

## **Theory of NP-Completeness**

- **NP-complete** is a class of problems with a certain property.
- Since it is a very theoretical area, it can also be complex.
- We won't get into the finer details of the theory, but will try to give an overall picture of what it is all about.
- Be aware that I will gloss over some important details/concepts, and make some simplifying assumptions.
- Before we talk about NP-complete problems, we first need to discuss several things, including
  - **decision problems,**
  - the class **P**, and
  - the class **NP**

## Optimization/Decision Problems

- **Optimization Problems**

- An **optimization problem** is one which asks “What is the optimal solution to problem  $X$ ?”
- **Examples:**
  - \* 0-1 knapsack/fractional knapsack
  - \* matrix chain multiplication
  - \* minimum spanning tree.

- **Decision Problems**

- A **decision problem** is a problem that asks “Is there a solution to problem  $X$  satisfying property  $Y$ ?”
- **Examples:**
  - \* Does graph  $G$  have a minimum spanning tree of weight  $\leq W$ ?
  - \* Can I multiply matrices  $(A_1, \dots, A_n)$  with  $\leq M$  operations?

## Optimization/Decision Example

- **Example**

- **Problem:** 0-1 knapsack

- **Instance:** A list of integer profits  $P = (p_1, \dots, p_n)$  and integer weights  $W = (w_1, \dots, w_n)$ , and capacity  $M$  (an integer).

- **Feasible Solution:** A vector  $x = (x_1, x_2, \dots, x_n)$ , where  $x_i \in \{0, 1\}$  for  $1 \leq i \leq n$ , and  $\sum_{i=1}^n w_i x_i \leq M$ .

- **Optimal Solution:** A feasible solution which maximizes **profit**.

That is, the quantity  $P = \sum_{i=1}^n p_i x_i$ .

- **Question:** Is the optimal profit  $\geq Q$ ?

- The optimization problem tries to find the optimal solution.

- The decision problem tries to answer the “yes/no” question.

## Why Decision Problems

- When we discuss the various classes of problems (P, NP, NP-Complete), we restrict our attention to decision problems.
- This simplifies the study of complexity.
- It turns out that this restriction is not really much of a restriction.
- There are simple ways of using the solution to a decision problem to get a solution to a related optimization problem.
- **Example:** If we can solve the 0-1 knapsack decision problem, we can solve the 0-1 knapsack optimization problem by using a **binary search** type algorithm.
- Before we go further, we will see another important example of a decision problem.

## Satisfiability (SAT)

- **Notation:** Let  $\vee$  denote logical *or* of 2 boolean variables, and  $\neg$  denote *negation* of a boolean variable. A **clause** is the logical *or* of boolean variables and/or their negations.
- **Problem:** Satisfiability (SAT)
  - **Instance:** A set of boolean variables  $U = \{x_1, x_2, \dots, x_m\}$ , and a set of clauses,  $C = \{c_1, c_2, \dots, c_s\}$ .
  - **Question:** Is there a satisfying truth assignment? In other words, can boolean values be assigned to the variables  $x_i$  so that each of the clauses is true?

## SAT Examples

- Let  $U = \{u, v, w, x\}$ , and

$$C = \{u \vee v \vee \neg x, \neg u \vee \neg v \vee w, u \vee \neg w \vee \neg x\}.$$

If we set  $u = x = \textit{false}$ , and  $v = w = \textit{true}$ , then it is easy to see that each of the clauses above is true, so the answer is “yes”.

- Let  $U = \{u, v, w\}$ , and

$$C = \{u \vee v, \neg u \vee \neg v, u \vee w, \neg u \vee \neg w, v \vee w, \neg v \vee \neg w\}.$$

Then the answer is “no”. But how do we know?

