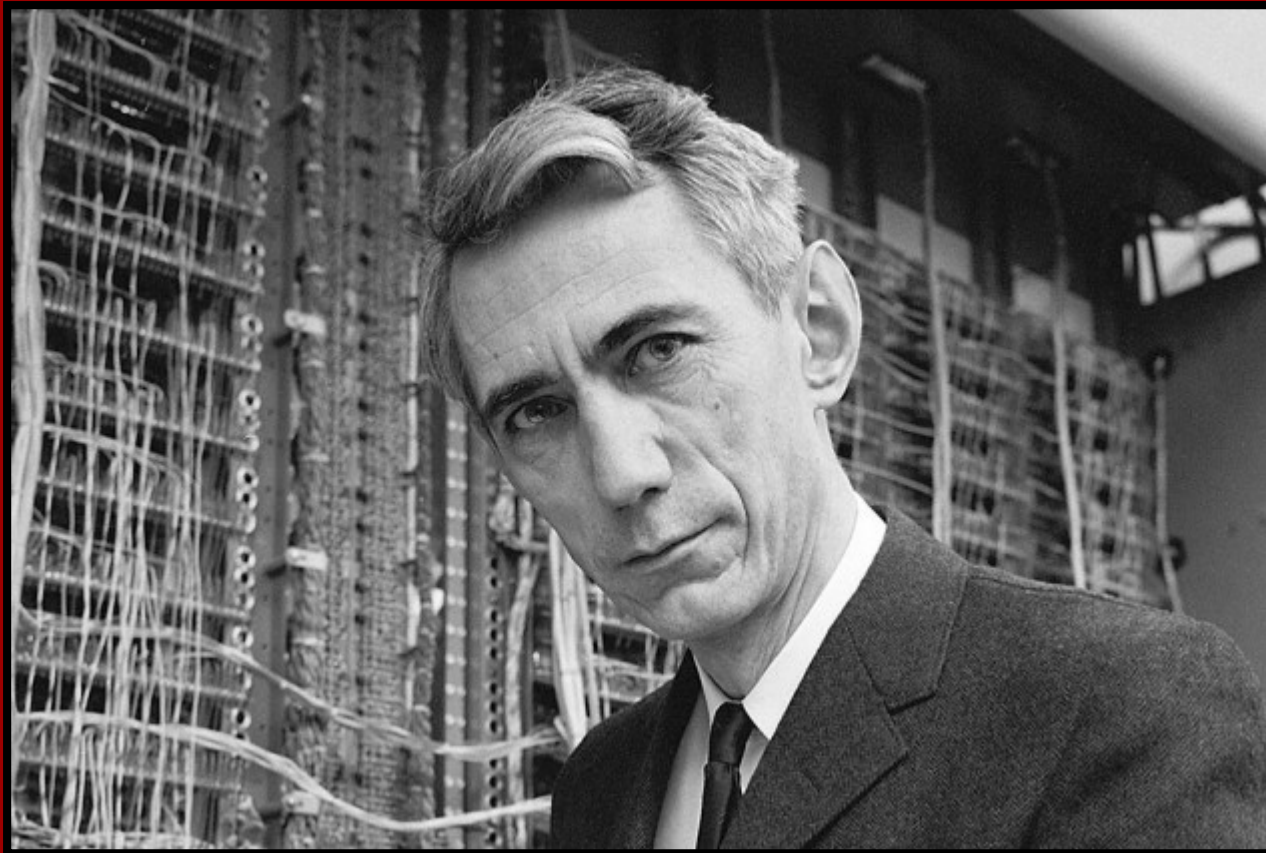


15-251: Great Theoretical Ideas in Computer Science
Fall 2018, Lecture 14

Boolean Formulas and Circuits

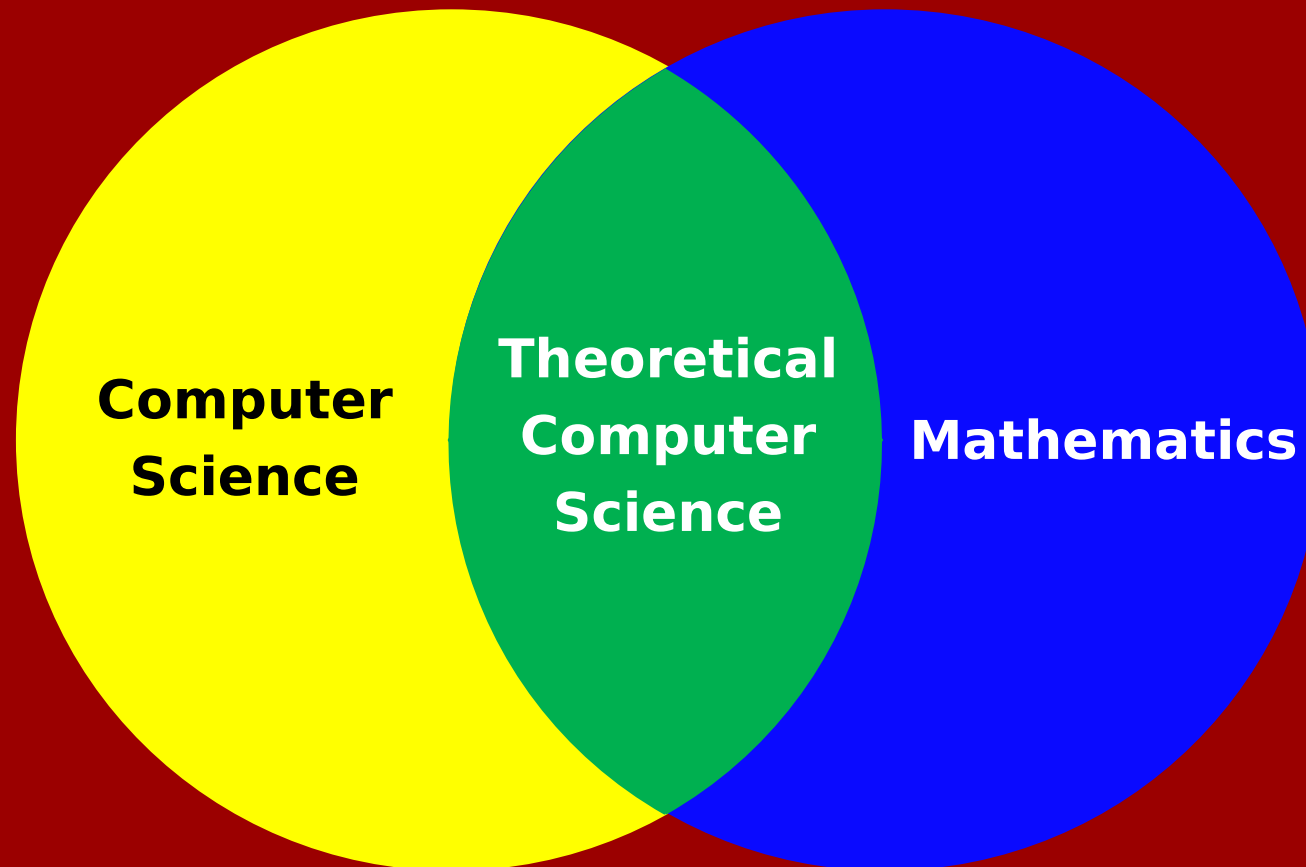


Today

- 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



P versus NP

The most famous unsolved problem in
Theoretical Computer Science

One of the most famous unsolved problems in
all of Computer Science, all of Mathematics



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List of unsolved problems in computer science

From Wikipedia, the free encyclopedia

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.

Contents [\[hide\]](#)

- [1 Computational complexity](#)
- [2 Algorithms](#)
- [3 Programming language theory](#)
- [4 Other problems](#)
- [5 External links](#)

Computational complexity [\[edit\]](#)

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



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List of unsolved problems in mathematics

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- 1 Lists of unsolved problems in mathematics
 - 1.1 Millennium Prize Problems
- 2 Other still-unsolved problems
 - 2.1 Additive number theory
 - 2.2 Algebra
 - 2.3 Algebraic geometry
 - 2.4 Algebraic number theory
 - 2.5 Analysis



Millennium Prize Problems [\[edit\]](#)

Of the original seven [Millennium Prize Problems](#) set by the [Clay Mathematics Institute](#), six have yet to be solved, as of August 2015:^[7]

- **P versus NP**
- [Hodge conjecture](#)
- [Riemann hypothesis](#)
- [Yang–Mills existence and mass gap](#)
- [Navier–Stokes existence and smoothness](#)
- [Birch and Swinnerton-Dyer conjecture](#)

The seventh problem, the [Poincaré conjecture](#), has been solved.^[8] The [smooth four-dimensional Poincaré conjecture](#)—that is, whether a four-dimensional topological sphere can have two or more inequivalent [smooth structures](#)—is still unsolved.^[9]

P versus **NP**

The most famous unsolved problem in
Theoretical Computer Science

One of the most famous unsolved problems in
all of Computer Science, 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 15–16

Boolean formulas

You've seen these before in Concepts.

Stuff like this:

$$((\neg x \rightarrow y) \wedge ((x \vee z) \leftrightarrow y))$$

$x, y, z, \dots$ Boolean variables, values 0/1 (or T/F)

$\neg, \wedge, \vee, \dots$ Boolean connectives (or operations)

A	B	$\neg A$	$(A \wedge B)$	$(A \vee B)$	$(A \rightarrow B)$	$(A \leftrightarrow B)$
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:

$$((\neg x \rightarrow y) \wedge ((x \vee z) \leftrightarrow y))$$

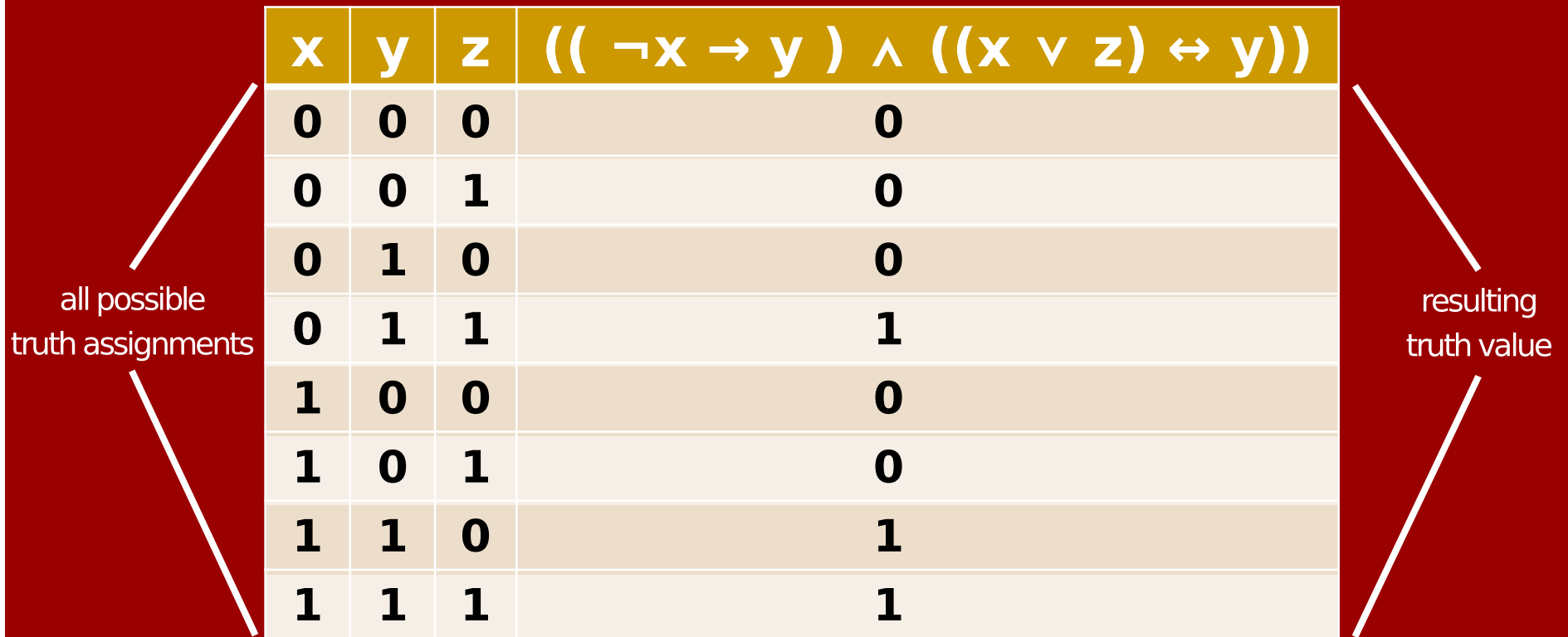
$x, y, z, \dots$ Boolean variables, values 0/1 (or T/F)

$\neg, \wedge, \vee, \dots$ 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



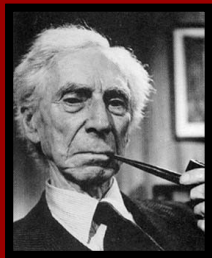
x	y	z	$((\neg x \rightarrow y) \wedge ((x \vee z) \leftrightarrow y))$
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	0
1	1	0	1
1	1	1	1

Satisfiable: At least one 1 in truth table

Unsatisfiable: No 1's in truth table

Tautology: All 1's in truth table

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



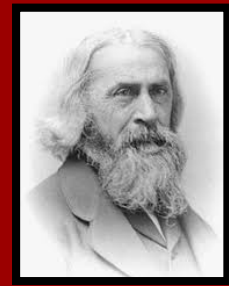
Russell?



