

## COMPLEXITY THEORY

**Lecture 19: Circuit Complexity** 

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### Motivation

One might imagine that P  $\neq$  NP, but **S**<sub>AT</sub> is tractable in the following sense: for every  $\ell$  there is a very short program that runs in time  $\ell^2$  and correctly treats all instances of size  $\ell$ . – Karp and Lipton, 1982

### Some questions:

- Even if it is hard to find a universal algorithm for solving all instances of a problem, couldn't it still be that there is a simple algorithm for every fixed problem size?
- What can complexity theory tell us about parallel computation?
- Are there any meaningful complexity classes below LogSpace? Do they contain relevant problems?

→ circuit complexity provides some answers

Intuition: use circuits with logical gates to model computation

# Computing with Circuits

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### **Boolean Circuits**

**Definition 19.1:** A Boolean circuit is a finite, directed, acyclic graph where

- each node that has no predecessor is an input node
- each node that is not an input node is one of the following types of logical gate:
  - AND with two input wires
  - OR with two input wires
  - NOT with one input wire
- one or more nodes are designated output nodes

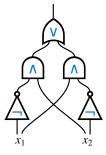
The outputs of a Boolean circuit are computed in the obvious way from the inputs.  $\rightarrow$  circuits with k inputs and  $\ell$  outputs represent functions  $\{0,1\}^k \rightarrow \{0,1\}^\ell$ 

We often consider circuits with only one output.

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## Example 1

#### XOR function:



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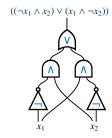
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## Alternative Ways of Viewing Circuits (1)

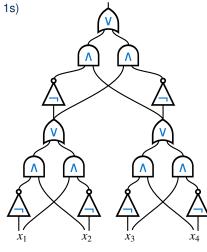
### Propositional formulae

- propositional formulae are special circuits: each non-input node has only one outgoing wire
- each variable corresponds to one input node
- each logical operator corresponds to a gate
- each sub-formula corresponds to a wire



## Example 2

Parity function with four inputs: (true for odd number of 1s)



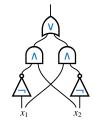
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## Alternative Ways of Viewing Circuits (2)

### Straight-line programs

- are programs without loops and branching (if, goto, for, while, etc.)
- that only have Boolean variables
- and where each line can only be an assignment with a single Boolean operator

 $\rightarrow$  *n*-line programs correspond to *n*-gate circuits



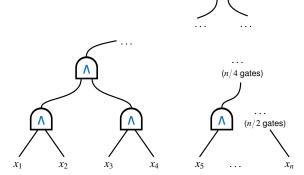
 $z_1 := \neg x_1$  $z_2 := \neg x_2$  $z_3 := z_1 \land x_2$  $z_4 := z_2 \land x_1$  slide 6 of 25

**05** return  $z_3 \lor z_4$ 

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### Example: Generalised AND

The function that tests if all inputs are 1 can be encoded by combining binary AND gates:



- works similarly for OR gates
- number of gates:n − 1
- we can use n-way AND and OR (keeping the real size in mind)

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# Circuit Complexity

To measure difficulty of problems solved by circuits, we can count the number of gates needed:

**Definition 19.4:** The size of a circuit is its number of gates.

Let  $f: \mathbb{N} \to \mathbb{R}^+$  be a function. A circuit family C is f-size bounded if each of its circuits  $C_n$  is of size at most f(n).

Size(f(n)) is the class of all languages that can be decided by an O(f(n))-size bounded circuit family.

**Example 19.5:** Our circuits for generalised AND show that  $\{1^n \mid n \ge 1\} \in \text{Size}(n)$ .

## Solving Problems with Circuits

Circuits are not universal: they have a fixed number of inputs! How can they solve arbitrary problems?

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Definition 19.2: A circuit family is an infinite list C = C_1, C_2, C_3, \ldots where each C_i is a Boolean circuit with i inputs and one output. We say that C decides a language L (over \{0,1\}) if w \in L if and only if C_n(w) = 1 for n = |w|.
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**Example 19.3:** The circuits we gave for generalised AND are a circuit family that decides the language  $\{1^n \mid n \geq 1\}$ .

