

CS 341: ALGORITHMS

Lecture 14: graph algorithms V – single source shortest path

Readings: see website

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DIJKSTRA'S ALGORITHM

Single-source shortest path in a graph with **non-negative** edge weights

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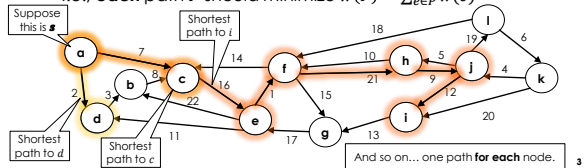
PROBLEM: SINGLE SOURCE SHORTEST PATHS (SSSP)

- Input: graph $G = (V, E)$ and a **non-negative** weight function $w(e)$ defined for every edge e
- Problem: for every node $v \neq s$, output a path $s \rightsquigarrow v$ with the **smallest total weight** (among all paths $s \rightsquigarrow v$)

Let's study **directed** G .
Can also be defined for undirected G ...

"Shortest" means minimum weight

I.e., each path P should minimize $w(P) = \sum_{e \in P} w(e)$

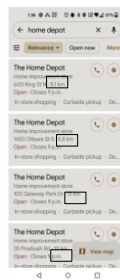


And so on... one path for each node.

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APPLICATION: DRIVING DISTANCE TO MANY POSSIBLE DESTINATIONS

- Single source: from where you are
- Shortest paths: to **all** destinations
- Display a subset of destinations
- Include the optimal distances computed using SSSP algorithm
- Other heuristics... traffic? Lights?
- Weights** can combine many factors



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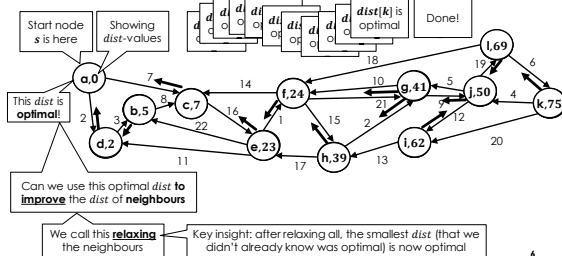
[video clip]

- Game AI: path finding with waypoints
- Divide game world into **linear paths**, then send game characters in **straight lines** between waypoints
- If some linear paths are much faster/slower, use **weighted** SSSP

Otherwise use BFS to find shortest sequence of waypoints (with **fewest** waypoints)

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DIJKSTRA'S ALGORITHM ILLUSTRATIVE EXAMPLE



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```

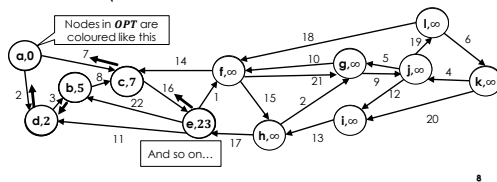
1 Dijkstra(adj[1..n], s)
2 pred[1..n] = [null, null, ..., null]
3 dist[1..n] = [inf, inf, ..., inf]
4 pq = new priority queue
5
6 dist[s] = 0
7 for u = 1..n
8   pq.enqueue(u, dist[u])
9
10 while pq is not empty
11   u = pq.dequeueMin()
12   for v in adj[u]
13     if dist[u] + w(u,v) < dist[v]
14       dist[v] = dist[u] + w(u,v)
15       pred[v] = u
16       pq.changePriority(v, dist[v])
17
18 return pred, dist
    
