

Directed Graphs (Digraphs) and Graphs

Definitions Graph ADT Traversal algorithms DFS

Lecturer: Georgy Gimel'farb

COMPSCI 220 Algorithms and Data Structures

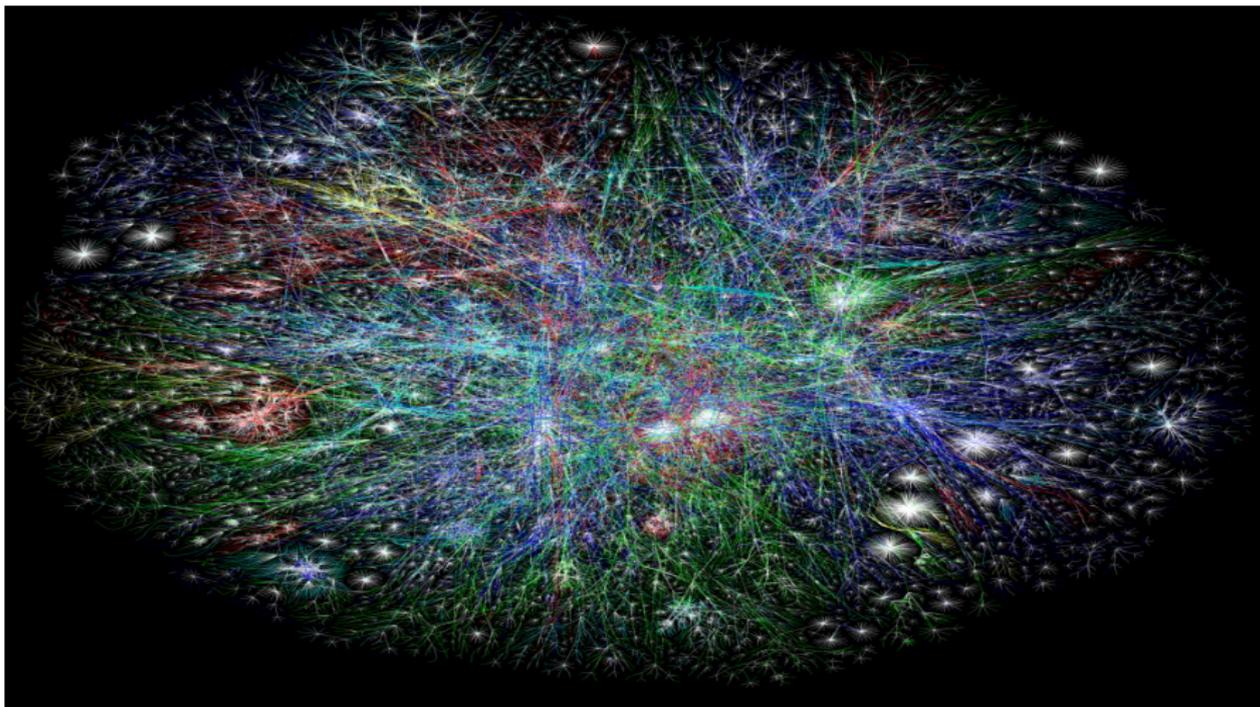
- ① Basic definitions
- ② Digraph Representation and Data Structures
- ③ Digraph ADT Operations
- ④ Graph Traversals and Applications
- ⑤ Depth-first Search in Digraphs

Graphs in Life: World Air Routes



<http://milenomics.com/2014/05/partners-alliances-partner-awards/>

Graphs in Life: Global Internet Connections



<http://www.opte.org/maps/>

Graphs in Life: Social Networks (Facebook)



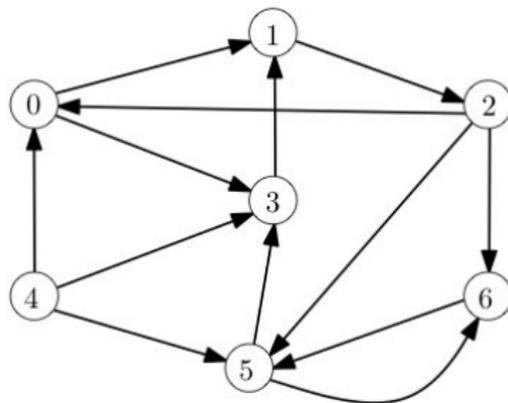
<http://robotmonkeys.net/wp-content/uploads/2010/12/social-nets-then-and-now-fb-cities-airlines-data.jpg>

