

COMP 251

Algorithms & Data Structures (Winter 2021)

Graphs - Introduction

School of Computer Science
McGill University

Slides of (Comp321 ,2021), Langer (2014), Kleinberg & Tardos, 2005 & Cormen et al., 2009, Jaehyun Park' slides CS 97SI, Top-coder tutorials, T-414-AFLV Course, Programming Challenges books, slides from D. Plaisted (UNC) and Comp251-Fall McGill.

Outline

- Graphs.
 - Introduction.
 - BFS.
 - DFS.
 - Strong Connected Components / Topological Sort.
 - Network Flow 1.
 - Network Flow 2.
 - Shortest Path.
 - Minimum Spanning Trees.
 - Bipartite Graphs.

Graphs - Definition

- An abstract way of representing connectivity using nodes (also called vertices) and edges.
- m edges connect some pairs of nodes.
 - Edges can be either one-directional (directed) or bidirectional.
- Nodes and edges can have some auxiliary information.

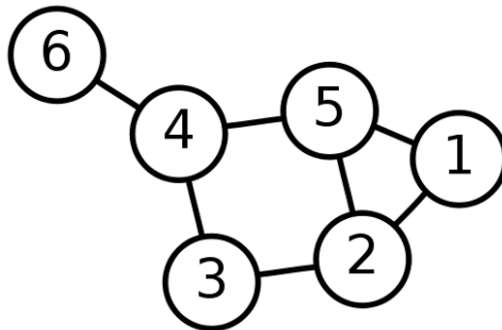
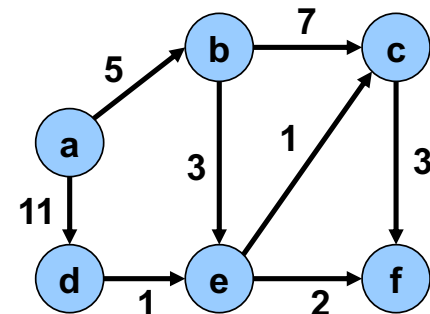
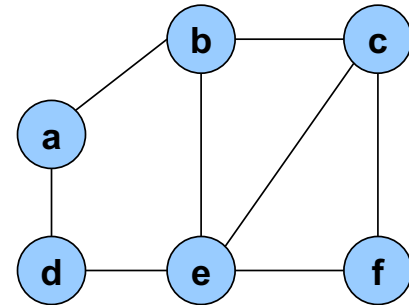


Figure from Wikipedia

Graphs - Definition

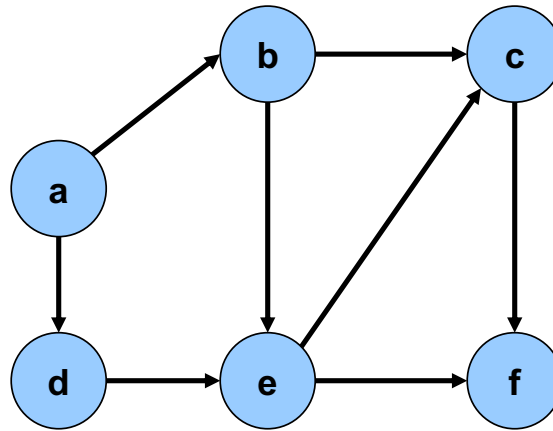
- *Graph* $G = (V, E)$
 - V = set of vertices
 - E = set of edges $\subseteq (V \times V)$
- Types of graphs
 - Undirected: edge $(u, v) = (v, u)$; for all v , $(v, v) \notin E$ (No self loops.)
 - Directed: (u, v) is edge from u to v , denoted as $u \rightarrow v$. Self loops are allowed.
 - Weighted: each edge has an associated weight, given by a weight function $w : E \rightarrow \mathbf{R}$.
 - Dense: $|E| \approx |V|^2$.
 - Sparse: $|E| \ll |V|^2$.
- $|E| = O(|V|^2)$



Graphs - Properties

- If $(u, v) \in E$, then vertex v is adjacent to vertex u .
- Adjacency relationship is:
 - Symmetric if G is undirected.
 - Not necessarily so if G is directed.
- If G is (strongly) connected:
 - There is a path between every pair of vertices.
 - $|E| \geq |V| - 1$.
 - Furthermore, if $|E| = |V| - 1$, then G is a tree.

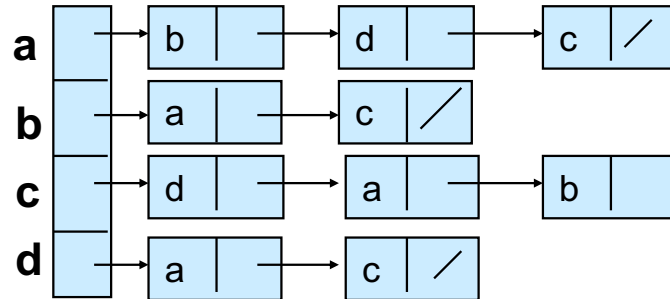
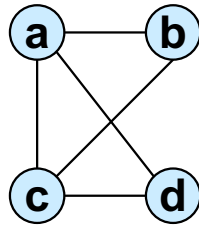
Graphs - Vocabulary



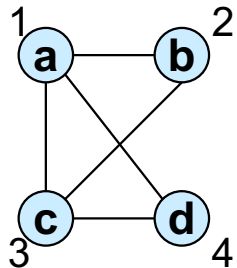
- Ingoing edges of u : $\{ (v,u) \in E \}$ (e.g. $\text{in}(e) = \{ (b,e), (d,e) \}$)
- Outgoing edges of u : $\{ (u,v) \in E \}$ (e.g. $\text{out}(d) = \{ (d,e) \}$)
- In-degree(u): $|\text{in}(u)|$ // Number of predecessors
- Out-degree(u): $|\text{out}(u)|$ // Number of successors

Graphs - Representation

- Two standard ways.
 - Adjacency Lists.



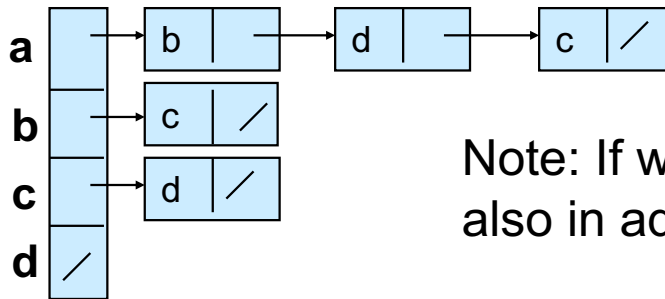
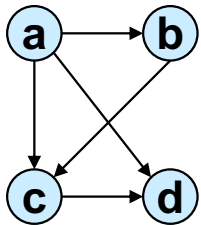
- Adjacency Matrix.



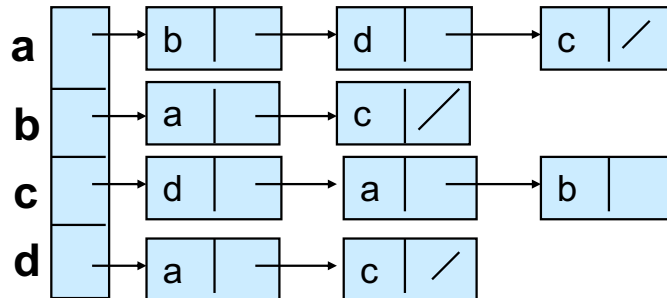
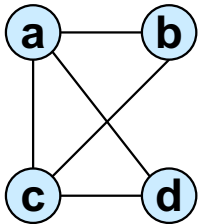
| | 1 | 2 | 3 | 4 |
|---|---|---|---|---|
| 1 | 0 | 1 | 1 | 1 |
| 2 | 1 | 0 | 1 | 0 |
| 3 | 1 | 1 | 0 | 1 |
| 4 | 1 | 0 | 1 | 0 |

Graphs – Representation – Adjacency List

- Consists of an array Adj of $|V|$ lists.
- One list per vertex.
- For $u \in V$, $Adj[u]$ consists of all vertices adjacent to u .



