## COMP 360 - Fall 2015 - Assignment 4

Due: 6:00 pm Nov 24th.

General rules: In solving these questions you may collaborate with other students but each student has to write his/her own solution. There are in total 110 points, but your grade will be considered out of 100. You should drop your solutions in the assignment drop-off box located in the Trottier Building.

- 1. (10 Points) Show that the following problem is in PSPACE:
  - Input: A CNF  $\phi$ .
  - Output: The number of the truth assignments that satisfy  $\phi$ .
- 2. (10 Points) Given a set P of n points on the plane, consider the problem of finding the smallest circle containing all the points in P. Show that the following is a 2-factor approximation algorithm for this problem. Pick a point x in P, and set r to be the distance of the farthest point in P from x. Output the circle centered at x with radius r.
- 3. (10 Points) Consider the following optimization version of the Subset-Sum problem: Given positive integers  $\{w_1, \ldots, w_n\}$  and a positive integer m. We want to find a set  $S \subseteq \{1, \ldots, n\}$  such that  $\sum_{i \in S} w_i \leq m$  and is maximized. Show that the following is a  $\frac{1}{2}$ -factor approximation algorithm:
  - Set  $S := \emptyset$ .
  - Sort the numbers such that  $w_1 \geq w_2 \geq \ldots \geq w_n$ .
  - For i = 1, ..., n:
  - if it is possible add i to S without violating  $\sum_{i \in S} w_i \leq m$ , then add i to S.
- 4. Consider the MAX-SAT problem: Given a CNF formula  $\phi$  on variables  $x_1, \ldots, x_n$ , find a truth assignment to the variables that maximizes the total number of satisfied clauses.
  - (a) (10 Points) Show that the following is a  $\frac{1}{2}$ -factor approximation algorithm for MAX-SAT (meaning that: the output of the algorithm is always at least half of the optimum): Let  $\sigma_{\text{true}}$  be the truth assignment that assigns True to every variable, and  $\sigma_{\text{false}}$  be the truth assignment that assigns False to every variable. Compute the number of clauses satisfied by  $\sigma_{\text{true}}$  and  $\sigma_{\text{false}}$ , and output the better assignment.
  - (b) (5 Points) Give a tight example: An input instance where this algorithm performs as bad as the  $\frac{1}{2}$  factor.
- 5. (10 Points) Problem 10 of Chapter 11: Suppose you are given an  $n \times n$  grid graph G. Associated with each node v is an integer weight  $w(v) \geq 0$ . You may assume that all the weights are distinct. Your goal is to choose an independent set S of nodes of the grid, so that the sum of the weights of the nodes in S is as large as possible. (The sum of the weights of the nodes in S will be called its total weight.) Consider the following greedy algorithm for this problem.

- Start with  $S := \emptyset$ .
- While some node remains in G:
  - Pick a node v of maximum weight.
  - Add v to S.
  - Delete v and its neighbors from G
- Endwhile.

Show that this algorithm returns an independent set of total weight at least  $\frac{1}{4}$  times the maximum total weight of any independent set in the grid graph G.

- 6. Consider a directed bipartite graph G = (V, E). We want to eliminate all the directed cycles of length 4 by removing a smallest possible set of vertices.
  - (a) (5 points) Let  $C_4$  denote the set of all cycles of length 4 in the graph. Show that the following integer program models the problem:

$$\min \sum_{v \in V} x_v 
\text{s.t.} \quad \sum_{u \in C} x_u \ge 1 \qquad \forall C \in C_4 
x_u \in \{0, 1\} \qquad u \in V$$

(b) (5 points) Why does the optimal solution to the following relaxation provides a lower bound for the optimal answer to the above integer linear program? In other words why it is not necessary to have the constraints  $x_u \leq 1$  in the relaxation?

- (c) (10 points) Give a simple 4-factor approximation algorithm for the problem based on rounding the solution to the above linear program.
- (d) (15 points) Let L and R denote the set of the vertices in the two parts of the bipartite graph. (Every edge has one endpoint in L and one endpoint in R). Let  $x^*$  denote an optimal solution to the linear program in Part (b). We round  $x^*$  in the following way: For every  $u \in V$ ,
  - if  $u \in R$  and  $x_u^* \ge 1/2$ , set  $\widehat{x}_u = 1$ .
  - if  $u \in L$  and  $x_u^* > 0$ , set  $\hat{x}_u = 1$ .
  - Otherwise set  $\hat{x}_u = 0$ .

Show that  $\hat{x}$  is a feasible solution to the integer linear program.

(e) (10 points) Consider the dual of the relaxation:

$$\begin{array}{ll} \max & \sum_{C \in \mathcal{C}_4} \ y_C \\ \text{s.t.} & \sum_{C \in \mathcal{C}_4, u \in C} \ y_C \leq 1 \\ & y_C \geq 0 \end{array} \qquad \forall u \in V$$

and let  $y^*$  be an optimal solution to the dual. Use the complementary slackness to prove the following statement: For every  $C \in \mathcal{C}_4$  either we have  $|\{u : \widehat{x}_u = 1\}| \leq 3$  or  $y_C^* = 0$ .

(f) (10 points) Use the complementary slackness and the previous parts to show that our rounding algorithm is a 3-factor approximation algorithm.

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