CS 221: Computational Complexity

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Problem Set 1

Assigned: Sun. Feb. 9, 2014 Due: Fri. Feb. 21, 2014 (5 PM sharp)

- You must *type* your solutions. LaTeX, Microsoft Word, and plain ascii are all acceptable. Submit your solutions *via email* to cs221-hw@seas.harvard.edu. If you use LaTeX, please submit both the compiled file (.pdf) and the source (.tex). Please name your files PS1-yourlastname.*.
- Strive for clarity and conciseness in your solutions, emphasizing the main ideas over low-level details. Do not despair if you cannot solve all the problems! Difficult problems are included to stimulate your thinking and for your enjoyment, not to overwork you. *'ed problems are extra credit.

Problem 1. (A Universal NTM) Show that there is a universal nondeterministic Turing machine whose running time when simulating a nondeterministic TM N (encoded by a string α) on input x, is at most $c_{\alpha} \cdot \text{Time}_{N}(x)$ for some constant c_{α} depending only on the encoding α . (Hint: use the "guess and verify" approach to designing efficient nondeterministic algorithms.)

Problem 2. (An Average-Case Time Hierarchy) Let $f, g : \mathbb{N} \to \mathbb{N}$ be such that $f(n) \log f(n) = o(g(n))$ and g is time-constructible. Show that there is a language $L \in \mathbf{DTIME}(g(n))$ with the property that for every TM M running in time f(n), there is a constant $\varepsilon_M > 0$ such that for all sufficiently large n, M errs in deciding L on at least an ε_M fraction of inputs of length n.

Problem 3. (A Tighter Time Hierarchy Theorem) Prove that for every constant $\varepsilon > 0$, $\mathbf{DTIME}(n \log^{\varepsilon} n) \not\subseteq \mathbf{DTIME}(n)$. (Hint: use translation. first try to handle the case that $\varepsilon > 1/2$.)

Problem 4. (LINEAR PROGRAMMING) A linear program consists of a collection of variables x_1, \ldots, x_n , a linear objective function $\sum_i c_i x_i$ (specified by the vector $\vec{c} \in \mathbb{Q}^n$), and a collection of constraints each of which is a linear inequality $\sum_i a_i x_i \leq b$ (specified by $\vec{a} \in \mathbb{Q}^n$ and $b \in \mathbb{Q}$). To solve a linear program is to find a vector $\vec{x} \in \mathbb{Q}^n$ maximizing the objective function subject to the given constraints. In vector notation, we maximize $\vec{c} \cdot \vec{x}$ subject to $A\vec{x} \leq \vec{b}$, where A is the matrix whose rows are the constraint vectors \vec{a} and the inequality is componentwise.

The decisional version of this problem is $LP = \{(\vec{c}, A, \vec{b}, K) : \exists \vec{x} \in \mathbb{Q}^n \text{ s.t. } A\vec{x} \leq \vec{b}, \vec{c} \cdot \vec{x} \geq K\}.$ The ellipsoid and interior point algorithms show that $LP \in \mathbf{P}$; you may use this below.

1. Prove that LP is **P**-complete under logspace mapping reductions. (Remark: Integer Programming, the variant of Linear Programming where all numbers in the problem are integers and we solve for integer solutions, is actually **NP**-complete.)

2. Show that a language L has polynomial-sized circuits if and only if there is a sequence of linear programs $P_n = (A_n, \vec{b}_n)$ (with no objective function) with poly(n) variables and poly(n) constraints and entries from $\{-1,0,1\}$ such that for every input $w \in \{0,1\}^n$, $w \in L$ if and only if P_n has a feasible solution \vec{x} whose first n coordinates equal w. Thus another approach to proving $\mathbf{P} \neq \mathbf{NP}$ is to prove a superpolynomial lower bound on the size of linear programs whose feasible solutions project to some \mathbf{NP} language.

Problem 5. (FACTORING) The FACTORING problem is: given a number n, find its prime factorization. There is no polynomial-time algorithm known for this problem; indeed, much of public-key cryptography relies on its presumed hardness. In this problem, you will explore the complexity of FACTORING. Throughout, you may use the fact that deciding primality is in \mathbf{P} .

- 1. Show that FACTORING can be cast as an **NP** search problem (in the sense of Problem 3 on Problem Set 0), and hence can be solved in polynomial time if P = NP.
- 2. Show that if Factoring is NP-hard under Cook reductions, then NP = coNP.
- 3. (*) Give an explicit algorithm for Factoring such that the running time of this algorithm is polynomial if and only if Factoring can be solved in polynomial time. (Hint: think diagonalization.) Compare the asymptotic running time of your algorithm to the running time of the fastest possible algorithm for Factoring. Is your algorithm practical?