

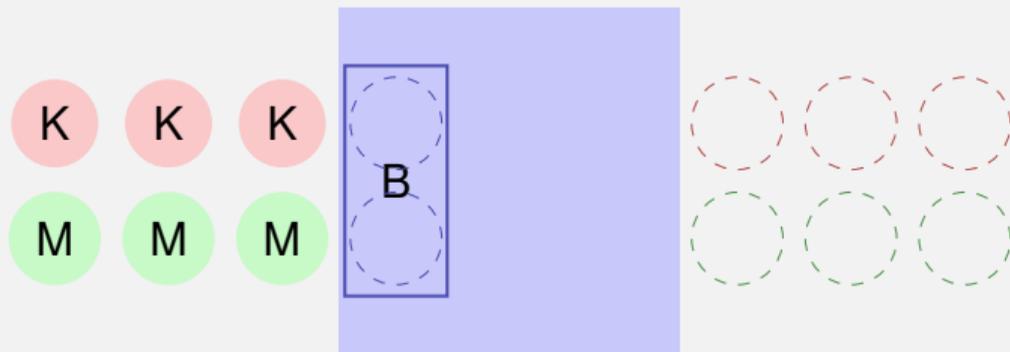
# 14. Shortest Paths

Motivation, Dijkstra's algorithm on distance graphs, Bellman-Ford Algorithm

[Ottman/Widmayer, Kap. 9.5.1-9.5.2 Cormen et al, Kap. 24.1-24.3]

# River Crossing (Missionaries and Cannibals)

Problem: Three cannibals and three missionaries are standing at a river bank. The available boat can carry two people. At no time may at any place (banks or boat) be more cannibals than missionaries. How can the missionaries and cannibals cross the river as fast as possible? <sup>23</sup>



<sup>23</sup>There are slight variations of this problem. It is equivalent to the jealous husbands problem.

# Problem as Graph

Enumerate permitted configurations as nodes and connect them with an edge, when a crossing is allowed. The problem then becomes a shortest path problem.

