

# CS151 Complexity Theory

Lecture 7  
April 20, 2004

## Outline

- 3 examples of the power of randomness
  - communication complexity
  - polynomial identity testing
  - complexity of finding unique solutions
- randomized complexity classes
- Adelman's Theorem

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## Communication complexity

two parties: Alice and Bob  
function  $f: \{0,1\}^n \times \{0,1\}^n \rightarrow \{0,1\}$   
Alice holds  $x \in \{0,1\}^n$ ; Bob holds  $y \in \{0,1\}^n$

- Goal: compute  $f(x, y)$  while communicating as few bits as possible between Alice and Bob
- count number of bits exchanged (computation free)
- at each step: one party sends bits that are a function of held input and received bits so far

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## Communication complexity

- simple function (equality):  
$$EQ(x, y) = 1 \text{ iff } x = y$$
- simple protocol:
  - Alice sends  $x$  to Bob ( $n$  bits)
  - Bob sends  $EQ(x, y)$  to Alice (1 bit)
  - total:  $n + 1$  bits
  - (works for any predicate  $f$ )

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## Communication complexity

- Can we do better?
  - deterministic protocol?
  - probabilistic protocol?
    - at each step: one party sends bits that are a function of held input and received bits so far and the result of some coin tosses
    - required to output  $f(x, y)$  with high probability over all coin tosses

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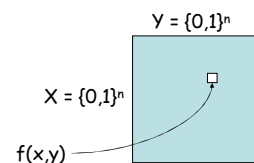
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## Communication complexity

**Theorem:** no deterministic protocol can compute  $EQ(x, y)$  while exchanging fewer than  $n+1$  bits.

- Proof:
  - “input matrix”:



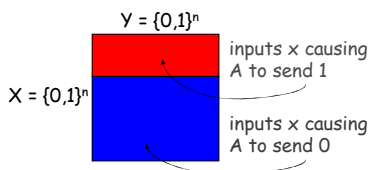
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## Communication complexity

- assume without loss of generality 1 bit sent at a time
- A sends 1 bit depending only on x:



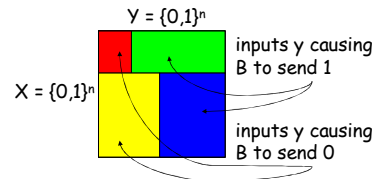
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## Communication complexity

- B sends 1 bit depending only on y and received bit:



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## Communication complexity

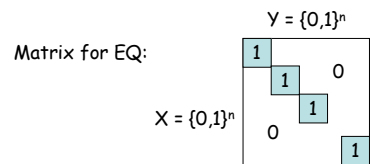
- at end of protocol involving  $k$  bits of communication, matrix is partitioned into at most  $2^k$  combinatorial rectangles
- bits sent in protocol are the same for every input  $(x, y)$  in given rectangle
- conclude:  $f(x,y)$  must be constant on each rectangle

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## Communication complexity



- any partition into combinatorial rectangles with constant  $f(x,y)$  must have  $2^n + 1$  rectangles
- protocol that exchanges  $\leq n$  bits can only create  $2^n$  rectangles, so must exchange at least  $n+1$  bits.

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## Communication complexity

- protocol for EQ employing randomness?
  - Alice picks random prime  $p$  in  $\{1 \dots 4n^2\}$ , sends:
    - $p$
    - $(x \bmod p)$
  - Bob sends:
    - $(y \bmod p)$
  - players output 1 if and only if:
 
$$(x \bmod p) = (y \bmod p)$$

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## Communication complexity

- $O(\log n)$  bits exchanged
  - if  $x = y$ , always correct
  - if  $x \neq y$ , incorrect if and only if:
    - $p$  divides  $|x - y|$
  - # primes in range is  $\geq 2n$
  - # primes dividing  $|x - y|$  is  $\leq n$
  - probability incorrect  $\leq 1/2$
- Randomness gives an exponential advantage!!

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## Polynomial identity testing

- Given: polynomial  $p(x_1, x_2, \dots, x_n)$  over field  $\mathbf{F}$
- Is  $p$  identically zero?
  - i.e., is  $p(\mathbf{x}) = 0$  for all  $\mathbf{x} \in \mathbf{F}^n$
  - (assume  $|\mathbf{F}|$  larger than degree...)
- “polynomial identity testing” because given two polynomials  $p, q$ , we can check the identity  $p \equiv q$  by checking if  $(p - q) \equiv 0$

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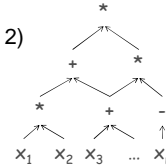
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## Polynomial identity testing

- polynomial  $p(x_1, x_2, \dots, x_n)$  given as arithmetic circuit:

- multiplication (fan-in 2)
- addition (fan-in 2)
- negation (fan-in 1)



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## Polynomial identity testing