Note: If weighted, store weights also in adjacency lists.



Adjacency List – Storage Requirement

- For directed graphs:

- Sum of lengths of all adj. lists is

$$\sum_{v \in V} \text{out-degree}(v) = |E|$$

← No. of edges leaving v

- Total storage: $\Theta(V+E)$

- For undirected graphs:

- Sum of lengths of all adj. lists is

$$\sum_{v \in V} \text{degree}(v) = 2|E|$$

← No. of edges incident on v . Edge (u,v) is incident on vertices u and v .

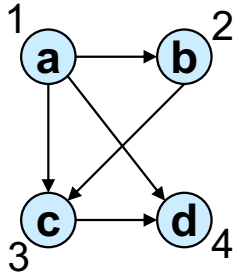
- Total storage: $\Theta(V+E)$

Adjacency List – Pros and Cons

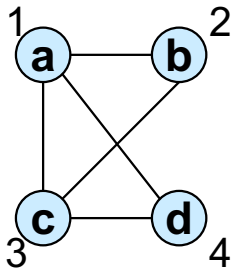
- Pros
 - Space-efficient, when a graph is sparse.
 - Can be modified to support many graph variants.
- Cons
 - Determining if an edge $(u,v) \in E$ is not efficient.
 - Have to search in u 's adjacency list. $\Theta(\text{degree}(u))$ time.
 - $\Theta(V)$ in the worst case.

Graphs – Representation – Adjacency Matrix

- $|V| \times |V|$ matrix A .
- Number vertices from 1 to $|V|$ in some arbitrary manner.
- A is then given by:
$$A[i, j] = a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$



| | 1 | 2 | 3 | 4 |
|---|---|---|---|---|
| 1 | 0 | 1 | 1 | 1 |
| 2 | 0 | 0 | 1 | 0 |
| 3 | 0 | 0 | 0 | 1 |
| 4 | 0 | 0 | 0 | 0 |



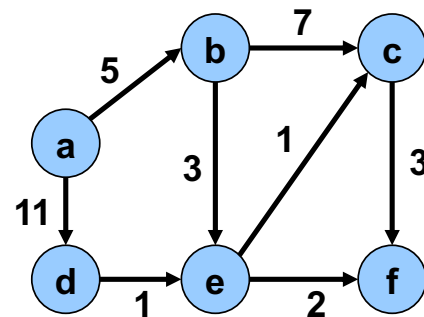
| | 1 | 2 | 3 | 4 |
|---|---|---|---|---|
| 1 | 0 | 1 | 1 | 1 |
| 2 | 1 | 0 | 1 | 0 |
| 3 | 1 | 1 | 0 | 1 |
| 4 | 1 | 0 | 1 | 0 |

$A = A^T$ for undirected graphs.

Adjacency Matrix – Storage-Time Requirement

- **Space:** $\Theta(V^2)$.
 - Not memory efficient for large sparse graphs.
- **Time:** to list all vertices adjacent to u : $\Theta(V)$.
- **Time:** to determine if $(u, v) \in E$: $\Theta(1)$.
- Can store weights instead of bits for weighted graph.

| | a | b | c | d | e | f |
|---|---|---|---|----|---|---|
| a | 0 | 5 | 0 | 11 | 0 | 0 |
| b | 0 | 0 | 7 | 0 | 3 | 0 |
| c | 0 | 0 | 0 | 0 | 0 | 3 |
| d | 0 | 0 | 0 | 0 | 1 | 0 |
| e | 0 | 0 | 1 | 0 | 0 | 2 |
| f | 0 | 0 | 0 | 0 | 0 | 0 |



Graphs— Traversal-Searching

- Is there a path from s to v in G ?
- Searching a graph:
 - Systematically follow the edges of a graph to visit the vertices of the graph.
- Used to discover the structure of a graph.
- Standard graph-searching algorithms.
 - Breadth-first Search (BFS).
 - Depth-first Search (DFS).

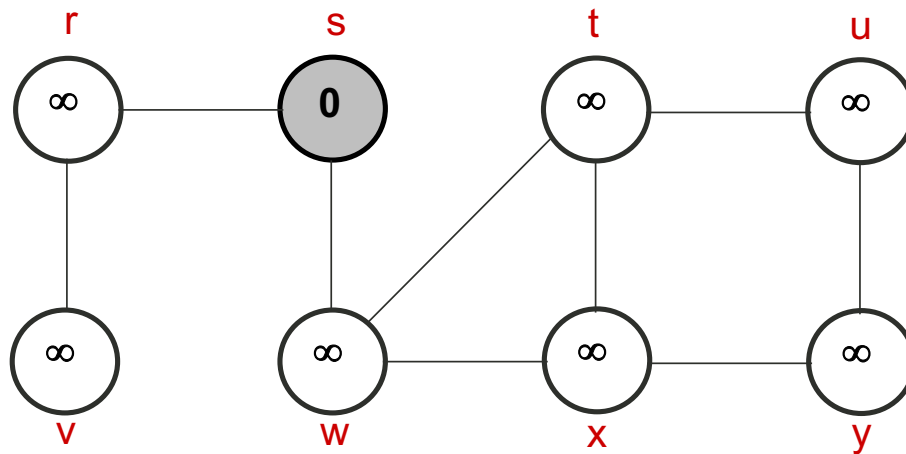
Graphs— Breadth-First Search

- Expands the frontier between discovered and undiscovered vertices uniformly across the breadth of the frontier.
 - The algorithm discovers all vertices at distance k from s before discovering any vertices at distance $k + 1$.
 - A vertex is “discovered” the first time it is encountered during the search. A vertex is “finished” if all vertices adjacent to it have been discovered.
 - It uses a first-in, first-out queue Q to manage the progress.
- Colors the vertices to keep track of progress.
 - White – Undiscovered.
 - Gray – Discovered but not finished.
 - Black – Finished.
 - Colors are required only to reason about the algorithm. Can be implemented without colors.

Graphs— Breadth-First Search

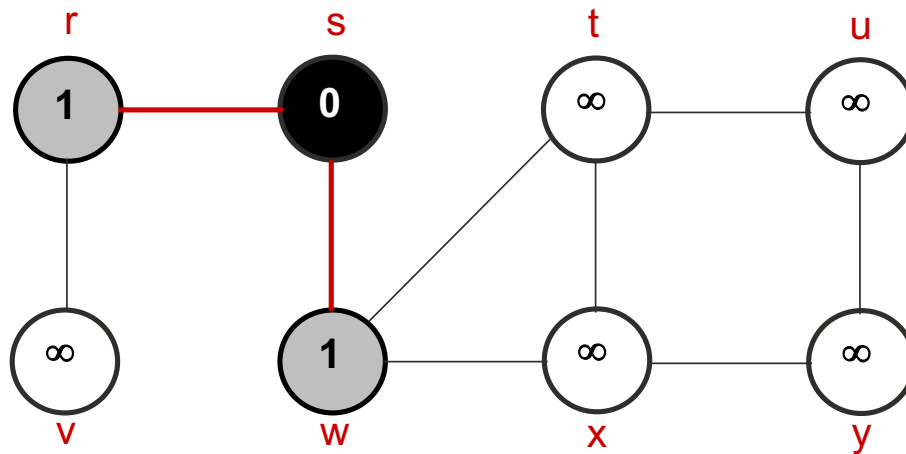
- **Input:** Graph $G = (V, E)$, either directed or undirected, and **source vertex** $s \in V$.
- **Output:**
 - $d[v]$ = distance (smallest # of edges, or shortest path) from s to v , for all $v \in V$. $d[v] = \infty$ if v is not reachable from s .
 - $\pi[v] = u$ such that (u, v) is last edge on shortest path $s \rightsquigarrow v$.
 - u is v 's predecessor.
 - Builds breadth-first tree with root s that contains all reachable vertices.

