Algorithms for Data Processing Lecture IV: Graph Algorithms – Shortest Path and Dijkstra's Algorithm

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Algorithms for Data Processing

Shortest Path

- While BFS is able to find the shortest *s*-*t* path when edges have uniform cost, we need different algorithms when edges have a distinct traversal cost.
- We consider here the following scenario:
 - A directed graph G = (V, E) with a designated start node *s*.
 - We assume that *s* has a path to every other node in *G*.
 - Each edge *e* has an associated *traversal cost* $\ell_e \ge 0$.
 - For a path *P*, the *length* of *P*-denoted $\ell(P)$ -is the sum of the lengths of all edges $e \in P$.
- In 1959, Edsger Dijkstra proposed a very simple greedy algorithm to solve the shortest-paths problem.

Edsger Dijkstra

"What's the shortest way to travel from Rotterdam to Groningen? It is the algorithm for the shortest path, which I designed in about 20 minutes. One morning I was shopping in Amsterdam with my young fiancée, and tired, we sat down on the café terrace to drink a cup of coffee and I was just thinking about whether I could do this, and I then designed the algorithm for the shortest path." — Edsger Dijsktra





A. Artale

Algorithms for Data Processing

Single-pair shortest path problem

Problem. Given a digraph G = (V, E), edge lengths $\ell_e \ge 0$, source $s \in V$, and destination $t \in V$, find a shortest directed path from s to t.



Algorithms for Data Processing

Single-source shortest path problem

Problem. Given a digraph G = (V, E), edge lengths $\ell_e \ge 0$, source $s \in V$, find a shortest directed path from s to every node.



Dijkstra's algorithm (for single-source shortest paths problem)

Greedy approach. Maintain a set of explored nodes *S* for which algorithm has determined $d[u] = \text{length of a shortest } s \rightarrow u \text{ path.}$

- Initialize $S \leftarrow \{s\}, d[s] \leftarrow 0$.
- Repeatedly choose unexplored node $v \notin S$ which minimizes

$$f(v) = \min_{e = (u,v): u \in S} d[u] + \ell_e$$
 the length of a shortest path from s to some node u in explored part S,

followed by a single edge e = (u, v)



Dijkstra's algorithm (for single-source shortest paths problem)

Greedy approach. Maintain a set of explored nodes *S* for which algorithm has determined d[u] = length of a shortest *s*→*u* path.

- Initialize $S \leftarrow \{s\}, d[s] \leftarrow 0$.
- Repeatedly choose unexplored node $v \notin S$ which minimizes

$$\pi(v) = \min_{e = (u,v): u \in S} d[u] + \ell_e$$

add *v* to *S*, and set $d[v] \leftarrow \pi(v)$.

the length of a shortest path from *s* to some node *u* in explored part *S*, followed by a single edge e = (u, v)

• To recover path, set *pred*[*v*] ← *e* that achieves min.



Shortest Path: An Example from the book Algorithm Design



Figure 4.7 A snapshot of the execution of Dijkstra's Algorithm. The next node that will be added to the set *S* is x, due to the path through u.

Dijkstra's algorithm: proof of correctness

Invariant. For each node $u \in S$: d[u] = length of a shortest $s \rightarrow u$ path.

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Pf. [ by induction on |S| ]
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Base case: |S| = 1 is easy since S = \{s\} and d[s] = 0.
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Inductive hypothesis: Assume true for $|S| \ge 1$.

- Let v be next node added to S, and let (u, v) be the final edge.
- A shortest $s \rightarrow u$ path plus (u, v) is an $s \rightarrow v$ path of length $\pi(v)$.
- Consider any other $s \rightarrow v$ path *P*. We show that it is no shorter than $\pi(v)$.
- Let e = (x, y) be the first edge in *P* that leaves *S*, and let *P'* be the subpath from *s* to *x*.
- The length of *P* is already ≥ π(v) as soon as it reaches y:

$$p' \rightarrow p$$

 $s \rightarrow p$
 $s \rightarrow q$
 $s \rightarrow q$
 $s \rightarrow q$

$$\ell(P) \ge \ell(P') + \ell_e \ge d[x] + \ell_e \ge \pi(y) \ge \pi(v) \bullet$$

$$\uparrow \qquad \uparrow \qquad \uparrow \qquad \uparrow$$
non-negative inductive definition Dijkstra chose v
lengths hypothesis of $\pi(y)$ instead of y

Dijkstra Algorithm

We assume that G = (V, E, L), where L contains the cost for each edge.

