15-251: Great Theoretical Ideas in Computer Science Spring 2017, Lecture 15

Boolean Formulas and Circuits



Today

- Briefly mention the "P versus NP" problem
- Remind you of Boolean formulas
- Tell you about Boolean circuits
- Relate circuit size to algorithmic efficiency
- See why circuits are a good approach to P vs. NP
- See why circuits are a bad approach to P vs. NP

P versus NP

The most famous unsolved problem in Theoretical Computer Science

Computer Science Theoretical Computer Science

Mathematics

P versus NP

One if not **the** most famous unsolved problems in all of Computer Science and all of Mathematics

Computer Science Theoretical Computer Science

Mathematics



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This article is a **list of unsolved problems in computer science**. A problem in computer science is considered unsolved when an expert in the field (i.e., a computer scientist) considers it unsolved or when several experts in the field disagree about a solution to a problem.

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Computational complexity [edit]

• P versus NP problem often written as "P = NP," which is technically not correct for the problem or those below)



P versus NP

One if not **the** most famous unsolved problems in all of Computer Science and all of Mathematics

I can state it for you in ten minutes

Warning: You won't get the full, glorious perspective on why "**P** versus **NP**" is so important until Lectures 17–19

Boolean formulas You've seen these before in Concepts:

 $((\neg x \rightarrow y) \land ((x \lor z) \leftrightarrow y))$

x, y, z, ... Boolean *variables*, values 0/1 (or T/F) \neg , \land , \lor , ... Boolean *connectives* (or *operations*)

Α	В	٦A	(A∧B)	(A v B)	(A→B)	(A⇔B)
0	0	1	0	0	1	1
0	1	1	0	1	1	0
1	0	0	0	1	0	0
1	1	0	1	1	1	1

Boolean formulas You've seen these before in Concepts. Stuff like this: $((\neg x \rightarrow y) \land ((x \lor z) \leftrightarrow y))$

x, y, z, ... Boolean *variables*, values 0/1 (or T/F) \neg , \land , \lor , ... Boolean *connectives* (or *operations*)

Truth assignment: 0/1 value for each variable

A formula is **satisfiable** if there's a truth assignment to the variables making the whole formula true

Truth tables



Satisfiable:At least one 1 in truth tableUnsatisfiable:No 1's in truth tableTautology:All 1's in truth table

An unsolved problem in Computer Science/Mathematics: Who invented truth tables?



Russell?



Wittgenstein?



Post?



Peirce?



Łukasiewicz?



Jevons?



Ladd–Franklin?

Another unsolved problem in Computer Science/Mathematics:

What is the intrinsic complexity of SAT?

SAT: Given as input a Boolean formula, decide if it is satisfiable or not.

Question:Is SAT decidable?Answer:Yes.

Say the input formula is G.

Brute-Force-Algorithm(G):

Enumerate all truth assignments α . For each α , compute the truth value it gives G. If any of them satisfy G, then ACCEPT, else REJECT.

Remark: RAM pseudocode should have some more detail, but I expect you could fill it in.

Say the input formula is G.

Brute-Force-Algorithm(G):

Enumerate all truth assignments α . For each α , compute the truth value it gives G. If any of them satisfy G, then ACCEPT, else REJECT.

Say the input length (encoding size) of G is N. Say the # of variables in G is n. (Note: $n \le N$.)

(Although we usually write n for input length, for SAT it's super-traditional to use it for # of variables.)

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Say the input length (encoding size) of G is N. Say the # of variables in G is n. (Note: $n \le N$.) # of truth assignments? 2^n Running time of **Brute-Force**: $\Omega(2^n)$

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Say the input length (encoding size) of G is N. Say the # of variables in G is n. (Note: $n \le N$.) Running time of **Brute-Force**: O(2ⁿ·N) Running time of **Brute-Force**: $\Omega(2^n)$

An unsolved problem in **Computer Science/Mathematics** What is the intrinsic complexity of SAT? **SAT:** Given as input a Boolean formula, decide if it is satisfiable or not.

We saw SAT is decidable in O(2^N·N) time. Is SAT decidable in polynomial O(N^c) time? This is precisely the P versus NP problem!

The **P** versus **NP** problem

Is SAT decidable in polynomial O(N^c) time?

Warning: You won't get the full, glorious perspective on why "P versus NP" is so important until Lectures 17–19

The **P** versus **NP** problem

Is SAT decidable in polynomial O(N^c) time?

Most(?) people believe the answer is NO.

Why is it so hard to prove this?

Polynomial-time algorithms can do so many amazing, surprising things!

Very hard to prove efficient algorithm don't exist.

