

# Computational complexity theory

## Introduction to computational complexity theory

- Complexity (computability) theory deals with two aspects:
  - Algorithm's complexity.
  - Problem's complexity.
  
- References
  - S. Cook, « The complexity of Theorem Proving Procedures », 1971.
  - Garey and Johnson, « Computers and Intractability, A guide to the theory of NP-completeness », 1979.
  - J. Carlier et Ph. Chrétienne « Problèmes d'ordonnements : algorithmes et complexité », 1988.

## Basic Notions

- Some problem is a “question” characterized by parameters and needs an answer.
  - Parameters description;
  - Properties that a solutions must satisfy;
  - An instance is obtained when the parameters are fixed to some values.
- An algorithm: a set of instructions describing how some task can be achieved or a problem can be solved.
- A program : the computational implementation of an algorithm.

## Algorithm's complexity (I)

- There may exists several algorithms for the same problem
- Raised questions:
  - Which one to choose ?
  - How they are compared ?
  - How measuring the efficiency ?
  - What are the most appropriate measures, running time, memory space ?

## Algorithm's complexity (II)

- Running time depends on:
  - The data of the problem,
  - Quality of program...,
  - Computer type,
  - Algorithm's efficiency,
  - etc.
- Proceed by analyzing the algorithm:
  - Search for some  $n$  characterizing the data.
  - Compute the running time in terms of  $n$ .
  - Evaluating the number of elementary operations, (*elementary operation = simple instruction of a programming language*).

## Algorithm's evaluation (I)

- Any algorithm is composed of two main stages: initialization and computing one
- The complexity parameter is the size data  $n$  (binary coding).

Definition:

Let be  $n > 0$  and  $T(n)$  the running time of an algorithm expressed in terms of the size data  $n$ ,  $T(n)$  is of  $O(f(n))$  iff  $\exists n_0$  and some constant  $c$  such that:

$$\forall n \geq n_0, \text{ we have } T(n) \leq c f(n).$$

*If  $f(n)$  is a polynomial then the algorithm is of polynomial complexity.*

- Study the complexity in the worst case.
- Study the complexity in average :  $tm(n)$  is the mean value of execution time when the data and the associated distribution law are given.

## Algorithm's evaluation (II)

### example

Given  $N$  numbers  $a_1, a_2, \dots, a_n$  in  $\{1, \dots, K\}$ .

The algorithm MIN finds the minimum value.

Algorithm MIN

```

Begin
  for i=1 to n read  $a_i$ ;
  B:= $a_1$ , j:=1;
  for i=2 to n do :
    if B=1 then j:=n;
    elseif  $a_i < B$  then
      begin
        j:=i;
        B:= $a_i$ ;
      end;
  write : the min value is  $a_j$ ;
End.
```

## Algorithm's evaluation (III)

- Study the complexity of algorithm MIN.

Proposition 1. The complexity of Algorithm MIN is in  $O(n)$ .

Proposition 2. if numbers  $a_i$  are independent random values and if any  $a_i$  has some probability  $1/K$  to be equal to 1, the complexity in average of algorithm MIN, apart the reading data phase, is in  $O(1)$ .

## Importance of polynomial algorithms (I)

f(n)	n=10	n=20	n=30	n=40	n=50
n	0.00001 sec	0.00002sec	0.00003 sec	0.00004 sec	0.00005 sec
n <sup>2</sup>	0.0001 sec	0.0004 sec	0.0009 sec	0.0016 sec	0.0025 sec
n <sup>3</sup>	0.001 sec	0.008 sec	0.027 sec	0.064 sec	0.125 sec
2 <sup>n</sup>	0.001 sec	1 sec	17.9 min	12.7 days	35.7 years
3 <sup>n</sup>	0.059 sec	58 min	6.5 years	3.855 centuries	2*10 <sup>8</sup> centuries

An elementary operation is run in one microsecond.

## Importance of polynomial algorithms (II)

f(n)	Todays computers	100 times faster	1000 times faster
n	N1	100 N1	1000 N1
n <sup>2</sup>	N2	10 N2	31.6 N2
n <sup>3</sup>	N3	4.64 N3	10N3
2 <sup>n</sup>	N4	N4 +6.64	N4 + 9.97
3 <sup>n</sup>	N5	N5 + 4.19	N5 + 6.29

Problem's sizes solved in one hour run time

## Computational complexity theory

- The decision problem is some mathematical question requiring some answer yes or no.
- Computational Complexity Theory is concerned with the question: for which decision problems do efficient algorithms exist ?

