



Chapter 1

Introduction



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.

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

↓sort

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

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

Insertion Sort

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

Insertion sort works on the above sequence of data elements as follows:

Sorted Sequence

11,

7, 11

7, 11, 14

1, 7, 11, 14

✓ 1, 5, 7, 11, 14

1, 5, 7, 9, 11, 14

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

Unsorted Sequence

7, 14, 1, 5, 9, 10

14, 1, 5, 9, 10

1, 5, 9, 10

5, 9, 10

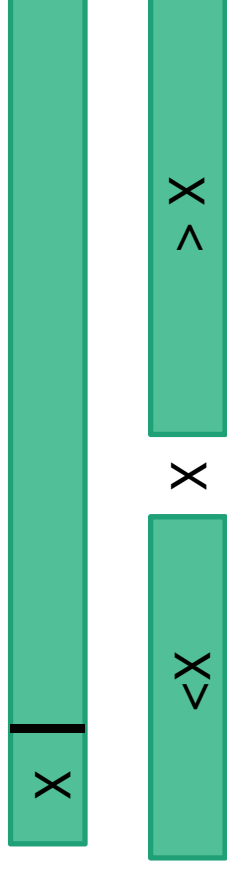
9, 10

10

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) (10) (17, 14, 26, 21).

We now have to sort two sequences:

(5, 1, 8, 7, 3)  (1, 3) (5) (7, 8)

and (17, 14, 26, 21).  (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

(5, 1, 8, 7, 3) (10) (17, 14, 26, 21).

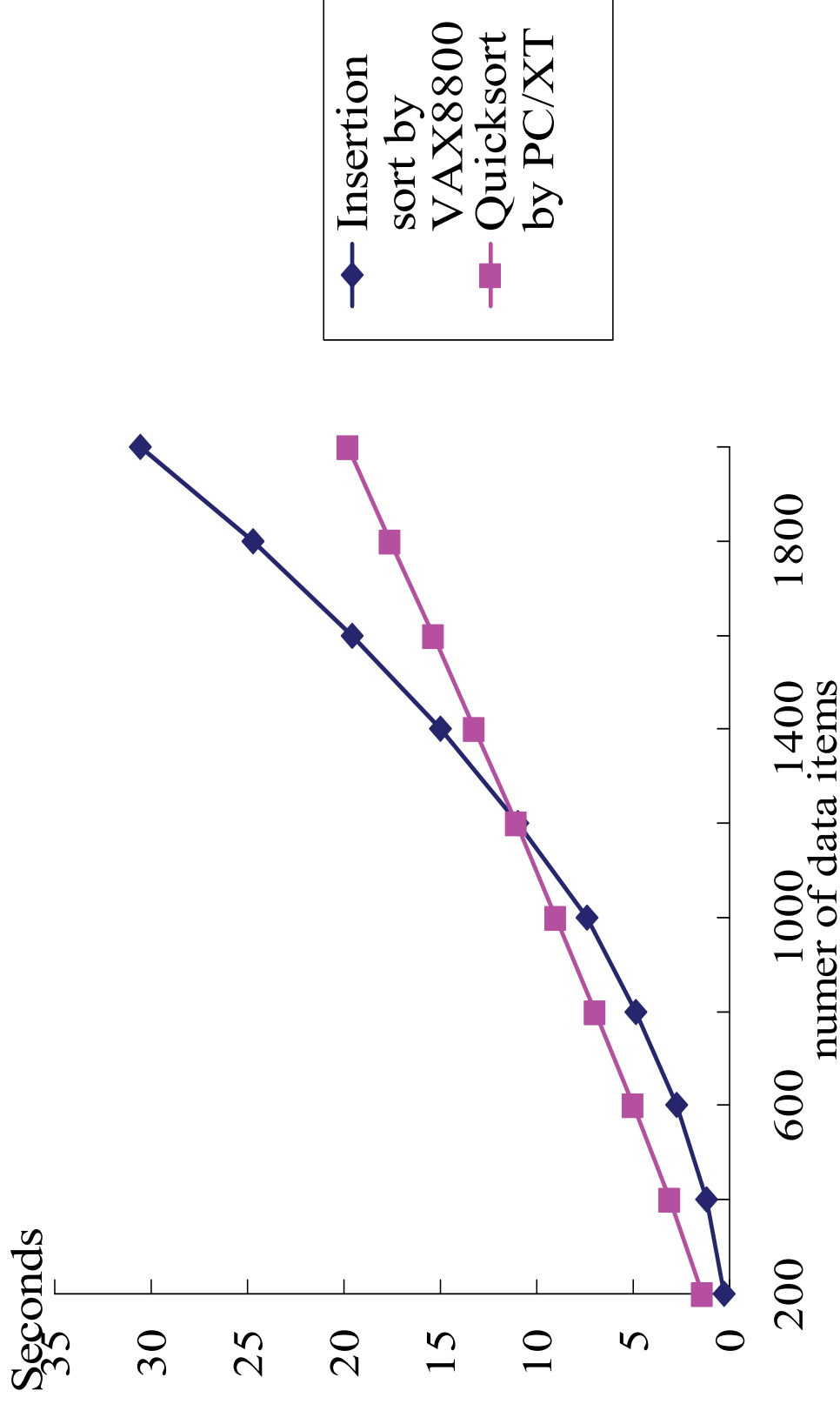
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(5, 1, 8, 7, 3)  (1, 3) (5) (7, 8)

