

# 算法设计与分析

## Lecture 5: Divide-and-Conquer

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# Divide-and-Conquer

- The **divide-and-conquer (分治)** algorithm divides an instance of a problem into two or more small instances.
  - The small instance belongs to the same problem as the original instance.
  - Assume that the small instance is easy to solve.
  - Combine solutions to the small instances to solve the original instance.
  - If the small instance is still difficult, divide again until it is easy.
- The divide-and-conquer is a **top-down** approach.
  - Recursion is usually adopted.



# Divide-and-Conquer

The divide-and-conquer paradigm involves three steps at each level of the recursion:

- **Divide** the problem instance into a number of small instances.
- **Conquer** the small instances by solving them recursively. If the sizes of small instances are small enough, just solve them without recursion.
- **Combine (optional)** the solutions to the small instances into the solution for the original instance.



# Analyzing Divide-and-Conquer Algorithms

- When an algorithm contains a recursive call to itself, its running time can often be described by a **recursion equation (递归方程)**.
- We can easily solve them by the methods we have learned in Lecture 4.



# Analyzing Divide-and-Conquer Algorithms

- If the instance size is small enough, say  $n \leq c$  for some constant  $c$ , we can simply assume that the straightforward solution takes constant time  $\Theta(1)$ .
- The running time of a divide-and-conquer algorithm is based on the three steps of the basic paradigm:

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq c \\ aT(n/b) + D(n) + C(n) & \text{otherwise} \end{cases}$$

- $D(n)$ : the cost of dividing into small instances.
- $aT(n/b)$ : conquer  $a$  small instances with each size  $n/b$ .
- $C(n)$ : the cost of combining the solutions of small instances.
- $D(n)$  and  $C(n)$  are usually merged into a function  $f(n)$  for analysis convenience.





# MERGESORT

# Mergesort

- **Mergesort (合并排序)** combines two sorted arrays into one sorted array.
- Given an array with  $n$  elements, Mergesort involves the following steps:
  1. **Divide** the array into two subarrays each with  $n/2$  elements.
  2. **Conquer** each subarray by sorting it. Unless the array is sufficiently small, use recursion to do this.
  3. **Combine** the solutions to the subarrays by merging them into a single sorted array.



