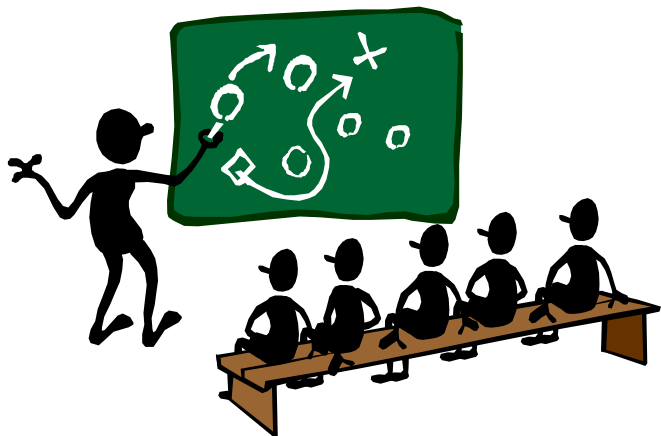


# Algorithms Chapter 1 & 2

## Getting Started



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# Algorithms

- Algorithm
  - is any **well-defined** computational procedure
  - takes some value or set of values as **input**
  - produces some value or set of values as **output**
  - ➔ is a sequence of computational steps that **transfer** the input into the output
- An algorithm is a tool for solving a well-specified **computational problem**
- The problem statement
  - specifies the desired input/output relationship
  - ➔ the algorithm is a specific procedure for achieving that I/O relationship

# Sorting Problem

- Sorting problem
  - input: a sequence of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$
  - output: a permutation  $\langle a'_1, a'_2, \dots, a'_n \rangle$   
of the input sequence such that  $a'_1 \leq a'_2 \leq \dots \leq a'_n$
- For example
  - input:  $\langle 31, 41, 59, 26, 41, 58 \rangle$  ← an instance of the problem
  - output:  $\langle 26, 31, 41, 41, 58, 59 \rangle$
- An **instance** of the problem
  - consists of the input **needed** to compute a solution (output) to the problem

# Correctness of Algorithm

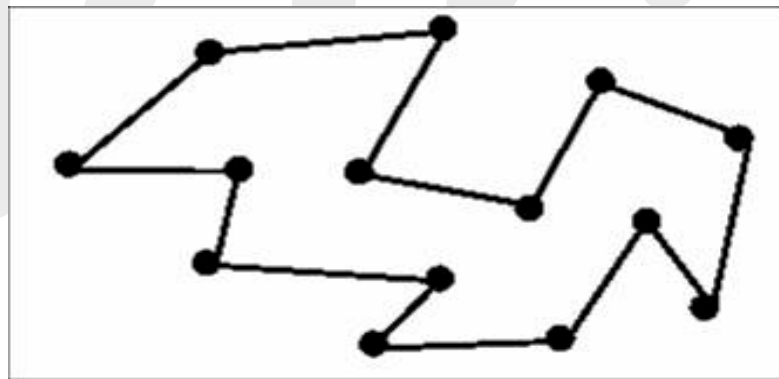
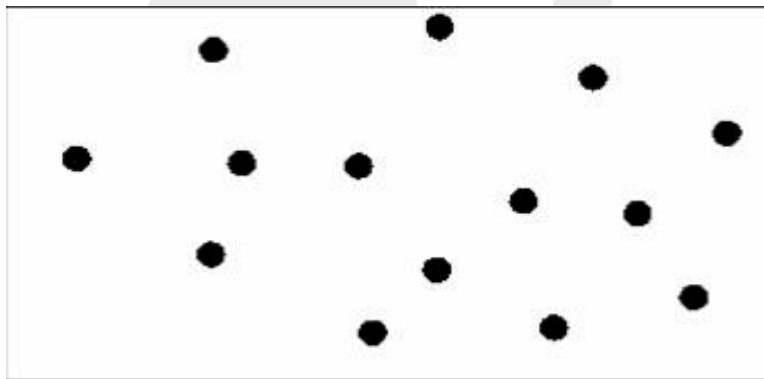
- Correct algorithm
  - halts with the correct output for **every** input instance
- A correct algorithm solves the given computational problem
- Incorrect algorithm
  - might halt with an incorrect output
  - might not halt for some input instances

# Beyond the Correctness

- Make the algorithm correct first
- Reality
  - computers are not infinitely fast
  - memory is not infinitely large
- In addition to correctness, a good algorithm should be efficient in both time and space
- Correctness vs. Efficiency ?

