## Introduction to Theoretical Computer Science

## Exercise Sheet: Week 7 – Solutions

(1) The basic claim was that polynomial problems are 'easy', and non-polynomial problems are hard. Consider  $f(n) = n^{10^{10}}$ , and  $g(n) = 10^{n/10^{10}}$ . Show that  $f(n) \in o(g(n))$ . (Recall this means that  $\forall \epsilon > 0. \exists n_0. \forall n > n_0. |f(n)| \le \epsilon |g(n)|$ .) (Hint: take logs, and remember that you only have to care about large enough n.) Where does g catch up with f?

Given  $\epsilon$ , we want to find where  $f(n) \leq \epsilon g(n)$ , i.e.  $n^{10^{10}} \leq \epsilon \cdot 10^{n/10^{10}}$ . This is not soluble in closed form, so let's just find n big enough. Taking logs,  $10^{10} \log n \leq \log \epsilon + n/10^{10}$ . Divide by  $\log n$  to get  $10^{10} \leq \log \epsilon / \log n + (n/\log n)/10^{10}$ . If we take  $n \geq 1/\epsilon$ , then  $-1 \leq \log \epsilon / \log n \leq 0$ , so now we just need  $(n/\log n) \geq (10^{10}+1)\cdot 10^{10}$ , so taking  $n/\log n \geq \max(1/\epsilon, 10^{21})$  will certainly do.

The point where g(n) = f(n) is best found numerically, and is about  $2.133 \times 10^{21}$ , pretty much where the above proof put it for  $\epsilon = 1$ .

**Bonus:** Where does the statement  $f(n) \in o(g(n))$  fit in the arithmetical hierarchy that we discussed unofficially? (Trick question!)

It's a trick question because as presented  $\epsilon$  is a real, so this is not an arithmetical statement. However, we can replace  $\forall \epsilon > 0....\epsilon...$  by  $\forall N > 0....^1/N...$  to get an equivalent arithmetical statement, which is in  $\Pi_3^0$ .

(2) We defined the class P in terms of polynomially bounded machines. Explain how to implement this definition. That is, given a register machine M (taking input R in  $R_0$  as usual), explain how to construct a machine M' which takes inputs R and k, and behaves like M except that it halts after  $(\lg R)^k$  steps of M's execution.

M' first computes ( $\lg R$ )<sup>k</sup>, by repeated division by 2 (and division by 2 is itself repeated subtraction) and brute force exponentiation (note that we don't care how long this takes since k is a constant), and stashes it in a dedicated register  $R_{-1}$  (or whatever). M' now jumps to a program that is the program of M, except that every instruction I is replaced by the sequence DECJZ(-1,HALT); I, with labels adjusted appropriately.

(3) Show that the Halting problem is not NP-complete. (This is obvious ... but can you prove it?)

Suppose it is. Then there is a non-deterministic Turing/Register Machine M and a polynomial time bound f(n) in the size of the input machine such that M determines the halting answer before time f(n). Let k be the maximum branching degree of the control graph of M – this is at most 2 for an NRM, or (number of states) for an NTM. Then we can simulate M deterministically by interleaving, in time  $O(k^{f(n)})$ , and thus solve the halting problem.

The following is a reasonably tricky algorithm design problem.

(4) 2-SAT is the following problem: given a set of boolean variables  $X_i$ , and a formula  $\phi = \bigwedge_{1 \leq j \leq n} (\alpha_j \vee \beta_j)$ , where each  $\alpha_j, \beta_j$  is a literal, i.e., either a variable or a negated variable, is there a satisfying assignment for  $\phi$ ?

Show that 2-SAT is polynomial (unlike SAT or 3-SAT). (Quite difficult. Hint: look for two clauses that contain a variable and its negation (e.g.  $(X \vee Y)$  and  $(Z \vee \neg Y)$ ), merge them into a single clause, and add it to the formula.)

A proof by using resolution: If a variable only occurs positively in  $\phi$ , we may as well set it to true and forget about it (removing any clauses in which it occurs); if only negatively, set it to false, and forget about it (removing it **from** any clause in which it occurs, leaving just the other disjunct). So we're left with variables that occur both positively and negatively. Clearly, if we now have a clause (Y) and a clause  $(\neg Y)$ , we can't satisfy  $\phi$ . So suppose we still have two-literal disjuncts, so, for example,  $(\alpha \vee Y)$  and  $(\beta \vee \neg Y)$ . If these are jointly satisfiable, then we can also satisfy  $\alpha \vee \beta$ , so resolve by adding  $(\alpha \vee \beta)$  to  $\phi$  (and conversely, if they're not jointly satisfiable, adding the clause does no harm). Repeat until there are no such pairs remaining unresolved. If at any point we end up with a clause (Y) and a clause  $(\neg Y)$  for some Y, we can't satisfy; otherwise we can. This takes polynomial time (about  $n^4$ ).

An alternative proof via graph theory: Every clause can be written as an implication  $(\alpha \Rightarrow \beta)$  where  $\alpha, \beta$  are literals. Consider a graph where the vertices are literals and directed edges describe these implications. The formula is satisfiable if we never have that  $\alpha \Rightarrow \neg \alpha$  for any literal  $\alpha$ . This is the case if and only if, in every strongly connected component of the graph a literal does not appear both positively and negatively.