

Polynomial Church-Turing thesis

A decision problem can be solved in polynomial time by using a reasonable sequential model of computation if and only if it can be solved in polynomial time by a Turing machine.

P is the class of these decision problems

Search Problems: **NP**

L is in **NP** iff there is a language L' in **P** and a polynomial p so that:

$$\forall x : x \in L \Leftrightarrow [\exists y \in \{0, 1\}^* : |y| \leq p(|x|) \wedge \langle x, y \rangle \in L']$$

Intuition

$$\forall x : x \in L \Leftrightarrow [\exists y \in \{0, 1\}^* : |y| \leq p(|x|) \wedge \langle x, y \rangle \in L']$$

- The y -strings are the ***possible solutions*** to the instance x .
- We require that solutions are not too long and that it can be checked efficiently if a given y is indeed a solution or ***witness*** (we have a “simple” search problem)

Reductions

- A **reduction** r of L_1 to L_2 is a polynomial time computable map so that

$$\forall x: x \in L_1 \text{ iff } r(x) \in L_2$$

- We write $L_1 \leq L_2$ if L_1 reduces to L_2 .
- **Intuition:** Efficient software for L_2 can also be used to efficiently solve L_1 .

NP-hardness

- A language L is called **NP-hard** iff

$$\forall L' \in \mathbf{NP}: L' \leq L$$

- **Intuition:** Software for L is strong enough to be used to solve any simple search problem.
- **Proposition:** If some **NP-hard** language is in \mathbf{P} , then $\mathbf{P}=\mathbf{NP}$.

NPC

- A language $L \in \mathbf{NP}$ that is **NP**-hard is called **NP**-complete.
- **NPC** := the class of **NP**-complete problems.
- **Proposition:**
$$L \in \mathbf{NPC} \Rightarrow [L \in \mathbf{P} \text{ iff } \mathbf{P}=\mathbf{NP}].$$

- L is in **NP** means:

There is a language L' in **P** and a polynomial p so that

$$\forall x : x \in L \Leftrightarrow [\exists y \in \{0, 1\}^* : |y| \leq p(|x|) \wedge \langle x, y \rangle \in L']$$

- $L_1 \leq L_2$ means:

For some polynomial time computable map r :

$$\forall x: x \in L_1 \text{ iff } r(x) \in L_2$$

- L is **NP-hard** means:

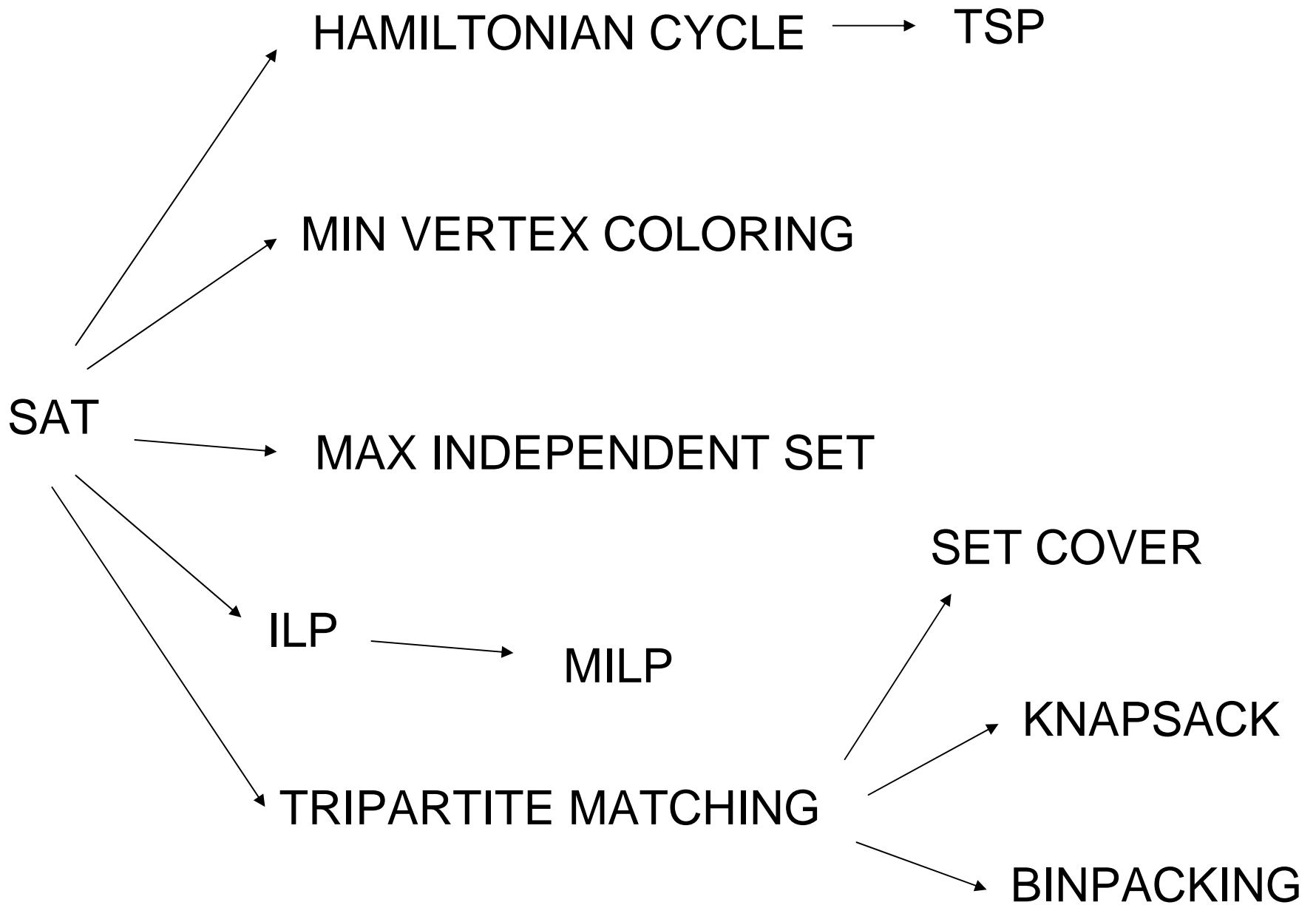
$$\forall L' \in \mathbf{NP}: L' \leq L$$

