

Lecture Note 3

CSCI 6470 Algorithms (Fall 2024)

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Chapter 3. Algorithms on graphs

Topics to be discussed:

- ▶ Basics and representations of graphs
- ▶ Depth-first search and applications
- ▶ Shortest path algorithms
- ▶ priority queue

1. Fundamentals of graphs

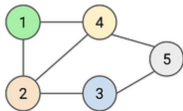
Terminologies:

- vertex, edge, graph, degree, neighbor, weight, directed edge,
- di-graph, subgraph, tree, path, cycle,
- connected component, strongly connected component,
- complete graph, planar graph, non-planar graph, bi-partite graph

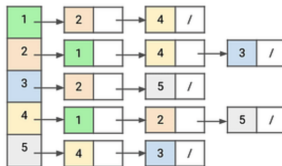
1. Fundamentals of graphs

Computer representations of graphs:

adjacency list, adjacency matrix



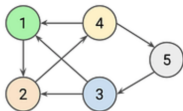
Undirected Graph



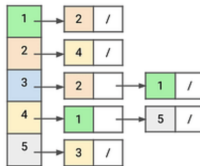
Adjacency List Representation

	1	2	3	4	5
1	0	1	0	1	0
2	1	0	1	1	0
3	0	1	0	0	1
4	1	1	0	0	1
5	0	0	1	1	0

Adjacency Matrix Representation



Directed Graph



	1	2	3	4	5
1	0	1	0	0	0
2	0	0	0	1	0
3	1	1	0	0	0
4	1	0	0	0	1
5	0	0	1	0	0

1. Fundamentals of graphs

Recursive definition for trees

- set pair $(\{x\}, \emptyset)$ is a **tree**;
- if (V, E) is a **tree**, vertex $u \in V$, and vertex $v \notin V$, then $(V \cup \{v\}, E \cup \{(v, u)\})$ is a **tree**.

Trees, created with these rules, are without a root. But the first vertex created can be designated as the root.

But why would non-biological trees need a root?

1. Fundamentals of graphs

Recursive definition for graphs

- set pair $(\{x\}, \emptyset)$ is a **graph**;
- if (V, E) is a **graph**, subset $U \subseteq V$, and vertex $v \notin V$, then (V', E') is a **graph**, where
$$V' = V \cup \{v\}, E' = E \cup \{(v, u) : u \in U\}.$$

Proper definitions of graphs may incur some structural views on graphs and help solve various computational problems on graphs.

2. Depth-first search

Based on the recursive definition, a given graph (V', E') can be decomposed as

- (1) a subgraph (V, E) ,
- (2) a vertex $v \in V' - V$ (also written as $V' \setminus V$);
- (3) a subset $U \subseteq V$;
- (4) $\forall u \in U$, edges $(v, u) \in E' - E$ (also written as $E' \setminus E$).

Graph traversal by exploiting the recursive definition of graphs.

- traverse a graph:
 visit vertex v and then recursively **visit** u , for all $(v, u) \in E$.
- two different traversal methods: DFS and BFS,
 depending on which vertex is to visit next

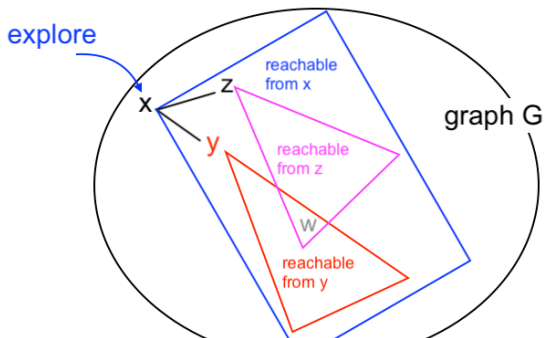
2. Depth-first search

Assume graph G that was created with x being any vertex

Explore all vertices reachable from vertex x :

```
function explore(G: graph; x: vertex)
```

1. `visited(x) = true;`
2. for each edge (x, y) in G
3. if not `visited(y)`
4. `explore(G, y);`



2. Depth-first search

Adding time stamps:

