# Prob. Algorithms - Home Assignment \#1 

December 13, 2004

1.     - Need to show: $R P \subseteq N P$. We show that any language $R P$ is also in $N P$.

Let $L$ be a language and let $A$ be a randomized algorithm such that for every $x$
$-\operatorname{Prob}_{\Omega}\left[\right.$ Time $\left._{A}(x) \leq p(|x|)\right]=1$

- If $x \in L$ then $\operatorname{Prob}_{\Omega}\left[A(x)=1_{L}(x)\right] \geq 1-1 / 3$.
- If $x \notin L$ then $\operatorname{Prob}_{\Omega}\left[A(x)=1_{L}(x)\right]=1$.
where $\Omega$ is the set of all possible coin tosses' results.
We define the following nondeterminsitc algorithm $A^{\prime}: A^{\prime}$ simulates $A$ except that for any coin flipped by $A, A^{\prime}$ makes a nondeterministic choice from the set $\{0,1\}$. If the nondeterministic choice is 0 then the simulation of $A$ proceeds as if a coin toss has been resulted with "tails". If the nondeterministic choice is 1 then the simultation of $A$ proceeds as if a coin toss has been resulted with "heads". The simulation of $A$ continues and $A^{\prime}$ accepts an input iff $A$ accepts it.
We have to show that all computational paths of $A^{\prime}$ are of polynomial length and that the following two statements hold:
(a) $x \in L$ implies some computational path of $A^{\prime}$ accepts $x$.
(b) $x \notin L$ implies all computational path of $A^{\prime}$ rejects $x$.

Since $A$ runs in time polynomial in the input length, with probability 1 , it follows that for all possible coin tosses, $A$ runs in time polynomial in the input's length, and this implies that all computational paths of $A^{\prime}$ (each of which corresponds to a series of coin tosses,) are of length polynomial in the input's length.
To see that the two other statements are valid, first notice that for all coin tosses, $A$ always reject $x \notin L$. That means that for every possible coins tosses, $A$ rejects $x \notin L$. This in turn implies that all computational paths of $A^{\prime}$ rejects $x$, if $x \notin L$. Lastly, note that if $x \in L$ then there exists a series of coin tosses which makes $A$ accept $x$. In fact, most (a fraction of $(1-\epsilon)$ ) such coin tosses leads $A$ to accept $x$. Hence, since every series of coin tosses in $A$ is a nondeterministic path in $A^{\prime}$, we have that most computational paths of $A^{\prime}$ accepts $x$, and in particular, there exists a computational path for $A^{\prime}$ which accepts $x$, for every $x \in L$.

- Need to show: $Z P P \subseteq R P \subseteq B P P$.
- We show: $Z P P \subseteq R P$. We show that every language $L \in Z P P$ is in $R P$. Let $L$ be a language in $Z P P$. Let $A$ be an algorithm deciding $L$ within expected time $p(|x|)$ and with no errors. We devise an algorithm $A^{\prime}$, as follows. Let $t=3$. Given an input $x$, the algorithm $A^{\prime}$ simulates $A$ on the input $x$, except that whenever the simulation takes time more than $t p(|x|), A^{\prime}$ stops the simulation and enters a rejecting state (i.e., says $x$ is not in L.) If the simulation returns some answer before the time limit is reached, then $A^{\prime}$ reports that answer which was reported by $A$.
By construction of $A^{\prime}, A^{\prime}$ always runs in time at most $3 p(|x|)$.
The only possibility for $A^{\prime}$ to make an error is if $x \in L$ and the simulation of $A$ took more than $3 p(|x|)$ time. By Markov's inequality, $\operatorname{Pr}[$ time more than $3 p(|x|)] \leq$ $1 / 3$ Hence, for $x \in L, A^{\prime}$ returns the right answer with probability at least 1-1/3.
- We show: $R P \subseteq B P P$. This follows from definition.
- Need to show: $R P_{1-1 /|x|}=R P_{1 / e|x|}$

Note: clearly showing the above is enough as $R P_{1 / 3}$ is covered by that case.
Let $A$ be an algorithm deciding $L$. Assume $A$ has the following properties:
(a) $\operatorname{Pr}\left[\operatorname{Time}_{A}(x) \leq p(|x|)\right]=1$.
(b) If $x \in L$ then $\operatorname{Pr}\left[A(x)=1_{L}(x)\right] \geq 1-(1-1 /|x|)=1 /|x|$.
(c) If $x \notin L$ then $\operatorname{Pr}\left[A(x)=1_{L}(x)\right]=1$.

