



# Shortest Path Algorithm

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B2-206



# Menu

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- Introduction to graphs, definitions.
- Finding a path.
- Shortest-path problem & algorithm.
  - Bellman-Ford.
  - Dijkstra.

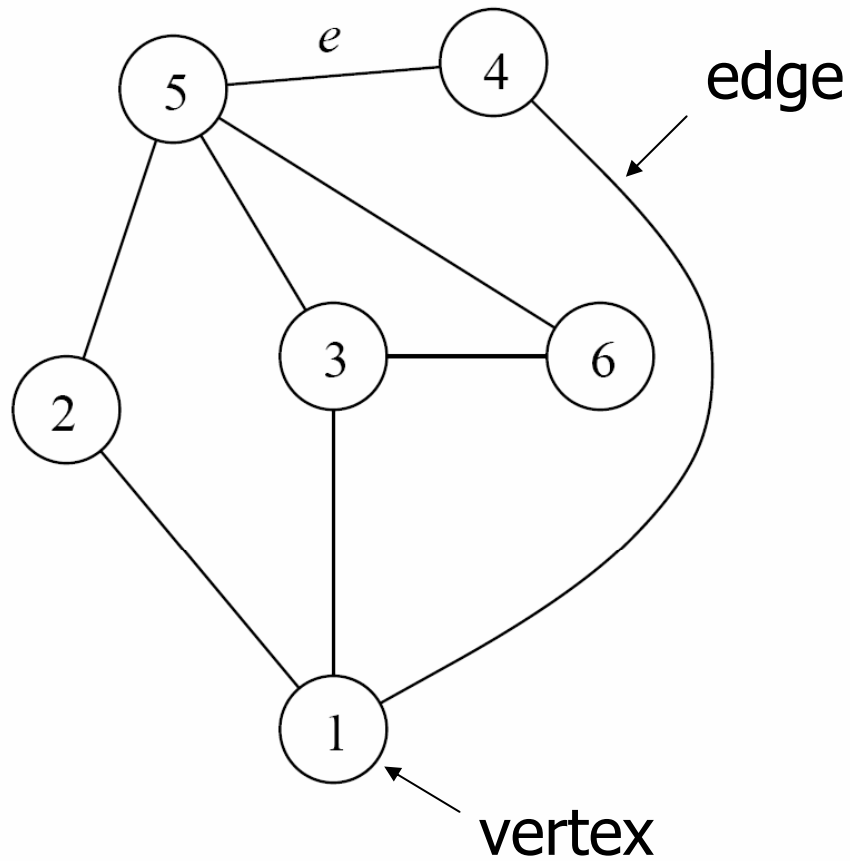


# Graphs – Definition

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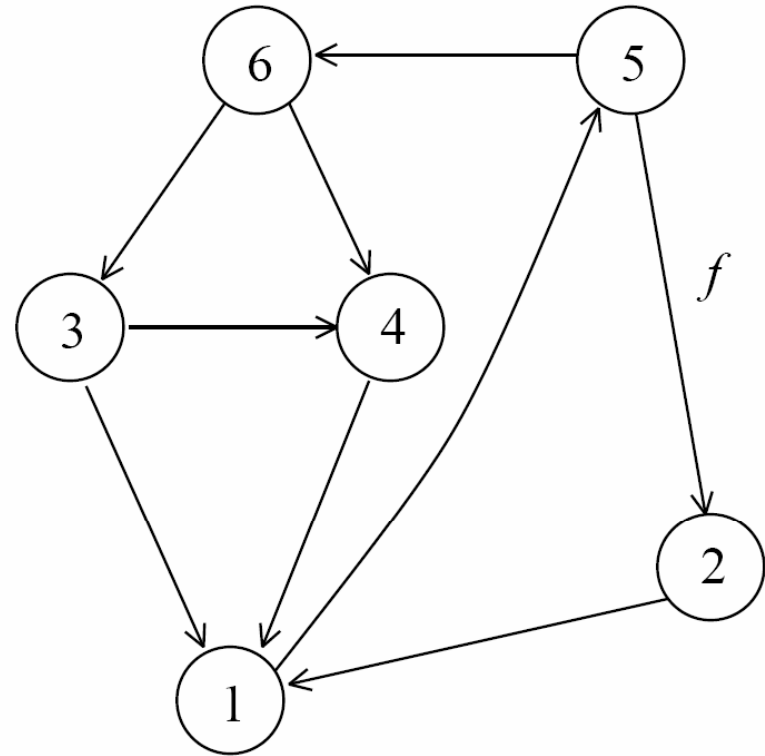
- A graph is a pair  $(V, E)$ 
  - $V$  finite set of vertices.
  - $E$  finite set of edges.  
 $e \in E$  is a pair  $(u, v)$  of vertices.  
Ordered pair  $\rightarrow$  directed graph.  
Unordered pair  $\rightarrow$  undirected graph.

$V=$   
 $E=$



(a)

$V=$   
 $E=$



(b)

**Figure 10.1** (a) An undirected graph and (b) a directed graph.



# Graphs – Edges

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- Directed graph:
  - $(u, v) \in E$  is incident **from**  $u$  and incident **to**  $v$ .
  - $(u, v) \in E$ : vertex  $v$  is adjacent to  $u$ .
- Undirected graph:
  - $(u, v) \in E$  is incident **on**  $u$  and  $v$ .
  - $(u, v) \in E$ : vertices  $u$  and  $v$  are adjacent to each other.