Breadth-First Search - Example



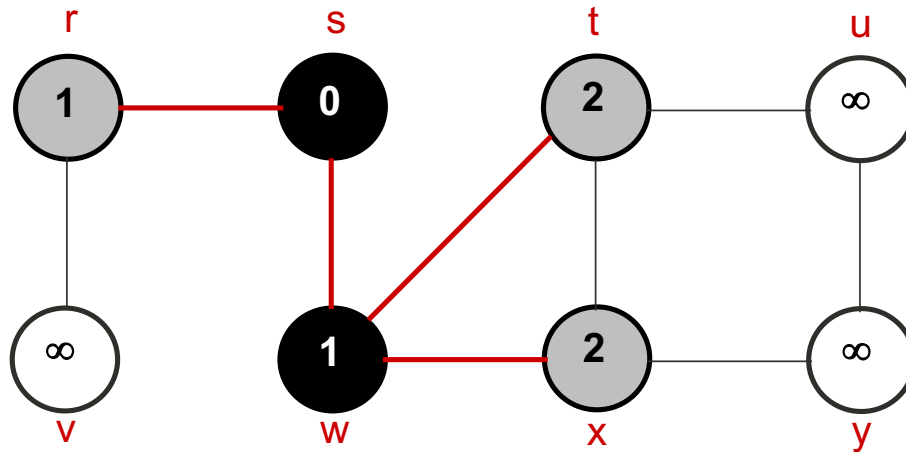
Q: s
0

Breadth-First Search - Example



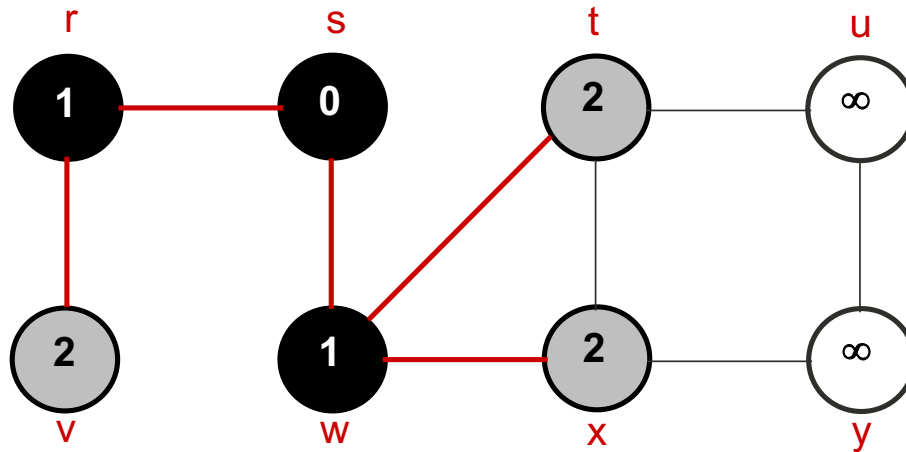
| | | |
|-----------|---|---|
| Q: | w | r |
| | 1 | 1 |

Breadth-First Search - Example



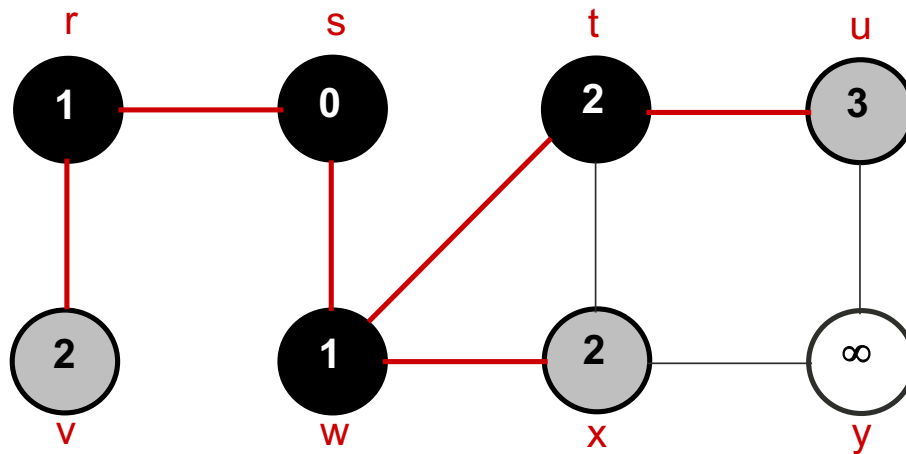
| | | | |
|----|---|---|---|
| Q: | r | t | x |
| | 1 | 2 | 2 |

Breadth-First Search - Example



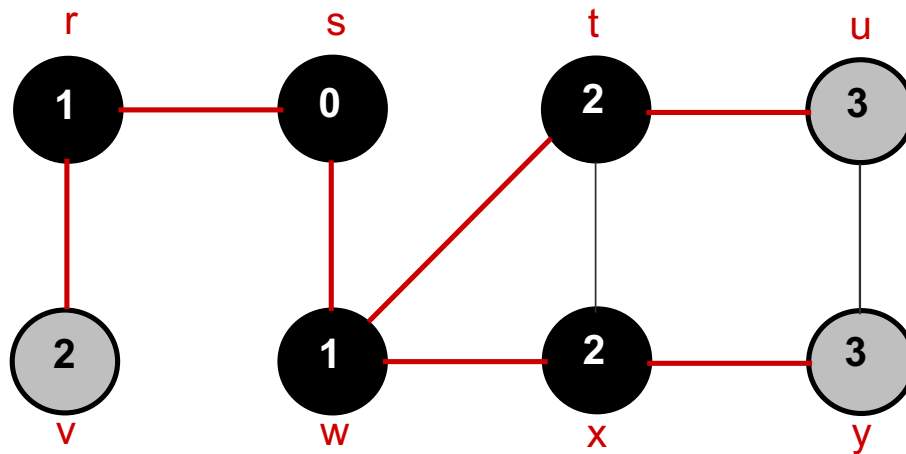
| | | | |
|-----------|---|---|---|
| Q: | t | x | v |
| | 2 | 2 | 2 |

Breadth-First Search - Example



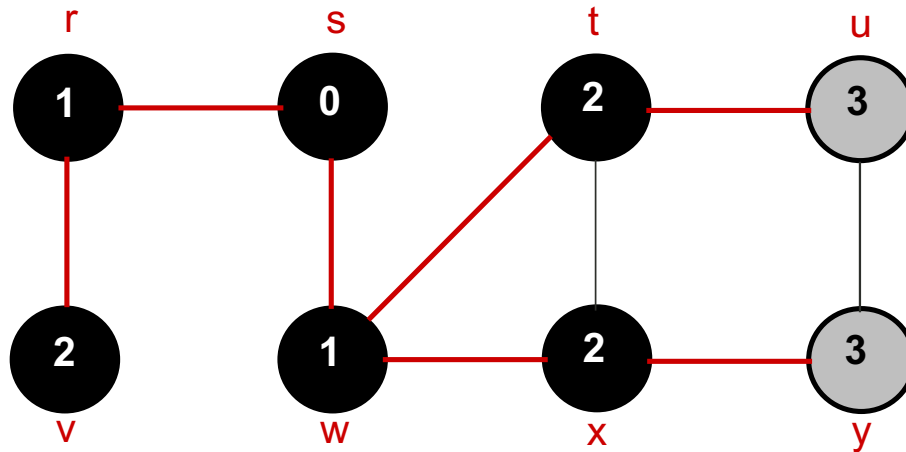
| | | | |
|-----------|---|---|---|
| Q: | x | v | u |
| | 2 | 2 | 3 |

