

Analysis and Design of Algorithms

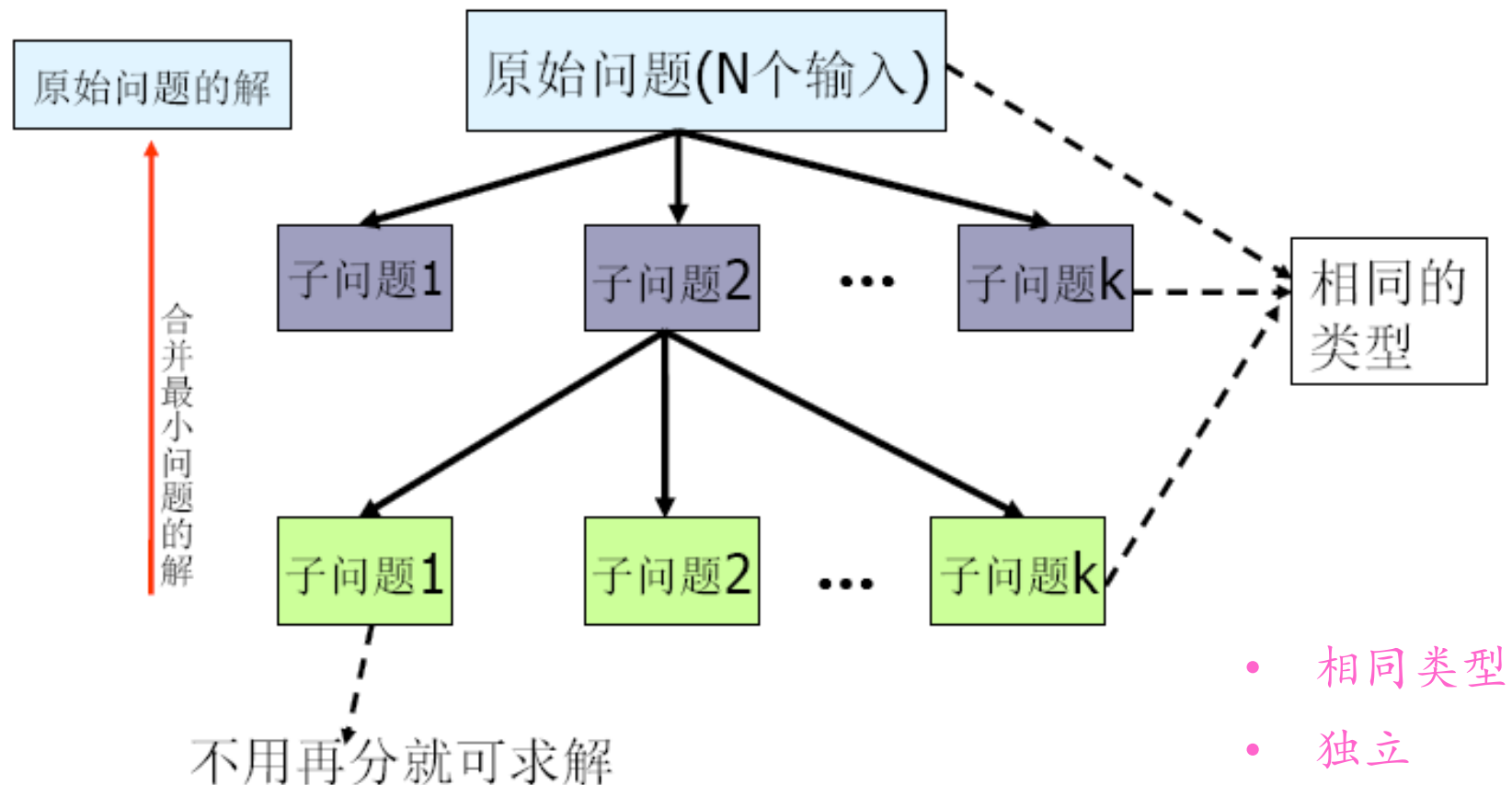
Chapter 5: Divide and Conquer



School of Software Engineering © Yanling Xu



Divide and Conquer Approach



Divide and Conquer Approach

■ *Three Steps of The Divide and Conquer Approach*

- ✦ *Divide* the problem into two or more smaller subproblems
- ✦ *Conquer* the subproblems by solving them recursively
- ✦ *Combine* the solutions to the subproblems into the solutions to the original problem

Divide and Conquer Approach

■ *Algorithm*

```
divide-and-conquer(P)
{
  if ( | P | <= n0) adhoc(P);           //解决小规模的问题
  divide P into smaller subinstances P1,P2,...,Pk ; //分解问题
  for (i=1,i<=k,i++)
    yi=divide-and-conquer(Pi);         //递归的解各子问题
  return merge(y1,...,yk);             //将各子问题的解合并为原问题的解
}
```

☆ 人们从大量实践中发现，在用分治法设计算法时，最好使子问题的规模大致相同。即将一个问题分成大小相等的 k 个子问题的处理方法是行之有效的。这种使子问题规模大致相等的做法是出自一种**平衡(balancing)子问题**的思想，它几乎总是比子问题规模不等的做法要好。

Divide and Conquer Approach

■ **Algorithm analysis—general divide-and-conquer recurrence**

- *A problem's instance of size n is divided into a instances of size n/b (assuming n is a power of b)*
- *a of the problems needs to be solved*
- *$f(n)$ is a function that counts for the time spent on dividing the problem into smaller ones and on combining their solutions*

$$T(n) = \begin{cases} O(1) & n = 1 \\ aT(n/b) + f(n) & n > 1 \end{cases}$$

Divide and Conquer Approach

■ **Master Theorem**

$$T(n) = aT(n/b) + f(n), \quad \text{where } f(n) \in \Theta(n^k), \quad k \geq 0$$

1. $a < b^k$ $T(n) \in \Theta(n^k)$
2. $a = b^k$ $T(n) \in \Theta(n^k \log n)$
3. $a > b^k$ $T(n) \in \Theta(n^{\log_b a})$

The same results hold with O instead of Θ

Multiplication of Large Integers

■ Idea of Multiplication of Large Integers

Consider the problem of multiplying two (large) n -digit integers represented by arrays of their digits such as:

A = 12345678901357986429 B = 87654321284820912836

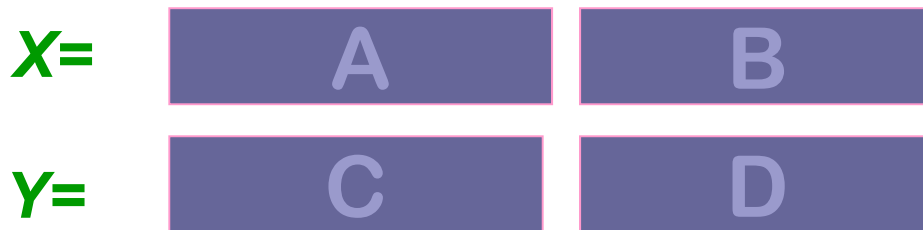
✦ *brute-force algorithm*

$$\begin{array}{ccccccc} & & a_1 & a_2 & \dots & a_n \\ & & b_1 & b_2 & \dots & b_n \\ (d_{10}) & d_{11} & d_{12} & \dots & d_{1n} \\ (d_{20}) & d_{21} & d_{22} & \dots & d_{2n} \\ \dots & \dots & \dots & \dots & \dots & \dots \\ (d_{n0}) & d_{n1} & d_{n2} & \dots & d_{nn} \end{array}$$

- Efficiency: n^2 one-digit multiplications

Multiplication of Large Integers

■ First Divide-and-Conquer Algorithm



✦ if $X = A 10^{n/2} + B$, and $Y = C 10^{n/2} + D$ --divide X, Y into two parts
where X and Y are n -digit, A, B, C, D are $n/2$ -digit numbers

$$X * Y = AC \cdot 10^n + (AD + BC) \cdot 10^{n/2} + BD$$

- If $n=2^k$, recurrence
- Could stop
 - when $n=1$;
 - or n is small enough to multiply the numbers of that size directly

Multiplication of Large Integers

■ First Divide-and-Conquer Algorithm

✦ Analysis:

