

Lecture 11

Shortest Paths

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Lecture Goals

- In this lecture we study **shortest-paths** problems. We begin by analyzing some basic properties of shortest paths and a generic algorithm for the problem.
- We introduce and analyze **Dijkstra's algorithm** for shortest-paths problems with nonnegative weights.
- Next, we consider an even faster **algorithm for DAGs**, which works even if the weights are negative.
- We conclude with the **Bellman–Ford–Moore** algorithm for edge-weighted digraphs with no negative cycles.

Shortest Paths in an Edge-weighted Digraph

Given an edge-weighted digraph, find the shortest path from s to t .

edge-weighted digraph

| | |
|-----|------|
| 4→5 | 0.35 |
| 5→4 | 0.35 |
| 4→7 | 0.37 |
| 5→7 | 0.28 |
| 7→5 | 0.28 |
| 5→1 | 0.32 |
| 0→4 | 0.38 |
| 0→2 | 0.26 |
| 7→3 | 0.39 |
| 1→3 | 0.29 |
| 2→7 | 0.34 |
| 6→2 | 0.40 |
| 3→6 | 0.52 |
| 6→0 | 0.58 |
| 6→4 | 0.93 |



shortest path from 0 to 6

| | |
|-----|------|
| 0→2 | 0.26 |
| 2→7 | 0.34 |
| 7→3 | 0.39 |
| 3→6 | 0.52 |

Variants

❖ Which vertices?

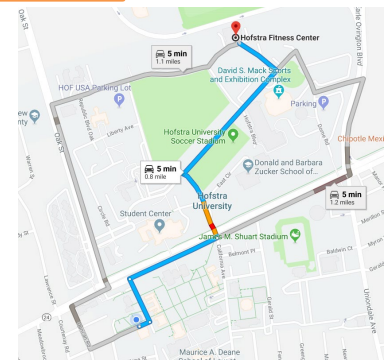
- Single source: from one vertex s to every other vertex.
- Source-sink: from one vertex s to another t .
- All pairs: between all pairs of vertices.

❖ Nonnegative weights?

❖ Cycles?

- Negative cycles.

Can we use BFS?



Simplifying assumption: Each vertex is reachable from s .

Weighted Directed Edge API

```
public class DirectedEdge
```

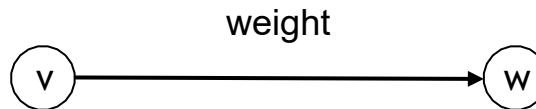
```
    DirectedEdge(int v, int w, double weight) //weighted edge v->w
```

```
    int from() // vertex v
```

```
    int to() // vertex w
```

```
    double weight() // the weight
```

```
    String toString() // string representation
```



Idiom for processing an edge *e*: `int v = e.from(), w = e.to();`

Weighted Edge: Java Implementation

```
public class DirectedEdge
{
    private final int v, w;
    private final double weight;

    public DirectedEdge(int v, int w, double weight)
    {
        this.v = v;
        this.w = w;
        this.weight = weight;
    }

    public int from()
    { return v; }

    public int to()
    { return w; }

    public int weight()
    { return weight; }
}
```

Edge-Weighted Graph API

```
public class EdgeWeightedDigraph
```

```
    EdgeWeightedDigraph(int V) // edge-weighted digraph with V vertices
```

```
    void addEdge(DirectedEdge e) // add weighted directed edge e
```

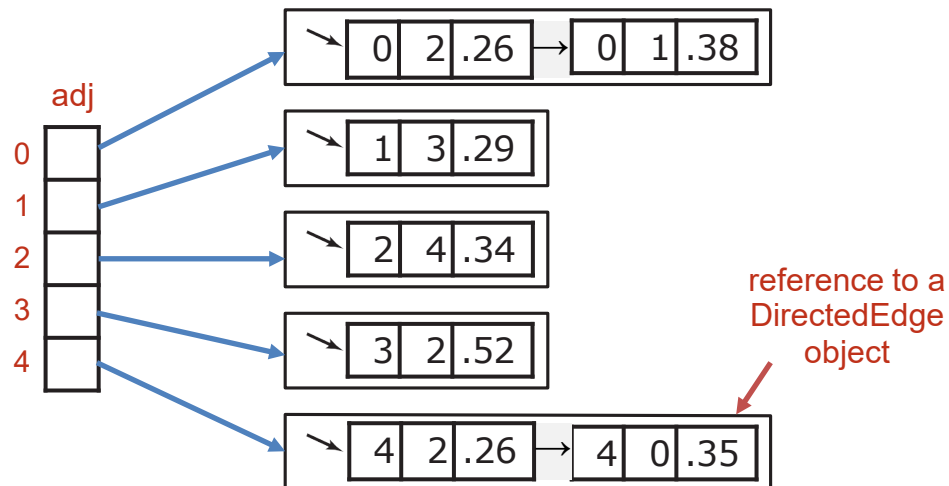
```
    Iterable<DirectedEdge> adj(int v) // edges pointing from v
```

```
    Iterable<DirectedEdge> edges() // all edges in this graph
```

```
    int V() // number of vertices
```

```
    int E() // number of edges
```

```
    String toString() // string representation
```