Boolean formulas as binary trees



Boolean formulas as binary trees Variables at the *leaves*



Boolean formulas as binary trees Variables at the *leaves* Connectives at the *internal nodes* Connectives have *fan-in* 2 (except ¬ has fan-in 1)



Boolean formula conventions

 The "size" of a formula is the # of leaves (which is also # of variable-appearances).



Boolean formula conventions

- The "size" of a formula is the # of leaves (which is also # of variable-appearances).
- Sometimes →, ↔, other connectives allowed.
 Sometimes just ¬, ∧, ∨.
 This is "without (much) loss of generality".
- $((((a \land b) \land c) \land d) \cdots \land z)$ is often written as $(a \land b \land c \land d \land \cdots \land z)$, similarly for \lor .

Doesn't affect "size" but does affect "depth".



$(a \land b \land c \land d)$



"Allowing unlimited fan-in"



Every n-variable formula yields a truth table.

Two different n-variable formulas can have the same truth table.



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Every n-variable formula yields a truth table.

Two different n-variable formulas can have the same truth table.



If two n-variable formulas have the same truth table, we call them **equivalent**.

X	У	(x → y)	X	У	(¬x ∨ y)
0	0	1	0	0	1
0	1	1	0	1	1
1	0	0	1	0	0
1	1	1	1	1	1
X	У	(x ∨ y)	X	У	¬(¬x ∧ ¬y)
X 0	У 0	(x v y) 0	X 0	У 0	¬(¬x ∧ ¬y) 0
X 0 0	У 0 1	(x v y) 0 1	X 0 0	У 0 1	¬(¬x ∧ ¬y) 0 1
X 0 0 1	y 0 1 0	(x v y) 0 1 1	X 0 0 1	У 0 1 0	<pre>¬(¬x ∧ ¬y) 0 1 1</pre>

If two n-variable formulas have the same truth table, we call them **equivalent**.

Boolean functions

We also think of an n-bit truth table as a **Boolean function**, $f: \{0,1\}^n \rightarrow \{0,1\}$.

We think of any formula having that truth table as "computing" that Boolean function.

A Boolean function f : $\{0,1\}^3 \rightarrow \{0,1\}$ can be specified by a truth table. E.g.:

X	У	Z	f(x,y,z)
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	1
1	1	1	1

Or it can be specified by words. E.g.: "f(x,y,z) = 1 iff at least two input bits are 1"

Question:

How many Boolean functions (truth tables) are there on n variables?

Answer: 2^{2ⁿ}

We know each Boolean formula on n variables "computes" one such function.

Question:

Is every Boolean function (truth table) computed by some Boolean formula?

Is every truth table computed by some formula?

x ₁	x_2	Х ₃	x ₄	f
0	0	0	0	0
0	0	0	1	0
0	0	1	0	0
0	0	1	1	0
0	1	0	0	0
0	1	0	1	0
0	1	1	0	0
0	1	1	1	0
1	0	0	0	0
1	0	0	1	0
1	0	1	0	0
1	0	1	1	0
1	1	0	0	0
1	1	0	1	0
1	1	1	0	0
1	1	1	1	1

 $x_1 \land x_2 \land x_3 \land x_4$

Is every truth table computed by some formula?

$$x_1$$
 x_2 x_3 x_4 f00001000100010001000010100101001100011000110010010101001100011010111001110011100

 $\neg x_1 \land \neg x_2 \land \neg x_3 \land \neg x_4$

Is every truth table computed by some formula?

$$X_1$$
 X_2 X_3 X_4 f000000000100001100010000010100011000011100011100100100101100110100110100111000111000111000111000

 $x_1 \land \neg x_2 \land x_3 \land \neg x_4$


 $\neg x_1 \land x_2 \land x_3 \land x_4$



We can similarly do any truth table with exactly one 1.







We have just given a "proof by example" o of:

Theorem:

Every Boolean function (truth table) over n variables can be computed by a formula.

Actually, we missed a case... ...the Boolean function which is always 0. Well, it's computed by $(x_1 \land \neg x_1)$.

Theorem:

Every Boolean function (truth table) over n variables can be computed by a formula.

In fact, by a big \vee of \wedge 's of (possibly negated) variables.



Theorem:

Every Boolean function (truth table) over n variables can be computed by a DNF formula of size $\leq 2^{n} \cdot n$.

Same statement but with a "CNF formula": a big \land of \lor 's of (possibly negated) variables.

Why?? "De Morgan formulas"!



Wait, aren't these circuits?



Yes they are, but *circuits* are **more general** than *formulas*.

Below is a *circuit*, but it's not a *formula*.



What's the difference?

Below is a *circuit*, but it's not a *formula*.



What's the difference? Circuits can have fan-out > 1.

Anatomy of a circuit



Anatomy of a circuit



No "loops" allowed! ("directed acyclic graph") There is (at least) one "evaluation ordering".

