# Theory of practical algorithms: exercises

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### 1 Skeleton dimension

A graph is  $\gamma$ -doubling if any ball of radius r is covered by  $\gamma$  balls of radius r/2. More precisely, for any node u and radius r > 0, there exist nodes  $u_1, \ldots, u_k$ with  $k \leq \gamma$  such that B(u, r) is included in  $\cup_{i=1}^k B(u_i, r/2)$ . Show that a graph with skeleton dimension k is 2k + 1-doubling.

### 2 Hopsets

Given a connected weighted undirected graph G, we define the *h*-hop distance  $d_G^h(u, v)$  between u and v as the minimum weight of a path from u to v with h edges at most. (Each edge of a path is called a "hop", it thus corresponds to the distance using at most h hops.) The usual distance from u to v is thus  $d_G(u, v) = d_G^{n-1}(u, v)$ . We define a *h*-hopset of G as a set H of edges such that  $d_{G\cup H}^h(u, v) = d_G(u, v)$  where each edge uv of H is considered to have weight  $d_G(u, v)$ .

a) What notion seen in course is tightly related to the notion of 2-hopset?

b) Suppose that G is a path of length n, propose a 2-hopset of G with as few edges as you can (we do not care about multiplicative constants).

c) Same question for a 3-hopset.

d) Same question for a 4-hopset.

## 3 Hierarchical hub labeling from contraction hierarchies

Given an unweighted undirected graph G and a contraction order  $\pi = u_1 \prec \cdots \prec u_n$  of the nodes of G, we consider the contraction hierarchies  $E^+ = E_n^+$  of directed edges where  $E_0^+ := \{vw \in E(G) \mid v \prec w\}$  and  $E_i^+ := E_{i-1}^+ \cup \{vw \mid u_i \prec v \prec w \text{ and } u_iv \in E_{i-1}^+$  and  $u_iw \in E_{i-1}^+\}$ . We define the hub-set  $H_u$  of u as the set of nodes visited by a BFS from u in the directed graph  $G\uparrow = (V(G), E^+)$ .

a) Give the main argument for proving that  $(H_u)_{u \in V(G)}$  has the following covering property: for all pair  $u, v \in V(G)$  and for any shortest path P between u and v, we have  $H_u \cap H_v \cap P \neq \emptyset$ .

b) Recall how the distance between u and v can be computed from  $H_u$  and  $H_v$  if appropriate information is associated to elements of  $H_u$  and  $H_v$ .

c) Give the main argument for proving that  $(H_u)_{u \in V(G)}$  is hierarchical, i.e. the relation "is hub of is transitive: for all triple  $u, v, w \in V(G)$ , if  $v \in H_u$  and  $w \in H_v$ , then  $w \in H_u$ .

#### Solution :

a) The node  $u_i \in P$  with *i* maximal is visited by both BFS.

b) Associate d(u, a) to  $a \in H_u$ ; d(u, v) can be obtained as  $\min_{a \in H_u \cap H_v} d(u, a) + d(v, a)$  according to a) and triangle inequality.

c) If v is visited by the BFS from u, then all edges considered in a BFS from v will be (or where already) considered in the BFS from u, implying  $H_v \subseteq H_u$ .

## 4 Connection Scan Algorithm Revisited

We consider a bus network given by an array C of connections. More precisely, each connection  $c \in C$  represents the elementary travel of a bus departing from a stop c.from at time c.dep and arriving at the next stop c.to at time c.arr > c.dep. C is sorted by increasing departure time. For simplicity, we assume that all departure times are distinct  $(c.dep \neq c'.dep$  for  $c \neq c')$ . A sequence  $c_1, \ldots, c_k$  of connections is a feasible journey if  $c_i.to = c_{i+1}.from$  and  $c_i.arr \leq c_{i+1}.dep$  for all  $i = 1 \ldots k - 1$  (no walk is considered). When a traveler starts at a given stop src, we store at every stop u an estimation  $\tau(u)$  of arrival time at u. (Initially,  $\tau(src)$  stores the starting time at src and  $\tau(u) = \infty$  for  $u \neq src$ .)

a) **Earliest arrival:** Propose a linear time algorithm for computing the earliest arrival time at a given stop dst.

b) **Last departure:** Propose a linear time algorithm for computing the last departure time from *src* such that *dst* can be reached at a given time  $\tau_{arr}$ .

c) **Profile:** Propose an algorithm for computing all interesting departure times from *src* to *dst* where a departure time  $\tau_{dep}$  is interesting if there is a journey reaching *dst* at some time  $\tau_{arr}$  such that no other journey with departure time  $\tau > \tau_{dep}$  arrives at  $\tau_{arr}$  or before.

