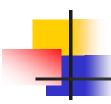


## Introduction

## 學習目標

- Why to study algorithm?
- How to analysis of algorithms?
- How to evaluate the goodness of algorithms?
- 本書著名的中演算法研究問題介紹。



## Why to study algorithms?

- It is commonly believed that in order to obtain high speed computation, it suffices to have a very high speed computer. This is, however, not entirely true.
- A good algorithm implemented on a slow computer may perform much better than a bad algorithm implemented on a fast computer.

## **Sorting problem:**

 To sort a set of elements into increasing or decreasing order.

↓sort

1, 5, 7, 9, 10, 11, 14

- Sorting Algorithm:
  - Insertion sort
  - Quick sort
  - Etc.

## **Insertion Sort**

#### Sorted Sequence

11,
7, 11
7, 11, 14
1, 7, 11. 14
1, 5, 7, 9, 11, 14
1, 5, 7, 9, 10, 11, 14

#### Unsorted sequence

## QuickSort

- Quicksort would use the first data element, say x, to divide all data elements into three subsets:
  - those smaller than x,



those larger than x, and



- those equal to x.
- Divide approach & recursive algorithm

$$(5, 1, 8, 7, 3) \Rightarrow (1, 3) (5) (7, 8)$$
  
 $(17, 14, 26, 21) \Rightarrow (14) (17) (26, 21)$ 

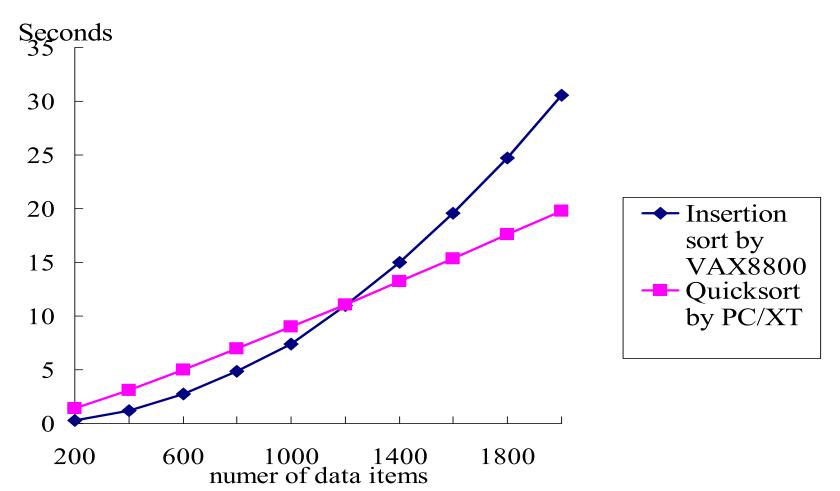
# Quicksort Example

Input: 10, 5, 1, 17, 14, 8, 7, 26, 21, 3

Steps: 10, 5, 1, 17, 14, 8, 7, 26, 21, 3

## Comparison of algorithms

Comparison of two algorithms implemented on two computers (average of ten times)



## Analysis of algorithms

- Measure the goodness of algorithms
  - efficiency
  - asymptotic notations: e.g. O(n²)
  - worst case
  - average case
  - best case
  - amortized analysis (均攤分析法)
- Measure the difficulty of <u>problems</u>
  - NP-complete ( $\ge$ O(2<sup>n</sup>)) or polynomial solvable
  - Undecidable
  - lower bound
- Is the algorithm optimal?

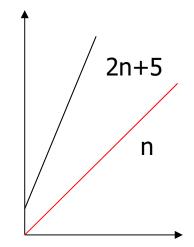
## **Asymptotic notations**

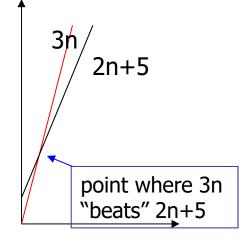
- **Def:** f(n) = O(g(n)) "at most" "upper bound"
- f(n) is less than or equal to g(n) up to a constant factor for large values of n

$$\exists c, n_0 \rightarrow |f(n)| \leq c|g(n)| \forall n \geq n_0$$

e.g. 
$$f(n) = 3n^2 + 2$$
  
 $g(n) = n^2$   
 $\Rightarrow n_0 = 2, c = 4$ 

$$\therefore f(n) = O(n^2)$$





• e.g. 
$$f(n) = n^3 + n = O(n^3)$$

• e. g. 
$$f(n) = 3n^2 + 2 = O(n^3)$$
 or  $O(n^{100})$ 

2n+5 is **O**( n )