G_1 : x (input) G_2 : y (input) G_3 : z (input) G_4 : \land (of G_1 , G_2) G_5 : \neg (of G_4) G_6 : \vee (of G_2 , G_3) G_7 : \land (of G_4 , G_6) Ζ G_8 : \vee (of G_5 , G_7)

Any set of gates can be designated as "output"; if unspecified, the "last" gate is the single output.

Evaluation ordering

Evaluation ordering



G₂: y (input) G₃: z (input) G_4 : \land (of G_1, G_2) G_5 : \neg (of G_4) G_6 : \vee (of G_2 , G_3) G_7 : \land (of G_4 , G_6) $G_8: \vee (of G_5, G_7)$

"Size" of a circuit: # of *non-input* gates. (In this example, 5.)

Circuits as programming languages

This is a great way to specify a circuit. No picture required!

Looks like code in a programming language!

 G_1 : x (input) G₂: y (input) G₃: z (input) G_4 : \land (of G_1 , G_2) G_5 : \neg (of G_4) $G_6: \vee (of G_2, G_3)$ G_7 : \land (of G_4 , G_6) G_8 : \vee (of G_5 , G_7)

Looks like circuit size \approx running time...

Circuits:

Super-simple. Looks like a programming language. Circuit complexity (size) is very concrete. Circuits can compute any Boolean function.

Why didn't we use circuits (instead of Turing Machines) to define computation?!

Good question, we'll come back to that...

Definitional question: What gates are "allowed" in circuits? ∧ with fan-in 2 Almost always allowed: \vee with fan-in 2 - with fan-in 1 0 with fan-in 0 Usually allowed: 1 with fan-in 0 any fan-in 2 gate; e.g., Sometimes allowed: \equiv (equals), \oplus (XOR) with any fan-in Often allowed:

v with **any** fan-in

Doesn't make a big difference, but always ask.

Let's build a circuit for 10-bit PALINDROMES $f: \{0,1\}^{10} \rightarrow \{0,1\}$ $f(x_1,x_2,x_3,x_4,x_5,x_6,x_7,x_8,x_9,x_{10}) = 1$

if and only if input string is same as its reverse

Let's be liberal, allow all gates on previous slide.





What if we only allow fan-in 2 gates?







What if we only allow fan-in 2 gates?







Circuits:

Super-simple. Look like a programming language. Circuit complexity (size) is very concrete. Circuits can compute any Boolean function.

Why didn't we use circuits (instead of Turing Machines) to define computation?!

Inspirational quotation from a famous man:

"An algorithm is a finite answer to an infinite number of questions"



Stephen Kleene

Less Inspirational quote

"Circuits are an **infinite** answer to an infinite number of questions ©"





Consider the language PALINDROMES $\subseteq \{0,1\}^*$ How can we compute it using circuits? Well, for length-10 inputs we had something...



Consider the language PALINDROMES $\subseteq \{0,1\}^*$ How can we compute it using circuits? Well, for length-10 inputs we had something...



Consider the language PALINDROMES $\subseteq \{0,1\}^*$ How can we compute it using circuits? Well, for length-10 inputs we had something... For length-11 inputs we had something else... For length-12 inputs we had something else... For length-0 inputs we had something else... For length-1 inputs we had something else...



This is called a "family of circuits".

It's a fine mathematical concept, but we don't like it to use it to *define* "computation", because it's *infinite*.

(It sort of *begs the question*: In real life, how do you get C_n ? Hopefully, there's an **algorithm** that on input n, outputs C_n ...)



Definition:

- A family of circuits C is an infinite sequence $C_0, C_1, C_2, ...$ where C_n is a circuit with n inputs. We say C decides $L \subseteq \{0,1\}^*$ if for all $n \in \mathbb{N}$, C_n decides $L_n = L \cap \{0,1\}^n$.
- The size of *C* is the function $S : \mathbb{N} \to \mathbb{N}$ defined by $S(n) = size of C_n$.

Example

Let \mathcal{C} be the family of circuits where C_n is...

Then *C* decides the language PALINDROMES and has size O(n); more precisely,

 $S(n) = \begin{cases} n-1 & \text{if } n \text{ is even} \\ n-2 & \text{if } n \text{ is odd} \end{cases}$

*only for $n \ge 4$; special cases for n=0,1,2,3

G ₁ :	x ₁	(input)
G ₂ :	x ₂	(input)
G _n :	x _n	(input)
E ₁ :	=	(of G ₁ , G _n)
E ₂	=	(of G ₂ , G _{n-1})
E ₃ :	=	(of G ₃ , G _{n-2})
E _{ln/2}]:	=	(of $G_{ln/2}$, $G_{n/2+1}$)
A ₁ :	^	(of E ₁ , E ₂)
A ₂ :	^	(of A ₁ , E ₃)
A ₃ :	^	(of A ₂ , E ₄)
A _{ln/2J-1} :	^	(of $A_{ln/2J-1}$, $E_{ln/2J}$)

Recall: Every n-bit Boolean function computable by a formula/circuit of size O(2ⁿ·n).

