

15-251: Great Theoretical Ideas in Computer Science

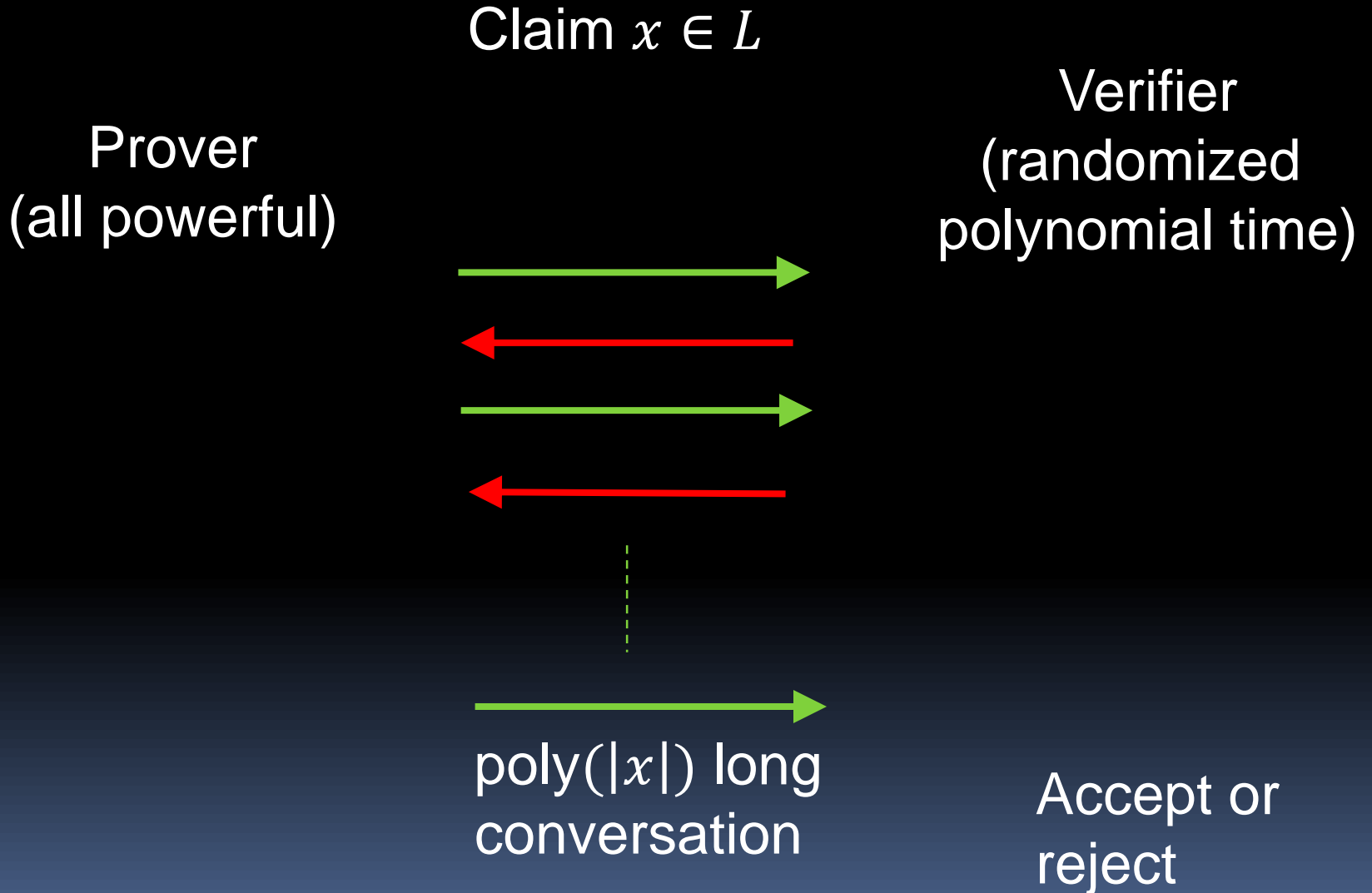
Fall 2016 Lecture 29

December 8, 2016

Epilogue:

**Interactive Proof for $\overline{3SAT}$ and
Couple of open problems**

Interactive Proofs



Prover any function $P(x,c)$ = what to say next if c is conversation so far.