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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. $O(n^2)$
 - worst case
 - average case
 - best case
 - **amortized analysis** (均攤分析法)
- Measure the difficulty of problems
 - NP-complete ($\geq O(2^n)$) or polynomial solvable
 - Undecidable
 - lower bound
- Is the algorithm optimal?



Asymptotic notations

■ Def: $f(n) = O(g(n))$

"at most"

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

■ e.g. $f(n) = 3n^2 + 2$

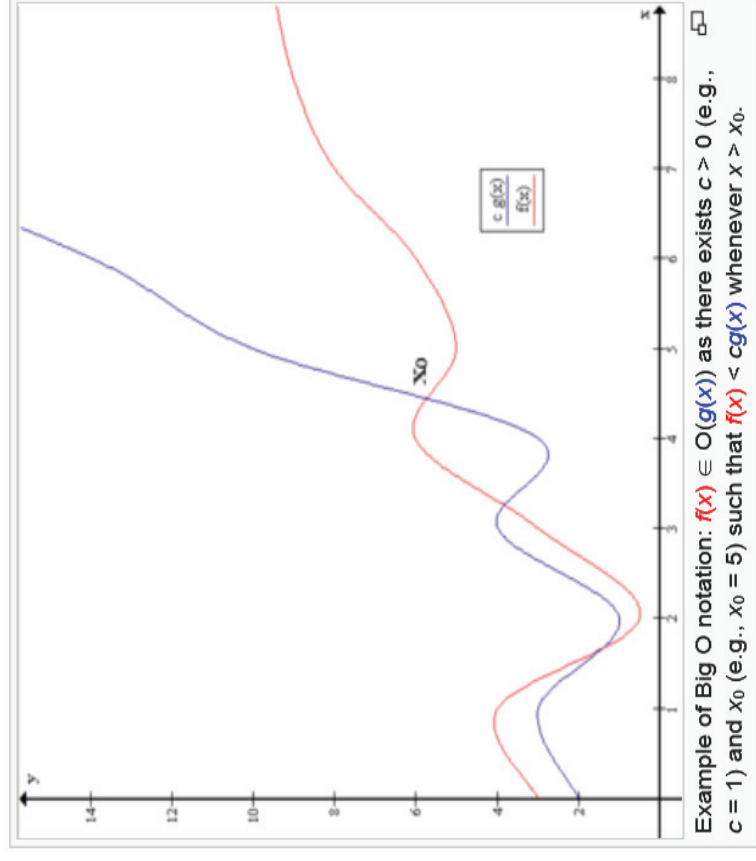
$$g(n) = n^2$$

$$\Rightarrow n_0=2, c=4$$

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

■ e.g. $f(n) = n^3 + n = O(n^3)$

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





Asymptotic notations

- Def : $f(n) = \Omega(g(n))$ “at least”, “lower bound”
 $\exists c, \text{ and } n_0, \ni |f(n)| \geq c|g(n)| \quad \forall n \geq n_0$
e. g. $f(n) = 3n^2 + 2 = \Omega(n^2)$ or $\Omega(n)$
- Def : $f(n) = \Theta(g(n))$
 $\exists c_1, c_2, \text{ and } n_0, \ni c_1|g(n)| \leq |f(n)| \leq c_2|g(n)| \quad \forall n \geq n_0$