- L is in **NPC** means:

$L \in \mathbf{NP}$ and L is **NP-hard**

How to establish **NP**-hardness

- Thousands of natural problems are **NP**-complete:
- **Empirical fact:** Most natural problems in **NP** are either in **P** or **NP**-hard.
- **Lemma:** If L_1 is **NP**-hard and $L_1 \leq L_2$ then L_2 is **NP**-hard.
- We need to establish **one** problem to be **NP**-hard, the rest follows using chains of reductions. Cook (1972) established SAT to be **NP**-hard.



Boolean functions

- $f: \{\text{false}, \text{true}\}^n \rightarrow \{\text{false}, \text{true}\}$
- Example: $\text{XOR}(x_1, x_2) = x_1 \oplus x_2$
- $\text{XOR}(\text{true}, \text{true}) = \text{false}$
- In this course $0 = \text{false}$, $1 = \text{true}$.
- $\text{XOR}(1, 1) = 0$

How to represent Boolean functions on a computer

- Tables.
- Formulae.

Tables

A	B	Z
0	0	0
0	1	1
1	0	1
1	1	0

- More compact representation as Boolean String: “0110”.
- A function $f: \{0,1\}^n \rightarrow \{0,1\}$ can be represented as a table using ...

..... 2^n bits

Boolean Formulae

- X_1, X_2, \dots, X_n are formulae.
- If f is a formula then $\neg f$ is a formula.
- If f_1 and f_2 are formulae then $(f_1) \wedge (f_2)$ and $(f_1) \vee (f_2)$ are formulae.
- Sometimes we leave out parentheses....

Formulae represent Boolean functions

- “ $(x_1 \wedge \neg x_2) \vee (x_2 \wedge \neg x_1)$ ” represents the function XOR.
- Sometimes formula-representation is much more compact than table representation. Sometimes not.
- It's never much *less* compact.

