

# Lecture 7

## Data Structures (DAT037)

Ramona Enache

(with slides from Nick Smallbone and  
Nils Anders Danielsson)

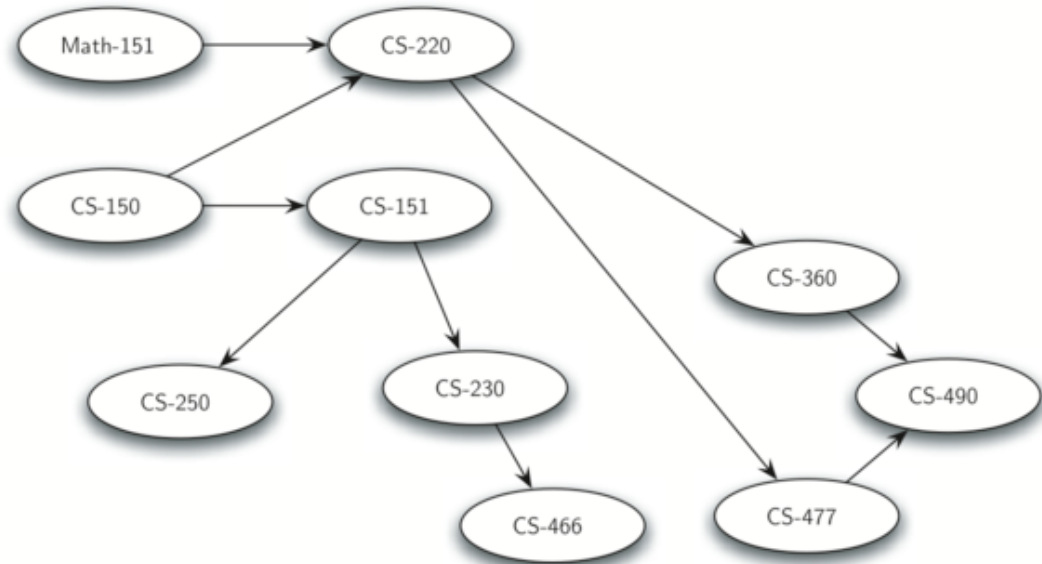
# Directed Acyclic Graphs

A **DAG** is a directed graph without cycles.

That means:

once you follow an edge there is no way back to the source node

(we can say that one node  
is after another in the graph)



# Topologic Sort

An example:

- nodes are tasks, and an edge  $(u, v)$  means “task  $u$  must be done before task  $v$ ”
- the graph is a DAG,

It means that there are no impossible dependencies between tasks

A **topological sort** gives a valid order to do the tasks in

# Implementing Topologic Sort

r = new empty list

**while**  $V \neq \emptyset$

**do**

**if** any  $v \in V$  with  $\text{indegree}(v) = 0$

**then** r.add-last(v)

remove v from G

**else**

raise error: cycle found

**return** r // Nodes, topologically sorted.

# Implementing Topologic Sort

```
r = new empty list
while  $V \neq \emptyset$ 
  do
    if any  $v \in V$  with  $\text{indegree}(v) = 0$ 
      then  $r.\text{add-last}(v)$ 
        remove  $v$  from  $G$ 
    else
      raise error: cycle found
return  $r$  // Nodes, topologically sorted.
```



How can we avoid  
removing  $v$  from  $G$  ?

# Implementing Topologic Sort

```
r = new empty list
d = map from vertices to their indegrees // null for nodes in r.
repeat  $|V|$  times
    if  $d[v] == 0$  for some  $v$ 
        then  $r.add\_last(v)$ 
             $d[v] = \text{null}$ 
            for each direct successor  $v'$  of  $v$ 
                do
                    decrease  $d[v']$  by 1
    else
        raise error: cycle found
return  $r$  // Nodes, topologically sorted.
```

# Implementing Topologic Sort

## Complexity Analysis

(Need more info because of pseudocode representation).

- Nodes:  $0, 1, \dots, |V|-1$
- *adjacent*: array with  $V$  positions

*adjacent*[ $i$ ] – contains list of neighbours for node  $i$

- *r*: dynamic array,
- *d*: array.