Verifier is a **poly-time** function
 $V(x,r,c)$ = what to say next if c is conversation so far (r =verifier's random coins)

$P \leftrightarrow V(x,r)$ denotes one conversation.
Total bits in conversation can't exceed some $\text{poly}(|x|)$. Verifier must accept/reject at end of conversation.

A language L belongs to **IP** iff

There is a polynomial time verifier $V(x,r)$, $|r| < \text{poly}(|x|)$;
all $P \leftrightarrow V(x,r)$ conversations are $\text{poly}(|x|)$ bounded
and accept/reject.

- Completeness

- $x \in L$: \exists Prover P , $\Pr_r[P \leftrightarrow V(x,r) \text{ accepts}] \geq \frac{3}{4}$

- Soundness

- $x \notin L$: \forall Provers P , $\Pr_r[P \leftrightarrow V(x,r) \text{ accepts}] \leq \frac{1}{4}$

We saw last lecture that graph non-isomorphism has an interactive proof, even though we don't know if it is in NP.

What about problems we *surely believe to not be in NP*, such as complements of 3COLOR, 3SAT, or other NP-complete problems?

In particular, can one (interactively) prove that a 3SAT formula is NOT satisfiable?

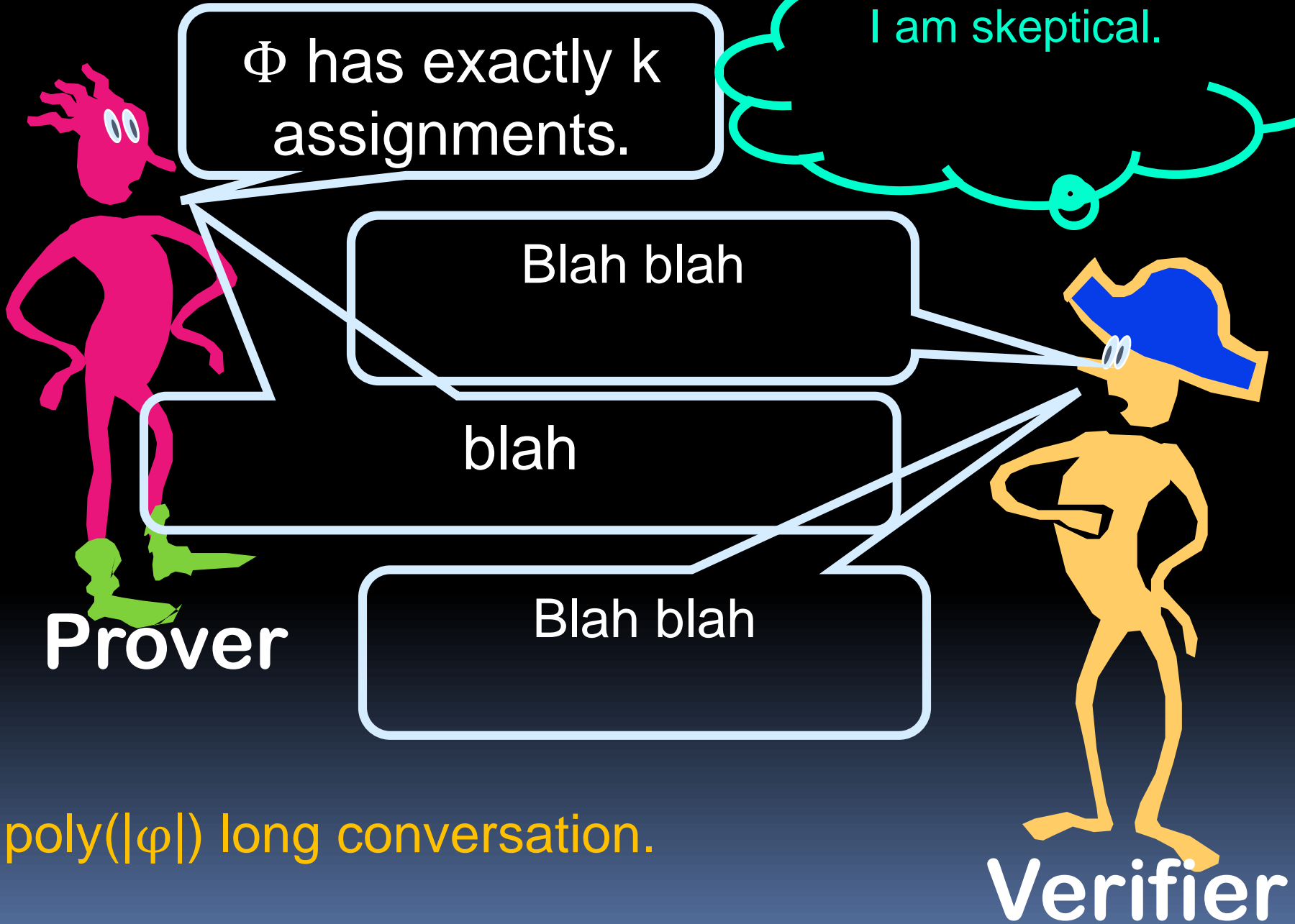
A more general problem: #SAT

How many satisfying assignments does φ have?

$3SAT = \{\varphi \mid \varphi \text{ is a satisfiable 3-CNF formula.}\}$

$\#3SAT = \{(\varphi, k) \mid \varphi \text{ is a 3CNF formula that has exactly } k \text{ satisfying assignments}\}$

