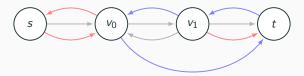
You have a graph G=(V,E). Some of the edges are red, some are white and some are blue. You are given two distinct vertices s and t and want to find a walk $[s \to t]$ such that:

- a white edge must be taken after a red edge only.
- a blue edge must be taken after a white edge only.
- and a red edge may be taken after a blue edge only.
- must start on red edge.



Develop an algorithm to find a path with these edge constraints.

ECE-374-B: Lecture 17 - Bellman-Ford and Dynamic Programming on Graphs

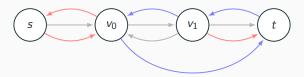
Instructor: Abhishek Kumar Umrawal

October 24, 2023

University of Illinois at Urbana-Champaign

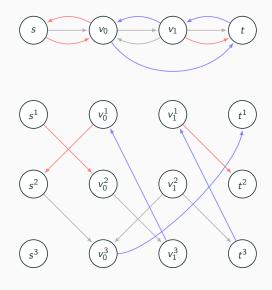
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Shortest Paths with Negative

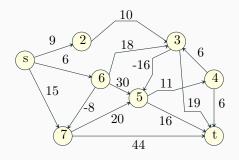
Length Edges

Why Dijkstra's algorithm fails with negative edges

Single-Source Shortest Paths with Negative Edge Lengths

Single-Source Shortest Path Problems Input: A directed graph G = (V, E) with arbitrary (including negative) edge lengths. For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.

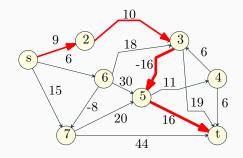
- Given nodes s, t find shortest path from s to t.
- Given node s find shortest path from s to all other nodes.



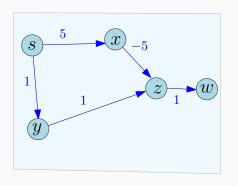
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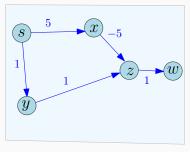


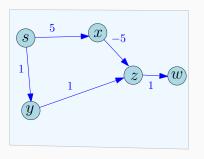
What are the distances computed by Dijkstra's algorithm?

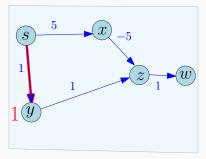


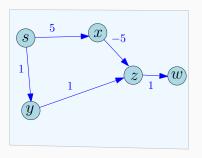
What are the final (shortest) distances as computed by Dijkstra algorithm starting from s?

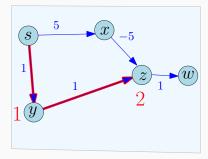
- (a) s = 0, x = 5, y = 1, z = 0, w = 1.
- (b) s = 0, x = 5, y = 1, z = 2, w = 3.
- (c) IDK.

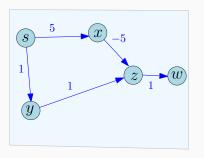


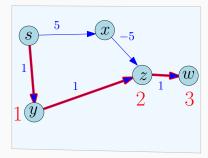


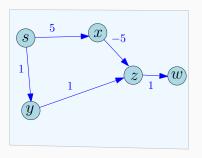


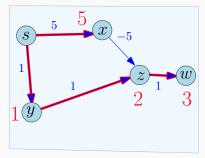


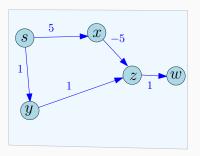


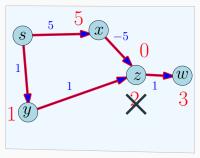


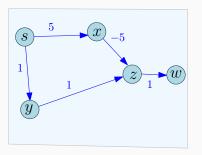


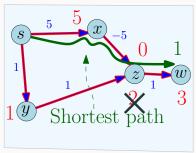




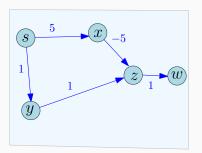


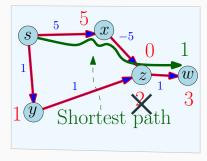






With negative length edges, Dijkstra's algorithm can fail.





False assumption: Dijkstra's algorithm is based on the assumption that if $s \to v_0 \to v_1 \to v_2 \dots \to v_k$ is a shortest path from s to v_k then $dist(s, v_i) \le dist(s, v_{i+1})$ for $0 \le i < k$. Holds true only for non-negative edge lengths.