# Implementing Topologic Sort

// d (indegree) is a map from nodes to the indegrees that it would have  
// easier for deleting a node from the graph – indegree would be null

d = new array of size  $|V|$

**for** i in  $[0, \dots, |V|-1]$

**do**

d[i] = 0

**for** i in  $[0, \dots, |V|-1]$

**do**

**for** each direct successor j of i

**do**

d[j]++

$O(|V|)$

$O(|E|)$



# Implementing Topologic Sort

**r** = new empty list

$O(1)$

**d** = map from vertices to their indegrees // null for nodes in **r**.

$O(|V| + |E|)$

**repeat**  $|V|$  times

$O(|V|)$

**if**  $d[v] == 0$  for some  $v$

$O(|V|)$

**then** **r.add-last**( $v$ )

$O(1)$

$d[v] = \text{null}$

$O(1)$

**for** each direct successor  $v'$  of  $v$

$O(|E|)$

**do**

decrease  $d[v']$  by 1

$O(1)$

**else**

raise error: cycle found

**return** **r** // Nodes, topologically sorted.

Total :  $O(|V|^2 + |E|)$

# Implementing Topologic Sort

```
r = new empty list
d = map from vertices to their indegrees
q = queue with all nodes of indegree 0
while q is non-empty
do
    v = q.dequeue()
    r.add-last(v)
    for each direct successor v' of v
        do
            decrease d[v'] by 1
            if d[v'] = 0 then q.enqueue(v')
if r.length() < |V| then raise error: cycle found
return r // Nodes, topologically sorted.
```

# Question

What is the time complexity of this solution if adjacency lists are used to represent edges ?

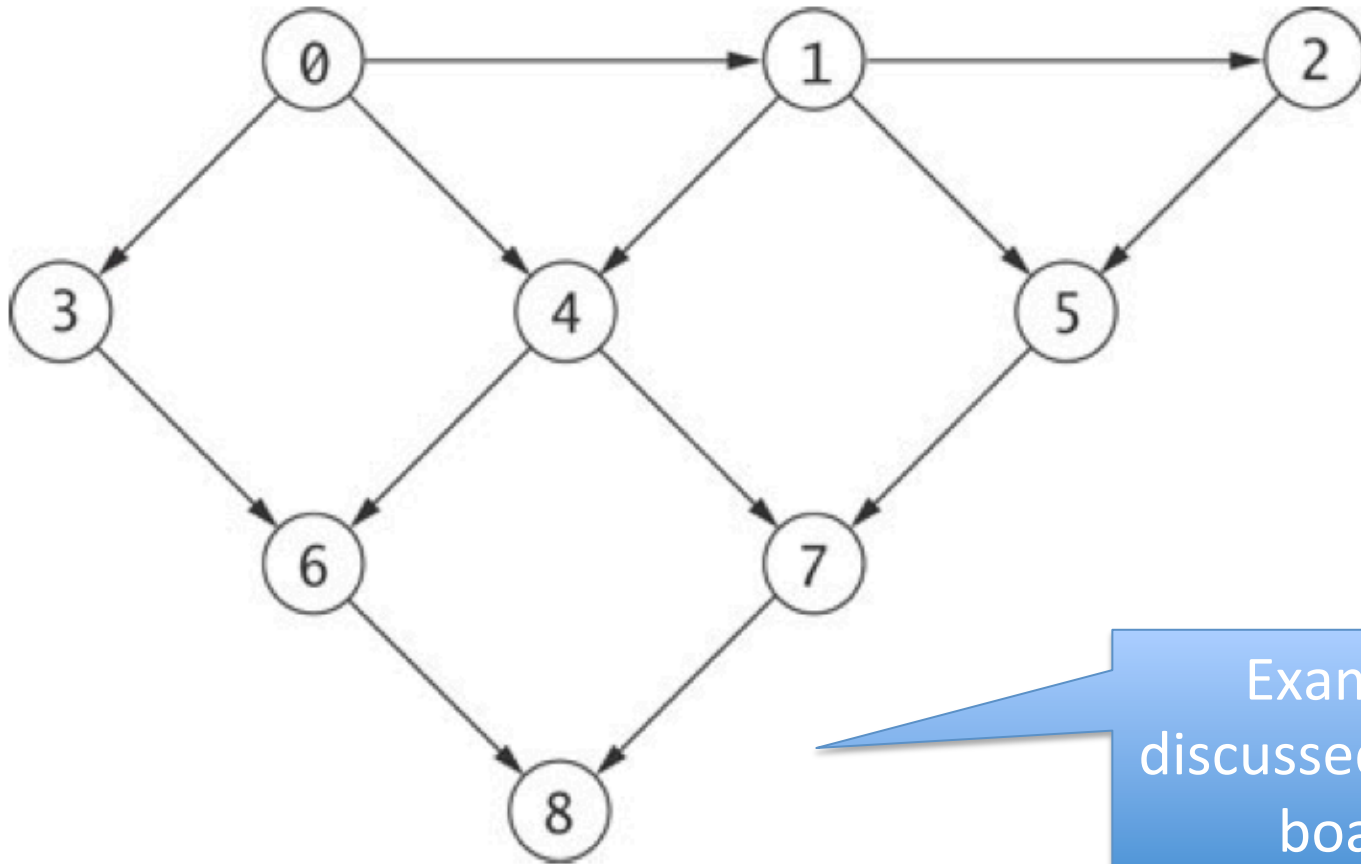
1.  $O(|V|)$
2.  $O(|E|)$
3.  $O(|V| + |E|)$
4.  $O(|V|^2)$