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# Examples

Many simple operations can be performed by circuits of polynomial size:

- Boolean functions such as parity (=sum modulo 2), sum modulo n, or majority
- Arithmetic operations such as addition, subtraction, multiplication, division (taking two fixed-arity binary numbers as inputs)
- Many matrix operations

See exercise for some more examples

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# **Polynomial Circuits**

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### Quadratic Circuits for Deterministic Time

**Theorem 19.7:** For  $f(n) \ge n$ , we have  $\mathsf{DTime}(f) \subseteq \mathsf{Size}(f^2)$ .

#### Proof sketch (see also Sipser, Theorem 9.30)

 We can represent the DTime computation as in the proof of Theorem 16.10: as a list of configurations encoded as words

$$*\sigma_1 \cdots \sigma_{i-1} \langle q, \sigma_i \rangle \sigma_{i+1} \cdots \sigma_m *$$

of symbols from the set  $\Omega = \{*\} \cup \Gamma \cup (Q \times \Gamma)$ .

- $\rightarrow$  Tableau (i.e., grid) with  $O(f^2)$  cells.
- We can describe each cell with a list of bits (wires in a circuit).
- We can compute one configuration from its predecessor by O(f) circuits (idea: compute the value of each cell from its three upper neighbours as in Theorem 16.10)
- Acceptance can be checked by assuming that the TM returns to a unique configuration position/state when accepting

### **Polynomial Circuits**

A natural class of problems to consider are those that have polynomial circuit families:

**Definition 19.6:**  $P_{\text{poly}} = \bigcup_{d \ge 1} \text{Size}(n^d)$ .

**Note:** A language is in  $P_{\rm poly}$  if it is solved by some polynomial-sized circuit family. There may not be a way to compute (or even finitely represent) this family.

How does P<sub>/poly</sub> relate to other classes?

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## From Polynomial Time to Polynomial Size

From  $\mathsf{DTime}(f) \subseteq \mathsf{Size}(f^2)$  we get:

**Corollary 19.8:**  $P \subseteq P_{poly}$ .

This suggests another way of approaching the P vs. NP question:

If any language in NP is not in  $P_{/poly}$ , then  $P \neq NP$ . (but nobody has found any such language yet)

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#### **CIRCUIT-SAT**

Input: A Boolean Circuit *C* with one output.

Problem: Is there any input for which *C* returns 1?

#### Theorem 19.9: CIRCUIT-SAT is NP-complete.

**Proof:** Inclusion in NP is easy (just guess the input).

For NP-hardness, we use that NP problems are those with a P-verifier:

- The DTM simulation of Theorem 19.7 can be used to implement a verifier (input: (w#c) in binary)
- We can hard-wire the *w*-inputs to use a fixed word instead (remaining inputs: *c*)
- The circuit is satisfiable iff there is a certificate for which the verifier accepts w

**Note:** It would also be easy to reduce **SAT** to **CIRCUIT-SAT**, but the above yields a proof from first principles.

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Is 
$$P = P_{poly}$$
?

We showed  $P \subseteq P_{poly}$ . Does the converse also hold?

#### No!

Theorem 19.11: P<sub>poly</sub> contains undecidable problems.

**Proof:** We define the unary Halting problem as the (undecidable) language:

**UHALT** :=  $\{1^n \mid \text{the binary encoding of } n \text{ encodes a pair } \langle \mathcal{M}, w \rangle$ where  $\mathcal{M}$  is a TM that halts on word  $w\}$ 

For a number  $1^n \in \mathbf{UHALT}$ , let  $C_n$  be the circuit that computes a generalised AND of all inputs. For all other numbers, let  $C_n$  be a circuit that always returns 0. The circuit family  $C_1, C_2, C_3, \ldots$  accepts  $\mathbf{UHALT}$ .

Note: Interestingly, UHALT is also not hard for NP (since it is a so-called sparse language)

A New Proof for Cook-Levin

Theorem 19.10: 3SAT is NP-complete.

**Proof:** Membership in NP is again easy (as before).

For NP-hardness, we express the circuit that was used to implement the verifier in Theorem 19.9 as propositional logic formula in 3-CNF:

- Create a propositional variable X for every wire in the circuit
- Add clauses to relate input wires to output wires, e.g., for AND gate with inputs X₁ and X₂ and output X₃, we encode (X₁ ∧ X₂) ↔ X₃ as:

$$(\neg X_1 \vee \neg X_2 \vee X_3) \wedge (X_1 \vee \neg X_3) \wedge (X_2 \vee \neg X_3)$$

- Fixed number of clauses per gate = linear size increase
- Add a clause (X) for the output wire X

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### **Uniform Circuit Families**

 $P_{\rm poly}$  too powerful, since we do not require the circuits to be computable. We can add this requirement:

**Definition 19.12:** A circuit family  $C_1, C_2, C_3, \ldots$  is log-space-uniform if there is a log-space computable function that maps words  $1^n$  to (an encoding of)  $C_n$ .

