

# *Analysis and Design of Algorithms*

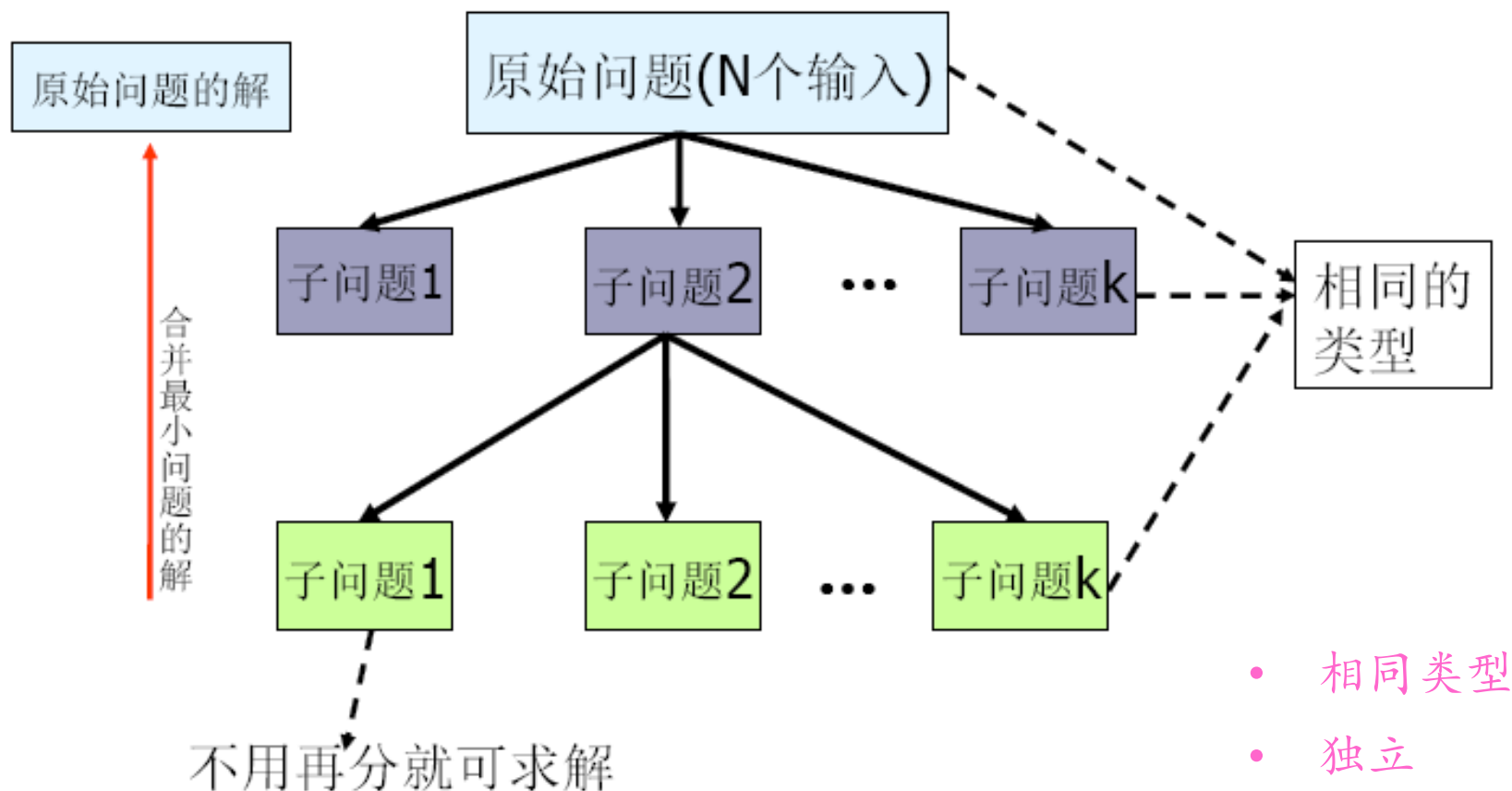
## Chapter 5: Divide and Conquer



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# Divide and Conquer Approach

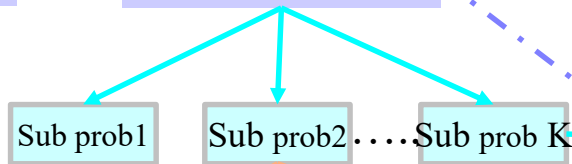


- 相同类型
- 独立
- 规模均衡

# Divide and Conquer Approach

Solutions → Original Problem(N input)

Combine the solutions



Share the same type



Sub prob1 Sub prob2 ....Sub prob M

- Share the same type
- Dependent
- Similar size of input

Can't be divided any more



分而治之，  
逐个击破。

# *Divide and Conquer Approach*

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## ■ **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*

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## ■ *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  $b$** )*
- *$a$  is the number of the problems needs of be solved*
- *$f(n)$  is a function that counts for the time spent on dividing the problem into smaller ones and on combing their solutions*

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

# Multiplication of Large Integers

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## ■ Idea of Multiplication of Large Integers :karatsuba

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

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## ▣ 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

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

**Idea of the algorithm:**

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

# Multiplication of Large Integers

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## ■ 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

① If  $n=2^k$ , then

$$\begin{aligned} C(2^k) &= 4C(2^{k-1}) = 4[4C(2^{k-2})] = 4^2C(2^{k-2}) \\ &= \dots = 4^k C(2^{k-k}) = 4^k \end{aligned}$$

Given  $n=2^k$ , we have  $k=\log_2 n$ ;

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

② If  $n \neq 2^k$ , by Smoothness Rule,  $C(n) \in O(n^2)$



# *Multiplication of Large Integers*


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## *Second Divide-and-Conquer Algorithm*

$$X * Y = A * C \cdot 10^n + (A * D + B * C) \cdot 10^{n/2} + B * D$$

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

Since we have:  $(A + B) * (C + D) = AC + (AD + BC) + BD$

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

  $X * Y = A * C \cdot 10^n + [(A + B) * (C + D) - A * C - B * D] \cdot 10^{n/2} + B * D$

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

# Multiplication of Large Integers

## ■ Second Divide-and-Conquer Algorithm

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

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

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

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

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

$$k = \log_2 n$$

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

Solution:

$$T(n) = O(3^{\log 2^n}) = O(n^{\log 2^3}) \approx O(n^{1.585}) \checkmark \text{ Promotion}$$

# *Multiplication of Large Integers*

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☆  $X = a * 2^{n/2} + b; Y = c * 2^{n/2} + d;$

