

# NP-Completeness and Boolean Satisfiability

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# Time-Bounded Turing Machines

- A Turing Machine that, given an input of length  $n$ , always halts within  $T(n)$  moves is said to be  $T(n)$ -time bounded.
  - The TM can be multitape.
  - Sometimes, it can be nondeterministic.
- The deterministic, multitape case corresponds roughly to an  $O(T(n))$  running-time algorithm.

# The class P

- If a DTM  $M$  is  $T(n)$ -time bounded for some polynomial  $T(n)$ , then we say  $M$  is polynomial-time (polytime) bounded.
- And  $L(M)$  is said to be in the class P.
- **Important Point:** When we talk of P, it doesn't matter whether we mean by a computer or by a TM.

# Polynomial Equivalence of Computers and TM's

- A multitape TM can simulate a computer that runs for time  $O(T(n))$  in at most  $O(T^2(n))$  of its own steps.
- If  $T(n)$  is a polynomial, so is  $T^2(n)$ .

# Examples of Problems in P

- Is  $w$  in  $L(G)$ , for a given CFG  $G$ ?
  - Input  $w$ .
  - Use CYK algorithm, which is  $O(n^3)$ .
- Is there a path from node  $x$  to node  $y$  in graph  $G$ ?
  - Input =  $x$ ,  $y$ , and  $G$ .
  - Use Dijkstra's algorithm, which is  $O(n \log n)$  on a graph of  $n$  nodes and arcs.

# Running Times Between Polynomials

- You might worry that something like  $O(n \log n)$  is not a polynomial.
- However, to be in  $P$ , a problem only needs an algorithm that runs in time *less than* some polynomial.
- Surely  $O(n \log n)$  is less than the polynomial  $O(n^2)$ .

# A Tricky Case: Knapsack

- The **Knapsack problem** is: given positive integers  $i_1, i_2, \dots, i_n$ , can we divide them in two sets with **equal** sums?
- Perhaps we can solve this problem in polytime by a **dynamic-programming** algorithm:
  - Maintain a table of all the differences we can achieve by partitioning the **first  $j$  integers**.

- **Basis:**  $j=0$ . Initially, the table has **true** for **0** and **false** for all other differences.
- **Induction:** To consider  $i_j$ , start with a new table initially **all false**.
- Then set  $k$  to **true** if, in the old table, there is a value  $m$  that was true, and  $k$  is either  $m+i_j$  or  $m-i_j$ .



- Suppose we measure running time in terms of the sum of the integers, say  $m$ .
- Each table only needs space  $O(m)$  to represent all the positive and negative differences we could achieve.
- Each table can be constructed in time  $O(n)$ .

- Since  $n \leq m$ , we can build the final table in  $O(m^2)$  time.
- From that table, we can see if  $0$  is achievable and solve the problem.

# Subtlety: Measuring Input Size

- **Input size** has a specific meaning: the length of the **representation** of the problem as it is **input to a TM**.
- For the Knapsack problem, you **cannot** always write the input in a number of characters that is polynomial in either the **number of** or **sum of** the integers.

# Knapsack - Bad Case

- Suppose we have  $n$  integers, each of which is around  $2^n$ .
- We can write integers in **binary**, so the input takes  $O(n^2)$  space to write down.
- But the tables require space  $O(n2^n)$ .
- They therefore require **at least** that order of time to construct.

- Thus, the proposed **polynomial** algorithm actually takes time  $O(n^2 2^n)$  on an input of length  $O(n^2)$ .
- Or, since we like to use  $n$  as the input size, it takes time  $O(n^{2^{\text{sqrt}(n)}})$  on an input of length  $n$ .
- In fact, it appears **no algorithm** solves Knapsack in polynomial time.

# Redefining Knapsack

- We are free to describe another problem, call it **Pseudo-Knapsack**, where integers are represented **unary**.
- Pseudo-Knapsack **is** in **P**.

# The class NP

- The running time of a **nondeterministic TM** is the maximum number of steps taken along any branch.
- If that time bound is **polynomial**, the NTM is said to be **polynomial-time bounded**.
- And its language/problem is said to be in the class **NP**.