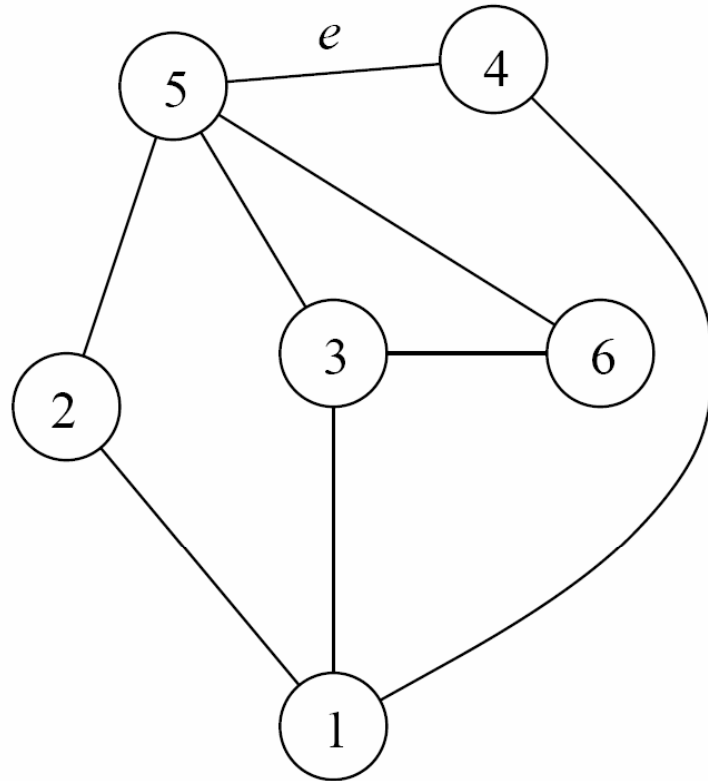


# Graphs – Paths

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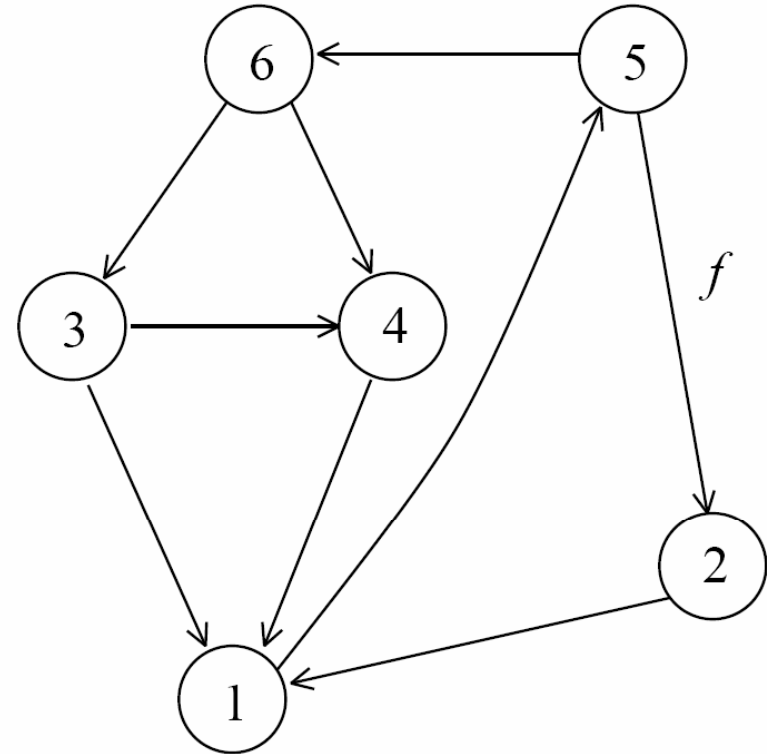
- A path is a sequence of adjacent vertices.
  - **Length** of a path = number of edges.
  - Path from  $v$  to  $u \Rightarrow u$  is **reachable** from  $v$ .
  - **Simple** path: All vertices are distinct.
  - A path is a **cycle** if its starting and ending vertices are the same.
  - **Simple cycle**: All intermediate vertices are distinct.

Simple path:  
Simple cycle:  
Non simple cycle:



(a)

Simple path:  
Simple cycle:  
Non simple cycle:



(b)

**Figure 10.1** (a) An undirected graph and (b) a directed graph.



# Graphs

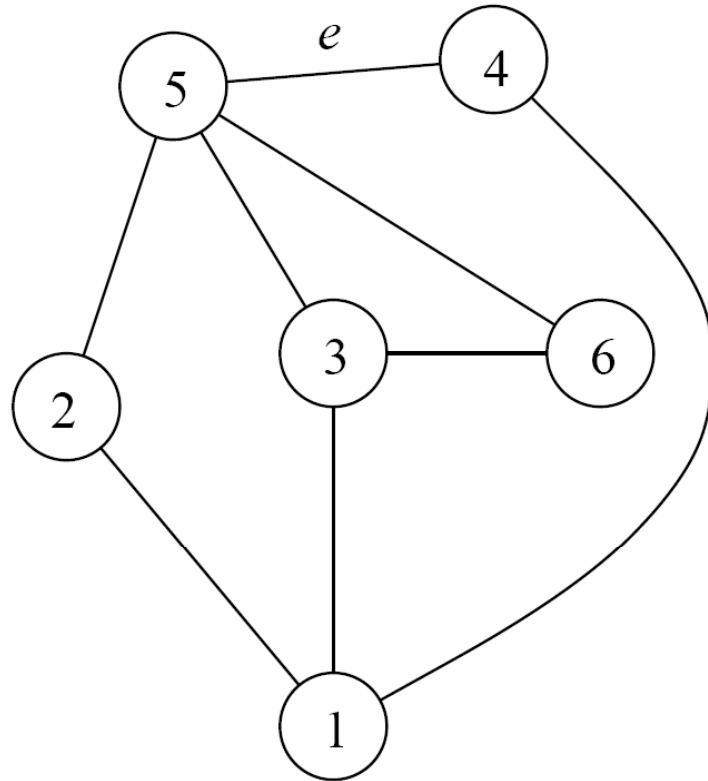
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- **Connected** graph:  $\exists$  path between any pair.
- $G'=(V',E')$  **sub-graph** of  $G=(V,E)$  if  $V' \subseteq V$  and  $E' \subseteq E$ .
- Sub-graph of  $G$  **induced** by  $V'$ : Take all edges of  $E$  connecting vertices of  $V' \subseteq V$ .
- **Complete** graph: Each pair of vertices adjacent.
- **Tree**: connected acyclic graph.

