

CSCE-629 Analysis of Algorithms

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Solutions to Assignment #2

(Prepared with TA Qin Huang)

1. Design algorithms for $\text{Min}(H)$, $\text{Insert}(H, a)$, and $\text{Delete}(H, i)$, where the set H is stored in a heap, a is the element to be inserted into the heap H , and i is the index of the element in the heap H to be deleted. Analyze the complexity of your algorithms.

Solutions.

A heap H is a complete binary tree with possibly some of its rightmost leaves missing. Thus, a heap of n nodes has a height bounded by $\log n$. Moreover, because of its special structure, a heap H of n elements can be represented by an array $A[1 \dots \text{max}]$ and an index n so that $A[1 \dots n]$ holds the n elements of H . If we place the elements in the heap level by level, starting from the root, and following the order from left to right, into the array $A[1 \dots n]$, then it is not difficult to verify that for an element $A[i]$ in the heap, $A[2i]$ and $A[2i + 1]$ are the two children of $A[i]$, and $A[\lfloor i/2 \rfloor]$ is the father of $A[i]$.

A min-heap $A[1 \dots n]$ is a heap that satisfies the condition that the value of each element in the heap is not larger than the values of its children, i.e., $A[i] \leq A[2i]$ and $A[i] \leq A[2i + 1]$ for all $1 \leq i \leq \lfloor n/2 \rfloor$. It's easy to verify that for all i , the value $A[i]$ is not larger than that of any of its descendants. In particular, the root $A[1]$ has the minimum value in the heap.

Algorithm 1 $\text{Min}(A[1 \dots \text{max}], n)$

1: return $(A[1])$;

Algorithm 2 $\text{Insert}(A[1 \dots \text{max}], n; a)$

1: $n = n + 1$; $A[n] = a$; $i = n$;
2: **while** $A[\lfloor i/2 \rfloor] > A[i]$ and $i \geq 2$ **do**
3: swap $A[\lfloor i/2 \rfloor]$ with $A[i]$;
4: $i = \lfloor i/2 \rfloor$;
5: **end while**

$\text{Min}(A, n)$ takes time $O(1)$, $\text{Insert}(A, n; a)$ and $\text{Delete}(A, n; i)$ takes time $O(\log n)$, i.e., the height of the tree. \square

Algorithm 3 Delete($A[1 \dots \text{max}], n; i$)

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1:  $A[i] = A[n]; n = n - 1;$ 
2: if  $i \geq 2$  and  $A[i] < A[\lfloor i/2 \rfloor]$  then
3:   while  $i \geq 2$  and  $A[\lfloor i/2 \rfloor] > A[i]$  do
4:     swap  $A[i]$  with  $A[\lfloor i/2 \rfloor]$ ; /*pushing up*/
5:      $i = \lfloor i/2 \rfloor;$ 
6:   end while
7: else
8:   while  $(2i \leq n)$  and  $(A[i] > A[2i] \text{ or } A[i] > A[2i + 1])$  do
9:     if  $(2i = n)$  or  $(A[2i] < A[2i + 1])$  then
10:      swap  $A[i]$  and  $A[2i]$ ;  $i = 2i$ ;
11:    else
12:      swap  $A[i]$  and  $A[2i + 1]$ ;  $i = 2i + 1$ ;
13:    end if
14:  end while
15: end if
```

Remark. In the following questions, you can assume that your graphs are connected.

2. Write the psuedo-code for the Dijkstra's algorithm that solves the SINGLE-SOURCE SHORTEST PATH problem. Analyze the complexity of the algorithm (you can assume that the algorithm uses a heap for fringes and you can use your results in Question 1 directly). Give a formal proof that the algorithm works correctly when the edge weights are all non-negative.

Solutions.

Let $G = (V, E)$ be a weighted, directed graph, with weight function $w : E \rightarrow R$. Suppose the source is s .

Time complexity: let $n = |V|, m = |E|$. First to note that $\text{Insert}(F, v)$, $\text{Min}(F)$ and $\text{Delete}(F, v)$ takes time $O(\log n)$, where F can be implemented as a 2-3 tree. Lines 1-3 takes time $O(n)$; lines 4-8 takes time $O(\deg(s) \log n = O(m \log n)$, where $\deg(s)$ is the out-degree of s . The while loops $n - 1$ times, each time line 10 takes $O(\log n)$ time, and so the total time line 10 takes is $O(n \log n)$. During the whole while loops, lines 11 - 19 takes time $O(m \log n)$ because each directed edge is visited at most once. In total, this algorithm takes time $O((m + n) \log n)$.

Correctness proof: let $\delta(s, u)$ be the distance of the shortest path from s to u . It suffices to show that for each vertex $u \in V$, we have $\text{dist}[u] = \delta(s, u)$ at the time when the status of u becomes *in-tree* or when u is removed from F . Initially, $\text{dist}[s] = 0$ and thus $\text{dist}[s] = \delta(s, s)$.

For the purpose of contradiction, let u be the first vertex for which $\text{dist}[u] > \delta(s, u)$. Let P be a shortest path from s to u . Prior to removing u from F , P must consist of at least one vertex other than u whose status is *fringe* or *unseen*; otherwise if all the vertices on P except u have the status *in-tree*, $\text{dist}[u]$ must be updated as the minimum distance, i.e., $\text{dist}[u] = \delta(s, u)$, contradiction. Let y be the first vertex along P such that $\text{status}[y] \neq \text{in-tree}$. Let x be the vertex that appears before y along the path P . Then, $\text{status}[x] = \text{in-tree}$ and $\text{dist}[x] = \delta(s, x)$. It follows that $\text{status}[y] = \text{fringe}$ and y is in F . Thus, $\text{dist}[y] \leq \text{dist}[x] + w(x, y)$. As x and y are on the shortest path P and all edges are non-negative, we have $\text{dist}[x] + w(x, y) \leq \delta(s, u)$. Since it's u that is chosen to be removed from F not y , then $\text{dist}[u] \leq \text{dist}[y]$. As a result, $\text{dist}[u] \leq \text{dist}[x] + w(x, y) \leq \delta(s, u)$, contradiction. Therefore, $\text{dist}[u] = \delta(s, u)$ for all the vertices. \square

Algorithm 4 Dijkstra's algorithm($G = (V, E), s, w$)

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1: for each  $v \in V$  do
2:    $status[v] = \text{unseen}$ ;
3: end for
4:  $status[s] = \text{in-tree}$ ;  $dist[s] = 0$ ;  $F = \emptyset$ ;
5: for each edge  $(s, v) \in E$  do
6:    $status[v] = \text{fringe}$ ;  $dad[v] = s$ ;
7:    $dist[v] = w(s, v)$ ;  $\text{Insert}(F, v)$ ;
8: end for
9: while  $F \neq \emptyset$  do
10:   $u = \text{Min}(F)$ ;  $status[u] = \text{in-tree}$ ;  $\text{Delete}(F, u)$ ;
11:  for each edge  $(u, v) \in E$  do
12:    if  $status[v] = \text{unseen}$  then
13:       $status[v] = \text{fringe}$ ;  $dad[v] = u$ ;  $dist[v] = dist[u] + w(u, v)$ ;
14:       $\text{Insert}(F, v)$ ;
15:    else if  $status[v] = \text{fringe}$  and  $dist[u] + w(u, v) < dist[v]$  then
16:       $\text{Delete}(F, v)$ ;  $dad[v] = u$ ;
17:       $dist[v] = dist[u] + w(u, v)$ ;  $\text{Insert}(F, v)$ ;
18:    end if
19:  end for
20: end while
21: return  $dad[], dist[]$ ;
```

