# CIS 511 Homework 7

# Stephen Phillips, Dagaen Golomb

April 15, 2015

## Problem 1

Show that  $STRONGLY - CONNECTED = \{\langle G \rangle \mid G \text{ is a graph that is strongly connected} \}$  is NL-complete

Since NL = co-NL, and we know that  $\overline{STRONGLY} - \overline{CONNECTED}$  is in NL, then  $STRONGLY - \overline{CONNECTED}$  is in co-NL, hence NL.

#### Problem 2

Show that 2SAT is NL-complete.

We show that  $\overline{2SAT}$  is in co-NL, which means that since NL = co-NL that 2SAT is in NL. Given an instance of 2SAT in CNF form, we can interpret the clauses as implications:  $(a+b) \iff (\bar{a} \implies b)$ . We can 'derive' one implication from another by using :  $(u \implies v)(v \implies w) \iff (u \implies w)$ . Since there are only two variables per clause, if the formula is unsatisfiable, then we must be able to derive both  $(x \implies \bar{x})$  and  $(\bar{x} \implies x)$ , since if there was a chain of implications that led to a contradiction then we could reduce the chain using derivations to get to these implications. This in turn leads to:

$$(x \Longrightarrow \bar{x})(\bar{x} \Longrightarrow x) \Longrightarrow (\bar{x} + \bar{x})(x + x) \Longrightarrow \bar{x}x \Longrightarrow 0$$

We can do this by non-deterministically following the implications that derive  $(x \implies \bar{x})$  then again non-deterministically choose the ones that derive  $(\bar{x} \implies x)$ . If we find this line of implications, which we only need to store  $O(\log n)$  bits for the intermediate implication we are using to derive the above.

We know that  $\overline{PATH}$  is co-NL-complete, hence NL-complete. Therefore we want to show  $\overline{PATH} \leq_L 2SAT$ . To do this we consider given a graph G we create the graph  $G^{\mathcal{R}}$ , the graph with all the edges reversed. We know that there is a path from s to t in G if and only if there is a path from t to s in  $G^{\mathcal{R}}$ . We use this in the reduction.

We do this by, given an instance of  $\overline{PATH}$ , meaning a graph  $\langle G, s, t \rangle$ , we create a boolean formula  $\phi$  from the edges. So  $\forall (u,v) \in E$ , we create clause  $(u \Longrightarrow v) = (\bar{u}+v)$  and clause  $(v' \Longrightarrow u') = (\bar{v'}+u')$  (corresponding to  $G^{\mathcal{R}}$ , different variables), then and all these clauses together. Except replace the variables for the nodes s and t that from the instance with literals x and  $\bar{x}$  respectively, and also replace the variables s' and t' with t and t, respectively. This can be done in log space since we only need to keep track of the two variables in the clause we are currently reading, a constant amount of space.

Why does this mapping work? If there is a path from s to t, then we can derive  $x \implies \bar{x}$  by following the implications on the path from s to t. Similarly on the reversed graph we can follow the path from t to s to derive  $\bar{x} \implies x$ . Hence we can always derive both  $x \implies \bar{x}$  and  $\bar{x} \implies x$  when there is a path from s to t. If this is the case then we can derive our unsatisfiability when we showed 2SAT was in NL. Hence if there is a path, this formula is never satisfiable. If there is not a path from s to t, the formula is always satisfiable. We examine several cases:

1. t has a path to s but not the other way around. That means in the G part of the formula we have a clause with  $\bar{x} \implies x$ , so we set x to true, and all variables that t can reach to true. For the variables corresponding to  $G^{\mathcal{R}}$  we set to false since the implication  $x \implies \bar{x}$  is in that set of clauses. Since we can vary the variables independently we can set both to true, and we can therefore set all the clauses to true.

2. s and t have no paths to each other at all. Then we are dealing with separate connected components, and similarly to last case we can just set x to a value and set the implications in the chain appropriately.