## **Asymptotic notations**

- $\underline{\mathbf{Def}}$ :  $\mathbf{f(n)} = \Omega(\mathbf{g(n)})$  "at least", "lower bound"  $\exists$  c, and  $\mathbf{n_0}$ ,  $\ni$   $|\mathbf{f(n)}| \ge \mathbf{c|g(n)}| \forall$   $\mathbf{n} \ge \mathbf{n_0}$ e. g.  $\mathbf{f(n)} = 3\mathbf{n^2} + 2 = \Omega(\mathbf{n^2})$  or  $\Omega$  (n)
- $\underline{\mathbf{Def}}$ :  $\mathbf{f(n)} = \Theta(\mathbf{g(n)})$   $\exists c_1, c_2, \text{ and } n_0, \ni c_1|\mathbf{g(n)}| \le |\mathbf{f(n)}| \le c_2|\mathbf{g(n)}| \ \forall \ n \ge n_0$ e. g.  $\mathbf{f(n)} = 3\mathbf{n}^2 + 2 = \Theta(\mathbf{n}^2)$
- Def : f(n) ~ o(g(n))

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}\to 1$$

e.g. 
$$f(n) = 3n^2 + n = o(3n^2)$$

### Problem size n and function

	10	$10^{2}$	$10^{3}$	104
log <sub>2</sub> n	3.3	6.6	10	13.3
n	10	102	$10^{3}$	104
nlog <sub>2</sub> n	$0.33 \times 10^2$	$0.7x10^3$	104	$1.3 \times 10^5$
$n^2$	102	104	$10^{6}$	108
2 <sup>n</sup>	1024	$1.3 \times 10^{30}$	>10100	>10100
n!	$3x10^{6}$	>10100	>10100	>10100

**Time Complexity Functions** 

## Common computing time functions

- Time complexity classes:
  - $O(1) < O(\log n) < O(n) < O(n \log n) < O(n^2) < O(n^3)$ <  $O(2^n) < O(n!) < O(n^n)$
  - Exponential algorithm: O(2<sup>n</sup>)
  - polynomial algorithm: e.g. O(n²), O(nlogn), ...
- Algorithm  $A : O(n^3)$ , algorithm B : O(n)
  - Should Algorithm B run faster than A?
    NO!
  - It is true only when n is large enough!



### Introduction

- How do we measure the goodness of an algorithm?
- How do we measure the difficulty of a problem?
- How do we know that an algorithm is optimal for a problem?
- How can we know that there does not exist any other better algorithm to solve the same problem?

# The goodness of an algorithm

- Time complexity (more important)
- Space complexity (memory size)
- For a parallel algorithm :
  - time-processor product
  - $O(\log n)$  time, O(n) processors  $\rightarrow O(n \log n)$
- For a VLSI circuit :
  - area-time (AT, AT²), A is the area of the VLSI



An undecidable problem is a decision problem (output **True** or **False**) for which it is known to be impossible to construct a single algorithm that always leads to a correct yes-or-no answer.

#### Example:

- Halting problem: "Given a description of an arbitrary computer program, decide whether the program finishes running or continues to run forever."
- This is equivalent to the problem of deciding, given a program and an input, whether the program will eventually halt when run with that input, or will run forever.
- Alan Turing proved in 1936.

## 0/1 Knapsack problem 0/1背包問題

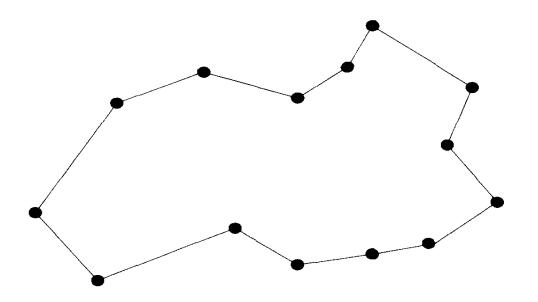
	P <sub>1</sub>	P <sub>2</sub>	P <sub>3</sub>	P <sub>4</sub>	P <sub>5</sub>	\$4 12 kg
Value	4	2	10	1	2	15 Kg \$2 2 kg
Weight	12	1	4	1	2	\$2 1k9 \$1 1k9
					!	\$10 4kg

- M (total weight limitation)=15;
- 0/1 constraint;
- best solution (maximal sum of value)?
- This problem is <u>NP-complete</u>.
- As the number of items becomes very large, it is very hard to find an optimal solution.

## Traveling salesperson problem (TSP)

- Given: A set of n planar points

  Find: A <u>closed tour</u> which includes all points exactly once such that its total length is minimized.
- **■** This problem is **NP-complete**.