We give an algorithm $A^{\prime}$ deciding $L$, with the following properties:
(a) $\operatorname{Pr}\left[\operatorname{Time}_{A}(x) \leq|x|^{2} p(|x|)\right]=1$.
(b) If $x \in L$ then $\operatorname{Pr}\left[A(x)=1_{L}(x)\right] \geq 1-1 / e^{|x|}$.
(c) If $x \notin L$ then $\operatorname{Pr}\left[A(x)=1_{L}(x)\right]=1$.

The algorithm $A^{\prime}$, given an input $x$, simulates $A$ on $x$ for $|x|^{2}$ times. If for some simulation of $A$ on $x, A$ accepts $x$ then $A^{\prime}$ accepts $x$. Otherwise, $A^{\prime}$ rejects $x$.
We next show that indeed $\operatorname{Pr}\left[A(x)=1_{L}(x)\right] \geq 1-1 / e^{|x|}$. If $x \in L$ then any one simulation of $A$ on $x$ is bound to give an error answer with probability at most $1-1 /|x|$. The probability that all $|x|^{2}$ simulations will give the wrong answer on an input $x \in L$ is at most

$$
\operatorname{Pr}[A \text { errors on } x \in L \text { for } t \text { times }] \leq(1-1 /|x|)^{|x|^{2}}
$$

Since

$$
(1-1 /|x|)^{|x|^{2}} \leq e^{-|x|}
$$

the probability that $A$ rejects an input $x \in L$ is at most $e^{-|x|}$. In other words, if $A^{\prime}$ simulates $A$ for $|x|^{2}$ times and if $x \in L$ then with probability at least $1-1 / e^{|x|}$, one of the $t=|x|^{2}$ simulations of $A$ by $A^{\prime}$ will accept the input $x$.

Note the obvious: if $x \notin L$ then $A$ rejects $x$ with probability 1 , since so does $A^{\prime}$. Since $A^{\prime}$ has the property that $\operatorname{Pr}\left[A(x)=1_{L}(x)\right] \geq 1-1 / e^{|x|}$, and it runs in time $|x|^{2} p(|x|)$, we have showed that $R P_{1-1 /|x|} \subseteq R P_{1 / e^{|x|}}$.

- Need to show: $B P P_{1 / 2-1 /|x|}=B P P_{1 / 2^{|x|}}$

Let $L$ be a language and let $A$ be an algorithm deciding $L$ having the following properties:
(a) $\operatorname{Pr}\left[\operatorname{Time}_{A}(x) \leq p(|x|)\right]=1$.
(b) $\operatorname{Pr}\left[A(x)=1_{L}(x)\right] \geq 1-1 / 2+1 /|x|$.

We devise an algorithm $A^{\prime}$ deciding $L$ with the following properties:
(a) $\operatorname{Pr}\left[\operatorname{Time}_{A^{\prime}}(x) \leq|x|^{3} p(|x|)\right]=1$
(b) $\operatorname{Pr}\left[A^{\prime}(x)=1_{L}(x)\right] \geq 1-1 / 2^{|x|}$.

Given an instance $x, A^{\prime}$ simulates $A$ on $x$ for $|x|^{3}$ times. If the majority of the answers yielded by simulations of $A$ on $x$ resulted with accepting $x$, then $A^{\prime}$ accepts $x$. Otherwise, $A^{\prime}$ rejects $x$.
We first claim that