☆  $X * Y = (a * 2^{n/2} + b) * (c * 2^{n/2} + d) = ac * 2^n + (a * d + b * c) * 2^{n/2} + bd$

☆ 因为  $a * d + b * c = (a + c)(b + d) - ac - bd$

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

☆ 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。

# Definition of Matrix Multiplication

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## ❏ Idea

✦ *brute-force alg.:*  $O(n^3)$

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

For  $C_{ij}$ ,  **$n$**  multiplications and  **$n-1$**  summations

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

# Strassen's Matrix Multiplication

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## ❏ Idea

✦ *Divide and Conquer— idea1*

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

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

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

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

# Strassen's Matrix Multiplication

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## ❏ 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

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## ■ Analysis of Strassen's Matrix Multiplication

✦ Number of multiplications:

$$M(n) = 7M(n/2),$$
$$M(1) = 1$$

If  $n=2^k$ , then

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

■ 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

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## ✦ 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

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## ✦ 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

$K$

选取一个下标 $m$ , 可得到三个子问题:

vs

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

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

$I_2 = (1, a_m, K)$

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

- 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  $K > A[m]$

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

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- *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

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## ■ 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

Basic operation:

## ■ Binary Search – a Recursive Algorithm

while循环中while中元素的比较运算

three-way comparison

```
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)
```

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

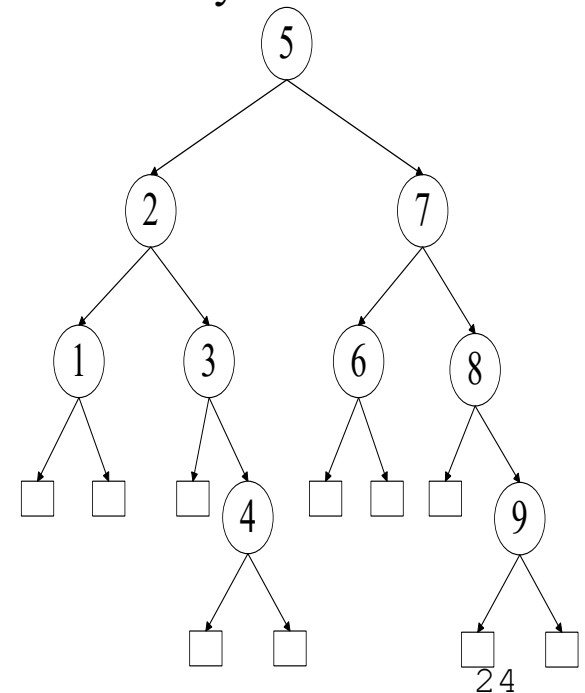
# Binary Search

## ■ 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

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## ■ Analysis of Binary Search

### ✦ 区分以下情况进行分析

1. 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
2. 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：内结点和外结点

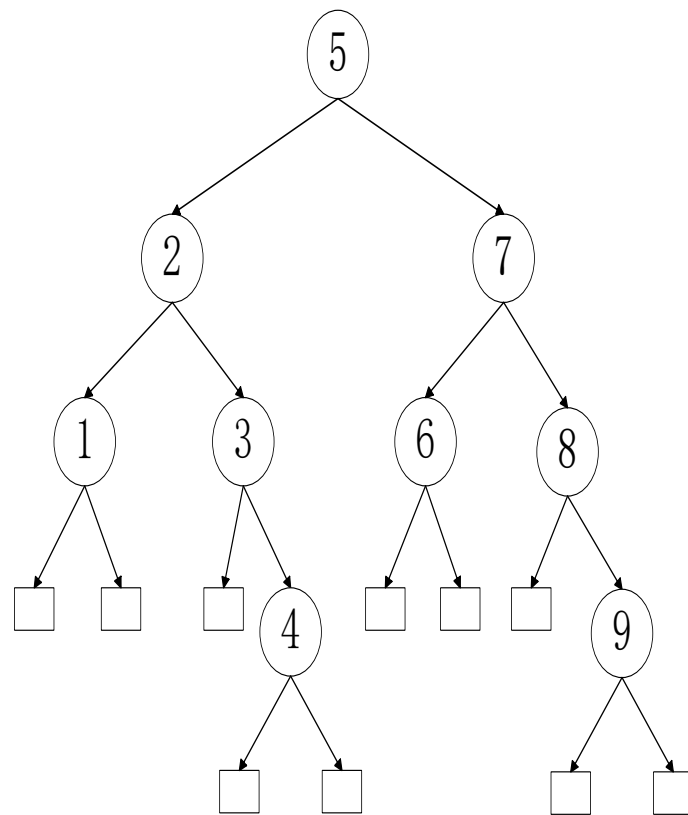
#### • 内结点：

- 代表一次元素比较
- 用 圆形结点表示
- 存放一个  $\text{mid}$ 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

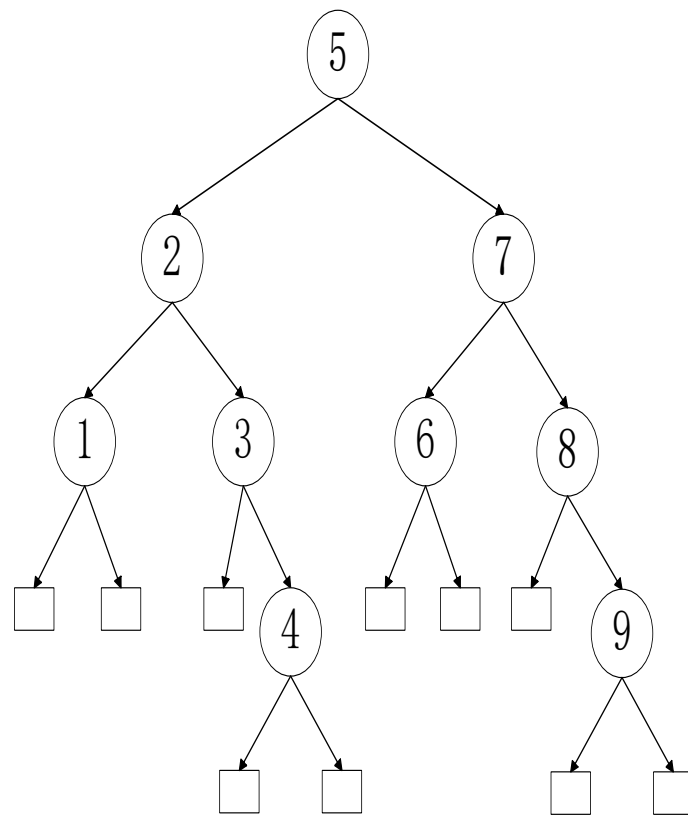
### ■ 路径：代表检索中元素的比较序列



# 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**外结点**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

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## ■ 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