Note: We could also define similar notions of uniformity for other complexity classes.

**Theorem 19.13:** The class of all languages that are accepted by a log-space-uniform circuit family of polynomial size is exactly P.

**Proof sketch:** A detailed analysis shows that our earlier reduction of polytime DTMs to circuits is log-space-uniform. Conversely, a polynomial-time procedure can be obtained by first computing a suitable circuit (in log-space) and then evaluating it (in polynomial time).

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### Turing Machines That Take Advice

One can also describe  $P_{poly}$  using TMs that take "advice":

**Definition 19.14:** Consider a function  $a: \mathbb{N} \to \mathbb{N}$ . A language **L** is accepted by a Turing Machine  $\mathcal{M}$  with a bits of advice if there is a sequence of advice strings  $\alpha_0, \alpha_1, \alpha_2, \dots$  of length  $|\alpha_i| = a(i)$  and  $\mathcal{M}$  accepts inputs of the form  $(w\#\alpha_{|w|})$  if and only if  $w \in \mathbf{L}$ .

 $P_{/\mathrm{poly}}$  is equivalent to the class of problems that can be solved by a PTime TM that takes a polynomial amount of "advice" (where the advice can be a description of a suitable circuit).

(This is where the notation  $P_{poly}$  comes from.)

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## P<sub>/polv</sub> and ExpTime

We showed  $P \subseteq P_{poly}$ . Does ExpTime  $\subseteq P_{poly}$  also hold? Nobody knows.

### Theorem 19.16 (Meyer's Theorem):

If ExpTime  $\subseteq P_{\text{poly}}$  then ExpTime  $= PH = \Sigma_2^p$ .

See [Arora/Barak, Theorem 6,20] for a proof sketch.

**Corollary 19.17:** If ExpTime  $\subseteq P_{/poly}$  then  $P \neq NP$ .

**Proof:** If ExpTime  $\subseteq P_{poly}$  then ExpTime  $= \Sigma_2^p$  (Meyer's Theorem).

So the Polynomial Hierarchy doesn't collapse completely, and  $P \neq NP$ .

By the Time Hierarchy Theorem,  $P \neq ExpTime$ , so  $P \neq \Sigma_2^p$ . П

# P<sub>poly</sub> and NP

We showed  $P \subseteq P_{poly}$ . Does  $NP \subseteq P_{poly}$  also hold? Nobody knows.

**Theorem 19.15 (Karp-Lipton Theorem):** If NP  $\subseteq$  P<sub>/poly</sub> then PH =  $\Sigma_2^p$ .

#### Proof sketch (see Arora/Barak Theorem 6.19):

- if  $NP \subseteq P_{/poly}$  then there is a polysize circuit family solving Sat
- Using this, one can argue that there is also a polysize circuit family that computes the lexicographically first satisfying assignment (k output bits for k variables)
- A  $\Pi_2$ -QBF formula  $\forall \vec{X}. \exists \vec{Y}. \varphi$  is true if, for all values of  $\vec{X}$ ,  $\varphi(\vec{X})$  is satisfiable.
- In  $\Sigma_2^P$ , we can: (1) guess the polysize circuit for SAT, (2) check for all values of  $\vec{X}$  if its output is really a satisfying assignment (to verify the guess)

- This solves  $\Pi_2^P$ -hard problems in  $\Sigma_2^P$
- But then the Polynomial Hierarchy collapses at  $\Sigma_2^P$ , as claimed.

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## How Big a Circuit Could We Need?

We should not be surprised that  $P_{poly}$  is so powerful: exponential circuit families are already enough to accept any language

**Exercise:** show that every Boolean function over *n* variables can be expressed by a circuit of size  $\leq n2^n$ .

It turns out that these exponential circuits are really needed:

**Theorem 19.18 (Shannon 1949 (!)):** For every n, there is a function  $\{0,1\}^n \rightarrow$  $\{0,1\}$  that cannot be computed by any circuit of size  $2^n/(10n)$ .

In fact, one can even show: almost every Boolean function requires circuits of size  $> 2^n/(10n)$  – and is therefore not in P<sub>/poly</sub>

Is any of these functions in NP? Or at least in Exp? Or at least in NExp? Nobody knows.

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# Summary and Outlook

Circuits provide an alternative model of computation

Nonuniform circuit families are very powerful, and even polynomial circuits can solve undecidable problems

Log-space-uniform polynomial circuits capture P.

Most boolean functions cannot be expressed by polynomial circuits, yet we don't know of any such function that is even in NExp

### What's next?

- Circuits for parallelism
- Complexity classes (strictly!) below P
- Randomness

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