Shortest Paths with Negative Lengths

Lemma

Let G be a directed graph with arbitrary edge lengths. If $s = v_0 \rightarrow v_1 \rightarrow v_2 \rightarrow \ldots \rightarrow v_k$ is a shortest path from s to v_k then for $1 \le i < k$:

• $s = v_0 \rightarrow v_1 \rightarrow v_2 \rightarrow \ldots \rightarrow v_i$ is a shortest path from s to v_i

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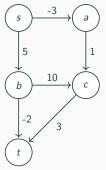
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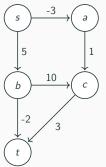
Cannot explore nodes in increasing order of distance! We need other strategies.

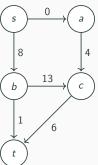
Why can't we just re-normalize the edge lengths!?

Why can't we simply add a weight to each edge so that the shortest length is 0 (or positive).

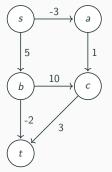


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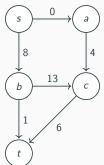




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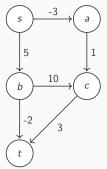


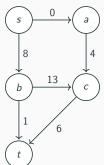
Shortest Path: $s \rightarrow a \rightarrow c \rightarrow t$



Shortest Path: $s \rightarrow b \rightarrow t$

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Shortest Path: $s \to a \to c \to t$ Shortest Path: $s \to b \to t$ Adding weights to edges penalizes paths with more edges.

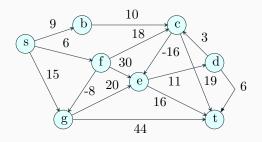
But wait! Things get worse:

Negative cycles

Negative Length Cycles

Definition

A cycle C is a negative length cycle if the sum of the edge lengths of C is negative.

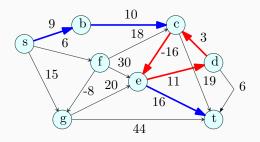


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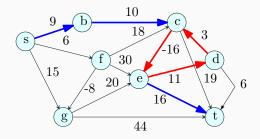


a

Negative Length Cycles

Definition

A cycle C is a negative length cycle if the sum of the edge lengths of C is negative.



What is the shortest path distance between s and t?

Reminder: Paths have to be simple ...

Shortest Paths and Negative Cycles

Given G = (V, E) with edge lengths and s, t. Suppose

- G has a negative length cycle C, and
- s can reach C and C can reach t.

Shortest Paths and Negative Cycles

Given G = (V, E) with edge lengths and s, t. Suppose

- G has a negative length cycle C, and
- s can reach C and C can reach t.

Question: What is the shortest distance from *s* to *t*?

Possible answers: Define shortest distance to be:

- undefined, that is $-\infty$, OR
- the length of a shortest $\underline{\text{simple}}$ path from s to t.

Really bad news about negative edges, and shortest path ...

Lemma

If there is an efficient algorithm to find a shortest simple $s \to t$ path in a graph with negative edge lengths, then there is an efficient algorithm to find the <u>longest</u> simple $s \to t$ path in a graph with positive edge lengths.

Finding the $s \to t$ longest path is difficult. **NP-Hard**!

Restating problem of Shortest path

with negative edges

Alternatively: Finding Shortest Walks

Given a graph G = (V, E):

- A path is a sequence of distinct vertices v_1, v_2, \dots, v_k such that $(v_i, v_{i+1}) \in E$ for $1 \le i \le k-1$.
- A walk is a sequence of vertices v₁, v₂,..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1. Vertices are allowed to repeat.

Define dist(u, v) to be the length of a shortest walk from u to v.

- If there is a walk from u to v that contains negative length cycle then $dist(u, v) = -\infty$.
- Else there is a path with at most n-1 edges whose length is equal to the length of a shortest walk and dist(u, v) is finite.

Helpful to think about walks.

Shortest Paths with Negative Edge Lengths - Problems

Algorithmic Problems

<u>Input</u>: A directed graph G = (V, E) with edge lengths (could be negative). For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.

Questions:

- Given nodes s, t, either find a negative length cycle C that s
 can reach or find a shortest path from s to t.
- Given node s, either find a negative length cycle C that s can reach or find shortest path distances from s to all reachable nodes.
- Check if *G* has a negative length cycle or not.

Shortest Paths with Negative Edge Lengths - In Undirected Graphs

Note: With negative lengths, shortest path problems and negative cycle detection in undirected graphs cannot be reduced to directed graphs by bi-directing each undirected edge. Why?