## Decision problems: SAT

- Satisfiability is the problem of determining if the variables of a given boolean formula can be assigned in such a way as to make the formula evaluate to TRUE.
- In complexity theory, the Boolean satisfiability problem (SAT) is a decision problem, whose instance is a Boolean expression written using only AND, OR, NOT, variables, and parentheses.
- The question is: given the expression, is there some assignment of *TRUE* and *FALSE* values to the variables that will make the entire expression true?
- A formula of propositional logic is said to be *satisfiable* if logical values can be assigned to its variables in a way that makes the formula true.

## Travelling salesman problem

- Given a weighted graph  $G=(X,E,v)$

$X$  = Vertices (= Cities)

$E$  = Edges (pair of cities)

$v$  = Distances between cities

- *Question: is-there a tour that visits all cities exactly once and its length(weight) is less than a given number  $B$  ?*

## Partition problems

- Partition problem
  - Data: given a set  $A=\{a_i \mid i \in I\}$  of  $n$  integer numbers.
  - Question : is-there some partition of  $A$  in two subsets  $A_1$  and  $A_2$  of equal weight ?
- Tripartition problem
  - Data: given a set  $A=\{a_i \mid i \in I\}$  of  $n=3q$  integer numbers such that  $\sum_{i \in I} a_i = qB$  and  $B/4 < a_i < B/2$  and  $B$  some positive integer.
  - Question : Is-there some partition of  $A$  in  $q$  subsets of cardinal 3 and weight  $B$ ?

## Some equivalent problems

- Vertex covering.
  - Data : a graph  $G=(V,E)$  with  $V$  a set of vertex,  $E$  a set of edges and some positive integer  $B \leq |V|$ .
  - Question : Is-there some subset  $V' \subseteq V$  such that  $|V'| \leq B$ , and for each edge  $(i,j) \in E$ ,  $i \in V'$  or  $j \in V'$  ?
- Independent set
  - Data : a graph  $G=(V,E)$  with the set of vertex  $V$  and  $E$  the set of edges and some positive integer  $B \leq |V|$ .
  - Question : Is-there some  $V' \subseteq V$   $|V'| \geq B$  such that for any  $(i,j) \in E$ ,  $i \notin V'$  or  $j \notin V'$  ?
- Maximal clique.
  - Data : a graph  $G=(V,E)$  where  $V$  is the vertex set and  $E$  the set of edges, and a positive integer  $B$ .
  - Question : Is-there some  $V' \subseteq V$  such that the corresponding sub-graph is complete and of size greater or equal to  $B$ ?

## Exact cover

- Exact cover by 3-sets (*X3C Problem*).
  - Data : Given a finite  $X$  with  $|X|=3q$  and a collection  $C$  of 3-elements subsets of  $X$ ,
  - Question : is-there some exact cover of  $X$ , that means is-there a sub-collection  $C' \subseteq C$  such that  $\forall x \in X$ , there is a only one  $c \in C'$  with  $x \in c$  ?



## Scheduling problems

- One machine problem
  - Data : a set  $I$  of  $n$  independent and indivisible tasks; for each task  $i \in I$  we have its duration  $p_i$ , availability date  $r_i$  and its deadline  $d_i$ .
  - Question : is-there some scheduling  $\phi$  of these  $n$  tasks in a single machine that satisfies the availability and deadline dates?
- Two processors problem
  - Data : a set  $I$  of  $n$  independent and indivisible tasks , durations  $p_i$  and a positive integer  $B$ .
  - Question : Is-there a scheduling  $\phi$  of these  $n$  tasks on two processors of duration less or equal to  $B$  ?
- NTRM
  - Data : a set  $I = \{1, 2, \dots, n\}$  of tasks of durations 1 and deadlines  $d_1, d_2, \dots, d_n$ , a partial order  $<$  on  $I$ , and a positive integer  $B$ .
  - Question : Is-there a scheduling  $\phi$  of these tasks on one machine satisfying the partial order and such that the number of delayed tasks is  $\leq B$ .

## Complexity theory: basic notions

- Why using the decision problems?
  - To introduce a simple formalism and making possible and easier the comparison between problems.
- *The complexity theory relies on Turing Machine...*
  - ....