- Basic operation: one-digit multiplication

- $$T(n) = \begin{cases} O(1) & n = 1 \\ 4T(n/2) + O(n) & n > 1 \end{cases}$$

Solution: $T(n) = O(n^2)$ ✖NO Promotion

If $n=2^k$, then

$$C(2^k)=4C(2^{k-1})=4[4C(2^{k-2})]=4^2C(2^{k-2})$$

$$= \dots = 4^k C(2^{k-k}) = 4^k$$

$$k = \log_2 n$$

$$C(2^k) = C(n) = 4^{\log n} = 2^{2\log n} = n^2$$

For $n \neq 2^k$ by Smoothness Rule, $O(n^2)$

如果在推导中不忽略 $O(n)$, 则

$$C(2^k) = 4C(2^{k-1}) + 2^k = 4[4C(2^{k-2}) + 2^{k-1}] + 2^k$$

$$= 4^2C(2^{k-2}) + 4 \cdot 2^{k-1} + 2^k = 4^2C(2^{k-2}) + 2^{k+1} + 2^k$$

$$= 4^k C(2^{k-k}) + 2^{k+k-1} + \dots + 2^k$$

$$= 2^k(2^k + 2^{k-1} + \dots + 2^0)$$

$$= 2^k(1 - 2^{k+1}) / (1 - 2) = 2^{2k+1} - 2^k$$

Multiplication of Large Integers

■ Second Divide-and-Conquer Algorithm

- ✦ The idea is to decrease the number of multiplications from 4 to 3:

$$(A + B) * (C + D) = AC + (AD + BC) + BD$$

$$\text{i.e., } (AD + BC) = (A + B) * (C + D) - AC - BD$$

which requires only 3 multiplications at the expense of (4-1) extra add/sub.

✦ Analysis:

Recurrence for the number of multiplications $T(n)$:

$$T(n) = \begin{cases} O(1) & n = 1 \\ 3T(n/2) + O(n) & n > 1 \end{cases}$$

If $n=2^k$, then

$$C(2^k) = 3C(2^{k-1}) = 3[3C(2^{k-2})] = 3^2C(2^{k-2}) \\ = \dots = 3^kC(2^{k-k}) = 3^k$$

$$k = \log_2 n$$

$$C(2^k) = C(n) = 3^{\log_2 n} = n^{\log_2 3} \approx n^{1.585}$$

Solution: $T(n) = O(3^{\log_2 n}) = O(n^{\log_2 3}) \approx O(n^{1.585})$ ✓ Promotion

Multiplication of Large Integers

☆ $XY = ac \cdot 2^n + ((a+c)(b+d) - ac - bd) \cdot 2^{n/2} + bd$

两个XY的复杂度都是 $O(n^{\log 3})$ ，但考虑到 $a+c, b+d$ 可能得到 $m+1$ 位的结果，使问题的规模变大，故不选择第2种方案。

☆ 如果将大整数分成更多段，用更复杂的方式把它们组合起来，将有可能得到更优的算法。

☆ 最终的，这个思想导致了快速傅利叶变换(Fast Fourier Transform)的产生。该方法也可以看作是一个复杂的分治算法。

如果在推导中不忽略 $O(n)$ ，则

$$\begin{aligned} C(2^k) &= 3C(2^{k-1}) + 2^k = 3[3C(2^{k-2}) + 2^{k-1}] + 2^k = 3^2C(2^{k-2}) + 3 \cdot 2^{k-1} + 2^k \\ &= \dots = 3^k C(2^{k-k}) + 3^{k-1} \cdot 2^1 + 3^{k-2} \cdot 2^2 + \dots + 3 \cdot 2^{k-1} + 2^k \\ &= 3^k \cdot 2^0 + 3^{k-1} \cdot 2^1 + 3^{k-2} \cdot 2^2 + \dots + 3 \cdot 2^{k-1} + 2^k \quad \text{其等比 } q = 2/3 \text{ 首相 } a_1 = 3^k \cdot 2^0 \\ &= a_1(1 - q^{k+1}) / (1 - q) = 3^{k+1} - 2^{k+1} = 3 \cdot 3^{\log_2 n} - 2 \cdot n = 3 \cdot n^{\log_2 3} - 2 \cdot n \quad \text{前者是主项} \end{aligned}$$

Strassen's Matrix Multiplication

❏ *Idea*

✦ *brute-force alg.:* $O(n^3)$

$$C_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

For C_{ij} , n multiples and $n-1$ additions

So for n elements in C , $T(n) = O(n^3)$

Strassen's Matrix Multiplication

■ Idea

✦ *Divide and Conquer— idea1*

divide A, B, and C into 4 equal-size sub-matrix,

$$\begin{aligned} \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} &= \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} \\ &= \begin{bmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{bmatrix} \end{aligned}$$

$$T(n) = \begin{cases} O(1) & n = 1 \\ 8T(n/2) + O(n^2) & n > 1 \end{cases}$$

$$T(n) = O(n^3)$$

Strassen's Matrix Multiplication

■ Idea

✦ *Divide and Conquer — idea2 to reduce the times for multiply*

Strassen observed [1969] that the product of two matrices can be computed as follows

$$\begin{aligned} \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} &= \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} \\ &= \begin{bmatrix} M_5 + M_4 - M_2 + M_6 & M_1 + M_2 \\ M_3 + M_4 & M_5 + M_1 - M_3 - M_7 \end{bmatrix} \end{aligned}$$

$$M_1 = A_{11}(B_{12} - B_{22})$$

$$M_2 = (A_{11} + A_{12})B_{22}$$

$$M_3 = (A_{21} + A_{22})B_{11}$$

$$M_4 = A_{22}(B_{21} - B_{11})$$

$$M_5 = (A_{11} + A_{22})(B_{11} + B_{22})$$

$$M_6 = (A_{12} - A_{22})(B_{21} + B_{22})$$

$$M_7 = (A_{11} - A_{21})(B_{11} + B_{12})$$

Strassen's Matrix Multiplication

■ Analysis of Strassen's Matrix Multiplication

✦ *Number of multiplications:*

$$M(n) = 7M(n/2),$$

$$M(1) = 1$$

If $n=2^k$, then

$$M(n)=7M(n/2)=7^2M(n/2^2)..... \\ =7^kM(1)=7^K$$

■ Solution:

$$M(n) = 7^{\log_2 n} = n^{\log_2 7} \approx n^{2.807} \quad \text{vs. } n^3 \text{ of brute-force alg.}$$

- If n is not a power of 2, matrices can be padded with zeros.
- ☆ Practical implementation of Strassen's alg. usually switch to brute-force method after matrix sizes become smaller than some “crossover point”

Strassen's Matrix Multiplication

■ *Analysis of Strassen's Matrix Multiplication*

✦ *Number of additions:*

$$A(n) = 7A(n/2) + 18(n/2)^2, \quad A(1) = 0$$

■ Solution:

$$A(n) = n^{\log_2 7}$$

Strassen's Matrix Multiplication

✦ Standard vs Strassen: Practical:

	N	Multiplications	Additions
Standard alg.	100	1,000,000	990,000
Strassen's alg.	100	411,822	2,470,334
Standard alg.	1000	1,000,000,000	999,000,000
Strassen's alg.	1000	264,280,285	1,579,681,709
Standard alg.	10,000	10^{12}	$9.99 \cdot 10^{11}$
Strassen's alg.	10,000	$0.169 \cdot 10^{12}$	10^{12}

Strassen's Matrix Multiplication

✦ More algorithms for matrix Multiplication:

- Algorithms with better asymptotic efficiency are known but they are even more complex.

