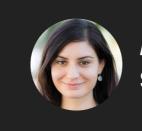
CS062 DATA STRUCTURES AND ADVANCED PROGRAMMING

23: Shortest Paths

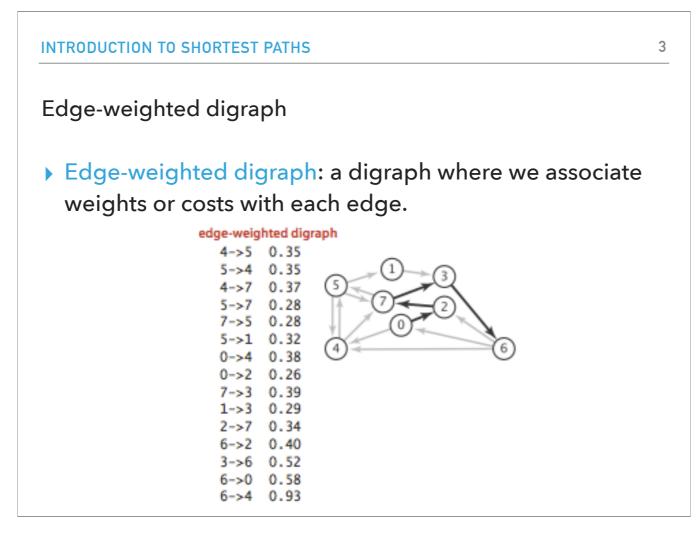


Alexandra Papoutsaki she/her/hers

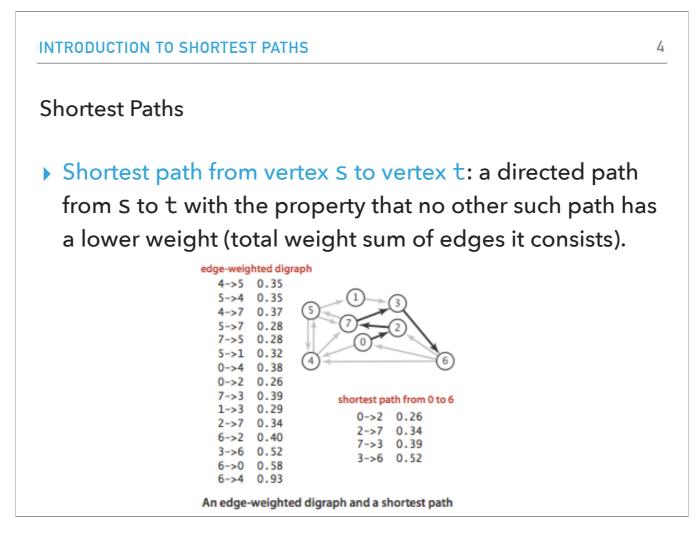
So far, we have seen how to represent and traverse graphs through DFS and BFS.



Today we will see an algorithm that will allow us to calculate the shortest path from one vertex to every other vertex in the graph.



We will work with edge-weighted digraphs, that is digraphs where we associate weights or costs with each edge. For example, the graph on the right only shows vertices and edges that connect them, but we also maintain a list of the weight for each edge.



A shortest path from vertex s to vertex t is a directed path from s to t with the property that no other such path has a lower weight, i.e., the total weight sum of edges it consists. (it's ok to have another path with the same total weight). For example, a shortest path from vertex 0 to vertex 6 in the graph above could take us from 0 to 2 to 7 to 3 to 6 with a total cost of 1.51 and no other path will have a lower weight than that.

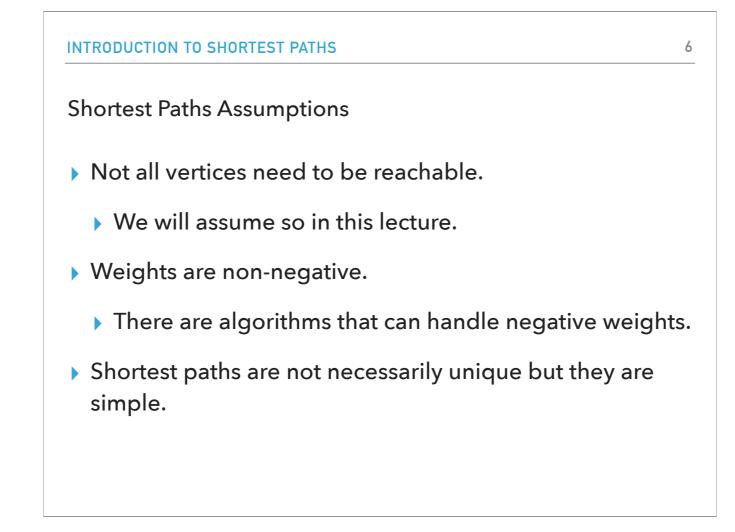
5

Shortest Path variants

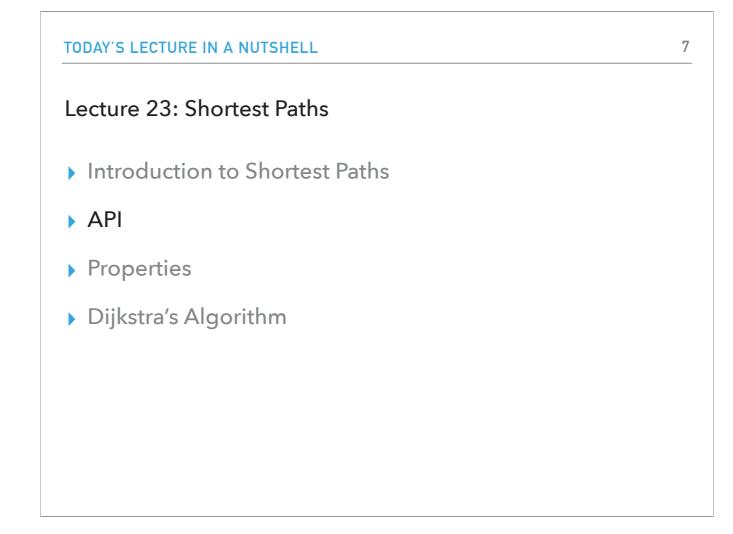
- Single source: from one vertex S to every other vertex.
- Single sink: from every vertex to one vertex t.
- Source-sink: from one vertex S to another vertex t.
- All pairs: from every vertex to every other vertex.
- > What version is there in your navigation app?

The problem of finding shortest paths has a lot of variants. For example: Single source: from one vertex s to every other vertex. Single sink: from every vertex to one vertex t. Source-sink: from one vertex s to another vertex t. All pairs: from every vertex to every other vertex.

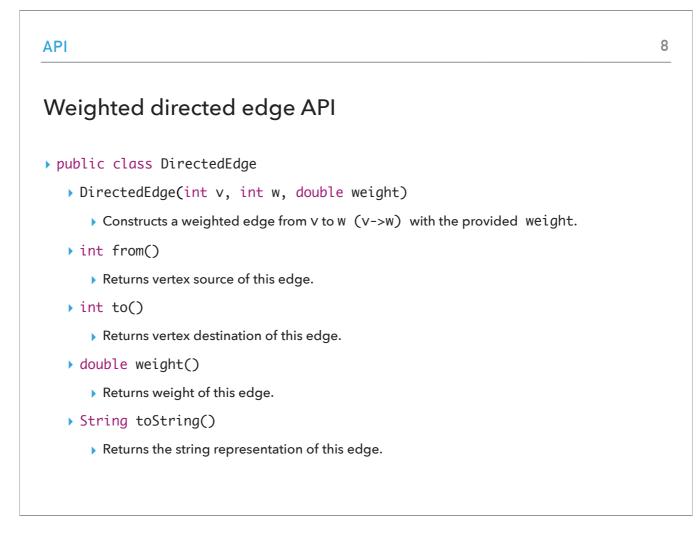
What version do you think your navigation app follows? That's right, source-sink, for the most part.



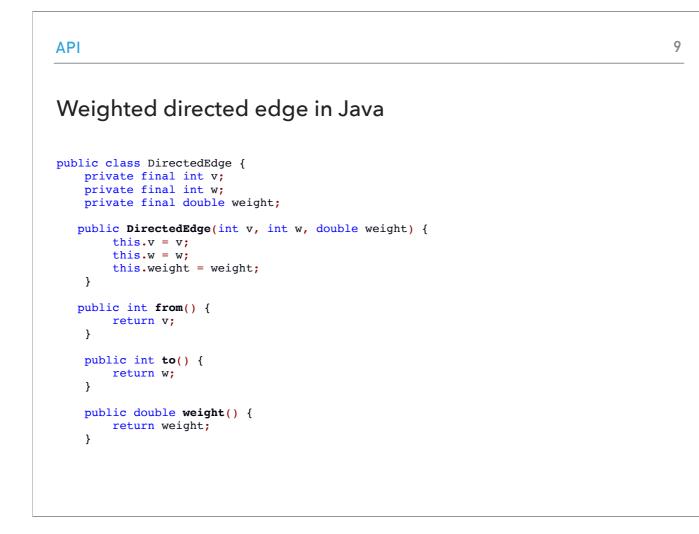
Let's state some assumptions for our shortest paths problem. Not all vertices need to be reachable (remember reachability refers to the ability to get from one vertex to another within a graph). Also we will assume that weights are non-negative (although there are algorithms that can handle negative weights). And finally, although shortest paths are not necessarily unique they have to be simple (a path in a graph which does not have repeating vertices).