So this instance of 2SAT is satisfiable if and only if there is no path from s to t in G.

#### Problem 3

Give an example of an NL-complete context free language.

### Problem 4

Define  $pad: \Sigma^* \times \mathcal{N} \to \Sigma^* \#^*$  as  $pad(s,l) = s \#^l$ . Define the language pad(A,f) for language A and function  $f: \mathcal{N} \to \mathcal{N}$  as

$$pad(A, f) = \{pad(s, f(|s|)) \mid s \in A\}$$

Show that if  $A \in TIME(n^6)$  then  $pad(A, n^2) \in TIME(n^3)$ 

Is this not just that padding extends the length of the string? Now if you run the TM for A on the first part of the string, it will run in time  $O(|s|^6)$ , and since the length of our input is  $n = |s|^2 + |s| = O(|s|^2)$  this becomes  $O(|s|^6) = O((|s|^2)^3) = O(n^3)$  (?)

#### Problem 5

Prove using pad from previous problem that if  $NEXPTIME \neq EXPTIME$  then  $P \neq NP$ .

To do this we show the converse, that  $P = NP \implies NEXPTIME = EXPTIME$ . Given a language  $L \in NEXPTIME$ , we know that there is a non-deterministic Turing machine M that runs in  $2^{O(f(n))}$  where f(n) is a polynomial, bounded by say  $cn^k$ . Now we construct the language

$$pad(L, 2^{cn^k}) = \left\{ x \#^{2^{c|x|^k}} \mid x \in L \right\}$$

as the definition of pad above. This language is in NP, since we construct a machine N that first checks that the string is in the right form of  $x\#^{2^{c|x|^k}}$ , then simulates M on x, which runs in  $O(2^{c|x|^k})$  time, which is linear in the size of the input (due to the padding). Since this runs in non-deterministic polynomial time in the size of the input, by our assumption P = NP, we know that there exists a deterministic polynomial time algorithm that accepts  $pad(L, 2^{cn^k})$ , let us denote it N'.

We construct a machine A that accepts L in deterministic exponential time as follows. First we generate the string  $y = x \#^{2^{c|x|^k}}$  in exponential time. Then we run N'(y), which runs in polynomial time in the length of the input. We return whatever N' returns. This is deterministic since N' is deterministic. It is still in exponential time. Let us say N' has time complexity bounded by  $c'n^{k'}$ , then the runtime of this machine A is  $c'\left(2^{c|x|^k} + |x|\right)^{k'} = O(2^{ck'|x|^k})$  which is still exponential. Therefore, we have constructed a deterministic machine that accepts in exponential time L for any L in NEXPTIME.

This technically only shows that  $P = NP \implies NEXPTIME \subseteq EXPTIME$ , but since  $EXPTIME \subseteq NEXPTIME$  by definition, this is sufficient.

#### Problem 6

Show that for an n variable polynomial P, with degree at most d, and total degree of t. We showed in class that if you pick  $r_1, \ldots, r_n$  uniformly and independently at random in a set S then

$$\mathbf{Pr}[P(r_1, r_2, \dots, r_n) = 0] \le \frac{nd}{|S|}$$

We now want to strengthen this result to:

$$\mathbf{Pr}[P(r_1, r_2, \dots, r_n) = 0] \le \frac{t}{|S|}$$

### Problem 7

Show if  $NP \subseteq BPP$ , then NP = RP

#### Problem 8

Define a ZPP-machine as a probabilistic Turing Machine that can output 3 things: accept, reject, and ?. A ZPP-machine M decided a language A if for every  $x \in L$  it accepts with probability at least 2/3, and rejects with probability 0, for  $x \notin L$  it rejects with probability 2/3 and accepts with probability 0, and it outputs ? on any input with probability at most 1/3. Let ZPP be the set of all languages that decided by ZPP machines. Formally, a language  $L \in ZPP$  if there is a probabilistic turing machine that satisfies

$$\forall x \in L, \mathbf{Pr}[Z(x) = 1] \ge 2/3, \mathbf{Pr}[Z(x) = 0] = 0, \mathbf{Pr}[Z(x) = ?] \le 1/3$$
  
 $\forall x \notin L, \mathbf{Pr}[Z(x) = 0] > 2/3, \mathbf{Pr}[Z(x) = 1] = 0, \mathbf{Pr}[Z(x) = ?] < 1/3$ 

Show that  $ZPP = RP \cap \text{co-}RP$ .