Problem can be solved efficiently in undirected graphs but algorithms are different and significantly more involved than those for directed graphs. One need to compute T-joins in the relevant graph. Pretty painful stuff.

Shortest path via number of hops

Shortest Paths and Recursion

- Compute the shortest path distance from s to t recursively?
- What are the smaller sub-problems?

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Lemma

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Sub-problem idea: paths of fewer hops/edges

Single-source problem: fix source s. Assume that all nodes can be reached by s in GAssume G has no negative-length cycle (for now).

d(v, k): shortest walk length from s to v using at most k edges.

Single-source problem: fix source *s*.

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Note: dist(s, v) = d(v, n - 1).

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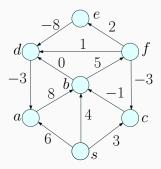
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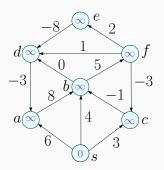
Note: dist(s, v) = d(v, n - 1). Recursion for d(v, k):

$$d(v,k) = \min egin{cases} \min_{u \in V} (d(u,k-1) + \ell(u,v)). \\ d(v,k-1) \end{cases}$$

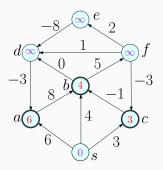
Base case: d(s,0) = 0 and $d(v,0) = \infty$ for all $v \neq s$.



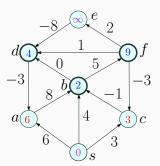
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	S	s a	s a b	s a b c	s a b c d	s a b c d e



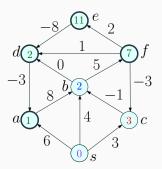
round	S	а	b	С	d	е	f
0	0	∞	∞	∞	∞	∞	∞



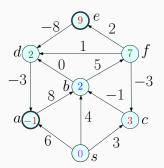
round	S	а	b	С	d	е	f
0	0	∞	∞	∞	∞	∞	∞
1	0	6	4	3	∞	∞	∞



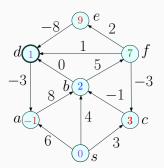
0	∞	∞				
_			∞	∞	∞	∞
0	6	4	3	∞	∞	∞
0	6	2	3	4	∞	9



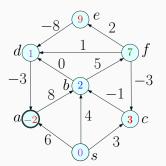
round	S	а	b	С	d	е	f
0	0	∞	∞	∞	∞	∞	∞
1	0	6	4	3	∞	∞	∞
2	0	6	2	3	4	∞	9
3	0	1	2	3	2	11	7



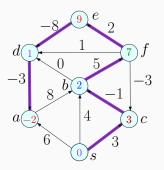
round	S	а	b	С	d	е	f
0	0	∞	∞	∞	∞	∞	∞
1	0	6	4	3	∞	∞	∞
2	0	6	2	3	4	∞	9
3	0	1	2	3	2	11	7
4	0	-1	2	3	2	9	7



round	S	а	b	С	d	е	f
0	0	∞	∞	∞	∞	∞	∞
1	0	6	4	3	∞	∞	∞
2	0	6	2	3	4	∞	9
3	0	1	2	3	2	11	7
4	0	-1	2	3	2	9	7
5	0	-1	2	3	1	9	7



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round	S	а	b	С	d	е	†
0	0	∞	∞	∞	∞	∞	∞
1	0	6	4	3	∞	∞	∞
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0	0	∞	∞	∞	∞	∞	∞
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Create in(G) list from adj(G)
for each u \in V do
     d(u,0) \leftarrow \infty
d(s,0) \leftarrow 0
for k = 1 to n - 1 do
     for each v \in V do
           d(v,k) \leftarrow d(v,k-1)
           for each edge (u, v) \in in(v) do
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Running time:

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Running time: O(n(n+m))

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Running time: O(n(n+m)) Space:

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Running time: O(n(n+m)) Space: $O(m+n^2)$

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Running time: O(n(n+m)) Space: $O(m+n^2)$

Note: Space can be reduced to O(m+n) as any row in our table depends only on itself and the previous row.

Bellman-Ford Algorithm: Cleaner version

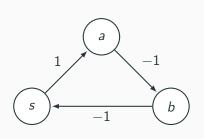
```
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                d(v) = \min\{d(v), d(u) + \ell(u, v)\}\
for each v \in V do
          dist(s, v) \leftarrow d(v)
```

Running time: O(n(m+n)) Space: O(m+n)

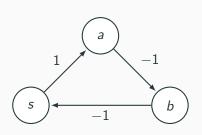
Exercise: Argue that this (cleaner) version achieves the same results the one on the previous slide.