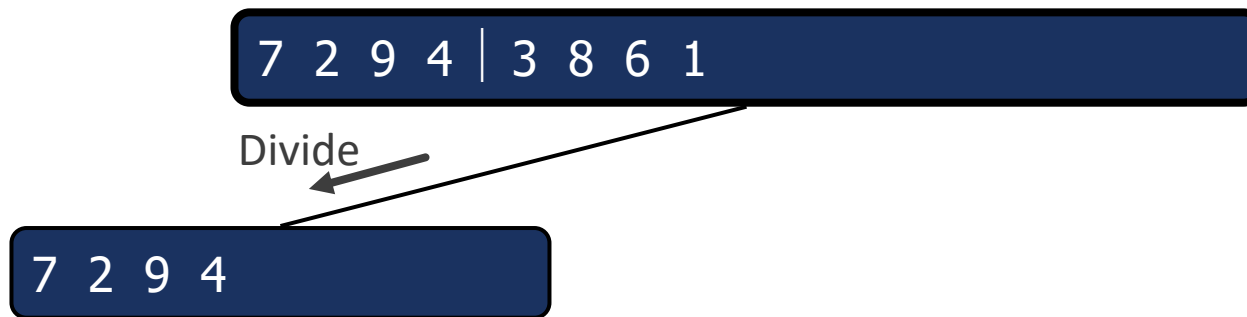
# Mergesort Example

7 2 9 4 3 8 6 1

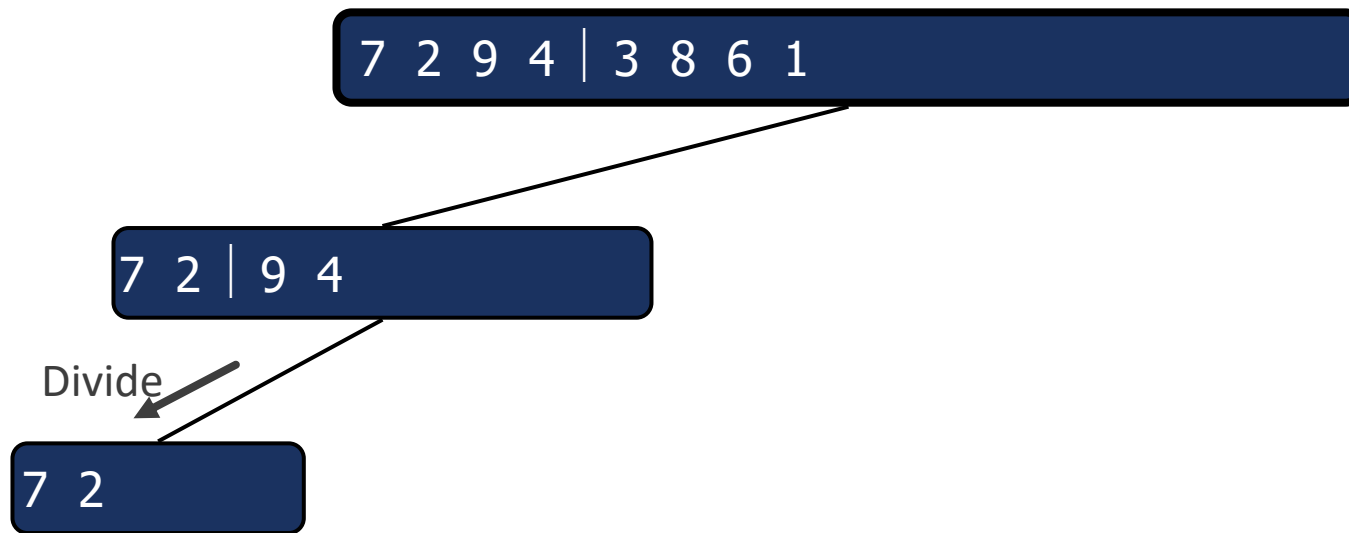




# Mergesort Example



# Mergesort Example



# Mergesort Example

7 2 9 4 | 3 8 6 1

7 2 | 9 4

7 | 2

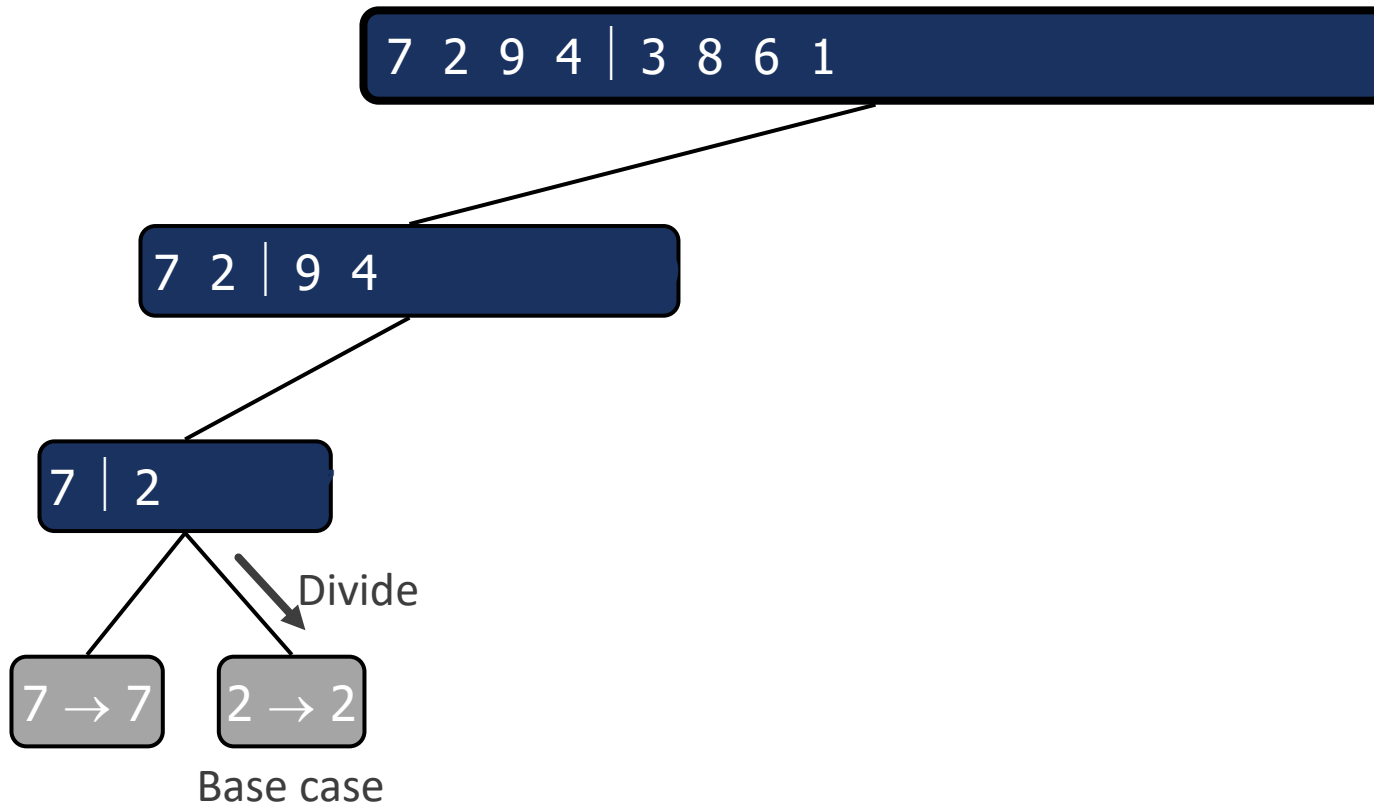
Divide ↙

7 → 7

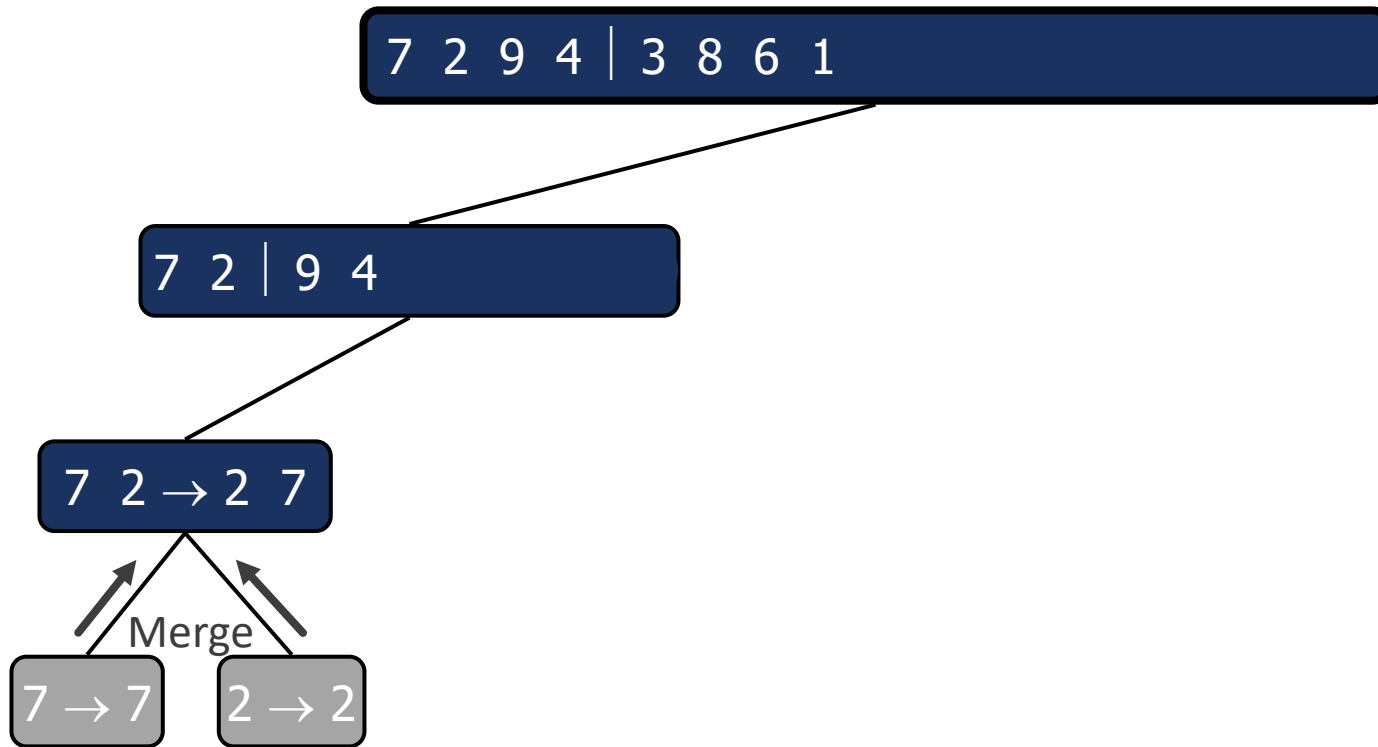
Base case



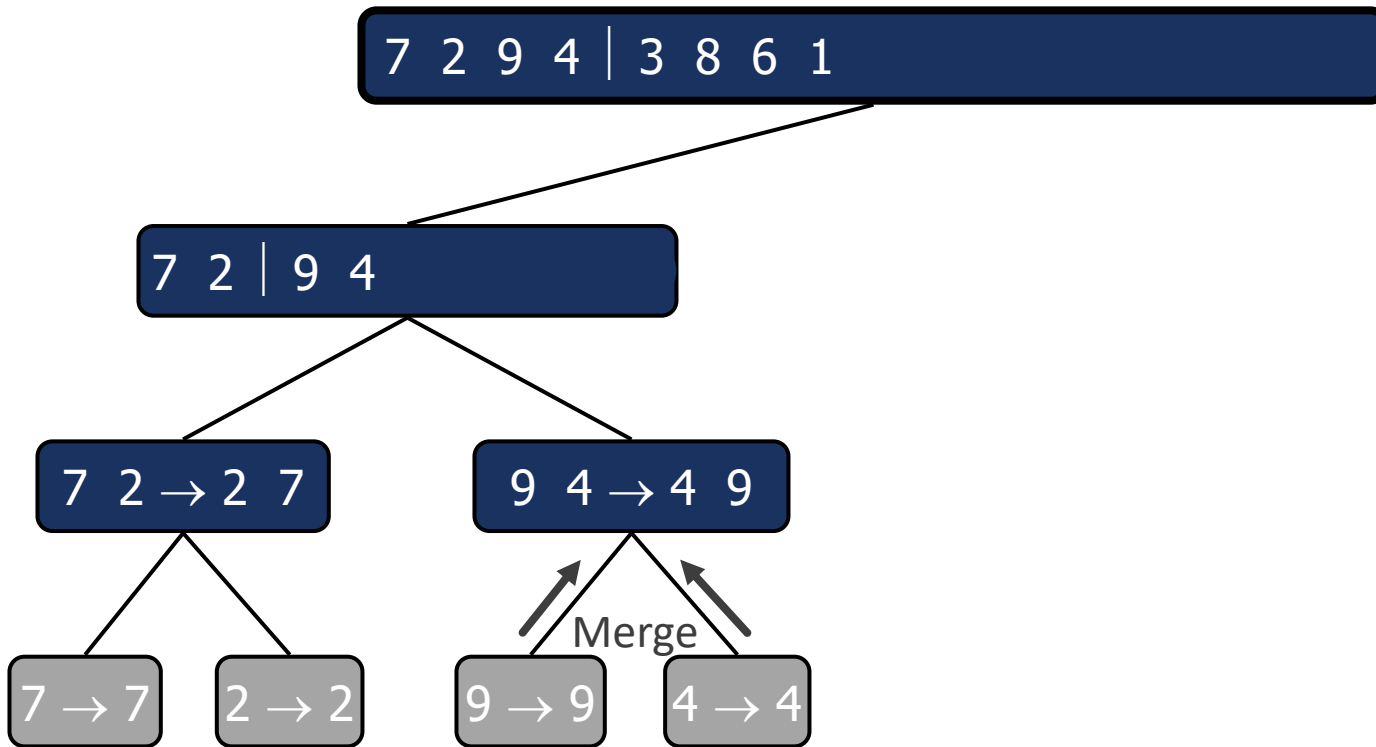
# Mergesort Example



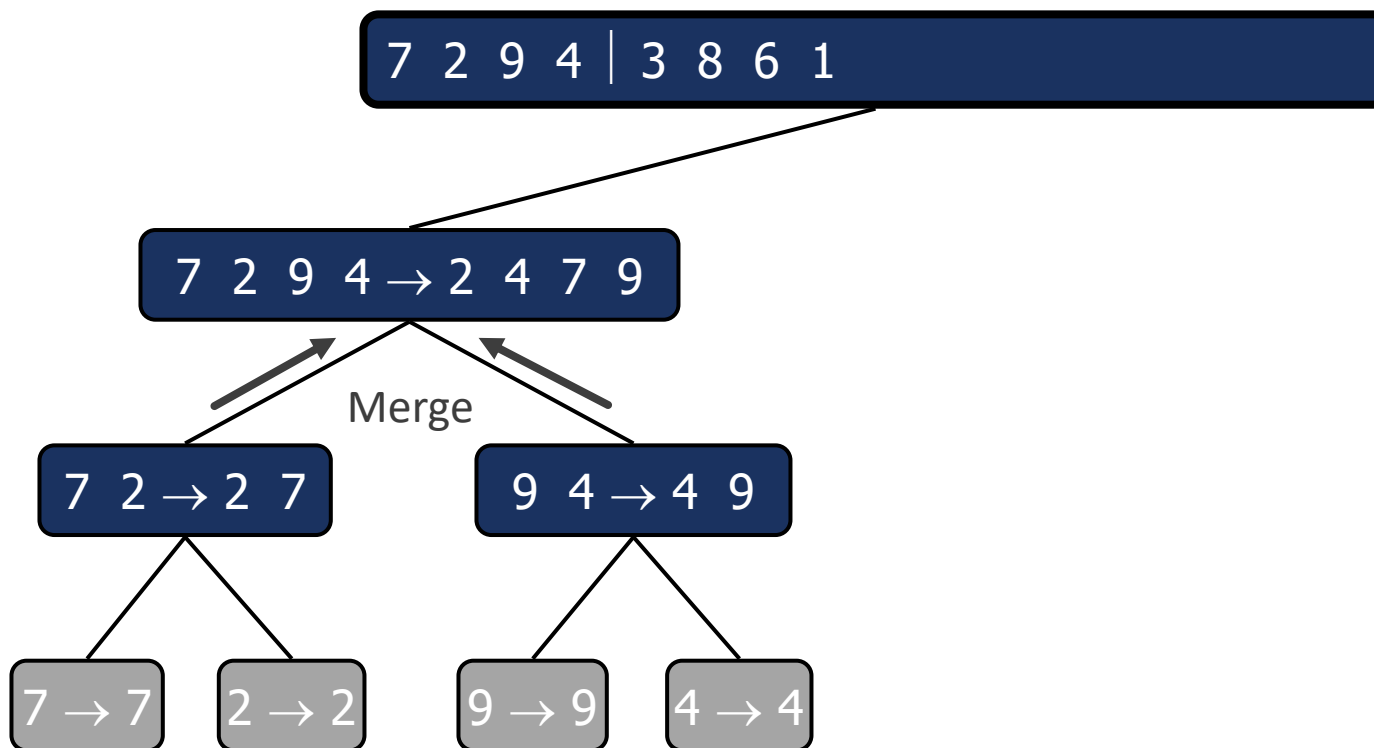
# Mergesort Example



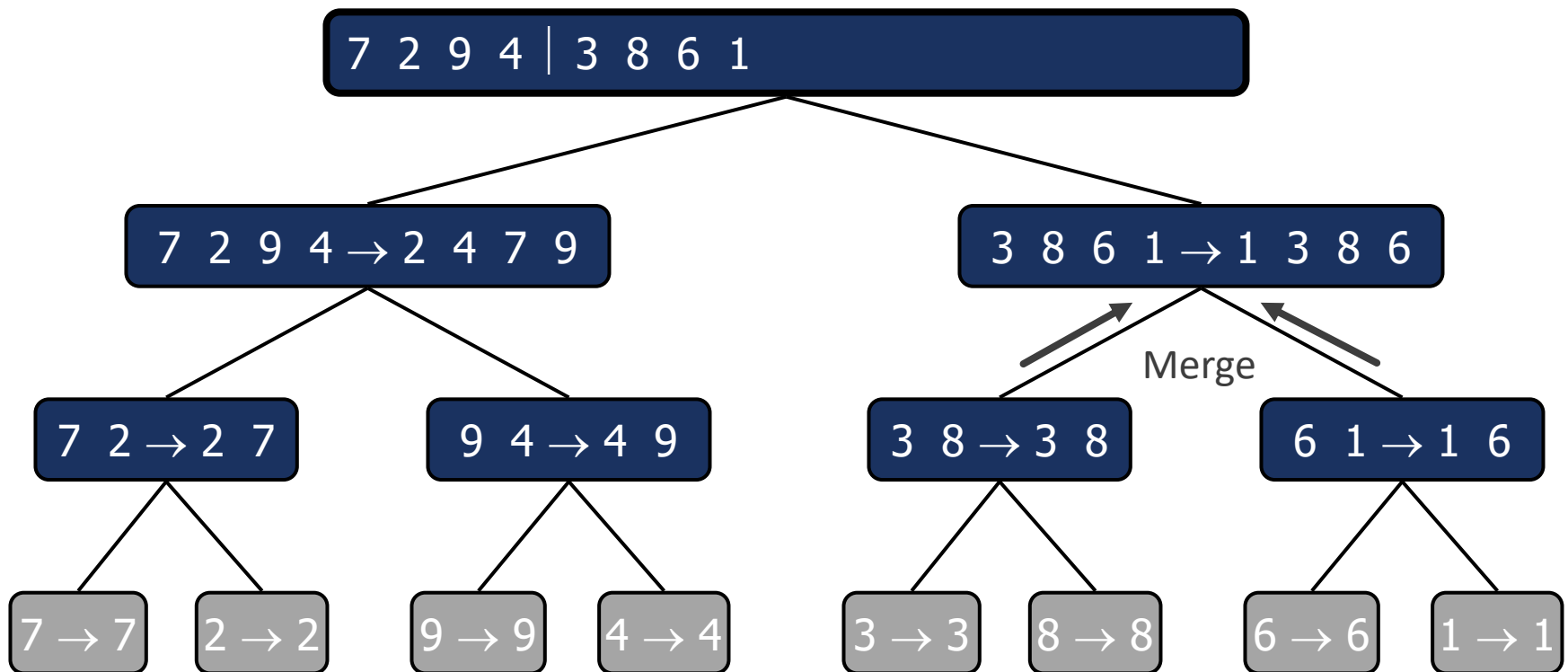
# Mergesort Example



# Mergesort Example

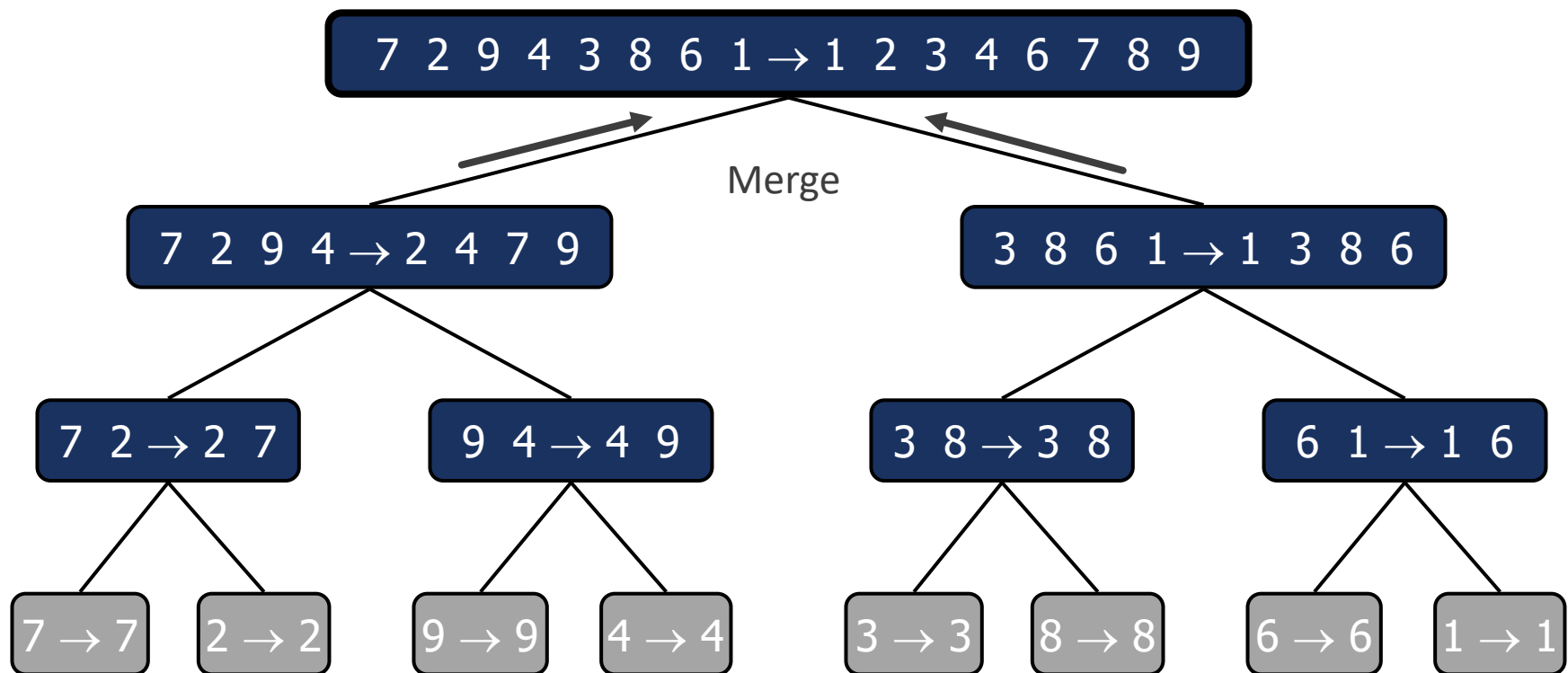