## The class NP

Alternatively:

We distinguish the following complexity classes:

- Class P : some problem is in P if it can be solved in polynomial time to the size of data by a determinist algorithm.
- A problem is said to be in **NP** if and only if for a guessed solution there exists a polynomial time algorithm verifying the solution.
- *Class NP : it groups all decision problems such that an answer yes can be decided by a non-determinist algorithm in polynomial time to the size of data.*
  - **Polynomial time checking**

## NP-completeness

Paper of Stephen Cook, «The complexity of Theorem Proving Procedures», 1971.

- Defines the polynomial reduction
- Defines the decision problems and the class NP.
- Shows that the problem SAT is at least as difficult as all the others in NP → NP-complete

## Complexity theory polynomial reduction

- Polynomial reduction allows to compare NP problems in terms of computational complexity
- *Definition:  $P_1$  is polynomially reduced to  $P_2$  ( $P_1 \leq P_2$ ) if  $P_1$  is polynomial or there exists a polynomial algorithm  $A$  that builds for any  $d_1$  of  $P_1$  some data  $d_2$  of  $P_2$  such that  $d_1$  has answer YES iff  $d_2 = A(d_1)$  has answer YES.*
- Some problem is said NP-Complete if it is in NP, and any NP-Complete problem can be polynomially reduced to this problem.
- The polynomial reduction defines a pre-order relation on NP.
- Cook's Theorem : SAT is NP-Complete.

## NP-completeness (I)

- The general method:
  - 1) show first that  $\Pi \in \text{NP}$
  - 2) show that there exists  $P' \in \text{NP-complete}$  such that  $P' \leq \Pi$ .
- The following decision problems are NP-complete.
  - TSP,
  - Partition problems,
  - Exact cover
  - Clique

## NP-completeness (II)

- Three main techniques are used to show the NP-completeness of combinatorial problems.
  - Restriction
    - examples : minimal covering, knapsack, etc.
  - Local replacement
    - examples : 3SAT, X4C...
  - Component design
    - example : NTRM...

## NP-completeness (III)

Show the following results :

- Proposition 1. Two processors problem is NP-complete.
  - hint: *partition*  $\propto$  *two processors*.
- Proposition 2. One machine problem is NP-complete.
  - hint: *partition*  $\propto$  *one machine problem*.

## Exercises

- 1) Show that the knapsack problem is NP-complete.
  - Data: a finite set  $X$ , for all  $x_i \in X$ , there is a weight  $s(x_i)$ , a value  $v(x_i)$ . There are also a number  $B \in \mathbb{Z}^+$  and a number  $K \in \mathbb{Z}^+$ .
  - Question : Is-there any  $X' \subseteq X$  such that  $\sum_{x_i \in X'} s(x_i) \leq B$  and  $\sum_{x_i \in X'} v(x_i) \geq K$ .  
*hint: use partition.*
  
- 2) Show that the problem X4C, is NP-complete. (*hint: X3C*).
  
- 3) Show that the minimum sum of squares problem is NP-complete.
  - data: a finite set  $A$  of positive integer numbers ( $a \in \mathbb{Z}^+$ ) and two positive integers  $K$  and  $J$ .
  - Question : is-there any partition of  $A$  into  $K$  disjoint sets  $A_1, A_2, \dots, A_K$  such that  $\sum_{i=1}^K (\sum_{a \in A_i} a)^2 \leq J$ .  
*hint: bipartition, tripartition...*

## NP-completeness : conjecture

- Fundamental Conjecture:  $P \neq NP$ .

**You win 1000000 USD if You show that  $P=NP$   
or  $P \neq NP$ .**

## Pseudo-polynomial algorithms (I)

- Dynamic programming
  - Idea : breaking down the initial problem in a sequence of simpler problems, solving the  $n$ -th problem can be done by recurrence on this of  $(n-1)$ -th one.
  - WESS problem (weight of a subset)
    - Data : a finite set  $A$  composed of  $n$  elements  $a_i \in \mathbb{Z}^+$  and a positive integer  $K$ .
    - Question : is-there a subset of  $A$  of weight  $K$  ?

## Pseudo-polynomial algorithms (II)

### Algorithm WESS

```

Begin
  for k=0 to  $\sum_{a_i \in A} a_i$ 
    for j=1 to n do :
      WEIGHT(k,j):=false;
    end for;
  end for;
  set WEIGHT(0,0) := true;
  for j=1 to n
    for k=0 to  $\sum_{a_i \in A} a_i$  do :
      if ( $k \geq a_j$  and WEIGHT(k- $a_j$ , j-1)=true) or WEIGHT(k, j-1)=true
        then WEIGHT(k, j)=true;
      end for;
    end for;
  end.
  