# Traveling Salesman Problem (TSP)

- **Input:** A set of points (cities)  $P$  together with a distance  $d(p, q)$  between any pair  $p, q \in P$
- **Output:** What is the shortest circular route that starts and ends at a given point and visits all the points

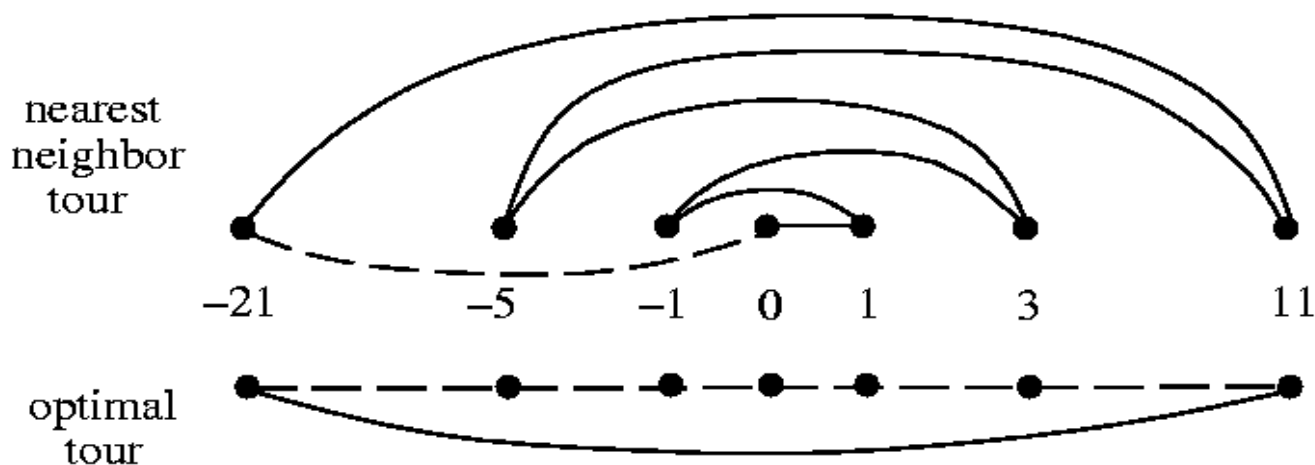


- Correct and efficient algorithms?

# Nearest Neighbor Tour

1. pick and visit an initial point  $p_0$
2.  $P \leftarrow p_0$
3.  $i \leftarrow 0$
4. **while** there are unvisited points **do**
5.     visit  $p_i$ 's nearest unvisited point  $p_{i+1}$
6.      $i \leftarrow i + 1$
7. return to  $p_0$  from  $p_i$

- Simple to implement and very efficient, but **incorrect!**



# A Correct but Inefficient Algorithm

```
1.  $d \leftarrow \infty$ 
2. for each of the  $n!$  permutations  $\pi_i$  of the  $n$  points
3.   if  $(\text{cost}(\pi_i) \leq d)$  then
4.      $d \leftarrow \text{cost}(\pi_i)$ 
5.      $T_{min} \leftarrow \pi_i$ 
6. return  $T_{min}$ 
```