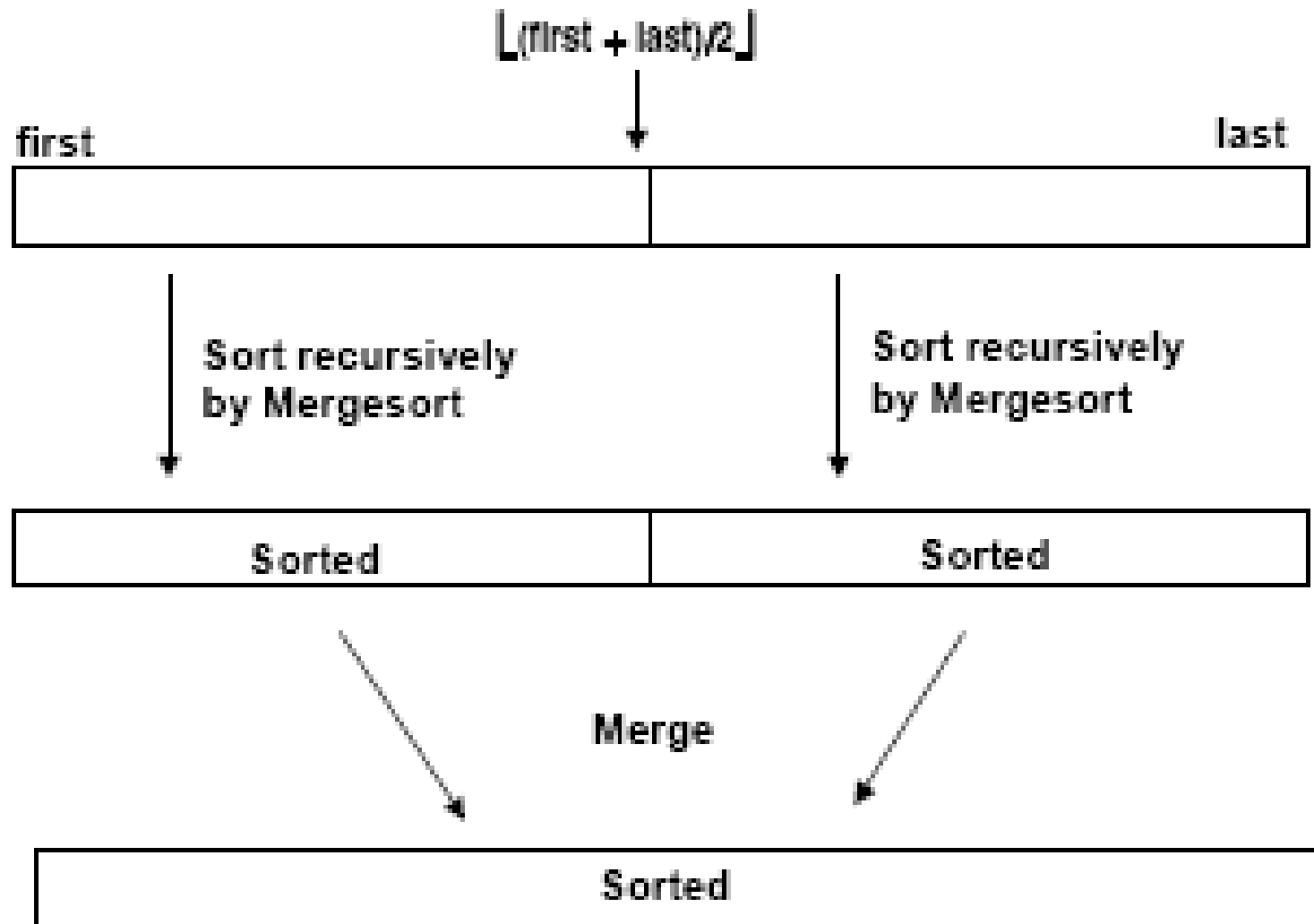
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## ■ Idea of Mergesort

- ✦ **Step1 Divide:** divide array  $A[0..n-1]$  in two *about equal* halves and make copies of each half in arrays  $B$  and  $C$
- ✦ **Step 2 Conquer:**
  - If number of elements in  $B$  and  $C$  is 1, directly solve it (go to step 3)
  - Sort arrays  $B$  and  $C$  recursively (go to step 1)
- ✦ **Step 3 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  $A$ .

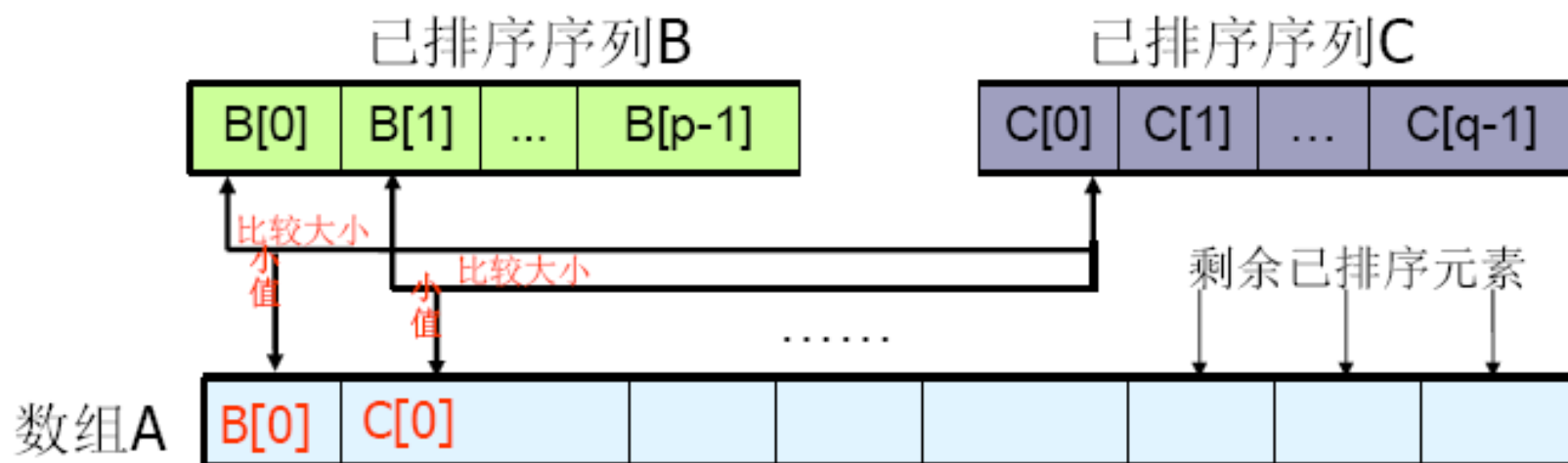
# Mergesort

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# 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

