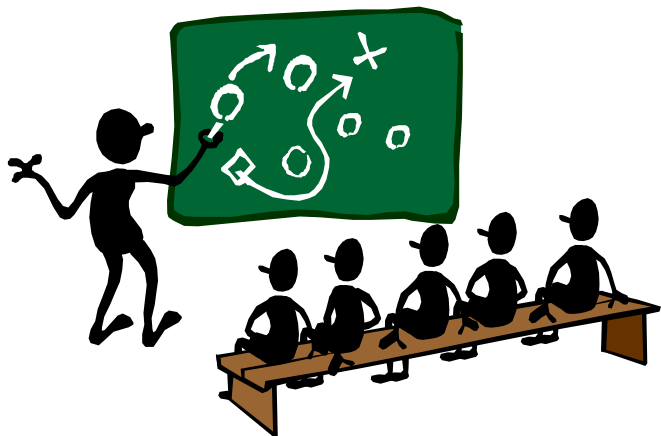


Algorithms Chapter 1 & 2

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

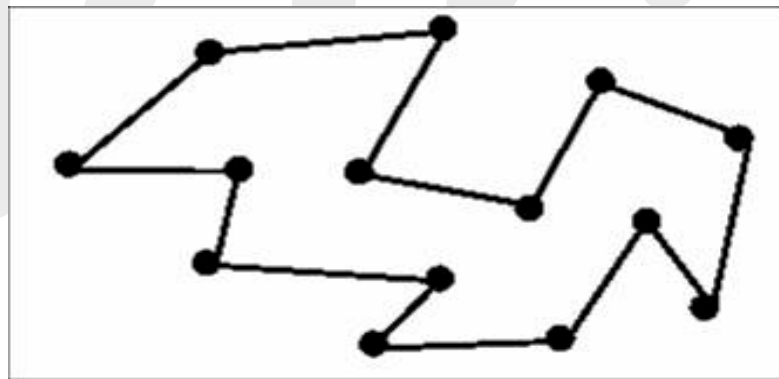
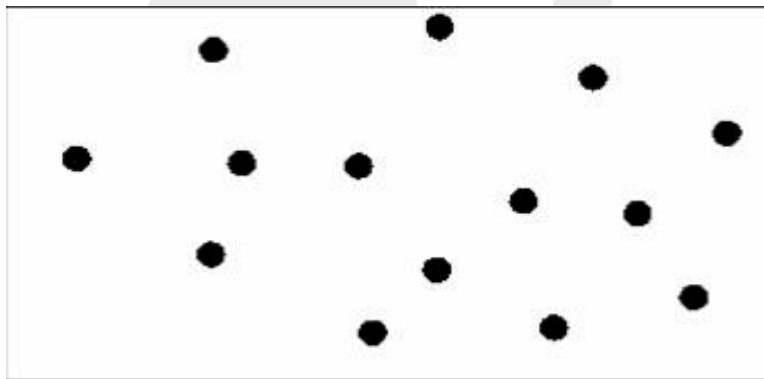
- Correct algorithm
 - halts with the correct output for **every** input instance
- A correct algorithm solves the given computational problem
- Incorrect algorithm
 - might halt with an incorrect output
 - might not halt for some input instances