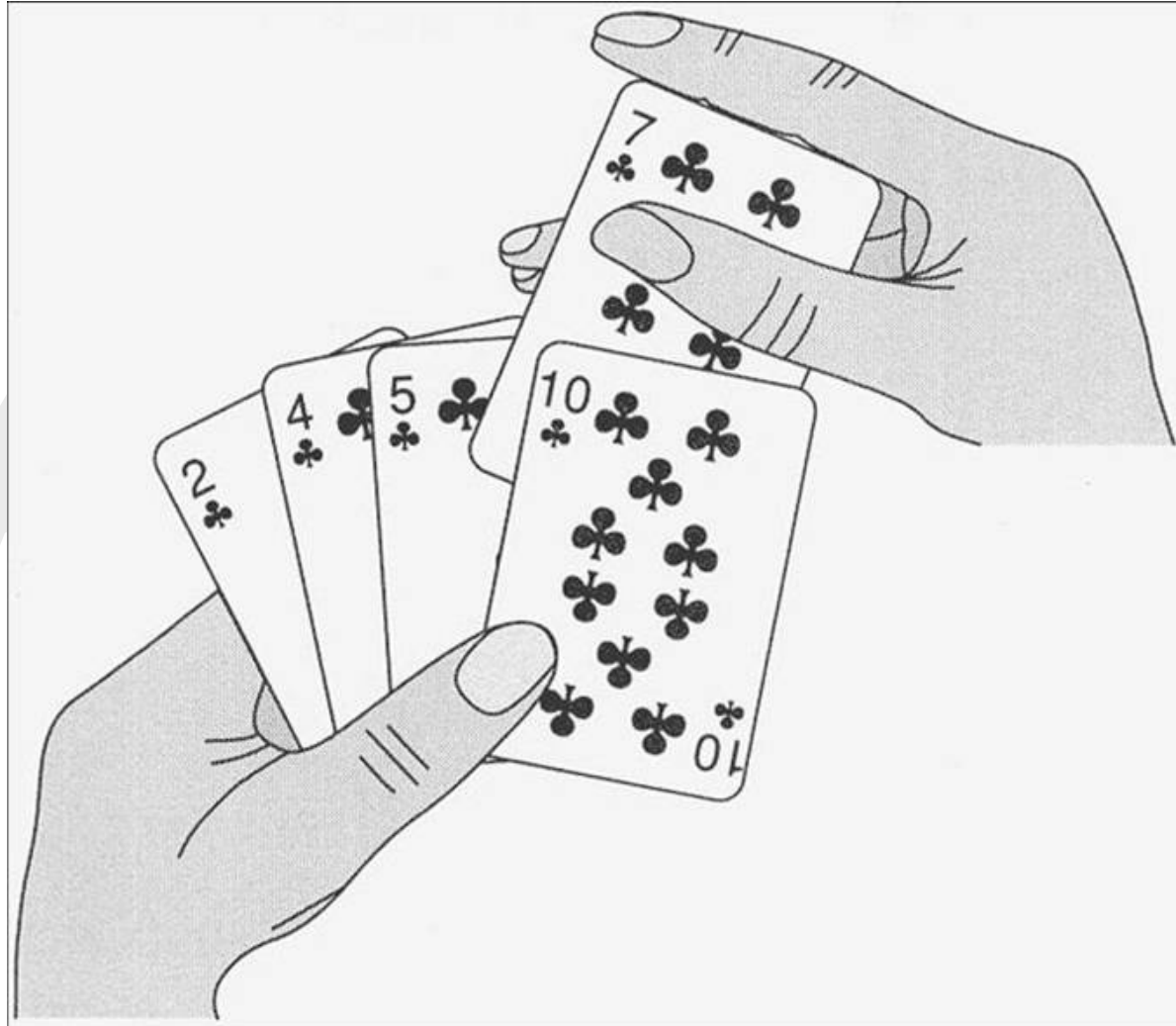
- **Correctness?** Try all possible orderings of the points →  
Guarantee to end up with the shortest possible tour
- **Efficiency?** Try  **$n!$**  possible routes!
  - 120 routes for 5 points; 3,628,800 routes for 10 points
  - no known efficient and correct algorithm for TSP!
  - TSP is **NP-complete**
- **Think Deep:** What will you do?