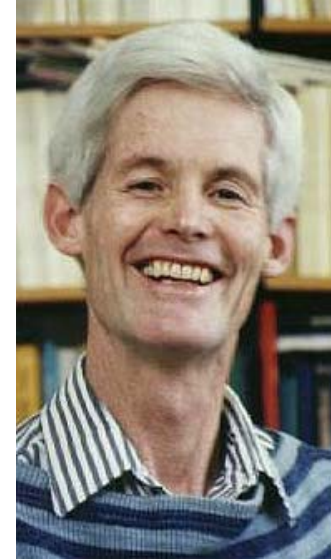
Two special classes of Boolean formulae

- $f(x_1, x_2) = x_1 \oplus x_2$
- **DNF:** $f(x_1, x_2) = (x_1 \wedge \neg x_2) \vee (x_2 \wedge \neg x_1)$
- **CNF:** $f(x_1, x_2) = (\neg x_1 \vee \neg x_2) \wedge (x_1 \vee x_2)$
- A CNF is any conjunction of **clauses** (disjunctions of literals). A DNF is any disjunction of **terms** (conjunctions of literals).
- Any function on n variables can be described by a CNF (DNF) formula containing at most 2^n clauses (terms), each containing at most n literals.

SAT

- SAT: Given a Boolean function in CNF representation, is there a way to assign truth values to the variables so that the function evaluates to true?
- SAT: Given a CNF, is it true that it does **not** represent the constant-0 function?
- Input: $(\neg x_1 \vee \neg x_2) \wedge (x_1 \vee x_2)$
- Output: **Yes.**
- Input: $(\neg x_1 \vee \neg x_2) \wedge (x_1 \vee x_2) \wedge (x_1 \vee \neg x_2) \wedge (\neg x_1 \vee x_2)$
- Output: **No.**

SAT

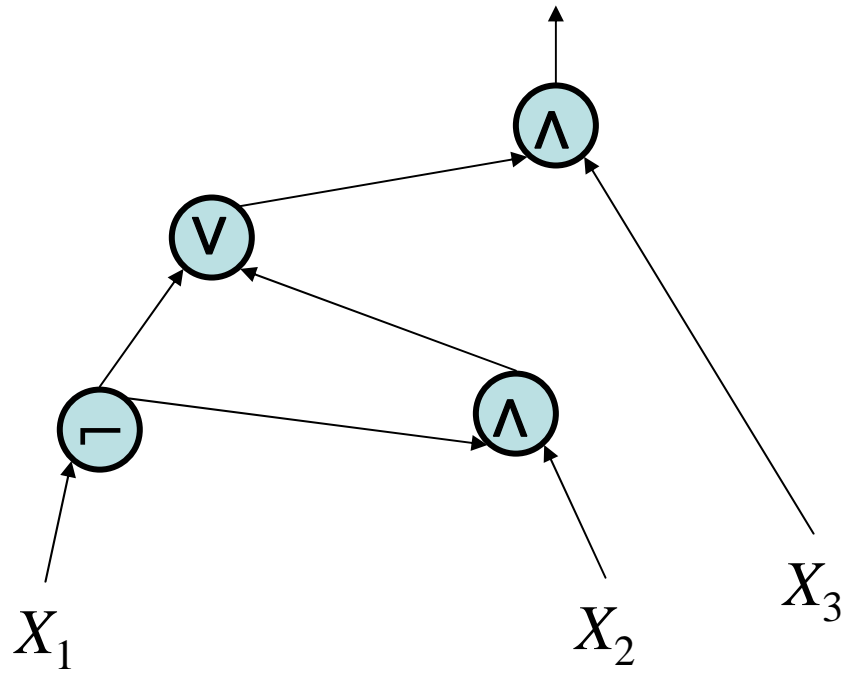


- SAT is in **NP**.
- Cook's theorem (1972): SAT is **NP**-hard.
- Hence, SAT is **NP**-complete. It is a universal search problem and we do *not* think it has a polynomial time algorithm. If we find one, we get \$1.000.000.

DNF?

- Suppose we ask the same question of DNF formulas, rather than CNF formulas.
- Can you get a million dollars for solving this problem?

