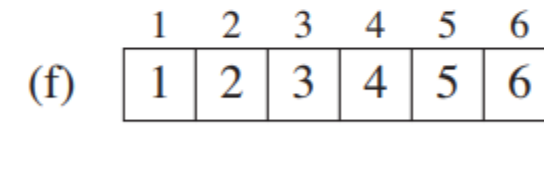
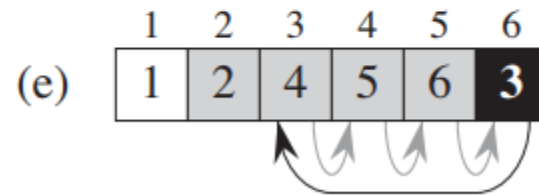
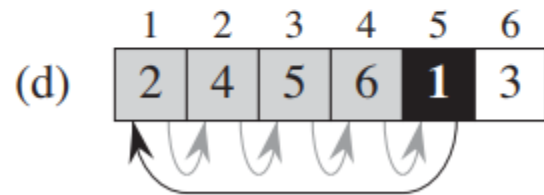
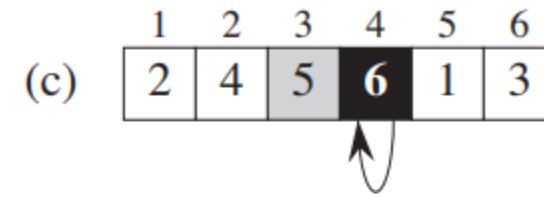
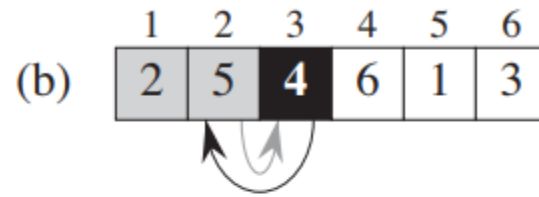
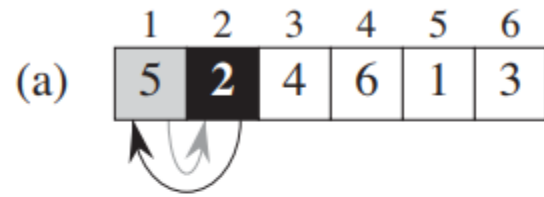


Recap: Insertion Sort



We're going to analyze its time-efficiency

Insertion Sort.

input size. $n=6$.

start with 1

output;
ascending,
ordered
list

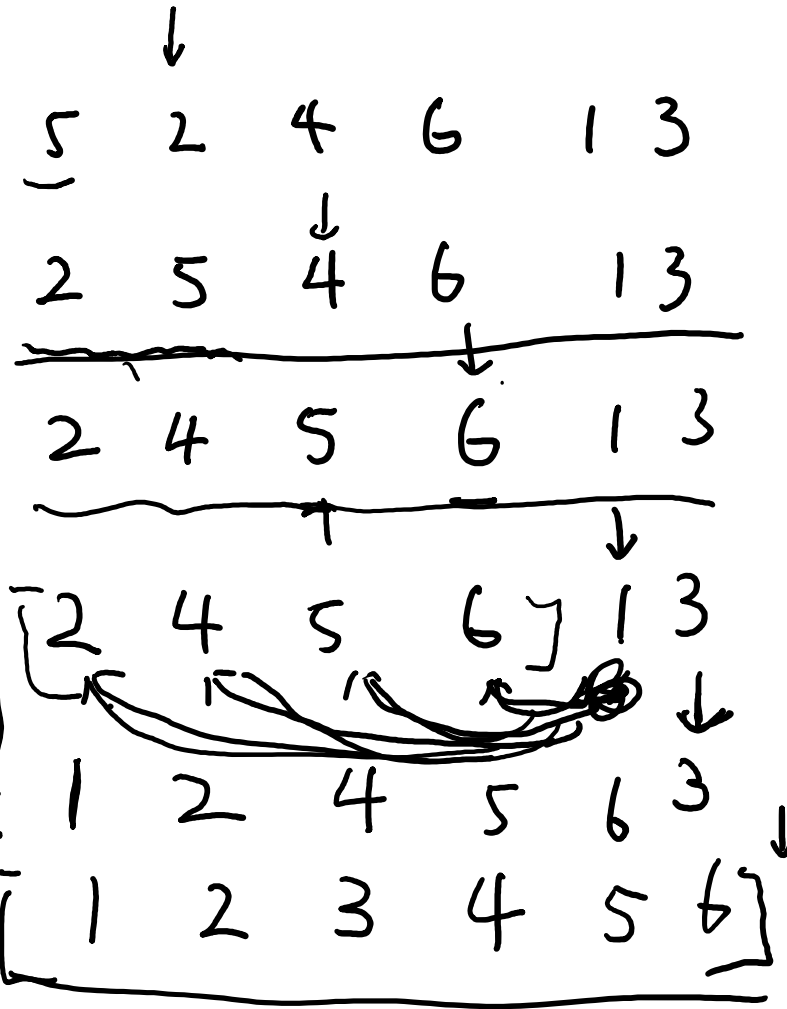
1	2	3	4	5	6
5	2	4	6	1	3

outer loop: $\underline{2} \dots \underline{n}$.