## Example: NP

- The Knapsack problem is definitely in **NP**, **even** using the conventional binary representation of integers.
- Use nondeterminism to **guess** one of the subsets.
- Sum the two subsets and compare.



- Originally a curiosity of Computer Science, mathematicians now recognize as one of the most important open problems the question  $P = NP$ ?
- There are **thousands** of problems that are in **NP** but appear not to be in **P**.
- But no proof that they aren't really in **P**.

# Complete Problems

- One way to address the  $P = NP$  question is to identify complete problems for  $NP$ .
- An **NP-complete problem** has the property that if it is in  $P$ , then every problem in  $NP$  is also in  $P$ .
- Defined formally via polytime reductions.

# Complete Problems: Intuition

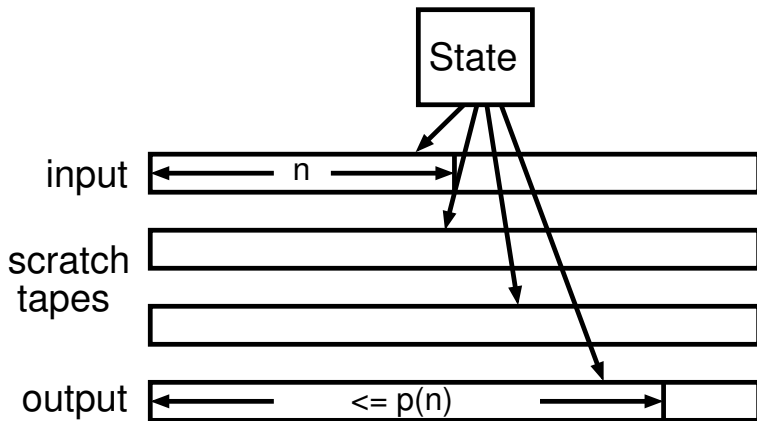
- A complete problem for a class **embodies every problem** in the class, even if it does not appear so.
- **Strange but true:** Knapsack embodies **every polytime NTM computation**.

- **Goal:** Find a way to show problem  $\mathcal{X}$  to be **NP-complete** by reducing **every** language/problem in **NP** to  $\mathcal{X}$  in such a way that if we had a **deterministic polynomial time algorithm** for  $\mathcal{X}$ , then we could **construct** a deterministic polynomial time algorithm for **any** problem in **NP**.

# Polytime Reductions

- We need the notion of a **polytime transducer** - a TM that:
  - ① Takes an input of length  $n$ .
  - ② Operates **deterministically** for some polynomial time  $p(n)$ .
  - ③ Produces an output on a separate **output tape**.
- **Note:** output length is at most  $p(n)$ .

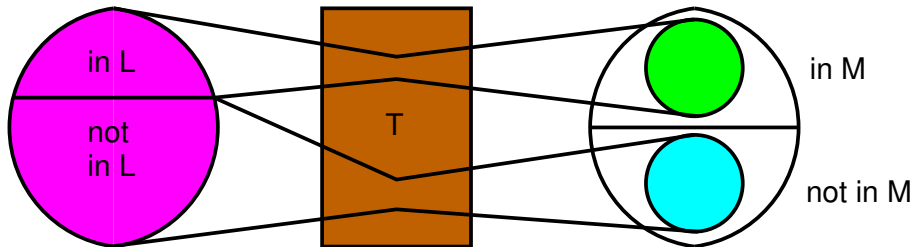
# Polytime Reductions



# Polytime Reductions

- Let  $L$  and  $M$  be languages.
- Say  $L$  is **polytime reducible** to  $M$  if there is a polytime transducer  $T$  such that for every input  $w$  to  $T$ , the output  $x = T(w) \in M$  if and only if  $w \in L$ .

# Picture of Polytime Reductions





# NP-complete Problems

- A problem/language  $M$  is said to be **NP-complete** if for every language  $L \in NP$ , there is a polytime reduction from  $L$  to  $M$ .
- **Fundamental Property:** If  $M$  has a polytime algorithm, then  $L$  also has a polytime algorithm.
  - i.e., if  $M \in P$ , then every  $L \in NP$  is also in  $P$ , or  $P = NP$ .

