



Problem Set 4

Linear Programming Relaxations and Rounding

Approximation Algorithms (MA5517)

Dr. Jannik Matuschke | M.Sc. Marcus Kaiser

This problem set will be discussed in the tutorials on November 20th/21st, 2018.

Problem 4.1 (Weighted Sum of Completion Times with Precedence Constraints)

We consider a scheduling problem on a single machine. There are n jobs with processing times $p_j \geq 0$ and non-negative weights $w_j \geq 0$ for all $j \in [n]$. In addition, we are given a partial order \preceq on the jobs. For a schedule, which is an ordering of the jobs, the completion time $C_j \geq 0$ of job $j \in [n]$ is the point in time when j is completed. The goal is to find a total ordering of the jobs that extends \preceq and minimizes the weighted sum of completion time $\sum_{j \in [n]} w_j C_j$.

- i) Smith's rule states to process the jobs in non-increasing order w.r.t. the ratio of their weight to their processing time. Prove that Smith's rule gives an optimal schedule for the problem without precedence constraints.
- ii) Show that the following linear program is a relaxation of the problem with precedence constraints.

$$\begin{aligned} \min \quad & \sum_{j \in [n]} w_j C_j \\ \text{s.t.} \quad & \sum_{j \in J} p_j C_j \geq \frac{1}{2} \sum_{j \in J} p_j^2 + \frac{1}{2} \left(\sum_{j \in J} p_j \right)^2 && \text{for all } J \subseteq [n] \\ & C_j - C_i \geq p_j && \text{for all } i, j \in [n] : i \prec j \\ & C \in \mathbb{R}_{\geq 0}^n \end{aligned}$$

Hint: Use (i).

- iii) Let $C^* \in \mathbb{R}_{\geq 0}^n$ be an optimal solution to the LP-relaxation. Consider the schedule defined by ordering the jobs in non-decreasing order w.r.t. C_j^* . Prove that this schedule is a 2-approximation.
- iv)* Deduce a 2-approximation algorithm for minimizing the weighted sum of completion times for scheduling on a single machine with precedence constraints. You may use that submodular^a functions can be minimized in polynomial time.

Problem 4.2 (Pipage Rounding)

In the MAXIMUMCUT problem, we are given as input an undirected graph $G = (V, E)$ with non-negative weights $w_e \geq 0$ for all $e \in E$. The goal is to find a set of vertices W so as to maximize the weight of the edges that leave W . We consider a variant of the problem in which the size of W has to agree with an integer $k \leq |V|/2$ that is part of the input.

- i) Show that the variant of MAXIMUMCUT is modeled by the following integer program with non-linear objective function based on $f : [0, 1]^2 \rightarrow \mathbb{R}$, $(y, z) \mapsto y + z - 2yz$.

$$\begin{aligned} \max \quad & \sum_{\{u,v\} \in E} w_{\{u,v\}} f(x_u, x_v) \\ \text{s.t.} \quad & \mathbf{1}^\top x = k \\ & x \in \{0, 1\}^V \end{aligned}$$

^aFor a finite set E , a function $f : 2^E \rightarrow \mathbb{R}$ is called submodular if $f(A \cap B) + f(A \cup B) \leq f(A) + f(B)$ for all $A, B \subseteq E$.

- ii) Show that the following program with objective function based on $g : [0, 1]^2 \rightarrow \mathbb{R}$, $(y, z) \mapsto \min(y + z, 2 - y - z)$ represents a relaxation of the above. Further, prove that $f(y, z) \geq \frac{1}{2}g(y, z)$ holds for any $y, z \in [0, 1]$.

$$\begin{aligned} \max \quad & \sum_{\{u,v\} \in E} w_{\{u,v\}} g(x_u, x_v) \\ \text{s.t.} \quad & \mathbf{1}^\top x = k \\ & x \in [0, 1]^V \end{aligned}$$

- iii) Let x be a solution to the relaxation. Argue that, for two fractional elements in x , it is possible to increase one by $\varepsilon \neq 0$ and decrease the other by ε such that $\sum_{\{u,v\} \in E} w_{\{u,v\}} f(x_u, x_v)$ does not decrease and one of the two variables becomes integer.
- iv) Use the arguments above to devise a $1/2$ -approximation algorithm for the given variant of MAXIMUMCUT problem.

Problem 4.3 (Maximum Directed Cut Problem)

In the MAXIMUMDIRECTEDCUT problem, we are given as input a directed graph $G = (V, A)$. Each arc $a \in A$ has non-negative weight $w_a \geq 0$. The goal is to partition V into two sets W and $V \setminus W$ so as to maximize the total weight of the arcs going from W to $V \setminus W$.

- i) Show that the following integer program models the maximum directed cut problem and linearize it, i.e., find an equivalent (mixed-)integer linear program.

$$\begin{aligned} \max \quad & \sum_{(u,v) \in A} w_{(u,v)} \min(x_u, 1 - x_v) \\ \text{s.t.} \quad & x \in \{0, 1\}^V \end{aligned}$$

- ii) Consider a randomized rounding algorithm for MAXIMUMDIRECTEDCUT that solves a linear programming relaxation of the mixed-integer program and puts vertex $v \in V$ in W with probability $1/4 + x_v/2$. Show that this gives a randomized $1/2$ -approximation algorithm for MAXIMUMDIRECTEDCUT.