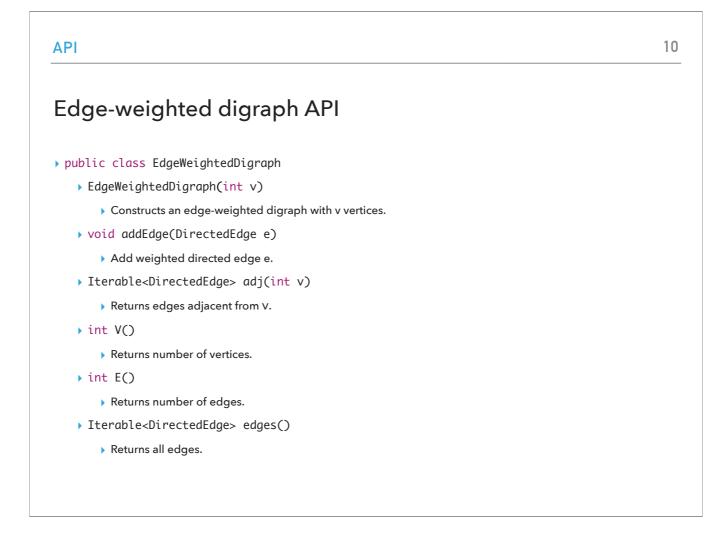
How would we go about modeling this problem?



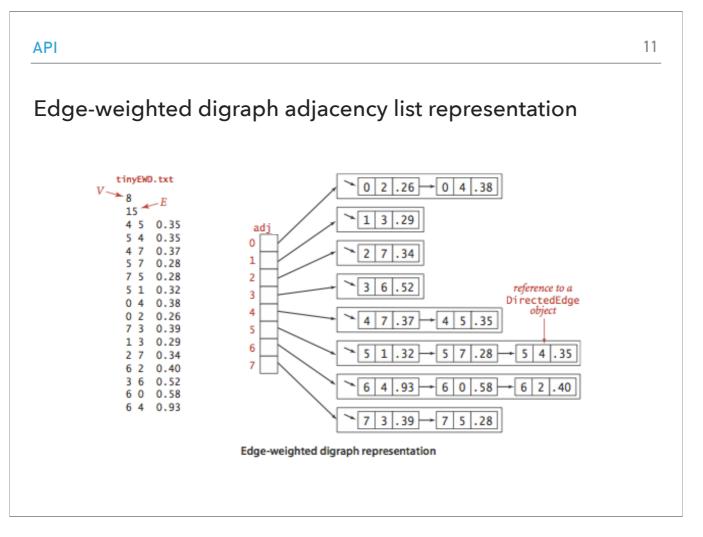
We will need to introduce the concept of a directed edge, and we will do so with a DirectedEdge class. Its constructor will create a weighted edge from v to w (v->w) with the provided weight. We should have a weight of getting the source and destination of the edge as well as its weight. And it would be convenient to have a string representation of the edge.



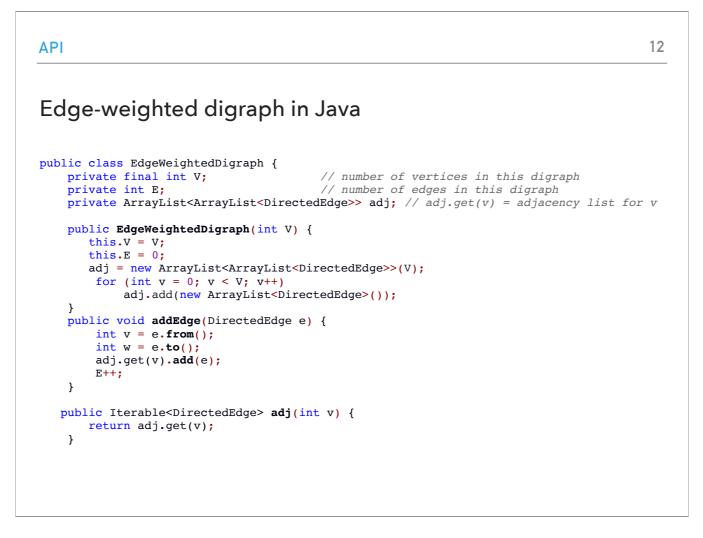
This is truly simple to code!



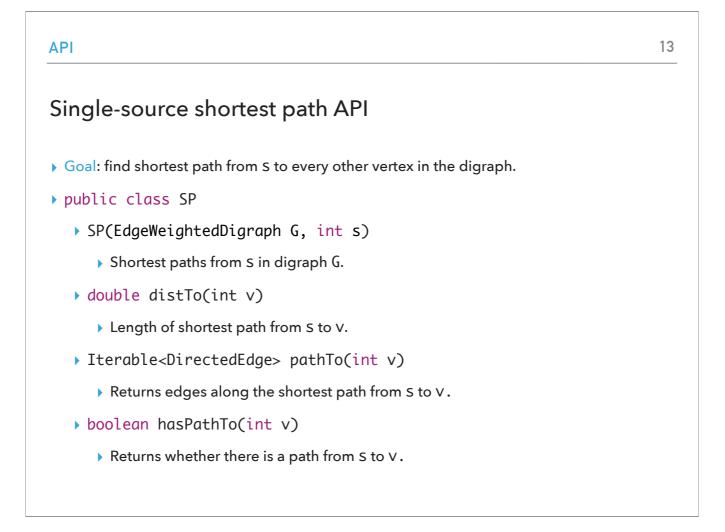
Now we can use the concept of a weighted directed edge to specify the API for an edge-weight digraph. We would need to provide a way of constructing such a digraph with v vertices. A way to add weighted directed edges as well as get back edges adjacent to v or all edges.



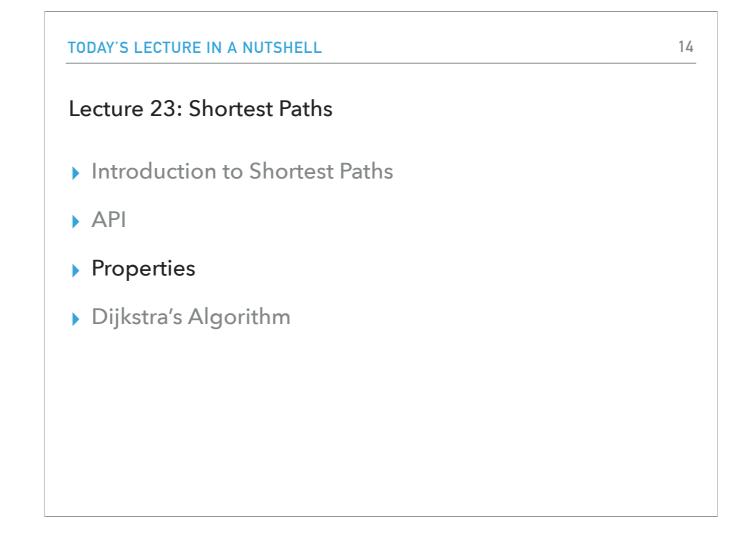
We will again follow the adjacency list representation for edge-weighted digraphs, where now we will hold references to a DirectedEdge object.



And that's how we would code such a class in Java. You will notice that the only difference from undirected and directed graphs is that we are working with DirectedEdges instead of Integers.



That brings us to the API for the single-source shortest path problem which states that our goal is to find the shortest path from s to EVERY other vertex in the digraph. We can imagine a class SP (for shortest paths) whose constructor takes an edge weighted digraph and the index of the starting vertex and calculates the shortest paths from s to every other vertex in the digraph. Some convenient methods to provide would be given a vertex, what is the length of the shortest path from s to that vertex, as well as a method that returns edges along such a path and a method that confirms whether such a path exists at all.



Now let's talk about some of the properties of the shortest paths problem.



