Shortest Paths in a Graph

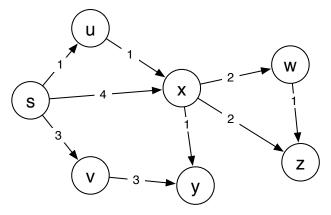
Slides by Carl Kingsford

Feb. 5, 2014

Based on/Reading: Chapter 4.5 of Kleinberg & Tardos

Shortest Paths in a Weighted, Directed Graph

Given a directed graph G with lengths $\ell_e > 0$ on each edge e:



Goal: Find the shortest path from a given node *s* to every other node in the graph.

Shortest Paths

Shortest Paths. Given directed graph G with n nodes, and non-negative lengths on each edge, find the n shortest paths from a given node s to each v_i .

- ▶ Dijkstra's algorithm (1959) solves this problem.
- ▶ If we have an undirected graph, we can replace each undirected edge by 2 directed edges:



▶ If all the edge lengths are = 1, how can we solve this?

Shortest Paths

Shortest Paths. Given directed graph G with n nodes, and non-negative lengths on each edge, find the n shortest paths from a given node s to each v_i .

- Dijkstra's algorithm (1959) solves this problem.
- ▶ If we have an undirected graph, we can replace each undirected edge by 2 directed edges:

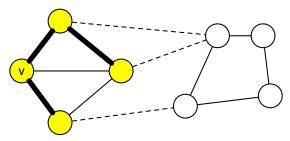


▶ If all the edge lengths are = 1, how can we solve this? BFS

General Tree Growing

Dijkstra's algorithm is just a special case of tree growing:

- ▶ Let T be the current tree T, and
- ▶ Maintain a list of frontier edges: the set of edges of *G* that have one endpoint in *T* and one endpoint not in *T*:



▶ Repeatedly choose a frontier edge (somehow) and add it to T.

Tree Growing

- ► The function nextEdge(G, S) returns a frontier edge from S.
- ▶ updateFrontier(G, S, e) returns the new frontier after we add edge e to T.

nextEdge for Shortest Path

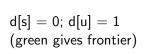
► Let *u* be some node that we've already visited (it will be in *S*).

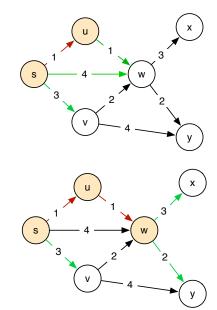
▶ Let d(u) be the length of the s-u path found for node $u \in S$.

nextEdge: return the frontier edge (u, v) for which d(u) + length(u, v) is minimized.

▶ The "d(u)" term is the difference from Prim's algorithm.

Example





d[w] = 2

Proof of Correctness

Theorem. Let T be the set of nodes explored at some point during the algorithm. For each $u \in T$, the path to u found by Dijkstra's algorithm is the shortest.

Proof. By induction on the size of T. Base case: When |T| = 1, the only node in T is s, for which we've obviously found the shortest path.

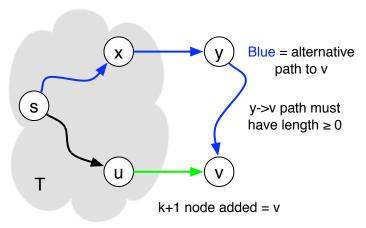
Induction Hypothesis: Assume theorem is true when $|T| \le k$.

Let v be the $(k+1)^{st}$ node added using edge (u, v).

Let P_v be the path chosen by Dijkstra's to v and let P be any other path from s to v.

Then we have the situation on the next slide.

Proof, cont.



The path to v chosen by Dijkstra's is of length \leq the alternative blue path.

Shortest Paths \implies Tree

Theorem. There is some optimal set of shortest paths from source s such that their union forms a tree.

Proof. Dijkstra's algorithm is correct and produces a tree.

Implementation of Dijkstra

```
1: for u \in V do dist[u] \leftarrow \infty
 2: H \leftarrow MakeHeap()
 3: u \leftarrow s \# (s \text{ is an arbitrary start vertex})
 4: while u \neq null do
        for v \in NEIGHBORS(u) do
 5:
             # If the distance is smaller than before, we have to update
 6:
             if dist[u] + d(u,v) < dist[v] then
 7:
                 dist[v] \leftarrow dist[u] + d(u,v)
 8:
 9.
                 if v \notin H then
                     INSERT(H, v, dist[v])
10:
                 else
11:
                     REDUCEKEY(H, v, dist[v]) # Sift up for new key
12:
                 parent[v] \leftarrow u
13:
        u \leftarrow DELETEMIN(H)
14:
15: return parent
```

Running time of Dijkstra's Algorithm

Same as Prim's MST algorithm:

- Every edge is processed in the for loop at most once.
- In response to that processing, we may either
 - 1. do nothing; O(1),
 - 2. insert a item into the heap of at most |V| items; $O(\log |V|)$, or
 - 3. reduce the key of an item in a heap of at most |V| items; $O(\log |V|)$
- ▶ Total time is therefore: $O(|E| \log |V|)$.