Breadth-First Search - Example



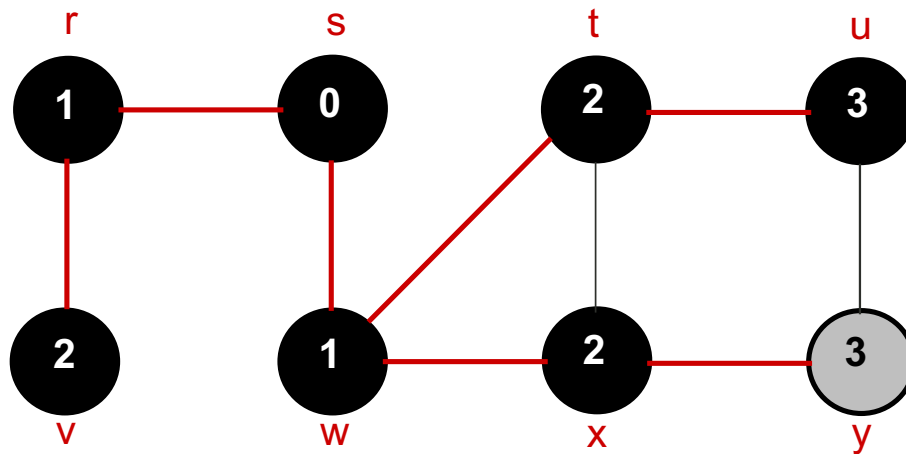
| | | | |
|----|---|---|---|
| Q: | v | u | y |
| | 2 | 3 | 3 |

Breadth-First Search - Example



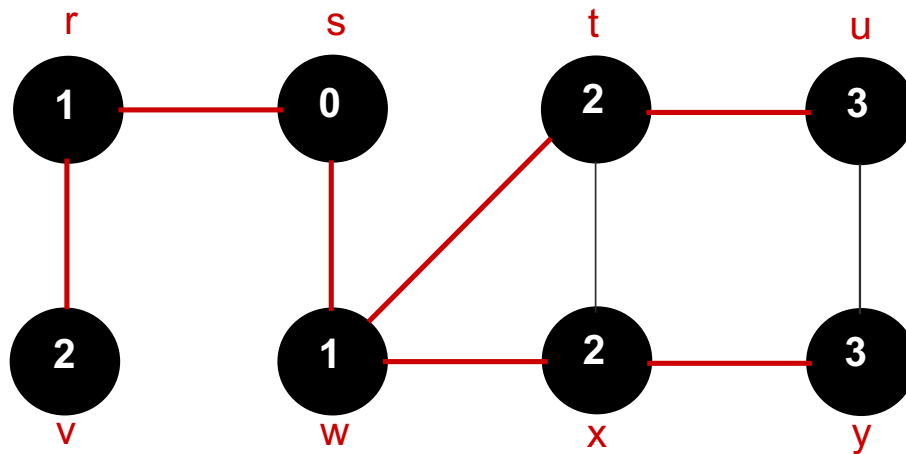
| | | |
|----|---|---|
| Q: | u | y |
| | 3 | 3 |

Breadth-First Search - Example



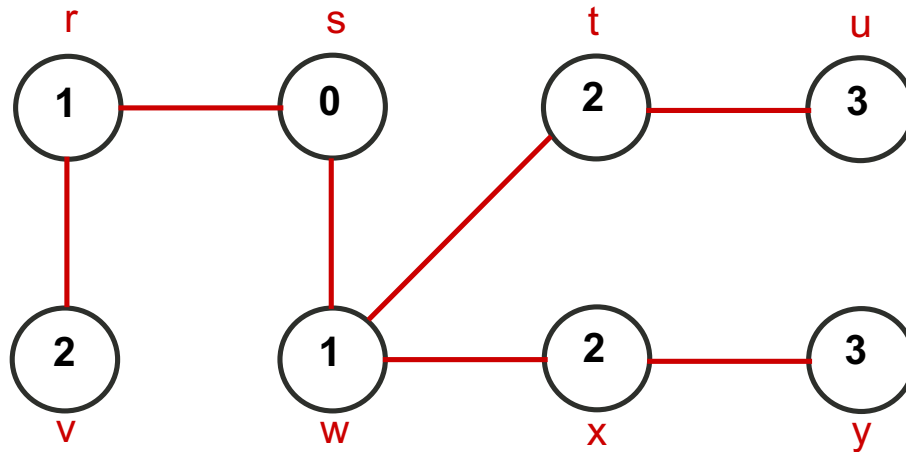
Q: y
3

Breadth-First Search - Example



Q: \emptyset

Breadth-First Search - Example



BF Tree

Breadth-First Search - Analysis

BFS(G, s)

```
1  for each vertex  $u \in G.V - \{s\}$ 
2       $u.color = \text{WHITE}$ 
3       $u.d = \infty$ 
4       $u.\pi = \text{NIL}$ 
5   $s.color = \text{GRAY}$ 
6   $s.d = 0$ 
7   $s.\pi = \text{NIL}$ 
8   $Q = \emptyset$ 
9  ENQUEUE( $Q, s$ )
10 while  $Q \neq \emptyset$ 
11      $u = \text{DEQUEUE}(Q)$ 
12     for each  $v \in G.Adj[u]$ 
13         if  $v.color == \text{WHITE}$ 
14              $v.color = \text{GRAY}$ 
15              $v.d = u.d + 1$ 
16              $v.\pi = u$ 
17             ENQUEUE( $Q, v$ )
18      $u.color = \text{BLACK}$ 
```

Initialization $O(V)$

Traversal Loop:

After initialization, each vertex is enqueued and dequeued at most once, and each operation takes $O(1)$. So, total time for queuing is $O(V)$.

The adjacency list of each vertex is scanned at most once. The sum of lengths of all adjacency lists is $\Theta(E)$.

