Shortest path algorithms nal Linguistics III Data Structures and Algorithms for Com-(ISCL-BA-07)

Çağrı Çöltekin ccoltekin@sfs.uni-tuebingen.de

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Weighted graphs

- $\ast\,$ A $weighted\,graph$ is a graph, where each edge is associated with a weight . Weights can be any numeric value, but some algorithms require
- Non-negative weights
 Euclidean' weights: weights that are proper distance metrics
- Weights often indicate distance or cost, but they can also represent positive relations (e.g., affinity between nodes)
 - Weight of a path is the sum of wights of the edges on the path

Shortest path

- · Finding shortest paths on a weighted (directed) graph is one of the most common problems in many fields

 • Applications include

- Navigation
 Navigation
 Navigation
 Routing in computer networks
 Optimal construction of electronic circuits, VLSI chips
 Robotics, transportation, finance, ...

Shortest paths on unweighted graphs

- A BPS search tree gives the short path from the source node to all other nodes . The BPS is not enough on weighted
- graphs Shortest-cost path may be longer in



Shortest paths on weighted graphs

- · Different versions of the problem:
- Single source shortest path: find shortest path from a source node to all others
 Single target (sometimes called sink) shortest path: find shortest path from all
 - nodes to a target node
- Source to target: from a particular source node to a particular target node
 All pairs: shortest paths between all pairs of nodes
 Restrictions on weights:

 - Euclidean weights
 Non-negative weights
 Arbitrary weights

Dijkstra's algorithm

- Dijkstra's algorithm is a 'weighted' version of the BPS
 The algorithm finds shortest path from a single source node to all connected
- · Weights has to be non-neg
- It is a greedy algorithm that grows a 'cloud' of nodes for which we know the shortest paths from the source node
- * The new nodes are included in the cloud in order of their shortest paths from the source node

Dijkstra's algorithm

- · We maintain a list D of mini know distances to each node
- · At each step we take closest node out of Q
 update the distances of all no
- Can be more efficient if Q is implemented using a (adaptable) priority queue
- 1: D[s] ← 0 for each node $v \neq s$ do $D[v] \leftarrow \infty$
- 4: Q ← nodes 5: while Q is not empty do Remove node u with min D[u] from Qfor each edge (u, v) do
- if D[u] + w(u, v) < D[v] then $D[v] \leftarrow D[u] + w(u, v)$
- 10: D contains the shortest distances from s

Dijkstra's algorithm







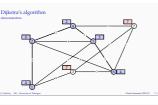






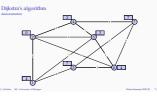
















- In general, complexity is $O(t_{ind_min}n + t_{update_key}m)$ With list-based implementat $Q: O(m+n^2) = O(n^2)$

- With a priority que $O((m+n) \log n)$
- $i \colon D[s] \leftarrow 0$

for each node ν ≠ s do
 D[ν] ← ∞

- 3: $D[v] \leftarrow \infty$ 4: $Q \leftarrow nodes$ 5: while Q is not empty do 6: Remove node u with min D[u] from Q7: for each edge (u,v) do 8: if D[u] + w(u,v) < D[v] then 9: $D[v] \leftarrow D[u] + w(u,v)$
- 10: D contains the shortest distances from s

Shortest-path tree

Dijkstra's algorithm

- The way we introduced, the Dijkstra's algorithm does not give the shortest-path

- $\begin{array}{c} E: T \leftarrow \varnothing \\ 2: \mbox{ for } u \in D \{s\} \mbox{ do } \\ 3: \mbox{ for each edge}(v,u) \mbox{ do } \\ 4: \mbox{ fo}(p) = -D[u] + w(v,u) \mbox{ then } \\ T \leftarrow T \cup \{v,u\} \end{array}$ Similar to traversal algorithms, we can extract it from distances D Running time is O(n²) (or O(n + m))

Shortest-paths on DAGs

- \ast The shortest path can be found more efficiently, if the graph is a DAG
- . The algorithm is similar to Dijkstra's, but simpler and faster
- . Only difference is we follow a topological order
- . The algorithm will also work with negative edge weights

