

COMP319 Algorithms 1

Lecture 13

Graph Search

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Textbook Chapters 23 and 24

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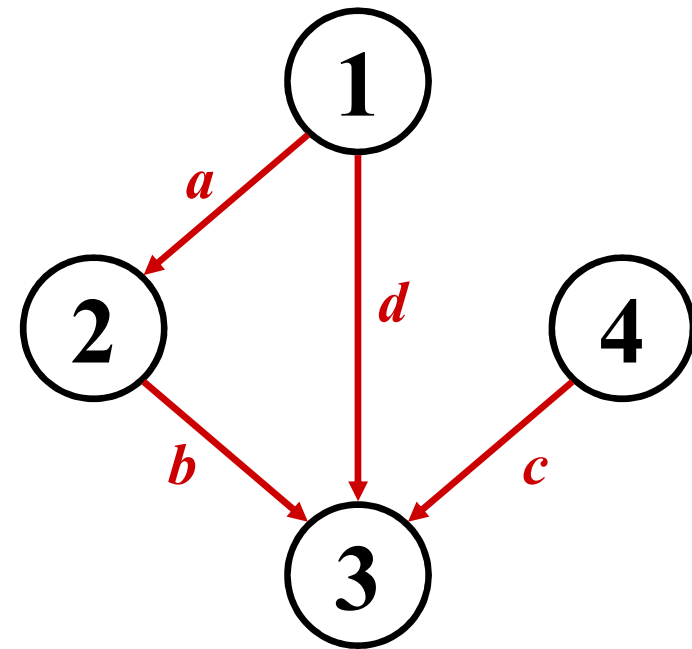
Graph definition

Implementation of graphs

GRAPH ALGORITHMS

Graphs

- A graph $G = (V, E)$
 - V = set of vertices (nodes)
 - E = set of edges = subset of $V \times V$
(Cartesian product of two vertices)
 - Thus $|E| = O(|V|^2)$



Graph Variations

- Variations:
 - A *connected graph* has a path from every vertex to every other
 - In an *undirected graph*:
 - Edge (u,v) = edge (v,u)
 - No self-loops
 - In a *directed graph*:
 - Edge (u,v) goes from vertex u to vertex v , notated $u \rightarrow v$

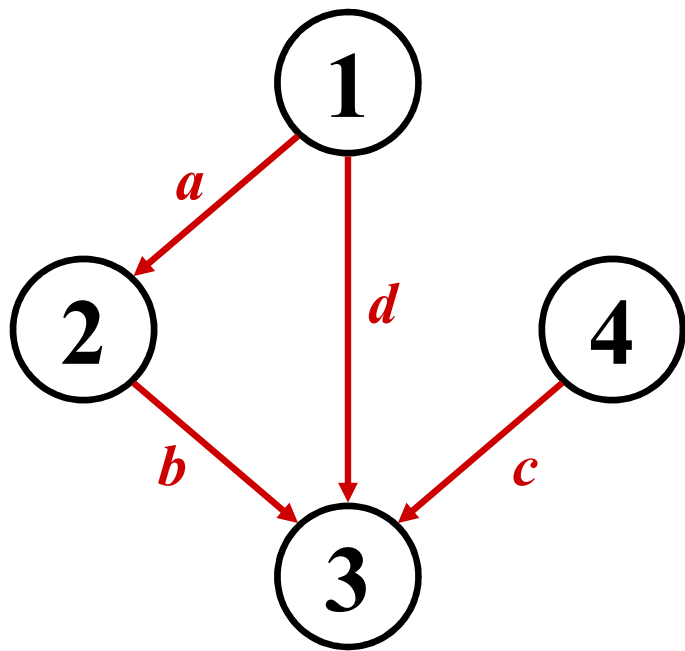
Graph Variations

- More variations:
 - A *weighted graph* associates weights with either the edges or the vertices
 - E.g., a road map: edges might be weighted w/ distance
 - A *multigraph* allows multiple edges between the same vertices
 - E.g., the call graph in a program (a function can get called from multiple points in another function)

Representing Graphs

- We will typically express running times in terms of $|E|$ and $|V|$ (often dropping the $|$'s)
 - If $|E| \approx |V|^2$ the graph is *dense*
 - If $|E| \approx |V|$ the graph is *sparse*
 - If you know you are dealing with dense or sparse graphs, different data structures may make sense
- Assume $V = \{1, 2, \dots, n\}$, an *adjacency matrix* represents the graph as a $n \times n$ matrix A :
 - $A[i, j] = 1$ if edge $(i, j) \in E$ (or weight of edge)
= 0 if edge $(i, j) \notin E$

Graphs: Adjacency Matrix



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

Graphs: Adjacency Matrix

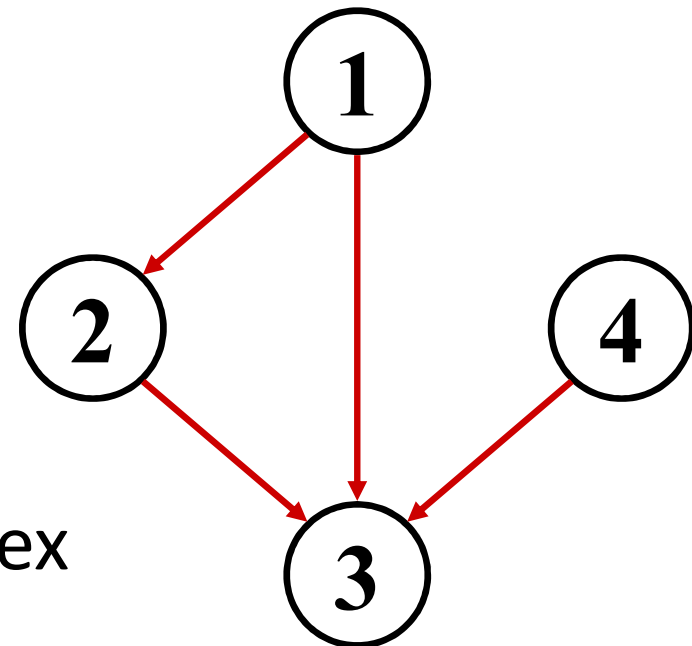
- *How much storage does the adjacency matrix require?*
- A: $O(V^2)$
- *What is the minimum amount of storage needed by an adjacency matrix representation of an undirected graph with 4 vertices?*
- A: 6 bits
 - Undirected graph \rightarrow matrix is symmetric
 - No self-loops \rightarrow don't need diagonal

Graphs: Adjacency Matrix

- The adjacency matrix is a dense representation
 - Usually too much storage for large graphs
 - But can be very efficient for small graphs
- Most large interesting graphs are sparse
 - E.g., planar graphs, in which no edges cross, have $|E| = O(|V|)$ by Euler's formula
 - For this reason the *adjacency list* is often a more appropriate representation

Graphs: Adjacency List

- Adjacency list: for each vertex $v \in V$, store a list of vertices adjacent to v
- Example:
 - $\text{Adj}[1] = \{2,3\}$
 - $\text{Adj}[2] = \{3\}$
 - $\text{Adj}[3] = \{\}$
 - $\text{Adj}[4] = \{3\}$
- Variation: can also keep a list of edges coming *into* vertex



Graphs: Adjacency List

- How much storage is required?
 - The *degree* of a vertex v = # incident edges
 - Directed graphs have in-degree, out-degree
 - For directed graphs, # of items in adjacency lists is
$$\sum \text{out-degree}(v) = |E|$$
takes $\Theta(V + E)$ storage (*Why?*)
 - For undirected graphs, # items in adj lists is
$$\sum \text{degree}(v) = 2 |E| \quad (\textit{handshaking lemma})$$
also $\Theta(V + E)$ storage
- So: Adjacency lists take $O(V+E)$ storage

Graph Searching

- Given: a graph $G = (V, E)$, directed or undirected
- Goal:
 - Find methods (algorithms) that explore every vertex and every edge
- Ultimately: build a **TREE** on the graph
 - Pick a vertex as the root
 - Choose certain edges to produce a tree
 - Note: might also build a *forest* if graph is not connected

BREADTH-FIRST SEARCH

Breadth-First Search

- “Explore” a graph, turning it into a tree
 - One vertex at a time
 - Expand frontier of explored vertices across the *breadth* of the frontier
- Builds a tree over the graph
 - Pick a *source vertex* to be the root
 - Find (“discover”) its children, then their children, etc.

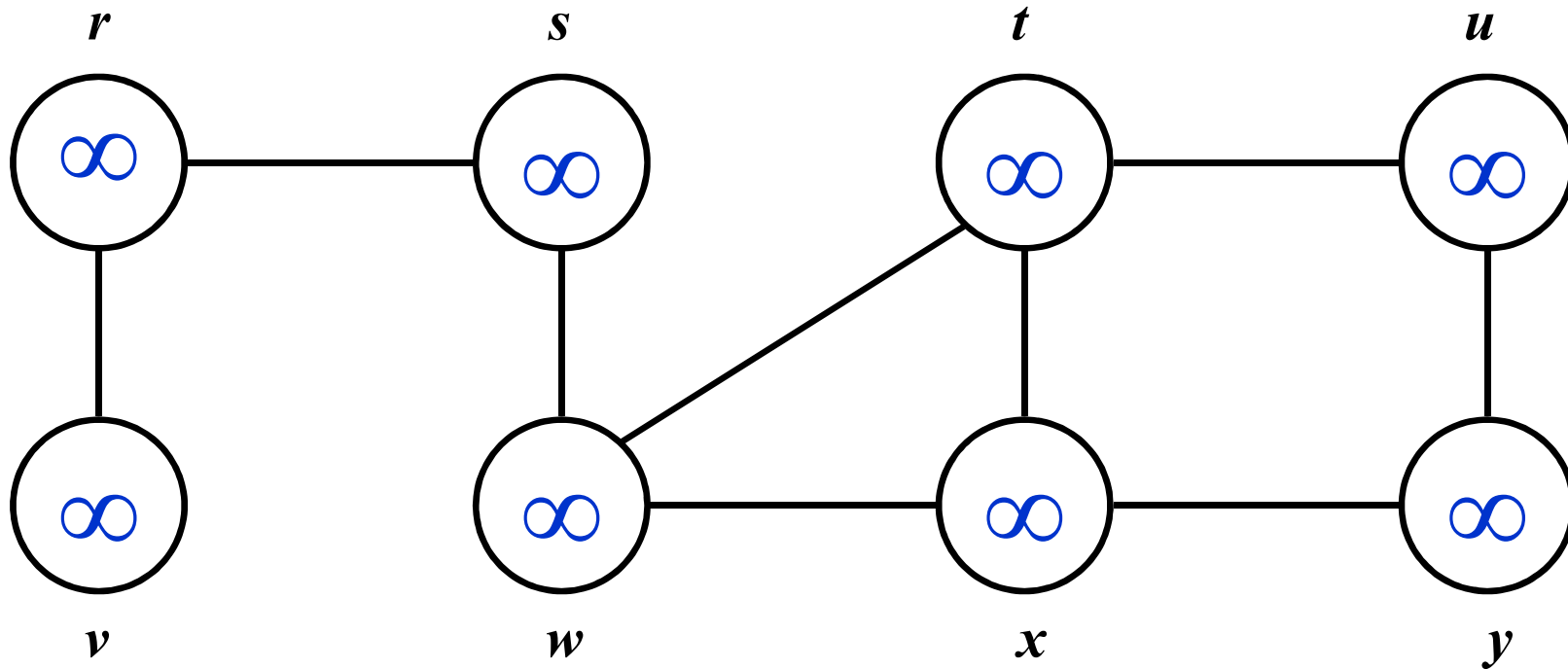
Breadth-First Search

- Again will associate vertex “colors” to guide the algorithm
 - White vertices have not been discovered
 - All vertices start out white
 - Grey vertices are discovered but not fully explored
 - They may be adjacent to white vertices
 - Black vertices are discovered and fully explored
 - They are adjacent only to black and gray vertices
- Explore vertices by scanning adjacency list of grey vertices

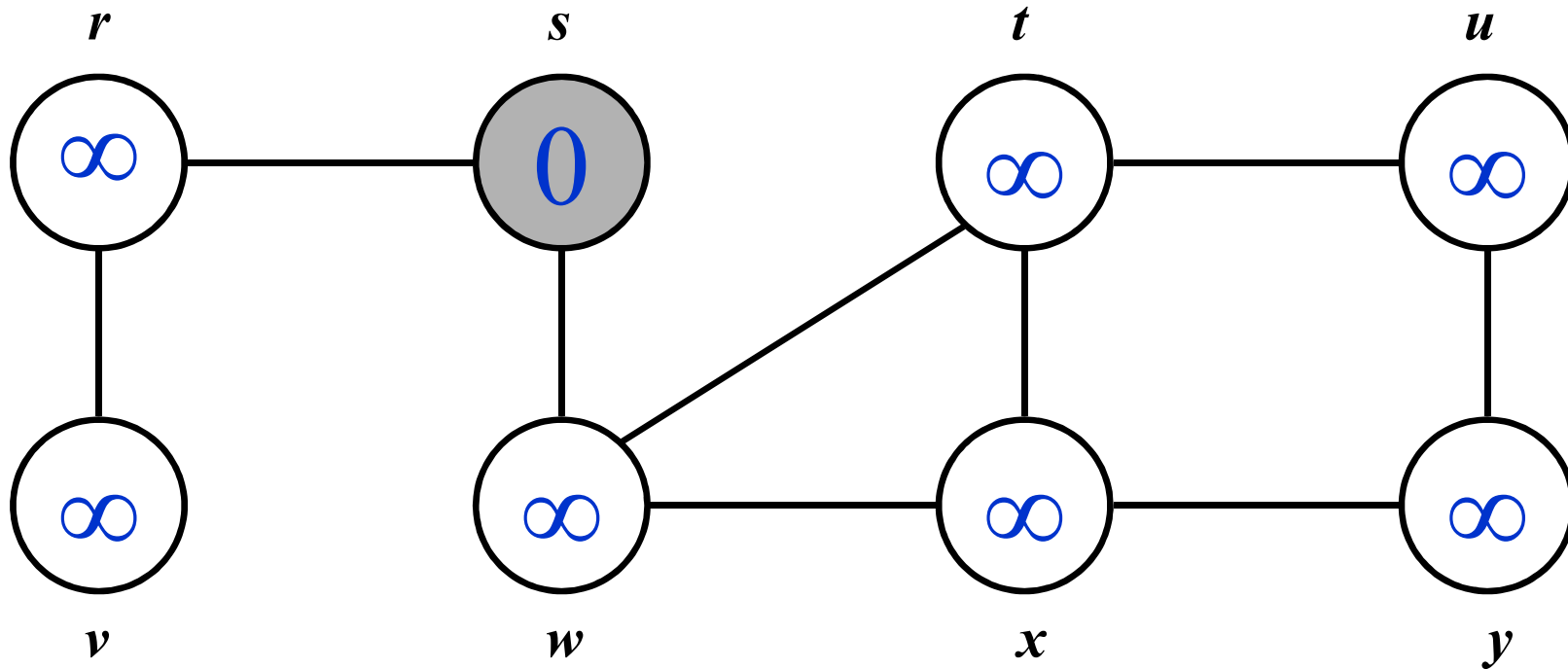
Breadth-First Search

```
BFS(G, s) {  
    initialize vertices;  
    Q = {s};           // Q is a queue (duh); initialize to s  
    while (Q not empty) {  
        u = RemoveTop(Q);  
        for each v ∈ u->adj {  
            if (v->color == WHITE)  
                v->color = GREY;  
                v->d = u->d + 1;  
                v->p = u;           What does v->d represent?  
                Enqueue(Q, v);    What does v->p represent?  
        }  
        u->color = BLACK;  
    }  
}
```

