

算法基础

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第三讲 递归式

内容提要:

- 代换法
- 迭代法
- 递归树法
- 主方法



对算法运行时间的分析，一般都是通过每条指令指定的代价 * 指令执行次数来计算的。比如，插入算法的运行时间分析。

$$T(n) = c_1n + c_2(n-1) + c_4(n-1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n-1)$$

	INSERTION-SORT(A)	cost	times
1	for(j = 2; j <=length[A]; j++)	c₁	n
2	{ key = A[j]	c₂	n-1
3	// Insert A[j] into the sorted sequence A[1 ..j-1]	0	n-1
4	i = j-1	c₄	n-1
5	while(i > 0 && A[i] > key)	c₅	$\sum_{j=2}^n t_j$
6	A[i+1] = A[i]	c₆	$\sum_{j=2}^n (t_j - 1)$
7	i = i-1	c₇	$\sum_{j=2}^n (t_j - 1)$
8	}		
9	A[i+1] = key	c₈	n-1
10	}		



- 当一个算法包含对其自身的递归调用时，其运行时间通常可以用递归式来表示。
- 递归式是一组等式或不等式，它所描述的函数是用在更小的输入下该函数的值来定义的。如，归并排序：

$$T(n) = \begin{cases} 1 & , \text{if } n = 1 \\ 2T(n/2) + n & , \text{if } n > 1 \end{cases} \quad (4.1)$$

	cost
MERGE-SORT (<i>A, p, r</i>)	<i>T</i>(<i>n</i>)
1 if <i>p</i> < <i>r</i>	
2 Then <i>q</i> ←	
3 MERGE-SORT (<i>A, p, q</i>)	<i>T</i>(<i>n</i>/2)
4 MERGE-SORT (<i>A, q+1, r</i>)	<i>T</i>(<i>n</i>/2)
5 MERGE (<i>A, p, q, r</i>)	<i>n</i>



□ 在表达和求解递归式时一般都会忽略一些技术性细节:

$$T(n) = \begin{cases} 1 & , \text{if } n = 1 \\ 2T(n/2) + n & , \text{if } n > 1 \end{cases} \quad (4.1)$$

1) 假设函数自变量为整数, 忽略上取整和下取整。

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

2) 忽略递归式的边界条件, 并假设对于小的n值, T(n)是常量。

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

初始值只会影响常数因子, 并不会影响函数增长的阶。



□ 递归表达式求解方法:

- 1) **代换法**: 先猜有某个解存在, 用数学归纳法证明猜测的正确性;
- 2) **迭代法**: 把递归式转化为求和表达式, 然后求和式的界;
- 3) **递归树法**: 直观地表达了迭代法
- 4) **主方法**: 给出了求解 $T(n) = aT(n/b) + f(n)$ 这种形式递归式的简单方法。





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代换法

□ 代换法求解递归式要点:

1) 猜测解的形式

2) 用数学归纳法找出使解真正有效的常数

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

(1) Guess: $T(n) = n \lg n + n$

(2) Induction

Basic: $n = 1 \Rightarrow n \lg n + n = 1 = T(n)$

Inductive step: Inductive hypothesis is that $T(k) = k \lg k + k$ for all $k < n$.

Use the inductive hypothesis of $T(n/2)$ to prove $T(n)$

$$\begin{aligned} T(n) &= 2T(n/2) + n \\ &= 2((n/2) \lg(n/2) + (n/2)) + n \quad (\text{by inductive hypothesis}) \\ &= n \lg(n/2) + n + n = n(\lg n - \lg 2) + 2n = n \lg n + n. \end{aligned}$$



代换法

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2) 用数学归纳法找出使解真正有效的常数

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(1) Guess: $T(n) = n \lg n + n$

(2) Induction

Basic: $n = 1 \Rightarrow n \lg n + n = 1 = T(n)$

Inductive step: Inductive hypothesis is that $T(k) = k \lg k + k$ for all $k < n$.

Use the inductive hypothesis of $T(n/2)$ to prove $T(n)$.

这个方法非常有效，但是只适合于求解一些比较容易猜测的情形！



代换法

- The substitution method can be used to establish either **upper** (O) or **lower bounds** (Ω) on a recurrence.
- example, determining an upper bound on the recurrence

$$T(n) = 2T(\lfloor n/2 \rfloor) + n, \quad (4.4)$$

(1) **Guessing** that the solution is $T(n) = O(n \lg n)$.

(2) **Proving** $T(n) \leq cn \lg n$ for a some constant $c > 0$.

