

Probabilistic Computing

Computation Complexity Seminar

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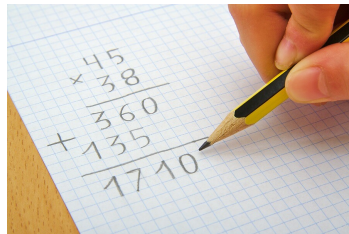
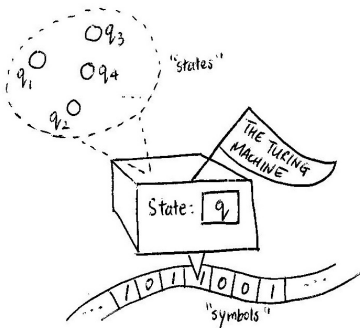
September 16, 2021

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- 1 Probabilistic Turing Machine
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- 3 Robustness
- 4 Randomized Reduction
- 5 Randomized Space-Bounded Computation
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Introduction

Idea of Turing Machine



Intelligence, Uncertainty and Error

$$L = \{w_1, w_2, \dots\}$$
$$w \in L \text{ or } w \notin L$$

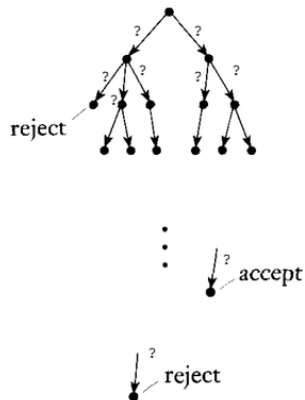
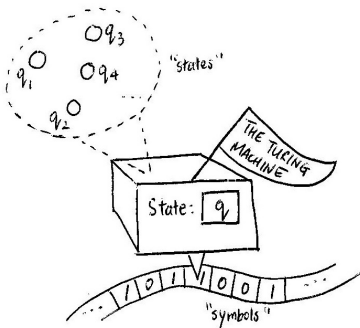
Uncertainty and permission to make error, speed up the computations.



Power of Randomness



Probabilistic Turing Machine (Algorithm)



An assumption

- We access to a fair coin, or adequate number of random bites.
(we will see other types of randomness, like unfair coin, fulfill our requirements too)



Example: Primality testing

$$PRIMES = \{p \in \mathbb{N} \mid p \text{ is prime}\}$$

Probabilistic Algorithm

On input p :

if p is even then

 | if $p = 2$, accept and reject otherwise;

end

Sample k independent random numbers $a_1, \dots, a_k \in \{1, \dots, p-1\}$;

for $i = 1$ to k do

 | Calculate $a_i^{p-1} \bmod p$. Reject if it is not 1;

 | Factorize $p-1$ as $p-1 = s \cdot 2^l$ in which s is odd;

 | Calculate series $y_0 = (a_i^{s \cdot 2^0} \bmod p), \dots, y_l = (a_i^{s \cdot 2^l} \bmod p)$;

 | If y_0, \dots, y_l wasn't like $(\dots, -1, 1, 1, \dots, 1)$, Reject;

end

If no reject occurred, accept;

Some lemmas

Lemma

If p is prime, then for all $a \in \{1, \dots, p-1\}$, there is $a^{p-1} \bmod p = 1$.

For odd composite p :

- $\exists a \in \{1, \dots, p-1\}$ such that $a^{p-1} \bmod p \neq 1$ (at least for half of such a 's).
- $\forall a \in \{1, \dots, p-1\}$: $a^{p-1} \bmod p = 1 \rightarrow$ Carmichael numbers.

Lemma

If p is a Carmichael numbers, $\exists q \notin \{1, -1\}$ such that $q^2 \bmod p = 1$.

Some lemmas

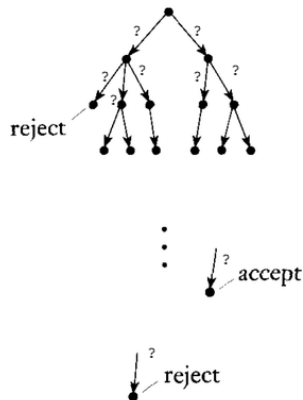
Lemma

For every odd composite p , for at least half of $\{1, \dots, p-1\}$, the last non-one element of series y_0, y_1, \dots, y_l in which $y_t = (a^{s \cdot 2^t} \bmod p)$, is not -1 .

