

COMP251: Elementary graph algorithms

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Based on (Cormen *et al.*, 2002)

Based on slides from D. Plaisted (UNC)

Announcements

- Assignment 1: Due tonight (no extension)
- Assignment 2: Released next week.
- Midterm: November 1 at 10am on CrowdMark.
- Final: Dec. 12 at 2pm. **In Person!**

Outline

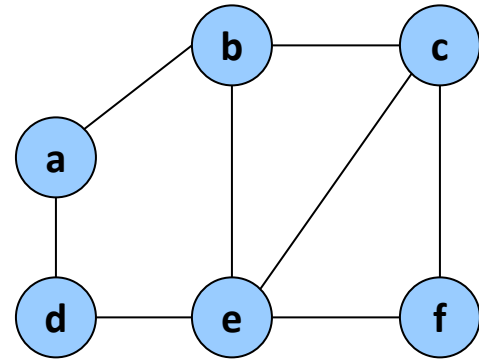
- Vocabulary, definition, and properties of graphs
- Exploring graphs:
 - Breadth First Search (BFS)
 - Depth First Search (DFS)
- Parenthesis theorem

Why DFS & BFS again?

We will cover many algorithms on graphs based on these techniques.

Graphs

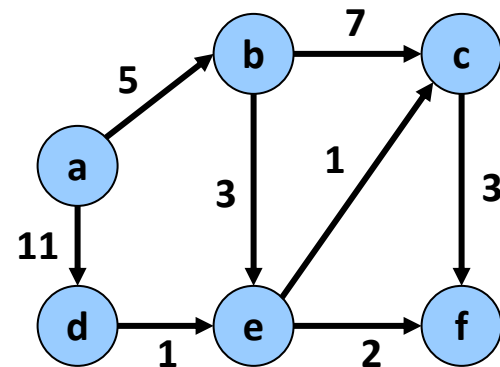
- *Graph* $G = (V, E)$
 - V = set of vertices
 - E = set of edges $\subseteq (V \times V)$



- 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: $|E| \ll |V|^2$.

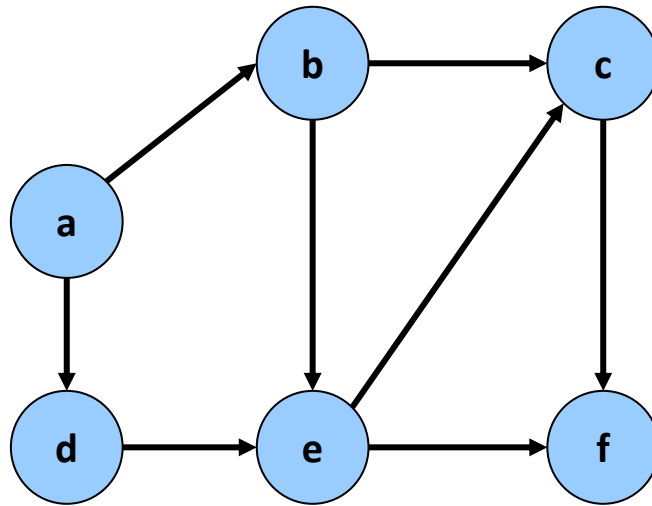
- $|E| = O(|V|^2)$



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 connected:
 - There is a path between every pair of vertices.
 - $|E| \geq |V| - 1$.
 - Furthermore, if $|E| = |V| - 1$, then G is a tree.

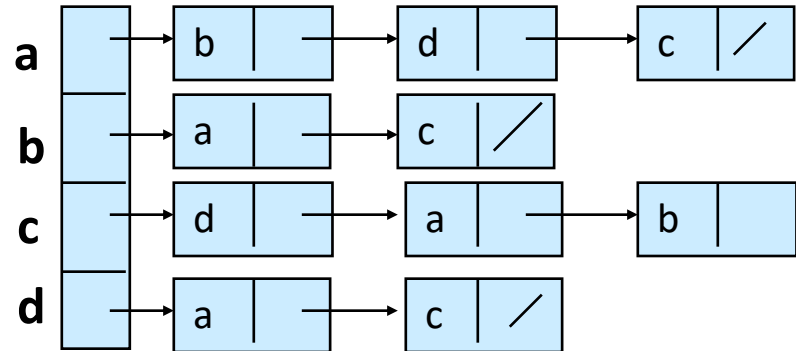
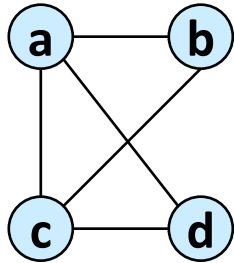
Vocabulary



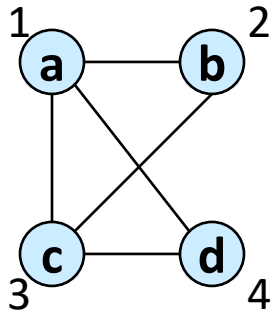
- Ingoing edges of u : $\{ (v, u) \in E \}$
e.g. $in(e) = \{ (b, e), (d, e) \}$
- Outgoing edges of u : $\{ (u, v) \in E \}$
e.g. $out(d) = \{ (d, e) \}$
- In-degree(u): $| in(u) |$
- Out-degree(u): $| out(u) |$

Representation of Graphs

- Two standard ways.
 - Adjacency Lists.



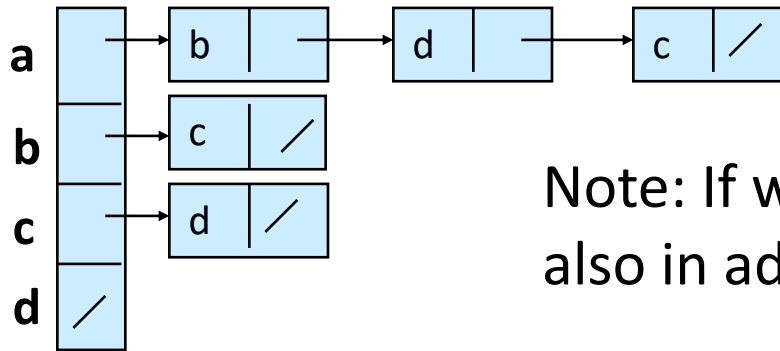
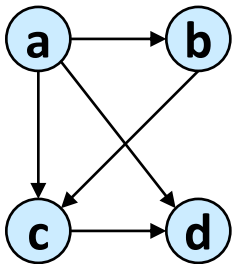
- Adjacency Matrix.



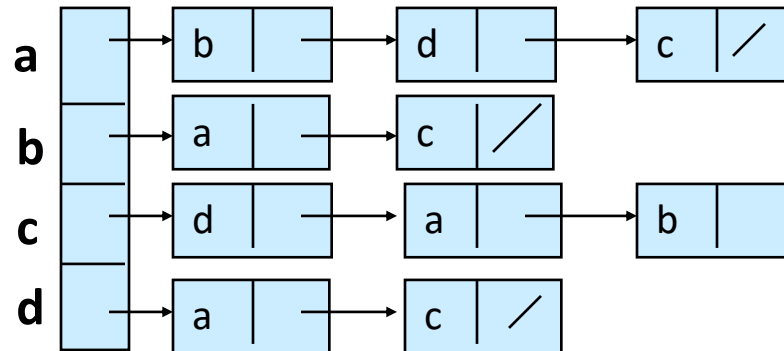
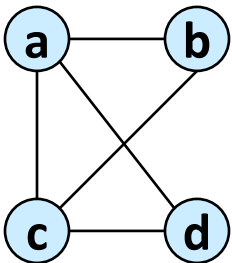
	1	2	3	4
1	0	1	1	1
2	1	0	1	0
3	1	1	0	1
4	1	0	1	0

Adjacency Lists

- 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.