Data structures for single-source shortest paths

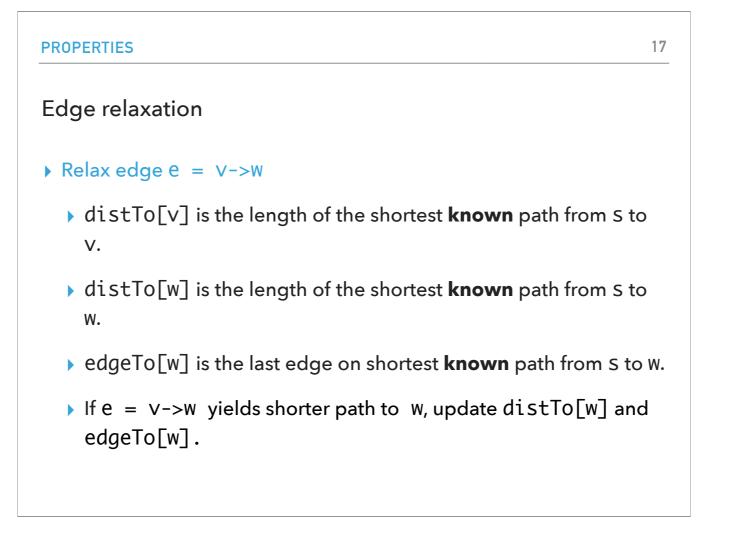
- Goal: find shortest path from S to every other vertex in the digraph.
- Shortest-paths tree (SPT): a subgraph which will be a directed tree rooted at S which will contain all the vertices reachable from S and every tree path in the SPT is a shortest path in the digraph.
- Representation of shortest paths with two vertex-indexed arrays.
 - Edges on the shortest-paths tree: edgeTo[v] is the last edge on a shortest path from S to V.
 - Distance to the source: distTo[v] is the length of the shortest path from S to V.

In our effort to find the shortest path from s to every other vertex in the digraph, we will calculate the shortest-paths tree (or SPT for short), which will be a directed tree rooted at s which will contain all the vertices reachable from s and every tree path in the SPT is a shortest path in the digraph.

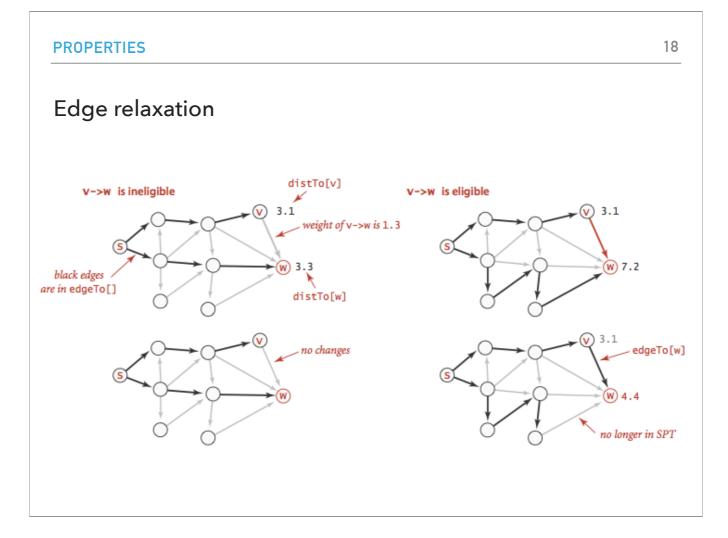
We will keep two vertex-indexed arrays. edgeTo will keep the edges on the shortest-paths tree, with edgeTo[v] being the LAST edge on the shortest path from s to v, and disco which will keep the distance to the source, that is distTo[v] will be the length of the shortest path from s to v.

				edge-weighted digra		
PROPERTIES			0.35	16		
			0.35			
			0.37			
			0.28			
			0.32			
<pre>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); } </directededge></directededge></directededge></pre>			0.38			
			0.26			
			0.39			
return path;		1->3	0.29			
}			0.34			
			0.40			
			0.52			
			0.58			
		0-24	0.55			
		edgeTo[]	distTo[]			
	0	null	0			
$ (1) \rightarrow (3) $	1	5->1 0.32	1.05			
Start and	2	0->2 0.26	0.26			
	3	7->3 0.39	0.99			
	4	0->4 0.38	0.38			
	5	4->5 0.35	0.73			
(4) (6)	6	3->6 0.52	1.51			
	7	2->7 0.34	0.60			

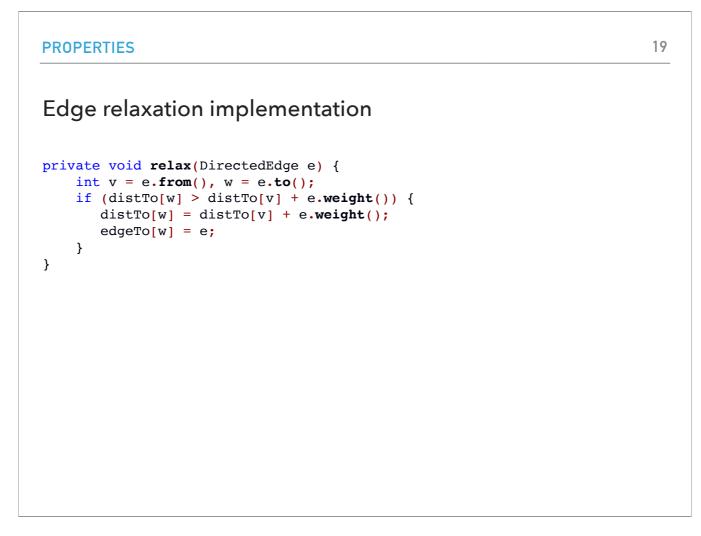
For example, here, we have an edge weighted digraph and you can see the edgeTo and distTo contents for all shortest paths from 0. To help us out return back a full shorts path from 0 to any vertex, we will use a pathTo method that uses a stack to give us back the edges from a destination vertex to the source vertex s in an iterable structure.



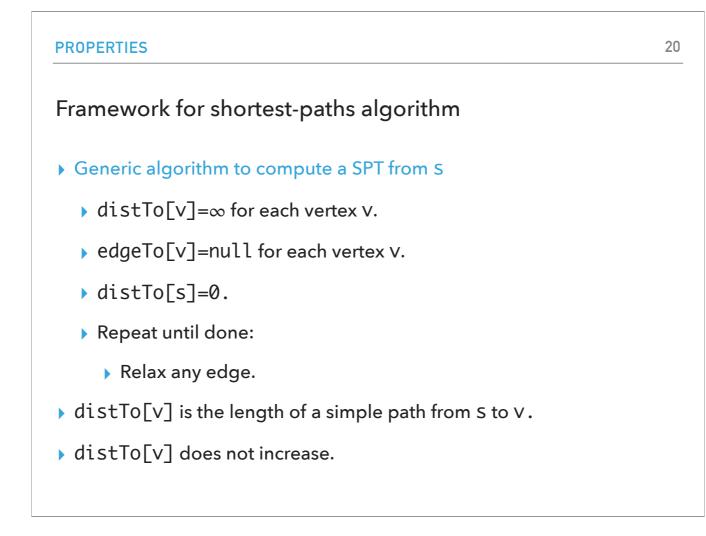
To calculate the contents of distTo and edgeTo, we will need the concept of edge relaxation. Remember, distTo[v] will store the length of the shortest path from s to v SO FAR. And similarly, edgeTo[w] will be the last edge of the shortest path from s to w SO FAR. Our ultimate goal, will be that when we are done with our algorithm, we will have found the overall shortest path. To get there, every time we encounter an edge e from v to w, we will ask whether it yields a shorter path from s to w. If yes, we will need to update distTo[w] and edgeTo[w].



There are two possible outcomes of an edge-relaxation operation. Either the edge is ineligible (as in the example at left, where the distTo[w] so far is 3.3 and going from s to w through v would actually increase the weight to 3.1+1.3=4.4) and no changes are made, or the edge v->w leads to a shorter path to w (as in the example at the right, where the best known path from s to w so far costs us 7.2, but if we were to go through v we would instead pay 4.4 which is cheaper) and we updated edgeTo[e] and distTo[e]

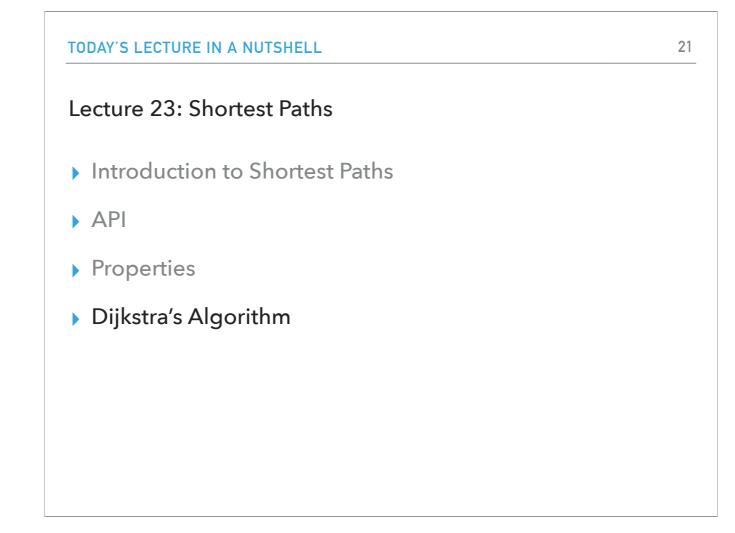


Implementing edge relaxation is extremely simple. Given an edge, we get the source and destination vertices and we update the distTo and edgeTo arrays for the destination only if the path through the source is shorter.