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

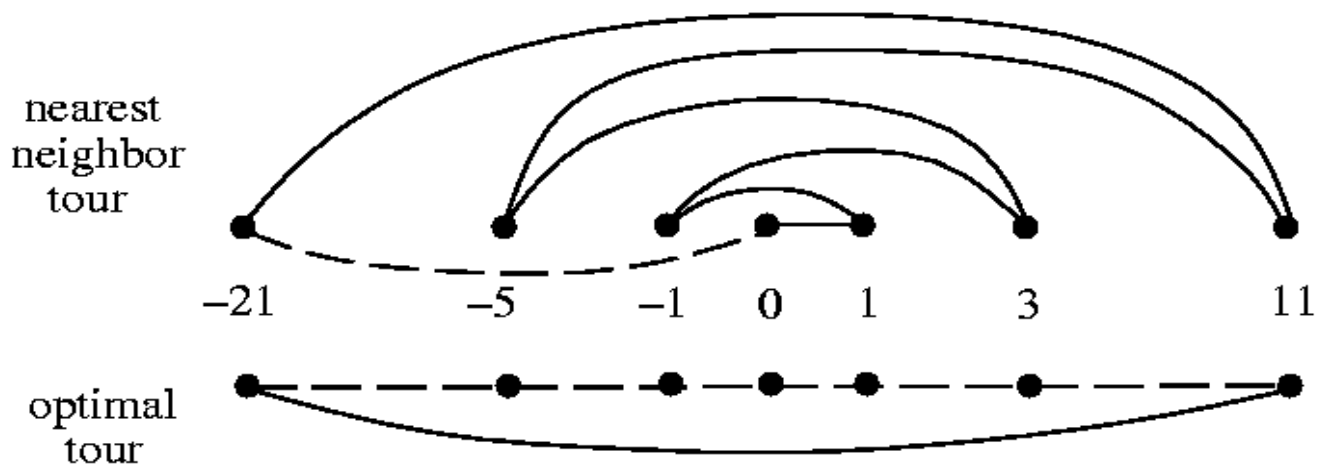


- Correct and efficient algorithms?

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

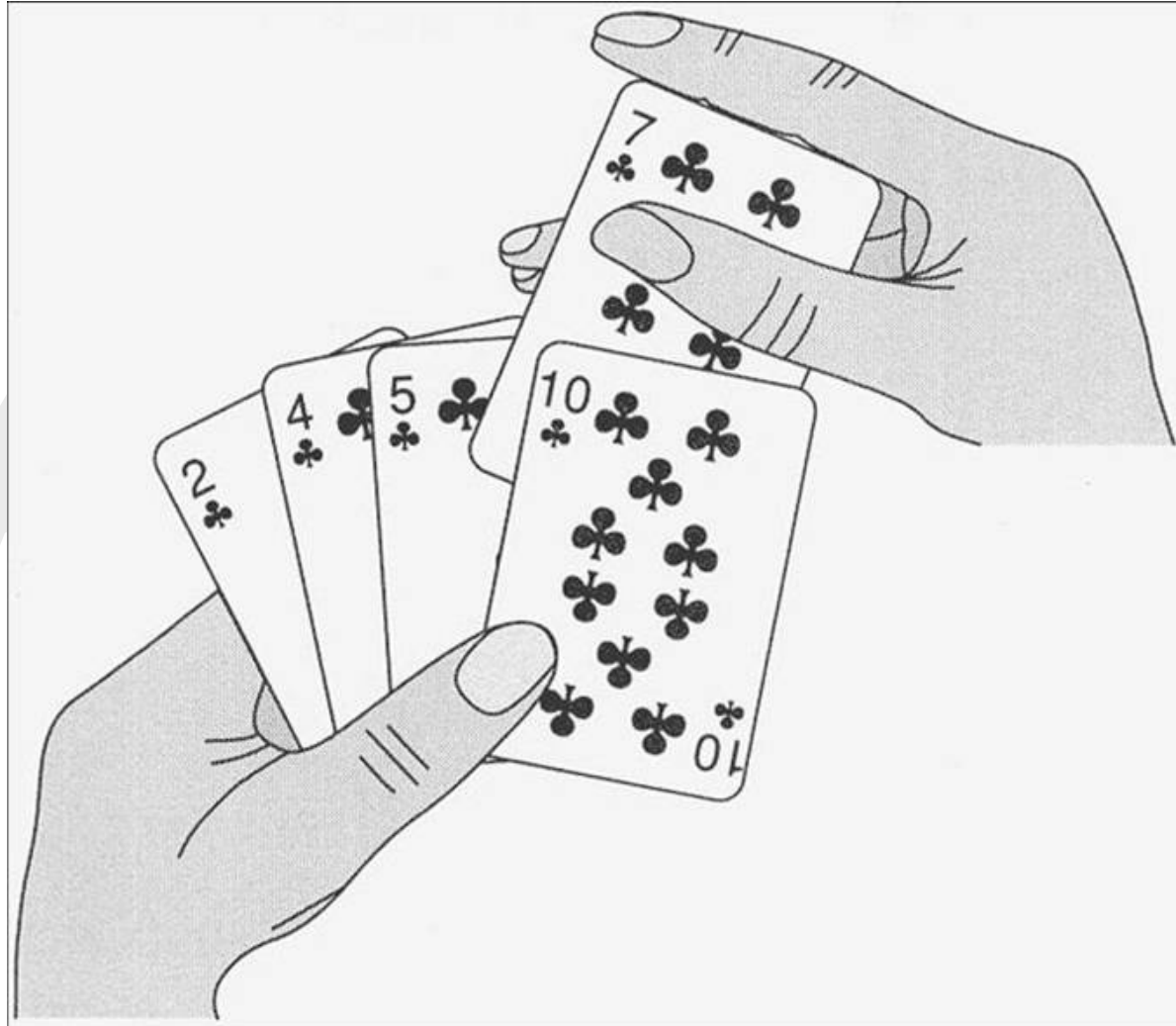
```
1.  $d \leftarrow \infty$ 
2. for each of the  $n!$  permutations  $\pi_i$  of the  $n$  points
3.   if  $(\text{cost}(\pi_i) \leq d)$  then
4.      $d \leftarrow \text{cost}(\pi_i)$ 
5.      $T_{min} \leftarrow \pi_i$ 
6. return  $T_{min}$ 
```