## Example

	links	rechts
Missionare	3	0
Kannibalen	3	0
Boot	x	

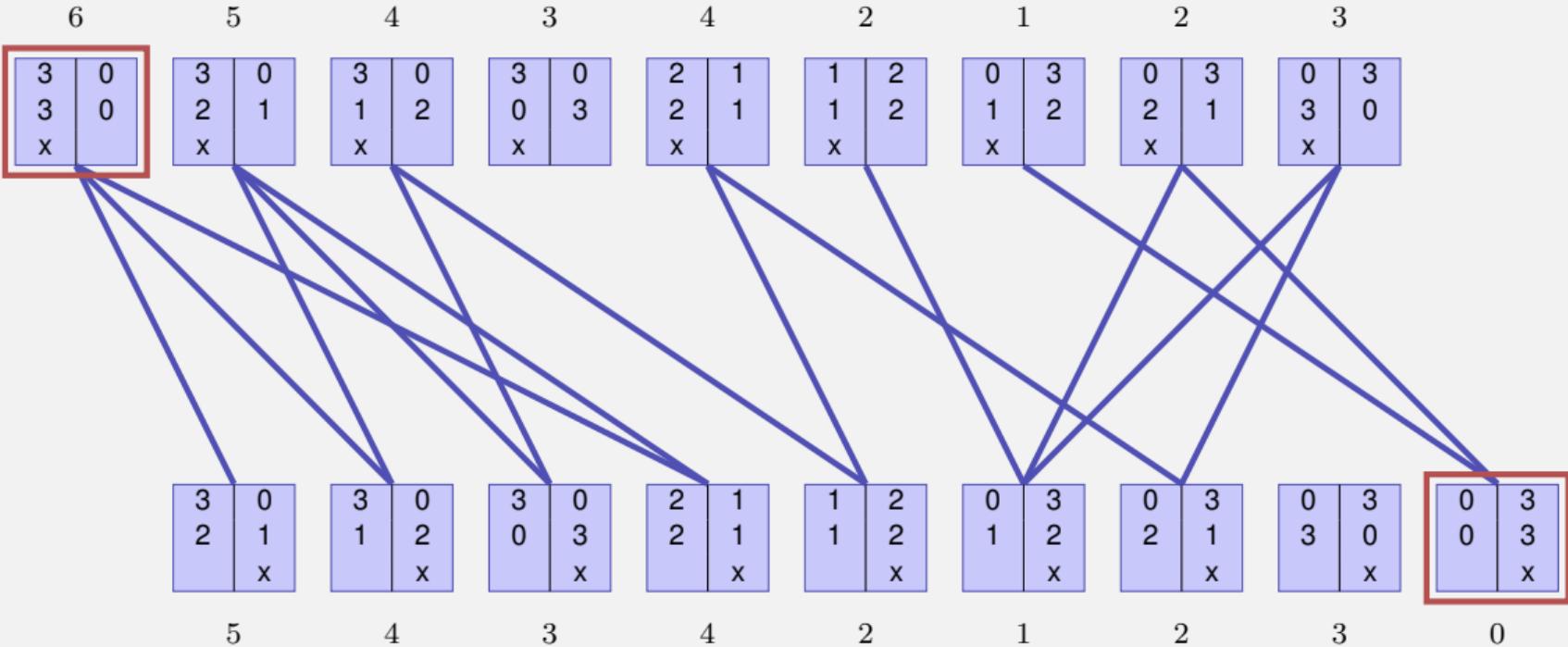
6 Personen am linken Ufer

Überfahrt möglich

	links	rechts
Missionare	2	1
Kannibalen	2	1
Boot		x

4 Personen am linken Ufer

# The whole problem as a graph

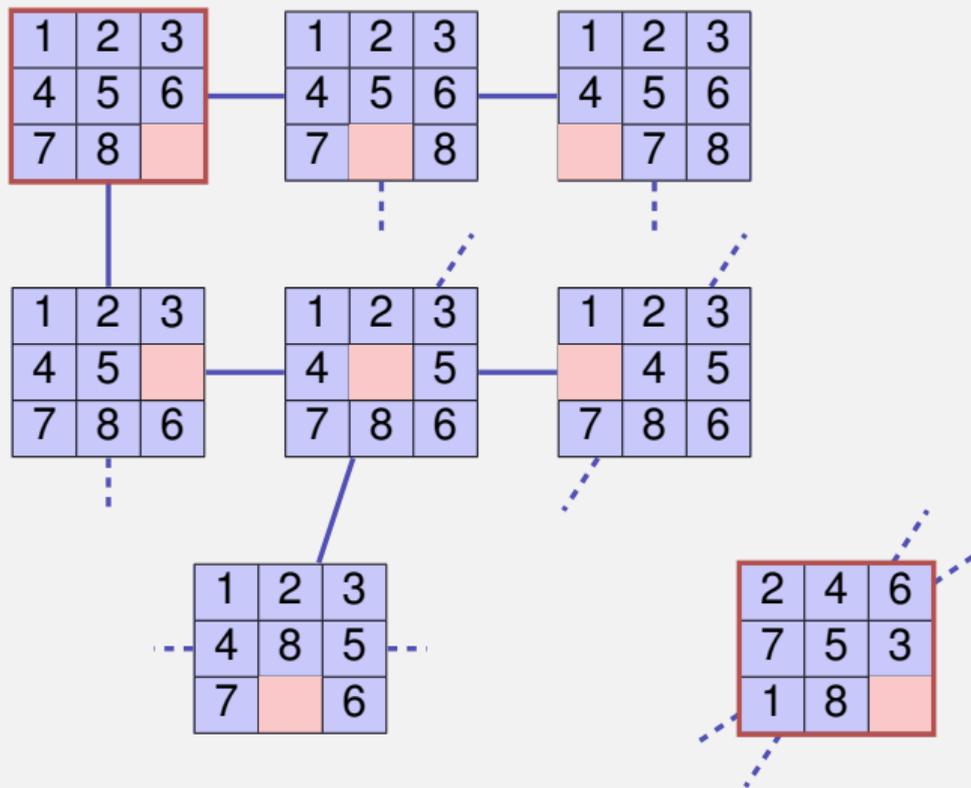


# Another Example: Mystic Square

Want to find the fastest solution for

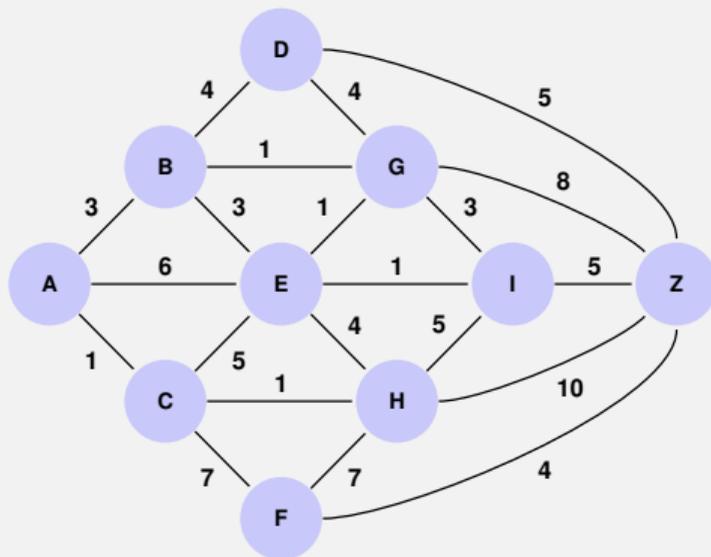


# Problem as Graph



# Route Finding

Provided cities A - Z and Distances between cities.

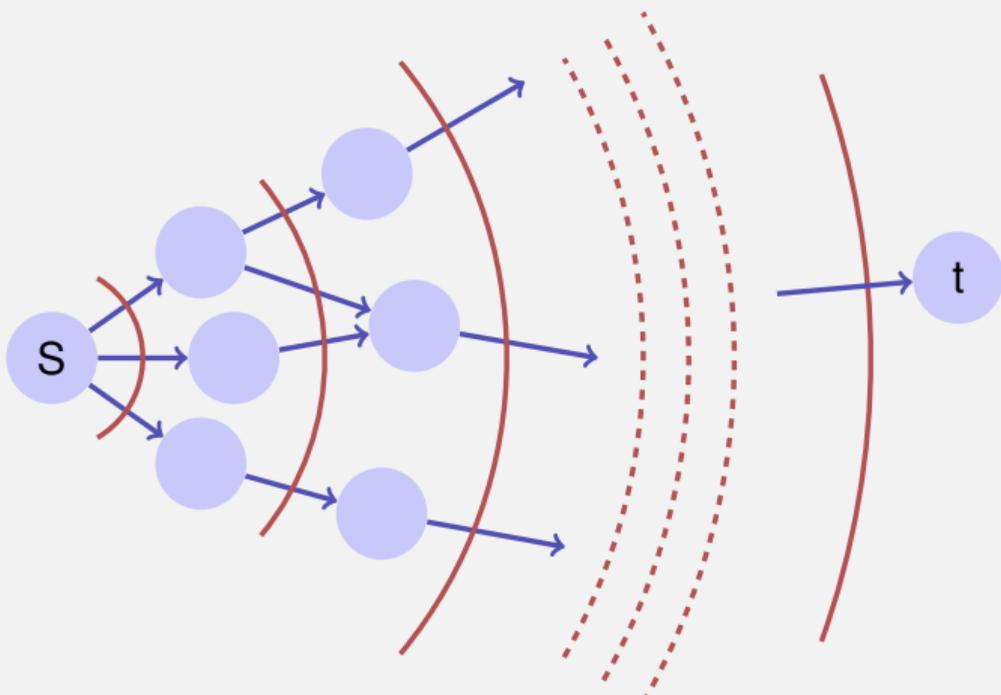


What is the shortest path from A to Z?

# Simplest Case

Constant edge weight 1 (wlog)

Solution: Breadth First Search



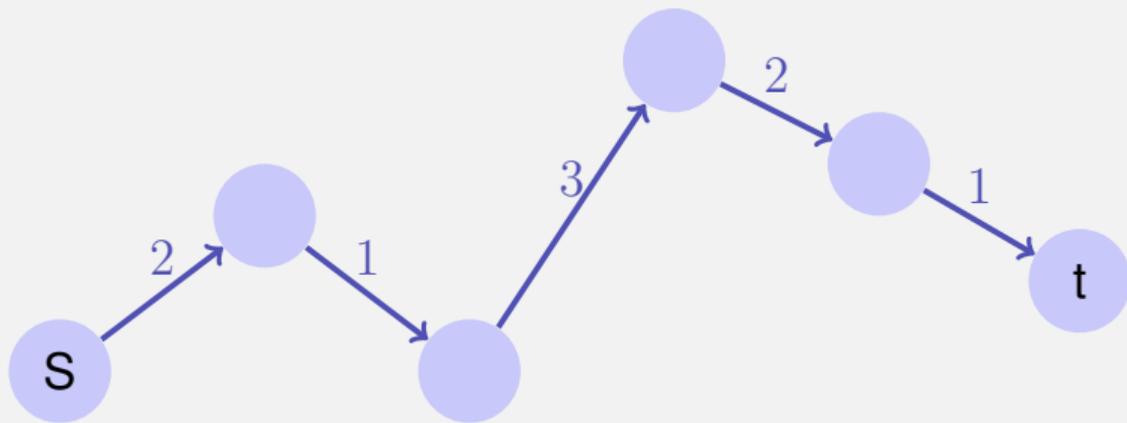
# Weighted Graphs

**Given:**  $G = (V, E, c)$ ,  $c : E \rightarrow \mathbb{R}$ ,  $s, t \in V$ .

**Wanted:** Length (weight) of a shortest path from  $s$  to  $t$ .

**Path:**  $p = \langle s = v_0, v_1, \dots, v_k = t \rangle$ ,  $(v_i, v_{i+1}) \in E$  ( $0 \leq i < k$ )

**Weight:**  $c(p) := \sum_{i=0}^{k-1} c((v_i, v_{i+1}))$ .



Path with weight 9

# Shortest Paths

**Notation:** we write

$$u \overset{p}{\rightsquigarrow} v \quad \text{oder} \quad p : u \rightsquigarrow v$$

