(\lfloor n/2 \rfloor) + n$

✓ Step 1 : guess the solution

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

✓ Step 2 : prove through mathematical induction

i. Require us to prove that $T(n) \leq cn \lg n$ for an appropriate choice of the constant $c > 0$

ii. Assume that the bound holds for all positive $m < n$, $\left(m = \frac{n}{2}\right)$

iii. $T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg(\lfloor n/2 \rfloor)$

iv.
$$\begin{aligned} T(n) &\leq 2(c \lfloor n/2 \rfloor \lg(\lfloor n/2 \rfloor)) + n \\ &\leq cn \lg(n/2) + n \\ &= cn \lg n - cn \lg 2 + n \\ &= cn \lg n - cn + n \\ &\leq cn \lg n, \end{aligned}$$

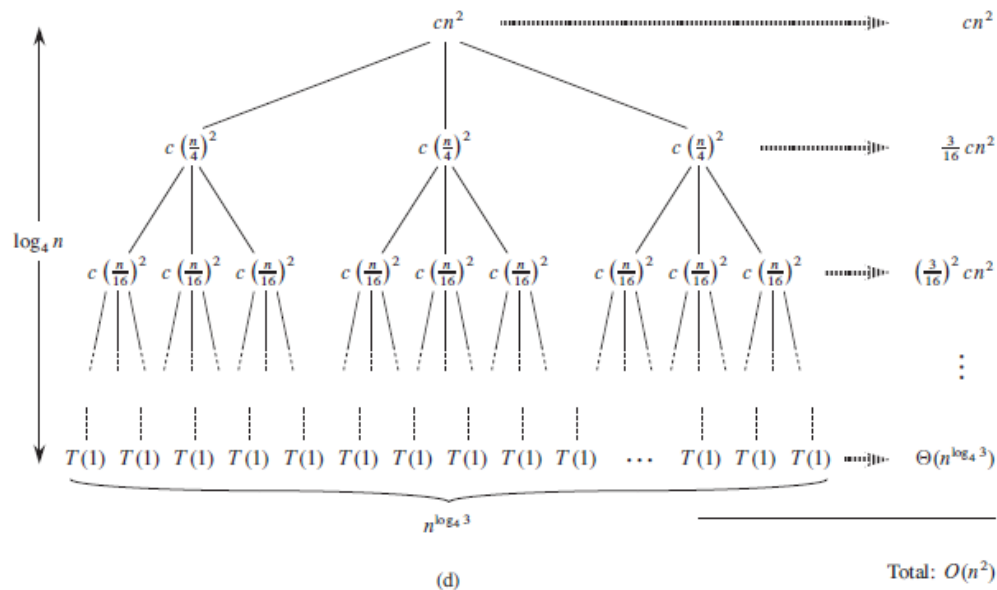
v. The last step holds as long as $c \geq 1$

vi. Require us to show that the solution holds for the boundary condition

❖ The recursion-tree method for solving recurrences

• Concept

- We sum all nodes to determine the total cost of the recursion
- Recursion tree is used to generate a good guess, which we can verify by the substitution method



- ✓ Each node represents the cost of a single subproblem
- ✓ Sum the costs within each level of the tree
- ✓ Sum all the per-level costs to determine the total cost of all levels of the recursion

❖ The recursion-tree method for solving recurrences

- Concept (cont'd)

- We can often tolerate a small amount of “sloppiness”

- ✓ Because we can verify a good guess by the substitution method later

- ✓ $T(n) = 3T(\lfloor n/4 \rfloor) + \theta(n^2) \rightarrow T(n) = 3T(n/4) + cn^2$

- Make a good guess for the recurrence (example)

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

- ✓ $T(n) = 3T(n/4) + cn^2$ (sloppiness that we can tolerate)

- ✓ Assume that n is an exact power of 4

- i. Another example of tolerable sloppiness

- ii. All subproblem sizes are integers

❖ The recursion-tree method for solving recurrences

- Make a good guess for the recurrence (cont'd)

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

- ✓ Part (a)

