# Föreläsning 21 Directed and weighted graphs

# TDDD86: DALP

Utskriftsversion av Föreläsing i Datastrukturer, algoritmer och programmeringsparadigm 05 December 2023

IDA, Linköpings universitet

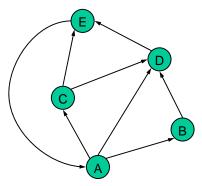
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#### **Directed graphs** 1

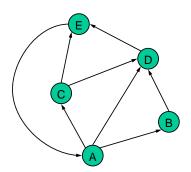
# Introduction

• In a directed graph, all edges are directed



# Properties

- A graph G = (V, E) s.t. each edge as a direction:
  - With edge (a, b) you can go from a to b but not from b to a.
- If *G* is simple (no parallel edges or loops) then  $m \le n \cdot (n-1)$ , hence  $m \in O(n^2)$ .

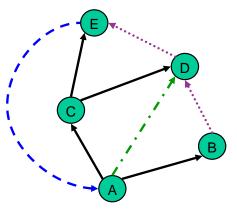


#### Some algorithmic graph problems

- Path. Is their a directed path from *s* to *t*?
- Shorted path. what is the shorted directed path from *s* to *t*?
- Strong connectivity. Is there a directed path between all pairs of nodes?
- Topological sort. Is it possible to draw the graph such that all the edges point in the same direction?
- Transitive closure. For which nodes *v* and *w* there is at least one path from *v* to *w*?
- Page Rank. How important is a web page?

#### **Directed DFS**

- We can specialize the DFS and BFS graph traversing algorithms to directed graphs
- · In the directed DFS algorithm there are 4 kinds of edges
  - "discovery"-edges
  - back-edges
  - forward-edges
  - crossing edges
- A directed DFS from node *s* lists the nodes that are reachable from *s*



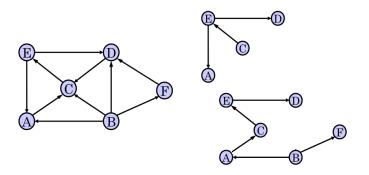
# 2 Connectivity

# Reachability

- DFS-tree rooted in *v*: nodes reachable from *v* via directed paths
- Not all nodes are reachable from node C, but all nodes are reachable from node B

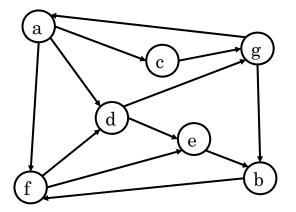
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# Strongly connected graphs

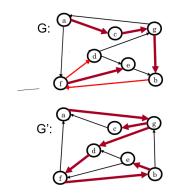
Each node is reachable from each other node



# Algorithm to decide whether a graph is strongly connected

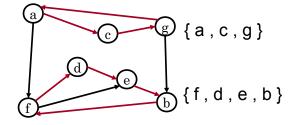
- Choose a node v in G
- // Can we reach all nodes from v? Perform DFS from v in G
  - If a node *w* remains unvisited, answer "no"
- Obtain G' from G by reversing all edges
- // Can we reach all nodes from v in G'? Perform DFS v in G'
  - If a node *w* remains unvisited, answer "no"
  - Otherwise answer "yes"
- Execution time: O(n+m)

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# Strongly connected components

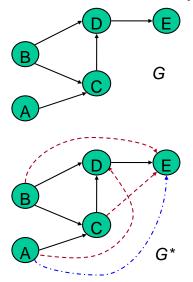
- A maximal sub-graph where each node is reachable from each other node in the sub-graph
- Can also be obtained in O(n+m) time complexity by using DFS in several steps
- The DFS based Kosaraju's algorithm:
  - Call DFS and enumerate vertices in post-order
  - Call DFS on transposed graph (i.e., reversed edges)



# 3 Transitive closure

#### Transitive closure

- Given a directed graph G, the transitive closure of G is the directed graph  $G^*$  where
  - $G^*$  has the same nodes as G
  - if G has a directed path from u to  $v (u \neq v)$  then  $G^*$  has a directed edge from u to v
- The transitive closure make explicit reachability in a directed graph