# Mergesort Example

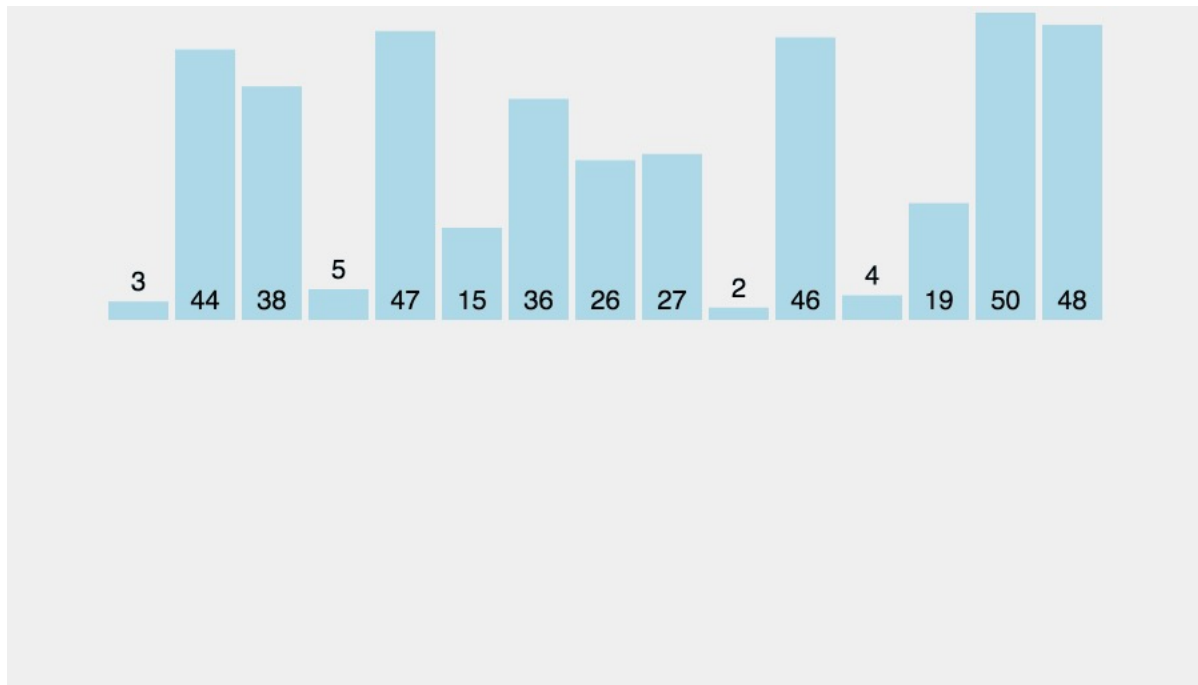




# Mergesort Example



# Mergesort Visualized Demo



# Mergesort

- Call MergeSort( $A, 1, \text{len}[A]$ ) for the sorting problem.
- Recursive call with different array index:
  - $p$ : starting index
  - $q$ : middle index
  - $r$ : end index
- Exit condition:  $p = r$ , there is only one element.

```
MergeSort( $A, p, r$ )
```

```
1  if  $p < r$  then
```

```
2       $q \leftarrow \lfloor (p + r) / 2 \rfloor$ 
```

```
3      MergeSort( $A, p, q$ )
```

```
4      MergeSort( $A, q + 1, r$ )
```

```
5      Merge( $A, p, q, r$ )
```



# Mergesort

Merge( $A, p, q, r$ )