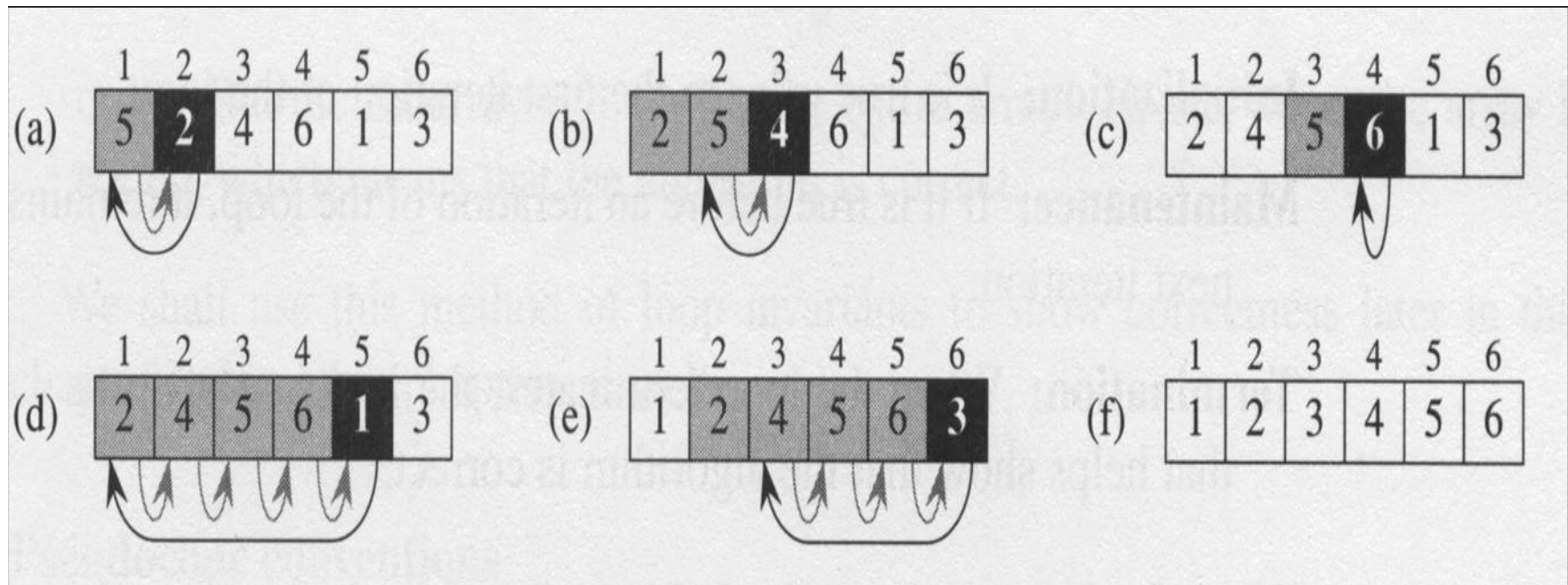
# Sorting Problem

- Sorting problem
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- For example
  - input:  $\langle 31, 41, 59, 26, 41, 58 \rangle$
  - output:  $\langle 26, 31, 41, 41, 58, 59 \rangle$
- The numbers under sorting are also known as **keys**
- How to sort?

# Sorting a Hand of Cards



# Insertion Sort



**Insertion sort is an efficient algorithm for sorting a small number of elements**

# Pseudocode

- **Insertion sort**

Insertion-Sort( $A$ )

1 **for**  $j \leftarrow 2$  **to**  $\text{length}[A]$

2   **do**  $\text{key} \leftarrow A[j]$

3       \* Insert  $A[j]$  into the sorted sequence  $A[1 .. j - 1]$

4        $i \leftarrow j - 1$

5       **while**  $i > 0$  and  $A[i] > \text{key}$

6           **do**  $A[i + 1] \leftarrow A[i]$



7            $i \leftarrow i - 1$

8        $A[i + 1] \leftarrow \text{key}$

# Analyzing Algorithms

- Analyzing an algorithm has come to mean predicting the resources that the algorithm requires
  - CPU time
  - memory space
- Assumption
  - single processor  
(not multiprocessor, not parallel computing)
  - RAM as memory (not disk)

# Running Time vs. Input Size

- The **running time** needed by Insertion-Sort depends on the **input size**
  - 10 numbers vs. 1 million numbers
- In general
  - input size   $\rightarrow$  running time 
  - the running time is typically described as a function of the input size
- How to calculate running time?
  - count the number of primitive operations (steps)
  - machine-independent

