

# Assignment 4 solutions

## Exercise 7.4

$M'(x)$  runs  $M(x)$   $k$  independent times and accepts  $x$  iff  $M(x) = 1$  at least once. Obviously, if  $x \notin L$  we have  $M(x) = 0$  for all  $k$  times, and  $Pr[M'(x) = 0] = 1$ . If  $x \in L$ ,

$$Pr[M'(x) = 0] = Pr[M(x) = 0 \forall k \text{ times}] \leq (1 - n^{-c})^k. \quad (1)$$

If we set  $k := dn^c \ln n$  then, since for large enough  $n$  we have  $(1 - n^{-c})^{n^c} \leq 1/e$  (1) implies  $Pr[M'(x) = 0] \leq n^{-d}$ .

## Exercise 7.6

- (a) First assume that  $M$  exists, and its running time is at most  $n^c$ . We construct a new TM  $M'$ , which runs  $M(x)$  repeatedly, until it gets something other than ?. If  $M'(x)$  terminates, then it terminates with  $L(x)$  by definition. To calculate its *expected* running time, we note that it runs  $M(x)$   $i$  times iff  $M(x) = ?$  for the first  $i - 1$  times, and  $M(x) \neq ?$  the  $i$ -th time. Then the running time is  $in^c$ , and the probability that this event happens is

$$Pr[\text{repeat } M(x) \text{ } i \text{ times}] \leq \frac{1}{2^{i-1}} Pr[M(x) \neq ?] \leq \frac{1}{2^{i-1}}.$$

So, the expected running time for  $M'(x)$  is at most  $\sum_{i=1}^{\infty} \frac{in^c}{2^{i-1}} = 2n^c \sum_{i=1}^{\infty} \frac{i}{2^i} \leq 2Dn^c = O(n^c)$ , where  $D$  is the constant of exercise 7.2. Therefore  $L \in ZPP$ .

Now assume that  $L \in ZPP$ . Then there is a probabilistic TM  $M'$  that outputs the correct answer *if* it terminates, and runs in *expected* polynomial time. If  $T_{M'}(n)$  is the running time of  $M'$ , then  $E[T_{M'}(n)] = \sum_{i=1}^{\infty} i Pr[T_{M'}(n) = i] = n^c$ . We apply Markov's inequality (Lemma A.7 for  $k := l/E[T_{M'}(n)]$ ) for some  $l$  (to be determined), and get

$$Pr[T_{M'}(n) \geq l] \leq \frac{E[T_{M'}(n)]}{l} = \frac{n^c}{l}. \quad (2)$$

If we set  $l := 2n^c$ , the probability of (2) is  $\leq 1/2$ . Let  $M$  be the probabilistic TM that runs  $M'(x)$  for at most  $l = 2n^c$  steps. If it terminates at some point with an answer  $M'(x) = 0$  or  $1$ , then output the answer, else output ?. Because of (2),  $Pr[M(x) = ?] \leq 1/2$ .

- (b) See lecture notes.

## Exercise 7.8

- (a) As described at the beginning of the proof for Theorem 7.14, we have a polynomial-time probabilistic TM  $M$  that, given input  $|x| = n$ , uses  $m = p(n)$  random bits for some polynomial  $p$  so that  $Pr_r[3SAT(M(x, r)) \neq \overline{3SAT}(x)] \leq \frac{1}{2^{n-1}}$ . Continuing with the argument of Theorem 7.14, there is a random string  $r_0$  with  $|r_0| = m$ , such that  $3SAT(M(x, r)) \neq \overline{3SAT}(x)$ ,  $\forall x$ .

- (b) Modify the proof in (a) to show how it works if the input is not just  $x$ , but  $(x, u)$ , where you know that  $u \in \{0, 1\}^{q(|x|)}$  for a polynomial  $q$ . (*Hint:* The only change is that  $r_0$  will now be of size polynomial in  $q(|x|)$ , which is still polynomial in  $|x|$ .)
- (c)  $\Sigma_4^P$  languages have polynomial-time TM  $M'$  s.t.

$$\exists u_1 \forall u_2 \exists u_3 \forall u_4 : M'(x, u_1, u_2, u_3, u_4) = 1.$$

The inner statement  $\forall u_4 : M'(x, u_1, u_2, u_3, u_4) = 1$  is an *coNP* statement, which can be reduced (deterministically in poly-time) to a  $\overline{3SAT}$  statement for fixed  $u_1, u_2, u_3$ , i.e., it is equivalent to  $\overline{3SAT}(R(x, u_1, u_2, u_3))$ . We guess the random string  $r_0$  that works for any  $u_1, u_2, u_3$ , to reduce *deterministically* this  $\overline{3SAT}$  statement to  $3SAT(M(R(x, u_1, u_2, u_3), r_0))$ , i.e., the statement becomes

$$\exists r_0 \exists u_1 \forall u_2 \exists u_3 \exists v_4 : M''(M(R(x, u_1, u_2, u_3), r_0), v_4) = 1$$

for a polynomial certifier  $M''$  of  $3SAT$ , which is a  $\Sigma_3^P$  statement.

### Exercise 7.10

Consider the directed graph that consists of a path  $s \rightarrow v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_{n-2} \rightarrow t$ , plus the edges  $(v_i, s)$ ,  $i = 1, \dots, n-2$ . Each edge can be taken with probability  $1/2$ , except for  $(s, v_1)$  that is taken with probability 1. Then the expected time to reach  $t$  can be calculated recursively

$$E[s \rightarrow t] = E[s \rightarrow v_{n-2}] + \frac{1}{2}1 + \frac{1}{2}E[s \rightarrow v_{n-2}], E[s \rightarrow v_1] = 1$$

or

$$\frac{1}{2}E_n = E_{n-1} + \frac{1}{2}, E_1 = 1$$

which implies  $E_n = \Omega(2^n)$ .