Boolean Circuits



Circuits

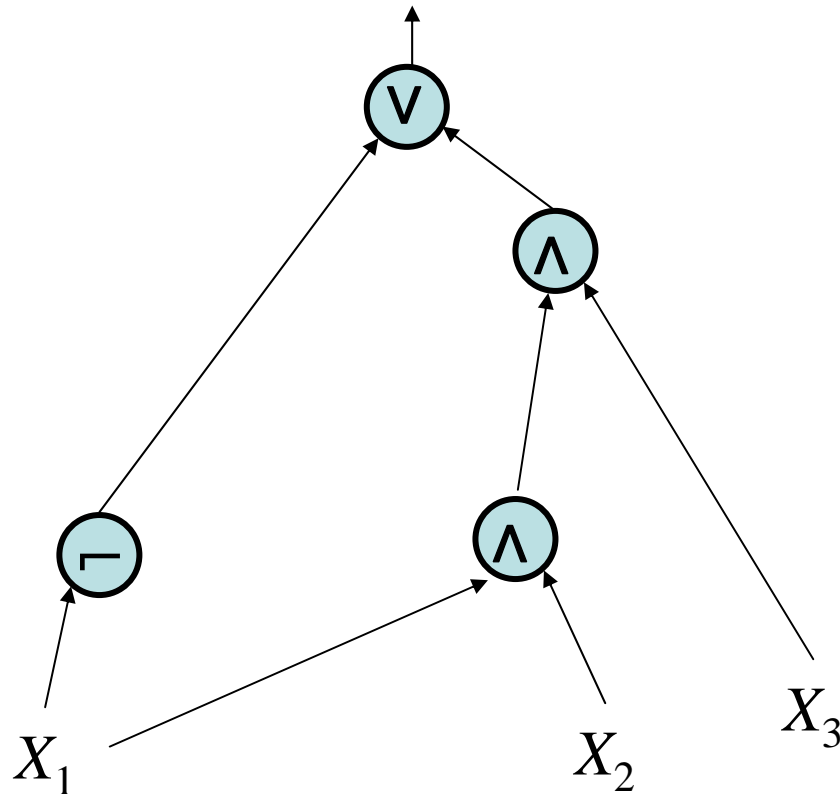
- A circuit C is a directed **acyclic** graph.
- Nodes in C are called **gates**.

Types of gates:

- Variable (or **input**) gates of indegree 0, labeled X_1, X_2, \dots, X_n
 - **Constant** gates of indegree 0, labeled 0,1
 - **AND**-, **OR**-gates of indegree 2,
 - **NOT**-, **COPY**-gates of indegree 1.
 - m distinguished gates are **output** gates.
- The circuit C computes a Boolean function
 $C: \{0,1\}^n \rightarrow \{0,1\}^m$

A formula can be viewed as a
special kind of circuit

$$f(X_1, X_2, X_3) = \neg X_1 \vee (X_1 \wedge X_2 \wedge X_3)$$



Are formulas and circuits in fact the same thing?

- **Not quite!**
- Given a circuit, we can write down an equivalent formula, but it may become *much* bigger.
- A circuit is allowed to *reuse* the result of a sub-computation without doing the computation again!

Multi-output circuits

- Any function $f: \{0,1\}^n \rightarrow \{0,1\}^m$ is represented by some circuit.

Proof:

- Any function $f: \{0,1\}^n \rightarrow \{0,1\}$ can be described by a Boolean formula.
- A formula can also be viewed as a circuit.
- Combine m such circuits.

Tables – Formulae - Circuits

- For a function $f: \{0,1\}^n \rightarrow \{0,1\}$, let $s_t = 2^n$ be the size (in bits) of its representation as a table.
- Let s_f be the size (in bits) of its smallest representation as a formula. Then,
 - $|s_f| \leq 10 |s_t|^2$ (formulae are not much less compact than tables).
 - For all n , there is a function so that $|s_f| \leq 10 (\log |s_t|)^2$ (for some functions, formulae are *much* more compact)
- Let s_c be the size (in bits) of its smallest representation as a circuit. Then,
 - $|s_c| \leq 10 |s_f|^2$ (circuits are not much less compact than formulae).
 - Is it true that there for all n is a function so that circuits are *much* more compact (e.g., $|s_c| \leq 10 (\log |s_f|)^2$)? **This is open!**
- **You won't get \$1.000.000 for solving this, but you'll still get pretty famous**

