## NP-Completeness

Reference: Computers and Intractability: A
Guide to the Theory of NP-Completeness
by Garey and Johnson,
W.H. Freeman and Company, 1979.

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# General Problems, Input Size and Time Complexity

• Time complexity of algorithms:

polynomial time algorithm ("efficient algorithm") v.s.

exponential time algorithm ("inefficient algorithm")

| <b>f</b> ( <b>n</b> ) \ <b>n</b> | 10          | 30          | 50          |
|----------------------------------|-------------|-------------|-------------|
| n                                | 0.00001 sec | 0.00003 sec | 0.00005 sec |
| n <sup>5</sup>                   | 0.1 sec     | 24.3 sec    | 5.2 mins    |
| 2 <sup>n</sup>                   | 0.001 sec   | 17.9 mins   | 35.7 yrs    |

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## "Hard" and "easy' Problems

- Sometimes the dividing line between "easy" and "hard" problems is a fine one. For example
  - Find the shortest path in a graph from X to Y. (easy)
  - Find the longest path in a graph from X to Y. (with no cycles) (hard)
- View another way as "yes/no" problems
  - Is there a simple path from X to Y with weight  $\leq$  M? (easy)
  - Is there a simple path from X to Y with weight  $\geq$  M? (hard)
  - First problem can be solved in polynomial time.
  - All known algorithms for the second problem (could) take exponential time .

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• <u>Decision problem</u>: The solution to the problem is "yes" or "no". Most optimization problems can be phrased as decision problems (still have the same time complexity).

#### Example:

Assume we have a decision algorithm X for 0/1 Knapsack problem with capacity M, i.e. Algorithm X returns "Yes" or "No" to the question

"Is there a solution with profit  $\geq P$  subject to knapsack capacity  $\leq M$ ?"

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We can repeatedly run algorithm X for various profits(P values) to find an optimal solution. Example: Use binary search to get the optimal profit, maximum of  $\lg \sum p_i$  runs. (where M is the capacity of the knapsack optimization problem) Min Bound **Optimal Profit** Max Bound Search for the optimal solution  $\sum p_i$ 0 CS 331 NP-Completeness Young D&A of Algo.

### The Classes of P and NP

- The class P and Deterministic Turing Machine
  - Given a decision problem X, if there is a polynomial time Deterministic Turing Machine program that solves X, then X is belong to P
  - Informally, there is a polynomial time algorithm to solve the problem

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- The class NP and Non-deterministic Turing Machine
  - Given a decision problem X.

    If there is a polynomial time Non-deterministic

    Turing machine program that solves X, then X

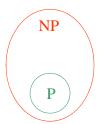
    belongs to NP
  - Given a decision problem X.
    For every instance I of X,
    (a) guess solution S for I, and
    (b) check "is S a solution to I?"
    If (a) and (b) can be done in polynomial time, then X belongs to NP.

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Obvious : P ⊆ NP, i.e. A
 (decision) problem in P does
 not need "guess solution".
 The correct solution can be
 computed in polynomial time.



- Some problems which are in NP, but may not in P:
  - 0/1 Knapsack Problem
  - PARTITION Problem : Given a finite set of positive integers Z.

Question: Is there a subset Z' of Z such that Sum of all numbers in Z' = Sum of all numbers in Z-Z'?

i.e. 
$$\sum Z' = \sum (Z-Z')$$

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• One of the most important open problem in theoretical compute science :

Is P=NP?

Most likely "No".

Currently, there are many known (decision) problems in NP, and there is no solution to show anyone of them in P.

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## **NP-Complete Problems**

- Stephen Cook introduced the notion of NP-Complete Problems.
  - This makes the problem "P = NP?" much more interesting to study.
- The following are several important things presented by Cook:

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- 1. Polynomial Transformation (" ∝ ")
  - $L1 \propto L2$ :

There is a polynomial time transformation that transforms arbitrary instance of L1 to some instance of L2.

- If L1  $\propto$  L2 then L2 is in P implies L1 is in P (or L1 is not in P implies L2 is not in P)
- If L1  $\propto$  L2 and L2  $\propto$  L3 then L1  $\propto$  L3

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- 2. Focus on the class of NP **decision** problems only. Many intractable problems, when phrased as decision problems, belong to this class.
- 3. L is **NP-Complete**

if (#1)  $L \in NP$  & (#2) for all other  $L' \in NP$ ,  $L' \propto L$ 

- If an NP-complete problem can be solved in polynomial time then all problems in NP can be solved in polynomial time.
- If a problem in NP cannot be solved in polynomial time then all problems in NP-complete cannot be solved in polynomial time.
- Note that an NP-complete problem is one of those hardest problems in NP.

L is **NP-Hard** if

(#2 of NP-Complete) for all other  $L' \in NP$ ,  $L' \propto L$ 

Note that an NP-Hard problem is a problem which is as hard as an NP-Complete problem and it's not necessary a decision problem.

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- So, if an NP-complete problem is in P then P=NP
- if P!= NP then all NP-complete problems are in NP-P
- 4. Question: How can we obtain the first NP-complete problem L?

**Cook Theorem : SATISFIABILITY is NP-Complete. (The first NP-Complete problem)** 

Instance: Given a set of variables, U, and a collection of clauses, C, over U.

Question: Is there a truth assignment for U that satisfies all clauses in C?

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$$\mathbf{U} = \{\mathbf{x}_1, \, \mathbf{x}_2\}$$

$$C_1 = \{(x_1, \neg x_2), (\neg x_1, x_2)\}$$

$$= (\mathbf{x}_1 \mathbf{OR} \neg \mathbf{x}_2) \mathbf{AND} (\neg \mathbf{x}_1 \mathbf{OR} \mathbf{x}_2)$$

if 
$$x_1 = x_2 = True \rightarrow C_1 = True$$

$$C_2 = (x_1, x_2) (x_1, \neg x_2) (\neg x_1) \rightarrow \text{not satisfiable}$$

"
$$\neg x_i$$
" = "not  $x_i$ " "OR" = "logical or" "AND" = "logical and"

This problem is also called "CNF-Satisfiability" since the expression is in CNF – Conjunctive Normal Form (the product of sums).

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• With the Cook Theorem, we have the following property:

#### Lemma:

If L1 and L2 belong to NP, L1 is NP-complete, and L1 ∝ L2 then L2 is NP-complete.

i.e. L1,  $L2 \in NP$  and for all other  $L' \in NP$ ,  $L' \propto L1$  and  $L1 \propto L2 \rightarrow L' \propto L2$ 

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- So now, to prove a (decision) problem L to be NP-complete, we need to
  - show L is in NP
  - select a known NP-complete problem L'
  - construct a polynomial time transformation f from L' to L
  - prove the correctness of f (i.e. L' has a solution if and only if L has a solution) and that f is a polynomial transformation

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• P: (Decision) problems solvable by deterministic algorithms in polynomial time

• NP: (Decision) problems solved by non-deterministic algorithms in polynomial time

 A group of (decision) problems, including all of the ones we have discussed (Satisfiability, 0/1 Knapsack, Longest Path, Partition) have an additional important property:

If any of them can be solved in polynomial time, then they all can!

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• These problems are called NP-complete problems.

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