govote.at  
Code 771118

# Implementing Topologic Sort

```
r = new empty list  $O(1)$ 
d = map from vertices to their indegrees  $O(|V| + |E|)$ 
q = queue with all nodes of indegree 0  $O(|V|)$ 
while q is non-empty  $O(|V|)$ 
do
    v = q.dequeue()  $O(1)$ 
    r.add-last(v)  $O(1)$ 
    for each direct successor v' of v  $O(E)$  in total
    do
        decrease d[v'] by 1  $O(1)$ 
        if d[v'] = 0 then q.enqueue(v')  $O(1)$ 
if r.length() < |V| then raise error: cycle found  $O(1)$ 
return r // Nodes, topologically sorted.  $O(1)$ 
```

# Example of Topologic Sort



# Depth-first Search – Again!

- **Depth-first search** is an alternative search order that's easier to implement
- To do a DFS starting from a node:
  - visit the node
  - recursively DFS all adjacent nodes (skipping any already-visited nodes)
- Much simpler 😊

# Applications of DFS

- Checking if a graph has a cycle
- Checking if the graph is connected
- Alternative algorithm for topological sort
- ...

# Topologic Sort with DFS

$L \leftarrow$  Empty list that will contain the sorted nodes

**while** there are unmarked nodes

**do**

    select an unmarked node  $n$

    visit( $n$ )

**function** visit(node  $n$ )

**if**  $n$  has a temporary mark **then** stop (not a DAG)

**if**  $n$  is not marked (i.e. has not been visited yet)

**then** mark  $n$  temporarily **for each** node  $m$  with an edge from  $n$  to  $m$  **do** visit( $m$ ) mark  $n$  permanently unmark  $n$  temporarily add  $n$  to *head* of  $L$



# Strong Connection in Graphs

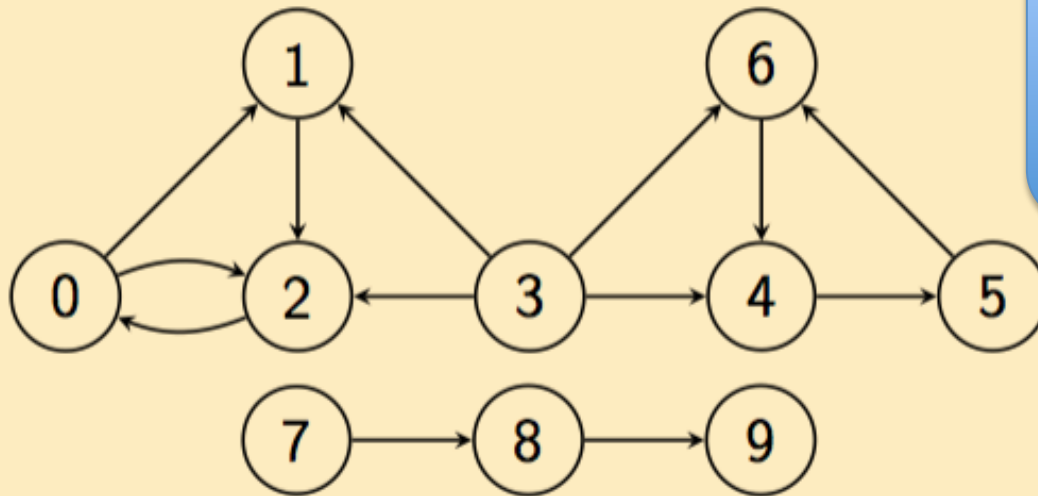
A directed graph where there is a path between any two nodes in both directions is called **strongly connected**.