Edge-Weighted Digraph: Adjacency-Lists Implementation

```
public class EdgeWeightedDigraph
{
    private final int V;
    private final List<DirectedEdge>[] adj;

    public EdgeWeightedDigraph (int V)
    {
        this.V = V;
        adj = (List<DirectedEdge>[]) new ArrayList[V];
        for (int v = 0; v < V; v++)
            adj[v] = new ArrayList<DirectedEdge>();
    }

    public void addEdge(DirectedEdge e)
    {
        int v = e.from();
        adj[v].add(e);
    }

    public Iterable<DirectedEdge> adj(int v)
    {
        return adj[v];
    }
}
```

← add edge $e = v \rightarrow w$ to
only v 's adjacency lists

Single-source Shortest Paths API

Goal. Find the shortest path from s to every other vertex.

```
public class SP
```

```
    SP(EdgeWeightedGraph G, int s) // shortest paths from s in graph G
```

```
    double distTo(int v) // length of shortest path from s to v
```

```
    Iterable<DirectedEdge> pathTo(int v) // shortest path from s to v
```

```
SP sp = new SP(G, s);
for (int v = 0; v < G.V(); v++)
{
    StdOut.printf("%d to %d (0.2f): ", s, v, sp.distTo(v));
    for (DirectedEdge e : sp.pathTo(v)) {
        StdOut.print(e + " ");
        StdOut.println();
    }
}
```

```
0 to 0 (0.00):
0 to 1 (1.05): 0->4 0.38 4->5 0.35 5->1 0.32
0 to 2 (0.26): 0->2 0.26
0 to 3 (0.99): 0->2 0.26 2->7 0.34 7->3 0.39
0 to 4 (0.38): 0->4 0.38
0 to 5 (0.73): 0->4 0.38 4->5 0.35
0 to 6 (1.51): 0->2 0.26 2->7 0.34 7->3 0.39 3->6 0.52
0 to 7 (0.60): 0->2 0.26 2->7 0.34
```

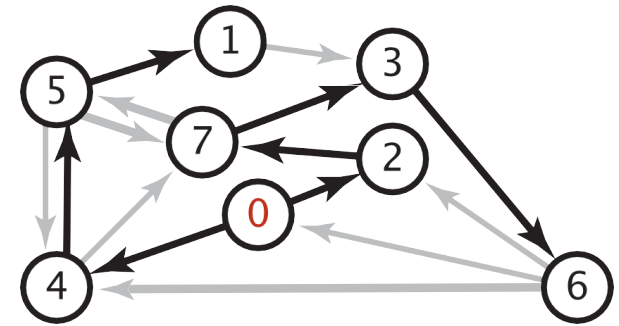

Data Structures for Single-source Shortest Paths

Goal. Find the shortest path from s to every other vertex.

Observation. A shortest-paths tree (SPT) solution exists.

Consequence. Can represent the SPT with two vertex-indexed arrays:

- $\text{distTo}[v]$ is length of shortest path from s to v .
- $\text{edgeTo}[v]$ is last edge on shortest path from s to v .



shortest-paths tree from 0

| | edgeTo[] | distTo[] |
|---|-----------|----------|
| 0 | null | 0 |
| 1 | 5->1 0.32 | 1.05 |
| 2 | 0->2 0.26 | 0.26 |
| 3 | 7->3 0.37 | 0.97 |
| 4 | 0->4 0.38 | 0.38 |
| 5 | 4->5 0.35 | 0.73 |
| 6 | 3->6 0.52 | 1.49 |
| 7 | 2->7 0.34 | 0.60 |