Storage Requirement

- For directed graphs:

- Sum of lengths of all adj. lists is

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

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

No. of edges leaving v



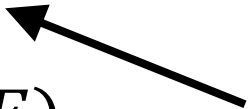
- For undirected graphs:

- Sum of lengths of all adj. lists is

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

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

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



Pros and Cons: adj list

- Pros

- Space-efficient, when a graph is sparse.
- Can be modified to support many graph variants.

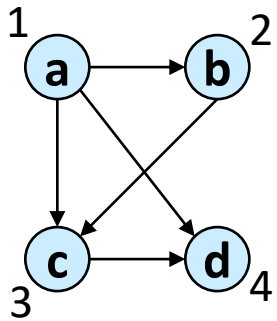
- Cons

- 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.

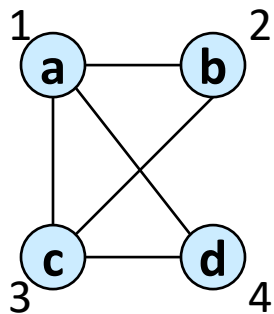
Adjacency Matrix

- $|V| \times |V|$ 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



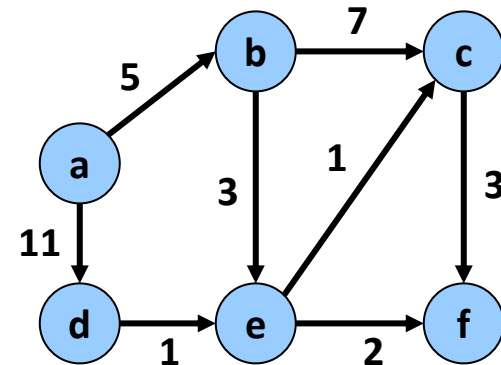
	1	2	3	4
1	0	1	1	1
2	1	0	1	0
3	1	1	0	1
4	1	0	1	0

$A = A^T$ for undirected graphs.

Space and Time

- **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



Graph-searching Algorithms (COMP250)

- 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).

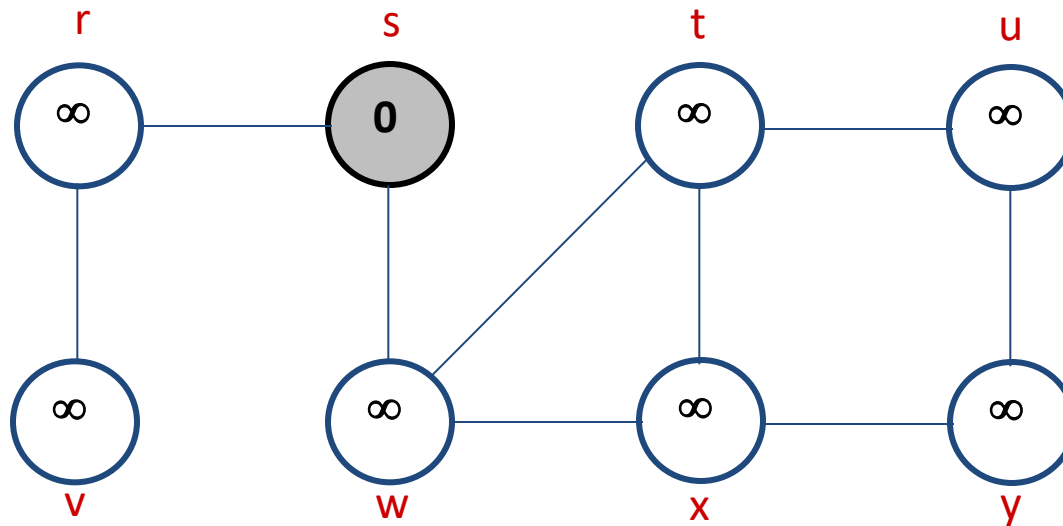
Breadth-first Search

- Expands the frontier between discovered and undiscovered vertices uniformly across the breadth of the frontier.
 - 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.
- 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.

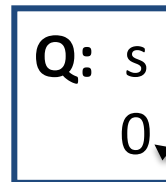
Breadth-first Search

- **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.

Example (BFS)



We use a priority queue to determine the next vertices to visit.



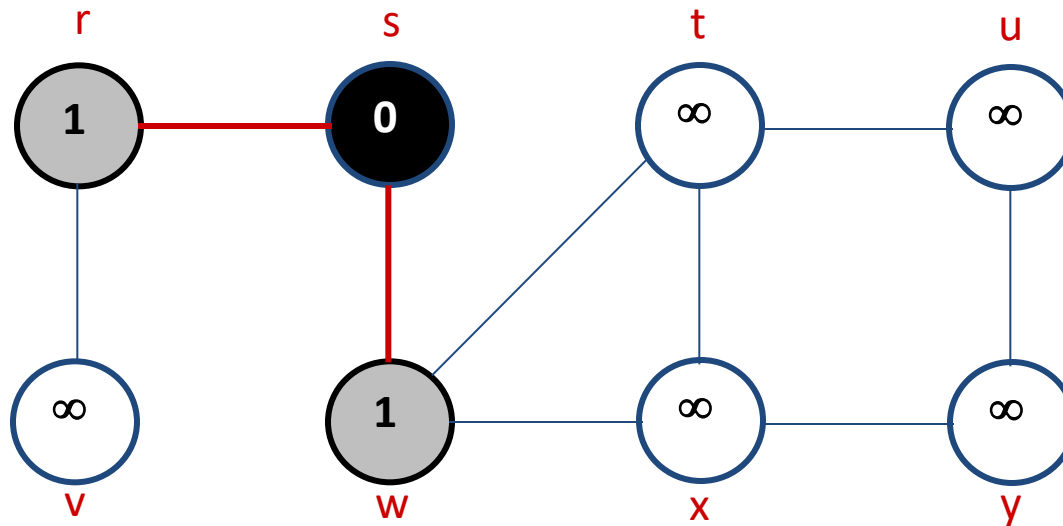
The first vertex we add in the queue is the source.

Priority = distance from the source.
Lower the distance, higher the priority.