- try all  $|\mathbf{F}|^n$  inputs?
  - may be exponentially many
- multiply out symbolically, check that all coefficients are zero?
  - may be exponentially many coefficients
- can randomness help?
  - i.e., flip coins, allow small probability of wrong answer

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## Polynomial identity testing

**Lemma** (Schwartz-Zippel): Let

$$p(x_1, x_2, \dots, x_n)$$

be a total degree  $d$  polynomial over a field  $\mathbf{F}$  and let  $S$  be any subset of  $\mathbf{F}$ . Then if  $p$  is not identically 0,

$$\Pr_{r_1, r_2, \dots, r_n \in S} [p(r_1, r_2, \dots, r_n) = 0] \leq d/|S|.$$

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## Polynomial identity testing

- Proof:
  - induction on number of variables  $n$
  - base case:  $n = 1$ ,  $p$  is univariate polynomial of degree at most  $d$
  - at most  $d$  roots, so
 
$$\Pr[p(r_1) = 0] \leq d/|S|$$

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## Polynomial identity testing

- write  $p(x_1, x_2, \dots, x_n)$  as
 
$$p(x_1, x_2, \dots, x_n) = \sum_i (x_1)^i p_i(x_2, \dots, x_n)$$
- $k = \max. i$  for which  $p_i(x_2, \dots, x_n)$  not id. zero
- by induction hypothesis:

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

- whenever  $p_k(r_2, \dots, r_n) \neq 0$ ,  $p(x_1, r_2, \dots, r_n)$  is a univariate polynomial of degree  $k$

$$\Pr[p(r_1, r_2, \dots, r_n) = 0 \mid p_k(r_2, \dots, r_n) \neq 0] \leq k/|S|$$

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## Polynomial identity testing

- $\Pr[ p_k(r_2, \dots, r_n) = 0 ] \leq (d-k)/|S|$   
 $\Pr[ p(r_1, r_2, \dots, r_n) = 0 \mid p_k(r_2, \dots, r_n) \neq 0 ] \leq k/|S|$   
 – conclude:  
 $\Pr[ p(r_1, \dots, r_n) = 0 ] \leq (d-k)/|S| + k/|S| = d/|S|$
- Note: can add these probabilities because
- $$\Pr[E_1] = \Pr[E_1|E_2]\Pr[E_2] + \Pr[E_1|\neg E_2]\Pr[\neg E_2] \leq \Pr[E_2] + \Pr[E_1|\neg E_2]$$

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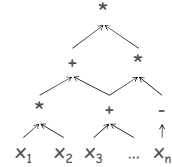
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## Polynomial identity testing

- Given: polynomial  $p(x_1, x_2, \dots, x_n)$  over field  $F$

- Is  $p$  identically zero?



- Note: degree  $d$  is at most the size of input

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## Polynomial identity testing

- randomized algorithm: pick a subset  $S \subset F$  of size  $2d$ 
  - pick  $r_1, r_2, \dots, r_n$  from  $S$  uniformly at random
  - if  $p(r_1, r_2, \dots, r_n) = 0$ , answer “yes”
  - if  $p(r_1, r_2, \dots, r_n) \neq 0$ , answer “no”
- if  $p$  identically zero, never wrong
- if not, Schwartz-Zippel ensures probability of error at most  $1/2$

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## Unique solutions

- a positive instance of SAT may have many satisfying assignments
- maybe the difficulty comes from not knowing which to “work on”
- if we knew # satisfying assignments was 1 or 0, could we zero in on the 1 efficiently?

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## Unique solutions

**Question:** given polynomial-time algorithm that works on SAT instances with at most 1 satisfying assignment, can we solve general SAT instances efficiently?

- Answer: yes
  - but only if “efficiently” allows randomness

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## Unique solutions

**Theorem** (Valiant-Vazirani): there is a randomized poly-time procedure that given a 3-CNF formula

$$\varphi(x_1, x_2, \dots, x_n)$$

outputs a 3-CNF formula  $\varphi'$  such that

- if  $\varphi$  is not satisfiable then  $\varphi'$  is not satisfiable
- if  $\varphi$  is satisfiable then with probability at least  $1/(8n)$   $\varphi'$  has exactly one satisfying assignment

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## Unique solutions

- **Proof:**
  - given subset  $S \subset \{1, 2, \dots, n\}$ , there exists a 3-CNF formula  $\theta_S$  on  $x_1, x_2, \dots, x_n$  and additional variables such that:
    - $|\theta_S| = O(n)$
    - $\theta_S$  is satisfiable iff an even number of variables in  $\{x_i\}_{i \in S}$  are true
    - for each such setting of the  $x_i$  variables, this satisfying assignment is unique
    - not difficult; details omitted