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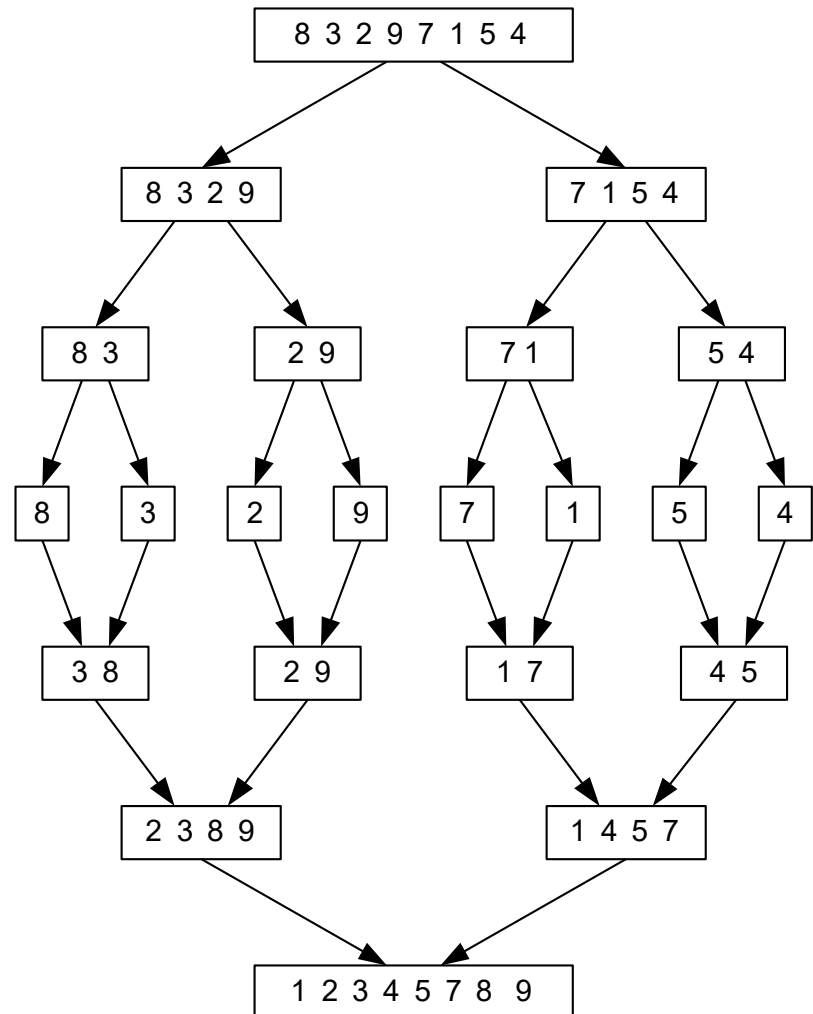
## ■ The Mergesort Algorithm ('cont)

**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

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## ■ **Analysis of Mergesort**

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

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

$$C(1) = 0$$

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

✦  **$C(n) = \Theta(n \log n)$**

# 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.

✦ **Conquer:**

- 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。

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

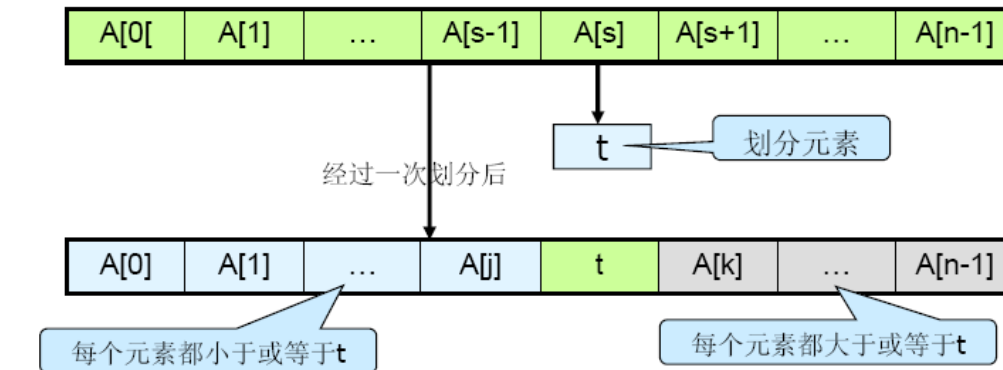
- $A[l..s-1]$  中所有元素小于等于  $A[s+1..r]$  中任何元素，所以这两个集合可独立进行划分

✦ **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) Pseudo Code of the Algorithm

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



3. After exchanging  $p$  with the last element in the first sublist (i.e.,  $\leq$  sublist), the pivot  $p$  is now in its final position
4. Sort the two sublists recursively using quicksort (i.e. **go to step 1**).

# Quicksort

---

## ■ Idea of Quicksort ('cont)

### ✦ Strategy for pivot selection

1. *Randomly selected*
2. *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

1. 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
2. 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

$i, j$  not crossed,  $i < j$   
 $\rightarrow i$



$\rightarrow i++; j--$

$i, j$  crossed,  $i > j$

$\rightarrow i$

$j \leftarrow$



Partition achieved

pointing to the same,  $i = j$

$\rightarrow i = j \leftarrow$



Partition achieved,  $s = i = j$



# 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* A : **65** 70 75 80 85 60 55 50 45

	(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



划分元素定位于此

# Quicksort

---

## ■ **Analysis of Quicksort**

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

# 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* —  $\Theta ( n \log n )$

$$C_b(n) = 2C_b(n/2) + \Theta (n) \quad //2 \text{ subproblems of size } n/2 \text{ each}$$