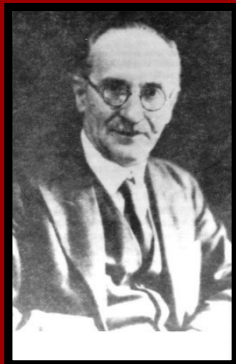
Wittgenstein?



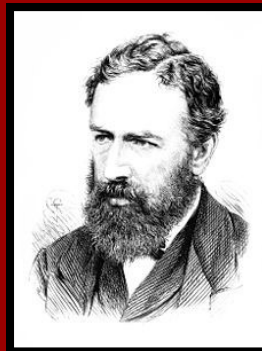
Post?



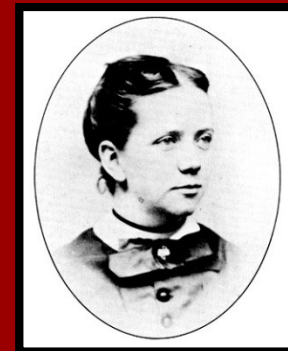
Peirce?



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.

SAT is decidable

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.

SAT is decidable

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: $n \leq N$.)

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

SAT is decidable

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: $n \leq N$.)

of truth assignments? 2^n

$\therefore$ running time of **Brute-Force**: $\Omega(2^n)$

SAT is decidable

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: $n \leq N$.)

Running time of **Brute-Force**: $O(2^n \cdot N)$

$\therefore$ 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 $O(2^N \cdot N)$ 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 15–16

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

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

Most(?) people believe the answer is NO.

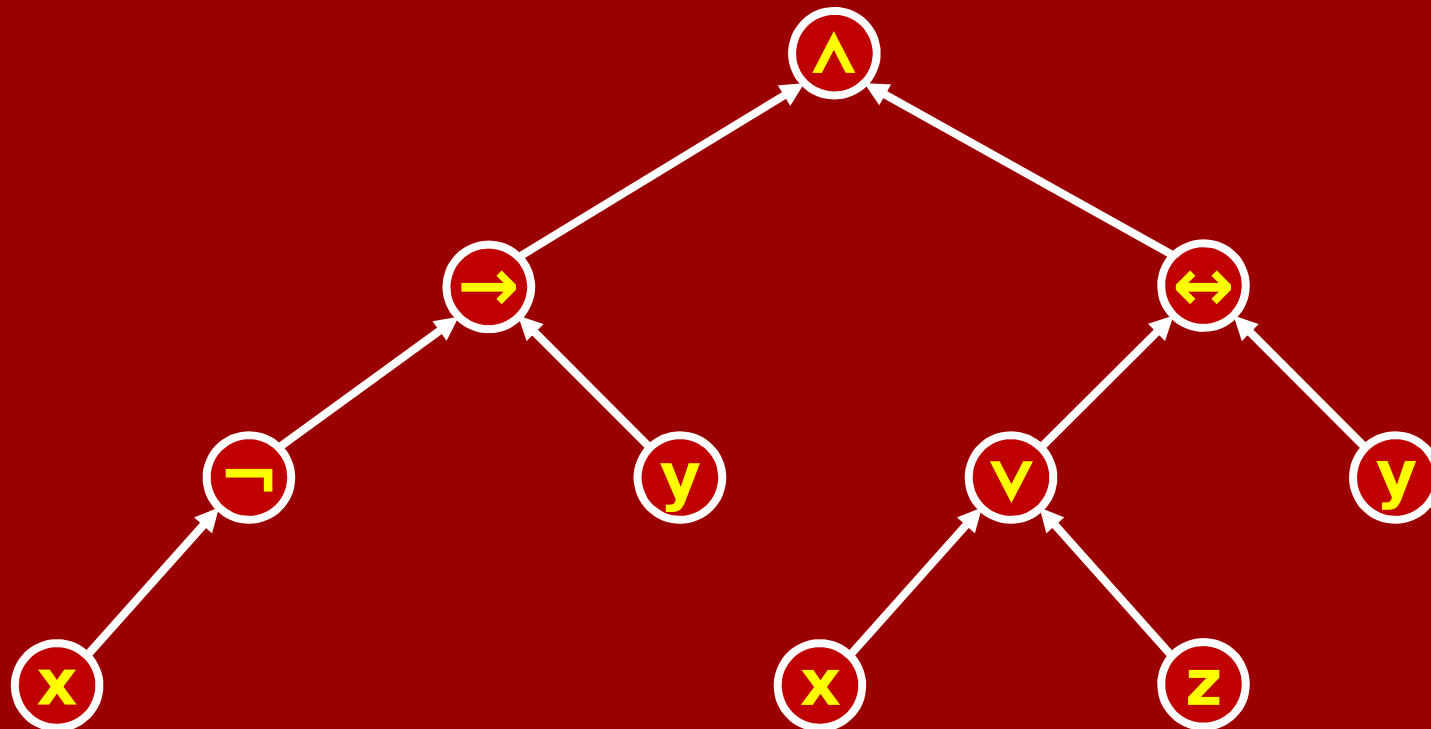
Why is it so hard to prove this?

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

Very hard to prove efficient algorithm don't exist.

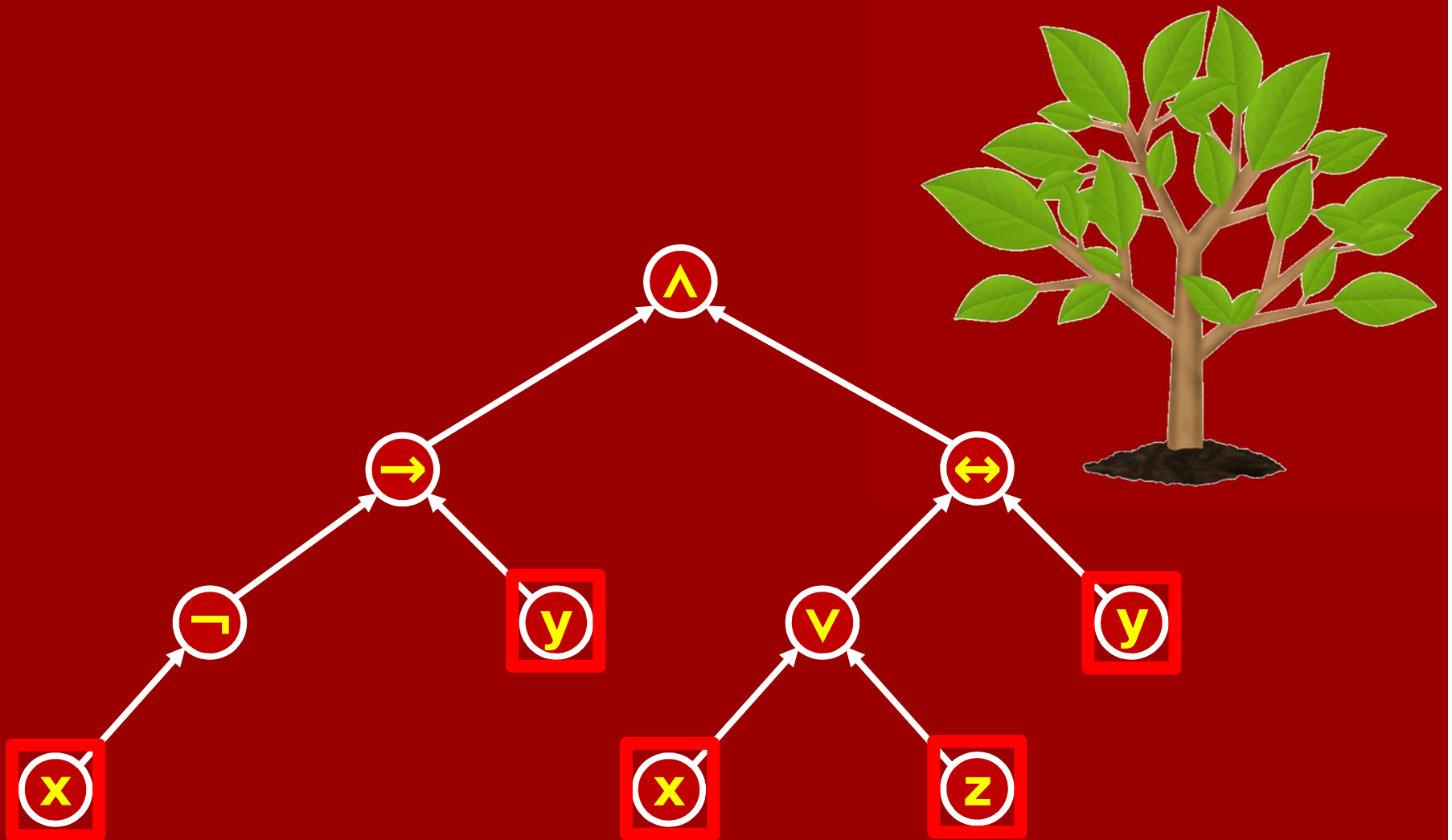
Boolean formulas as binary trees

$$((\neg x \rightarrow y) \wedge ((x \vee z) \leftrightarrow y))$$



Boolean formulas as binary trees

Variables at the leaves

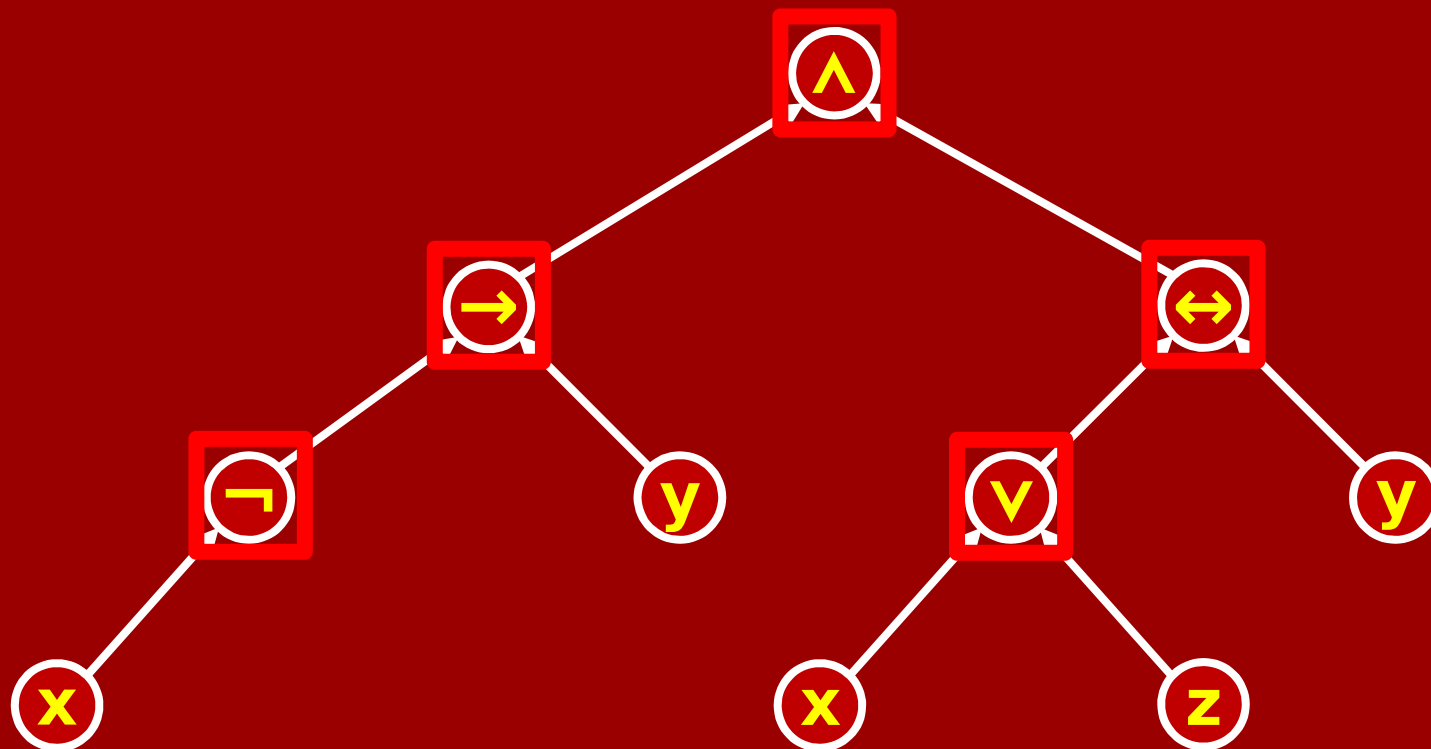


Boolean formulas as binary trees

Variables at the leaves

Connectives at the internal nodes

Connectives have fan-in 2 (except $\neg$ has fan-in 1)