时间	复杂度	作者
<1969	$O(n^3)$	
1969	$O(n^{2.81})$	<u>Strassen</u>
1978	$O(n^{2.79})$	Pan
1979	$O(n^{2.7799})$	<u>Bini, Lotti etc.</u>
1981	$O(n^{2.55})$	<u>Schonhage</u>
1984	$O(n^{2.52})$	Victor Pan
1987	$O(n^{2.48})$	<u>Strassen</u>
1987	$O(n^{2.376})$	Coppersmith and <u>Winograd</u>

Binary Search

问题描述:

已知一个按非降次序排列的元素表 a_0, a_1, \dots, a_{n-1} , 判定某个给定元素 K 是否在该表中出现。

■ Idea of Binary Search

若是, 则找出该元素在表中的位置并返回其所在位置的下标 j ; 否则, 返回值 -1。

Very efficient algorithm for searching in **sorted array** with non-decreasing order

K

vs

$A[0] \dots A[m] \dots A[n-1]$

选取一个下标 m , 可得到三个子问题:

$I_1 = (m, a_0, \dots, a_{m-1}, K)$

$I_2 = (1, a_m, K)$

$I_3 = (n-m-1, a_{m+1}, \dots, a_{n-1}, K)$

- If $K = A[m]$, stop (successful search);
- otherwise, continue searching by the same method
 - in $A[0..m-1]$ if $K < A[m]$
 - and in $A[m+1..n-1]$ if $K > A[m]$

问题的规模缩小到一定的程度就可以容易地解决;

该问题可以分解为若干个规模较小的相同问题;

分解出的子问题的解可以合并为原问题的解;

分解出的各个子问题是相互独立的。

如果对所求解的问题 (或子问题) 所选的下标 m 都是中间元素的下标, $m = \lfloor (l+r)/2 \rfloor$, 则由此产生的算法就是二分检索算法。

Binary Search

- *Example*

- $K=70$

0	1	2	3	4	5	6	7	8	9	10	11	12
3	14	27	31	39	42	55	70	74	81	85	93	98
<i>l</i>						<i>m</i>						<i>r</i>
							<i>l</i>		<i>m</i>			<i>r</i>
							<i>l,m</i>	<i>r</i>				

Binary Search

■ Example

If $A(1:9)=(-15, -6, 0, 7, 9, 23, 54, 82, 101)$

search in A: $k=101, -14, 82$ 。

Searching process:

k=101			k=-14			k=82		
low	high	mid	low	high	mid	low	high	mid
1	9	5	1	9	5	1	9	5
6	9	7	1	4	2	6	9	7
8	9	8	1	1	1	8	9	8
9	9	9	2	1				
找到			找不到			找到		

successful search

unsuccessful
search

successful search

Binary Search

■ Binary Search – an Iterative Algorithm

```
ALGORITHM BinarySearch(A[0..n-1], K)
   $l \leftarrow 0$ ;  $r \leftarrow n-1$ 
  while  $l \leq r$  do                                //  $l$  and  $r$  crosses over  $\rightarrow$  can't find K
     $m \leftarrow \lfloor (l+r)/2 \rfloor$ 
    if  $K = A[m]$  return  $m$                           // the key is found
    else if  $K < A[m]$   $r \leftarrow m-1$               // the key is on the left half of the array
    else  $l \leftarrow m+1$                           // the key is on the right half of the array
  return -1
```

Binary Search

■ Binary Search – a Recursive Algorithm

ALGORITHM BinarySearchRecur($A[0..n-1]$, l , r , K)

if $l > r$

 return -1

else

$m \leftarrow \lfloor (l + r) / 2 \rfloor$

 if $K = A[m]$

 return m

 else if $K < A[m]$

 return BinarySearchRecur($A[0..n-1]$, l , $m-1$, K)

 else

 return BinarySearchRecur($A[0..n-1]$, $m+1$, r , K)

Basic operation:

while循环中 k 与 A 中元素的比较运算

three-way comparison

Binary Search

■ Analysis of Binary Search

✦ *Basic operation:* key comparison (three-way comparison)

✦ *Worst-case* (successful or fail) :

$$\begin{aligned}C_w(n) &= C_w(\lfloor n/2 \rfloor) + 1, \\C_w(1) &= 1\end{aligned}$$

■ Solution:

$$C_w(n) = \Theta(\log n)$$

✦ *Best-case:*

■ successful

$$C_b(n) = 1$$

一次找到, 即 $K = A[n/2]$

for $n = 2^k$,

$$C(2^k) = C(2^{k-1}) + 1 \quad \text{for } k > 0$$

$$C(2^0) = 1$$

backward substitutions:

$$C(2^k) = C(2^{k-1}) + 1$$

$$= [C(2^{k-2}) + 1] + 1$$

$$= C(2^{k-2}) + 2 = \dots = C(2^{k-i}) + i \quad \dots$$

$$= C(2^{k-k}) + k = 1 + k$$

then, $C(n) = \log_2 n + 1$

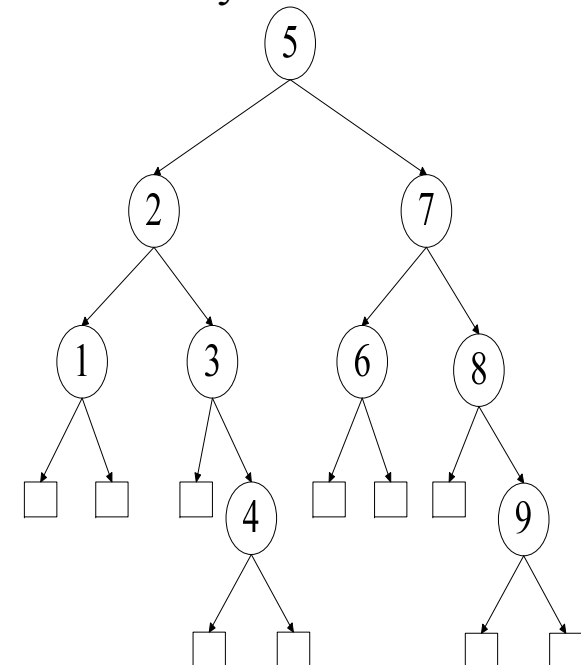
Binary Search

■ Analysis of Binary Search

✦ Average-case:

- Consider the searching process as a binary tree. In looking at the binary tree, we see that there are i comparisons needed to search 2^{i-1} elements on level i of the tree.
- For a list with $n = 2^k - 1$ elements, there are k levels in the binary tree.
- The average case for all successful search:

$$A(n) = \frac{1}{n} \sum_{i=1}^k i 2^{i-1} \approx \log(n+1) - 1$$



Binary Search

■ Analysis of Binary Search

✦ 区分以下情况进行分析

- 成功检索：指所检索的 K 恰好在 A 中出现
由于 A 中共有 n 个元素，故成功检索恰好有 n 种可能的情况
- 不成功检索：指 K 不出现在 A 中
根据取值，不成功检索共有 $n+1$ 种可能的情况 (取值区间)
 $K < A(1)$ 或 $A(i) < K < A(i+1)$, $1 \leq i < n-1$ 或 $K > A(n)$