#### Computing the transitive closure

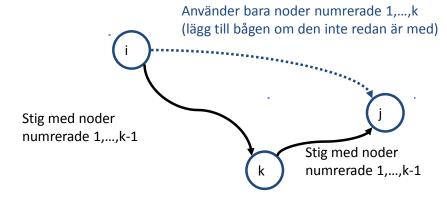
- We could execute DFS from each node  $v_1, \ldots, v_n$ , hence  $O(n \cdot (n+m))$
- A dynamic programming alternative: Floyd-Warshalls algorithm

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#### Transitive closure with Floyd-Warshall

- Identify the nodes with  $1, 2, \ldots, n$ .
- In phase k, only consider paths that use nodes in 1, 2, ..., k as intermediary nodes:



# Floyd-Warshall algorithm

• The Floyd-Warshall algorithm enumerates the nodes in *G* as  $v_1, \ldots, v_n$  and computes the series of directed graphs  $G_0, \ldots, G_n$ 

$$-G_0 = G$$

-  $G_k$  has a directed edge  $(v_i, v_j)$  if G has a directed path from  $v_i$  to  $v_j$  with intermediary nodes in the set  $\{v_1, \ldots, v_k\}$ 

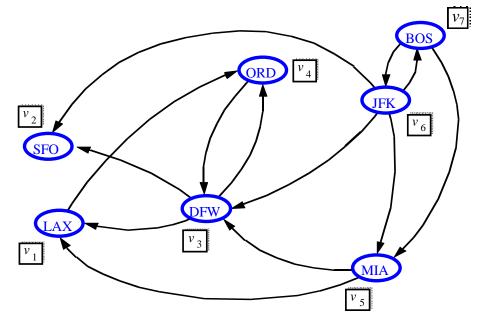
- We get  $G_n = G^*$
- At iteration k the graph  $G_k$  is computed from  $G_{k-1}$
- Execution time:  $O(n^3)$  if areAdjacent is O(1)

## The Floyd-Warshall algorithm

function FLOYDWARSHALL(G)  $G_0 \leftarrow G$ for  $k \leftarrow 1$  to n do  $G_k \leftarrow G_{k-1}$ for  $i \leftarrow 1$  to n  $(i \neq k)$  do for  $j \leftarrow 1$  to n  $(j \neq i, k)$  do if  $G_{k-1}$ .AREADJACENT $(v_i, v_k)$  then if  $G_{k-1}$ .AREADJACENT $(v_k, v_j)$  then if  $\neg G_k$ .AREADJACENT $(v_i, v_j)$  then  $G_k$ .INSERTDIRECTEDEDGE $(v_i, v_j, k)$ 

return G<sub>n</sub>

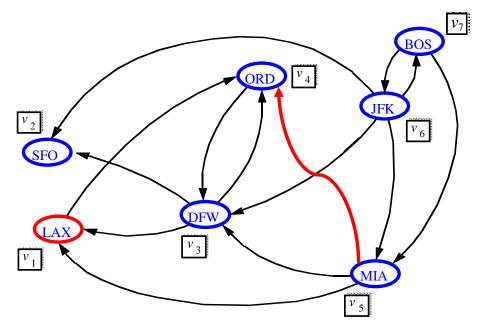
#### Example: Floyd-Warshall



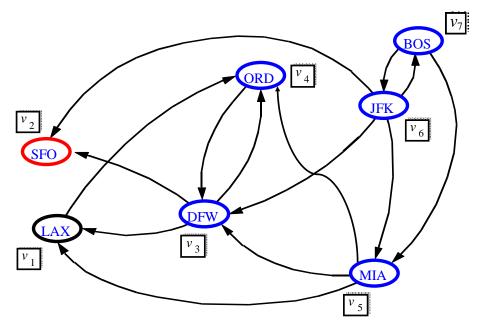
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Floyd-Warshall, iteration 1