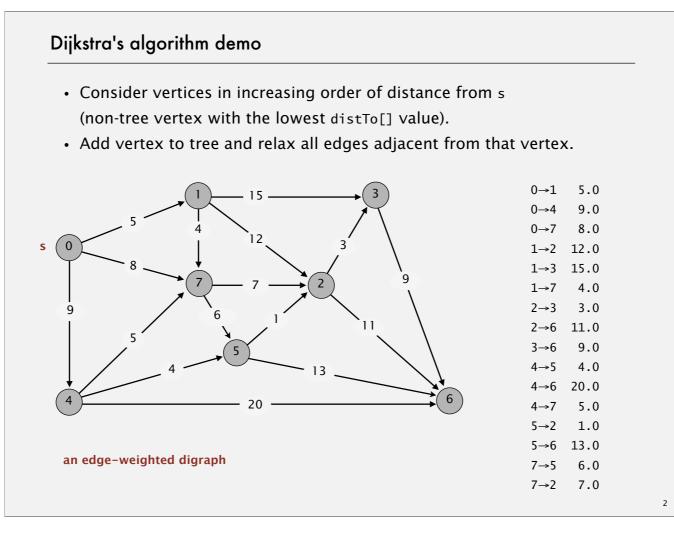


Putting it all together, it's not hard to imagine a generic algorithm to compute a SPT from s. We will initiate distTo for every vertex as a very large number, which here I will represent as infinity, and edgeTo for every vertex as null. The distTo[s] will be obviously 0 :)

We will repeatedly relax any edge until no change is registered. Remember, distTo[v] is the length of a simple path from s to v and distTo[v] does not increase.

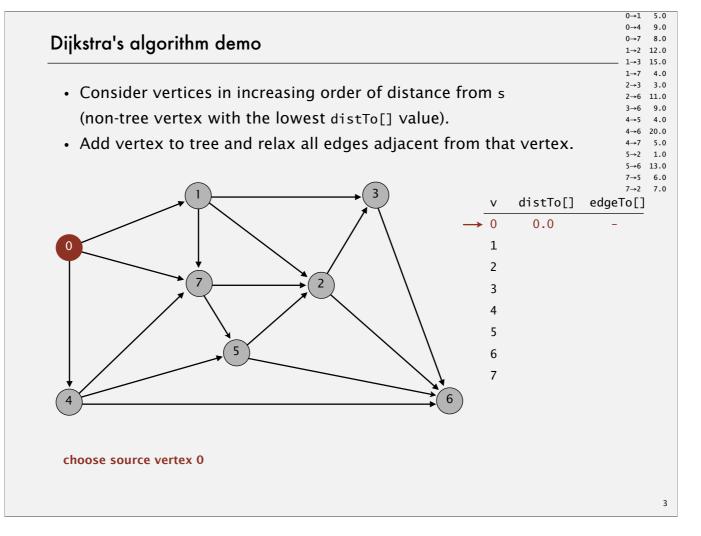


All that can be done systematically using Dijkstra's algorithm.

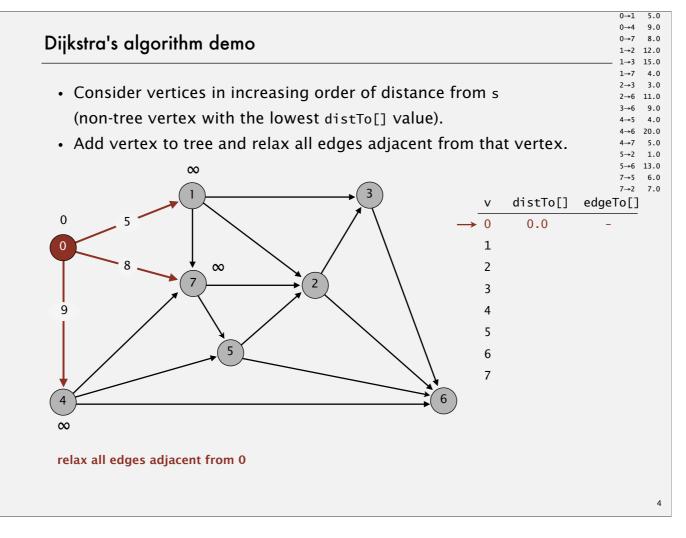


The algorithm considers vertices in increasing order of distance from s (let's say for this graph, vertex 0) by looking into the lowest disco values for vertices that are not part of the SPT. It then adds the vertex to the SPT and relaxes all edges adjacent from that vertex.

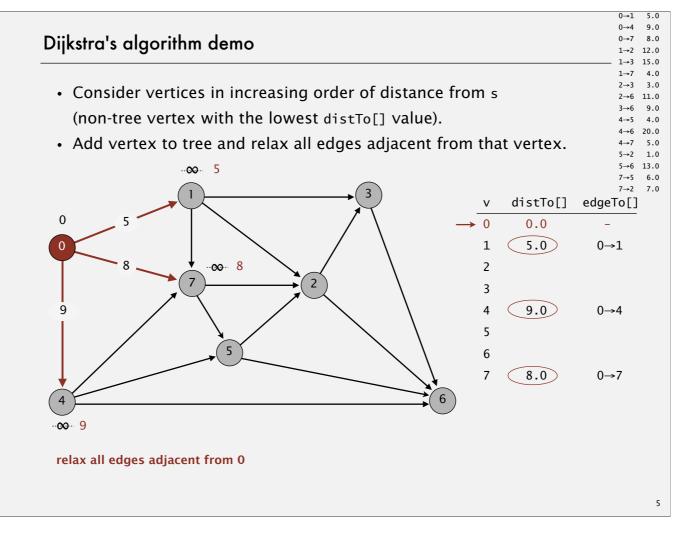
Let's take this graph as an example. On the right, you can see the edges and weights which are also indicated on the graph itself.



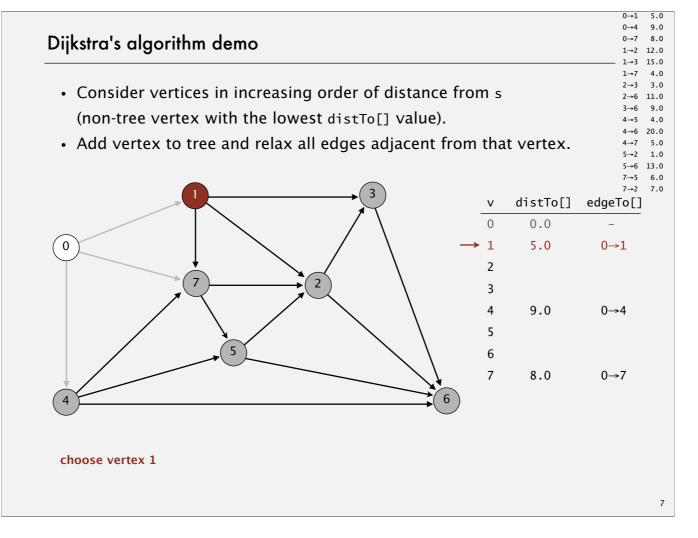
We start at the source vertex 0, add the vertex to the SPT (marked in red), and set the distTo[0] as 0.0 and the edgeTo[0] as null.



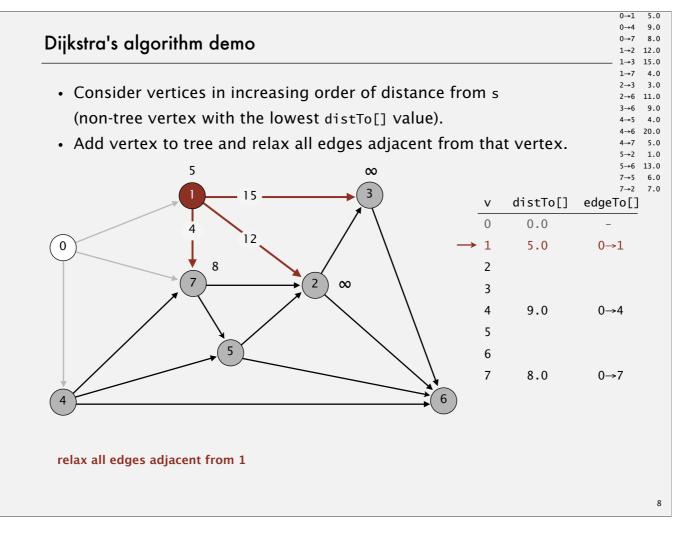
We then consider all adjacent vertices to 0 and relax them. Initially, all have been initialized to infinity. So the distance to them from 0 will be updated since they can all be relaxed.