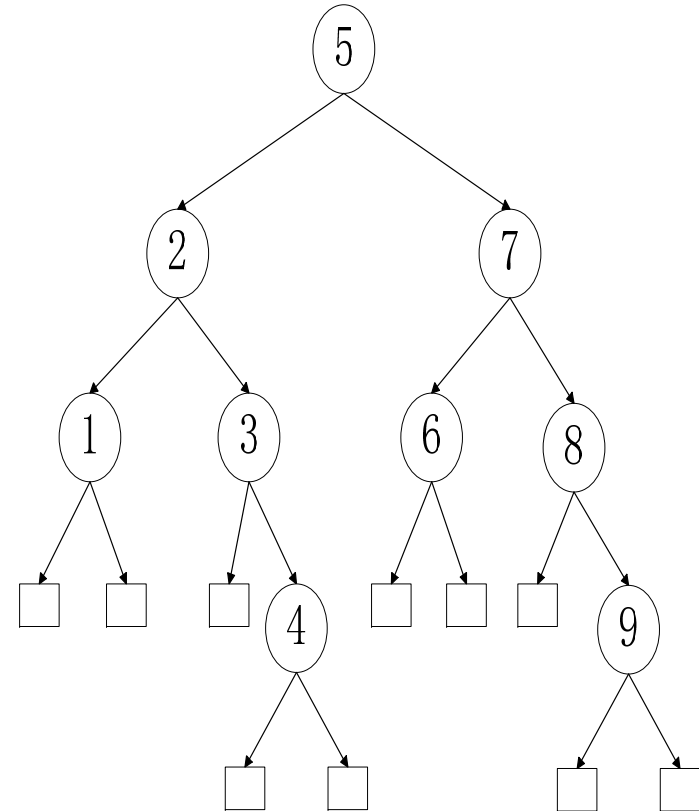
Binary Search

✦ 二元比较树

算法执行过程的主体是 k 与一系列中间元素 $A(\text{mid})$ 比较。

用一棵二元树描述该过程，称为二元比较树

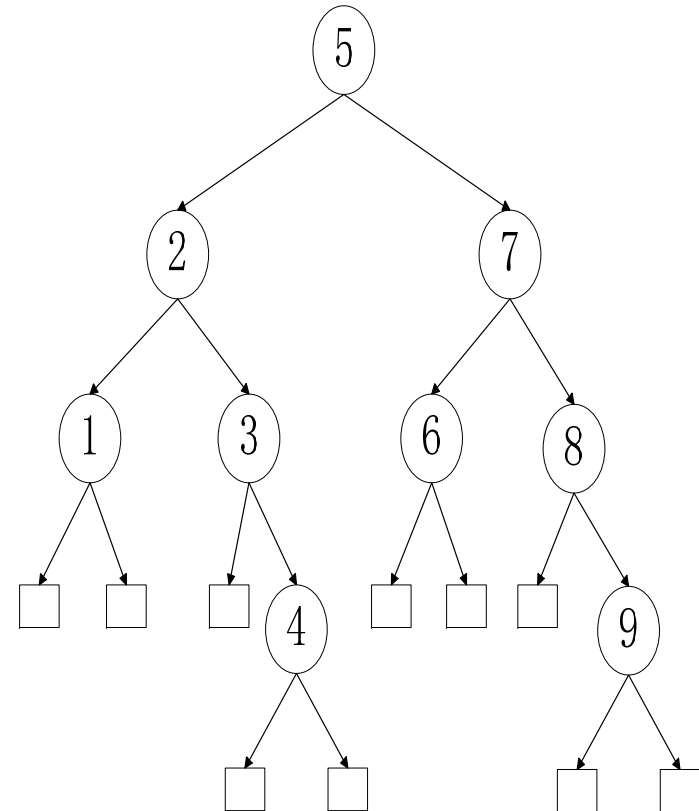
- 结点：
 - 内结点：
 - 代表一次元素比较
 - 用 圆形结点表示
 - 存放一个 mid 值(下标)
 - 代表成功检索情况
 - 外结点：
 - 用方形结点表示，
 - 表示不成功检索情况
- 路径：代表检索中元素的比较序列



Binary Search

- 二元比较树的查找过程
 - 若 K 在 A 中出现，则算法的执行过程在一个圆形的内结点处结束
 - 若 K 不在 A 中出现，则算法的执行过程在一个方形的内结点处结束

■ 注：外结点不代表元素的比较，因为比较过程在该外结点的上一级的内结点处结束。



Binary Search

■ Analysis of Binary Search

逐个查找法的复杂度AverageCase

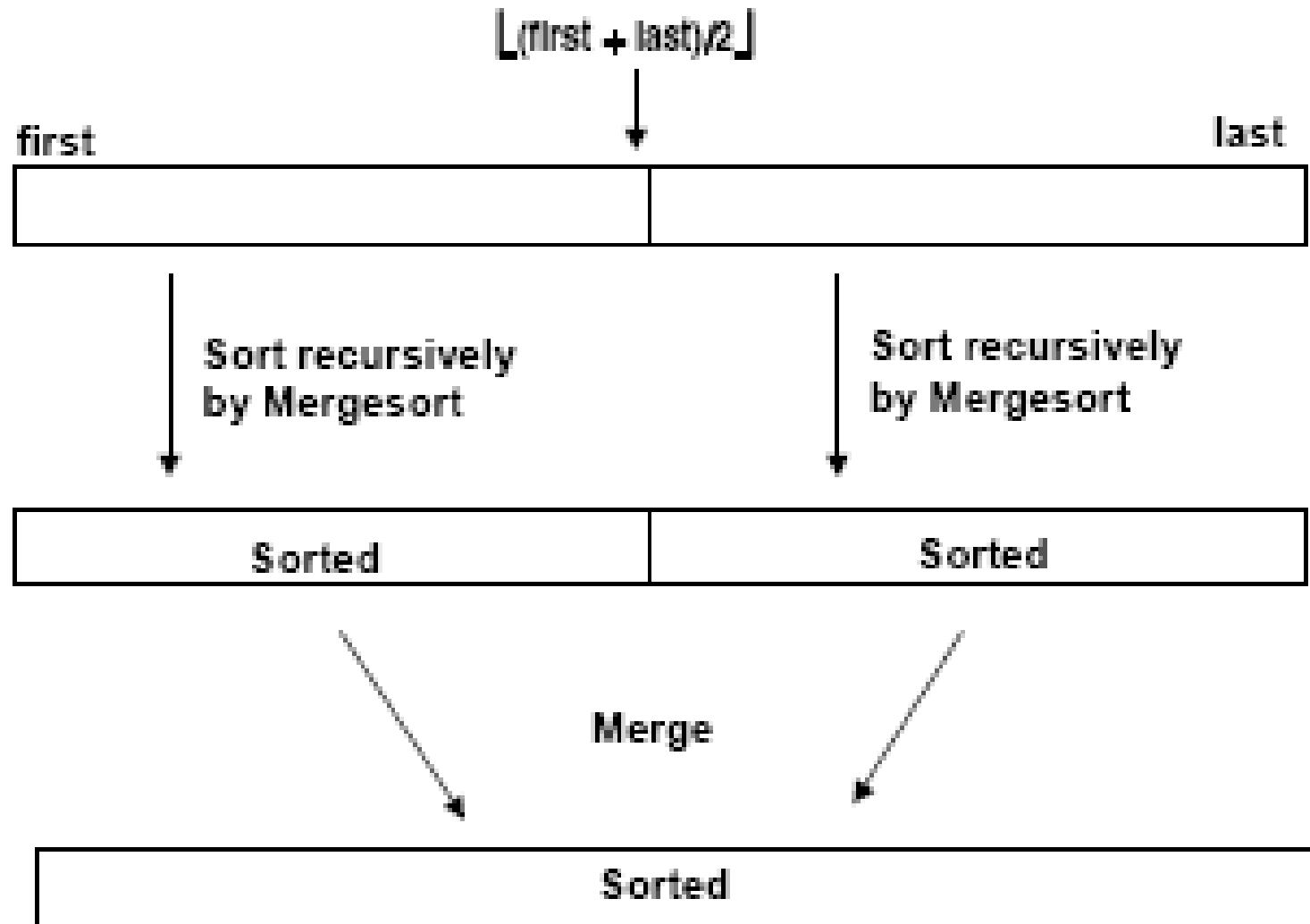
$$\begin{aligned} T_{avg}(n) &= \sum_{size(I)=n} p(I)T(I) \\ &= \left(1 \cdot \frac{p}{n} + 2 \cdot \frac{p}{n} + 3 \cdot \frac{p}{n} + \dots + n \cdot \frac{p}{n} \right) + n \cdot (1 - p) \\ &= \frac{p}{n} \sum_{i=1}^n i + n(1 - p) = \frac{p(n+1)}{2} + n(1 - p) \end{aligned}$$