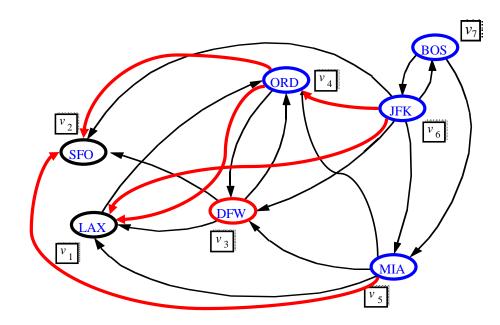


Floyd-Warshall, iteration 2

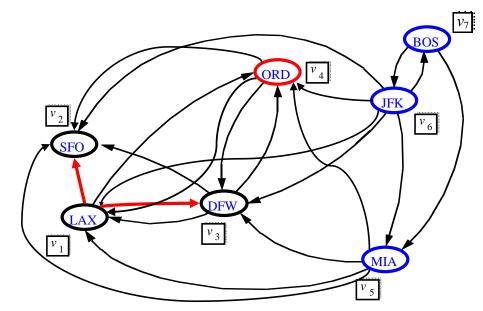


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Floyd-Warshall, iteration 3



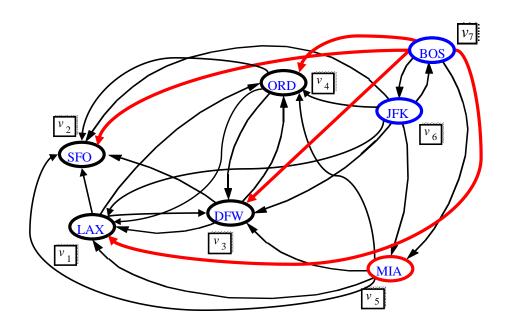
Floyd-Warshall, iteration 4



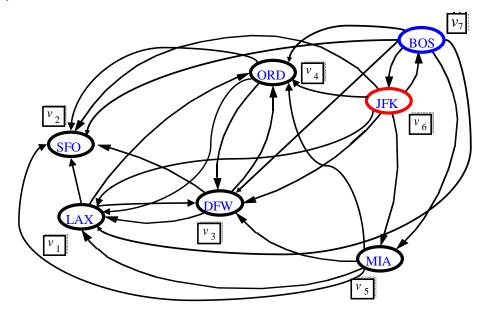
Floyd-Warshall, iteration 5

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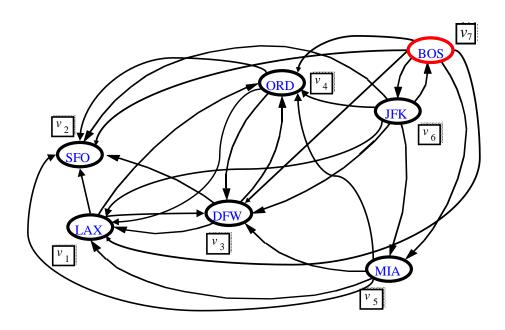
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Floyd-Warshall, iteration 6



Floyd-Warshall, end

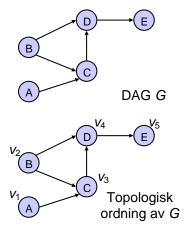


# 4 Topological sorting

# Directed acyclic graphs and topological sorting

- A directed acyclic graph (DAG) is a directed graph that does not have any directed cycle
- A topological sorting of a graph is a total ordering  $v_1, ..., v_n$  of the nodes such that each edge  $(v_i, v_j)$  satisfies i < j
- Example: Existence of a plan for tasks that depend on each other.

Proposition 1. A graph can be topologically sorted iff it is a DAG



# An algorithm for topological sorting

**procedure** TOPOLOGICALSORT(G)  $H \leftarrow G$   $n \leftarrow G.NUMVERTICES$  **while** H is non-empty **do** let v be a node without outgoing edges mark v with n  $n \leftarrow n-1$ remove v from H

Execution time: O(n+m). How...?

# Algorithm for topological sorting via DFS

**procedure** TOPOLOGICALDFS(G)  $n \leftarrow G.NUMVERTICES$ mark all nodes and edges as *UNEXPLORED* like in DFS **for all**  $v \in G.VERTICES()$  **do**   $\triangleright$  temporary copy of *G* 

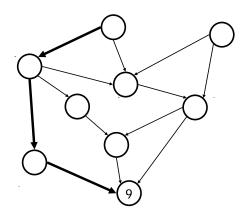
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if GETLABEL(v) = UNEXPLORED then TOPOLOGICALDFS(G, v) procedure TOPOLOGICALDFS(G, v) SETLABEL(v, VISITED) for all  $e \in G$ .INCIDENTEDGES(v) do if GETLABEL(e) = UNEXPLORED then  $w \leftarrow OPPOSITE(v, e)$ if GETLABEL(w) = UNEXPLORED then SETLABEL(e, DISCOVERY) TOPOLOGICALDFS(G, w) else e is a crossing edge or a forward edge mark v with the topological number n $n \leftarrow n-1$ 

