

Chapter 2

Getting Started

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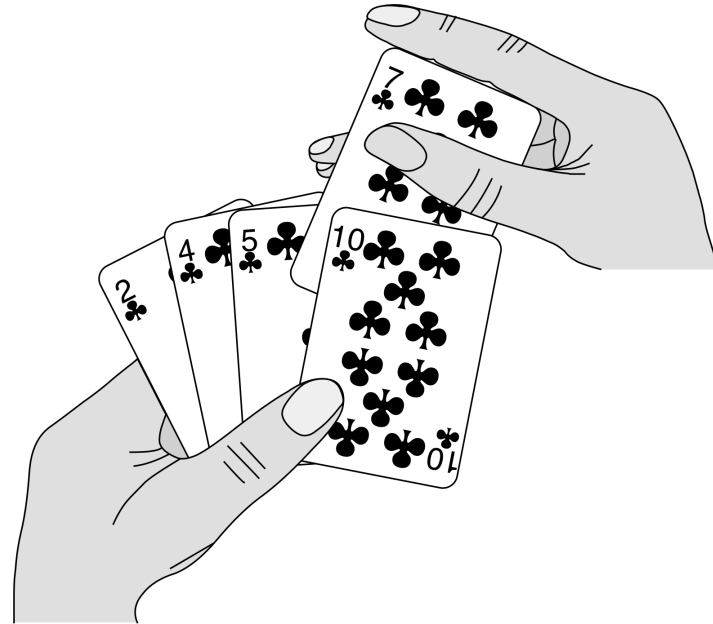
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90th Anniversary of NCKU

2.1 Insertion sort

- **Example:** 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
$$a'_1 \leq a'_2 \leq \dots \leq a'_n$$
- The number that we wish to sort are known as the *keys*.



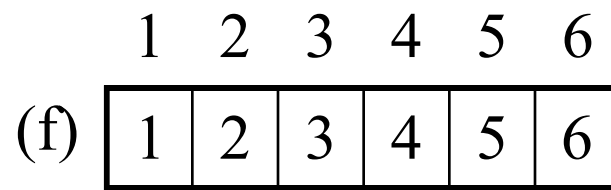
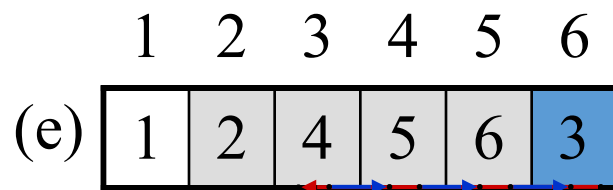
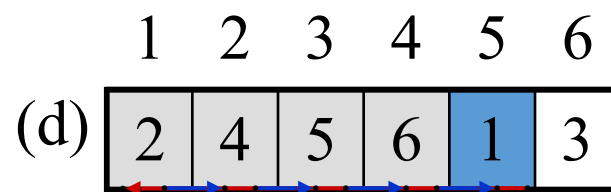
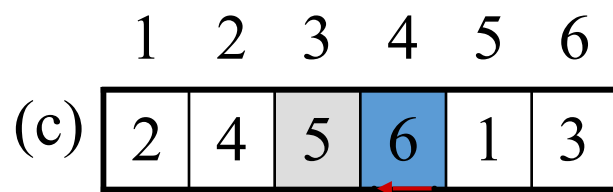
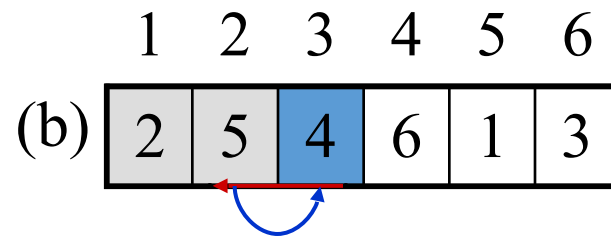
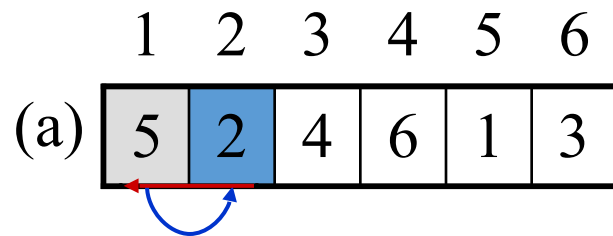
Pseudocode

- 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 $A[1, \dots, j-1]$
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. $A[i+1] \leftarrow \text{key}$

The operation of Insertion-Sort



- ***Sorted in place:***

- The numbers are rearranged within the array A , with at most a constant number of them sorted outside the array at any time.