$$
\operatorname{Pr}\left[A^{\prime}(x)=1_{L}(x)\right] \geq 1-1 / 2^{|x|}
$$

We show this claim is valid. $A^{\prime}$ makes an "error" if
$-x \notin L$ and the number of accepts by the $|x|^{3}$ simulations of $A$ on $x$ is more than $|x|^{3} / 2$.
$-x \in L$ and the number of rejects by the $|x|^{3}$ simulations of $A$ on $x$ is more than $|x|^{3} / 2$.
Given $x \in L$, the expected number of accepts by simulations of $A$ on $x$ is $|x|^{3}(1 / 2+$ $1 /|x|)=|x|^{3} / 2+|x|^{2}$. Let $E=|x|^{3} / 2+|x|^{2}$.
Hence, the probability that $A^{\prime}$ does not accept $x \in L$ is the probability that the number of simulations of $A$ on $x$ accepting $x$ is less than $|x|^{3} / 2$. By Chernoff, this is at most

$$
\begin{aligned}
\operatorname{Pr}\left[\# \text { accepts of } A \text { at most }|x|^{3} / 2\right] & = \\
\operatorname{Pr}\left[\mid \# \text { accepts of } A-E\left|\geq|x|^{2}\right]\right. & \leq \\
2^{-\frac{|x|^{2}}{|x|^{3} / 2+|x|^{2}} \frac{|x|^{3} / 2+|x|^{2}}{6}} & \approx \\
2^{-|x|} &
\end{aligned}
$$

The analyzis for $x \notin L$ is similiar. Hence, we have shown that $A^{\prime}$ accepts $x \in L$ with probability at least $1-2^{-|x|}$.
It is also clear that

$$
\operatorname{Pr}\left[\operatorname{Time}_{A^{\prime}}(x) \leq|x|^{3} p(|x|)\right]=1
$$

Hence: $B P P_{1 / 2-1 /|x|}=B P P_{1 / 2|x|}$.

- Need to show: $Z P P=R P \cap c o R P$.

Let $L$ be a language in $R P \cap \operatorname{coRP}$. Then there are two algorithms deciding $L$ : a first one, $A_{1}$, in $R P$ and a second, $A_{2}$, in coRP. The first algorithm has the property that if it accepts an instance then surely the instance is in $L$. The second algorithm on the other hand, has the property that if it rejects an instance then that instance is not in $L$ with probability 1.
Given two such algorithms, $A_{1}$ and $A_{2}$, we build an algorithm $A$ which decides $L$ with probability 1 and in expected polynomial (in $|x|$ ) time. The algorithm $A$, given an input $x$, simulates both $A_{1}$ and $A_{2}$, independantly, each for an unbounded number of times - until either $A_{1}$ accepts or $A_{2}$ rejects. With high probability (approaching exponentialy fast to 1, ) since $x$ is either in $L$ or not, one of the simulations - either the one of $A_{1}$ or the one of $A_{2}$ will accept, or reject $x$ after only a polynomial number of simulations. Hence, the expected time algorithm $A$ has in polynomial.
Conversly, let $A$ be an algorithm deciding a language $L$ in expected poylnomial time and with probability of success 1 . We've already showed above that this implies $L$ is in $R P$. A symmetric argument also shows that $L$ is in coRP.
2. Let $A$ be an array of length $n$. Denote the sorted elements of $A$ by $a_{1}, a_{2}, \ldots, a_{n}$. Given an integers $a, t, k$, we are interested in finding an $a_{k^{\prime}}$ such that $\left|k-k^{\prime}\right| \leq a n / \sqrt{t}$. Consider the following randomized algorithm:
(a) Pick uniformly at random $t$ elements $b_{1}, b_{2}, \ldots, b_{t}$ from the array.
(b) Sort the elements $b_{1}, \ldots, b_{t}$. Assume w.l.o.g., $b_{i}<b_{j} \Leftrightarrow i<j$.
(c) Return $b_{k}$ (simply by indexing the sorted array of $b_{1}, \ldots, b_{t}$.)