```

**Proposition** : Algorithm WESS is of complexity  $O(n \sum_{a_i \in A} a_i)$  and assigns true to WEIGHT( $K$ ,  $n$ ) if there exists some subsets of  $n$  numbers  $a_1, a_2, \dots, a_n$  of weight  $K$ .

## Exercise: AN INVESTMENT PROBLEM

An example. 6 million is at our disposal, to invest in 3 regions. The following table shows the benefits given by the invested sums.

	Region I	Region II	Region III
1 million	0.2	0.1	0.4
2 million	0.5	0.2	0.5
3 million	0.9	0.8	0.6
4 million	1	1	0.7

1) Determine the optimal investment policy for the three regions using a "dynamic programming" method. The idea is to associate a graph with levels to the data. Level 0 contains only the vertex  $(0,0)$ , (because no money has been invested yet). Level 1 contains the vertices  $(1,0)$   $(1,1)$   $(1,2)$   $(1,3)$   $(1,4)$ , which correspond to the cumulated amounts invested in region 1. Level  $i$  contains the vertices  $(i, 0)$ ,  $(i, 1)$ ,  $(i, 2)$ ,  $(i, 3)$ ,  $(i, 4)$ ,  $(i, 5)$ ,  $(i, 6)$ , which correspond to the sums invested in the regions  $1 \dots i$  ( $i = 2, 3$ ). The arcs are placed between the levels  $i$  and  $i + 1$ , valued by the sums invested in the region  $i + 1$ . The last vertex is  $(3,6)$ . The goal is to seek a maximum value path in this graph.

2) The general case. More generally, we have  $B$  million to invest in  $n$  regions. We shall set  $f_i(y)$ , the optimal profit for a cumulative investment of a sum  $y$  in the regions  $1, 2, \dots, i$ . We have  $f_0(0) = 0$ . Determine a recurrence formula connecting  $f_i$  to  $f_{i-1}$  for  $i$  from 1 to  $n$ .

3) What is the complexity of the dynamic programming method in function of  $n$  and  $B$ , and complexity of enumerating all possible solutions.

## Strong NP-completeness (I)

- *Binary code* is the system of representing text or computer processor instructions by the use of the *Binary* number system's two-*binary* digits "0" and "1".
- Binary code of  $a \rightarrow \log_2(a)$  places
  - example : the size of  $2^{31}$  is  $\log_2(2^{31}) + 1 = 32$ , thus we need 32 bits to code it in machine;
- In computational complexity theory, a numeric algorithm runs in pseudo-polynomial time if its running time is polynomial in the *numeric value* (unary code) of the input.

## Strong NP-completeness (II)

- An NP-complete problem with known pseudo-polynomial time algorithms is called weakly NP-complete.
- An NP-complete problem is called strongly NP-complete if any pseudo-polynomial time algorithm solving it implies the existence of a polynomial time algorithm solving it.
  - Examples : TSP, tripartition, clique, sat.
  - What to say about bipartition, two-processeurs ? And about one machine ?

## Combinatorial optimization problems

- *Research problems*
  - A research problem consists in a set of instances  $D_M$  and of solutions  $S_M(D)$ . Solving it for instance  $D \in D_M$  comes to show that  $S_M(D)$  is empty or provide a solution from  $S_M(D)$ .
  - In an optimization problem,  $S_M(D)$  is the set of « optimal solutions ».
  - A decision problem can be seen as a special case of a research problem: the set of solution is empty if the answer is NON and otherwise  $S_M(D) = \{\text{YES}\}$ .



## Turing reduction

- Turing Reduction : a research problem P1 is polynomially reduced to a research problem P2 by Turing reduction ( $P1 \leq_T P2$ ), if it exists algorithm A1 using as a subroutine an algorithm A2 solving P2 such that A1 is polynomial-time if any call to A2 is taken to be of  $O(1)$ .
  - A1 is polynomial-time if A2 is polynomial-time.
  - Turing reduction defines a preorder relation on the research problems.
  - Turing reduction generalizes the polynomial reduction to research problems.
- Definition : some problem P2 is said NP-hard if it exists P1 NP-complete such that ( $P1 \leq_T P2$ ).

END