Example (BFS)

Color code:

- White: not visited yet
- Gray: visited but neighborhood not fully explored
- Black: Complete

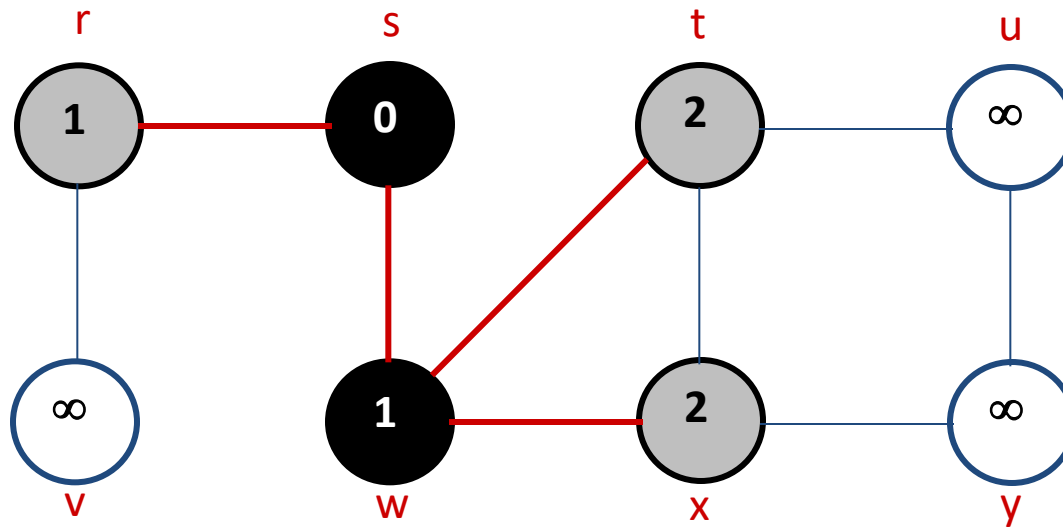


We store in the queue the vertices in the neighborhood of the current vertex.

Q: w r
1 1

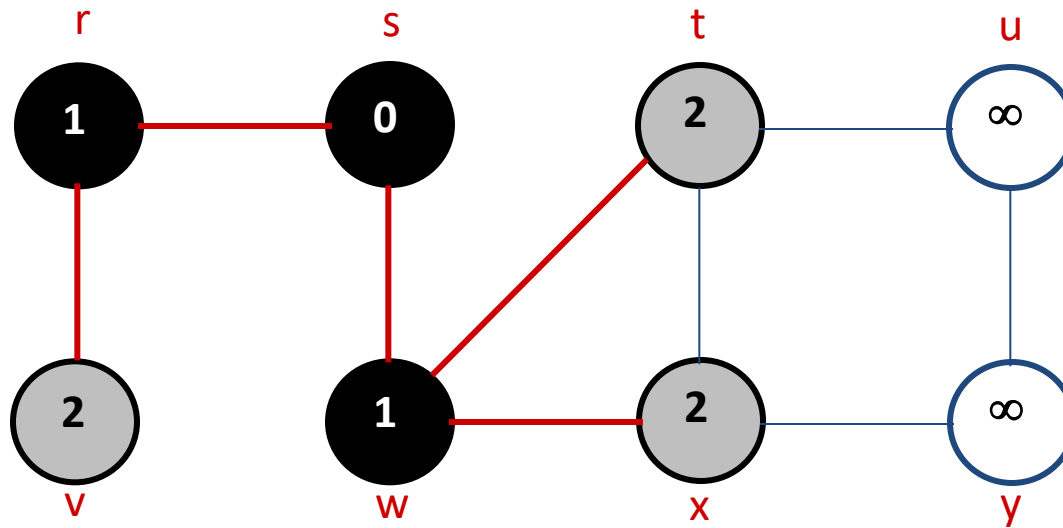
The vertices are indexed with the number of edges from the source.

Example (BFS)



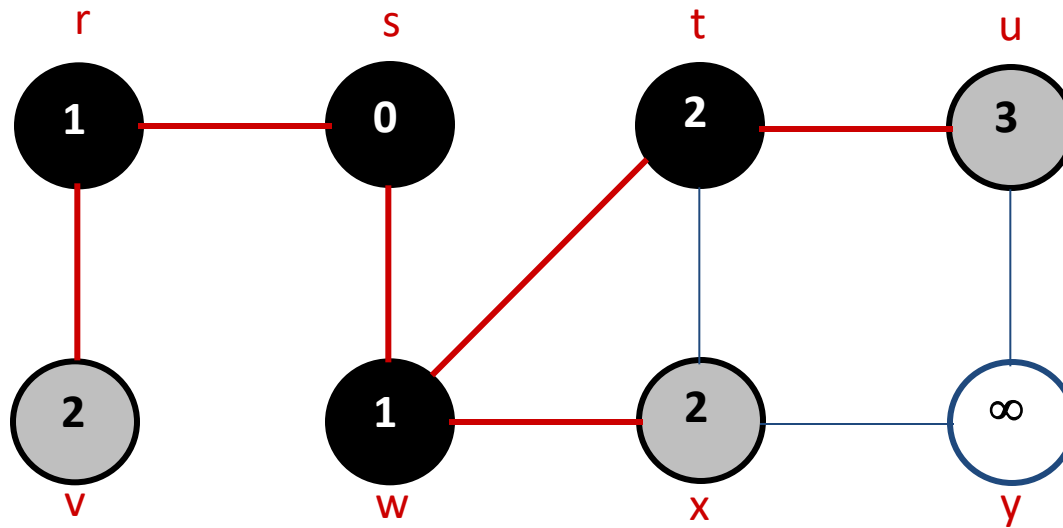
Q: r t x
1 2 2

Example (BFS)



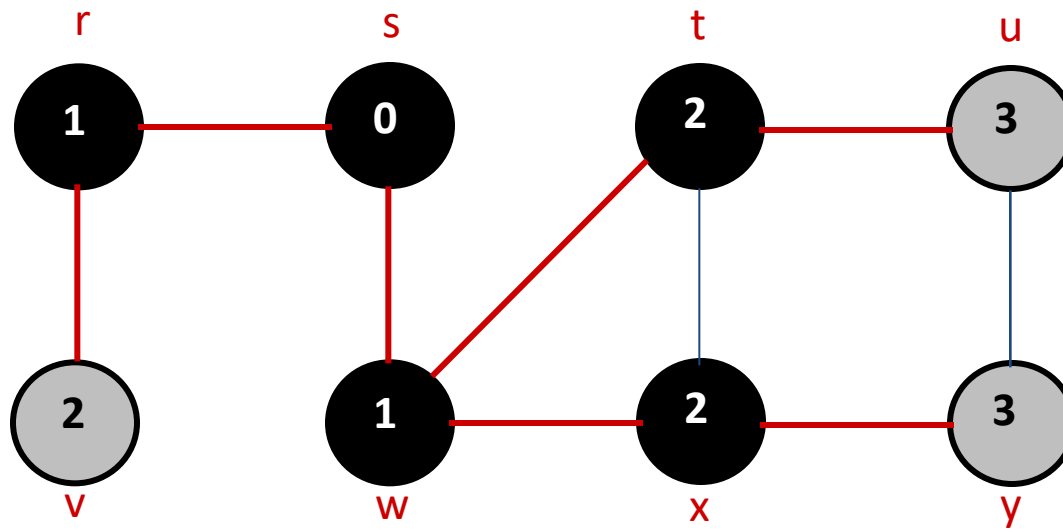
Q: t x v
2 2 2

Example (BFS)



