

# **Modern Complexity Theory (CS1.405)**

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# Intractability



- Assume that we grant the TM the ability to solve the satisfiability problem in a single step, for any size Boolean formula.
- Imagine an attached "black box" that gives the machine this
  capability. We call the black box an oracle to emphasize that it
  does not necessarily correspond to any physical device.
- The machine could use the oracle to solve any NP problem in polynomial time, regardless of whether P equals NP, because every NP problem is polynomial time reducible to the satisfiability problem (SAT).
- Such a TM is said to be computing relative to the satisfiability problem (SAT); hence the term relativization.



#### **Definition**

An oracle for a language A is a device that is capable of reporting whether any string w is a member of A. An oracle Turing machine  $M^A$  is a modified Turing machine that has the additional capability of querying an oracle for A. Whenever  $M^A$  writes a string on a special oracle tape, it is informed whether that string is a member of A in a single computation step.



### Definition (The Class $P^A$ )

 $P^A$  be the class of languages decidable with a polynomial time oracle deterministic Turing machine that uses oracle A. In other words,  $P^A := \{L | L \text{ is decided by a polynomial time oracle} \}$ 

deterministic Turing machine (ODTM) that uses oracle A}.

# Definition (The Class NPA)

 $NP^A$  be the class of languages decidable with a polynomial time oracle non-deterministic Turing machine that uses oracle A. In other words,  $NP^A := \{L|L \text{ is decided by a polynomial time oracle non-deterministic Turing machine (ONTM) that uses oracle <math>A$ }.



#### **Theorem**

 $NP \subseteq P^{SAT}$ .

#### **Theorem**

 $coNP \subset P^{SAT}$ .

 Polynomial-time computation relative to the satisfiability problem (SAT) contains all of NP.
 In other words,

$$NP \subset P^{SAT}$$

Furthermore,

$$coNP \subseteq P^{SAT}$$

because  $P^{SAT}$ , being a deterministic complexity class, is closed under complementation.



#### **Theorem**

- 1. An oracle A exists whereby  $P^A \neq NP^A$ .
- 2. An oracle B exists whereby  $P^B = NP^B$ .

**PROOF IDEA** Exhibiting oracle B is easy. Let B be any PSPACE-complete problem such as TQBF.

We exhibit oracle A by construction. We design A so that a certain language  $L_A$  in  $\mathrm{NP}^A$  provably requires brute-force search, and so  $L_A$  cannot be in  $\mathrm{P}^A$ . Hence we can conclude that  $\mathrm{P}^A \neq \mathrm{NP}^A$ . The construction considers every polynomial time oracle machine in turn and ensures that each fails to decide the language  $L_A$ .

**PROOF** Let B be TQBF. We have the series of containments

$$\mathrm{NP}^{\mathit{TQBF}} \overset{1}{\subseteq} \mathrm{NPSPACE} \overset{2}{\subseteq} \mathrm{PSPACE} \overset{3}{\subseteq} \mathrm{P}^{\mathit{TQBF}}.$$

Containment 1 holds because we can convert the nondeterministic polynomial time oracle TM to a nondeterministic polynomial space machine that computes the answers to queries regarding TQBF instead of using the oracle. Containment 2 follows from Savitch's theorem. Containment 3 holds because TQBF is PSPACE-complete. Hence we conclude that  $P^{TQBF} = NP^{TQBF}$ .



#### **Problem**

If  $NP = P^{SAT}$ , then NP = coNP.

**Solution**:  $P^{SAT}$  is a deterministic class, so it is closed by complementation. We are also given that  $NP = P^{SAT}$ .

Note that,  $A \in P$  if and only if (iff)  $A \in P^{SAT}$ .

Then,  $A \in P^{SAT}$  iff  $\bar{A} \in P^{SAT}$ .

Since  $NP = P^{SAT}$ ,  $\bar{A} \in NP$  iff  $A \in coNP$ .

As a result,  $A \in NP$  iff  $A \in coNP$ , that is,  $NP \subseteq coNP$  and  $coNP \subseteq NP$ .

Hence, NP = coNP.



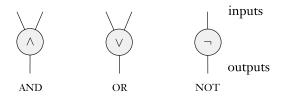
#### Motivation

- Computers are built from electronic devices wired together in a design called a digital circuit.
- We can also simulate theoretical models, such as Turing machines, with the theoretical counterpart to digital circuits, called Boolean circuits.
- Two purposes are served by establishing the connection between TMs and Boolean circuits.
  - First, researchers believe that circuits provide a convenient computational model for attacking the P versus NP and related questions.
  - Second, circuits provide an alternative proof of the Cook–Levin theorem that SAT is NP-complete.