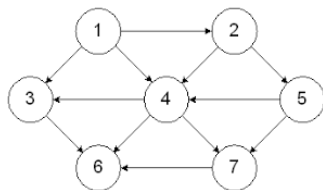
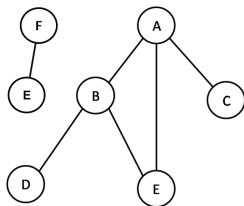
```
function explore(G: graph; x: vertex)
  1. visited(x) = true;
  2. pre(x) = time_stamp;           // pre-visit work
  3. time_stamp = time_stamp + 1;
  4. for each edge (x,y) in G
  5.   if not visited(y)
  6.     parent(y)=x;               // record tree edge
  7.     explore(G, y);
  8. post(x) = time_stamp;         // post-visit work
  9. time_stamp = time_stamp + 1;
```

Main body

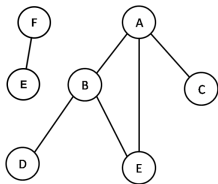
```
function main(G: graph);
  1. for every vertex v in G
  2.   visited(v) = false;
  3. for every vertex v in G
  4.   if not visited(v)
  5.     explore(G, v);
```

2. Depth-first search

Examples for DFS



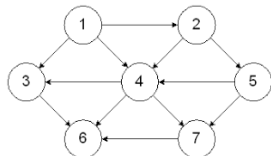
2. Depth-first search



- DFS on a graph yields a **DF-search tree**:
- properties of $\text{pre}(x)$ and $\text{post}(x)$ values
paired parentheses (look familiar?)
- type of edges in DFS tree
 - tree edges
 - back edges

2. Depth-first search

DFS on directed graphs



- types of edges in DFS tree
 - tree edges
 - back edges
 - forward edges
 - cross edges
- DFS on directed acyclic graphs (DAGs)

Theorem If there is a path $x \rightsquigarrow y$, then $\text{post}(x) > \text{post}(y)$

2. Depth-first search

DFS algorithm time complexity

- for loops in both functions `main` and `explore` visit every vertex (once), thus $O(|V|)$;
- for every vertex, all edges shared with its neighbors are checked at most twice (**why?**), thus $O(|E|)$;
- the total time is $O(|V| + |E|)$.

2. Depth-first search

Applications of DFS

- determine if the input graph is connected;
- determine if two given vertices are connected on input graph;
- determine if the input graph contains a cycle;
- find a topological order of vertices on the input DAG;
- find the strong connected components of the input graph;
- find single-source shortest paths on the input DAG;

All have time complexity: $O(|E| + |V|)$

2. Depth-first search

Determine if the input graph contains a cycle;

The graph has a cycle if, when encountered with edge (x,y) in function `explore`, (x,y) is a back edge, i.e.,

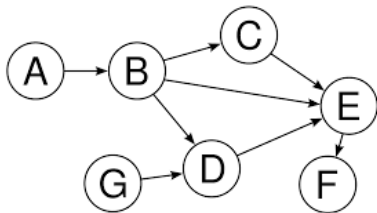
- for non-directed graphs,
 - (1) `visited(y) = true`, and
 - (2) `y \neq parent(x)`
- for directed graphs,
 - (1) `visited(y) = true`, and
 - (2) `pre(y) < pre(x)` and `post(x)` is not defined yet

2. Depth-first search

Topological Sort problem

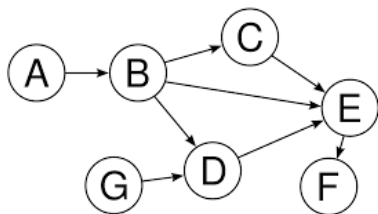
Input: directed acyclic graph $G(V, E)$,

Output: vertices of V in order: v_1, v_2, \dots, v_n such that
 $\forall i < j, (v_j, v_i) \notin E$.



Some topological sorted orders:

2. Depth-first search



- DFS can be used to solve this problem
remember the following?

Theorem If there is a path $x \rightsquigarrow y$, then $\text{post}(x) > \text{post}(y)$

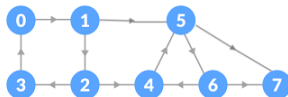
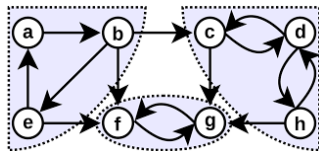
Conclusion: Reversed order of post values is a topological sort order.

