

# Computation and Reasoning Laboratory National Technical University of Athens

2014-2015

## 2st Part

Oracles - Polynomial Hierarchy - Randomization - Nonuniform Complexity - Interaction - Counting Complexity

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# Bibliography

## Textbooks

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# Contents

- Introduction
- Turing Machines
- Undecidability
- Complexity Classes
- **Oracles & Optimization Problems**
- Randomized Computation
- Non-Uniform Complexity
- Interactive Proofs
- Counting Complexity

# Oracle TMs and Oracle Classes

## Definition

A Turing Machine  $M^?$  with *oracle* is a multi-string deterministic TM that has a special string, called **query string**, and three special states:  $q_?$  (*query state*), and  $q_{YES}$ ,  $q_{NO}$  (*answer states*). Let  $A \subseteq \Sigma^*$  be an arbitrary language. The computation of oracle machine  $M^A$  proceeds like an ordinary TM except for transitions from the query state:

From the  $q_?$  moves to either  $q_{YES}$ ,  $q_{NO}$ , depending on whether the current query string is in  $A$  or not.

- The answer states allow the machine to use this answer to its further computation.
- The computation of  $M^?$  with oracle  $A$  on input  $x$  is denoted as  $M^A(x)$ .

# Oracle TMs and Oracle Classes

## Definition

Let  $\mathcal{C}$  be a time complexity class (deterministic or nondeterministic).

Define  $\mathcal{C}^A$  to be the class of all languages decided by machines of the same sort and time bound as in  $\mathcal{C}$ , only that the machines have now oracle  $A$ . Also, we define:  $\mathcal{C}_1^{\mathcal{C}_2} = \bigcup_{L \in \mathcal{C}_2} \mathcal{C}_1^L$ .

For example,  $\mathbf{P}^{\mathbf{NP}} = \bigcup_{L \in \mathbf{NP}} \mathbf{P}^L$ . Note that  $\mathbf{P}^{\mathbf{SAT}} = \mathbf{P}^{\mathbf{NP}}$ .

## Theorem

There exists an oracle  $A$  for which  $\mathbf{P}^A = \mathbf{NP}^A$

## Proof

Take  $A$  to be a **PSPACE**-complete language. Then:

$$\mathbf{PSPACE} \subseteq \mathbf{P}^A \subseteq \mathbf{NP}^A \subseteq \mathbf{NPSPACE} \subseteq \mathbf{PSPACE}. \quad \square$$

# Oracle TMs and Oracle Classes

## Theorem

There exists an oracle  $B$  for which  $\mathbf{P}^B \neq \mathbf{NP}^B$

## Proof:

Th.14.5, p.340-342 [1]

- We will find a language  $L \in \mathbf{NP}^B \setminus \mathbf{P}^B$ .
- Let  $L = \{1^n \mid \exists x \in B \text{ with } |x| = n\}$ .
- $L \in \mathbf{NP}^B$  (why?)
- We will define the oracle  $B \subseteq \{0, 1\}^*$  such that  $L \notin \mathbf{P}^B$ :

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- $L \in \mathbf{NP}^B$  (why?)
- We will define the oracle  $B \subseteq \{0, 1\}^*$  such that  $L \notin \mathbf{P}^B$ :
- Let  $M_1^?, M_2^?, \dots$  an enumeration of all PDTMs with oracle, such that every machine appears *infinitely many* times in the enumeration.
- We will define  $B$  iteratively:  $B_0 = \emptyset$ , and  $B = \bigcup_{i \geq 0} B_i$ .
- In  $i^{\text{th}}$  stage, we have defined  $B_{i-1}$ , the set of all strings in  $B$  with length  $< i$ .
- Let also  $X$  the set of **exceptions**.

**Proof** (*cont'd*):

- We simulate  $M_i^B(1^i)$  for  $i^{\log i}$  steps.
- How do we answer the oracle questions “ $Is x \in B$ ”?



**Proof** (*cont'd*):

- We simulate  $M_i^B(1^i)$  for  $i^{\log i}$  steps.
- How do we answer the oracle questions “Is  $x \in B$ ”?
- **If**  $|x| < i$ , we look for  $x$  in  $B_{i-1}$ .
- $\rightarrow$  **If**  $x \in B_{i-1}$ ,  $M_i^B$  goes to  $q_{YES}$   
 $\rightarrow$  **Else**  $M_i^B$  goes to  $q_{NO}$
- **If**  $|x| \geq i$ ,  $M_i^B$  goes to  $q_{NO}$ , and  $x \rightarrow X$ .

**Proof** (*cont'd*):

- We simulate  $M_i^B(1^i)$  for  $i^{\log i}$  steps.
- How do we answer the oracle questions “Is  $x \in B$ ”?
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 $\rightarrow$  **Else**  $M_i^B$  goes to  $q_{NO}$
- **If**  $|x| \geq i$ ,  $M_i^B$  goes to  $q_{NO}$ , and  $x \rightarrow X$ .
- Suppose that after at most  $i^{\log i}$  steps the machine *rejects*.
  - Then we define  $B_i = B_{i-1} \cup \{x \in \{0, 1\}^* : |x| = i, x \notin X\}$   
 so  $1^i \in L$ , and  $L(M_i^B) \neq L$ .
  - Why  $\{x \in \{0, 1\}^* : |x| = i, x \notin X\} \neq \emptyset$  ? ?
- If the machine *accepts*, we define  $B_i = B_{i-1}$ , so that  $1^i \notin L$ .
- If the machine fails to halt in the allotted time, we set  
 $B_i = B_{i-1}$ , but we know that the same machine will appear in  
 the enumeration with an index sufficiently large. □

# The Limits of Diagonalization

- As we saw, an oracle can transfer us to an alternative computational “*universe*”.  
(We saw a universe where  $\mathbf{P} = \mathbf{NP}$ , and another where  $\mathbf{P} \neq \mathbf{NP}$ )
- Diagonalization is a technique that relies in the facts that:
  - **TMs are (effectively) represented by strings.**
  - **A TM can simulate another without much overhead in time/space.**
- So, diagonalization or any other proof technique relies only on these two facts, holds also for *every* oracle.
- Such results are called **relativizing results**.  
E.g.,  $\mathbf{P}^A \subseteq \mathbf{NP}^A$ , for every  $A \in \{0, 1\}^*$ .
- The above two theorems indicate that  $\mathbf{P}$  vs.  $\mathbf{NP}$  is a **nonrelativizing** result, so diagonalization and any other relativizing method doesn't suffice to prove it.

# The Classes $\mathbf{P}^{\mathbf{NP}}$ and $\mathbf{FP}^{\mathbf{NP}}$

- $\mathbf{P}^{\mathbf{SAT}}$  is the class of languages decided in pol time with a SAT oracle.
  - Polynomial number of queries
  - Queries computed adaptively
- SAT is **NP**-complete  $\Rightarrow \mathbf{P}^{\mathbf{SAT}} = \mathbf{P}^{\mathbf{NP}}$
- $\mathbf{FP}^{\mathbf{NP}}$  is the class of **functions** that can be computed by a pol-time TM with a SAT oracle.
- We will try to determine the complexity of the Traveling Salesman Problem (TSP):
- Goal:  $\text{MAX OUTPUT} \leq_m^P \text{MAX-WEIGHT SAT} \leq_m^P \text{TSP}$

# FP<sup>NP</sup>-complete Problems

## MAX OUTPUT Definition

Given NTM  $N$ , with input  $1^n$ , which halts after  $\mathcal{O}(n)$ , with output a string of length  $n$ . Which is the largest output, of any computation of  $N$  on  $1^n$ ?

## Theorem

MAX OUTPUT is **FP<sup>NP</sup>**-complete.

## Proof:

- MAX OUTPUT  $\in$  **FP<sup>NP</sup>**.
- Let  $F : \Sigma^* \rightarrow \Sigma^* \in$  **FP<sup>NP</sup>**  $\Rightarrow \exists$  poly-time TM  $M^?$ , s.t.  
 $M^{\text{SAT}}(x) = F(x)$
- We'll show:  $F \leq_m^p$  MAX OUTPUT:

# FP<sup>NP</sup>-complete Problems

## Proof (cont'd):

- Reductions  $R$  and  $S$  (*log space computable*) s.t.:
  - $\forall x, R(x)$  is a instance of MAX OUTPUT
  - $S(\text{max output of } R(x)) \rightarrow F(x)$

NTM  $N$ :

Let  $n = p^2(|x|)$ ,  $p(\cdot)$ , is the poly bound of SAT.

$N(1^n)$  generates  $x$  on a string.

$M^{\text{SAT}}$  query state ( $\phi_1$ ):

- If  $z_1 = 0$  ( $\phi_1$  unsat'd), then continue from  $q_{NO}$ .
- If  $z_1 = 1$  ( $\phi_1$  sat'd), then guess assignment  $T_1$ :
  - If test succeeds, continue from  $q_{YES}$ .
  - If test fails, output= $0^n$  and **halt**. (Unsuccessful computation)

Continue to all guesses ( $z_i$ ), and **halt**, with output= $\underbrace{z_1 z_2 \dots 00}_n$

(*Successful computation*)

# FP<sup>NP</sup>-complete Problems

## Proof (cont'd):

We claim that the successful computation that outputs the largest integer, correspond to a correct simulation:

Let  $j$  the smallest integer, s.t.:  $z_j = 0$ , while  $\phi_j$  was satisfiable.

Then,  $\exists$  another successful computation of  $N$ , s.t.:  $z_j = 1$ .

The computations agree to the first  $j - 1$  digits,  $\Rightarrow$  the  $2^{nd}$  represents a larger number.

The  $S$  part:  $F(x)$  can be read off the end of the largest output of  $N$ . □

# $FP^{NP}$ -complete Problems

## MAX-WEIGHT SAT Definition

Given a set of clauses, each with an integer weight, find the truth assignment that satisfies a set of clauses with the most total weight.



# FP<sup>NP</sup>-complete Problems

## MAX-WEIGHT SAT Definition

Given a set of clauses, each with an integer weight, find the truth assignment that satisfies a set of clauses with the most total weight.

## Theorem

MAX-WEIGHT SAT is **FP<sup>NP</sup>**-complete.

## Proof:

MAX-WEIGHT SAT is in **FP<sup>NP</sup>**: By *binary search*, and a SAT oracle, we can find the largest possible total weight of satisfied clauses, and then, by setting the variables 1-1, the truth assignment that achieves it.

MAX OUTPUT  $\leq_m^P$  MAX-WEIGHT SAT:

# FP<sup>NP</sup>-complete Problems

## Proof (cont.):

- $NTMN(1^n) \rightarrow \phi(N, m)$ :  
Any satisfying truth assignment of  $\phi(N, m) \rightarrow$  legal comp. of  $N(1^n)$
- Clauses are given a huge weight ( $2^n$ ), so that any t.a. that aspires to be optimum satisfy all clauses of  $\phi(N, m)$ .
- Add more clauses:  $(y_i): i = 1, ..n$  with weight  $2^{n-i}$ .
- Now, optimum t.a. must *not* represent any legal computation, but this which produces the *largest* possible output value.
- *S* part: From optimum t.a. of the resulting expression (or the weight), we can recover the optimum output of  $N(1^n)$ .



# $\text{FP}^{\text{NP}}$ -complete Problems

And the main result:

Theorem

TSP is  $\text{FP}^{\text{NP}}$ -complete.

# The Class $\mathbf{P}^{\mathbf{NP}[\log n]}$

## Definition

$\mathbf{P}^{\mathbf{NP}[\log n]}$  is the class of all languages decided by a polynomial time oracle machine, which on input  $x$  asks a total of  $\mathcal{O}(\log |x|)$  SAT queries.

- $\mathbf{FP}^{\mathbf{NP}[\log n]}$  is the corresponding class of functions.

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## CLIQUE SIZE Definition

Given a graph, determine the size of his *largest* clique.

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## CLIQUE SIZE Definition

Given a graph, determine the size of his *largest* clique.

## Theorem

CLIQUE SIZE is  $\mathbf{FP}^{\mathbf{NP}[\log n]}$ -complete.

# Conclusion

- ①  $TSP_{(D)}$  is **NP**-complete.
- ②  $TSP$  is **FP<sup>NP</sup>**-complete.

And now,

- $P^{NP} \rightarrow NP^{NP}$  ?
- Oracles for  $NP^{NP}$  ?

