# Lecture 14: Single-Source Shortest Paths 

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October 31, 2023

## Overview

- Dijkstra's Algorithm
- Single-Source Shortest Paths
- Weighted Shortest Paths as a BFS
- Dijkstra's Algorithm
- Acknowledgements


## Single-Source Shortest Paths

- Input: Weighted directed graph $G(V, E, w)$, where $w: E \rightarrow \mathbb{R}_{>0}$, vertex $s \in V$

Adjacency list.

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- Think of graph as the network of roads in a province
- Edge weights account for how long it takes to drive through that edge (i.e. traffic)


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Succinct representation of the output.

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- Can we modify our graph so that it becomes unweighted? OUI et NON! Let's look at that now...


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- Making above intuition algorithmic:
- Make unweighted graph $H(U, F)$ from $G$ as follows
(1) For each $e:=(u, v) \in E$, create path of length $w(e)$ from $u \rightarrow v$ in $H$ Add new vertices and edges appropriately.
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- Problem: running time of above algorithm will be linear in $H$, but $O(|U|+|F|)=O\left(n+\sum_{e \in E} w(e)\right)$


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(2) While $Q \neq \emptyset$ :
- let $u \in V \backslash R$ be closest vertex to $R, e=(v, u)$ be edge such that $D[v]+w(e)$ minimizes distance $s \rightarrow u$
- extract $u$ from $Q$
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- Correctness: follows from our BFS process
- Runtime: need to find closest vertex \& update distances fast.

How can we do that?
Via priority-queue (min heap).
Using such a priority-queue, runtime is given by $O((n+m) \log n)$.

## Dijkstra - Full Implementation

- Full Algorithm
(1) Initialization:
- $T=\emptyset$, (edges of our shortest-path)
- $R=\{s\}$, (set of "reached vertices")
- $p[u]=N U L L$ for all $u \in V$ (parents)
- $D[u]=\infty$ for all $u \in V \backslash\{s\}, D[s]=0$
(distance to $s$ )
- $Q=V$ priority-queue (min heap $w /$ values given by $D$ )
(2) While $Q \neq \emptyset$ :
- $u=\operatorname{EXTRACT}-\operatorname{MIN}(Q)$
- For $v \in N_{\text {out }}(u)$ :
if $D[u]+w((u, v))<D[v]$, then:
set $D[v]=D[u]+w((u, v))$, $p[v]=u$, DECREASE-KEY $(Q, v)$
- $T \leftarrow T+(p[u], u), R \leftarrow R \cup\{u\}$
(3) return $T$


## Runtime of Dijkstra's Algorithm

- Each vertex is enqueued once and dequeued once
- When vertex is dequeued, check outgoing edges and update distances (if needed)
- All queue operations implemented in $O(\log n)$ time by min-heap
- Total runtime:

$$
O\left(\left(n+\sum_{u \in V} \operatorname{deg}_{\text {out }}(u)\right) \log n\right)=O((n+m) \log n)
$$

## Correctness of Dijkstra's Algorithm

- Similar analysis than the one we did for MST
- Proof of correctness by induction.
- Claim: for any $u \in R, D[u]$ is the shortest path distance from $s \rightarrow u$


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(5) Since $D[u]$ is the distance of some $s \rightarrow u$ path, we know it is at least the shortest path distance.
(0) For the converse, consider any $s \rightarrow u$ path $P$. Since $s \in R$ and $u \notin R$, there is edge $(x, v) \in P$ where $x \in R$ and $v \notin R$.
( $R$ is a cut)

$$
\Rightarrow \ell(P) \geq D[x]+w((x, v))
$$

as $D[x]$ is shortest $s \rightarrow x$ distance, since $x \in R$

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- Thus, we have

$$
\ell(P) \geq D[u]
$$

Since the above holds for any $s \rightarrow u$ path, $D[u]$ is the shortest path distance.

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- Since such edges (by construction) satisfy

$$
D[u]=D[p[u]]+w(p[u], u)
$$

and we just proved that $D[u]$ is the shortest $s \rightarrow u$ path distance, this tree stores all shortest path distances from $s$ !

## Acknowledgement

- Based on Prof. Lau's Lecture 9
https://cs.uwaterloo.ca/~lapchi/cs341/notes/L09.pdf
- Also based on [Kleinberg Tardos 2006, Chapter 4]
- For refresher on min heaps, see [CLRS 2009, Chapter 6.5]


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