A graph can be split into **strongly connected components** (subgraphs that are strongly connected) with the help of DFS

# Strongly Connected Components

## Question

How many strongly connected components are here ?



govote.at  
Code 510490

# Strongly Connected Components

$S \leftarrow$  empty stack  
**while**  $S$  does not contain all vertices  
  **do**

Kosaraju's  
algorithm

    choose an arbitrary vertex  $v$  not in  $S$   
    dfs( $v$ )

      - for each vertex  $u$  found, push  $u$  onto  $S$ .

reverse the directions of all arcs to obtain the transpose graph.

**while**  $S$  is nonempty  
  pop the top vertex  $v$  from  $S$ .  
  dfs( $v$ ) in the transposed graph

The set of visited vertices will give the strongly connected component containing  $v$ ; record this and remove all these vertices from the graph  $G$  and the stack  $S$ .

# Shortest paths

## Typical problems

- Find the shortest path between
- $u$  and  $v$
- Find the shortest path between  $u$  and all other nodes from the graph
- Find the shortest path between any two nodes from the graph



# Shortest Path – Based on BFS

```
d = new array of size |V|, initialised to  $\infty$ 
p = new array of size |V|, initialised to null
q = new empty queue
q.enqueue(s)
d[s] = 0
while q is non-empty
  do
    v = q.dequeue()
    for each direct successor v' of v
      do
        if d[v'] =  $\infty$  then
          d[v'] = d[v] + 1
          p[v'] = v
          q.enqueue(v')
return (d, p)
```

# Shortest Path based on BFS

$d$  = new array of size  $|V|$ , initialised to  $\infty$

$O(|V|)$

$p$  = new array of size  $|V|$ , initialised to null

$O(|V|)$

$q$  = new empty queue

$q.enqueue(s)$

$d[s] = 0$

**while**  $q$  is non-empty

$O(|V|)$

**do**

$v = q.dequeue()$

**for** each direct successor  $v'$  of  $v$

$O(|E|)$

**do**

**if**  $d[v'] = \infty$  **then**

$d[v'] = d[v] + 1$

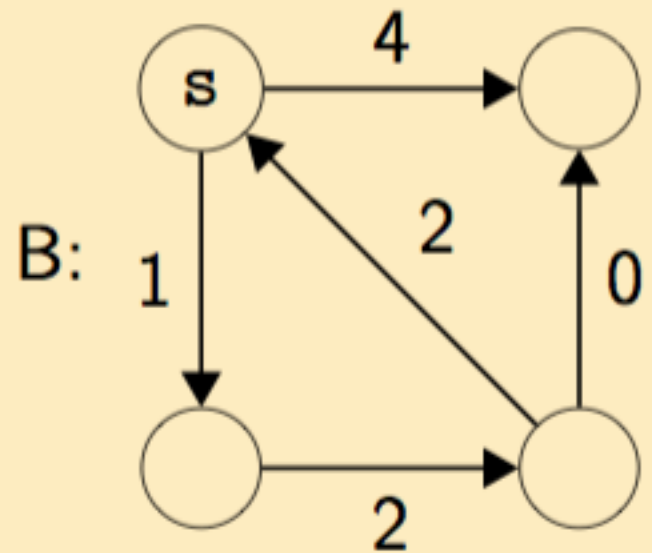
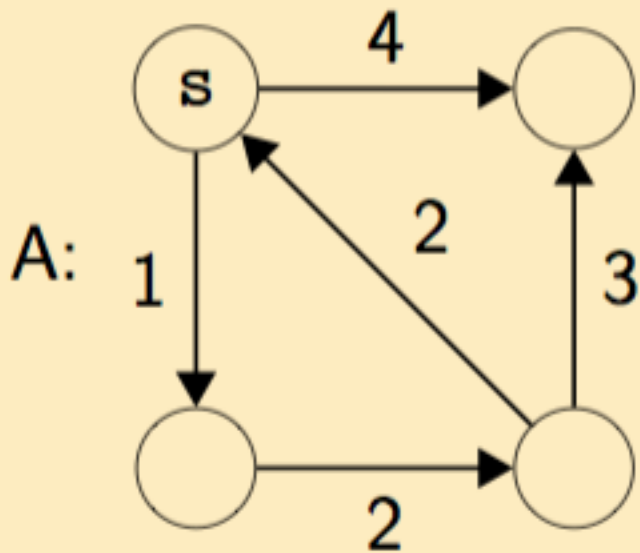
$p[v'] = v$