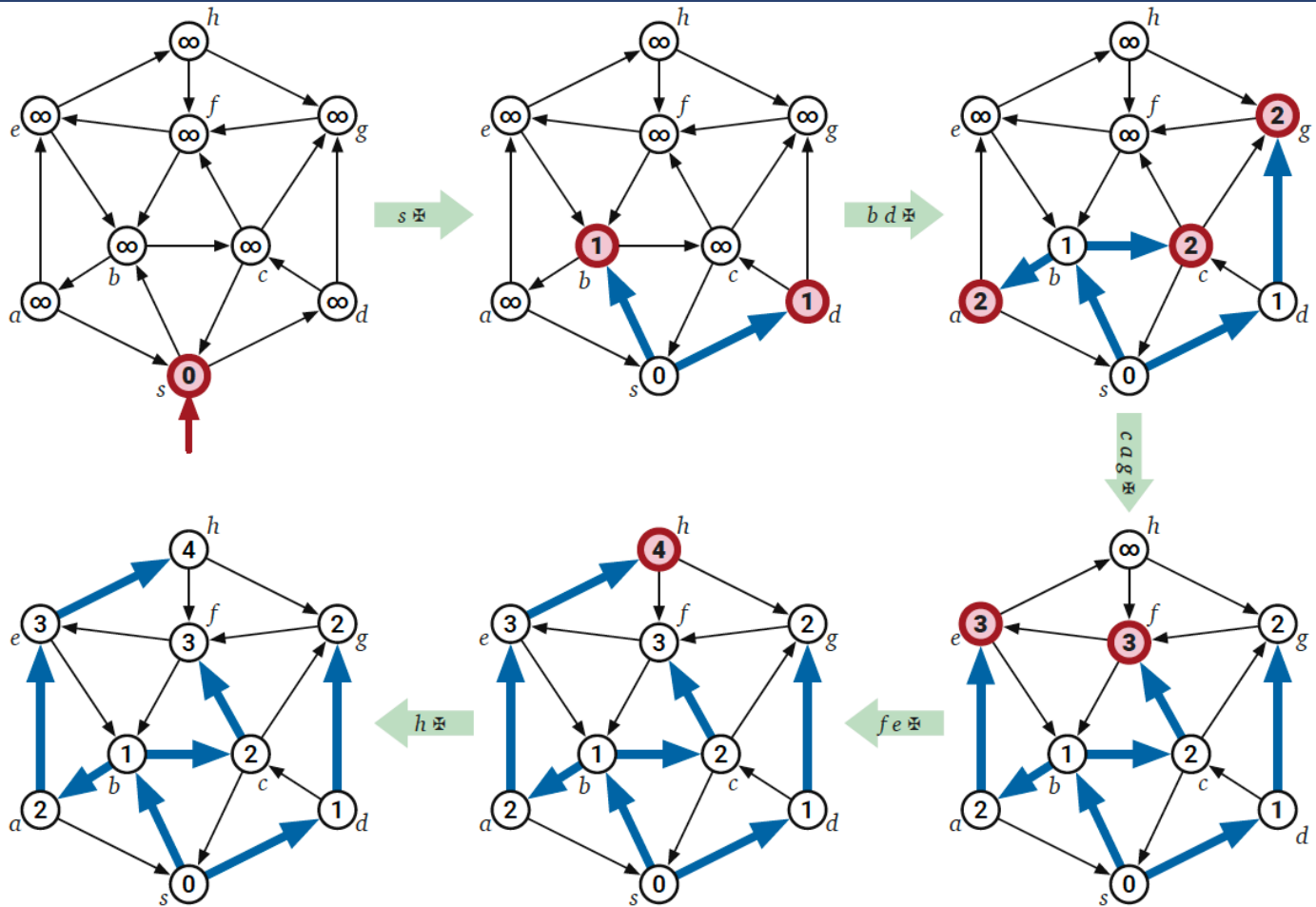
Breadth-First Search - Analysis

- Initialization takes $O(V)$.
- Traversal Loop
 - After initialization, each vertex is enqueued and dequeued at most once, and each operation takes $O(1)$. So, total time for queuing is $O(V)$.
 - The adjacency list of each vertex is scanned at most once. The sum of lengths of all adjacency lists is $\Theta(E)$.
- Summing up over all vertices \Rightarrow total running time of BFS is $O(V+E)$, linear in the size of the adjacency list representation of graph.

Breadth-First Search - Application

- Suppose we are given an unweighted directed graph $G = (V, E)$ with two special vertices, and we want to find the shortest path from a source vertex s to a target vertex t .
 - Special case of shortest path.
 - All edges have weight 1, and the length of a path is just the number of edges.

Breadth-First Search - Application



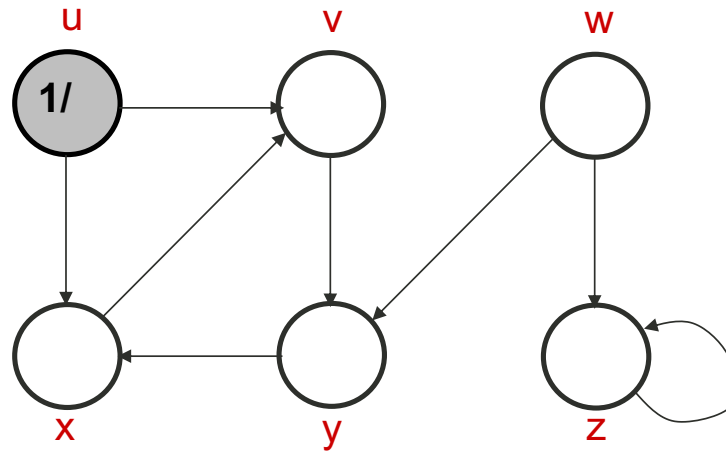
Depth-First Search

- Explore edges out of the most recently discovered vertex v .
- When all edges of v have been explored, backtrack to explore other edges leaving the vertex from which v was discovered (its *predecessor*).
- “Search as deep as possible first.”
- Continue until all vertices reachable from the original source are discovered.
- If any undiscovered vertices remain, then one of them is chosen as a new source and search is repeated from that source.

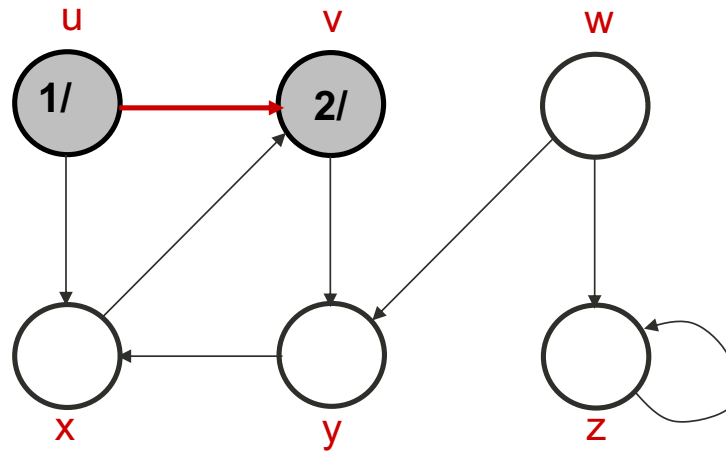
Depth-First Search

- **Input:** $G = (V, E)$, directed or undirected. No source vertex given.
- **Output:**
 - 2 **timestamps** on each vertex. Integers between 1 and $2|V|$.
 - $d[v]$ = **discovery time** (v turns from white to gray)
 - $f[v]$ = **finishing time** (v turns from gray to black)
 - $\pi[v]$: predecessor of $v = u$, such that v was discovered during the scan of u 's adjacency list.
- Uses the same coloring scheme for vertices as BFS.
- Builds depth-first forest comprising several depth-first trees.