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Dijkstra(G,s)
Let S be the set of explored nodes, and;
\pi an array of n elements such that \pi[u] is the shortest s-u path;
S = \{s\}; \pi[s] = 0; \text{ pred}[s] = null;
while S \neq V do
     min-dist = \infty:
    for each v \in V \setminus S with (u, v) \in E and u \in S do
         for each (u, v) \in E with u \in S do
        if (\pi[u] + \ell(u, v) < \min\text{-dist}) then

 \lim_{v \to \infty} \min\text{-dist} = \pi[u] + \ell(u, v); 
 \max_{v \to \infty} \max[node] = u 
    S = S \cup \{node\}; \pi[node] = min-dist.
```

Dijkstra Algorithm – Complexity

- Each While iteration adds a new node *v* to the set *S*.
- Then, there are n 1 iterations of the While-loop for a graph with n nodes.
- Each iteration considers each node $v \notin S$, and goes through all the edges between S and v to determine the minimum distance vertex. This takes O(m).
- Thus, the Algorithm runs in $O(m \cdot n)$.

Critical optimization 1. For each unexplored node $v \notin S$: explicitly maintain $\pi[v]$ insted of recomputing them in each iteration



$$\pi(v) = \min_{e = (u,v) : u \in S} d[u] + \ell_e$$

- For each $v \notin S$: $\pi(v)$ can only decrease (because set *S* increases).
- More specifically, suppose *u* is added to *S* and there is an edge e = (u, v) leaving *u*. Then, it suffices to update:

Critical optimization 2. Use a priority queue (PQ) to choose an unexplored node that minimizes $\pi[v]$.

Brief Introduction to Priority Queues

- Priority Queue (PQ). Data Structure where elements have a priority value, or key, and we access just the element with *highest priority*.
- Data Structure. Balanced Binary Tree, where the root contains the element with highest priority.
 - Heap Order. The key of any element is at least as large as the key of its parent node.
- Cost of managing a PQ.
 - Extraction. *O*(1), since we can access only the root element, which has the highest priority.
 - ► Addition and Deletion: $O(\log n)$.

From the Book: Chapter 2



Figure 2.3 Values in a heap shown as a binary tree on the left, and represented as an array on the right. The arrows show the children for the top three nodes in the tree.



Figure 2.4 The Heapify-up process. Key 3 (at position 16) is too small (on the left). After swapping keys 3 and 11, the heap violation moves one step closer to the root of the tree (on the right).

Dijkstra's algorithm: efficient implementation

Implementation.

- Algorithm maintains $\pi[v]$ for each node v.
- Priority queue stores unexplored nodes, using $\pi[\cdot]$ as priorities.
- Once *u* is deleted from the PQ, $\pi[u] = \text{length of a shortest } s \rightarrow u \text{ path.}$

```
DIJKSTRA (V, E, \ell, s)
pred[s] \leftarrow null; \pi[s] \leftarrow 0
FOREACH v \neq s: \pi[v] \leftarrow \infty, pred[v] \leftarrow null;
Create an empty priority queue pq.
FOREACH v \in V: INSERT(pq, v, \pi[v]).
WHILE (IS-NOT-EMPTY(pq))
   u \leftarrow \text{DEL-MIN}(pq).
   FOREACH edge e = (u, v) \in E leaving u:
      IF (\pi[v] > \pi[u] + \ell_e)
          \pi[v] \leftarrow \pi[u] + l_{a}:
          DECREASE-KEY(pq, v, \pi[v]).
          pred[v] \leftarrow u.
```

Dijkstra's algorithm: which priority queue?

Performance. Depends on PQ: *n* INSERT, *n* DELETE-MIN, $\leq m$ DECREASE-KEY.

- Array implementation optimal for dense graphs. $\longleftarrow \Theta(n^2)$ edges
- Binary heap much faster for sparse graphs. ← ⊖(n) edges
- 4-way heap worth the trouble in performance-critical situations.

priority queue	INSERT	Delete-Min	Decrease-Key	total
unordered array	<i>O</i> (1)	O(n)	<i>O</i> (1)	$O(n^2)$
binary heap	$O(\log n)$	$O(\log n)$	$O(\log n)$	$O(m \log n)$
d-way heap (Johnson 1975)	$O(d \log_d n)$	$O(d \log_d n)$	$O(\log_d n)$	$O(m \log_{m/n} n)$
Fibonacci heap (Fredman-Tarjan 1984)	<i>O</i> (1)	$O(\log n)^{\dagger}$	$O(1)^{\dagger}$	$O(m + n \log n)$
integer priority queue (Thorup 2004)	<i>O</i> (1)	$O(\log \log n)$	<i>O</i> (1)	$O(m + n \log \log n)$

Thank You!