#3SAT \in IP



$\text{poly}(|\varphi|)$ long conversation.

A Great Theoretical Idea

Arithmetization: From Boolean formulae to a *polynomial over a finite field*.

- $\text{Arith}(T) = 1, \text{Arith}(F) = 0$
- $\text{Arith}(x) = x$
- $\text{Arith}(\neg\theta_1) = 1 - \text{Arith}(\theta_1)$
- $\text{Arith}(\theta_1 \wedge \theta_2) = \text{Arith}(\theta_1) \text{Arith}(\theta_2)$
- $\text{Arith}(\theta_1 \vee \theta_2) = 1 - (1 - \text{Arith}(\theta_1))(1 - \text{Arith}(\theta_2))$

Claim (proof by induction): For any Boolean formula θ , θ and $\text{Arith}(\theta)$ agree on all 0/1 assignments.

Also, $\text{degree}(\text{Arith}(\theta)) = O(\text{size}(\theta))$

Example

Suppose

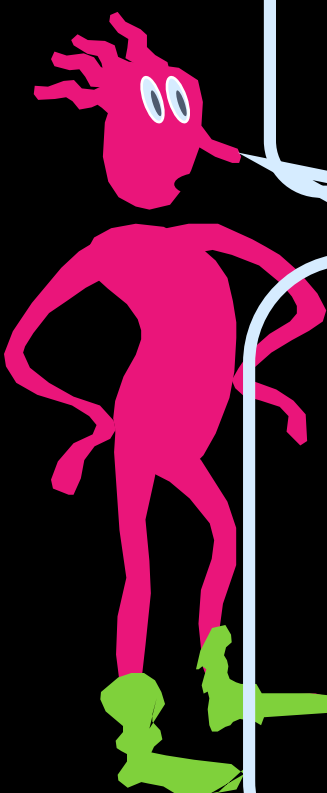
$$\phi = (x_1 \vee \neg x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_2 \vee \neg x_4)$$

Arith(ϕ) =


$$(1 - (1 - x_1)x_2(1 - x_3)) \cdot (1 - x_1x_3(1 - x_4)) \cdot (1 - x_2x_4)$$

Product of m terms if there are m clauses.

Each has degree at most 3 if ϕ is a 3SAT instance



Claim: $\Phi(x_1, x_2, \dots, x_n)$ has exactly k assignments.



First I choose a prime $p > 2^n$ and let us encode $\Phi(x_1, x_2, \dots, x_n)$ as $P = \text{Arith}(\Phi)$ over F_p so that P and Φ are identical on all 2^n 0/1 assignments.