e. g. $f(n) = 3n^2 + 2 = \Theta(n^2)$
- Def : $f(n) \sim o(g(n))$

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

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



Problem size n and function

	10	10^2	10^3	10^4
$\log_2 n$	3.3	6.6	10	13.3
n	10	10^2	10^3	10^4
$n \log_2 n$	0.33×10^2	0.7×10^3	10^4	1.3×10^5
n^2	10^2	10^4	10^6	10^8
2^n	1024	1.3×10^{30}	$> 10^{100}$	$> 10^{100}$
$n!$	3×10^6	$> 10^{100}$	$> 10^{100}$	$> 10^{100}$

Time Complexity Functions

Common computing time functions

- Time complexity classes:
 - $O(1) < O(\log n) < O(n) < O(n \log n) < O(n^2) < O(n^3) < O(2^n) < O(n!)$
 - Exponential algorithm: $O(2^n)$
 - polynomial algorithm: e.g. $O(n^2)$, $O(n \log n)$, ...
 - 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 , AT^2), A is the area of the VLSI



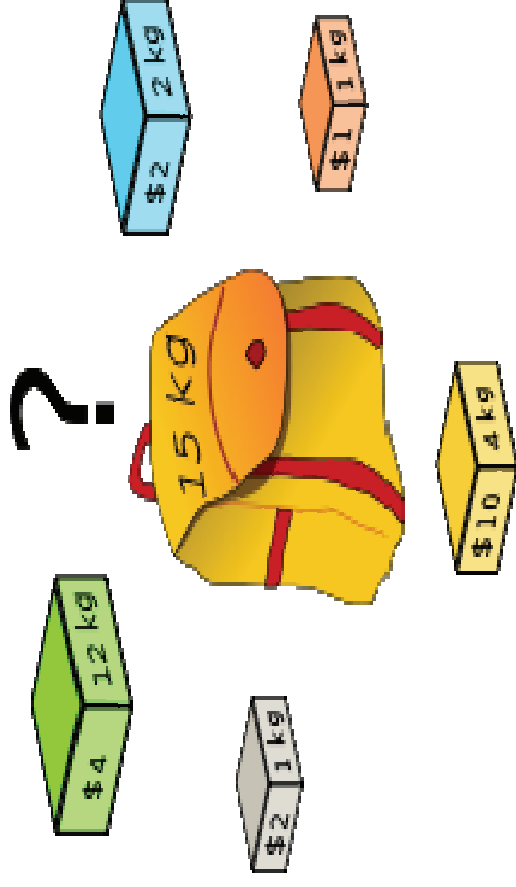
Undecidable problems

- An undecidable problem is a **decision problem** (output T or F) 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.