Bellman-Ford: Detecting negative

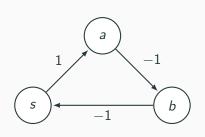
cycles



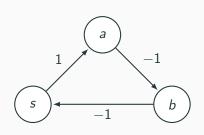
round	S	а	b



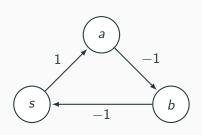
round	S	а	b
0	0	∞	∞



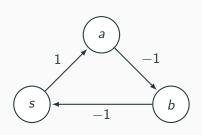
round	S	а	b
0	0	∞	∞
1	0	1	∞



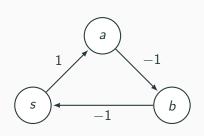
round	S	а	b
0	0	∞	∞
1	0	1	∞
2	0	1	0



round	S	а	b
0	0	∞	∞
1	0	1	∞
2	0	1	0
3	-1	1	0



round	S	а	b
0	0	∞	∞
1	0	1	∞
2	0	1	0
3	-1	1	0
4	-1	0	0



round	S	а	b
0	0	∞	∞
1	0	1	∞
2	0	1	0
3	-1	1	0
4	-1	0	0
5	-1	0	-1

Correctness: detecting negative length cycle

Lemma

Suppose G has a negative cycle C reachable from s. Then there is some node $v \in C$ such that d(v,n) < d(v,n-1).

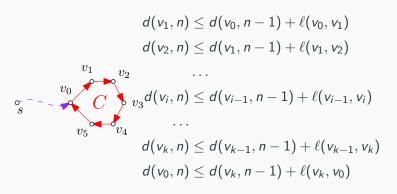
Correctness: detecting negative length cycle

Lemma

Suppose G has a negative cycle C reachable from s. Then there is some node $v \in C$ such that d(v, n) < d(v, n - 1).

Proof.

Suppose not. Let $C = v_1 \rightarrow v_2 \rightarrow \ldots \rightarrow v_h \rightarrow v_1$ be negative length cycle reachable from s. $d(v_i, n-1)$ is finite for $1 \le i \le h$ since C is reachable from s. By assumption $d(v, n) \ge d(v, n - 1)$ for all $v \in C$; implies no change in n^{th} iteration; $d(v_i, n-1) = d(v_i, n)$ for $1 \le i \le h$. This means $d(v_i, n-1) \le d(v_{i-1}, n-1) + \ell(v_{i-1}, v_i)$ for $2 \le i \le h$ and $d(v_1, n-1) \le d(v_n, n-1) + \ell(v_n, v_1)$. Adding up all these inequalities results in the inequality $0 \le \ell(C)$ which contradicts the assumption that $\ell(C) < 0$.



$$d(v_1, n) \leq d(v_0, n) + \ell(v_0, v_1)$$
 $d(v_2, n) \leq d(v_1, n) + \ell(v_1, v_2)$
 \dots
 v_3
 $d(v_i, n) \leq d(v_{i-1}, n) + \ell(v_{i-1}, v_i)$
 \dots
 $d(v_k, n) \leq d(v_{k-1}, n) + \ell(v_{k-1}, v_k)$
 $d(v_0, n) \leq d(v_k, n) + \ell(v_k, v_0)$

$$d(v_{1}, n) \leq d(v_{0}, n) + \ell(v_{0}, v_{1})$$

$$d(v_{2}, n) \leq d(v_{1}, n) + \ell(v_{1}, v_{2})$$

$$\vdots$$

$$v_{3}$$

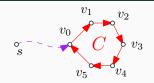
$$d(v_{i}, n) \leq d(v_{i-1}, n) + \ell(v_{i-1}, v_{i})$$

$$\vdots$$

$$d(v_{k}, n) \leq d(v_{k-1}, n) + \ell(v_{k-1}, v_{k})$$

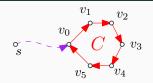
$$d(v_{0}, n) \leq d(v_{k}, n) + \ell(v_{k}, v_{0})$$

$$\sum_{i=0}^{k} d(v_{i}, n) \leq \sum_{i=0}^{k} d(v_{i}, n) + \sum_{i=1}^{k} \ell(v_{i-1}, v_{i}) + \ell(v_{k}, v_{0})$$