Therefore are rejected with error less than 2^{-k} .

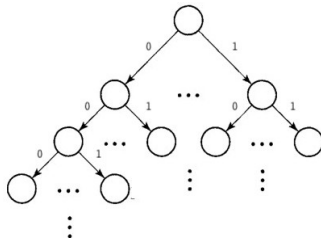
Probabilistic Turing Machine

Probabilistic Turing Machine (PTM)



Probabilistic Turing Machine

$$\langle Q, \Sigma, \Gamma, (\delta_0, \delta_1), q_0, q_a, q_r \rangle$$

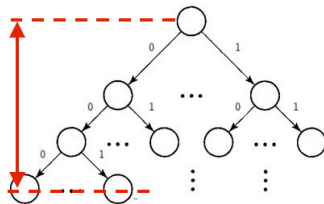


Tosses a coin at each node to use δ_0 or δ_1

Probabilistic Turing Machine

- $\delta_0 = \delta_1 \Rightarrow PTM$ would be a DTM
- $Pr[M \text{ accepts } x] = \sum_{b \text{ is accepting branch}} Pr[b] = \frac{\# \text{accepting branches}}{\# \text{all branches}}$
(considering same height for all branches)
- No non-determinism (\Rightarrow practically implementable)

PTM computation time



Length of the longest computation branch

Bounded-error Probabilistic Decider (BP-Desider)

For a language L , PTM M is called a BP-Decider for L by error ϵ , if on every $x \in \Sigma^*$:

$$x \in L \Rightarrow \Pr[M \text{ accepts } x] \geq 1 - \epsilon$$

$$x \notin L \Rightarrow \Pr[M \text{ rejects } x] \geq 1 - \epsilon$$

i.e. $\Pr[M(x) = L(x)] \geq 1 - \epsilon$.

Bounded-error Probabilistic TIME (BPTIME) classes

$$BPTIME(T(n)) = \{L \subseteq \Sigma^* \mid \exists M \in PTMs \text{ such that } \forall x : Pr[M(x) = L(x)] \geq \frac{2}{3} \\ \text{and } M \text{ decides } x \text{ on } O(T(|x|)) \text{ time}\}$$

We will see later that $\frac{2}{3}$ could be replaced with any number $p > \frac{1}{2}$.

Bounded-error Probabilistic Polynomial-time (BPP) class

$$BPP = \bigcup_{k=0}^{\infty} BPTIME(n^k)$$

BPP equivalent definition

$$BPP = \bigcup_{k=0}^{\infty} BPTIME(n^k)$$

Equivalently, $L \in BPP$ iff there exists some TM M (uses a bit stream r to decide about which transition function to chose) and a polynomial $P : \mathbb{N} \rightarrow \mathbb{N}$ which for every $x \in \Sigma^*$:

$$Pr_{r \in_R \{0,1\}^{P(|x|)}} [M(x, r) = L(x)] \geq \frac{2}{3}$$

BPP relationships

- $P \subseteq BPP$ (and is guessed $BPP = P$, though an open problem yet)
- $BPP \subseteq EXP$ (even $BPP \stackrel{?}{=} NEXP$ is open!)
- No proved relation with NP

Example

$$ZEROP = \{\langle p \rangle \mid p \text{ is a polynomial and } p(x_1, \dots, x_n) \equiv 0\}$$

(Equivalent to polynomials equality, and same with RO branching program)

Probabilistic Algorithm

Lemma

if $p \neq 0$, $\deg(p) = d$ and $S \subseteq \mathbb{N}$ to be a finite set, for a_1, \dots, a_n randomly sampled from S :

$$\Pr[p(a_1, \dots, a_n) \neq 0] \geq 1 - \frac{d}{|S|}$$

For $d \leq 2^m$ and $S = \{1, \dots, 10 \cdot 2^m\}$, $1 - \frac{d}{|S|} \geq 0.9$.

($m = |\langle p \rangle|$ and $\langle p \rangle$ is represented like a circuit with $\{\times, +, -\}$ instead of $\{\wedge, \vee, \neg\}$)

Problem: terms like $a^d = O((10 \cdot 2^m)^{2^m})$.

Example (cont.)

Idea: do calculations in mod $k \in \{1, 2, \dots, 2^{2m}\}$.

Lemma

Number y has at most $\log_2 y$ prime factors.