- ***Loop invariant:***

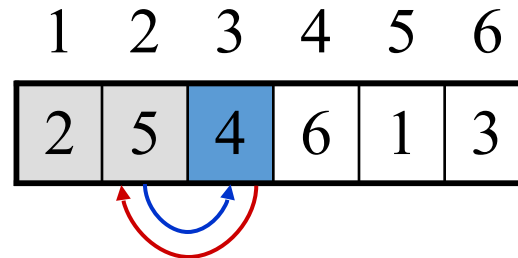
- At the start of each iteration of the for loop of line 1-8, the subarray $A[1, \dots, j - 1]$ consists of the elements originally in $A[1, \dots, j - 1]$ but in sorted order.

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- We use loop invariants to help us understand why an algorithm is correct. We must show three things about a loop invariant:
 - **Initialization:** It is true prior to the first iteration of the loop.
 - **Maintenance:** If it is true before an iteration of the loop, it remains true before the next iteration.
 - **Termination:** When the loop terminates, the invariant gives us a useful property that helps show that the algorithm is correct.

- Let us see how these properties hold for insertion sort.

- **Initialization:** When $j = 2$, $A[1, \dots, j - 1] = A[1]$ is true.

- **Maintenance:**



The subarray $A[1, \dots, j]$ then consists of the elements originally in $A[1, \dots, j]$, but in sorted order. Incrementing j for the next iteration of the for loop then preserves the loop invariant.

- **Termination:** The for loop to terminate is that $j > A.length = n$. When $j = n + 1$, the subarray $A[1, \dots, n]$ then consists of the elements originally in $A[1, \dots, n]$, but in sorted order.

2.2 Analyzing algorithms

- Analyzing an algorithm has come to mean predicting the resources that the algorithm requires.
 - **Resources:** memory, communication, bandwidth, logic gate, **time**.
 - **Assumption:** one processor, **RAM**
 - constant-time instruction: **arithmetic** (add, subtract, multiply, divide, remainder, floor, ceiling); **data movement** (load, store, copy); **control** (conditional and unconditional branch, subroutine call and return)
 - Data type: integer and floating point
 - Limit on the size of each word of data

2.2 Analyzing algorithms

- The best notion for **input size** depends on the problem being studied.
- The **running time** of an algorithm on a particular input is the number of primitive operations or “steps” executed. It is convenient to define the notion of step so that it is as machine-independent as possible.



Analysis of 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 $A[1, \dots, j-1]$
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. $A[i+1] \leftarrow \text{key}$

cost

times

c_1

n

c_2

$n-1$

0

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_j : the number of times the while loop test in line 5 is executed for the value of j .



Analysis of insertion sort

- The running time

$$T(n) = c_1n + c_2(n - 1) + c_4(n - 1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n - 1)$$

- $t_j = 1$, for $j = 2, 3, \dots, n$: Linear function on n

$$\begin{aligned} T(n) &= c_1n + c_2(n - 1) + c_4(n - 1) + c_5(n - 1) + c_6(n - 1) + c_7(n - 1) + c_8(n - 1) \\ &= (c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8) \end{aligned}$$

Analysis of insertion sort

- $t_j = j$, for $j = 2, 3, \dots, n$: Quadratic function on n

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

Noting that, $\sum_{j=2}^n j = \frac{n(n+1)}{2} - 1$ and $\sum_{j=2}^n (j-1) = \frac{n(n-1)}{2}$

Worst-case and average-case analysis

- Usually, we concentrate on finding only on the *worst-case running time*.
- Reason:
 - It is an upper bound on the running time.
 - The worst case occurs fair often.
 - The average case is often as bad as the worst case. For example, the insertion sort. Again, quadratic function.

Order of growth

- In some particular cases, we shall be interested in *average-case*, or *expect* running time of an algorithm.
- It is the *rate of growth*, or *order of growth*, of the running time that really interests us.