# The Polynomial Hierarchy

## Polynomial Hierarchy Definition

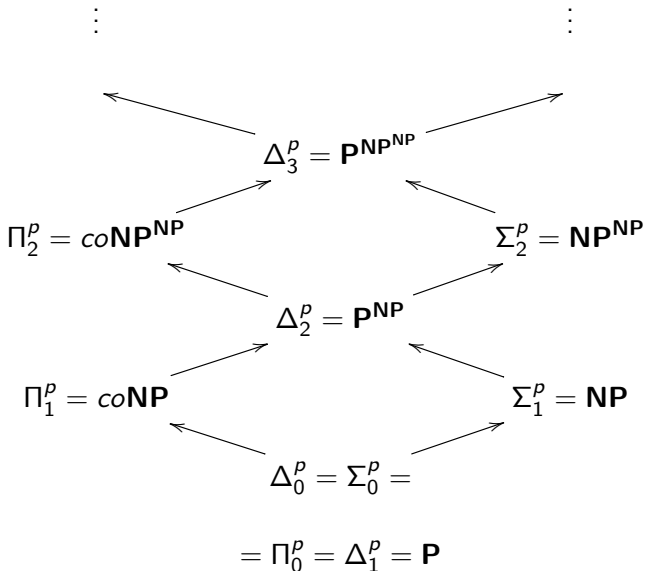
- $\Delta_0^P = \Sigma_0^P = \Pi_0^P = \mathbf{P}$
- $\Delta_{i+1}^P = \mathbf{P}^{\Sigma_i^P}$
- $\Sigma_{i+1}^P = \mathbf{NP}^{\Sigma_i^P}$
- $\Pi_{i+1}^P = \mathbf{coNP}^{\Sigma_i^P}$
- 

$$\mathbf{PH} \equiv \bigcup_{i \geq 0} \Sigma_i^P$$

- $\Sigma_0^P = \mathbf{P}$
- $\Delta_1^P = \mathbf{P}$ ,  $\Sigma_1^P = \mathbf{NP}$ ,  $\Pi_1^P = \mathbf{coNP}$
- $\Delta_2^P = \mathbf{P}^{\mathbf{NP}}$ ,  $\Sigma_2^P = \mathbf{NP}^{\mathbf{NP}}$ ,  $\Pi_2^P = \mathbf{coNP}^{\mathbf{NP}}$



## The Polynomial Hierarchy



# Basic Theorems

## Theorem

Let  $L$  be a language, and  $i \geq 1$ .  $L \in \Sigma_i^P$  iff there is a polynomially balanced relation  $R$  such that the language  $\{x; y : (x, y) \in R\}$  is in  $\Pi_{i-1}^P$  and

$$L = \{x : \exists y, s.t. : (x, y) \in R\}$$

## Proof (by Induction)

- For  $i = 1$

$\{x; y : (x, y) \in R\} \in \mathbf{P}$ , so  $L = \{x | \exists y : (x, y) \in R\} \in \mathbf{NP} \checkmark$

- For  $i > 1$

If  $\exists R \in \Pi_{i-1}^P$ , we must show that  $L \in \Sigma_i^P \Rightarrow$

$\exists$  NTM with  $\Sigma_{i-1}^P$  oracle: NTM( $x$ ) guesses a  $y$  and asks  $\Pi_{i-1}^P$  oracle whether  $(x, y) \notin R$ .

# Basic Theorems

## Proof (cont.)

- If  $L \in \Sigma_i^P$ , we must show the existence of  $R$ .

$L \in \Sigma_i^P \Rightarrow \exists$  NTM  $M^K$ ,  $K \in \Sigma_{i-1}^P$ , which decides  $L$ .

$K \in \Sigma_{i-1}^P \Rightarrow \exists S \in \Pi_{i-2}^P : (z \in K \Leftrightarrow \exists w : (z, w) \in S)$

We must describe a relation  $R$  (we know:  $x \in L \Leftrightarrow$  accepting comp of  $M^K(x)$ )

Query Steps: “yes”  $\rightarrow z_i$  has a certificate  $w_i$  st  $(z_i, w_i) \in S$ .

So,  $R(x) =$  “ $(x, y) \in R$  iff  $y$  records an accepting computation of  $M^?$  on  $x$ , together with a certificate  $w_i$  for each **yes** query  $z_i$  in the computation.”

We must show  $\{x; y : (x, y) \in R\} \in \Pi_{i-1}^P$ .

# Basic Theorems

## Corollary

Let  $L$  be a language, and  $i \geq 1$ .  $L \in \Pi_i^P$  iff there is a polynomially balanced relation  $R$  such that the language  $\{x; y : (x, y) \in R\}$  is in  $\Sigma_{i-1}^P$  and

$$L = \{x : \forall y, |y| \leq |x|^k, \text{ s.t. } : (x, y) \in R\}$$

## Corollary

Let  $L$  be a language, and  $i \geq 1$ .  $L \in \Sigma_i^P$  iff there is a polynomially balanced, polynomially-time decidable  $(i + 1)$ -ary relation  $R$  such that:

$$L = \{x : \exists y_1 \forall y_2 \exists y_3 \dots Q y_i, \text{ s.t. } : (x, y_1, \dots, y_i) \in R\}$$

where the  $i^{\text{th}}$  quantifier  $Q$  is  $\forall$ , if  $i$  is even, and  $\exists$ , if  $i$  is odd.

# Basic Theorems

## Theorem

If for some  $i \geq 1$ ,  $\Sigma_i^P = \Pi_i^P$ , then for all  $j > i$ :

$$\Sigma_j^P = \Pi_j^P = \Delta_j^P = \Sigma_i^P$$

Or, the polynomial hierarchy *collapses* to the  $i^{\text{th}}$  level.

## Proof

It suffices to show that:  $\Sigma_i^P = \Pi_i^P \Rightarrow \Sigma_{i+1}^P = \Sigma_i^P$

Let  $L \in \Sigma_{i+1}^P \Rightarrow \exists R \in \Pi_i^P: L = \{x \mid \exists y : (x, y) \in R\}$

Since  $\Pi_i^P = \Sigma_i^P \Rightarrow R \in \Sigma_i^P$

$(x, y) \in R \Leftrightarrow \exists z : (x, y, z) \in S, S \in \Pi_{i-1}^P.$

Thus,  $x \in L \Leftrightarrow \exists y; z : (x, y, z) \in S, S \in \Pi_{i-1}^P$ , which means

$L \in \Sigma_i^P.$

# Basic Theorems

## Corollary

If  $\mathbf{P}=\mathbf{NP}$ , or even  $\mathbf{NP}=\mathbf{coNP}$ , the Polynomial Hierarchy collapses to the first level.

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## QSAT<sub>i</sub>; Definition

Given expression  $\phi$ , with Boolean variables partitioned into  $i$  sets  $X_i$ , is  $\phi$  satisfied by the overall truth assignment of the expression:

$$\exists X_1 \forall X_2 \exists X_3 \dots Q X_i \phi$$

, where  $Q$  is  $\exists$  if  $i$  is *odd*, and  $\forall$  if  $i$  is *even*.

## Theorem

For all  $i \geq 1$  QSAT<sub>i</sub> is  $\Sigma_i^P$ -complete.

# Basic Theorems

## Theorem

If there is a **PH**-complete problem, then the polynomial hierarchy collapses to some finite level.

## Proof

Let  $L$  is **PH**-complete.

Since  $L \in \mathbf{PH}$ ,  $\exists i \geq 0 : L \in \Sigma_i^P$ .

But any  $L' \in \Sigma_{i+1}^P$  reduces to  $L$ . Since PH is closed under reductions, we imply that  $L' \in \Sigma_i^P$ , so  $\Sigma_i^P = \Sigma_{i+1}^P$ .



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## Theorem

### **PH** $\subseteq$ **PSPACE**

- **PH**  $\stackrel{?}{=} \mathbf{PSPACE}$  (**Open**). If it was, then **PH** has complete problems, so it collapses to some finite level.

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# Warmup: Polynomial Identity Testing

- ① Two polynomials are equal if they have the same coefficients for corresponding powers of their variable.
- ② A polynomial is *identically zero* if all its coefficients are equal to the additive identity element.
- ③ How we can test if a polynomial is identically zero?
- ④ We can choose uniformly at random  $r_1, \dots, r_n$  from a set  $S \subseteq \mathbb{F}$ .
- ⑤ We are wrong with a probability at most:

Theorem (Schwartz-Zippel Lemma)

Let  $Q(x_1, \dots, x_n) \in \mathbb{F}[x_1, \dots, x_n]$  be a multivariate polynomial of total degree  $d$ . Fix any finite set  $S \subseteq \mathbb{F}$ , and let  $r_1, \dots, r_n$  be chosen independently and uniformly at random from  $S$ . Then:

$$\Pr[Q(r_1, \dots, r_n) = 0 | Q(x_1, \dots, x_n) \neq 0] \leq \frac{d}{|S|}$$



# Warmup: Polynomial Identity Testing

## Proof:

(By Induction on  $n$ )

- For  $n = 1$ :  $\Pr[Q(r) = 0 | Q(x) \neq 0] \leq d/|S|$
- For  $n$ :

$$Q(x_1, \dots, x_n) = \sum_{i=0}^k x_1^i Q_i(x_2, \dots, x_n)$$

where  $k \leq d$  is the *largest* exponent of  $x_1$  in  $Q$ .

$\deg(Q_k) \leq d - k \Rightarrow \Pr[Q_k(r_2, \dots, r_n) = 0] \leq (d - k)/|S|$

Suppose that  $Q_k(r_2, \dots, r_n) \neq 0$ . Then:

$$q(x_1) = Q(x_1, r_2, \dots, r_n) = \sum_{i=0}^k x_1^i Q_i(r_2, \dots, r_n)$$

$\deg(q(x_1)) = k$ , and  $q(x_1) \neq 0$ !

# Warmup: Polynomial Identity Testing

**Proof** (cont'd):

The base case now implies that:

$$\Pr[q(r_1) = Q(r_1, \dots, r_n) = 0] \leq k/|S|$$

Thus, we have shown the following two equalities:

$$\Pr[Q_k(r_2, \dots, r_n) = 0] \leq \frac{d-k}{|S|}$$

$$\Pr[Q_k(r_1, r_2, \dots, r_n) = 0 | Q_k(r_2, \dots, r_n) \neq 0] \leq \frac{k}{|S|}$$

Using the following identity:  $\Pr[\mathcal{E}_1] \leq \Pr[\mathcal{E}_1|\overline{\mathcal{E}}_2] + \Pr[\mathcal{E}_2]$  we obtain that the requested probability is no more than the sum of the above, which proves our theorem!  $\square$





# BPP Class

## Definition (Alternative Definition of BPP)

A language  $L \in \mathbf{BPP}$  if there exists a poly-time TM  $M$  and a polynomial  $p \in \text{poly}(n)$ , such that for every  $x \in \{0, 1\}^*$ :

$$\Pr_{r \in \{0,1\}^{p(n)}} [M(x, r) = L(x)] \geq \frac{2}{3}$$

- $\mathbf{P} \subseteq \mathbf{BPP}$
- $\mathbf{BPP} \subseteq \mathbf{EXP}$
- The “ $\mathbf{P}$  vs  $\mathbf{BPP}$ ” question.

# Quantifier Characterizations

- Proper formalism (*Zachos et al.*):

## Definition (Majority Quantifier)

Let  $R : \{0, 1\}^* \times \{0, 1\}^* \rightarrow \{0, 1\}$  be a predicate, and  $\varepsilon$  a rational number, such that  $\varepsilon \in (0, \frac{1}{2})$ . We denote by  $(\exists^+ y, |y| = k)R(x, y)$  the following predicate:

“There exist at least  $(\frac{1}{2} + \varepsilon) \cdot 2^k$  strings  $y$  of length  $m$  for which  $R(x, y)$  holds.”

We call  $\exists^+$  the *overwhelming majority* quantifier.

- $\exists_r^+$  means that the fraction  $r$  of the possible certificates of a certain length satisfy the predicate for the certain input.

# Quantifier Characterizations

## Definition

We denote as  $\mathcal{C} = (Q_1/Q_2)$ , where  $Q_1, Q_2 \in \{\exists, \forall, \exists^+\}$ , the class  $\mathcal{C}$  of languages  $L$  satisfying:

- $x \in L \Rightarrow Q_1y R(x, y)$
- $x \notin L \Rightarrow Q_2y \neg R(x, y)$

- **P** =  $(\forall/\forall)$
- **NP** =  $(\exists/\forall)$
- **coNP** =  $(\forall/\exists)$
- **BPP** =  $(\exists^+/\exists^+) = \text{coBPP}$

# RP Class

- In the same way, we can define classes that contain problems with one-sided error:

## Definition

The class **RTIME** $[T(n)]$  contains every language  $L$  for which there exists a PTM  $M$  running in  $\mathcal{O}(T(|x|))$  time such that:

- $x \in L \Rightarrow \Pr[M(x) = 1] \geq \frac{2}{3}$
- $x \notin L \Rightarrow \Pr[M(x) = 0] = 1$

We define

$$\mathbf{RP} = \bigcup_{c \in \mathbb{N}} \mathbf{RTIME}[n^c]$$

- Similarly we define the class **coRP**.