3. Develop a linear-time (i.e., $O(m)$ -time) algorithm that solves the SINGLE-SOURCE SHORTEST PATH problem for graphs whose edge weights are positive integers bounded by 10. (**Hint.** You can either modify Dijkstra's algorithm or consider using Breath-First-Search.)

Solutions.

For each edge (u, v) in the graph, let $w(u, v)$ be its weight. We construct a new graph as follows: in the given graph, for each edge (u, v) , insert $w(u, v) - 1$ vertices into (u, v) , say $x_1, x_2, \dots, x_{w(u, v)-1}$, such that (u, v) becomes a path $(u, x_1, x_2, \dots, x_{w(u, v)-1}, v)$. The number of vertices of the new graph is at most $n + 9m$, and the number of edges is at most $10m$.

Use Breath-First-Search to solve the Single-Source Shortest Path problem. The time required is $O(m + n)$. \square

4. Consider an extended version of the SHORTEST-PATH problem. Suppose that you want to traverse from city s to city t . In addition, for some reason, you also need to pass through cities x , y , and z (in any order) during your trip. Your objective is to minimize the cost of the trip. The problem can be formulated as a graph problem as follows: Given a positively weighted directed graph G and five vertices s, t, x, y, z , find a path from s to t that contains the vertices x, y, z such that the path is the shortest over all paths from s to t that contain x, y, z , assuming that these paths are allowed to contain repeated vertices and edges. Develop an $O(m \log n)$ -time algorithm for this problem.

Solutions.

The problem requires that the path starts at s , ends at t , and passes through all three vertices x, y , and z . However, we do not know the order in which the path passes through the vertices

x , y , and z . Thus, we consider all possible paths that use a possible order of the vertices x , y , and z . Note that the number of possible orders is bounded by 6. Therefore, the algorithm can go as follows:

1. run Dijkstra's algorithm starting from s , we can obtain the $d(s, x), d(s, y), d(s, z), d(s, t)$;
2. run Dijkstra's algorithm starting from x , we obtain $d(x, y), d(x, z), d(x, t)$;
3. run Dijkstra's algorithm starting from y , we compute $d(y, z), d(y, x), d(y, t)$;
4. run Dijkstra's algorithm starting from z , we compute $d(z, x), d(z, y), d(z, t)$;
5. compute the length of the six possible paths, i.e., (s, x, y, z, t) , (s, x, z, y, t) , (s, y, x, z, t) , (s, y, z, x, t) , (s, z, x, y, t) , and (s, z, y, x, t) , and select the one with the minimum distance. For example, given (s, x, y, z, t) , the distance is $d(s, x) + d(x, y) + d(y, z) + d(z, t)$.

Since the above algorithm runs Dijkstra's algorithm four times, and Dijkstra's algorithm runs in time $O(m \log n)$, we conclude that the proposed algorithm solves the given problem and runs in time $O(m \log n)$. \square