http://en.wikipedia.org/wiki/Halting_problem

0/1 Knapsack problem 背包問題

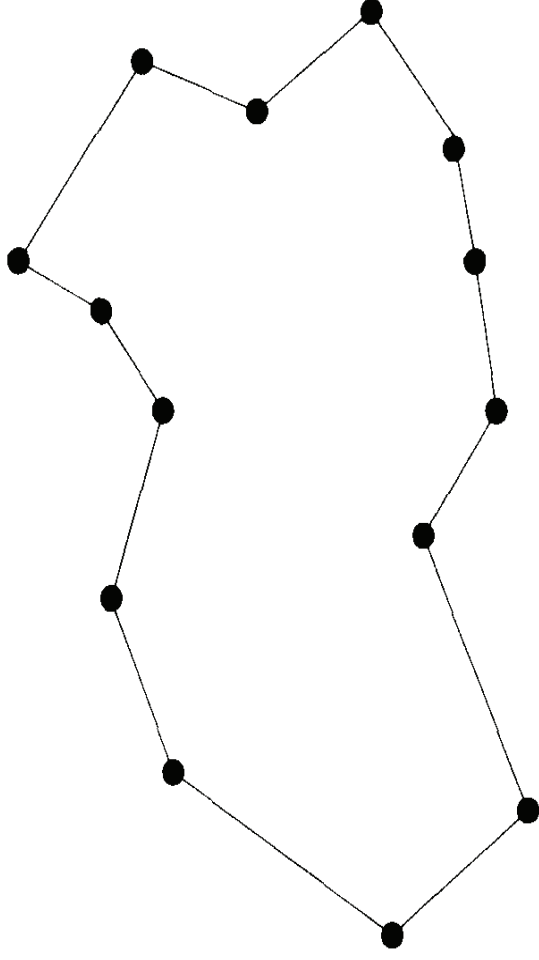
	P ₁	P ₂	P ₃	P ₄	P ₅
Value	4	2	10	1	2
Weight	12	1	4	1	2



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

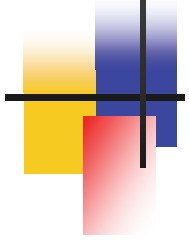
Traveling salesperson problem (TSP)

- Given: A set of n planar points
- Find: A closed tour which includes all points exactly once such that its total length is minimized.
- This problem is **NP-complete**.