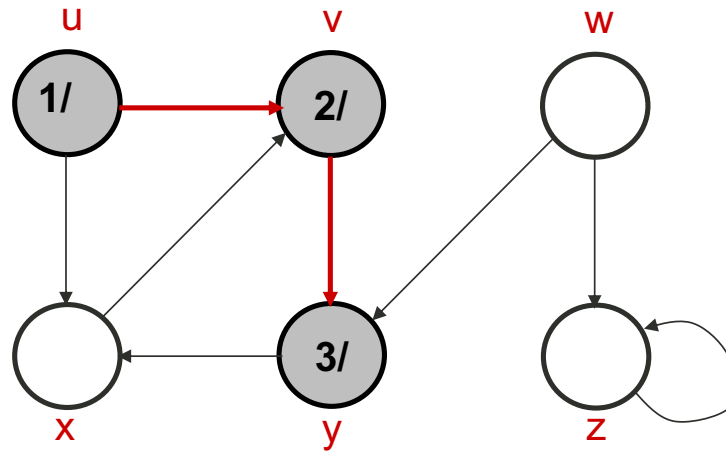
Depth-First Search - Example



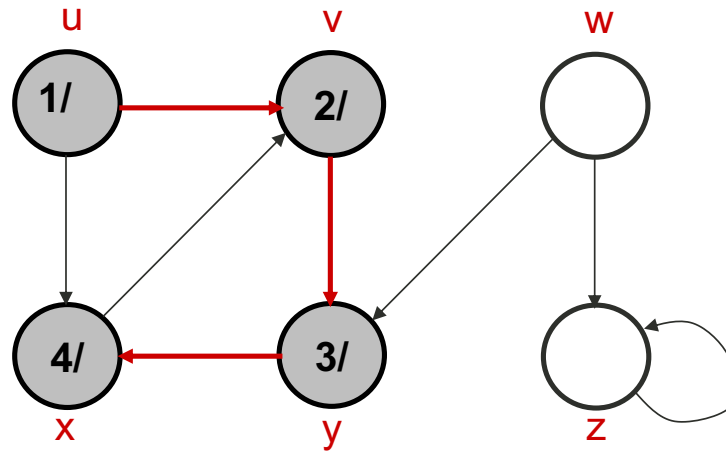
Depth-First Search - Example



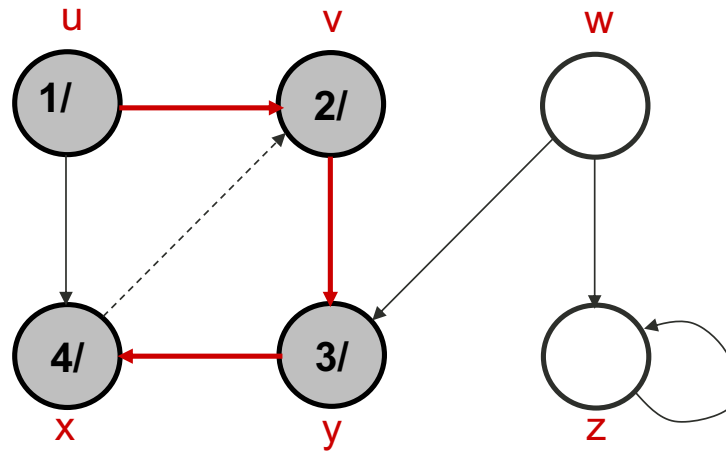
Depth-First Search - Example



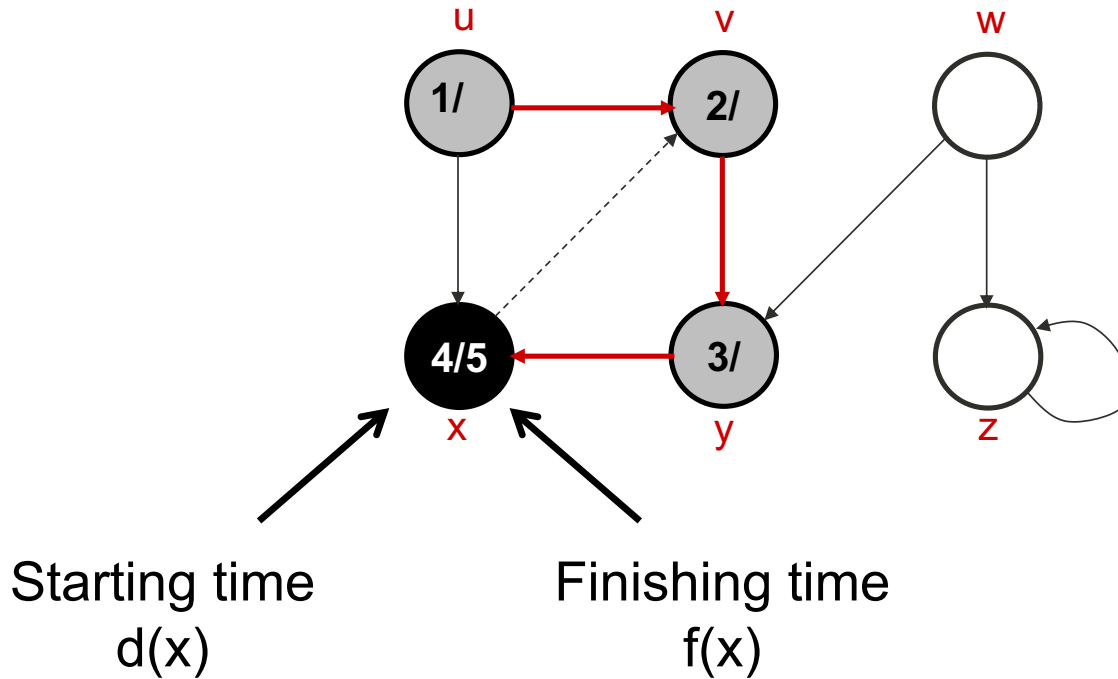
Depth-First Search - Example



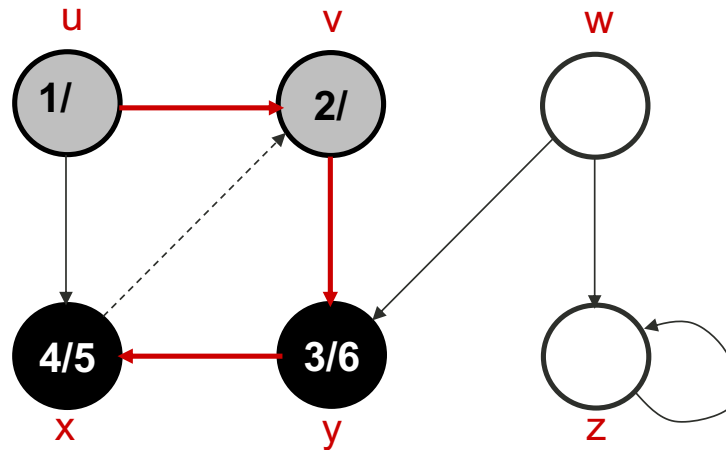
Depth-First Search - Example



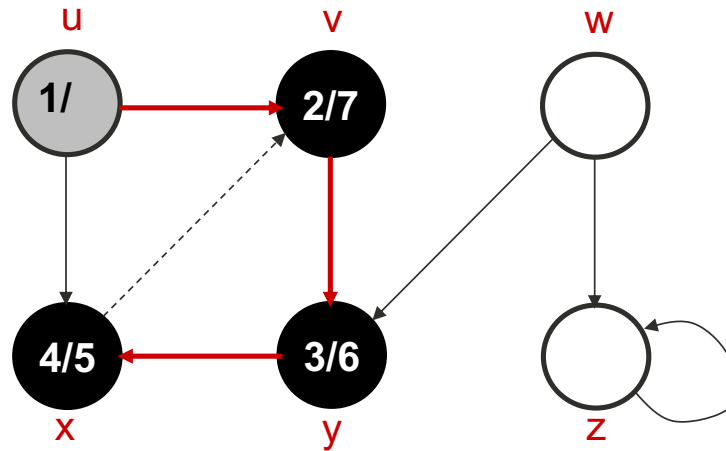
Depth-First Search - Example



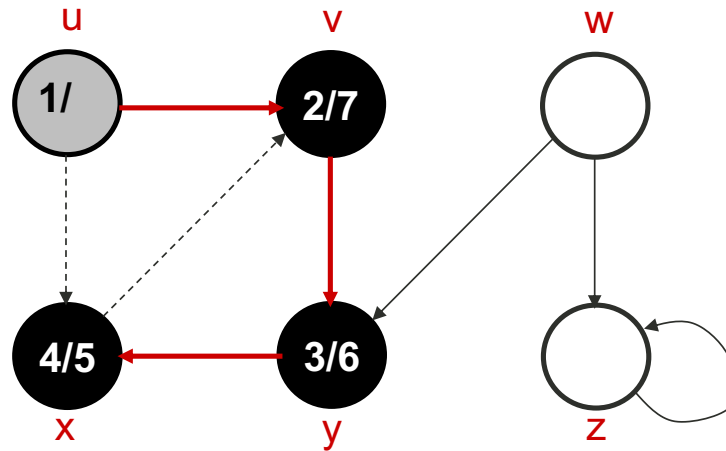
Depth-First Search - Example



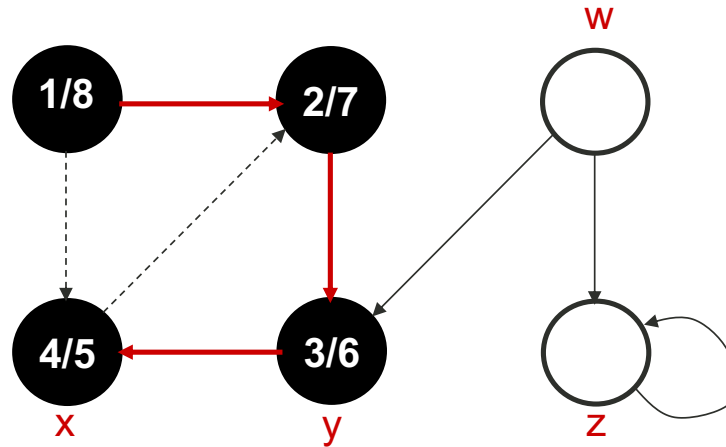
Depth-First Search - Example



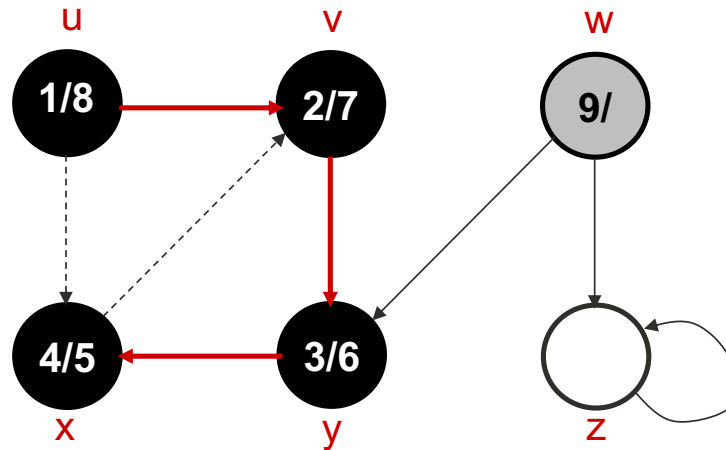
Depth-First Search - Example



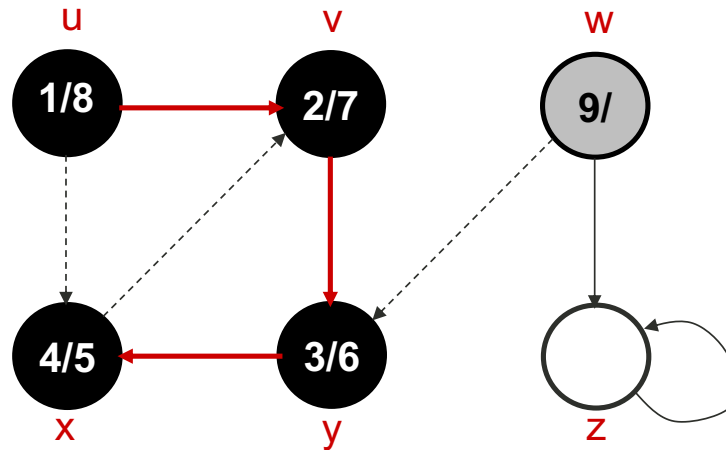
Depth-First Search - Example



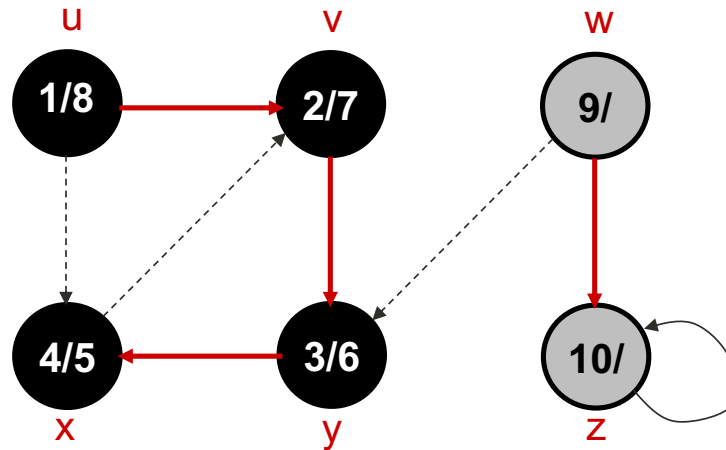
Depth-First Search - Example



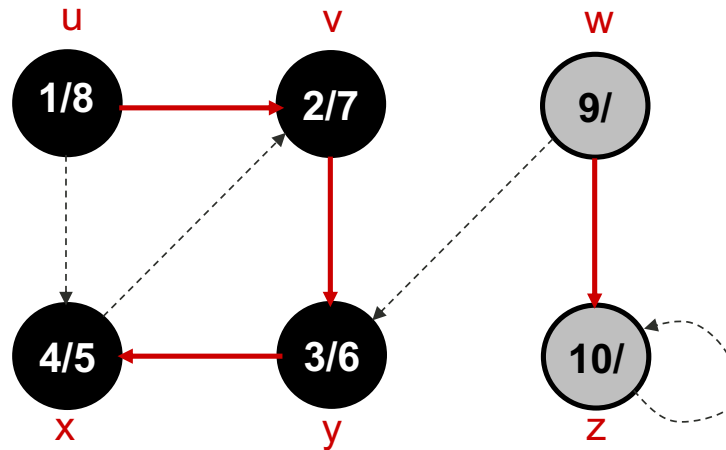
Depth-First Search - Example



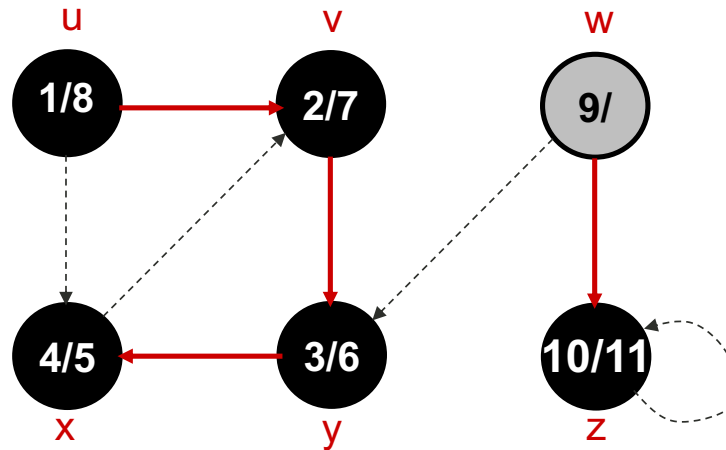
Depth-First Search - Example



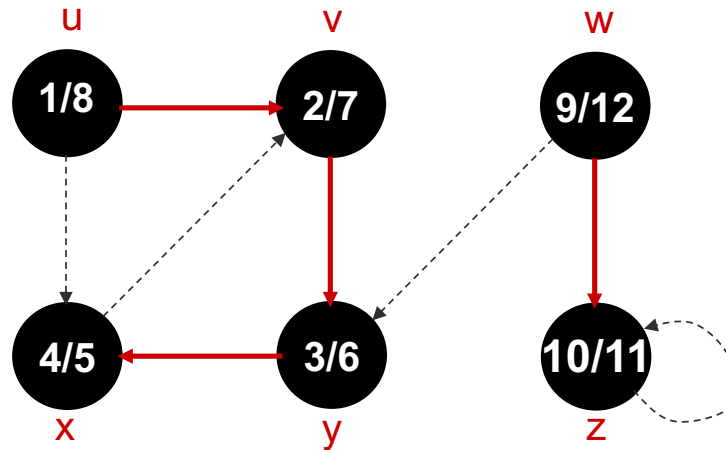
Depth-First Search - Example



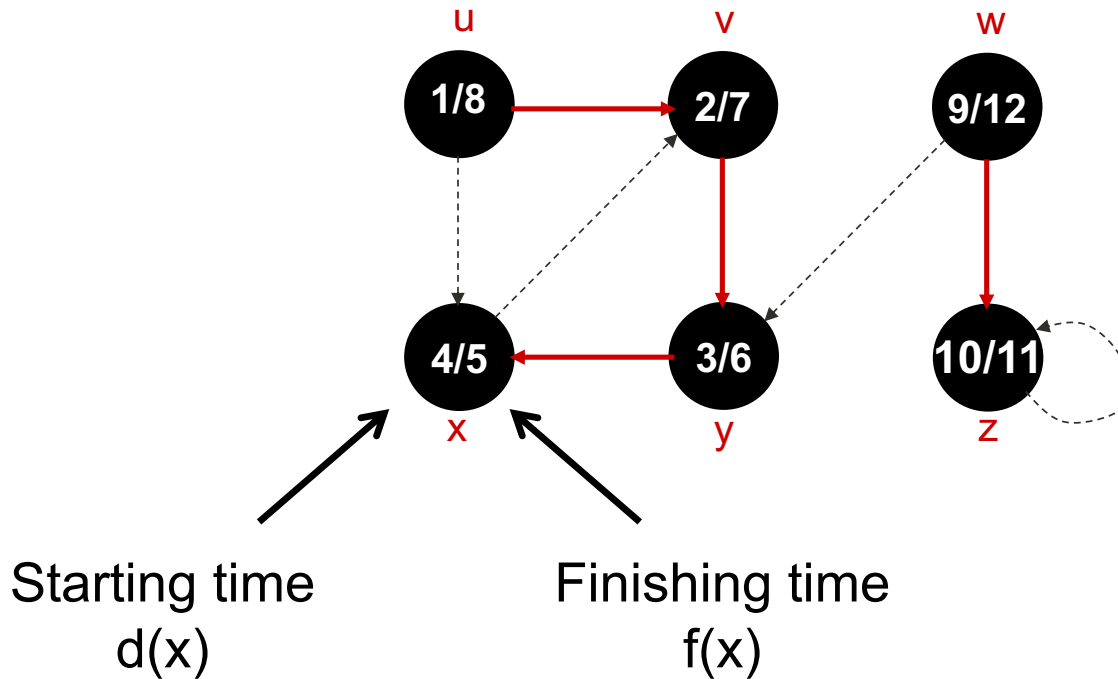
Depth-First Search - Example



Depth-First Search - Example



Depth-First Search - Example



Depth-First Search

DFS(G)

1. **for** each vertex $u \in V[G]$
2. **do** $color[u] \leftarrow \text{white}$
3. $\pi[u] \leftarrow \text{NIL}$
4. $time \leftarrow 0$