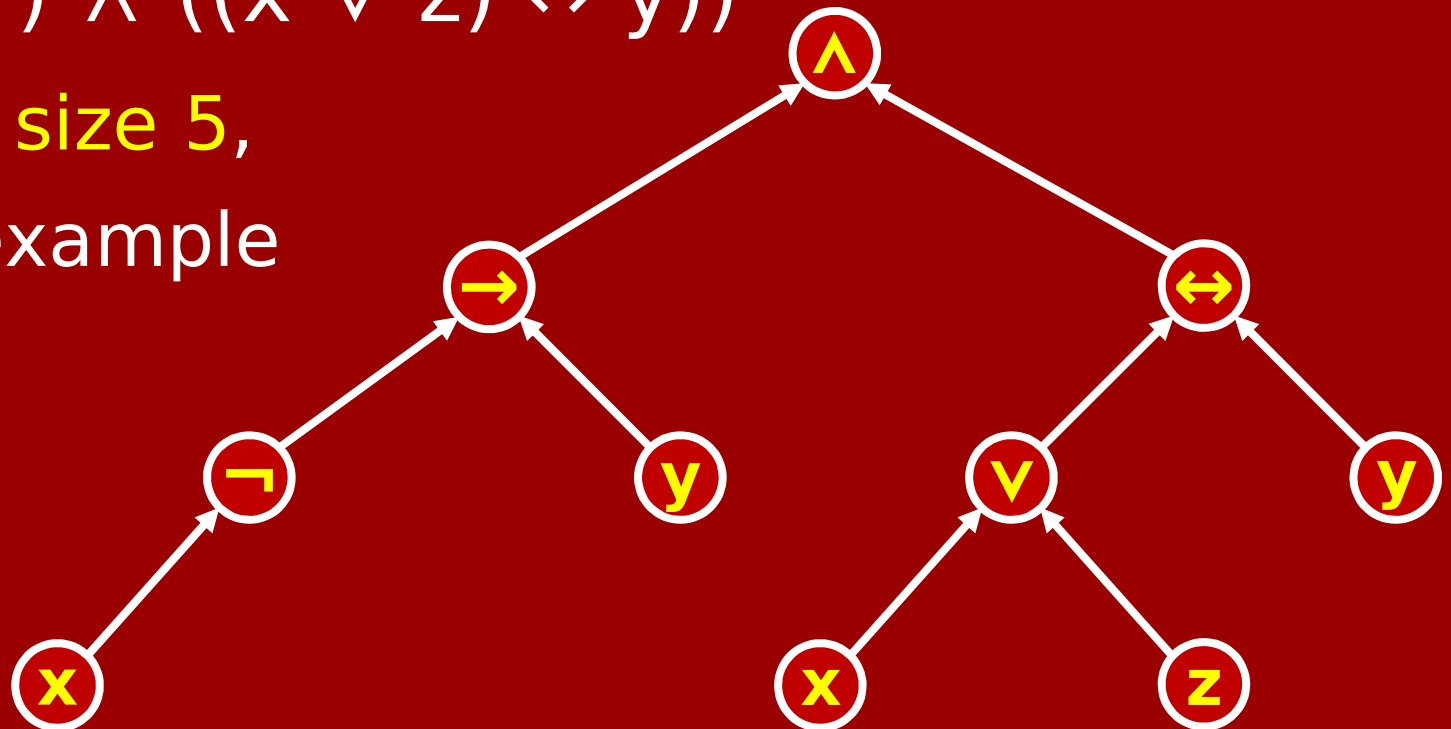


Boolean formula conventions

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

$((\neg x \rightarrow y) \wedge ((x \vee z) \leftrightarrow y))$

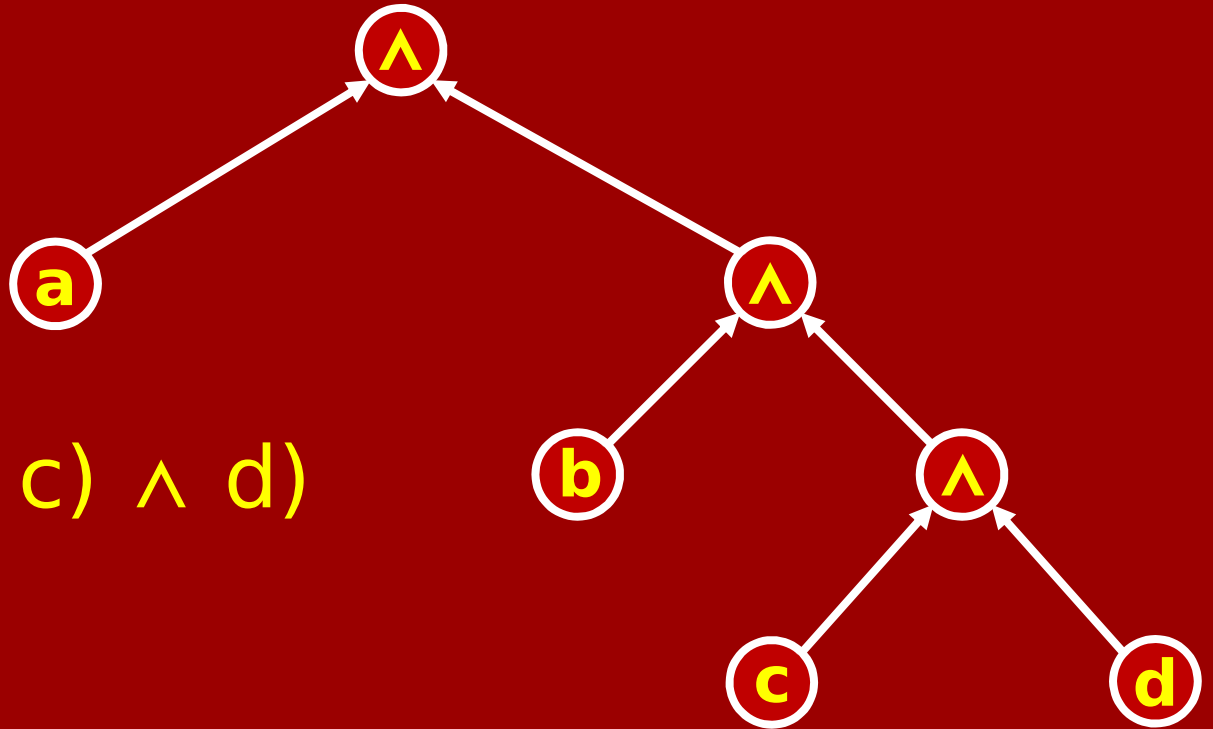
has size 5,
for example



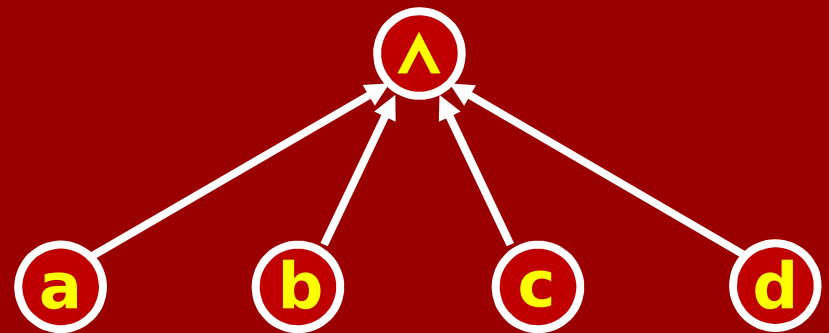
Boolean formula conventions

- The “size” of a formula is the # of leaves (which is also # of variable-appearances).
- Sometimes $\rightarrow$, $\leftrightarrow$, other connectives allowed. Sometimes just $\neg$, $\wedge$, $\vee$. (“De Morgan formulas”) This is “without (much) loss of generality”.
- $((((a \wedge b) \wedge c) \wedge d) \cdots \wedge z)$ is often written as $(a \wedge b \wedge c \wedge d \wedge \cdots \wedge z)$, similarly for $\vee$. Doesn't affect “size” but does affect “depth”.

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

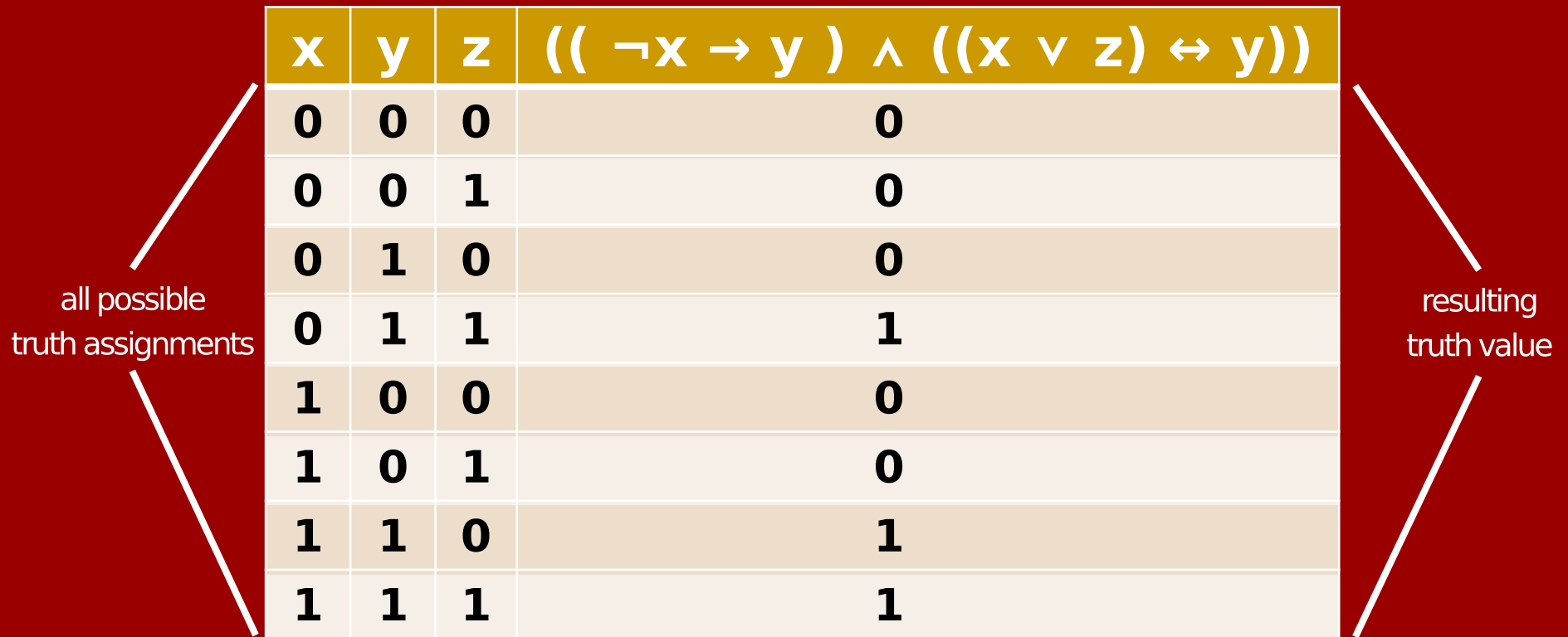


$(a \wedge b \wedge c \wedge d)$



“Allowing unlimited fan-in”

More on truth tables



x	y	z	$((\neg x \rightarrow y) \wedge ((x \vee z) \leftrightarrow y))$
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	0
1	1	0	1
1	1	1	1

Every n-variable formula yields a truth table.

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

More on truth tables

x	y	z	$(x \wedge y) \vee (y \wedge z)$
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	0
1	1	0	1
1	1	1	1

Every n-variable formula yields a truth table.

Two different n-variable formulas
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More on truth tables

x	y	z	$(y \wedge (x \vee z))$
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	0
1	1	0	1
1	1	1	1

Every n-variable formula yields a truth table.

Two different n-variable formulas
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More on truth tables

x	y	z	$(y \wedge (x \vee z))$
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	0
1	1	0	1
1	1	1	1

all possible truth assignments

resulting truth value

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

More on truth tables

x	y	$(x \rightarrow y)$
0	0	1
0	1	1
1	0	0
1	1	1

$\equiv$

x	y	$(\neg x \vee y)$
0	0	1
0	1	1
1	0	0
1	1	1

x	y	$(x \vee y)$
0	0	0
0	1	1
1	0	1
1	1	1

$\equiv$

x	y	$\neg(\neg x \wedge \neg y)$
0	0	0
0	1	1
1	0	1
1	1	1

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	y	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^{2^n}

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	x_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

$$x_1 \wedge x_2 \wedge x_3 \wedge x_4$$

Is every truth table computed by some formula?

x_1	x_2	x_3	x_4	f
0	0	0	0	1
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	0

$$\neg x_1 \wedge \neg x_2 \wedge \neg x_3 \wedge \neg x_4$$

Is every truth table computed by some formula?

x_1	x_2	x_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	1
1	0	1	1	0
1	1	0	0	0
1	1	0	1	0
1	1	1	0	0
1	1	1	1	0

$$x_1 \wedge \neg x_2 \wedge x_3 \wedge \neg x_4$$

Is every truth table computed by some formula?

x_1	x_2	x_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	1
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	0

$$\neg x_1 \wedge x_2 \wedge x_3 \wedge x_4$$

Is every truth table computed by some formula?

x_1	x_2	x_3	x_4	f
0	0	0	0	
0	0	0	1	
0	0	1	0	
0	0	1	1	
0	1	0	0	
0	1	0	1	
0	1	1	0	
0	1	1	1	
1	0	0	0	
1	0	0	1	
1	0	1	0	
1	0	1	1	
1	1	0	0	
1	1	0	1	
1	1	1	0	
1	1	1	1	

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

Is every truth table computed by some formula?

x_1	x_2	x_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	1
1	0	0	0	0
1	0	0	1	0
1	0	1	0	1
1	0	1	1	0
1	1	0	0	0
1	1	0	1	0
1	1	1	0	0
1	1	1	1	0

What if there
are **two** 1's?