Directed Graph, or Digraph: Definition

A **digraph** $G = (V, E)$ is a finite nonempty set V of **nodes** together with a (possibly empty) set E of ordered pairs of nodes^o of G called **arcs**.

$$V = \{ 0, 1, 2, 3, 4, 5, 6 \}$$

$$E = \{ (0, 1), (0, 3), (1, 2), (2, 0), (2, 5), (2, 6), (3, 1), (4, 0), (4, 3), (4, 5), (5, 3), (5, 6), (6, 5) \}$$

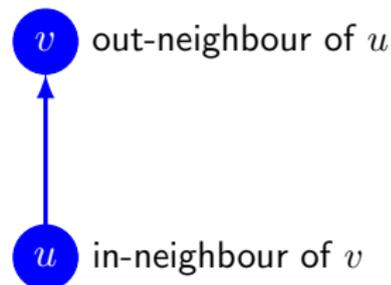


^o) Set E is a **neighbourhood**, or **adjacency relation** on V .

Digraph: Relations of Nodes

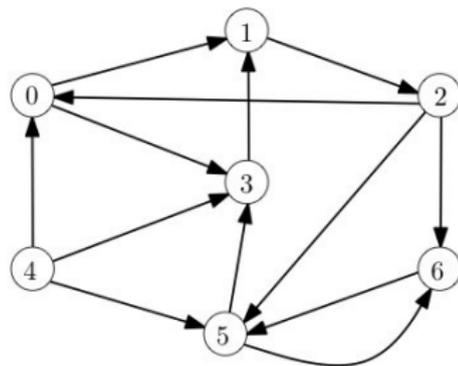
If $(u, v) \in E$,

- v is **adjacent** to u ;
- v is an **out-neighbour** of u , and
- u is an **in-neighbour** of v .



Examples:

- Nodes (points) 1 and 3 are adjacent to 0.
- 1 and 3 are out-neighbours of 0.
- 0 is an in-neighbour of 1 and 3.
- Node 1 is adjacent to 3.
- 1 is an out-neighbour of 3.
- 3 is an in-neighbour of 1. ...
- 5 is an out-neighbour of 2, 4, and 6.

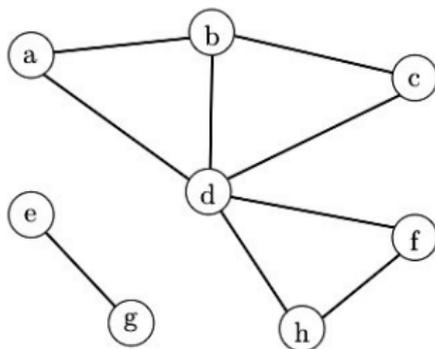


(Undirected) Graph: Definition

A **graph**^o $G = (V, E)$ is a finite nonempty set V of **vertices** together with a (possibly empty) set E of unordered pairs of vertices of G called **edges**.

$$V = \{ a, b, c, d, e, f, g, h \}$$

$$E = \{ \{a, b\}, \{a, d\}, \{b, d\}, \{b, c\}, \\ \{c, d\}, \{d, f\}, \{d, h\}, \{f, h\}, \\ \{e, g\} \}$$



o) The symmetric digraph: each arc (u, v) has the opposite arc (v, u) .

Such a pair is reduced into a single undirected edge that can be traversed in either direction.

Order, Size, and In- / Out-degree

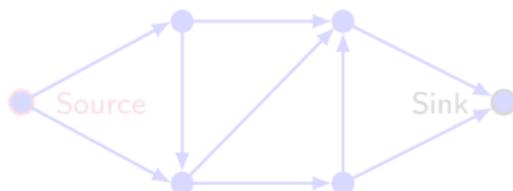
The **order** of a digraph $G = (V, E)$ is the number of nodes, $n = |V|$.

The **size** of a digraph $G = (V, E)$ is the number of arcs, $m = |E|$.

For a given n , Sparse digraphs: $|E| \in O(n)$ Dense digraphs: $|E| \in \Theta(n^2)$
 $m = 0$ ————— $n(n-1)$

The **in-degree** or **out-degree** of a node v is the number of arcs entering or leaving v , respectively.