$$C_b(1) = 0$$

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

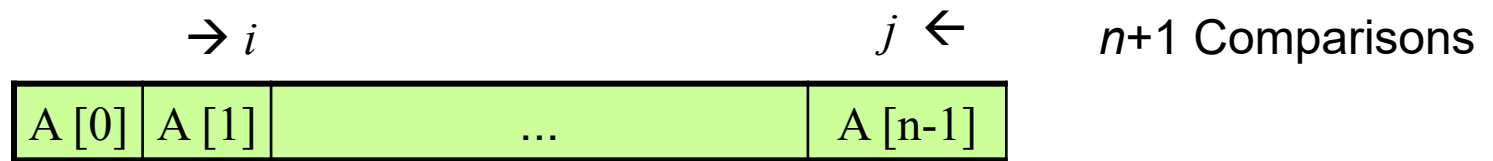
# Quicksort

## ■ 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 —  $O(n \log n)$

# 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。

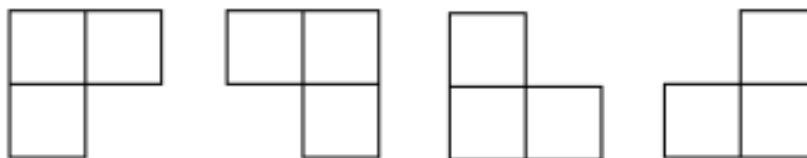
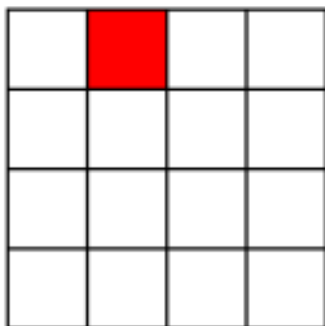
```
template<class Type>
int RandomizedPartition (Type a[], int p, int r)
{
    int i = Random(p,r);
    Swap(a[i], a[p]);
    return Partition (a, p, r);
}
```



# 棋盘覆盖 (Chessboard Cover) 问题

---

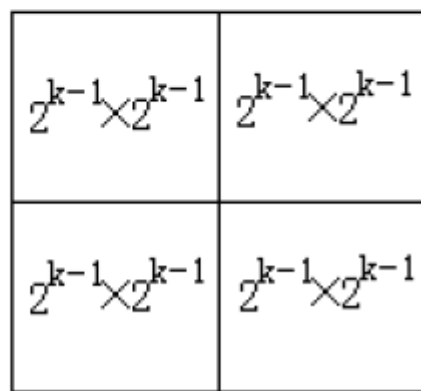
在一个 $2^k \times 2^k$ 个方格组成的棋盘中，恰有一个方格与其它方格不同，称该方格为一特殊方格，且称该棋盘为一特殊棋盘。在棋盘覆盖问题中，要用图示的4种不同形态的L型骨牌覆盖给定的特殊棋盘上除特殊方格以外的所有方格，且任何2个L型骨牌不得重叠覆盖。



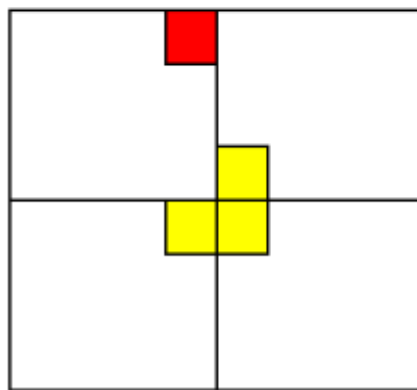
# 棋盘覆盖问题

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 $1 \times 1$ 。



(a)



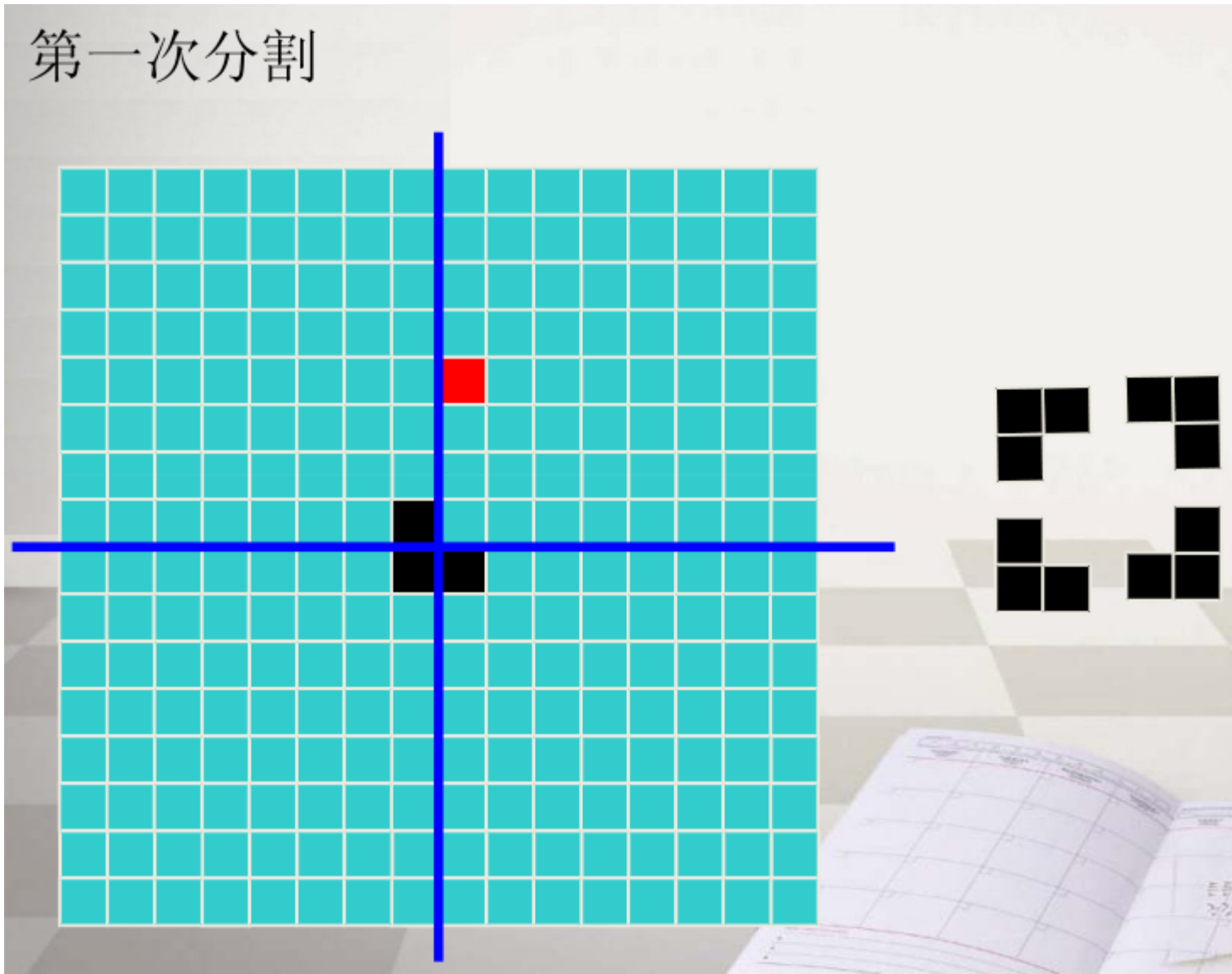
(b)