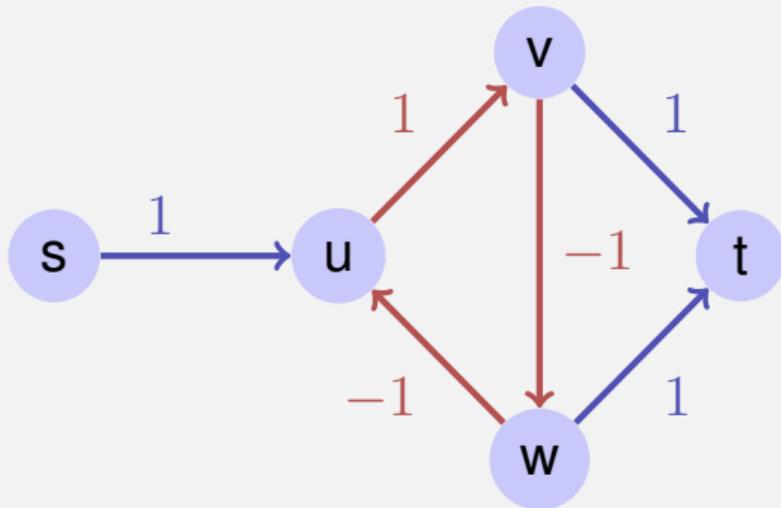
and mean a path  $p$  from  $u$  to  $v$

**Notation:**  $\delta(u, v)$  = weight of a shortest path from  $u$  to  $v$ :

$$\delta(u, v) = \begin{cases} \infty & \text{no path from } u \text{ to } v \\ \min\{c(p) : u \overset{p}{\rightsquigarrow} v\} & \text{otherwise} \end{cases}$$

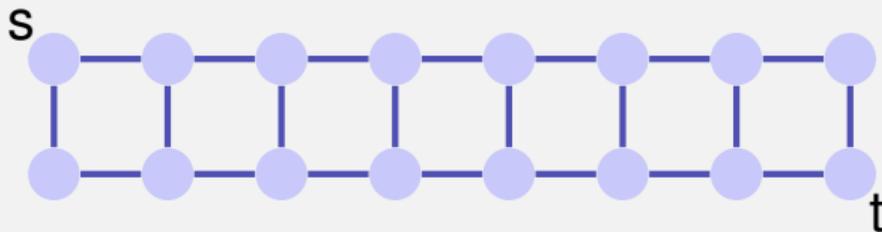
# Observations (1)

It may happen that a shortest paths does not exist: negative cycles can occur.



## Observations (2)

There can be exponentially many paths.



(at least  $2^{|V|/2}$  paths from  $s$  to  $t$ )

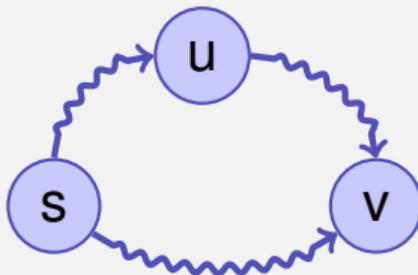
⇒ To try all paths is too inefficient

# Observations (3)

## *Triangle Inequality*

For all  $s, u, v \in V$ :

$$\delta(s, v) \leq \delta(s, u) + \delta(u, v)$$

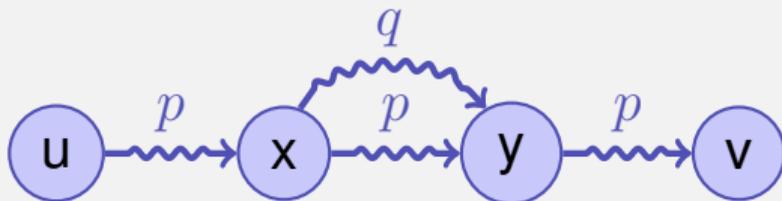


A shortest path from  $s$  to  $v$  cannot be longer than a shortest path from  $s$  to  $v$  that has to include  $u$

# Observations (4)

## *Optimal Substructure*

Sub-paths of shortest paths are shortest paths. Let  $p = \langle v_0, \dots, v_k \rangle$  be a shortest path from  $v_0$  to  $v_k$ . Then each of the sub-paths  $p_{ij} = \langle v_i, \dots, v_j \rangle$  ( $0 \leq i < j \leq k$ ) is a shortest path from  $v_i$  to  $v_j$ .



If not, then one of the sub-paths could be shortened which immediately leads to a contradiction.

# Observations (5)

## Shortest paths do not contain cycles

- 1 Shortest path contains a negative cycle: there is no shortest path, contradiction
- 2 Path contains a positive cycle: removing the cycle from the path will reduce the weight. Contradiction.
- 3 Path contains a cycle with weight 0: removing the cycle from the path will not change the weight. Remove the cycle (convention).

# Ingredients of an Algorithm

Wanted: shortest paths from a starting node  $s$ .

- Weight of the shortest path found so far

$$d_s : V \rightarrow \mathbb{R}$$

*At the beginning:*  $d_s[v] = \infty$  for all  $v \in V$ .

*Goal:*  $d_s[v] = \delta(s, v)$  for all  $v \in V$ .

- Predecessor of a node

$$\pi_s : V \rightarrow V$$

Initially  $\pi_s[v]$  undefined for each node  $v \in V$

# General Algorithm

- 1 Initialise  $d_s$  and  $\pi_s$ :  $d_s[v] = \infty$ ,  $\pi_s[v] = \text{null}$  for each  $v \in V$
- 2 Set  $d_s[s] \leftarrow 0$
- 3 Choose an edge  $(u, v) \in E$

Relaxiere  $(u, v)$ :

if  $d_s[v] > d_s[u] + c(u, v)$  then

$d_s[v] \leftarrow d_s[u] + c(u, v)$

$\pi_s[v] \leftarrow u$

- 4 Repeat 3 until nothing can be relaxed any more.  
(until  $d_s[v] \leq d_s[u] + c(u, v) \quad \forall (u, v) \in E$ )