parent-link representation

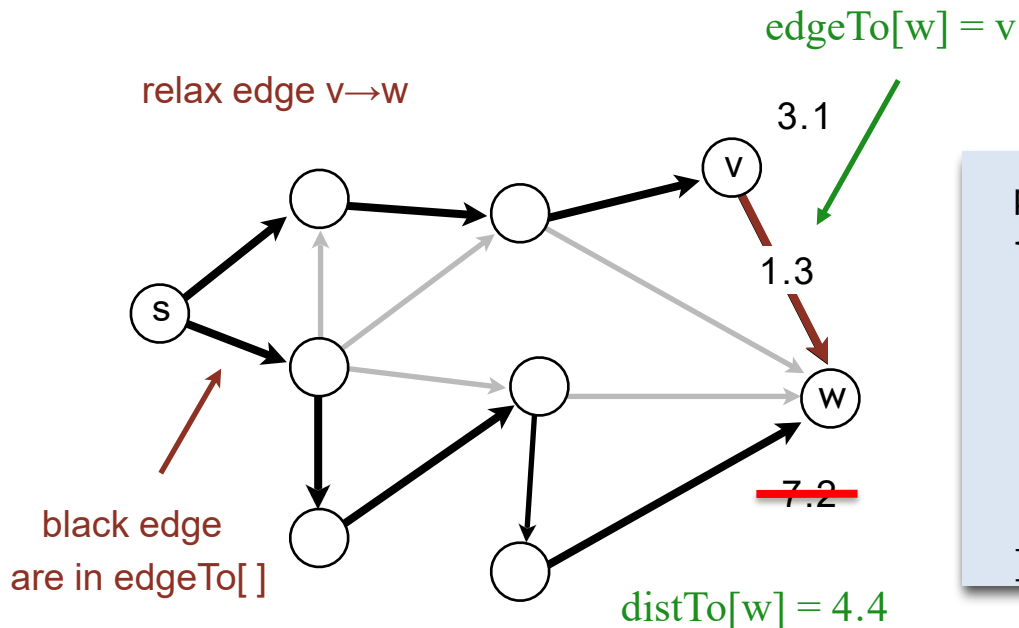
```
public double distTo(int v)
{ return distTo[v]; }

public Iterable<DirectedEdge> pathTo(int v)
{
    Stack<DirectedEdge> path = new Stack<DirectedEdge>();
    for (DirectedEdge e = edgeTo[v]; e != null; e = edgeTo[e.from()])
        path.push(e);
    return path;
}
```

Edge Relaxation

Relax edge $e = v \rightarrow w$. (basic of building SPT)

- $\text{distTo}[v]$ is length of shortest **known** path from s to v .
- $\text{distTo}[w]$ is length of shortest **known** path from s to w .
- $\text{edgeTo}[w]$ is last edge on shortest **known** path from s to w .
- If $e = v \rightarrow w$ gives shorter path to w through v , update $\text{distTo}[w]$ and $\text{edgeTo}[w]$.



```
private void relax(DirectedEdge e)
{
    int v = e.from(), w = e.to();
    if (distTo[w] > distTo[v] + e.weight())
    {
        distTo[w] = distTo[v] + e.weight();
        edgeTo[w] = e;
    }
}
```

Generic Shortest-paths Algorithm

Generic algorithm (to compute SPT from s)

For each vertex v : $\text{distTo}[v] = \infty$.

For each vertex v : $\text{edgeTo}[v] = \text{null}$.

$\text{distTo}[s] = 0$.

Repeat until done:

- Relax any edge.

Proposition. Generic algorithm computes SPT (if it exists) from s .

Pf.

- Throughout algorithm, $\text{distTo}[v]$ is the length of a simple path from s to v (and $\text{edgeTo}[v]$ is last edge on path).
- Each successful relaxation decreases $\text{distTo}[v]$ for some v .
- The entry $\text{distTo}[v]$ can decrease at most a finite number of times.

