- Difficulty of Algorithms Research
  - ◆ Model 建模
  - ◆ Specify 描述
  - ◆ Correctness 正确性
    - Verify 验证
    - Proof 证明

Design 设计

Correctness Analysis 正确性分析

Computing Analysis 可计算性分析

- ◆ Complex 复杂度 (Efficiency 有效性)
  - Actual computing 实际可计算性
- Recurrence is a basic method to analyze algorithm

### **Algorithms analysis**

sum

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1)$$

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

INSERTION-SORT(A)		cost	times
1 $for(j = 2; j \le length[A]; j++)$		$c_1$	$\boldsymbol{n}$
$2 \qquad \{ \qquad key = A[j] $		$c_2$	<i>n</i> -1
3 // Insert $A[j]$ i	nto the sorted sequence $A[1j-1]$	0	<i>n</i> -1
4   i = j-1		$c_4$	<i>n</i> -1
5 while $(i > 0 \&$	& A[i] > key	$c_5$	$\sum\nolimits_{j=2}^{n}t_{j}$
$ 6 \qquad \qquad \{ \qquad A[i+1] =$	A[i]	<i>c</i> <sub>6</sub>	$\sum_{j=2}^{n} (t_j - 1)$
i = i-1		<i>c</i> <sub>7</sub>	$\sum_{j=2}^{n} (t_j - 1)$
8 }			<b></b> <i>J</i> _2 <i>J</i>
9   A[i+1] = key		$c_8$	<i>n</i> -1
10 }			

#### **Algorithms analysis**

## recursion

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

		cost
MERGE-SORT(A, p, r)		T(n)
1 if	p < r	
2	$q \leftarrow \lfloor (p+r)/2 \rfloor$	
3	MERGE-SORT(A, p, q)	T(n/2)
4	MERGE-SORT(A, q+1, r)	T(n/2)
5	MERGE(A, p, q, r)	n

#### **Algorithms analysis**

## recursion

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

```
f(n)
{
    if(n<=2)
       return 1;
    else
      return f(n-1)+f(n-2);
}</pre>
```

#### **Algorithms analysis**

## recursion

$$h(n) = h(0)*h(n-1) + h(1)*h(n-2) + ... + h(n-1)h(0)$$

## 该递推关系的解为:

$$h(n) = C(2n, n)/(n+1)$$
  
(n = 1, 2, 3,...)

# E Zexal的二叉树 (签到)

时间限制: 1000ms 内存限制: 65536kb

通过率: 200/209 (95.69%) 正确率: 200/596 (33.56%)

#### 题目

知识点: 树, 数论, dp, 递归(都可以做)

上学期我们学习了二叉树,也都知道3个结点的二叉树有5种, 现给你二叉树的结点个数n,要你输出不同形态二叉树的种数。

#### 输入

第一个数为一个整数n(n <= 30)

#### 输出

对于每组数据,输出一行,不同形态二叉树的种数。

#### 输入样例

3

#### 输出样例

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# A recurrence is an equation or inequality in terms of

- one or more base cases, and
- itself, with smaller arguments.

## **Examples:**

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

(3) 
$$T(n) = \begin{cases} 0, & \text{if } n=2, \\ T(\sqrt{n})+1, & \text{if } n>2. \end{cases}$$
  
Solution:  $T(n) = \lg \lg n$ .

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

(4) 
$$T(n) = \begin{cases} 1, & \text{if } n=1, \\ T(n/3) + T(2n/3) + n, & \text{if } n > 1. \end{cases}$$
  
Solution:  $T(n) = \Theta(n \lg n)$ 

How to obtain asymptotic " $\Theta$ " or "O" bounds on the recurrence solution?

- Substitution method (置換法): guesses a bound and then use mathematical induction to prove our guess correct.
- Iteration method (迭代法): converts the recurrence into a summation and then relies on techniques for bounding summations to solve the recurrence.
- Recursion-tree method (a kind of iteration method)
- Master method (主方法,母函数法): provides bounds for recurrences of the form T(n) = aT(n/b) + f(n), where  $a \ge 1$ , b > 1, and f(n) is a given function.

