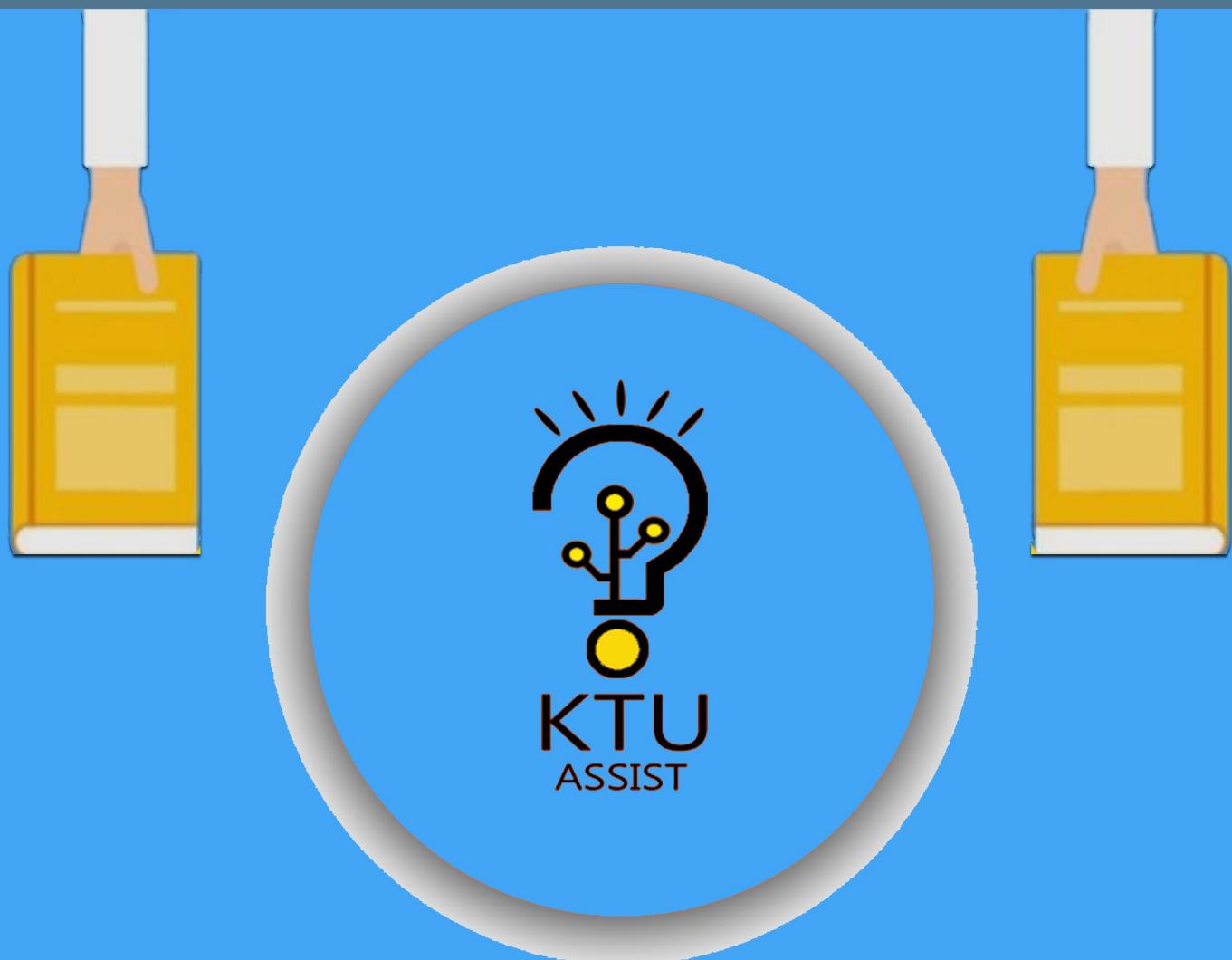


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STUDY MATERIALS



a complete app for ktu students

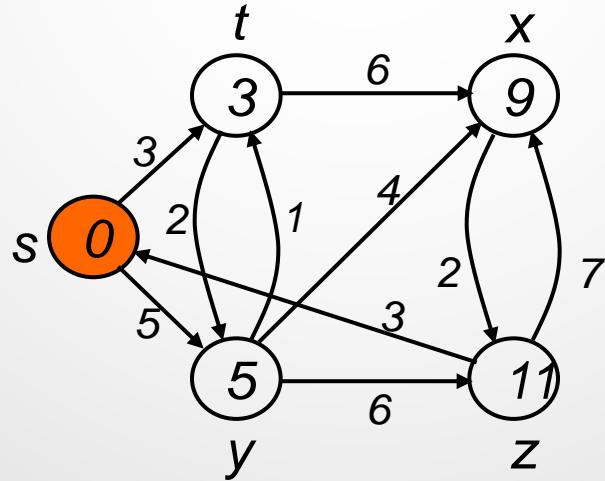
Get it on Google Play

[www.ktuassist.in](http://www.ktuassist.in)

# Single-Source Shortest Path

# Shortest Path Problems

- Directed weighted graph.
- Path length is sum of weights of edges on path.
- The vertex at which the path begins is the source vertex.
- The vertex at which the path ends is the destination vertex.



# Shortest Path Variants

- ▶ Single-source single-destination (1-1): Find the shortest path from source  $s$  to destination  $v$ .
- ▶ Single-source all-destination(1-Many): Find the shortest path from  $s$  to each vertex  $v$ .
- ▶ Single-destination shortest-paths (Many-1): Find a shortest path to a given *destination* vertex  $t$  from each vertex  $v$ .
- ▶ All-pairs shortest-paths problem (Many-Many): Find a shortest path from  $u$  to  $v$  for every pair of vertices  $u$  and  $v$ .

# Shortest Path Variants



<b>Single-Source Single-Destination (1-1)</b>	<b>Single-Source All-Destination (1-M)</b>
<ul style="list-style-type: none"><li>- No good solution that beats 1-M variant</li><li>- Thus, this problem is mapped to the 1-M variant</li></ul>	<ul style="list-style-type: none"><li>- Need to be solved (several algorithms)</li><li>- We will study this one</li></ul>
<b>All-Sources Single-Destination (M-1)</b>	<b>All-Sources All-Destinations (M-M)</b>
<ul style="list-style-type: none"><li>- Reverse all edges in the graph</li><li>- Thus, it is mapped to the (1-M) variant</li></ul>	<ul style="list-style-type: none"><li>- Need to be solved (several algorithms)</li><li>- We will study it (if time permits)</li></ul>

# Single-Source Shortest Path

- ▶ Problem: given a weighted directed graph  $G$ , find the minimum-weight path from a given source vertex  $s$  to another vertex  $v$ 
  - ▶ “Shortest-path” = minimum weight
  - ▶ Weight of path is sum of edges
  - ▶ E.g., a road map: what is the shortest path from Kunnakulam to Thrissur?

# Shortest Path Properties

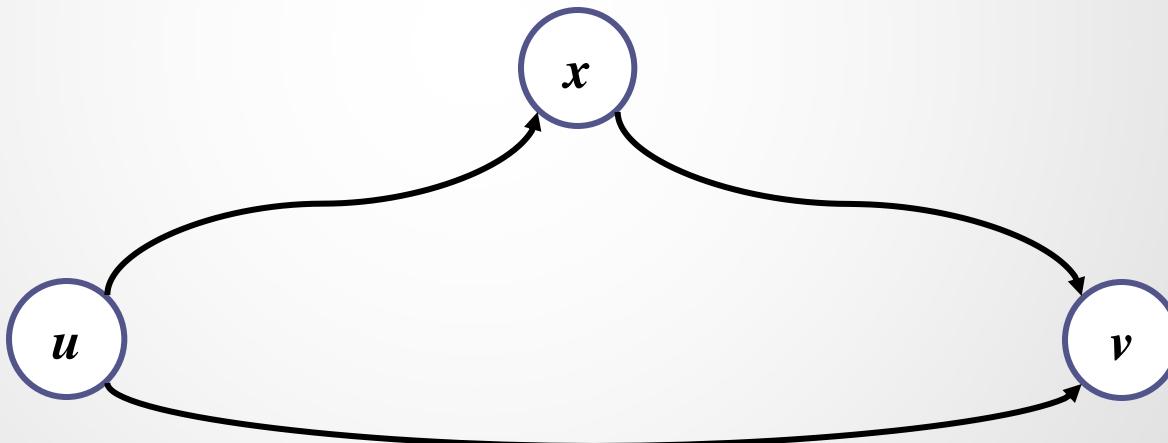
- ▶ *Optimal substructure*: the shortest path consists of shortest subpaths:



- ▶ suppose some subpath is not a shortest path
  - ▶ There must then exist a shorter subpath
  - ▶ Could substitute the shorter subpath for a shorter path
  - ▶ But then overall path is not shortest path. Contradiction

# Shortest Path Properties

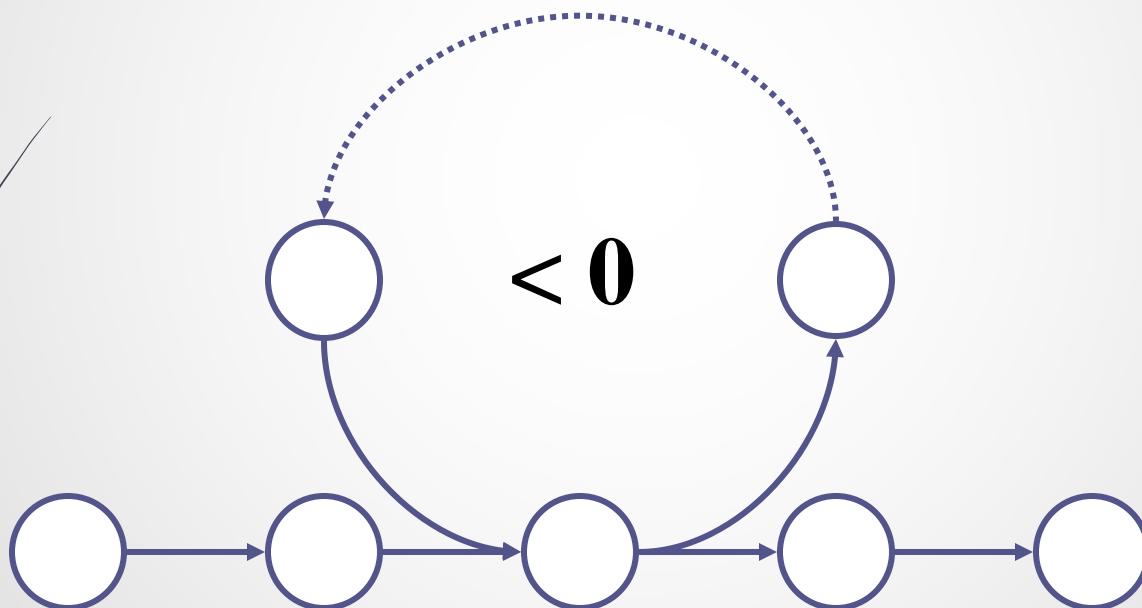
- ▶ Define  $\delta(u,v)$  to be the weight of the shortest path from  $u$  to  $v$
- ▶ *Triangle inequality:*  $\delta(u,v) \leq \delta(u,x) + \delta(x,v)$



*This path is no longer than any other path*

# Shortest Path Properties

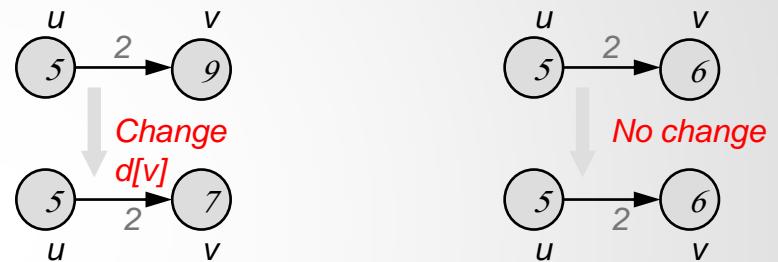
- In graphs with negative weight cycles, some shortest paths will not exist



# Relaxation

Algorithms keep track of  $d[v]$ ,  $\pi[v]$ . **Initialized** as follows:

```
Initialize( $G, s$ )
  for each  $v \in V[G]$  do
     $d[v] := \infty;$ 
     $\pi[v] := NIL$ 
  od;
   $d[s] := 0$ 
```



These values are *changed* when an edge  $(u, v)$  is **relaxed**:

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

# Bellman-Ford Algorithm

Can have negative-weight edges.

Will “detect” reachable negative-weight cycles.