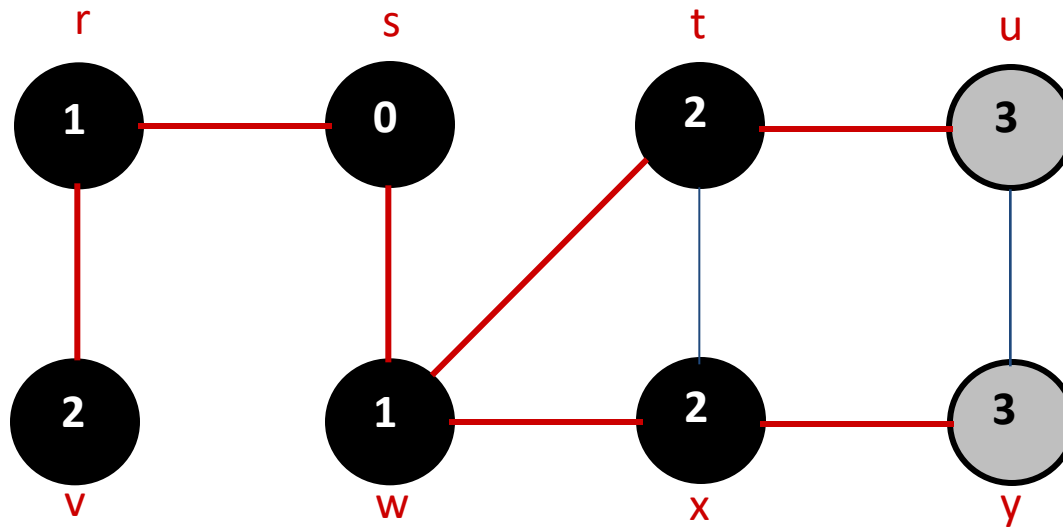
Q: x v u
2 2 3

Example (BFS)



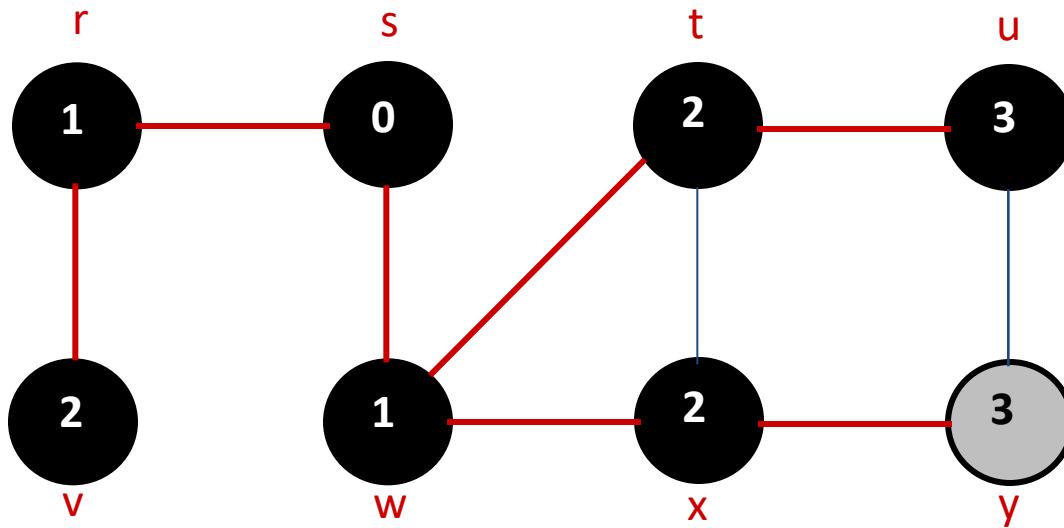
Q: v u y
2 3 3

Example (BFS)



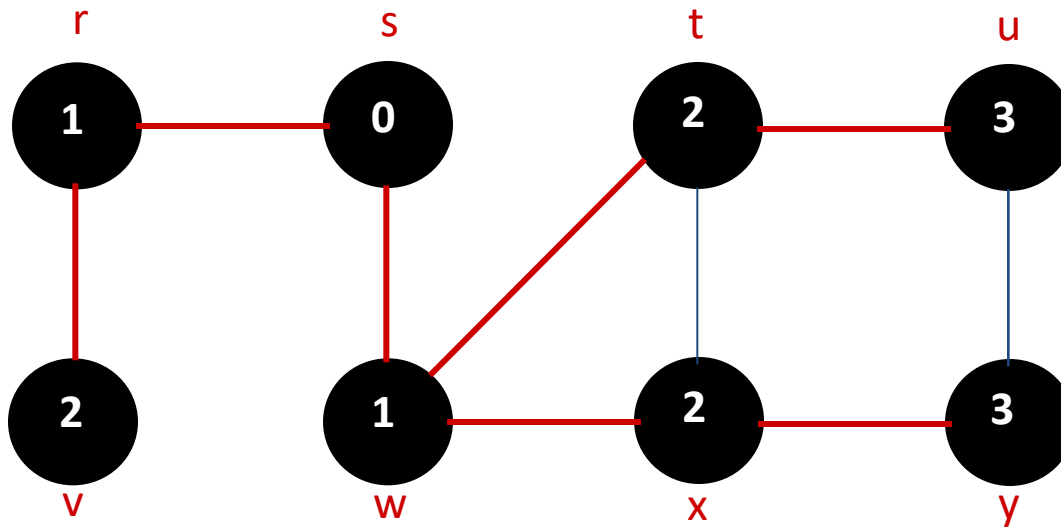
Q: u y
3 3

Example (BFS)



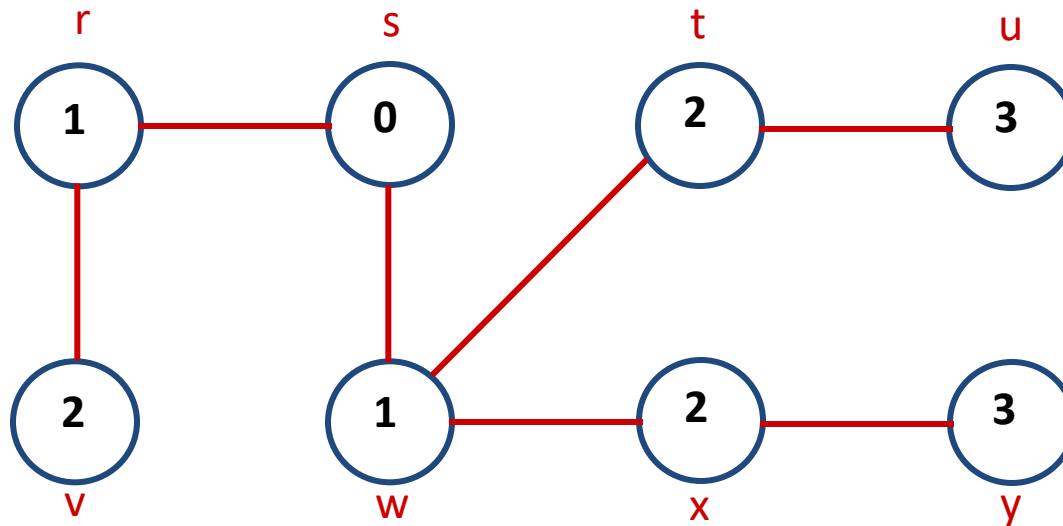
Q: y
3

Example (BFS)



Q: \emptyset

Example (BFS)



The index of each vertex stores the length of the shortest path to reach them (Note: unweighted graph!).

BF Tree

Analysis of BFS

- 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.

Depth-first Search (DFS)

- 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.