# It is Safe to Relax

At any time in the algorithm above it holds

$$d_s[v] \geq \delta(s, v) \quad \forall v \in V$$

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In the relaxation step:

$$\delta(s, v) \leq \delta(s, u) + \delta(u, v) \quad \text{[Triangle Inequality].}$$

$$\delta(s, u) \leq d_s[u] \quad \text{[Induction Hypothesis].}$$

$$\delta(u, v) \leq c(u, v) \quad \text{[Minimality of } \delta \text{]}$$

$$\Rightarrow d_s[u] + c(u, v) \geq \delta(s, v)$$

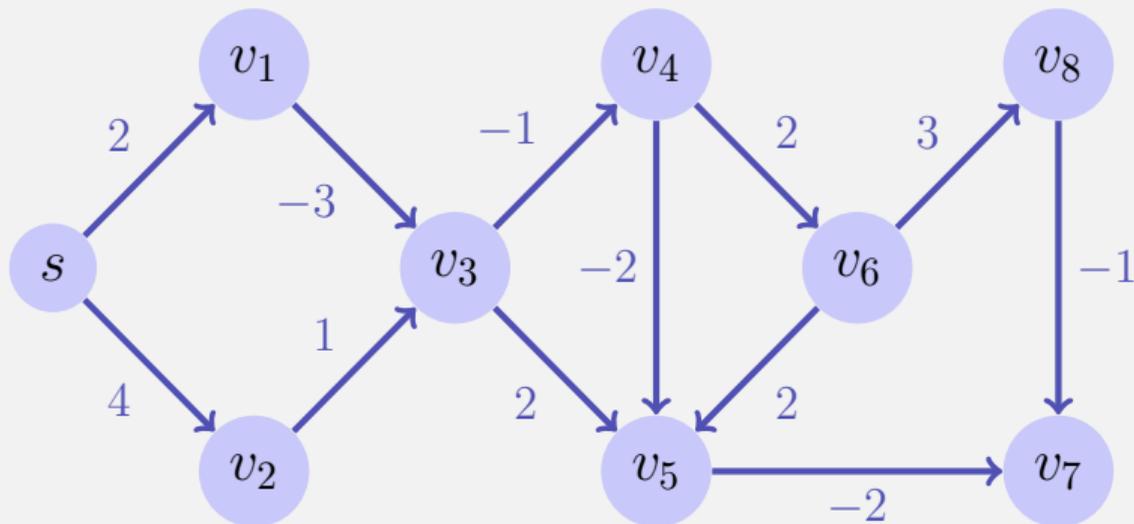
$$\Rightarrow \min\{d_s[v], d_s[u] + c(u, v)\} \geq \delta(s, v)$$

# Central Question

How / in which order should edges be chosen in above algorithm?

# Special Case: Directed Acyclic Graph (DAG)

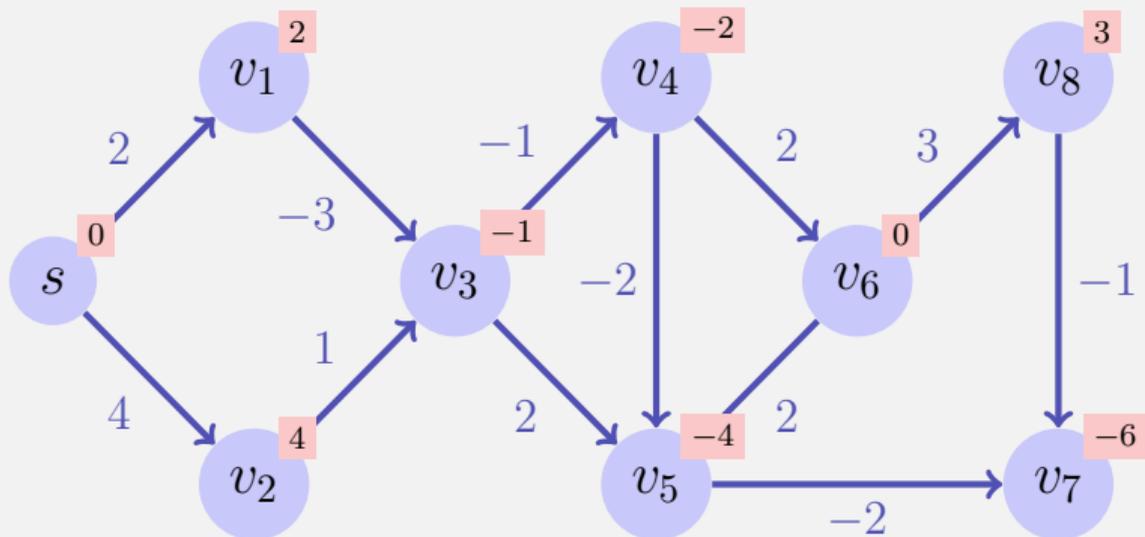
DAG  $\Rightarrow$  topological sorting returns optimal visiting order



Top. Sort:  $\Rightarrow$  Order  $s, v_1, v_2, v_3, v_4, v_6, v_5, v_8, v_7$ .

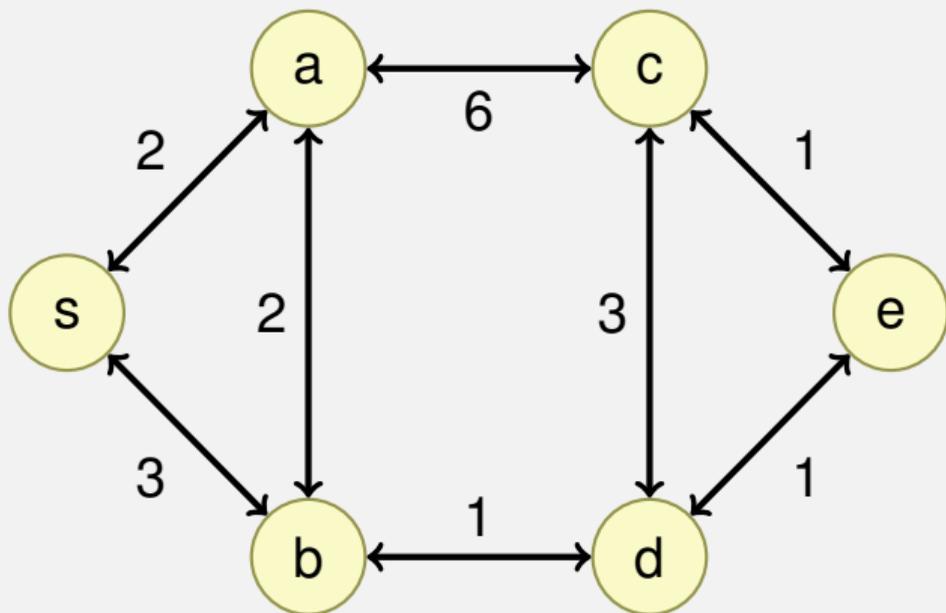
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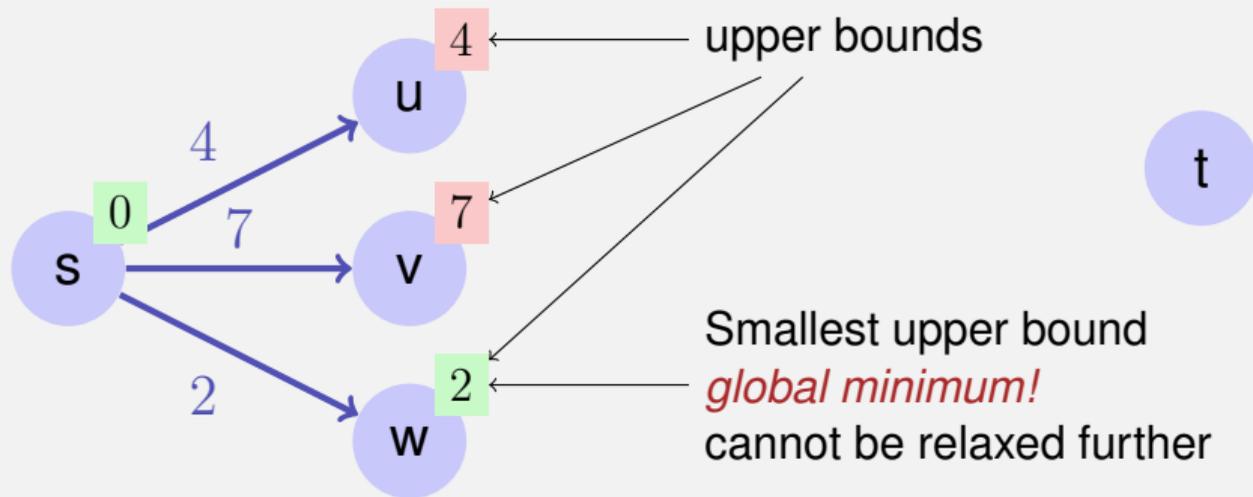
Top. Sort:  $\Rightarrow$  Order  $s, v_1, v_2, v_3, v_4, v_6, v_5, v_8, v_7$ .