2.3 Designing algorithms

- There are many ways to design algorithms:
 - **Incremental approach:** having sorted the subarray $A[1 \dots j - 1]$, we inserted the single element $A[j]$ into its proper place, yielding the sorted subarray $A[1 \dots j]$.
Ex. insertion sort
 - **Divide-and-conquer:** merge sort
 - recursive:
 - divide
 - conquer
 - combine

Pseudocode

- Merge sort

Merge(A, p, q, r)

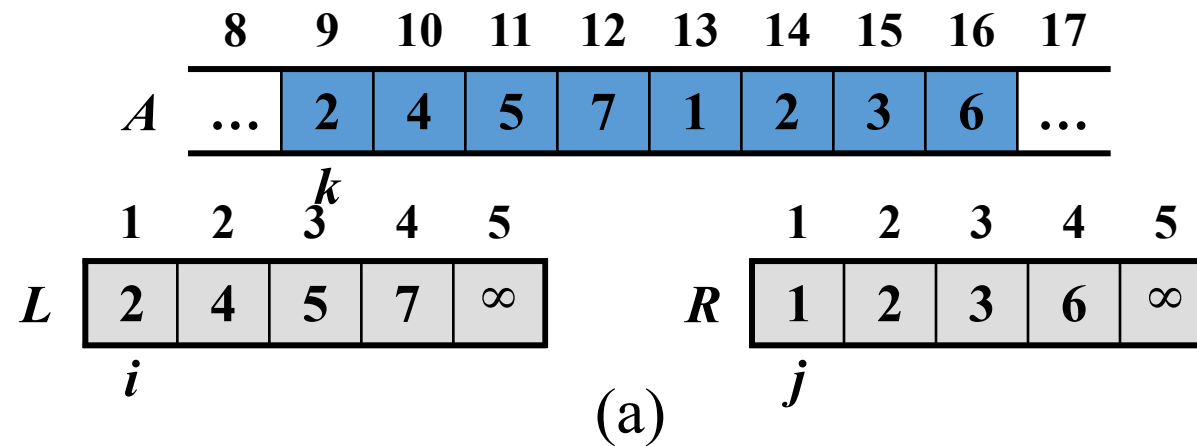
1. $n_1 \leftarrow q - p + 1$
2. $n_2 \leftarrow r - q$
3. create array $L[1, \dots, n_1 + 1]$ and $R[1, \dots, n_2 + 1]$
4. **for** $i \leftarrow 1$ **to** n_1
5. **do** $L[i] \leftarrow A[p + i - 1]$
6. **for** $j \leftarrow 1$ **to** n_2
7. **do** $R[j] \leftarrow A[q + j]$
8. $L[n_1 + 1] \leftarrow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

Pseudocode

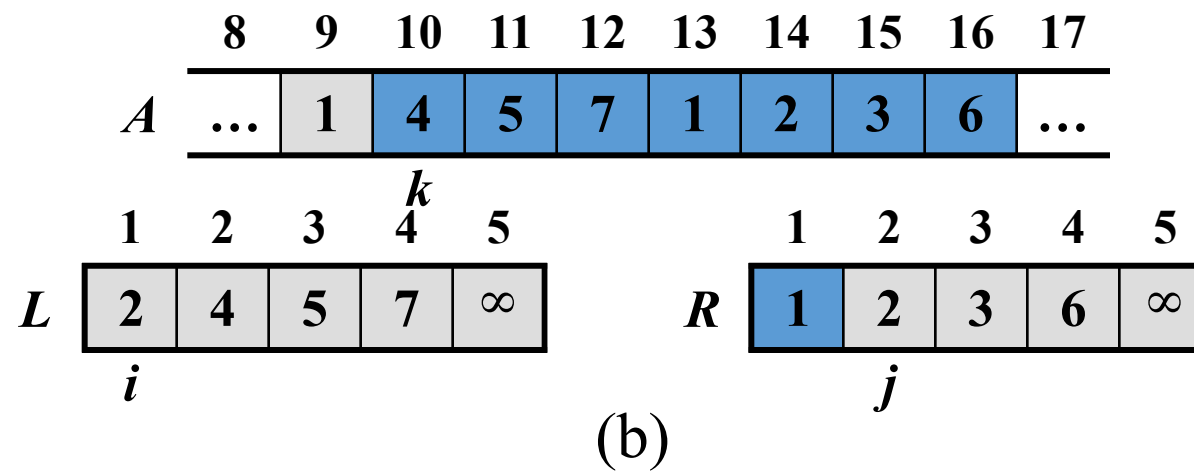
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10.  $i \leftarrow 1$ 
11.  $j \leftarrow 1$ 
12. for  $k \leftarrow p$  to  $r$ 
13.   do if  $L[i] \leq R[j]$ 
14.     then  $A[k] \leftarrow L[i]$ 
15.          $i \leftarrow i + 1$ 
16.     else  $A[k] \leftarrow R[j]$ 
17.          $j \leftarrow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