# Analysis of Insertion Sort

Insertion-sort( $A$ )	cost	times
1 <b>for</b> $j \leftarrow 2$ <b>to</b> $\text{length}[A]$	$c_1$	$n$
2 <b>do</b> $\text{key} \leftarrow A[j]$	$c_2$	$n - 1$
3         * Insert $A[j]$ into the sorted sequence $A[1 \dots j - 1]$	$0$	
4 $i \leftarrow j - 1$	$c_4$	$n - 1$
5 <b>while</b> $i > 0$ and $A[i] > \text{key}$	$c_5$	$\sum_{j=2}^n t_j$
6 <b>do</b> $A[i + 1] \leftarrow A[i]$	$c_6$	$\sum_{j=2}^n (t_j - 1)$
7 $i \leftarrow i - 1$	$c_7$	$\sum_{j=2}^n (t_j - 1)$
8 $A[i + 1] \leftarrow \text{key}$	$c_8$	$n - 1$

$t_j$ : the number of times the while loop test in Line 5 is executed for the value of  $j$

# Best-Case Analysis

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

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

- $T(n)$  does not **solely** depend on  $n$
- Best case: the array is already sorted

–  $t_j = 1$  for  $j = 2, 3, \dots, n$

➡ 
$$\begin{aligned} T(n) &= c_1n + c_2(n-1) + c_4(n-1) + c_5(n-1) + c_8(n-1) \\ &= (c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8) \end{aligned}$$

–  $T(n)$  is a **linear** function of  $n$



# Worst-Case Analysis

- Worst case: the array is sorted in **reverse** order

–  $t_j = j$

➔ 
$$\begin{aligned} T(n) &= c_1 n + c_2(n-1) + c_4(n-1) + c_5\left(\frac{n(n+1)}{2} - 1\right) + \\ &\quad c_6\left(\frac{n(n-1)}{2}\right) + c_7\left(\frac{n(n-1)}{2}\right) + c_8(n-1) \\ &= \left(\frac{c_5 + c_6 + c_7}{2}\right)n^2 - \left(c_1 + c_2 + c_4 + \frac{c_5 - c_6 - c_7}{2} + c_8\right)n \\ &\quad - (c_2 + c_4 + c_5 + c_8) \end{aligned}$$

- $T(n)$  is a **quadratic** function of  $n$

# Best/Worst/Average Cases

- We usually concentrate on finding the worst-case running time
  - the longest running time for any input size of  $n$
- Why?
  - worst case gives an **upper bound**
  - for some algorithms, the worst case occurs fairly often
    - e.g., searches in databases
  - the average case is often roughly as bad as the worst case
    - e.g., for insertion sort,  $T(n)$  is still a quadratic function of  $n$

# Order of Growth

- Why is the running time typically described as a function of the input size?
  - to know how the running time grows as the input size grows
- It is the **rate of growth**, or **order of growth**, of the running time that really interests us
- For insertion sort
  - the worst-case running time,  $\Theta(n^2)$
- Assume Algo1 with  $\Theta(n^2)$  and Algo2 with  $\Theta(n^3)$ 
  - Algo1 is considered more efficient than Algo2
  - i.e., for a large enough  $n$ , Algo1 runs faster than Algo2

# Divide-and-Conquer

- Many algorithms are recursive
  - they call themselves recursively one or more times to deal with subproblems
- Divide-and-conquer paradigm
  - **divide** the problem into a number of subproblems
  - **conquer** the subproblems by solving them **recursively**
  - **combine** the solutions to the subproblems into the solution to the original problem

# Merge Sort

- Divide-and-conquer
  - divide: divide the  $n$ -element sequence into 2 subsequences of  $n/2$  elements each
  - conquer: sort the 2 subsequences **recursively**
  - combine: merge the 2 sorted subsequences to produce the answer

# MERGE(A, p, q, r) (1/2)

- MERGE(A, p, q, r)
    - given that  $A[p \dots q]$  and  $A[q+1 \dots r]$  are 2 sorted subarrays
    - MERGE(A, p, q, r) produce a sorted  $A[p \dots r]$  subarray
- 