# Assumption (preliminary)



All weights of  $G$  are *positive*.

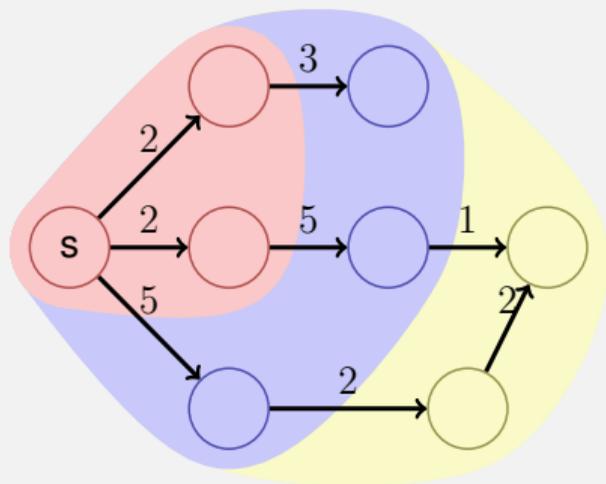
# Observation (Dijkstra)



# Basic Idea

Set  $V$  of nodes is partitioned into

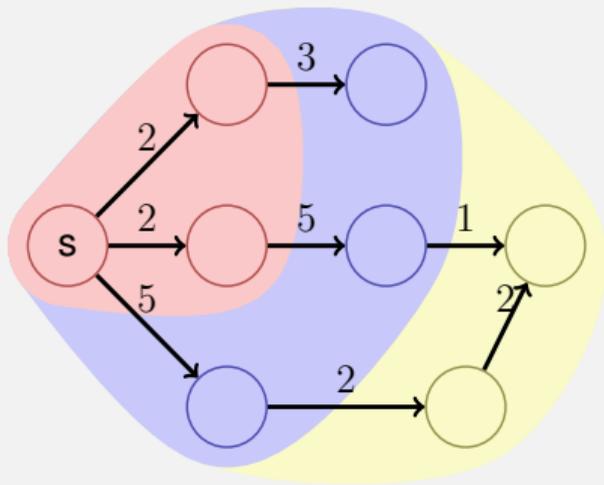
- the set  $M$  of nodes for which a shortest path from  $s$  is already known,
- the set  $R = \bigcup_{v \in M} N^+(v) \setminus M$  of nodes where a shortest path is not yet known but that are accessible directly from  $M$ ,
- the set  $U = V \setminus (M \cup R)$  of nodes that have not yet been considered.



# Induction

Induction over  $|M|$ : choose nodes from  $R$  with smallest upper bound. Add  $r$  to  $M$  and update  $R$  and  $U$  accordingly.

Correctness: if within the “wavefront” a node with minimal weight  $w$  has been found then no path over later nodes (providing weight  $\geq d$ ) can provide any improvement.



# Algorithm Dijkstra( $G, s$ )

**Input:** Positively weighted Graph  $G = (V, E, c)$ , starting point  $s \in V$ ,

**Output:** Minimal weights  $d$  of the shortest paths and corresponding predecessor node for each node.

**foreach**  $u \in V$  **do**

$d_s[u] \leftarrow \infty$ ;  $\pi_s[u] \leftarrow \text{null}$

$d_s[s] \leftarrow 0$ ;  $R \leftarrow \{s\}$

**while**  $R \neq \emptyset$  **do**

$u \leftarrow \text{ExtractMin}(R)$

**foreach**  $v \in N^+(u)$  **do**

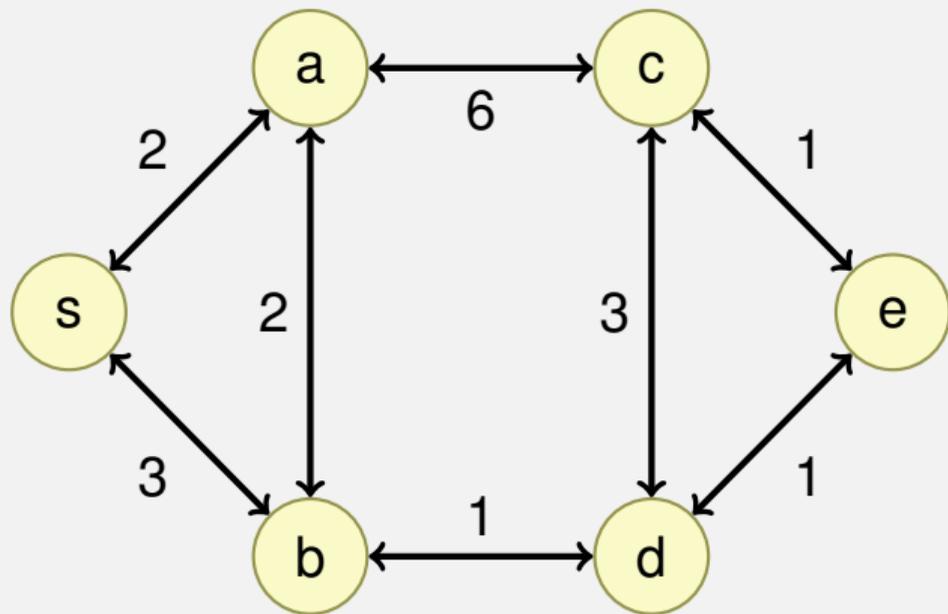
**if**  $d_s[u] + c(u, v) < d_s[v]$  **then**