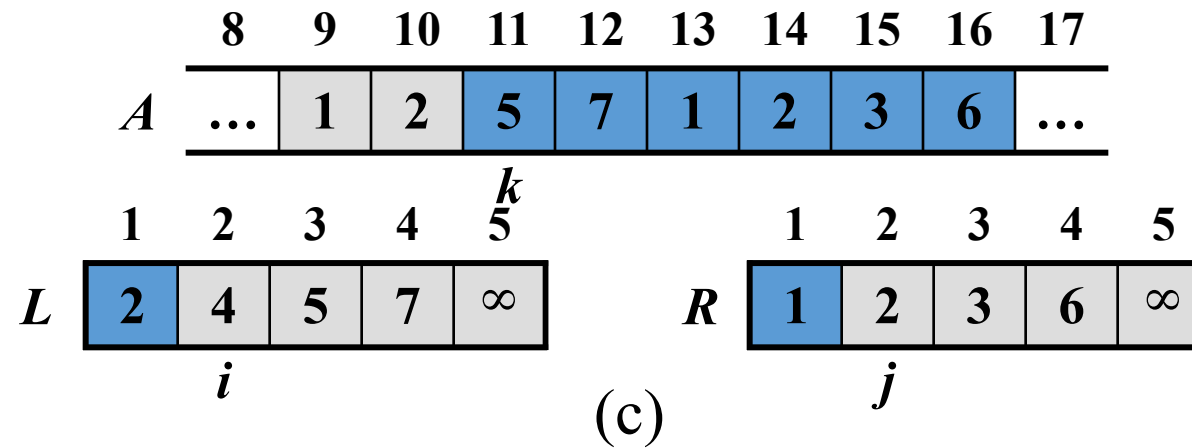
The operation of lines 10-17



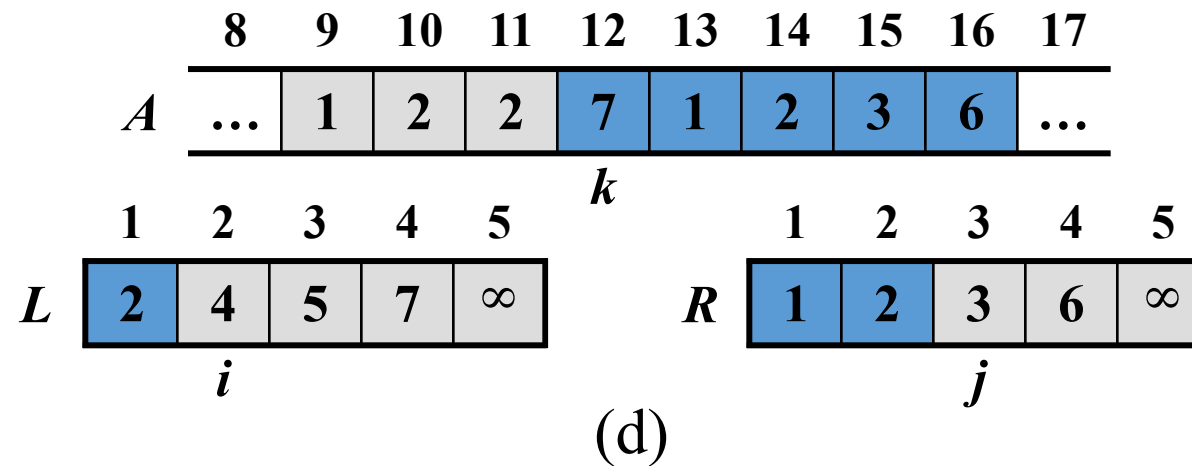
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