So for $y = O((10 \cdot 2^m)^{2^m})$, number of prime factors would be at most $(m + \log_2 10) \cdot 2^m = o(\frac{2^{2m}}{2m}) < \frac{2^{2m}}{4m}$.

Lemma

There is at least $\frac{2^{2m}}{2m}$ prime numbers among $\{1, 2, \dots, 2^{2m}\}$.

Therefore at least $\frac{2^{2m}}{4m}$ prime numbers, different to prime factors of y , exist in $\{1, 2, \dots, 2^{2m}\}$. Hence $Pr[k \nmid y = p(a_1, \dots, a_n)] \geq \frac{1}{4m}$.

RP, coRP and ZPP Classes

Randomized TIME (RTIME) classes

$RTIME(T(n))$ is the class of all languages like L for which there exists some PTM M such that for every input $x \in \Sigma^*$:

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

$$x \notin L \Rightarrow Pr[M(x) = 0] = 1$$

and running time of M is $O(T(|x|))$.

(Again $\frac{2}{3}$ could be replaced with any positive number $0 < p < 1$)

Note: we can be sure of result when we see $M(x) = 1$.

Randomized Polynomial-time (RP) class

$$RP = \bigcup_{k=0}^{\infty} RTIME(n^k)$$

coRP class

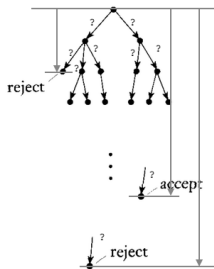
$$coRP = \{L \mid \bar{L} \in RP\}$$

RP and coRP relationships

- $RP \subseteq NP$
- $coRP \subseteq coNP$
- $RP, coRP \subseteq BPP$
- $P \subseteq RP, coRP$

Zero-sided error TIME (ZTIME) classes

$ZTIME(T(n))$ is the class of all languages like L for which there exists some PTM M such that for every input $x \in \Sigma^*$, $M(x) = L(x)$ and expected running time is $O(T(|x|))$.



Zero-sided error Probabilistic Polynomial-time (ZPP) class

$$ZPP = \bigcup_{k=0}^{\infty} ZTIME(n^k)$$

ZPP (other definition)

ZPP is the class of all languages like L for which there exists some polynomial-time PTM M such that for every input $x \in \Sigma^*$, $M(x) \in \{0, 1, \perp\}$ and $Pr[M(x) = \perp] < \frac{1}{2}$ and:

$$x \in L \Rightarrow M(x) \neq 0$$

$$x \notin L \Rightarrow M(x) \neq 1$$

Proof idea

- $ZPP_2 \subseteq ZPP_1$:

Repeat M until anything else than \perp appears.

$$\sigma = \Pr[M(x) \neq \perp] \geq \frac{1}{2}$$

$$\mathbb{E}[T] = \sigma \cdot 1 + (1 - \sigma) (\mathbb{E}[T] + 1) \implies \mathbb{E}[T] = \frac{1}{\sigma} \leq 2$$

- $ZPP_1 \subseteq ZPP_2$:

Halt after running $3T(n)$ and output \perp .

Markov Inequality

$$\Pr[T > a] \leq \frac{\mathbb{E}[T]}{a}$$

ZPP vs RP and coRP

Theorem

$$ZPP = RP \cap coRP$$

- $ZPP \subseteq RP$ ($coRP$):

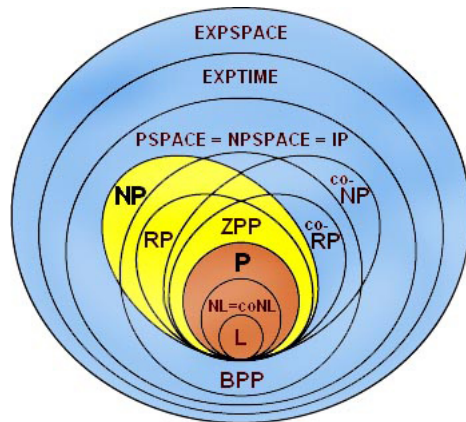
Halt after running $3T(n)$ and output 0 (1).

- $RP \cap coRP \subseteq ZPP$:

M_1 no mistake on reject, and M_2 no mistake on accept.

If $M_1(x) = M_2(x) = 1$, output 1; if $M_1(x) = M_2(x) = 0$, output 0; and otherwise ($M_1(x) = 0$ and $M_2(x) = 1$) output \perp .