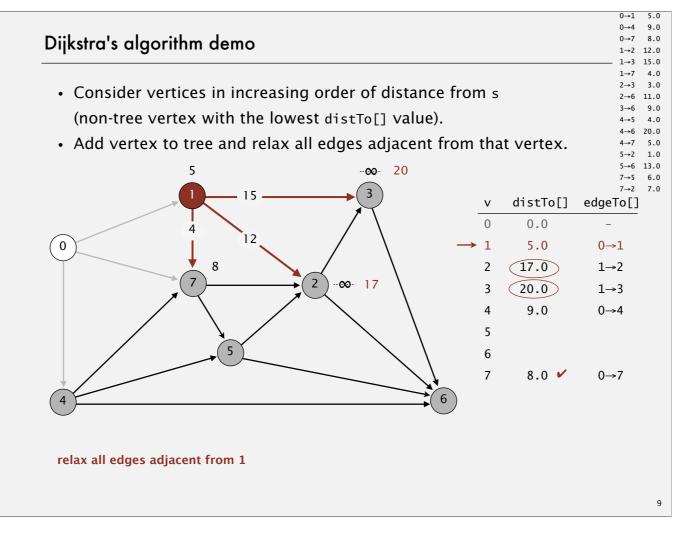
We will be updating the table as we go. Now we will consider all vertices that have not been added to the SPT yet (everything but 0 so far) and pick the one with the smallest distance to s.



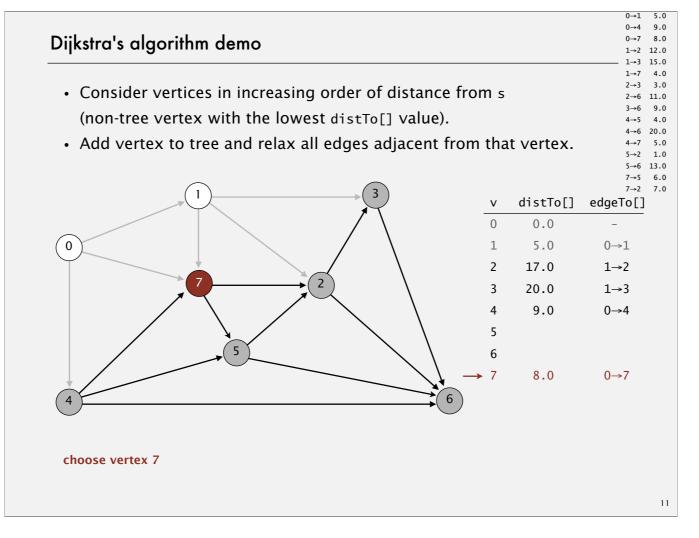
That would be vertex 1 with a total distance of 5.



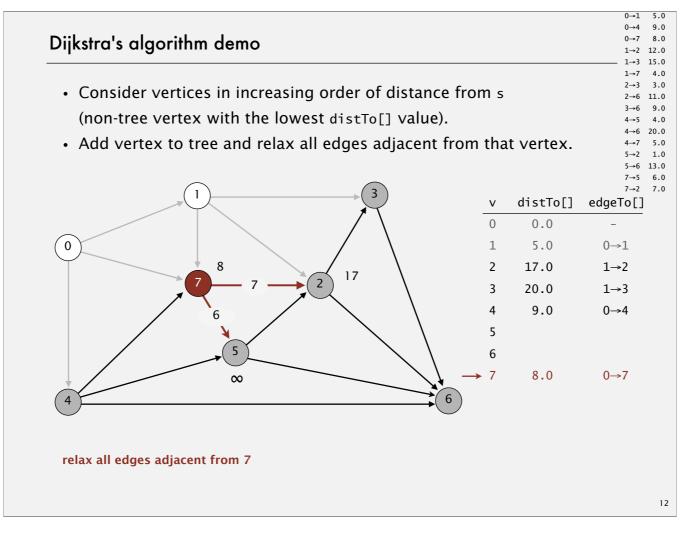
We can add 1 to the SPT. 1's adjacent vertices are 2, 3, and 7. We will need to relax them by comparing their current distance (infinity, infinity, and 8, respectively) with the cost we would have to pay if we were to go from 0 to them through 1 (i.e. 5+15=20, 5+12=17, and 5+4=9).



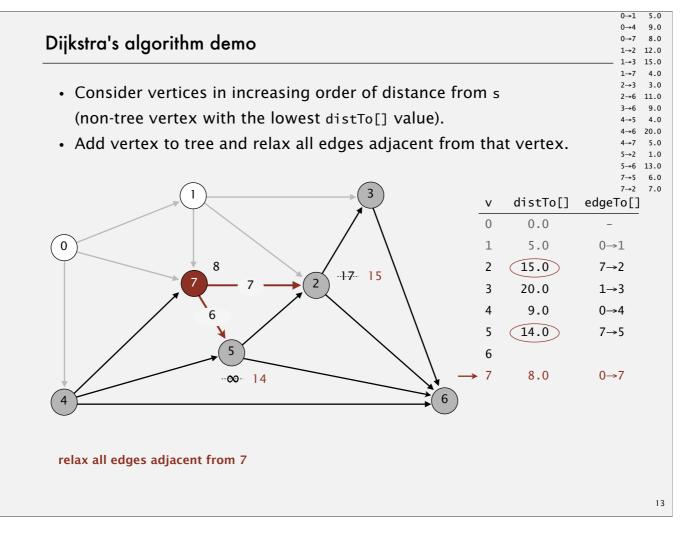
We will update the table for vertices 2 and 3 but not for 7 since paying 9 would be higher than the current cost of going directly to it (8). We will consider all vertices that are not on SPT yet (all except for 0 and 1) and pick the one with the shortest distance.



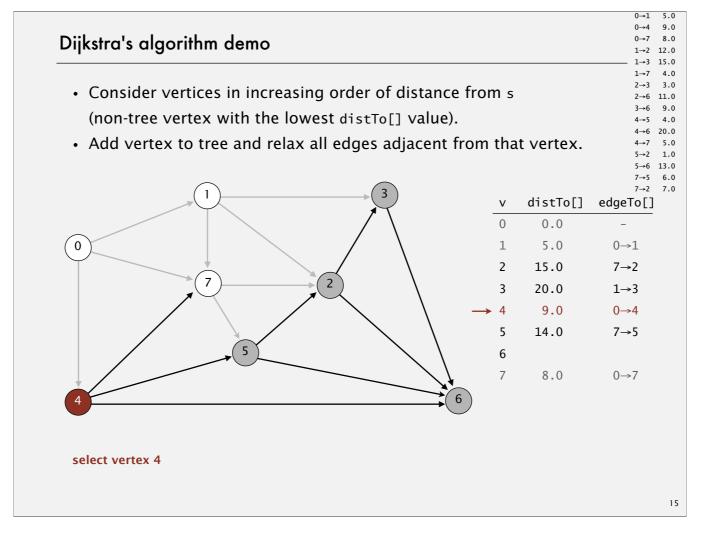
That would be vertex 7 with a cost of 8. Don't forget to update edgeTo as the last edge to the best known shortest path so far from 0 to that edge.



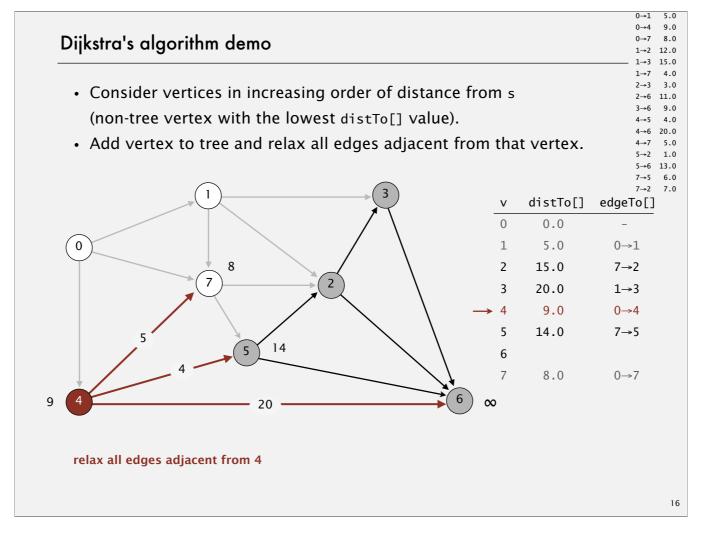
Consider the adjacent vertices of 7. These would be 2 and 5 with a cost of 8+7=15 and 8+6=14.