2. Depth-first search

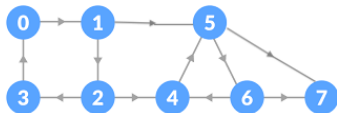
Strongly Connected Components (SCC) problem

Input: directed graph $G(V, E)$,

Output: strongly connected components for G .



2. Depth-first search



Idea:

2. Depth-first search

SCC algorithm:

- DFS on G ;
- generated G^T , transpose of G (reversed edges directions);
- DFS on G^T from vertex v with the highest $\text{post}(v)$ value;

2. Depth-first search

More about graph traversal algorithms

- non-recursive version of DFS; using (?)
- breadth first search (BFS); explores vertices in the order of their distance from the source vertex
- non-recursive BFS? using (?)
- recursive version ?
- time complexity

Taken-home exercises III(A)

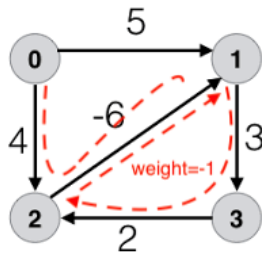
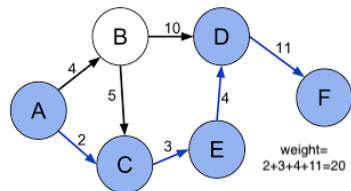
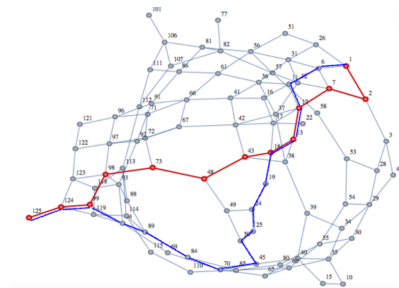
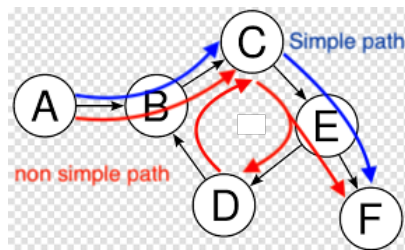
1. Run DFS on non-directed graphs (of at least 6 vertices). Show (a) DFS tree (forest), (b) time stamps $\text{pre}(v)$ and $\text{post}(v)$ for every vertex v ; (c) group edges from the searched graph into 2 categories.
2. DFS on non-directed graphs (of at least 6 vertices). Show (a) DFS tree (forest), (b) time stamps $\text{pre}(v)$ and $\text{post}(v)$ for every vertex v ; (c) group edges from the searched graph into 4 categories.
3. Assume DFS is run on a directed graph (vertices are searched in the alphabetical order if there are options). Record the 4 types of edges from the searched graph. Now if we remove directions of edges from the graph and run the DFS search. The search yields only two types of edges for the graph. What types of edges have forward and cross edges become respectively?
4. DFS algorithm visits every vertex (possibly more than once). Consider a modification of DFS on non-directed graphs so that it can compute the number of paths from the start vertex s to every other vertex in the graph. *Hint*: set a counter for every vertex which will be incremented by 1 every time the vertex is examined by the DFS.

Taken-home exercises III(A)

- 5 Would the modification to DFS in Question 4 be the same on directed graphs? Explain.
- 6 Modify DFS so that it can be used to determine if a graph contains a cycle. Try the modification for directed graphs and then for non-directed graphs.
- 7 Use DFS to solve the topological sort problem on directed acyclic graphs of at least 10 vertices.
- 8 Use DFS to solve the topological sort problem on a directed graph example that contains cycle. Explain where and why the method may fail.
- 9 DFS is called twice in STRONGLY CONNECTED COMPONENT (SCC) algorithm. What is the purpose of the first DFS call? What about the second DFS call?
- 10 Run the SCC algorithm on directed graphs of at least 10 vertices.