# Quantifier Characterizations

- **RP**  $\subseteq$  **NP**, since every accepting “branch” is a certificate!
- **RP**  $\subseteq$  **BPP**, **coRP**  $\subseteq$  **BPP**
- **RP** =  $(\exists^+/\forall)$

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Theorem (Decisive Characterization of BPP)

$$\mathbf{BPP} = (\exists^+/\exists^+) = (\exists^+\forall/\forall\exists^+) = (\forall\exists^+/\exists^+\forall)$$

# Quantifier Characterizations

## Proof:

- Let  $L \in \mathbf{BPP}$ . Then, by definition, there exists a polynomial-time computable predicate  $Q$  and a polynomial  $q$  such that for all  $x$ 's of length  $n$ :

$$x \in L \Rightarrow \exists^+ y Q(x, y)$$

$$x \notin L \Rightarrow \exists^+ y \neg Q(x, y)$$

## Swapping Lemma

- i  $\forall y \exists^+ z R(x, y, z) \Rightarrow \exists^+ C \forall y \bigvee_{z \in C} R(x, y, z)$
- ii  $\forall z \exists^+ y R(x, y, z) \Rightarrow \forall C \exists^+ y \bigwedge_{z \in C} R(x, y, z)$

- By the above Lemma:  $x \in L \Rightarrow \exists^+ z Q(x, z) \Rightarrow \forall y \exists^+ z Q(x, y \oplus z) \Rightarrow \exists^+ C \forall y [\exists(z \in C) Q(x, y \oplus z)]$ , where  $C$  denotes (as in the Swapping's Lemma formulation) a set of  $q(n)$  strings, each of length  $q(n)$ .

# Quantifier Characterizations

## Proof (cont'd):

- On the other hand,  $x \notin L \Rightarrow \exists^+ y \neg Q(x, z) \Rightarrow \forall z \exists^+ y \neg Q(x, y \oplus z) \Rightarrow \forall C \exists^+ y [\forall (z \in C) \neg Q(x, y \oplus z)]$ .
- Now, we only have to assure that the appeared predicates  $\exists z \in C Q(x, y \oplus z)$  and  $\forall z \in C \neg Q(x, y \oplus z)$  are computable in polynomial time
- Recall that in Swapping Lemma's formulation we demanded  $|C| \leq p(n)$  and that for each  $v \in C : |v| = p(n)$ . This means that we seek if a string of polynomial length *exists*, or if the predicate holds *for all* such strings in a set with polynomial cardinality, procedure which can be surely done in polynomial time.

# Quantifier Characterizations

## Proof (cont'd):

- Conversely, if  $L \in (\exists^+ \forall / \forall \exists^+)$ , for each string  $w$ ,  $|w| = 2p(n)$ , we have  $w = w_1 w_2$ ,  $|w_1| = |w_2| = p(n)$ . Then:  
 $x \in L \Rightarrow \exists^+ y \forall z R(x, y, z) \Rightarrow \exists^+ w R(x, w_1, w_2)$   
 $x \notin L \Rightarrow \forall y \exists^+ z R(x, y, z) \Rightarrow \exists^+ w \neg R(x, w_1, w_2)$
- So,  $L \in \mathbf{BPP}$ .  $\square$
- The above characterization is *decisive*, in the sense that if we replace  $\exists^+$  with  $\exists$ , the two predicates are still complementary (i.e.  $R_1 \Rightarrow \neg R_2$ ), so they still define a complexity class.
- In the above characterization of **BPP**, if we replace  $\exists^+$  with  $\exists$ , we obtain very easily a well-known result:

Corollary (Sipser-Gács Theorem)

$$\mathbf{BPP} \subseteq \Sigma_2^P \cap \Pi_2^P$$

# BPP and PH

Theorem (Sipser-Gács)

$$\mathbf{BPP} \subseteq \Sigma_2^P \cap \Pi_2^P$$

**Proof** (*Lautemann*)

Because  $\text{coBPP} = \mathbf{BPP}$ , we prove only  $\mathbf{BPP} \subseteq \Sigma_2^P$ .

Let  $L \in \mathbf{BPP}$  ( $L$  is accepted by “clear majority”).

For  $|x| = n$ , let  $A(x) \subseteq \{0, 1\}^{p(n)}$  be the set of *accepting* computations.

We have:

- $x \in L \Rightarrow |A(x)| \geq 2^{p(n)} \left(1 - \frac{1}{2^n}\right)$
- $x \notin L \Rightarrow |A(x)| \leq 2^{p(n)} \left(\frac{1}{2^n}\right)$

Let  $U$  be the set of all bit strings of length  $p(n)$ .

For  $a, b \in U$ , let  $a \oplus b$  be the XOR:

$a \oplus b = c \Leftrightarrow c \oplus b = a$ , so “ $\oplus b$ ” is 1-1.



# BPP and PH

## Proof (cont.)

For  $t \in U$ ,  $A(x) \oplus t = \{a \oplus t : a \in A(x)\}$  (translation of  $A(x)$  by  $t$ ). We imply that:  $|A(x) \oplus t| = |A(x)|$

If  $x \in L$ , consider a *random* (drawing  $p^2(n)$  bits) sequence of translations:  $t_1, t_2, \dots, t_{p(n)} \in U$ .

For  $b \in U$ , these translations *cover*  $b$ , if  $b \in A(x) \oplus t_j$ ,  $j \leq p(n)$ .

$$b \in A(x) \oplus t_j \Leftrightarrow b \oplus t_j \in A(x) \Rightarrow \Pr[b \notin A(x) \oplus t_j] = \frac{1}{2^n}$$

$$\Pr[b \text{ is not covered by any } t_j] = 2^{-np(n)}$$

$$\Pr[\exists \text{ point that is not covered}] \leq 2^{-np(n)} |U| = 2^{-(n-1)p(n)}$$

# BPP and PH

## Proof (cont.)

So,  $T = (t_1, \dots, t_{p(n)})$  has a positive probability that it covers all of  $U$ .

If  $x \notin L$ ,  $|A(x)|$  is exp small, and (for large  $n$ ) there's not  $T$  that cover all  $U$ .

$(x \in L) \Leftrightarrow (\exists T \text{ that cover all } U)$

So,

$$L = \{x \mid \exists (T \in \{0, 1\}^{p^2(n)}) \forall (b \in U) \exists (j \leq p(n)) : b \oplus t_j \in A(x)\}$$

which is precisely the form of languages in  $\Sigma_2\mathbf{P}$ .

The last existential quantifier  $(\exists(j \leq p(n))\dots)$  affects only polynomially many possibilities, so it doesn't "count" (can be tested in polynomial time by trying all  $t_j$ 's).



# ZPP Class

- And now something completely different:
- What is the random variable was the running time and not the output?

# ZPP Class

- And now something completely different:
- What is the random variable was the running time and not the output?
- We say that  $M$  has expected running time  $T(n)$  if the expectation  $\mathbf{E}[T_{M(x)}]$  is at most  $T(|x|)$  for every  $x \in \{0, 1\}^*$ . ( $T_{M(x)}$  is the running time of  $M$  on input  $x$ , and it is a **random variable!**)

## Definition

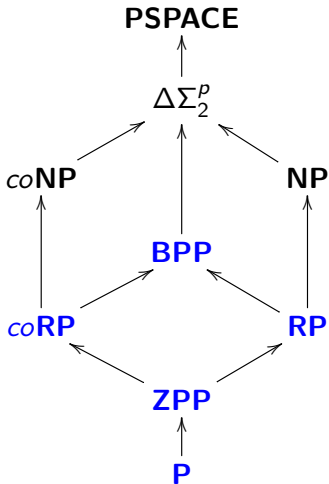
The class **ZTIME** $[T(n)]$  contains all languages  $L$  for which there exists a machine  $M$  that runs in an expected time  $\mathcal{O}(T(|x|))$  such that for every input  $x \in \{0, 1\}^*$ , whenever  $M$  halts on  $x$ , the output  $M(x)$  it produces is exactly  $L(x)$ . We define:

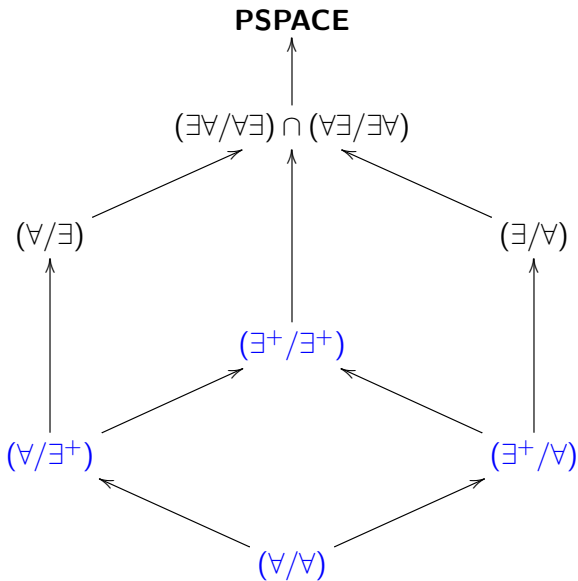
$$\mathbf{ZPP} = \bigcup_{c \in \mathbb{N}} \mathbf{ZTIME}[n^c]$$

# ZPP Class

- The output of a **ZPP** machine is always correct!
- The problem is that we aren't sure about the running time.
- We can easily see that **ZPP** = **RP**  $\cap$  **coRP**.
- The next Hasse diagram summarizes the previous inclusions:  
 (Recall that  $\Delta\Sigma_2^P = \Sigma_2^P \cap \Pi_2^P = \mathbf{NP}^{\mathbf{NP}} \cap \mathbf{coNP}^{\mathbf{NP}}$ )

# BPP and PH





# Error Reduction for BPP

Theorem (Error Reduction for BPP)

*Let  $L \subseteq \{0, 1\}^*$  be a language and suppose that there exists a poly-time PTM  $M$  such that for every  $x \in \{0, 1\}^*$ :*

$$\Pr[M(x) = L(x)] \geq \frac{1}{2} + |x|^{-c}$$

*Then, for every constant  $d > 0$ ,  $\exists$  poly-time PTM  $M'$  such that for every  $x \in \{0, 1\}^*$ :*

$$\Pr[M'(x) = L(x)] \geq 1 - 2^{-|x|^d}$$



**Proof:** The machine  $M'$  does the following:

- Run  $M(x)$  for every input  $x$  for  $k = 8|x|^{2c+d}$  times, and obtain outputs  $y_1, y_2, \dots, y_k \in \{0, 1\}$ .
- If the majority of these outputs is 1, return 1
- Otherwise, return 0.

We define the r.v.  $X_i$  for every  $i \in [k]$  to be 1 if  $y_i = L(x)$  and 0 otherwise.

$X_1, X_2, \dots, X_k$  are independent Boolean r.v.'s, with:

$$\mathbf{E}[X_i] = \Pr[X_i = 1] \geq p = \frac{1}{2} + |x|^{-c}$$

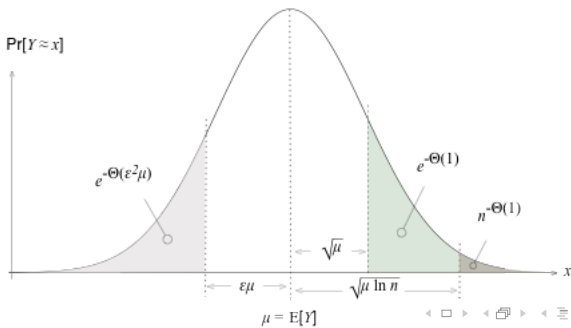
Applying a Chernoff Bound we obtain:

$$\Pr \left[ \left| \sum_{i=1}^k X_i - pk \right| > \delta pk \right] < e^{-\frac{\delta^2}{4} pk} = e^{-\frac{1}{4|x|^{2c}} \frac{1}{2} 8|x|^{2c+d}} \leq 2^{-|x|^d}$$



# Intermission: Chernoff Bounds

- *How many* samples do we need in order to estimate  $\mu$  up to an error of  $\pm\varepsilon$  with *probability* at least  $1 - \delta$ ?
- Chernoff Bound tells us that this number is  $\mathcal{O}(\rho/\varepsilon^2)$ , where  $\rho = \log(1/\delta)$ .
- The probability that  $k$  is  $\rho\sqrt{n}$  far from  $\mu n$  decays **exponentially** with  $\rho$ .