◆ Assume that this bound holds for $n/2$, that is, that

$T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \lg(\lfloor n/2 \rfloor)$. Substituting into the recurrence yields

$$T(n) = 2T(\lfloor n/2 \rfloor) + n \leq 2(c \lfloor n/2 \rfloor \lg(\lfloor n/2 \rfloor)) + n$$

$$\leq cn \lg(n/2) + n = cn \lg n - cn \lg 2 + n = cn \lg n - cn + n \leq cn \lg n,$$

where the last step holds as long as $c \geq 1$.





代换法

- Mathematical induction now requires us to show that our solution **holds for the boundary conditions**.
- Typically, the boundary conditions are suitable as **base cases** for the inductive proof.

$$T(n) = 2T(\lfloor n/2 \rfloor) + n, \quad (4.4)$$

$$T(n) = O(n \lg n) \quad , \quad T(n) \leq c n \lg n$$

- This requirement can sometimes lead to **problems**.
- Assume that $T(1) = 1$ is the sole boundary condition of the recurrence. Then, we can't choose c large enough, since $T(1) \leq c \cdot 1 \lg 1 = 0$, which is at odds with $T(1) = 1$. **The case of our inductive proof fails to hold.** (递归结果与初始情况矛盾, 即递归证明失败?)



代换法

$$T(n) = 2T(\lfloor n/2 \rfloor) + n; \quad T(n) = O(n \lg n), \quad T(n) \leq c n \lg n$$

- An inductive hypothesis inconsistent with specific boundary condition, How to overcome the difficulty?
(如何克服递归结果与边界条件不一致的问题?)
 - ◆ asymptotic notation only requires us to prove $T(n) \leq cn \lg n$ for $n \geq n_0$, where n_0 is a constant.
 - ◆ to remove the difficult boundary condition $T(1) = 1$
 - ◆ Impose $T(2)$ and $T(3)$ as boundary conditions for the inductive proof.
 - ◆ From the recurrence, we derive $T(2) = 4$ and $T(3) = 5$.
 - ◆ The inductive proof that $T(n) \leq cn \lg n$ can now be completed by choosing any $c \geq 2$ so that $T(2) \leq c 2 \lg 2$ and $T(3) \leq c 3 \lg 3$.





代换法

做一个好的猜测:

- Unfortunately, there is **no general way to guess the correct solutions to recurrences.** (猜想不是一种方法)
- **Guessing a solution takes experience and, occasionally, creativity.** (why we study the course? It's a training for us to get experience, to catch occasion, to have creativity.)
- Fortunately, though, there are some **heuristics (recursion tress)** that can help you become a good guesser.





代换法

- If a recurrence is **similar to one you have seen before**, then guessing a similar solution is reasonable. For example,

$$T(n) = 2T(\lfloor n/2 \rfloor + 17) + n,$$

- ◆ which looks difficult because of the added “17”.
- ◆ Intuitively, this additional term cannot substantially affect the solution to the recurrence. (该附加项不会从本质上影响递归解)
- ◆ When n is large, the difference between $T(n/2)$ and $T(n/2 + 17)$ is not that large. Consequently, we make the guess that $T(n) = O(n \lg n)$, which you can verify as correct by using the substitution method.





代换法

一些细微的问题:

- Sometimes, guess **correctly**, but somehow **the math doesn't seem to work out in the induction.**

For example $T(n) = T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + 1.$

- Guess the solution is $O(n)$, then try to show that $T(n) \leq cn$ for an appropriate constant c . Substituting .., then

$$T(n) \leq c \lfloor n/2 \rfloor + c \lceil n/2 \rceil + 1 = cn + 1,$$

which does not imply $T(n) \leq cn$ for any choice of c .





代换法

一些细微的问题:

- Sometimes, guess correctly, but somehow the math doesn't seem to work out in the induction.

For example $T(n) = T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + 1$.

guess $T(n) = cn$, thne $T(n) \leq c \lfloor n/2 \rfloor + c \lceil n/2 \rceil + 1 = cn + 1$,
contradiction.