Claim 1 Let $b_{k}=a_{k}^{\prime}$. Then with probability larger than $1-2^{-\Omega\left(a^{2}\right)},\left|k-k^{\prime}\right| \leq a n / \sqrt{t}$.
Proof: Let $\delta n=k$. We have to show that with probability larger than $2^{-\Omega\left(a^{2}\right)}$, we have that $b_{k}=a_{k^{\prime}}$ and $\left|k^{\prime}-k\right| \leq a n / \sqrt{t}$.
Let $X_{i}$ be the indicator ( $0 / 1$ ) random variable that the $i$-th element ( $i \in[t]$ ) we choose is one of the first $\delta n-\frac{a n}{\sqrt{t}}$ elements in the array. Let $Y_{i}$ be the indicator ( $0 / 1$ ) random variable that the $i$-th element we choose is one of the last $\delta n+\frac{a n}{\sqrt{t}}$ elements in the array. Fix $X=\sum_{i} X_{i}$ and $Y=\sum_{i} Y_{i}$.
Clearly, the algorithm fails if either $X \geq \delta n=k$, or if $Y \geq t-\ldots$. We show that the probability of the union of the above two events is small.
We have

$$
P\left[X_{i}\right]=\frac{\delta n-a n / \sqrt{t}}{n}=\delta-a / \sqrt{t}
$$

Hence,

$$
E[X]=\delta t-a \sqrt{t}
$$

We now bound from above, using Chernoff's inequality, the probability that $X \geq \delta n$.

$$
\begin{aligned}
\operatorname{Pr}[X \geq \delta t] & =\operatorname{Pr}[|X-E[X]| \geq a \sqrt{t}] \\
& =\operatorname{Pr}\left[|X-E[X]| \geq\left(\frac{a}{\delta \sqrt{t}}\right) \delta t\right] \\
& \leq 2^{-\frac{a^{2}}{\delta^{2} t} \frac{\delta t}{6}} \\
& =2^{-\Omega\left(a^{2}\right)}
\end{aligned}
$$

The same analysis leads to a similiar upper bound on the event $Y \geq \ldots$. Hence, we conclude that with probability greater than

$$
1-2^{-\Omega\left(a^{2}\right)}
$$

$b_{k}=a_{k^{\prime}}$ with $\left|k^{\prime}-k\right| \leq a n / \sqrt{t}$.
3. (a) Let $x=x_{1}, x_{2}, \ldots, x_{n}, y=y_{1}, y_{2}, \ldots, y_{n}$, where $x_{i}, y_{i}$ are chosen uniformly and independantly of each other, in random from $\{0,1\}$. Let $z_{i}$ be the random variable which equals 1 if $x_{i}=y_{i}$, and equals 0 otherwise. Clearly, since the $x_{i}$ 's are independant of each other (and so does the $y_{i}$ 's), the $z_{i}$ 's are independant. Let $Z=\sum_{i=1}^{n} z_{i}$. Clearly, $Z=d_{H}(x, y)$. We have that $E[Z]=n / 2$, as the probability of $z_{i}$ to be 1 is exactly half (as $x_{i}, y_{i}$ are uniform.) Since the $z_{i}$ 's are independant $0 / 1$ variables, we apply Chernoff's inequlity to obtain

$$
\begin{aligned}
\operatorname{Prob}_{\Omega}\left[d_{H}(x, y)<n / 4\right] & =\operatorname{Prob}_{\Omega}[Z<n / 4] \\
& =\operatorname{Prob}_{\Omega}\left[|Z-n / 2| \geq \frac{1}{2} n / 2\right] \\
& \leq 2^{-\frac{0.5^{2}(n / 2)}{6}} \\
& =2^{-\Omega(n)}
\end{aligned}
$$

(b) Let $c$ be the constant in the upper bound $2^{-\Omega(n)}$ we proved above. We choose uniformly at random $l=2^{c n / 2}$ strings in $\{0,1\}^{n}$.
From what we've proved above, we know that any two strings $x, y$ of the $l$ strings we chose, satisfy $d_{H}(x, y)<n / 4$ with probability at most $2^{-c n}$,
Those, the expected number of strings in $l$ which satisfy $d_{H}(x, y)<n / 4$ is

$$
2^{c n / 2} 2^{-c n}<1
$$

Hence, by the pigeon hole property of the expectation, there exist a set of $l=2^{c n / 2}$ strings in $\{0,1\}^{n}$ such that no two satisfy $d_{H}(x, y)<n / 4$.