## SAT Examples Continued

- Let  $U = \{u, v, w\}$ , and

$$C = \{u \vee v, \neg u \vee \neg v, u \vee w, \neg u \vee \neg w, v \vee w, \neg v \vee \neg w\}.$$

It seem that the only way to know that there is no solution is to try them all:

$u$	0	0	0	0	1	1	1	1
$v$	0	0	1	1	0	0	1	1
$w$	0	1	0	1	0	1	0	1
$u \vee v$	0	0	1	1	1	1	1	1
$\neg u \vee \neg v$	1	1	1	1	1	1	0	0
$u \vee w$	0	1	0	1	1	1	1	1
$\neg u \vee \neg w$	1	1	1	1	1	0	1	0
$v \vee w$	0	1	1	1	0	1	1	1
$\neg v \vee \neg w$	1	1	1	0	1	1	1	0

## **YES versus NO**

- Notice that in the case of **SAT**:
  - answering “yes” is easy—we just find one assignment that works, but
  - answering “no” requires an exhaustive computation of all possible solutions.
- This is true for many decision problems.
- This is the basis of the theory of NP-completeness.
- We shall see more about this shortly.



## The Class P

- A **polynomial-time algorithm** is one which runs in time  $O(n^k)$ , for some constant  $k$ , where  $n$  is the *size of the input*.
- A **deterministic algorithm** is (essentially) one which always returns the correct answer.
- The algorithms we have studied in this course are all deterministic, and most of them are polynomial-time.
- **Definition:** **P** denotes the collection of decision problems which have *deterministic polynomial-time algorithms*.
- **Examples:**
  - Is the largest element in the array  $A$  larger than  $m$ ?
  - Does the graph  $G$  have a spanning tree with weight at most  $w$ ?

## Example of an problem in P

- **Example**

- **Problem:** Fractional knapsack

- **Instance:** A list of integer profits  $P = (p_1, \dots, p_n)$  and integer weights  $W = (w_1, \dots, w_n)$ , and capacity  $M$  (an integer).

- **Feasible Solution:** A vector  $x = (x_1, x_2, \dots, x_n)$ , where  $0 \leq x_i \leq 1$  for  $1 \leq i \leq n$ , and 
$$\sum_{i=1}^n w_i x_i \leq M.$$

- **Optimal Solution:** A feasible solution which maximizes **profit**.

That is, the quantity  $P = \sum_{i=1}^n p_i x_i$ .

- **Question:** Is the optimal profit  $\geq Q$ ?

- **Solution:** Solve optimization problem with the polynomial-time greedy algorithm, and test whether the optimal profit is larger than  $Q$ .

## Certificates

- A **certificate** is a set of data representing a solution to an instance of a (decision) problem.
- If a particular instance of a problem is a “yes” instance, then there is some certificate (set of data) that meets the requirements.
- Thus, certificates can be used to prove that a particular instance of a problem is a “yes” instance.
- A certificate is **valid** if it provides proof that the particular instance of a decision problem is a “yes” instance.
- To prove that a particular instance of a decision problem is a “yes” instance, one can generate certificates until a valid one is found.
- This may or may not yield an efficient algorithm.

## Certificate Example

- For **SAT**, a certificate is a truth assignment for the variables.

- We saw that when  $U = \{u, v, w, x\}$ , and

$$C = \{u \vee v \vee \neg x, \neg u \vee \neg v \vee w, u \vee \neg w \vee \neg x\}.$$

the certificate  $\{u = x = \text{true}, v = w = \text{false}\}$  is a proof that  $\{U, C\}$  is a “yes” instance of **SAT**.

- We also saw that if  $U = \{u, v, w\}$ , and

$$C = \{u \vee v, \neg u \vee \neg v, u \vee w, \neg u \vee \neg w, v \vee w, \neg v \vee \neg w\},$$

then none of the 8 certificates are valid. Since  $\{U, C\}$  is a “no” instance of **SAT**, this is what we should expect.

## Certificate Example

- For the 0-1 knapsack problem, a certificate is a list of integers  $(x_1, x_2, \dots, x_n)$  such that for  $1 \leq i \leq n$ ,  $x_i \in \{0, 1\}$ , and

$$\sum_{i=1}^n w_i x_i \leq M.$$

- A certificate is valid if it has profit  $\geq Q$ .

## Non-Deterministic Algorithms

- To prove that an instance of a decision problem is a “yes” instance, we only need to find one valid certificate.
- A **non-deterministic algorithm N** is one which
  - Guesses a certificate (The nondeterministic stage)
  - Checks the validity of the certificate (The deterministic stage)
  - Returns “yes” if the certificate is valid and “no” otherwise
- Notice that if a non-deterministic algorithm produces the answer “false” it *does not necessarily* mean that the instance of the problem is a “no” instance.
- It only means that the certificate it guessed was not a valid certificate.