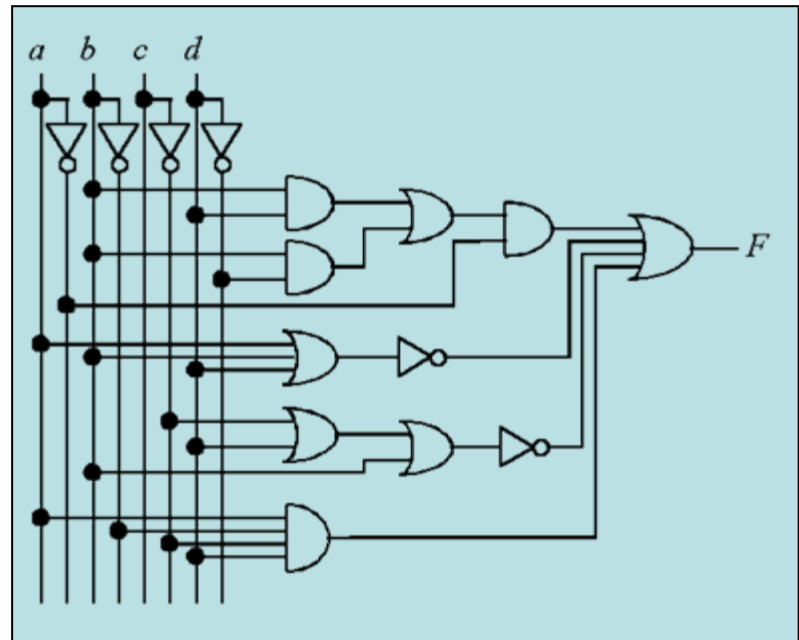
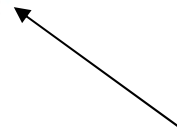
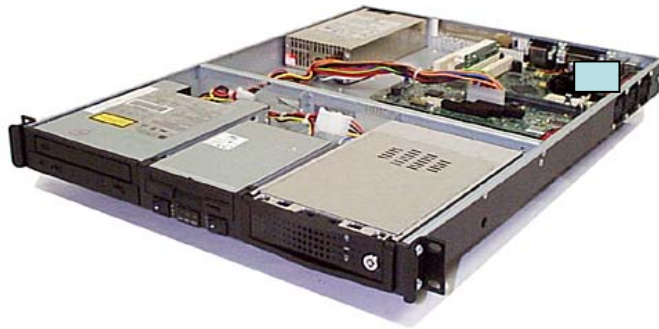
Circuits vs. Turing Machines

- Circuits can be given as *inputs* to algorithms but they can also be seen as computational *devices* themselves!
- Like Turing Machines, circuits
 $C: \{0,1\}^n \rightarrow \{0,1\}$ solve *decision* problems on $\{0,1\}^n$.
- Unlike Turing machines, circuits takes inputs of a *fixed* input length n only.

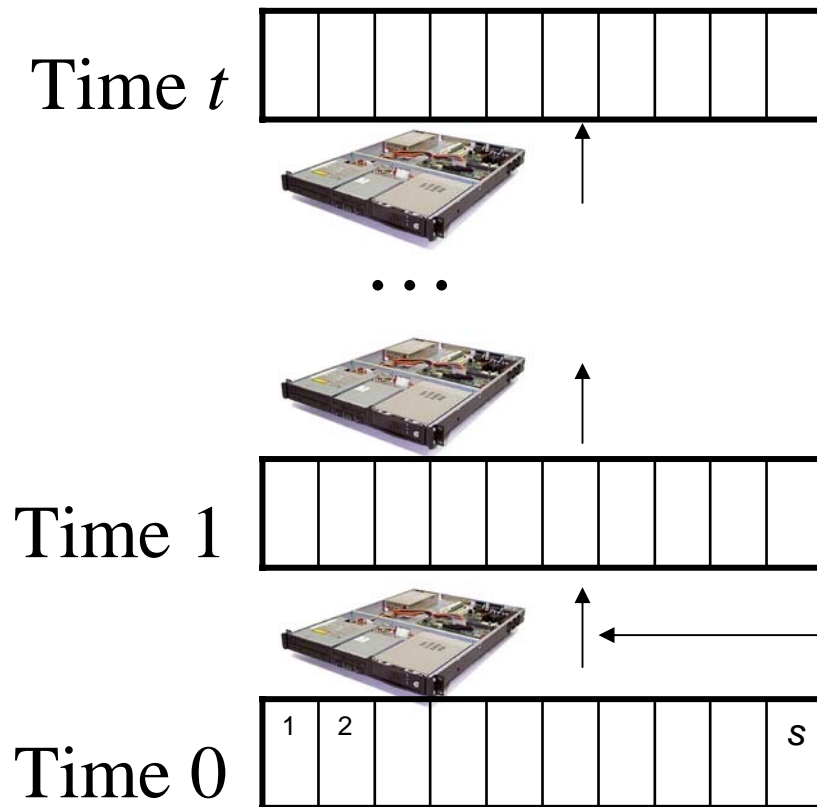
Theorem

- Given Turing Machine M running in **time** at most $p(n)$ on inputs of length n , where p is a polynomial.
- For every n , there is a circuit C_n with at most $O(p(n)^2)$ **gates** so that
$$\forall x \in \{0,1\}^n: C_n(x)=1 \text{ iff } M \text{ accepts } x.$$
- The map $1^n \rightarrow C_n$ is polynomial time computable.