- A node of in-degree 0 – a **source**.
- A node of out-degree 0 – a **sink**.
- This example: the order $|V| = 6$ and the size $|E| = 9$.



Order, Size, and In- / Out-degree

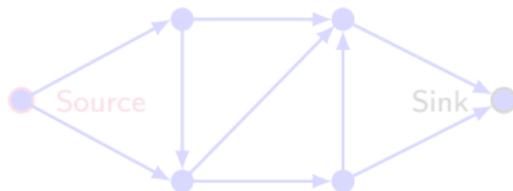
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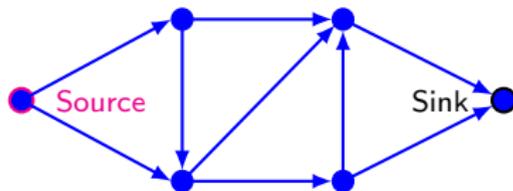
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Walk, Path, and Cycle

A **walk** in a digraph $G = (V, E)$:

a sequence of nodes $v_0 v_1 \dots v_n$, such that (v_i, v_{i+1}) is an arc in G , i.e., $(v_i, v_{i+1}) \in E$, for each i ; $0 \leq i < n$.

- The **length** of the walk $v_0 v_1 \dots v_n$ is the number n of arcs involved.
- A **path** is a walk, in which no node is repeated.
- A **cycle** is a walk, in which $v_0 = v_n$ and no other nodes are repeated.
- By convention, a cycle in a graph is of length at least 3.
- It is easily shown that if there is a walk from u to v , then there is at least one path from u to v .

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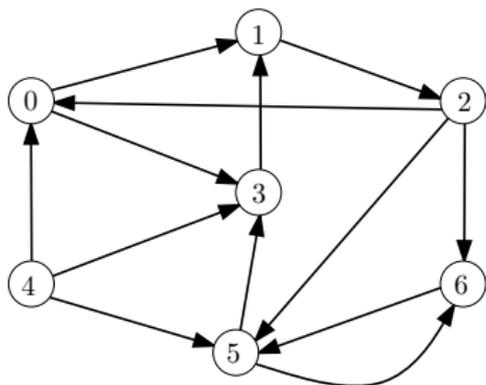
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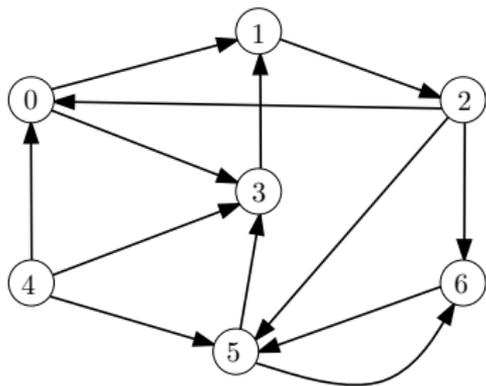
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Walks, Paths, and Cycles in a Digraph: an Example



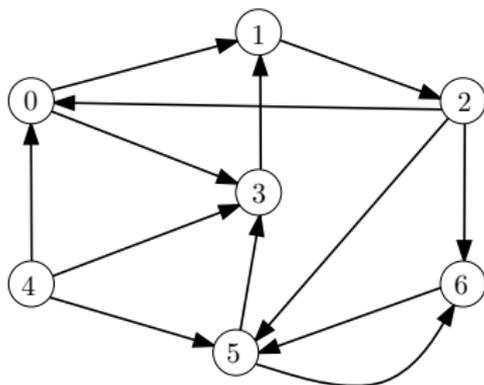
Sequence	Walk?	Path?	Cycle?
0 2 3	no	no	no
3 1 2	yes	yes	no
1 2 6 5 3 1	yes	no	yes
4 5 6 5	yes	no	no
4 3 5	no	no	no