← $(\neg x_1 \wedge x_2 \wedge x_3 \wedge x_4)$

$\vee$

← $(x_1 \wedge \neg x_2 \wedge x_3 \wedge \neg x_4)$

Is every truth table computed by some formula?

x_1	x_2	x_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	1
1	0	0	0	0
1	0	0	1	0
1	0	1	0	1
1	0	1	1	0
1	1	0	0	0
1	1	0	1	0
1	1	1	0	0
1	1	1	1	0

What if there
are **three** 1's?

← $(\neg x_1 \wedge x_2 \wedge x_3 \wedge x_4)$

$\vee$

← $(x_1 \wedge \neg x_2 \wedge x_3 \wedge \neg x_4)$

Is every truth table computed by some formula?

x_1	x_2	x_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	1
1	0	0	0	0
1	0	0	1	0
1	0	1	0	1
1	0	1	1	0
1	1	0	0	0
1	1	0	1	0
1	1	1	0	1
1	1	1	1	0

What if there
are **three** 1's?

← $(\neg x_1 \wedge x_2 \wedge x_3 \wedge x_4)$

$\vee$

← $(x_1 \wedge \neg x_2 \wedge x_3 \wedge \neg x_4)$

$\vee$

← $(x_1 \wedge x_2 \wedge x_3 \wedge \neg x_4)$

We have just done “proof by example” 😊
for the following result (proper proof in Notes):

Theorem:

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

Actually, we missed a case...

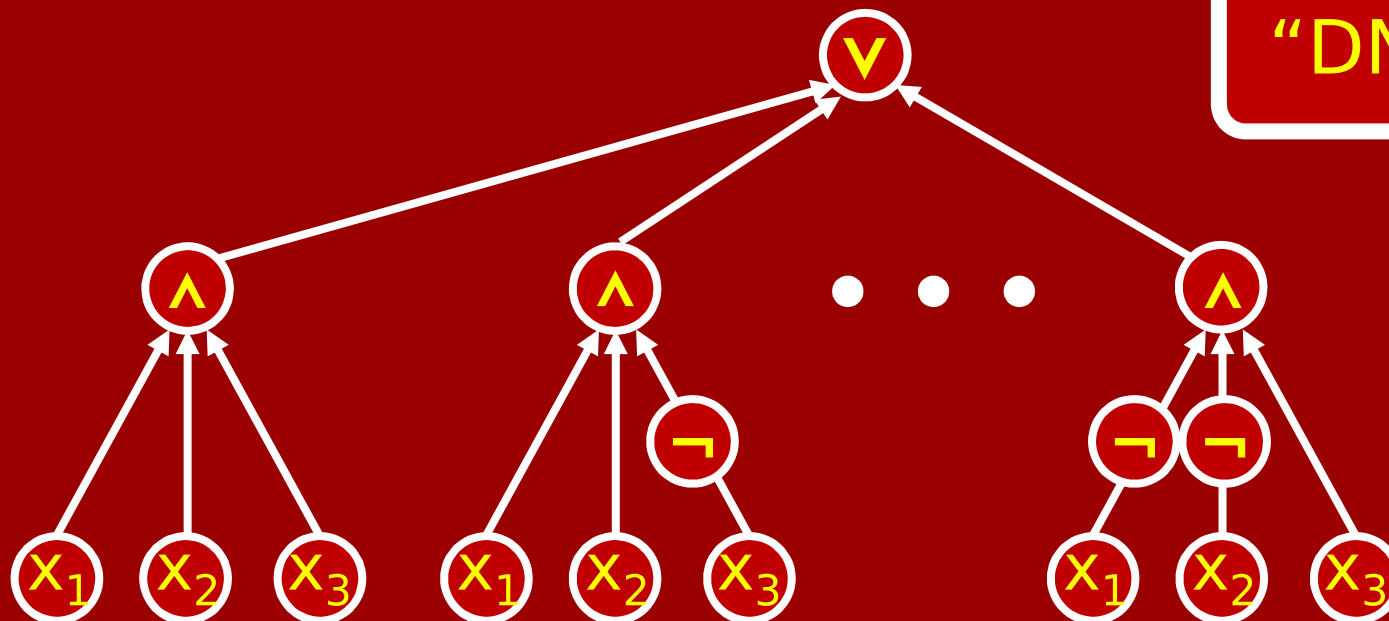
...the Boolean function which is always 0.

Well, it's computed by $(x_1 \wedge \neg x_1)$.

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.



“DNF formula”

Size $\leq \underline{2^n \cdot n}$

Theorem:

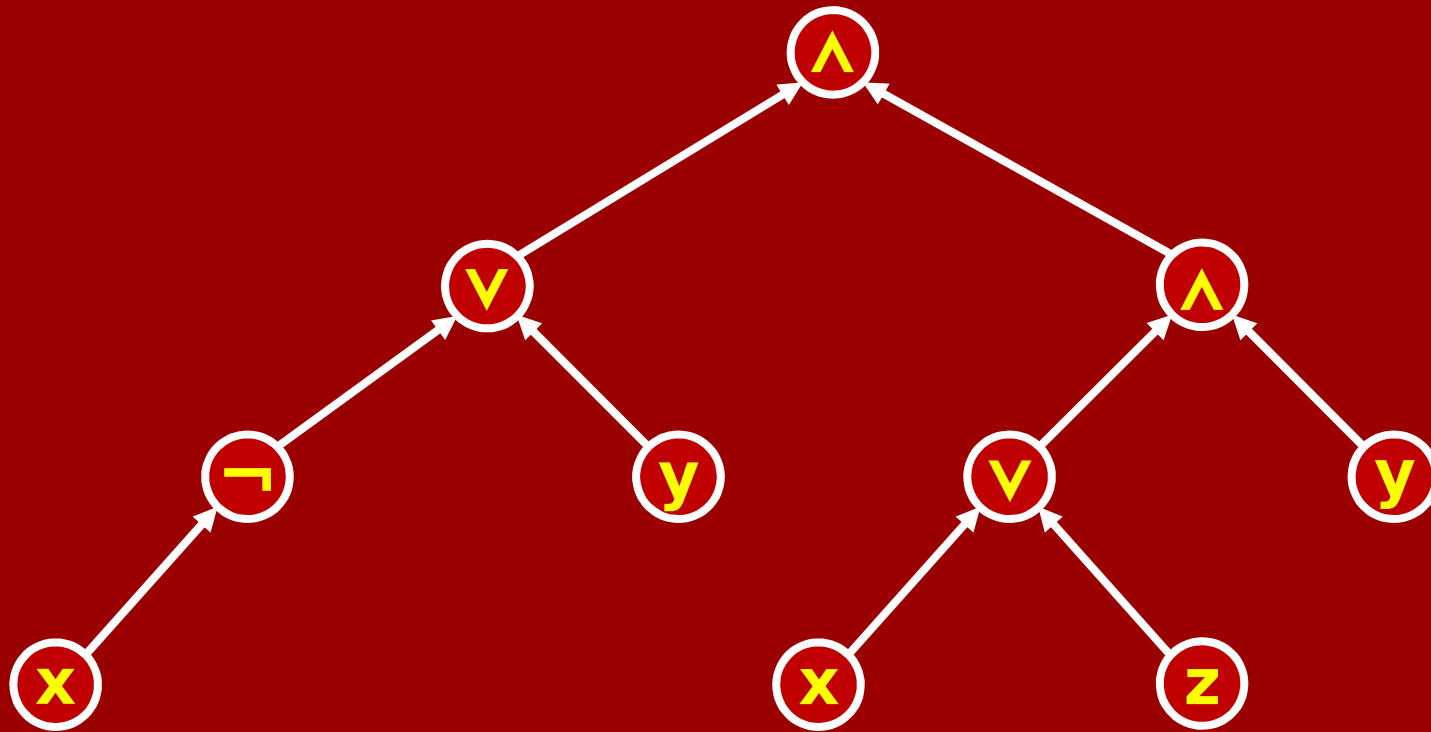
Every Boolean function (truth table) over n variables can be computed by a DNF formula of size $\leq 2^n \cdot n$.

Exercise:

Same statement but with a “CNF formula”:
a big $\wedge$ of $\vee$'s of (possibly negated) variables.

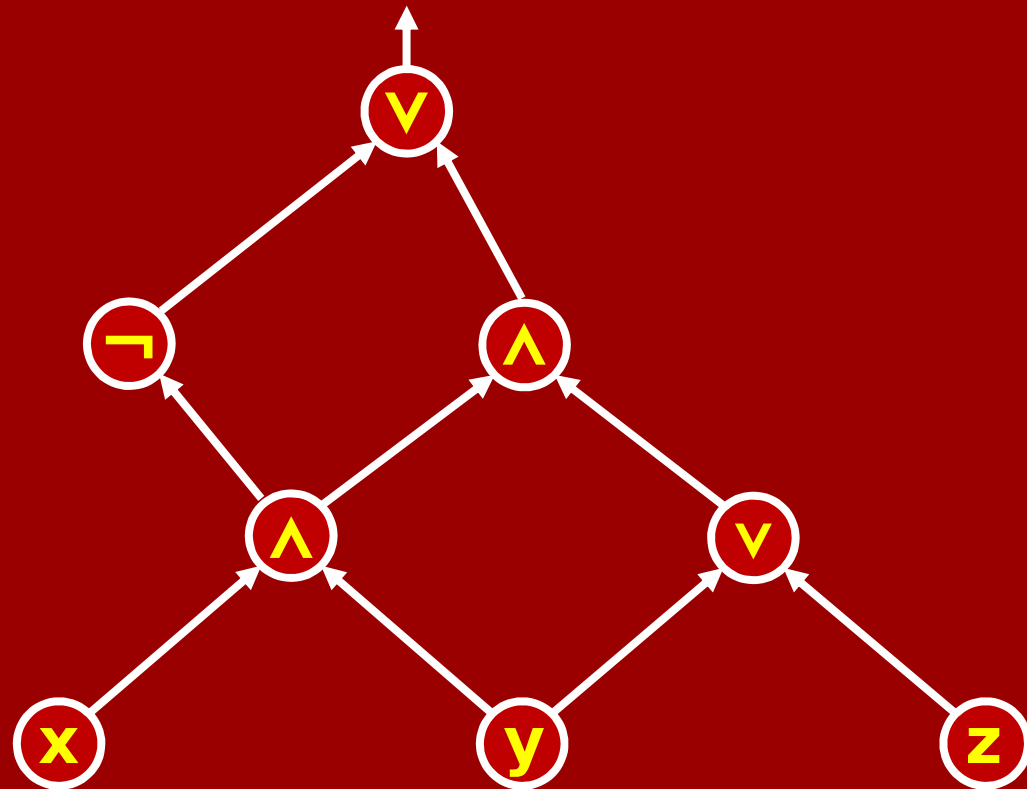
Circuits

Wait, aren't these formulas?