3. Shortest path problems

Paths on graphs:



3. Shortest path problems

Problems about paths on graphs:

- **Reachability** : given $G = (V, E)$, and vertices $s, t \in V$;
asked if there is a path: $s \rightsquigarrow t$, from s to t ;
- **Cycle**: given $G = (V, E)$,
asked if there is a cycle in the graph. **How exactly is this done?**
- **s - t shortest path**: given $G = (V, E)$, vertices $s, t \in V$;
find a shortest path $s \rightsquigarrow t$;
- **Single source shortest path**: given $G = (V, E)$,
 $\forall v \in V$, find a shortest path $s \rightsquigarrow v$;
- **All pair shortest path**: given $G = (V, E)$,
 $\forall u, v \in V$, find a shortest path $u \rightsquigarrow v$;
[To be discussed in the subject of dynamic programming]

3. Shortest path problems

Shortest Distance problem:

Input: di-graph $G = (V, E)$, lengths $w : E \rightarrow R$; source $s \in V$

Output: $dist(v)$, shortest distance s to every vertex $v \in V$.

- Is the following problem easier?

Given source s and target t , find a shortest path from s to t ;

- Subpath on a shortest path is a shortest path

if $p_{s,r,t} : s \rightsquigarrow r \rightsquigarrow t$ is a shortest path from s to t ,

then subpath $p_{s,r} : s \rightsquigarrow r$ is a shortest path from s to r . [proof?]

3. Shortest path problems

Shortest Distance problem:

Input: di-graph $G = (V, E)$, lengths $l : E \rightarrow R$; source $s \in V$

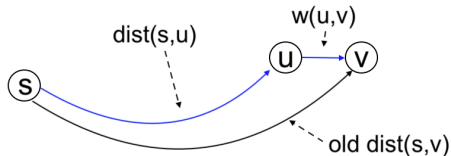
Output: $dist(v)$, shortest distance s to every vertex $v \in V$.

- Finding a shortest path from s to t implies finding shortest paths from source s to all other vertices.
 - shortest distance from s to one vertex may impact on another;
- The question becomes “which vertices should be considered before others in computing their shortest distances from s ?”
- That is: is there an order of vertices whose shortest distances can be computed correctly and efficiently?

3. Shortest path problems

Shortest Distance problem:

- updating path distances,



$$\text{dist}(v) = \min \begin{cases} \text{dist}(v) \\ \text{dist}(u) + w(u, v) \end{cases}$$

called **relaxation of edges**

3. Shortest path problems

Shortest Distance problem on 3 different types of graphs:

- G is a DAG: relying on DFS
edges are relaxed according to topo sort order;
- G does not contain negative edges: Dijkstra's algorithm
edges (u, v) are relaxed only after $dist(u)$ is shortest;
- G may contain negative cycles: Bellman-Ford algorithm
edges are relaxed in an arbitrary order

3. Shortest path problems

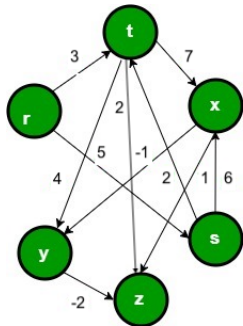
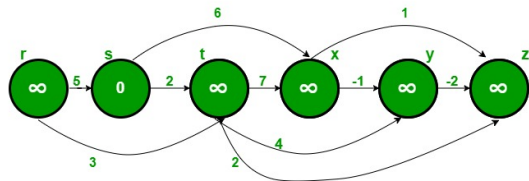
Input: DAG $G = (V, E)$, edge lengths $l : E \rightarrow R$, and $s \in V$,

Output: $\forall u \in V$, the smallest distance $\text{dist}(u)$ from s to u .

```
function dag-shortest-paths(G, l, s)
1. for all u in V
2.   dist(u) = infinity;
3.   prev(u) = nil; // predecessor of u in the path
4. dist(s) = 0;
5. topological sort V;
6. for all u in V in the sorted order
7.   for all edge (u, v) in E
8.     if dist(v) > dist(u) + l(u, v);
9.       dist(v) = dist(u) + l(u, v);
10.      prev(v) = u;
```