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



**sentinels**

# MERGE(A, p, q, r) (2/2)

10  $i \leftarrow 1$

11  $j \leftarrow 1$

12 **for**  $k \leftarrow p$  **to**  $r$

13     *do if*  $L[i] \leq R[j]$

14         **then**  $A[k] \leftarrow L[i]$

15          $i \leftarrow i + 1$

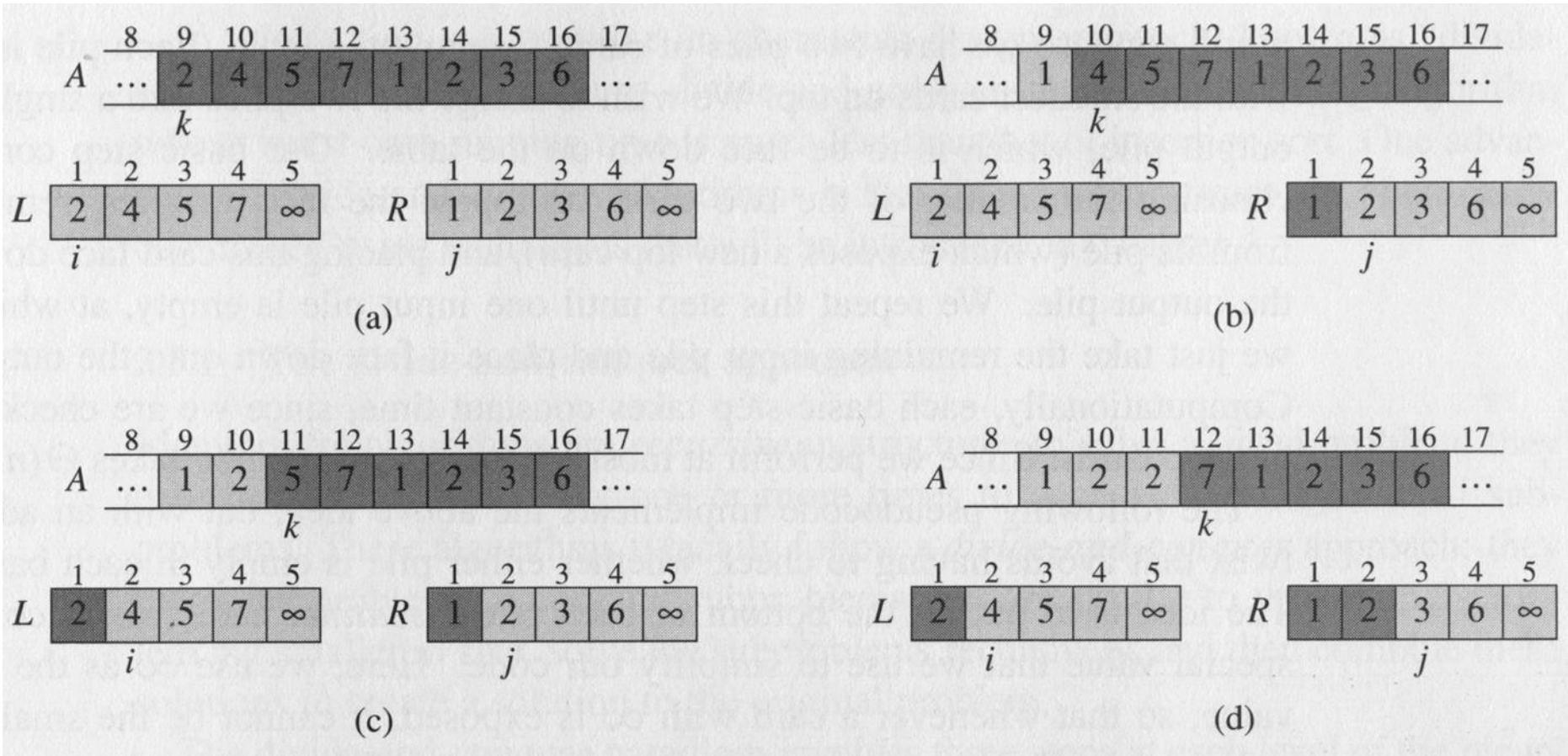