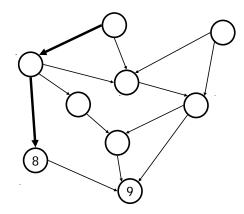
# Example: Topological sorting

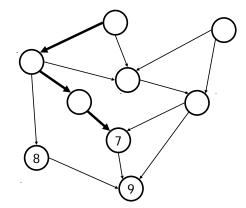
Example: Topological sorting



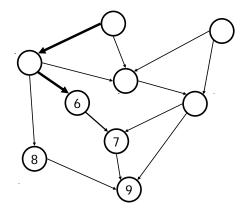
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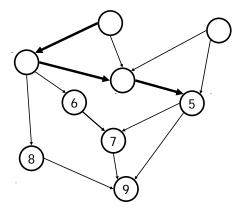
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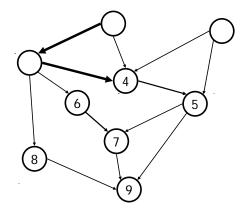
Example: Topological sorting

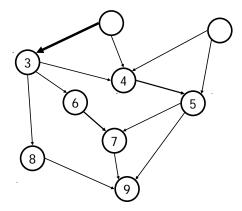




Example: Topological sorting

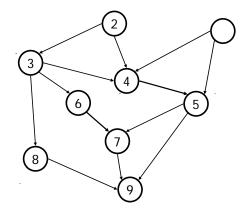
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Example: Topological sorting

21.34



2 1 3 4 5 6 7 8 9

#### Weighted graphs 5

# Weighted graphs

- In a weighted graph, each edge is associated a numerical *weight*. Weights can represent distances, costs, etc.

# Google maps

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# Continentals fly routes in USA (august 2010)



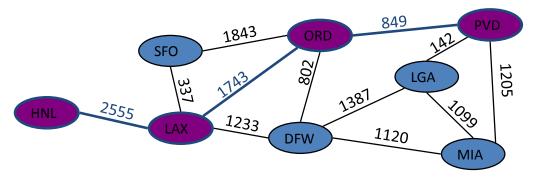
# 6 Shortest paths

# The shortest path problem

- Given a weighted graph and two nodes *u* and *v*, find a path between *u* and *v* with minimal total weight.
  - Length of a path is the sum of the weights of its edges

Example

Shortest path between Providence and Honolulu



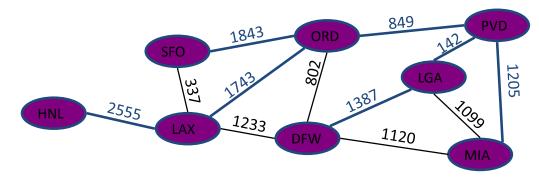
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#### Properties of shortest paths

- A sub-path of a shortest path is also a shortest path
- There is a tree of shortest paths from a start node to all other nodes

#### Example

A tree of shortest paths from Providence



Weighted Floyd-Warshalls algorithm

function WEIGHTEDFLOYDWARSHALL(G(N, E, w)) for  $i \leftarrow 1$  to N do for  $j \leftarrow 1$  to N do dist $(i, j) \leftarrow w(i, j)$ for  $i \leftarrow 1$  to N do dist $(i, i) \leftarrow 0$ for  $k \leftarrow 1$  to N do for  $i \leftarrow 1$  to N do for  $j \leftarrow 1$  to N do if  $(\text{dist}(v_i, v_j) > \text{dist}(v_i, v_k) + \text{dist}(v_k, v_j))$  then dist $(v_i, v_j) \leftarrow \text{dist}(v_i, v_k) + \text{dist}(v_k, v_j)$ return dist