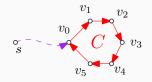
$$\sum_{i=0}^k d(v_i, n) \leq \sum_{i=0}^k d(v_i, n) + \sum_{i=1}^k \ell(v_{i-1}, v_i) + \ell(v_k, v_0)$$

$$0 \leq \sum_{i=1}^{k} \ell(v_{i-1}, v_i) + \ell(v_k, v_0).$$



$$\sum_{i=0}^k d(v_i, n) \leq \sum_{i=0}^k d(v_i, n) + \sum_{i=1}^k \ell(v_{i-1}, v_i) + \ell(v_k, v_0)$$

$$0 \leq \sum_{i=1}^{k} \ell(v_{i-1}, v_i) + \ell(v_k, v_0) = \operatorname{len}(C).$$



$$\sum_{i=0}^k d(v_i, n) \leq \sum_{i=0}^k d(v_i, n) + \sum_{i=1}^k \ell(v_{i-1}, v_i) + \ell(v_k, v_0)$$

$$0 \leq \sum_{i=1}^{k} \ell(v_{i-1}, v_i) + \ell(v_k, v_0) = \operatorname{len}(C).$$

C is a not a negative cycle. Contradiction.

Negative cycles can not hide

Lemma restated

If G does not has a negative length cycle reachable from $s \implies \forall v \colon d(v,n) = d(v,n-1).$

Also, d(v, n-1) is the length of the shortest path between s and v.

Put together are the following:

Lemma

G has a negative length cycle reachable from $s \iff$ there is some node v such that d(v,n) < d(v,n-1).

Bellman-Ford: Negative Cycle Detection - final version

```
for each u \in V do
    d(u) \leftarrow \infty
d(s) \leftarrow 0
for k = 1 to n - 1 do
     for each v \in V do
          for each edge (u, v) \in in(v) do
               d(v) = \min\{d(v), d(u) + \ell(u, v)\}\
(* One more iteration to check if distances change *)
for each v \in V do
     for each edge (u, v) \in in(v) do
          if (d(v) > d(u) + \ell(u, v))
               Output ''Negative Cycle''
for each v \in V do
    \operatorname{dist}(s,v) \leftarrow d(v)
```

Variants on Bellman-Ford

Finding the Paths and a Shortest Path Tree

How do we find a shortest path tree in addition to distances?

- For each v the d(v) can only get smaller as algorithm proceeds.
- If d(v) becomes smaller it is because we found a vertex u such that $d(v) > d(u) + \ell(u, v)$ and we update $d(v) = d(u) + \ell(u, v)$. That is, we found a shorter path to v through u.
- For each v have a prev(v) pointer and update it to point to u
 if v finds a shorter path via u.
- At end of algorithm prev(v) pointers give a shortest path tree oriented towards the source s.

Negative Cycle Detection

Negative Cycle Detection

Given directed graph G with arbitrary edge lengths, does it have a negative length cycle?

Negative Cycle Detection

Negative Cycle Detection

Given directed graph G with arbitrary edge lengths, does it have a negative length cycle?

- Bellman-Ford checks whether there is a negative cycle C that
 is reachable from a specific vertex s. There may negative
 cycles not reachable from s.
- Run Bellman-Ford |V| times, once from each node u?

Negative Cycle Detection

- Add a new node s' and connect it to all nodes of G with zero length edges. Bellman-Ford from s' will fill find a negative length cycle if there is one. Exercise: why does this work?
- Negative cycle detection can be done with one Bellman-Ford invocation.

Shortest Paths in DAGs

Shortest Paths in a DAG

Single-Source Shortest Path Problems

Input A directed acyclic graph G = (V, E) with arbitrary (including negative) edge lengths. For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.

- Given nodes s, t find shortest path from s to t.
- Given node *s* find shortest path from *s* to all other nodes.

Shortest Paths in a DAG

Single-Source Shortest Path Problems

Input A directed acyclic graph G = (V, E) with arbitrary (including negative) edge lengths. For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.

- Given nodes s, t find shortest path from s to t.
- Given node s find shortest path from s to all other nodes.