inner loop:

best case: 1 comparing.

worst case: \underline{j} -th element.
4 comparing



iter 1.

iter 2.

iter 3

iter 4

iter 5

Time-consuming on input size n

Abstract 1: the size of input n , almost all algorithm run longer on larger input, so the time function is $T(n)$

Abstract 2: the time spending on each operation vary when using different machine and compiler, we use symbol c_i to denote the time cost. It would make us focus on counting the number of times each operator is executed

INSERTION-SORT(A)

```

1  for  $j = 2$  to  $A.length$ 
2       $key = A[j]$ 
3      // Insert  $A[j]$  into the sorted sequence  $A[1..j-1]$ .
4       $i = j - 1$ 
5      while  $i > 0$  and  $A[i] > key$ 
6           $A[i+1] = A[i]$ 
7           $i = i - 1$ 
8       $A[i+1] = key$ 
    
```

worst
case
(see next page
for detail)

cost

c_1

c_2

c_4

c_5

c_6

c_7

c_8

times

n

$n-1$

$n-1$

$2+3+\dots+n$

$1+2+\dots+n-1$

$1+2+\dots+n-1$

$n-1$

$(2, 3, \dots, n+1)$

Line 5

worst case:

$j=2$	$i=1, 0$	2	} sum
$j=3$	$i=2, 1, 0$	3	
\vdots			
$j=n$	$i=n-1, \dots, 0$	n	} $2+3+\dots+n$ $= \frac{n(n+1)}{2} - 1$

Line 6

worst case

$j=2$	$i=1$	1	} sum
$j=3$	$i=2, 1$	2	
\vdots			
$j=n$	$i=n-1 \dots 1$	$n-1$	} $1+2+\dots+(n-1)$ $= \frac{(n-1) \cdot n}{2}$

Polynomial

$$T(n) = C_1 \cdot n + C_2(n-1) + C_4(n-1) + C_8(n-1) + \\ C_5 \cdot \left(\frac{n(n+1)}{2} - 1 \right) + C_6 \cdot \frac{n(n-1)}{2} + C_7 \cdot \frac{n(n-1)}{2}$$

$$= C_1 \cdot n + (C_2 + C_4 + C_8)(n-1) + C_5 \frac{(n+2)(n-1)}{2} \\ + C_6 \cdot \frac{n(n-1)}{2} + C_7 \frac{n(n-1)}{2}$$

$$= C_1 n + C_2(n-1) + \frac{n+1}{2} [C_5(n+2) + C_6 n + C_7 n]$$

$$T(n) = c_1 n + c_2 (n-1) + \frac{n-1}{2} [c_3 (n+2) + c_6 \cdot n + c_7 \cdot n]$$

To count each operation is difficult and unnecessary, apply Approximation

Approximation 1:

count the number of basic operators

After Applying Approximation 1;

$$T(n) = \frac{(n-1)(n+2)}{2} \cdot c_5$$

basic operators: most time-consuming operation in the inner-most loop, e.g. > (in line 5) is the basic operator in insertion-sort algorithm

Approximation 2: ignore the terms in lower order

Apply it

$$T(n) = c_5 \cdot n^2$$

Order of growth: ignore the multiplicative coefficient.

$$T(n) \in O(n^2)$$

why terms in lower-order can be ignored

TABLE 2.1 Values (some approximate) of several functions important for analysis of algorithms

n	$\log_2 n$	n	$n \log_2 n$	n^2	n^3	2^n	$n!$
10	3.3	10^1	$3.3 \cdot 10^1$	10^2	10^3	10^3	$3.6 \cdot 10^6$
10^2	6.6	10^2	$6.6 \cdot 10^2$	10^4	10^6	$1.3 \cdot 10^{30}$	$9.3 \cdot 10^{157}$
10^3	10	10^3	$1.0 \cdot 10^4$	10^6	10^9		
10^4	13	10^4	$1.3 \cdot 10^5$	10^8	10^{12}		
10^5	17	10^5	$1.7 \cdot 10^6$	10^{10}	10^{15}		
10^6	20	10^6	$2.0 \cdot 10^7$	10^{12}	10^{18}		

large input size

$$T_1(n^2) : \begin{cases} T(10000) = 100000000 \\ T(100000) = 100010000 \end{cases} \Rightarrow \begin{matrix} 10000 \\ \text{can be} \\ \text{ignored} \end{matrix}$$

$$T_2(n^2 + n) :$$

Summary:

Analysing the time-efficiency of an algorithm == find its order of growth.

Path of finding:

Count the number of times basic operation is executed, write down its function $f(n)$ in polynomial, then return its order of growth by keeping the term with highest order and removing the multiplicative coefficient.