$d_s[v] \leftarrow d_s[u] + c(u, v)$

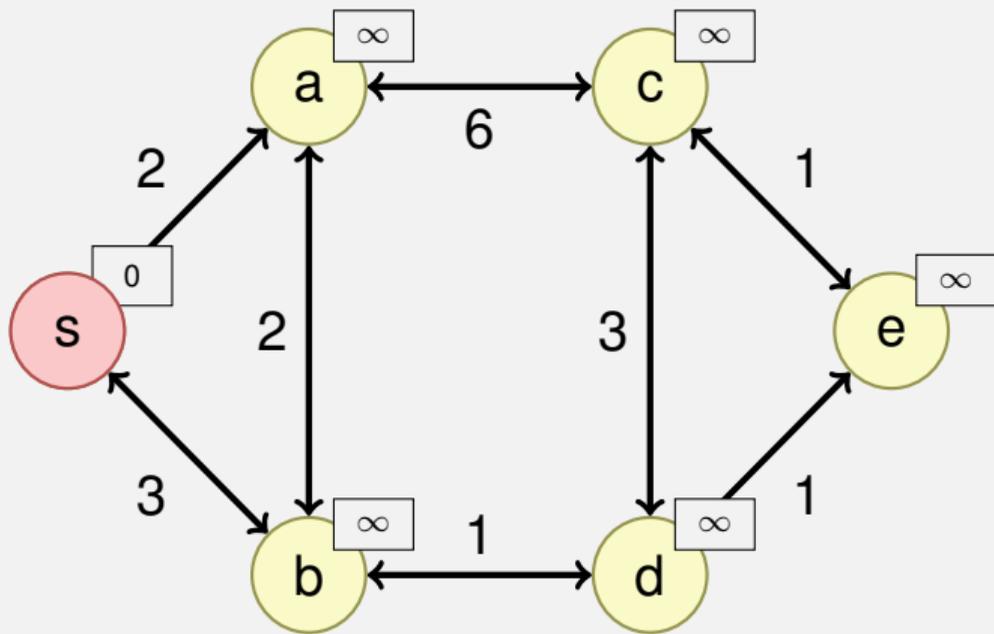
$\pi_s[v] \leftarrow u$

$R \leftarrow R \cup \{v\}$

# Example



# Example

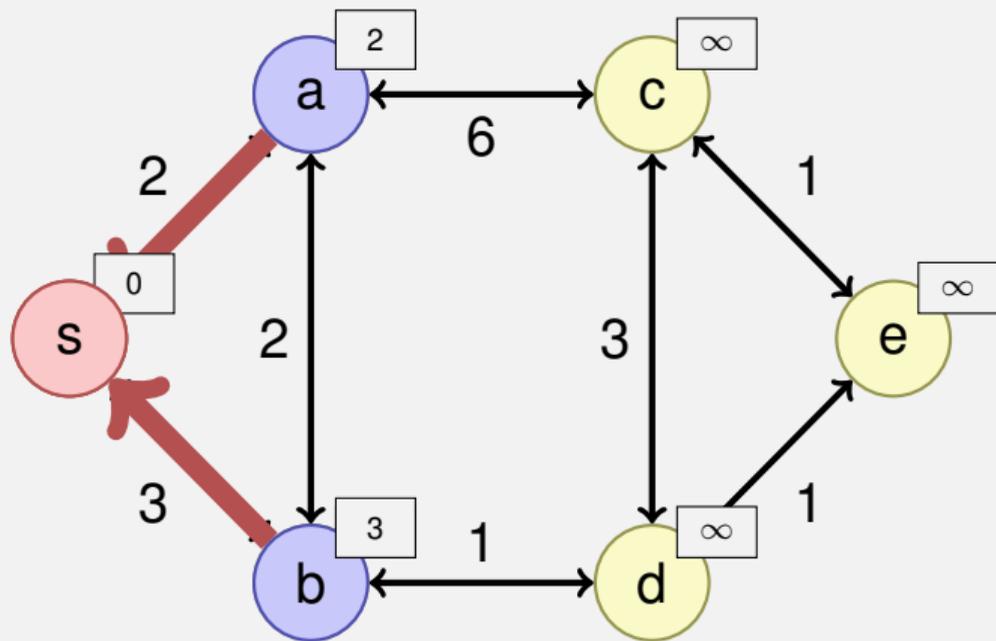


$$M = \{s\}$$

$$R = \{\}$$

$$U = \{a, b, c, d, e\}$$

# Example

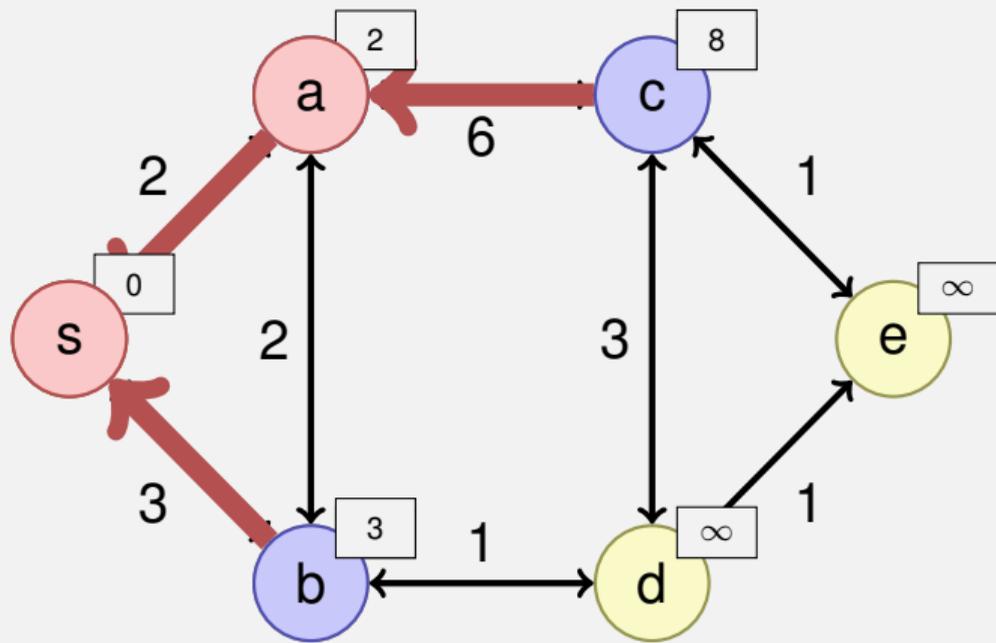


$$M = \{s\}$$

$$R = \{a, b\}$$

$$U = \{c, d, e\}$$

# Example

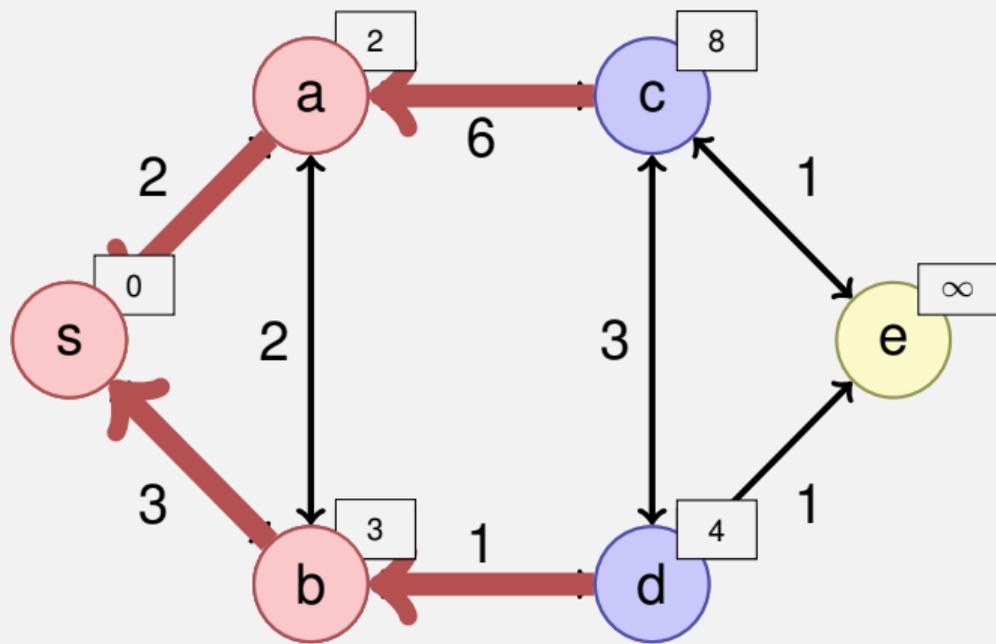


$$M = \{s, a\}$$

$$R = \{b, c\}$$

$$U = \{d, e\}$$

# Example

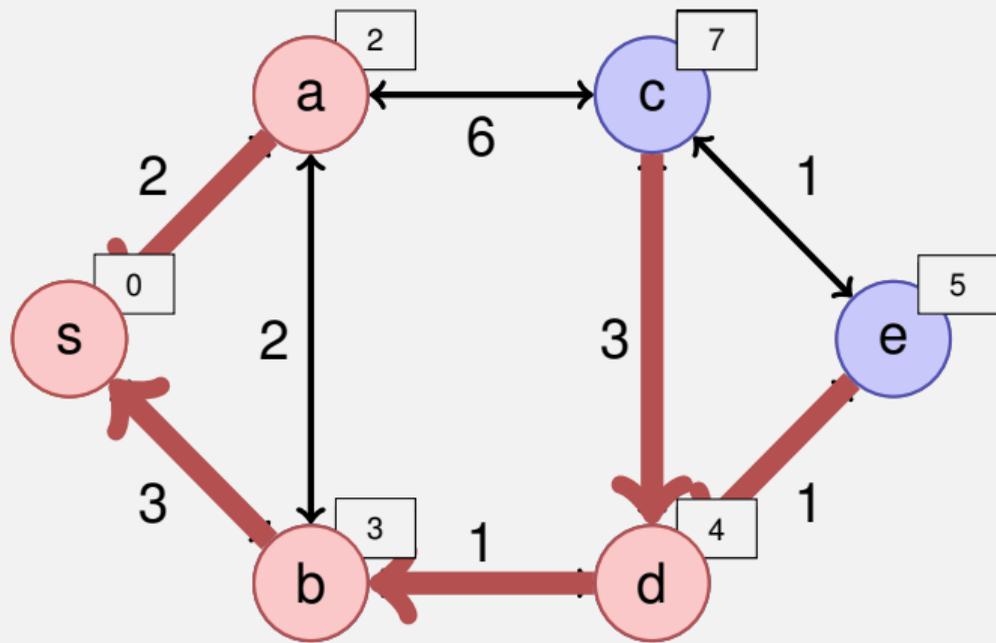


$$M = \{s, a, b\}$$

$$R = \{c, d\}$$

$$U = \{e\}$$

# Example

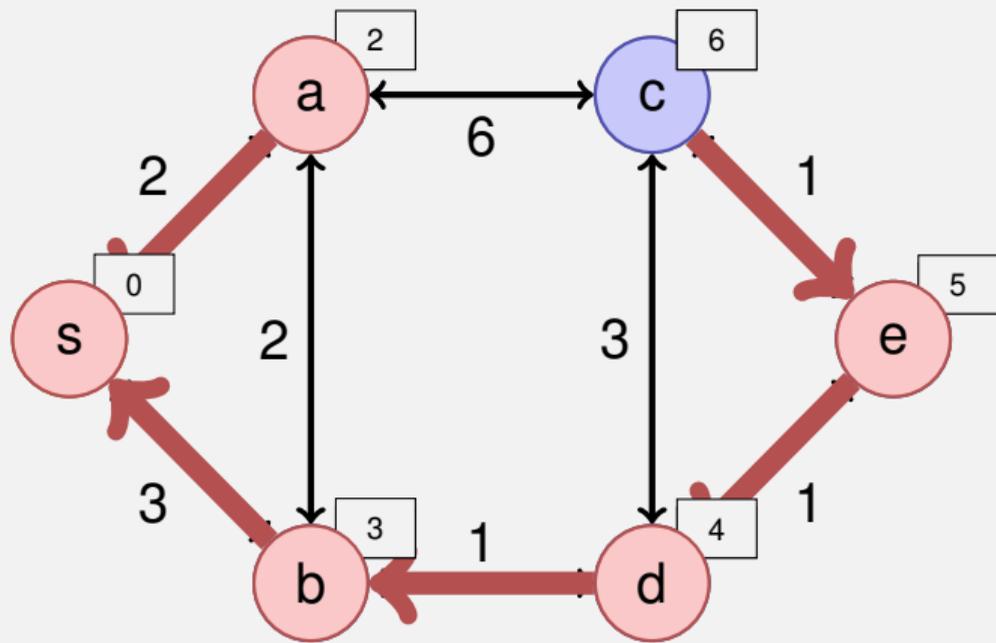


$$M = \{s, a, b, d\}$$

$$R = \{c, e\}$$

$$U = \{\}$$

# Example

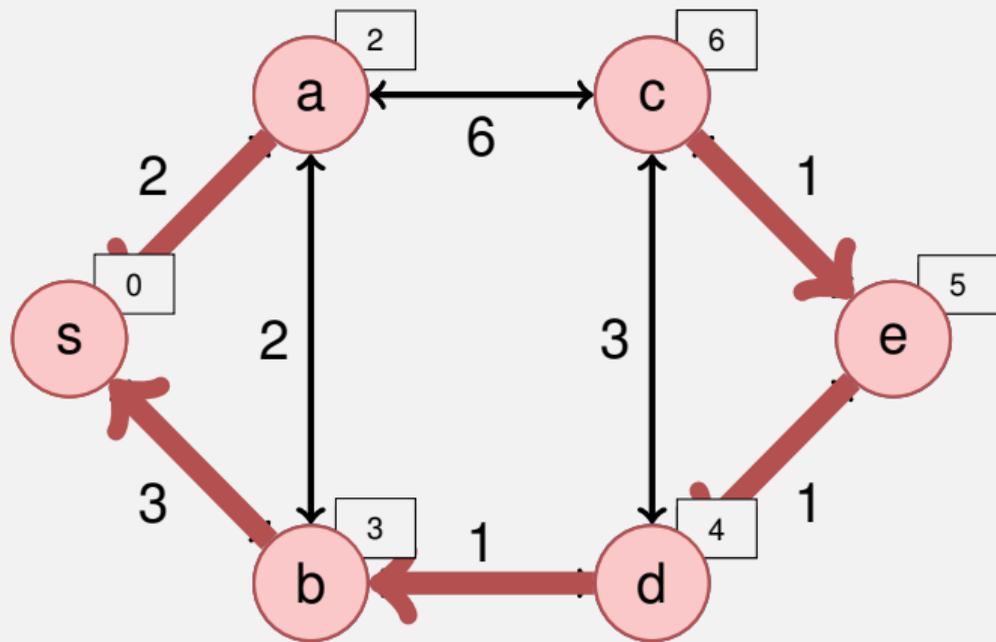


$$M = \{s, a, b, d, e\}$$

$$R = \{c\}$$

$$U = \{\}$$

# Example



$$M = \{s, a, b, d, e, c\}$$

$$R = \{\}$$

$$U = \{\}$$

# Implementation: Data Structure for $R$ ?

Required operations:

- Insert (add to  $R$ )
- ExtractMin (over  $R$ ) and DecreaseKey (Update in  $R$ )

**foreach**  $v \in N^+(u)$  **do**

**if**  $d_s[u] + c(u, v) < d_s[v]$  **then**

$d_s[v] \leftarrow d_s[u] + c(u, v)$

$\pi_s[v] \leftarrow u$

**if**  $v \in R$  **then**

            DecreaseKey( $R, v$ )

        // Update of a  $d(v)$  in the heap of  $R$

**else**

$R \leftarrow R \cup \{v\}$

        // Update of  $d(v)$  in the heap of  $R$

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     $\pi_s[v] \leftarrow u$ 
    if  $v \in R$  then
      DecreaseKey( $R, v$ )           // Update of a  $d(v)$  in the heap of  $R$ 
    else