- **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?

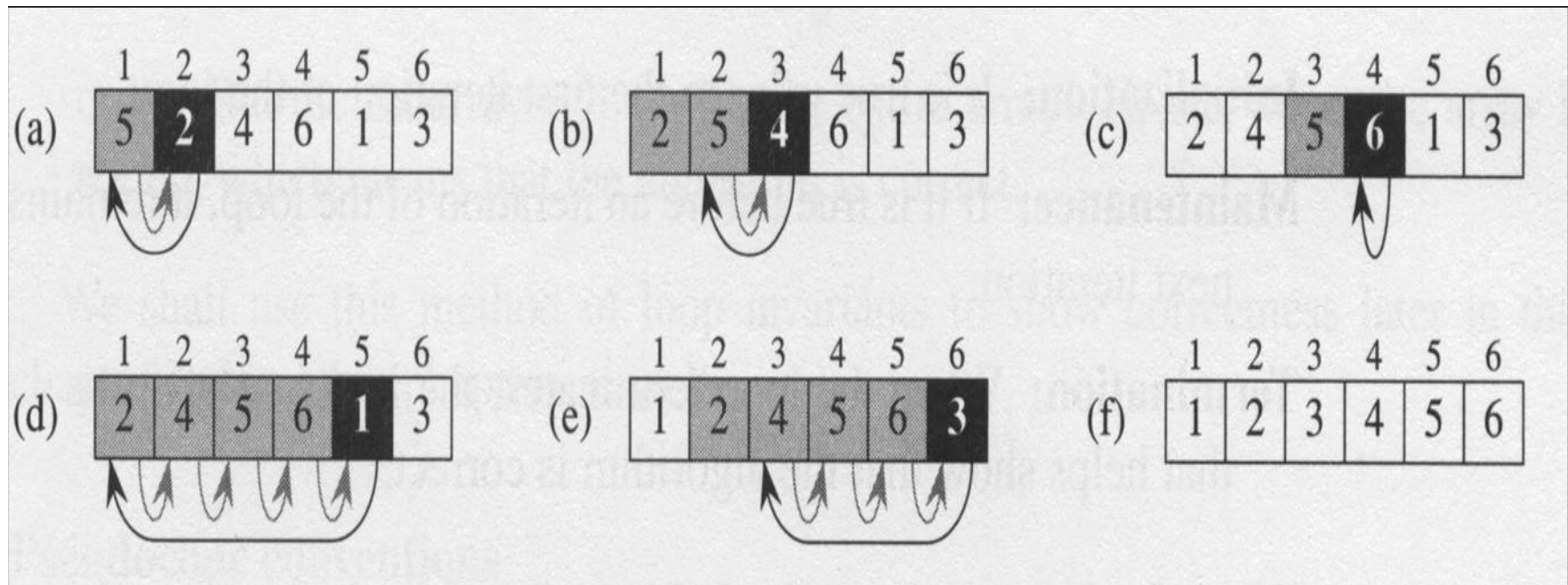
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- For example
 - input: $\langle 31, 41, 59, 26, 41, 58 \rangle$
 - output: $\langle 26, 31, 41, 41, 58, 59 \rangle$
- The numbers under sorting are also known as **keys**
- How to sort?