Breadth-First Search: Example

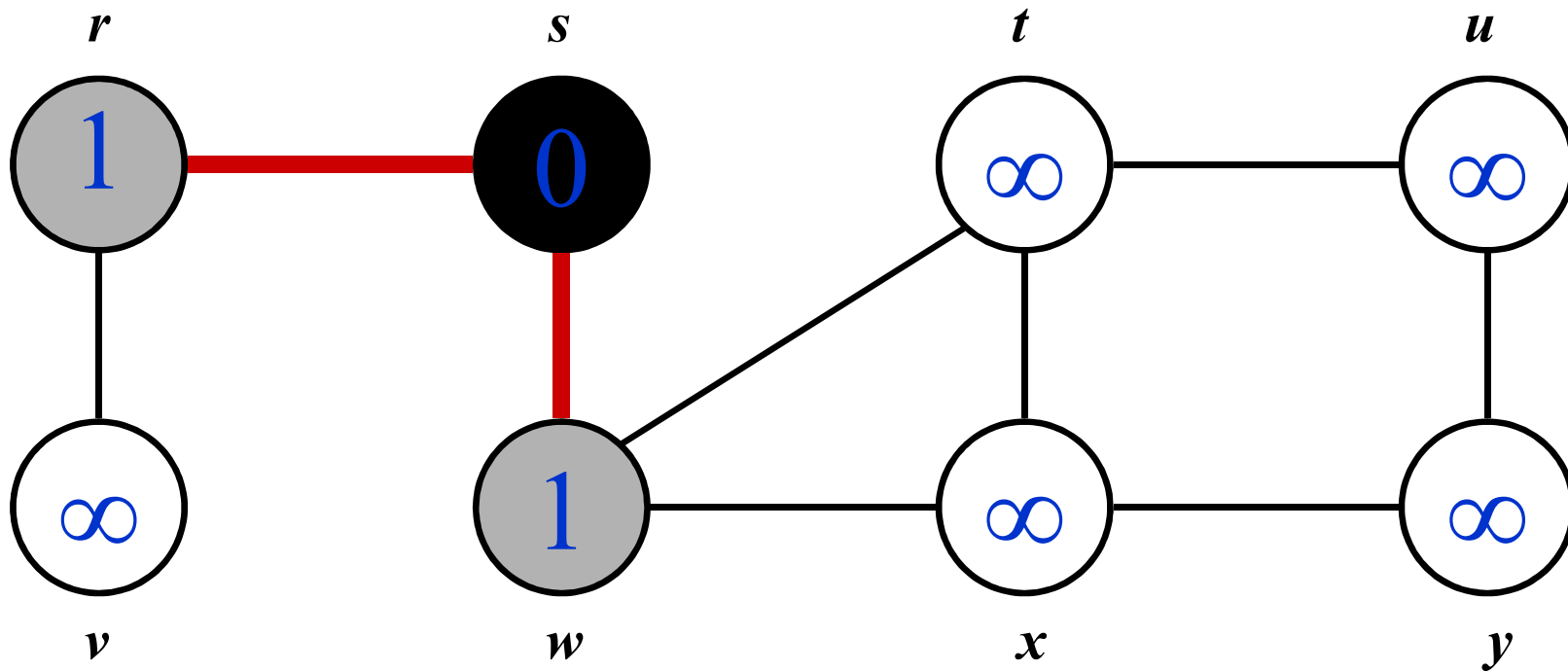


Breadth-First Search: Example



Q : s

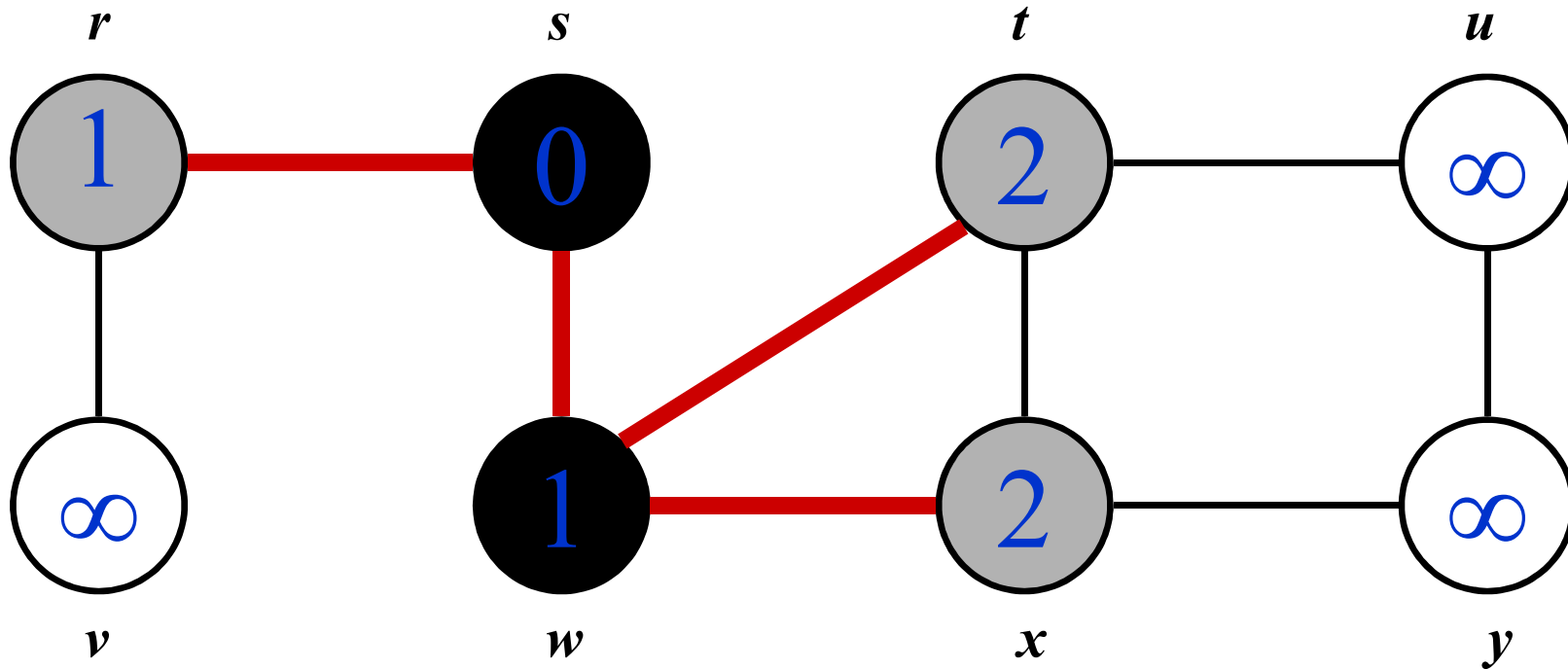
Breadth-First Search: Example



Q :

w	r
-----	-----

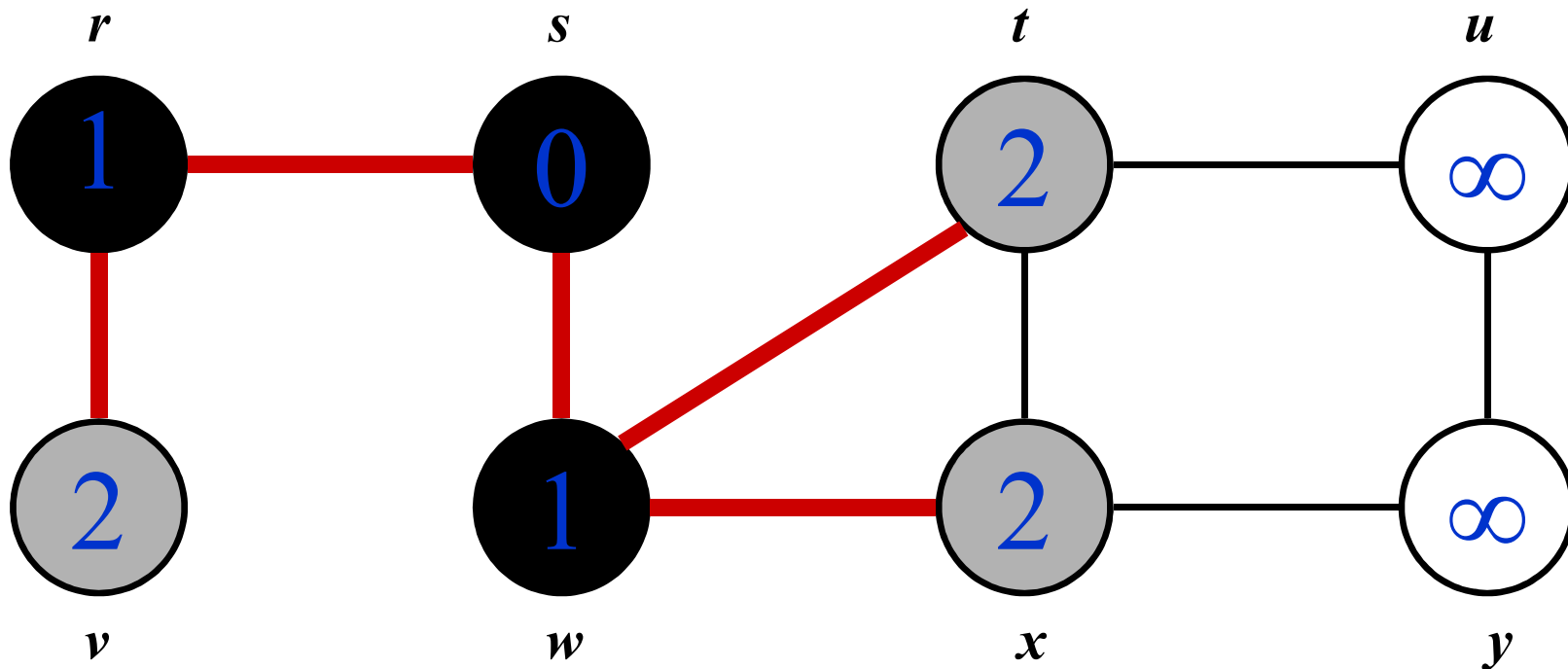
Breadth-First Search: Example



Q :

r	t	x
-----	-----	-----

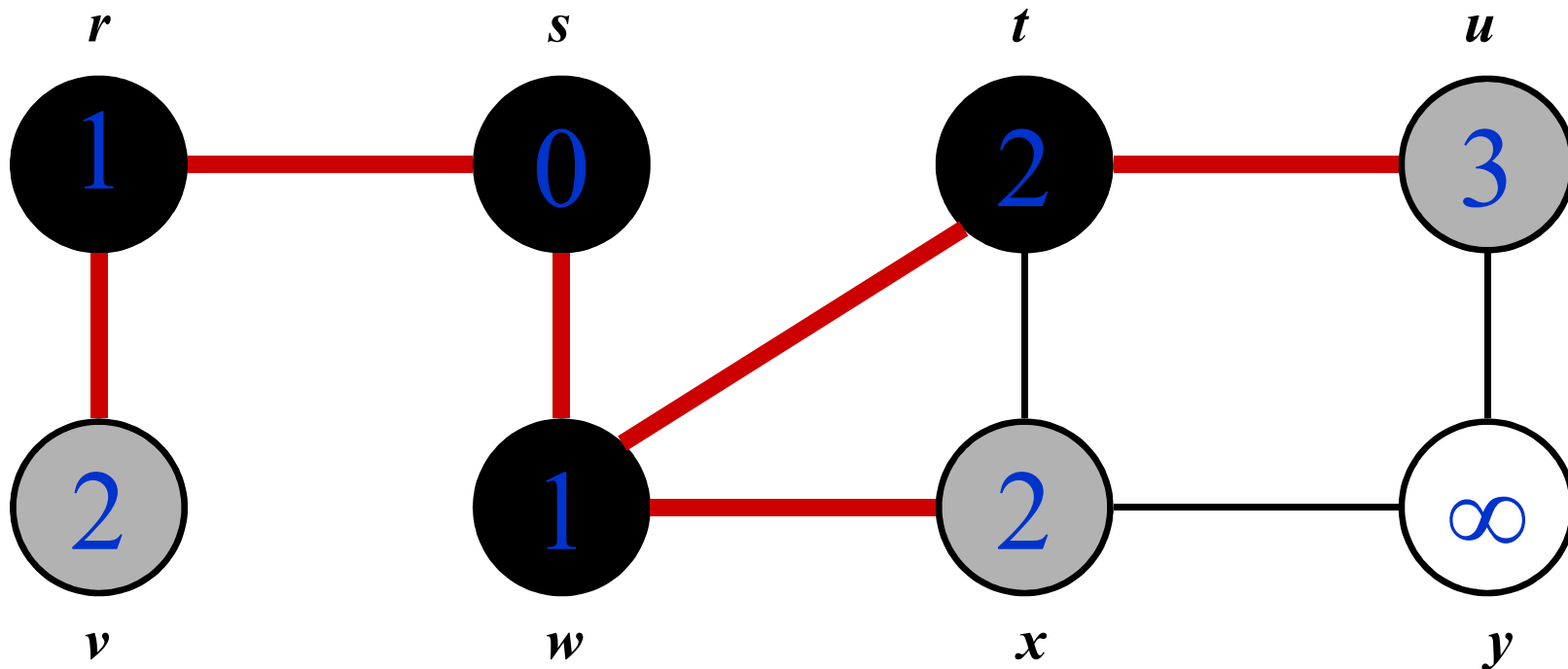
Breadth-First Search: Example



Q :

t	x	v
-----	-----	-----

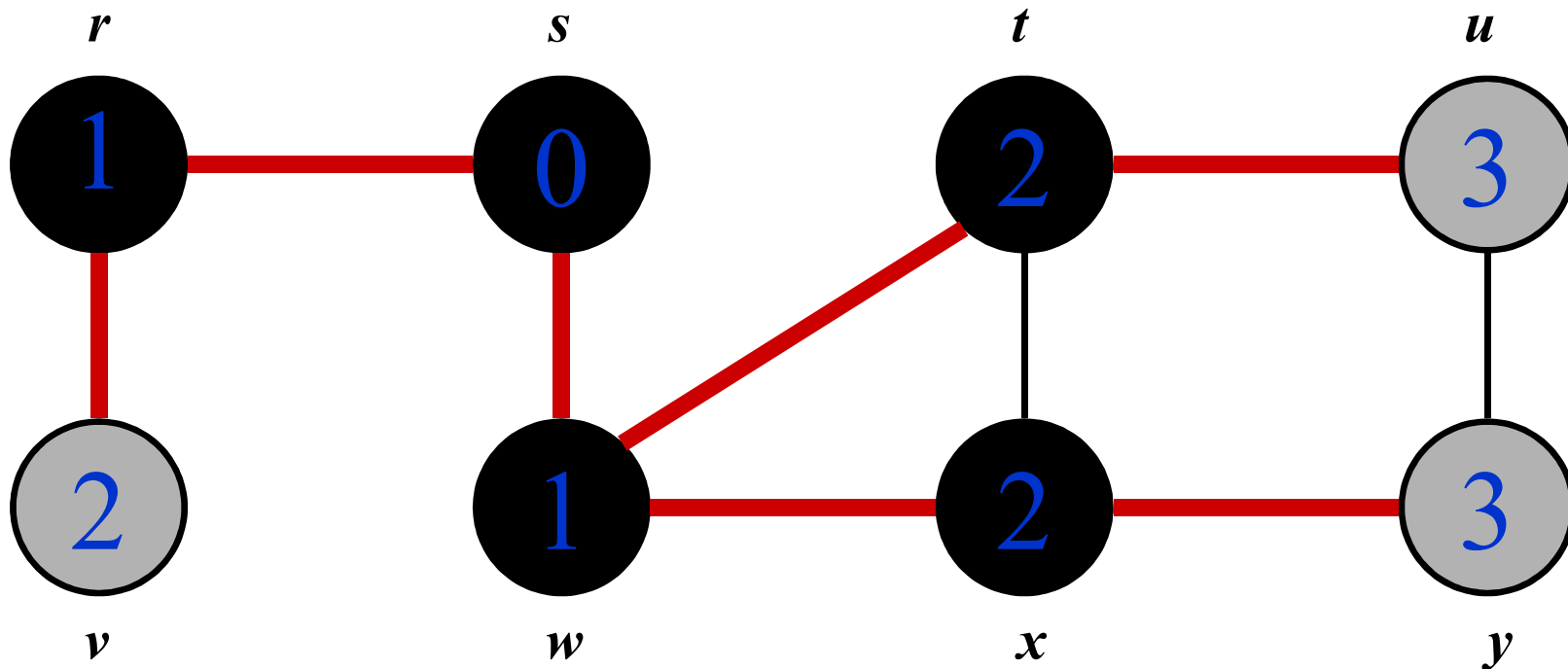
Breadth-First Search: Example



Q :