52

# 棋盘覆盖问题

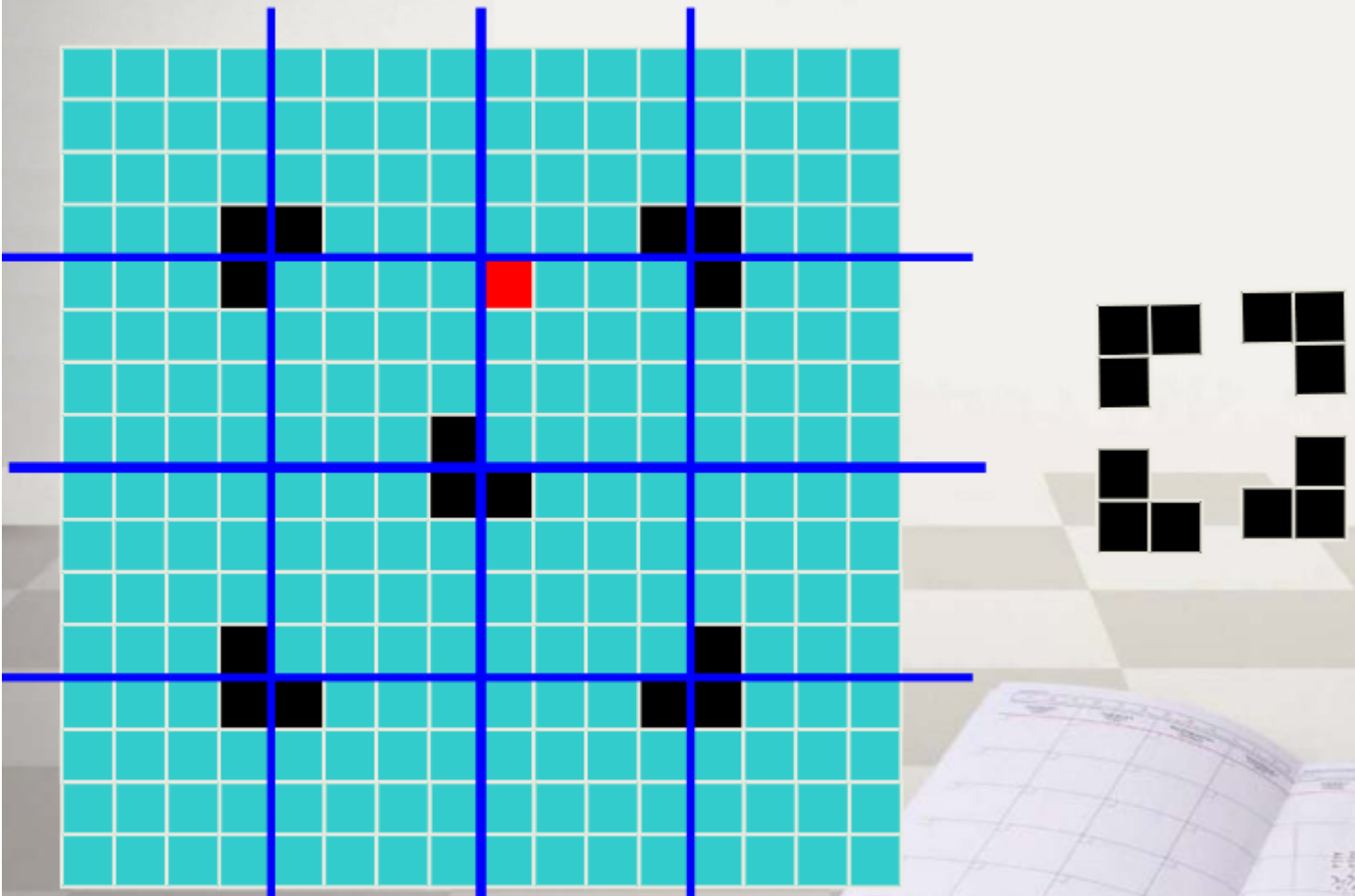
---

第一次分割



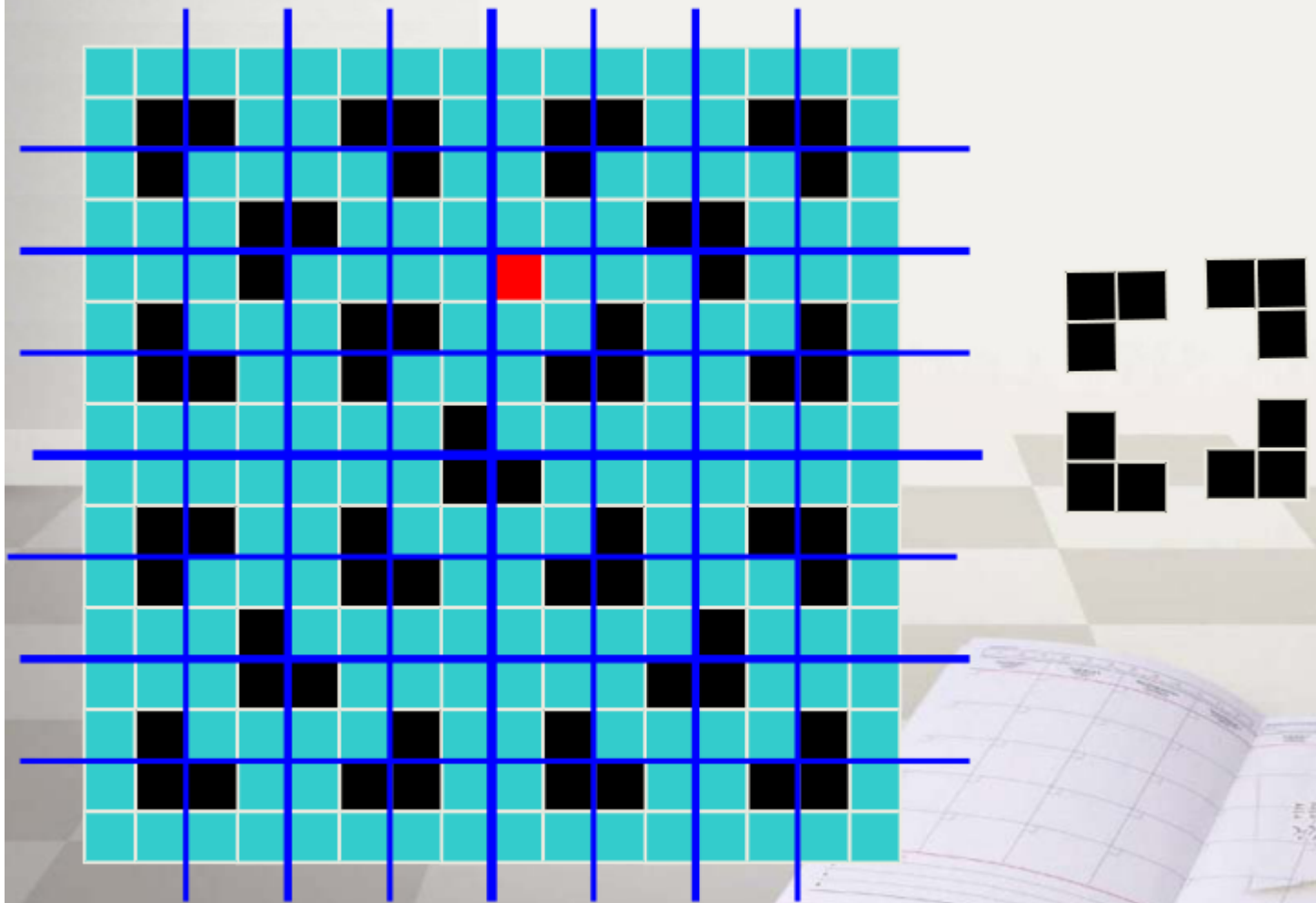
# 棋盘覆盖问题

第二次分割



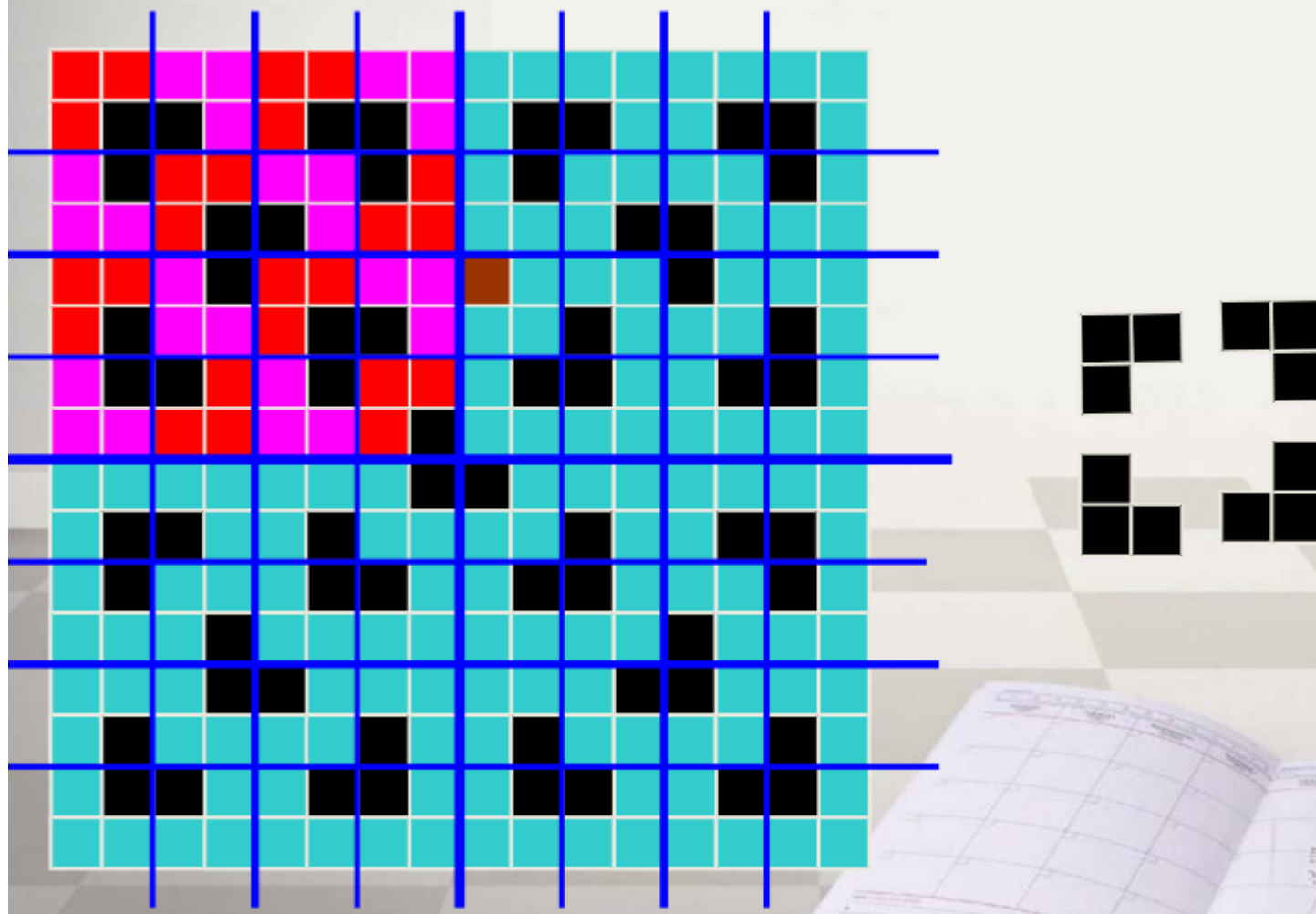
# 棋盘覆盖问题

第三次分割



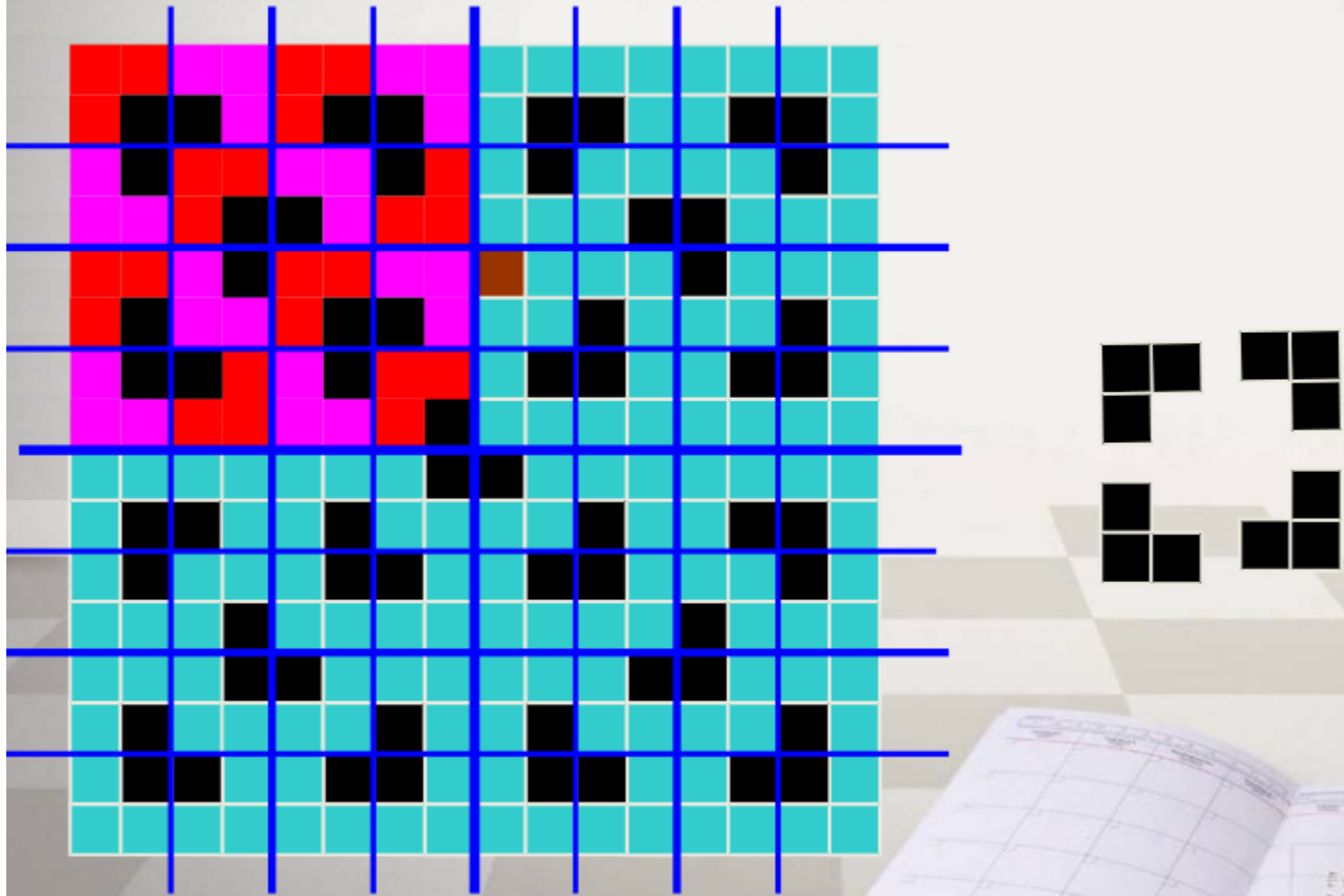
# 棋盘覆盖问题

第四次分割为 $1 \times 1$ 棋盘



# 棋盘覆盖问题

第一次分割后子棋盘的覆盖结果



# 棋盘覆盖问题

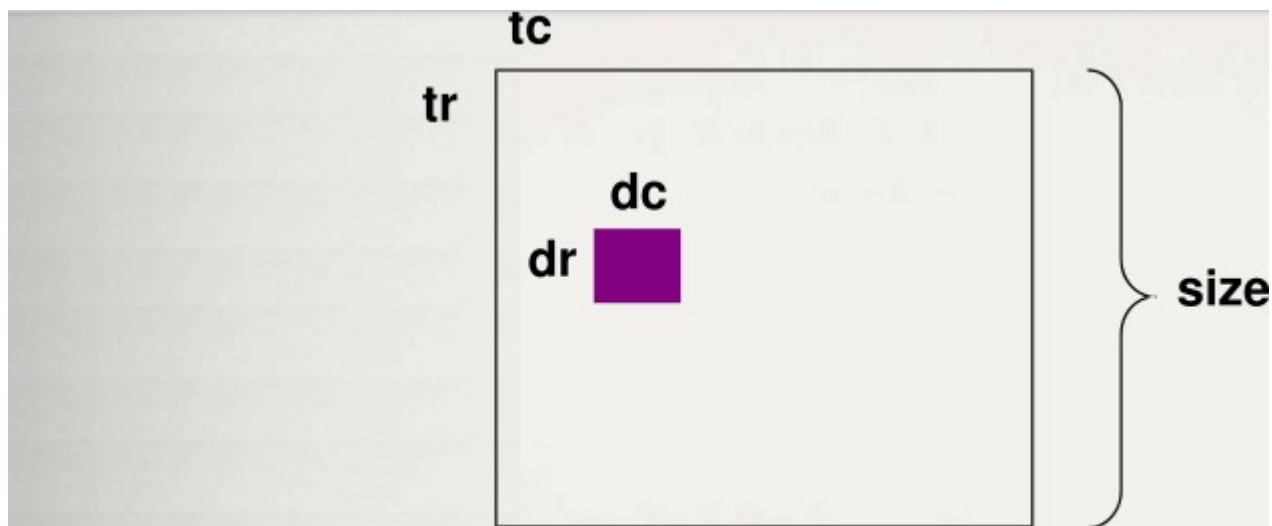
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分析算法所用数据及数据结构：

- (1) 棋盘：可以用一个二维数组`board[size][size]`表示一个棋盘，其中， $\text{size}=2^k$ 。为了在递归处理的过程中使用同一个棋盘，将数组`board`设为全局变量；