Prover

I'll now prove that

$$\sum_{(b_1, b_2, \dots, b_n) \in \{0, 1\}^n} P(b_1, b_2, \dots, b_n) = k$$

Let's define some polynomials

$P(x_1, x_2, \dots, x_n) = \text{Arith}(\phi)$; n -variate polynomial

$$P_0 = \sum_{b_1, \dots, b_n \in \{0,1\}} P(b_1, b_2, \dots, b_n); \text{ To prove } P_0 = k$$

$$P_1(x) = \sum_{b_2, \dots, b_n \in \{0,1\}} P(x, b_2, \dots, b_n); \text{ Univariate polynomial in } F_p[x]$$

Prover can send $P_1(x)$ and verifier can check $P_1(0) + P_1(1) = k$.

But what if prover sends bogus $P_1(x)$?

To combat this, Verifier picks random $r_1 \in F_p$, and challenges prover to “prove” the value of $P_1(r_1)$

Key point: If prover sends polynomial $Q_1(x) \neq P_1(x)$, then $Q_1(r_1) \neq P_1(r_1)$ with prob. $\geq 1 - O(\text{size}(\phi)/p)$

The power of polynomials

Two n -variable formulae ϕ and ψ can differ on just one out of 2^n assignments. So can't catch their difference by checking at a random assignment.

Two low-degree polynomials $P \neq Q$ must differ on significant fraction of the domain.

This property was very useful for “error-correction” and is now handy again.

Amazing reach of this simple fact about polynomials: **a nonzero degree d polynomial has at most d roots over a field.**

$P(x_1, x_2, \dots, x_n) = \text{Arith}(\phi)$; n -variate polynomial

$$P_1(x) = \sum_{b_2, \dots, b_n \in \{0,1\}} P(x, b_2, \dots, b_n); \text{ Univariate polynomial in } F_p[x]$$

Prover asked to send low-degree polynomial $P_1(x)$ (by listing its coefficients); verifier checks $P_1(0) + P_1(1) = k$, picks random $r_1 \in F_p$, and challenges prover to prove the claimed value of $P_1(r_1)$.

Lies beget lies: If prover sends poly $Q_1(x) \neq P_1(x)$, then $Q_1(r_1) \neq P_1(r_1)$ with high probability.

Next round:

$$P_2(x) = \sum_{b_3, \dots, b_n \in \{0,1\}} P(r_1, x, b_3, \dots, b_n); \text{ Univariate poly. in } F_p[x]$$

Prover asked to send polynomial $P_2(x)$ (by listing its coefficients); verifier checks $P_2(0) + P_2(1) = P_1(r_1)$, picks random $r_2 \in F_p$, and challenges prover to prove the claimed value of $P_2(r_2)$.

Next round:

$$P_2(x) = \sum_{b_3, \dots, b_n \in \{0,1\}} P(r_1, x, b_3, \dots, b_n); \text{ Univariate poly. in } F_p[x]$$

Prover sends low-degree polynomial $P_2(x)$ (by listing its coefficients); verifier checks $P_2(0) + P_2(1) = P_1(r_1)$, picks random $r_2 \in F_p$, and challenges prover to prove the claimed value of $P_2(r_2)$.

Lies beget more lies: If prover sends polynomial $Q_2(x) \neq P_2(x)$, then $Q_2(r_2) \neq P_2(r_2)$ with high prob.

Round i invariant: Verifier has chosen r_1, r_2, \dots, r_{i-1} .

Prover must commit to a polynomial, which for honest behavior should be

$$P_i(x) = \sum_{b_{i+1}, \dots, b_n \in \{0,1\}} P(r_1, \dots, r_{i-1}, x, b_{i+1}, \dots, b_n)$$

Verifier checks $P_i(0) + P_i(1) = P_{i-1}(r_{i-1})$, picks random $r_i \in F_p$, and tasks prover with backing up the claimed value of $P_i(r_i)$.

Final round: Verifier has chosen r_1, r_2, \dots, r_{n-1} .

Prover sends a univariate low-degree polynomial, supposedly

$$P_n(x) = P(r_1, \dots, r_{n-1}, x)$$

Verifier checks $P_n(0) + P_n(1) = P_{n-1}(r_{n-1})$,

picks random $r_n \in F_p$,

and checks $P_n(r_n) = P(r_1, r_2, \dots, r_n)$.