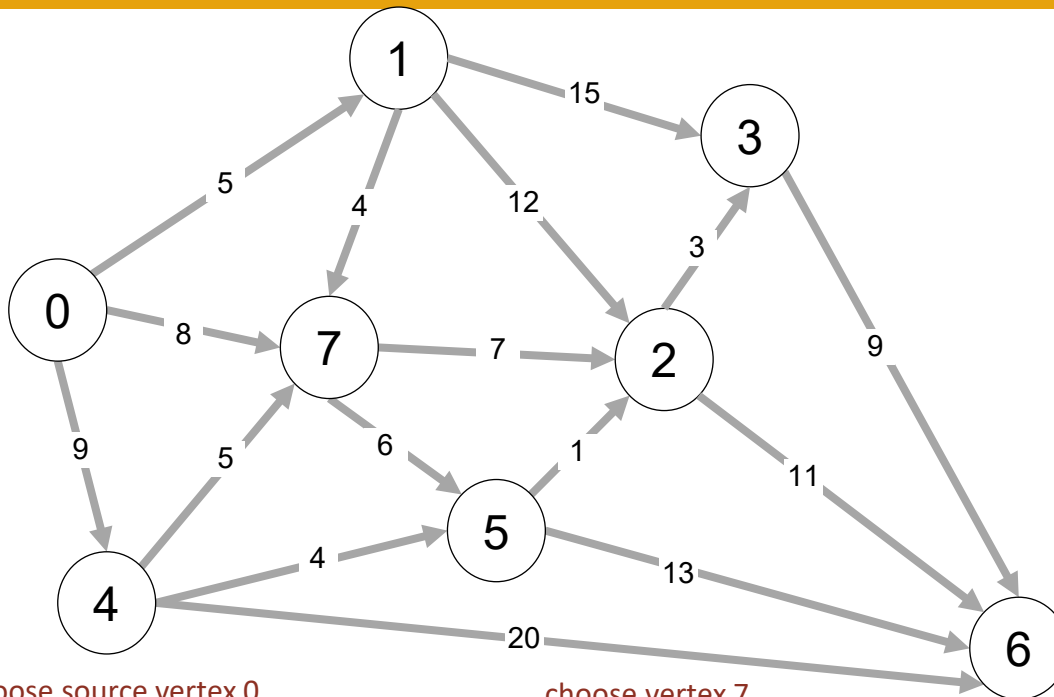
Efficient implementations. How to choose which edge to relax?

- Ex 1. Dijkstra's algorithm. (**nonnegative weights, directed cycles**).
- Ex 2. Topological sort algorithm. (**no directed cycles**).
- Ex 3. Bellman–Ford algorithm. (**no negative cycles**).

Dijkstra's Algorithm

- Consider vertices in increasing order of distance from s
 - (non-tree vertex with the lowest distTo[] value).
- Add vertex to tree and relax all edges pointing from that vertex.

choose vertex 5
 relax all edges adjacent from 5
 choose vertex 2
 relax all edges adjacent from 2
 choose vertex 3
 relax all edges adjacent from 3
 choose vertex 6
 relax all edges adjacent from 6



choose source vertex 0
 relax all edges adjacent from 0
 choose vertex 1
 relax all edges adjacent from 1

choose vertex 7
 relax all edges adjacent from 7
 choose vertex 4
 relax all edges adjacent from 4



| v distTo[] | | | |
|------------|--------------|---------------|------------------|
| 0 | ∞ | 0 | |
| 1 | ∞ | 5 | |
| 2 | ∞ | 17 | 15 14 |
| 3 | ∞ | 20 | 17 |
| 4 | ∞ | 9 | |
| 5 | ∞ | 14 | 13 |
| 6 | ∞ | 29 | 26 25 |
| 7 | ∞ | 8 | |

| v edgeTo[] | | | |
|------------|--------------|--------------|----------------|
| 0 | - | | |
| 1 | - | 0 | |
| 2 | - | 1 | 7 5 |
| 3 | - | 1 | 2 |
| 4 | - | 0 | |
| 5 | - | 7 | 4 |
| 6 | - | 4 | 5 2 |
| 7 | - | 0 | |

Dijkstra's Algorithm: Correctness Proof

Proposition. Dijkstra's algorithm computes a SPT in any edge-weighted digraph with nonnegative weights.

Pf.