# Proof that Polytime Reductions Work

- Suppose  $M$  has an algorithm of polynomial time  $q(n)$ .
- Let  $L$  have a polytime transducer  $T$  to  $M$ , taking polynomial time  $p(n)$ .
- The output of  $T$ , given an input of length  $n$ , is at most of length  $p(n)$ .
- The algorithm for  $M$  on the output of  $T$  takes time at most  $q(p(n))$ .

# Proof that Polytime Reductions Work

- We now have a polytime algorithm for  $L$ :
  - ① Given  $w$  of length  $n$ , use  $T$  to produce  $x$  of length  $\leq p(n)$ , taking time  $\leq p(n)$ .
  - ② Use the algorithm for  $M$  to tell if  $x \in M$  in time  $\leq q(p(n))$ .
  - ③ Answer for  $w$  is whatever the answer for  $x$  is.
- Total time  $\leq p(n) + q(p(n)) =$  a polynomial.

- Boolean, or propositional-logic expressions are built from variables and constants using the operators **AND**, **OR**, and **NOT**.
  - Constants are **true** and **false**, represented by **1** and **0**, respectively.
  - We'll use **concatenation** for **AND**, **+** for **OR**, **-** for **NOT**.

## Example: Boolean Expressions

- $(x + y)(-x + -y)$  is **true** only when variables **x** and **y** have **opposite truth values**.
- **Note:** parentheses can be used **at will**, and are needed to modify the precedence order **NOT** (highest), **AND**, **OR**.

# The Satisfiability Problem (SAT)

- Study of boolean functions generally is concerned with the set of **truth assignments** (assignments of **0** or **1** to each of the variables) that make the function **true**.
- NP-completeness needs only a **simpler question** (SAT): does there exist a truth assignment making the function **true**?

# Example: SAT

- $(x + y)(-x + -y)$  is satisfiable.
- There are, in fact, two satisfying truth assignments:
  - ①  $x=0; y=1.$
  - ②  $x=1; y=0.$
- $x(-x)$  is not satisfiable.

- There is a multitape NTM that can decide if a Boolean formula of length  $n$  is satisfiable.
- The NTM takes  $O(n^2)$  time along **any path**.
- Use **nondeterminism** to guess a truth assignment on a second tape.
- **Replace** all variables by guessed truth values.
- Evaluate the formula for this assignment.
- Accept if true.

## Cook's Theorem

SAT is **NP-complete**.



- We have one NP-complete problem: SAT.
- In the future, we shall do polytime reductions of SAT to other problems, thereby showing them to be NP-complete.
- **Why?** If we polytime reduce SAT to  $\mathcal{X}$ , and  $\mathcal{X} \in \mathbf{P}$ , then so is SAT, and therefore so is all of NP.

# Conjunctive Normal Form

- A Boolean formula is in **Conjunctive Normal Form (CNF)** if it is the AND of **clauses**.
- Each **clause** is the OR of **literals**.
- Each **literal** is either a variable or the **negation** of a variable.

## CSAT

Is a boolean formula in CNF satisfiable?

# NP-completeness of CSAT

- You can convert **any formula** to CNF.
- It may **exponentiate** the size of the formula and therefore take time to write down that is exponential in the size of the original formula, but these numbers are all **fixed** for a given NTM **M** and **independent** of **n**.

- If a boolean formula is in CNF and every clause consists of exactly  $k$  literals, we say that the boolean formula is an instance of  $k$ -SAT.
  - Say the formula is in  $k$ -CNF.
- **Example:** 3-SAT formula:

$$(x + y + z)(x + -y + z)(x + y + -z)(x + -y + -z)$$

- **Every** boolean formula has an equivalent CNF formula.
  - But the size of the CNF formula may be **exponential** in the size of the original.
- **Not every** boolean formula has a  $k$ -SAT equivalent.
- **2-SAT** is in **P**, **3-SAT** is **NP-complete**.