Mergesort

■ Idea of Mergesort

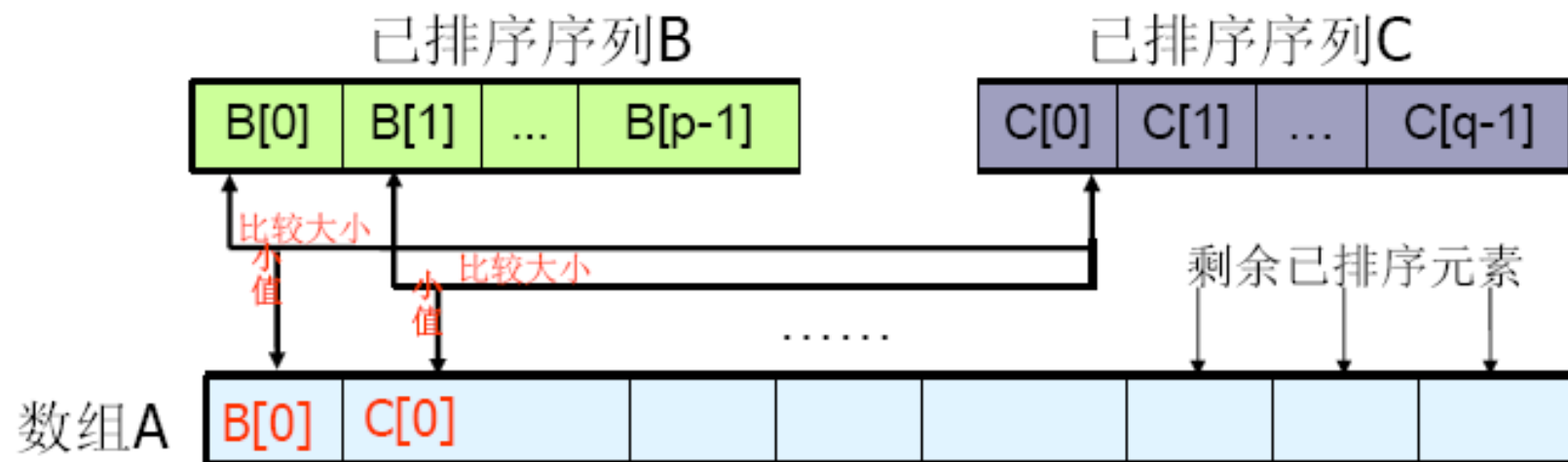
- ✦ *Divide:* divide array $A[0..n-1]$ in two *about equal* halves and make copies of each half in arrays B and C
- ✦ *Conquer:*
 - If number of elements in B and C is 1, directly solve it
 - Sort arrays B and C recursively
- ✦ *Combine:* Merge sorted arrays B and C into a single sorted A
 - Repeat the following until no elements remain in one of the arrays:
 - compare the first elements in the remaining unprocessed portions of the arrays B and C
 - copy the smaller of the two into A , while incrementing the index indicating the unprocessed portion of that array
 - Once all elements in one of the arrays are processed, the remaining unprocessed elements from the other array are copied into the end of A .

Mergesort



Mergesort

MERGE



Mergesort

■ The Mergesort Algorithm

ALGORITHM *Mergesort*($A[0..n - 1]$)

//Sorts array $A[0..n - 1]$ by recursive mergesort
//Input: An array $A[0..n - 1]$ of orderable elements
//Output: Array $A[0..n - 1]$ sorted in nondecreasing order

if $n > 1$

 copy $A[0..\lfloor n/2 \rfloor - 1]$ to $B[0..\lfloor n/2 \rfloor - 1]$

 copy $A[\lfloor n/2 \rfloor..n - 1]$ to $C[0..\lceil n/2 \rceil - 1]$

Mergesort($B[0..\lfloor n/2 \rfloor - 1]$)

Mergesort($C[0..\lceil n/2 \rceil - 1]$)

Merge(B, C, A)

中分点 $\text{mid} \leftarrow \lfloor (\text{low} + \text{high}) / 2 \rfloor$

Mergesort

■ The Mergesort Algorithm ('cont)

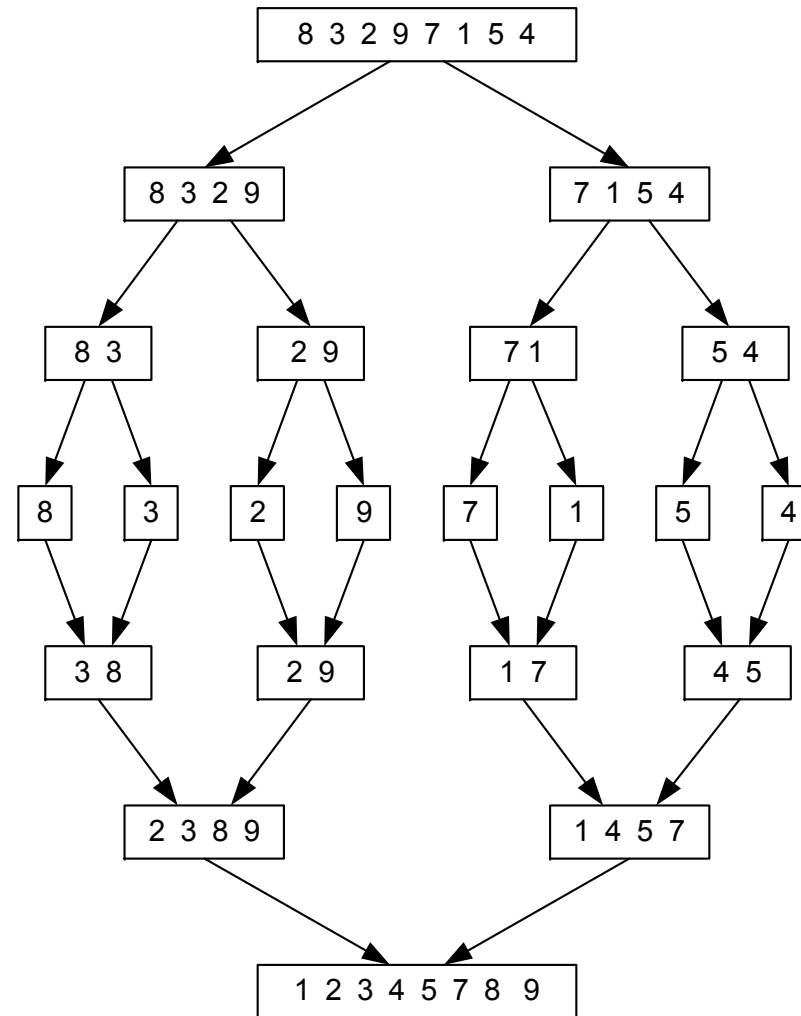
ALGORITHM *Merge*($B[0..p-1]$, $C[0..q-1]$, $A[0..p+q-1]$)

//Merges two sorted arrays into one sorted array
//Input: Arrays $B[0..p-1]$ and $C[0..q-1]$ both sorted
//Output: Sorted array $A[0..p+q-1]$ of the elements of B and C
 $i \leftarrow 0$; $j \leftarrow 0$; $k \leftarrow 0$
while $i < p$ **and** $j < q$ **do**
 if $B[i] \leq C[j]$
 $A[k] \leftarrow B[i]$; $i \leftarrow i + 1$
 else $A[k] \leftarrow C[j]$; $j \leftarrow j + 1$
 $k \leftarrow k + 1$
if $i = p$
 copy $C[j..q-1]$ to $A[k..p+q-1]$
else copy $B[i..p-1]$ to $A[k..p+q-1]$

Mergesort

■ *Example:*

- 8 3 2 9 7 1 5 4



Mergesort

■ Analysis of Mergesort

✦ *Number of basic operations (key comparisons):*

$$C(n) = 2C(n/2) + C_{\text{merge}}(n) \text{ for } n > 1$$

$$C(1) = 0$$

$$\text{Where } C_{\text{merge}}(n) = n - 1$$

✦ **$C(n) = \Theta(n \log n)$**

If $n = 2^k$

divide : $D(n) = \Theta(1)$

conquer: $T(n) = 2T(n/2)$, $T(1) = \Theta(1)$

merge: $T(n) = \Theta(n)$

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

$$= 4T(n/4) + 2\Theta(n/2) + \Theta(n)$$

$$= 8T(n/8) + 4\Theta(n/4) + 2\Theta(n/2) + \Theta(n)$$

$$= 2^{\log n} T(1) + \Theta(n) + \Theta(n) \dots + \Theta(n) \text{ [一共 } \log n \text{ 个 } \Theta(n)]$$

$$= \Theta(n) + \log n \Theta(n) = \Theta(n \log n)$$

If $2^k < n < 2^{k+1}$, 则有 $T(n) \leq T(2^{k+1})$ 。

Quicksort

■ Idea of Quicksort

----- Partition

- 快速分类是一种基于划分的分类方法;

✦ **Divide:** Partition array $A[l..r]$ into 2 subarrays, $A[l..s-1]$ and $A[s+1..r]$ such that each element of the first array is $\leq A[s]$ and each element of the second array is $\geq A[s]$. (computing the index of s is part of partition.)