Simplification of algorithms for DAGs

- No cycles and hence no negative length cycles! Hence can find shortest paths even for negative length edges
- Can order nodes using topological sort

Algorithm for DAGs

- Want to find shortest paths from s. Ignore nodes not reachable from s.
- Let $s = v_1, v_2, v_{i+1}, \dots, v_n$ be a topological sort of G

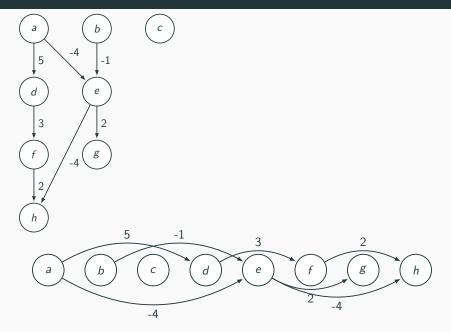
Algorithm for DAGs

- Want to find shortest paths from s. Ignore nodes not reachable from s.
- Let $s = v_1, v_2, v_{i+1}, \dots, v_n$ be a topological sort of G

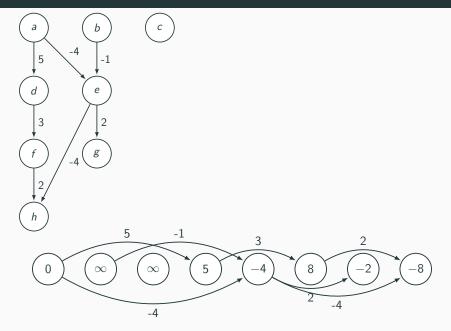
Observation:

- shortest path from s to v_i cannot use any node from v_{i+1}, \ldots, v_n
- can find shortest paths in topological sort order.

Shortest Paths for DAGs - Example



Shortest Paths for DAGs - Example



Algorithm for DAGs

```
\begin{aligned} &\text{for } i=1 \text{ to } n \text{ do} \\ &\quad d(s,v_i)=\infty \\ &d(s,s)=0 \end{aligned} &\text{for } i=1 \text{ to } n-1 \text{ do} \\ &\quad \text{for each edge } (v_i,v_j) \text{ in } \mathrm{Adj}(v_i) \text{ do} \\ &\quad d(s,v_j)=\min\{d(s,v_j),d(s,v_i)+\ell(v_i,v_j)\} \end{aligned} &\text{return } d(s,\cdot) \text{ values computed}
```

Correctness: induction on i and observation in previous slide. Running time: O(m+n) time algorithm! Works for negative edge lengths and hence can find <u>longest</u> paths in a <u>DAG</u>.

All Pairs Shortest Paths

Shortest Path Problems

Shortest Path Problems

Input A (undirected or directed) graph G = (V, E) with edge lengths (or costs). For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.

- Given nodes s, t find shortest path from s to t.
- Given node s find shortest path from s to all other nodes.
- Find shortest paths for <u>all</u> pairs of nodes.

SSSP: Single-Source Shortest Paths

Single-Source Shortest Path Problems

Input A (undirected or directed) graph G = (V, E) with edge lengths. For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.

- Given nodes s, t find shortest path from s to t.
- Given node s find shortest path from s to all other nodes.

SSSP: Single-Source Shortest Paths

Single-Source Shortest Path Problems

- **Input** A (undirected or directed) graph G = (V, E) with edge lengths. For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.
- Given nodes s, t find shortest path from s to t.
- Given node s find shortest path from s to all other nodes.
- **Dijkstra's algorithm** for non-negative edge lengths. Running time: $O((m+n)\log n)$ with heaps and $O(m+n\log n)$ with advanced priority queues.
- **Bellman-Ford algorithm** for arbitrary edge lengths. Running time: O(nm).

All-Pairs Shortest Paths - Using known algorithms...

All-Pairs Shortest Path Problem

Input A (undirected or directed) graph G = (V, E) with edge lengths. For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.

• Find shortest paths for all pairs of nodes.

All-Pairs Shortest Paths - Using known algorithms...

All-Pairs Shortest Path Problem

- **Input** A (undirected or directed) graph G = (V, E) with edge lengths. For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.
- Find shortest paths for all pairs of nodes.

Apply single-source algorithms n times, once for each vertex.

- Non-negative lengths. $O(nm \log n)$ with heaps and $O(nm + n^2 \log n)$ using advanced priority queues.
- Arbitrary edge lengths: $O(n^2m)$. $\Theta(n^4)$ if $m = \Omega(n^2)$.

All-Pairs Shortest Paths - Using known algorithms...

All-Pairs Shortest Path Problem

Input A (undirected or directed) graph G = (V, E) with edge lengths. For edge e = (u, v), $\ell(e) = \ell(u, v)$ is its length.

Find shortest paths for all pairs of nodes.

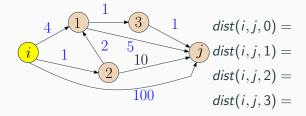
Apply single-source algorithms n times, once for each vertex.