5. **for** each vertex $u \in V[G]$
6. **do if** $color[u] = \text{white}$
7. **then** DFS-Visit(u)

Uses a global timestamp **time**.

DFS-Visit(u)

1. $color[u] \leftarrow \text{GRAY}$ # White vertex u has been discovered
2. $time \leftarrow time + 1$
3. $d[u] \leftarrow time$
4. **for** each $v \in Adj[u]$
5. **do if** $color[v] = \text{WHITE}$
6. **then** $\pi[v] \leftarrow u$
7. DFS-Visit(v)
8. $color[u] \leftarrow \text{BLACK}$ # Blacken u ; it is finished.
9. $f[u] \leftarrow time \leftarrow time + 1$

Depth-First Search - Analysis

- Loops on lines 1-2 & 5-7 take $\Theta(V)$ time, excluding time to execute DFS-Visit.
- DFS-Visit is called once for each white vertex $v \in V$ when it's painted gray the first time. Lines 3-6 of DFS-Visit is executed $|\text{Adj}[v]|$ times. The total cost of executing DFS-Visit is $\sum_{v \in V} |\text{Adj}[v]| = \Theta(E)$
- Total running time of DFS is $\Theta(V+E)$.

Depth-First Search - Analysis

RECURSIVEDFS(v):

if v is unmarked

mark v

for each edge vw

RECURSIVEDFS(w)

ITERATIVEDFS(s):

PUSH(s)

while the stack is not empty

$v \leftarrow \text{POP}$

if v is unmarked

mark v

for each edge vw

PUSH(w)

Whatever First Search

BFS and DFS differ essentially only in that one uses a queue and the other uses a stack.

WHATEVERFIRSTSEARCH(s):

```
  put  $s$  into the bag