- Implication: $A[s]$ will be in its final position in the sorted array.

- 快速分类: 通过反复地对待排序集合进行划分达到分类目的的分类算法。

✦ **Conquer:**

- Sort the two subarrays $A[l..s-1]$ and $A[s+1..r]$ by recursive calls to quicksort

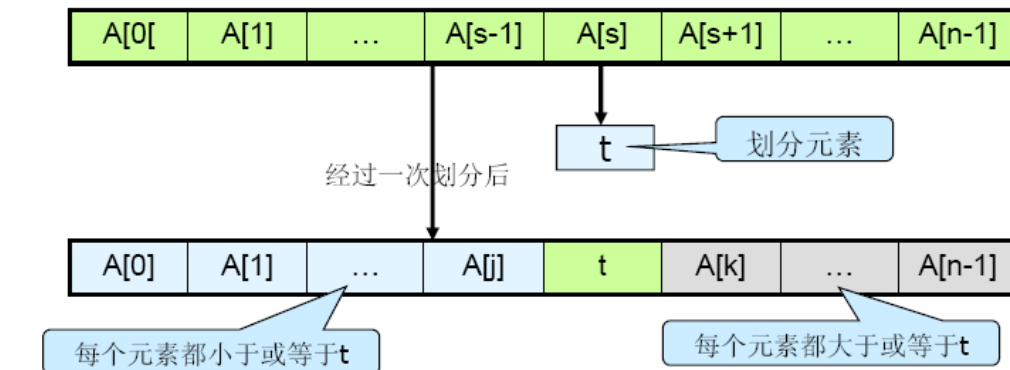
- $A[l..s-1]$ 中所有元素小于等于 $A[s+1..r]$ 中任何元素, 所以这两个集合可独立进行划分

✦ **Combine:** No work is needed, because $A[s]$ is already in its correct place after the partition is done, and the two subarrays have been sorted.

Quicksort

■ Idea of Quicksort ('cont)

- ✦ Select a pivot w.r.t. whose value we are going to divide the list. (typically, $p = A[l]$)
- ✦ Rearrange the list so that
 - all elements in the first s positions are smaller than or equal to the pivot
 - all elements in the remaining $n-s$ positions are larger than or equal to p



- ✦ Exchange the pivot with the last element in the first sublist (i.e., \leq sublist)
 - the pivot is now in its final position
- ✦ Sort the two sublists recursively using quicksort.

Quicksort

■ Idea of Quicksort ('cont)

✦ *Strategy for pivot selection*

- *Randomly selected*
- *Simplest Strategy: selecting the array's first element $A[l]$*

Quicksort

■ Idea of Quicksort ('cont)

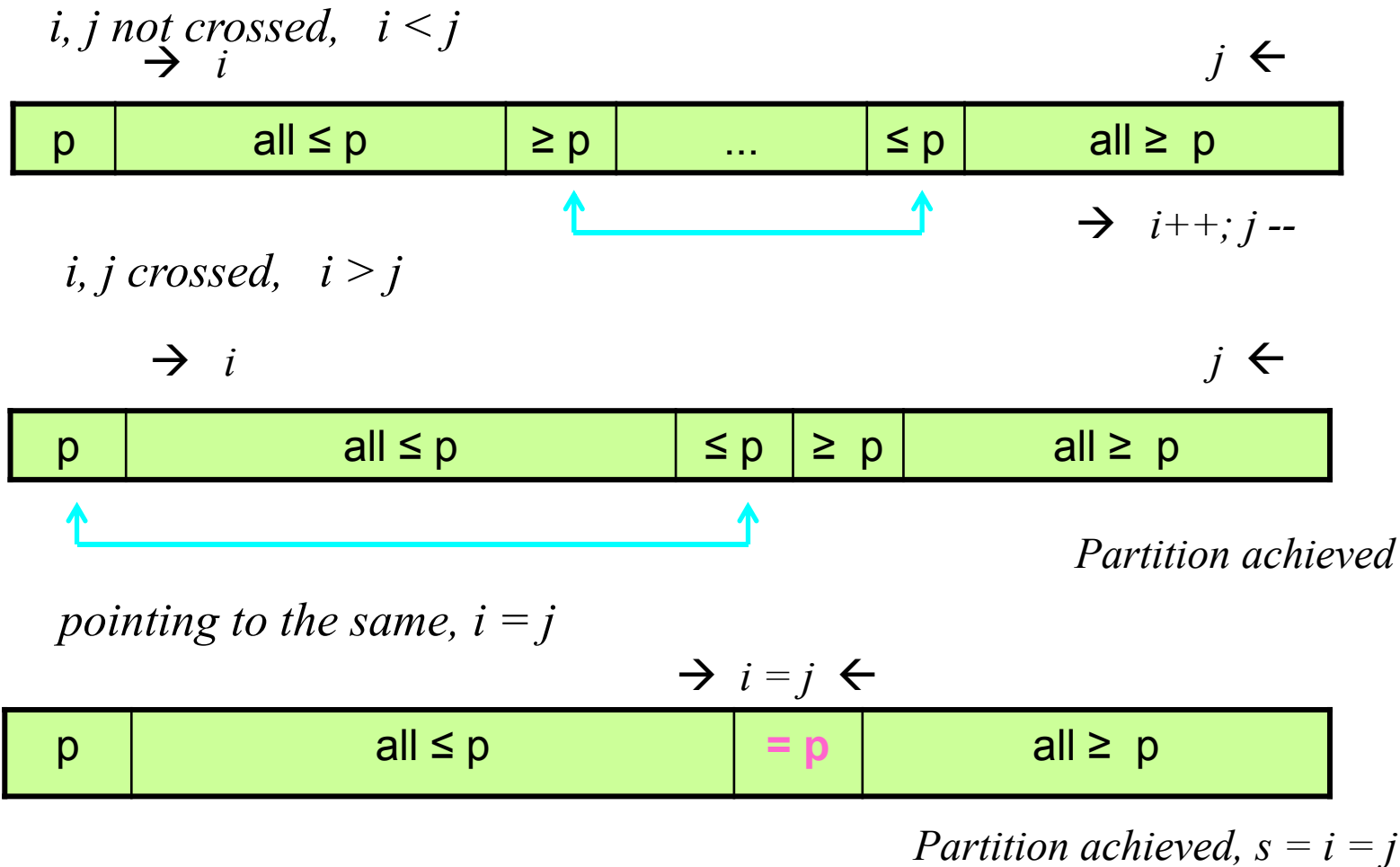
✦ Procedure for rearranging elements in a partition

----- based on two-scans of the subarray

- *Left-to-right scan: index i , starts with the second element,*
 - *Wants elements smaller than the pivot to be in the first part*
 - *Skip over elements that are smaller than the pivot*
 - *Stop on encountering the first element greater than or equal to pivot*
- *Right-to-left scan: index j , starts with the last element,*
 - *Wants elements larger than the pivot to be in the second part of the subarray*
 - *Skip over elements that are larger than the pivot*
 - *Stop on encountering the first element smaller than or equal to pivot*

Quicksort

- three cases for scan stopping



Quicksort

■ The Quicksort Algorithm

```
ALGORITHM Quicksort( $A[l..r]$ )  
//Sorts a subarray by quicksort  
//Input: A subarray  $A[l..r]$  of  $A[0..n-1]$ , defined by its left and right  
//       indices  $l$  and  $r$   
//Output: The subarray  $A[l..r]$  sorted in nondecreasing order  
if  $l < r$   
     $s \leftarrow \text{Partition}(A[l..r])$  //  $s$  is a split position  
    Quicksort( $A[l..s-1]$ )  
    Quicksort( $A[s+1..r]$ )
```

Quicksort

■ The Quicksort Algorithm - Partitioning

```
template<class Type>
int Partition (Type a[], int l, int r)
{
    int i = l, j = r + 1;
    Type x=a[l];
    // 将< x的元素交换到左边区域
    // 将> x的元素交换到右边区域
    while (true) {
        while (a[++i] <x);
        while (a[--j] >x);
        if (i >= j) break;
        Swap(a[i], a[j]);
    }
    //将x交换到它在排序序列中应在的位置上
    a[l] = a[j];
    a[j] = x;
    return j;
}
```