Pseudo-code

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)

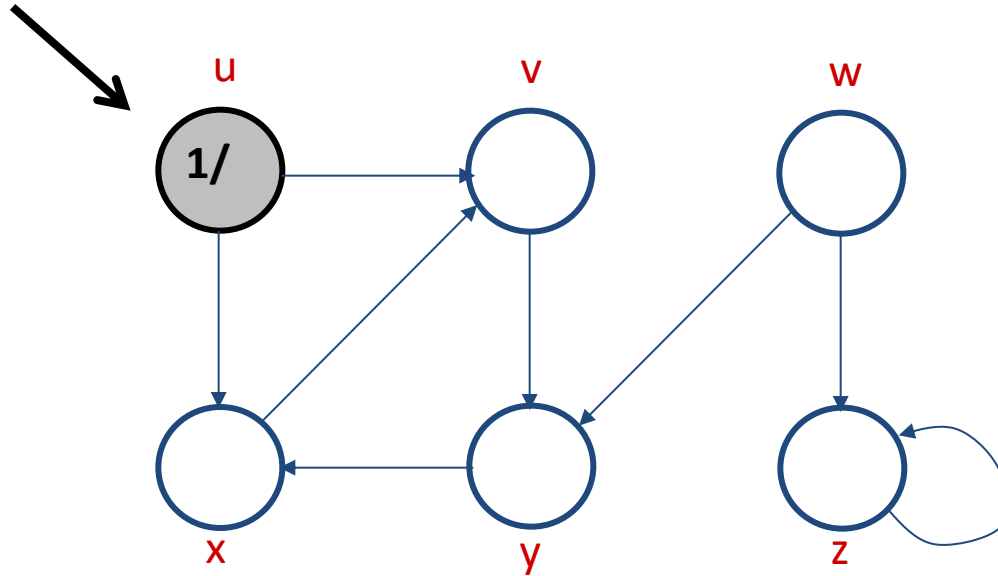
Uses a global timestamp *time*.

DFS-Visit(u)

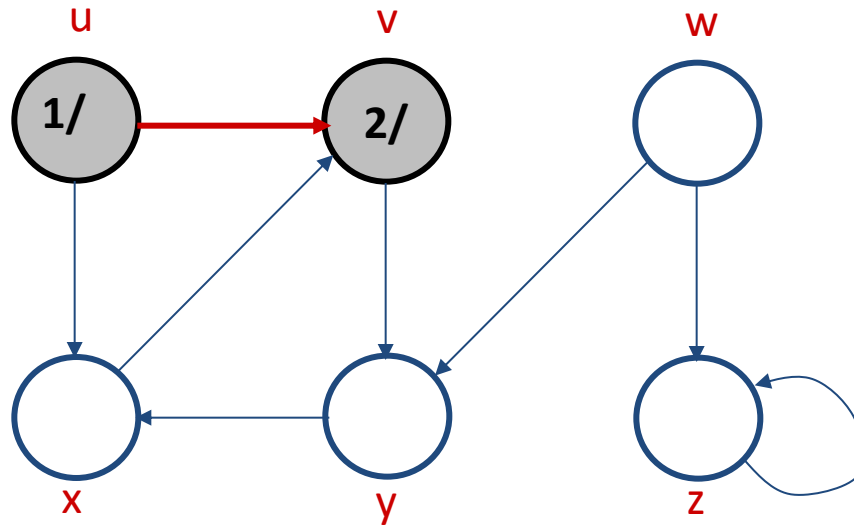
1. $color[u] \leftarrow \text{GRAY}$ # White vertex u has been discovered
2. $time \leftarrow time + 1$
3. $d[u] \leftarrow time$
4. **for** each $v \in Adj[u]$
5. **do if** $color[v] = \text{WHITE}$
6. **then** $\pi[v] \leftarrow u$
7. DFS-Visit(v)
8. $color[u] \leftarrow \text{BLACK}$ # Blacken u ;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

Example (DFS)

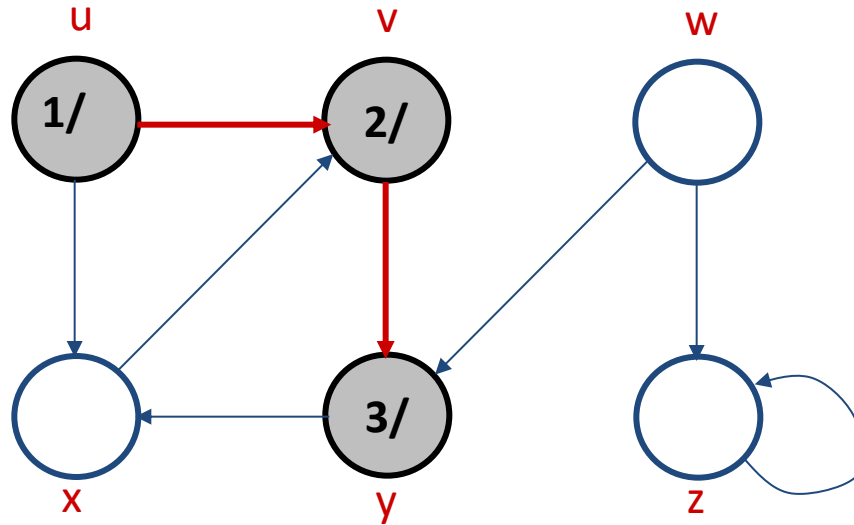
Starting time
 $d[u]$



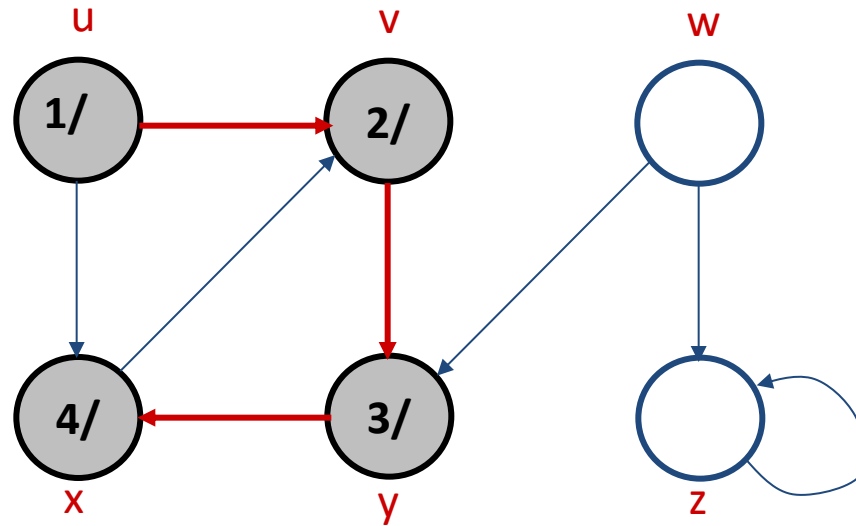
Example (DFS)



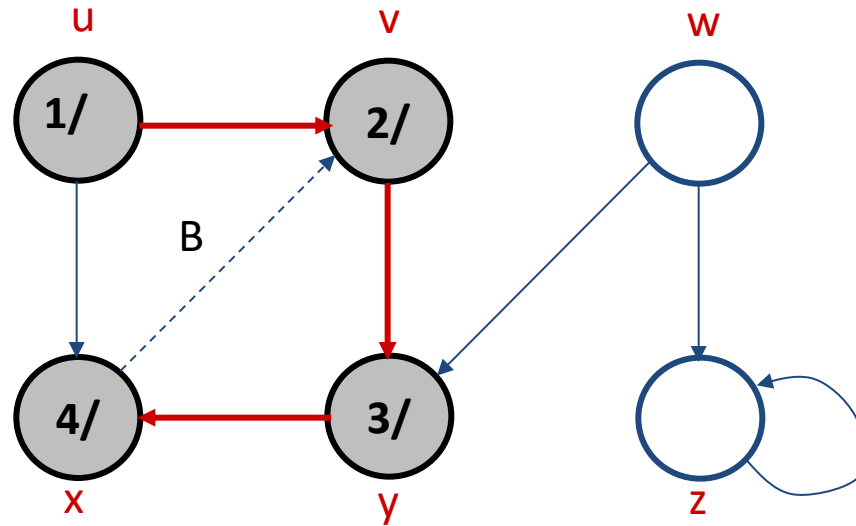
Example (DFS)