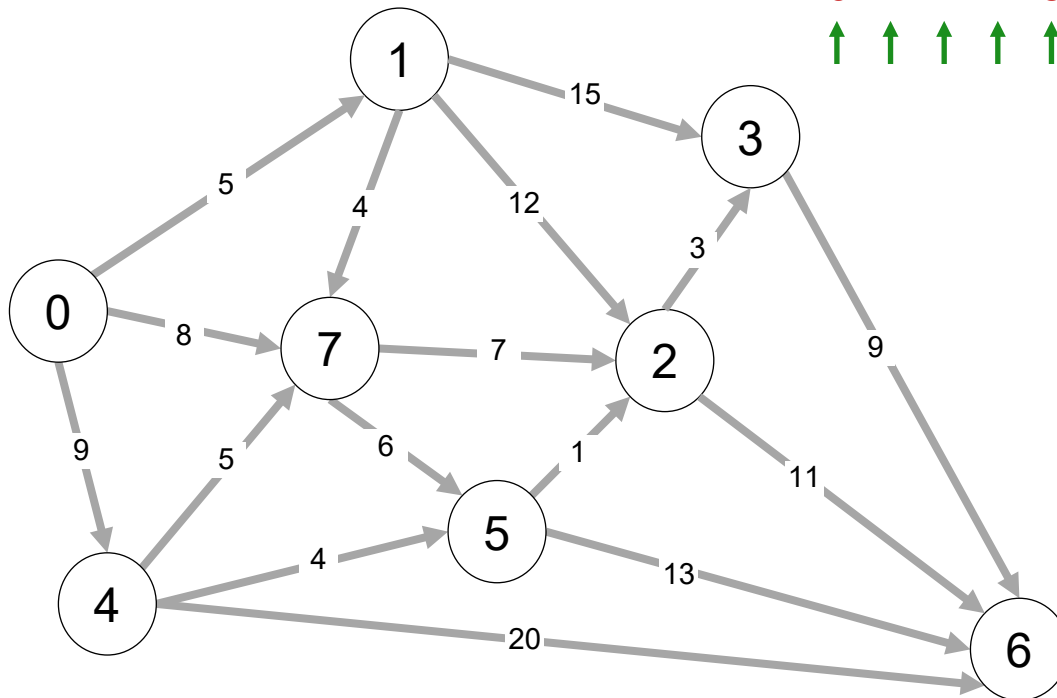
- Each edge $e = v \rightarrow w$ is relaxed exactly once (when v is relaxed),
 - leaving $\text{distTo}[w] \leq \text{distTo}[v] + e.\text{weight}()$.
- Inequality holds until algorithm terminates because:
 - $\text{distTo}[w]$ cannot increase  $\text{distTo}[\]$ values are monotone decreasing
 - $\text{distTo}[v]$ will not change  we choose lowest $\text{distTo}[\]$ value at each step (and edge weights are nonnegative)
- Thus, upon termination, shortest-paths optimality conditions hold.

Shortest Paths in Edge-weighted DAG

Suppose that an edge-weighted digraph has no directed cycles. Is it easier to find shortest paths than in a general digraph?

Yes!

- Consider vertices in topological order.
- Relax all edges pointing from that vertex



0 1 4 7 5 2 3 6
↑ ↑ ↑ ↑ ↑ ↑ ↑ ↑

| v | distTo[] | |
|---|--------------|--------------------------------|
| 0 | ∞ | 0 |
| 1 | ∞ | 5 |
| 2 | ∞ | 17 15 14 |
| 3 | ∞ | 20 17 |
| 4 | ∞ | 9 |
| 5 | ∞ | 13 |
| 6 | ∞ | 29 26 25 |
| 7 | ∞ | 8 |

| v | edgeTo[] | |
|---|--------------|-----------------------------|
| 0 | - | |
| 1 | - | 0 |
| 2 | - | 1 7 5 |
| 3 | - | 1 2 |
| 4 | - | 0 |
| 5 | - | 4 |
| 6 | - | 4 5 2 |
| 7 | - | 0 |

Shortest Paths in Edge-weighted DAG: Correctness Proof

Proposition. Topological sort algorithm computes SPT in any edge-weighted DAG in time proportional to $E + V$.

edge weights
can be negative!

Pf.

- Each edge $e = v \rightarrow w$ is relaxed exactly once (when v is relaxed),
 - leaving $\text{distTo}[w] \leq \text{distTo}[v] + e.\text{weight}()$.
- Inequality holds until algorithm terminates because:
 - $\text{distTo}[w]$ cannot increase ← $\text{distTo}[\]$ values are monotone decreasing
 - $\text{distTo}[v]$ will not change ← because of topological order, no edge pointing to v will be relaxed after v is relaxed
- Thus, upon termination, shortest-paths optimality conditions hold.

Longest Paths in Edge-weighted DAG

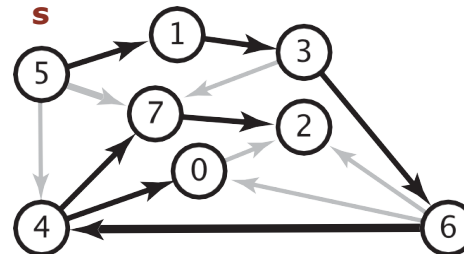
Formulate as a shortest paths problem in edge-weighted DAGs.