<http://www.tsp.gatech.edu/>

<http://www.math.uwaterloo.ca/tsp/index.html>



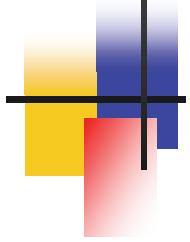
Partition problem

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

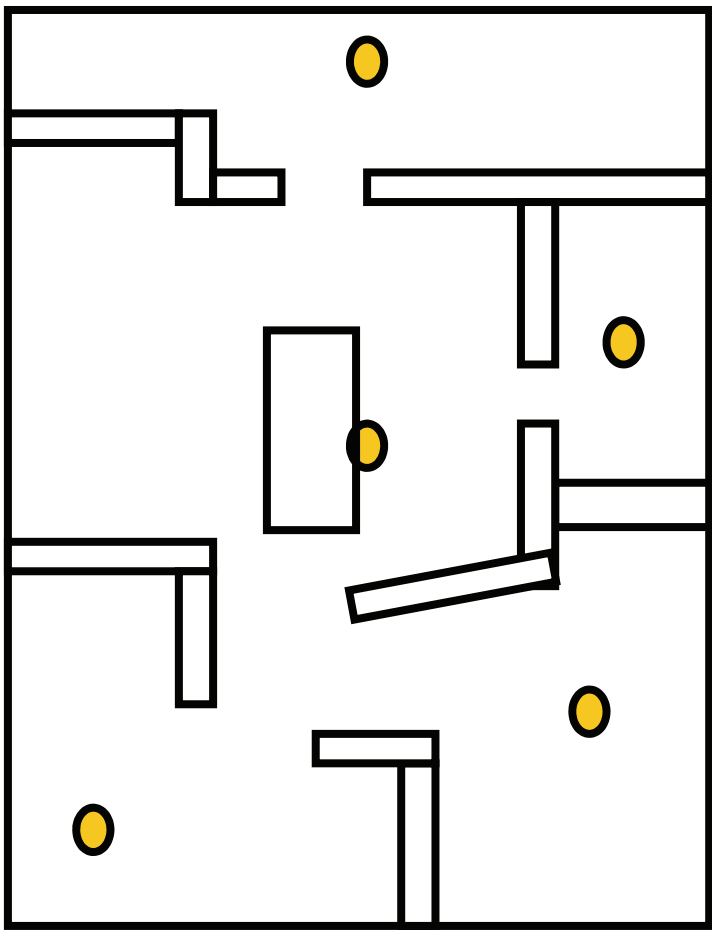
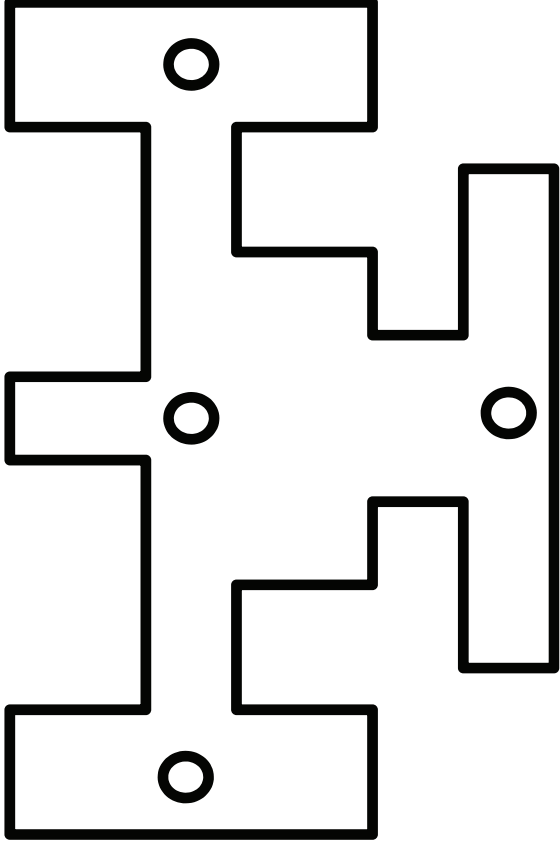
$$\sum_{i \in S_1} i = \sum_{i \in S_2} i$$

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

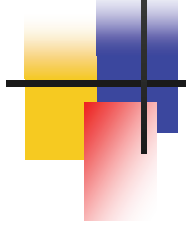
- e.g. $S = \{1, 7, 10, 4, 6, 3, 8, 13\}$
 - $S_1 = \{1, 10, 4, 8, 3\}$
 - $S_2 = \{7, 6, 13\}$
- This problem is NP-complete.



Art gallery problem

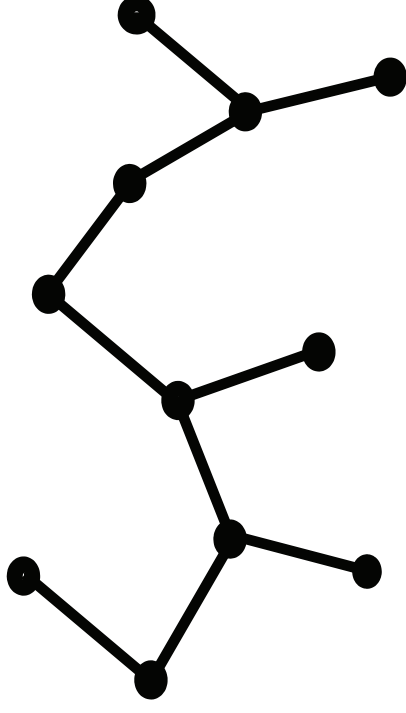


- Given: an art gallery
- Determine: **min # of guards** and their placements
such that the entire art gallery can be monitored.
- **NP-complete**