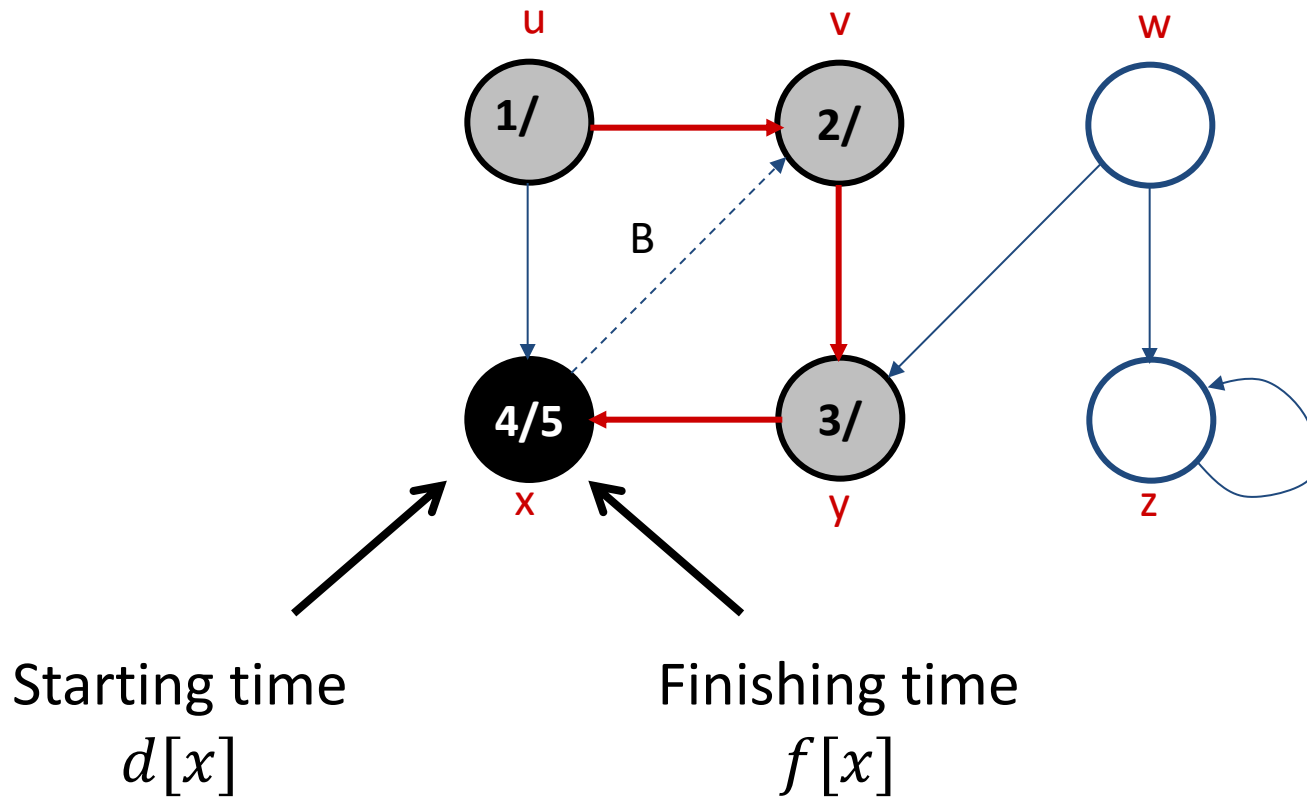
Example (DFS)



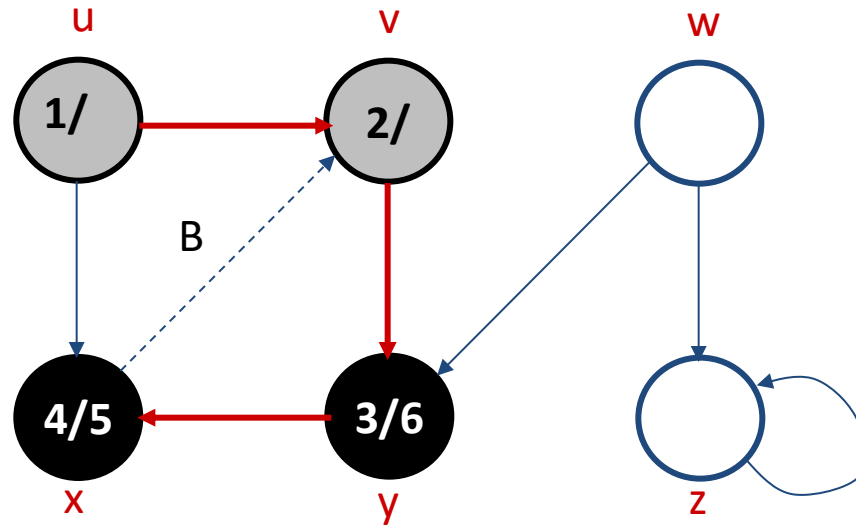
Example (DFS)



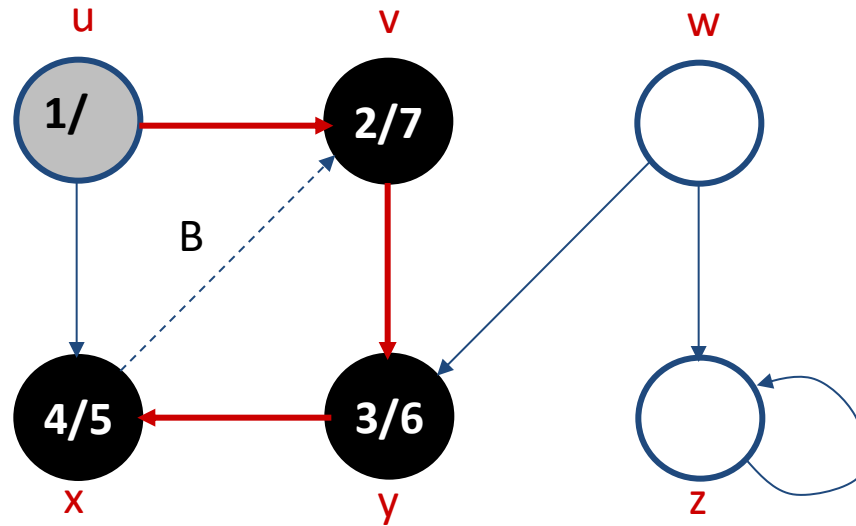
Example (DFS)