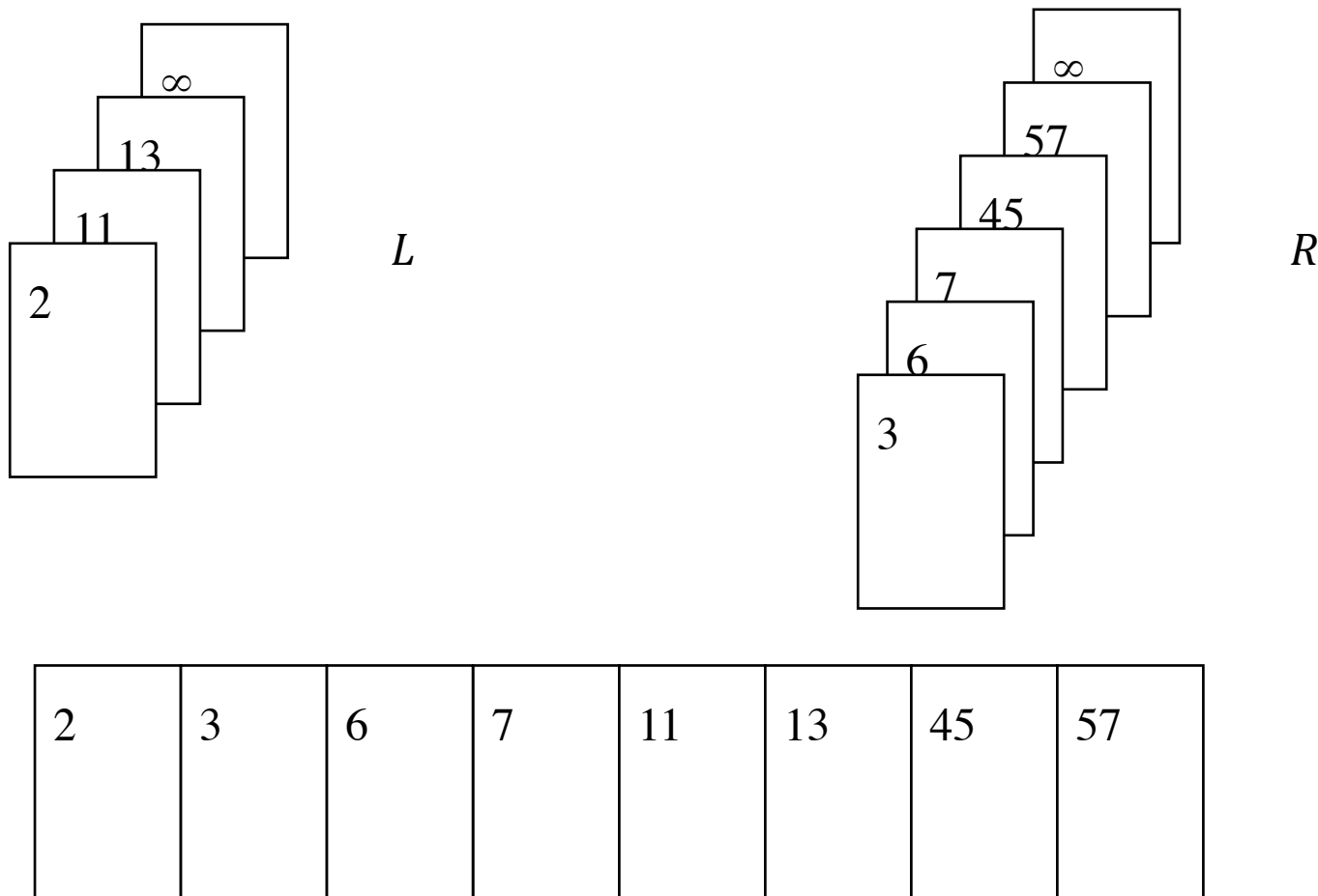
```
1   $n_1 \leftarrow q - p + 1$ 
2   $n_2 \leftarrow r - q$ 
3  for  $i \leftarrow 1$  to  $n_1$  do
4       $L[i] \leftarrow A[p + i - 1]$ 
5  for  $j \leftarrow 1$  to  $n_2$  do
6       $R[j] \leftarrow A[q + j]$ 
7   $L[n_1 + 1] \leftarrow \infty$ 
8   $R[n_2 + 1] \leftarrow \infty$ 
```

```
9   $i \leftarrow 1$ 
10  $j \leftarrow 1$ 
11 for  $k \leftarrow p$  to  $r$  do
12     if  $L[i] \leq R[j]$  then
13          $A[k] \leftarrow L[i]$ 
14          $i \leftarrow i + 1$ 
15     else  $A[k] \leftarrow R[j]$ 
16          $j \leftarrow j + 1$ 
```

- Line 1-6:  $L$  and  $R$  are used to store two sorted subarrays with size  $n_1$  and  $n_2$ .
- Line 7-8: Assign infinity at the end of  $L$  and  $R$  for comparison convenience.
- Line 9-16: For each index from  $p$  to  $r$ , compare one by one and increase the index of the array with smaller element.



# Mergesort



# The Correctness of Merge

```
11 for  $k \leftarrow p$  to  $r$  do
12     if  $L[i] \leq R[j]$  then
13          $A[k] \leftarrow L[i]$ 
14          $i \leftarrow i + 1$ 
15     else  $A[k] \leftarrow R[j]$ 
16          $j \leftarrow j + 1$ 
```

- Loop invariant:
  - At the start of each iteration of the for loop in Lines 12-17, the subarray  $A[p \dots k - 1]$  contains the  $k - p$  smallest elements of  $L$  and  $R$ , in sorted order.
  - $L[i]$  and  $R[j]$  are the smallest elements of their arrays that have not been copied back into  $A$ .
- We show that this loop invariant holds
  - prior to the first iteration of the loop;
  - after the  $k$ th iteration of the loop;
  - when the loop terminates.



## The Correctness of Merge

```
11 for  $k \leftarrow p$  to  $r$  do
12     if  $L[i] \leq R[j]$  then
13          $A[k] \leftarrow L[i]$ 
14          $i \leftarrow i + 1$ 
15     else  $A[k] \leftarrow R[j]$ 
16          $j \leftarrow j + 1$ 
```

### Initialization

- Prior to the first iteration of the loop, we have  $k = p$ , so that the subarray  $A[p \dots p - 1]$  is empty.
- Both  $L[i]$  and  $R[j]$  are the smallest elements of their arrays that have not been copied back into  $A$ .



# The Correctness of Merge

```
11 for  $k \leftarrow p$  to  $r$  do
12     if  $L[i] \leq R[j]$  then
13          $A[k] \leftarrow L[i]$ 
14          $i \leftarrow i + 1$ 
15     else  $A[k] \leftarrow R[j]$ 
16          $j \leftarrow j + 1$ 
```

## Maintenance

- Hypothesis: Before the  $k$ th iteration, the loop invariant holds.
- Let us first suppose that  $L[i] \leq R[j]$ . Then  $L[i]$  is the smallest element not yet copied back into  $A$ .
- Because  $A[p \dots k - 1]$  contains the  $k - p$  smallest elements, after Line 13 copies  $L[i]$  into  $A[k]$ , the subarray  $A[p \dots k]$  will contain the  $k - p + 1$  smallest elements.
- Before the next iteration,  $k$  and  $i$  are increased by 1.
  - $A[p \dots k]$  contain the  $k - p + 1$  smallest elements.
  - $L[i + 1]$  and  $R[j]$  are the smallest elements of their arrays that have not been copied back into  $A$ .
- Therefore, before the  $(k + 1)$ th iteration, the loop invariant holds.





# The Correctness of Merge

```
11 for  $k \leftarrow p$  to  $r$  do
12     if  $L[i] \leq R[j]$  then
13          $A[k] \leftarrow L[i]$ 
14          $i \leftarrow i + 1$ 
15     else  $A[k] \leftarrow R[j]$ 
16          $j \leftarrow j + 1$ 
```

## Termination

- At termination,  $k = r + 1$ .
- By the loop invariant, the subarray  $A[p \dots k - 1]$ , which is  $A[p \dots r]$ , contains the  $k - p = r - p + 1$  smallest elements of  $L$  and  $R$ , in sorted order.
- Therefore, the loop invariant holds. Merge correctly merges two sorted arrays into one sorted array.



# Time Complexity of Mergesort

- No matter how different the input is, Merge always does  $r - p + 1 = n$  times of key comparison.
  - For sorting algorithms, we usually only count the number of key comparisons.

- So the recursion equation is:

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

- By the master method case 2, we have  $f(n) = n = \Theta(n) = \Theta(n^{\log_2 2})$ .
- Therefore,  $T(n) = \Theta(n \lg n)$  for best-, worst- and average-case.

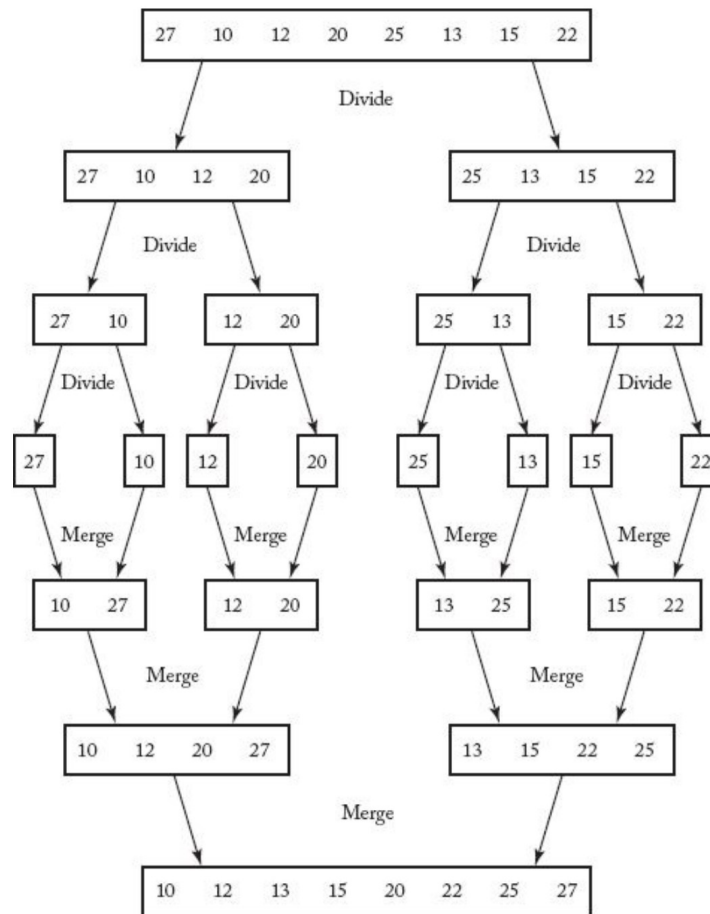


# Classroom Exercise

- Write each step of Mergesort to sort the following array:  
 $\langle 27, 10, 12, 20, 25, 13, 15, 22 \rangle$



# Classroom Exercise





# QUICKSORT

# Quicksort

- Mergesort splits the array first, and then combines them by merging.
- Can we roughly sort the array first, and then split it?
  - E.g. put small elements on the left, and large element on the right.
  - If we can do in this way, we don't need to merge.