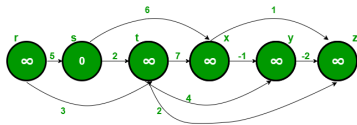
3. Shortest path problems

Shortest Distance problem on DAG

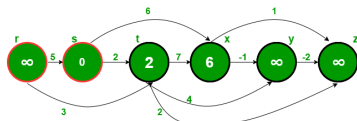


3. Shortest path problems

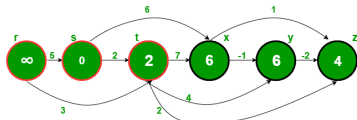
Shortest Distance problem on DAG



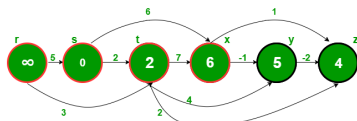
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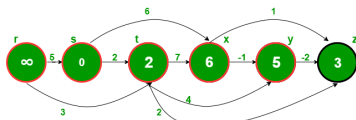
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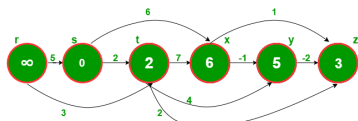
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(f)



(g)



(h)

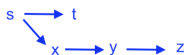
3. Shortest path problems

Shortest Distance problem on DAG

algorithm steps on the previous example

dist table								prev table					
r	s	t	x	y	z		step	r	s	t	x	y	z
inf	0	inf	inf	inf	inf		initial	nil	nil	nil	nil	nil	nil
inf	0	2	6	inf	inf		u = s	nil	nil	s	s	nil	nil
inf	0	2	6	6	4		u = t	nil	nil	s	s	t	t
inf	0	2	6	5	4		u = x	nil	nil	s	s	x	t
inf	0	2	6	5	3		u = y	nil	nil	s	s	x	y
inf	0	2	6	5	3		u = z	nil	nil	s	s	x	y

Shortest paths tree



shortest paths from s to all other vertices



3. Shortest path problems

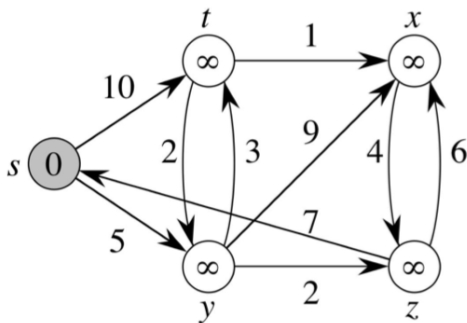
Shortest Distance problem: Dijkstra's algorithm on general directed graphs, without negative weights

Input: $G = (V, E)$, edge lengths $l : E \rightarrow R_{\geq 0}$, and $s \in V$,

Output: $\forall u \in V$, the smallest distance $\text{dist}(u)$ from s to u .

```
function Dijkstra(G, l, s)
1. for all u in V
2.   dist(u) = infinity;
3.   prev(u) = nil;
4. dist(s) = 0;
5. H = makequeue(V);
6. While H is not empty // body including lines 7 - 11
7.   u = dequeue(H);
8.   for all edges (u, v) in E
9.     if dist(v) > dist(u) + l(u, v);
10.      dist(v) = dist(u) + l(u, v);
11.      prev(v) = u;
12. return (prev)
```

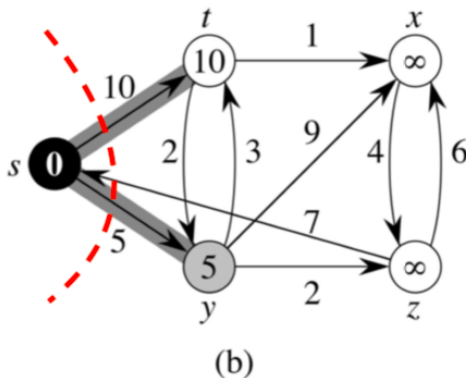
3. Shortest path problems



(a)

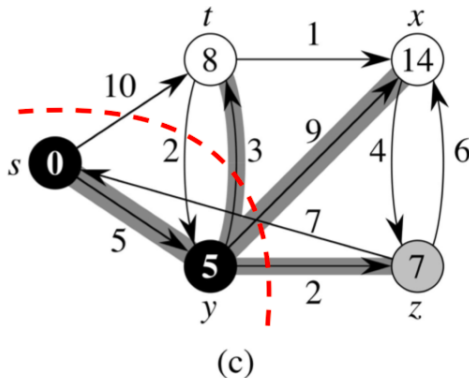
Note: the gray-colored vertex is to be dequeued while the black-colored vertices are in set S .

