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# Bullet Notes on Computational Complexity

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## 1 Probabilistic computation

### 1.1 The model

- The model of non-deterministic TM will also serve as our model of randomized algorithm.
- Let  $M = (Q, \Gamma, \delta_0, \delta_1, q_0, q_F)$  be an NTM.
- Think of the machine as making an independent and fair coin-flip at each step.
- Let  $c_i \in \{0, 1\}$  be the outcome of the  $i$ -th coin-flip.
- We think of the machine as taking the random computation path given by the sequence  $\delta_{c_1}, \dots, \delta_{c_t}$ .
- Fix an input  $x \in \Gamma^*$  and let  $t = T_M(x)$ .
- For  $c = (c_1, c_2, \dots, c_t) \in \{0, 1\}^t$ , which we think of as  $t$  independent fair coin-flips:
- Notation:  $M_c(x)$ : the output of  $M$  on input  $x$  when it uses  $\delta_{c_i}$  in its  $i$ -th step.
- Notation:  $\Pr[E] := |E|/2^t$  for any *event*  $E \subseteq \{0, 1\}^t$ .
- For example, for  $y \in \Gamma^*$  and  $E = \{c \in \{0, 1\}^t : M_c(x) = y\}$ , what is  $\Pr[E]$ ?
- It's the probability that (a random computation path of)  $M$  on input  $x$  outputs  $y$ .
- In this interpretation, if  $L \in \text{NP}$  and  $M$  is NTM that decides  $L$  with  $t_M(n) \leq p(n)$ , then:
- For every input  $x$  we have
- 1) if  $x \in L$ , then  $\Pr[M(x) = 1] > 0$ ,
- 2) if  $x \notin L$ , then  $\Pr[M(x) = 1] = 0$ .
- But note that in case  $x \in L$ , the probability  $\Pr[M(x) = 1]$  can be as small as  $1/2^{p(|x|)}$ .
- If  $p(n)$  is, say,  $n^2$ , this could be useless even for rather small  $n$ .
- For  $n = 200$ , less likely than the planet exploding spontaneously in the next second.
- We want NTMs that for  $x \in L$  achieve good probability of acceptance.
- Say 0.99, instead of  $1/2^{40000}$ .

## 1.2 Example: primality testing

- COMPOSITENESS/PRIMALITY:
- Input: An  $m$  bit number  $n$ .
- Output: Is  $n$  composite or prime?
- The naive algorithm that tests  $x|n$  for all  $x \leq \lfloor \sqrt{n} \rfloor$  takes exponential time:  $O(2^{m/2})$ .
- Recall Miller-Rabin test for compositeness:
  1. given an odd  $n \geq 3$ ,
  2. write  $n - 1 = 2^s q$  where  $q$  is odd; i.e., repeatedly divide  $n - 1$  by 2 until odd,
  3. choose a random  $a \in \{1, \dots, n - 1\}$ ,
  4. compute  $x_0 := a^q \pmod n$  by modular repeated squaring,
  5. for  $i = 0, 1, 2, \dots, s - 1$ , compute  $x_{i+1} := x_i^2 \pmod n$ ,
  6. if we find some  $i \in \{0, \dots, s - 1\}$  with  $x_i \neq \{+1, -1\}$  and  $x_{i+1} = 1$ , then output 'composite'.
  7. else output 'probably prime'.
- The test is sound:
- If  $n$  is prime, then the only two roots of  $x^2 - 1 \pmod n$  are  $+1$  and  $-1$ .
- This follows from the fact that  $\mathbb{Z}/n\mathbb{Z}$  is a field when  $n$  is prime.
- The test succeeds with good probability:
- If  $n$  is not prime, then at least  $3/4$  fraction of  $a \in \{1, \dots, n - 1\}$  output 'composite'.
- This follows from a little bit of basic number theory (beyond the scope here).
- The probability of error can be made exponentially small in  $t$  by repeating the test  $t$  many times.
- How is this implemented in a NTM?
- Modular arithmetic has efficient algorithms; for exponentiation we use repeated squaring.
- But how is  $a \in \{1, \dots, n - 1\}$  chosen uniformly at random?
- It is not:
  - 1) find  $k$  such that  $2^k \leq n - 1 < 2^{k+1}$ ,
  - 2) flip  $k + 1$  many independent coin-flips and interpret it as a number  $x \in \{0, \dots, 2^{k+1} - 1\}$ ,
  - 3) if  $x \notin \{1, \dots, n - 1\}$ , repeat; else, output  $x$ .
- The probability that nothing is output after  $t$  iterations is  $\leq (1 - (n - 1)/2^{k+1})^t \leq (1/2)^t$ .
- Conditioned on success:  $x$  is uniformly distributed in  $\{1, \dots, n - 1\}$ .