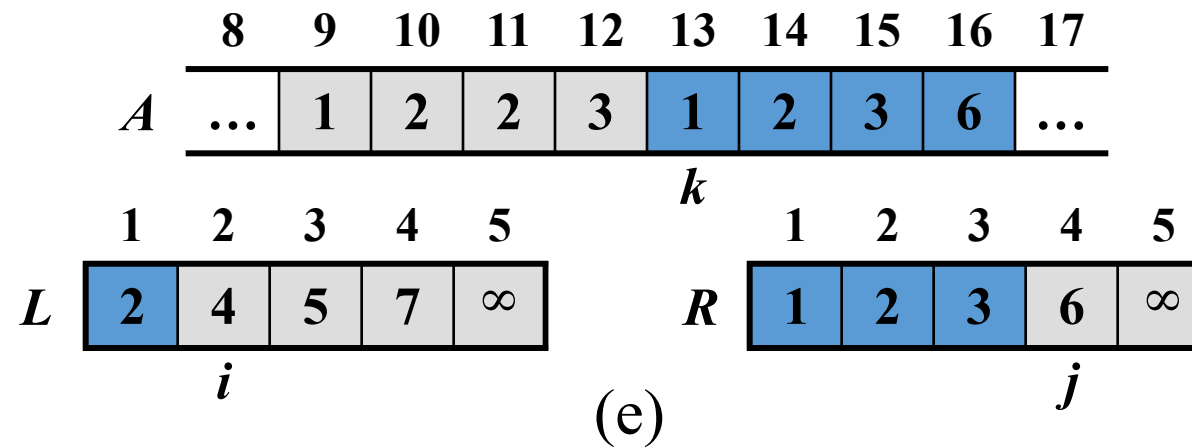
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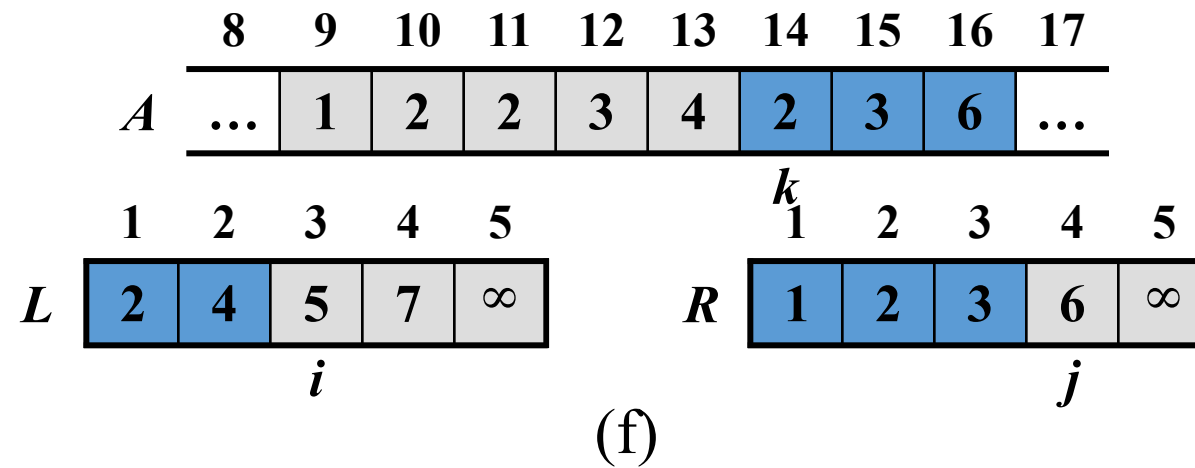
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