#### Definition (Boolean circuit)

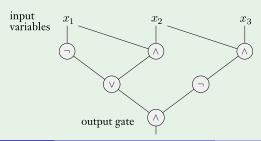
A **Boolean circuit** is a collection of **gates** and **inputs** connected by wires. Cycles are not permitted. Gates take three forms: 1) AND gates, 2) OR gates, and 3) NOT gates



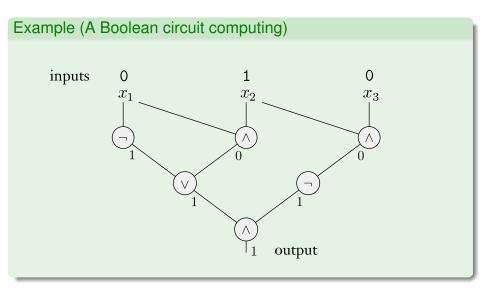


#### Example (A Boolean circuit)

The wires in a Boolean circuit carry the Boolean values 0 and 1. The gates are simple processors that compute the Boolean functions AND, OR, and NOT. The AND function outputs 1 if both of its inputs are 1 and outputs 0 otherwise. The OR function outputs 0 if both of its inputs are 0 and outputs 1 otherwise. The NOT function outputs the opposite of its input; in other words, it outputs a 1 if its input is 0 and a 0 if its input is 1. The inputs are labeled  $x_1, \ldots, x_n$ . One of the gates is designated the *output gate*. The following figure depicts a Boolean circuit.









#### Boolean circuit computing

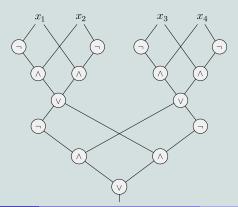
We use functions to describe the input/output behavior of Boolean circuits. To a Boolean circuit C with n input variables, we associate a function  $f_C: \{0,1\}^n \longrightarrow \{0,1\}$ , where if C outputs b when its inputs  $x_1, \ldots, x_n$  are set to  $a_1, \ldots, a_n$ , we write  $f_C(a_1, \ldots, a_n) = b$ . We say that C computes the function  $f_C$ . We sometimes consider Boolean circuits that have multiple output gates. A function with k output bits computes a function whose range is  $\{0,1\}^k$ .



### Boolean circuit computing

Example (A Boolean circuit that computes the parity function)

The *n*-input *parity function*  $parity_n : \{0,1\}^n \to \{0,1\}$  outputs 1 if an odd number of 1s appear in the input variables.





#### Definition (Circuit family)

A *circuit family* C is an infinite list of circuits, say  $(C_0, C_1, C_2, ...)$ , where  $C_n$  has n input variables. Then, C decides a language A over the alphabet  $\{0,1\}$  if for every string w,

$$w \in A \operatorname{iff} C_n(w) = 1,$$

where n is the length of w.

- The size of a circuit is the number of gates that it contains.
- Two circuits are equivalent if they have the same input variables and output the same value on every input assignment.
- A circuit is **size minimal** if no smaller circuit is equivalent to it.
- The problem of minimizing circuits has obvious engineering applications but is very difficult to solve in general.



- A circuit family is minimal if every  $C_i$  on the list is a minimal circuit.
- The *size complexity* of a circuit family  $(C_0, C_1, C_2,...)$  is the function  $f: N \to N$ , where f(n) is the size of  $C_n$ .
- The *depth* of a circuit is the length (number of wires) of the longest path from an input variable to the output gate.
- We define depth minimal circuits and circuit families, and the depth complexity of circuit families, as we did with circuit size.



#### Definition

- The circuit complexity of a language is the size complexity of a minimal circuit family for that language.
- The circuit depth complexity of a language is defined similarly, using depth instead of size.

#### **Theorem**

Let  $t: N \to N$  be a function, where  $t(n) \ge n$ . If a language  $A \in TIME(t(n))$ , then A has circuit complexity  $O(t^2(n))$ .



We say that a Boolean circuit is **satisfiable** if some setting of the input causes the circuit to output 1.