#### Dijkstra's algorithm

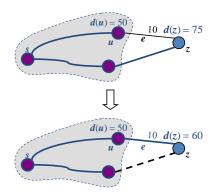
- Distance from a node v to a node s is the length of a shortest path between s and v
- Dijkstra's algorithm computes the distance from a given start node s to all other nodes v in the graph
- Assumptions:
  - the graph is connected
  - graph has no loops or parallel edges
  - weights are non-negative
- We build a "cloud" of nodes, starting from *s*, that will cover all nodes
- We mark each node v in the cloud or neighbor to it with d(v), which represents the distance between v and s
- At each step:
  - Extend the cloud to the node u that was outside the cloud and which has the minimal distance d(u)
  - update distances of nodes that are neighbor to u

#### Extension step

- Consider an edge e = (u, z) s.t.:
  - u has just been added to the cloud
  - -z is not part of the cloud yet
- Edge e updates d(z) with :
  - $d(z) \leftarrow \min\{d(z), d(u) + weight(e)\}$

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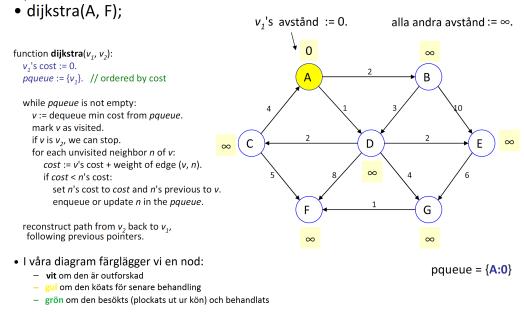


# Dijkstra pseudocode function dijkstra( $v_1, v_2$ ): initialize every vertex to have a cost of infinity. set $v_1$ 's cost to 0. pqueue := { $v_1$ , with priority 0}. // ordered by cost while pqueue is not empty:

v := dequeue vertex from pqueue with minimum priority.
mark v as visited.
if v is v<sub>2</sub>, we can stop.
for each unvisited neighbor n of v:
 cost := v's cost + weight of edge (v, n).
 if cost < n's cost:
 set n's cost to cost, and n's previous to v.
 enqueue n in the pqueue with priority of cost,
 or update its priority if it was already in the pqueue.</pre>

reconstruct path from  $v_2$  back to  $v_1$ , following previous pointers.

#### Example

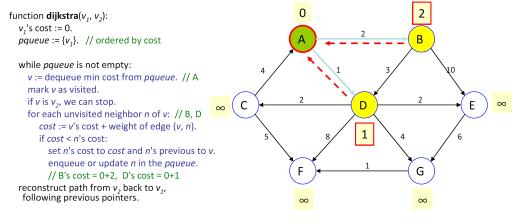


## Example

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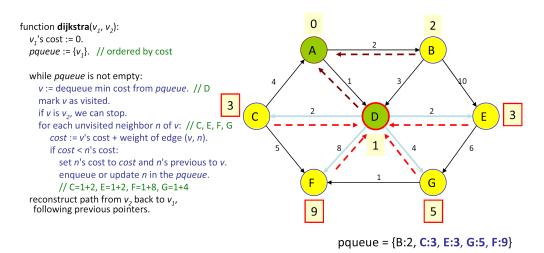
# • dijkstra(A, F);



pqueue = {**D:1**, **B:2**}

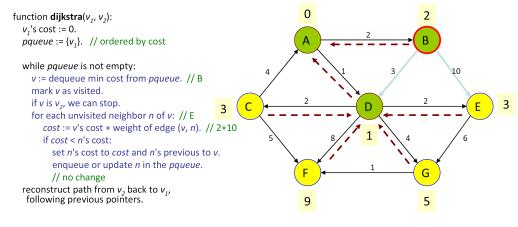
#### Example

• dijkstra(A, F);



Example

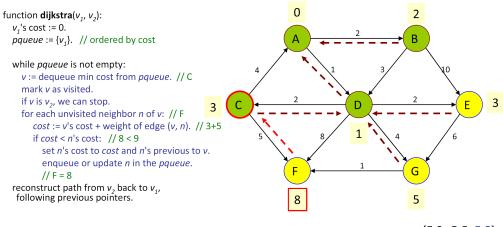
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pqueue = {C:3, E:3, G:5, F:9}

#### Example

dijkstra(A, F);