- Negate all weights.
 - Find shortest paths.
 - Negate weights in result.
- equivalent: reverse sense of equality in relax()

longest paths input

shortest paths input

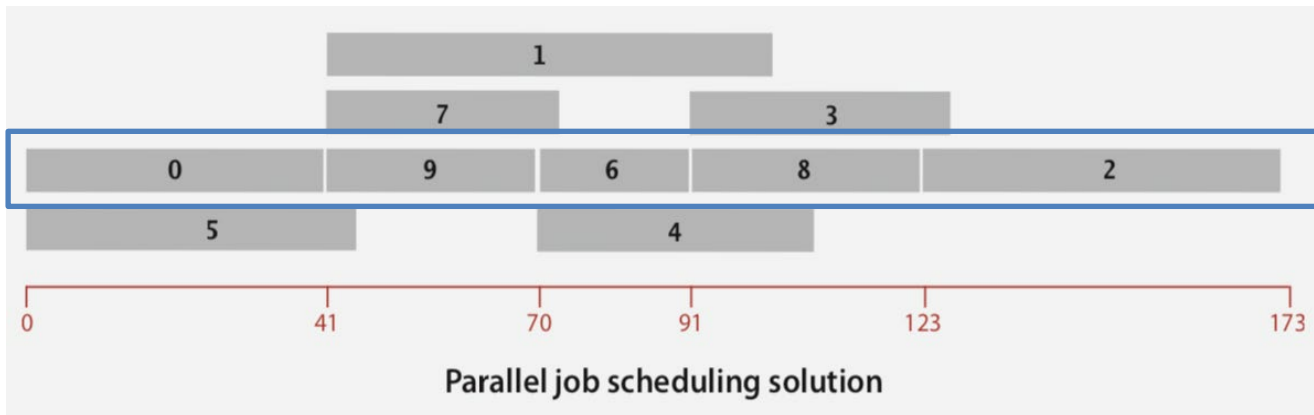
| | | | |
|------|------|------|-------|
| 5->4 | 0.35 | 5->4 | -0.35 |
| 4->7 | 0.37 | 4->7 | -0.37 |
| 5->7 | 0.28 | 5->7 | -0.28 |
| 5->1 | 0.32 | 5->1 | -0.32 |
| 4->0 | 0.38 | 4->0 | -0.38 |
| 0->2 | 0.26 | 0->2 | -0.26 |
| 3->7 | 0.39 | 3->7 | -0.39 |
| 1->3 | 0.29 | 1->3 | -0.29 |
| 7->2 | 0.34 | 7->2 | -0.34 |
| 6->2 | 0.40 | 6->2 | -0.40 |
| 3->6 | 0.52 | 3->6 | -0.52 |
| 6->0 | 0.58 | 6->0 | -0.58 |
| 6->4 | 0.93 | 6->4 | -0.93 |



Key point. Topological sort algorithm works even with negative weights.

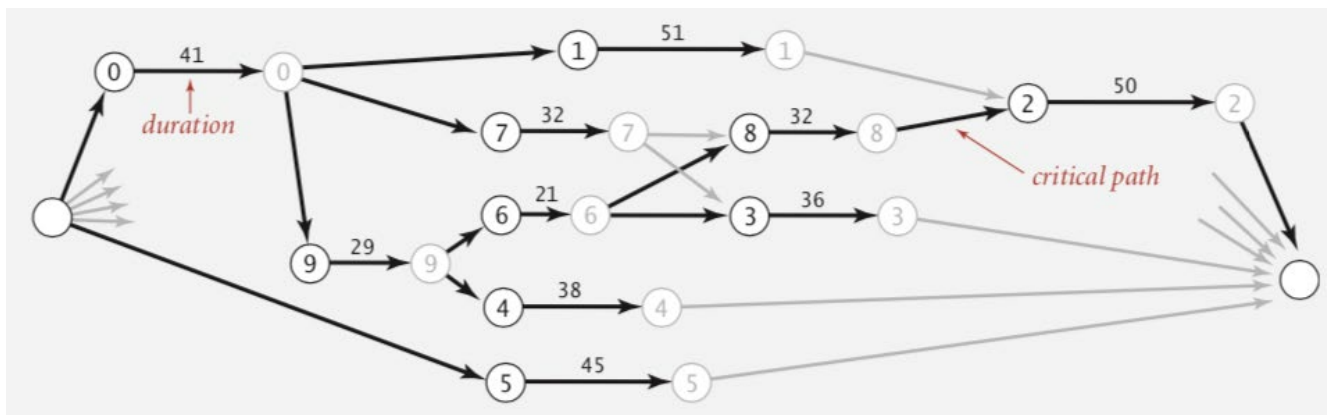
Longest Paths in Edge-weighted DAG: Application

Parallel job scheduling. Given a set of jobs with durations and precedence constraints, schedule the jobs (by finding a start time for each) so as to achieve the minimum completion time, while respecting the constraints.