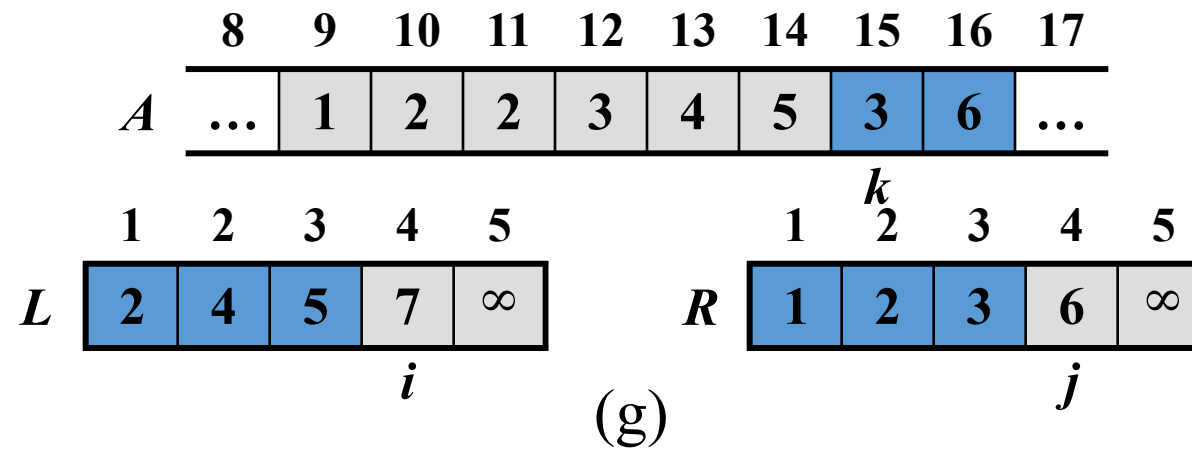
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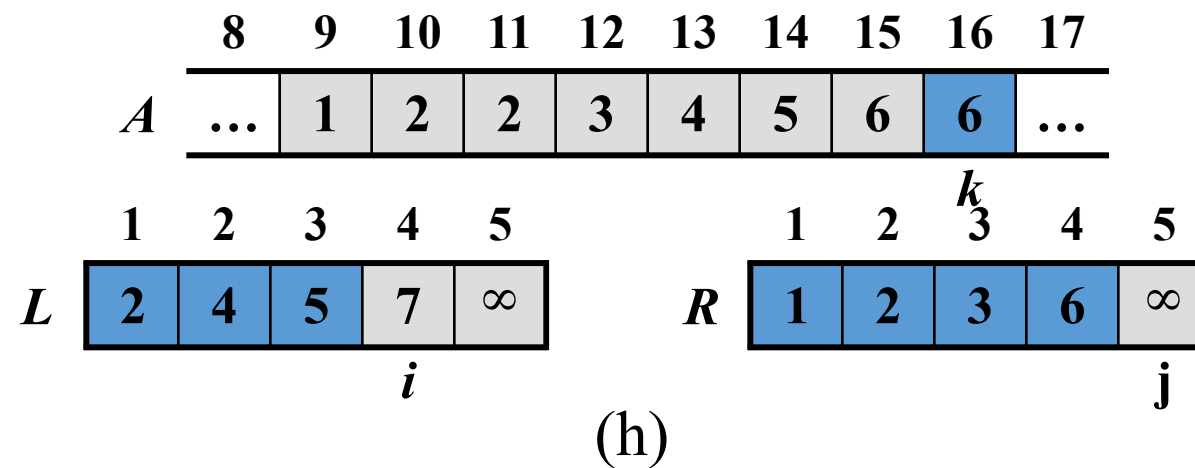
The operation of lines 10-17