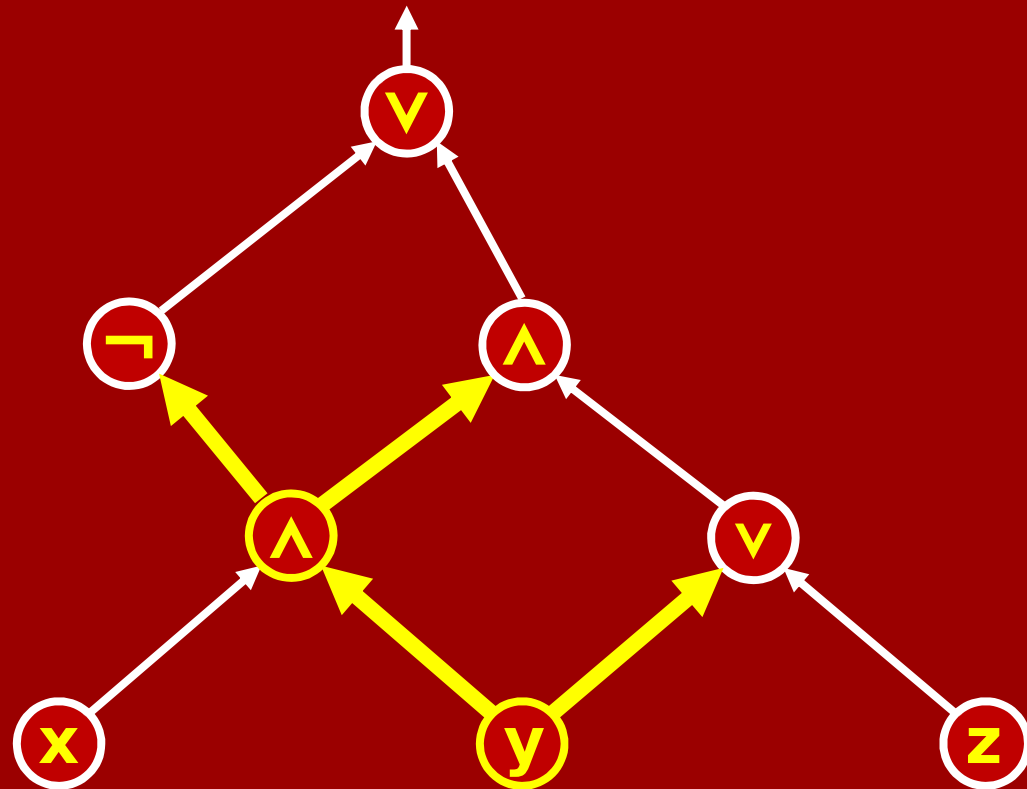
Yes they are, but circuits are
more general than formulas.

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



What's the difference?

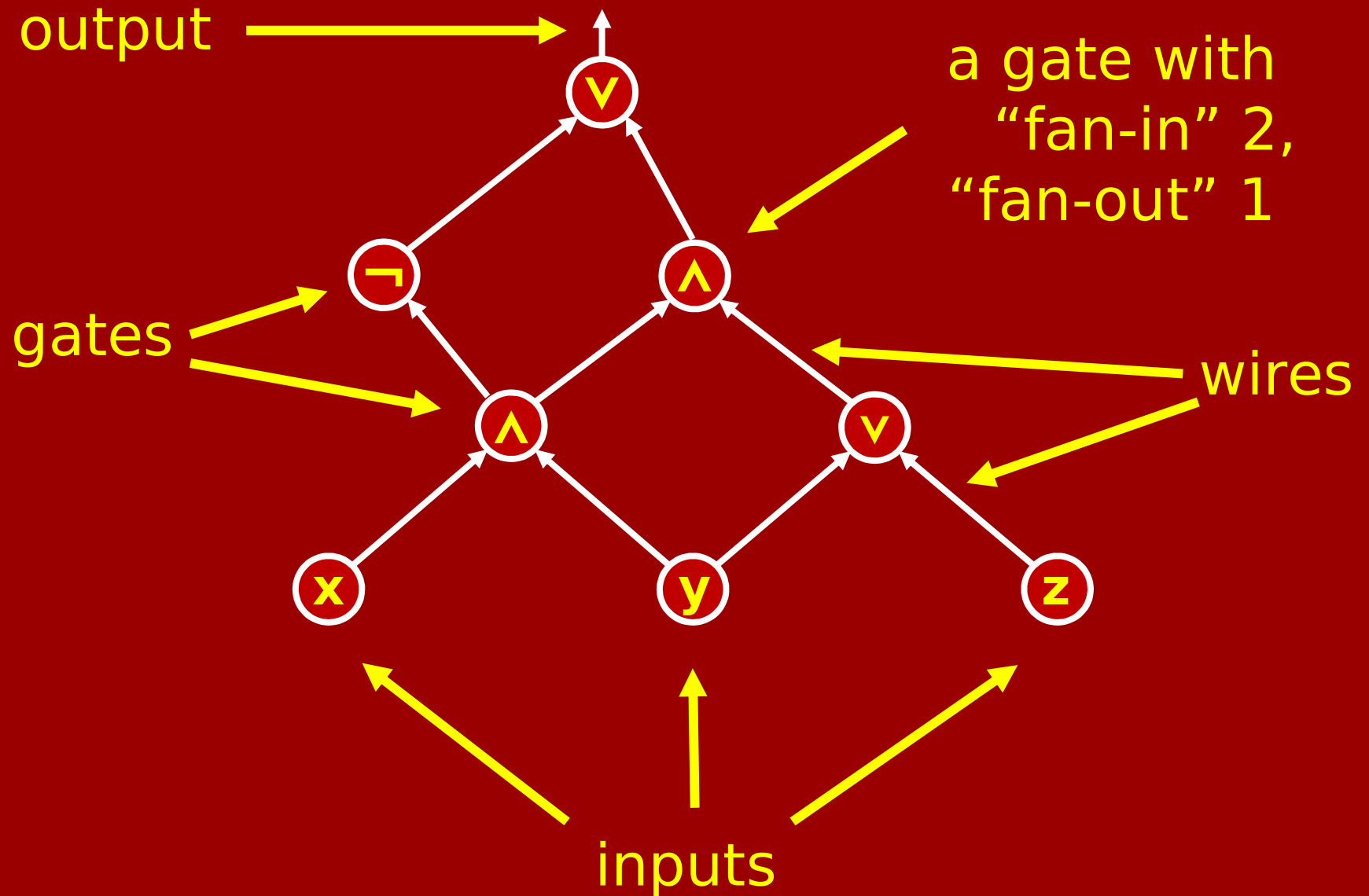
Below is a circuit, but it's not a formula.



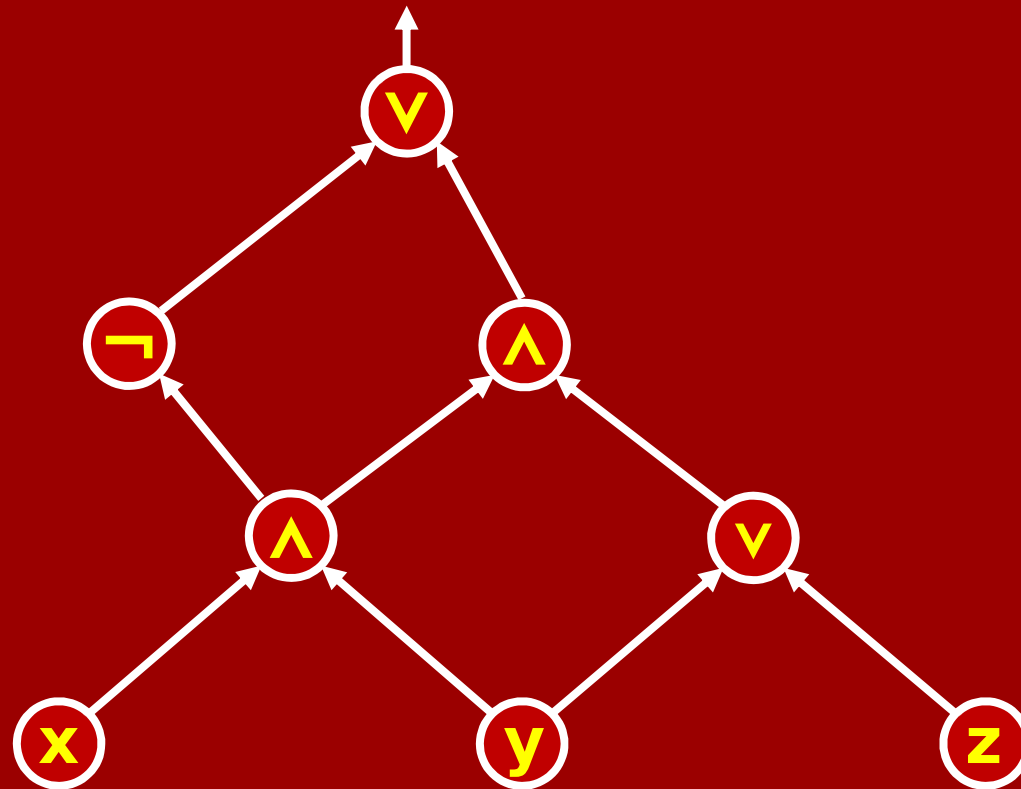
What's the difference?

Circuits can have **fan-out** > 1.

Anatomy of a circuit



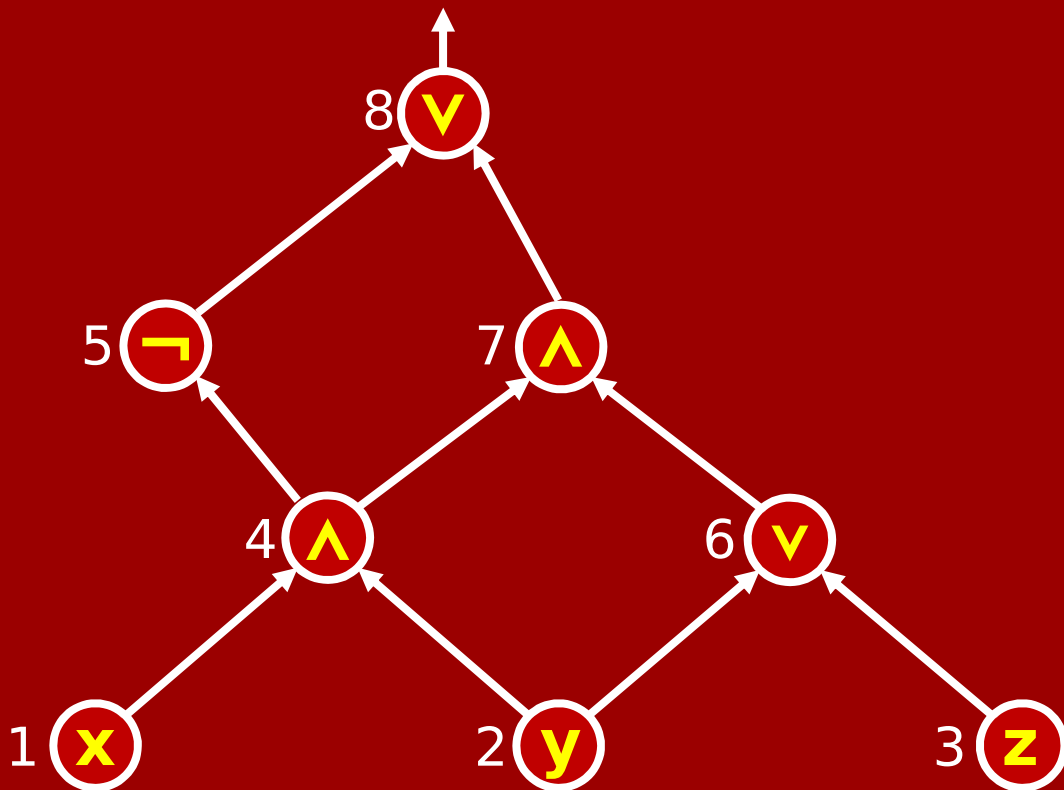
Anatomy of a circuit



No “loops” allowed! (“directed acyclic graph”)

There is (at least) one “evaluation ordering”.

Evaluation ordering



G_1 : x (input)

G_2 : y (input)

G_3 : z (input)

G_4 : $\wedge$ (of G_1 , G_2)

G_5 : $\neg$ (of G_4)

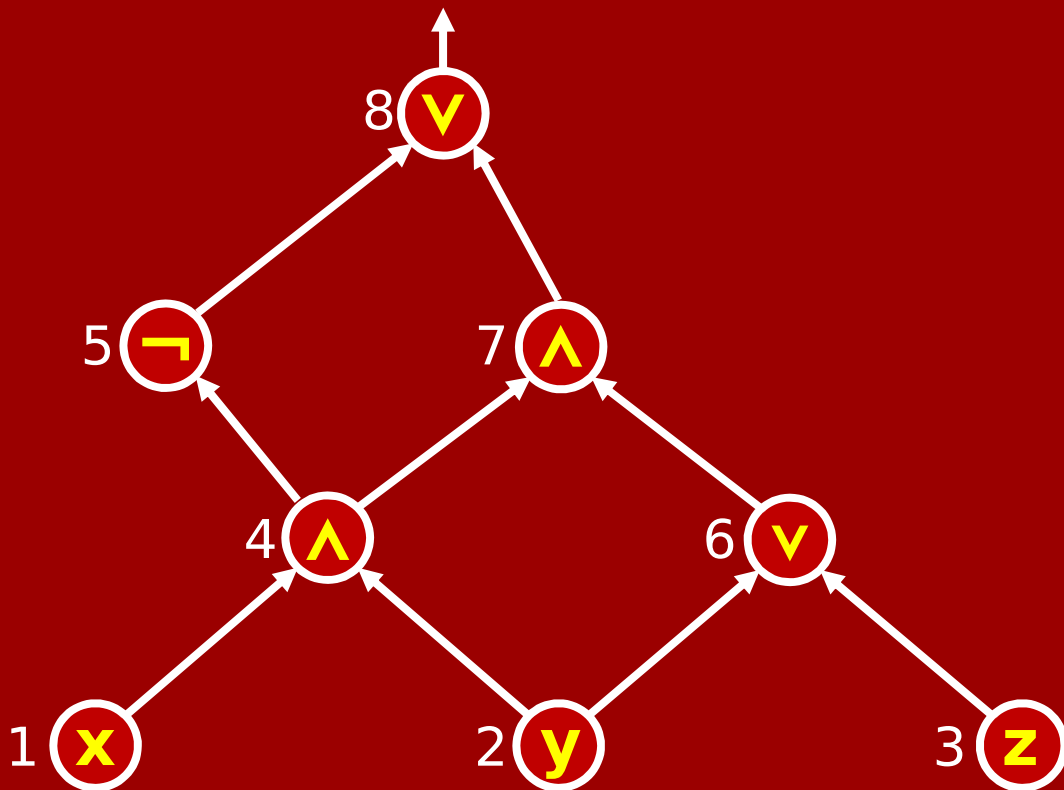
G_6 : $\vee$ (of G_2 , G_3)

G_7 : $\wedge$ (of G_4 , G_6)

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



G_1 : x (input)

G_2 : y (input)

G_3 : z (input)

G_4 : $\wedge$ (of G_1 , G_2)

G_5 : $\neg$ (of G_4)

G_6 : $\vee$ (of G_2 , G_3)

G_7 : $\wedge$ (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)

G_2 : y (input)

G_3 : z (input)

G_4 : $\wedge$ (of G_1, G_2)

G_5 : $\neg$ (of G_4)

G_6 : $\vee$ (of G_2, G_3)

G_7 : $\wedge$ (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?