- Usually, it is that the **inductive assumption isn't strong enough to prove the detailed bound. How to overcome?**
(递归假设条件不强)
- **Revising the guess by subtracting a lower-order term often permits the math to go through.** (减去低阶项)



代换法

一些细微的问题:

$$T(n) \leq T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + 1 \quad \text{Solution: } T(n) = O(n)$$

- try a larger guess $T(n) = O(n^2)$, which can work.
- But the guess that the solution is $T(n) = O(n)$ is correct.
- Intuitively, our **guess is nearly right**: we're only off by the constant 1, a lower-order term.
- Nevertheless, mathematical induction doesn't work!
- **Subtracting a lower-order term** from our previous guess. New guess is $T(n) \leq cn - b$, where $b \geq 0$ is constant, then

$$T(n) \leq (c \lfloor n/2 \rfloor - b) + (c \lceil n/2 \rceil - b) + 1 = cn - 2b + 1 \leq cn - b,$$

as long as $b \geq 1$. As before, the constant c must be chosen large enough to handle the boundary conditions.



代换法

一些细微的问题:

- Most people find the idea of subtracting a lower-order term counterintuitive. (违反直觉)
- After all, if the math doesn't work out, shouldn't we be increasing our guess?
- The key to understand this step is to remember that we are using mathematical induction: we can **prove something stronger** for a given value by **assuming something stronger** for smaller values. (假设更强的条件, 可证明更强的结论)





代换法

避免陷阱:

- It is easy to err in the use of asymptotic notation.

For example, in the recurrence (4.4)

$$T(n) = 2T(\lfloor n/2 \rfloor) + n \quad (4.4)$$

we can falsely prove $T(n) = O(n)$ by guessing $T(n) \leq cn$ and then arguing

$$\begin{aligned} T(n) &= 2T(\lfloor n/2 \rfloor) + n \leq 2(c \lfloor n/2 \rfloor) + n \\ &\leq cn + n \\ &= O(n), \quad \Leftarrow \text{wrong !!!} \end{aligned}$$

since c is a constant. The error is that we haven't proved the exact form of the inductive hypothesis, that is, that $T(n) \leq cn$.





代换法

改变变量:

- **algebraic manipulation (代数变换)** : sometimes solve an unknown recurrence similar to one you have seen before.

Example, $T(n) = 2T(\lfloor \sqrt{n} \rfloor) + \lg n,$

which looks difficult. Simplify the recurrence with a **change of variables**. For convenience, we shall not worry about rounding off values, such as \sqrt{n} , to be integers.

Let $m = \lg n$, then $T(2^m) = 2T(2^{m/2}) + m.$

Thus rename $S(m) = T(2^m) \Rightarrow S(m) = 2S(m/2) + m,$

which is very much like recurrence (4.4) and has the same solution: $S(m) = O(m \lg m)$. Changing back from $S(m)$ to $T(n)$, we obtain $T(n) = T(2^m) = S(m) = O(m \lg m) = O(\lg n \lg \lg n)$.





代换法

□ 代换法求解步骤小结:

- (1) 做一个好的猜测
- (2) 证明一般情况成立
- (3) 处理边界条件，可以进行边界扩展

□ 注意事项:

- (1) 对更小的值做更强的假设
- (2) 避免陷阱
- (3) 适当时进行变量替换





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迭代法

- **Substitution: It is difficult to come up with a good guess**
- **The iteration method**
 - ◆ **doesn't require us to guess the answer**
 - ◆ **may require more algebra** (迭代法对代数能力的要求较高)
 - ◆ **to expand (iterate) the recurrence and express it as a summation of terms, and the initial conditions**
 - ◆ **to evaluate summations.** (不断迭代展开为级数, 并求和)

For example, $T(n) = 3T(\lfloor n/4 \rfloor) + n$

$$\begin{aligned} T(n) &= n + 3T(\lfloor n/4 \rfloor) \\ &= n + 3(\lfloor n/4 \rfloor + 3T(\lfloor n/16 \rfloor)) \\ &= n + 3(\lfloor n/4 \rfloor + 3(\lfloor n/16 \rfloor + 3T(\lfloor n/64 \rfloor))) \\ &= n + 3\lfloor n/4 \rfloor + 9\lfloor n/16 \rfloor + 27T(\lfloor n/64 \rfloor), \end{aligned}$$





迭代法

$$\begin{aligned} T(n) &= n + 3T(\lfloor n/4 \rfloor) = n + 3(\lfloor n/4 \rfloor + 3T(\lfloor n/16 \rfloor)) \\ &= n + 3(\lfloor n/4 \rfloor + 3(\lfloor n/16 \rfloor + 3T(\lfloor n/64 \rfloor))) \\ &= n + 3\lfloor n/4 \rfloor + 9\lfloor n/16 \rfloor + 27T(\lfloor n/64 \rfloor), \end{aligned}$$