16     **else**  $A[k] \leftarrow R[j]$

17          $j \leftarrow j + 1$

**Time:  $\Theta(n)$ ,  $n$  = number of elements**

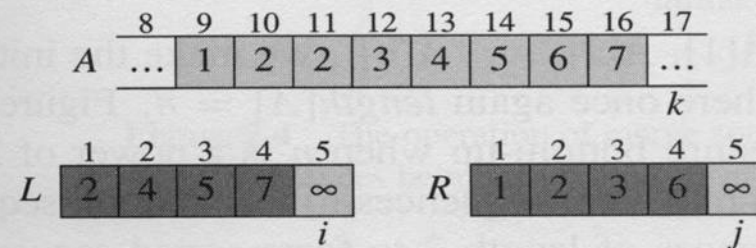
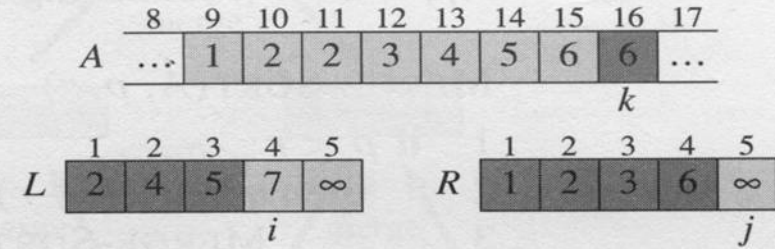
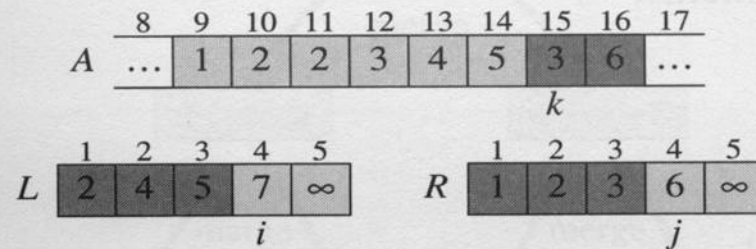
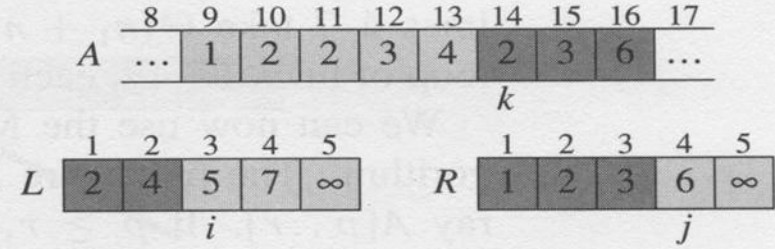
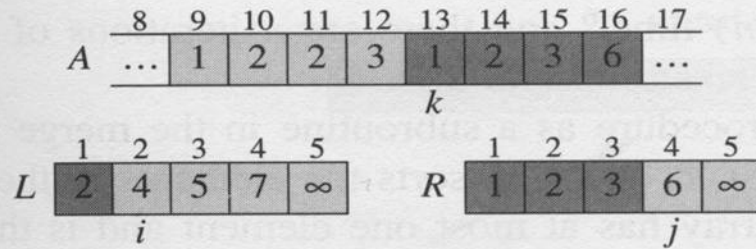
# Illustration of MERGE (1/2)