Overview



Any BPP-complete?

- Probably no! (hard to find any such language)
- Semantic definition of *BP – Decider*: at least $\frac{2}{3}$ branches of M must be equal (checking this characteristic is undecidable).
- Example:
 $L = \{\langle M, x, 1^t \rangle \mid M \text{ accepts } x \text{ with probability more than } \frac{2}{3} \text{ in at most } t \text{ steps}\}$
 U_{BPP} : simulator $\Rightarrow L \in BPP - hard$
 but U_{BPP} does not hold the characteristic for ill M 's.

Any hierarchy theorem for BPP?

- Probably not and again hard
- Even unknown: $BPTIME(n) \stackrel{?}{=} BPTIME(n^{(\log n)^{10}})$

Robustness

Robustness

$$L \in \text{BPP} \iff \begin{cases} \Pr[M(x) = L(x)] \geq \frac{2}{3} \\ \Pr[M(x) \neq L(x)] \leq \frac{1}{3} \end{cases}$$

What's so special about $\frac{1}{3}$?

The magnitude of failure

$$L \in \text{BPP} \iff \begin{cases} \Pr[M(x) = L(x)] \geq \frac{2}{3} \\ \Pr[M(x) \neq L(x)] \leq \frac{1}{3} \end{cases}$$

What's so special about $\frac{1}{3}$? Nothing! any value less than $\frac{1}{2}$ works.

The magnitude of failure

$\forall c > 0 :$

$$BPP_{\frac{1}{2}+n^{-c}} := \{L \mid \exists \text{ poly-time PTM } M, \forall x : Pr[M(x) = L(x)] \geq \frac{1}{2} + |x|^{-c}\}$$

The magnitude of failure

$\forall c > 0 :$

$$BPP_{\frac{1}{2}+n^{-c}} := \{L \mid \exists \text{ poly-time PTM } M, \forall x : Pr[M(x) = L(x)] \geq \frac{1}{2} + |x|^{-c}\}$$

- n^{-c} is noticeable
- success probability $\geq \frac{1}{2} + n^{-c}$
- $BPP = BPP_{\frac{2}{3}}$

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- n^{-c} is noticeable
- success probability $\geq \frac{1}{2} + n^{-c}$
- $BPP = BPP_{\frac{2}{3}}$

$$BPP_{\frac{2}{3}} \stackrel{?}{=} BPP_{\frac{3}{4}} \stackrel{?}{=} BPP_{\frac{501}{1000}}$$

Amplification lemma

Theorem (Error reduction for BPP)

$\forall c > 0, \exists$ poly-time PTM $M, \forall x : Pr[M(x) = L(x)] \geq \frac{1}{2} + |x|^{-c}$

\Downarrow

$\forall d > 0, \exists$ poly-time PTM $M', \forall x : Pr[M'(x) = L(x)] \geq 1 - 2^{-|x|^d}$

Amplification lemma

Theorem (Error reduction for *BPP*)

$\forall c > 0, \exists$ poly-time *PTM* $M, \forall x : Pr[M(x) = L(x)] \geq \frac{1}{2} + |x|^{-c}$

\Downarrow

$\forall d > 0, \exists$ poly-time *PTM* $M', \forall x : Pr[M'(x) = L(x)] \geq 1 - 2^{-|x|^d}$

- error probability $\leq 2^{-|x|^d}$ (negligible)

Amplification lemma

Theorem (Error reduction for *BPP*)

$$\forall c > 0, \exists \text{ poly-time PTM } M, \forall x : \Pr[M(x) = L(x)] \geq \frac{1}{2} + |x|^{-c}$$

\Downarrow

$$\forall d > 0, \exists \text{ poly-time PTM } M', \forall x : \Pr[M'(x) = L(x)] \geq 1 - 2^{-|x|^d}$$

Proof idea :

- M' simulates M many times
- Output the majority value

Error reduction Algorithm

M' : on input x

1. Simulate M on x for $8|x|^{2c+d}$ times.
2. If the majority of outputs are 1, *output* 1; otherwise *output* 0.

$$BPP_{\frac{1}{2}+n^{-c}} = BPP$$

Theorem

$$\forall c > 0 : BPP_{\frac{1}{2}+n^{-c}} = BPP$$

$$BPP_{\frac{1}{2}+n^{-c}} = BPP$$

Theorem

$$\forall c > 0 : BPP_{\frac{1}{2}+n^{-c}} = BPP$$

- Proof : Clearly $BPP = BPP_{\frac{2}{3}} \subseteq BPP_{\frac{1}{2}+n^{-c}}$

$$BPP_{\frac{1}{2}+n^{-c}} \subseteq BPP_{1-2^{-n^d}} \subseteq BPP_{\frac{2}{3}} = BPP$$