### **Technicalities**

In practice, we neglect certain technical details when we state and solve recurrences. (忽略技术细节)

- 1) Assumption of integer arguments to functions
- Normally, T(n) is only defined when n is an integer
- Example, the worst-case running time of MERGE-SORT

$$T(n) = \begin{cases} \Theta(1) &, \text{ if } n=1, \\ T(\lceil n/2 \rceil) + T(\lceil n/2 \rceil) + \Theta(n) &, \text{ if } n > 1, \end{cases}$$
(4.2)

### **Technicalities**

## 2) Ignore boundary conditions

• Omit statements of the boundary conditions of recurrences, assume that T(n) is constant for small n, that is  $T(n) = \Theta(1)$  for sufficiently small n.

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

- Example, state recurrence (4.1) as  $T(n) = 2T(n/2) + \Theta(n)$ , (Omit  $T(1) = \Theta(1)$ ) (4.3) without explicitly giving values for small n.
  - The reason is that although changing the value of *T*(1) changes the solution to the recurrence, the order of growth is unchanged. (改变边界值,可能改变递归式的解,但不改变解的函数增长率)

### **Technicalities**

neglect certain technical details

- 1) Assumption of integer arguments to functions
- 2) Ignore boundary conditions
- 3) Omit floors, ceilings

 These details don't affect the asymptotic bounds of many recurrences encountered in the analysis of algorithms (忽略细节不会影响算法的渐近分析)

# **Key points**

- 1) guessing the form of the solution;
- 2) using mathematical induction to show the solution works.

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)

$$T(n) = 2T(n/2) + n$$
  
=  $2((n/2)\lg(n/2) + (n/2)) + n$  (by inductive hypothesis)  
=  $n\lg(n/2) + n + n = n(\lg n - \lg 2) + 2n = n\lg n + n$ .

# Key points

- 1) guessing the form of the solution;
- 2) using mathematical induction to show the solution works.

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

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**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).

• The method is powerful, but it can be applied only in cases when it is easy to guess the form of the answer.

• The substitution method is powerful, but it can be applied only in cases when it is easy to guess the form of the answer.

$$h(n) = h(0)*h(n-1) + h(1)*h(n-2) + ... + h(n-1)h(0)$$



$$h(n) = C(2n, n)/(n+1)$$
  
(n = 1, 2, 3,...)

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#### 输入

第一个数为一个整数n(n <= 30)

#### 输出

对于每组数据,输出一行,不同形态二叉树的种数。

#### 输入样例

3

#### 输出样例

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

$$T(n) = 2T(\lfloor n/2 \rfloor) + n, \tag{4.4}$$

- (1) Guessing that the solution is  $T(n) = O(n \lg n)$ .
- (2) Proving  $T(n) \le cn \lg n$  for a some constant c > 0.
- **Assume that this bound holds for** n/2, that is, that  $T(\lfloor n/2 \rfloor) \le c \lfloor n/2 \rfloor \lg(\lfloor n/2 \rfloor)$ . Substituting into the recurrence yields

$$T(n) = 2T(\lfloor n/2 \rfloor) + n \le 2(c\lfloor n/2 \rfloor \lg(\lfloor n/2 \rfloor)) + n$$
  
$$\le cn \lg(n/2) + n = cn \lg n - cn \lg 2 + n = cn \lg n - cn + n \le cn \lg n,$$

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

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

$$T(n) = 2T(\lfloor n/2 \rfloor) + n, \tag{4.4}$$

$$T(n) = O(n \lg n)$$
 ,  $T(n) \le c n \lg n$ 

- 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  $T(1) \le c$  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?)

# (1)To extend boundary conditions

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

• 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  $T(n) \le cn \lg n$  for  $n \ge n_0$ , 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) = 2\*T(1) + 2 = 4, T(3) = 2\*T(1) + 3 = 5.
- The inductive proof that T(n)≤cn lg n can now be completed by choosing any c≥2 so that  $T(2) = 4 \le c*2$  lg 2 and  $T(3) = 5 \le c$  3 lg 3.

# 4.1.1 Making a good guess

- 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.