       $R \leftarrow R \cup \{v\}$     // Update of  $d(v)$  in the heap of  $R$ 
```

MinHeap!

# DecreaseKey

- DecreaseKey: climbing in MinHeap in  $\mathcal{O}(\log |V|)$
- Position in the heap?

---

<sup>24</sup>For lazy deletion a pair of edge (or target node) and distance is required.

# DecreaseKey

- DecreaseKey: climbing in MinHeap in  $\mathcal{O}(\log |V|)$
- Position in the heap?
  - alternative (a): Store position at the nodes

---

<sup>24</sup>For lazy deletion a pair of edge (or target node) and distance is required.

# DecreaseKey

- DecreaseKey: climbing in MinHeap in  $\mathcal{O}(\log |V|)$
- Position in the heap?
  - alternative (a): Store position at the nodes
  - alternative (b): Hashtable of the nodes

---

<sup>24</sup>For lazy deletion a pair of edge (or target node) and distance is required.

# DecreaseKey

- DecreaseKey: climbing in MinHeap in  $\mathcal{O}(\log |V|)$
- Position in the heap?
  - alternative (a): Store position at the nodes
  - alternative (b): Hashtable of the nodes
  - alternative (c): re-insert node after successful relax operation and mark it "deleted" once extracted (Lazy Deletion).<sup>24</sup>

---

<sup>24</sup>For lazy deletion a pair of edge (or target node) and distance is required.

# Runtime

- $|V| \times$  ExtractMin:  $\mathcal{O}(|V| \log |V|)$
- $|E| \times$  Insert or DecreaseKey:  $\mathcal{O}(|E| \log |V|)$
- $1 \times$  Init:  $\mathcal{O}(|V|)$
- Overall:  $\mathcal{O}(|E| \log |V|)$ .

# General Weighted Graphs

Relaxing Step as before but with a return value:

**Relax**( $u, v$ ) ( $u, v \in V, (u, v) \in E$ )

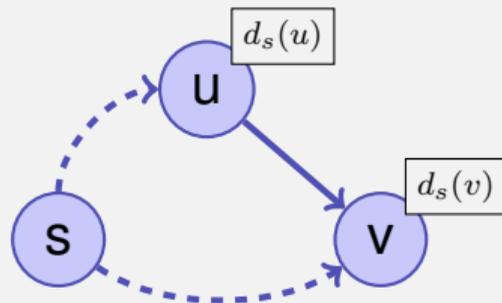
