

Exercises  
**Approximation Algorithms**  
Spring 2010  
Sheet 9

**Reminder: On May 5, lecture and tutorial are moved to AAC 132.**

**Exercise 1**

Recall that for the  $k$ -TSP problem, we are given a complete graph  $G = (V, E)$  with a metric cost function  $c : E \rightarrow \mathbb{Q}_+$  and a parameter  $k \in \{1, \dots, n\}$ . The goal is to find a minimum length tour, visiting *at least*  $k$  nodes.

- i) Show that if  $c$  is a tree metric (and you know the underlying tree  $T$ ), then one can find an optimum tour in polynomial time.
- ii) Give an expected  $O(\log n)$ -approximation algorithm for  $k$ -TSP in general metric graphs. Can you derandomize it?

**Solution:**

- i) We show the claim by dynamic programming. The dynamic program will be easier to state for a slightly more general problem, where the set of nodes  $V$  is partitioned into *required vertices*  $R$  and *Steiner nodes*  $S$  and we have to visit at least  $k$  many different required nodes at least once (by short-cutting one can again obtain a tour that does not revisit a node and does not visit more than  $k$  nodes). Then it would not make sense to use an edge  $(u, v)$  that is not contained in the tree — instead we can buy all edges on the  $u$ - $v$  path in  $T$  for the same price. In other words, it suffices to use only tree edges. We guess a node  $r \in V$  that is visited by the optimum tour and consider  $r$  as the root of the tree.

The degrees in  $T$  might be arbitrary. To make our life easier, by inserting new Steiner nodes and cost 0 edges, we turn  $T$  into a tree with out-degree  $\leq 2$ . Furthermore, by adding Steiner nodes, we may assume that required nodes have exactly  $\{0, 1\}$  children and Steiner nodes have degree 2 children.

For any node  $v \in V$ , and  $k' \in \{0, \dots, k\}$  we define table entries

$$A(v, k') = \text{cheapest tour in the subtree below } v, \text{ starting and ending at } v \text{ that visits at least } k' \text{ required nodes}$$

If  $v \in R$  and  $v$  has one child  $v_1$ , then

$$A(v, k) = A(v_1, k - 1) + 2 \cdot d^T(v, v_1)$$

If  $v \in S$  and has 2 children  $v_1, v_2$ , we use:

$$A(v, k) = \min \left\{ A(v_1, k) + 2 \cdot d^T(v, v_1), A(v_2, k) + 2 \cdot d^T(v, v_2), \min_{k=k_1+k_2} \left\{ A(v_1, k_1) + 2d^T(v, v_1) + A(v_2, k_2) + 2d^T(v, v_2) \right\} \right\}$$

ii) We use the theorem from the last slide of the tree embedding section to obtain trees  $T_1, \dots, T_q$  with cost  $d_i$ , weight  $\lambda_i$ . Suppose we choose a tree  $T$  from  $T_1, \dots, T_q$  (i.e.  $\Pr[T = T_i] = \lambda_i$ ), and let  $d^T$  be the induced tree metric. Then for any  $u, v \in V$

- $c(u, v) \leq d^T(u, v)$
- $E[d^T(u, v)] \leq O(\log n) \cdot c(u, v)$

Let  $OPT^{T_i}$  be the cost of the optimum  $k$ -TSP solution w.r.t. metric  $d^T$ . We claim that  $E[OPT^T] \leq O(\log n) \cdot OPT$ . This can be easily seen as follows: Let  $E^*$  be the edges of the optimum tour in  $G$ . Then the same set is still a valid tour w.r.t.  $d^T$ . And

$$E[OPT^T] \leq E \left[ \sum_{(u,v) \in E^*} d^T(u, v) \right] \stackrel{\text{linearity of expectation}}{=} \sum_{(u,v) \in E^*} E[d^T(u, v)] \leq O(\log n) \cdot c(u, v)$$

By i), we can compute a tour using edges  $E'$  of expected cost

$$E \left[ \sum_{(u,v) \in E'} d^T(u, v) \right] \leq E[OPT^T] \leq O(\log n) \cdot OPT.$$

Then at least one of the trees  $T_i$  (with induced tree metric  $d^{T_i}$ ) must have cost

$$\sum_{(u,v) \in E'} d^{T_i}(u, v) \leq E[OPT^T] \leq O(\log n) \cdot OPT$$

Since  $d^{T_i}(u, v) \geq c(u, v)$ , the tour  $E'$  is also not more expensive w.r.t. the original costs.