Almost always allowed:

- $\wedge$ with fan-in 2
- $\vee$ with fan-in 2
- $\neg$ with fan-in 1

Usually allowed:

- 0 with fan-in 0
- 1 with fan-in 0

Sometimes allowed:

- any fan-in 2 gate; e.g., $\equiv$ (equals), $\oplus$ (XOR)

Often allowed:

- $\wedge$ with **any** fan-in
- $\vee$ with **any** fan-in

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

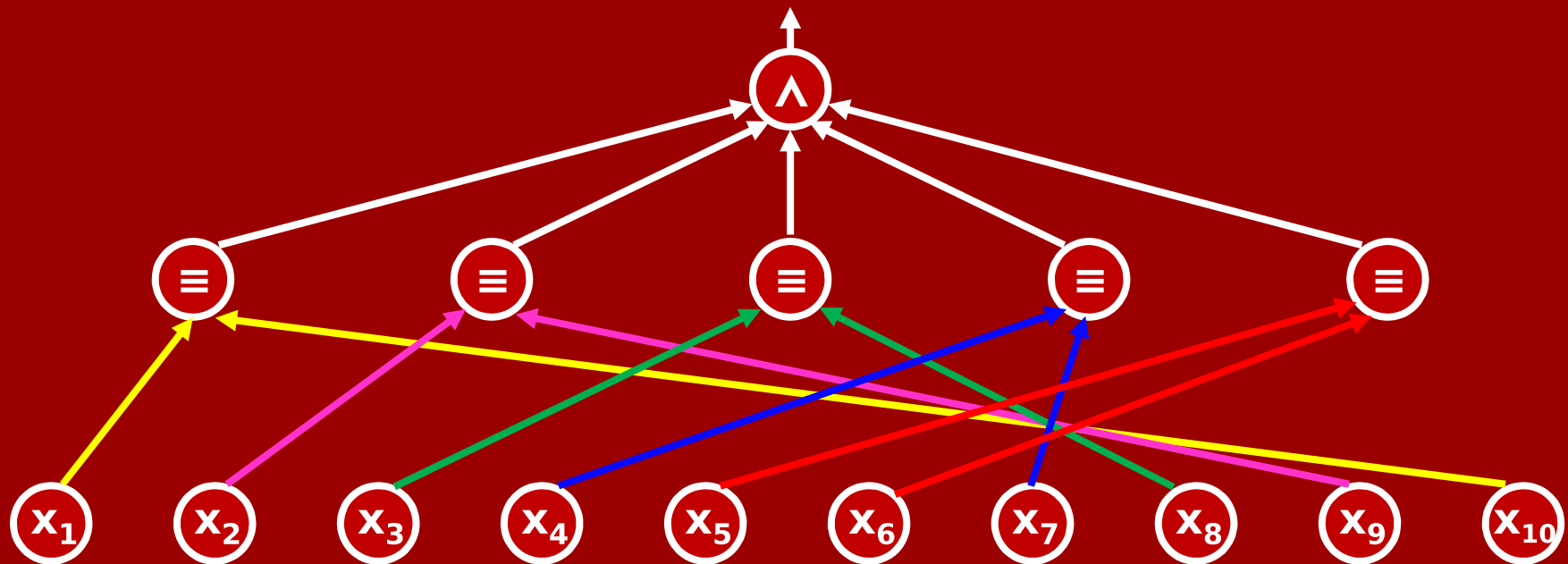
Let's build a circuit for 10-bit PALINDROMES

$$f : \{0,1\}^{10} \rightarrow \{0,1\}$$

$$f(x_1, x_2, x_3, x_4, x_5, x_6, x_7, x_8, x_9, x_{10}) = 1$$

if and only if input string is same as its reverse

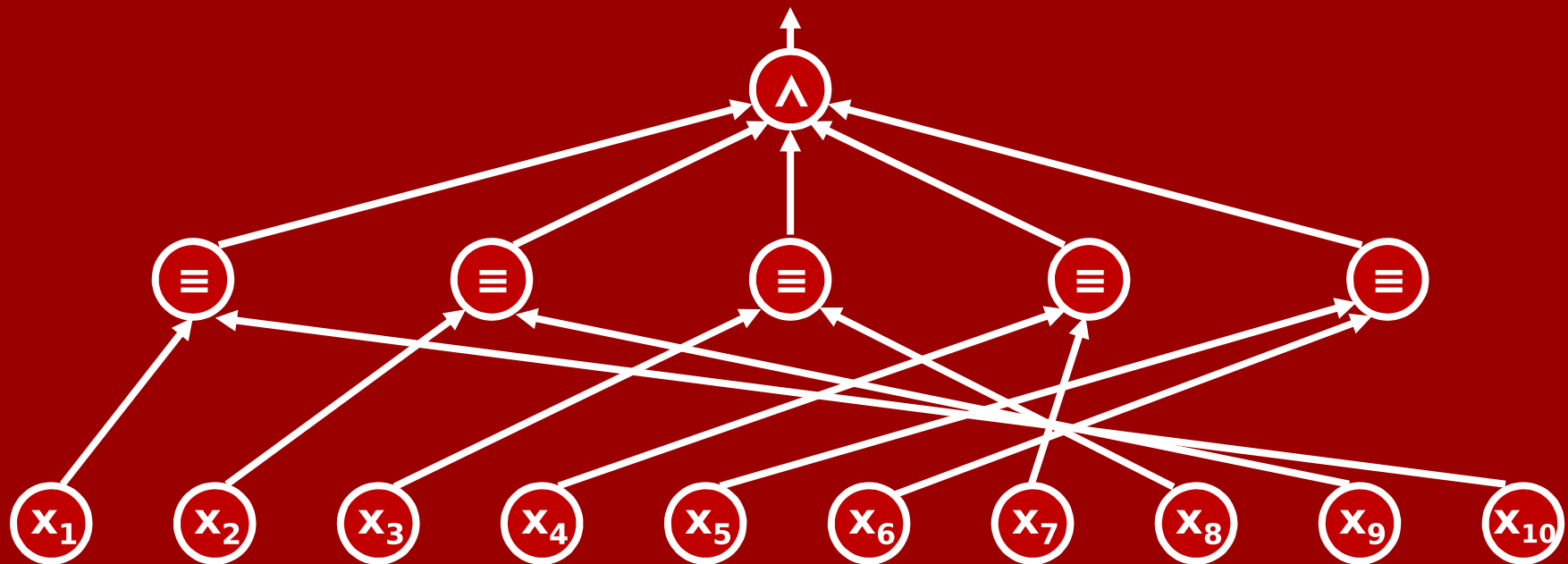
Let's be liberal, allow all gates on previous slide.



Size? 6

(Depth? 2)

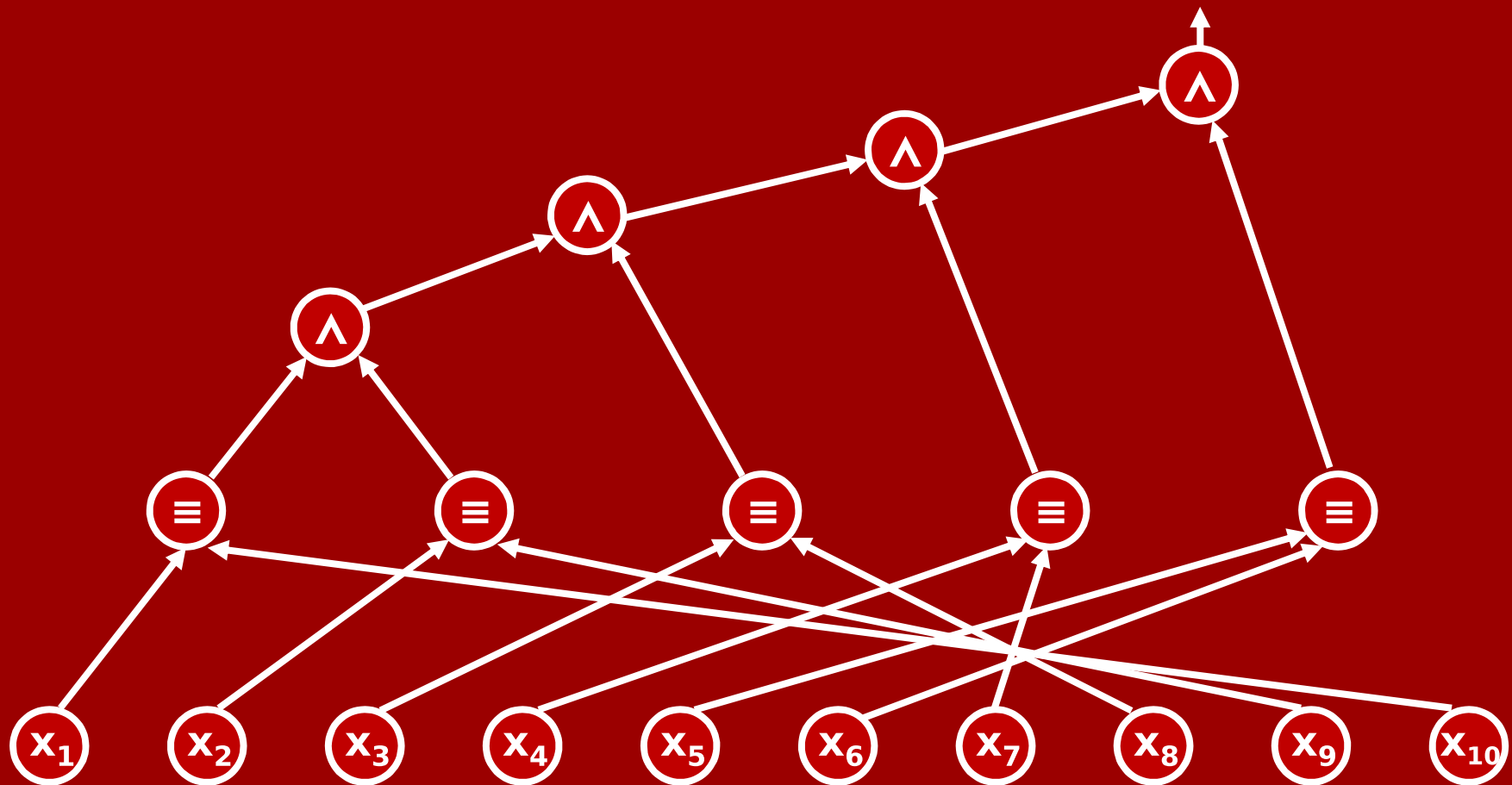
What if we only allow fan-in 2 gates?



Size? 9

(Depth? higher)

What if we only allow fan-in 2 gates?