x	v	u
-----	-----	-----

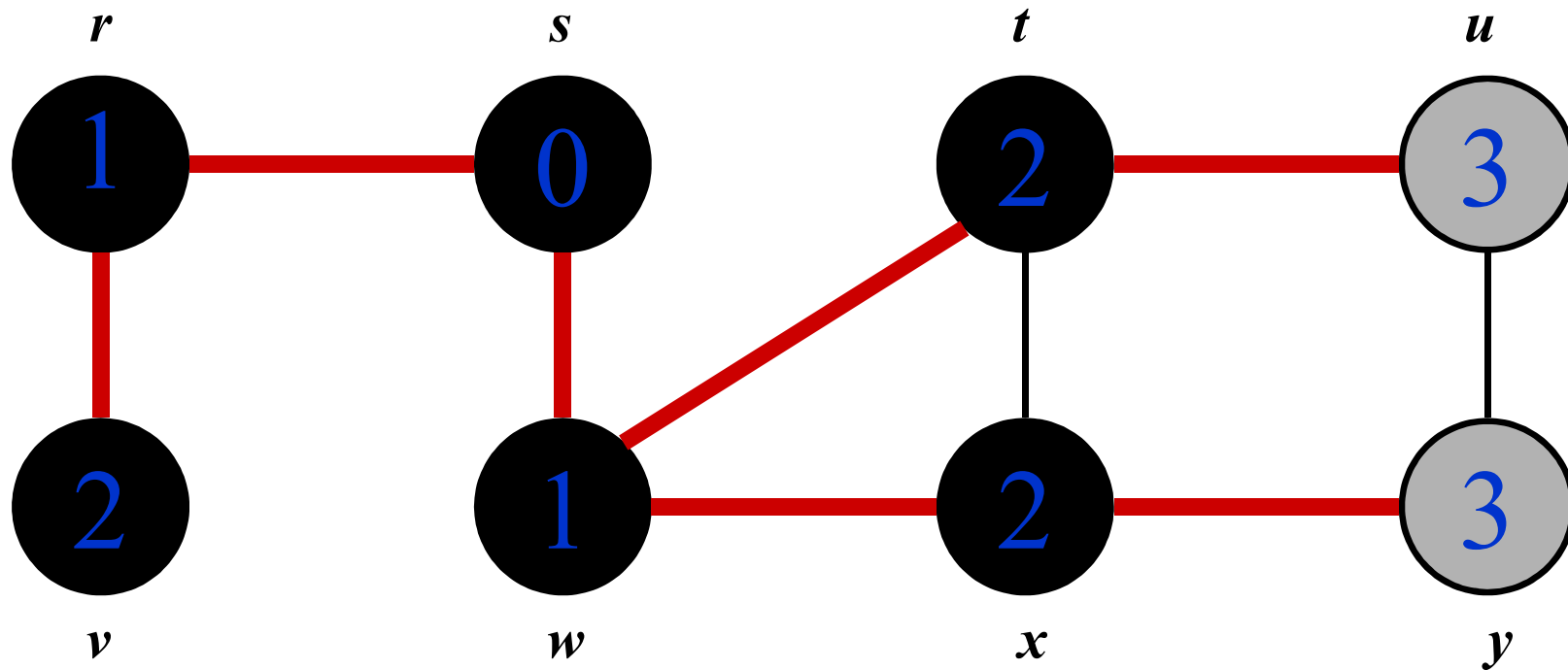
Breadth-First Search: Example



Q :

v	u	y
-----	-----	-----

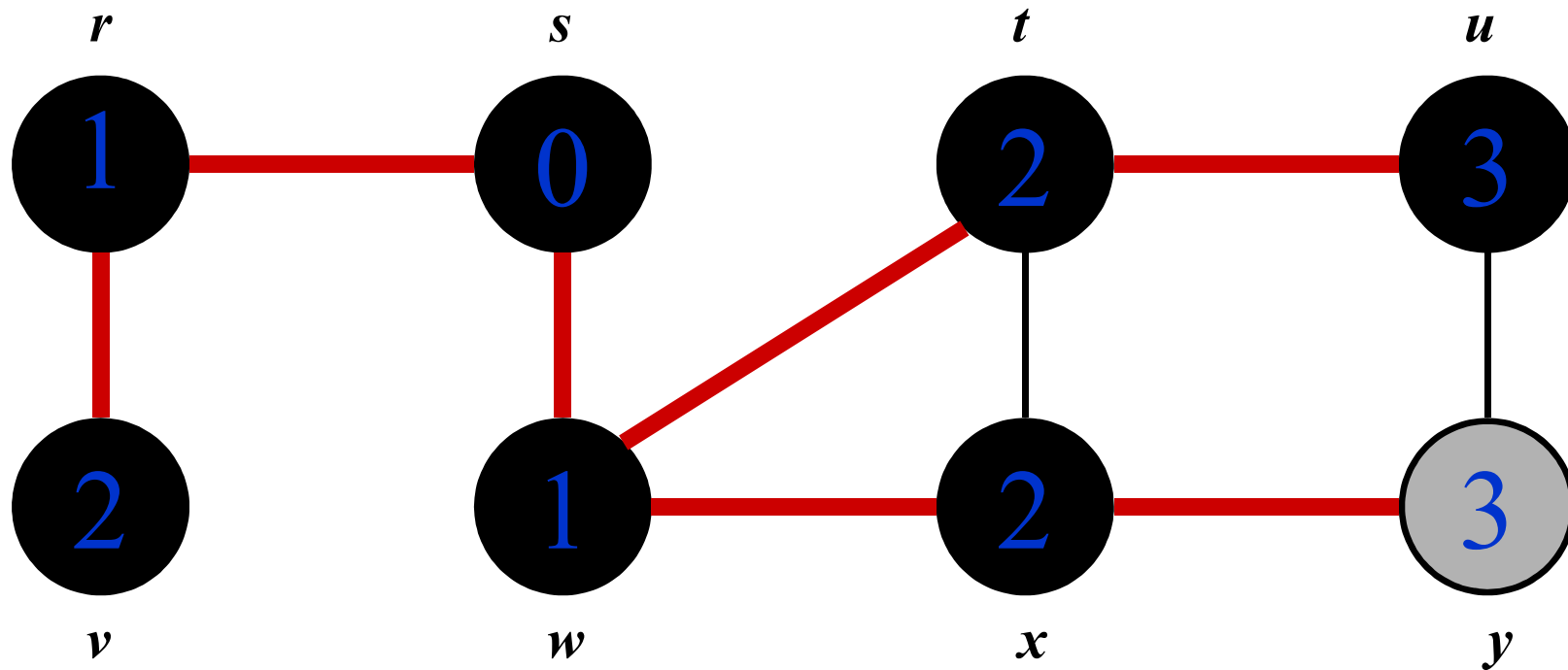
Breadth-First Search: Example



Q :

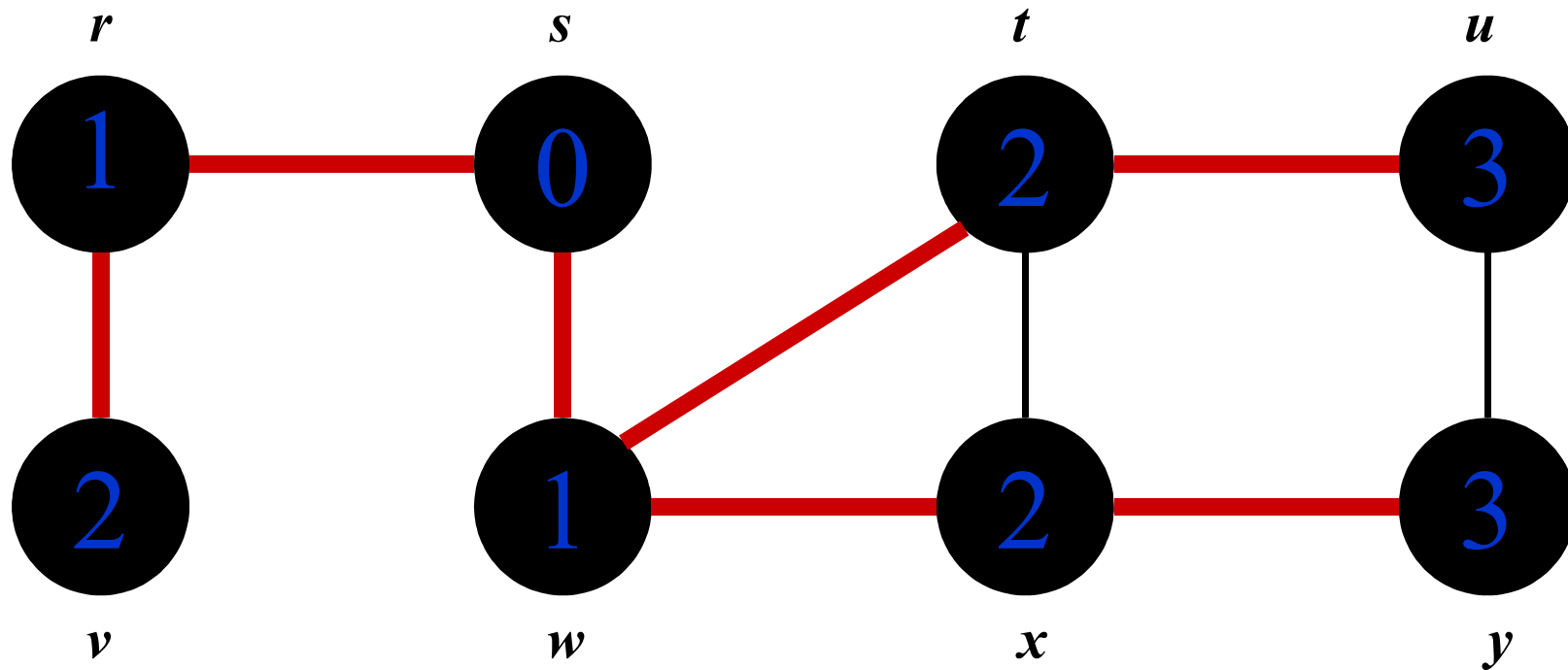
u	y
-----	-----

Breadth-First Search: Example



Q : y

Breadth-First Search: Example




$Q: \emptyset$


BFS: The Code Again

```

BFS (G, s) {
    initialize vertices;
    Q = {s};
    while (Q not empty) {
        u = RemoveTop(Q);
        for each v ∈ u->adj {
            if (v->color == WHITE)
                v->color = GREY;
                v->d = u->d + 1;
                v->p = u;
                Enqueue(Q, v);
            u->color = BLACK;
        }
    }
}

```

 *Touch every vertex: $O(V)$*

 *$u = \text{every vertex, but only once}$*
(Why?)

So $v = \text{every vertex}$ that appears in some other vert's adjacency list

What will be the running time?
Total running time: $O(V+E)$

BFS: The Code Again

```
BFS(G, s) {  
    initialize vertices;  
    Q = {s};  
    while (Q not empty) {  
        u = RemoveTop(Q);  
        for each v ∈ u->adj {  
            if (v->color == WHITE)  
                v->color = GREY;  
                v->d = u->d + 1;  
                v->p = u;  
                Enqueue(Q, v);  
        }  
        u->color = BLACK;  
    }  
}
```

*What will be the storage cost
in addition to storing the tree?*

Total space used:

$O(\max(\text{degree}(v))) = O(E)$

Breadth-First Search: Properties

- BFS calculates the *shortest-path distance* to the source node
 - Shortest-path distance $\delta(s,v)$ = minimum number of edges from s to v , or ∞ if v not reachable from s
 - Proof given in the book (p. 472-5)
- BFS builds *breadth-first tree*, in which paths to root represent shortest paths in G
 - Thus can use BFS to calculate shortest path from one vertex to another in $O(V+E)$ time

DEPTH-FIRST SEARCH

Depth-First Search

- *Depth-first search* is another strategy for exploring a graph
 - Explore “deeper” in the graph whenever possible
 - Edges are explored out of the most recently discovered vertex v that still has unexplored edges
 - When all of v 's edges have been explored, backtrack to the vertex from which v was discovered
- Vertices initially colored white
- Then colored gray when discovered
- Then black when finished

Depth-First Search: The Code

DFS (G)

```
{
    for each vertex u ∈ G->V
    {
        u->color = WHITE;
    }
    time = 0;
    for each vertex u ∈ G->V
    {
        if (u->color == WHITE)
            DFS_Visit(u);
    }
}
```

DFS_Visit(u)

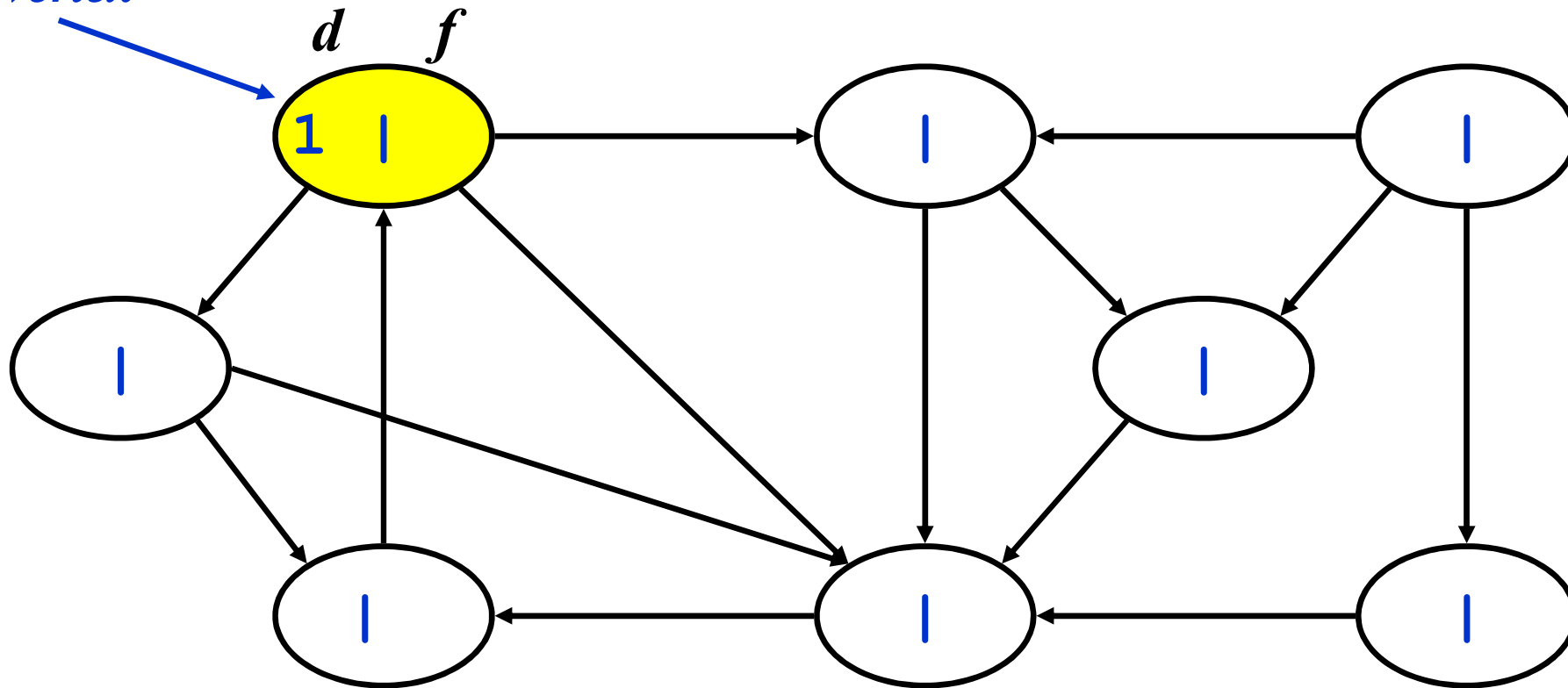
```
{
    u->color = GREY;
    time = time+1;
    u->d = time;
    for each v ∈ u->Adj[]
    {
        if (v->color == WHITE)
            DFS_Visit(v);
    }
    u->color = BLACK;
    time = time+1;
    u->f = time;
}
```

What does u->d represent?

What does u->f represent?