```

Annotations:

- Maintain nodes in priority order, ordered by smallest distance
- Enqueue all nodes with distance ∞ except for s with distance 0
- Eventually dequeue all nodes (no more enqueues)
- Each dequeued node u has optimal $dist$
- Relax neighbour v

CORRECTNESS: INTUITION

- Dijkstra's algorithm iteratively construct a set **OPT** of nodes for which we **know** the shortest path from s (initially **OPT** = { s })
- After each relaxation step, we grow **OPT** by adding the node in $V \setminus OPT$ with the smallest **dist**

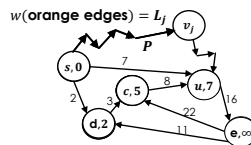


PROOF

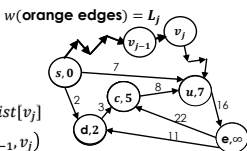
- Theorem:** At the end of the algorithm, for all u , $dist[u]$ is exactly the total weight of the shortest $s \rightsquigarrow u$ path
- We prove this in two parts
 - $dist[u] \leq$ the total weight of the shortest $s \rightsquigarrow u$ path (case \leq)
 - $dist[u] \geq$ the total weight of the shortest $s \rightsquigarrow u$ path (case \geq)

CASE \leq [ERICKSON THM.8.5]

- Let P be **any arbitrary** $s \rightsquigarrow u$ path $v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_\ell$ where $v_0 = s$ and $v_\ell = u$
- For any index j let L_j denote $w(v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_j)$
- We prove by induction: $dist[v_j] \leq L_j$ for all j



- Prove by induction: $\forall j : dist[v_j] \leq L_j$
- Base case: $dist[v_0] = dist[s] = 0 = L_0$
- Ind. step: **suppose** $\forall_{j>0} : dist[v_{j-1}] \leq L_{j-1}$
 - When dequeueMin() returns v_{j-1} : we **check** if $dist[v_{j-1}] + w(v_{j-1}, v_j) < dist[v_j]$
 - If so, we **set** $dist[v_j] = dist[v_{j-1}] + w(v_{j-1}, v_j)$
 - If not, $dist[v_j] \leq dist[v_{j-1}] + w(v_{j-1}, v_j)$
 - In **both cases**, $dist[v_j] \leq dist[v_{j-1}] + w(v_{j-1}, v_j)$
 - By I.H. $dist[v_{j-1}] \leq L_{j-1}$ so $dist[v_j] \leq L_{j-1} + w(v_{j-1}, v_j)$
 - And $L_{j-1} + w(v_{j-1}, v_j) = L_j$ by definition



This proves $dist[u] \leq L_u$, the weight of an arbitrary $s \rightsquigarrow u$ path.
 So $dist[u] \leq$ the weight of EVERY $s \rightsquigarrow u$ path, including the shortest $s \rightsquigarrow u$ path!

CASE \geq

- Let P' be the path $s \rightarrow \dots \rightarrow pred[pred[u]] \rightarrow pred[u] \rightarrow u$ i.e., the reverse of following **pred** pointers from u back to s
- We show $dist[u]$ is as long as **this path** (and hence as long as the **shortest** path)
- Denote the nodes in P' by v_0, v_1, \dots, v_ℓ where $v_0 = s$ and $v_\ell = u$
- Let $L_j = w(v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_j)$
- Prove by induction:** $\forall_{j>0} : dist[v_j] = L_j$
- Base case: $dist[v_0] = dist[s] = 0 = L_0$

CASE \geq

- $P' = v_0 \rightarrow \dots \rightarrow v_k = s \rightarrow \dots \rightarrow pred[pred[u]] \rightarrow pred[u] \rightarrow u$
 - $L_j = w(v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_j)$
 - Inductive step:** suppose $\forall j > 0 : dist[v_{j-1}] = L_{j-1}$
 - When we set $pred[v_j] = v_{j-1}$, we set $dist[v_j] = dist[v_{j-1}] + w(v_{j-1}, v_j)$
- Recall:** if $dist[u] + w(u, v) < dist[v]$
 $dist[v] = dist[u] + w(u, v)$
 $pred[v] = u$
- By I.H., $dist[v_j] = L_{j-1} + w(v_{j-1}, v_j)$
 - By definition $L_j = L_{j-1} + w(v_{j-1}, v_j)$
 - So $dist[v_j] = L_j$
- So $dist[u] \geq$ length of **shortest path** $s \rightarrow u$
 And length of P' is \geq length of **shortest path**
 So $dist[u] \geq$ length of **shortest path** $s \rightarrow u$
 So $dist[u]$ is both \leq and \geq to the length of the shortest $s \rightarrow u$ path!
 That means it's **equal** to the length of the shortest path!

```

1 Dijkstra(adj[1..n], s)
2 pred[1..n] = [null, null, ..., null]
3 dist[1..n] = [inf, inf, ..., inf]
4 pq = new priority queue
5
6 dist[s] = 0
7 for u = 1..n
8   pq.enqueue(u, dist[u])
9
10 while pq is not empty
11   u = pq.dequeueMin()
12   for v in adj[u]
13     if dist[u] + w(u,v) < dist[v]
14       dist[v] = dist[u] + w(u,v)
15       pred[v] = u
16       pq.changePriority(v, dist[v])
17
18 return pred, dist
    
```

RUNTIME

- Each node enqueued and dequeueMin'd once $O(n \log n)$
- For each dequeueMin, do $O(\log n)$ per neighbour $O(\log n)$ for **each edge**
- $O(m \log n)$ w/adjacency lists
- Total time $O((n + m) \log n)$**

Space complexity?

OUTPUTTING ACTUAL SHORTEST PATH(S)?

- To compute the actual shortest **path** $s \rightarrow t$
- Inspect $pred[t]$
 - If it is NULL, there is no such path
 - Otherwise, follow $pred$ pointers back to s , and return the **reverse** of that path

```

1 Dijkstra(adj[1..n], s)
2 pred[1..n] = [null, null, ..., null]
3 dist[1..n] = [inf, inf, ..., inf]
4 OPT = [false, false, ..., false]
5
6 dist[s] = 0
7 OPT[s] = true
8 numOpt = 1
9
10 while numOpt < n
11   choose u such that OPT[u] == false
12   and dist[u] is minimized
13   OPT[u] = true
14   numOpt = numOpt + 1
15   for v = adj[u]
16     if dist[u] + w(u,v) < dist[v]
17       dist[v] = dist[u] + w(u,v)
18       pred[v] = u
19
20 return pred, dist
    