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## Unique solutions

- set  $\varphi_0 = \varphi$
- for  $i = 1, 2, \dots, n$ 
  - pick random subset  $S_i$
  - set  $\varphi_i = \varphi_{i-1} \wedge \theta_{S_i}$
- output random  $\varphi_i$
- $T =$  set of satisfying assignments for  $\varphi$
- Claim: if  $|T| > 0$ , then
$$\Pr_{k \in \{0, 1, 2, \dots, n-1\}}[2^k \leq |T| \leq 2^{k+1}] \geq 1/n$$

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## Unique solutions

**Claim:** if  $2^k \leq |T| \leq 2^{k+1}$ , then the probability  $\varphi_{k+2}$  has exactly one satisfying assignment is  $\geq 1/8$

– fix  $t \in T$

$$\begin{array}{l} t = 0101 \boxed{00101} 0111 \\ t' = 1010 \boxed{11100} 0101 \end{array}$$

↖  $S_i$

- $\Pr[t \text{ "agrees with" } t' \text{ on } S_i] = 1/2$
- $\Pr[t \text{ agrees with } t' \text{ on } S_1, S_2, \dots, S_{k+2}] = (1/2)^{k+2}$

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## Unique solutions

- $\Pr[t \text{ agrees with some } t' \text{ on } S_1, \dots, S_{k+2}] \leq (|T|-1)(1/2)^{k+2} < 1/2$
- $\Pr[t \text{ satisfies } S_1, S_2, \dots, S_{k+2}] = (1/2)^{k+2}$
- $\Pr[t \text{ unique satisfying assignment of } \varphi_{k+2}] > (1/2)^{k+3}$
- sum over at least  $2^k$  different  $t \in T$  (disjoint events); claim follows.

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## Randomized complexity classes

- **model:** probabilistic Turing Machine
  - deterministic TM with additional read-only tape containing “coin flips”
- **BPP** (Bounded-error Probabilistic Poly-time)
  - $L \in \text{BPP}$  if there is a p.p.t. TM  $M$ :
    - $x \in L \Rightarrow \Pr_y[M(x,y) \text{ accepts}] \geq 2/3$
    - $x \notin L \Rightarrow \Pr_y[M(x,y) \text{ rejects}] \geq 2/3$
  - “p.p.t.” = probabilistic polynomial time

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## Randomized complexity classes

- **RP** (Random Polynomial-time)
  - $L \in \text{RP}$  if there is a p.p.t. TM  $M$ :
    - $x \in L \Rightarrow \Pr_y[M(x,y) \text{ accepts}] \geq 1/2$
    - $x \notin L \Rightarrow \Pr_y[M(x,y) \text{ rejects}] = 1$
- **coRP** (complement of Random Polynomial-time)
  - $L \in \text{coRP}$  if there is a p.p.t. TM  $M$ :
    - $x \in L \Rightarrow \Pr_y[M(x,y) \text{ accepts}] = 1$
    - $x \notin L \Rightarrow \Pr_y[M(x,y) \text{ rejects}] \geq 1/2$

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## Randomized complexity classes

These classes may capture “efficiently computable” better than **P**.

- “1/2” in **RP**, **coRP** definition unimportant
  - can replace by  $1/\text{poly}(n)$
- “2/3” in **BPP** definition unimportant
  - can replace by  $\frac{1}{2} + 1/\text{poly}(n)$
- Why? error reduction
  - we will see simple error reduction by repetition
  - more sophisticated error reduction later

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## Error reduction for RP

- given  $L$  and p.p.t TM  $M$ :
  - $x \in L \Rightarrow \Pr_y[M(x,y) \text{ accepts}] \geq \epsilon$
  - $x \notin L \Rightarrow \Pr_y[M(x,y) \text{ rejects}] = 1$
- new p.p.t TM  $M'$ :
  - simulate  $M$   $k/\epsilon$  times, each time with independent coin flips
  - accept if **any** simulation accepts
  - otherwise reject

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## Error reduction

$$x \in L \Rightarrow \Pr_y[M(x,y) \text{ accepts}] \geq \epsilon$$

$$x \notin L \Rightarrow \Pr_y[M(x,y) \text{ rejects}] = 1$$

- if  $x \in L$ :
  - probability a given simulation “bad”  $\leq (1 - \epsilon)$
  - probability all simulations “bad”  $\leq (1 - \epsilon)^{(k/\epsilon)} \leq e^{-k}$
  - $\Pr_y[M'(x, y') \text{ accepts}] \geq 1 - e^{-k}$
- if  $x \notin L$ :
  - $\Pr_y[M'(x, y') \text{ rejects}] = 1$