# Quicksort

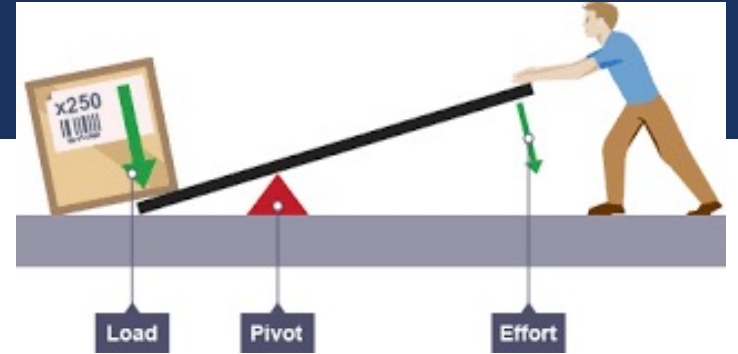
- Quicksort (快速排序) is developed by British computer scientist Charles Antony Richard Hoare (Tony Hoare) in 1962.
- You can know the main property of Quicksort by its name – quick!
- When implemented well, it can be about two or three times faster than Mergesort.



Tony Hoare in 2011



# Quicksort



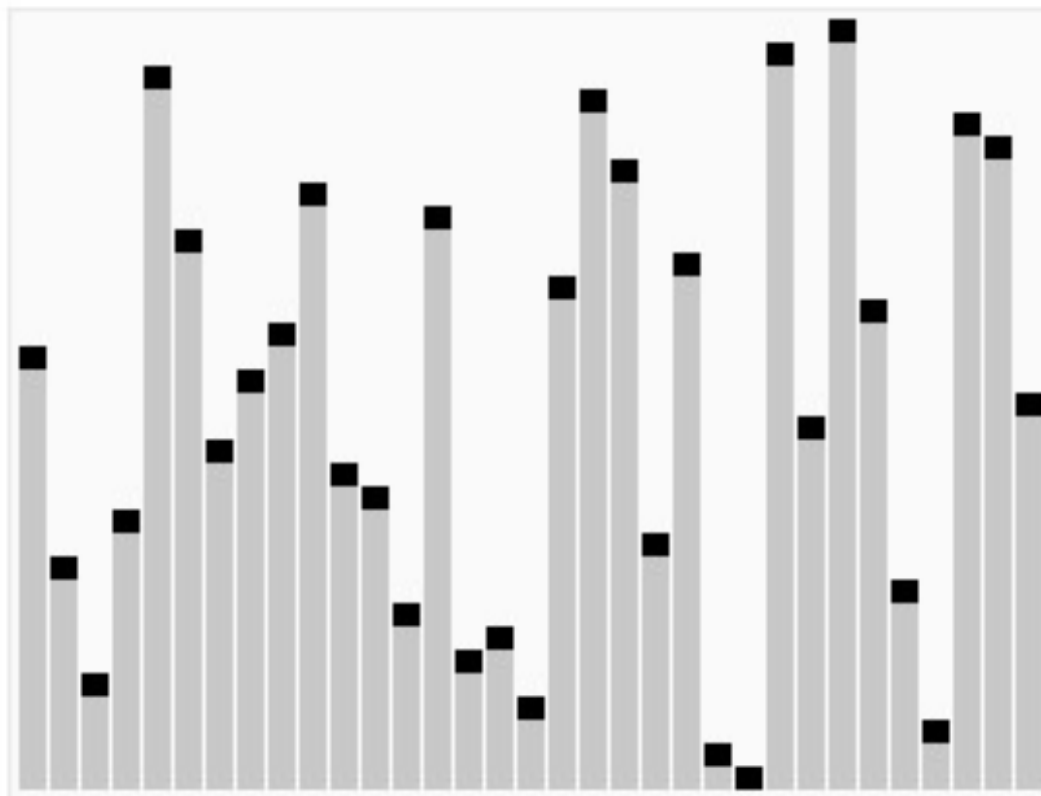
Steps:

- Randomly select a **pivot (支点)** element.
  - Conventional use the first or last item.
- Put all the elements smaller than the pivot element on its left, and all the elements greater than the pivot element on its right.
- Recursively sort the left subarray and right subarray.
  - Each subarray is sorted after recursion call. Therefore, there's no need to combine the results.





# Quicksort Visualized Demo



# Quicksort

- Call  $\text{QuickSort}(A, 1, \text{len}[A])$  for the sorting problem.
- Recursive call with different array index:
  - $p$ : starting index
  - $q$ : pivot index
  - $r$ : end index
- Exit condition:  $p = r$ , there is only one element.

```
QuickSort( $A, p, r$ )
1  if  $p < r$  then
2       $q \leftarrow \text{Partition}(A, p, r)$ 
3      QuickSort( $A, p, q - 1$ )
4      QuickSort( $A, q + 1, r$ )
```

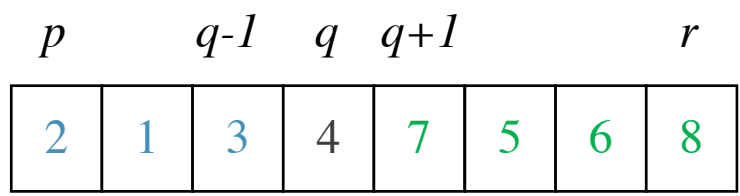
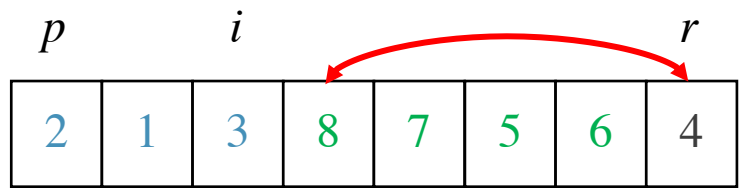
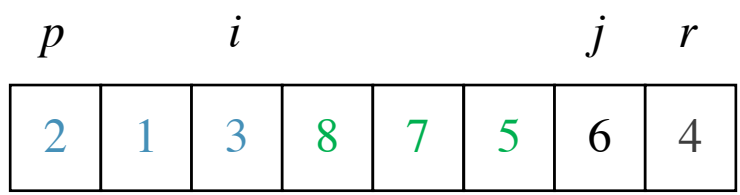
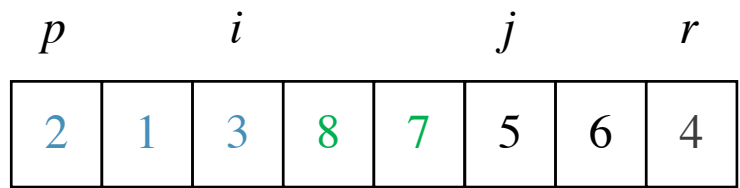
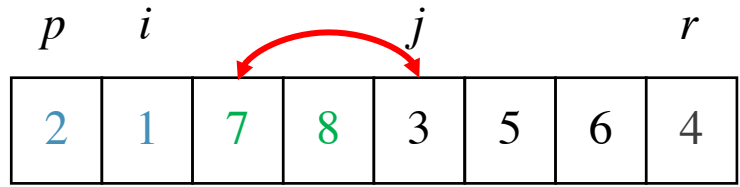
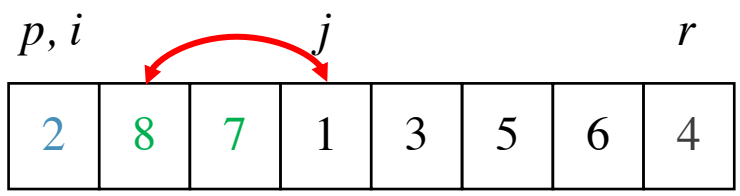
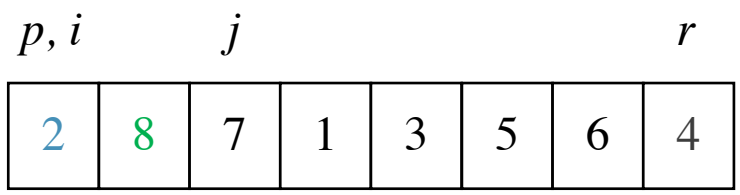
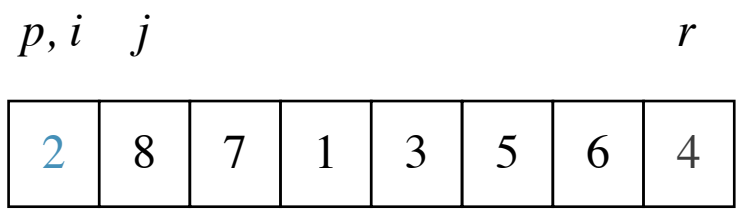
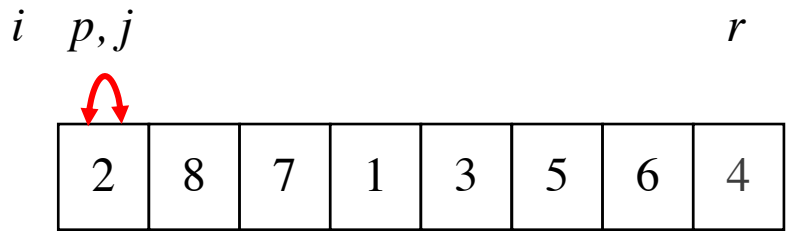


# Quicksort

- Line 1: Simply select the last element  $A[r]$  as the pivot.
- Line 2: Use  $i$  to store the index for switching.
- Line 3-6: iterate over  $j$  to find elements smaller than  $x$  and switch them to the front.
- Line 7: put pivot at the proper position.

```
Partition( $A, p, r$ )  
1   $pivot \leftarrow A[r]$   
2   $i \leftarrow p - 1$   
3  for  $j \leftarrow p$  to  $r - 1$  do  
4      if  $A[j] \leq pivot$  then  
5           $i \leftarrow i + 1$   
6           $A[i] \leftrightarrow A[j]$   
7   $A[i + 1] \leftrightarrow A[r]$   
8  return  $i + 1$ 
```





# Time Complexity of Quicksort

- In Partition, each element in  $A$  is compared with the pivot except itself.
- Therefore, the number of comparisons in Partition is  $n - 1$ .

```
Partition( $A, p, r$ )
1   $pivot \leftarrow A[r]$ 
2   $i \leftarrow p - 1$ 
3  for  $j \leftarrow p$  to  $r - 1$  do
4      if  $A[j] \leq pivot$  then
5           $i \leftarrow i + 1$ 
6           $A[i] \leftrightarrow A[j]$ 
7   $A[i + 1] \leftrightarrow A[r]$ 
8  return  $i + 1$ 
```



# Worst-Case Time Complexity of Quicksort

- The worst-case occurs when the array is already sorted (in either nondecreasing or nonincreasing order).
- In each recursion step, the pivot element is always the smallest or largest item.
  - Thus,  $n$  elements are divided into  $n - 1$  and 0 elements during recursive call.