## Examples

- **SAT**

- Let  $|U| = n$ , and  $|C| = m$ .
- If we are given a truth assignment for the  $n$  variables in  $U$ , we can check whether or not each of the  $m$  clauses is true or not in  $O(n)$  time in the worst-case.
- The total time to do the checking is  $O(nm)$ .
- If each of the clauses is true, we output “true”.
- Otherwise we output “false”.

- **Fractional knapsack**

- Given an assignment  $(x_1, x_2, \dots, x_n)$ , we can compute the profit  $P$  in  $O(n)$  steps, and compare it to  $Q$  in constant time.
- If  $P \geq Q$ , output “true”.
- Otherwise we output “false”.

## The Class NP

- **Definition:** NP denotes the collection of decision problems which have polynomial-time non-deterministic algorithms to solve them.
- In other words, a problem is in NP if every “yes” instance has some certificate that can be validated in polynomial time.
- Notice that a problem in NP does not necessarily have a deterministic polynomial algorithm to solve it.
- This is because the definition only assumes we can check the validity of certificates. It does not give us a method for finding valid certificates.
- For certain problems, there is no known method of finding valid certificates in polynomial-time.



## Problems in NP

- From our earlier example, it is clear that **SAT** and the fractional knapsack problem are both **NP**.
- The famous **traveling salesman problem** is in **NP**. Can you argue this?
  - **Problem:** Traveling salesman problem
  - **Instance:**  $n$  cities, and a cost  $c(i, j)$  to travel from city  $i$  to city  $j$ .
  - **Feasible Solution:** A route that takes the salesman through every city.
  - **Optimal Solution:** A feasible solution with minimal cost.
  - **Question:** Is there a route with cost  $\leq C$ ?
- As we will prove (indirectly), most of the algorithms we have discussed are in **NP**.

## More on NP

- As we just stated, for some problems in **NP**, there is no known way of producing a valid certificate in polynomial time.
- For instance, there is no known polynomial-time deterministic algorithm, or polynomial-time algorithm for finding a valid certificate for, the **traveling salesman problem**.
- In other words, it is unknown whether  $\mathbf{NP} \subseteq \mathbf{P}$  or not.
- It can be shown, however, that problems in **NP** can be solved in worst-case by exponential-time algorithms.
- As will be seen shortly,  $\mathbf{P} \subseteq \mathbf{NP}$ .
- It is important to remember that **NP** *does not* stand for *non-polynomial*.
- **NP** stands for **non-deterministic polynomial**, which is much different.

$$\mathbf{P} \subseteq \mathbf{NP}$$

- It is not very difficult to argue that  $\mathbf{P} \subseteq \mathbf{NP}$
- Since a problem in  $\mathbf{P}$  has a polynomial-time deterministic algorithm  $\mathbf{A}$ , we can compute a valid certificate  $C$  in polynomial time.
- We use this fact to write a nondeterministic algorithm  $\mathbf{N}$  as follows
  - Guess a certificate in the nondeterministic stage
  - Ignore the certificate in the deterministic stage
  - Instead, run  $\mathbf{A}$ , which returns “yes” or “no,” and return that value.
- Since  $\mathbf{A}$  is polynomial, so is the nondeterministic algorithm.
- If an instance is a “no” instance,  $\mathbf{A}$  (and hence  $\mathbf{N}$ ) will return “no,” the correct answer.
- If an instance is a “yes” instance,  $\mathbf{A}$  (and hence  $\mathbf{N}$ ) will return “yes,” the correct answer.