Number of comparisons:
 $n + 1$ (if indices i, j , cross over)
 n (if indices i, j , coincide)

Quicksort

Example

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	i	j
A:	65	70	75	80	85	60	55	50	45	$+\infty$	2	9
											
A:	65	45	75	80	85	60	55	50	70	$+\infty$	3	8
											
A:	65	45	50	80	85	60	55	75	70	$+\infty$	4	7
											
A:	65	45	50	55	85	60	80	75	70	$+\infty$	5	6
											
A:	65	45	50	55	60	85	80	75	70	$+\infty$	6	5
					j.....i							
A:	60	45	50	55	65	85	80	75	70	$+\infty$		
					↑							

交换
划分
元素



划分元素位于此

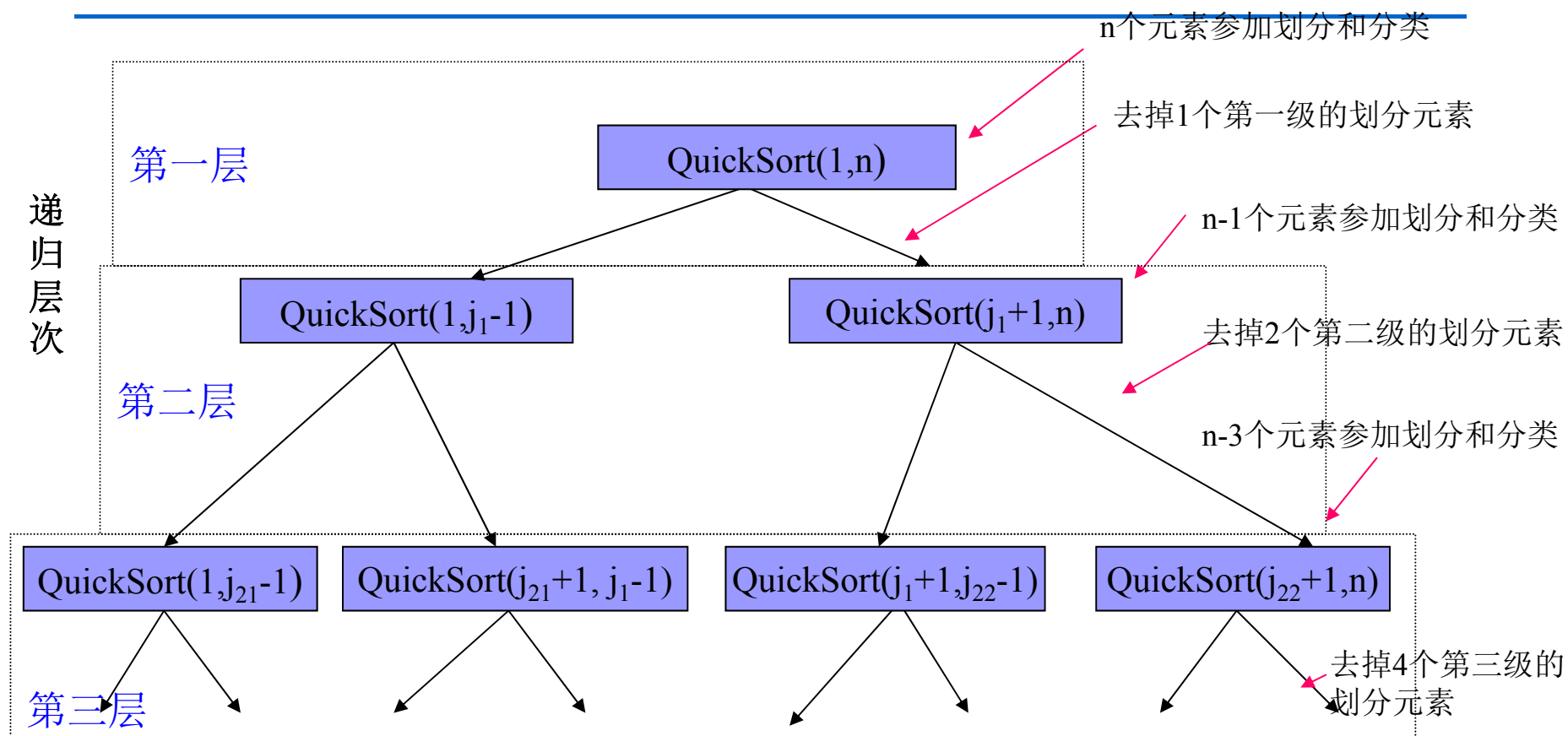
- 经过一次“划分”，实现了对集合元素的调整：以划分元素为界，左边子集合的所有元素均小于等于右边子集合的所有元素。
- 按同样的策略对两个子集合进行分类处理。当子集合分类完毕后，整个集合的分类也完成了。这一过程避免了子集合的归并操作。

Quicksort

■ **Analysis of Quicksort**

- ✦ *basic operation: key comparison*
- ✦ *Based on whether the partitioning is balanced.*

Quicksort



设在任一级递归调用上，调用PARTITION处理的所有元素总数为 r ，则，初始时 $r=n$ ，以后的每级递归上，由于删去了上一级的划分元素，故 r 比上一级至少1：

理想情况，第一级少1，第二级少2，第三级少4， ...；

最坏情况，每次仅减少1（每次选取的划分元素刚好是当前集合中最小或最大者）

Quicksort

■ **Analysis of Quicksort**

Number of comparisons for a partition:

$n + 1$ (if indices i, j , cross over)

n (if indices i, j , coincide)

✦ *Best case:* split in the middle — $\Theta (n \log n)$

$C_b(n) = 2C_b(n/2) + \Theta (n)$ //2 subproblems of size $n/2$ each

$C_b(1) = 0$

for $n = 2^k$, backward substitutions, could get it

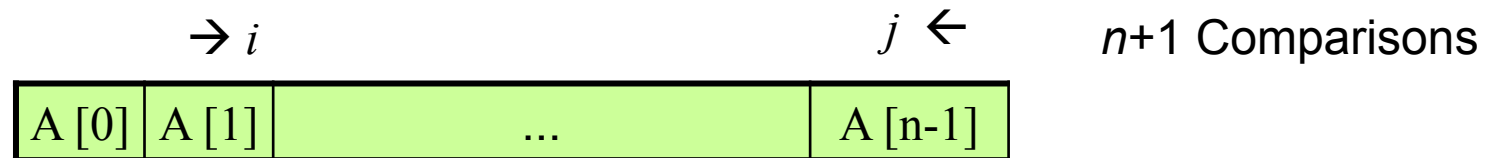
Quicksort

■ Analysis of Quicksort

✦ *Worst case: sorted array! — $\Theta(n^2)$*

$$C_w(n) = C_w(n-1) + \Theta(n) \quad //2 \text{ subproblems of size } 0 \text{ and } n-1$$

A[0...n-1] is a strictly increasing array, and A[0] is used as pivot, the left-to-right scan stops on A[1], right-to-left scan goes all the way to A[0],



$$C_w = (n+1) + n + \dots + 3 = (n+1)(n+2)/2 - 3 = \Theta(n^2)$$

Quicksort

■ Analysis of Quicksort

✦ *Average case: random arrays — $O(n \log n)$*

Partition split in each position $s \in [0, n-1]$, with the same probability $1/n$

$$\begin{aligned}C_{avg}(n) &= \frac{1}{n} \sum_{s=0}^{n-1} [(n+1) + C_{avg}(s) + C_{avg}(n-1-s)] \\C_{avg}(0) &= 0 \\C_{avg}(1) &= 0\end{aligned}$$

Solution

$$C_{avg}(n) \approx 2n \ln n \approx 1.38n \log_2 n$$

- ☆ quicksort makes only 38% more comparisons than in the best case
- ☆ its innermost loop is so efficient that it runs faster than mergesort on randomly ordered arrays

