

# **Chapter 11**

# **Introduction to Computational Complexity**

# Introduction to Computational Complexity

- A decision problem is decidable if there is an algorithm that can answer it in principle
- In this chapter, we try to identify the problems for which there are *practical* algorithms
  - Ones that can answer reasonable-size instances in a reasonable amount of time
- The *satisfiability problem* is decidable, but the known algorithms aren't much of an improvement on the brute-force algorithm that takes exponential time

# The Time Complexity of a Turing Machine, and the Set $P$

- The set  $P$  is the set of problems that can be decided by a TM in *polynomial time*, as a function of the instance size. (Brute-force algorithms tend to be exponential)
- $NP$  is defined similarly, except that we allow the use of a *nondeterministic* TM
- Most people assume that  $NP$  is a larger set, but no one has been able to demonstrate that  $P \neq NP$
- We discuss  $NP$ -complete problems, which are hardest problems in  $NP$ , and show that the satisfiability problem is one of these

# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- A TM deciding a language  $L \subseteq \Sigma^*$  solves a decision problem: Given  $x \in \Sigma^*$ , is  $x \in L$ ?
  - A measure of the size of the problem is the length of the input string  $x$

# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- Definition 11.1: Suppose  $T$  is a TM with input alphabet  $\Sigma$  that eventually halts on every input string
  - The *time complexity* of  $T$  is the function  $\tau_T : \mathbb{N} \rightarrow \mathbb{N}$ , where  $\tau_T(n)$  is defined by considering, for every input string of length  $n$  in  $\Sigma^*$ , the number of moves  $T$  makes on that string before halting, and letting  $\tau_T(n)$  be the maximum of these numbers
  - When we refer to a TM with a certain time complexity, it will be understood that it halts on every input

# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- Definition 11.4: If  $f$  and  $g$  are partial functions from  $\mathbb{N}$  to  $\mathbb{R}^+$  ; that is, both functions have values that are nonnegative real numbers wherever they are defined
  - We say that  $f = O(g)$ , or  $f(n) = O(g(n))$  (which we read “ $f$  is big-oh of  $g$ ” or “ $f(n)$  is big-oh of  $g(n)$ ”) if, for some positive numbers  $C$  and  $N$ ,  $f(n) \leq C g(n)$  for every  $n \geq N$
  - For example, every polynomial of degree  $k$  with positive leading coefficient is  $O(n^k)$

# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- An instance of the *satisfiability problem* is a Boolean expression
  - It involves Boolean variables  $x_1, x_2, \dots, x_n$  and the logical connectives  $\wedge, \vee$ , and  $\neg$
  - It is in conjunctive normal form (the conjunction of several clauses, each of which is a disjunction)
- Is there an assignment of truth values to the variables that satisfies the expression (makes it true)?
  - This problem is clearly decidable
    - We could simply try every possible assignment of values to variables

# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- The *traveling salesman problem* considers  $n$  cities that a salesman must visit, with a distance specified for every pair of cities
  - It's simplest to formulate this as an optimization problem
    - Determine the order that minimizes the total distance traveled
  - We can turn this into a decision problem by introducing a variable  $k$  and asking whether there is an order in which the cities could all be visited by traveling no more than distance  $k$



# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- There's a brute-force solution to this problem too
  - Consider all  $n!$  possible permutations of the cities
- With current hardware we can solve very large problems, if the problems require time  $O(n)$
- We can still solve largish problems if they take time  $O(n^2)$  or even  $O(n^3)$
- Exponential problems are another story
  - If the problem really requires time proportional to  $2^n$ , then doubling the speed of the machine only allows us to increase the size of the problem by 1!

# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- Showing that a brute-force approach takes a long time does not necessarily mean that the problem is complex
  - The satisfiability problem and the traveling salesman problem are assumed to be hard, not because the brute-force approach takes exponential time, but because no one has found a way of solving either problem that *doesn't* take at least exponential time

# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- What constitutes a *tractable* problem?
  - The most common answer is those that can be solved in polynomial time on a TM or other computer
  - One reason for this characterization is that it is relatively robust, as problems that can be solved in polynomial time on any computer can be solved in polynomial time on a TM as well, and vice-versa