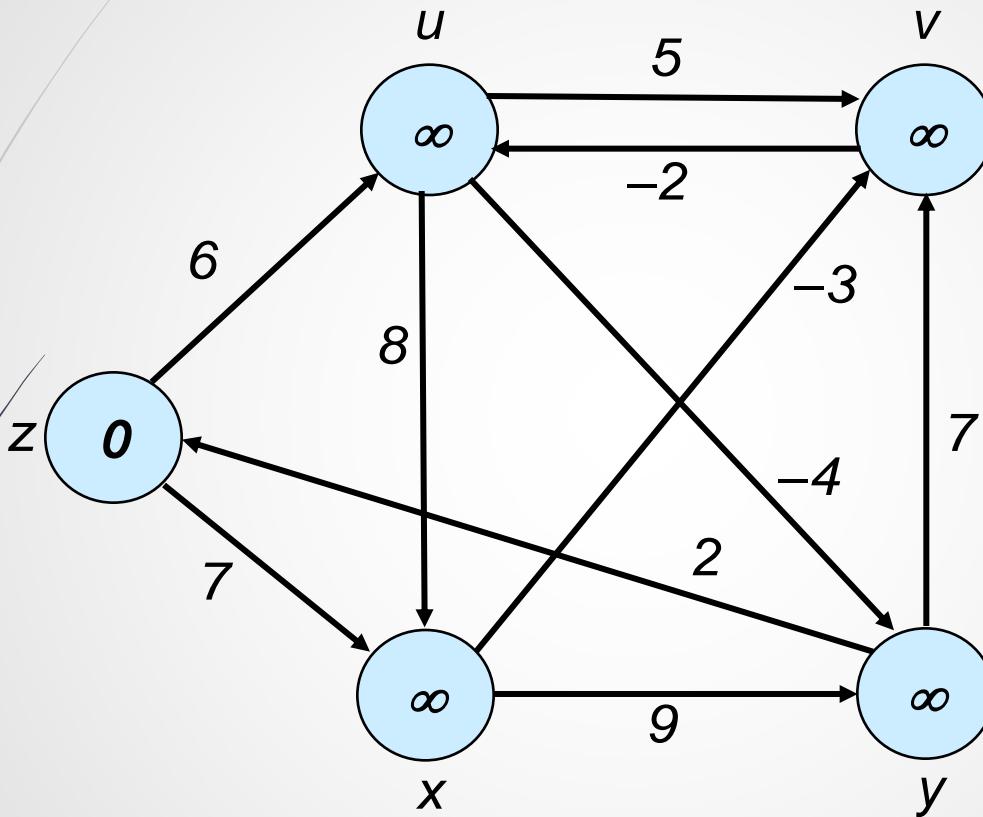
```
Initialize( $G, s$ );  
for  $i := 1$  to  $|V[G]| - 1$  do  
    for each  $(u, v)$  in  $E[G]$  do  
        Relax( $u, v, w$ )  
    od  
od;  
for each  $(u, v)$  in  $E[G]$  do  
    if  $d[v] > d[u] + w(u, v)$  then  
        return false  
    fi  
od;  
return true
```

Time  
Complexity  
is  $O(VE)$ .

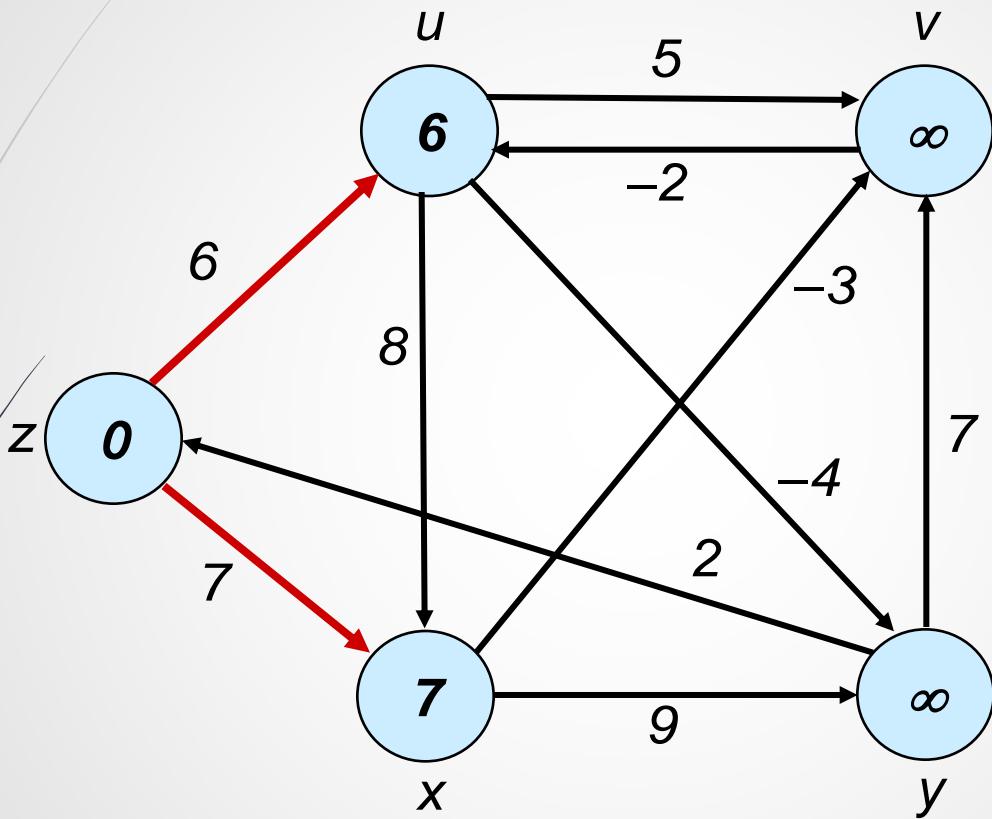
## Contd...

- So if Bellman-Ford has not converged after  $V(G) - 1$  iterations, then there cannot be a shortest path tree, so there must be a negative weight cycle.