- i. initial state of the recurrence

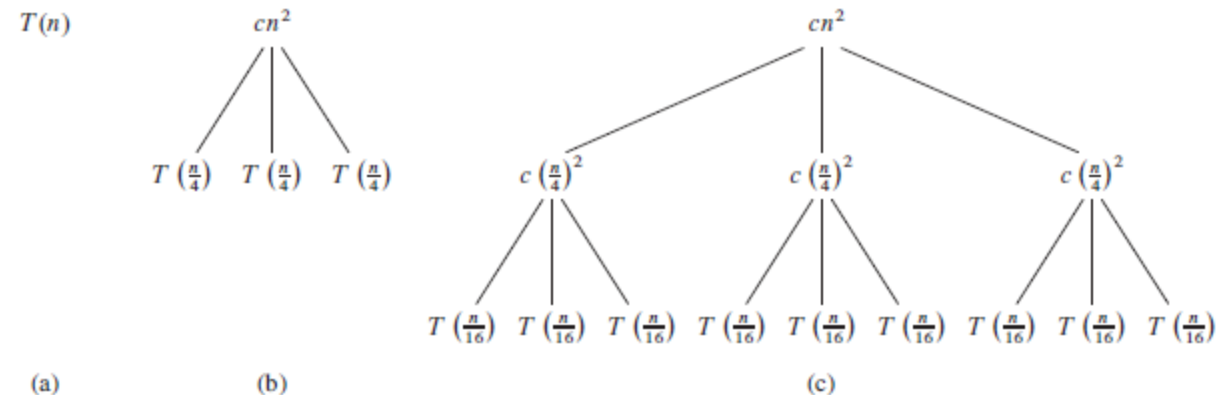
- ✓ Part (b)

- i. The cn^2 term at the root represents the cost at the top level of recursion

- ii. The three subtrees of the root represent the costs incurred by the subproblems of size $n/4$

- ✓ Part (c)

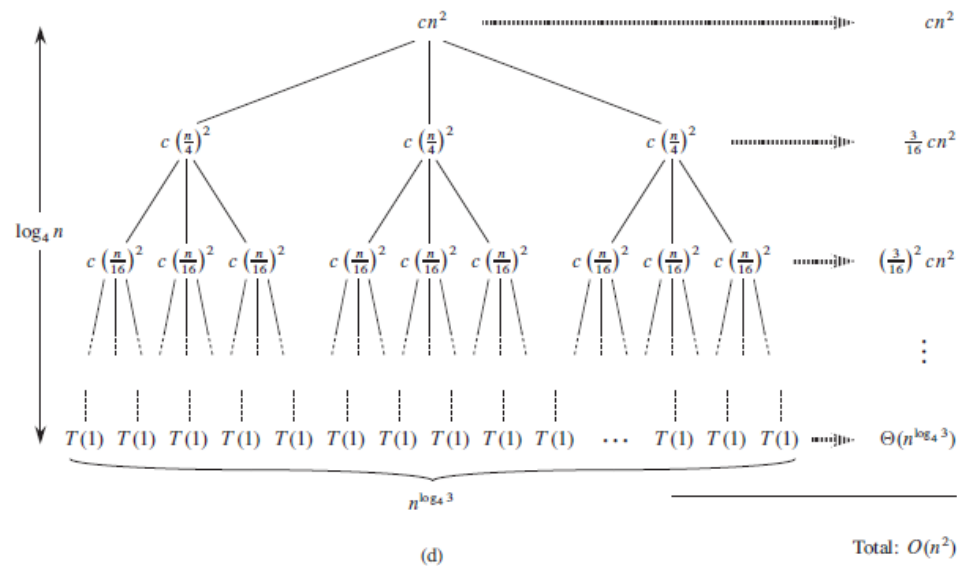
- i. The cost for each of the three children of the root is $c(\frac{n}{4})^2$
- ii. The three subtrees of the root represent the costs incurred by the subproblems of size $n/16$



❖ The recursion-tree method for solving recurrences

- Make a good guess for the recurrence (cont'd)

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



- ✓ Subproblem size for a node at depth i : $\frac{n}{4^i}$
- ✓ The tree has $\log_4 n + 1$ levels
 - Bottom node size = $\frac{n}{4^i} = 1$
 - $i = \log_4 n$
- ✓ The number of nodes at depth $i = 3^i$
- ✓ The cost of each node at level $i = c(\frac{n}{4^i})^2$
- ✓ The total cost of level $i = 3^i c(\frac{n}{4^i})^2 = cn^2 (\frac{3}{16})^i$

- ✓ The bottom level has $3^{\log_4 n} = n^{\log_4 3}$ nodes, each contributing cost $T(1)$ is a constant
- ✓ A total cost of bottom level = $n^{\log_4 3} T(1) = \theta(n^{\log_4 3})$

❖ The recursion-tree method for solving recurrences

- Make a good guess for the recurrence (cont'd)