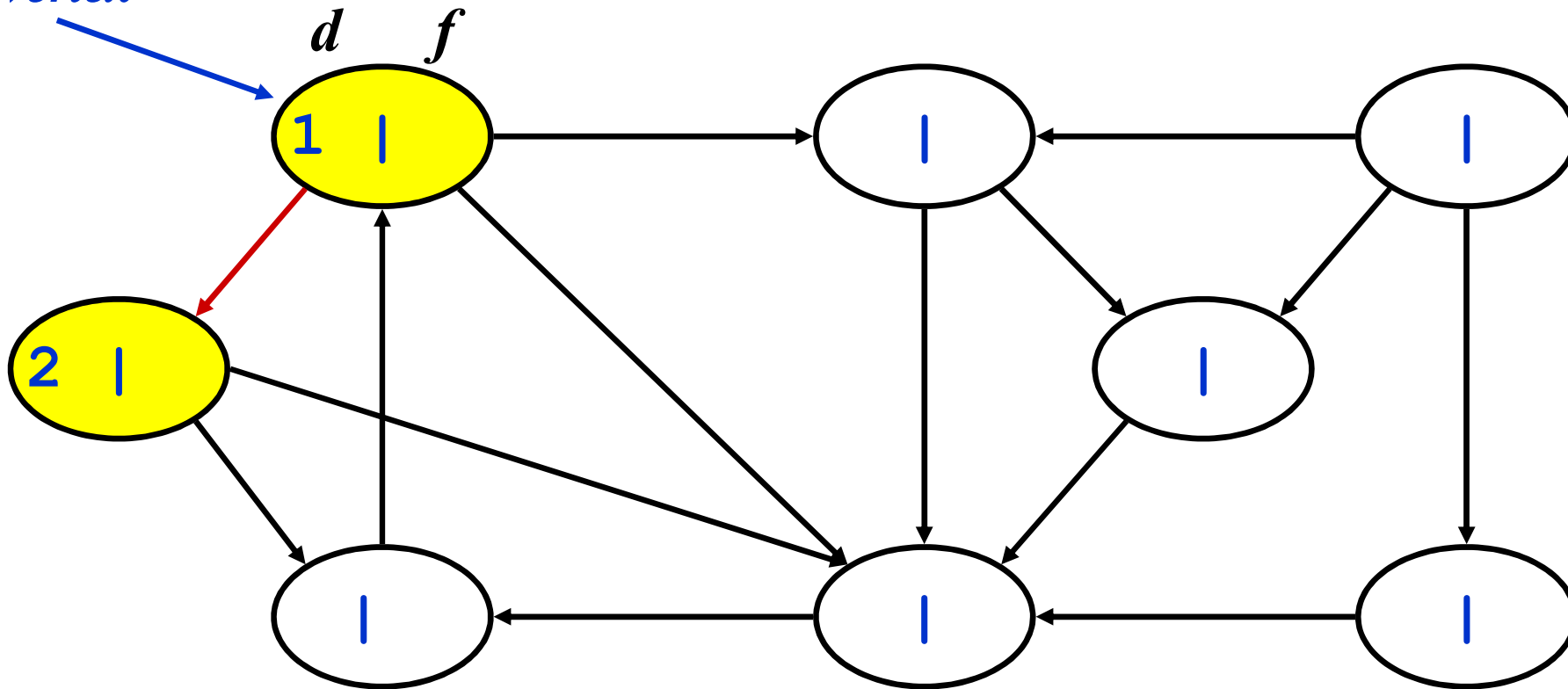
DFS Example

*source
vertex*



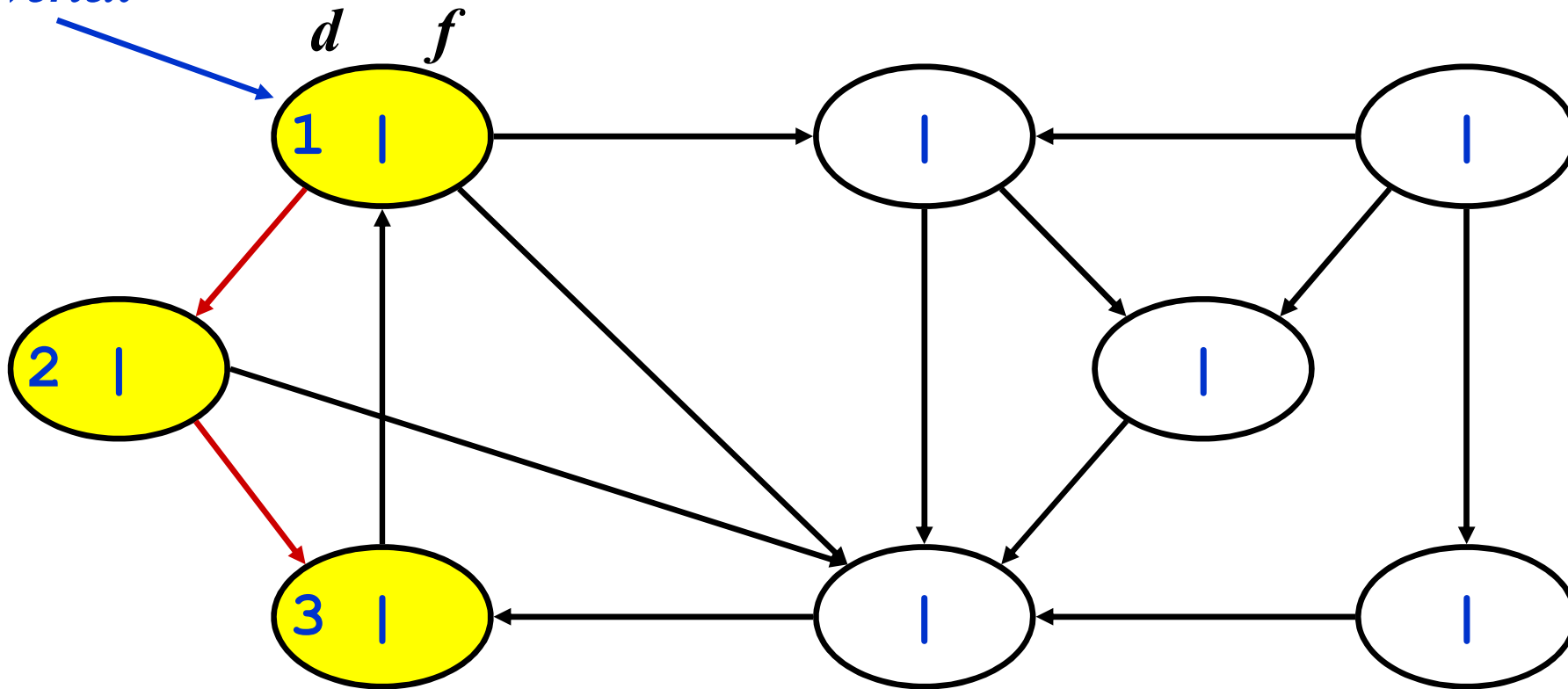
DFS Example

*source
vertex*



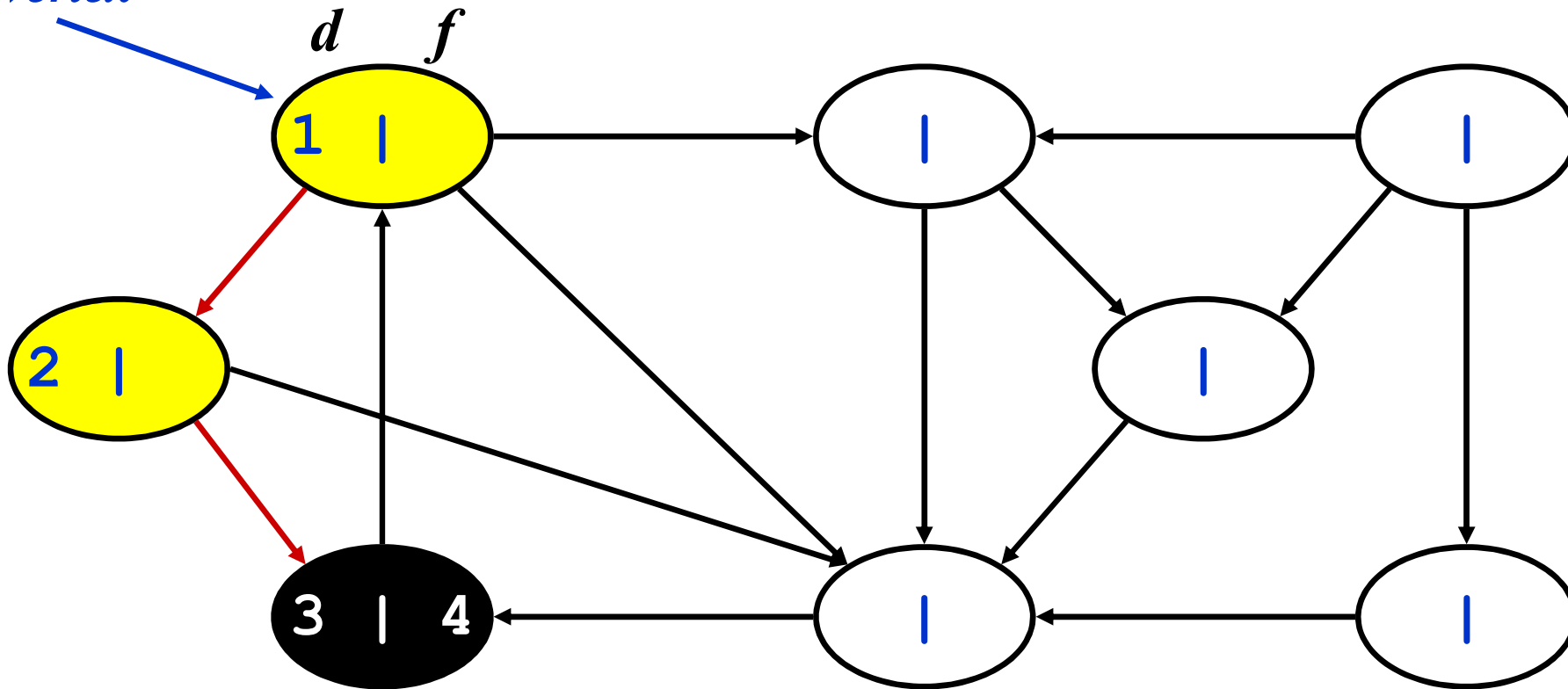
DFS Example

*source
vertex*



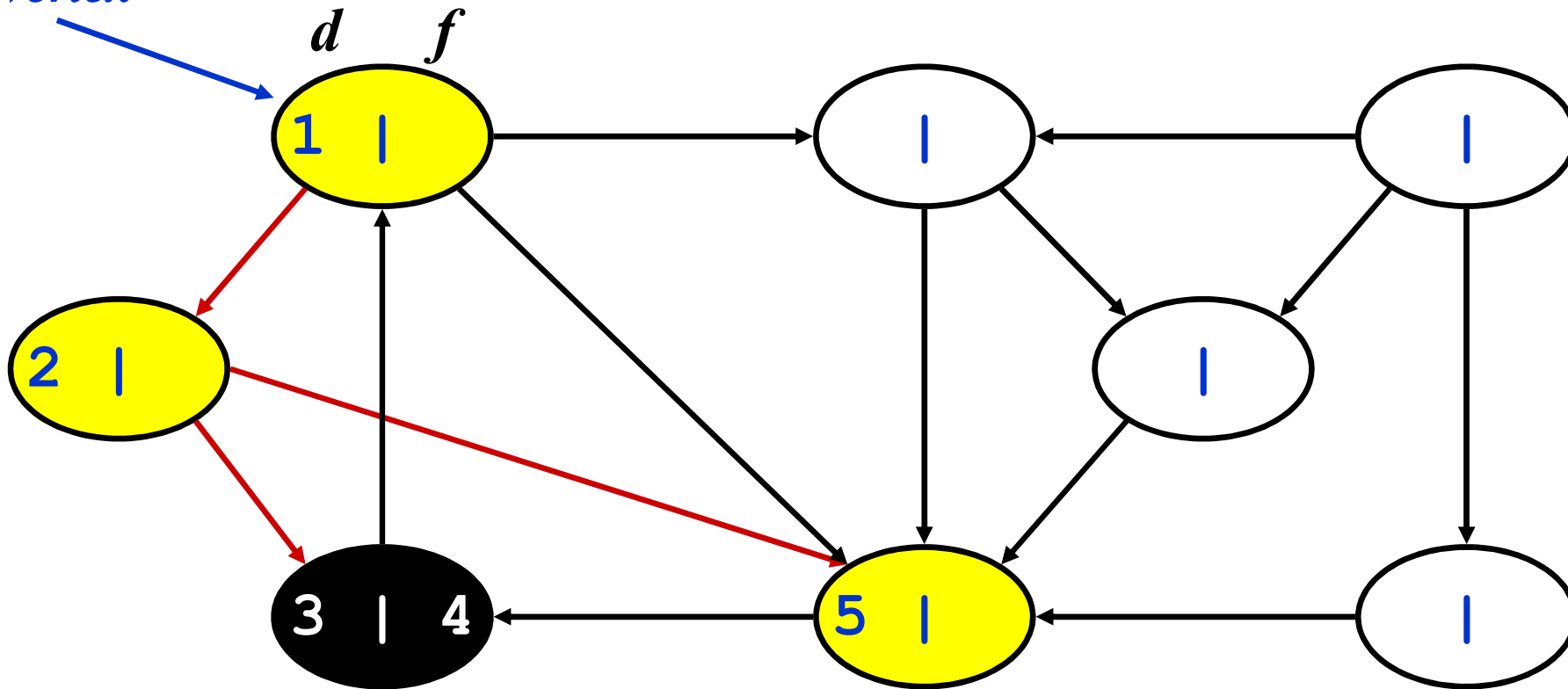
DFS Example

*source
vertex*



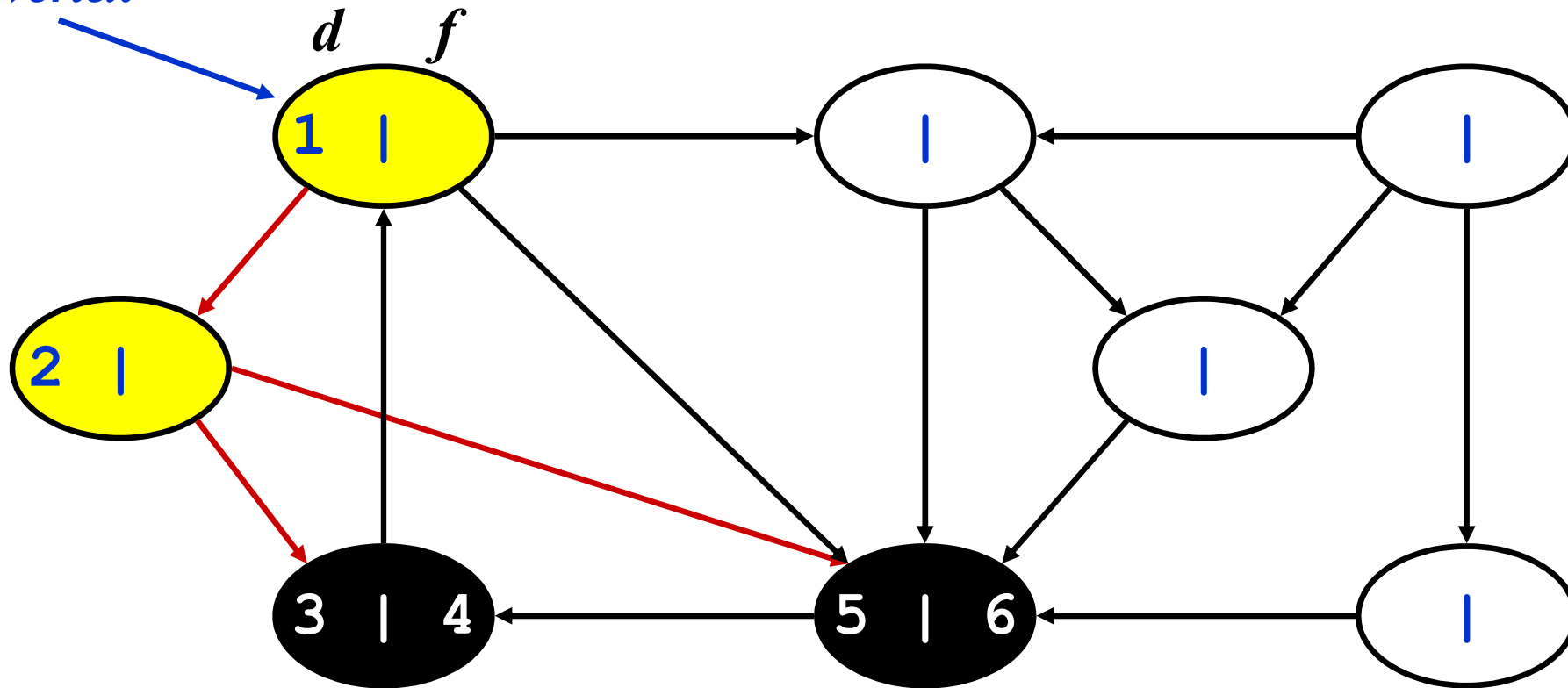
DFS Example

*source
vertex*