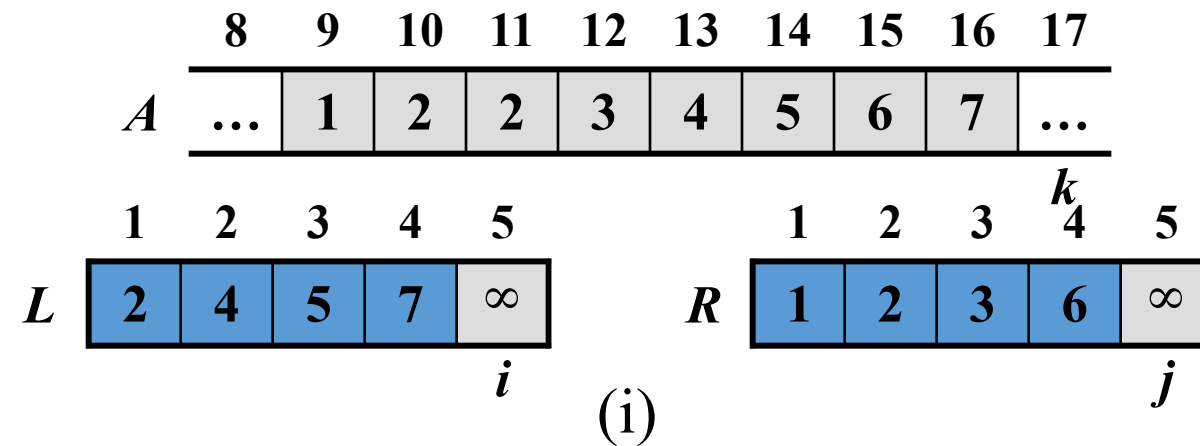
The operation of lines 10-17



The operation of lines 10-17



The operation of lines 10-17



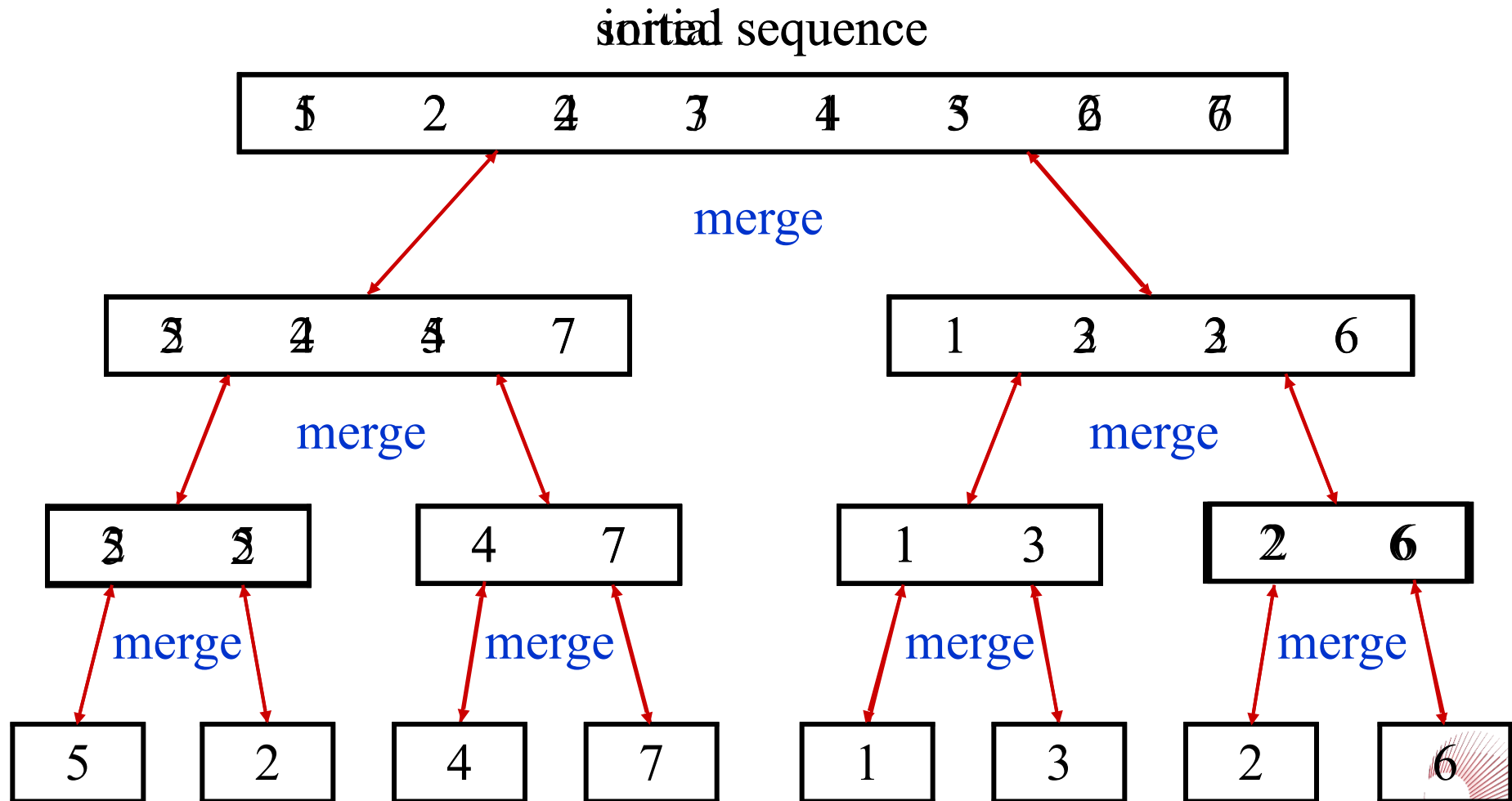
Pseudocode

MERGE-SORT(A, p, 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)



The operation of Merge sort



Analysis of Merge sort

- Analyzing divide-and-conquer algorithms

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

- See Chapter 4.