- Non-negative lengths. $O(nm \log n)$ with heaps and $O(nm + n^2 \log n)$ using advanced priority queues.
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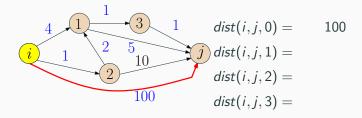
Can we do better?

All Pairs Shortest Paths: A recursive solution

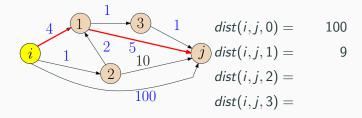
- Number vertices arbitrarily as v_1, v_2, \ldots, v_n
- dist(i, j, k): length of shortest walk from v_i to v_j among all walks in which the largest index of an intermediate node is at most k (could be $-\infty$ if there is a negative length cycle).



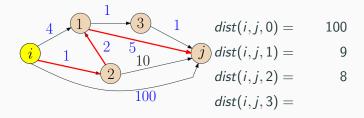
- Number vertices arbitrarily as v_1, v_2, \ldots, v_n
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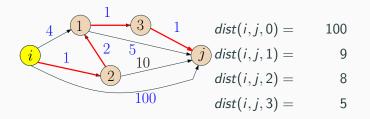
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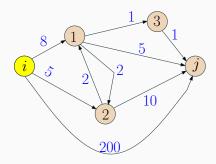
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- Number vertices arbitrarily as v_1, v_2, \ldots, v_n
- dist(i, j, k): length of shortest walk from v_i to v_j among all walks in which the largest index of an intermediate node is at most k (could be $-\infty$ if there is a negative length cycle).

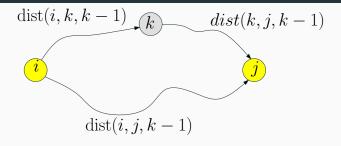


For the following graph, dist(i, j, 2) is...



- 1. 9
- 2. 10
- 3. 11
- 4. 12
- 5. 15

All-Pairs: Recursion on index of intermediate nodes



$$dist(i, j, k) = min$$

$$\begin{cases} dist(i, j, k - 1) \\ dist(i, k, k - 1) + dist(k, j, k - 1) \end{cases}$$

Base case: $dist(i, j, 0) = \ell(i, j)$ if $(i, j) \in E$, otherwise ∞

Correctness: If $i \to j$ shortest walk goes through k then k occurs only once on the path — otherwise there is a negative length

All-Pairs: Recursion on index of intermediate nodes

If i can reach k and k can reach j and dist(k, k, k - 1) < 0 then G has a negative length cycle containing k and $dist(i, j, k) = -\infty$.

Recursion below is valid only if $dist(k, k, k - 1) \ge 0$. We can detect this during the algorithm or wait till the end.

$$dist(i,j,k) = min$$

$$\begin{cases} dist(i,j,k-1) \\ dist(i,k,k-1) + dist(k,j,k-1) \end{cases}$$

Floyd-Warshall algorithm

$$d(i, j, k) = \min \begin{cases} d(i, j, k - 1) \\ d(i, k, k - 1) + d(k, j, k - 1) \end{cases}$$

```
for i = 1 to n do
      for i = 1 to n do
           d(i, j, 0) = \ell(i, j)
 (* \ell(i,j) = \infty \text{ if } (i,j) \notin E, 0 \text{ if } i = j *)
for k = 1 to n do
      for i = 1 to n do
            \quad \mathbf{for}\ j=1\ \mathbf{to}\ n\ \mathbf{do}
                 d(i,j,k) = \min \begin{cases} d(i,j,k-1), \\ d(i,k,k-1) + d(k,j,k-1) \end{cases}
for i = 1 to n do
      if (dist(i, i, n) < 0) then
            Output \exists negative cycle in G
```

$$d(i, j, k) = \min \begin{cases} d(i, j, k - 1) \\ d(i, k, k - 1) + d(k, j, k - 1) \end{cases}$$

```
\begin{array}{l} \mbox{ for } i = 1 \mbox{ to } n \mbox{ do} \\ \mbox{ for } j = 1 \mbox{ to } n \mbox{ do} \\ \mbox{ } d(i,j,0) = \ell(i,j) \\ \mbox{ (* $\ell(i,j)$} = \infty \mbox{ if } (i,j) \notin E, \mbox{ 0 if } i = j \mbox{ *)} \\ \mbox{ for } k = 1 \mbox{ to } n \mbox{ do} \\ \mbox{ for } i = 1 \mbox{ to } n \mbox{ do} \\ \mbox{ } for \mbox{ } j = 1 \mbox{ to } n \mbox{ do} \\ \mbox{ } d(i,j,k) = \min \begin{cases} d(i,j,k-1), \\ d(i,k,k-1) + d(k,j,k-1) \end{cases} \\ \mbox{ for } i = 1 \mbox{ to } n \mbox{ do} \\ \mbox{ if } (dist(i,i,n) < 0) \mbox{ then} \\ \mbox{ Output } \exists \mbox{ negative cycle in } G \end{array}
```