  while the bag is not empty
    take  $v$  from the bag
    if  $v$  is unmarked
      mark  $v$ 
      for each edge  $vw$ 
        put  $w$  into the bag
```

Bag = Stack : Depth-First (Shortest Path)

Bag = Queue : Breadth-First (Topological Sort)

Bag = Priority Queue : Best-First (Minimum Spanning Tree)

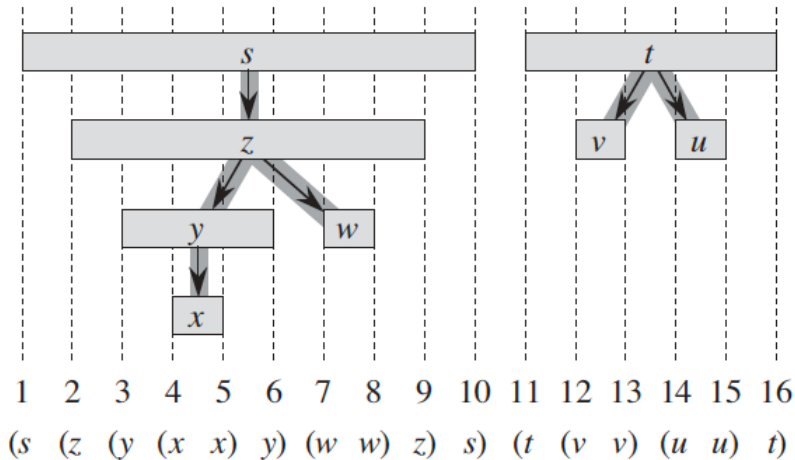
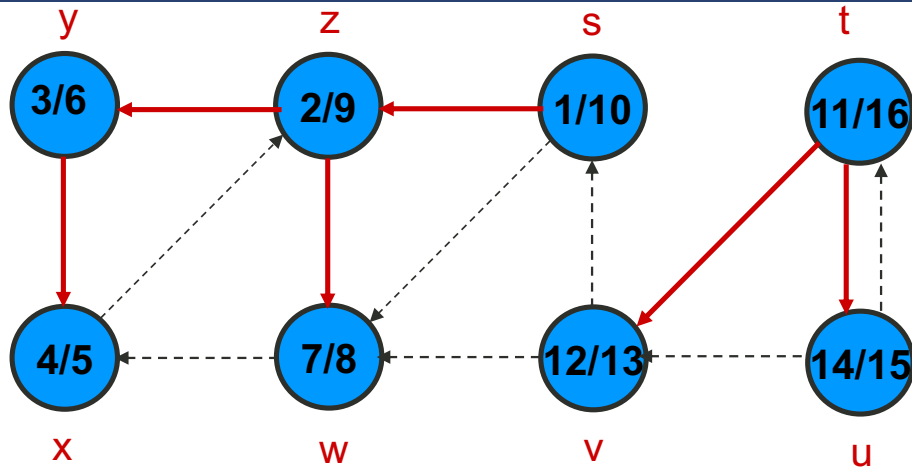
Depth-First Search - Application

- Depth-first search is often a subroutine in another algorithm.
 - Depth-first search yields valuable information about the structure of a graph.
 - The most basic property of depth-first search is that the predecessor subgraph G does indeed form a forest of trees.
 - Another important property of depth-first search is that discovery and finishing times have **parenthesis structure**.
 - If we represent the discovery of vertex u with a left parenthesis “(” and represent its finishing by a right parenthesis “)”, then the history of discoveries and finishings makes a well-formed expression in the sense that the parentheses are properly nested.

OK: ({ }) []
1 2 3 4 5 6

Not OK: ({) }
1 2 3 4

Parenthesis Property



Outline

- Graphs.
 - Introduction.
 - Strong Connected Components / Topological Sort.
 - Network Flow 1.
 - Network Flow 2.
 - Shortest Path.
 - Minimum Spanning Trees.
 - Bipartite Graphs.