# Intermission: Chernoff Bounds

$$\Pr \left[ \sum_{i=1}^n X_i \geq (1 + \delta)\mu \right] \leq \left[ \frac{e^\delta}{(1 + \delta)^{1+\delta}} \right]^\mu$$

$$\Pr \left[ \sum_{i=1}^n X_i \leq (1 - \delta)\mu \right] \leq \left[ \frac{e^{-\delta}}{(1 - \delta)^{1-\delta}} \right]^\mu$$

Other useful form is:

$$\Pr \left[ \left| \sum_{i=1}^n X_i - \mu \right| \geq c\mu \right] \leq 2e^{-\min\{c^2/4, c/2\} \cdot \mu}$$

- This probability is bounded by  $2^{-\Omega(\mu)}$ .

# Error Reduction for BPP

- From the above we can obtain the following interesting corollary:

## Corollary

For  $c > 0$ , let  $\mathbf{BPP}_{1/2+n^{-c}}$  denote the class of languages  $L$  for which there is a polynomial-time PTM  $M$  satisfying  $\Pr[M(x) = L(x)] \geq 1/2 + |x|^{-c}$  for every  $x \in \{0, 1\}^*$ . Then:

$$\mathbf{BPP}_{1/2+n^{-c}} = \mathbf{BPP}$$

- Obviously,  $\exists^+ = \exists^+_{1/2+\varepsilon} = \exists^+_{2/3} = \exists^+_{3/4} = \exists^+_{0.99} = \exists^+_{1-2^{-p(|x|)}}$

# Semantic vs. Syntactic Classes

- Every NPTM defines some language in **NP**:  
 $x \in L \Leftrightarrow \# \text{accepting paths} \neq 0$
- We can get an effective enumeration of all NPTMs, each deciding an **NP** language.
- But not every NPTM decides a language in **RP**:  
 e.g., the NPTM that has *exactly one* accepting path.
- In this case, there is no way to tell whether the machine will always halt with the certified output. We call these classes **semantic**.
- So we have:
  - **Syntactic Classes** (like **P**, **NP**)
  - **Semantic Classes** (like **RP**, **BPP**,  **$\text{NP} \cap \text{coNP}$** , **TFNP**)

# Complete Problems for BPP?

- Any syntactic class has a “free” complete problem:

$$\{\langle M, x \rangle : M \in \mathcal{M} \ \& \ M(x) = \text{“yes”}\}$$

where  $\mathcal{M}$  is the class of TMs of the variant that defines the class

- In semantic classes, this complete language is usually *undecidable* (Rice’s Theorem).
- The defining property of **BPTIME** machines is **semantic**!
- If finally **P = BPP**, then **BPP** will have complete problems!!
- For the same reason, in semantic classes we cannot prove Hierarchy Theorems using Diagonalization.

# The Class PP

## Definition

A language  $L \in \mathbf{PP}$  if there exists an NPTM  $M$ , such that for every  $x \in \{0, 1\}^*$ :  $x \in L$  if and only if *more than half* of the computations of  $M$  on input  $x$  accept.

- Or, equivalently:

## Definition

A language  $L \in \mathbf{PP}$  if there exists a poly-time TM  $M$  and a polynomial  $p \in \text{poly}(n)$ , such that for every  $x \in \{0, 1\}^*$ :

$$x \in L \Leftrightarrow \left| \left\{ y \in \{0, 1\}^{p(|x|)} : M(x, y) = 1 \right\} \right| \geq \frac{1}{2} \cdot 2^{p(|x|)}$$

# The Class PP

- The defining property of **PP** is **syntactic**, any NPTM can define a language in **PP**.
- Due to the lack of a gap between the two cases, we cannot amplify the probability with polynomially many repetitions, as in the case of **BPP**.
- **PP** is closed under complement.
- A breakthrough result of R. Beigel, N. Reingold and D. Spielman is that **PP** is closed under *intersection*!



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- **PP** is closed under complement.
- A breakthrough result of R. Beigel, N. Reingold and D. Spielman is that **PP** is closed under *intersection*!
- The syntactic definition of **PP** gives the possibility for *complete problems*:
- Consider the problem MAJSAT:  
Given a Boolean Expression, is it true that the majority of the  $2^n$  truth assignments to its variables (that is, at least  $2^{n-1} + 1$  of them) satisfy it?

# The Class PP

## Theorem

*MAJSAT is **PP**-complete!*

- MAJSAT is not likely in **NP**, since the (*obvious*) certificate is not very succinct!

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## Theorem

$$\mathbf{NP} \subseteq \mathbf{PP} \subseteq \mathbf{PSPACE}$$

### Proof:

It is easy to see that **PP**  $\subseteq$  **PSPACE**:

We can simulate any **PP** machine by enumerating all strings  $y$  of length  $p(n)$  and verify whether **PP** machine accepts. The **PSPACE** machine accepts if and only if there are more than  $2^{p(n)-1}$  such  $y$ 's (by using a counter).

# The Class PP

## Proof (cont'd):

Now, for  $\mathbf{NP} \subseteq \mathbf{PP}$ , let  $A \in \mathbf{NP}$ . That is,  $\exists p \in \text{poly}(n)$  and a poly-time and balanced predicate  $R$  such that:

$$x \in A \Leftrightarrow (\exists y, |y| = p(|x|)) : R(x, y)$$

Consider the following TM:

*M* accepts input  $(x, by)$ , with  $|b| = 1$  and  $|y| = p(|x|)$ , if and only if  $R(x, y) = 1$  or  $b = 1$ .

- If  $x \in A$ , then  $\exists$  at least one  $y$  s.t.  $R(x, y)$ .  
Thus,  $\Pr[M(x) \text{ accepts}] \geq 1/2 + 2^{-(p(n)+1)}$ .
- If  $x \notin A$ , then  $\Pr[M(x) \text{ accepts}] = 1/2$ .



# Other Results

Theorem

*If  $\mathbf{NP} \subseteq \mathbf{BPP}$ , then  $\mathbf{NP} = \mathbf{RP}$ .*

# Other Results

## Theorem

If  $\mathbf{NP} \subseteq \mathbf{BPP}$ , then  $\mathbf{NP} = \mathbf{RP}$ .

### Proof:

- $\mathbf{RP}$  is closed under  $\leq_m^P$ -reducibility.
- It suffices to show that if  $\mathbf{SAT} \in \mathbf{BPP}$ , then  $\mathbf{SAT} \in \mathbf{RP}$ .
- Recall that  $\mathbf{SAT}$  has the **self-reducibility** property:  
 $\phi(x_1, \dots, x_n): \phi \in \mathbf{SAT} \Leftrightarrow (\phi|_{x_1=0} \in \mathbf{SAT} \vee \phi|_{x_1=1} \in \mathbf{SAT})$ .
- $\mathbf{SAT} \in \mathbf{BPP}$ :  $\exists$  PTM  $M$  computing  $\mathbf{SAT}$  with error probability bounded by  $2^{-|\phi|}$ .
- We can use the *self-reducibility* of  $\mathbf{SAT}$  to produce a truth assignment for  $\phi$  as follows:

# Other Results

## Proof (cont'd):

Input: A Boolean formula  $\phi$  with  $n$  variables

**If**  $M(\phi) = 0$  **then** reject  $\phi$ ;

**For**  $i = 1$  **to**  $n$

→ **If**  $M(\phi|_{x_1=\alpha_1, \dots, x_{i-1}=\alpha_{i-1}, x_i=0}) = 1$  **then** let  $\alpha_i = 0$

→ **Elseif**  $M(\phi|_{x_1=\alpha_1, \dots, x_{i-1}=\alpha_{i-1}, x_i=1}) = 1$  **then** let  $\alpha_i = 1$

→ **Else** reject  $\phi$  and halt;

**If**  $\phi|_{x_1=\alpha_1, \dots, x_n=\alpha_n} = 1$  **then** accept  $F$

**Else** reject  $F$



## Other Results

### Proof (cont'd):

Input: A Boolean formula  $\phi$  with  $n$  variables

**If**  $M(\phi) = 0$  **then reject**  $\phi$ ;

**For**  $i = 1$  **to**  $n$

→ **If**  $M(\phi|_{x_1=\alpha_1, \dots, x_{i-1}=\alpha_{i-1}, x_i=0}) = 1$  **then let**  $\alpha_i = 0$

→ **Elseif**  $M(\phi|_{x_1=\alpha_1, \dots, x_{i-1}=\alpha_{i-1}, x_i=1}) = 1$  **then let**  $\alpha_i = 1$

→ **Else reject**  $\phi$  **and halt**;

**If**  $\phi|_{x_1=\alpha_1, \dots, x_n=\alpha_n} = 1$  **then accept**  $F$

**Else reject**  $F$

- Note that  $M_1$  accepts  $\phi$  only if a t.a.  $t(x_i) = \alpha_i$  is found.
- Therefore,  $M_1$  never makes mistakes if  $\phi \notin \text{SAT}$ .
- If  $\phi \in \text{SAT}$ , then  $M$  rejects  $\phi$  on each iteration of the loop w.p.  $2^{-|\phi|}$ .
- So,  $\Pr[M_1 \text{ accepting } x] = (1 - 2^{-|\phi|})^n$ , which is greater than  $1/2$  if  $|\phi| \geq n > 1$ .  $\square$

# Relativized Results

## Theorem

Relative to a random oracle  $A$ ,  $\mathbf{P}^A = \mathbf{BPP}^A$ . That is,

$$\Pr_{A \in \{0,1\}^*} [\mathbf{P}^A = \mathbf{BPP}^A] = 1$$

Also,

- $\mathbf{BPP}^A \subsetneq \mathbf{NP}^A$ , relative to a *random* oracle  $A$ .
- There exists an  $A$  such that:  $\mathbf{P}^A \neq \mathbf{RP}^A$ .
- There exists an  $A$  such that:  $\mathbf{RP}^A \neq \mathbf{coRP}^A$
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- There exists an  $A$  such that:  $\mathbf{RP}^A \neq \mathbf{NP}^A$ .

## Corollary

There exists an  $A$  such that:

$$\mathbf{P}^A \neq \mathbf{RP}^A \neq \mathbf{NP}^A \not\subseteq \mathbf{BPP}^A$$

# Contents

- Introduction
- Turing Machines
- Undecidability
- Complexity Classes
- Oracles & Optimization Problems
- Randomized Computation
- **Non-Uniform Complexity**
- Interactive Proofs
- Counting Complexity

# Boolean Circuits

- A Boolean Circuit is a natural model of *nonuniform* computation, a generalization of hardware computational methods.
- A non-uniform computational model allows us to use a different “algorithm” to be used for every input size, in contrast to the standard (or *uniform*) Turing Machine model, where the same T.M. is used on (infinitely many) input sizes.
- Each circuit can be used for a fixed input size, which limits or model.

## Definition (Boolean circuits)

For every  $n \in \mathbb{N}$  an  $n$ -input, single output Boolean Circuit  $C$  is a directed acyclic graph with  $n$  sources and *one* sink.

- All nonsource vertices are called *gates* and are labeled with one of  $\wedge$  (and),  $\vee$  (or) or  $\neg$  (not).
- The vertices labeled with  $\wedge$  and  $\vee$  have *fan-in* (i.e. number or incoming edges) 2.
- The vertices labeled with  $\neg$  have *fan-in* 1.
- The *size* of  $C$ , denoted by  $|C|$ , is the number of vertices in it.
- For every vertex  $v$  of  $C$ , we assign a value as follows: for some input  $x \in \{0, 1\}^n$ , if  $v$  is the  $i$ -th input vertex then  $val(v) = x_i$ , and otherwise  $val(v)$  is defined recursively by applying  $v$ 's logical operation on the values of the vertices connected to  $v$ .
- The *output*  $C(x)$  is the value of the output vertex.
- The *depth* of  $C$  is the length of the longest directed path from an input node to the output node.

- To overcome the fixed input length size, we need to allow families (or sequences) of circuits to be used:

### Definition

Let  $T : \mathbb{N} \rightarrow \mathbb{N}$  be a function. A  $T(n)$ -size circuit family is a sequence  $\{C_n\}_{n \in \mathbb{N}}$  of Boolean circuits, where  $C_n$  has  $n$  inputs and a single output, and its size  $|C_n| \leq T(n)$  for every  $n$ .