# 4.1.1 Making a good guess

• 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.

# 4.1.1 Making a good guess

• Another way to make a good guess is to prove loose upper and lower bounds on the recurrence and then reduce the range of uncertainty.

(寻找松的上、下渐近界,范围缩小,逐步逼近)

## For example

$$T(n) = 2T(|n/2|) + n$$
 (4.4)

- we might start with a lower bound of  $T(n) = \Omega(n)$ , since we have the term n in the recurrence,
- and we can prove an initial upper bound  $T(n) = O(n^2)$ .
- Then, we can gradually lower the upper bound and raise the lower bound until we converge on the correct, asymptotically tight solution of  $T(n)=\Theta(n\lg n)$ .

 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  $T(n) \le cn$  for an appropriate constant c. Substituting ..., then

$$T(n)=T(n/2)+T(n/2)+1$$
  
 $\leq c \cdot n/2 + c \cdot n/2 + 1 = cn+1,$ 

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

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

$$T(n) \le T(\mid n/2 \mid) + T(\lceil n/2 \rceil) + 1$$
 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!
- (2)Subtracting a lower-order term from our previous guess. New guess is  $T(n) \le cn b$ , where  $b \ge 0$  is constant, then

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

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

- Most people find the idea of subtracting a lowerorder 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)

# 4.1.3 (3)Avoiding pitfalls (陷阱)

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

For example, in the recurrence (4.4)

$$T(n) = 2T(\lfloor n/2 \rfloor) + n \tag{4.4}$$

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

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

$$\le cn + n$$

$$= O(n), \qquad \iff wrong !!!$$

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

# 4.1.4 Changing variables

• algebraic manipulation: sometimes solute 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  $S(m) = T(2^m) = > S(m) = 2S(m/2) + m$ , 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  $T(n) = T(2^m) = S(m) = O(m \lg m) = O(\lg n \lg \lg n)$ .

### 4.1 The substitution method: Notes

- (1) To extend boundary conditions
- (2) Subtracting a lower-order term

(3) Avoiding pitfalls

- 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, 
$$T(n) = 3T(\lfloor n/4 \rfloor) + n$$
  
 $T(n) = n + 3T(\lfloor n/4 \rfloor)$   
 $= n + 3(\lfloor n/4 \rfloor + 3T(\lfloor n/16 \rfloor))$   
 $= n + 3(\lfloor n/4 \rfloor + 3(\lfloor n/16 \rfloor + 3T(\lfloor n/64 \rfloor)))$   
 $= n + 3(\lfloor n/4 \rfloor + 9(\lfloor n/16 \rfloor + 27T(\lfloor n/64 \rfloor)))$ 

$$T(n) = n + 3T(\lfloor n/4 \rfloor) = n + 3(\lfloor n/4 \rfloor + 3T(\lfloor n/16 \rfloor))$$

$$= n + 3(\lfloor n/4 \rfloor + 3(\lfloor n/16 \rfloor + 3T(\lfloor n/64 \rfloor)))$$

$$= n + 3\lfloor n/4 \rfloor + 9\lfloor n/16 \rfloor + 27T(\lfloor n/64 \rfloor),$$

- How far must we iterate the recurrence?
  - The *i*th term in the series is  $3^{i}T(\lfloor n/4^{i} \rfloor)$ .
  - The iteration halts when  $\lfloor n/4^i \rfloor = 1$ . By continuing the iteration until this point and using the bound  $\lfloor n/4^i \rfloor \le n/4^i$ , we get a decreasing geometric series:  $T(n) \le n + 3n/4 + 9n/16 + 27n/64 + ... + 3^i T(n/4^i)$   $\le n \sum_{i=0}^{\infty} (3/4)^i + \Theta(n^{\log_4 3}) = 4n + o(n) = O(n)$ .

 $(n/4^{i} = 1 \implies i = \log_{4} n \implies 3^{i} = 3^{\log_{4} n} = n^{\log_{4} 3})$ 

- 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, (迭代次数)
  - 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,接着用置换法进行证明。)

- 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.

### \* 4.3 The recursion-tree method

- 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.

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