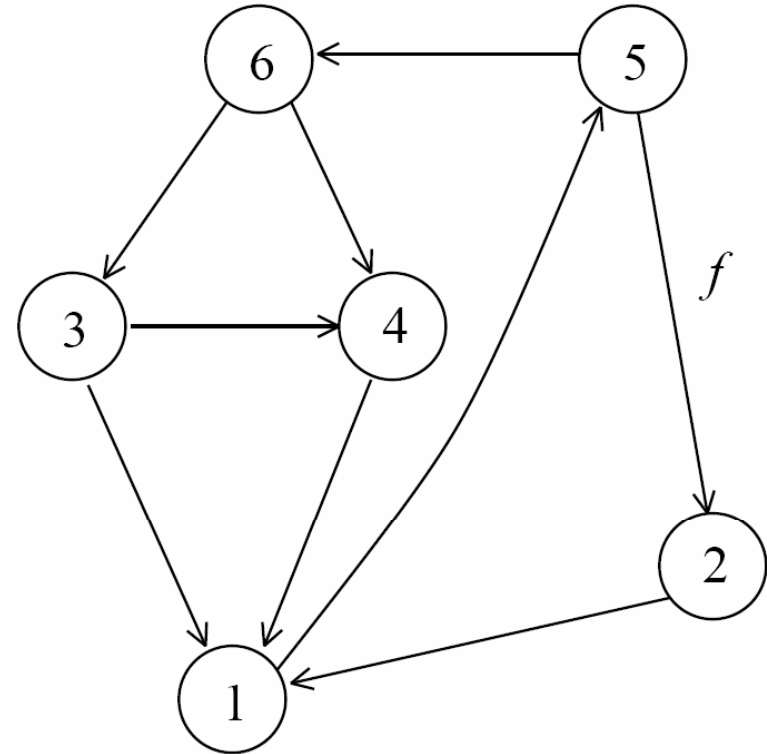


Sub-graph:

Induced sub-graph:



(a)



(b)

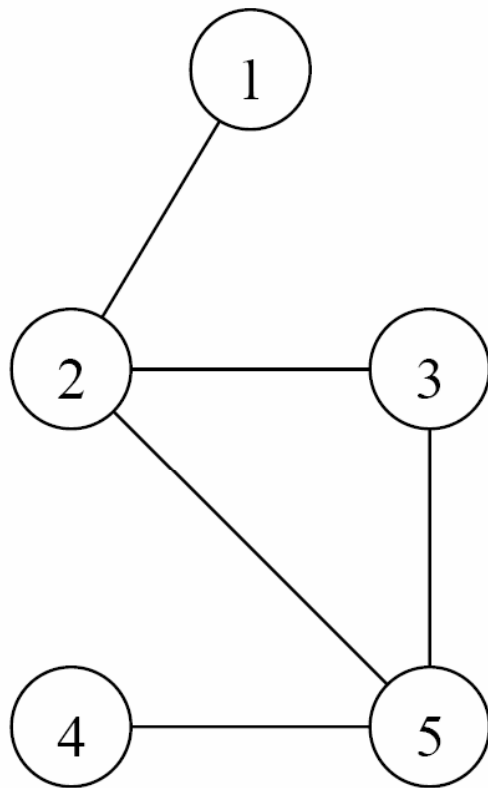
**Figure 10.1** (a) An undirected graph and (b) a directed graph.



# Graph Representation

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- Sparse graph ( $|E|$  much smaller than  $|V|^2$ ):
  - Adjacency list representation.
- Dense graph:
  - Adjacency matrix.
- For weighted graphs  $(V, E, w)$ : weighted adjacency list/matrix.



$$a_{i,j} = \begin{cases} 1 & \text{if } (v_i, v_j) \in E \\ 0 & \text{otherwise} \end{cases}$$

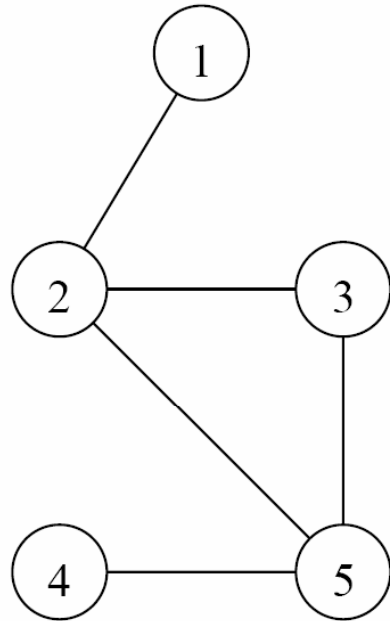
$$A = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 1 & 1 & 0 \end{bmatrix}$$

$|V|^2$  entries

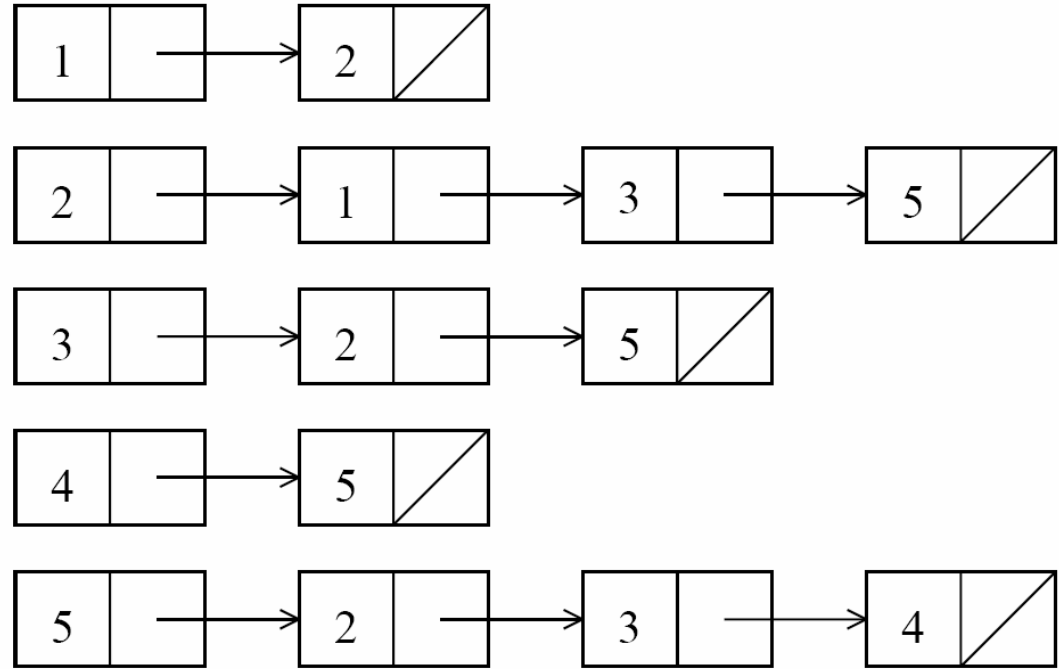
**Figure 10.2** An undirected graph and its adjacency matrix representation.