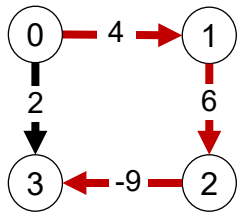
| job | duration | must complete before | | |
|-----|----------|----------------------|---|---|
| 0 | 41.0 | 1 | 7 | 9 |
| 1 | 51.0 | 2 | | |
| 2 | 50.0 | | | |
| 3 | 36.0 | | | |
| 4 | 38.0 | | | |
| 5 | 45.0 | | | |
| 6 | 21.0 | 3 | 8 | |
| 7 | 32.0 | 3 | 8 | |
| 8 | 32.0 | 2 | | |
| 9 | 29.0 | 4 | 6 | |

Use **longest path** from the source to schedule each job.

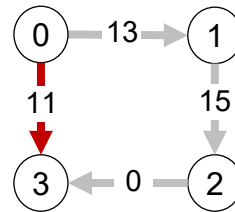


Shortest Paths with Negative weights

Dijkstra. Doesn't work with negative edge weights.



Dijkstra selects vertex 3 immediately after 0.
But shortest path from 0 to 3 is $0 \rightarrow 1 \rightarrow 2 \rightarrow 3$.

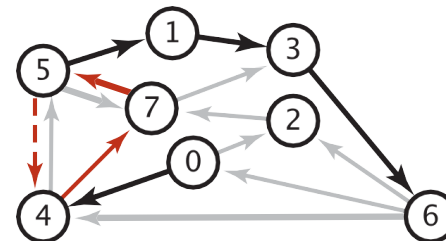


Adding 9 to each edge weight changes the
shortest path from $0 \rightarrow 1 \rightarrow 2 \rightarrow 3$ to $0 \rightarrow 3$.

Conclusion.
Need a different algorithm.

Re-weighting. Add a constant to every edge weight doesn't work.

- A **negative cycle** is a directed cycle whose sum of edge weights is negative.
- A SPT exists **iff** no negative cycles, assuming all vertices reachable from s



negative cycle $(-0.66 + 0.37 + 0.28)$

$5 \rightarrow 4 \rightarrow 7 \rightarrow 5$

shortest path from 0 to 6

$0 \rightarrow 4 \rightarrow 7 \rightarrow 5 \rightarrow 4 \rightarrow 7 \rightarrow 5 \dots \rightarrow 1 \rightarrow 3 \rightarrow 6$

| | |
|------|-------|
| 4->5 | 0.35 |
| 5->4 | -0.66 |
| 4->7 | 0.37 |
| 5->7 | 0.28 |
| 7->5 | 0.28 |
| 5->1 | 0.32 |
| 0->4 | 0.38 |
| 0->2 | 0.26 |
| 7->3 | 0.39 |
| 1->3 | 0.29 |
| 2->7 | 0.34 |
| 6->2 | 0.40 |
| 3->6 | 0.52 |
| 6->0 | 0.58 |
| 6->4 | 0.93 |

Bellman-Ford Algorithm

Bellman-Ford algorithm

For each vertex v : $\text{distTo}[v] = \infty$.

For each vertex v : $\text{edgeTo}[v] = \text{null}$.

$\text{distTo}[s] = 0$.

Repeat $V-1$ times:

- Relax each edge.

```
for (int i = 1; i < G.V(); i++)  
    for (int v = 0; v < G.V(); v++)  
        for (DirectedEdge e : G.adj(v))  
            relax(e);
```