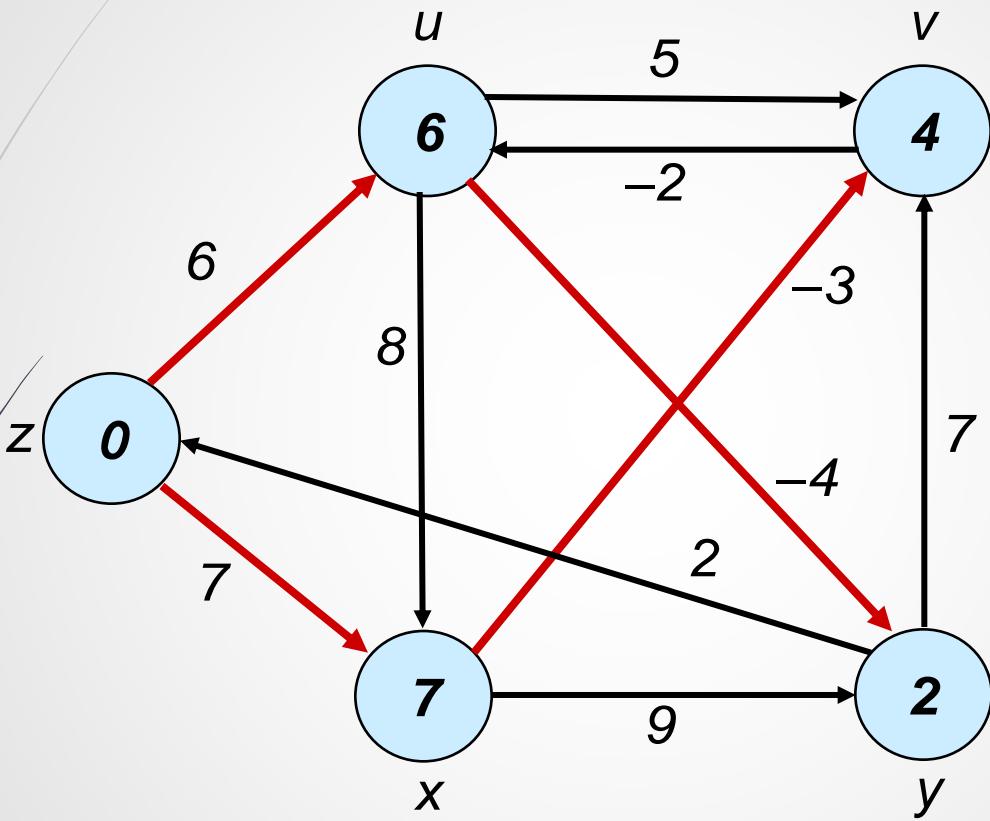
# Example



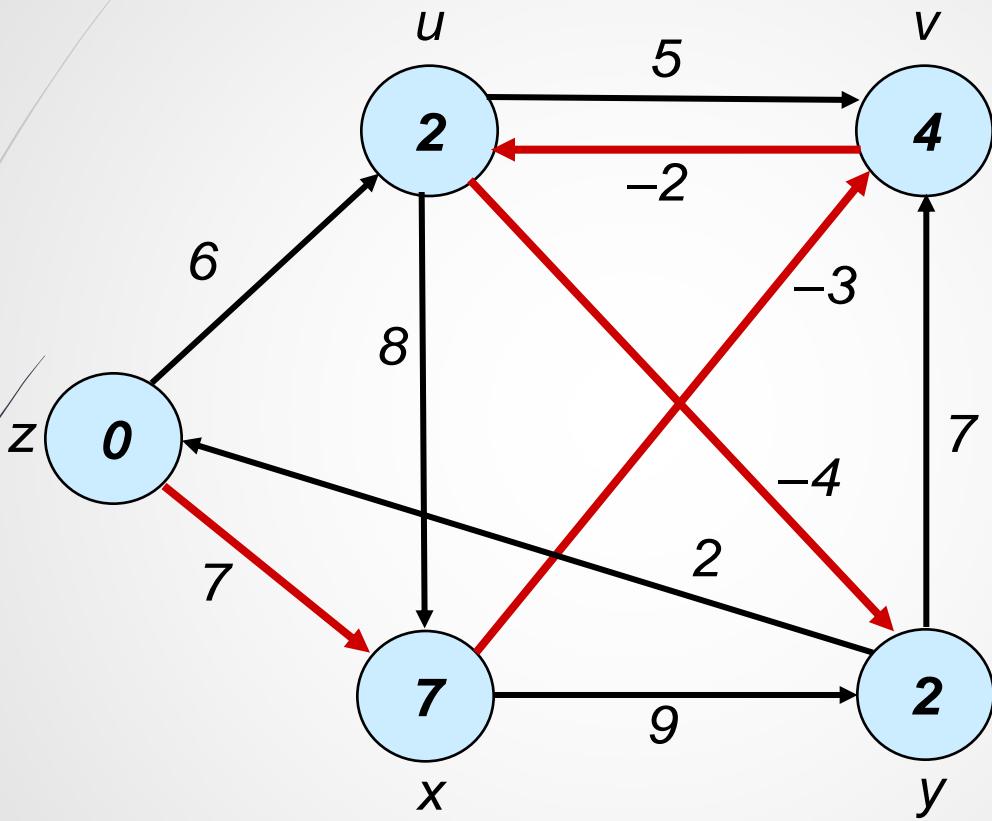
# Example



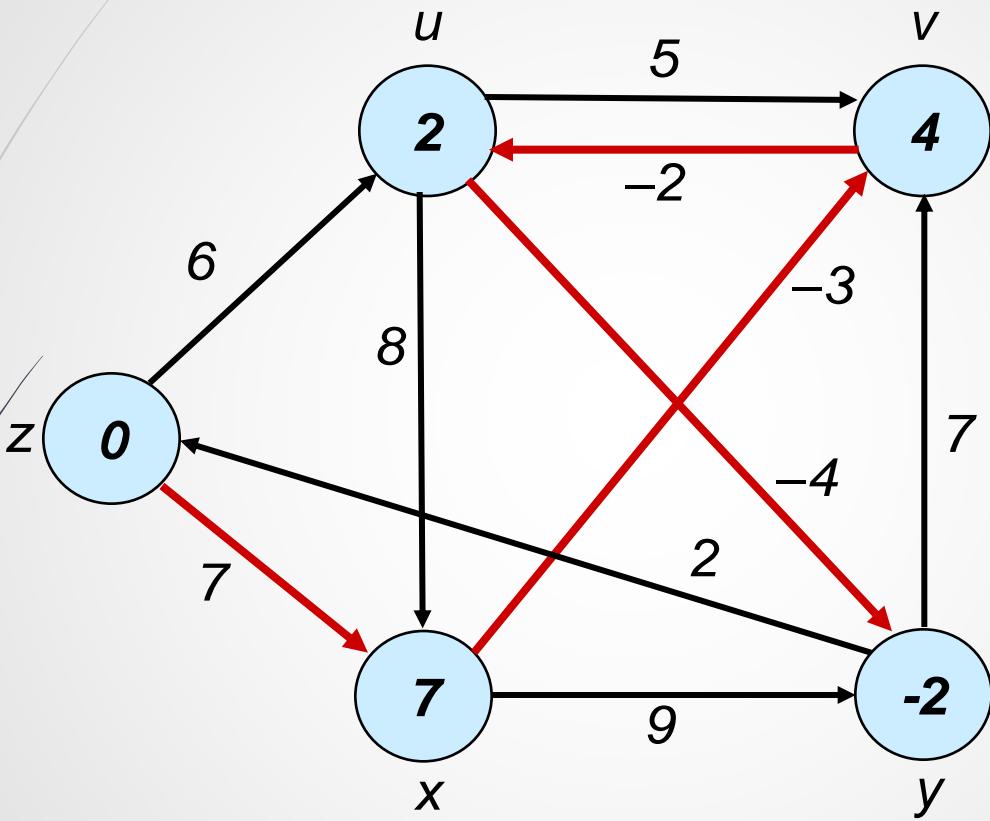
## Example



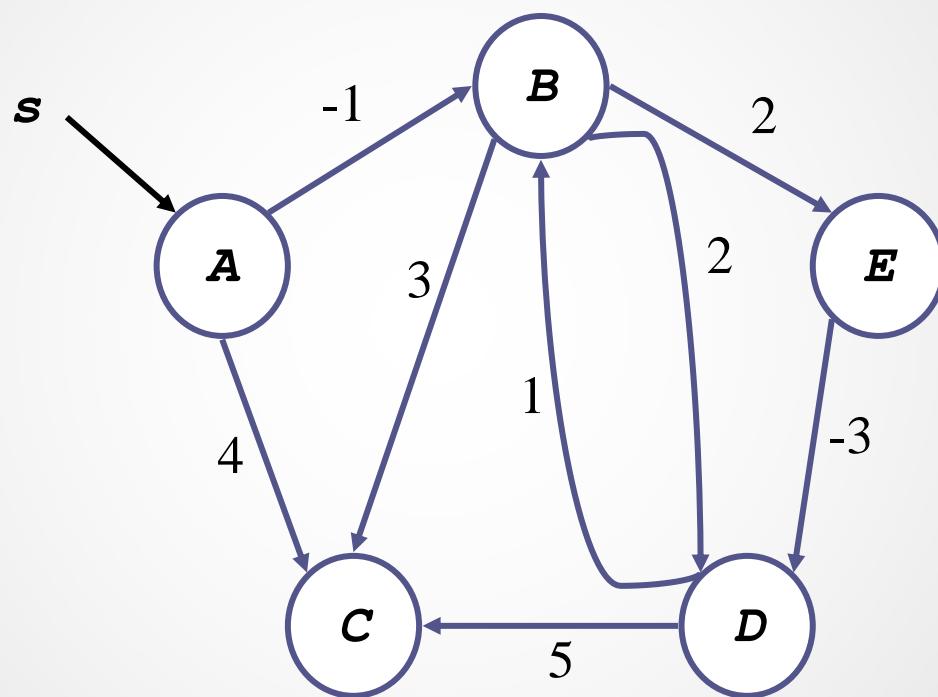
## Example



## Example



# Bellman-Ford Algorithm



# Dijkstra's Algorithm For Shortest Paths

- ▶ Non-negative edge weight
- ▶ Like BFS: If all edge weights are equal, then use BFS, otherwise use this algorithm
- ▶ Use  $Q = \text{min-priority queue}$  keyed on  $d[v]$  values

# Dijkstra's Algorithm

Assumes **no negative-weight edges**.

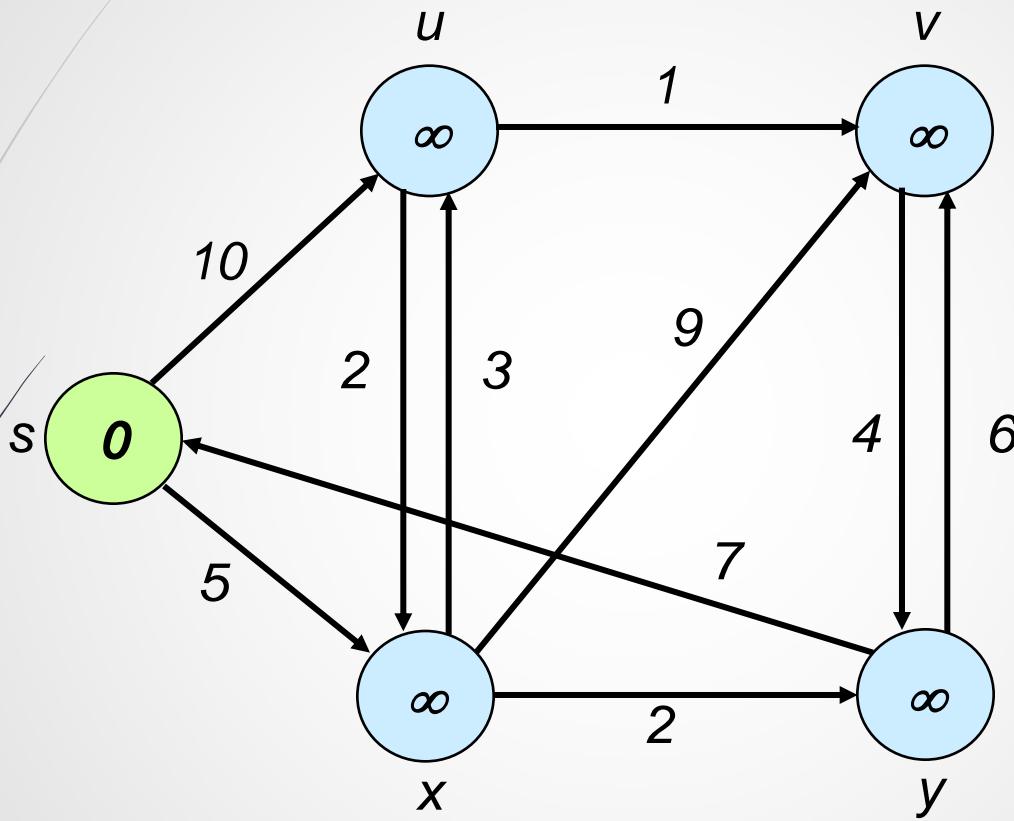
Maintains a set  $S$  of vertices whose SP from  $s$  has been determined.

Repeatedly selects  $u$  in  $V-S$  with minimum SP estimate (*greedy choice*).

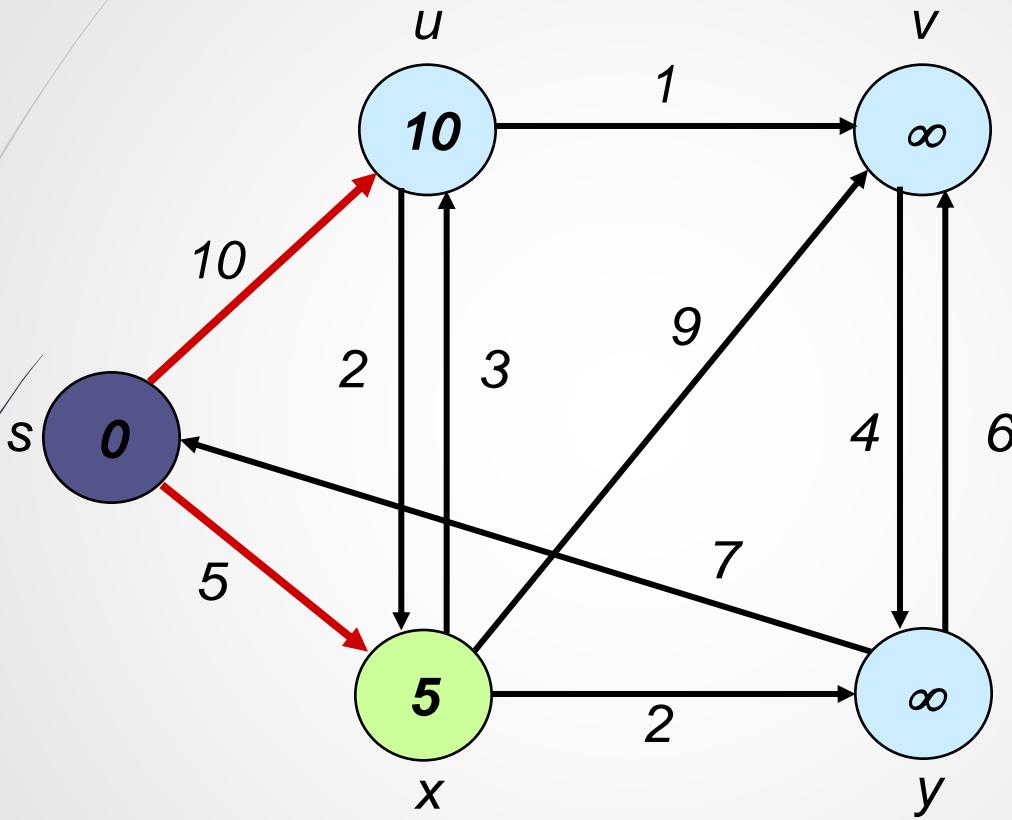
Store  $V-S$  in **priority queue  $Q$** .

```
Initialize( $G, s$ );  
 $S := \emptyset$ ;  
 $Q := V[G]$ ;  
while  $Q \neq \emptyset$  do  
     $u := Extract\text{-}Min(Q)$ ;  
     $S := S \cup \{u\}$ ;  
    for each  $v \in Adj[u]$  do  
         $Relax(u, v, w)$   
    od  
od
```