```

AN ALTERNATIVE IMPLEMENTATION

- Instead of using a **priority queue**
- Find the minimum $dist[]$ node to add to OPT via **linear search**
- Runtime?** $O(n^2)$
- Better or worse than $O((n + m) \log n)$?**

WEBSITE DEMONSTRATING DIJKSTRA'S ALG

<https://www.cs.usfca.edu/~galles/visualization/Dijkstra.html>

BELLMAN-FORD

Single-source shortest path in a graph with possibly **negative** edge weights but **no negative cycles**

Shortest Paths and Negative Weight Cycles

Subsequent algorithms we will be studying will solve shortest path problems as long as there are no cycles having negative weight.

If there is a negative weight cycle, then there is no shortest path (why?).

There is still a shortest simple path, but there are apparently no known efficient algorithms to find the shortest simple paths in in graphs containing negative weight cycles.

If there are no negative weight cycles, we can assume WLOG that shortest paths are simple paths (any path can be replaced by a simple path having the same weight).

Negative weight edges in an undirected graph are not allowed, as they would give rise to a negative weight cycle (consisting of two edges) in the associated directed graph.

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BELLMAN-FORD

The *Bellman-Ford algorithm* solves the single source shortest path problem in any directed graph without negative weight cycles.

The algorithm is very simple to describe:

Repeat $n - 1$ times: *relax* every edge in the graph (where *relax* is the updating step in Dijkstra's algorithm).

```

1 BellmanFord(n, E[1..m], s)
2   pred[1..n] = new array filled with null
3   D[1..n] = new array filled with infinity
4   D[s] = 0
5   for i = 1..n
6     for (u,v,w) in E
7       if D[u] + w < D[v]
8         D[v] = D[u] + w
9         pred[v] = u
10  return D, pred
    
```

Annotations:

- $O(n)$ (lines 1-4)
- $O(n)$ (lines 5-6)
- $O(m)$ inner iterations per outer iteration (lines 6-9)
- $O(1)$ work per inner iteration (lines 7-9)
- Total $O(nm)$ (lines 1-10)

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BEST CASE EXECUTION

```

1 BellmanFord(n, E[1..m], s)
2   pred[1..n] = new array filled with null
3   D[1..n] = new array filled with infinity
4   D[s] = 0
5   for i = 1..n
6     for (u,v,w) in E
7       if D[u] + w < D[v]
8         D[v] = D[u] + w
9         pred[v] = u
10  return D, pred
    
```

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WORST CASE EXECUTION

```

1 BellmanFord(n, E[1..m], s)
2   pred[1..n] = new array filled with null
3   D[1..n] = new array filled with infinity
4   D[s] = 0
5   for i = 1..n
6     for (u,v,w) in E
7       if D[u] + w < D[v]
8         D[v] = D[u] + w
9         pred[v] = u
10  return D, pred
    
```

Dijkstra's is similar, but consistently achieves good ordering using its priority queue

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WHY BELLMAN-FORD WORKS

- Not going to prove this (by induction), but the crucial lemma is:
 - After i iterations of the outer for-loop,
 - if $D[u] \neq \infty$, it is equal to the weight of some path $s \rightsquigarrow u$; and
 - if there is a path $P = (s \rightsquigarrow u)$ with at most i edges, then $D[u] \leq w(P)$
 - So, after $n - 1$ iterations, if \exists path P with at most $n - 1$ edges, then $D[u] \leq w(P)$. (Note: any more edges would create a cycle.)
 - So, if u is reachable from s , then $D[u]$ is the length of the shortest simple path (no cycles) from s to u

Of course every simple path has at most $n - 1$ edges

So what if we do another iteration, and some $D[u]$ improves?

There is a negative cycle!

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A MORE DETAILED IMPLEMENTATION

```

1 BellmanFordCheck(n, E[1..m], s)
2   pred[1..n] = new array filled with null
3   D[1..n] = new array filled with infinity
4   D[s] = 0
5   for i = 1..n
6     changed = false
7     for (u,v,w) in E
8       if D[u] + w < D[v]
9         D[v] = D[u] + w
10        pred[v] = u
11        changed = true
12   if not changed
13     exit loop
14   if i == n // assert: changed == true
15     return NEGATIVE_CYCLE
16   return (D, pred)
    
```

- With early stopping
- and checking for negative cycles

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BONUS SLIDE

Why can't you just modify a graph with negative weights by: finding the minimum edge weight W_{min} , and adding that to each edge, so you no longer have negative edges and can run Dijkstra's algorithm?

Exercise: can you find a graph for which this will cause Dijkstra's algorithm to return the **wrong answer**?

Solution:

- Consider a graph with 5 nodes: s, a, b, c, t
- And edges s→a with weight -10, b→t with weight 10
s→b weight -1, b→c weight -1, c→t weight -1
- What happens if you modify this graph as proposed, then run Dijkstra's to find the shortest path from s to t? 25