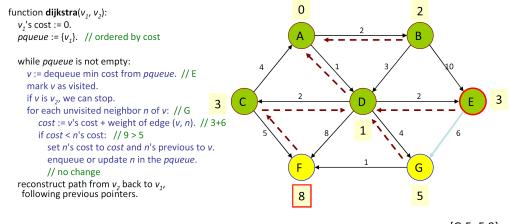


pqueue = {E:3, G:5, **F:8**}

Example

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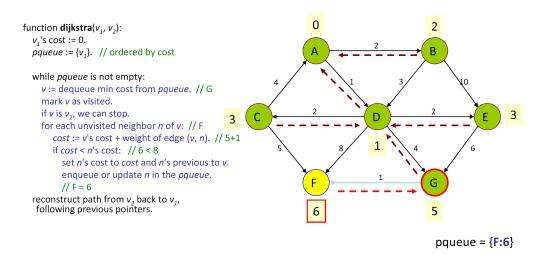
# • dijkstra(A, F);



pqueue = {G:5, F:8}

#### Example

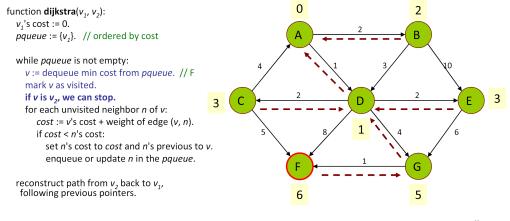
dijkstra(A, F);



Example

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# • dijkstra(A, F);



pqueue = {}

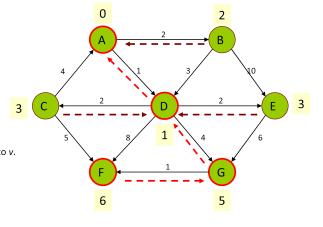
# Example



function **dijkstra**( $v_1$ ,  $v_2$ ):  $v_1$ 's cost := 0. *pqueue* := { $v_1$ }. // ordered by cost

while pqueue is not empty: v := dequeue min cost from pqueue. mark v as visited. if v is v<sub>2</sub>, we can stop. for each unvisited neighbor n of v: cost := v's cost + weight of edge (v, n). if cost < n's cost: set n's cost and n's previous to v. enqueue or update n in the pqueue.

reconstruct path from  $v_2$  back to  $v_1$ , following previous pointers. // path = {A, D, G, F}



#### Analysis of Dijkstra algorithm

- incidentEdges is called once for each node
- markings are fetched/updated for node z O(deg(z)) times
- to fetch/update a marking takes O(1) time
- Each node is inserted once and removed once from the priority queue, where each insertion and removal takes  $O(\log n)$  time
- The key of a node in the priority queue is updated at most deg(w) times, where each update may take at most  $O(\log n)$  time
- function **dijkstra**( $v_1$ ,  $v_2$ ): initialize every vertex to have a cost of infinity. set  $v_1$ 's cost to 0.

#### $pqueue := \{v_1, with priority 0\}$ . // ordered by cost

while *pqueue* is not empty:

- v := dequeue vertex from *pqueue* with minimum priority. mark v as visited.
- if v is  $v_2$ , we can stop. for each unvisited neighbor n of v:
- or each unvisited neighbor n of v: cost := v's cost + weight of edge (v, n).
- cost := v s cost + weight of edge (v, n)if cost < n's cost:
- set *n*'s cost to *cost*, and *n*'s previous to *v*.
- enqueue *n* in the *pqueue* with priority of *cost*, or update its priority if it was already in the *pqueue*.
- reconstruct path from  $v_2$  back to  $v_1$ , following previous pointers.
- Dijkstra's algorithm has execution time of  $O((n+m)\log n)$  given the graph is represented using adjacent lists
- Execution time can also be expressed as  $O(m \log n)$  since we assume it is connected

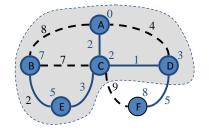
# Why does it work

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Dijkstra's algorithm is a greedy algorithm. It greedily adds nodes in increasing distances to the source.

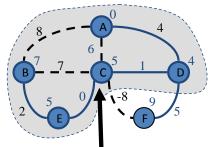
- Suppose the algorithm does not find all shortest distances. Let *F* be the first node that got a wrong shortest distance.
- Any node *D* preceding *F* along a shortest path must have obtained a correct shortest distance and added to the cloud at some point.
- But then the edge (D, F) must have been *updated* when such a D was added!
- In other words, since  $d(F) \ge d(D)$ , then the distance to F should have been correct.