- (2) 子棋盘：当整个棋盘用二维数组`board[size][size]`表示，其中的子棋盘由棋盘左上角的下标`tr`、`tc`和棋盘大小`s`表示；
- (3) 特殊方格：用`board[dr][dc]`表示特殊方格，**dr**和**dc**是该特殊方格在二维数组`board`中的下标；
- (4) L型骨牌：将所有L型骨牌从1开始连续编号，用一个全局变量`t`表示。



# 棋盘覆盖问题



棋盘覆盖问题中的数据结构

算法的输入参数是：

tr: 棋盘左上角方格的行号

tc: 棋盘左上角方格的列号

dr: 特殊方格所在的行号

dc: 特殊方格所在的列号

size:  $\text{size} = 2^k$ , 棋盘规格为  $2^k \times 2^k$

# 棋盘覆盖问题

---

```
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
```

# 棋盘覆盖问题

---

```
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);}
```

```
}
```

# 棋盘覆盖问题

---

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

如何推导出所需要的骨牌个数呢？

$$(4^k - 1)/3$$

# Summary

---

1. 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。
2. 分治法的时间效率满足： $T(n)=aT(n/b)+f(n)$
3. 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。
4. 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。
5. 折半查找是一种对有序数组进行查找的算法，效率为 $\log n$
6.  $n$ 位大整数乘法的分治算法，大约需要做 $n^{1.585}$ 次乘法。
7. Stressan 算法也是分治算法

# 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。