Unfair Coin

- Fair coin : $Pr[Head] = Pr[Tail] = \frac{1}{2}$
- ρ -biased coin :
$$\begin{cases} Pr[Head] = \rho \\ Pr[Tail] = 1 - \rho \end{cases} \quad 0 \leq \rho \leq 1$$
- Turing Machine + Fair coin VS. Turing Machine + Unfair coin

Unfair Coin

Theorem

- ρ -biased coin :
$$\begin{cases} Pr[Head] = \rho \\ Pr[Tail] = 1 - \rho \end{cases}$$
- The i th bit of ρ is computable in $poly(i)$ time

\implies ρ -biased coin can be simulated with a standard PTM in expected time $O(1)$

Unfair Coin

P : on input $\rho = 0.\rho_1\rho_2\cdots$

For $i = 1, 2, \dots$:

Generate random bit b_i

1. If $b_i > \rho_i$, output *Tail*.
2. If $b_i < \rho_i$, output *Head*.

Unfair Coin

P : on input $\rho = 0.\rho_1\rho_2\cdots$

For $i = 1, 2, \dots$:

Generate random bit b_i

1. If $b_i > \rho_i$, output *Tail*.

2. If $b_i < \rho_i$, output *Head*.

example : let $\rho = \frac{3}{4} = (0.11000\cdots)_2$

$$\left\{ \begin{array}{l} 0 \rightarrow Head \quad \left(\frac{1}{2}\right) \\ 1 \left\{ \begin{array}{l} 0 \rightarrow Head \quad \left(\frac{1}{4}\right) \\ 1 \left\{ \begin{array}{l} 0 \rightarrow \cdots Tail \quad \left(\frac{1}{8}\right) \\ 1 \rightarrow Tail \quad \left(\frac{1}{8}\right) \end{array} \right. \end{array} \right. \end{array} \right.$$

Unfair Coin

P : on input $\rho = 0.\rho_1\rho_2\cdots$

For $i = 1, 2, \dots$:

Generate random bit b_i

1. If $b_i > \rho_i$, output *Tail*.

2. If $b_i < \rho_i$, output *Head*.

$$\rho = \rho_1 \cdot 2^{-1} + \rho_2 \cdot 2^{-2} + \cdots$$

$$Pr[Head] = \sum_{i=1}^{\infty} \rho_i \frac{1}{2^i} = \rho$$

$$E[T] = \sum_{i=1}^{\infty} E[T \mid i_{stop} = i] \cdot Pr[i_{stop} = i] = \sum_{i=1}^{\infty} i^c \frac{1}{2^i} = O(1)$$

Unfair Coin

Theorem

- Fair coin: $Pr[Head] = Pr[Tail] = \frac{1}{2}$
- *PTM* M has access to a ρ -biased coin.

\implies Fair coin can be simulated by M in expected time $O\left(\frac{1}{\rho(1-\rho)}\right)$

Unfair Coin

1. Generate two bits b_1 and b_2
2. If $b_1 = \textit{Head}$ and $b_2 = \textit{Tail}$, output *Head*
3. If $b_1 = \textit{Tail}$ and $b_2 = \textit{Head}$, output *Tail*
4. If $b_1 = b_2$ Generate two fresh random bits then go to 2.

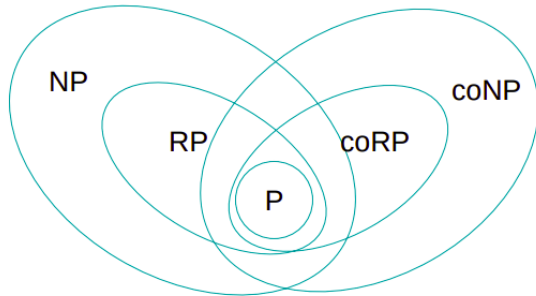
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4. If $b_1 = b_2$ Generate two fresh random bits then go to 2.

$$\Pr[\text{Halt}] \text{ in each step} = \rho(1 - \rho) \implies E[T] = \frac{1}{\rho(1-\rho)} = O(1)$$