We now consider a connected symmetric footpath graph G whose vertex set V contains all stops. We let  $d_G(u, v) = d_G(v, u)$  denote the walking time from u to v. A traveler can now walk from a stop to any other (unrestricted walking). We still assume that a traveler arriving at stop u at time  $\tau$  can catch any connection c such that c.from = u and  $\tau \leq c.dep$ . For representing these walking transfers, we assume that a hub labeling is given: each vertex  $u \in V$  is associated to a set  $H(u) \subseteq V$  of hubs. Each hub  $x \in H(u)$  is associated with its walking time  $d_G(u, x)$  from u. The hub sets have the following covering property: for any pair u, v of vertices there exists a common hub  $x \in H(u) \cap H(v)$  lying on a shortest waling path from u to v:  $d_G(u, v) = d_G(u, x) + d_G(x, v)$ . We let  $\Delta = \max_{u \in V} |H(u)|$  denote the maximum size of a hub set.

a') **Unrestricted walking:** Propose a modification of the algorithm proposed in a) so that journeys can include walking transfer from a stop to any other. How does the complexity of your algorithm increase with  $\Delta$ ?

### Solution :

a) Use Connection Scan without transfers: for each connection  $c \in C$  (in increasing order of departure time), if  $\tau(c.from) \leq c.dep$  then set  $\tau(c.to) := \min(\tau(c.to), c.arr)$ . This requires constant time per connection, yielding linear complexity. All the connections of the best trip to dst must be visited in order and  $\tau(dst)$  is the correct arrival time after all these connections have been scanned.

b) Same algorithm but reversed: Initialize  $\tau(dst) = \tau_{arr}$  and  $\tau(u) = -\infty$  for  $u \neq dst$ . For each connection  $c \in C$  in reverse order of departure time, if  $c.arr \leq \tau(c.to)$  then set  $\tau(c.dep) := \max(\tau(c.dep), c.dep)$ .

c) We proceed similarly as in b) but storing interesting departures at each stop with associated arrival times. For each connection  $c \in C$  in reverse order of departure time, let  $\tau_{dep}$  be the first interesting departure time after *c.arr* at *c.to* and let  $\tau_{arr}$  be the associated arrival time at *dst* (if there is one). Store *c.dep* at *c.from* with associated arrival time  $\tau_{arr}$  if it is interesting, i.e. no pair  $\tau'_{dep}, \tau'_{arr}$  stored at *c.from* satisfies  $\tau'_{dep} > c.dep$  and  $\tau'_{arr} \leq \tau_{arr}$ . Dichotomic search can be used if pairs are stored by decreasing departure. The complexity is then within a logarithmic factor from linear.

a') For  $x \in H(src)$ ,  $\tau(x) := \min(\tau(x), \tau(src) + d_G(src, x))$ . For each connection  $c \in C$  in increasing order of departure:

- for each  $x \in H(c.from)$ , set  $\tau(c.from) := \min(\tau(c.from), \tau(x) + d_G(x, c.from))$ ,
- if  $\tau(c.from) \leq c.dep$  then set  $\tau(c.to) := \min(\tau(c.to), c.arr),$
- for each  $x \in H(c.to)$ , set  $\tau(x) := \min(\tau(x), \tau(c.to) + d_G(c.to, x))$ .

For  $x \in H(dst)$ ,  $\tau(dst) := \min(\tau(dst), \tau(x) + d_G(x, dst))$ . The covering property ensures correctness. The complexity is linear in  $\Delta$ .