The *circuit-satisfiable* problem tests whether a circuit is satisfiable. Define

CIRCUIT-SAT :=  $\{\langle C \rangle | C \text{ is a satisfiable Boolean circuit} \}$ .

#### **Theorem**

CIRCUIT-SAT is NP-complete.

#### Proof.

We must demonstrate the following two things:

- CIRCUIT-SAT is in NP.
- (NP-hard) any language A in NP is reducible to CIRCUIT-SAT.



#### Part 1. CIRCUIT-SAT is in NP.

We design the following polynomial-time verifier (DTM, V) that can decide CIRCUIT-SAT.

#### Algorithm: Polynomial-time verifier (DTM, V) for CIRCUIT-SAT

**Input:**  $(\langle C \rangle, \beta)$ , where  $\beta$  is a certificate (an assignment of the setting of the n Boolean variables, say  $x_1, x_2, \cdots, x_n$ )

Output: Accept/Reject

- 1: **if**  $\beta$  does not contain n Boolean variables assignment **then**
- 2: return "reject"
- 3: **end if**
- 4: Evaluate the circuit C on  $\beta$ .
- 5: if the circuit output is 1 then
- 6: return "accept"
- 7: **else**
- 8: return "reject"
- 9: end if

#### Part 2. CIRCUIT-SAT is NP-hard.



We show that any language A in NP is poly-time reducible to CIRCUIT-SAT. We must give a polynomial time reduction  $f: \Sigma^* \to \Sigma^*$  that maps strings to circuits, where  $f(w) = \langle C \rangle$  implies that  $w \in A$  iff Boolean circuit C is satisfiable.

Because A is in NP, it has a polynomial time verifier (DTM) V whose input has the form  $\langle x,c\rangle$  where c may be the certificate showing that the assignment x is in A. In order to construct f, we need to obtain the circuit simulating V.

We now fill in the inputs to the circuit that correspond to x with the symbols of w. The only remaining inputs to the circuit correspond to the certificate c. We call this circuit C and output it.

If C is satisfiable, a certificate exists, so w is in A. Conversely, if w is in A, a certificate exists, so C is satisfiable.

**Time Complexity:** If the running time of the polynomial-time verifier V is  $n^k$  for some positive integer k, so the size of the circuit constructed is  $O((n^k)^2) = O(n^{2k})$  by the circuit complexity theorem. Since the structure of the circuit is quite simple (actually, it is highly repetitious), so the running time of the reduction is  $O(n^{2k})$ , which is poly-time.



#### Definition (P-complete)

A language B is P-complete if

- lacktriangledown  $B \in P$ , and
- 2 every A in P is log space reducible to B, that is,

$$A \leq_L B, \forall A \in P$$
.

[P-hard]



### P-completeness

For a circuit C and input setting x, we represent C(x) to be the value of C on x. Define

CIRCUIT-VALUE :=  $\{\langle C, x \rangle | C \text{ is a Boolean circuit and } C(x) = 1\}.$ 

#### **Theorem**

CIRCUIT-VALUE is P-complete.

#### Proof.

**Part 1.** CIRCUIT-VALUE is in P. We need to design a DTM, say, M that can decide CIRCUIT-VALUE in poly-time.

Part 2. Any A in P is log space reducible to CIRCUIT-VALUE, that is,

$$A \leq_L CIRCUIT-VALUE, \forall A \in P.$$





#### Part 1.

Algorithm: Polynomial-time DTM, M for CIRCUIT-VALUE Input: (C, x), where C is a Boolean circuit and x an assignment of the setting of the *n* Boolean variables, say  $x_1, x_2, \dots, x_n$ 

Output: Accept/Reject

- 1: Evaluate the circuit C on x.
- 2: if the circuit output is 1 then
- return "accept"
- 4: else
- return "reject"
- 6: **end** if



#### Part 2. Required to Prove (RTP):

$$A \leq_L CIRCUIT-VALUE, \forall A \in P$$

Let A be any language in P. We convert an input string  $w \in A$  to a Boolean circuit C with w such that  $w \in A$  if and only if  $f(w) = \langle C \rangle$  is satisfiable Boolean circuit on w, where  $f : \Sigma^* \to \Sigma^*$  is log-space reduction function.

On input w, the reduction produces a circuit C that simulates the polynomial time deterministic Turing machine (DTM) for A. The input to the circuit C is w itself. The reduction can be carried out in log space because the circuit C it produces has a simple and repetitive structure.



# Thank You!!!