#### Why does it require non-negative weights?

Dijkstra's algorithm is a greedy algorithm. It greedily adds nodes in increasing distances to the source.

• If a node with a negative incident edge is added later to the cloud, it would jeopardize the distances to nodes that were earlier added.



C's sanna avstånd är 1, men finns redan i molnet med d(C)=5!

#### Observations

- Dijkstra's algorithm works by incrementally computing shortest path to potential intermediary nodes.
  - Most such paths are in the wrong direction.

			<u>5?</u>	4	5?	6?				
	6?	5?	4	3	4	5	6?			
6?	5	4	3	2	3	4	5?			
5?	4	3	2	1	2	3	4	5?		
4	3	2	1		1	2	3	4		
<b>5</b> ?	4	3	2	1	2	3	4	5?		
	5?	4	3	2	3	4	5	6?		
	6?	5	4	3	4	5?	6?			
		6?	5?	4	5?					

- The algorithm explores in all directions;
  - Could we tips the algorithm to first explore more promising directions?

#### Heuristics

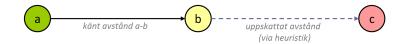
- heuristic: Speculation, estimation or a qualified guess on how the search for a solution should proceed.
  - Example: Estimate the distance between two locations in a map using a direct line.
- for the following algorithm, an admissible heuristic is one that does not over-estimate the distance.
  - Ok if the heuristic under-estimates the distance (as above with the maps).

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#### A\*-algorithm

• A\*("A-star): a modified version of Dijkstra's algorithm that uses a heuristic to direct the search.



- Suppose we are looking for a path from source node *a* to target node *c* 
  - Each intermediary node *b* has two costs:
  - The known exact cost from *a* to *b*
  - A heuristic based estimation of the cost from b to the target node c.
- Idea: Execute Dijkstra's algorithm but adopt the following priority in the priority queue:
  - priority(b) = cost(a, b) + Heuristic(b, c)
  - Explore based on the smallest estimated cost

#### Example: Labyrinth heuristic

- A possible heuristic to find paths in a labyrinth:
  - $H(p_1, p_2) = abs(p_1.x p_2.x) + abs(p_1.y p_2.y)$  // dx + dy
  - Idea: Explore neighbors with low value of (cost + heuristic)

6	5	5	4	3	4
5		4	З	2	3
4	ļ	3	2	1	2
а	I	2	1	с	1
4	Ļ	3	2	1	2
5	5	4	3	2	3

# Recall: pseudo-code for Dijkstra's algorithm

```
function dijkstra(v_1, v_2):
initialize every vertex to have a cost of infinity.
set v_1's cost to 0.
pqueue := {v_1, with priority 0}. // ordered by cost
```

```
while pqueue is not empty:
    v := dequeue vertex from pqueue with minimum priority.
    mark v as visited.
    if v is v<sub>2</sub>, we can stop.
    for each unvisited neighbor n of v:
        cost := v's cost + weight of edge (v, n).
        if cost < n's cost:
            set n's cost to cost, and n's previous to v.
            enqueue n in the pqueue with priority of cost,
            or update its priority if it was already in the pqueue.
```

reconstruct path from  $v_2$  back to  $v_1$ , following previous pointers.

Pseudo-code for the A\*-algorithm

function **astar**( $v_1$ ,  $v_2$ ): initialize every vertex to have a cost of infinity. set  $v_1$ 's cost to 0. pqueue := { $v_1$ , at priority  $H(v_1, v_2)$ }.

while pqueue is not empty: v := dequeue vertex from pqueue with minimum priority. mark v as visited. if v is v<sub>2</sub>, we can stop. for each unvisited neighbor n of v: cost := v's cost + weight of edge (v, n). if cost < n's cost : set n's cost to cost, and n's previous to v. enqueue n in the pqueue with priority of (cost + H(n, v<sub>2</sub>)), or update its priority to be (cost + H(n, v<sub>2</sub>)) if it was already in the pqueue.

reconstruct path from  $v_2$  back to  $v_1$ , following previous pointers.

Observe that only nodes' priorities are influenced by the heuristic, not their costs.