### 1.3 Example: polynomial identity testing

- POLYNOMIAL IDENTITY TESTING:
- Input: An arithmetic expression  $F$  with variables  $x_1, \dots, x_n$  and constants  $+1$  and  $-1$ ,
- Output: Does  $F$  compute the identically zero polynomial in  $\mathbb{R}[x_1, \dots, x_n]$ ?
- Example 1:  $(x_1 + x_2)(x_2 - x_3) - x_1x_2 + x_1x_3 - x_2^2 + x_2x_3$  is identically zero.
- Example 2:  $x_1x_3x_5 \cdots x_{n-1} - \prod_{i=1}^{n/2} (x_{2i-1} - x_{2i}) + x_2x_4x_6 \cdots x_n$  is not identically zero.
- The second example shows that naively expanding the expression could take exponential time.
- The “mindless” test:
  1. given an arithmetic expression  $F(x_1, \dots, x_n)$  with  $n$  variables and  $m$  operations,
  2. for  $i = 1, \dots, n$ , choose a random  $a_i \in \{1, \dots, 20m\}$ , independently,
  3. evaluate  $F(x_1/a_1, \dots, x_n/a_n)$ ,
  4. if output is not zero, output ‘not the identically zero polynomial’
  5. else, output ‘probably the identically zero polynomial’.
- The test is sound:
  - This is obvious: how could  $F(x_1, \dots, x_n)$  be identically 0 if  $F(x_1/a_1, \dots, x_n/a_n) \neq 0$ .
  - The test succeeds with high probability:
  - This relies on the following two facts.
- Lemma 1:
  - Assume  $F(x_1, \dots, x_n)$  is an arithmetic expression with  $n$  variables and  $m$  subexpressions.
  - Then  $F$  computes a polynomial  $p \in \mathbb{R}[x_1, \dots, x_n]$  of degree at most  $2m - 1$ .
  - Proof:
    - Let  $\text{size}(F)$  denote the number of subexpressions of  $F$ .
    - Let  $\deg(F)$  denote the degree of the polynomial computed by  $F$ .
    - Proof is by induction on  $\text{size}(F)$ .
    - Base case:  $F$  is a variable or a constant. Obvious.
    - Inductive cases:  $F = G + H$  or  $F = G \times H$  for subexpressions  $G$  and  $H$ .
    - Let  $m_1 = \text{size}(F)$  and  $m_2 = \text{size}(G)$ , so  $m = m_1 + m_2 + 1$ .
    - $F = G + H$ : Then  $\deg(F) = \max\{\deg(G), \deg(H)\} \leq \max\{2m_1 - 1, 2m_2 - 1\} \leq 2m - 1$ .

- $F = G \times H$ : Then  $\deg(F) = \deg(G) + \deg(H) \leq 2m_1 - 1 + 2m_2 - 1 = 2m - 1$ .
- QED

- Lemma 2 [Schwartz-Zippel Lemma]:
- Assume  $p \in \mathbb{R}[x_1, \dots, x_n]$  has degree at most  $d$  and let  $S \subseteq \mathbb{R}$ .
- Then  $p$  has at most  $d|S|^{n-1}$  many roots in  $S^n$ , unless it is identically zero.
- Proof: Induction on  $n$ .
- Base case:  $n = 1$ .
- This follows from the fundamental theorem of algebra:
- Over fields, non-zero univariate polynomials of degree  $d$  have at most  $d$  roots.
- Inductive case:  $n \geq 2$ .
- Write  $p = \sum_{i=0}^d p_i(x_1, \dots, x_{n-1})x_n^i$  and assume  $p$  is not identically zero.
- Then there exists a maximal  $d^* \in [d]$  so that  $p_{d^*}$  is not identically zero.
- Note that  $p_{d^*}$  has degree at most  $d - d^*$  and at most  $n - 1$  variables.
- Now fix a root  $(a_1, \dots, a_n) \in S^n$  of  $p$ .
- Then, either  $(a_1, \dots, a_{n-1})$  is a root of  $p_{d^*}$ , or it isn't and  $a_n$  is a root of  $\sum_{i=0}^{d^*} p_i(a_1, \dots, a_{n-1})x_n^i$ .
- Of the first type we have no more than  $(d - d^*)|S|^{n-2}|S|$  (by the induction hypothesis).
- Of the second type we have no more than  $|S|^{n-1}d^*$  (by the base case).
- This leaves a total of at most  $(d - d^*)|S|^{n-2}|S| + |S|^{n-1}d^* = d|S|^{n-1}$ .
- QED