# The Time Complexity of a Turing Machine, and the Set $P$ (cont'd.)

- Definition 11.5:  $P$  is the set of languages  $L$  such that for some TM  $T$  deciding  $L$  and some  $k \in \mathbb{N}$ ,  
 $\tau_T(n) = O(n^k)$
- The satisfiability and traveling salesman problems seem to be good candidates for real-life problems that are not in  $P$

# The Set $NP$ and Polynomial Verifiability

- The satisfiability problem seems like a hard problem
  - Testing a potential answer is easy, but there are an exponential number of potential answers
- We can approach this problem nondeterministically
  - We guess an answer (a particular truth assignment) and then test it deterministically
  - This can be done in polynomial time

# The Set $NP$ and Polynomial Verifiability (cont'd.)

- Definition 11.6: If  $T$  is an NTM with input alphabet  $\Sigma$  such that, for every  $x \in \Sigma^*$ , every possible sequence of moves of  $T$  on input  $x$  eventually halts, the time complexity  $\tau_T : \mathbb{N} \rightarrow \mathbb{N}$  is defined as follows:
  - Let  $\tau_T(n)$  be the maximum number of moves  $T$  can possibly make on any input string of length  $n$  before halting
  - As before, if we speak of an NTM as having a time complexity, we are assuming implicitly that no input string can cause it to loop forever

# The Set $NP$ and Polynomial Verifiability (cont'd.)

- Definition 11.7:  $NP$  is the set of languages  $L$  such that for some NTM  $T$  that cannot loop forever on any input, and some integer  $k$ ,  $T$  accepts  $L$  and

$$\tau_T(n) = O(n^k)$$

- We say that a language in  $NP$  can be accepted in *nondeterministic polynomial time*
- It is clear that  $P \subseteq NP$
- The *Sat* problem is in  $NP$  (the “guess-and-test” strategy is typical of problems in  $NP$ , and we can formalize this by constructing an appropriate NTM)

# The Set $NP$ and Polynomial Verifiability (cont'd.)

- Definition 11.10: If  $L \subseteq \Sigma^*$ , we say that a TM  $T$  is a *verifier* for  $L$  if:
  - $T$  accepts a language  $L_1 \subseteq \Sigma^*\{\$\}\Sigma^*$ ,  $T$  halts on every input, and
  - $L = \{x \in \Sigma^* \mid \text{for some } a \in \Sigma^*, x\$a \in L_1\}$  (we will call such a value  $a$  a *certificate* for  $x$ )
- A verifier  $T$  is a *polynomial-time verifier* if:
  - There is a polynomial  $p$  such that for every  $x$  and every  $a$  in  $\Sigma^*$ , the number of moves  $T$  makes on the input string  $x\$a$  is no more than  $p(|x|)$



# The Set $NP$ and Polynomial Verifiability (cont'd.)

- Theorem 11.11: For every language  $L \in \Sigma^*$ ,  $L \in NP$  if and only if  $L$  is polynomially verifiable
  - i.e., there is a polynomial-time verifier for  $L$
- Proof: See book
- A verifier for the satisfiability problem could take a specific truth assignment as a certificate; the traveling salesman problem could take a permutation of the cities as a certificate

# Polynomial-Time Reductions and *NP*-Completeness

- Just as we can show that a problem is decidable by reducing it to another one that is, we can show that a language is in  $P$  by reducing it to another that is
  - In the case of decidability, we only needed the reduction to be computable
  - Here we need the reduction function to be computable in polynomial time

# Polynomial-Time Reductions and *NP*-Completeness (cont'd.)

- Definition 11.12: If  $L_1$  and  $L_2$  are languages over respective alphabets  $\Sigma_1$  and  $\Sigma_2$ , a *polynomial-time reduction* from  $L_1$  to  $L_2$  is a function  $f: \Sigma_1^* \rightarrow \Sigma_2^*$  satisfying two conditions
  - First: for every  $x \in \Sigma_1^*$ ,  $x \in L_1$  if and only if  $f(x) \in L_2$
  - Second:  $f$  can be computed in polynomial time
    - i.e., there is a TM with polynomial time complexity that computes  $f$
- If there is a polynomial-time reduction from  $L_1$  to  $L_2$ , we write  $L_1 \leq_p L_2$  and say that  $L_1$  is polynomial-time reducible to  $L_2$ .