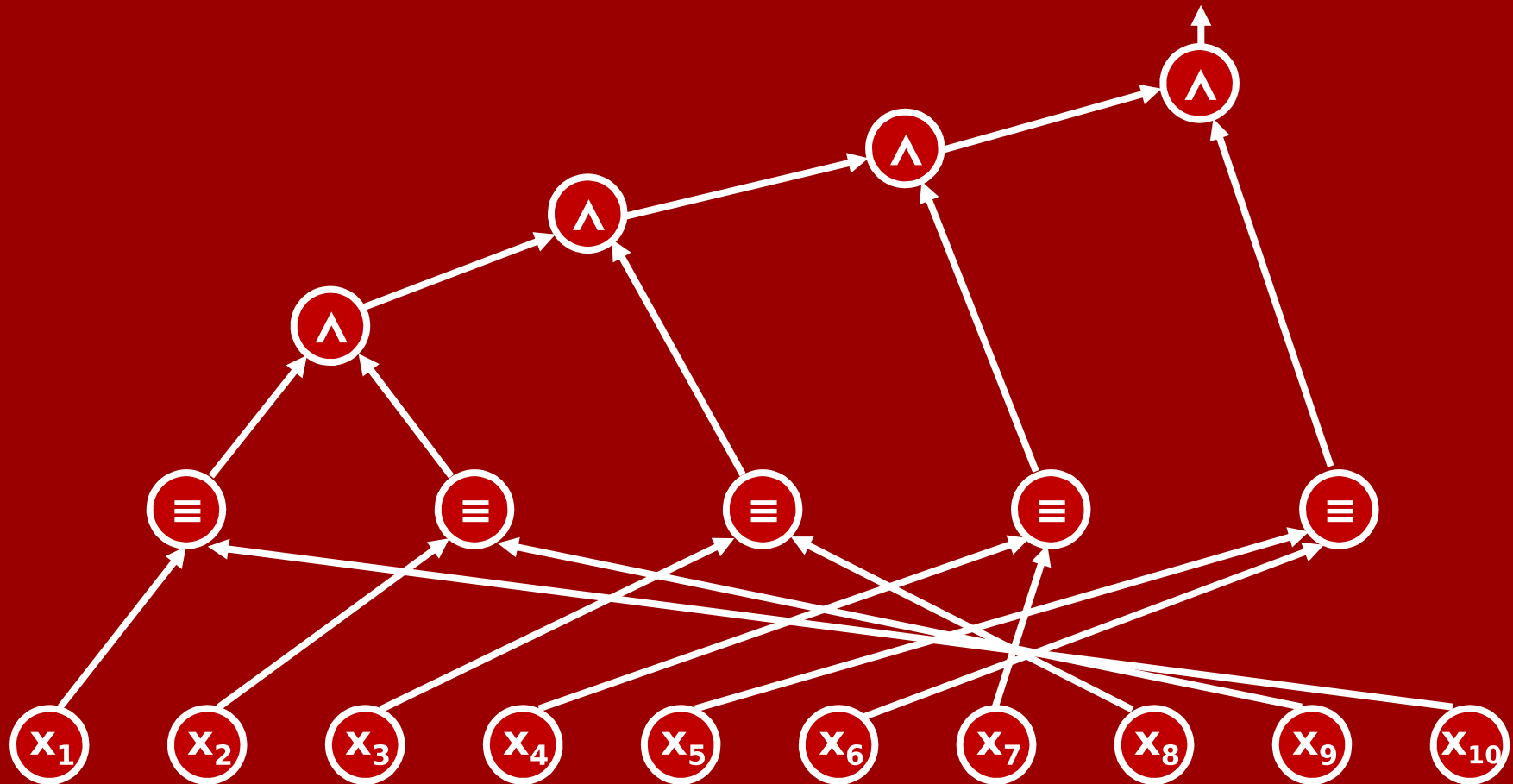


Circuit size for 10-bit inputs: 9

Circuit size for 11-bit inputs: 9

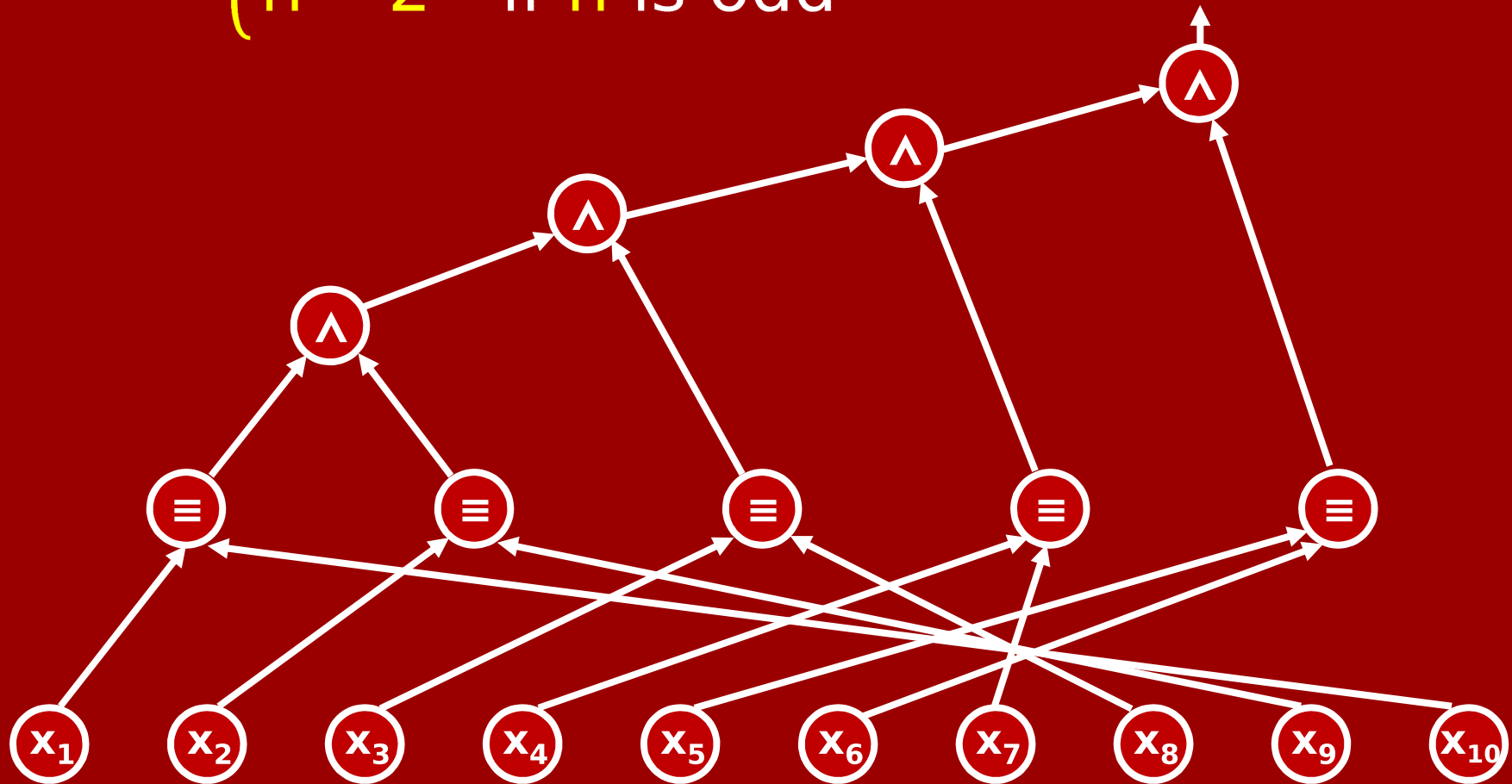
Circuit size for 12-bit inputs: 11

Circuit size for 12-bit inputs: 11



Continuing this pattern, we can get a circuit deciding n -bit inputs for PALINDROME having size...

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



Circuits:

Super-simple.

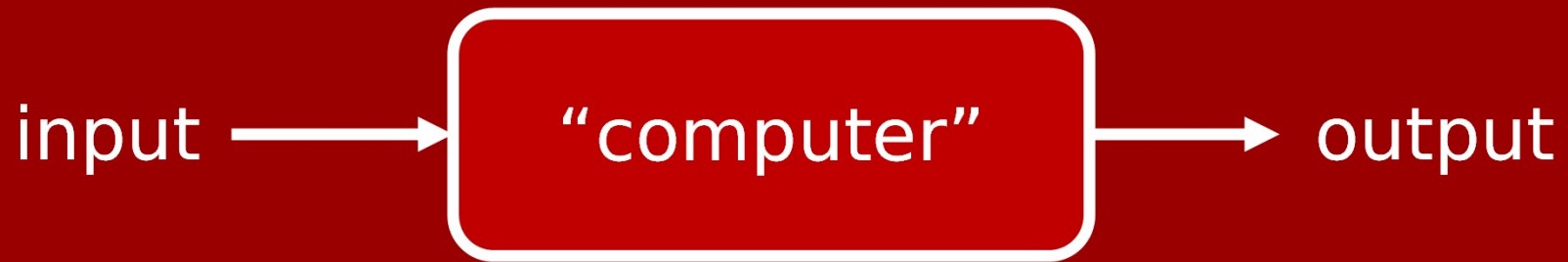
Look like a programming language.

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?!**

From Lecture 2...



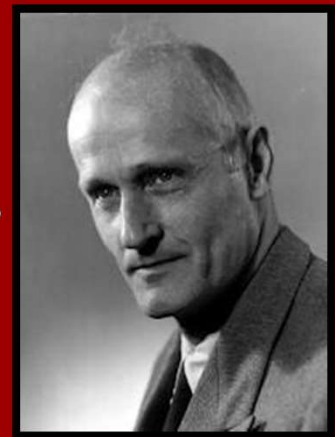
What is **computation**?

What is an **algorithm**?

How can we mathematically define them?

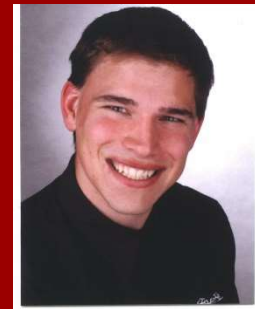
Inspirational quotation

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



Stephen Kleene

“Circuits are an **infinite**
answer to an infinite
number of questions ☹️”

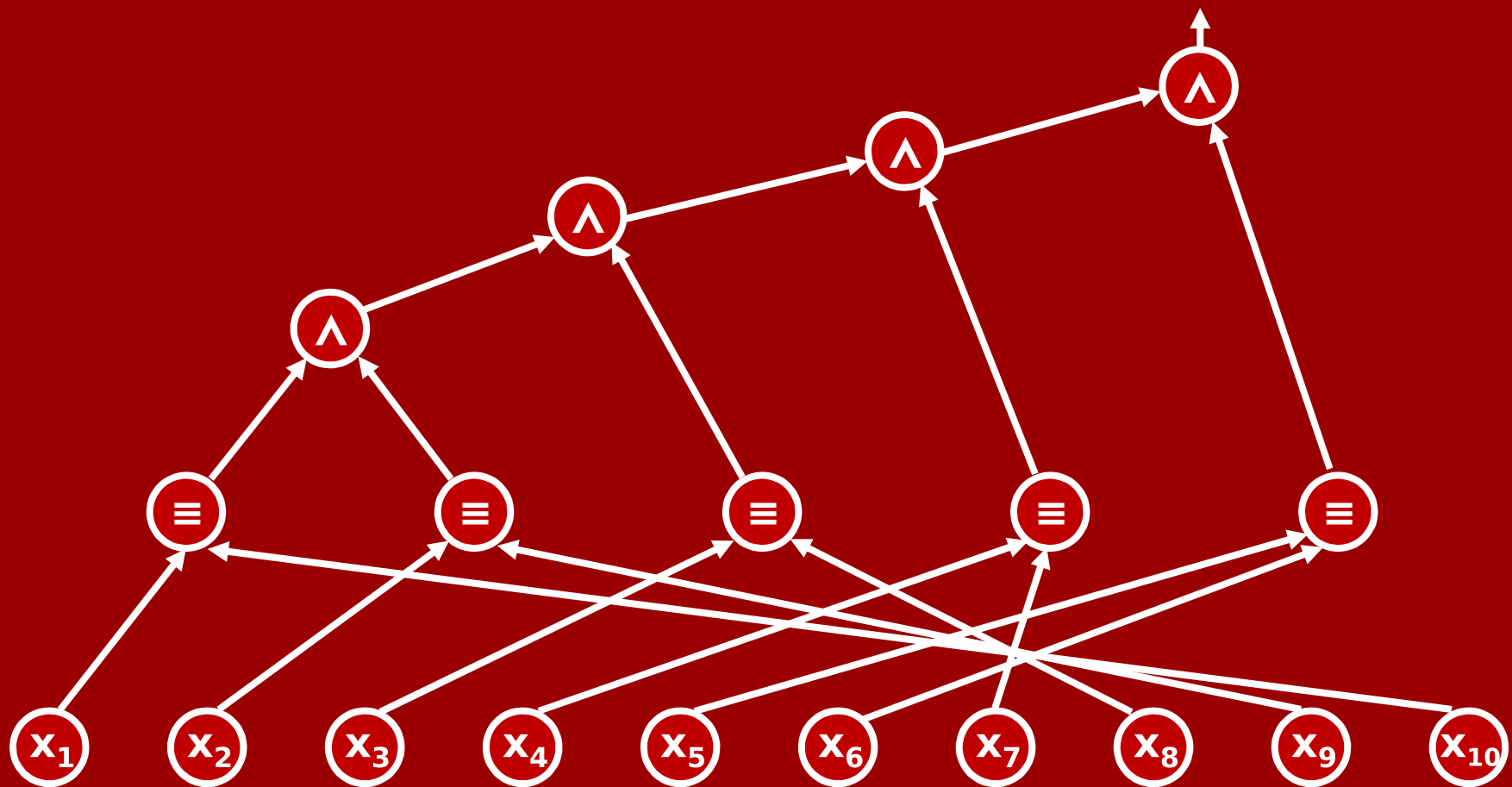


me

Consider the language **PALINDROMES** $\subseteq \{0,1\}^*$

How can we compute it using circuits?