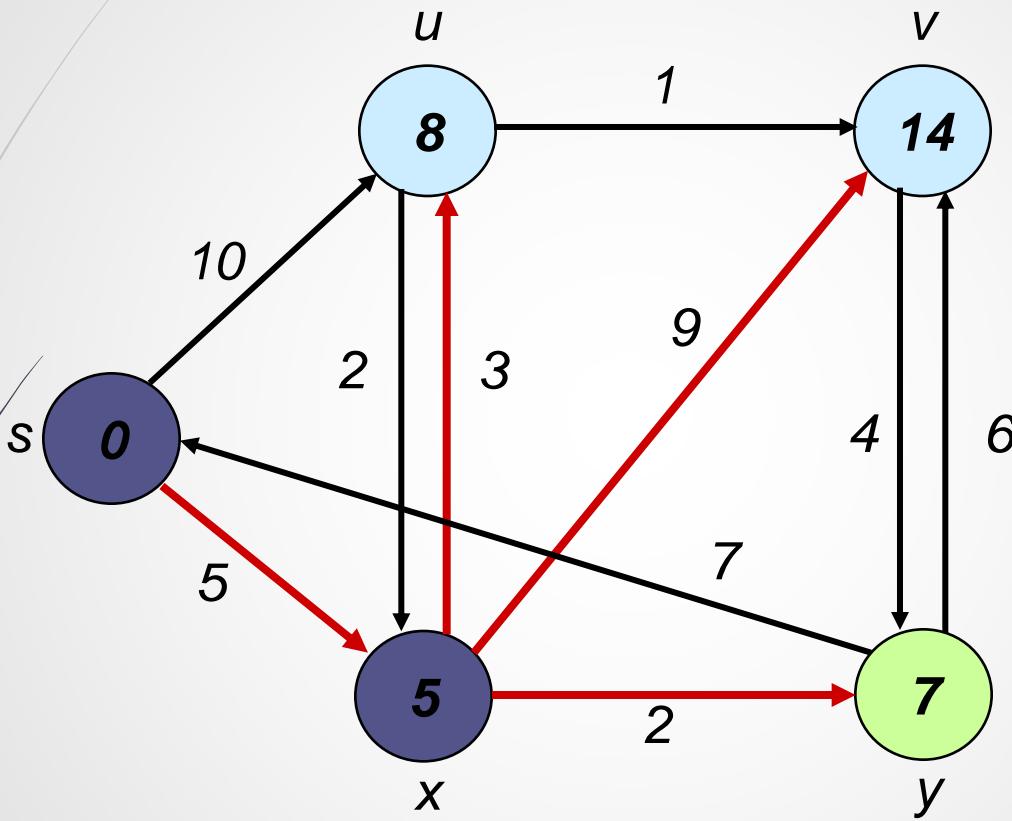
# Example



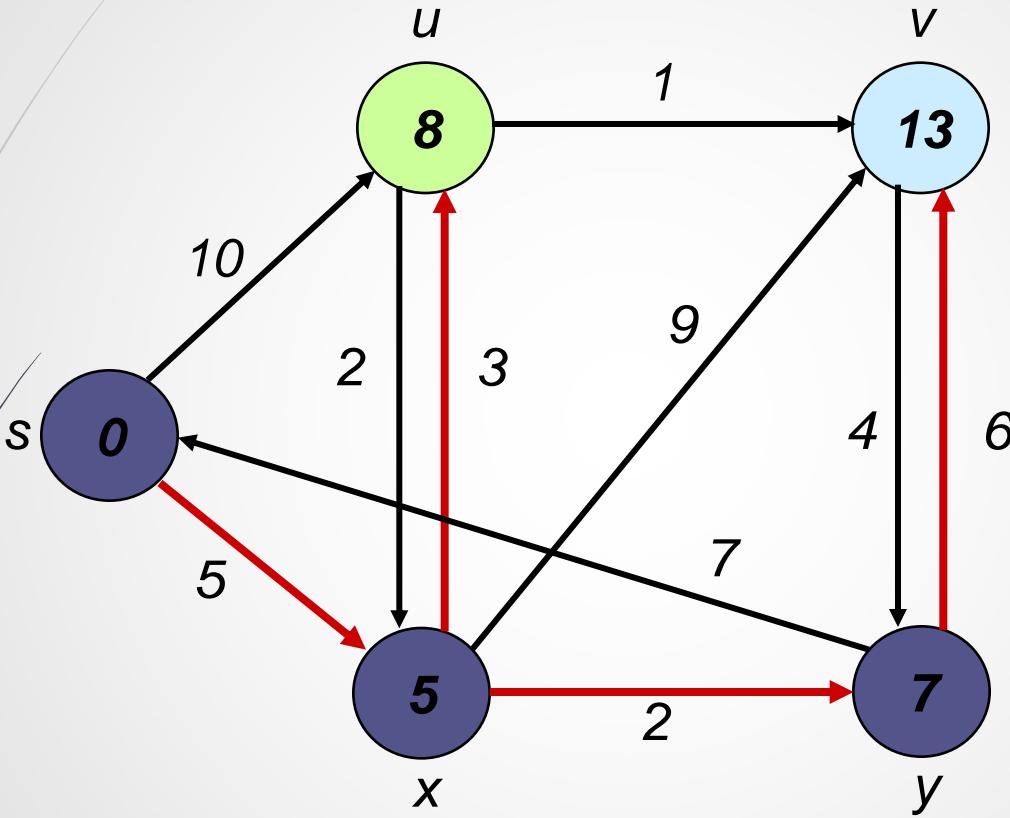
# Example



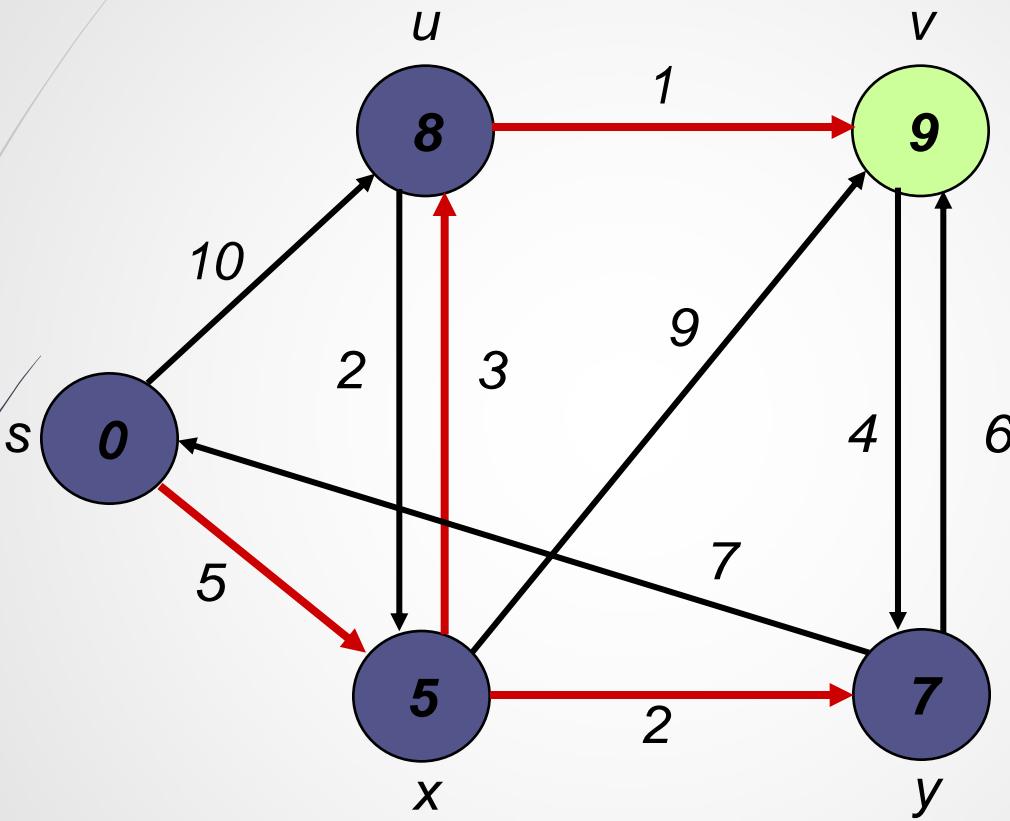
## Example



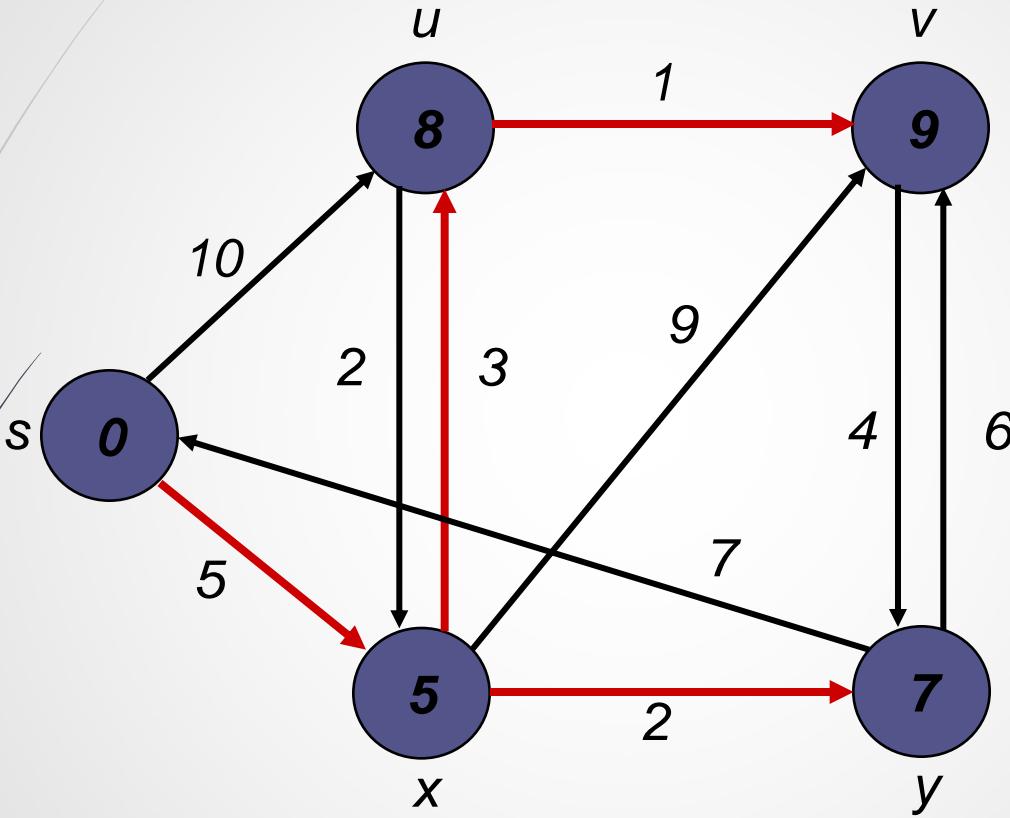
# Example



# Example

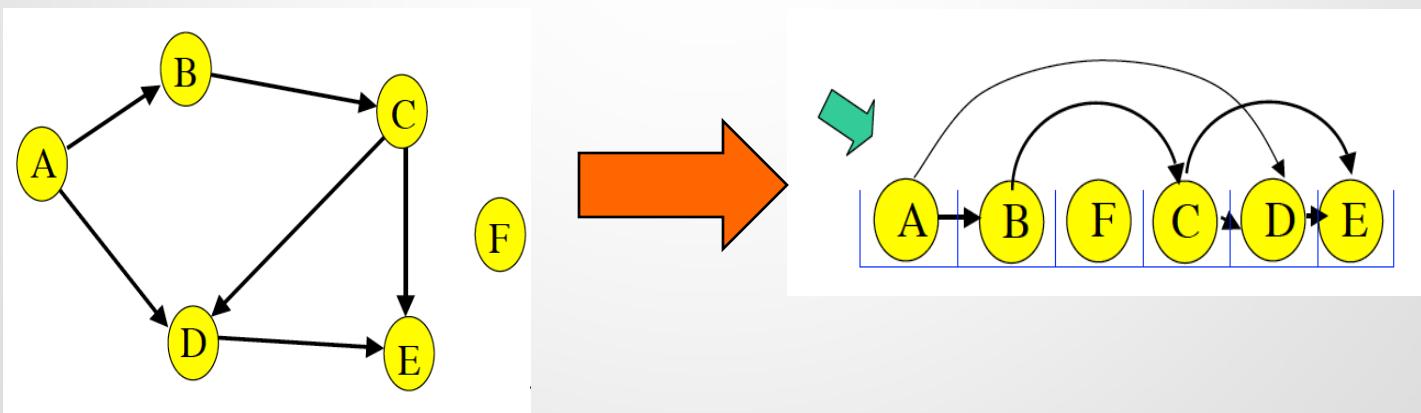


# Example



# Shortest paths in DAGs

- ▶ If there are no cycles → it is called a DAG
- ▶ In DAGs, nodes can be sorted in a linear order such that all edges are forward edges
  - ▶ Topological sort



# Single-Source Shortest Paths in DAGs

- ▶ Shortest paths are always *well-defined* in dags
  - no cycles => no negative-weight cycles even if there are negative-weight edges
- ▶ **Idea:** If we were lucky
  - To process vertices on each shortest path from left to right, we would be done in 1 pass