# Polynomial -Time Reductions and *NP*-Completeness (cont'd.)

- Theorem 11.13:
  - Polynomial-time reducibility is transitive:
    - If  $L_1 \leq_p L_2$  and  $L_2 \leq_p L_3$  then  $L_1 \leq_p L_3$
  - If  $L_1 \leq_p L_2$  and  $L_2 \in P$ , then  $L_1 \in P$
- Proof sketch:
  - For the first statement, simply use the composition of the reduction functions
  - For the second statement, simply combine the TM that accepts  $L_2$  and the one that computes the reduction  $f$

# Polynomial-Time Reductions and *NP*-Completeness (cont'd.)

- Definition 11.16: A language  $L$  is *NP-hard* if  $L_1 \leq_p L$  for every  $L_1 \in NP$ ;  $L$  is *NP-complete* if  $L \in NP$  and  $L$  is *NP-hard*
- Theorem 11.17:
  - If  $L$  and  $L_1$  are languages such that  $L$  is *NP-hard* and  $L \leq_p L_1$ , then  $L_1$  is also *NP-hard*
  - If  $L$  is any *NP-complete* language, then  $L \in P$  if and only if  $P = NP$
- Proof of Theorem 11.17: both parts follow from Theorem 11.13

# The Cook-Levin Theorem

- Theorem 11.18:
  - The language *Satisfiable* (or the corresponding decision problem *Sat*) is *NP*-complete
- Proof:
  - We know that *Satisfiable* is in *NP*, so we need to show that every language  $L \in NP$  is reducible to *Sat*
  - We do this by using a TM  $T$  that accepts  $L$ ; the reduction considers the details of  $T$  and takes a string  $x$  to a Boolean formula that is satisfiable if and only if  $x$  is accepted by  $T$
  - The details are complex and can be found in the book

# Some Other *NP*-Complete Problems

- Theorem 11.19:
  - The complete subgraph problem (Given a graph  $G$  and an integer  $k$ , does  $G$  have a complete subgraph with  $k$  vertices?) is *NP*-complete.
- Proof sketch:
  - By reduction from *Satisfiability*. For a Boolean expression  $x$  in conjunctive normal form, a graph can be constructed with vertices corresponding to occurrences of literals in  $x$ , and edges and an integer  $k$  chosen so that  $x$  is satisfiable if and only if the graph has a complete subgraph with  $k$  vertices

# Some Other *NP*-Complete Problems

- The problem *3-Sat* is the same as *Sat* except that every conjunct in the CNF expression is assumed to be the disjunction of three or fewer literals
- Theorem 11.20: *3-Sat* is *NP*-complete. Proof sketch:
  - *3-Sat* is in *NP* because *Sat* is
  - To get a reduction  $f$  from *Sat* to *3-Sat*, we let  $f(x)$  involve the variables in  $x$  as well as new ones
  - The trick is to incorporate the new variables so that
    - For every satisfying truth assignment to the variables of  $x$ , some assignment to the new variables makes  $f(x)$  true
    - For every nonsatisfying assignment to the variables of  $x$ , no assignment to the new variables makes  $f(x)$  true



# Some Other *NP*-Complete Problems (cont'd.)

- A *vertex cover* for a graph  $G$  is a set  $C$  of vertices such that every edge of  $G$  has an endpoint in  $C$
- The *vertex cover problem* is this: Given a graph  $G$  and an integer  $k$ , is there a vertex cover for  $G$  with  $k$  vertices?
- A  *$k$ -coloring* of  $G$  is an assignment to each vertex of one of the  $k$  colors so that no two adjacent vertices are colored the same
- The  *$k$ -colorability problem*: Given  $G$  and  $k$ , is there a  $k$ -coloring of  $G$ ?

# Some Other *NP*-Complete Problems (cont'd.)

- Theorem 11.21: The vertex cover problem is *NP*-complete
- Proof: We show that the problem is *NP*-hard by reducing the complete subgraph problem to it
  - The problem is clearly in *NP*
- Theorem 11.22: The *k-colorability* problem is *NP*-complete
- Proof: by reducing *3-Sat* to *k-colorability*
  - This problem is also clearly in *NP*

# Some Other $NP$ -Complete Problems (cont'd.)

- We now have five problems that are  $NP$ -complete
- There are thousands of others that are also known to be  $NP$ -complete
- Many real-life decision problems require some kind of solution
  - If a polynomial-time algorithm does not present itself, it is worth checking whether the problem is  $NP$ -complete
  - If so, finding such an algorithm will be as hard as proving that  $P = NP$