Quicksort

■ Improvements

- 快速排序算法的性能取决于划分的对称性。通过修改算法partition，可以设计出采用随机选择策略的快速排序算法。

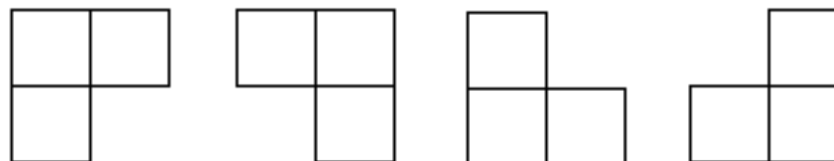
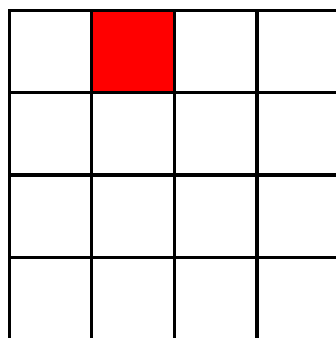
- 随机选取划分元素

在快速排序算法的每一步中，当数组还没有被划分时，可以在a[p:r]中随机选出一个元素作为划分基准，这样可以使划分基准的选择是随机的，从而可以期望划分是较对称的。

```
template<class Type>
int RandomizedPartition (Type a[], int p, int r)
{
    int i = Random(p,r);
    Swap(a[i], a[p]);
    return Partition (a, p, r);
}
```

棋盘覆盖

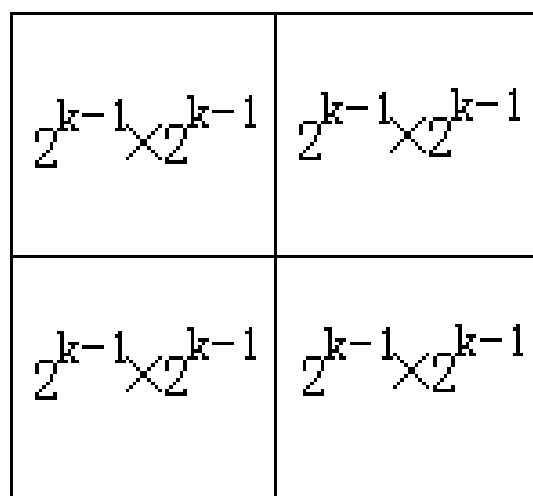
在一个 $2^k \times 2^k$ 个方格组成的棋盘中，恰有一个方格与其它方格不同，称该方格为一特殊方格，且称该棋盘为一特殊棋盘。在棋盘覆盖问题中，要用图示的4种不同形态的L型骨牌覆盖给定的特殊棋盘上除特殊方格以外的所有方格，且任何2个L型骨牌不得重叠覆盖。



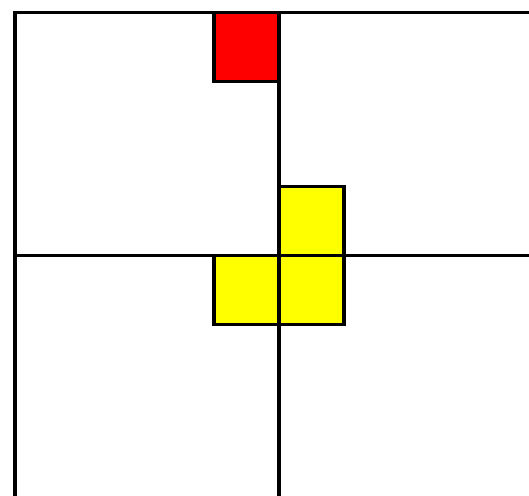
棋盘覆盖

分治策略：

特殊方格必位于4个较小子棋盘之一中，其余3个子棋盘中无特殊方格。为了将这3个无特殊方格的子棋盘转化为特殊棋盘，可以用一个L型骨牌覆盖这3个较小棋盘的会合处，从而将原问题转化为4个较小规模的棋盘覆盖问题。递归地使用这种分割，直至棋盘简化为棋盘 1×1 。

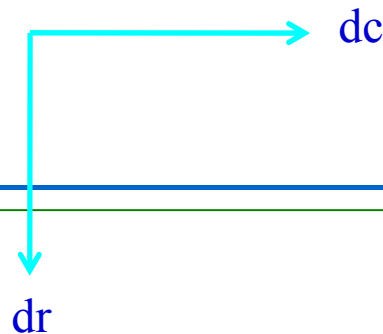


(a)



(b)

棋盘覆盖



```
void chessBoard(int tr, int tc, int dr, int dc, int size)
{
    if (size == 1) return;
    int t = tile++; // L型骨牌号
    s = size/2; // 分割棋盘
    // 覆盖左上角子棋盘
    if (dr < tr + s && dc < tc + s) // 特殊方格在此棋盘中
        chessBoard(tr, tc, dr, dc, s);
    else { // 此棋盘中无特殊方格,则用 t 号L型骨牌覆盖该子棋盘的右下角
        board[tr + s - 1][tc + s - 1] = t;
        chessBoard(tr, tc, tr+s-1, tc+s-1, s);}

    // 覆盖右上角子棋盘
    if (dr < tr + s && dc >= tc + s) // 特殊方格在此棋盘中
        chessBoard(tr, tc+s, dr, dc, s);
    else { // 此棋盘中无特殊方格,则用 t 号L型骨牌覆盖左下角
```

棋盘覆盖

```
board[tr + s - 1][tc + s] = t;
chessBoard(tr, tc+s, tr+s-1, tc+s, s);}

// 覆盖左下角子棋盘
if (dr >= tr + s && dc < tc + s)      // 特殊方格在此棋盘中
    chessBoard(tr+s, tc, dr, dc, s);
else {      //此棋盘中无特殊方格,则用 t 号L型骨牌覆盖右上角
    board[tr + s][tc + s - 1] = t;
    chessBoard(tr+s, tc, tr+s, tc+s-1, s);}

// 覆盖右下角子棋盘
if (dr >= tr + s && dc >= tc + s)      // 特殊方格在此棋盘中
    chessBoard(tr+s, tc+s, dr, dc, s);
else {      //此棋盘中无特殊方格,则用 t 号L型骨牌覆盖左上角
    board[tr + s][tc + s] = t;
    chessBoard(tr+s, tc+s, tr+s, tc+s, s);}
}
```

棋盘覆盖

✦ 复杂度分析

$$T(k) = \begin{cases} O(1) & k = 0 \\ 4T(k-1) + O(1) & k > 0 \end{cases}$$

$T(n)=O(4^k)$ 渐进意义下的最优算法

覆盖 $2^k \times 2^k$ 的棋盘所需L型骨牌个数为 $(4^k-1)/3$

Concluding

- ✦ 分治法是一种一般性的算法设计技术，他将问题的实例划分为若干个较小的实例(最好用有相同的规模)，对这些小的问题求解，然后合并这些解，得到原始问题的解。
- ✦ 分治法的时间效率满足： $T(n) = aT(n/b) + f(n)$
- ✦ 合并排序是一种分治排序算法，任何情况下，该算法的时间效率都是 $\Theta(n \log n)$ ，它的键值比较次数非常接近理论的最小值，缺点是需要大量的额外存储空间。
- ✦ 快速排序也是一种分治排序算法，具有出众的时间效率 $n \log n$ ，最差效率是平方级的。
- ✦ 折半查找是一种对有序数组进行查找的算法，效率为 $\log n$
- ✦ n 位大整数乘法的分治算法，大约需要做 $n^{1.585}$ 次乘法。

思考题

1. 设 $a[0:n-1]$ 是一个已排好序的数组。设计搜索算法，使得当搜索元素在数组中时， i 和 j 相同，均为 x 在数组中的位置；搜索元素 x 不在数组中时，返回小于 x 的最大元素的位置 i 和大于 x 的最小元素位置 j 。

并对自己的程序进行复杂性分析。

2. 给定2个大整数 u 和 v ，分别有 m 位和 n 位数字，且 $m \leq n$ 。用通常的乘法求 uv 的值需要 $O(mn)$ 时间。当 m 比 n 小得多时，试设计一个算法，在上述情况下用 $O(nm^{\log(3/2)})$ 时间求出 uv 值。