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

- ✓ Add up the costs over all levels to determine the cost for the entire tree

$$\begin{aligned} T(n) &= \underbrace{cn^2}_{\text{Root node cost}} + \underbrace{\frac{3}{16}cn^2}_{\text{Sum of geometric sequence}} + \underbrace{\left(\frac{3}{16}\right)^2 cn^2}_{\text{Sum of geometric sequence}} + \cdots + \underbrace{\left(\frac{3}{16}\right)^{\log_4 n - 1} cn^2}_{\text{Sum of geometric sequence}} + \underbrace{\Theta(n^{\log_4 3})}_{\text{Last level total cost}} \\ &= \sum_{i=0}^{\log_4 n - 1} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3}) \\ &= \frac{(3/16)^{\log_4 n} - 1}{(3/16) - 1} cn^2 + \Theta(n^{\log_4 3}) \quad (\text{by equation (A.5)}) \end{aligned}$$

Annotations in the diagram:

- Root node cost: points to cn^2
- Sum of geometric sequence: points to $\frac{(3/16)^{\log_4 n} - 1}{(3/16) - 1} cn^2$
- total cost of each level(1 ~ $\log_4 n - 1$): points to the summation term
- Last level total cost: points to $\Theta(n^{\log_4 3})$

- ✓ This formula looks somewhat messy → we can take advantage of small amounts of sloppiness
 - ✓ Use an infinite decreasing geometric series as an upper bound

❖ The recursion-tree method for solving recurrences

- Make a good guess for the recurrence (cont'd)

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

- ✓ Take advantage of small amounts of sloppiness & use infinite decreasing geometric series

$$\begin{aligned} T(n) &= \sum_{i=0}^{\log_4 n - 1} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3}) \\ &< \sum_{i=0}^{\infty} \left(\frac{3}{16}\right)^i cn^2 + \Theta(n^{\log_4 3}) \\ &= \frac{1}{1 - (3/16)} cn^2 + \Theta(n^{\log_4 3}) \\ &= \frac{16}{13} cn^2 + \Theta(n^{\log_4 3}) \\ &= O(n^2). \end{aligned}$$

- i. Line 1~2 : use infinite decreasing geometric series as an upper bound
- ii. Line 4~5 : take advantage of small amounts of sloppiness

- We have derived a guess of $T(n) = O(n^2)$ for our original recurrence $T(n) = 3T(\lfloor n/4 \rfloor) + \theta(n^2)$

- Now we can use the substitution method to verify that our guess was correct

❖ The recursion-tree method for solving recurrences

- Make a good guess for the recurrence (cont'd)

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

- ✓ Use the substitution method to verify our guess

- ✓ $T(n) = O(n^2)$ is an upper bound for the recurrence $T(n) = 3T(\lfloor n/4 \rfloor) + \theta(n^2)$

- ✓ We want to show that $T(n) \leq dn^2$ for some constant $d > 0$

$$\begin{aligned} T(n) &\leq 3T(\lfloor n/4 \rfloor) + cn^2 \\ &\leq 3d \lfloor n/4 \rfloor^2 + cn^2 \\ &\leq 3d(n/4)^2 + cn^2 \\ &= \frac{3}{16} dn^2 + cn^2 \\ &\leq dn^2, \end{aligned}$$

- i. Line 1~2 : $T(n) = O(n^2)$ is an upper bound

- ii. Line 2~3 : Floor function means upper bound

- iii. Line 4~5 : holds as long as $d \geq \left(\frac{16}{13}\right)c$

- ✓ So, there exists some constant $d > 0$, our guess is correct!!

❖ The master method for solving recurrences

- Introduction

- The master method provides a “cookbook” method for solving recurrences of the form $T(n) = aT(n/b) + f(n)$
 - ✓ Where $a \geq 1$ and $b > 1$ are constants
 - ✓ $f(n)$ is an asymptotically positive function
- The recurrence describes the running time of algorithms
 - ✓ Divide the problem of size n into a subproblems
 - ✓ Each of size n/b
 - ✓ a subproblems are solved recursively, each in time $T(n/b)$
 - ✓ $f(n)$: divide and combine steps in divide and conquer algorithm
 - ✓ $T(\lfloor n/b \rfloor)$ or $T(\lceil n/b \rceil)$ will not affect the asymptotic behavior of the recurrence, so replace the terms $T(n/b)$

❖ The master method for solving recurrences

- The master theorem

➤ To use the master method, you will need to memorize three cases

Theorem 4.1 (Master 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 $T(n)$ has the following asymptotic bounds:

1. If $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
2. If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \lg n)$.
3. If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$, and if $af(n/b) \leq cf(n)$ for some constant $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$. ■

✓ Case 2

- i. The two functions are the same size
- ii. We multiply by a logarithmic factor
- iii. The solution is $T(n) = \theta(n^{\log_b a} \lg n) = \theta(f(n) \lg n)$