First we show that  $ZPP \supseteq RP \cap \text{co-}RP$ . If a language L is in  $RP \cap \text{co-}RP$ , then there exists two machines  $R'_1$  and  $R'_2$  such that

$$\forall x \in L, \mathbf{Pr}[R'_1(x) = 1] \ge 1/2, \ \mathbf{Pr}[R'_2(x) = 1] = 1$$

and

$$\forall x \notin L, \mathbf{Pr}[R'_1(x) = 0] = 1, \ \mathbf{Pr}[R'_2(x) = 0] \ge 1/2$$

To make the analysis simpler, we use the fact that we can enhance the probabilities by running them several times, so we can create new machines  $R_1$  and  $R_2$  such that:

$$\forall x \in L, \mathbf{Pr}[R'_1(x) = 1] \ge 2/3, \ \mathbf{Pr}[R'_2(x) = 1] = 1$$

and

$$\forall x \notin L, \mathbf{Pr}[R'_1(x) = 0] = 1, \ \mathbf{Pr}[R'_2(x) = 0] \ge 2/3$$

So now we can construct a ZPP machine that accepts this language as follows.

# function Z(x)

Let  $y_1 = R_1(x)$ 

Let  $y_2 = R_2(x)$ 

If  $y_1 = y_2 = 0$ , reject

If  $y_1 = y_2 = 1$ , accept

Output?

### end function

Let us analyze this, using that the randomness of  $R_1$  and  $R_2$  are independent. For  $x \in L$ :

$$\begin{aligned} \mathbf{Pr}[Z(x) = 1] &= \mathbf{Pr}[R_1(x) = 1, R_2(x) = 1] \\ &= \mathbf{Pr}[R_1(x) = 1] \mathbf{Pr}[R_2(x) = 1] \\ &\geq (2/3)(1) = 2/3 \\ \mathbf{Pr}[Z(x) = 0] &= \mathbf{Pr}[R_1(x) = 0, R_2(x) = 0] \\ &= \mathbf{Pr}[R_1(x) = 0] \mathbf{Pr}[R_2(x) = 0] \\ &= \mathbf{Pr}[R_1(x) = 0](0) = 0 \\ \mathbf{Pr}[Z(x) =?] &= \mathbf{Pr}[R_1(x) = 0, R_2(x) = 1 \text{ or } R_1(x) = 1, R_2(x) = 0] \\ &= \mathbf{Pr}[R_1(x) = 0, R_2(x) = 1] + \mathbf{Pr}[R_1(x) = 1, R_2(x) = 0] \\ &= \mathbf{Pr}[R_1(x) = 0] \mathbf{Pr}[R_2(x) = 1] + \mathbf{Pr}[R_1(x) = 1] \mathbf{Pr}[R_2(x) = 0] \\ &\leq (1/3)(1) + \mathbf{Pr}[R_1(x) = 1](0) = 1/3 \end{aligned}$$