Undirected graph  $\Rightarrow$  symmetric adjacency matrix.

$|V| + |E|$  entries



$|V|$



**Figure 10.3** An undirected graph and its adjacency list representation.



## Finding a Path

- Straight-forward DFS or BFS algorithm.
- Specialized DFS version. (Call with  $S = \emptyset$ ).

```
DFS_find(G,s,t,S):
  if s ∈ S then
    return false
  fi
  push(S,s)
  if s = t then
    return true
  fi
  forall s → s' do
    if DFS_find(G,s',t,S) then
      return true
    fi
  done
  pop(s)
  return false
```



# Shortest Path Problem

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- Given a weighted directed graph  $G=(V,E)$  with weight function  $w : E \rightarrow \mathbf{R}$ , the **weight of a path**  $p = \langle v_0 \dots v_k \rangle$  is defined by

$$w(p) = \sum_{i=0}^k w(v_{i-1}, v_i)$$

- The **shortest-path weight** from  $u$  to  $v$  is defined by  $\delta(u,v) = \min\{w(p) : \text{there is a path from } u \text{ to } v\}$ ,  $\infty$  otherwise.
- A **shortest path** from vertex  $u$  to vertex  $v$  is then defined by any path with weight  $w(p) = \delta(u,v)$ .



# Variants

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- Single-source shortest-paths: from a source to every vertex.
- Single-destination shortest-paths: from every vertex to a destination.
- Single-pair shortest path: between a pair of vertices  $u$  and  $v$ .
- All-pairs shortest-paths: for all pairs of vertices.

# Optimal Sub-structure of Shortest Paths

- Shortest-paths algorithm rely on the property that a shortest path between 2 vertices *contains other shortest paths within it*.
- **Lemma:** Let  $p = \langle v_0 \dots v_k \rangle$  be a shortest path from  $v_0$  to  $v_k$ . For any  $i, j$   $0 \leq i \leq j \leq k$ ,  $p_{ij} = \langle v_i \dots v_j \rangle$  is a shortest path from  $v_i$  to  $v_j$ .
  - Proof technique: Suppose it is not the case and obtain a contradiction with the hypothesis.





# Negative Weight Cycles

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- Pose problems to define shortest-paths.
- We suppose we do not have negative weight cycles otherwise the shortest-path is not well-defined.
  - Stay in the cycle and get  $-\infty$  as the sum.
- Other cycles can be removed without loss of generality (if weight=0, otherwise not shortest).



# Representation

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- We want to compute shortest-path weights and the vertices on the path. For a graph  $G=(V,E)$ 
  - $\pi(v)$  is a predecessor of  $v \in V$ , or NIL.
  - $\pi$  values induce the predecessor sub-graph  $G_\pi=(V_\pi,E_\pi)$ .  
 $V_\pi=\{v \in V : \pi(v) \neq \text{NIL}\} \cup \{s\}$ . (+ source  $s$ )  
 $E_\pi=\{(\pi(v),v) \in E : v \in V_\pi - \{x\}\}$ .
  - The shortest-path algorithm computes  $\pi$  and the result is a "shortest-path tree".

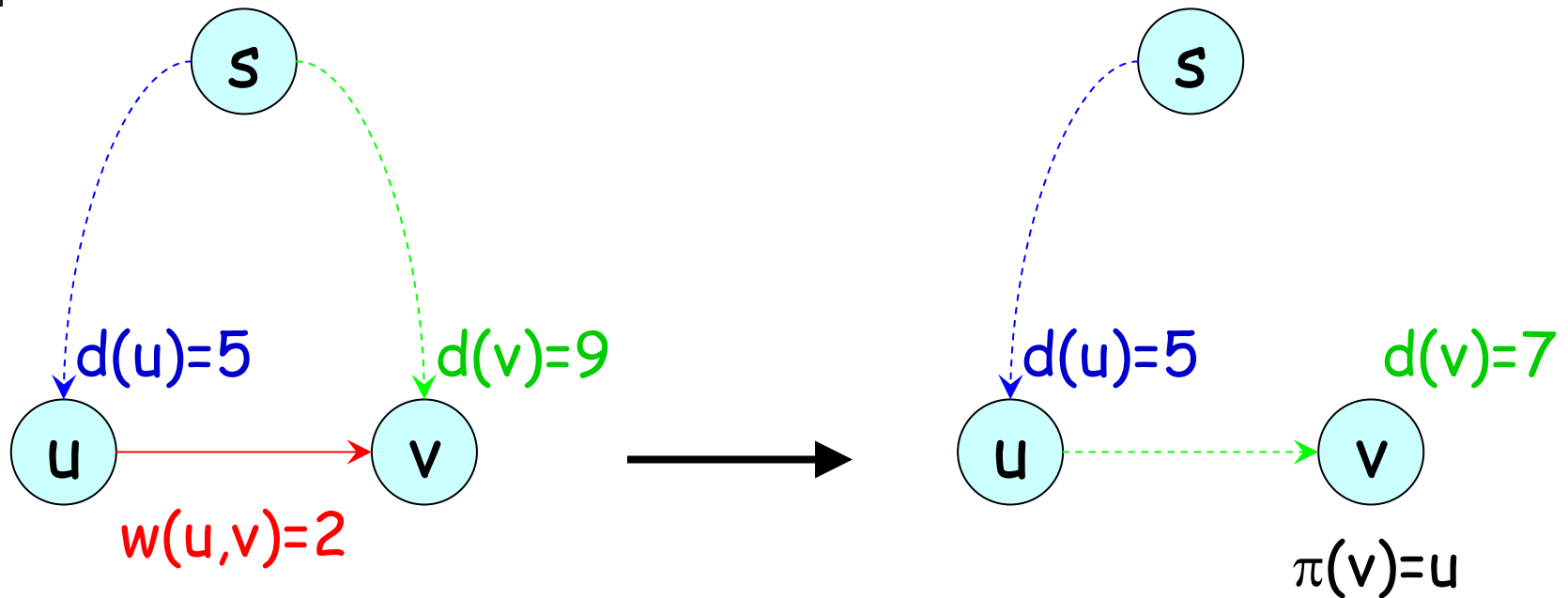
# Tightening – Relaxation (Historical Reasons)

- Attribute  $d(v)$  is the current known shortest path weight to  $v$ , i.e., it's an upper-bound on **the** shortest path weight.

```
Initialize_single_source( $G, s$ ):  
forall  $v \in V(G)$  do  
     $d(v) = \infty$   
     $\pi(v) = \text{NIL}$   
done  
 $d(s) = 0$ 
```

```
Relax( $u, v, w$ ):  
if  $d(v) > d(u) + w(u, v)$  then  
     $d(v) = d(u) + w(u, v)$   
     $\pi(v) = u$   
fi
```

# Tightening



Shorter to  $v$  via  $u$ :  
 $d(u)+w(u,v) < d(v)$ .