**if**  $d_s[u] + c(u, v) < d_s[v]$  **then**

$d_s[v] \leftarrow d_s[u] + c(u, v)$

$\pi_s[v] \leftarrow u$

**return true**

**return false**



Problem: cycles with negative weights can shorten the path, a shortest path is not guaranteed to exist.

# Dynamic Programming Approach (Bellman)

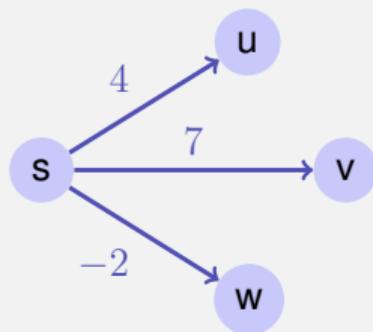
Induction over number of edges  $d_s[i, v]$ : Shortest path from  $s$  to  $v$  via maximally  $i$  edges.

$$d_s[i, v] = \min\{d_s[i - 1, v], \min_{(u,v) \in E} (d_s[i - 1, u] + c(u, v))\}$$

$$d_s[0, s] = 0, d_s[0, v] = \infty \quad \forall v \neq s.$$

# Dynamic Programming Approach (Bellman)

	$s$	$\dots$	$v$	$\dots$	$w$
$0$	$0$	$\infty$	$\infty$	$\infty$	$\infty$
$1$	$0$	$\infty$	$7$	$\infty$	$-2$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$n - 1$	$0$	$\dots$	$\dots$	$\dots$	$\dots$



Algorithm: Iterate over last row until the relaxation steps do not provide any further changes, maximally  $n - 1$  iterations. If still changes, then there is no shortest path.

# Algorithm Bellman-Ford( $G, s$ )

**Input:** Graph  $G = (V, E, c)$ , starting point  $s \in V$

**Output:** If return value true, minimal weights  $d$  for all shortest paths from  $s$ , otherwise no shortest path.

**foreach**  $u \in V$  **do**

$d_s[u] \leftarrow \infty; \pi_s[u] \leftarrow \text{null}$

$d_s[s] \leftarrow 0;$

**for**  $i \leftarrow 1$  **to**  $|V|$  **do**

$f \leftarrow \text{false}$

**foreach**  $(u, v) \in E$  **do**

$f \leftarrow f \vee \text{Relax}(u, v)$

**if**  $f = \text{false}$  **then return** true

**return** false;