- These infinite families of circuits are defined arbitrarily: There is **no** pre-defined connection between the circuits, and also we haven't any "guarantee" that we can construct them efficiently.
- Like each new computational model, we can define a complexity class on it by imposing some restriction on a *complexity measure*:

## Definition

We say that a language  $L$  is in **SIZE**( $T(n)$ ) if there is a  $T(n)$ -size circuit family  $\{C_n\}_{n \in \mathbb{N}}$ , such that  $\forall x \in \{0, 1\}^n$ :

$$x \in L \Leftrightarrow C_n(x) = 1$$

## Definition

**P**<sub>/poly</sub> is the class of languages that are decidable by polynomial size circuits families. That is,

$$\mathbf{P}_{/\text{poly}} = \bigcup_{c \in \mathbb{N}} \mathbf{SIZE}(n^c)$$

## Theorem (Nonuniform Hierarchy Theorem)

For every functions  $T, T' : \mathbb{N} \rightarrow \mathbb{N}$  with  $\frac{2^n}{n} > T'(n) > 10T(n) > n$ ,

$$\mathbf{SIZE}(T(n)) \subsetneq \mathbf{SIZE}(T'(n))$$



# Turing Machines that take advice

## Definition

Let  $T, a : \mathbb{N} \rightarrow \mathbb{N}$ . The class of languages decidable by  $T(n)$ -time Turing Machines with  $a(n)$  bits of advice, denoted

$$\mathbf{DTIME}(T(n)/a(n))$$

contains every language  $L$  such that there exists a sequence  $\{a_n\}_{n \in \mathbb{N}}$  of strings, with  $a_n \in \{0, 1\}^{a(n)}$  and a Turing Machine  $M$  satisfying:

$$x \in L \Leftrightarrow M(x, a_n) = 1$$

for every  $x \in \{0, 1\}^n$ , where on input  $(x, a_n)$  the machine  $M$  runs for at most  $\mathcal{O}(T(n))$  steps.

# Turing Machines that take advice

Theorem (Alternative Definition of  $\mathbf{P}_{/\text{poly}}$ )

$$\mathbf{P}_{/\text{poly}} = \bigcup_{c,d \in \mathbb{N}} \mathbf{DTIME}(n^c/n^d)$$

# Turing Machines that take advice

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**Proof:** ( $\subseteq$ ) Let  $L \in \mathbf{P}_{/poly}$ . Then,  $\exists \{C_n\}_{n \in \mathbb{N}} : C_{|x|} = L(x)$ .  
We can use  $C_n$ 's encoding as an advice string for each  $n$ .

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We can use  $C_n$ 's encoding as an advice string for each  $n$ .

( $\supseteq$ ) Let  $L \in \mathbf{DTIME}(n^c/n^d)$ . Then, since CVP is  $\mathbf{P}$ -complete, we construct for every  $n$  a circuit  $D_n$  such that, for  $x \in \{0, 1\}^n$ ,  $a_n \in \{0, 1\}^{a(n)}$ :

$$D_n(x, a_n) = M(x, a_n)$$

Then, let  $C_n(x) = D_n(x, a_n)$  (**We hard-wire the advice string!**)

Since  $a(n) = n^d$ , the circuits have polynomial size.  $\square$



# Uniform Families of Circuits

- We saw that  $\mathbf{P}_{/poly}$  contains an undecidable language.
- The root of this problem lies in the “weak” definition of such families, since it suffices that  $\exists$  a circuit family for  $L$ .
- We haven’t a way (or an algorithm) to construct such a family.
- So, may be useful to restric or attention to families we can construct efficiently:

## Theorem (P-Uniform Families)

*A circuit family  $\{C_n\}_{n \in \mathbb{N}}$  is  $\mathbf{P}$ -uniform if there is a polynomial-time T.M. that on input  $1^n$  outputs the description of the circuit  $C_n$ .*

- But...

## Theorem

*A language  $L$  is computable by a  $\mathbf{P}$ -uniform circuit family iff  $L \in \mathbf{P}$ .*

## Theorem

$$\mathbf{BPP} \subset \mathbf{P}_{/\text{poly}}$$

**Proof:** Recall that if  $L \in \mathbf{BPP}$ , then  $\exists$  PTM  $M$  such that:

$$\Pr_{r \in \{0,1\}^{\text{poly}(n)}} [M(x, r) \neq L(x)] < 2^{-n}$$

Then, taking the union bound:

$$\begin{aligned} \Pr[\exists x \in \{0,1\}^n : M(x, r) \neq L(x)] &= \Pr \left[ \bigcup_{x \in \{0,1\}^n} M(x, r) \neq L(x) \right] \leq \\ &\leq \sum_{x \in \{0,1\}^n} \Pr[M(x, r) \neq L(x)] < 2^{-n} + \dots + 2^{-n} = 1 \end{aligned}$$

So,  $\exists r_n \in \{0,1\}^{\text{poly}(n)}$ , s.t.  $\forall x \in \{0,1\}^n: M(x, r_n) = L(x)$ .

Using  $\{r_n\}_{n \in \mathbb{N}}$  as advice string, we have the non-uniform machine. □

## Relationship among Complexity Classes

## Theorem

*The following are equivalent:*

- ①  $A \in \mathbf{P}/\text{poly}$ .
- ② *There exists a sparse set  $S$  such that  $A \leq_T^P S$ .*

## Corollary

Every sparse set has polynomial-size circuits.





# Hierarchies for Semantic Classes with advice

- We have argued why we can't obtain Hierarchies for semantic measures using classical diagonalization techniques. But using small advice we can have the following results:

Theorem ([Bar02], [GST04])

For  $a, b \in \mathbb{R}$ , with  $1 \leq a < b$ :

$$\mathbf{BPTIME}(n^a)/1 \not\subseteq \mathbf{BPTIME}(n^b)/1$$

Theorem ([FST05])

For any  $1 \leq a \in \mathbb{R}$  there is a real  $b > a$  such that:

$$\mathbf{RTIME}(n^b)/1 \not\subseteq \mathbf{RTIME}(n^a)/\log(n)^{1/2a}$$

# Circuit Lower Bounds

- The significance of proving lower bounds for this computational model is related to the famous "**P** vs **NP**" problem, since:

$$\mathbf{NP} \setminus \mathbf{P}_{/poly} \neq \emptyset \Rightarrow \mathbf{P} \neq \mathbf{NP}$$

- But...after decades of efforts, The best lower bound for an **NP** language is  $5n - o(n)$ , proved very recently (2005).
- There are better lower bounds for some special cases, i.e. some restricted classes of circuits, such as: bounded depth circuits, monotone circuits, and bounded depth circuits with "counting" gates.

## Definition

Let  $PAR : \{0, 1\}^n \rightarrow \{0, 1\}$  be the *parity* function, which outputs the modulo 2 sum of an  $n$ -bit input. That is:

$$PAR(x_1, \dots, x_n) \equiv \sum_{i=1}^n x_i \pmod{2}$$

## Theorem

*For all constant  $d$ ,  $PAR$  has no polynomial-size circuit of depth  $d$ .*

- The above result (improved by Håstad and Yao) gives a relatively tight lower bound of  $\exp(\Omega(n^{1/(d-1)}))$ , on the size of  $n$ -input  $PAR$  circuits of depth  $d$ .

## Definition

For  $x, y \in \{0, 1\}^n$ , we denote  $x \preceq y$  if every bit that is 1 in  $x$  is also 1 in  $y$ . A function  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  is *monotone* if  $f(x) \leq f(y)$  for every  $x \preceq y$ .

## Definition

A Boolean Circuit is *monotone* if it contains only AND and OR gates, and no NOT gates. Such a circuit can only compute monotone functions.

## Theorem (Monotone Circuit Lower Bound for CLIQUE)

Denote by  $CLIQUE_{k,n} : \{0, 1\}^{\binom{n}{2}} \rightarrow \{0, 1\}$  the function that on input an adjacency matrix of an  $n$ -vertex graph  $G$  outputs 1 iff  $G$  contains a  $k$ -clique. There exists some constant  $\epsilon > 0$  such that for every  $k \leq n^{1/4}$ , there is no monotone circuit of size less than  $2^{\epsilon\sqrt{k}}$  that computes  $CLIQUE_{k,n}$ .

- So, we proved a significant lower bound ( $2^{\Omega(n^{1/8})}$ )
- The significance of the above theorem lies on the fact that there was some alleged connection between monotone and non-monotone circuit complexity (e.g. that they would be polynomially related). Unfortunately, Éva Tardos proved in 1988 that the gap between the two complexities is exponential.
- Where is the problem finally?  
Today, we know *that a result for a lower bound using such techniques would imply the inversion of strong one-way functions:*



# Contents

- Introduction
- Turing Machines
- Undecidability
- Complexity Classes
- Oracles & Optimization Problems
- Randomized Computation
- Non-Uniform Complexity
- **Interactive Proofs**
- Counting Complexity















# Interactive Proof for Graph Non-Isomorphism

## Definition

Two graphs  $G_1$  and  $G_2$  are *isomorphic*, if there exists a permutation  $\pi$  of the labels of the nodes of  $G_1$ , such that  $\pi(G_1) = G_2$ . If  $G_1$  and  $G_2$  are isomorphic, we write  $G_1 \cong G_2$ .

- GI: Given two graphs  $G_1, G_2$ , decide if they are isomorphic.
- GNI: Given two graphs  $G_1, G_2$ , decide if they are *not* isomorphic.

- Obviously,  $\text{GI} \in \mathbf{NP}$  and  $\text{GNI} \in \text{coNP}$ .
- This proof system relies on the Verifier's access to a *private* random source which cannot be seen by the Prover, so we confirm the crucial role the private coins play.

# Interactive Proof for Graph Non-Isomorphism

Verifier: Picks  $i \in \{1, 2\}$  uniformly at random.

Then, it permutes randomly the vertices of  $G_i$  to get a new graph  $H$ . It sends  $H$  to the Prover.

Prover: Identifies which of  $G_1, G_2$  was used to produce  $H$ . Let  $G_j$  be the graph. Sends  $j$  to  $V$ .

Verifier: Accept if  $i = j$ . Reject otherwise.





# Babai's Arthur-Merlin Games

## Definition (Extended (FGMSZ89))

An Arthur-Merlin Game is a pair of interactive TMs  $A$  and  $M$ , and a predicate  $R$  such that:

- On input  $x$ , exactly  $2q(|x|)$  messages of length  $m(|x|)$  are exchanged,  $q, m \in poly(|x|)$ .
- $A$  goes first, and at iteration  $1 \leq i \leq q(|x|)$  chooses u.a.r. a string  $r_i$  of length  $m(|x|)$ .
- $M$ 's reply in the  $i^{th}$  iteration is  $y_i = M(x, r_1, \dots, r_i)$  ( $M$ 's strategy).
- For every  $M'$ , a **conversation** between  $A$  and  $M'$  on input  $x$  is  $r_1 y_1 r_2 y_2 \dots r_{q(|x|)} y_{q(|x|)}$ .
- The set of all conversations is denoted by  $CONV_x^{M'}$ ,  
 $|CONV_x^{M'}| = 2^{q(|x|)m(|x|)}$ .

# Babai's Arthur-Merlin Games

## Definition (*cont'd*)

- The predicate  $R$  maps the input  $x$  and a conversation to a Boolean value.
- The set of accepting conversations is denoted by  $ACC_x^{R,M}$ , and is the set:

$$\{r_1 \cdots r_q \mid \exists y_1 \cdots y_q \text{ s.t. } r_1 y_1 \cdots r_q y_q \in CONV_x^M \wedge R(r_1 y_1 \cdots r_q y_q) = 1\}$$

- A language  $L$  has an Arthur-Merlin proof system if:
  - **There exists** a strategy for  $M$ , such that for all  $x \in L$ :  
 $\frac{ACC_x^{R,M}}{CONV_x^M} \geq \frac{2}{3}$  (*Completeness*)
  - **For every** strategy for  $M$ , and for every  $x \notin L$ :  $\frac{ACC_x^{R,M}}{CONV_x^M} \leq \frac{1}{3}$  (*Soundness*)

# Definitions

- So, with respect to the previous **IP** definition:

## Definition

For every  $k$ , the complexity class **AM** $[k]$  is defined as a subset to **IP** $[k]$  obtained when we restrict the verifier's messages to be *random bits*, and not allowing it to use any other random bits that are not contained in these messages.

We denote **AM**  $\equiv$  **AM** $[2]$ .

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- **Merlin** → **Prover**
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We denote  $\mathbf{AM} \equiv \mathbf{AM}[2]$ .

- **Merlin**  $\rightarrow$  **Prover**
- **Arthur**  $\rightarrow$  **Verifier**
- Also, the class **MA** consists of all languages  $L$ , where there's an interactive proof for  $L$  in which the prover first sending a message, and then the verifier is "tossing coins" and computing its decision by doing a deterministic polynomial-time computation involving the input, the message and the random output.