3. Shortest path problems



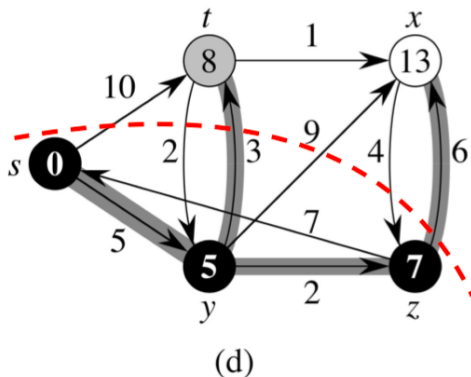
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3. Shortest path problems



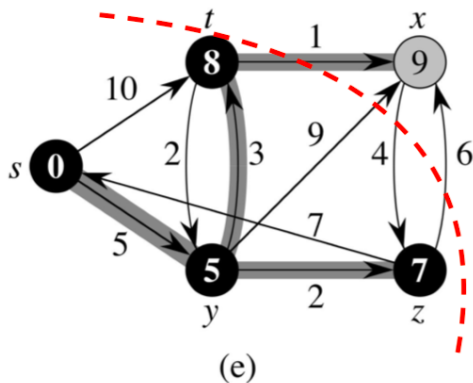
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3. Shortest path problems



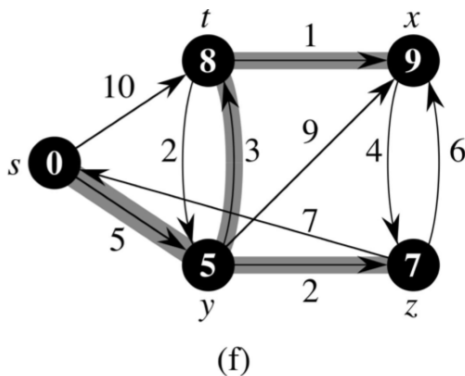
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3. Shortest path problems



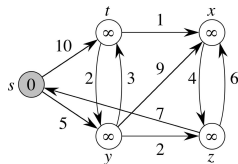
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3. Shortest path problems

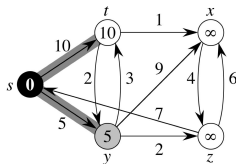


Note: the gray-colored vertex is to be dequeued while the black-colored vertices are in set S .

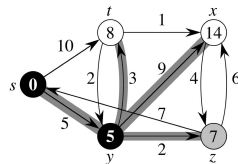
3. Shortest path problems



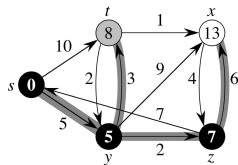
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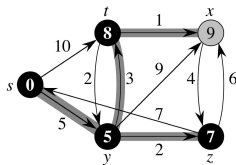
(b)



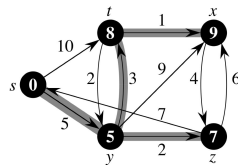
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(d)



(e)



(f)

Note: the gray-colored vertex is to be dequeued while the black-colored vertices are in set S .

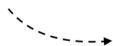
3. Shortest path problems

Shortest Distance problem on graphs without negative edges

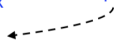
Dijkstra's algorithm steps on the previous example

dist table							prev table				
s	t	x	y	z		step	s	t	x	y	z
0	inf	inf	inf	inf		initial	nil	nil	nil	nil	nil
0	10	inf	5	inf		u = s	nil	s	nil	s	nil
0	8	14	5	7		u = y	nil	y	y	s	y
0	8	13	5	7		u = z	nil	y	z	s	y
0	8	9	5	7		u = t	nil	y	t	s	y
0	8	9	5	7		u = x	nil	y	t	s	y

Shortest paths tree



shortest paths from s to all other vertices



Taken-home exercises III(B)