# Single-Source Shortest-Paths Algorithms



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- Bellman-Ford.
  - General case with negative weights.
  - Detect if negative weight cycles are reachable.
- Dijkstra.
  - Requires positive weights.



# Bellman-Ford

Repeat →

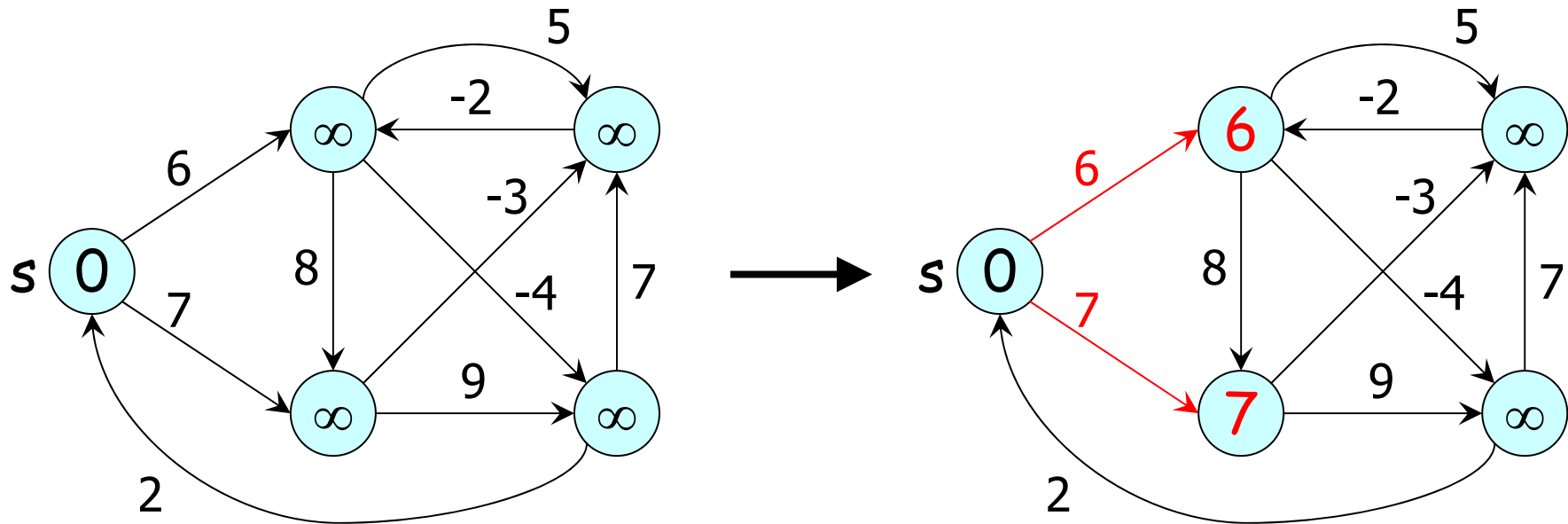
Relax all pairs  
of edges. →

$O(|V|*|E|)$

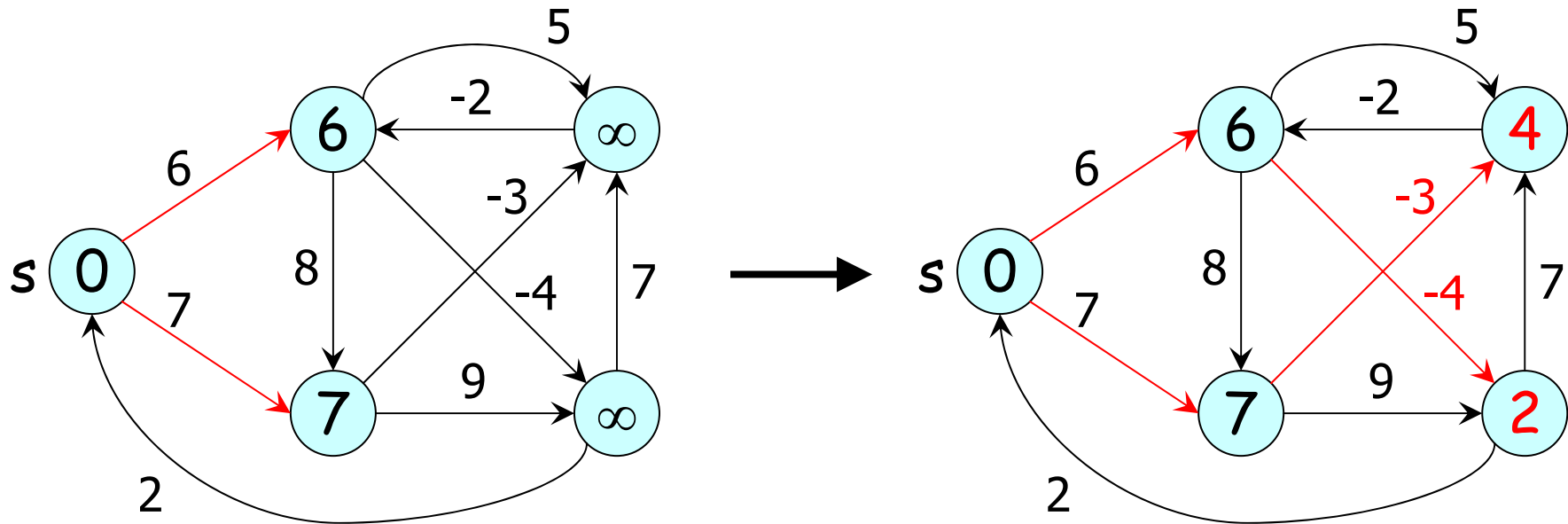
- Upon termination, either
- the algorithm converged to the shortest path,
  - or there is a negative cycle and it didn't converge.

```
Bellman_Ford(G,w,s):  
Initialize_single_source(G,s)  
for i = 1 to |V(G)|-1 do  
    forall (u,v) ∈ E(G) do  
        Relax(u,v,w)  
    done  
done  
forall (u,v) ∈ E(G) do  
    if d(v) > d(u)+w(u,v) then  
        return false  
    fi  
done  
return true
```