- The test succeeds with high probability (continued):
- Assume  $F$  is not identically zero.
- By Lemma 1, we have  $\deg(F) \leq 2m - 1$ .
- By Lemma 2 with  $S = \{1, \dots, 20m\}$  we have:
- For random and independent  $a_1, \dots, a_n \in S$ :
- The probability that  $F(x_1/a_1, \dots, x_n/a_n) = 0$  is at most  $(2m - 1)(20m)^{n-1}/(20m)^n \leq 1/10$ .
- Thus the algorithm outputs “probably not identically zero” with probability at least  $9/10$ .

## 1.4 Probabilistic complexity classes: BPP, RP, co-RP, ZPP

- BPP: Bounded-error Probabilistic Polynomial-time.
- Class of all  $L \subseteq \{0,1\}^*$  for which there exist NTM  $M$  and  $c \geq 0$  such that  $t_M = O(n^c)$  and:
- For all  $x \in \{0,1\}^*$ :
- 0)  $\Pr[ M(x) \in \{0,1\} ] = 1$ .
- 1) if  $x \in L$ , then  $\Pr[ M(x) = 1 ] \geq 3/4$  (hence  $\Pr[ M(x) = 0 ] \leq 1/4$ ),
- 2) if  $x \notin L$ , then  $\Pr[ M(x) = 1 ] \leq 1/4$  (hence  $\Pr[ M(x) = 0 ] \geq 3/4$ ).
  
- RP: Randomized Polynomial-time with 1-sided error and no false positives.
- Class of all  $L \subseteq \{0,1\}^*$  for which there exist NTM  $M$  and  $c \geq 0$  such that  $t_M = O(n^c)$  and:
- For all  $x \in \{0,1\}^*$ :
- 0)  $\Pr[ M(x) \in \{0,1\} ] = 1$ .
- 1) if  $x \in L$ , then  $\Pr[ M(x) = 1 ] \geq 1/2$  (hence  $\Pr[ M(x) = 0 ] \leq 1/2$ ),
- 2) if  $x \notin L$ , then  $\Pr[ M(x) = 1 ] = 0$  (hence  $\Pr[ M(x) = 0 ] = 1$ ).
  
- RP': Randomized Polynomial-time with 1-sided error and no false negatives.
- Class of all  $L \subseteq \{0,1\}^*$  for which there exist NTM  $M$  and  $c \geq 0$  such that  $t_M = O(n^c)$  and:
- For all  $x \in \{0,1\}^*$ :
- 0)  $\Pr[ M(x) \in \{0,1\} ] = 1$ .
- 1) if  $x \in L$ , then  $\Pr[ M(x) = 1 ] = 1$  (hence  $\Pr[ M(x) = 0 ] = 0$ ),
- 2) if  $x \notin L$ , then  $\Pr[ M(x) = 1 ] \leq 1/2$  (hence  $\Pr[ M(x) = 0 ] \geq 1/2$ ).
  
- ZPP: Zero-error Probabilistic Polynomial-time.
- Class of all  $L \subseteq \{0,1\}^*$  for which there exist NTM  $M$  and  $c \geq 0$  such that  $t_M = O(n^c)$  and:
- For all  $x \in \{0,1\}^*$ :
- 0)  $\Pr[ M(x) \in \{0,1,?\} ] = 1$  and  $\Pr[ M(x) = ? ] \leq 1/2$ .
- 1) if  $x \in L$ , then  $\Pr[ M(x) = 0 ] = 0$  (hence  $\Pr[ M(x) = 1 ] \geq 1/2$ ),
- 2) if  $x \notin L$ , then  $\Pr[ M(x) = 1 ] = 0$  (hence  $\Pr[ M(x) = 0 ] \geq 1/2$ ).
- I.e., algorithm is never wrong (zero-error), but may fail to give a definite (i.e. 0/1) answer.
  
- Obvious relationships:
- $\text{RP} = \text{co-RP}'$  and  $\text{RP}' = \text{co-RP}$ .
- $\text{RP} \subseteq \text{NP}$  and  $\text{co-RP} \subseteq \text{co-NP}$ .

- $\text{RP} \subseteq \text{BPP}$  and  $\text{co-RP} \subseteq \text{BPP}$ .
- More:
- Theorem:  $\text{ZPP} = \text{RP} \cap \text{co-RP}$
- Think about it: proof given in the next version of these notes.