# Properties of Arthur-Merlin Games

- $\mathbf{MA} \subseteq \mathbf{AM}$
- $\mathbf{MA}[1] = \mathbf{NP}$ ,  $\mathbf{AM}[1] = \mathbf{BPP}$
- $\mathbf{AM}$  could be intuitively approached as the probabilistic version of  $\mathbf{NP}$  (usually denoted as  $\mathbf{AM} = \mathcal{BP} \cdot \mathbf{NP}$ ).
- $\mathbf{AM} \subseteq \Pi_2^P$  and  $\mathbf{MA} \subseteq \Sigma_2^P \cap \Pi_2^P$ .
- $\mathbf{NP}^{\mathbf{BPP}} \subseteq \mathbf{MA}$ ,  $\mathbf{MA}^{\mathbf{BPP}} = \mathbf{MA}$ ,  $\mathbf{AM}^{\mathbf{BPP}} = \mathbf{AM}$  and  $\mathbf{AM}^{\Delta\Sigma_1^P} = \mathbf{AM}^{\mathbf{NP} \cap \text{coNP}} = \mathbf{AM}$
- If we consider the complexity classes  $\mathbf{AM}[k]$  (the languages that have Arthur-Merlin proof systems of a bounded number of rounds, they form an hierarchy:

$$\mathbf{AM}[0] \subseteq \mathbf{AM}[1] \subseteq \cdots \subseteq \mathbf{AM}[k] \subseteq \mathbf{AM}[k+1] \subseteq \cdots$$

- Are these inclusions proper ? ? ?





# Properties of Arthur-Merlin Games

- Proper formalism (*Zachos et al.*):

## Definition (Majority Quantifier)

Let  $R : \{0, 1\}^* \times \{0, 1\}^* \rightarrow \{0, 1\}$  be a predicate, and  $\varepsilon$  a rational number, such that  $\varepsilon \in (0, \frac{1}{2})$ . We denote by  $(\exists^+ y, |y| = k)R(x, y)$  the following predicate:

*“There exist at least  $(\frac{1}{2} + \varepsilon) \cdot 2^k$  strings  $y$  of length  $m$  for which  $R(x, y)$  holds.”*

We call  $\exists^+$  the *overwhelming majority* quantifier.

- $\exists_r^+$  means that the fraction  $r$  of the possible certificates of a certain length satisfy the predicate for the certain input.
- Obviously,  $\exists^+ = \exists_{1/2+\varepsilon}^+ = \exists_{2/3}^+ = \exists_{3/4}^+ = \exists_{0.99}^+ = \exists_{1-2^{-p(|x|)}}^+$

# Properties of Arthur-Merlin Games

## Definition

We denote as  $\mathcal{C} = (Q_1/Q_2)$ , where  $Q_1, Q_2 \in \{\exists, \forall, \exists^+\}$ , the class  $\mathcal{C}$  of languages  $L$  satisfying:

- $x \in L \Rightarrow Q_1 y R(x, y)$
- $x \notin L \Rightarrow Q_2 y \neg R(x, y)$
  
- So: **P** =  $(\forall/\forall)$ , **NP** =  $(\exists/\forall)$ , **coNP** =  $(\forall/\exists)$   
**BPP** =  $(\exists^+/\exists^+)$ , **RP** =  $(\exists^+/\forall)$ , **coRP** =  $(\forall/\exists^+)$

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## Arthur-Merlin Games

$$\mathbf{AM} = \mathcal{BP} \cdot \mathbf{NP} = (\exists^+\exists/\exists^+\forall)$$

$$\mathbf{MA} = \mathcal{N} \cdot \mathbf{BPP} = (\exists\exists^+/\forall\exists^+)$$

- Similarly: **AMA** =  $(\exists^+\exists\exists^+/\exists^+\forall\exists^+)$  etc.

# Properties of Arthur-Merlin Games

## Theorem

- i  $\mathbf{MA} = (\exists \forall / \forall \exists^+)$
- ii  $\mathbf{AM} = (\forall^+ \exists / \exists \forall^+)$

## Proof:

### Lemma

- $\mathbf{BPP} = (\exists^+ / \exists^+) = (\exists \forall^+ / \forall \exists^+) = (\forall \exists^+ / \exists \forall^+) \mathbf{(1)}$  (BPP-Theorem)
- $(\forall \exists^+ / \exists \forall^+) \subseteq (\forall \exists / \exists \forall^+) \mathbf{(2)}$

i)  $\mathbf{MA} = \mathcal{N} \cdot \mathbf{BPP} = (\exists \exists^+ / \forall \exists^+) \stackrel{\mathbf{(1)}}{=} (\exists \exists \forall^+ / \forall \exists \exists^+) \subseteq (\exists \forall \exists^+ / \forall \exists \forall^+) \subseteq (\exists \forall / \forall \exists^+) \mathbf{(2)}$

(the last inclusion holds by quantifier contraction). Also,

$(\exists \forall / \forall \exists^+) \subseteq (\exists \exists^+ / \forall \exists^+) = \mathbf{MA}$ .

ii) Similarly,

$\mathbf{AM} = \mathcal{BP} \cdot \mathbf{NP} = (\forall \exists^+ \exists / \exists^+ \forall) = (\forall \exists^+ \exists / \exists^+ \forall) \subseteq (\forall \exists^+ \exists / \exists^+ \forall) \subseteq (\forall \exists / \exists \forall^+) \mathbf{(2)}$ .

Also,  $(\forall \exists / \exists \forall^+) \subseteq (\forall \exists^+ \exists / \exists^+ \forall) = \mathbf{AM}$ .

# Properties of Arthur-Merlin Games

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i **MA** =  $(\exists \forall / \forall \exists^+)$

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## Proof:

### Lemma

• **BPP** =  $(\exists^+ / \exists^+)$  =  $(\exists \forall^+ / \forall^+ \exists)$  =  $(\forall^+ \exists / \exists \forall^+)$  (1) (BPP-Theorem)

•  $(\forall^+ \exists / \exists \forall^+)$   $\subseteq$   $(\exists \forall^+ / \forall^+ \exists)$  (2)

i) **MA** =  $\mathcal{N} \cdot \mathbf{BPP}$  =  $(\exists \exists^+ / \forall^+ \exists^+)$   $\stackrel{(1)}{=} (\exists \exists^+ \forall / \forall \exists^+ \exists^+)$   $\subseteq (\exists \forall / \forall \exists^+)$

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Also,  $(\forall^+ \exists / \exists \forall)$   $\subseteq (\forall^+ \exists \exists / \forall^+ \exists \forall)$  = **AM**.

# Properties of Arthur-Merlin Games

## Theorem

i **MA** =  $(\exists V/\forall E)^+$

ii **AM** =  $(\forall E/E^+)$

## Proof:

### Lemma

- **BPP** =  $(\exists^+/E^+) = (\exists V/A^+) = (A^+E/E^+)$  (1) (BPP-Theorem)

- $(\exists V/A^+) \subseteq (A^+E/E^+)$  (2)

i) **MA** =  $\mathcal{N} \cdot \mathbf{BPP} = (\exists E^+/V^+) \stackrel{(1)}{=} (\exists E^+/A^+) \subseteq (\exists V/A^+)$

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Also,  $(A^+E/E^+) \subseteq (A^+E/E^+) = \mathbf{AM}$ .

# Properties of Arthur-Merlin Games

## Theorem

i **MA** =  $(\exists A/\forall E)^+$

ii **AM** =  $(\forall A^+/\exists E)^+$

## Proof:

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- BPP** =  $(\exists^+/E^+) = (\exists A^+/A^+E) = (A^+E/\exists A)^+ \text{ (1) (BPP-Theorem)}$
- $(A^+E/\exists A)^+ \subseteq (\forall A^+/E^+) \text{ (2)}$

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Also,  $(\forall A^+ E/\exists A)^+ \subseteq (\forall A^+ \exists^+ E/\exists E^+) = \mathbf{AM}.$



# Properties of Arthur-Merlin Games

Theorem

$$\mathbf{MA} \subseteq \mathbf{AM}$$

**Proof:**

Obvious from (2):  $(\forall \epsilon / \exists \epsilon^+) \subseteq (\exists \epsilon / \forall \epsilon^+)$ .  $\square$

Theorem

- i  $\mathbf{AM} \subseteq \Pi_2^P$
- ii  $\mathbf{MA} \subseteq \Sigma_2^P \cup \Pi_2^P$

**Proof:**

i)  $\mathbf{AM} = (\forall \epsilon / \exists \epsilon^+) \subseteq (\forall \epsilon / \forall \epsilon) = \Pi_2^P$

ii)  $\mathbf{MA} = (\exists \epsilon / \forall \epsilon^+) \subseteq (\exists \epsilon / \exists \epsilon) = \Sigma_2^P$ , and

$\mathbf{MA} \subseteq \mathbf{AM} \Rightarrow \mathbf{MA} \subseteq \Pi_2^P$ . So,  $\mathbf{MA} \subseteq \Sigma_2^P \cup \Pi_2^P$ .  $\square$

# Properties of Arthur-Merlin Games

Theorem (Speedup Theorem)

For  $t(n) \geq 2$ :

$$\mathbf{AM}[2t(n)] = \mathbf{AM}[t(n)]$$

- **The Arthur-Merlin Hierarchy collapses at its second level:**

Theorem (Collapse Theorem)

For every  $k \geq 2$ :

$$\mathbf{AM} = \mathbf{AM}[k] = \mathbf{MA}[k + 1]$$

Example

$$\mathbf{MAM} = (A^+E/EAE) \stackrel{(1)}{\subseteq} (A^+EA/E+EE) \subseteq (A^+EAA/EA+EE) \subseteq (A^+EA/EAE) \stackrel{(2)}{\subseteq} (A^+E/EAE) = \mathbf{AM}$$

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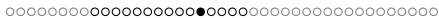
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## Theorem (Collapse Theorem)

For every  $k \geq 2$ :

$$\mathbf{AM} = \mathbf{AM}[k] = \mathbf{MA}[k + 1]$$

## Example

$$\mathbf{MAM} = (E \subseteq A^+E) \stackrel{(1)}{\subseteq} (EA \subseteq A^+EA) \subseteq (EA \subseteq A^+EA) \stackrel{(2)}{\subseteq} (EA \subseteq A^+EA) \subseteq (EA \subseteq A^+EA) = \mathbf{AM}$$

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Example

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# Properties of Arthur-Merlin Games

## Proof:

- The general case is implied by the generalization of BPP-Theorem **(1)** & **(2)**:
- $(Q_1 \exists^+ Q_2 / Q_3 \exists^+ Q_4) = (Q_1 \exists^+ \forall Q_2 / Q_3 \forall \exists^+ Q_4) = (Q_1 \forall \exists^+ Q_2 / Q_3 \exists^+ \forall Q_4)$  **(1')**
- $(Q_1 \exists \forall Q_2 / Q_3 \exists^+ Q_4) \subseteq (Q_1 \forall \exists \forall Q_2 / Q_3 \exists^+ \forall Q_4)$  **(2')**
- Using the above we can easily see that the Arthur-Merlin Hierarchy collapses at the second level. (*Try it!*)  $\square$



# Properties of Arthur-Merlin Games

Theorem (BHZ)

If  $\text{coNP} \subseteq \text{AM}$  (that is, if  $\text{GI}$  is **NP**-complete), then the Polynomial Hierarchy collapses at the second level, and  $\text{PH} = \Sigma_2^P = \text{AM}$ .

**Proof:** Our hypothesis states:  $(\forall/\exists) \subseteq (\forall/\exists)^+$

Then:

$$\Sigma_2^P = (\exists/\forall) \stackrel{\text{Hyp.}}{\subseteq} (\exists/\forall/\exists) \stackrel{(2)}{\subseteq} (\forall/\exists/\forall) = (\forall/\exists/\forall) = (\forall/\exists)^+ = \text{AM} \subseteq (\forall/\exists) = \Pi_2^P. \quad \square$$

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# Measure One Results

- $\mathbf{P}^A \neq \mathbf{NP}^A$ , for almost all oracles  $A$ .
- $\mathbf{P}^A = \mathbf{BPP}^A$ , for almost all oracles  $A$ .
- $\mathbf{NP}^A = \mathbf{AM}^A$ , for almost all oracles  $A$ .