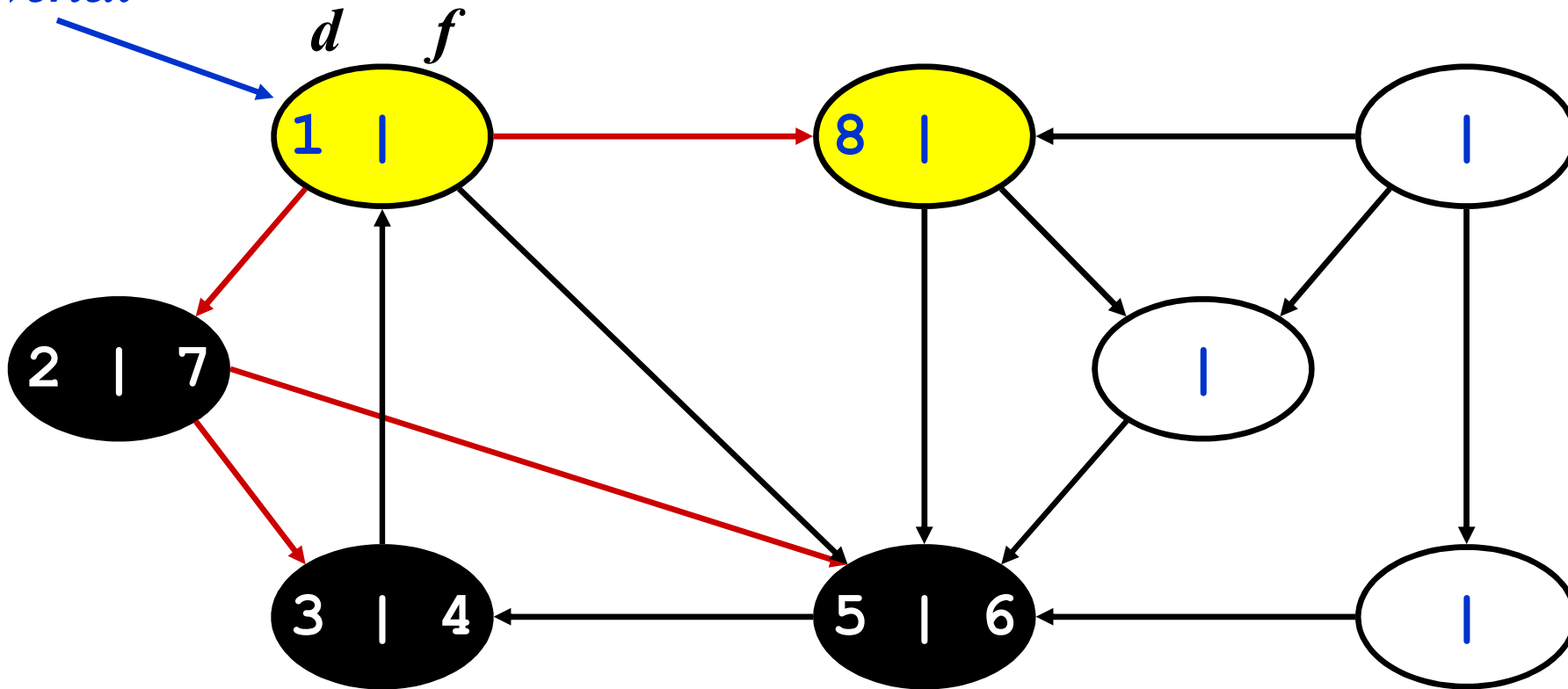
DFS Example

*source
vertex*



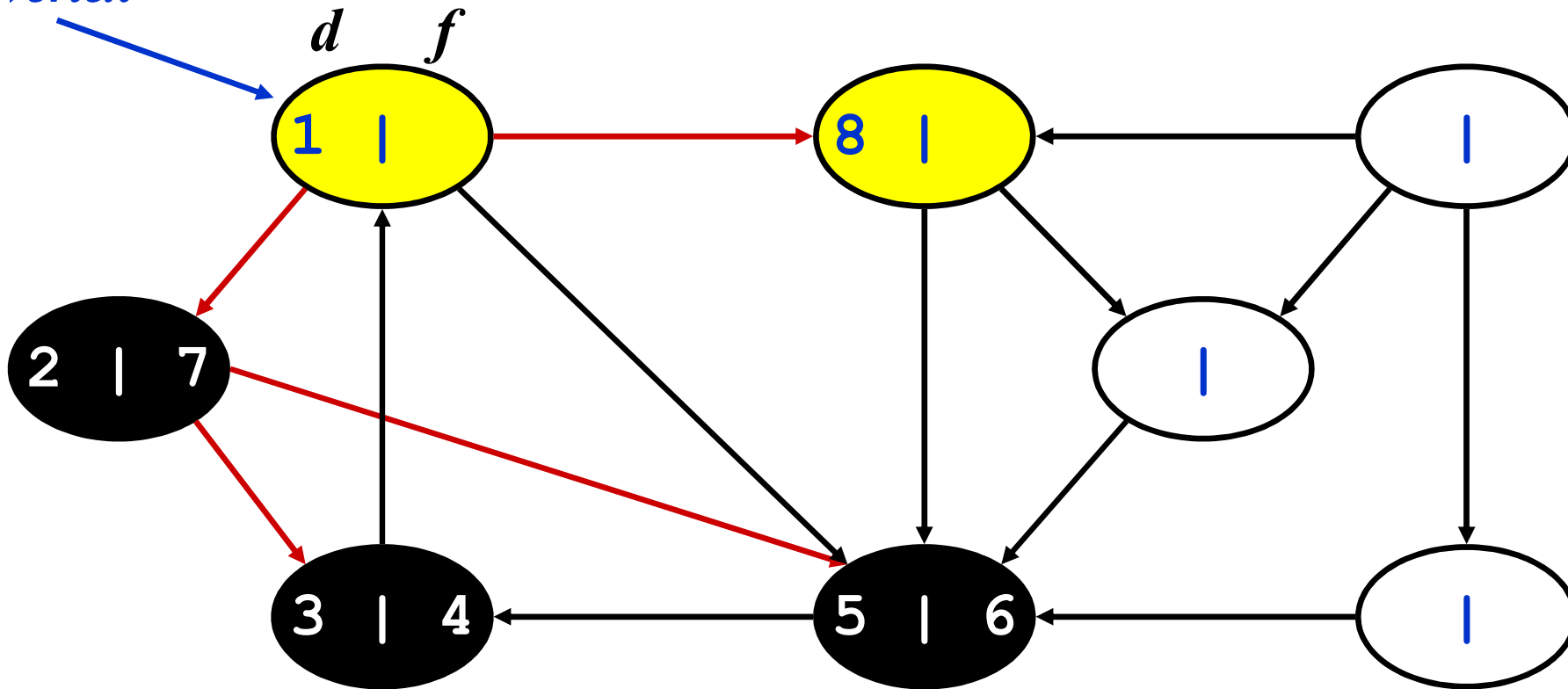
DFS Example

*source
vertex*



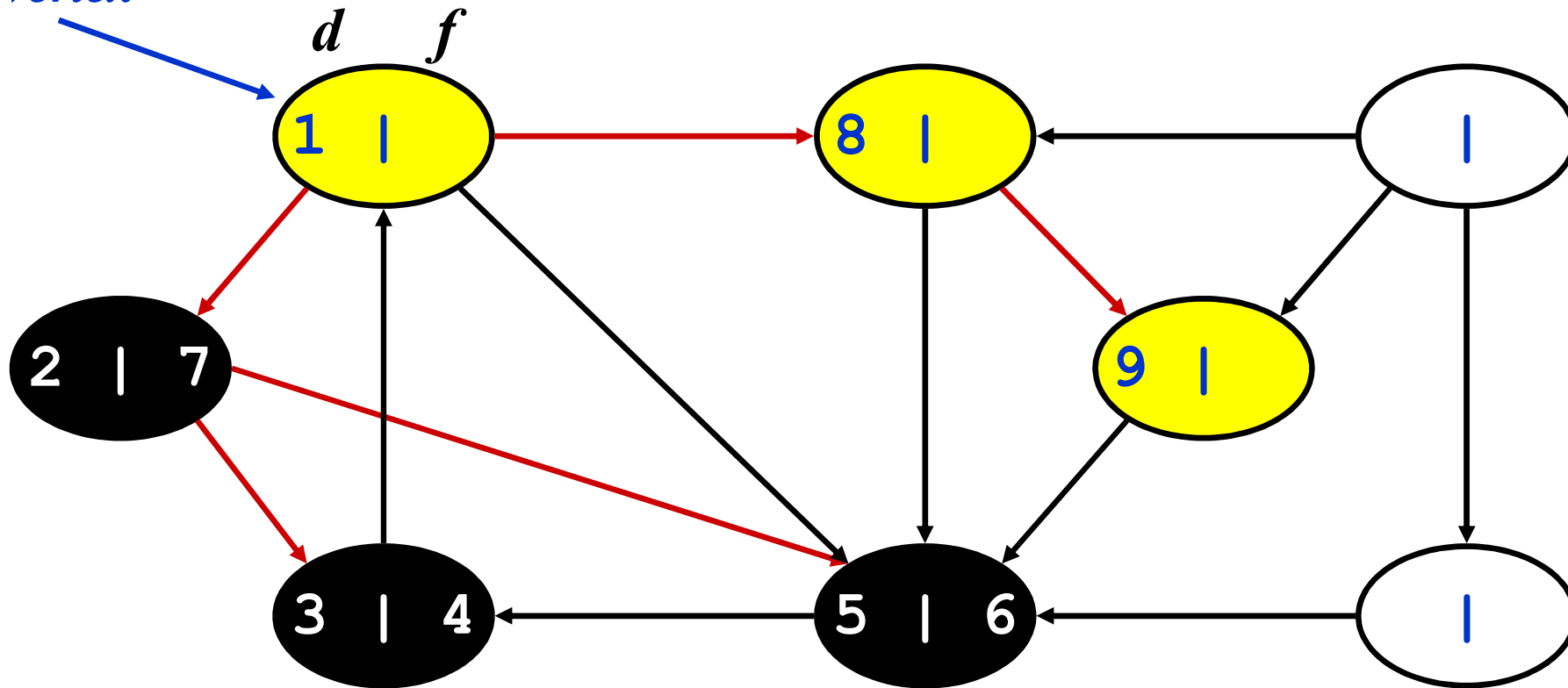
DFS Example

*source
vertex*



DFS Example

*source
vertex*

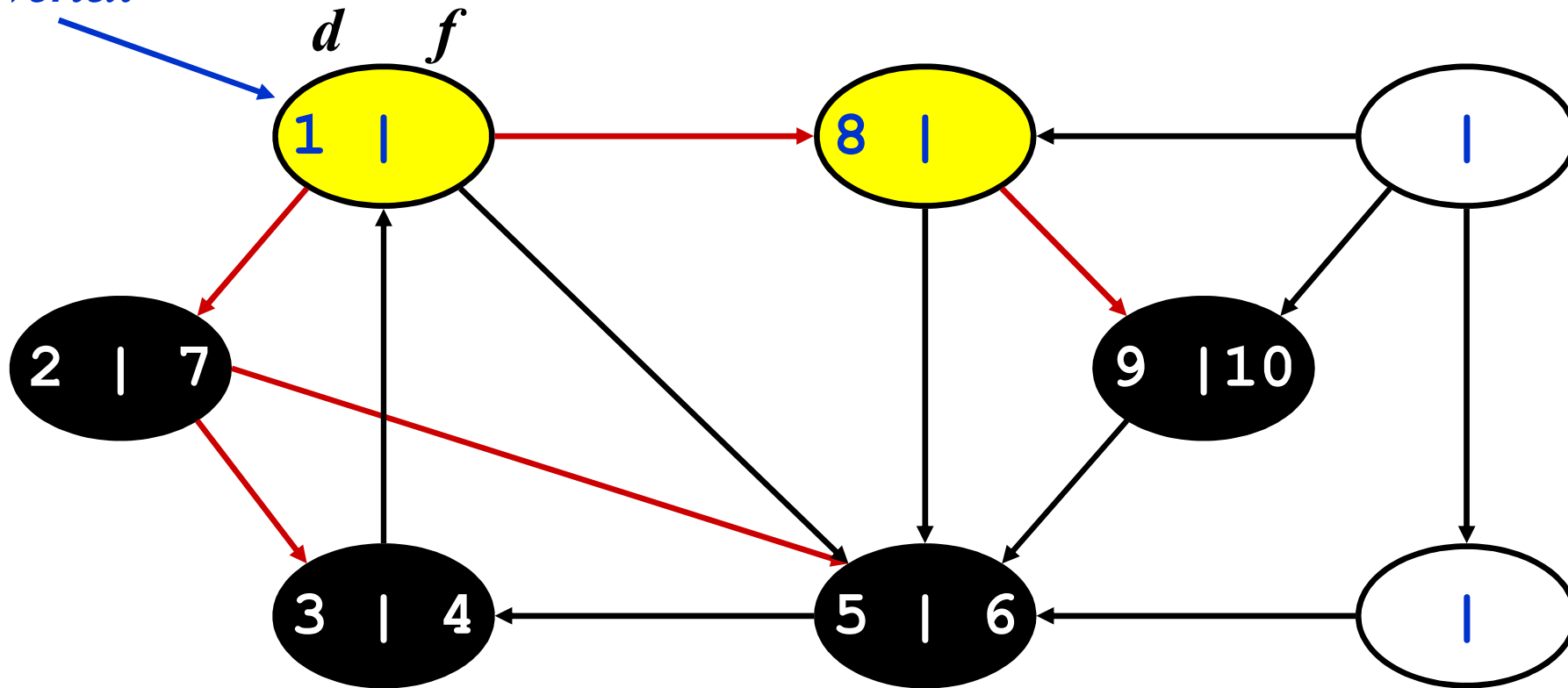


What is the structure of the yellow vertices?

What do they represent?