# Example

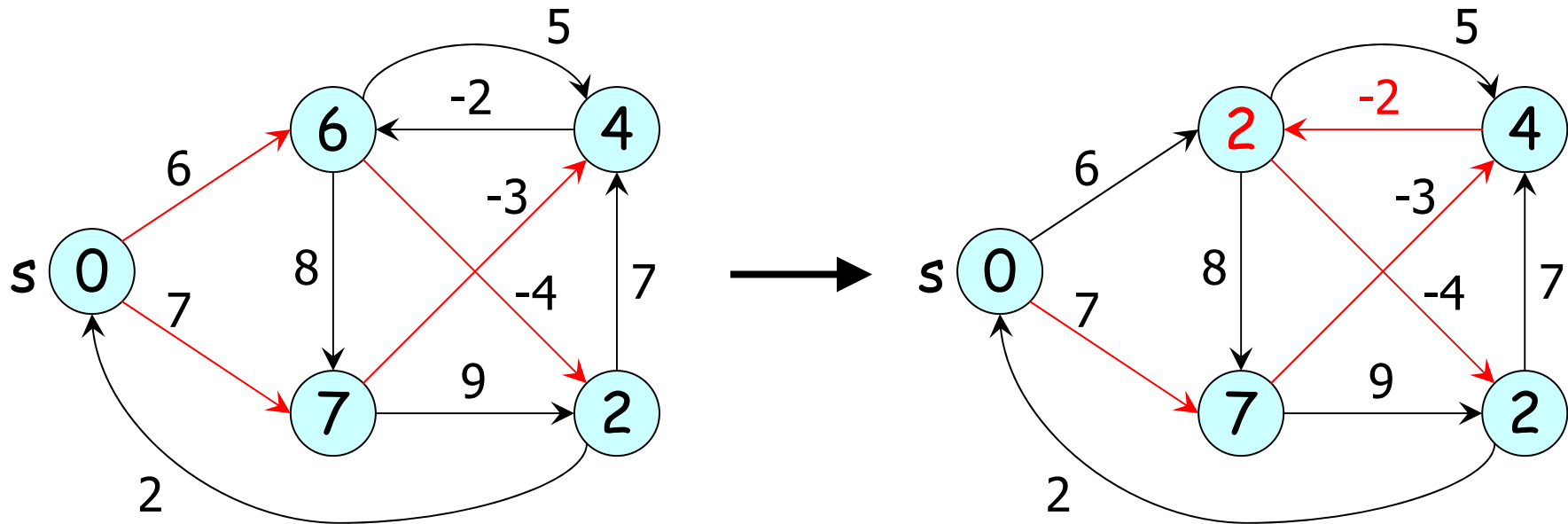


# Example

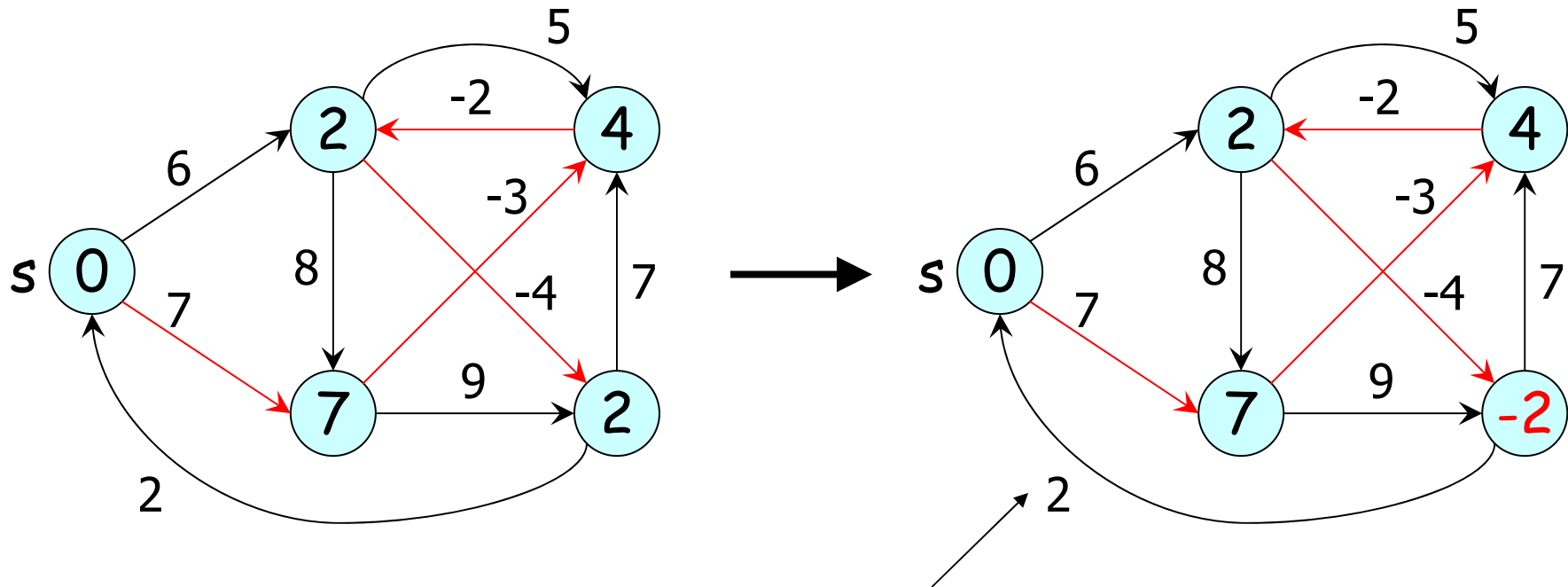




# Example



# Example



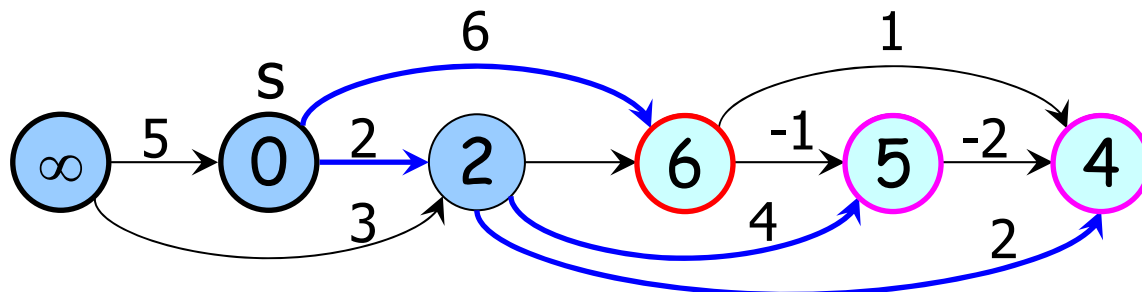
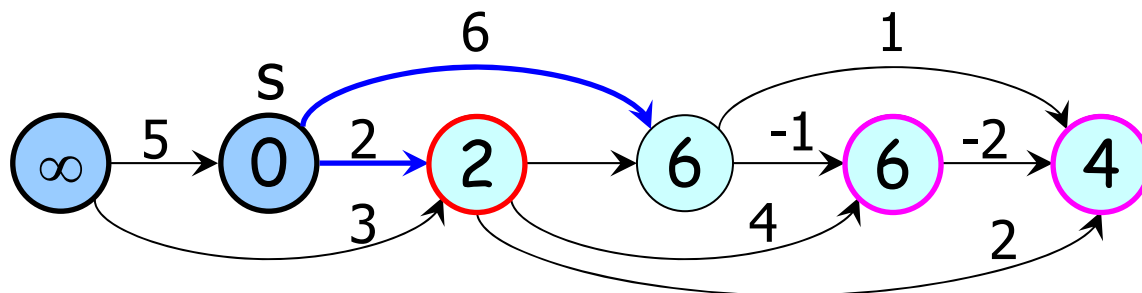
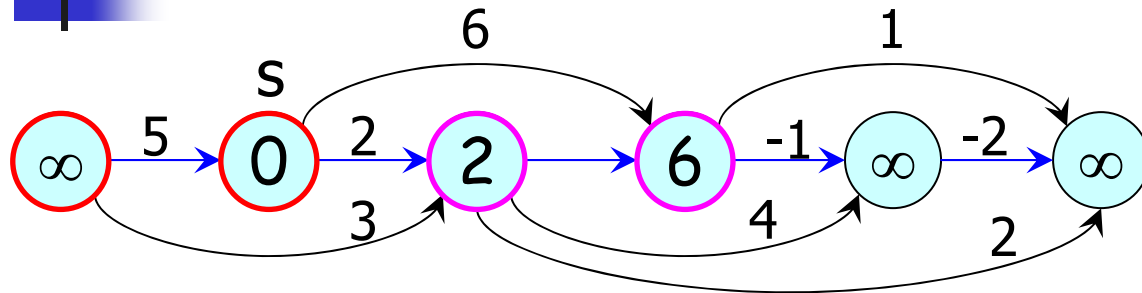
If 1 here, then  
we have a negative cycle!

# Special Case: Directed Acyclic Graphs

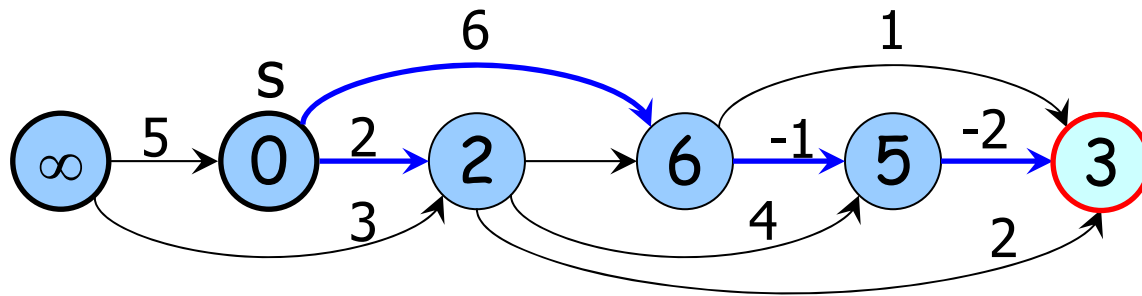
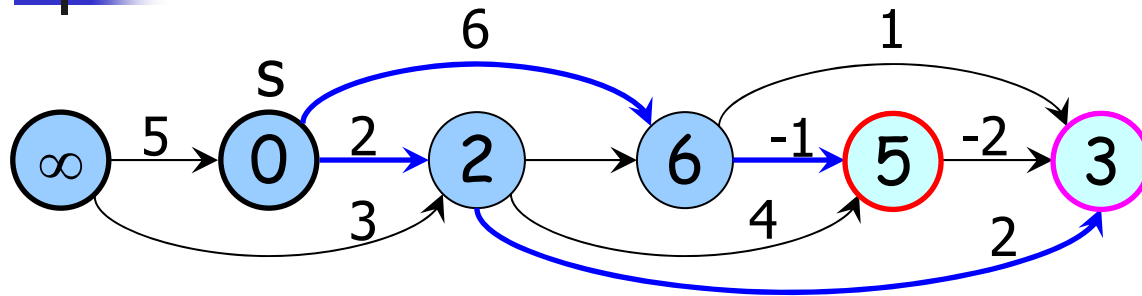
- Specialized algorithm: One pass over vertices in topologically sorted order.

```
DAG_shortest_path( $G, w, s$ ):  
Initialize_single_source( $G, s$ )  
forall  $u \in V$  in topological order do  
    forall  $v$  adjacent to  $u$  do  
        Relax( $u, v, w$ )  
    done  
done
```

# Example



# Example



# Dijkstra's Algorithm

*For non-negative weights*

- Maintains a set  $S$  of vertices with known shortest paths.
  - Select  $u \in V-S$  with minimum estimate.
  - Add  $u$  to  $S$ .
  - Relax edges leaving  $u$ .