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   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 for $j \leftarrow 2$ to $\text{length}[A]$	c_1	n
2 do $\text{key} \leftarrow A[j]$	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 while $i > 0$ and $A[i] > \text{key}$	c_5	$\sum_{j=2}^n t_j$
6 do $A[i + 1] \leftarrow A[i]$	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 $i \leftarrow 1$

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

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 $j \leftarrow j + 1$

Illustration of MERGE (1/2)

MERGE(A, 9, 12, 16)

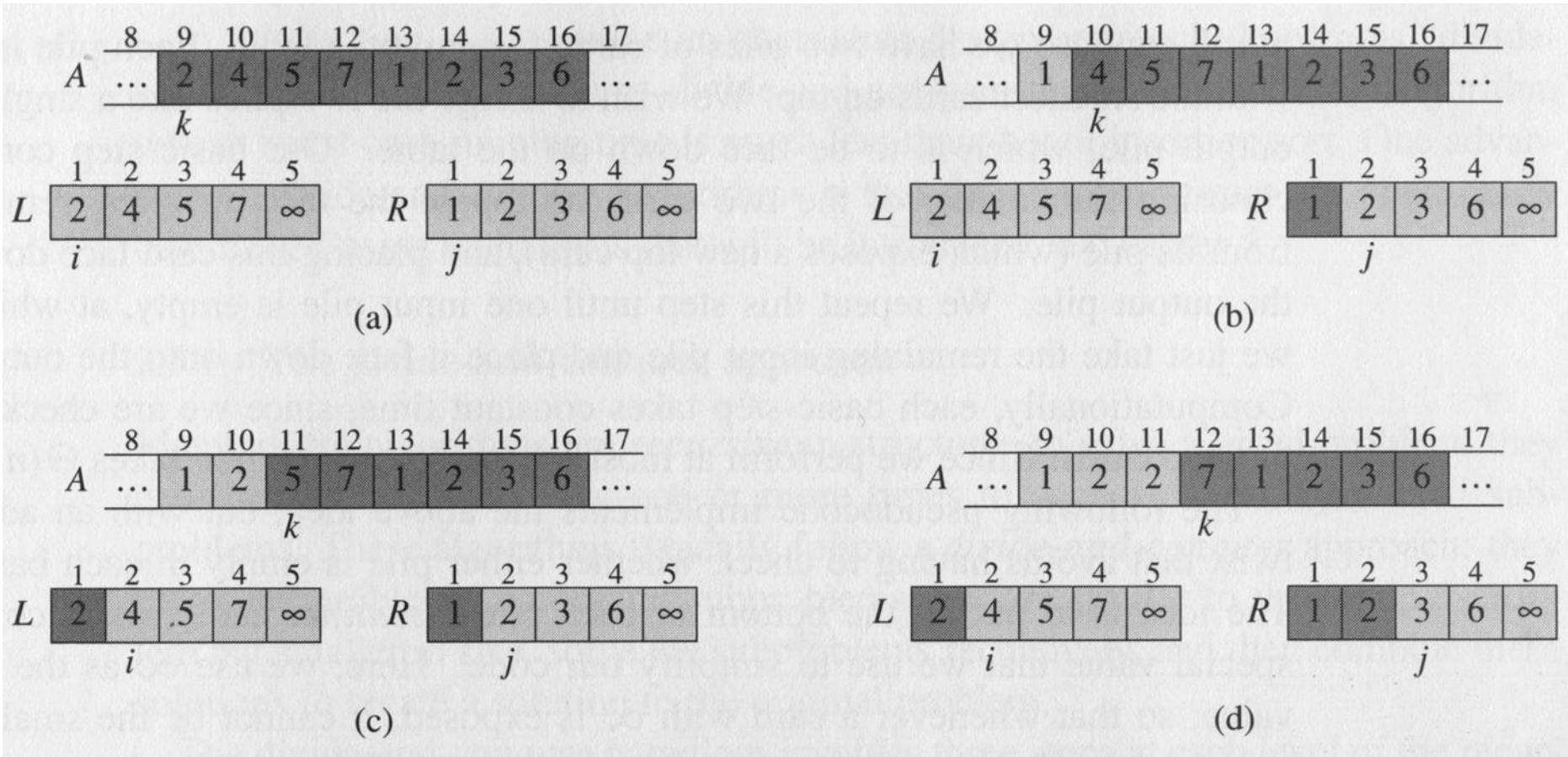
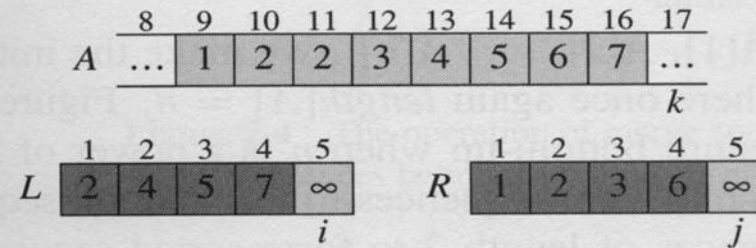
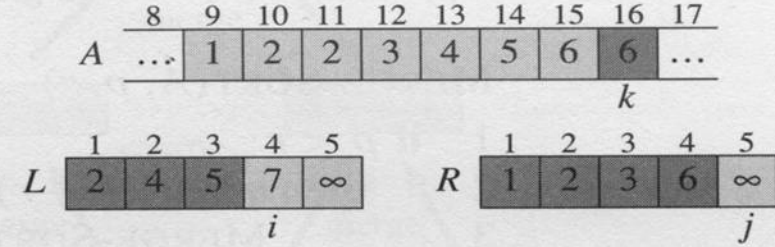
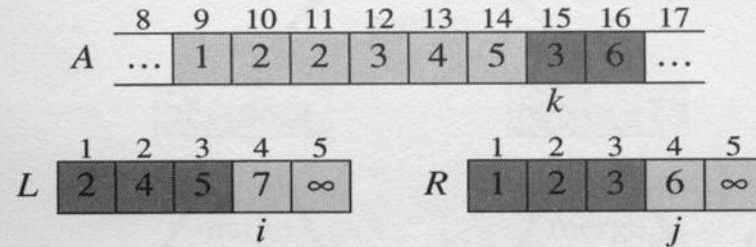
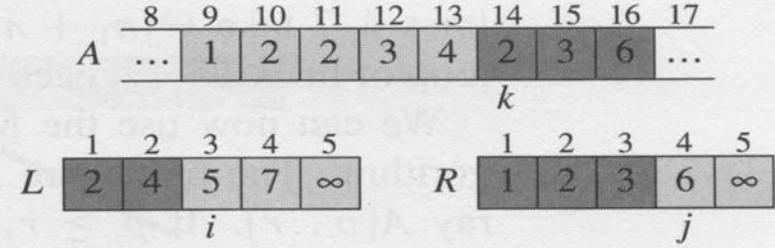
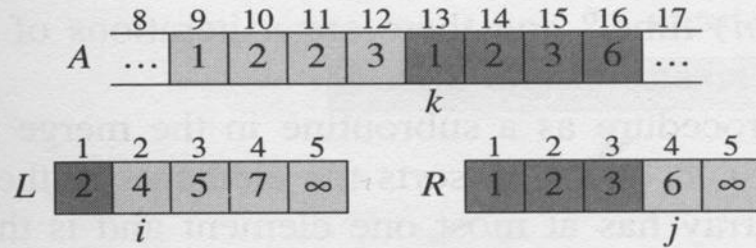