## Exercise 2

For STEINER FOREST, the input is a complete, undirected graph  $G = (V, E)$  with metric cost function  $c : E \rightarrow \mathbb{Q}_+$  and pairs  $(s_1, t_1), \dots, (s_k, t_k)$  ( $s_i, t_i \in V$ ). The goal is to find a min cost subgraph  $H$ , that connects each  $s_i$ - $t_i$  pair:

$$OPT = \min_{H \subseteq E} \left\{ \sum_{e \in H} c(e) \mid \forall i = 1, \dots, k : H \text{ connects } s_i \text{ and } t_i \right\}$$

(there is no need to connect  $s_i, t_j$  for  $i \neq j$ , hence  $H$  itself does not need to be connected. In fact, in general it will be a *forest*, that is a collection of trees). Consider the following linear program

$$\begin{aligned} \min \quad & \sum_{e \in E} x_e c_e \\ & \sum_{e \in \delta(S)} x_e \geq 1 \quad \forall i = 1, \dots, k \quad \forall S \subseteq V : s_i \in S, t_i \notin S \\ & x_e \geq 0 \end{aligned}$$

Here  $x_e$  can be interpreted as a variable that indicates whether  $e$  is included in  $H$  or not. Prove that the integrality gap of this LP is upperbounded by  $O(\log n)$ .

**Solution:**

Let  $x_e^*$  be an optimum fractional solution of cost  $OPT_f = \sum_{e \in E} x_e^* c_e$ . Let  $T$  be a random tree and  $d^T$  be the induced tree embedding with  $O(\log n)$ -distortion (which exists using a theorem from the lecture). Then the cost of  $x^*$  in the dominating metric is

$$E\left[\sum_{e \in E} d^T(e) \cdot x_e^*\right] = \sum_{e \in E} x_e^* E[d^T(e)] \leq \sum_{e \in E} x_e^* O(\log n) \cdot c(e) \leq O(\log n) \cdot OPT_f$$

Next, for any edge  $e = (u, v) \notin T$ , we install  $x_e^*$  units on each edge on the  $u$ - $v$  path in tree without that the cost increases. We obtain a new fractional solution  $y_e^*$  where  $y_e^* = 0$  if  $e \notin T$  (in fact this is a feasible solution, since any cut  $S$  with  $u \in S, v \notin S$  contains also at least one from any  $u$ - $v$  path). Consider now any edge  $e \in T$ . Removing  $e$  from  $T$  gives 2 subtrees  $T_1, T_2$  (i.e.  $T = T_1 \cup T_2 \cup \{e\}$ ). If one has  $s_i \in T_1$  and  $t_i \in T_2$  for some  $i$  (or vice versa), then the cut constraint for  $S := V(T_1)$  says that  $y_e^* \geq 1$ . Let us define

$$z_e^* := \begin{cases} 1 & \text{if } e \text{ separates an } s_i - t_i \text{ pair in } T \\ 0 & \text{otherwise} \end{cases},$$

which is a feasible and integral LP solution with  $z_e^* \leq y_e^*$ . Furthermore

$$E\left[\sum_{e \in E} d^T(e) \cdot z_e^*\right] \leq \sum_{e \in E} E[d^T(e)] y_e^* = \sum_{e \in E} E[d^T(e)] \cdot x_e^* = O(\log n) \cdot OPT_f$$

Of course,  $H := \{e \in E \mid z_e^* = 1\}$  is a feasible solution that is not more expensive w.r.t. the original cost  $c$  than  $O(\log n) \cdot OPT_f$ . By the principle of the probabilistic method, there must be one tree embedding and hence one concrete solution  $H$ , that really costs at most  $O(\log n) \cdot OPT_f$ . Hence the integrality gap is bounded by  $O(\log n)$ .

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