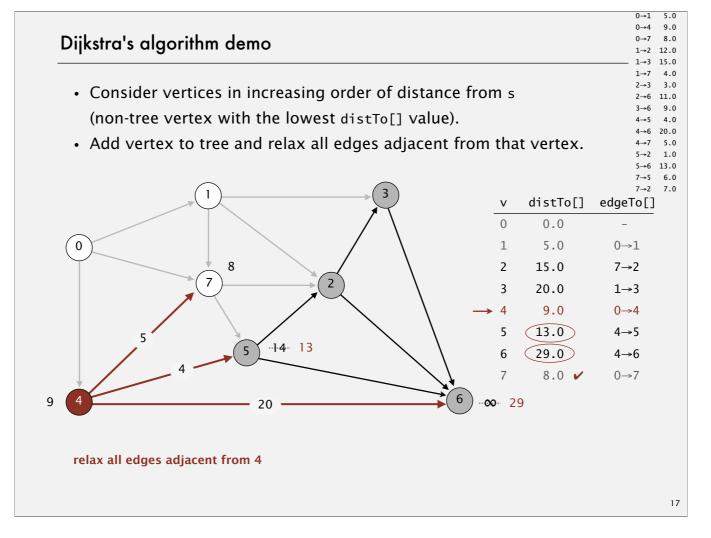
We will update both vertices with the shorter new total weights. We are ready to move on to the next vertex that is not part of the SPT and has the smallest distance.



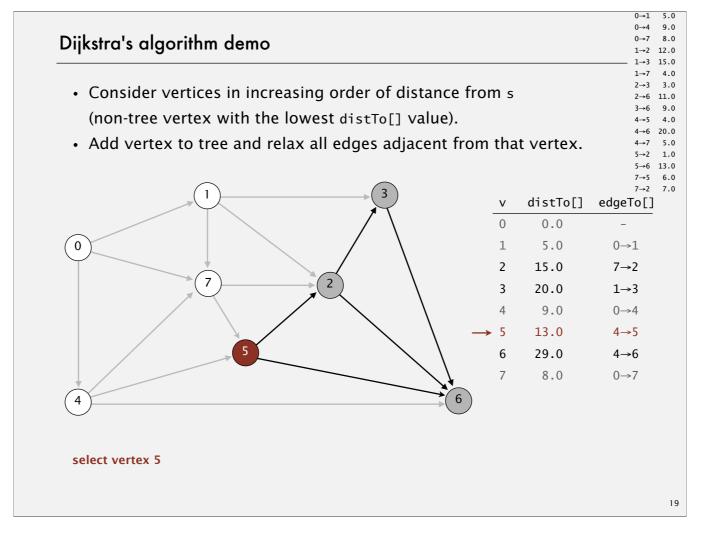
That would be vertex 4 for a cost of 9.



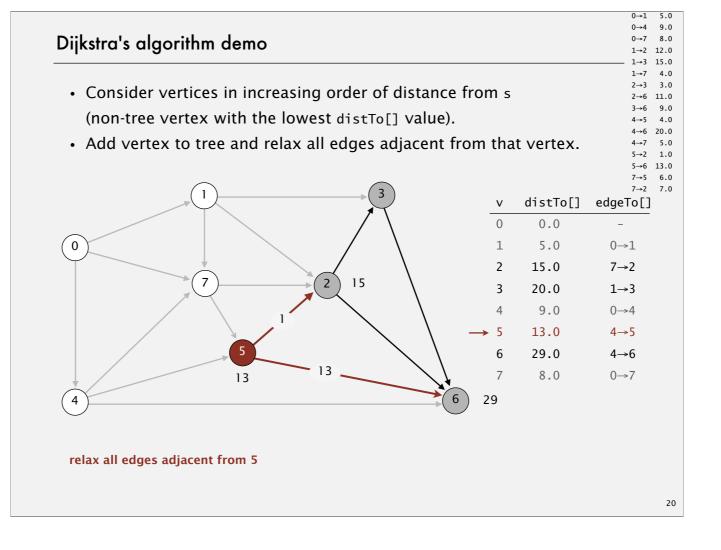
4's adjacent vertices are 5, 6, and 7. If we were to go from 0 to them through 4, we would pay 9+4=13, 9+20=29, and 9+5=14, respectively.



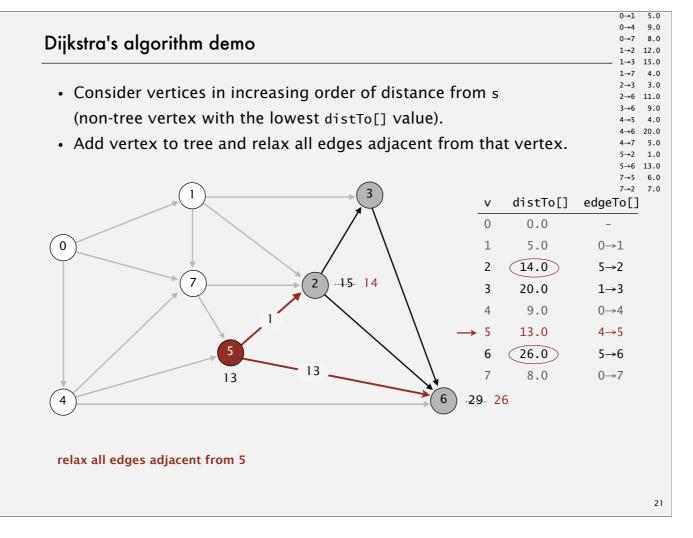
Which means we will update only vertices 5 and 6 (since the cost to 7 is higher than the existing 8). Vertex 4 is now part of the SPT.



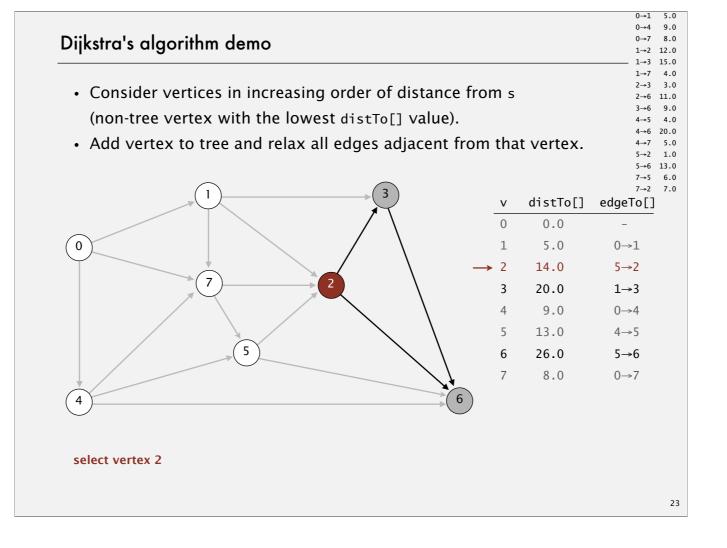
The next vertex would be 5.