Illustration of MERGE (2/2)

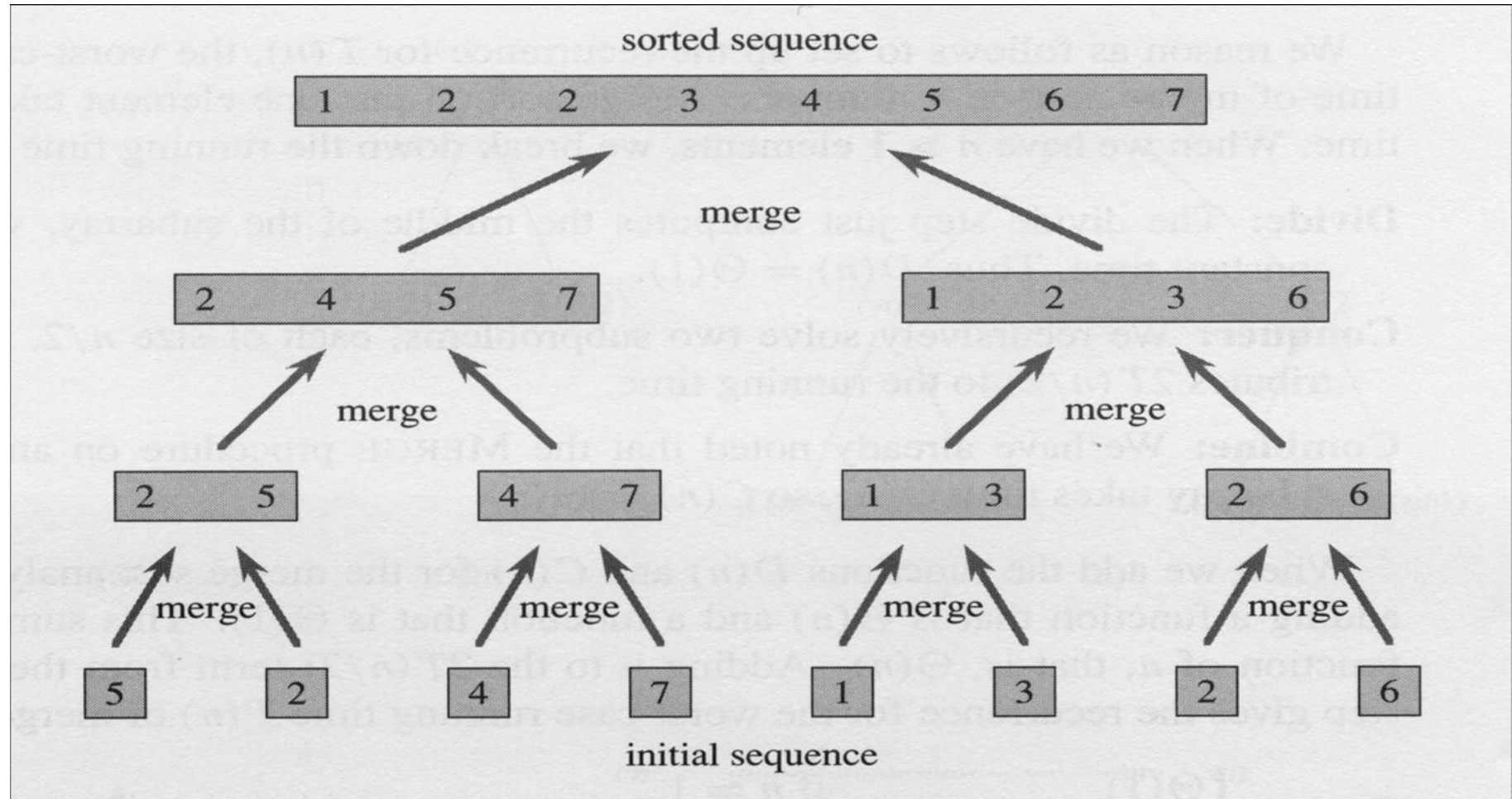


Merge Sort

- 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)

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