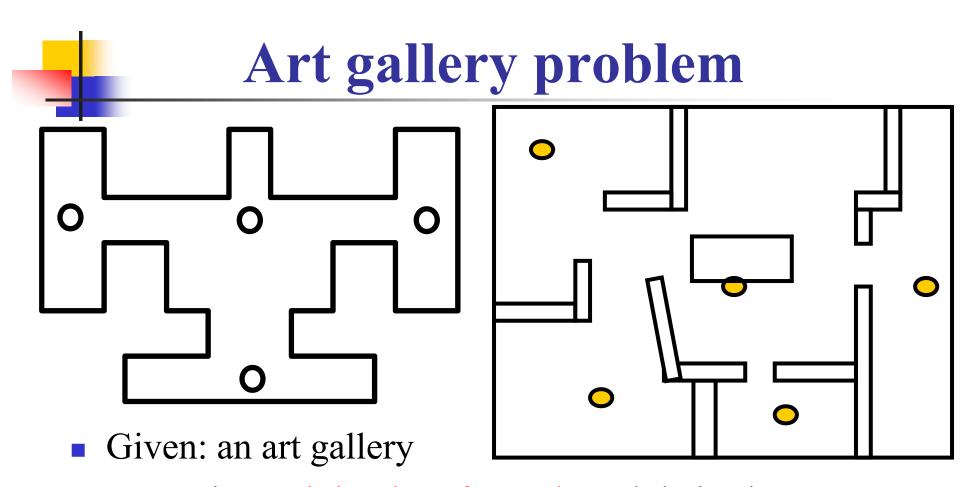
## Partition problem

• Given: A set of positive integers S Find:  $S_1$  and  $S_2$  such that  $S_1 \cap S_2 = \emptyset$ ,  $S_1 \cup S_2 = S$ ,

$$\sum_{i \in S_1} i = \sum_{i \in S_2} i$$

(partition into  $S_1$  and  $S_2$  such that the sum of  $S_1$  is equal to that of  $S_2$ )

- $\bullet$  e.g. S={1, 7, 10, 4, 6, 3, 8, 13}
  - $S_1 = \{1, 10, 4, 8, 3\}$
  - $S_2 = \{7, 6, 13\}$
- This problem is NP-complete.



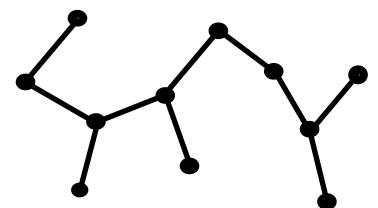
Determine: minimal # of guards and their placements such that the entire art gallery can be monitored.

NP-complete



## Minimum spanning tree

- graph: greedy method
- geometry(on a plane): divide-and-conquer
- # of possible spanning trees for n points: n<sup>n-2</sup>
   (Cayley's formula )
- $n=10 \rightarrow 10^8, n=100 \rightarrow 10^{196}$

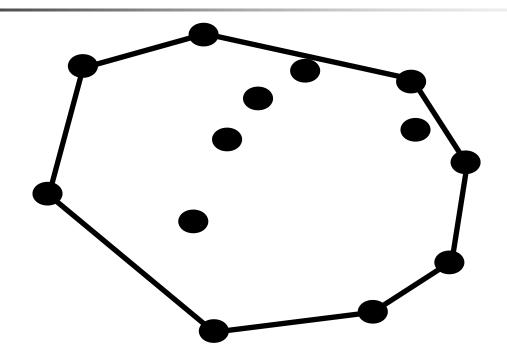


#### **Cayley's Formula**

https://www.youtube.com/watch?v=Ve447EOW8ww



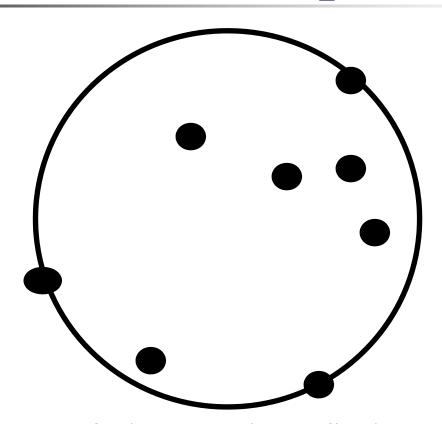
#### **Convex hull**



- Given a set of planar points, find a smallest convex polygon which contains all points.
- It is not obvious to find a convex hull by examining all possible solutions
- divide-and-conquer



## One-center problem



- Given a set of planar points, find a smallest circle which contains all points.
- Prune-and-search



## Question

## Question:

- Which problem is an NP-complete problem?
- (1) minimal spanning tree on 2-D plan
- (2) graph coloring problem for plane graph
- (3) Sorting problem for a set of distinct integers
- (4) Partition problem.

Ans. 4

## Question:

- Which problem is an Undecidable problem?
- (1) Travelling Salesman Problem (TSP)
- (2) Art Gallery Problem
- (3) Halting problem
- (4) Partition problem.

Ans. 3