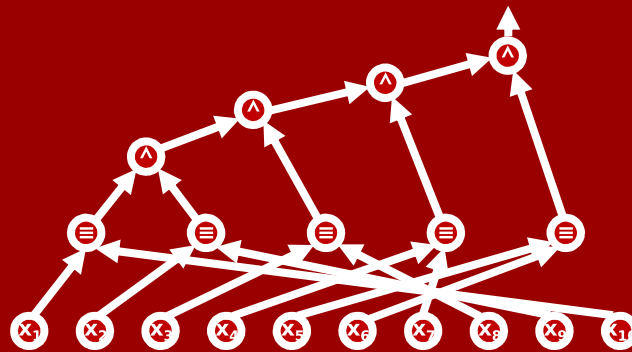
Well, for length-**10** inputs we had something...



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?

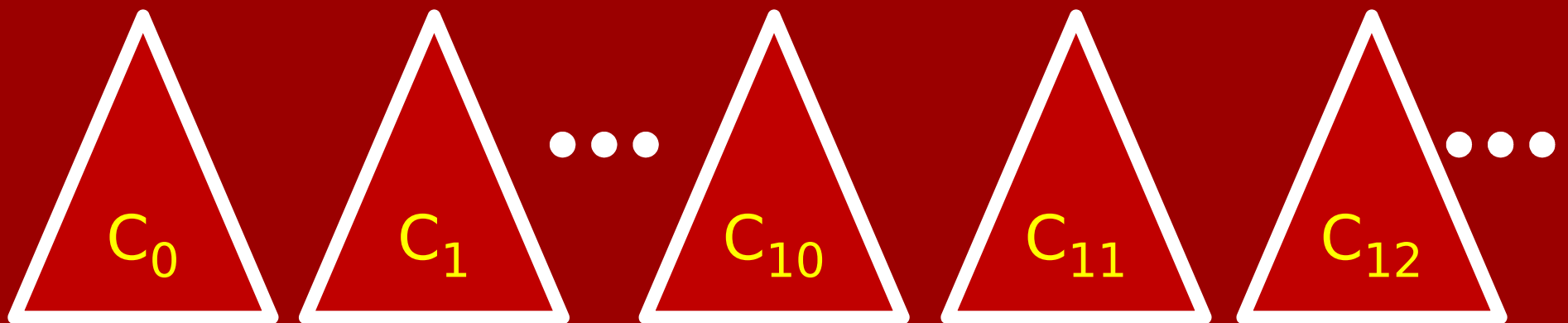
Well, for length-**10** inputs we had something...

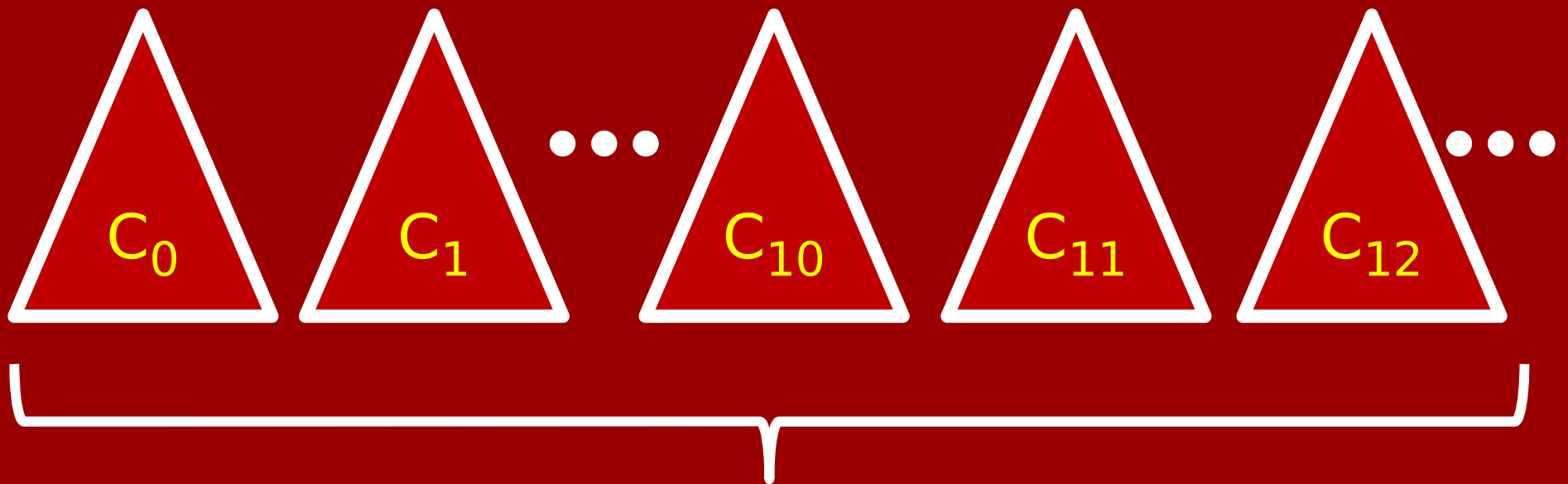
For length-**11** inputs we had something else...

For length-**12** inputs we had something else...

For length-**0** inputs we had something else...

For length-**1** inputs we had something else...

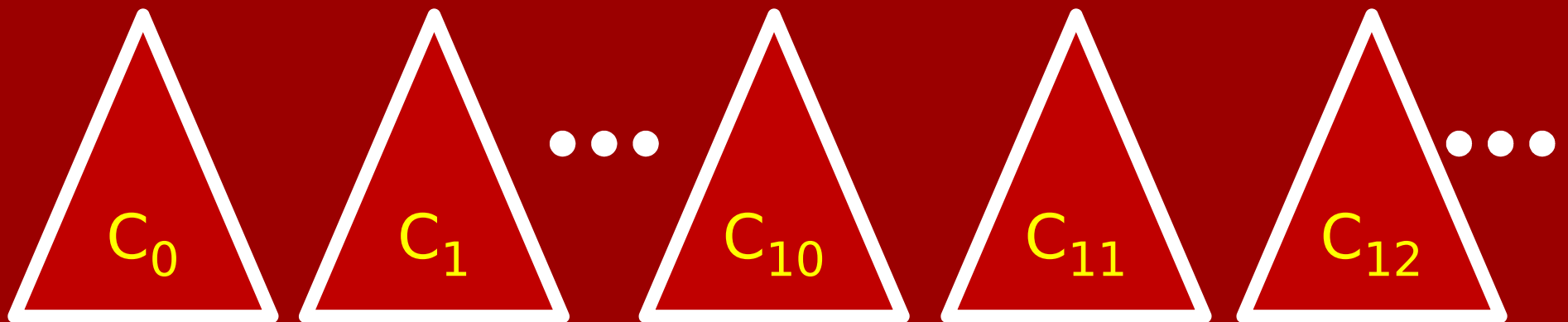




This is called a “family of circuits”.

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 ?
Probably there’s an **algorithm** that on input n , outputs C_n ...)



Definition:

A family of circuits $\mathcal{C}$ is an infinite sequence $C_0, C_1, C_2, \dots$ where C_n is a circuit with n inputs.

We say $\mathcal{C}$ 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 $\mathcal{C}$ is the function $S : \mathbb{N} \rightarrow \mathbb{N}$
defined by $S(n) = \text{size of } C_n$.

Example

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

Then $\mathcal{C}$ 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 $n \geq 4$;
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 :	$\equiv$	(of G_1, G_n)
E_2 :	$\equiv$	(of G_2, G_{n-1})
E_3 :	$\equiv$	(of G_3, G_{n-2})
	...	
$E_{\lfloor n/2 \rfloor}$:	$\equiv$	(of $G_{\lfloor n/2 \rfloor}, G_{\lfloor n/2 \rfloor+1}$)
A_1 :	$\wedge$	(of E_1, E_2)
A_2 :	$\wedge$	(of A_1, E_3)
A_3 :	$\wedge$	(of A_2, E_4)
	...	
$A_{\lfloor n/2 \rfloor-1}$:	$\wedge$	(of $A_{\lfloor n/2 \rfloor-1}, E_{\lfloor n/2 \rfloor}$)

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

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

Consequence:

Every language is computed by a family of circuits of size $O(2^n \cdot n)$.



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

Easy improvement:

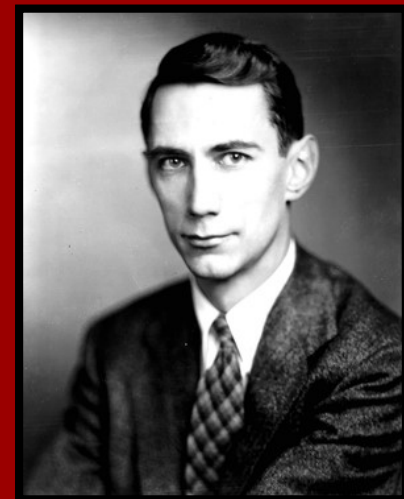
Every language is computed by a family of circuits of size $O(2^n)$.

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

Slightly trickier improvement:

Every language is computed by a family of circuits of size $O(2^n/n)$.

Proved by the great
Claude Shannon in 1949.



TM time versus circuit size

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.

If you don't like a challenge, but are still curious,
see the Notes online.

If you neither like a challenge nor are curious, ☹.

We'll need theorem when studying "NP-hardness."

TM time versus circuit size

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.

TM time versus circuit size

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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TM time versus circuit size

Corollary:

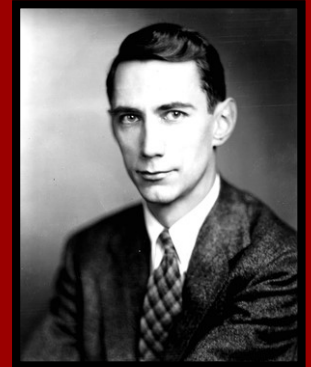
If you want to show some **L** is **not** solvable in polynomial time, suffices to show it is **not** solvable by polynomial-size circuit families.

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 $\wedge/\vee/\neg$ circuit of size $O(2^n/n)$



Shannon's Theorem 2:

Almost every n -bit Boolean function **requires** a circuit of size $\Omega(2^n/n)$

(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 $\Omega(2^n/n)$.

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 $[16 \cdot (n+s) \cdot (n+s)]^s$ possible circuits.

The $[\dots]$ quantity is $\leq 64s^2$ because $n+s \leq 2s$, and $64s^2 \leq (2^n)^2$ for large n . So there are at most $[(2^n)^2]^s = 2^{2ns} = 2^{(1/2) 2^n} = (1.41\dots)^{2^n}$ size- s circuits and most n -bit Boolean functions need a larger size.

Shannon's Theorem 2:

Almost every n -bit Boolean function **requires** a circuit of size $\Omega(2^n/n)$.

“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 $\geq 2n - 3$

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 $\geq 3n - 3$

Consider $n = 50$

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

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

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 $\geq 2n - 3$

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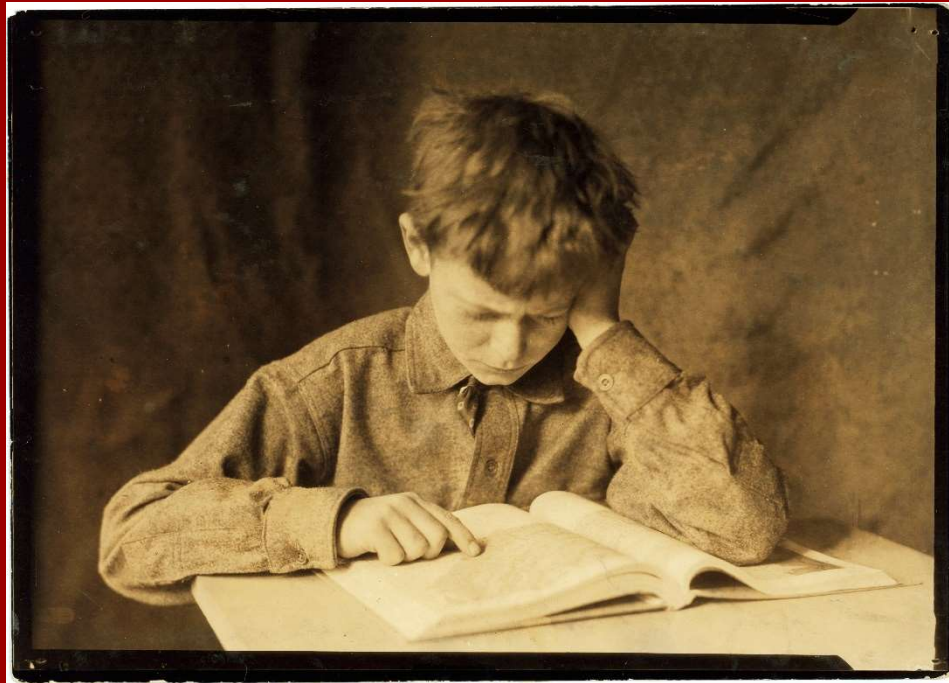
1984: N. Blum showed a certain pretty simple function requires size $\geq 3n - 3$

Good news!!

2016: 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 families & size

Theorems:

Every function can be computed by a DNF

Almost every function requires circuits of size $\Omega(2^n/n)$.