## Definition

$$\text{almost } \mathcal{C} = \left\{ L \mid \Pr_{A \in \{0,1\}^*} [L \in \mathcal{C}^A] = 1 \right\}$$

## Theorem

- i  $\text{almost } \mathbf{P} = \mathbf{BPP}$  [BG81]
- ii  $\text{almost } \mathbf{NP} = \mathbf{AM}$  [NW94]
- iii  $\text{almost } \mathbf{PH} = \mathbf{PH}$

# Measure One Results

## Theorem (Kurtz)

For almost every pair of oracles  $B, C$ :

- i  $\mathbf{BPP} = \mathbf{P}^B \cap \mathbf{P}^C$
- ii  $\mathbf{almostNP} = \mathbf{NP}^B \cap \mathbf{NP}^C$

## Indicative Open Questions

- Does exist an oracle separating  $\mathbf{AM}$  from  $\mathbf{almostNP}$ ?
- Is  $\mathbf{almostNP}$  contained in some finite level of Polynomial-Time Hierarchy?
- Motivated by [BHZ]: If  $\mathbf{coNP} \subseteq \mathbf{almostNP}$ , does it follow that  $\mathbf{PH}$  collapses?

# The power of Interactive Proofs

- As we saw, **Interaction** alone does not give us computational capabilities beyond **NP**.
- Also, **Randomization** alone does not give us significant power (we know that  $\mathbf{BPP} \subseteq \Sigma_2^P$ , and many researchers believe that  $\mathbf{P} = \mathbf{BPP}$ , which holds under some plausible assumptions).
- How much power could we get by their *combination*?
- We know that for fixed  $k \in \mathbb{N}$ ,  $\mathbf{IP}[k]$  collapses to

$$\mathbf{IP}[k] = \mathbf{AM} = \mathbf{BP} \cdot \mathbf{NP}$$

a class that is “close” to **NP** (under similar assumptions, the non-deterministic analogue of **P** vs. **BPP** is **NP** vs. **AM**.)

- If we let  $k$  be a polynomial in the size of the input, how much more power could we get?



# The power of Interactive Proofs

- Surprisingly:

**Theorem (L.F.K.N. & Shamir)**

$$\mathbf{IP = PSPACE}$$

# The power of Interactive Proofs

## Lemma 1

$$\mathbf{IP} \subseteq \mathbf{PSPACE}$$

# The power of Interactive Proofs

## Lemma 1

$$\mathbf{IP} \subseteq \mathbf{PSPACE}$$

### Proof:

- If the Prover is an **NP**, or even a **PSPACE** machine, the lemma holds.
- But what if we have an omnipotent prover?
- On any input, the Prover chooses its messages in order to *maximize the probability of V's acceptance!*
- We consider prover as an **oracle**, by assuming wlog that his responses are one bit at a time.
- Th protocol has polynomially many rounds (say  $N=n^c$ ), which bounds the messages and the random bits used.
- So, the protocol is described by a computation tree  $T$ :

# The power of Interactive Proofs

## Proof(cont'd):

- Vertices of  $T$  are  $V$ 's configurations.
- **Random Branches** (queries to the random tape)
- **Oracle Branches** (queries to the prover)
- For each fixed  $P$ , the tree  $T_P$  can be pruned to obtain only random branches.
- Let  $\Pr_{opt}[E \mid F]$  the conditional probability given that the prover *always behaves optimally*.
- The acceptance condition is  $m_N = 1$ .
- For  $y_i \in \{0, 1\}^N$  and  $z_i \in \{0, 1\}$  let:

$$R_i = \bigwedge_{j=1}^i m_j = y_j$$

$$S_i = \bigwedge_{j=1}^i l_j = z_j$$

# The power of Interactive Proofs

## Proof(cont'd):



$$\Pr_{\text{opt}}[m_N = 1 \mid R_{i-1} \wedge S_{i-1}] =$$

$$\sum_{y_i} \max_{z_i} \Pr_{\text{opt}}[m_N = 1 \mid R_i \wedge S_i] \cdot \Pr_{\text{opt}}[R_i \mid R_{i-1} \wedge S_{i-1}]$$

- $\Pr_{\text{opt}}[R_i \mid R_{i-1} \wedge S_{i-1}]$  is **PSPACE**-computable, by simulating  $V$ .
- $\Pr_{\text{opt}}[m_N = 1 \mid R_i \wedge S_i]$  can be calculated by DFS on  $T$ .
- The probability of acceptance is
 
$$\Pr_{\text{opt}}[m_N = 1] = \Pr_{\text{opt}}[m_N = 1 \mid R_0 \wedge S_0]$$
- The prover can calculate its optimal move at any point in the protocol in **PSPACE** by calculating  $\Pr_{\text{opt}}[m_N = 1 \mid R_i \wedge S_i]$  for  $z_i \in \{0, 1\}$  and choosing its answer to be the value that gives the maximum.

# Warmup: Interactive Proof for UNSAT

## Lemma 2

$$\text{PSPACE} \subseteq \text{IP}$$

- For simplicity, we will construct an Interactive Proof for UNSAT (a **coNP**-complete problem), showing that:

## Theorem

$$\text{coNP} \subseteq \text{IP}$$

- Let  $N$  be a prime.
- We will translate a **formula**  $\phi$  with  $m$  clauses and  $n$  variables  $x_1, \dots, x_n$  to a **polynomial**  $p$  over the field ( $\text{mod}N$ ) (where  $N > 2^n \cdot 3^m$ ), in the following way:

# Arithmetization

- Arithmetic generalization of a CNF Boolean Formula.

$$\begin{array}{lcl}
 \mathbf{T} & \longrightarrow & 1 \\
 \mathbf{F} & \longrightarrow & 0 \\
 \neg x & \longrightarrow & 1 - x \\
 \wedge & \longrightarrow & \times \\
 \vee & \longrightarrow & +
 \end{array}$$

## Example

$$\begin{array}{c}
 (x_3 \vee \neg x_5 \vee x_{17}) \wedge (x_5 \vee x_9) \wedge (\neg x_3 \vee x_4) \\
 \downarrow \\
 (x_3 + (1 - x_5) + x_{17}) \cdot (x_5 + x_9) \cdot ((1 - x_3) + (1 - x_4))
 \end{array}$$

- Each literal is of degree 1, so the polynomial  $p$  is of degree at most  $m$ .
- Also,  $0 < p < 3^m$ .

# Warmup: Interactive Proof for UNSAT

**Prover**

Sends primality proof for  $N$

→

**Verifier**

checks proof



# Warmup: Interactive Proof for UNSAT

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Sends primality proof for  $N$

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checks proof

$$q_1(x) = \sum p(x, x_2, \dots, x_n)$$

→

checks if  $q_1(0) + q_1(1) = 0$

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# Warmup: Interactive Proof for UNSAT

## Prover

Sends primality proof for  $N$

$$q_1(x) = \sum p(x, x_2, \dots, x_n)$$

$$q_2(x) = \sum p(r_1, x, x_3, \dots, x_n)$$

## Verifier

checks proof

checks if  $q_1(0) + q_1(1) = 0$

← sends  $r_1 \in \{0, \dots, N-1\}$

checks if  $q_2(0) + q_2(1) = q_1(r_1)$

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Sends primality proof for  $N$

$$q_1(x) = \sum p(x, x_2, \dots, x_n)$$

$$q_2(x) = \sum p(r_1, x, x_3, \dots, x_n)$$

$$q_n(x) = p(r_1, \dots, r_{n-1}, x)$$

## Verifier

checks proof

checks if  $q_1(0) + q_1(1) = 0$

← sends  $r_1 \in \{0, \dots, N-1\}$

checks if  $q_2(0) + q_2(1) = q_1(r_1)$

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⋮

checks if  $q_n(0) + q_n(1) = q_{n-1}(r_{n-1})$

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picks  $r_n \in \{0, \dots, N-1\}$

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Sends primality proof for  $N$

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## Verifier

checks proof

checks if  $q_1(0) + q_1(1) = 0$

← sends  $r_1 \in \{0, \dots, N-1\}$

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picks  $r_n \in \{0, \dots, N-1\}$

checks if  $q_n(r_n) = p(r_1, \dots, r_n)$

# Warmup: Interactive Proof for UNSAT

- If  $\phi$  is **unsatisfiable**, then

$$\sum_{x_1 \in \{0,1\}} \sum_{x_2 \in \{0,1\}} \cdots \sum_{x_n \in \{0,1\}} p(x_1, \dots, x_n) \equiv 0 \pmod{N}$$

and the protocol will succeed.

- Also, the arithmetization can be done in polynomial time, and if we take  $N = 2^{\mathcal{O}(n+m)}$ , then the elements in the field can be represented by  $\mathcal{O}(n+m)$  bits, and thus an evaluation of  $p$  in any point of  $\{0, \dots, N-1\}$  can be computed in polynomial time.
- We have to show that if  $\phi$  is satisfiable, then the verifier will **reject** with high probability.
- If  $\phi$  is satisfiable, then

$$\sum_{x_1 \in \{0,1\}} \sum_{x_2 \in \{0,1\}} \cdots \sum_{x_n \in \{0,1\}} p(x_1, \dots, x_n) \neq 0 \pmod{N}$$



- So,  $p_1(0) + p_1(1) \neq 0$ , so if the prover send  $p_1$  we 're done.
- If the prover send  $q_1 \neq p_1$ , then the polynomials will agree on at most  $m$  places. So,  $\Pr [p_1(r_1) \neq q_1(r_1)] \geq 1 - \frac{m}{N}$ .
- If indeed  $p_1(r_1) \neq q_1(r_1)$  and the prover sends  $p_2 = q_2$ , then the verifier will reject since  $q_2(0) + q_2(1) = p_1(r_1) \neq q_1(r_1)$ .
- Thus, the prover must send  $q_2 \neq p_2$ .
- **We continue in a similar way:** If  $q_i \neq p_i$ , then with probability at least  $1 - \frac{m}{N}$ ,  $r_i$  is such that  $q_i(r_i) \neq p_i(r_i)$ .
- Then, the prover must send  $q_{i+1} \neq p_{i+1}$  in order for the verifier not to reject.
- At the end, if the verifier has not rejected before the last check,  $\Pr [p_n \neq q_n] \geq 1 - (n - 1)\frac{m}{N}$ .
- If so, with probability at least  $1 - \frac{m}{N}$  the verifier will reject since,  $q_n(x)$  and  $p(r_1, \dots, r_{n-1}, x)$  differ on at least that fraction of points.
- **The total probability that the verifier will accept is at most  $\frac{nm}{N}$ .**

# Arithmetization of QBF

$$\begin{array}{l} \exists \longrightarrow \Sigma \\ \forall \longrightarrow \Pi \end{array}$$

Example

$$\forall x_1 \exists x_2 [(x_1 \wedge x_2) \vee \exists x_3 (\bar{x}_2 \wedge x_3)]$$

↓

$$\prod_{x_1 \in \{0,1\}} \sum_{x_2 \in \{0,1\}} \left[ (x_1 \cdot x_2) + \sum_{x_3 \in \{0,1\}} (1 - x_2) \cdot x_3 \right]$$

Theorem

*A closed QBF is true if and only if the value of its arithmetic form is non-zero.*

# Arithmetization of QBF

- If a QBF is true, its value could be quite large:

## Theorem

*Let  $A$  be a closed QBF of size  $n$ . Then, the value of its arithmetic form cannot exceed  $\mathcal{O}(2^{2^n})$ .*

- Since such numbers cannot be handled by the protocol, we reduce them modulo some -smaller- prime  $p$ :

## Theorem

*Let  $A$  be a closed QBF of size  $n$ . Then, there exists a prime  $p$  of length polynomial in  $n$ , such that its arithmetization*

$$A' \not\equiv 0 \pmod{p} \Leftrightarrow A \text{ is true.}$$

# Arithmetization of QBF

- A QBF with all the variables quantified is called **closed**, and can be evaluated to either True or False.
- An **open** QBF with  $k > 0$  free variables can be interpreted as a boolean function  $\{0, 1\}^k \rightarrow \{0, 1\}$ .
- Now, consider the language of all true quantified boolean formulas:

$$\text{TQBF} = \{\Phi \mid \Phi \text{ is a true quantified Boolean formula}\}$$

- It is known that TQBF is a **PSPACE**-complete language!
- So, if we have an interactive proof protocol recognizing TQBF, then we have a protocol for every **PSPACE** language.

# Protocol for TQBF

- Given a quantified formula

$$\Psi = \forall x_1 \exists x_2 \forall x_3 \cdots \exists x_n \phi(x_1, \dots, x_n)$$

we use *arithmetization* to construct the polynomial  $P_\phi$ . Then,  $\Psi \in \text{TQBF}$  if and only if

$$\prod_{b_1 \in \{0,1\}^*} \sum_{b_2 \in \{0,1\}^*} \prod_{b_3 \in \{0,1\}^*} \cdots \sum_{b_n \in \{0,1\}^*} P_\phi(b_1, \dots, b_n) \neq 0$$

PCPs

# Epilogue: Probabilistically Checkable Proofs

- But if we put a **proof** instead of a Prover?