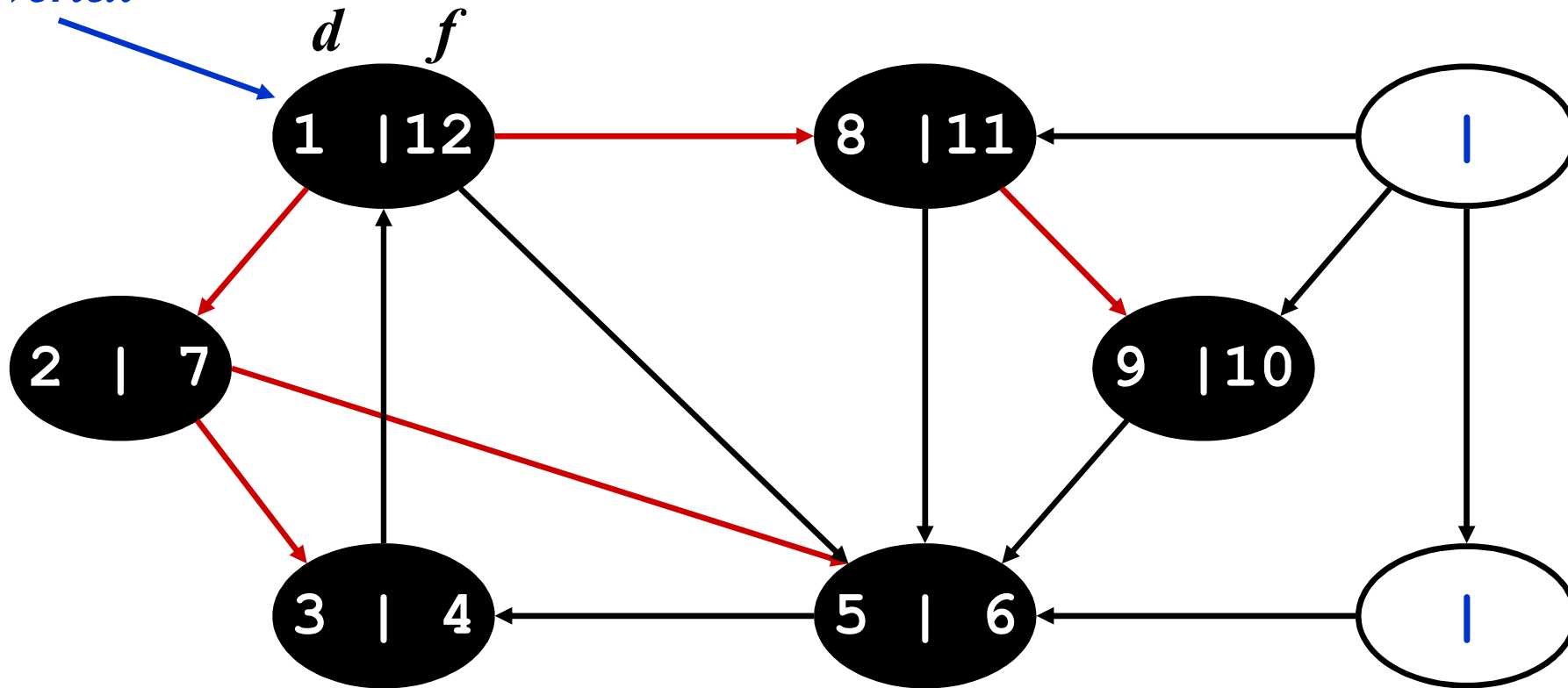
DFS Example

*source
vertex*

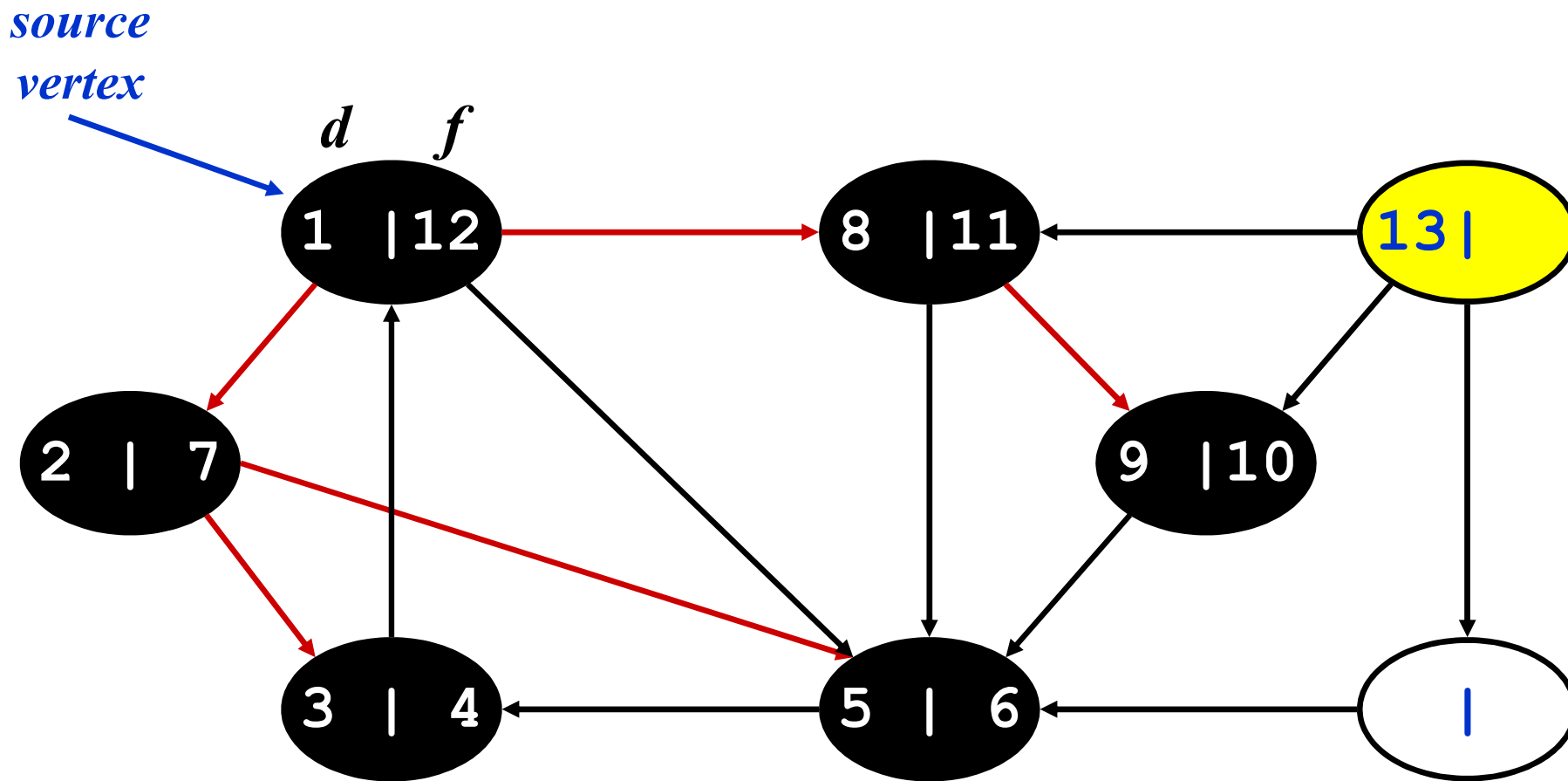


DFS Example

*source
vertex*

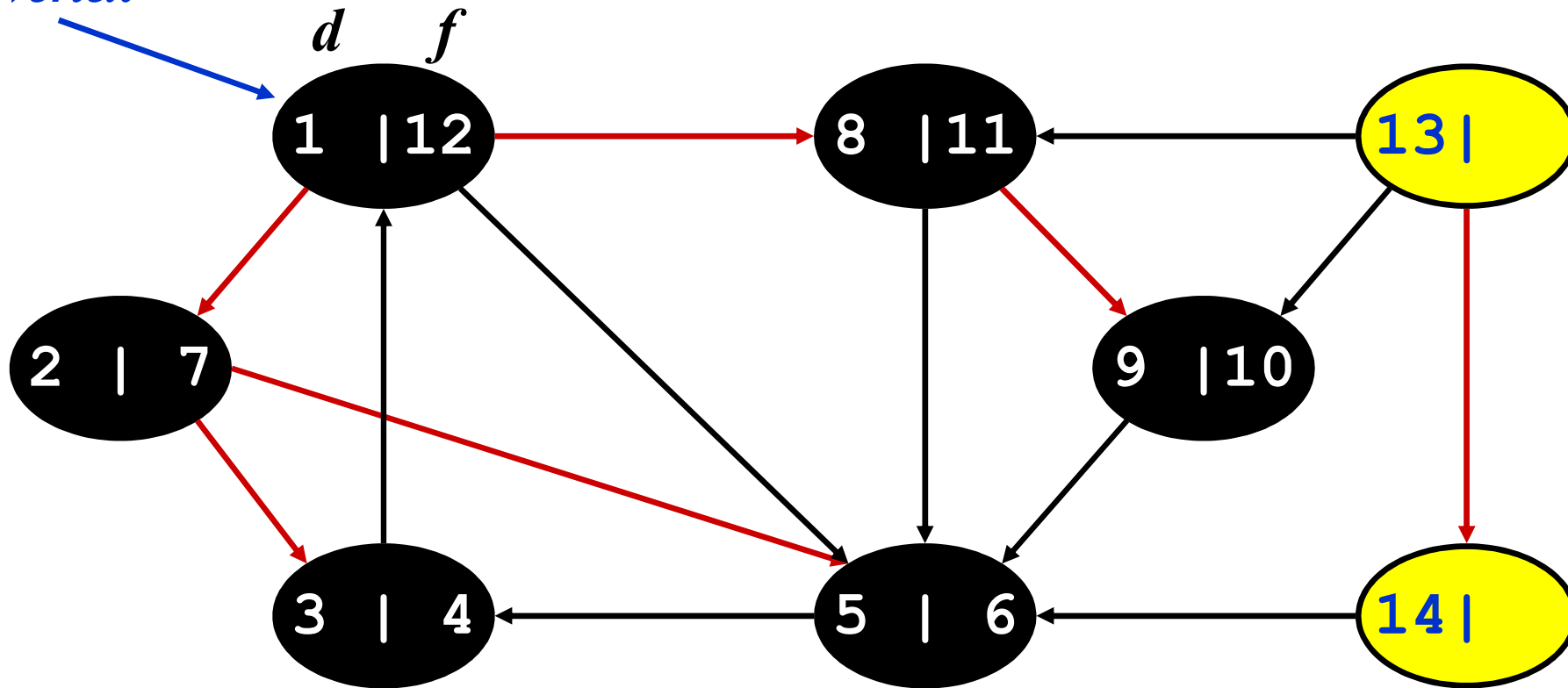


DFS Example



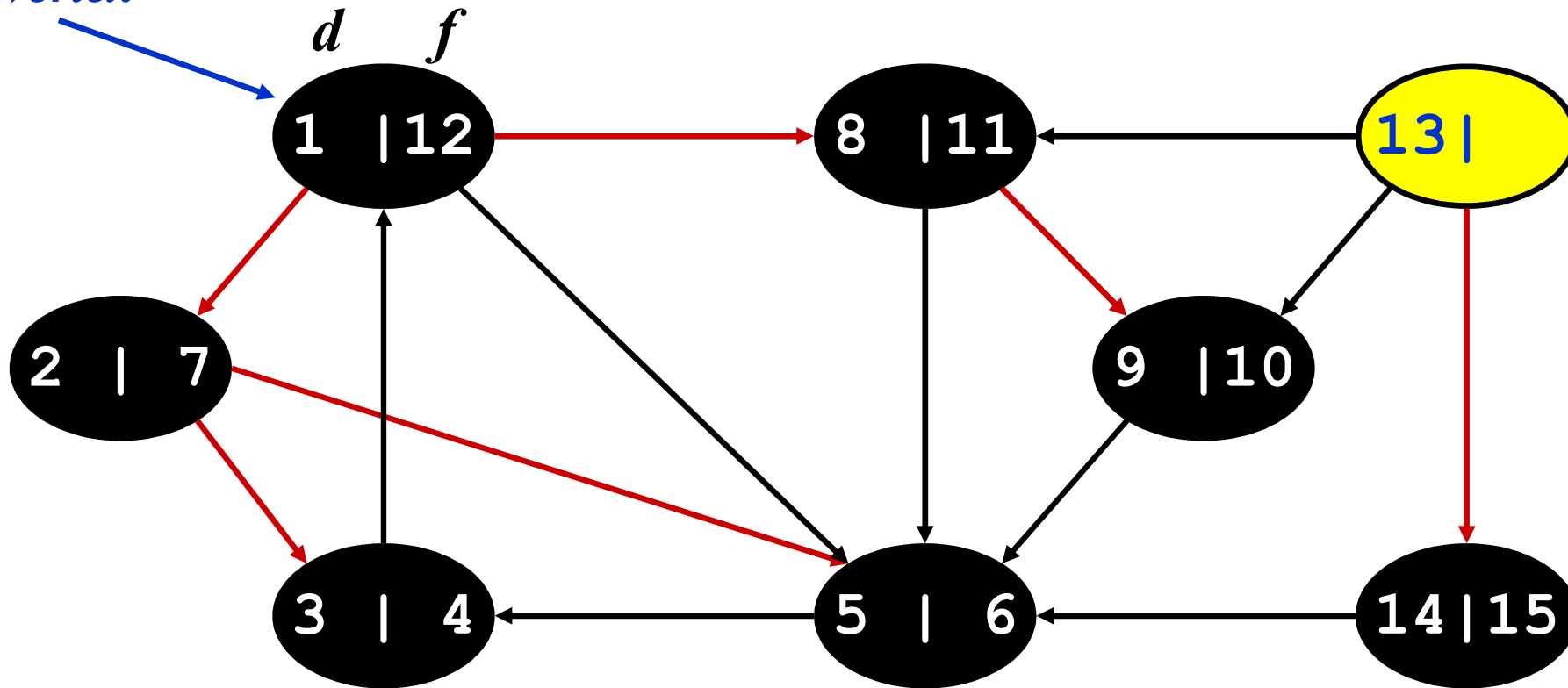
DFS Example

*source
vertex*



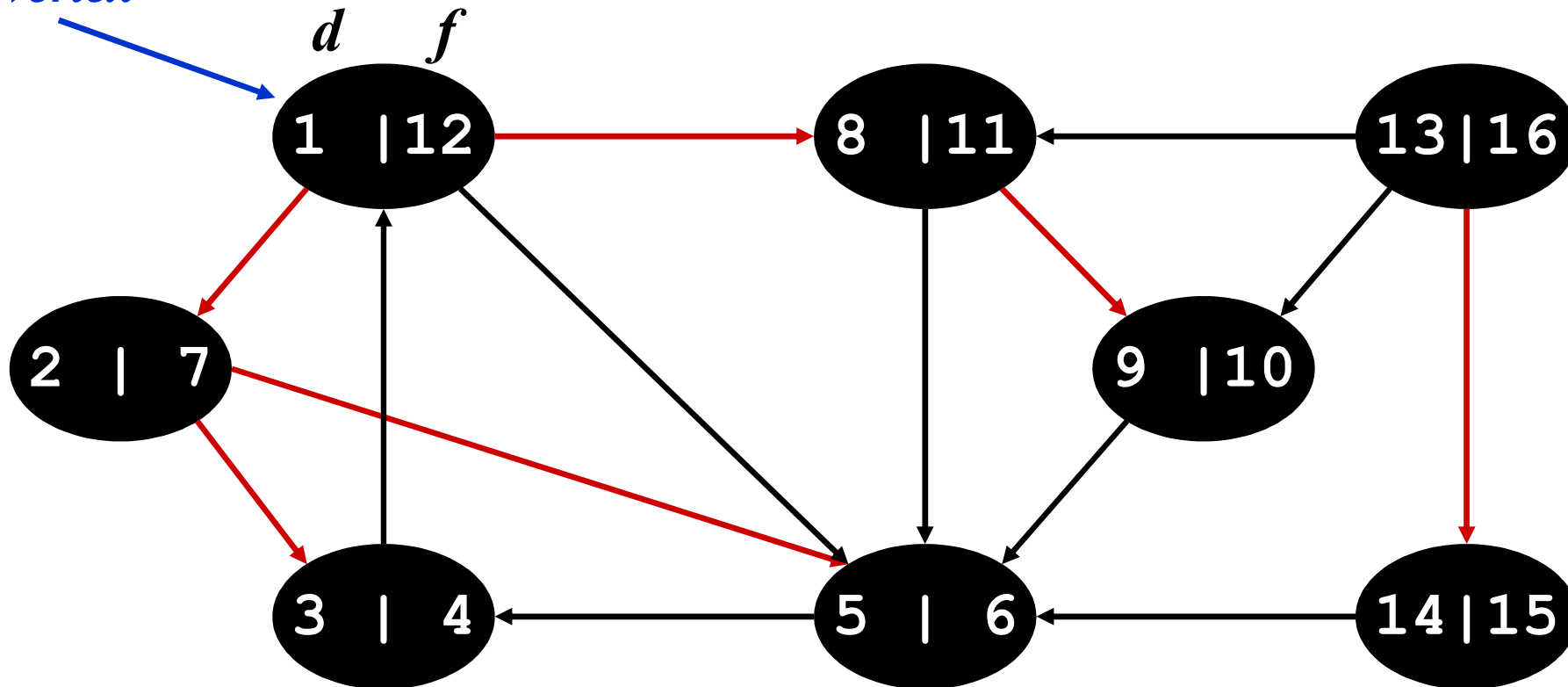
DFS Example

*source
vertex*



DFS Example

*source
vertex*



Depth-First Search: The Code

DFS (G)

```
{
    for each vertex  $u \in G \rightarrow V$ 
    {
         $u \rightarrow \text{color} = \text{WHITE};$ 
    }
    time = 0;
    for each vertex  $u \in G \rightarrow V$ 
    {
        if ( $u \rightarrow \text{color} == \text{WHITE}$ )
            DFS_Visit(u);
    }
}
```

DFS_Visit(u)

```
{
     $u \rightarrow \text{color} = \text{GREY};$ 
    time = time+1;
     $u \rightarrow d = \text{time};$ 
    for each  $v \in u \rightarrow \text{Adj}[]$ 
    {
        if ( $v \rightarrow \text{color} == \text{WHITE}$ )
            DFS_Visit(v);
    }
     $u \rightarrow \text{color} = \text{BLACK};$ 
    time = time+1;
     $u \rightarrow f = \text{time};$ 
}
```

Will all vertices eventually be colored black?

What will be the running time?

Depth-First Search: The Code

DFS (G)

```
{
    for each vertex u ∈ G->V
    {
        u->color = WHITE;
    }
    time = 0;
    for each vertex u ∈ G->V
    {
        if (u->color == WHITE)
            DFS_Visit(u);
    }
}
```

DFS_Visit(u)

```
{
    u->color = GREY;
    time = time+1;
    u->d = time;
    for each v ∈ u->Adj[]
    {
        if (v->color == WHITE)
            DFS_Visit(v);
    }
    u->color = BLACK;
    time = time+1;
    u->f = time;
}
```

*Running time: $O(n^2)$ because call DFS_Visit on each vertex,
and the loop over Adj[] can run as many as $|V|$ times*

Slide credit: 홍석원, 명지대학교; 김한준, 서울시립대학교; J. Lillis, UIC

Depth-First Search: The Code

```

DFS (G)
{
    for each vertex  $u \in G \rightarrow V$ 
    {
         $u \rightarrow \text{color} = \text{WHITE};$ 
    }
    time = 0;
    for each vertex  $u \in G \rightarrow V$ 
    {
        if ( $u \rightarrow \text{color} == \text{WHITE}$ )
            DFS_Visit( $u$ );
    }
}

```

```

DFS_Visit( $u$ )
{
     $u \rightarrow \text{color} = \text{GREY};$ 
    time = time+1;
     $u \rightarrow d = \text{time};$ 
    for each  $v \in u \rightarrow \text{Adj}[]$ 
    {
        if ( $v \rightarrow \text{color} == \text{WHITE}$ )
            DFS_Visit( $v$ );
    }
     $u \rightarrow \text{color} = \text{BLACK};$ 
    time = time+1;
     $u \rightarrow f = \text{time};$ 
}

```

BUT, there is actually a tighter bound.

How many times will DFS_Visit() actually be called?