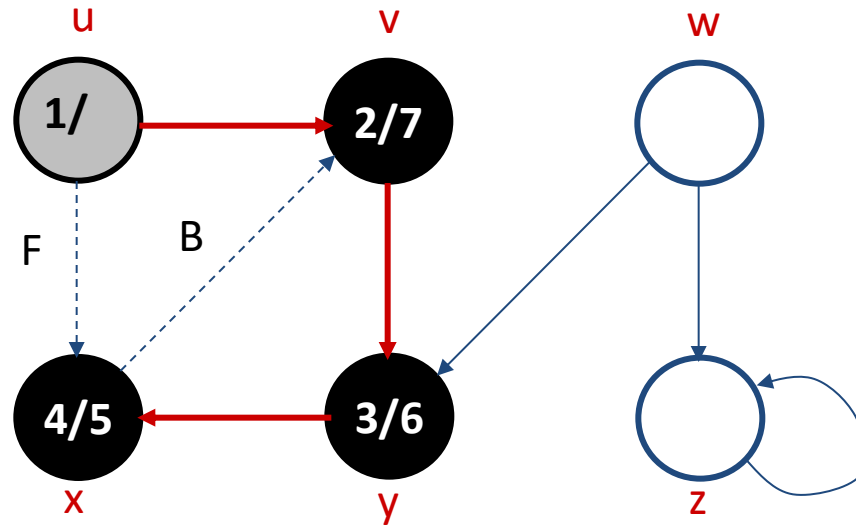
Example (DFS)



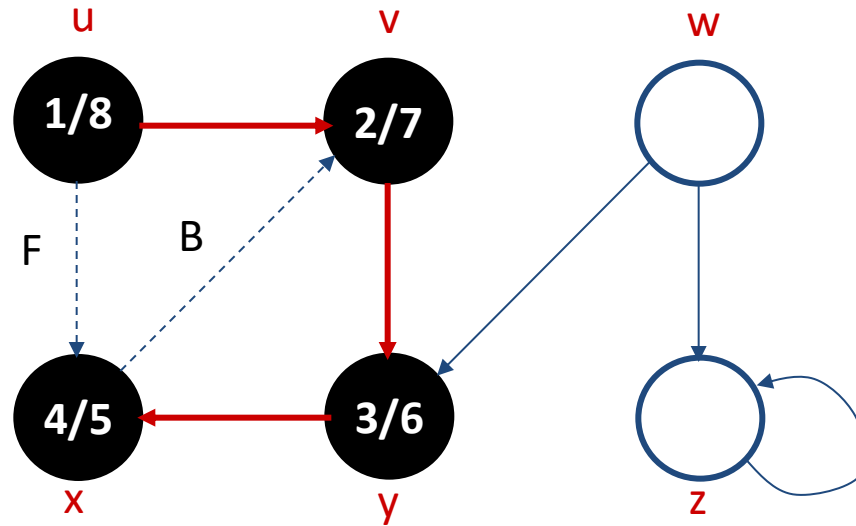
Example (DFS)



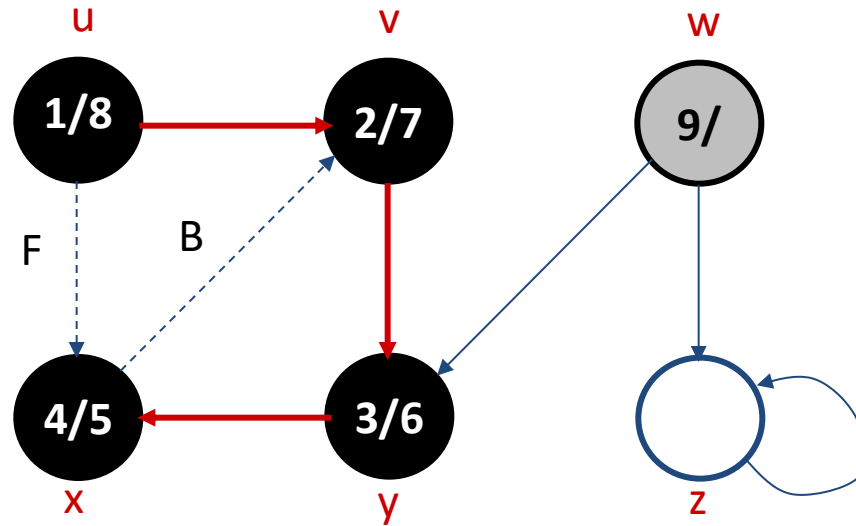
Example (DFS)



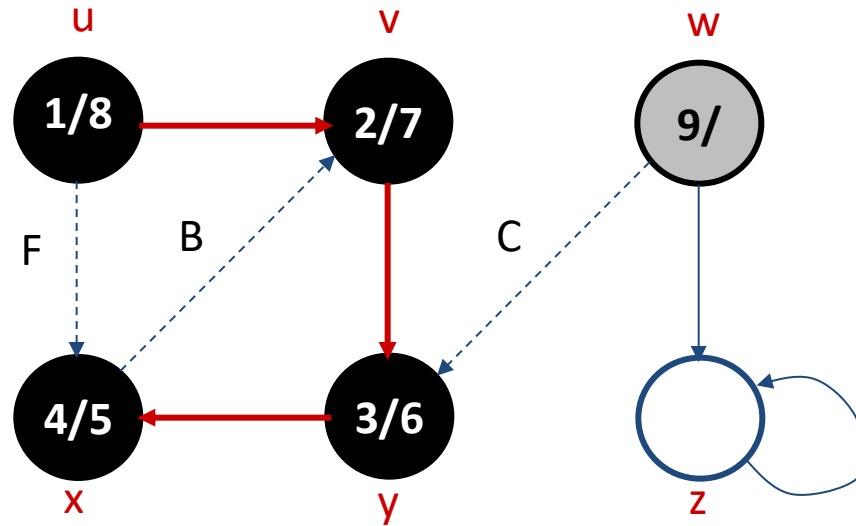
Example (DFS)



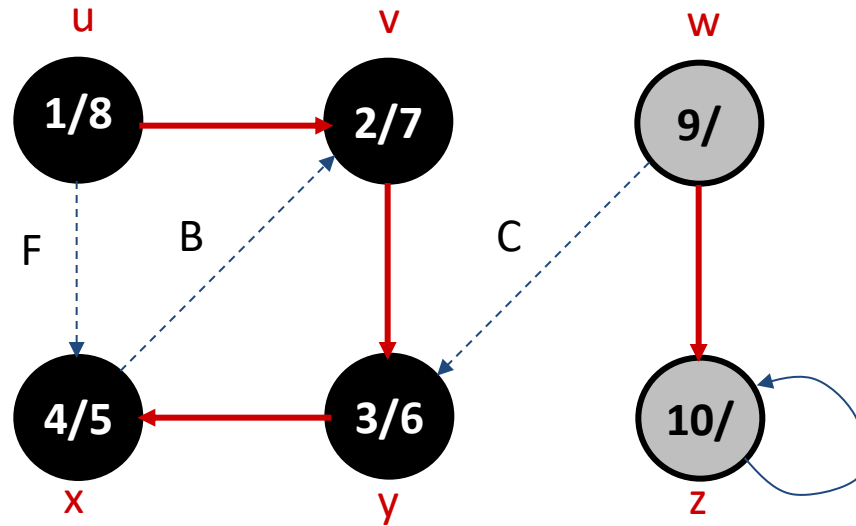
Example (DFS)



