

问题与反馈

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Exercise 4.2.1.4. Change the greedy strategy of the algorithm GMS to an arbitrary choice, i.e., without sorting p_1, p_2, \dots, p_n (removing Step 1 of GMS), assign the jobs in the on-line manner² as described in Step 2 and 3 of GMS. Prove that this simple algorithm, called GRAHAM'S ALGORITHM, is a 2-approximation algorithm for MS, too. \square

Exercise 4.2.1.5. Find, for every integer $m \geq 2$, an input instance I_m of MS such that $R_{\text{GMS}}(I) = \frac{\text{cost}(\text{GMS}(I))}{\text{Opt}_{\text{MS}}(I)}$ is as large as possible. \square

Exercise 4.2.3.3. Prove that the functions $dist$, $dist_k$, and $distance$ (defined in Example 4.2.3.2) are distance functions for TSP according to L_Δ . \square

$$\text{dist}(G, c) = \max \left\{ 0, \max \left\{ \frac{c(\{u, v\})}{c(\{u, p\}) + c(\{p, v\})} - 1 \mid u, v, p \in V(G), \right. \right. \\ \left. \left. u \neq v, u \neq p, v \neq p \right\} \right\},$$

$$\text{dist}_k(G, c) = \max \left\{ 0, \max \left\{ \frac{c(\{u, v\})}{\sum_{i=1}^m c(\{p_i, p_{i+1}\})} - 1 \mid u, v \in V(G) \text{ and} \right. \right.$$

$u = p_1, p_2, \dots, p_m = v$ is a simple path between u and v

of length at most k (i.e., $m + 1 \leq k$) $\left. \right\} \left. \right\}$

for every integer $k \geq 2$, and

$$\text{distance}(G, c) = \max\{\text{dist}_k(G, c) \mid 2 \leq k \leq |V(G)| - 1\}.$$

Definition 4.2.3.1. Let $U = (\Sigma_I, \Sigma_O, L, L_I, \mathcal{M}, \text{cost}, \text{goal})$ and $\bar{U} = (\Sigma_I, \Sigma_O, L, L, \mathcal{M}, \text{cost}, \text{goal})$ be two optimization problems with $L_I \subset L$. A **distance function for \bar{U} according to L_I** is any function $h_L : L \rightarrow \mathbb{R}^{\geq 0}$ satisfying the properties

- (i) $h_L(x) = 0$ for every $x \in L_I$, and
- (ii) h is polynomial-time computable.

Let h be a distance function for \bar{U} according to L_I . We define, for any $r \in \mathbb{R}^+$,

$$\mathbf{Ball}_{r,h}(L_I) = \{w \in L \mid h(w) \leq r\}.$$
⁶

Let A be a consistent algorithm for \bar{U} , and let A be an ε -approximation algorithm for U for some $\varepsilon \in \mathbb{R}^{>1}$. Let p be a positive real. We say that A is **p -stable according to h** if, for every real $0 < r \leq p$, there exists a $\delta_{r,\varepsilon} \in \mathbb{R}^{>1}$ such that A is a $\delta_{r,\varepsilon}$ -approximation algorithm for $U_r = (\Sigma_I, \Sigma_O, L, \mathbf{Ball}_{r,h}(L_I), \mathcal{M}, \text{cost}, \text{goal})$.

A is **stable according to h** if A is p -stable according to h for every $p \in \mathbb{R}^+$. We say that A is **unstable according to h** if A is not p -stable for any $p \in \mathbb{R}^+$.

For every positive integer r , and every function $f_r : \mathbb{N} \rightarrow \mathbb{R}^{>1}$ we say that A is **$(r, f_r(n))$ -quasistable according to h** if A is an $f_r(n)$ -approximation algorithm for $U_r = (\Sigma_I, \Sigma_O, L, \mathbf{Ball}_{r,h}(L_I), \mathcal{M}, \text{cost}, \text{goal})$.

Exercise 4.2.3.4. Let $U = (\Sigma_I, \Sigma_O, L, L_I, \mathcal{M}, cost, goal)$ and $\bar{U} = (\Sigma_I, \Sigma_O, L, L, \mathcal{M}, cost, goal) \in \text{NPO}$. Let h_{index} be defined as follows:

- (i) $h_{index}(w) = 0$ for every $w \in L_I$, and
- (ii) $h_{index}(u)$ is equal to the order of u according to the canonical order of words in Σ_I^* .

Prove that

- a) h_{index} is a distance function of \bar{U} according to L_I .
- b) For every δ -approximation algorithm A for U , if A is consistent for \bar{U} , then A is stable according to h_{index} . □

*An algorithm A is **consistent** for U if, for every $x \in L_I$, the output $A(x) \in \mathcal{M}(x)$. We say that an algorithm B **solves** the optimization problem U if*

Exercise 4.2.3.3 shows that it is not interesting to consider a distance function h with the property

$$|Ball_{r,h}(L_I)| - |Ball_{q,h}(L_I)| \text{ is finite}$$

for every $r > q$. The distance function h' investigated later has the following additional property (called the **property of infinite jumps**):

“If $Ball_{q,h'}(L_I) \subset Ball_{r,h'}(L_I)$ for some $q < r$, then $|Ball_{r,h'}(L_I)| - |Ball_{q,h'}(L_I)|$ is infinite.”

Exercise 4.2.3.5. Define two optimization problems $U = (\Sigma_I, \Sigma_O, L, L_I, \mathcal{M}, cost, goal)$ and $\bar{U} = (\Sigma_I, \Sigma_O, L, L, \mathcal{M}, cost, goal)$ from NPO with infinite $|L| - |L_I|$, and a distance function h for \bar{U} according to L_I such that:

- (i) h has the property of infinite jumps, and
- (ii) for every δ -approximation algorithm A for U , if A is consistent for \bar{U} , then A is stable according to h . □

Lemma 4.3.5.9. CHRISTOFIDES ALGORITHM *is stable according to distance.*