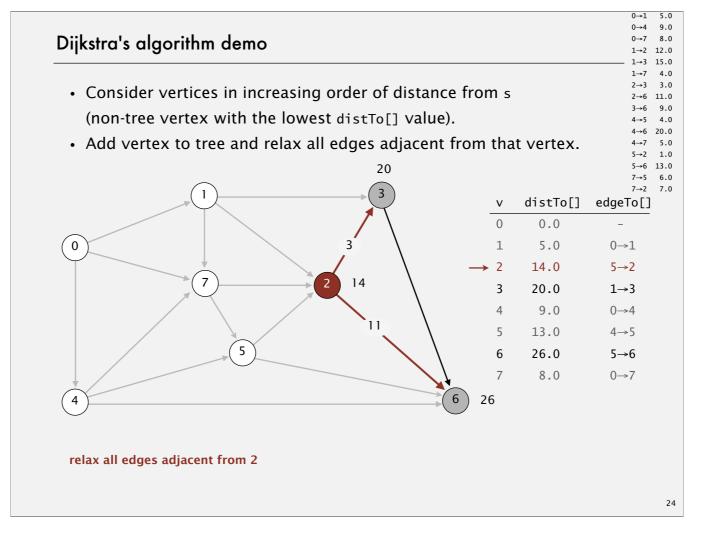
Its adjacent vertices are 2 and 6



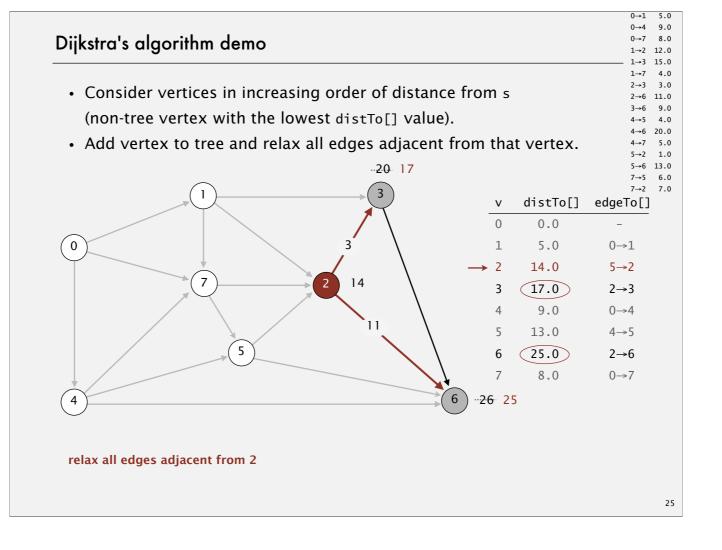
and both are relaxed to new total distances.



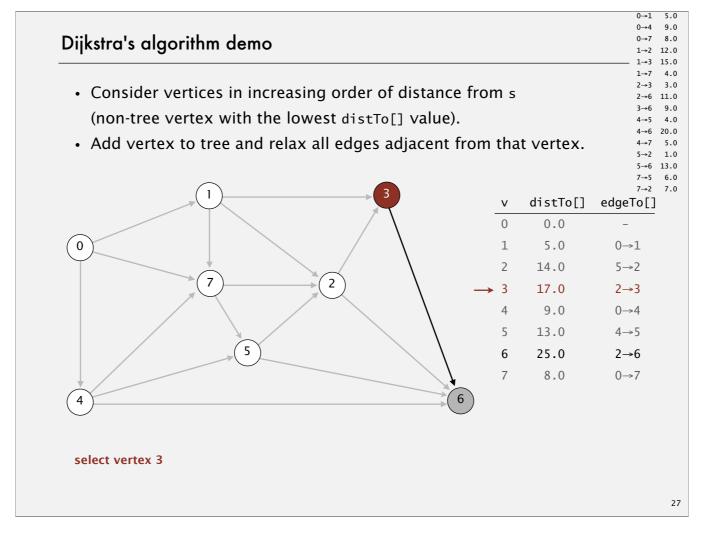
We then proceed with vertex 2.



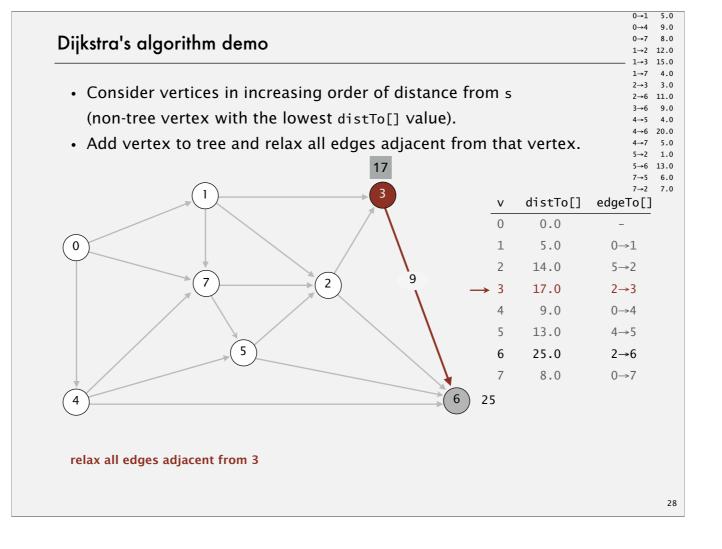
its adjacent vertices are 3 and 6



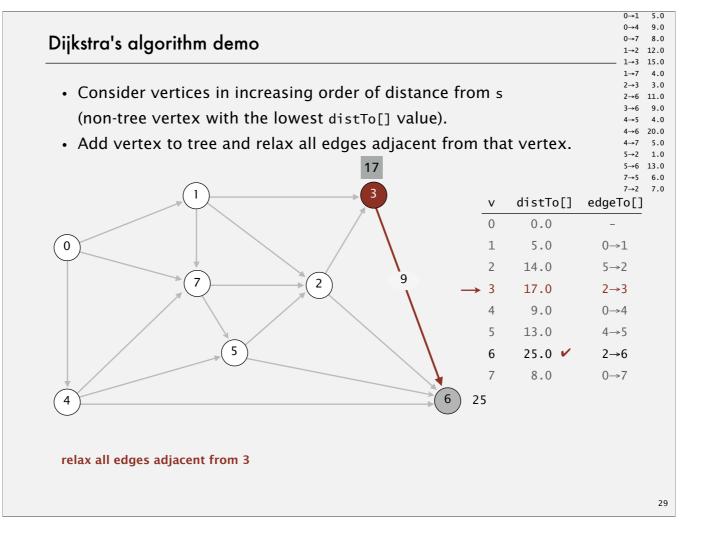
and are both relaxed.



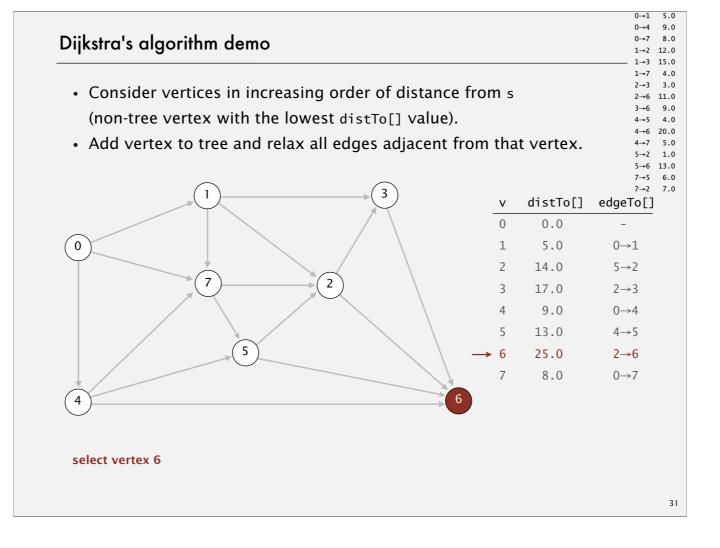
next vertex is 3



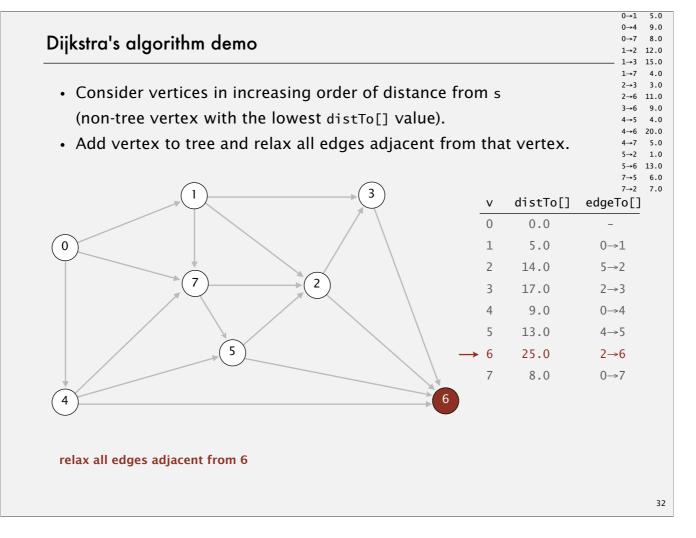
it has only one adjacent vertex, 6,



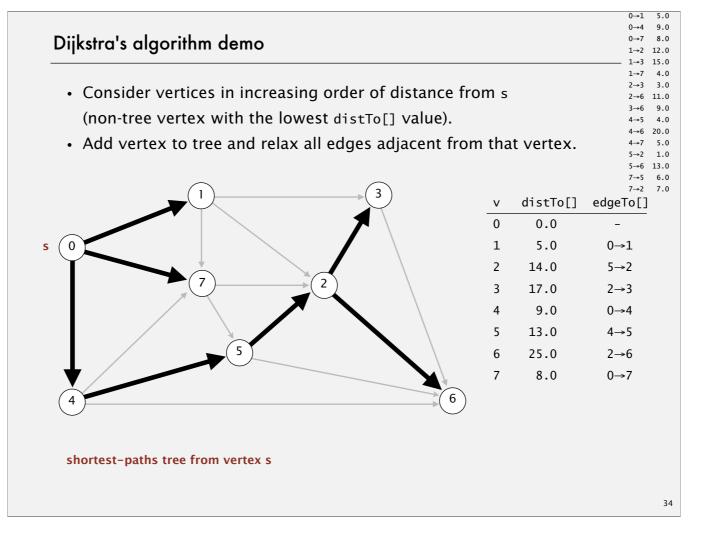
which won't be relaxed because the cost to go to it from 0 through 3 would be higher than go through 2.