$q.enqueue(v')$

**return**  $(d, p)$

# Question – Part 1

Test out the algorithm on the following graphs.



# Question – Part 2

What kind of graphs is the previous algorithm good for finding the shortest path:

- A. No graphs
- B. Undirected graphs
- C. Unweighted graphs
- D. Any graphs

govote.at  
Code 637013



# Dijkstra's Algorithm

$d$  = new array of size  $|V|$ , initialised to  $\infty$   
 $p$  = new array of size  $|V|$ , initialised to null  
 $k$  = new array of size  $|V|$ , initialised to false  
 $d[s] = 0$

**repeat**

**if** no  $v'$  satisfies  $!k[v] \ \&\& \ d[v'] < \infty$  **then break**

$v = v'$  with smallest  $d[v']$  that satisfies  $!k[v']$

$k[v] = \text{true}$

**for** each direct successor  $v'$  of  $v$


**do**

**if**  $(!k[v'])$  and  $d[v'] > d[v] + c(v, v')$  **then**

$d[v'] = d[v] + c(v, v')$

$p[v'] = v$

**return**  $(d, p)$



Edges are represented with adjacency lists

# Dijkstra's Algorithm

Complexity

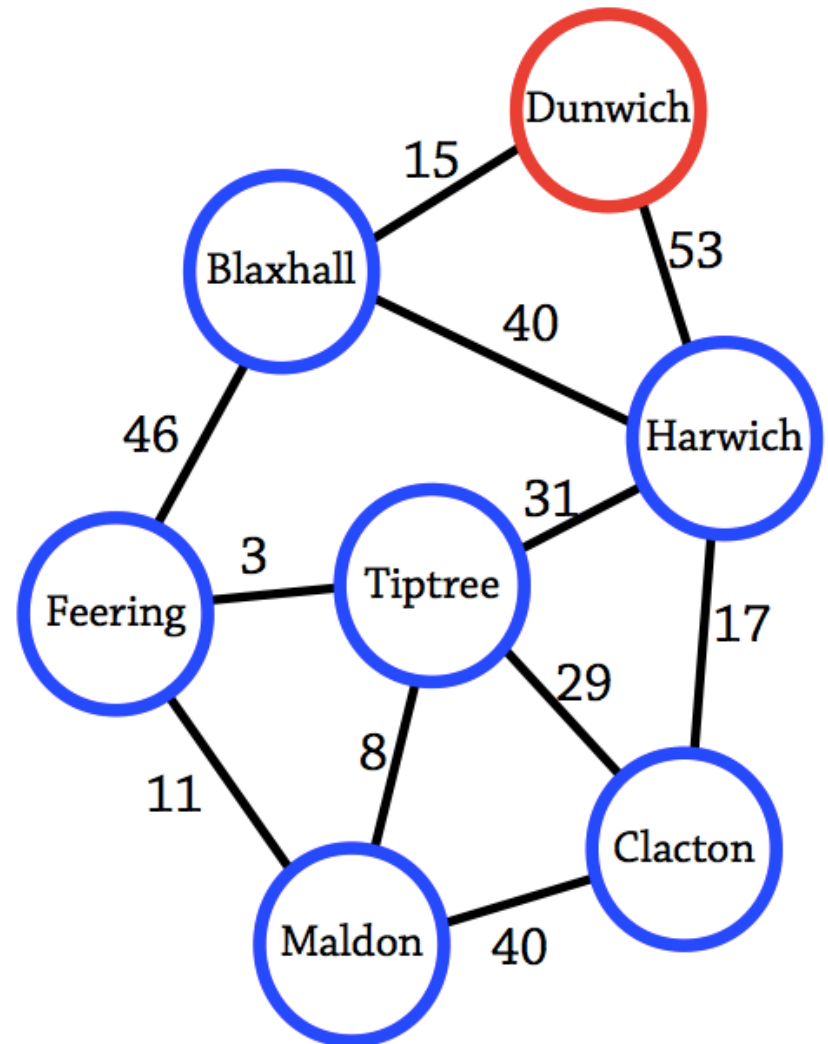
$$O(|V|^2 + |E|) = O(|V|^2)$$



Motivation in the book

# Dijkstra's Algorithm - Example

Demonstration on the board !

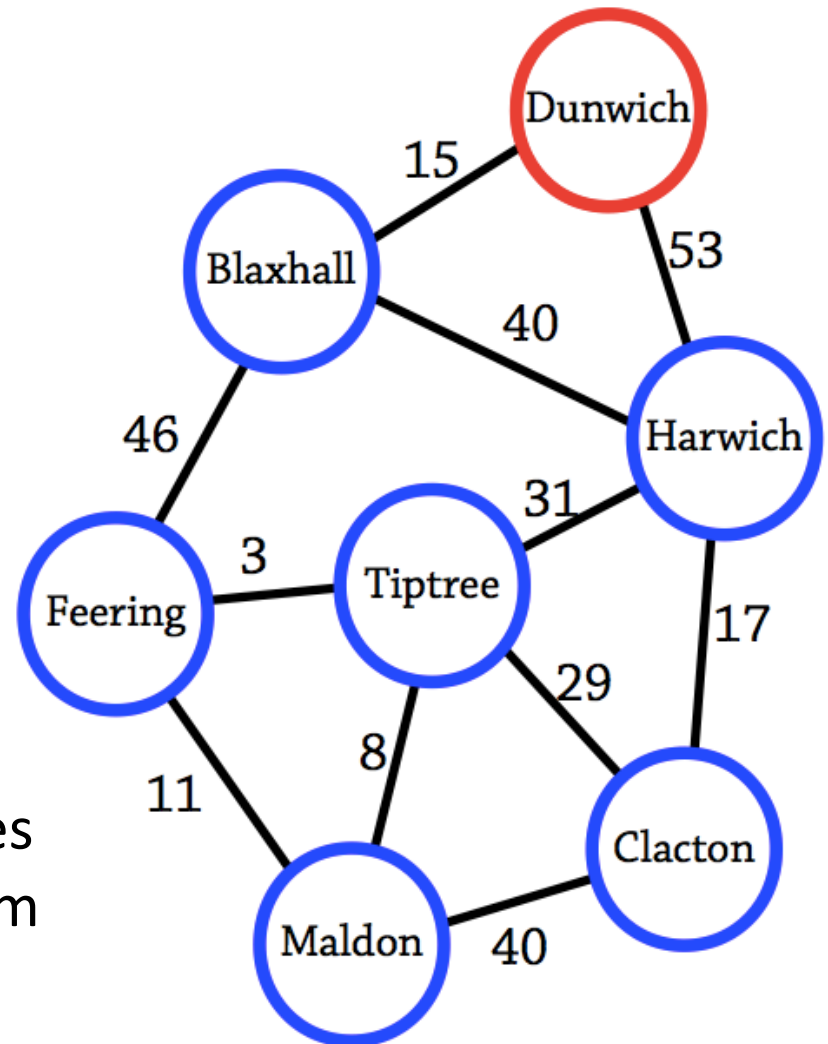


# Dijkstra's Algorithm - Example

S = {Dunwich → 0,  
Blaxhall → 15,  
Harwich → 53,  
Feering → 61,  
Tiptree → 64,  
Clacton → 70,  
Maldon → 72}

Finished 😊

Dijkstra's algorithm enumerates nodes in order of how far away they are from the start node



# Dijkstra's Algorithm

Where can we optimize the algorithm ?