1. Run algorithm dag-shortest-paths on DAGs of at least 5 vertices. Show changes of the `dist` and `prev` values when every vertex in the topological sort order is considered. **Put results in a table.**
2. What is the time complexity of algorithm dag-shortest-path in terms of $|V|$ and $|E|$ of the input DAG? Why?
3. Someone, when copying the pseudo code of dag-shortest-paths, was careless and swapped v with u in all places **except** `l(u,v)` in lines 7 - 10. Would the modified code be still correct to find shortest paths from the source s to all other vertices? Justify your answer.
4. Run algorithm Dijkstra on **non-directed graphs** of at least 6 vertices. Show (1) status of queue H , (2) **the vertex dequeued**, and (3) changes of `dist` and `prev` values for all vertices after every vertex is dequeued.
5. What is the time complexity of Dijkstra in terms of $|V|$ and $|E|$ of the input graph? Why?
6. Dijkstra may or **may not** work on graphs with negative edges. Show a small example of **non-directed** graph that contains one negative edge **but without negative cycles** on which Dijkstra would fail.
7. Vertices are considered in the topological order in dag-shortest-paths on DAGs. In what order are vertices dequeued in algorithm Dijkstra? Why does the order work on the graphs that Dijkstra handles?

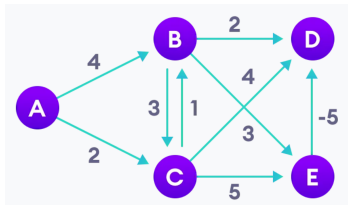
3. Shortest path problems

For graphs that may contain negative edges, edge relaxations cannot be done in specific order.

But every shortest path consists of at most $n - 1$ edges;
 $n - 1$ rounds of edge relaxations suffices

```
function Bellman-Ford(G=(V,E): graph; s: vertex)
1. for all u in V
2.   dist(u) = infinite;
3.   prev(u) = nil;
4. dist(s) = 0;
5. for k=1 to n-1
6.   for all edge (u,v) in E
7.     if dist(v) > dist(u) + l(u,v)      // l(u,v) is the
8.       then dist(v) = dist(u) + l(u,v); // weight of edge
9.       prev(v) = u;
10. for every an edge (u, v) in E
11.   if dist(u) + l(u,v) < dist(v)
12.     then return "G contains a negative cycle"
13. return (dist, prev)
```

3. Shortest path problems



The order of edges chosen to be relaxed at each round has impact on distances updated

Source: A

Round 1

Round 2

Round 3

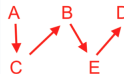
Round 4

A	B	C	D	E
0	∞	∞	∞	∞
0	4 (A) 3 (C)	2 (A)	6 (B) 2 (E)	7 (C)
0	3 (C)	2 (A)	1 (E)	6 (B)
0	3 (C)	2 (A)	1 (E)	6 (B)
0	3 (C)	2 (A)	1 (E)	6 (B)

Order of edges selected to relax : (A,B), (A,C), (B,C), (B,D), (B,E), (C,B), (C,D), (C,E), (E,D)

vertex within () is the prev vertex

Shortest-paths tree:



3. Shortest path problems

Bellman-Ford algorithm correctly detects negative cycles.

- Assume shortest path $p : s \rightsquigarrow v$.
- If after the $n - 1$ rounds of relaxations are done, $\exists(u, v)$ with $dist(u) + l(u, v) < dist(v)$
- then path $p : s \rightsquigarrow v$ consists of more than $n - 1$ edges; i.e., p contains at least $n + 1$ vertices, some vertex x occurs twice.

$$p : s \rightsquigarrow y \rightarrow x \rightsquigarrow x \rightarrow z \rightsquigarrow v$$

- the path cannot be shorter than $s \rightsquigarrow y \rightarrow x \rightarrow z \rightsquigarrow v$
unless cycle $x \rightsquigarrow x$ is negative.

3. Shortest path problems

Priority queue implementation using heap

what is a heap? physical implementation?

how to use a heap to realize a priority queue?

Time complexities of different shortest path algorithms

- dag-shortest-path: $O(|V| + |E|)$
- Dijkstra's: $O(|V| \log_2 |V| + E)$
- Bellman-Ford: $O(|V||E|)$.