- Analysis of merge sort

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

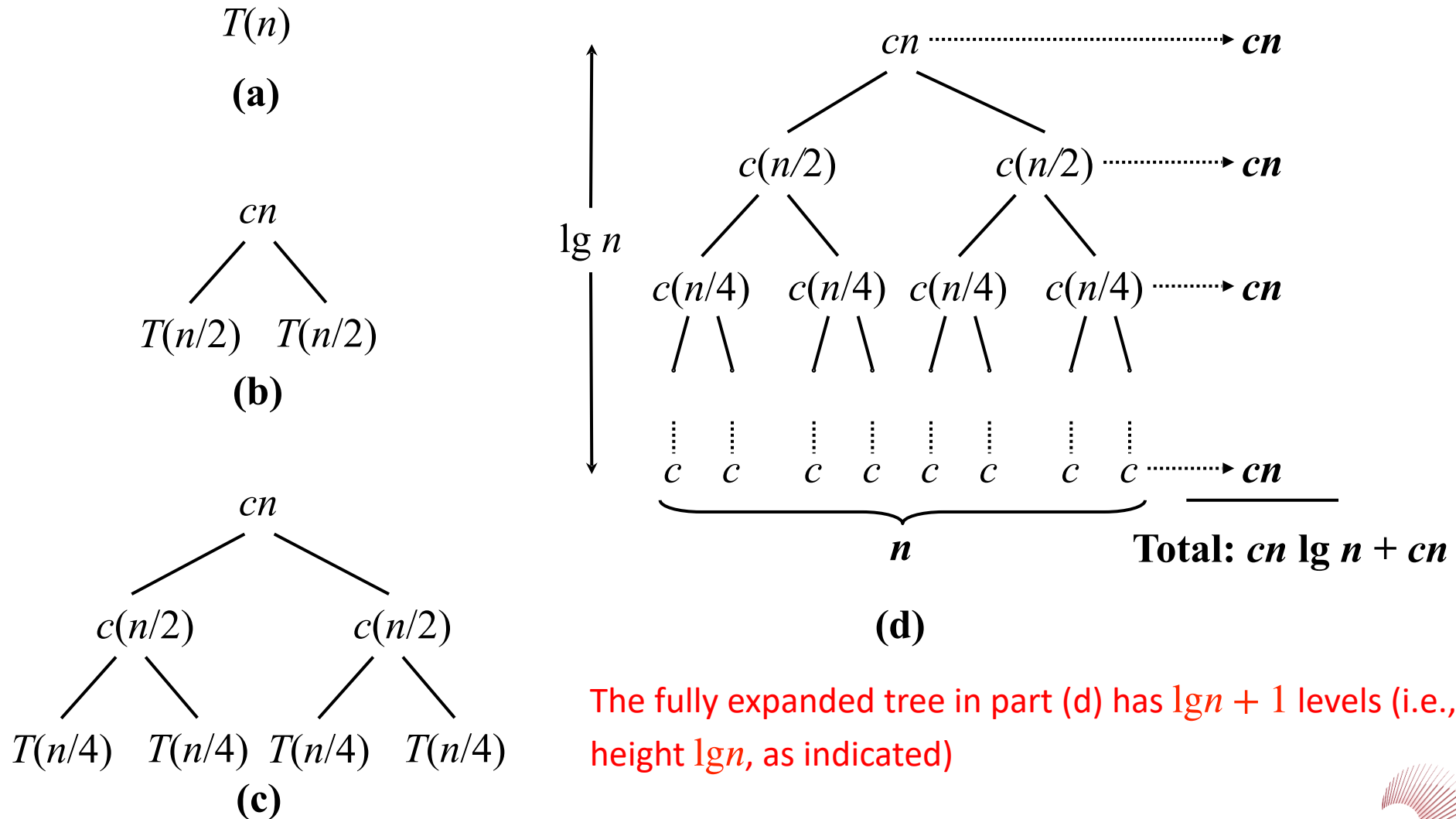
- $$T(n) = \Theta(n \lg n)$$

Analysis of Merge sort

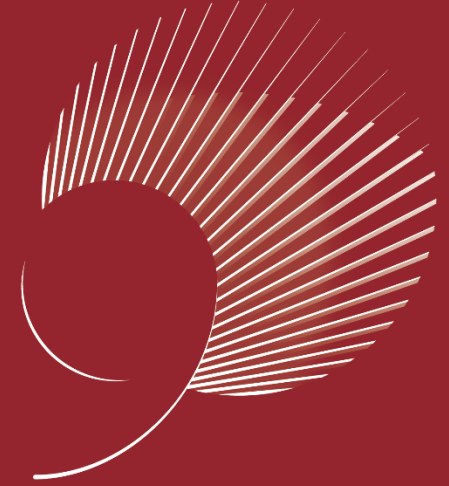
- $$T(n) = \begin{cases} c & \text{if } n = 1 \\ 2T\left(\frac{n}{2}\right) + cn & \text{if } n > 1 \end{cases}$$

where the constant c represents the time require to solve problem of size 1 as well as the time per array element of the divide and combine steps.

The construction of a recursion tree



- Outperforms insertion sort!



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