CS 599: Rounding Techniques in Approximation Algorithms

Homework 2: Dependent Randomized Rounding

Fall 2024

1 Problem 1: P=NP...?

1. In the traveling salesperson problem, we are given a complete graph G = (V, E) with edge costs $c : E \to \mathbb{R}_{\geq 0}$ that form a metric, i.e. $c_{\{u,w\}} \leq c_{\{u,v\}} + c_{\{v,w\}}$ for all u,v,w. Our goal is to find the minimum cost Hamiltonian cycle. Now where n = |V| and E(S) for $S \subseteq V$ is the set of edges with both endpoints in S, let

$$P_{\text{sub}} = \begin{cases} \sum_{e \in E} x_e = n \\ \sum_{e \in E(S)} x_e \le |S| - 1 & \forall S \subsetneq V \\ x_e \ge 0 & \forall e \in E \end{cases}$$

Prove that $P_{\text{sub}} \cap \{0,1\}^E$ is the set of all feasible solutions to the traveling salesperson problem, i.e. all Hamiltonian cycles in G.

- 2. Notice that P_{sub} is exactly the spanning tree polytope P_{st} except we have changed the n-1 to an n. So, it seems like we should be able to apply the proof from Lecture 4 to show that it has integral vertices. Either use this to prove P=NP or find a flaw in the argument from Lecture 4 when applied to P_{sub} instead of P_{st} (i.e. when we change the n-1 to an n).
- 3. In Lecture 3, we mentioned that it is NP-Hard to obtain a 1.001 approximation for weighted k-ECSS for any k. Either explain where the proof fails and give an explicit counterexample showing that the $1 + O(\sqrt{\frac{\log n}{k}})$ approximation does not work for weighted k-ECSS, or adapt the algorithm to prove that P=NP.

2 Problem 2: Scaling into Integral

Suppose P is a polytope in $[0,1]_{\geq 0}^n$ and \tilde{P} is the convex hull of $P \cap \mathbb{Z}^n$. Now, suppose that there is an $\alpha \geq 1$ such that given any point $x \in P$ there exists a randomized algorithm A that produces a random point $\tilde{x} \in \tilde{P}$ such that $\mathbb{E}\left[\tilde{x}_i\right] \leq \alpha x_i$ for every input, where the expectation is taken over the possible outputs \tilde{x} of A given x.

Given a polytope P, P^{\uparrow} is called the *dominant* of P and consists of all points x such that there exists $x' \in P$ for which $x' - x \in \mathbb{R}^n_{>0}$.

- 1. Prove that $\alpha \cdot P \subseteq \tilde{P}^{\uparrow}$ (where $\alpha \cdot P$ consists of all points in x scaled entry-wise by α).
- 2. Prove that if P is defined by the constraints $0 \le x_i \le 1$ and a collection of *covering constraints*, i.e. constraints of the form $a^Tx \ge b$ for $a \in \mathbb{R}^n_{\ge 0}$, $b \in \mathbb{R}_{\ge 0}$, then the integrality gap of the LP $\min c^Tx$ subject to $x \in P$ is at most α for any $c \in \mathbb{R}^n_{>0}$.

3 Problem 3: Randomized Pipage Rounding for Matroids

A *matroid* $M = (E, \mathcal{I})$ is defined by a collection of elements E and a collection of *independent sets* $\mathcal{I} \subseteq 2^E$ with $\emptyset \in \mathcal{I}$ and the properties:

- (i) **Downward Closed**: If $I \in \mathcal{I}$, then $J \in \mathcal{I}$ for every $J \subseteq I$.
- (ii) **Augmentation Property**: If $I, J \in \mathcal{I}$ and |I| < |J| then there exists some $e \in J$ such that $I \cup \{e\} \in \mathcal{I}$.

A *basis* of a matroid is any maximal independent set. The *rank* of a collection of elements $F \subseteq E$ is the maximum possible size of $I \cap F$ for any $I \in \mathcal{I}$. Let $r : 2^E \to \mathbb{Z}_{>0}$ be the rank function.

In this problem, we will consider the following polytope P_M for a matroid M:

$$P_{M} = \begin{cases} x(E) = r(E) \\ x(S) \le r(S) & \forall S \subseteq E \\ x_{e} \ge 0 & \forall e \in E \end{cases}$$

- 1. Argue that the set of forests of a graph forms a matroid M, and that in this case $P_M = P_{\rm st}$.
- 2. Adapt the proof in class for the spanning tree polytope to show that for every matroid M, P_M as defined above is the convex hull of its (integral) bases, i.e. it is the base polytope of M. You may use that the rank function r is submodular, i.e. $r(S) + r(T) \ge r(S \cup T) + r(S \cap T)$ for $S, T \subseteq E$. Then briefly argue that randomized pipage rounding works for any matroid with the same guarantees as for spanning trees.
- 3. Use the method of conditional expectation to derandomize randomized pipage rounding so that given $x \in P_M$ and any cost function $c : E \to \mathbb{R}^E$, we can find a point of cost at most c(x) in polynomial time.
- 4. When using given an independent distribution $\mu: 2^E \to \mathbb{R}^E_{\geq 0}$ with marginals x, we have used that $\mathbb{P}_{S \sim \mu}[|S \cap F| = 0] \leq e^{-x(F)}$ for any $F \subseteq E$. Show that this also holds for distributions with 0-negative correlation and thus for randomized pipage rounding.

Show that the same bound does *not* hold for all distributions μ with only negative correlation, i.e. $\mathbb{E}\left[\prod_{i \in S} X_i\right] \leq \prod_{i \in S} \mathbb{E}\left[X_i\right]$ for all $S \subseteq E$, by exhibiting a negatively correlated distribution for which this does not hold.

4 Problem 4: Lottery

Use the above to show that we can design a multi-item lottery as follows. Suppose we have a collection of n goods g_1, \ldots, g_n , a collection of m people w_1, \ldots, w_m , and a specified $0 < \epsilon < 1$. Now, we will allow people to buy up to one lottery ticket. Before they purchase a ticket, they will specify the subset of the goods that they would be happy winning, and if we sell them a ticket we must promise them that their chance of getting one of their chosen goods is at least ϵ .

Using that P_M has a polynomial time separation oracle for every M, implement a polynomial time lottery system that (i) can determine when a ticket can be faithfully sold (this will depend on the subset of goods desired), (ii) will output a random assignment that respects the guarantee, and (iii) is *fair to all groups* in the sense that for any group of k people, the probability at least one of them wins something is at least $1 - e^{-k\varepsilon}$.

5 Bonus Problems

- 1. Prove the parsimonious property from Lecture 6 using splitting off.
- 2. Show that the randomized rounding algorithm for the multi-commodity flow problem can be derandomized using the method of conditional expectation.
- 3. Give an example showing that the Chernoff bound upper tail does not hold for distributions with only *pairwise* negative correlation, i.e. $\mathbb{P}\left[e, f \in S\right] \leq \mathbb{P}\left[e \in S\right] \mathbb{P}\left[f \in S\right]$ (and not full negative correlation a.k.a the NCP as we proved for pipage rounding).
- 4. (***) Prove that swap rounding gives a $1 + O(1/\sqrt{k})$ for k-ECSM. (Note: This is quite difficult. I think I vaguely see how to prove this but I'm not 100% sure. This could potentially even be false, although I don't think so. This could be an interesting final project.)