- The recursion equation is:

$$T(n) = T(n - 1) + T(0) + n - 1$$

- Using recursion tree, we can easily get

$$T(n) = n(n - 1)/2 = \Theta(n^2)$$



# Worst-Case Time Complexity of Quicksort

- The closer the input array is to being sorted, the closer we are to the worst-case performance.
  - Because the pivot can't fairly separate two subarrays.
  - Recursion loses its power.
- How to wisely choose the pivot?
  - Random.
  - Median of  $A[1]$ ,  $A[\lfloor n/2 \rfloor]$ , and  $A[n]$ . Safe to avoid the worst-case but more comparisons are needed.
- What will be the best case?



# Best-Case Time Complexity of Quicksort

- The best-case occurs when the Partition almost evenly splits the array:
  - Array has odd number of elements: Both subarrays have  $\lfloor n/2 \rfloor$  elements.
  - Array has even number of elements: One subarray has  $\lfloor n/2 \rfloor$  elements and another has  $n/2$ .
- In both cases, the size of the subarray is no more than  $n/2$ .
- The recursion equation is:

$$T(n) = 2T(n/2) + n - 1 = O(n \lg n).$$





# Classroom Exercise

What is the best case input for Quicksort when  $n = 12$ ?



# Classroom Exercise

## Solution:

The best case occurs when each pivot evenly split the array:

1, 2, 4, 5, 3, 7, 8, 10, 12, 11, 9, 6

Think: how to write a best-case input generator for Quicksort?



# Average-Case Time Complexity of Quicksort

- The worst-case of Quicksort is no faster than insertion sort (also  $\Theta(n^2)$ ), and slower than Mergesort ( $\Theta(n \log n)$ ).
- The best-case of Quicksort is slower than insertion sort ( $\Theta(n)$ ).
- How dare it name itself “quick”?
  - The average-case behavior earns its name!



# Average-Case Time Complexity of Quicksort

- To analyze the average-case time complexity, we can add randomization.
  - Randomly permute the input array (uniform distributed input).
  - Randomly choose the pivot item.



# Average-Case Time Complexity of Quicksort

- By randomization, now the probability of pivot being any item in the array is  $1/n$ .

$$T(n) = \sum_{p=1}^n \frac{1}{n} [T(p-1) + T(n-p)] + n - 1$$

$$T(n) = \frac{2}{n} \sum_{p=1}^n T(p-1) + n - 1$$

$$nT(n) = 2 \sum_{p=1}^n T(p-1) + n(n-1) \text{ (multiply by } n\text{)}$$

$$(n-1)T(n-1) = 2 \sum_{p=1}^{n-1} T(p-1) + (n-1)(n-2) \text{ (apply to } n-1\text{)}$$



# Average-Case Time Complexity of Quicksort

$$\begin{aligned} nT(n) - (n-1)T(n-1) \\ = 2T(n-1) + 2(n-1) \text{ (subtraction)} \end{aligned}$$

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

■ Let  $a_n = \frac{T(n)}{n+1}$ ,

$$a_n = a_{n-1} + \frac{2(n-1)}{n(n+1)} = \sum_{i=1}^n \frac{2(i-1)}{i(i+1)} \approx 2 \sum_{i=1}^n \frac{1}{i} \approx 2 \ln n.$$

Harmonic series ↙

■ Therefore,  $T(n) \approx (n+1)2 \ln n = (n+1)2 \ln 2 \lg n \approx 1.38(n+1) \lg n = \Theta(n \lg n)$ .





# LARGE INTEGER MULTIPLICATION

# Large Integer Multiplication

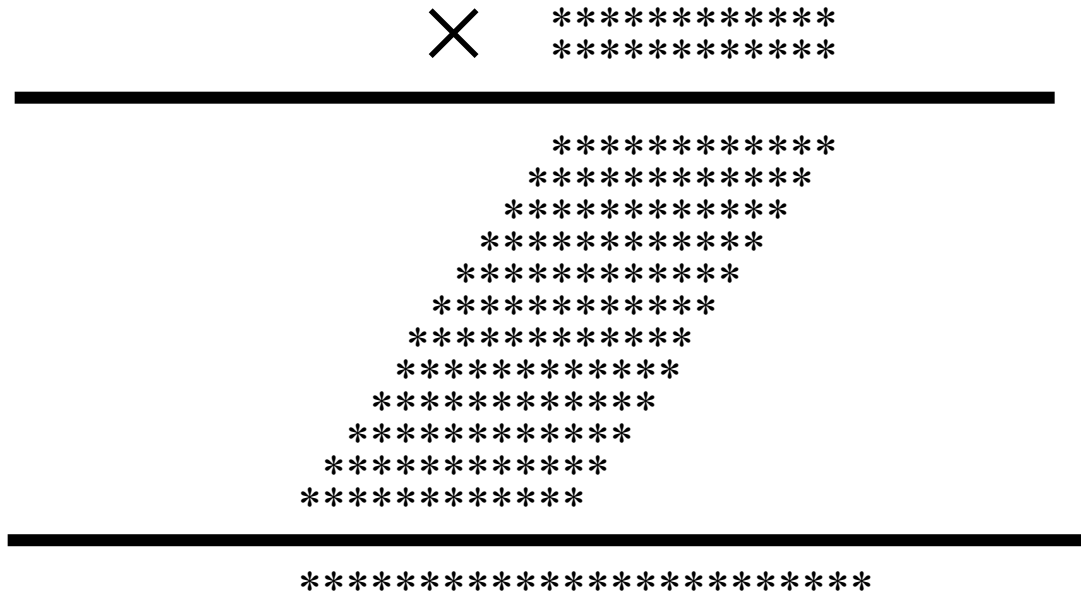
- Suppose that we need to do arithmetic operations on integers whose size is very large.
- In cryptography (密码学) and network security, encryption and decryption need to multiply very large numbers.
- How to do arithmetic for those large integers?





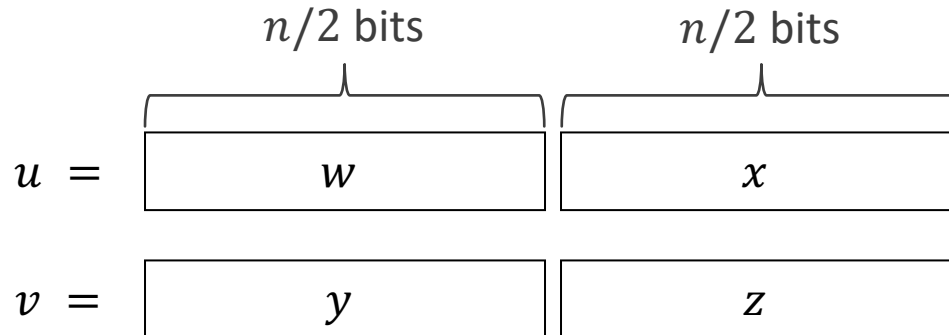
# Large Integer Multiplication

- Tradition algorithm is nothing but what we have learned in primary school.
- It takes  $\Theta(n^2)$  bit operations.



# Large Integer Multiplication

- Use  $n$ -bit binary representation for  $u$  and  $v$ .
- We can use divide-and-conquer: Each integer is divided into two parts of  $n/2$  bits each.



Example:

$$(110011)_2 = (110)_2 \times 2^3 + (011)_2$$

- Therefore, integers  $u$  and  $v$  can be represented as:

$$u = w2^{n/2} + x, \quad v = y2^{n/2} + z.$$

- Then, we have:

$$uv = (w2^{n/2} + x)(y2^{n/2} + z) = wy2^n + (wz + xy)2^{n/2} + xz.$$



# Large Integer Multiplication

- We need 4 recursive multiplications with size  $n/2$  to calculate

$$wy2^n + (wz + xy)2^{n/2} + xz$$

- Multiply with  $2^n$  is to simply shift by  $n$  bits to the left with cost  $\Theta(n)$ .
- 3 times of addition is also with cost  $\Theta(n)$ .
- The recursion equation is:

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

- By the master method, we get  $T(n) = \Theta(n^2)$ .
- It is still quadratic. Why?



# Large Integer Multiplication

- We decompose the instance of  $n$  into 4 small instances with size  $n/2$ .
- If we can decrease 4 to 3, by the master method we get  $T(n) = \Theta(n^{\lg 3})$ .
- As before, we need to calculate

$$wy, wz + xy, xz$$

- If instead we set

$$r = (w + x)(y + z) = wy + (wz + zy) + xz$$

we have

$$wz + xy = r - wy - xz$$

- Then, we only need to calculate

$$r, wy, xz$$



# Large Integer Multiplication

```
Multiply2int(u, v)
1  if |u| = |v| = 1 then return uv
2  else
3      Split u into w and x; split v into y and z
4       $A_1 \leftarrow \text{Multiply2int}(w, y)$ 
5       $A_2 \leftarrow \text{Multiply2int}(x, z)$ 
6       $A_3 \leftarrow \text{Multiply2int}(w + x, y + z)$ 
7      return  $A_1 2^n + (A_3 - A_1 - A_2) 2^{n/2} + A_2$ 
```

- The above method yields the following recursion equation:

$$T(n) = 3T(n/2) + cn.$$

- By the master method, we get  $T(n) = \Theta(n^{\lg 3}) \approx \Theta(n^{1.59})$ .



# Classroom Exercise

Write the pseudocode of binary search algorithm:

- Given a sorted array  $A$ ,
  - If  $x$  equals the middle item, quit.
  - Otherwise, compare  $x$  with the middle item.
    - If  $x$  is smaller, search the left subarray.
    - If  $x$  is greater, search the right subarray.
- What is the time complexity?



# Classroom Exercise

## Solution:

BinarySearch( $A, x$ )

```
1  $p \leftarrow 1$ 
2  $r \leftarrow n$ 
3  $k \leftarrow 0$ 
4 while  $p \leq r$  and  $k = 0$  do
5    $m \leftarrow \lfloor (p + r) / 2 \rfloor$ 
6   if  $A[m] = x$  then return  $m$ 
7   else if  $x < A[m]$  then  $r \leftarrow m - 1$ 
8     else  $p \leftarrow m + 1$ 
9 return 0
```

RecursiveBinarySearch( $A, p, r$ )

```
1 if  $p > r$  then return 0
2 else
3    $m \leftarrow \lfloor (p + r) / 2 \rfloor$ 
4   if  $A[m] = x$  then return  $m$ 
5   else if  $x < A[m]$  then
6     RecursiveBinarySearch( $A, p, r - 1$ )
7   else
8     RecursiveBinarySearch( $A, p + 1, r$ )
```

Divide-and-conquer algorithm is not necessarily implemented by recursion.



