## Shortest Paths in a Graph

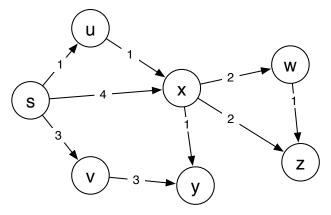
Slides by Carl Kingsford

Feb. 8, 2013

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:

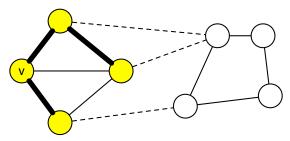


▶ If all the edge lengths are = 1, how can we solve this? BFS

## 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.

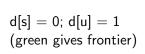
## nextEdge for Shortest Path

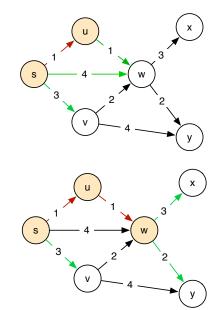
Let *u* be some node that we've already visited (it will be in *S*).

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

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

# Example





d[w] = 2

## **Analysis**

**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.

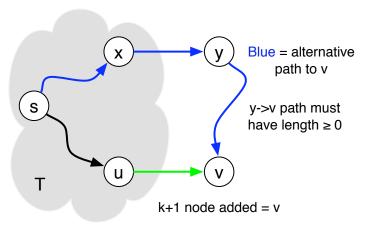
**Induction Hypothesis**: Assume theorem is true when  $|T| \le k$ .

Let v be the k+1 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[v] + 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

### 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:  $O(|E| \log |V|)$ .