## Proof: 2-SAT is in P (sketch)

- Pick an assignment for some variable, say  $x = \text{true}$ .
- Any clause with  $\neg x$  forces the **other** literal to be **true**.
  - **Example:**  $(\neg x + \neg y)$  forces  $y$  to be **false**.
- Keep seeing what other truth values are **forced** by variables with **known** truth values.

# Proof: 2-SAT is in P (sketch)

- One of **three** things can happen:
  - ① You reach a **contradiction** (e.g., **z** is forced to be both **true** and **false**).
  - ② You reach a point where **no more variables** have their truth value forced, but some clauses are **not yet** made true.
  - ③ You reach a **satisfying truth assignment**.

## Proof: 2-SAT is in P (sketch)

- **Case 1:** (**Contradiction**) There can be only a satisfying assignment if you use the **other truth value** for  $x$ .
  - **Simplify** the formula by replacing  $x$  by this truth value and **repeat** the process.
- **Case 3:** You found a satisfying assignment, so answer **yes**.



## Proof: 2-SAT is in P (sketch)

- **Case 2:** (You force values for **some** variables, but other variables and clauses are **not affected**.)
  - **Adopt** these truth values, **eliminate** the clauses that they satisfy, and **repeat**.
- In cases **1** and **2** you have spend  $O(n^2)$  time and have reduced the length of the formula by  $\geq 1$ , so  $O(n^3)$  total.

- This problem is NP-complete.
- Clearly it is in NP, since SAT is.
- It is not true that every boolean formula can be converted to an equivalent 3-CNF formula, even if we exponentiate the size of the formula.

- But we **don't need equivalence**.
- We need to reduce **every** CNF formula **F** to some 3-CNF formula that is satisfiable **if and only if F** is.
- Reduction involves introducing **new variables** into long clauses, so that we can split them apart.

# Reduction of CSAT to 3-SAT

- Let  $(x_1 + x_2 + \dots + x_n)$  be a clause in some CSAT instance, with  $n \geq 4$ .
  - **Note:** the  $x$ 's are **literals**, **not variables**; any of them could be **negated variables**.
- Introduce new variables  $y_1, \dots, y_{n-3}$  that appear in **no other clause**.

# Reduction of CSAT to 3-SAT

- Replace  $(x_1 + x_2 + \dots + x_n)$  by  $(x_1 + x_2 + y_1)(x_3 + y_2 + -y_1) \dots (x_{n-2} + y_{n-3} + -y_{n-4})(x_{n-1} + x_n + -y_{n-3})$ .
- If there is a satisfying assignment of the  $x$ 's for the CSAT instance, then one of the literals  $x_j$  must be made **true**.
- Assign  $y_i = \text{true}$  if  $j < i-1$  and  $y_j = \text{false}$  for larger  $j$ .

# Reduction of CSAT to 3-SAT

- We are **not** done.
- We also need to show that if the resulting 3SAT instance is satisfiable, then the **original** CSAT instance was **satisfiable**.

# Reduction of CSAT to 3-SAT

- Suppose  $(x_1 + x_2 + y_1)(x_3 + y_2 + -y_1) \dots (x_{n-2} + y_{n-3} + -y_{n-4})(x_{n-1} + x_n + -y_{n-3})$  is satisfiable, but none of the  $x$ 's is true.
- The first clause forces  $y_1 = \text{true}$ .
- Then the second clause forces  $y_2 = \text{true}$ .
- And so on ... all the  $y$ 's must be true.
- But then the last clause is false.

# Reduction of CSAT to 3-SAT

- There is a little more to the reduction, for handling clauses of 1 or 2 literals.
- Replace  $(x)$  by  $(x+y_1+y_2)(x+y_1+ -y_2)(x+ -y_1+y_2)(x+ -y_2+ -y_2)$ .
- Replace  $(w+x)$  by  $(w+x+y)(w+x+ -y)$ .



# CSAT to 3-SAT Running Time

- This reduction is surely **polynomial**.
- In fact, it is **linear** in the **length** of the CSAT instance.
- Thus, we have polytime-reduced CSAT to 3-SAT.
- Since CSAT is **NP-complete**, so is 3-SAT.