✓ compare the function $f(n)$ with the function $n^{\log_b a}$

✓ The larger of the two functions determines the solution to the recurrence

✓ Case 1

- i. The function $n^{\log_b a}$ is the larger
- ii. The solution is $T(n) = \theta(n^{\log_b a})$
- iii. The function $f(n)$ must be *polynomially* smaller $\rightarrow f(n)$ must be asymptotically smaller than $n^{\log_b a}$ by a factor of n^ϵ

❖ The master method for solving recurrences

- The master theorem (cont'd)

➤ To use the master method, you will need to memorize three cases

Theorem 4.1 (Master theorem)

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1. If $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
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3. If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$, and if $af(n/b) \leq cf(n)$ for some constant $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$. ■

✓ Case 3

- i. The function $f(n)$ is the larger
- ii. The solution is $T(n) = \theta(f(n))$
- iii. The function $f(n)$ must be *polynomially* larger
→ $f(n)$ must be asymptotically larger than $n^{\log_b a}$ by a factor of n^ϵ
- iv. Must satisfy the “regularity” condition that $af(n/b) \leq cf(n)$

- ✓ Three cases do not cover all the possibilities for $f(n)$
- ✓ Gap between cases 1 and 2 when $f(n)$ is smaller than $n^{\log_b a}$ but not *polynomially* smaller
- ✓ Gap between cases 2 and 3 when $f(n)$ is larger than $n^{\log_b a}$ but not *polynomially* larger
- ✓ The regularity condition in case 3 fails to hold
→ Cannot use the master method to solve the recurrence!!!

❖ The master method for solving recurrences

- Using the master method

- To use the master method, we simply determine which case of the master theorem applies and write down the answer

- $T(n) = 9T\left(\frac{n}{3}\right) + n$

- ✓ $a = 9, b = 3, f(n) = n$

- ✓ $f(n) = O(n^{\log_3 9 - \epsilon})$, where $\epsilon = 1$

- ✓ $n^{\log_b a} = n^{\log_3 9} = \theta(n^2)$

- ✓ We can apply case 1 of the master theorem and conclude that the solution is $T(n) = \theta(n^2)$

- $T(n) = T\left(\frac{2n}{3}\right) + 1$

- ✓ $a = 1, b = \frac{3}{2}, f(n) = 1$

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

- ✓ $f(n) = \theta(n^{\log_b a}) = \theta(1)$

- ✓ We can apply case 2 of the master theorem and conclude that the solution is $T(n) = \theta(\lg n)$

❖ The master method for solving recurrences

- Using the master method (cont'd)

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

✓ $a = 2, b = 2, f(n) = n \lg n$

✓ $n^{\log_b a} = n^{\log_2 2} = n$

✓ We might mistakenly think that case 3 should apply, since $f(n)$ is asymptotically larger than n

✓ The problem is that it is not *polynomially* larger

i. The ratio $f(n)/n^{\log_b a} = n \lg n / n = \lg n$

ii. It is asymptotically less than n^ϵ for any positive constant ϵ

✓ The recurrence falls into the gap between case 2 and case 3

➤ $T(n) = 7T(n/2) + \theta(n^2)$ (Strassen's algorithm)

✓ $a = 7, b = 2, f(n) = \theta(n^2)$

✓ $n^{\log_b a} = n^{\log_2 7} \rightarrow 2.80 < \lg 7 < 2.81$

✓ $f(n) = O(n^{\lg 7 - \epsilon})$ for $\epsilon = 0.8$

✓ We can apply case 1 of the master theorem and conclude that the solution is $T(n) = \theta(n^{\lg_2 7})$

Thank You!

❖ The substitution method for solving recurrences

• Example (cont'd)

➤ Show the solution holds for base case

✓ For $n = 1$

i. $T(n) \leq cn \lg n \rightarrow T(1) \leq c \lg 1 = 0$

ii. It is odds with $T(1) = 1 \leq 0$

iii. The base case of our inductive proof fails to hold.

✓ we can overcome this obstacle by only proving $T(n) \leq cn \lg n$ for $n \geq n_0$ where n_0 is a constant that we get to choose

i. Induction proves the statement for $n \geq 2$, $n_0 = 2$

ii. Base cases for induction proof $n = 2$ and $n = 3$

base case	recurrence relation	induction proof [$T(n) \leq cn \lg n$]
$n = 2$	$T(2) = 2T(1) + 2 = 4$	$4 \leq (2)(2) \lg(2)$
$n = 3$	$T(3) = 2T(1) + 3 = 5$	$5 \leq (2)(3) \lg(3)$

} $c = 2$