## Epilogue: Probabilistically Checkable Proofs

- But if we put a **proof** instead of a Prover?
- The alleged proof is a string, and the (probabilistic) verification procedure is given direct (**oracle**) access to the proof.
- The verification procedure can access only *few* locations in the proof!
- We parameterize these Interactive Proof Systems by two complexity measures:
  - **Query** Complexity
  - **Randomness** Complexity
- The effective proof length of a PCP system is upper-bounded by  $q(n) \cdot 2^{r(n)}$  (in the non-adaptive case).  
(How long can be in the adaptive case?)





# Main Results

- Obviously:

$$\mathbf{PCP}[0, 0] = ?$$

$$\mathbf{PCP}[0, poly] = ?$$

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- Obviously:

$$\mathbf{PCP}[0, 0] = \mathbf{P}$$

$$\mathbf{PCP}[0, \mathit{poly}] = \mathbf{NP}$$

$$\mathbf{PCP}[\mathit{poly}, 0] = \mathbf{?}$$





# Main Results

- The restriction that the proof length is at most  $q2^r$  is inconsequential, since such a verifier can look on at most this number of locations.
- We have that  $\mathbf{PCP}[r(n), q(n)] \subseteq \mathbf{NTIME}[2^{\mathcal{O}(r(n))}q(n)]$ , since a NTM could guess the proof in  $2^{\mathcal{O}(r(n))}q(n)$  time, and verify it deterministically by running the verifier for all  $2^{\mathcal{O}(r(n))}$  possible choices of its random coin tosses. If the verifier accepts for all these possible tosses, then the NTM accepts.

# Contents

- Introduction
- Turing Machines
- Undecidability
- Complexity Classes
- Oracles & Optimization Problems
- Randomized Computation
- Non-Uniform Complexity
- Interactive Proofs
- **Counting Complexity**

# Why counting?

- So far, we have seen two versions of problems:
  - Decision Problems (if a solution *exists*)
  - Function Problems (if a solution can be *produced*)
- A very important type of problems in Complexity Theory is also:
  - Counting Problems (*how many* solution exist)

## Example (#SAT)

Given a Boolean Expression, compute the number of different truth assignments that satisfy it.

- Note that if we can solve #SAT in polynomial time, we can solve SAT also.
- Similarly, we can define #HAMILTON PATH, #CLIQUE, etc.



# Basic Definitions

## Definition ( $\#\mathbf{P}$ )

A function  $f : \{0, 1\}^* \rightarrow \mathbb{N}$  is in  $\#\mathbf{P}$  if there exists a polynomial  $p : \mathbb{N} \rightarrow \mathbb{N}$  and a polynomial-time Turing Machine  $M$  such that for every  $x \in \{0, 1\}^*$ :

$$f(x) = |\{y \in \{0, 1\}^{p(|x|)} : M(x, y) = 1\}|$$

- The definition implies that  $f(x)$  can be expressed in  $\text{poly}(|x|)$  bits.
- Each function  $f$  in  $\#\mathbf{P}$  is equal to the number of paths from an initial configuration to an accepting configuration, or **accepting paths** in the configuration graph of a poly-time NDTM.
- $\mathbf{FP} \subseteq \#\mathbf{P} \subseteq \mathbf{PSPACE}$
- If  $\#\mathbf{P} = \mathbf{FP}$ , then  $\mathbf{P} = \mathbf{NP}$ .
- If  $\mathbf{P} = \mathbf{PSPACE}$ , then  $\#\mathbf{P} = \mathbf{FP}$ .

- In order to formalize a notion of completeness for  $\#\mathbf{P}$ , we must define proper reductions:

### Definition (Cook Reduction)

A function  $f$  is  $\#\mathbf{P}$ -complete if it is in  $\#\mathbf{P}$  and every  $g \in \#\mathbf{P}$  is in  $\mathbf{FP}^g$ .

- As we saw, for each problem in  $\mathbf{NP}$  we can define the associated counting problem: If  $A \in \mathbf{NP}$ , then  $\#A(x) = |\{y \in \{0, 1\}^{p(|x|)} : R_A(x, y) = 1\}| \in \#\mathbf{P}$

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- We now define a more strict form of reduction:

### Definition (Parsimonious Reduction)

We say that there is a parsimonious reduction from  $\#A$  to  $\#B$  if there is a polynomial time transformation  $f$  such that for all  $x$ :

$$|\{y : R_A(x, y) = 1\}| = |\{z : R_B(f(x), z) = 1\}|$$

# Completeness Results

## Theorem

$\#\text{CIRCUIT SAT}$  is  $\#\mathbf{P}$ -complete.

## Proof:

- Let  $f \in \#\mathbf{P}$ . Then,  $\exists M, p$ :  
 $f = |\{y \in \{0, 1\}^{p(|x|)} : M(x, y) = 1\}|$ .
- Given  $x$ , we want to construct a circuit  $C$  such that:

$$|\{z : C(z)\}| = |\{y : y \in \{0, 1\}^{p(|x|)}, M(x, y) = 1\}|$$

- We can construct a circuit  $\hat{C}$  such that on input  $x, y$  simulates  $M(x, y)$ .
- We know that this can be done with a circuit with size about the square of  $M$ 's running time.
- Let  $C(y) = \hat{C}(x, y)$ .

# Completeness Results

## Theorem

$\#SAT$  is  $\#P$ -complete.

## Proof:

- We reduce  $\#CIRCUIT SAT$  to  $\#SAT$ :
- Let a circuit  $C$ , with  $x_1, \dots, x_n$  input gates and  $1, \dots, m$  gates.
- We construct a Boolean formula  $\phi$  with variables  $x_1, \dots, x_n, g_1, \dots, g_m$ , where  $g_i$  represents the output of gate  $i$ .
- A gate can be completely described by simulating the output for each of the 4 possible inputs.
- In this way, we have reduced  $C$  to a formula  $\phi$  with  $n + m$  variables and  $4m$  clauses. □

# The Permanent

## Definition (PERMANENT)

For a  $n \times n$  matrix  $A$ , the permanent of  $A$  is:

$$\text{perm}(A) = \sum_{\sigma \in S_n} \prod_{i=1}^n A_{i, \sigma(i)}$$

- Permanent is similar to the determinant, but it seems more difficult to compute.
- Combinatorial interpretation: If  $A$  has entries  $\in \{0, 1\}$ , it can be viewed as the adjacency matrix of a bipartite graph  $G(X, Y, E)$  with  $X = \{x_1, \dots, x_n\}$ ,  $Y = \{y_1, \dots, y_n\}$  and  $\{x_i, y_j\} \in E$  iff  $A_{i,j} = 1$ .

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For a  $n \times n$  matrix  $A$ , the permanent of  $A$  is:

$$\text{perm}(A) = \sum_{\sigma \in S_n} \prod_{i=1}^n A_{i,\sigma(i)}$$

- Permanent is similar to the determinant, but it seems more difficult to compute.
- Combinatorial interpretation: If  $A$  has entries  $\in \{0, 1\}$ , it can be viewed as the adjacency matrix of a bipartite graph  $G(X, Y, E)$  with  $X = \{x_1, \dots, x_n\}$ ,  $Y = \{y_1, \dots, y_n\}$  and  $\{x_i, y_i\} \in E$  iff  $A_{i,j} = 1$ .
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# The Permanent

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- The term  $\prod_{i=1}^n A_{i,\sigma(i)}$  is 1 iff  $\sigma$  has a perfect matching.
- So, in this case  $\text{perm}(A)$  is the number of perfect matchings in the corresponding graph!



# Valiant's Theorem

Theorem (Valiant's Theorem)

*PERMANENT* is #P-complete.

- Notice that the decision version of PERMANENT is in **P** ! !

## Proof Idea:

- We reduce 3SAT to PERMANENT in two steps:
- Given  $\phi$ , we create an undirected graph  $G'$  with small weights, such that:

$$PERM(G') = 4^{3m} \cdot \#\phi$$

- In the second step, we convert  $G'$  to an undirected graph  $G$  such that  $PERM(G) = PERM(G') \pmod{M}$ , where  $M$  has polynomially many bits.
- The problem PERMANENT MODN reduces to PERMANENT.
- Finally, the permanent of the resulting matrix is  $4^{3m}$  times the number of sat. truth assignments of the original formula.

The Class  $\oplus\mathbf{P}$ The Class  $\oplus\mathbf{P}$ 

## Definition

A language  $L$  is in the class  $\oplus\mathbf{P}$  if there is a NDTM  $M$  such that for all strings  $x$ ,  $x \in L$  iff the *number of accepting paths* on input  $x$  is odd.

- The problems  $\oplus\text{SAT}$  and  $\oplus\text{HAMILTON PATH}$  are  $\oplus\mathbf{P}$ -complete.
- $\oplus\mathbf{P}$  is closed under complement.



# Valiant-Vazirani Theorem

## Proof:

### Definition

Let  $S \subseteq \{x_1, \dots, x_n\}$ . Hyperplane  $\eta_S$  is a CNF Boolean Formula, s.t. an even number among the variables in  $S$  are true.

- We will construct the formulas  $F_i$ :  $F_i = F_{i-1} \wedge \eta_{S_i}$ ,  $1 \leq i \leq n+1$  where  $S_i$  is a random generated subset of th variables, and  $F_0 = F$ .
- If  $F_i \in \text{USAT}$ , then we answer that  $F$  is satisfiable.
- If none of the  $F_i$ 's are in  $\text{USAT}$ , then we answer that  $F$  is probably unsatisfiable.
- We shall prove now that the probability to be wrong is  $< 7/8$  (and by repeating the algorithm 6 times we are  $< 1/2$ , as required:

# Valiant-Vazirani Theorem

## Lemma

*If the number of satisfying truth assignments of  $F$  is between  $2^k$  and  $2^{k+1}$ ,  $0 \leq k < n$ , then the probability that  $F_{k+2}$  has exactly one satisfying truth assignments is at least  $1/8$ .*

## Proof (of the Lemma):

- Let  $T$  the number of satisfying t.a.'s.
- Two t.a. **agree** on  $\eta_S$  if they both satisfy or falsify it.
- Fix  $t \in T$ .

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- Let  $T$  the number of satisfying t.a.'s.
- Two t.a. **agree** on  $\eta_S$  if they both satisfy or falsify it.
- Fix  $t \in T$ .
- A  $t' \in T$  agrees with  $t$  on all  $k+2$  first hyperplanes w.p.  $\frac{1}{2^{k+2}}$ .
- Summing up over all  $t' \in T \setminus t$ ,  $t$  agrees with *some*  $t'$  on the first  $k+2$  hyperplanes w.p. at most  $\frac{|T|-2}{2^{k+2}}$ .
- So, the probability that  $t$  **disagrees** with every other member of  $T$  on one of the first  $k+2$  hyperplanes is at least  $1/2$ .

# Valiant-Vazirani Theorem

- The probability that  $t$  satisfies all  $k + 2$  first hyperplanes is  $\frac{1}{2^{k+2}}$ , and with probability  $1/2$  it is the only one that does.
- So, with probability at least  $\frac{1}{2^{k+3}}$   $t$  is the unique satisfying t.a. of  $F_{k+2}$ .
- Since this holds for each  $t \in T$ , the probability that such an element of  $T$  exists is  $2^k \times \frac{1}{2^{k+3}} = \frac{1}{8}$ . □

## Proof (cont'd):

- If the number of satisfying t.a.'s of  $F$  is not zero, then it lies between  $2^k$  and  $2^{k+1}$ , for some  $k < n$ .
- So, at least one of the  $F_i$  will have probability at least  $1/8$  to be satisfied by a unique t.a. □

# Quantifiers vs Counting

- An important open question in the 80s concerned the relative power of Polynomial Hierarchy and  $\#\mathbf{P}$ .
- Both are natural generalizations of  $\mathbf{NP}$ , but it seemed that their features were not directly comparable to each other.
- But, in 1989, S. Toda showed the following theorem:



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- But, in 1989, S. Toda showed the following theorem:

Theorem (Toda's Theorem)

$$\mathbf{PH} \subseteq \mathbf{P}^{\#\mathbf{P}[1]}$$

# Toda's Theorem

- The proof consists of two main lemmas:

Lemma 1

$$\mathbf{PH} \subseteq \mathbf{BP} \cdot \oplus \mathbf{P}$$

Lemma 2

$$\mathbf{BP} \cdot \oplus \mathbf{P} \subseteq \mathbf{P}^{\#\mathbf{P}}$$