Intuition behind proof



The Tableau Method

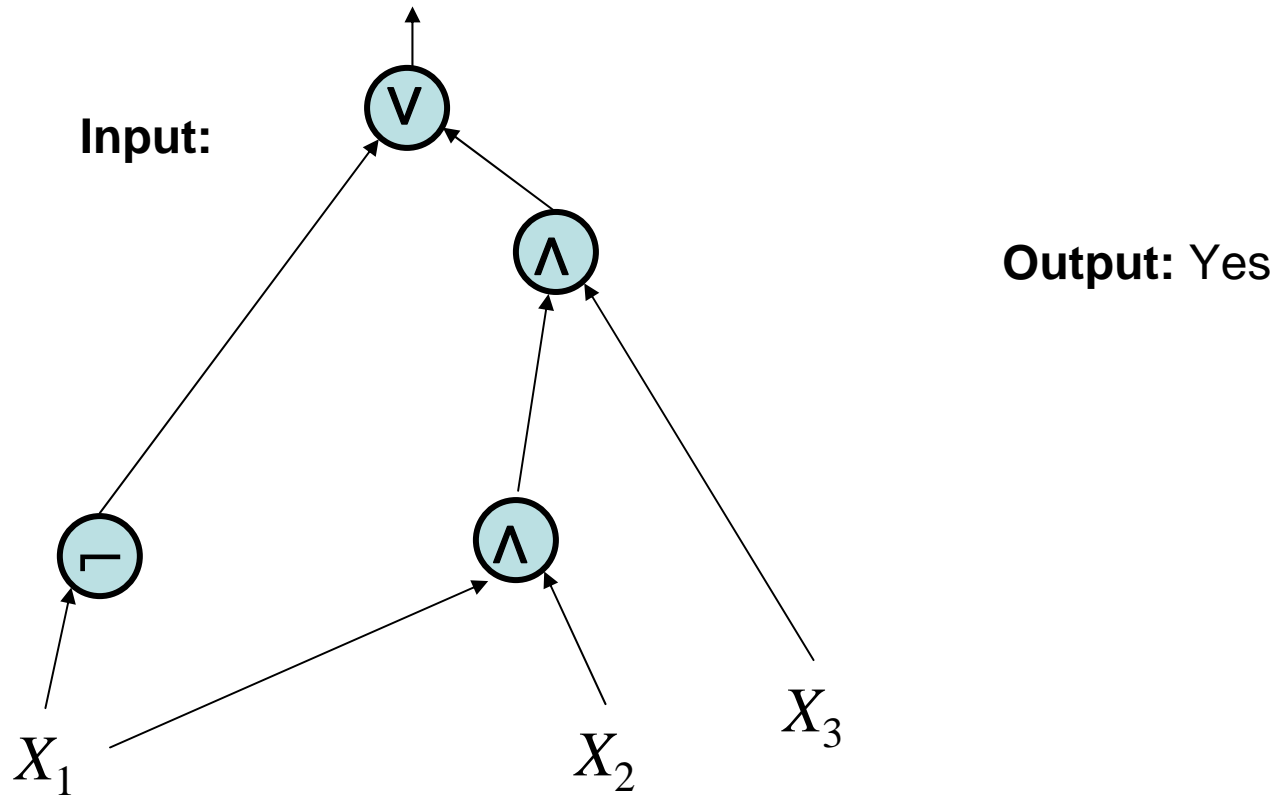


**Can be replaced by
acyclic Boolean
circuit of size $\approx s$**

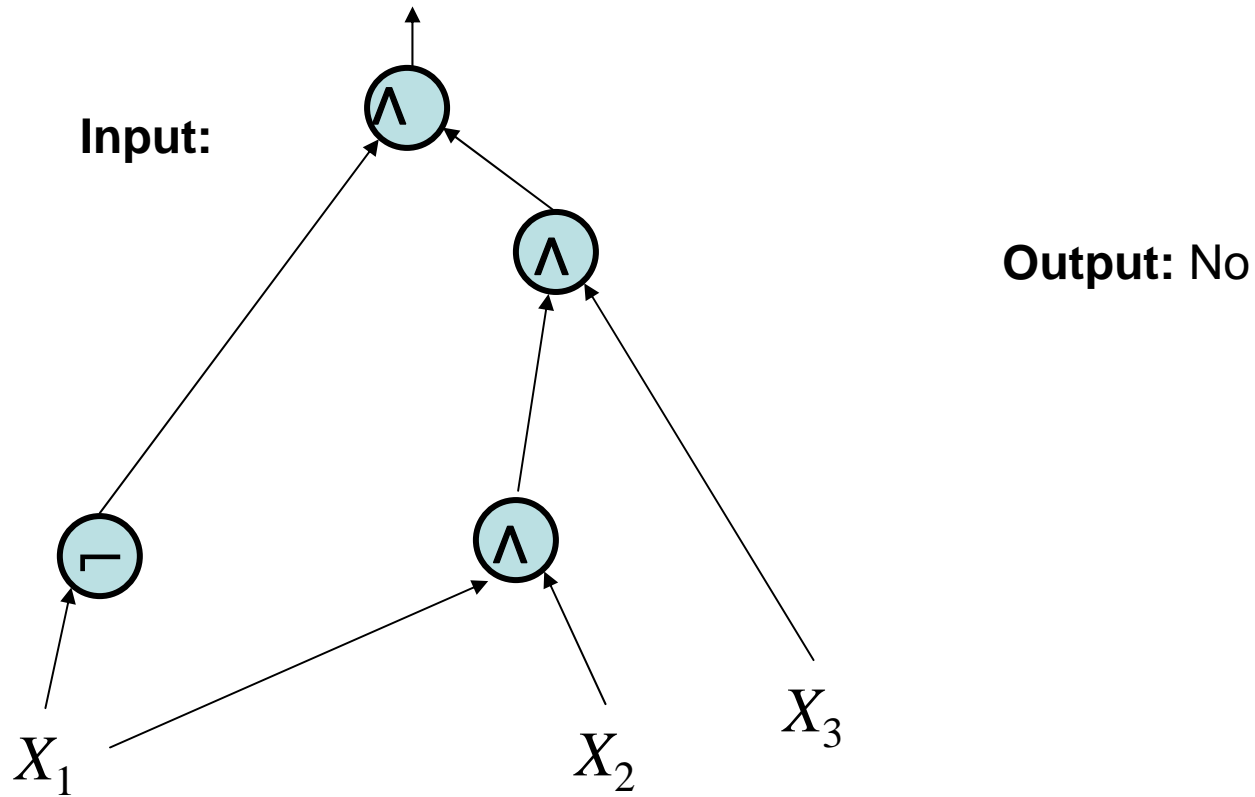
CIRCUIT SAT

- CIRCUIT SAT: Given a Boolean circuit, is there a way to assign truth values to the input gates variables, so that the output gate evaluates to true?
- Generalizes SAT, as CNFs are formulas and formulas are circuits.

Example



Example



CIRCUIT SAT

- CIRCUIT SAT is in **NP**.
- Proof of Cook's theorem:
 - CIRCUIT SAT reduces to SAT.
 - CIRCUIT SAT is **NP**-hard.
 - Hence, SAT is **NP**-hard.

CIRCUIT SAT \longrightarrow SAT

CIRCUIT SAT is NP-hard

- Given an arbitrary language L in NP we must show that L reduces to CIRCUIT SAT.
- This means: We must construct a polynomial time computable map r mapping instances of L to circuits, so that

$$\forall x: x \in L \Leftrightarrow r(x) \in \text{CIRCUIT SAT}$$

- The only thing we know about L is that there is a language L' in P and a polynomial p , so that:

$$\forall x: x \in L \Leftrightarrow [\exists y \in \{0, 1\}^* : |y| \leq p(|x|) \wedge \langle x, y \rangle \in L']$$

Theorem

- Given Turing Machine M running in **time** at most $p(n)$ on inputs of length n , where p is a polynomial.
- For every n , there is a circuit C_n of **size** at most $O(p(n)^2)$ so that
$$\forall x \in \{0,1\}^n: C_n(x)=1 \text{ iff } M \text{ accepts } x.$$
- The map $1^n \rightarrow C_n$ is polynomial time computable.