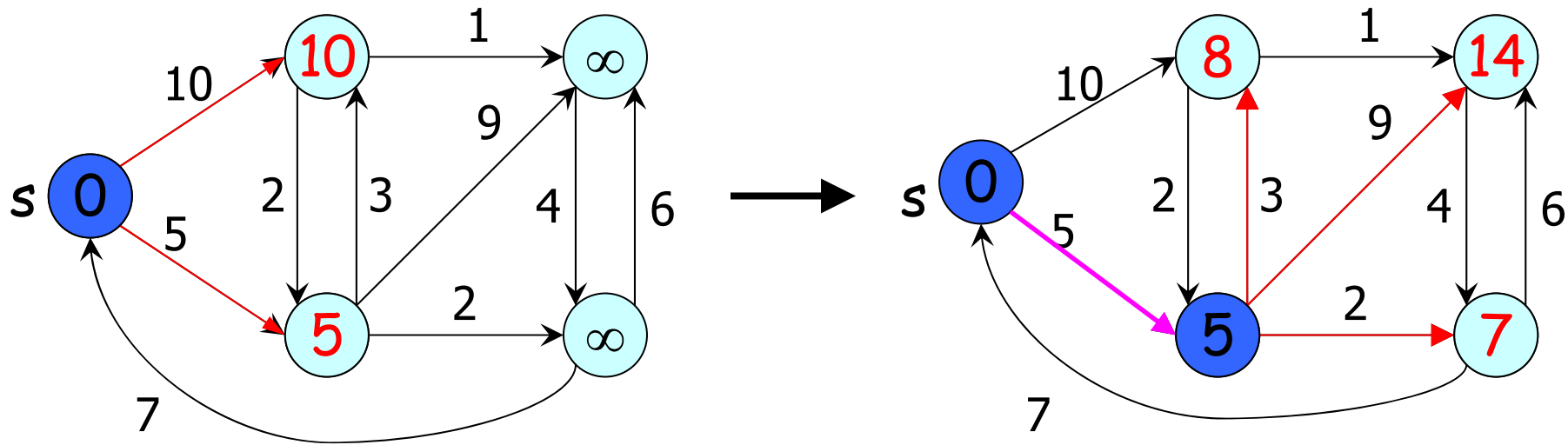
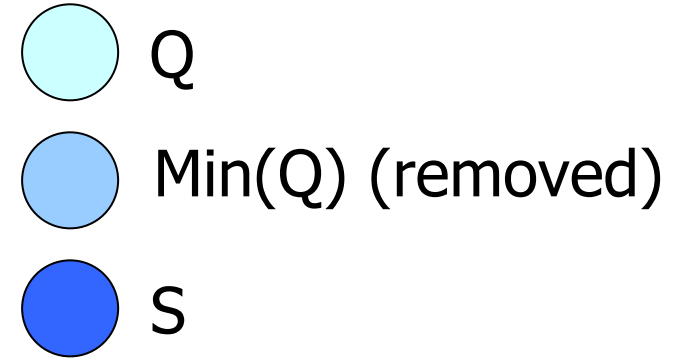


# Dijkstra's Algorithm

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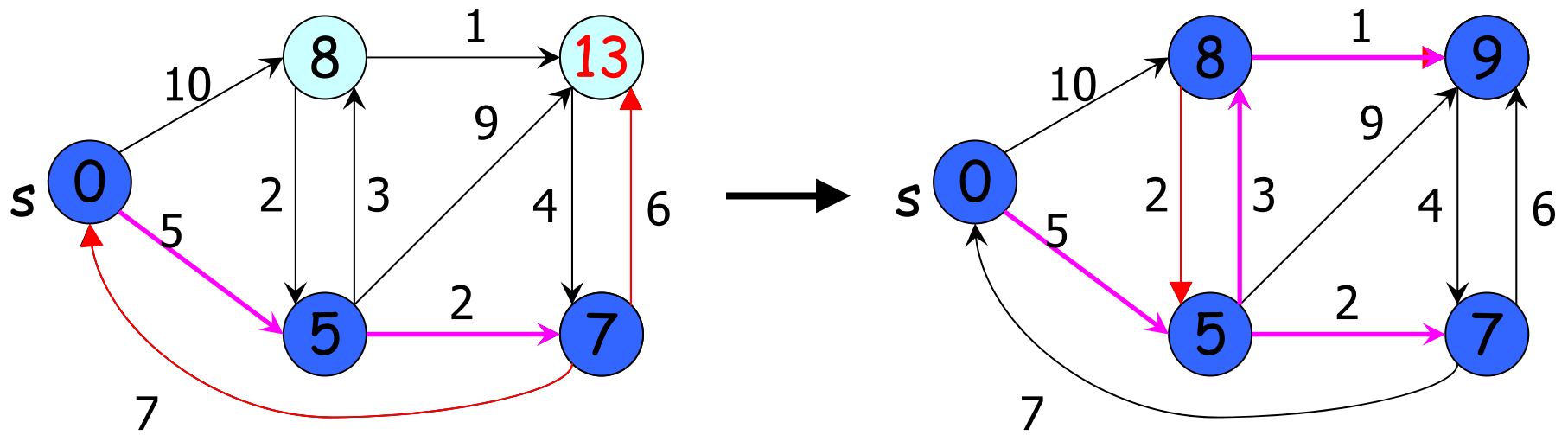
```
Dijkstra( $G, w, s$ ):  
Initialize_single_source( $G, s$ )  
 $S = \emptyset$   
 $Q = V(G)$  priority queue keyed by  $d$   
while  $Q \neq \emptyset$  do  
     $u = \text{get\_min}(Q)$   
     $S = S \cup \{u\}$   
    forall  $v$  adjacent to  $u$  do  
        Relax( $u, v, w$ )  
    done  
done
```

# Example





# Example





# Dijkstra's Algorithm

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- Efficiency: Depends on the priority queue. Can be  $O((V+E) \lg V)$ .
- Implementation:
  - Array  $d[]$  for “distance” from the source.
  - Array  $l[]$  for “last” vertex.
  - The priority queue.