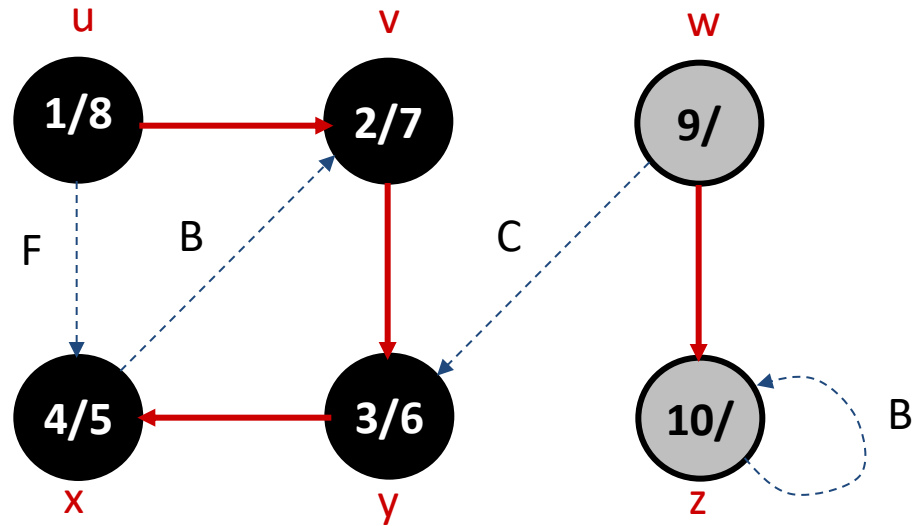
Example (DFS)



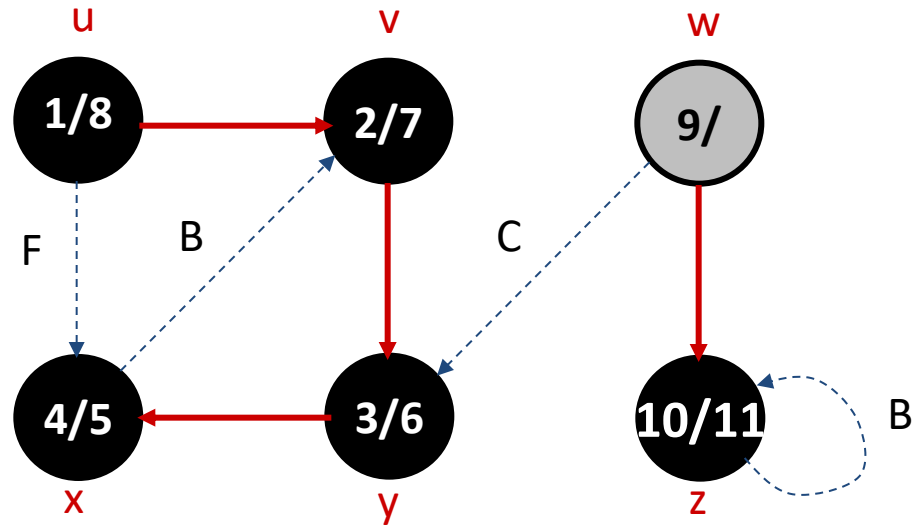
Example (DFS)



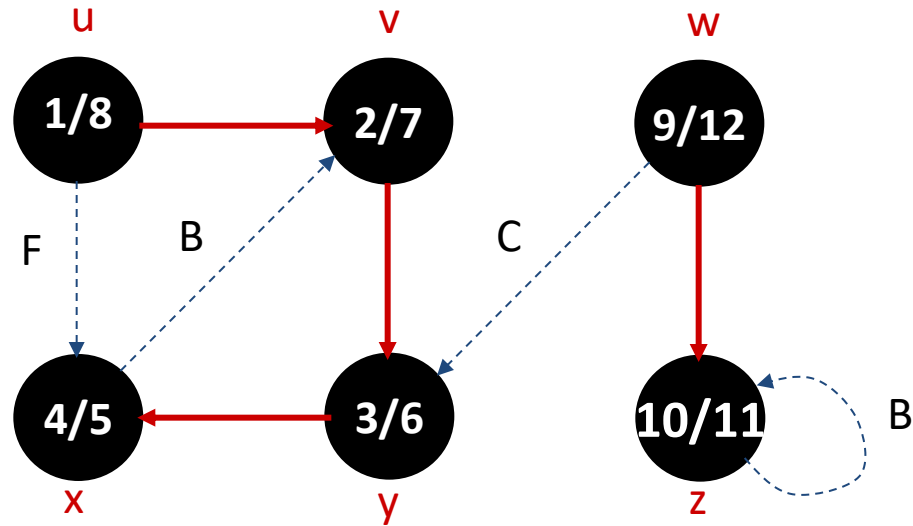
Example (DFS)



Example (DFS)



Example (DFS)



Analysis of DFS

- 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 $|Adj[v]|$ times. The total cost of executing DFS-Visit is $\sum_{v \in V} |Adj[v]| = \Theta(E)$
- Total running time of DFS is $\Theta(V + E)$.

Parenthesis Theorem

Theorem 1:

For all u, v , exactly one of the following holds:

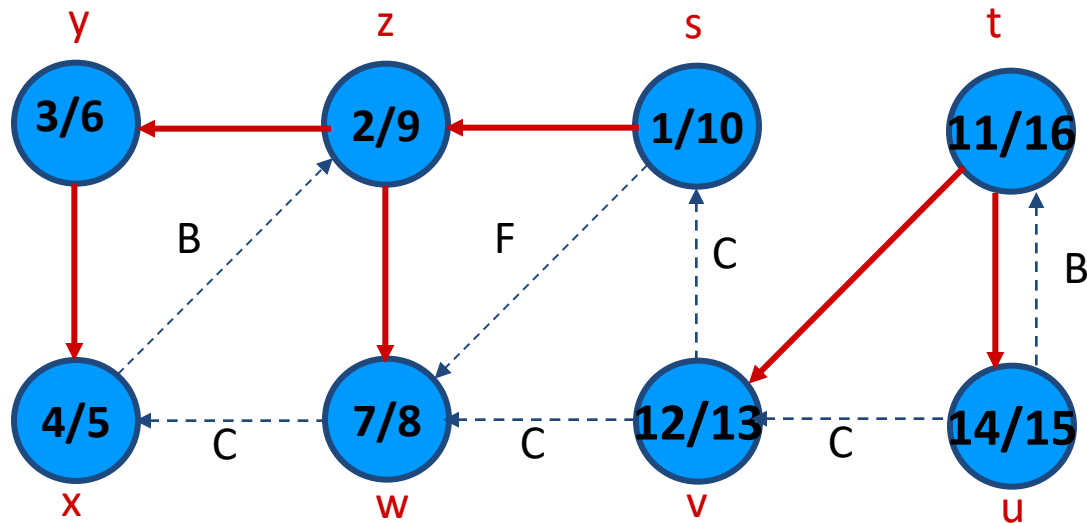
1. $d[u] < f[u] < d[v] < f[v]$ or $d[v] < f[v] < d[u] < f[u]$ and neither u nor v is a descendant of the other.
2. $d[u] < d[v] < f[v] < f[u]$ and v is a descendant of u .
3. $d[v] < d[u] < f[u] < f[v]$ and u is a descendant of v .

- ◆ So $d[u] < d[v] < f[u] < f[v]$ cannot happen.
- ◆ Like parentheses:
 - ◆ OK: $() [] ([]) [()]$
 - ◆ Not OK: $([]) [()]$

Corollary

v is a proper descendant of u if and only if $d[u] < d[v] < f[v] < f[u]$.

Example (Parenthesis Theorem)



(s (z (y (x x) y) (w w) z) s) (t (v v) (u u) t)