Minimum spanning tree

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

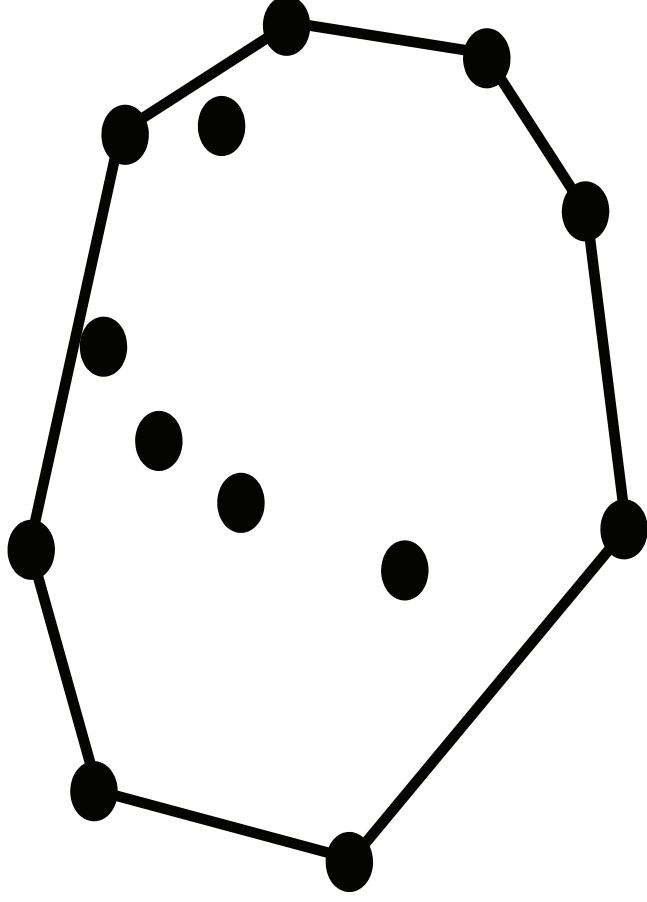


Cayley's Formula

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

<http://www.csie.ntu.edu.tw/~kmchao/tree07spr/counting.pdf>

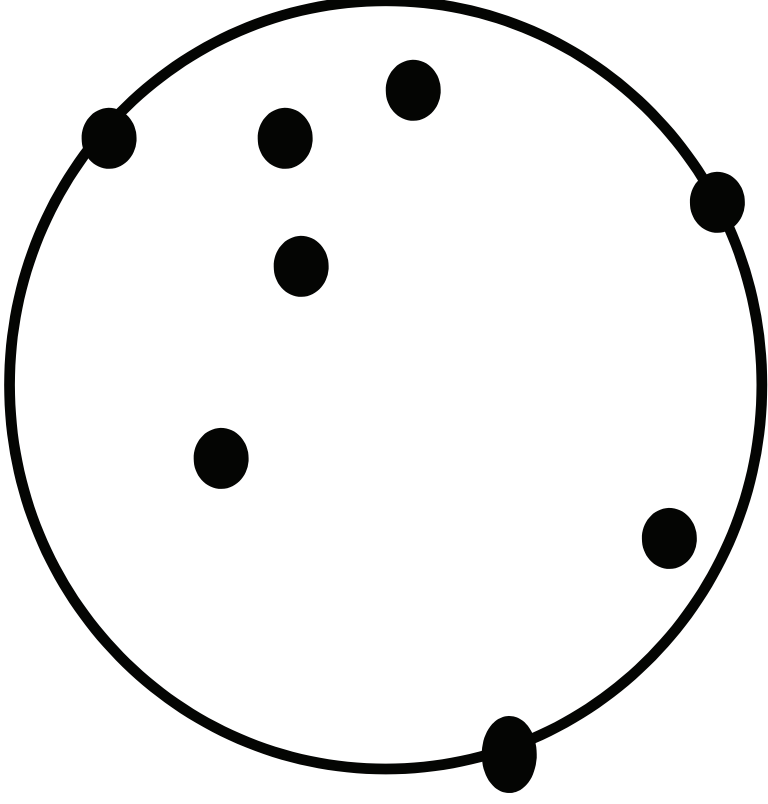
Convex hull



- 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