Slide credit: 홍석원, 명지대학교; 김한준, 서울시립대학교; J. Lillis, UIC

Depth-First Search: The Code

DFS (G)

```
{
    for each vertex u ∈ G->V
    {
        u->color = WHITE;
    }
    time = 0;
    for each vertex u ∈ G->V
    {
        if (u->color == WHITE)
            DFS_Visit(u);
    }
}
```

DFS_Visit(u)

```
{
    u->color = GREY;
    time = time+1;
    u->d = time;
    for each v ∈ u->Adj[]
    {
        if (v->color == WHITE)
            DFS_Visit(v);
    }
    u->color = BLACK;
    time = time+1;
    u->f = time;
}
```

So, running time of DFS = $O(V+E)$

Slide credit: 홍석원, 명지대학교; 김한준, 서울시립대학교; J. Lillis, UIC

Depth-First Sort Analysis

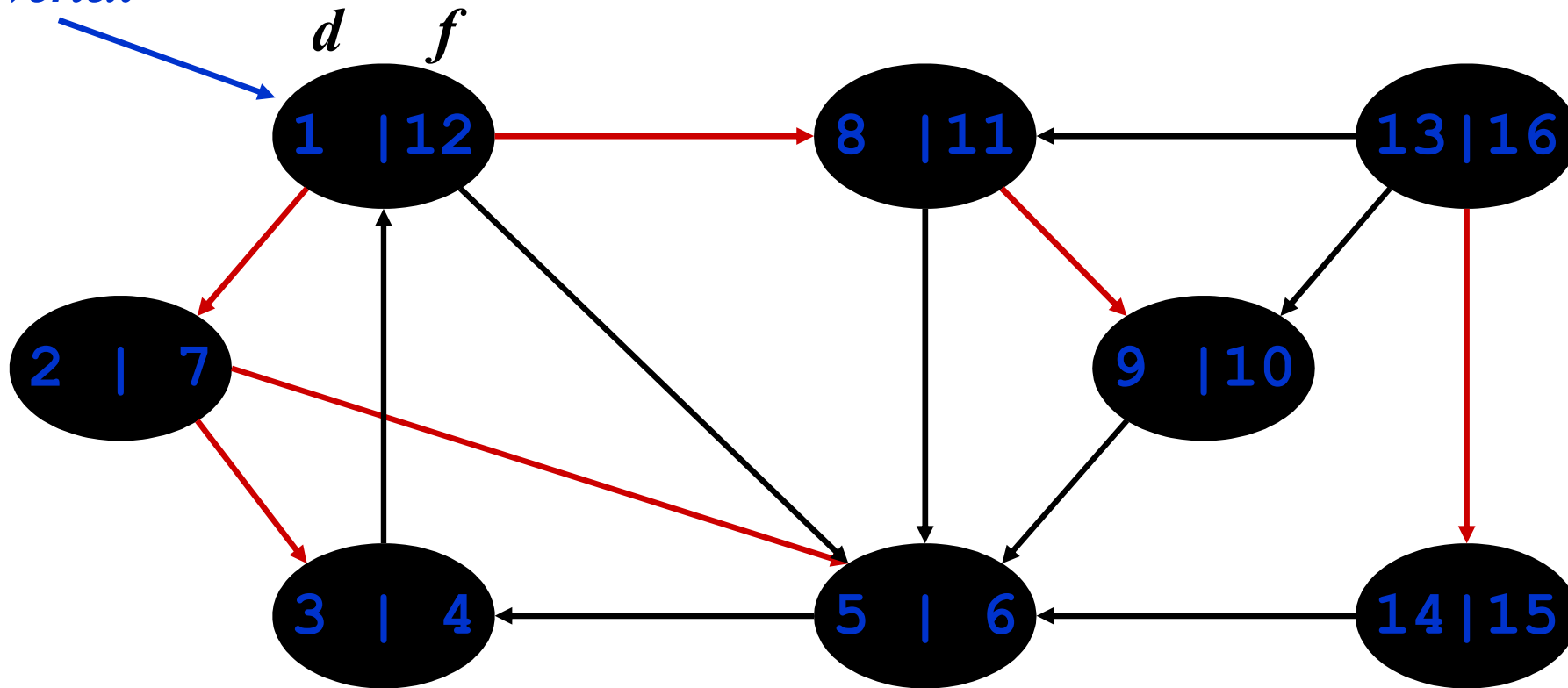
- This running time argument is an informal example of *amortized analysis*
 - “Charge” the exploration of edge to the edge:
 - Each loop in DFS_Visit can be attributed to an edge in the graph
 - Runs once/edge if directed graph, twice if undirected
 - Thus loop will run in $O(E)$ time, algorithm $O(V+E)$
 - ✳ Considered linear for graph, b/c adj list requires $O(V+E)$ storage
 - Important to be comfortable with this kind of reasoning and analysis

DFS: Kinds of edges

- DFS introduces an important distinction among edges in the original graph:
 - *Tree edge*: encounter new (white) vertex
 - The tree edges form a spanning forest
 - *Can tree edges form cycles? Why or why not?*

DFS Example

*source
vertex*



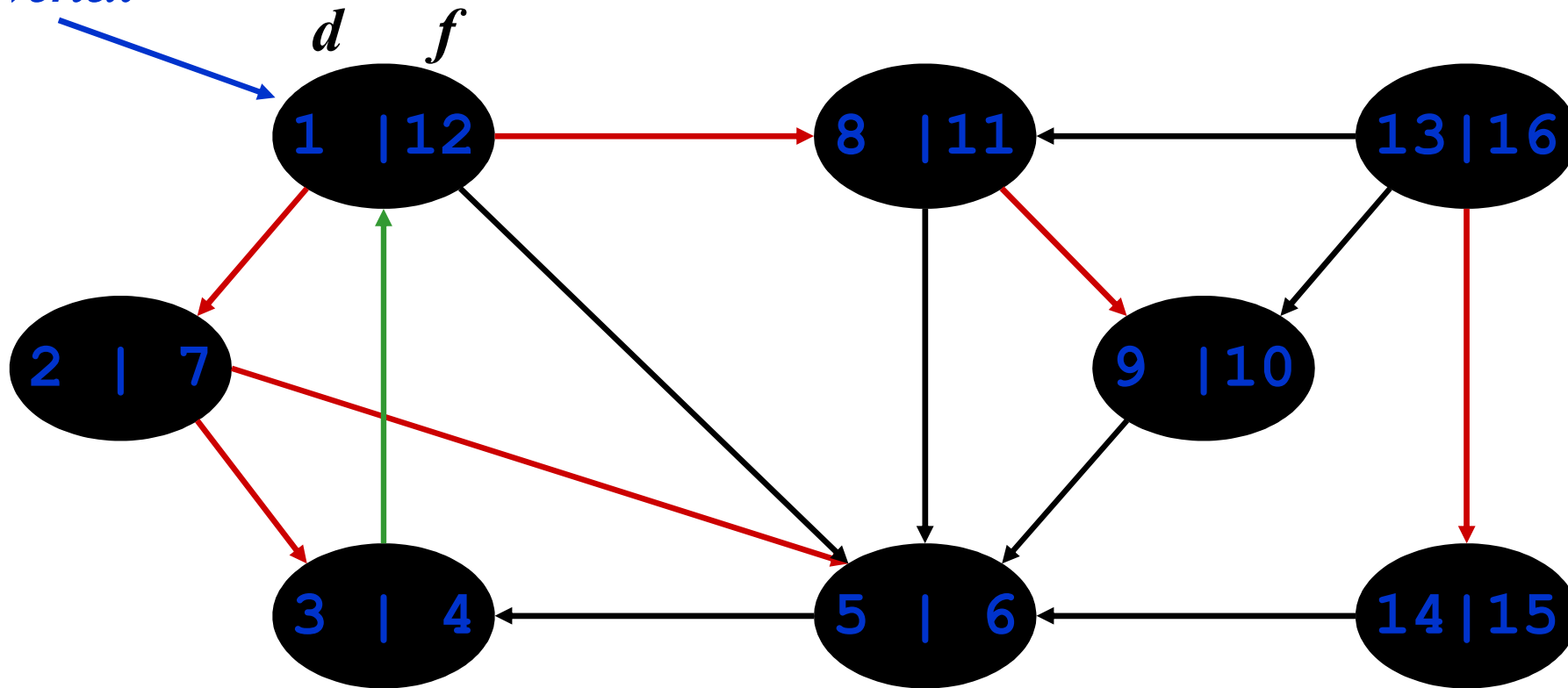
Tree edges

DFS: Kinds of edges

- DFS introduces an important distinction among edges in the original graph:
 - *Tree edge*: encounter new (white) vertex
 - *Back edge*: from descendent to ancestor
 - Encounter a grey vertex (grey to grey)

DFS Example

*source
vertex*

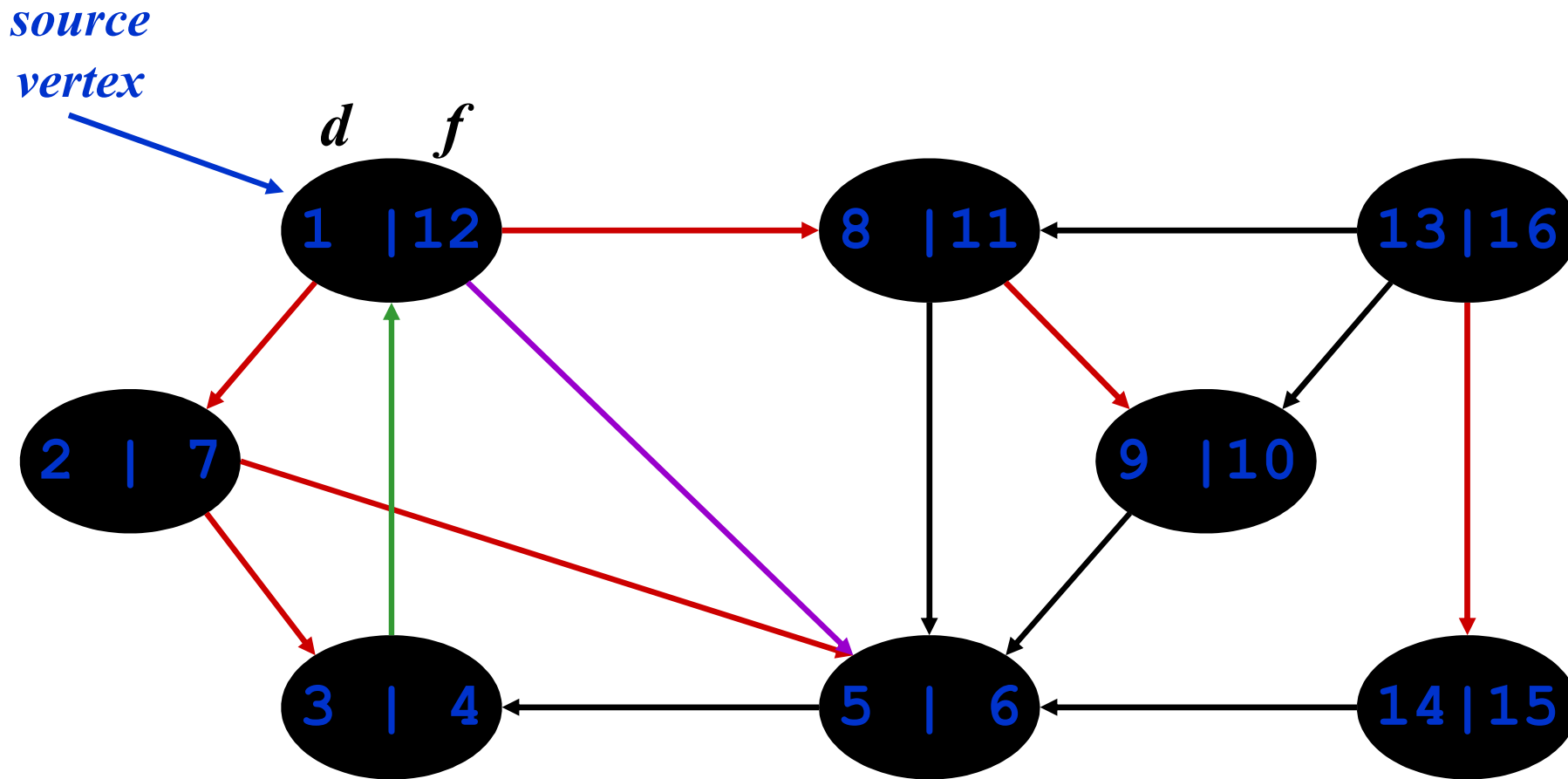


Tree edges Back edges

DFS: Kinds of edges

- DFS introduces an important distinction among edges in the original graph:
 - *Tree edge*: encounter new (white) vertex
 - *Back edge*: from descendent to ancestor
 - *Forward edge*: from ancestor to descendent
 - Not a tree edge, though
 - From grey node to black node

DFS Example



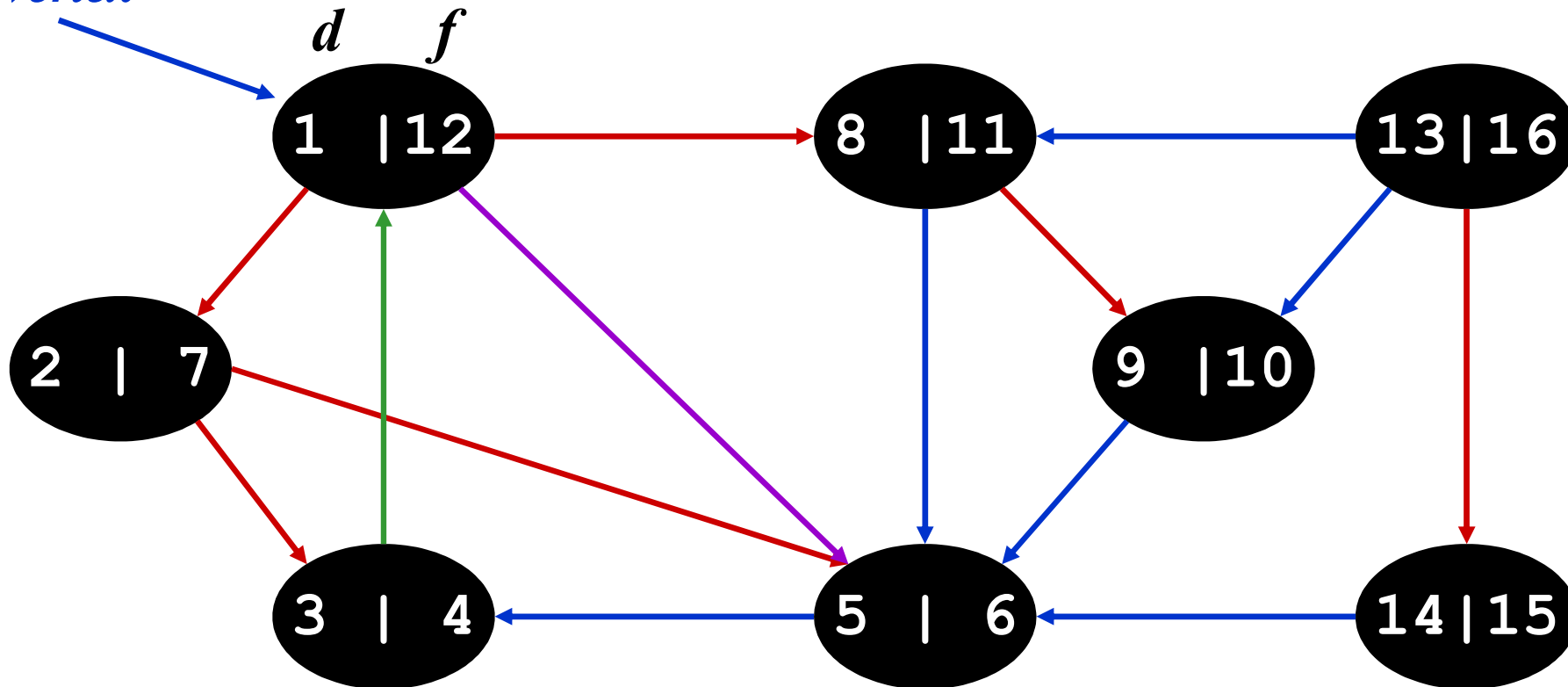
Tree edges *Back edges* *Forward edges*

DFS: Kinds of edges

- DFS introduces an important distinction among edges in the original graph:
 - *Tree edge*: encounter new (white) vertex
 - *Back edge*: from descendent to ancestor
 - *Forward edge*: from ancestor to descendent
 - *Cross edge*: between a tree or subtrees
 - From a grey node to a black node

DFS Example

*source
vertex*



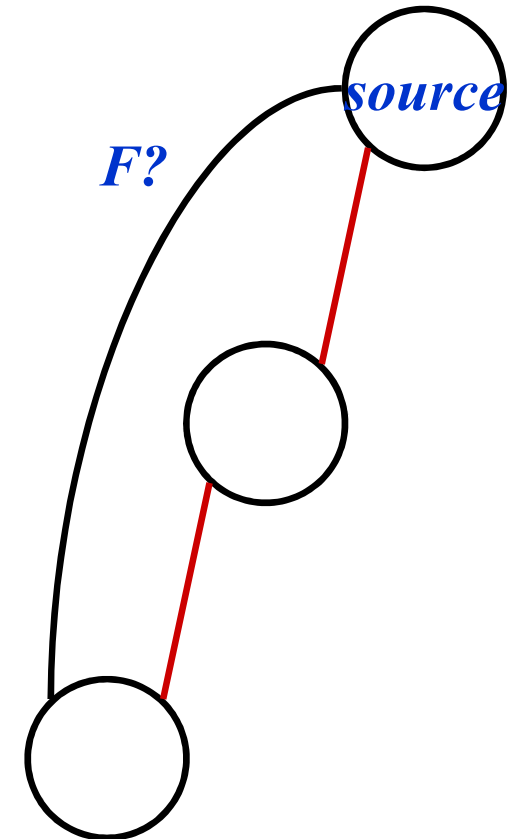
Tree edges Back edges Forward edges Cross edges

DFS: Kinds of edges

- DFS introduces an important distinction among edges in the original graph:
 - *Tree edge*: encounter new (white) vertex
 - *Back edge*: from descendent to ancestor
 - *Forward edge*: from ancestor to descendent
 - *Cross edge*: between a tree or subtrees
- Note: tree & back edges are important; most algorithms don't distinguish forward & cross