# Classroom Exercise

- The recursion equation is:

$$T(n) = T(n/2) + 1.$$

- By using master method case 2, we have  $T(n) = \Theta(\lg n)$ .







# MATRIX MULTIPLICATION

# Matrix Multiplication

- Given matrices  $A$  and  $B$  with size  $n \times n$ , compute the matrix product  $C = AB$ .
- The formula we have learned in linear algebra for doing this is:

$$C(i, j) = \sum_{k=1}^n A(i, k)B(k, j).$$

- Calculating each  $C(i, j)$  takes  $O(n)$ . Thus, calculating total  $n \times n$  elements in  $C$  takes  $O(n^3)$ .



# Matrix Multiplication

- Suppose we want to product  $C$  of two  $2 \times 2$  matrices,  $A$  and  $B$ ,  
That is,

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \times \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}.$$

- The divide-and-conquer version consists of computing  $C$  as defined by the following equation:

$$C = \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} & a_{11}b_{12} + a_{12}b_{22} \\ a_{21}b_{11} + a_{22}b_{21} & a_{21}b_{12} + a_{22}b_{22} \end{bmatrix}.$$



# Matrix Multiplication

- Will it be better?
- The cost of multiplying two  $n \times n$  matrices consists of:
  - 8 times the cost of multiplying two  $n/2 \times n/2$  matrices;
  - 4 times the cost of adding two  $n/2 \times n/2$  matrices.
- The recursion equation is:

$$T(n) = \begin{cases} \Theta(1) & n = 1 \\ 8T(n/2) + 4(n/2)^2 & n > 1 \end{cases}$$

- Make use of the master method,  $T(n) = \Theta(n^3)$ . And thus this method is no faster than the ordinary one.
- What can we do?



# Matrix Multiplication with Strassen's algorithm

- Strassen's algorithm (斯特拉森算法) reduces the number of multiplications from 8 to 7.
- Strassen determined that if we let

$$m_1 = (a_{11} + a_{22})(b_{11} + b_{22})$$

$$m_2 = (a_{21} + a_{22})b_{11}$$

$$m_3 = a_{11}(b_{12} - b_{22})$$

$$m_4 = a_{22}(b_{21} - b_{11})$$

$$m_5 = (a_{11} + a_{12})b_{22}$$

$$m_6 = (a_{21} - a_{11})(b_{11} + b_{12})$$

$$m_7 = (a_{12} - a_{22})(b_{21} + b_{22})$$

Less multiplication, at expense of more addition and subtraction.

the product  $C$  is given by

$$C = \begin{bmatrix} m_1 + m_4 - m_5 + m_7 & m_3 + m_5 \\ m_2 + m_4 & m_1 + m_3 - m_2 + m_6 \end{bmatrix}$$



# Matrix Multiplication with Strassen's algorithm

- To multiply two  $2 \times 2$  matrices, Strassen's method requires 7 multiplications and 18 additions/subtractions.
  - The standard method requires 8 multiplications and 4 additions/subtractions.
  - Use 14 more additions/subtractions to save 1 multiplication. Is that worthy?
- Recursion equation:

$$T(n) = \begin{cases} \Theta(1) & n = 1 \\ 7T(n/2) + 18(n/2)^2 & n > 1 \end{cases}$$

- Use the master method case 1,  $f(n) = \frac{18}{4}n^2 = O(n^{\log_2 7 - \epsilon}) \approx O(n^{2.81 - \epsilon})$  for  $\epsilon \approx 0.81$ .
- Therefore, we have  $T(n) = \Theta(n^{2.81})$ .





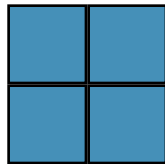
# DEFECTIVE CHESSBOARD

# Defective Chessboard

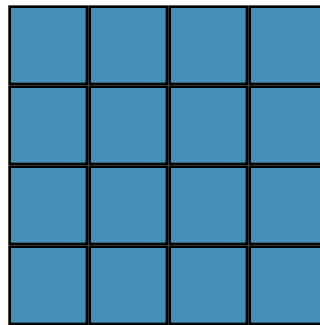
- A chessboard is an  $2^n \times 2^n$  grid, for  $n \geq 0$ :



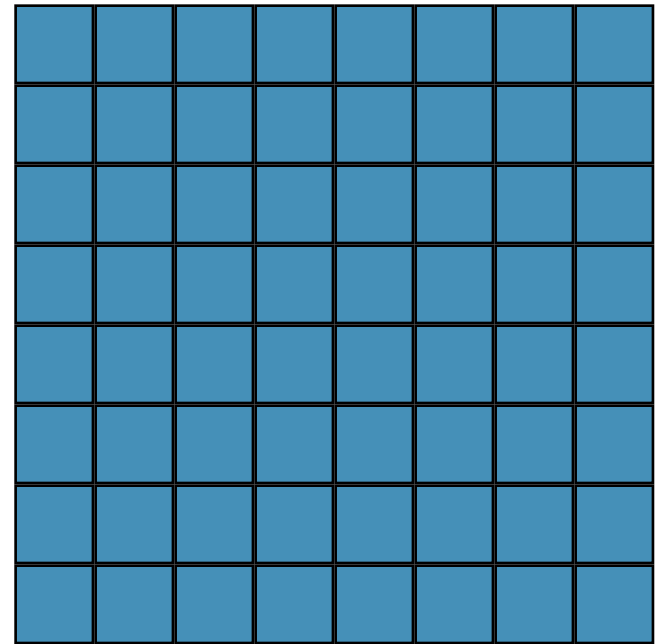
1×1



2×2



4×4



8×8



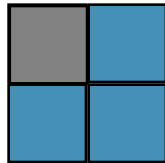


# Defective Chessboard

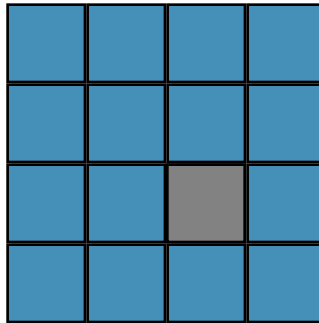
- A defective chessboard (残缺棋盘) is chessboard that has one unavailable (defective) position.



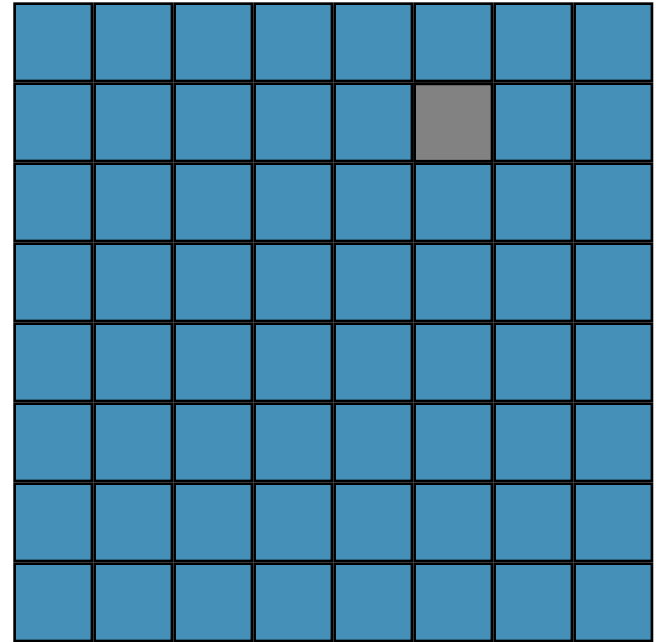
1×1



2×2



4×4

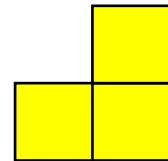
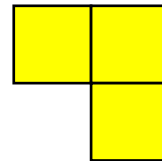
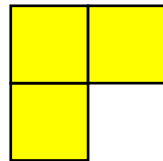
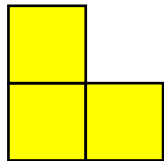


8×8



# Defective Chessboard

- A **triomino (三格板)** is an L shaped object that can cover three squares of a chessboard.
- A triomino has four orientations.
- You can use infinite number of triominoes.

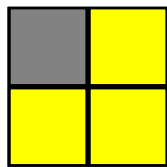


# Defective Chessboard

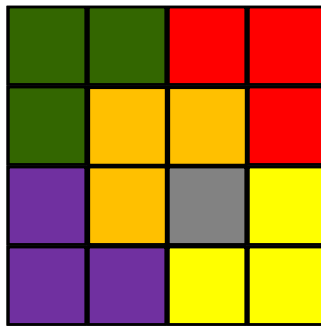
- Task: Place triominoes on an  $2^n \times 2^n$  ( $n \geq 1$ ) defective chessboard so that all  $2^n \times 2^n - 1$  nondefective positions are covered.
- Totally, we place  $(2^n \times 2^n - 1)/3$  triominoes.