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## Error reduction for BPP

- given  $L$ , and p.p.t. TM  $M$ :
  - $x \in L \Rightarrow \Pr_y[M(x,y) \text{ accepts}] \geq \frac{1}{2} + \epsilon$
  - $x \notin L \Rightarrow \Pr_y[M(x,y) \text{ rejects}] \geq \frac{1}{2} + \epsilon$
- new p.p.t. TM  $M'$ :
  - simulate  $M$   $k/\epsilon^2$  times, each time with independent coin flips
  - accept if **majority** of simulations accept
  - otherwise reject

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## Error reduction for BPP

- $X_i$  random variable indicating “correct” outcome in  $i$ -th simulation (out of  $m = k/\epsilon^2$ )
  - $\Pr[X_i = 1] \geq \frac{1}{2} + \epsilon$
  - $\Pr[X_i = 0] \leq \frac{1}{2} - \epsilon$
- $E[X_i] \geq \frac{1}{2} + \epsilon$
- $X = \sum_i X_i$
- $\mu = E[X] \geq (\frac{1}{2} + \epsilon)m$
- Chernoff:  $\Pr[X \leq m/2] \leq 2^{-\Omega(\epsilon^2 \mu)}$

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## Error reduction for BPP

$$x \in L \Rightarrow \Pr_y[M(x,y) \text{ accepts}] \geq \frac{1}{2} + \epsilon$$

$$x \notin L \Rightarrow \Pr_y[M(x,y) \text{ rejects}] \geq \frac{1}{2} + \epsilon$$

– if  $x \in L$

$$\Pr_{y'}[M'(x, y') \text{ accepts}] \geq 1 - (\frac{1}{2})^{\Omega(k)}$$

– if  $x \notin L$

$$\Pr_{y'}[M'(x, y') \text{ rejects}] \geq 1 - (\frac{1}{2})^{\Omega(k)}$$

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## Randomized complexity classes

One more important class:

- **ZPP** (Zero-error Probabilistic Poly-time)
  - $\mathbf{ZPP} = \mathbf{RP} \cap \mathbf{coRP}$
  - $\Pr_y[M(x,y) \text{ outputs "fail"}] \leq \frac{1}{2}$
  - otherwise outputs correct answer

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## Randomized complexity classes

- We have shown:
  - polynomial identity testing is in **coRP**
  - a poly-time algorithm for detecting unique solutions to SAT implies

$$\mathbf{NP} = \mathbf{RP}$$

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## Relationship to other classes

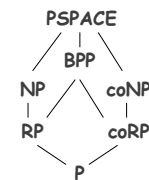
- all these classes contain **P**
  - they can simply ignore the tape with coin flips
- all are in **PSPACE**
  - can exhaustively try all strings  $y$
  - count accepts/rejects; compute probability
- $\mathbf{RP} \subset \mathbf{NP}$  (and  $\mathbf{coRP} \subset \mathbf{coNP}$ )
  - multitude of accepting computations
  - **NP** requires only one

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## Relationship to other classes



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## BPP

- How powerful is **BPP**?
- We have seen an example of a problem in **BPP** that we only know how to solve in **EXP**.

Is randomness a panacea for intractability?

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## BPP

- It is not known if  $\mathbf{BPP} = \mathbf{EXP}$  (or even **NEXP**)!
  - but there are strong hints that it does not
- Is there a deterministic simulation of **BPP** that does better than brute-force search?
  - yes, if allow non-uniformity

**Theorem** (Adelman):  $\mathbf{BPP} \subset \mathbf{P/poly}$

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## BPP and Boolean circuits

- Proof:
  - language  $L \in \mathbf{BPP}$
  - error reduction gives TM  $M$  such that
    - if  $x \in L$   
 $\Pr_y[M(x, y) \text{ accepts}] \geq 1 - (\frac{1}{2})^{|x|^2}$
    - if  $x \notin L$   
 $\Pr_y[M(x, y) \text{ rejects}] \geq 1 - (\frac{1}{2})^{|x|^2}$

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## BPP and Boolean circuits

- say “ $y$  is bad for  $x$ ” if  $M(x, y)$  gives incorrect answer
- for fixed  $x$ :  $\Pr_y[y \text{ is bad for } x] \leq (\frac{1}{2})^{|x|^2}$
- $\Pr_y[y \text{ is bad for some } x] \leq 2^{|x|}(\frac{1}{2})^{|x|^2} < 1$
- Conclude: there exists some  $y$  on which  $M(x, y)$  is always correct
- build circuit for  $M$ , hardwire this  $y$

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## BPP and Boolean circuits

- Does  $\mathbf{BPP} = \mathbf{EXP}$  ?
- Adelman's Theorem shows:  
 $\mathbf{BPP} = \mathbf{EXP}$  implies  $\mathbf{EXP} \subset \mathbf{P/poly}$

If you believe that randomness is all-powerful, you must also believe that non-uniformity gives an exponential advantage.

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## BPP

- Next:  
further explore the relationship between  
randomness  
and  
nonuniformity
- Main tool: pseudo-random generators

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