# Single-Source Shortest Paths in DAGs

**DAG-SHORTEST PATHS( $G, s$ )**

TOPOLOGICALLY-SORT the vertices of  $G$

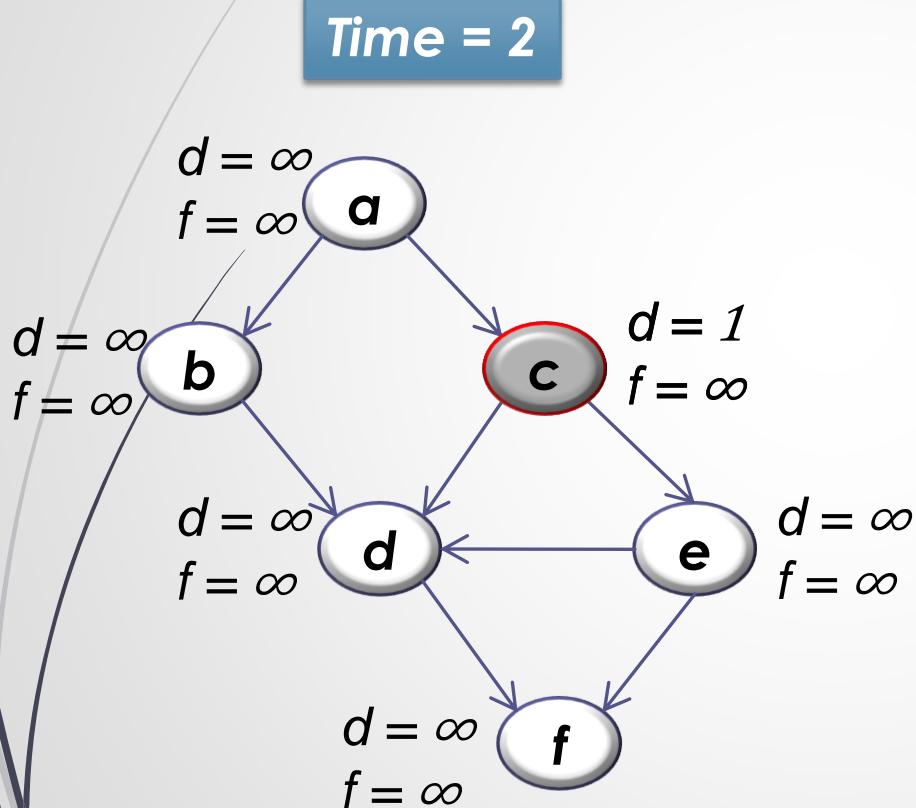
**INIT( $G, s$ )**

*for* each vertex  $u$  taken in topologically sorted order **do**

*for* each  $v$  in  $\text{Adj}[u]$  **do**

**RELAX( $u, v$ )**

# Topological sort

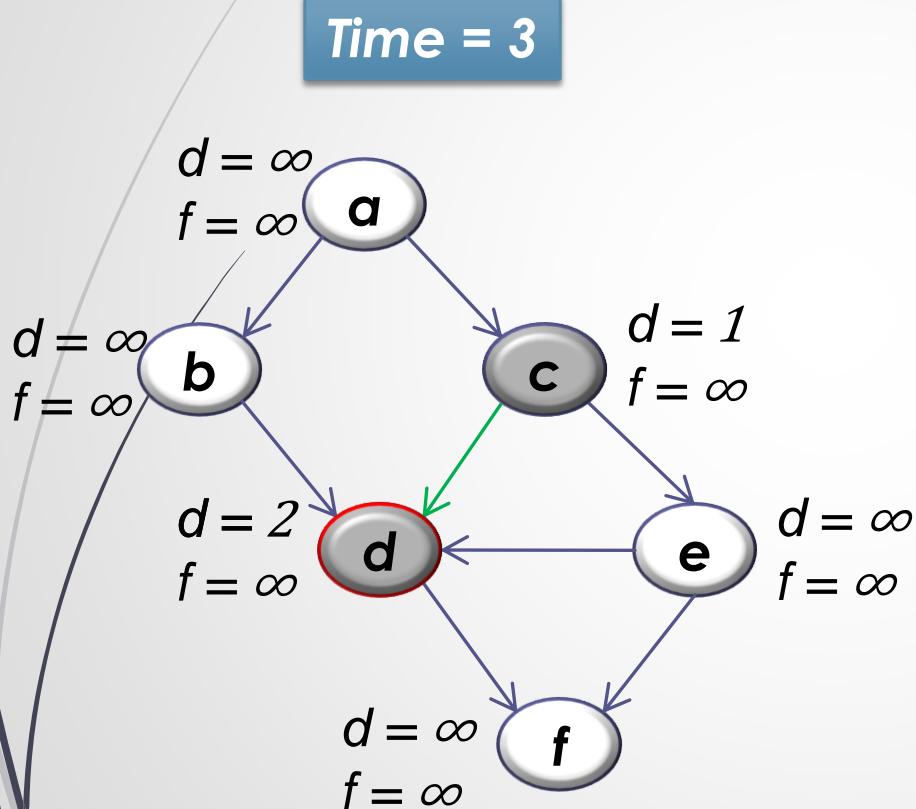


1) Call  $\text{DFS}(\mathbf{G})$  to compute the finishing times  $\mathbf{f}[\mathbf{v}]$

Let's say we start the DFS from the vertex **c**

Next we discover the vertex **d**

# Topological sort

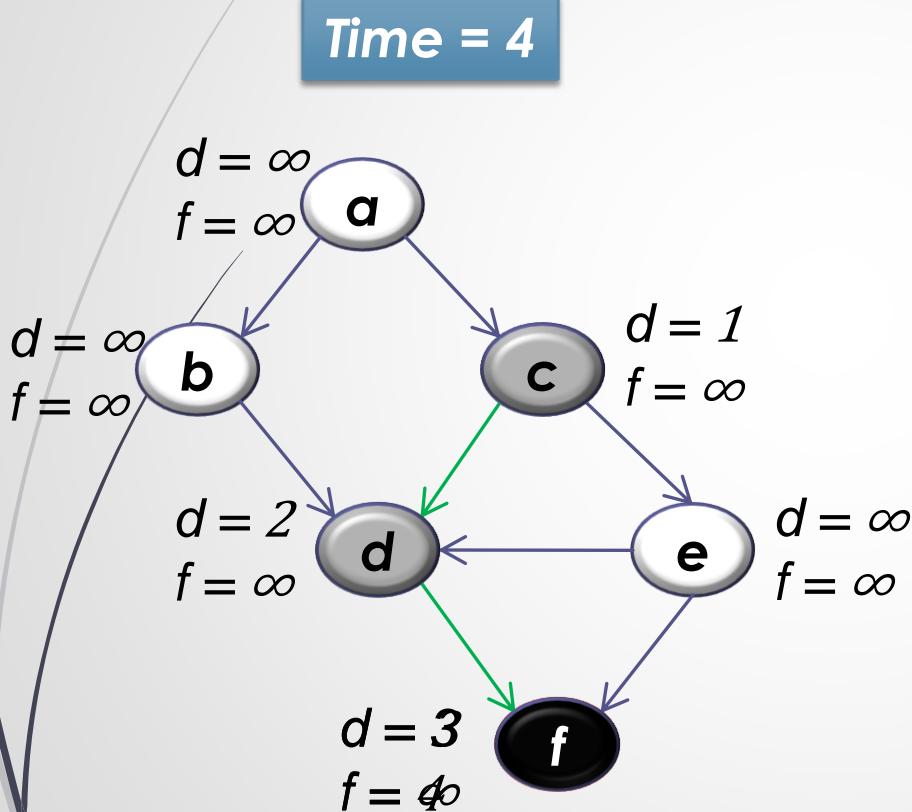


1) Call  $\text{DFS}(\mathbf{G})$  to compute the finishing times  $\mathbf{f}[\mathbf{v}]$

Let's say we start the DFS from the vertex **c**

Next we discover the vertex **d**

# Topological sort



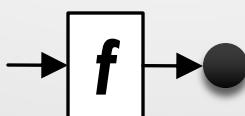
1) Call  $\text{DFS}(\mathbf{G})$  to compute the finishing times  $\mathbf{f}[\mathbf{v}]$

2) as each vertex is finished, insert it onto the **front** of a linked list

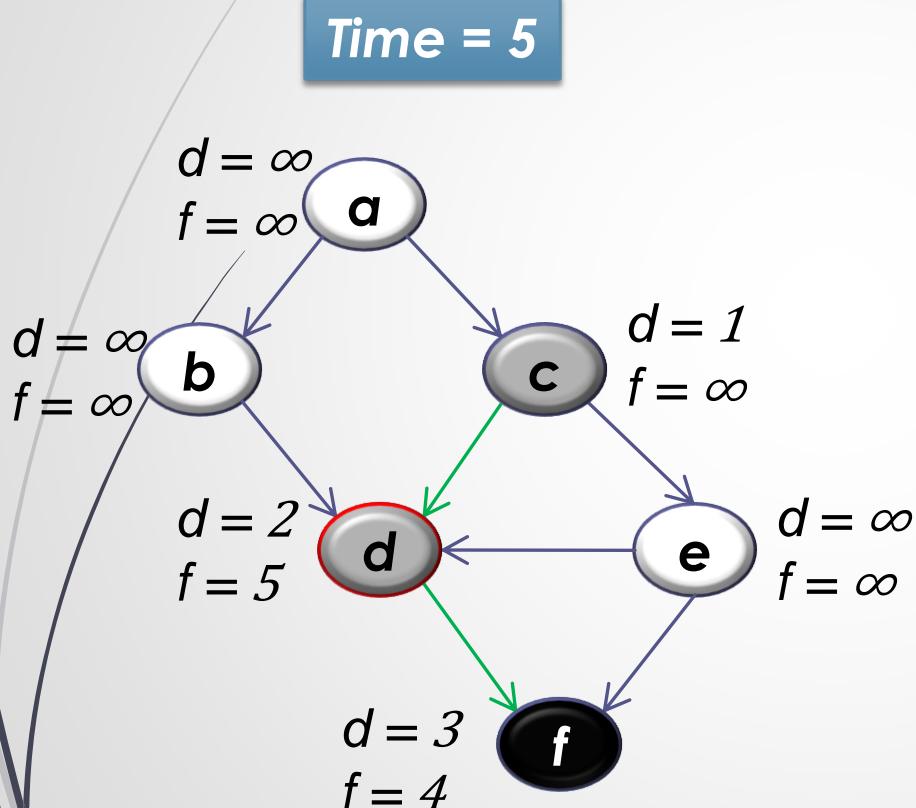
Next we discover vertex **d**

Next we discover vertex **f**

**f** is done, move back to **d**



# Topological sort



1) Call  $\text{DFS}(\mathbf{G})$  to compute the finishing times  $f[\mathbf{v}]$

Let's say we start DFS from the vertex **c**

Next we discover vertex **d**

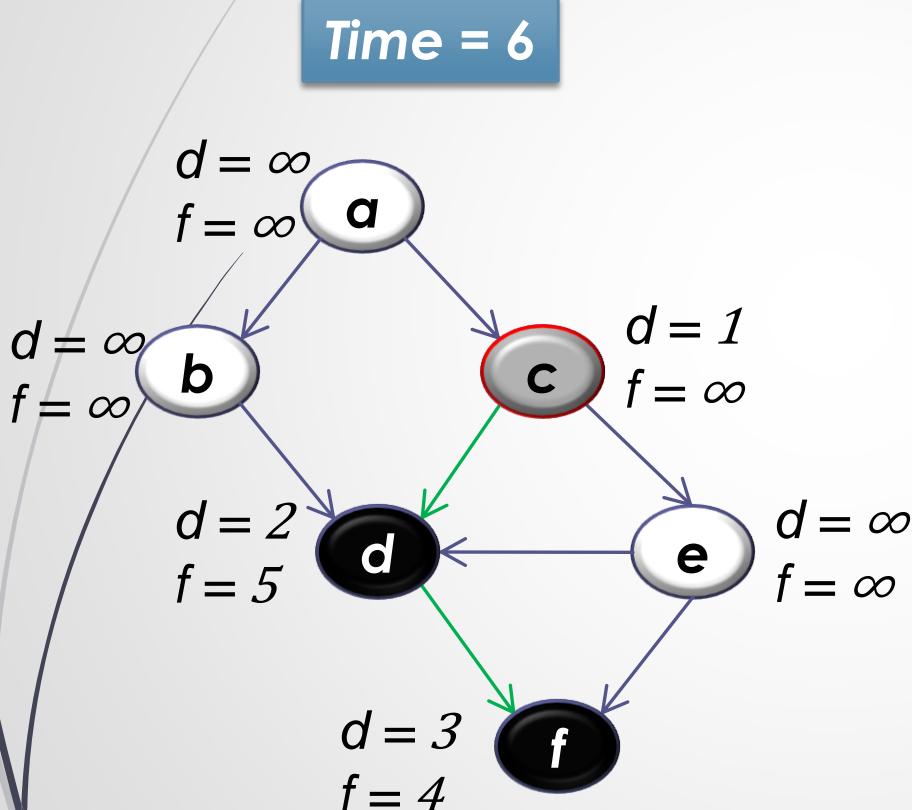
Next we discover the vertex **f**

**f** is done, move back to **d**

**d** is done, move back to **c**



# Topological sort



1) Call  $\text{DFS}(\mathbf{G})$  to compute the finishing times  $\mathbf{f}[\mathbf{v}]$