BPP

- We know :



BPP is in *PH*

Theorem

$$BPP \subseteq \Sigma_2^p \cap \Pi_2^p$$

Adleman's theorem

Theorem (Adleman's theorem)

$$BPP \subset P_{poly}$$

Adleman's theorem

Theorem (Adleman's theorem)

$$BPP \subset P_{/poly}$$

- Goal : $\exists r_0, M, \forall x : |x| = n \implies M(x, r_0) = L(x)$
- Proof idea : $\begin{cases} \text{error reduction} \\ \text{union bound} \end{cases}$

Adleman's theorem

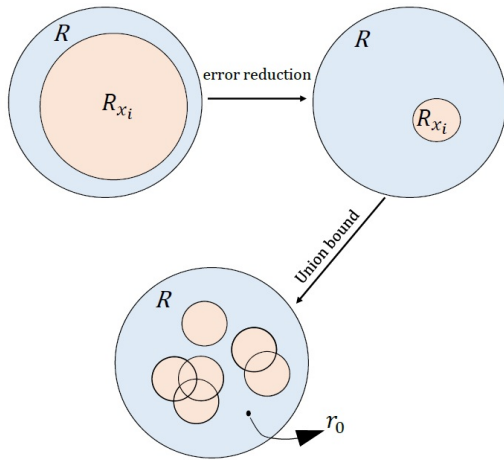
$$|x| = n \text{ and } |r| = m$$

$$R_{x_i} = \{r \mid M(x, r) \neq L(x)\}$$

$$\Pr_{r \in_R \{0,1\}^m} [M(x_i, r) \neq L(x_i)] < \epsilon$$

$$\implies \frac{|R_{x_i}|}{|R|} < \epsilon \implies |R_{x_i}| < \epsilon |R|$$

$$\sum_{i=1}^{2^n} |R_{x_i}| < |R|$$



Randomized Reduction

Randomized Reduction

Definition (Randomized reduction)

$$B \leq_r C := \exists \text{ polynomial-time PTM } M, \forall x : \Pr[C(M(x)) = B(x)] \geq \frac{2}{3}$$

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- $B \leq_p C := \exists M, \forall x : \Pr[C(M(x)) = B(x)] = 1$
- Not transitive
- $M(x) \in \{0, 1\}^*$

Randomized Reduction

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$$B \leq_r C := \exists \text{ polynomial-time PTM } M, \forall x : \Pr[C(M(x)) = B(x)] \geq \frac{2}{3}$$

- $B \leq_p C := \exists M, \forall x : \Pr[C(M(x)) = B(x)] = 1$
- Not transitive
- $M(x) \in \{0, 1\}^*$

Definition (transitive randomized reduction)

$$B \leq_r C := \\ \exists \text{ polynomial-time PTM } M, \forall x : \Pr[C(M(x)) = B(x)] \geq 1 - 2^{-|x|^d}$$

Randomized Reduction

Observation : $C \in BPP$ & $B \leq_r C \implies B \in BPP$

Randomized Reduction

Observation : $C \in BPP$ & $B \leq_r C \implies B \in BPP$

$$\exists M_C, \forall x : Pr[M_C(x) = C(x)] \geq \frac{2}{3} \rightarrow Pr[M'_C(x) = C(x)] \geq \frac{11}{12}$$

$$\exists M_r, \forall x : Pr[C(M_r(x)) = B(x)] \geq \frac{2}{3}$$

M_B : On input x

1. Simulate M_r on x to obtain $M_r(x)$
2. Simulate M'_C on $M_r(x)$, output $M'_C(M_r(x))$

$BP \cdot \mathcal{C}$

Definition : $BP \cdot \mathcal{C}$

$\forall \mathcal{C} : (\text{Class of languages})$

$L \in BP \cdot \mathcal{C} \iff \exists D \in \mathcal{C} \text{ \& polynomial-time PTM } M :$

$$\begin{cases} x \in L \implies \Pr[M(x) \in D] \geq \frac{2}{3} \\ x \notin L \implies \Pr[M(x) \notin D] \geq \frac{2}{3} \end{cases}$$

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Theorem

$$BP \cdot NP = \{L \mid L \leq_r SAT\}$$

- $NP = \{L \mid L \leq_p SAT\}$

$BP \cdot NP$ (Probabilistic NP)