## MERGE(A, 9, 12, 16)





# Illustration of MERGE (2/2)

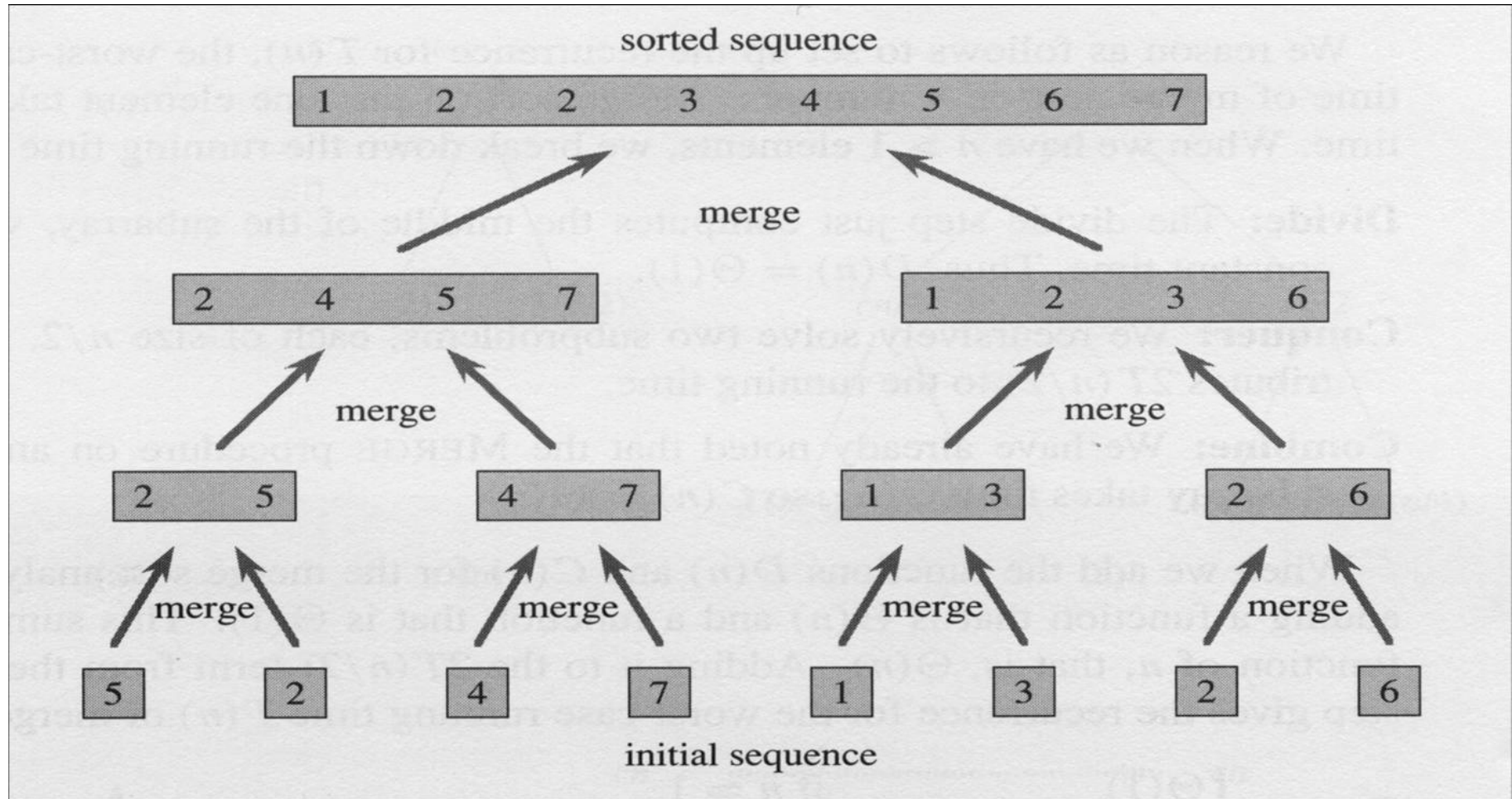


# Merge Sort

- MERGE-SORT( $A, p, r$ )
    - sort the elements in  $A[p \dots r]$
- 

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

# Illustration of MERGE-SORT



# Analysis of Merge Sort (1/2)

- The running time of a recursive algorithm can be expressed as a **recurrence** equation
  - discuss later in Chap 4
  - “**master theorem**” can be found in discrete math
- Analysis of merge sort

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

➡  $T(n) = \Theta(n \lg n)$  by the **master theorem**

**Hence the merge sort is better than the insertion sort when n is large**



# Analysis of Merge Sort (2/2)

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

