Aa Lecture 13: Ee HGgHh Li J KWeighted Shortest Paths 1234567890

These slides include material originally prepared by Dr. Ron Cytron, Dr. Jeremy Buhler, and Dr. Steve Cole.

Announcements

- Lab 13 released tomorrow Dijkstra's algorithm
- Studio 13 Thursday
- Exam 3 May 1st 10 am 12 pm

Adding Weights to Graphs

- A weighted graph assigns to each edge e a real-valued weight w(e)
- For today, we will assume that $w(e) \ge 0$.



Path Lengths

• The **length** (or total weight) of a path is the sum of its edges' weights



• **Problem**: given a starting vertex v, *find path of least total weight* from v to each other vertex in the graph.



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Why Solve Weighted Shortest Paths?

- Road map with distances between cities shortest route
- Routing network with cost to each hop cheapest way to send data
- State-space search in AI with action costs (A* search)



How Can We Solve Weighted Shortest Paths?

• Could reduce to unweighted problem:



- Apply to every edge; use BFS on resulting graph
- Only works if weights are integers
- Would be very expensive for graphs with large weights

Alternate Strategy – "Relaxation"

- Explore graph while maintaining, for each vertex v, length of shortest path to v seen *so far*. Store this shortest path estimate as **v.dist**.
- Whenever we follow an edge (v,u), check whether
 v.dist + w(v,u) < u.dist
- If so, we've found a new, shorter path to u via v.





• Update v.dist with new estimate and continue

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Who Gets to Relax, When?

- Need to decide which vertices to relax at each step.
- Also, need to know when we are done!
- **Proposal** (*E. Dijkstra* *): at each step, explore edges out of vertex v with *smallest* v.dist, and relax all its adjacent vertices.
- Stop when each vertex has had its outgoing edges explored once.

Dijkstra's Shortest Path Algorithm

- starting vertex v gets v.dist \leftarrow 0; all other u get u.dist $\leftarrow \infty$
- mark all vertices as unfinished
- while (any vertex unfinished)
- v ← unfinished vertex with smallest v.dist
- for each edge (v,u)
- if (v.dist + w(u,v) < u.dist)
- $u.dist \leftarrow v.dist + w(u,v)$
- mark v finished

// relax!





dist 0 unfinished В As for BFS/DFS, we finished explore a vertex's 4 Start @ A outgoing edges in some arbitrary order. ∞
























0

As with BFS, parent edges from Dijkstra's algo form a shortest-path tree from starting vertex.



- **Claim**: when we explore the edges out of vertex v, v has its correct shortest-path distance D(start, v) stored in current best estimate v.dist.
- **Pf**: by induction on order of exploration.
- **Bas**: starting vertex is explored first, with its correct shortest-path distance of 0.

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- Ind: suppose the algorithm is about to choose v for exploration.
- Assume that v.dist > D(start, v) (i.e. v's distance is wrong).

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- Ind: suppose the algorithm is about to choose v for exploration.
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- [we will derive a contradiction...hence, v.dist must be = D(start, v)]

- Ind: suppose the algorithm is about to choose v for exploration.
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- Consider a *shortest* path from start to v.



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• Let u be last finished (i.e., already explored) vertex on this path.

- By IH, u had its correct shortest-path distance when it was explored.
- Moreover, D(start, u) ≤ D(start, v), since u precedes v on shortest path to v.
- If edge u → v is on shortest path, then exploring u's outgoing edges assigns v its correct shortest-path distance D(start,v). → ←



- If edge u → v is on shortest path, then exploring u's outgoing edges assigns v its correct shortest-path distance D(start,v). → ←
- Otherwise, some other vertex x lies between u and v on this path, with D(start,x) ≤ D(start,v).



Since v does not have its correct shortest-path distance,
v.dist > x.dist, and so x would be explored next, not v. → ← QED 44

How Do We Track Next Vertex to Explore?

- Maintain collection of unfinished vertices
- At each step, must efficiently **find** vertex v in collection with smallest v.dist and **remove** it
- But vertices' distances may change repeatedly due to relaxation!
- Changes are all in one direction (*decrease*)

What Data Structure Can We Use To Track Distances to Unfinished Vertices?

Use a Priority Queue!

- Maintain priority queue PQ of unfinished vertices, keyed on **dist**
- Initially, every vertex is inserted into PQ w/its starting dist
- At each step, find next vertex to explore by PQ.extractMin()
- Decreasing v.dist is done using v's **Decreaser** object
- We assume a map D[] from vertices to their Decreasers

Dijkstra's Shortest Path Algorithm w/Prio Queue

- v.dist \leftarrow 0; D[v] \leftarrow PQ.insert(starting vertex v)
- For all other vertices u
- u.dist $\leftarrow \infty$; D[u] \leftarrow PQ.insert(u)
- while (PQ not empty)
- v ← PQ.extractMin()
- for each edge (v,u)
- if (v.dist + w(v, u) < u.dist)
- u.dist ← v.dist + w(v,u)
- D[u].decrease(u)

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Note that Lab 13 creates pairs (vertex, dist) rather than vertices with an internal distance field.































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- For each edge, we do one PQ decrease()
- Hence, total cost is $|V|(T_{insert} + T_{extractMin}) + |E| T_{decrease}$
- Times for PQ operations using binary heap are all Θ(log |V|)
- Hence, algorithm runs in time $\Theta((|V| + |E|) \log |V|)$



Hence, algorithm runs in time Θ((|V| + |E|) log |V|)










Why Must Edge Weights Be Non-Negative?

• With negative weights, Dijkstra's algorithm does not necessarily find a shortest (smallest total sum of edge weights) path.



D is *finalized* with distance 2.

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But *shortest* path is of length 3 + -2 = 1

Alternatives to Dijkstra

- If negative-weight edges are allowed...
- May use *Bellman-Ford* algorithm (Θ(|V||E|))
- If shortest-path distances are desired between every pair of vertices...
- May use *Floyd-Warshall* algorithm ($\Theta(|V|^3)$)
- (Other approaches may be better for sparse graphs)