## *The Question: Is $P=NP$ ?*

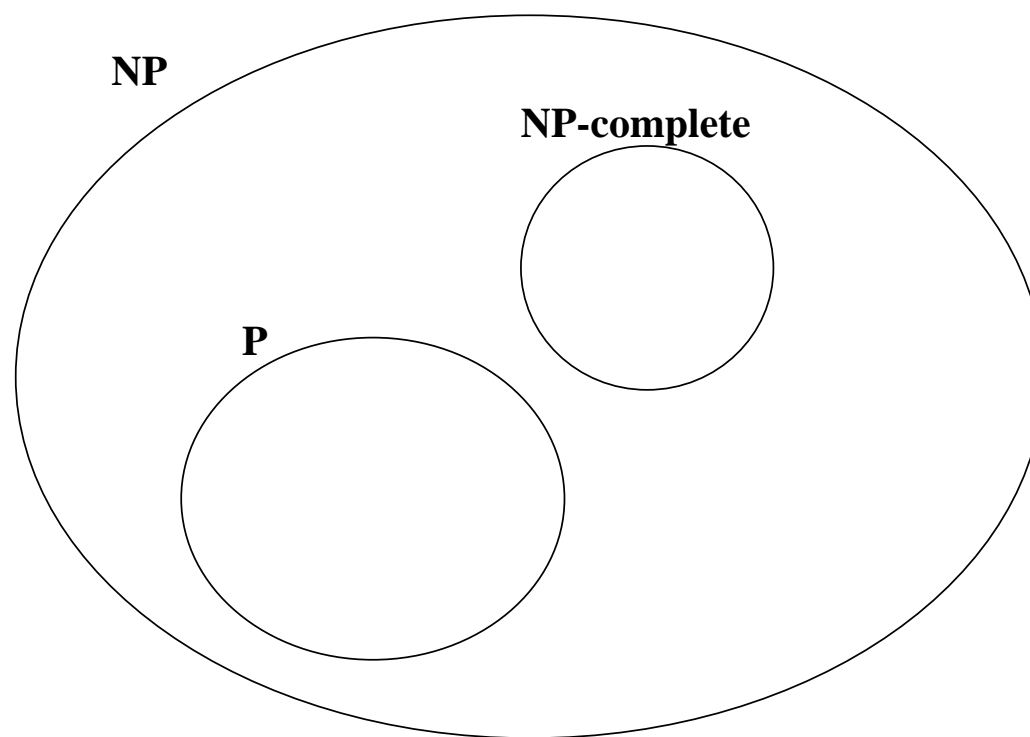
- The most important open question in theoretical computer science is the question “Is  $P=NP$ ?”
- Why is this so important?
  - If it can be proved that indeed  $P=NP$ , then we will have polynomial-time algorithms to solve thousands of problems for which the best-known algorithms are exponential.
  - If it can be shown that  $P \neq NP$ , then we will know that no polynomial-time algorithm exists for certain problems. That is, problems in  $NP-P$ .
- The question of which problems belong to the set  $NP-P$ , if any, has been studied extensively for several decades.
- This is where the class of **NP-complete** problems comes into play.

## NP-complete

- There is a lot more to the theory of NP-Completeness than we can cover here.
- It is important, however, to have a basic understanding of why **NP-complete** problems are so important.
- A problem  $D$  is said to be **NP-complete** if
  - $D \in \mathbf{NP}$ , and
  - For all  $D' \in \mathbf{NP}$ , there exists a polynomial time algorithm that maps “yes” (“no”) instances of  $D'$  to “yes” (“no”) instances of  $D$ .
- In other words, a solution to an **NP-complete** problem yields a solution to *any other* problem in **NP**. Why?
- Thus, if one can find a polynomial-time algorithm to solve an **NP-complete** problem, then it can be used to solve *all* problems in **NP**, and **P=NP**.

## **NP-complete Summary**

- Although it is not known whether or not  $P=NP$ , most people think  $P \neq NP$ .
- If  $P \neq NP$ , the **NP-complete** problems are the most likely problems to be in **NP-P**.
- Thus, the most likely scenario is as follows:



- If you can prove that  **$P=NP$**  or that  **$P \neq NP$** , you will be famous.

## A Final Note on NP-completeness

- There is one more important use of the theory of **NP-completeness**.
- If you are asked by your boss to write an efficient algorithm for some problem, there are several possibilities:
  - You will find a polynomial-time algorithm to solve the problem, and everyone will be happy.
  - You will be unable to find a polynomial-time algorithm, and your boss will fire you, and you will both be unhappy.
  - You will be unable to find a polynomial-time algorithm, but you will be able to show that the problem is **NP-complete**. Then your boss will be unhappy, but he can't fire you, because you have shown that nobody else can come up with a good solution either, so you will be happy.