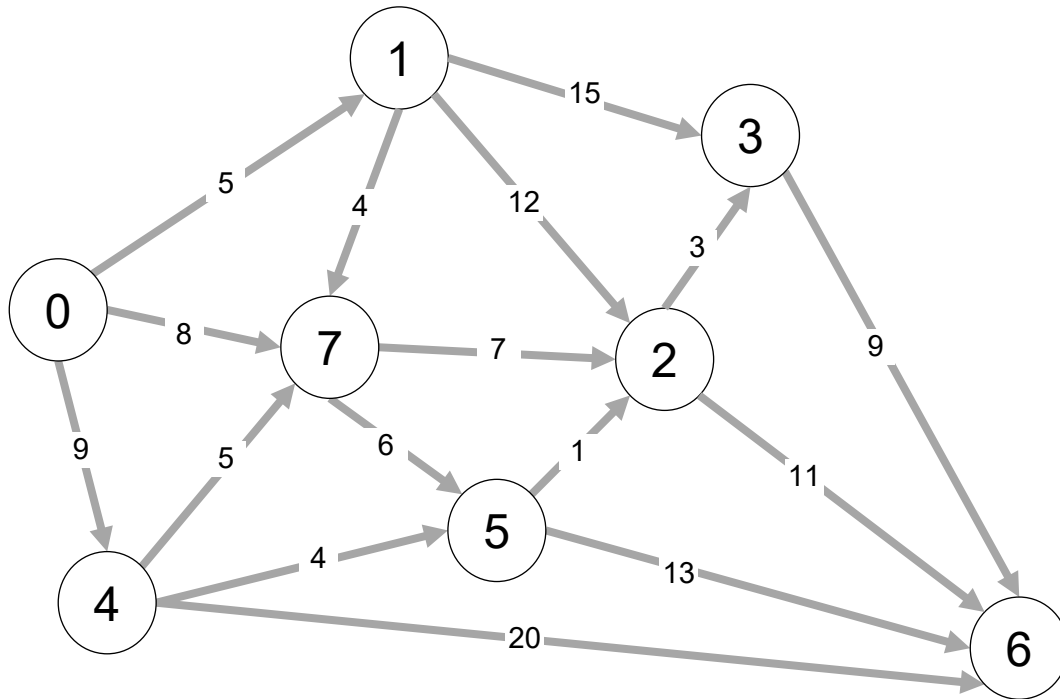
← pass i (relax each edge)

Bellman-Ford in 5 minutes — Step by step example

<https://www.youtube.com/watch?v=obWXjtg0L64>

Bellman-Ford Algorithm

Repeat $V - 1$ times: relax all E edges.



| v | distTo[] | |
|---|---|--|
| 0 | ∞ 0 | |
| 1 | ∞ 5 | |
| 2 | ∞ 17 14 | |
| 3 | ∞ 20 17 | |
| 4 | ∞ 9 | |
| 5 | ∞ 13 | |
| 6 | ∞ 28 26 25 | |
| 7 | ∞ 8 | |

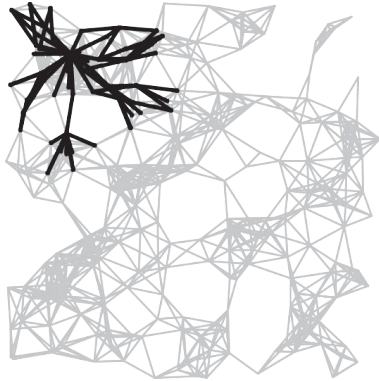
| v | edgeTo[] | |
|---|--|--|
| 0 | - | |
| 1 | - 0 | |
| 2 | - 1 5 | |
| 3 | - 1 2 | |
| 4 | - 0 | |
| 5 | - 4 | |
| 6 | - 2 5 2 | |
| 7 | - 0 | |

pass 1 pass 2 pass 3 (no further changes) pass 4-7 (no further changes)

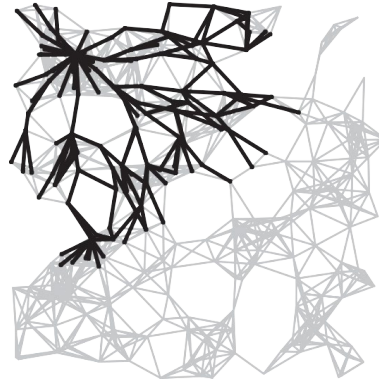
0→1 0→4 0→7 1→2 1→3 1→7 2→3 2→6 3→6 4→5 4→6 4→7 5→2 5→6 7→2 7→5

Bellman-Ford Algorithm Visualization

passes 4



7



10



13



SPT



Single Source Shortest-paths Implementation: Cost Summary

| algorithm | restriction | typical case | worst case | extra space |
|---|---------------------|--------------|------------|-------------|
| topological sort | no directed cycles | $E + V$ | $E + V$ | V |
| Dijkstra (binary heap) | no negative weights | $E \log V$ | $E \log V$ | V |
| Bellman-Ford | no negative cycles | $E V$ | $E V$ | V |
| Bellman-Ford (queue-based) (omitted) | | $E + V$ | $E V$ | V |

- **Remark 1.** Directed cycles make the problem harder.
- **Remark 2.** Negative weights make the problem harder.
- **Remark 3.** Negative cycles makes the problem intractable.

Backup Slides