Let's say we start the DFS from the vertex **c**

Next we discover vertex **d**

Next we discover vertex **f**

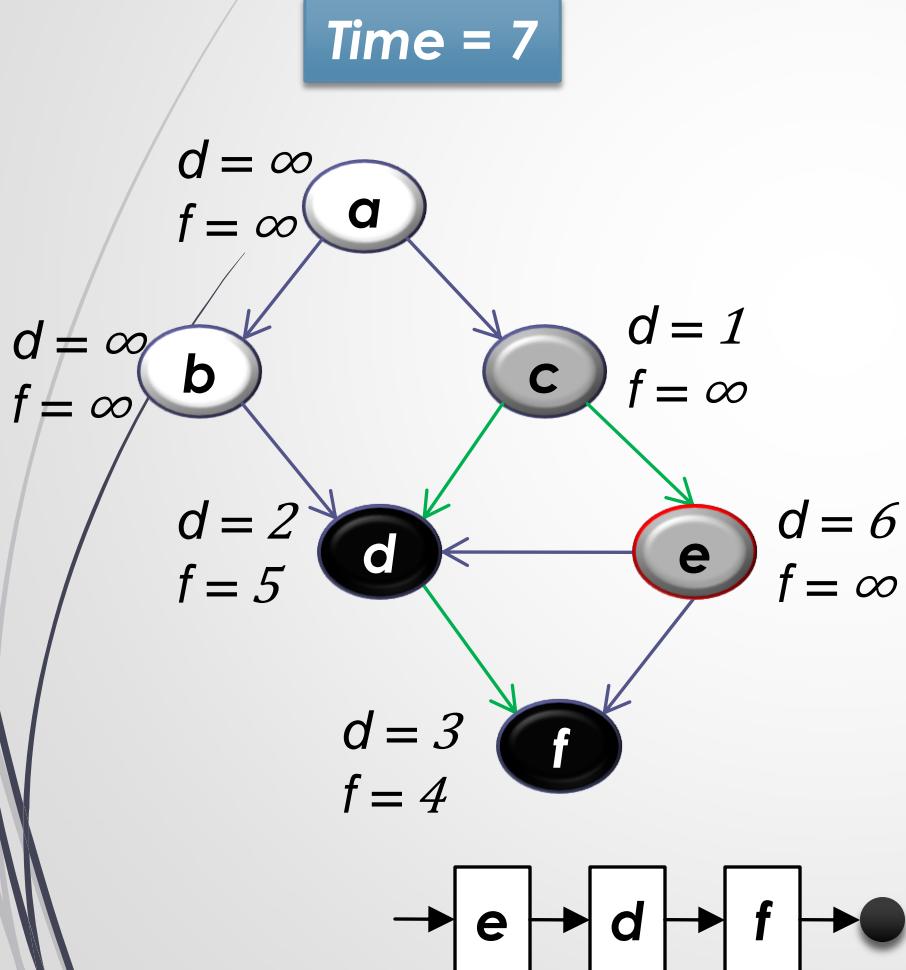
**f** is done, move back to **d**

**d** is done, move back to **c**

Next we discover vertex **e**



# Topological sort



1) Call  $\text{DFS}(\mathbf{G})$  to compute the finishing times  $\mathbf{f}[\mathbf{v}]$

Let's say we start the DFS from the vertex **c**

Next we discover vertex **d**

Both edges from **e** are **cross edges**

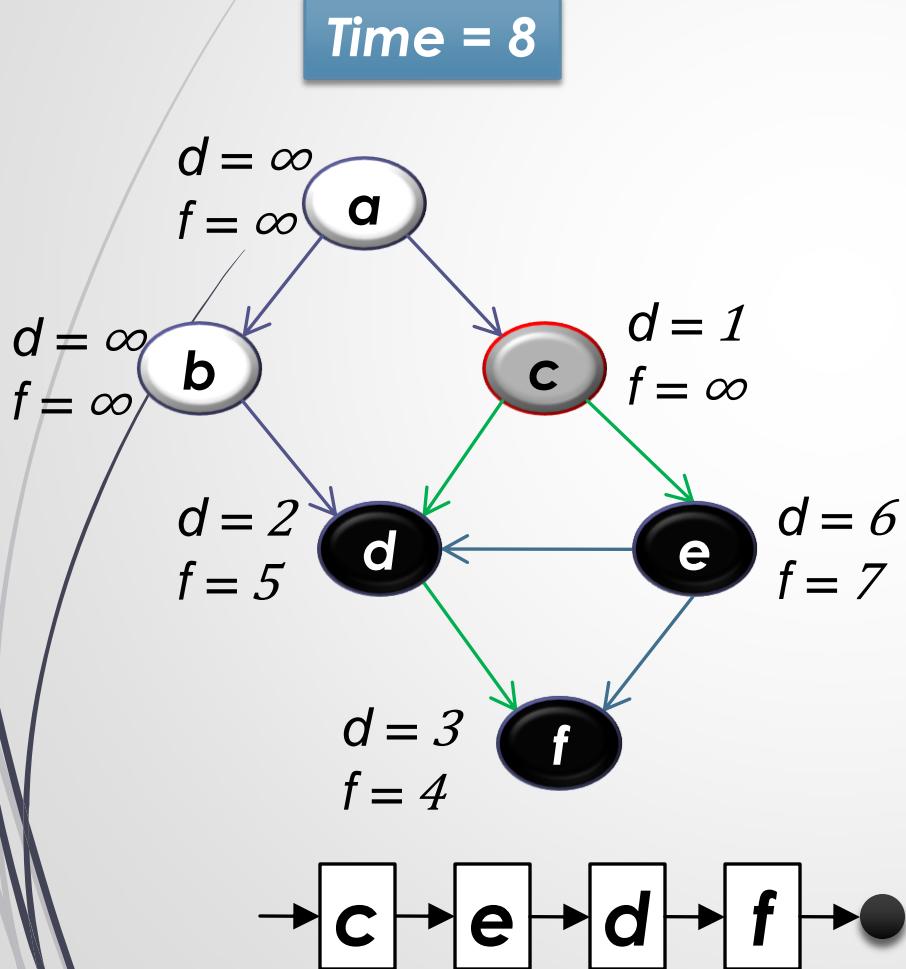
This is done, move back to **a**

**d** is done, move back to **c**

Next we discover vertex **e**

**e** is done, move back to **c**

# Topological sort



1) Call  $\text{DFS}(\mathbf{G})$  to compute the finishing times  $\mathbf{f}[\mathbf{v}]$

Let's say we start the DFS from the vertex **c**

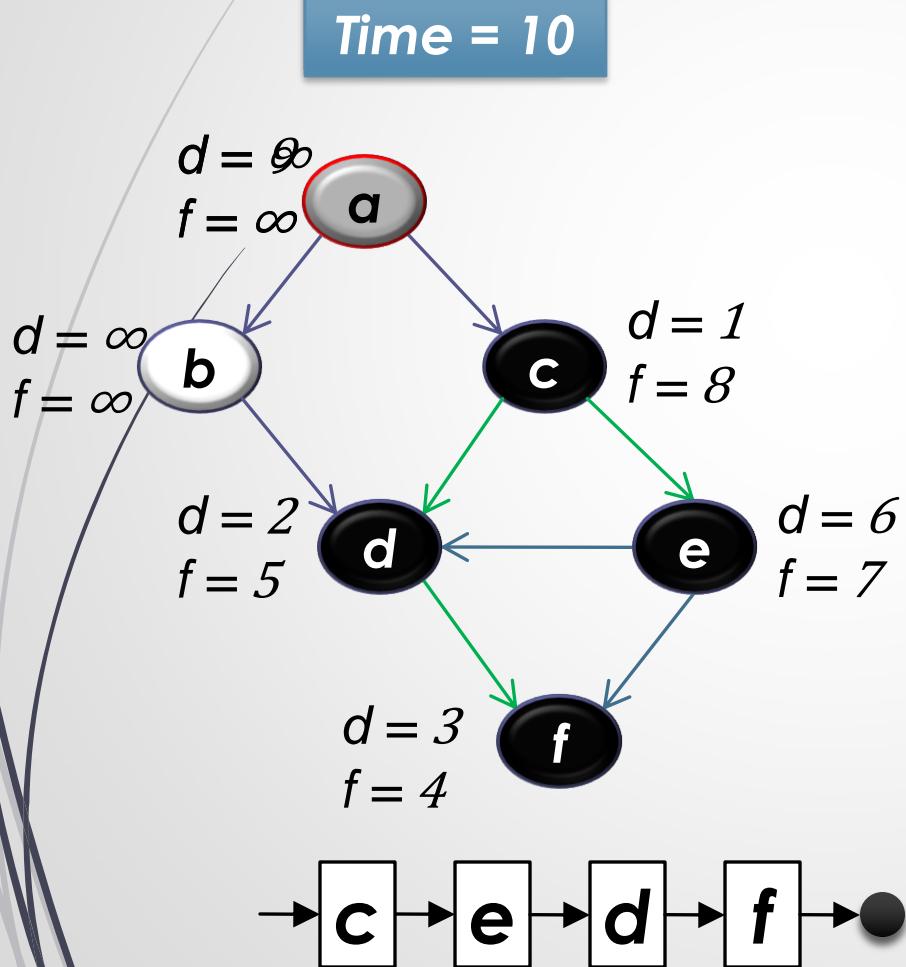
Next we discover the vertex **d**  
Just a note: If there was  $(\mathbf{c}, \mathbf{f})$  edge in the graph, it would be classified as a **forward edge** (in this particular DFS run)

Next we discover the vertex **e**

**e** is done, move back to **c**

**c** is done as well

# Topological sort



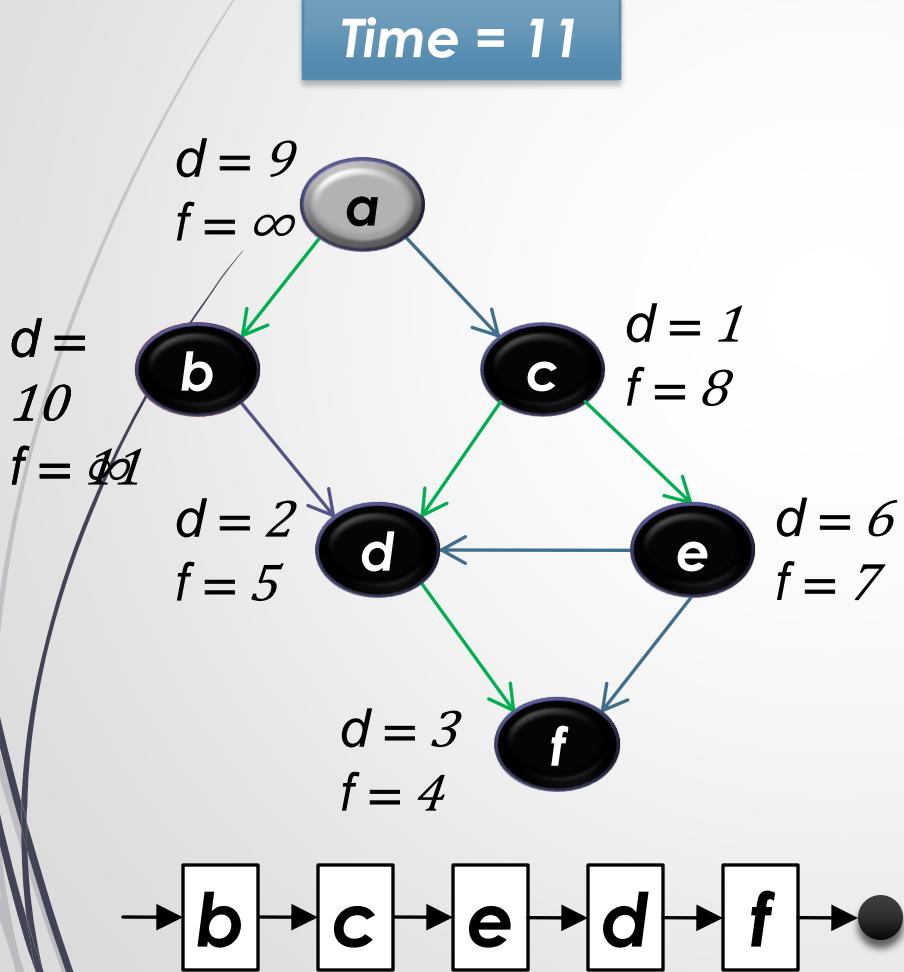
1) Call  $DFS(\mathbf{G})$  to compute the finishing times  $f[\mathbf{v}]$

Let's now call  $DFS$  visit from the vertex **a**

Next we discover vertex **c**, but **c** was already processed  
=> (**a,c**) is a cross edge

Next we discover the vertex **b**

# Topological sort



1) Call  $DFS(\mathbf{G})$  to compute the finishing times  $f[\mathbf{v}]$

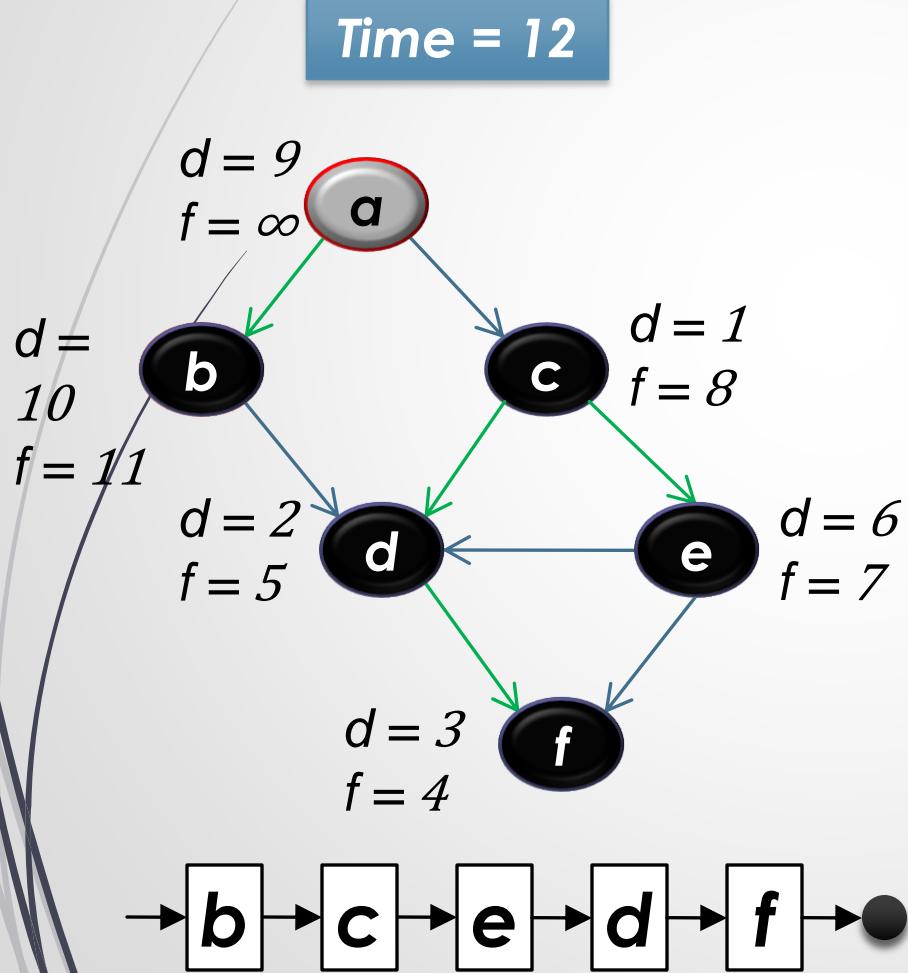
Let's now call  $DFS$  visit from the vertex **a**