Last vertex is 6



which happens to not have any adjacent vertices.



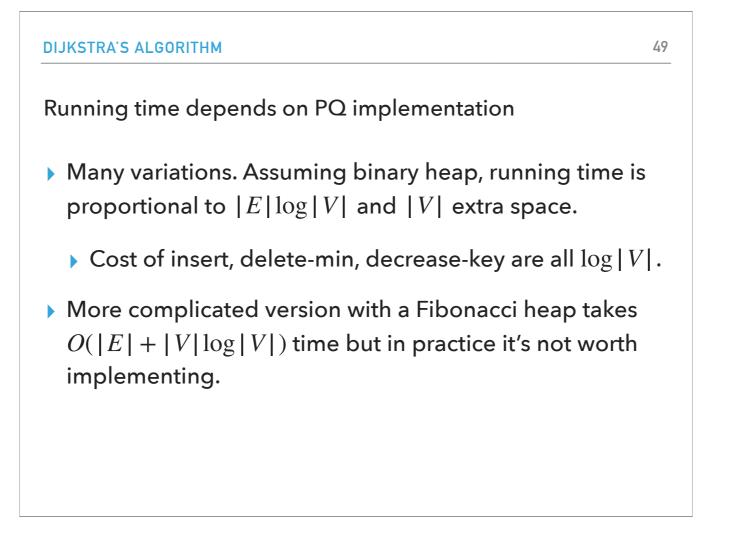
Here's our SPT from vertex s and the final arrays distTo and edgeTo.

Indexed min-priority queue (Section 2.4 in reco	mmended textbook
Associate an index between 0 and n-1 with each key in a priority queue.	
Insert a key associated with a given index.	
Delete a minimum key and return associated index.	
Decrease the key associated with a given index.	
public class IndexMinPQ <key comparable<key="" extends="">></key>	
<pre>> IndexMinPQ(int n)</pre>	
Create indexed PQ with indices 0,1,n-1	
<pre>void insert(int i, Key key)</pre>	
Associate key with index i.	
<pre>int delMin()</pre>	
Remove a minimal key and return its associated index.	
<pre>void decreaseKey(int i, Key key)</pre>	
Decrease the key with index i to the specified value.	

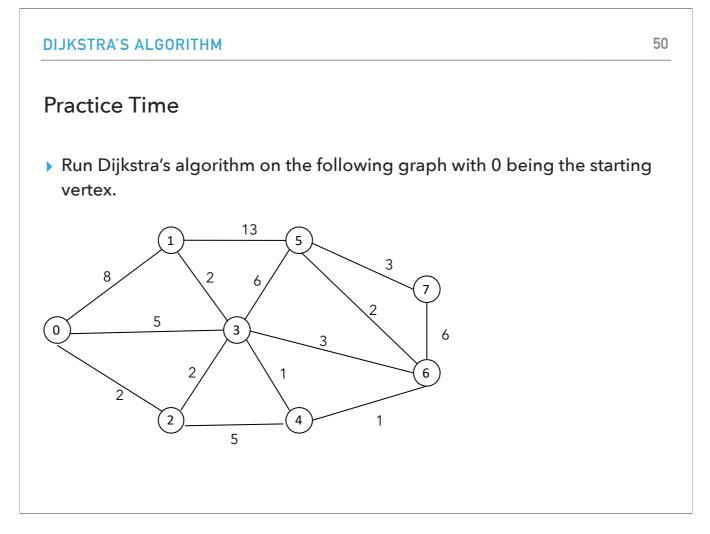
How do we go about implementing Dijkstra's algorithm? We will use a min priority queue (essentially a min-heap whose root is the minimum value) and work with inserting deleting and decreasing keys.



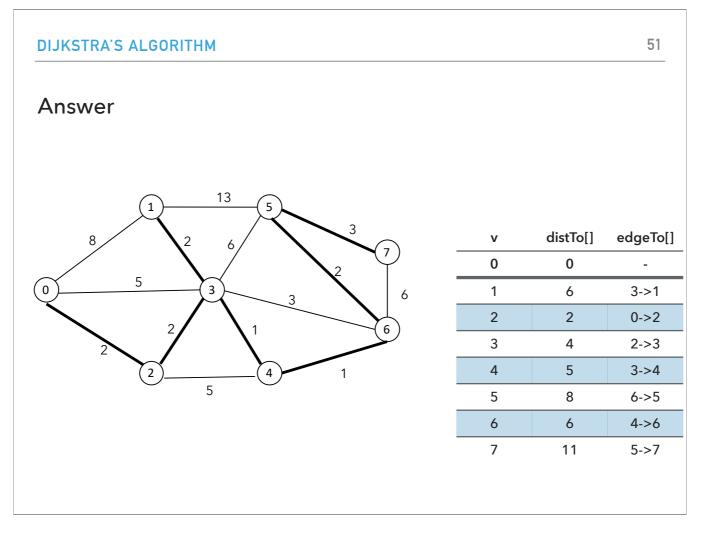
We won't really focus on the implementation (we used to have an assignment on this), but you can see that the implementation of Dijkstra's algorithm is not hard.



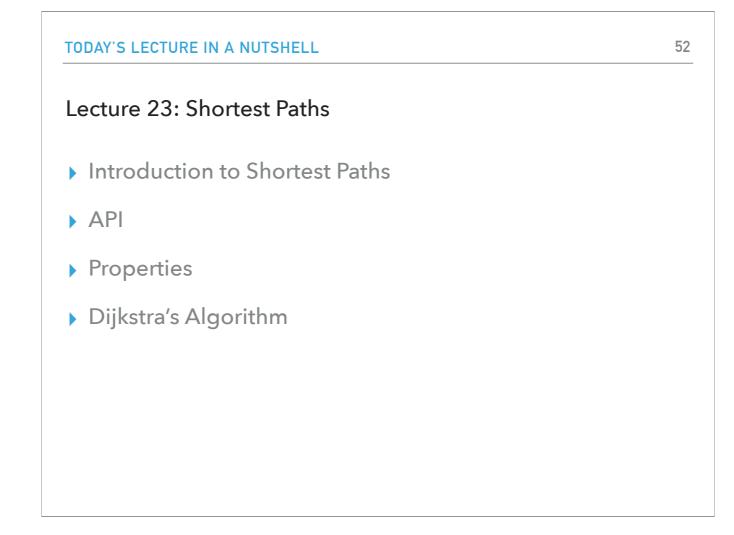
We won't do the analysis (stay tuned for CS140) but assuming you use a binary heap, the running time will be proportional to E log V and you will need V extra space (the cost for insertion, deletion of min and decreasing a key are all logarithmic). There is a fancier version that uses a Fibonacci heap and makes the algorithm faster but it's not worth implementing.



Let's see whether we understand how the algorithm works. Run Dijkstra's algorithm on the following graph with 0 being the starting vertex.



You should have ended up with this SPT and table.

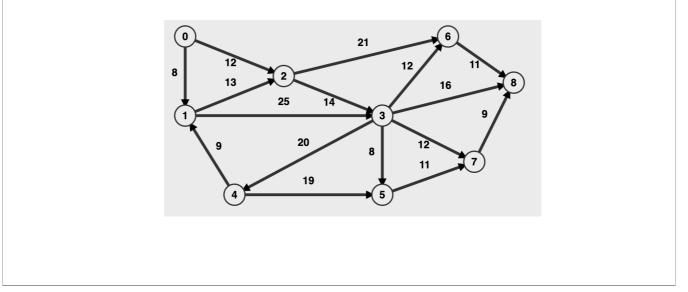


And that's all for today where we saw how to solve the shortest paths problem using Dijkstra's algorithm.

ASSIGNED READINGS AND PRACTICE PROBLEMS	53
Readings:	
Recommended Textbook: Chapter 4.4 (Pages 638-676)	
• Website:	
https://algs4.cs.princeton.edu/44sp/	
Visualization	
https://visualgo.net/en/sssp	

Problem

 Run Dijkstra's algorithm on the following graph with 0 being the starting vertex.



55

Answer

 Run Dijkstra's algorithm on the following graph with 0 being the starting vertex.