Theorem

$$BP \cdot NP = \{L \mid L \leq_r SAT\}$$

$\{L \mid L \leq_r SAT\} \subseteq BP \cdot NP$ by definition.

$$BP \cdot NP \subseteq \{L \mid L \leq_r SAT\} : \\ C \in BP \cdot NP \implies \exists D \in NP : C \leq_r D$$

$$\left. \begin{array}{l} C \leq_r D \\ D \leq_p SAT \end{array} \right\} C \leq_r D \leq_p SAT \implies C \leq_r SAT$$

$BP \cdot NP$ (Probabilistic NP)

Theorem

$$BP \cdot NP = \{L \mid L \leq_r SAT\}$$

$\{L \mid L \leq_r SAT\} \subseteq BP \cdot NP$ by definition.

$$BP \cdot NP \subseteq \{L \mid L \leq_r SAT\} : \\ C \in BP \cdot NP \implies \exists D \in NP : C \leq_r D$$

- $BPP \subseteq BP \cdot NP$
- $NP \subseteq BP \cdot NP$

$$\left. \begin{array}{l} C \leq_r D \\ D \leq_p SAT \end{array} \right\} C \leq_r D \leq_p SAT \implies C \leq_r SAT$$

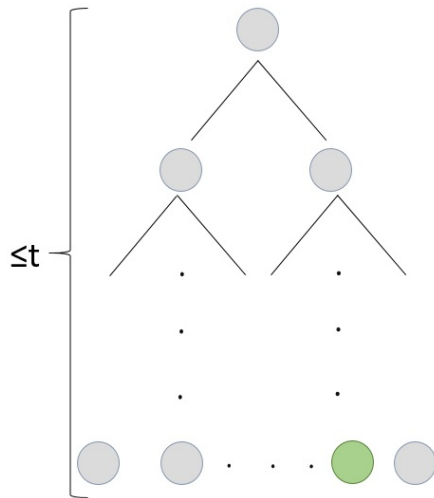
Randomized Space-Bounded Computation

$$L \in RL \iff \exists O(\log n)\text{-space } PTM M : \begin{cases} x \in L \implies Pr[M(x) = 1] \geq \frac{2}{3} \\ x \notin L \implies Pr[M(x) = 0] = 1 \end{cases}$$

$$L \in BPL \iff \exists O(\log n)\text{-space } PTM M : \begin{cases} x \in L \implies Pr[M(x) = 1] \geq \frac{2}{3} \\ x \notin L \implies Pr[M(x) = 0] \geq \frac{2}{3} \end{cases}$$

Definitional Issue

$RL \subseteq NL$ similar to $RP \subseteq NP$



Definitional Issue

$$NL \subseteq RL :$$

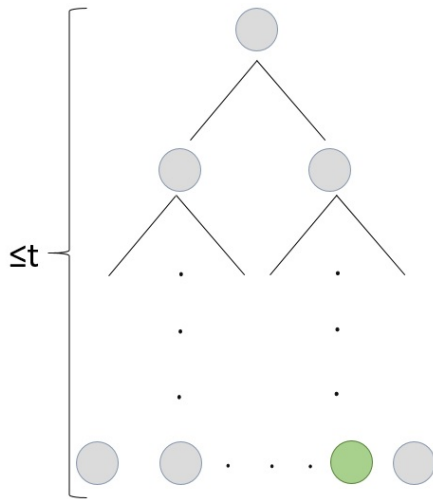
$$L \in NL \implies \exists \text{ poly-time } N : N(x) = L(x)$$

$$t = P(|x|)$$

$$Pr[\text{find an accepting path}] \geq 2^{-t}$$

$$E[T] = 2^t$$

$$O(\log \log n) \text{ counter}$$



UPATH






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 - G is an undirected graph
 - s and t are connected in G
- Recall $PATH$ is NL -complete
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- M : On input $\langle G, s, t \rangle$
1. Take a random walk of length $100n^4$ starting from s
 2. Accept iff the walk reaches t within $100n^4$ steps.

References

References

-  Computational complexity, a modern approach; Sanjeev Arora, Boaz Barak
-  Introduction to the theory of computation; Michael Sipser; Third edition
-  Computational complexity, a conceptual perspective; Oded Goldreich
-  Rafael Pass's lecture notes on Theory of Computing
-  Markov Chains and Random Walks

Thanks :))