Next we discover vertex **c**, but **c** was already processed  
=> (**a,c**) is a cross edge

Next we discover the vertex **b**

**b** is done as (**b,d**) is a cross edge => now move back to **c**

# Topological sort



1) Call  $DFS(\mathbf{G})$  to compute the finishing times  $\mathbf{f}[\mathbf{v}]$

Let's now call  $DFS$  visit from the vertex **a**

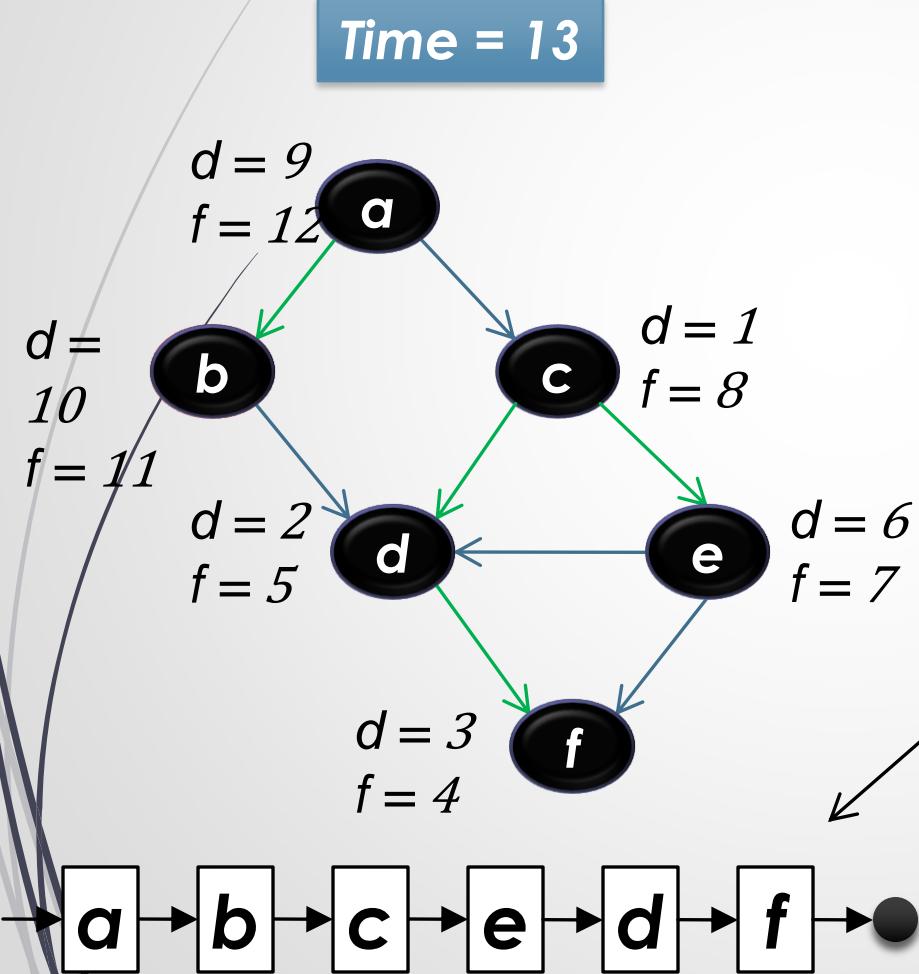
Next we discover the vertex **c**, but **c** was already processed  
=> (**a,c**) is a cross edge

Next we discover the vertex **b**

**b** is done as (**b,d**) is a cross edge => now move back to **c**

**a** is done as well

# Topological sort



1) Call  $\text{DFS}(\mathbf{G})$  to compute the finishing times  $\mathbf{f}[\mathbf{v}]$

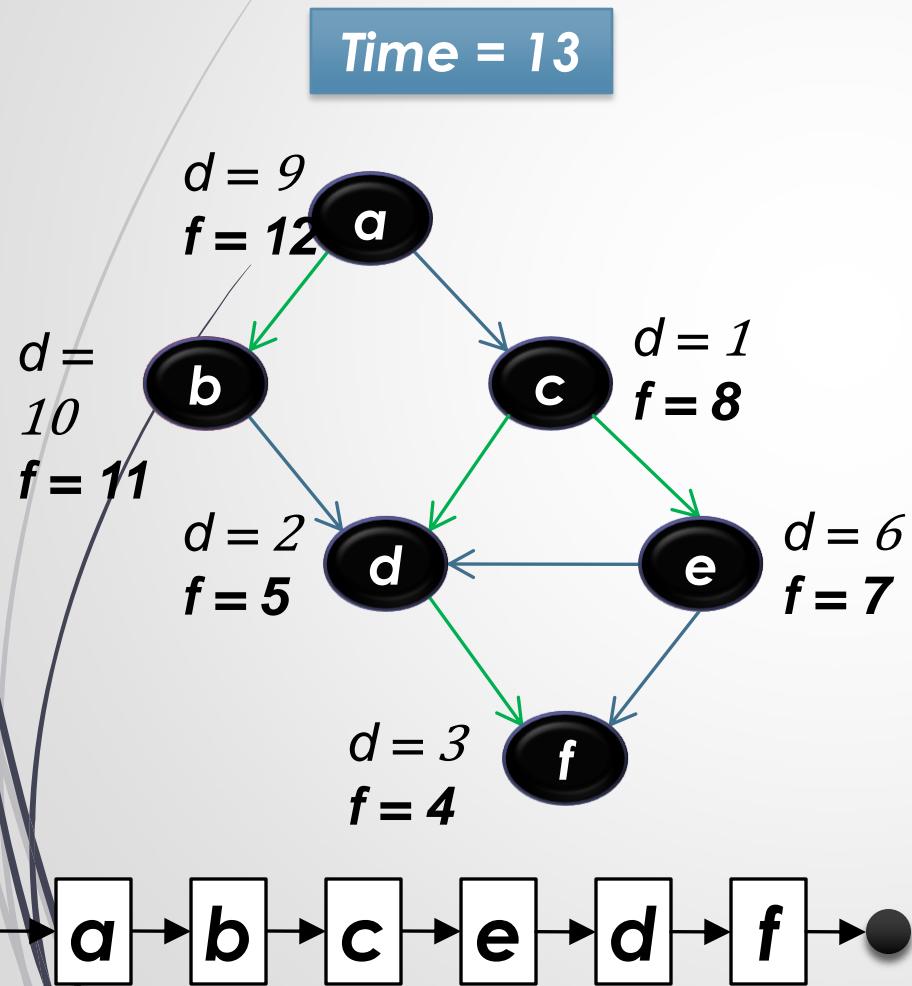
Let's now call DFS visit from the vertex **a**