Running Time:

$$d(i, j, k) = \min \begin{cases} d(i, j, k - 1) \\ d(i, k, k - 1) + d(k, j, k - 1) \end{cases}$$

```
for i = 1 to n do
      for i = 1 to n do
            d(i, j, 0) = \ell(i, j)
 (* \ell(i,j) = \infty \text{ if } (i,j) \notin E, 0 \text{ if } i = j *)
for k = 1 to n do
      for i = 1 to n do
            \quad \mathbf{for} \ j = 1 \ \mathsf{to} \ n \ \mathbf{do}
                  d(i,j,k) = \min \begin{cases} d(i,j,k-1), \\ d(i,k,k-1) + d(k,j,k-1) \end{cases}
for i = 1 to n do
     if (dist(i, i, n) < 0) then
            Output \exists negative cycle in G
```

Running Time: $\Theta(n^3)$. Space: $\Theta(n^3)$.

$$d(i, j, k) = \min \begin{cases} d(i, j, k - 1) \\ d(i, k, k - 1) + d(k, j, k - 1) \end{cases}$$

```
for i = 1 to n do
     for i = 1 to n do
           d(i, j, 0) = \ell(i, j)
 (* \ell(i,j) = \infty \text{ if } (i,j) \notin E, 0 \text{ if } i = j *)
for k = 1 to n do
     for i = 1 to n do
           for j = 1 to n do
                d(i,j,k) = \min \begin{cases} d(i,j,k-1), \\ d(i,k,k-1) + d(k,j,k-1) \end{cases}
for i = 1 to n do
     if (dist(i, i, n) < 0) then
           Output \exists negative cycle in G
```

Running Time: $\Theta(n^3)$. Space: $\Theta(n^3)$.

Correctness: via induction and recursive definition

Floyd-Warshall Algorithm: Finding the Paths

Question: Can we find the paths in addition to the distances?

Floyd-Warshall Algorithm: Finding the Paths

Question: Can we find the paths in addition to the distances?

- Create a n × n array Next that stores the next vertex on shortest path for each pair of vertices
- With array Next, for any pair of given vertices i, j can compute a shortest path in O(n) time.

Floyd-Warshall Algorithm - Finding the Paths

```
for i = 1 to n do
    for j = 1 to n do
         d(i,j,0) = \ell(i,j)
(* \ell(i,j) = \infty \text{ if } (i,j) \text{ not edge, 0 if } i = j *)
         Next(i, i) = -1
for k = 1 to n do
    for i = 1 to n do
         for i = 1 to n do
              if (d(i, j, k-1) > d(i, k, k-1) + d(k, j, k-1)) then
                   d(i, j, k) = d(i, k, k - 1) + d(k, j, k - 1)
                   Next(i, i) = k
for i = 1 to n do
    if (d(i, i, n) < 0) then
         Output that there is a negative length cycle in G
```

Exercise: Given *Next* array and any two vertices i, j describe an O(n) algorithm to find a i-j shortest path.

Summary of shortest path

algorithms

Summary of results on shortest paths

Single source		
No negative edges	Dijkstra	$O(n \log n + m)$
Edge lengths can be negative	Bellman Ford	O(nm)

All Pairs Shortest Paths

No negative edges	n * Dijkstra	$O(n^2 \log n + nm)$
No negative cycles	n * Bellman Ford	$O(n^2m) = O(n^4)$
No negative cycles	Johnson's ¹	$O(nm + n^2 \log n)$
No negative cycles	Floyd-Warshall	$O(n^3)$
Unweighted	Matrix multiplication ²	$O(n^{2.38}), O(n^{2.58})$

Summary of results on shortest paths

- (1): The algorithm for the case that there are no negative cycles, and doing all shortest paths, works by computing a potential function using **Bellman-Ford** and then doing **Dijkstra**. It is mentioned for the sake of completeness, but it outside the scope of the class.
- (2): https://resources.mpi-inf.mpg.de/departments/d1/teaching/ss12/AdvancedGraphAlgorithms/Slides14.pdf