- **How far must we iterate the recurrence?**

- ◆ The i th term in the series is $3^i \lfloor n/4^i \rfloor$.
- ◆ The iteration halts when $\lfloor n/4^i \rfloor = 1$. By continuing the iteration until this point and using the bound $\lfloor n/4^i \rfloor \leq n/4^i$, we get a decreasing geometric series:

$$\begin{aligned} T(n) &\leq n + 3n/4 + 9n/16 + 27n/64 + \dots + 3^i T(n/4^i) \\ &\leq n \sum_{i=0}^{\infty} (3/4)^i + \Theta(n^{\log_4 3}) = 4n + o(n) = O(n). \\ (n/4^i = 1 &\Rightarrow i = \log_4 n \Rightarrow 3^i = 3^{\log_4 n} = n^{\log_4 3}) \end{aligned}$$





迭代法

- The iteration method usually leads to lots of algebra. It can be a challenge. The key points:
 - ◆ the number of times the recurrence needs to be iterated to reach the boundary condition, (递归次数)
 - ◆ and the sum of the terms arising from each level of the iteration process. (级数求和)
- Sometimes, **in the process** of iterating a recurrence, you can guess the solution without working out all the math. Then, the iteration can be abandoned in favor of the substitution method, which usually requires less algebra.

(在展开递归式为迭代求和的过程中, 有时只需要部分展开, 然后根据其规律来猜想递归式的解, 接着用代换法进行证明。)





迭代法

- When a recurrence contains **floor** and **ceiling** functions, the math can become especially **complicated**.
- Often, it helps to assume that the recurrence is defined only on exact **powers of a number**.

Example, $T(n) = 3T(\lfloor n/4 \rfloor) + n$

if we had assumed that $n = 4^k$ for some integer k , the floor functions could have been conveniently omitted.





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递归树法

- Drawing out a recursion tree, is a straightforward way to devise a good guess, and to show the iteration method **intuitively**. (画递归树可以从直观上表示迭代法，也有助于猜想递归式的解)
- Recursion trees are particularly useful when the recurrence describes the running time of a divide-and-conquer algorithm.
- 递归树中，每一个节点都代表着递归函数调用集合中一个子问题的代价。将树中每一层内的代价相加得到一个每层代价的集合，再将每层的代价相加得到递归式所有层次的总代价。



递归树法

- 使用递归树产生好的猜测时，通常需要容忍小量的“不良量”：
 - ✓ 1) floor, ceiling忽略；
 - ✓ 2) n 经常假设为某个整数的幂次方；
- 例子见课本P41
- 关键：1) 树的深度如何确定？
2) 每个节点的代价—>每层的代价—>总代价





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主方法

- solving recurrences of the form

$$T(n) = aT(n/b) + f(n), \quad (4.5)$$

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

- The master method requires **memorization of three cases**, but then the solution of many recurrences can be determined quite easily, often **without pencil and paper**.
- n/b might not be an integer. Replacing each of $T(n/b)$ with either $T(\lfloor n/b \rfloor)$ or $T(\lceil n/b \rceil)$ doesn't affect the asymptotic behavior .
- Normally, it is convenient to **omit the floor and ceiling functions** when writing divide-and-conquer recurrences of this form.





主方法

□ Theorem 4.1 (主定理)

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.

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



主方法

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

$$T(n) = \begin{cases} \Theta(n^{\log_b a}), & f(n) = O(n^{(\log_b a) - \varepsilon}) \\ \Theta(n^{\log_b a} \lg n), & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)), & f(n) = \Omega(n^{(\log_b a) + \varepsilon}) \text{ and } af(n/b) \leq cf(n) \text{ for large } n \end{cases} \left. \begin{array}{l} \exists \varepsilon > 0 \\ , c < 1 \end{array} \right\}$$

- Comparing the function $f(n)$ with $n^{\log_b a}$. Intuitively, the solution is determined by **the larger of the two functions**.
 - ◆ Case 1, $n^{\log_b a}$ larger, then the solution is $T(n) = \Theta(n^{\log_b a})$.
 - ◆ Case 3, $f(n)$ larger, then the solution is $T(n) = \Theta(f(n))$.
 - ◆ Case 2, the two functions are the same size, we multiply by a logarithmic factor, and the solution is

$$T(n) = \Theta(n^{\log_b a} \lg n) = \Theta(f(n) \lg n) \text{ .}$$



主方法

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

$$T(n) = \begin{cases} \Theta(n^{\log_b a}), & f(n) = O(n^{(\log_b a) - \varepsilon}) \\ \Theta(n^{\log_b a} \lg n), & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)), & f(n) = \Omega(n^{(\log_b a) + \varepsilon}) \text{ and } af(n/b) \leq cf(n) \text{ for large } n \end{cases} \left. \begin{array}{l} \exists \varepsilon > 0 \\ , c < 1 \end{array} \right\}$$

主方法特殊情况:

- **Polynomially**

- ◆ Case 1, $f(n)$ must be polynomially smaller than $n^{\log_b a}$.
- ◆ Case 3, $f(n)$ must be polynomially larger than $n^{\log_b a}$.