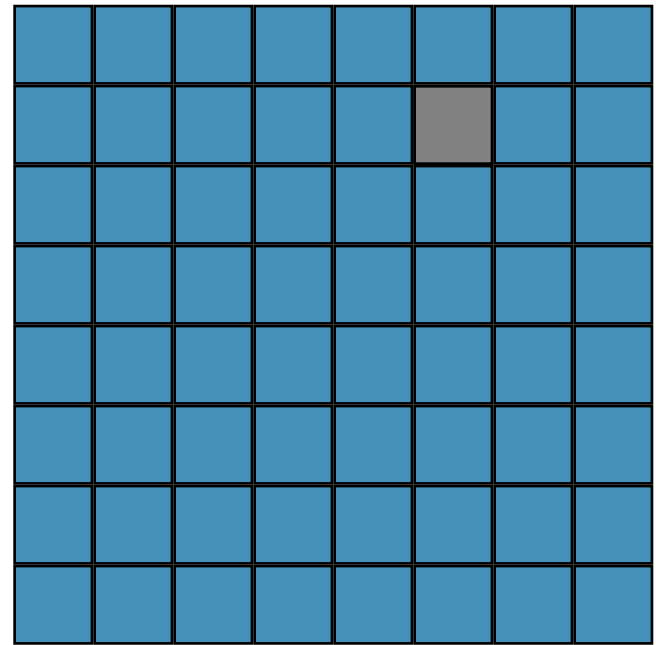
1×1



2×2



4×4

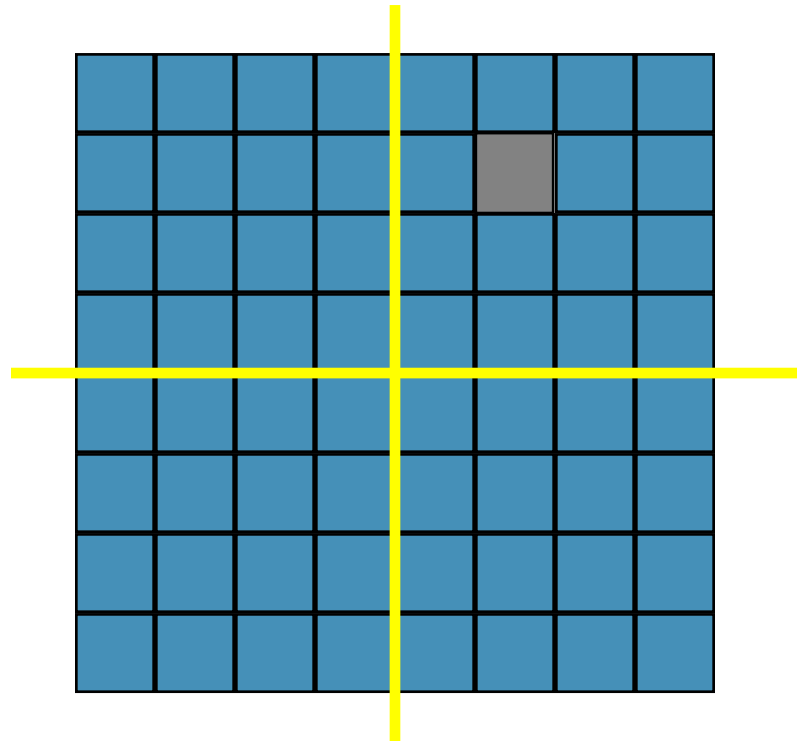


8×8



# Defective Chessboard

- How to use divide-and-conquer?

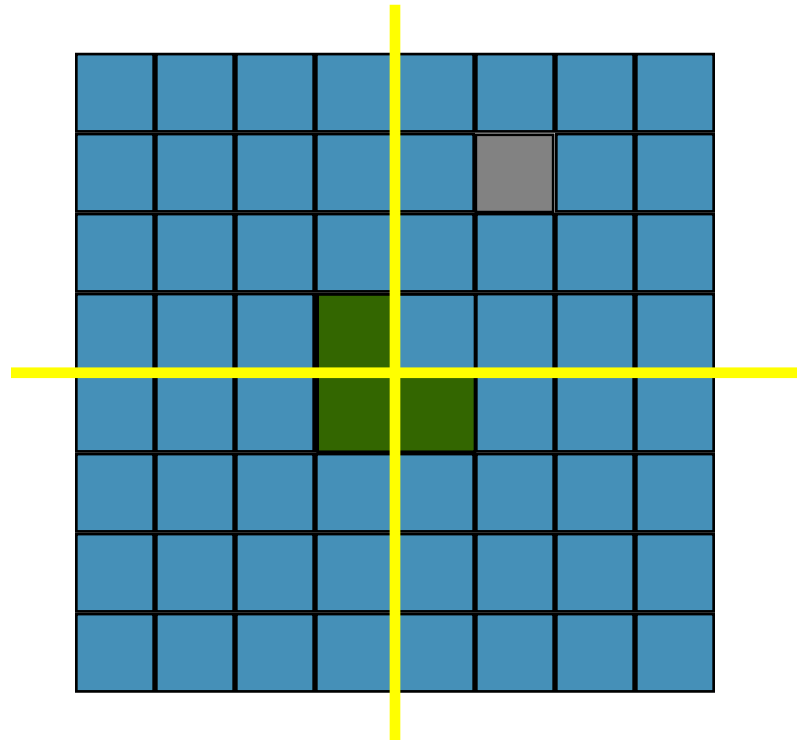


- If we divide it into 4  $2^{n-1} \times 2^{n-1}$  chessboard, only one is defective.



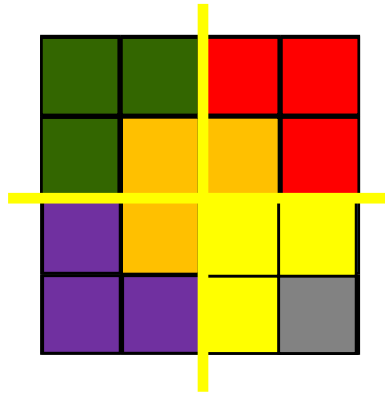
# Defective Chessboard

- Put one triomino at their common corner, which makes all 4 small chessboards have a defective position.



# Defective Chessboard

- Then, simply recursively solve this problem.



```
TileBoard(tr, tc, dr, dc, size)
```

```
1  if size = 1 return ok
```

```
2  tile ← tile + 1; t ← tile
```

```
3  s ← size/2
```

```
4  if dr < tr + s and dc < tc + s then
```

```
5      TileBoard(tr, tc, dr, dc, s)
```

```
6  else Board[tr + s - 1, tc + s - 1] ← t
```

```
7      TileBoard(tr, tc, tr + s - 1, tc + s - 1, s)
```

```
8  if dr < tr + s and dc ≥ tc + s then
```

```
9      TileBoard(tr, tc + s, dr, dc, s)
```

```
10 else Board[tr + s - 1, tc + s] ← t
```

```
11     TileBoard(tr, tc + s, tr + s - 1, tc + s, s)
```

```
12 if dr ≥ tr + s and dc < tc + s then
```

```
13     TileBoard(tr + s, tc, dr, dc, s)
```

```
14 else Board[tr + s, tc + s - 1] ← t
```

```
15     TileBoard(tr + s, tc, tr + s, tc + s - 1, s)
```

```
16 if dr ≥ tr + s and dc ≥ tc + s then
```

```
17     TileBoard(tr + s, tc + s, dr, dc, s)
```

```
18 else Board[tr + s, tc + s] ← t
```

```
19     TileBoard(tr + s, tc + s, tr + s, tc + s, s)
```

When *size* = 1, the board is 1×1.

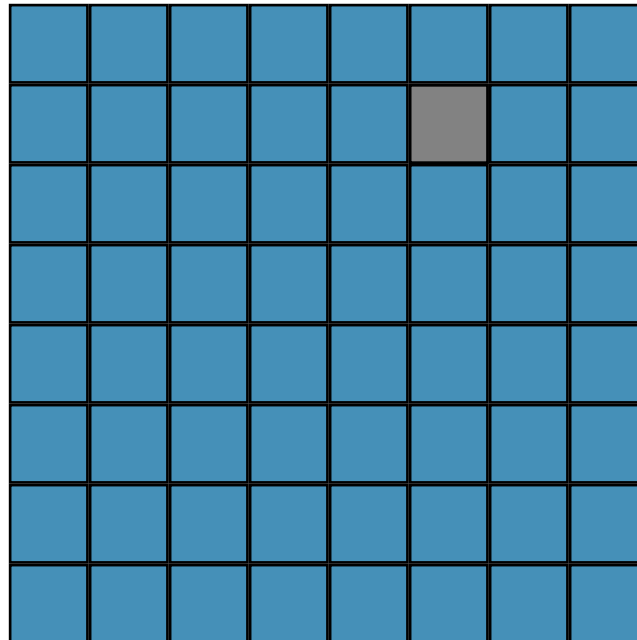
Check if defective position in this region.

Record triomino *t* on the board.

*tr*: row of left-upper square  
*tc*: column of left-upper square  
*dr*: row of defective square  
*dc*: column of defective square  
*tile*: accumulated triomino number  
*t*: current triomino number

# Classroom Exercise

Write down the triomino number in the following defective chessboard.





# Classroom Exercise

Solution:

3	3	4	4	8	8	9	9
3	2	2	4	8		7	9
5	2	6	6	10	7	7	11
5	5	6	1	10	10	11	11
13	13	14	1	1	18	19	19
13	12	14	14	18	18	17	19
15	12	12	16	20	17	17	21
15	15	16	16	20	20	21	21



# Time Complexity

- The recursion equation is:

$$T(n) = \begin{cases} \Theta(1) & n = 1 \\ 4T(n-1) + c & n > 1 \end{cases}$$

where

$$\begin{aligned} & 4T(n-1) + c \\ = & 4[4T(n-2) + c] + c \\ = & 4^2T(n-2) + 4c + c \\ = & 4^3T(n-3) + 4^2c + 4c + c \\ = & 4^{n-1}T(1) + 4^{n-2}c + \dots + 4c + c \\ = & \Theta(4^{n-1}) \end{aligned}$$





# DETERMINING THRESHOLD

# Determining Thresholds

- For matrix multiplication and large integer multiplication, when  $n$  is small, using standard algorithm will be even faster.
- For Mergesort, using recursive method on small array will also be slower than quadratic sorting algorithm like exchange sort.
- How to determine the threshold?



# Determining Thresholds

- If we have the recursive equation of Mergesort measured by computational time:

$$T(n) = 32n \lg n \mu s$$

and selection sort takes

$$T(n) = \frac{n(n-1)}{2} \mu s$$

- We can compare and get the threshold:

$$\frac{n(n-1)}{2} < 32n \lg n$$
$$n < 591.$$



# When Not to Use Divide-and-Conquer

- An instance of size  $n$  is divided into two or more instances each almost of size  $n$ .
  - $n$ th Fibonacci term:  $T(n) = T(n - 1) + T(n - 2) + 1$ .
  - Worst-case Quicksort is also not acceptable:  $T(n) = T(n - 1) + n - 1$ .
- An instance of size  $n$  is divided into almost  $n$  instances of size  $n/c$ , where  $c$  is a constant.
  - E.g.  $T(n) = T(n/2) + T(n/2) + \dots + T(n/2)$ .



# Conclusion

After this lecture, you should know:

- What is the key idea of divide-and-conquer.
- How to divide a big problem instance into several small instances.
- How to use recursion to design a divide-and-conquer algorithm.
- How Mergesort and Quicksort work and what are their complexity.



# Homework

- Page 63-65

5.1

5.3

5.8

5.10

5.18





# Experiment

- 5.19和5.20中选一题



# 谢谢

有问题欢迎随时跟我讨论



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