Verifier rejects if any of its checks across the n rounds fails; otherwise he accepts.

Completeness: If $\Phi(x_1, x_2, \dots, x_n)$ has exactly k assignments, then a prover playing honestly by the rules will satisfy all checks made by the verifier, and the verifier will accept with certainty.

Soundness Theorem: If number of satisfying assignments to $\Phi(x_1, x_2, \dots, x_n)$ doesn't equal k , then the verifier accepts with probability $\leq \text{poly}(n)/p \ll 1/2$

Proof idea: Let $Q_i(x)$ be poly. prover sends in round i
Since # sat. assignments to $\Phi = P_1(0) + P_1(1) \neq k$,
prover must lie about $P_1(x)$ in round 1, sending $Q_1 \neq P_1$.
(otherwise the check $Q_1(0)+Q_1(1)=k$ will fail)

Now $P_2(0)+P_2(1)=P_1(r_1)$ (by defn) & $P_1(r_1) \neq Q_1(r_1)$ w.h.p.
So prover is forced to lie in round 2, sending $Q_2 \neq P_2$
(otherwise the check $Q_2(0)+Q_2(1)=Q_1(r_1)$ will fail)

Continuing this argument, unless very lucky in an earlier round,
prover must send $Q_n(x) \neq P_n(x)$ in round n .

Verifier can compute $P_n(r_n) = P(r_1, r_2, \dots, r_n)$
(as he knows P) & will find $P(r_1, r_2, \dots, r_n) \neq Q_n(r_n)$ w.h.p.

Probability of accepting false claim

For verifier to accept, prover must get lucky in some round.

Let i be the earliest round where this happens, i.e., $P_i(r_i) = Q_i(r_i)$ even though $P_i(x) \neq Q_i(x)$

As P_i and Q_i are degree $\text{poly}(n)$ polys, this happens with probability $\leq \text{poly}(n)/p$

The probability that prover gets lucky in *some* round is at most n times bigger, and thus also $\leq \text{poly}(n)/p$

Summary

One can prove that a 3SAT formula is not satisfiable via an interactive proof!
(Note: verifier is efficient, prover has to work hard)

Via NP-completeness reductions, same holds for claim that graph is not 3-colorable, not Hamiltonian, etc.

In fact, power of interactive proofs extends to all problems solvable in polynomial *space* ($IP=PSPACE$)

Surprising efficacy of polynomials in unexpected places!!

Problem 1: Bin Packing

Given a set $A = \{a_1, a_2, \dots, a_n\}$ of positive integers, and a positive integer $B \geq \max a_i$, partition A into minimum number of subsets such that the sum of the elements in each subset is at most B .

Midterm 2: Bin Packing is NP-hard

(reduction from PARTITION to telling if two “bins” suffice)

Open: Is there a polytime algo to find a partition with $\text{OPT} + 1$ subsets, where OPT is the number of subsets in an optimal solution?

Best known: $(1+\epsilon) \text{OPT} + 1$, or $\text{OPT} + O(\log \text{OPT})$

Problem 2: Graph 3-Coloring

3COLOR = { $\langle G \rangle$ | G is 3-colorable } is NP-complete.

So, unless $P=NP$, there is no polynomial time algorithm that given as input a 3-colorable graph, finds a proper 3-coloring of it.

Also known (and harder to prove):

Finding a 4-coloring of a 3-colorable graph is NP-hard.

Open: What about 5-coloring a 3-colorable graph?

Most likely NP-hard, but we can't prove it!

Best algorithm uses $\approx n^{0.2}$ colors, where $n = \#$ vertices

Problem 3: Finding a satisfying assignment when there is an abundance of them

Suppose we are given a CNF formula ϕ on n variables that is promised to have $\geq 2^{n-1}$ satisfying assignments (i.e., at least $\frac{1}{2}$ the assignments satisfy every clause).

- Trivial to find a satisfying assignment of such an instance in randomized polynomial time.

Open: Is there a *deterministic* polynomial time algorithm for finding a satisfying assignment given such an instance of SAT?

Best known runtime: $n^{O((\log \log n)^2)}$ (2016)