~~WE HAVE THE RESULT!~~  
3) return the linked list of vertices

**b** is done as  $(\mathbf{b}, \mathbf{d})$  is a cross edge => now move back to **c**

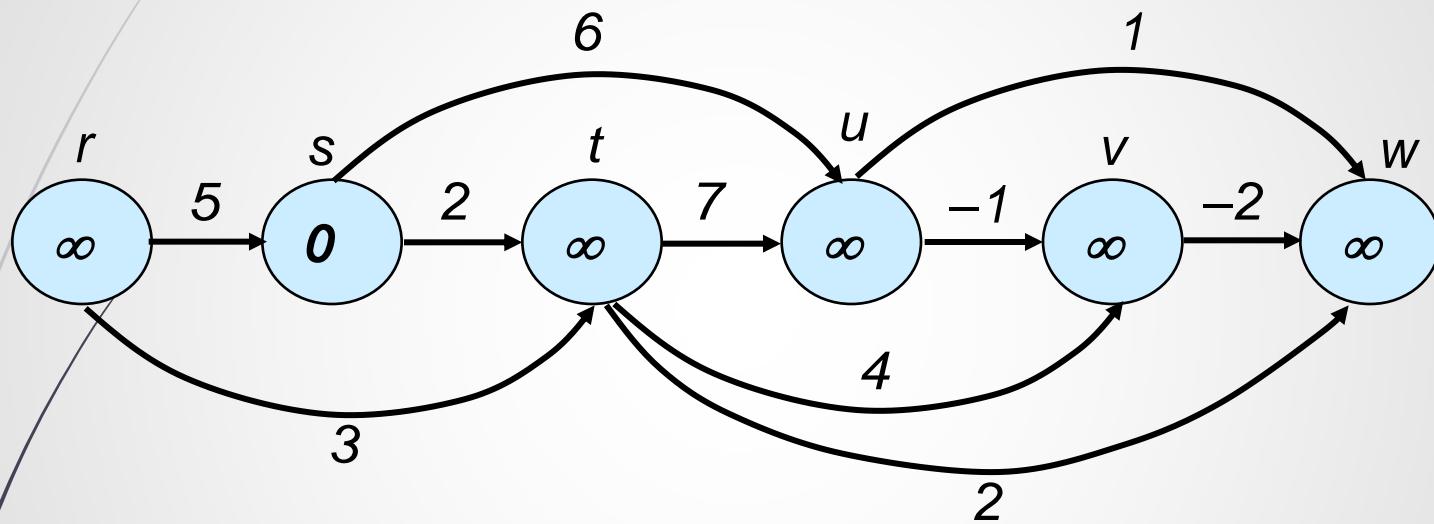
**a** is done as well

# Topological sort

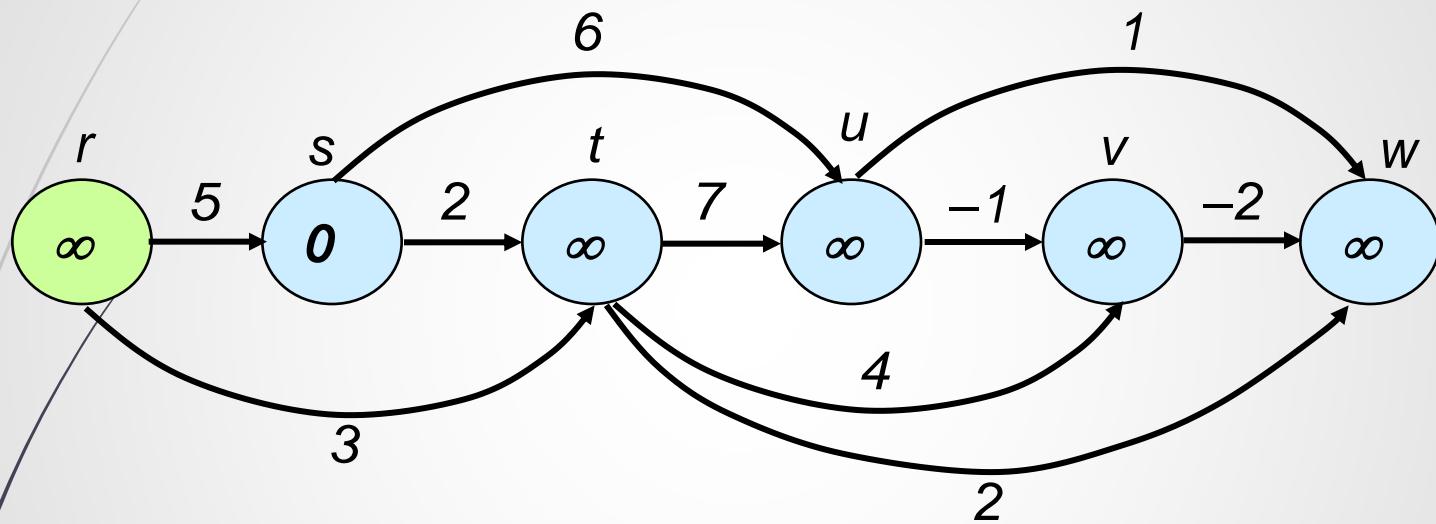


The linked list is sorted in **decreasing** order of finishing times  $f[]$

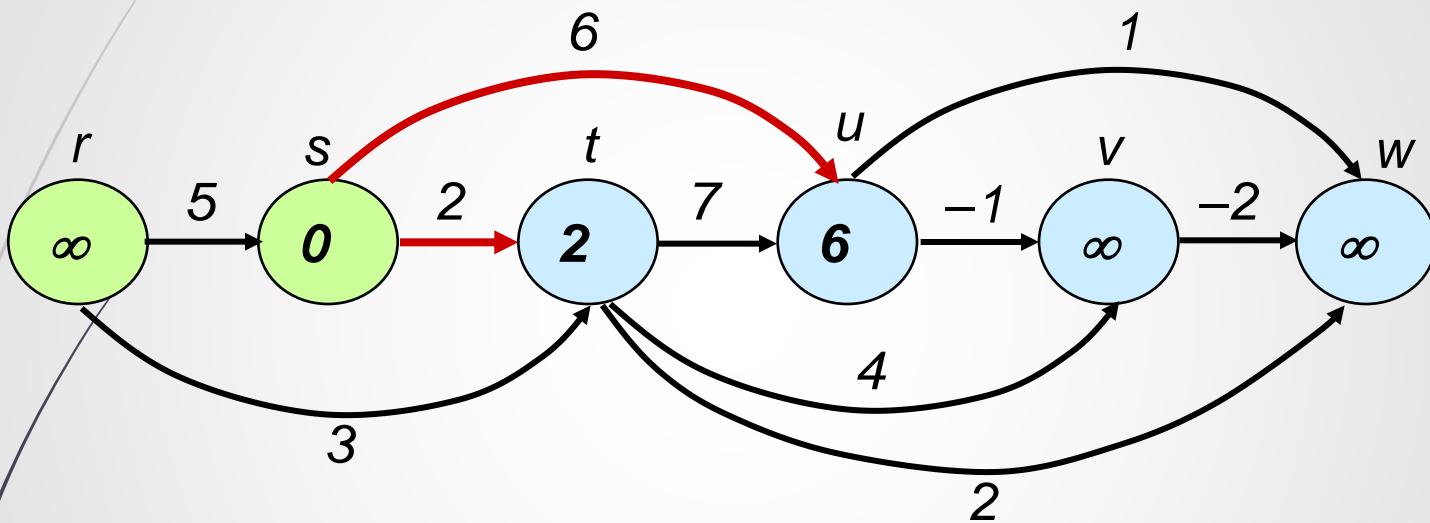
# Example



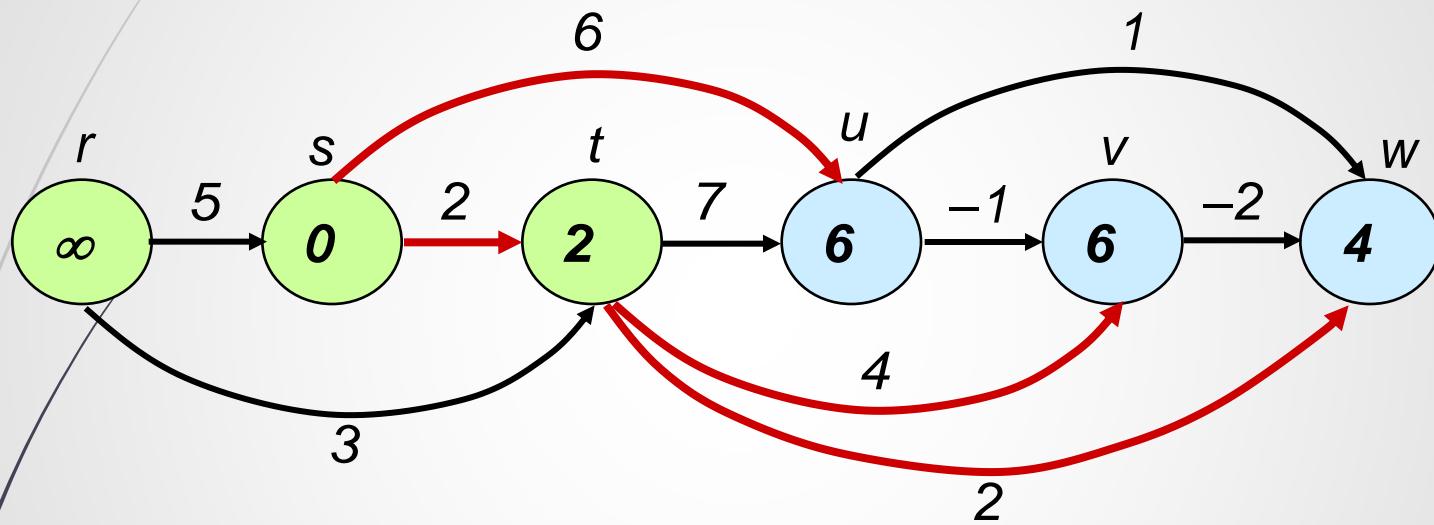
# Example



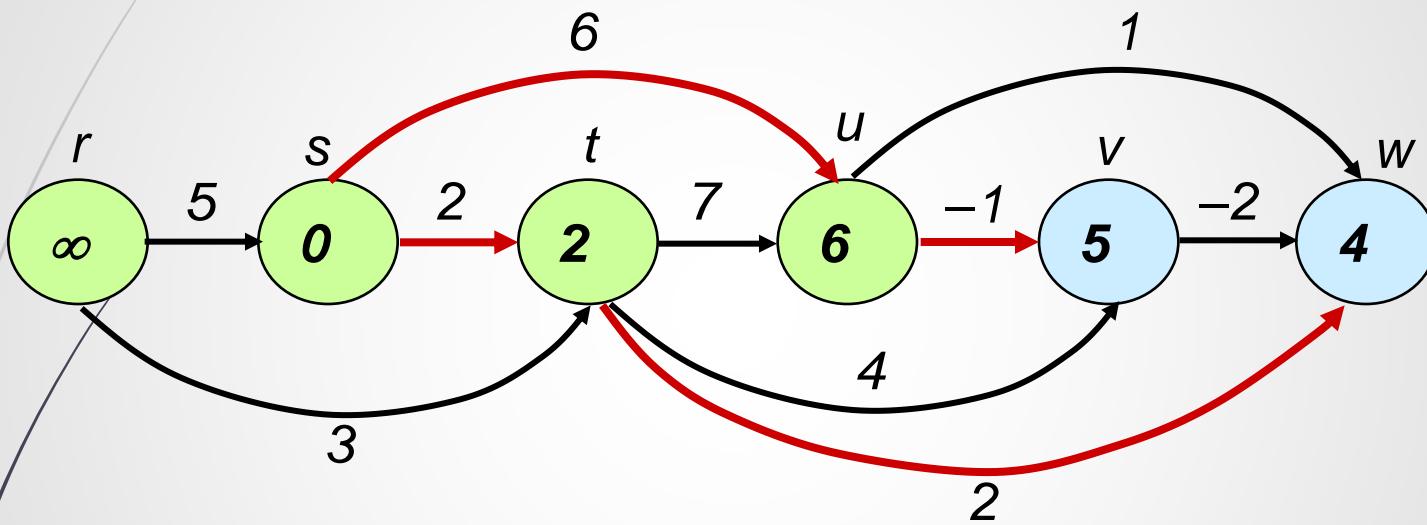
## Example



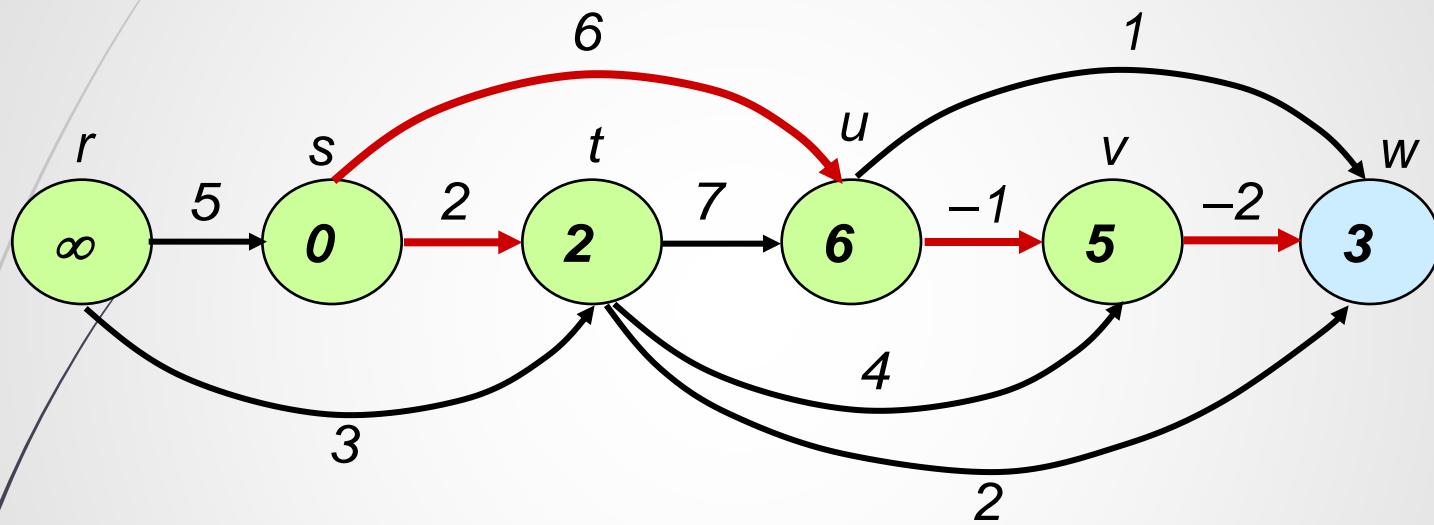
## Example



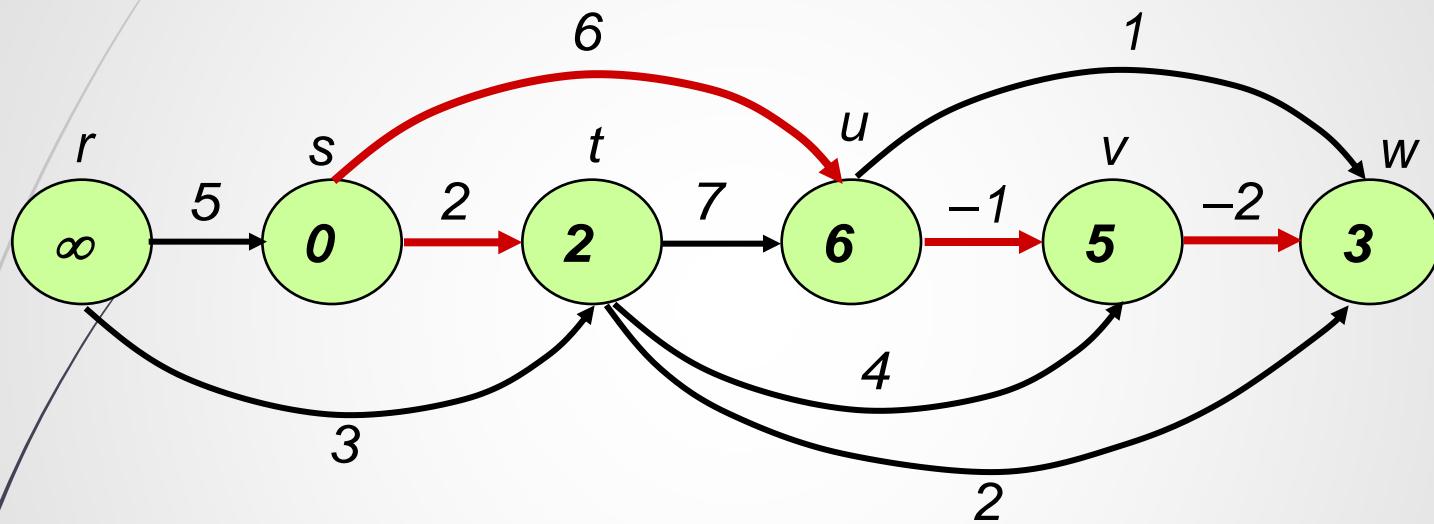
## Example



# Example



# Example



# END



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