- **Gap** Example: $f(n) = n \lg n$, $n^{\log_b a} = n$

- ◆ There is a gap between cases 1 and 2 when $f(n)$ is smaller than $n^{\log_b a}$ but not polynomially smaller.
- ◆ Similarly, there is a gap between cases 2 and 3 when $f(n)$ is larger than $n^{\log_b a}$ but not polynomially larger.





主方法

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

$$T(n) = \begin{cases} \Theta(n^{\log_b a}), & f(n) = O(n^{(\log_b a) - \varepsilon}) \\ \Theta(n^{\log_b a} \lg n), & f(n) = \Theta(n^{\log_b a}) \\ \Theta(f(n)), & f(n) = \Omega(n^{(\log_b a) + \varepsilon}) \text{ and } af(n/b) \leq cf(n) \text{ for large } n \end{cases} \left. \begin{array}{l} \exists \varepsilon > 0 \\ , c < 1 \end{array} \right\}$$

举例:

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

$$a = 9, b = 3, f(n) = n \Rightarrow n^{\log_b a} = n^{\log_3 9} = n^2 = \Theta(n^2)$$

$$\Rightarrow f(n) = O(n^{\log_3 9 - \varepsilon}), \text{ where } \varepsilon = 1 \Rightarrow T(n) = \Theta(n^2)$$

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

$$a = 1, b = 3/2, f(n) = 1 \Rightarrow n^{\log_b a} = n^{\log_{3/2} 1} = n^0 = 1$$

$$\Rightarrow f(n) = \Theta(n^{\log_3 a}) = \Theta(1) \Rightarrow T(n) = \Theta(\lg n)$$



主方法

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

$$T(n) = \left\{ \begin{array}{l} \Theta\left(n^{\log_b a}\right), f(n) = O\left(n^{(\log_b a) - \varepsilon}\right) \\ \Theta\left(n^{\log_b a} \lg n\right), f(n) = \Theta\left(n^{\log_b a}\right) \\ \Theta\left(f(n)\right), f(n) = \Omega\left(n^{(\log_b a) + \varepsilon}\right) \text{ and } af(n/b) \leq cf(n) \text{ for large } n \end{array} \right\} \begin{array}{l} \exists \varepsilon > 0 \\ , c < 1 \end{array}$$

举例： $T(n) = 3T(n/4) + n \lg n$

$$a = 3, b = 4, f(n) = n \lg n \Rightarrow n^{\log_b a} = n^{\log_4 3} = O(n^{0.793})$$

$\Rightarrow f(n) = \Omega(n^{(\log_4 3) + \varepsilon})$, where $\varepsilon \approx 0.2$, and for sufficiently large n ,

$$af(n/b) = 3(n/4) \lg(n/4) \leq (3/4)n \lg n = cf(n) \text{ for } c = 3/4$$

$\Rightarrow T(n) = \Theta(n \lg n)$



主方法

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

$$T(n) = \left\{ \begin{array}{l} \Theta\left(n^{\log_b a}\right), f(n) = O\left(n^{(\log_b a) - \varepsilon}\right) \\ \Theta\left(n^{\log_b a} \lg n\right), f(n) = \Theta\left(n^{\log_b a}\right) \\ \Theta\left(f(n)\right), f(n) = \Omega\left(n^{(\log_b a) + \varepsilon}\right) \text{ and } af(n/b) \leq cf(n) \text{ for large } n \end{array} \right. \left. \begin{array}{l} \varepsilon > 0 \\ c < 1 \end{array} \right.$$

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

$$a = 2, b = 2, f(n) = n \lg n \quad \Rightarrow \quad n^{\log_b a} = n,$$

but $f(n)/n = \lg n$, which is asymptotically less than n^ε for any positive constant c , that is $f(n)$ is not polynomially larger than n . Consequently, the recurrence falls into the gap between case 2 and case 3.



谢谢!

Q & A