> (I don't mean to alarm you, but this includes HALT!!)

Consequence: **Every** language is computed by a family of circuits of size O(2ⁿ·n). Recall: Every n-bit Boolean function computable by a formula/circuit of size O(2ⁿ·n).

Easy improvement: **Every** language is computed by a family of circuits of size O(2ⁿ). Recall: Every n-bit Boolean function computable by a formula/circuit of size O(2ⁿ·n).

Slightly trickier improvement: **Every** language is computed by a family of circuits of size O(2ⁿ/n).

> Proved by the great Claude Shannon in 1949.


Theorem:

Suppose there is a TM deciding L in time T(n). Then it can be converted into a circuit family deciding L with size $S(n) = O(T(n)^2)$.

If you like a challenge, try to prove this yourself.

We will need and use this when studying "NP-hardness".

Theorem:

Suppose there is a TM deciding L in time T(n). Then it can be converted into a circuit family deciding L with size $S(n) = O(T(n)^2)$.

Corollary:

Any L solvable in polynomial time on TMs (or in RAM model) has polynomial-size circuits.

Corollary:

If you want to show some L is **not** solvable in polynomial time, suffices to show it is **not** solvable by polynomial-size circuit families.

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If you want to show some L is **not** solvable in polynomial time, suffices to show it is **not** solvable by polynomial-size circuit families.

In the '80s, this was viewed as the approach that would solve **P** versus **NP**.

"Just" have to show that SAT doesn't have polynomial-size circuit families. Shannon's Theorem 1: **Every** n-bit Boolean function has an $\Lambda/\sqrt{\neg}$ circuit of size O(2ⁿ/n)



Shannon's Theorem 2:
 Almost every n-bit Boolean function
 requires a circuit of size Ω(2ⁿ/n)

(even when all fan-in 2 gates are allowed)

"Essentially all computational problems require exponential circuit complexity." Shannon's Theorem 2:
Almost every n-bit Boolean function requires a circuit of size Ω(2ⁿ/n).

Proof:

Let $s = (1/4) 2^n/n$. We'll show: There are $\leq (1.5)^{2^n}$ circuits of size s. But there are **way** more n-bit Boolean functions: 2^{2^n} .

Think of the "programming language" form of a size-s circuit. After the n input gates, we have s more lines. Each defined by a gate type (16 choices) and two previous lines (\leq n+s choices). So there are at most [16 · (n+s) · (n+s)]^S possible circuits.

The […] quantity is $\leq 64s^2$ because $n+s \leq 2s$, and $64s^2 \leq (2^n)^2$. So there are at most $[(2^n)^2]^s = 2^{2ns} = 2^{(1/2)2^n} = (1.41...)^{2^n}$ circuits. Shannon's Theorem 2:
Almost every n-bit Boolean function requires a circuit of size Ω(2ⁿ/n).

"Essentially all computational problems require exponential circuit complexity."

So... what's an example of one?

If **SAT** is an example, we resolve **P** versus **NP**!

Or... can we just find **any** explicit example?!

Challenge: Find an explicit n-bit function requiring large circuit size.

Shannon: Practically **all** functions need $\Omega(2^n/n)$.

1965: Kloss & Malyshev show a certain simple function requires size ≥ 2n - 3

1977: Paul & Stockmeyer show certain simple functions requires size $\geq 2.5n - 1.5$

1984: N. Blum showed a certain pretty simply function requires size ≥ 3n - 3



Shannon: Practically **all** functions **20 trillion**

1965: Kloss & Malyshev show a certain simple function requires size ≥ 2n - 3

1977: Paul & Stockmeyer show certain simple functions requires size $\geq 2.5n - 1.5$

1984: N. Blum showed a certain pretty simple function requires size \geq 147

1965: Kloss & Malyshev show a certain simple function requires size ≥ 2n -3
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functions requires size $\geq 2.5n - 1.5$

1984: N. Blum showed a certain pretty simple function requires size ≥ 3n - 3

Great news!!

Last year: Find, Golevnev, Hirsch, Kulikov showed a certain function requires size $\geq (3+1/86)n - O(n^{.8})$ This pretty much sums up where we are on **P** versus **NP**.

Study Guide



Definitions: Boolean formulas Truth tables Boolean functions The SAT problem Circuits Circuit familes & size

Theorems:

Every function can be computed by a DNF Almost every function requires circuits of size $\Omega(2^n/n)$.