1. Can't
2. Choose a better starting point
3. Use an adjacency matrix instead
4. Use a better structure to compute the next node

govote.at  
Code 308091

# Dijkstra's Algorithm – Take 2

$d, p, k$  – as before

$q$  = new empty **priority queue**

$d[s] = 0$

$q.insert(s, 0)$

**while**  $q$  is non-empty

**do**

$v = q.delete-min()$

**if**  $\neg k[v]$  **then**

$k[v] = \text{true}$

**for** each direct successor  $v'$  of  $v$  **do**

**if**  $(\neg k[v'])$  and  $d[v'] > d[v] + c(v, v')$  **then**

$d[v'] = d[v] + c(v, v')$

$p[v'] = v$

$q.insert(v', d[v'])$

**return**  $(d, p)$

# Dijkstra's Algorithm – Take 2



Motivation in the book

Complexity

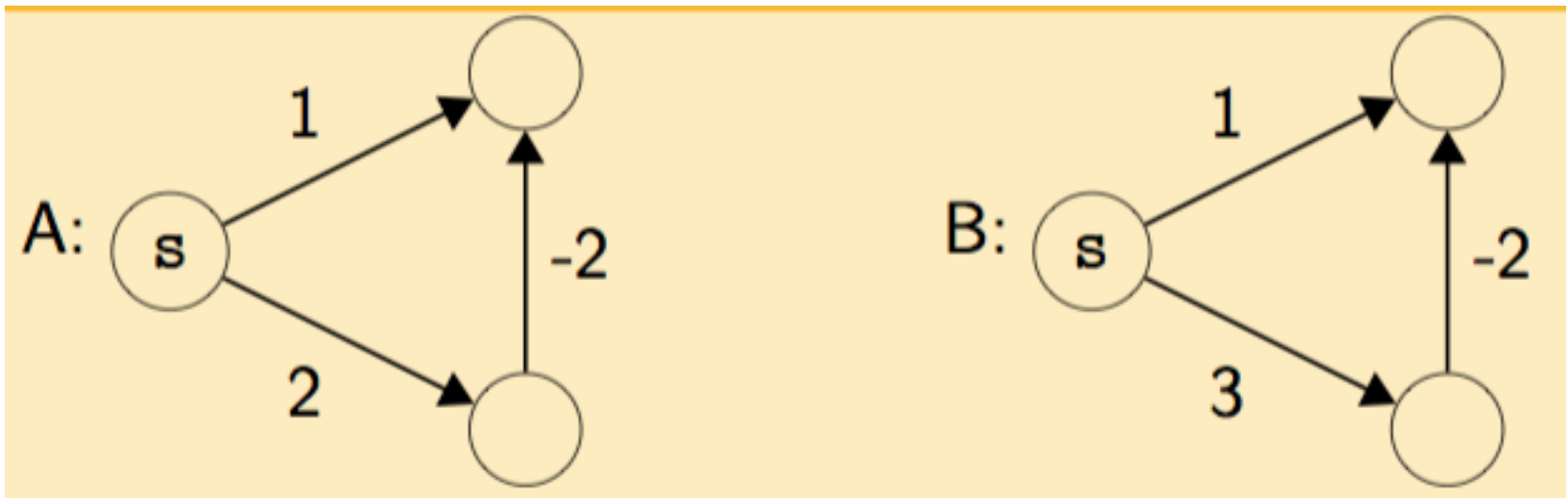
$$\begin{aligned} O(|E| \log |V| + |V| \log |V|) \\ = O(|E| \log |V|) \end{aligned}$$

Because of **decreaseKey** and **findMin** operations of the priority queue

# Dijkstra's Algorithm

Let's look at the following example.

What happens here if we apply Dijkstra's algorithm ?





# Dijkstra's Algorithm

Dijkstra's algorithm only works for positive weights !!

There is a more general algorithm for weighted graphs

- more complex 😞
- described in the book 😊

# To Do

Read from the book:

+ 9.3, 9.6

+ **better** reference for graphs:

*Wikipedia*

Introduction to Algorithms by Cormen, Leiserson, Rivest and Stein

Fun with graphs:

+ famous graph problem: **The Seven Bridges of Königsberg**

<http://www.mathsisfun.com/activity/seven-bridges-konigsberg.html>

Implement:

+ graphs + the algorithms from today in  
your favourite programming language

**Labs:**

+ 26<sup>th</sup> Nov – deadline Lab 2

+ 27<sup>th</sup> Nov – final deadline Lab 1