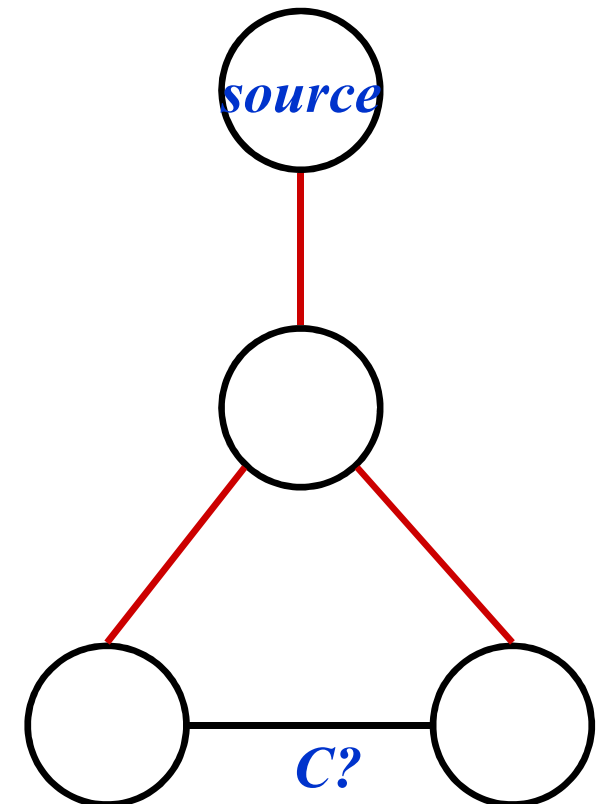
DFS: Kinds Of Edges

- Thm 23.9: If G is undirected, a DFS produces only tree and back edges
- Proof by contradiction:
 - Assume there's a forward edge
 - But F? edge must actually be a back edge (*why?*)



DFS: Kinds Of Edges

- Thm 23.9: If G is undirected, a DFS produces only tree and back edges
- Proof by contradiction:
 - Assume there's a cross edge
 - But C? edge cannot be cross:
 - must be explored from one of the vertices it connects, becoming a tree vertex, before other vertex is explored
 - So in fact the picture is wrong...both lower tree edges cannot in fact be tree edges



DFS And Graph Cycles

- Thm: An undirected graph is *acyclic* iff a DFS yields no back edges
 - If acyclic, no back edges (because a back edge implies a cycle)
 - If no back edges, acyclic
 - No back edges implies only tree edges (*Why?*)
 - Only tree edges implies we have a tree or a forest
 - Which by definition is acyclic
- Thus, can run DFS to find whether a graph has a cycle

DFS And Cycles

- *How would you modify the code to detect cycles?*

DFS (G)

```
{
    for each vertex u ∈ G->V
    {
        u->color = WHITE;
    }
    time = 0;
    for each vertex u ∈ G->V
    {
        if (u->color == WHITE)
            DFS_Visit(u);
    }
}
```

DFS_Visit(u)

```
{
    u->color = GREY;
    time = time+1;
    u->d = time;
    for each v ∈ u->Adj[]
    {
        if (v->color == WHITE)
            DFS_Visit(v);
    }
    u->color = BLACK;
    time = time+1;
    u->f = time;
}
```

DFS And Cycles

- *What will be the running time?*

DFS (G)

```
{
    for each vertex u ∈ G->V
    {
        u->color = WHITE;
    }
    time = 0;
    for each vertex u ∈ G->V
    {
        if (u->color == WHITE)
            DFS_Visit(u);
    }
}
```

DFS_Visit(u)

```
{
    u->color = GREY;
    time = time+1;
    u->d = time;
    for each v ∈ u->Adj[]
    {
        if (v->color == WHITE)
            DFS_Visit(v);
    }
    u->color = BLACK;
    time = time+1;
    u->f = time;
}
```

DFS And Cycles

- *What will be the running time?*
- A: $O(V+E)$
- We can actually determine if cycles exist in $O(V)$ time:
 - In an undirected acyclic forest, $|E| \leq |V| - 1$
 - So count the edges: if ever see $|V|$ distinct edges, must have seen a back edge along the way

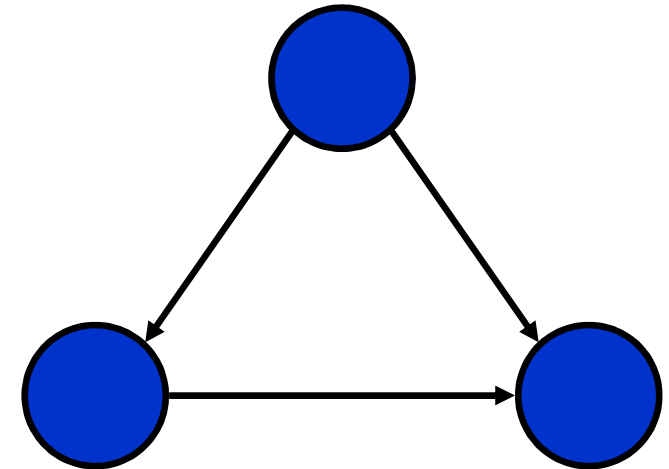
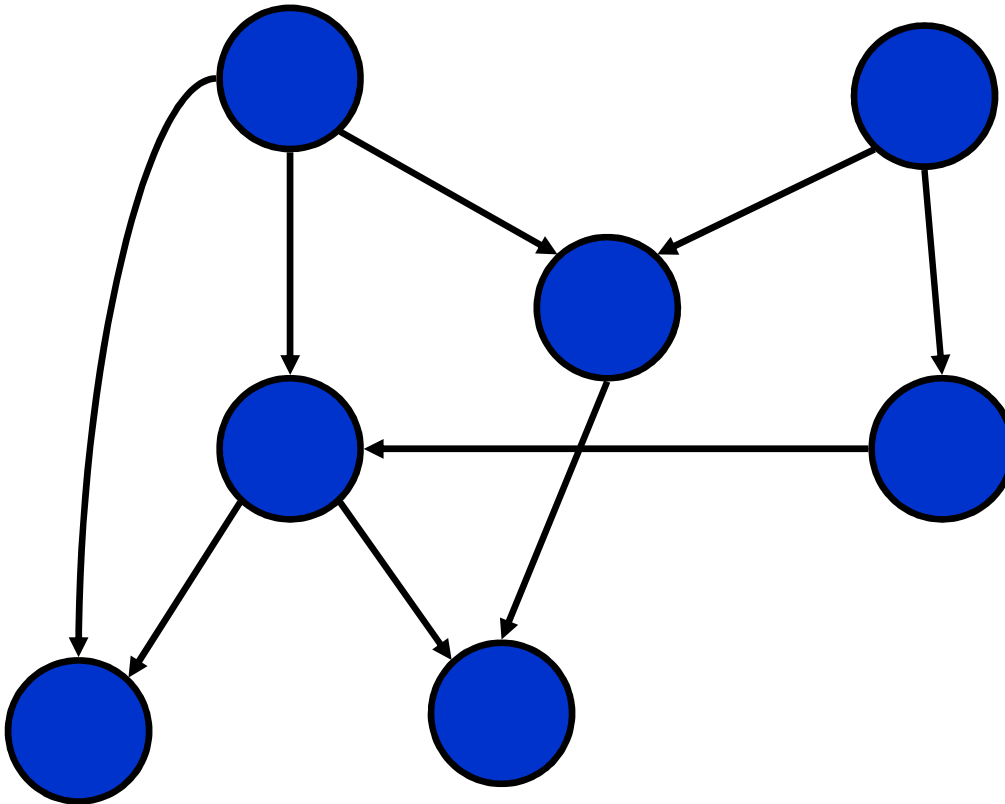
DFS And Cycles

- Running time: $O(V+E)$
- We can actually determine if cycles exist in $O(V)$ time:
 - In an undirected acyclic forest, $|E| \leq |V| - 1$
 - So count the edges: if ever see $|V|$ distinct edges, must have seen a back edge along the way
 - *Why not just test if $|E| < |V|$ and answer the question in constant time?*

DIRECTED ACYCLIC GRAPHS AND TOPOLOGICAL SORTING

Directed Acyclic Graphs

- A *directed acyclic graph* or *DAG* is a directed graph with no directed cycles:



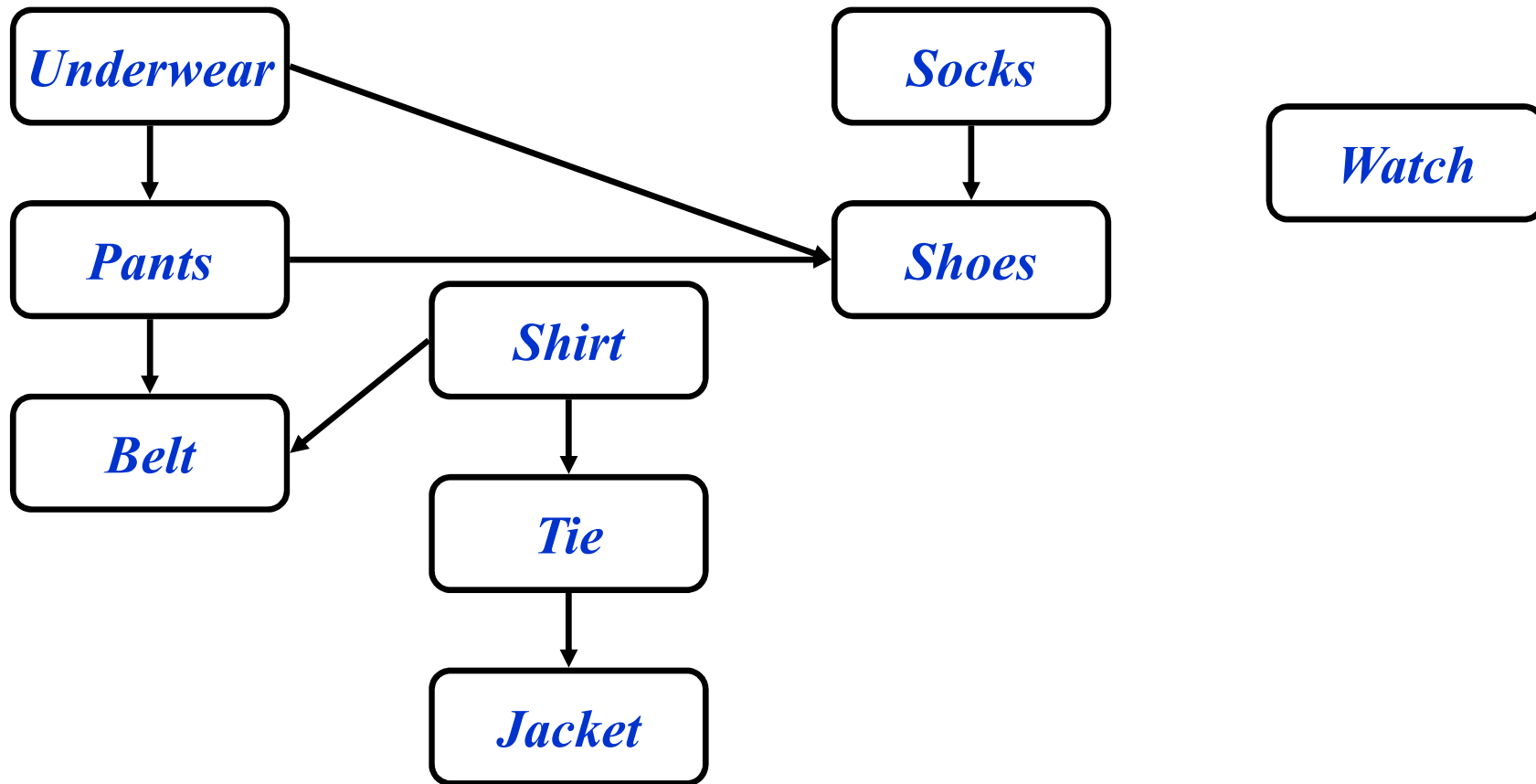
DFS and DAGs

- Argue that a directed graph G is acyclic iff a DFS of G yields no back edges:
 - Forward: if G is acyclic, will be no back edges
 - Trivial: a back edge implies a cycle
 - Backward: if no back edges, G is acyclic
 - Argue contrapositive: G has a cycle $\Rightarrow \exists$ a back edge
 - ✳ Let v be the vertex on the cycle first discovered, and u be the predecessor of v on the cycle
 - ✳ When v discovered, whole cycle is white
 - ✳ Must visit everything reachable from v before returning from DFS-Visit()
 - ✳ So path from $u \rightarrow v$ is yellow \rightarrow yellow, thus (u, v) is a back edge

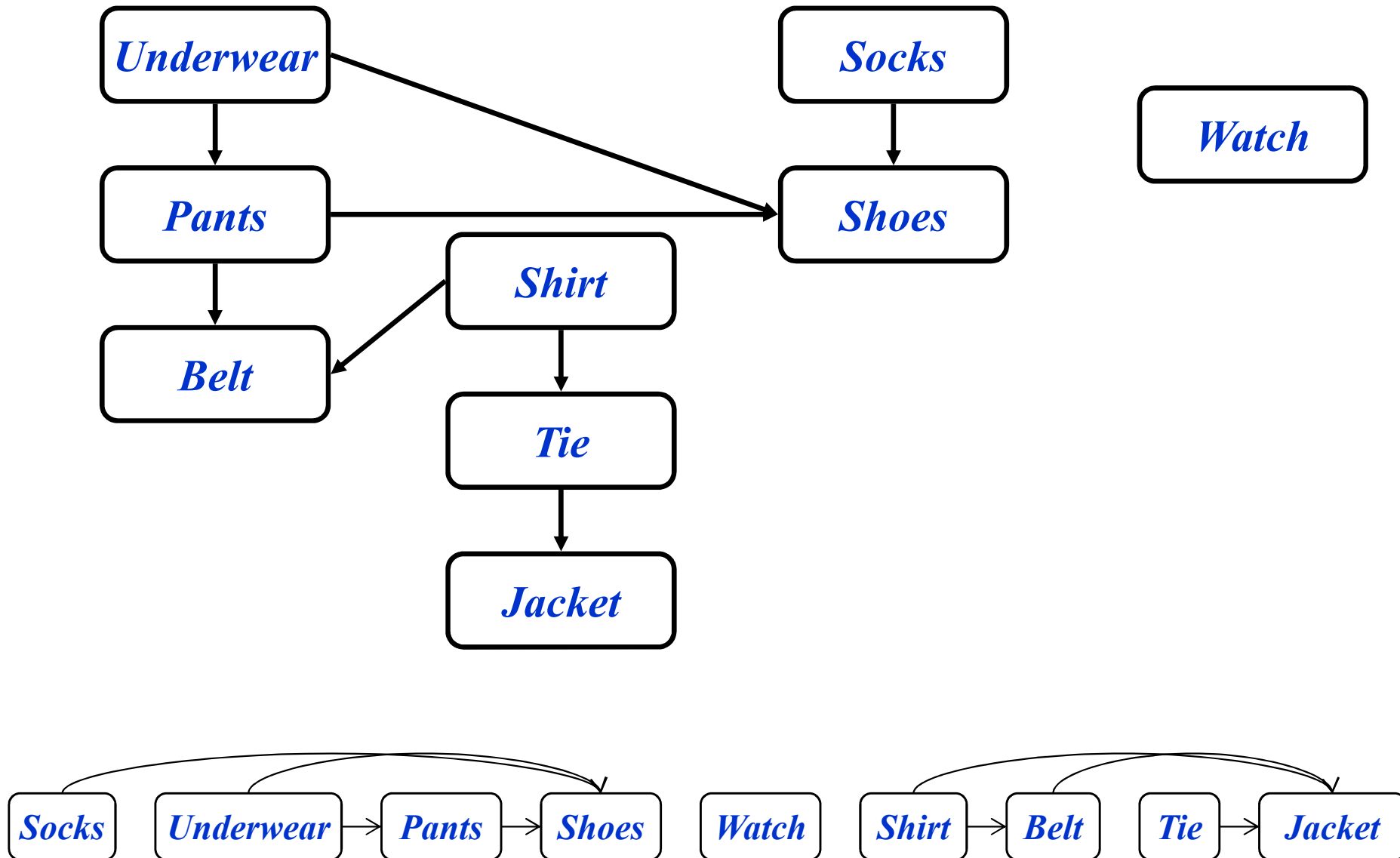
Topological Sort

- *Topological sort* of a DAG:
 - Linear ordering of all vertices in graph G such that vertex u comes before vertex v if edge $(u, v) \in G$
- Real-world example: getting dressed

Getting Dressed



Getting Dressed



Topological Sort Algorithm

Topological-Sort()

{

 Run DFS

 When a vertex is finished, output it

 Vertices are output in reverse
 topological order

}

- Time: $O(V+E)$
- Correctness: Want to prove that
$$(u,v) \in G \Rightarrow u \rightarrow f > v \rightarrow f$$

Correctness of Topological Sort

- Claim: $(u,v) \in G \Rightarrow u \rightarrow f > v \rightarrow f$
 - When (u,v) is explored, u is yellow
 - $v = \text{yellow} \Rightarrow (u,v)$ is back edge. Contradiction (*Why?*)
 - $v = \text{white} \Rightarrow v$ becomes descendent of $u \Rightarrow v \rightarrow f < u \rightarrow f$
(since must finish v before backtracking and finishing u)
 - $v = \text{black} \Rightarrow v$ already finished $\Rightarrow v \rightarrow f < u \rightarrow f$

Shortest path search

NEXT TOPICS

Next topic: Finding Shortest Paths

END OF LECTURE 13