So it satisfies all the conditions of ZPP in the case where  $x \in L$ . Now similarly for when  $x \notin L$ :

$$\begin{aligned} \mathbf{Pr}[Z(x) = 0] &= \mathbf{Pr}[R_1(x) = 0, R_2(x) = 0] \\ &= \mathbf{Pr}[R_1(x) = 0] \mathbf{Pr}[R_2(x) = 0] \\ &\geq (1)(2/3) = 2/3 \\ \mathbf{Pr}[Z(x) = 1] &= \mathbf{Pr}[R_1(x) = 1, R_2(x) = 1] \\ &= \mathbf{Pr}[R_1(x) = 1] \mathbf{Pr}[R_2(x) = 1] \\ &= (0) \mathbf{Pr}[R_2(x) = 1] = 0 \\ \mathbf{Pr}[Z(x) = ?] &= \mathbf{Pr}[R_1(x) = 0, R_2(x) = 1 \text{ or } R_1(x) = 1, R_2(x) = 0] \\ &= \mathbf{Pr}[R_1(x) = 0, R_2(x) = 1] + \mathbf{Pr}[R_1(x) = 1, R_2(x) = 0] \\ &= \mathbf{Pr}[R_1(x) = 0] \mathbf{Pr}[R_2(x) = 1] + \mathbf{Pr}[R_1(x) = 1] \mathbf{Pr}[R_2(x) = 0] \\ &\leq (1)(1/3) + (0) \mathbf{Pr}[R_2(x) = 0] = 1/3 \end{aligned}$$

So this also satisfies the conditions for ZPP, hence for any language in  $RP \cap \text{co-}RP$ , we can construct a ZPP machine.

Now we show that  $ZPP \subseteq RP \cap \text{co-}RP$ . First we show that  $ZPP \subseteq RP$ . For a language  $L \in ZPP$ , we have a machine Z that satisfies.

$$\forall x \in L, \mathbf{Pr}[Z(x) = 1] \ge 2/3, \mathbf{Pr}[Z(x) = 0] = 0, \mathbf{Pr}[Z(x) = ?] \le 1/3$$
  
 $\forall x \notin L, \mathbf{Pr}[Z(x) = 0] \ge 2/3, \mathbf{Pr}[Z(x) = 1] = 0, \mathbf{Pr}[Z(x) = ?] \le 1/3$ 

For this we need to construct a machine,  $R_1$ , that satisfies

$$\forall x \in L, \mathbf{Pr}[R_1(x) = 1] \ge 1/2$$
  
 $\forall x \notin L, \mathbf{Pr}[R_1(x) = 0] = 1$ 

We construct it as follows:

function  $R_1(x)$ Let y = Z(x)If y = 1, accept If y = 0 or y = ?, reject end function

The analysis fairly simple. For  $x \in L$ 

$$\Pr[R_1(x) = 1] = \Pr[Z(x) = 1] \ge 2/3 \ge 1/2$$

And for  $x \notin L$ 

$$\Pr[Z(x) = 0] = \Pr[Z(x) = 0 \text{ or } Z(x) = ?]$$
  
= 1 - \Pr[Z(x) = 1]  
= 1 - 0 = 1

Now we show that  $ZPP \subseteq RP$ . So we construct another machine  $R_2$  that satisfies

$$\forall x \in L, \mathbf{Pr}[R_2(x) = 1] = 1$$
  
 $\forall x \notin L, \mathbf{Pr}[R_2(x) = 0] \ge 1/2$ 

Similarly to last time, we construct the following machine:

```
\begin{array}{l} \textbf{function} \ R_2(x) \\ \text{Let} \ y = Z(x) \\ \text{If} \ y = 0, \ \text{reject} \\ \text{If} \ y = 1 \ \text{or} \ y = ?, \ \text{accept} \\ \textbf{end function} \end{array}
```

The analysis is similar to before. For  $x \in L$ 

$$Pr[Z(x) = 1] = Pr[Z(x) = 1 \text{ or } Z(x) = ?]$$
  
= 1 - Pr[Z(x) = 0]  
= 1 - 0 = 1

And for  $x \notin L$ 

$$\mathbf{Pr}[R_1(x) = 0] = \mathbf{Pr}[Z(x) = 0] \ge 2/3 \ge 1/2$$