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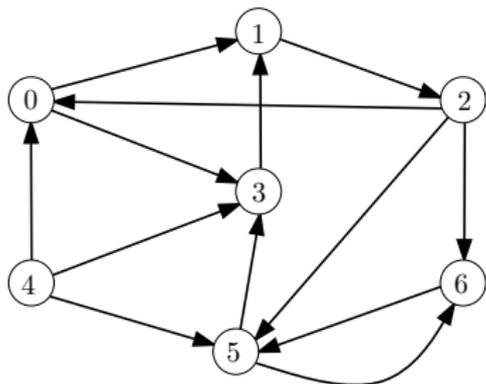
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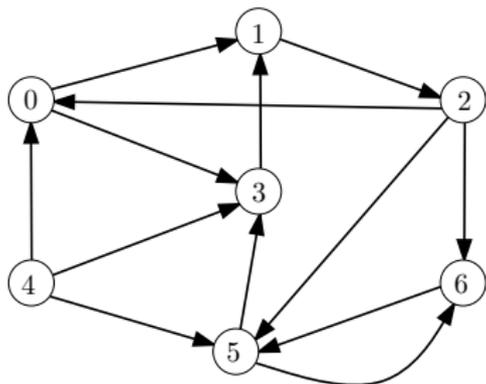
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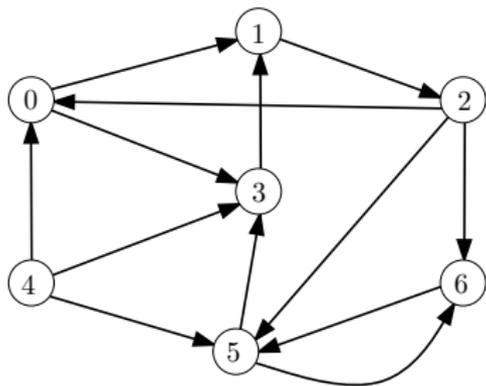
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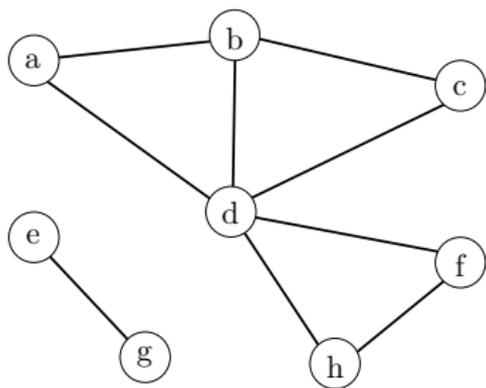
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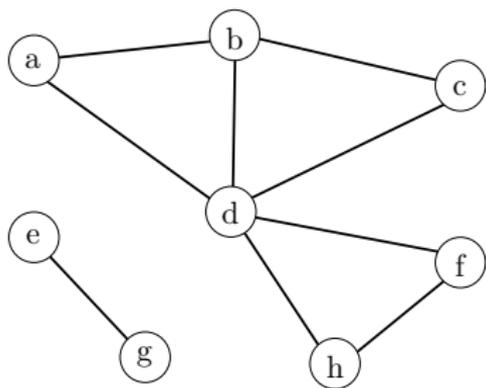
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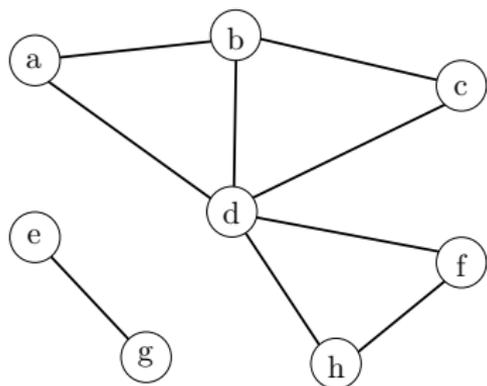
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<i>ege</i>	yes	no	no
<i>dbcd</i>	yes	no	yes
<i>dadf</i>	yes	no	no
<i>abdfh</i>	yes	yes	no

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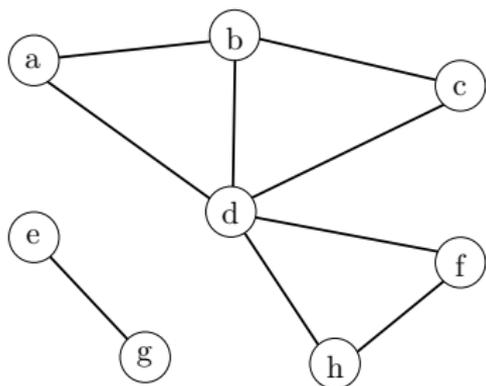
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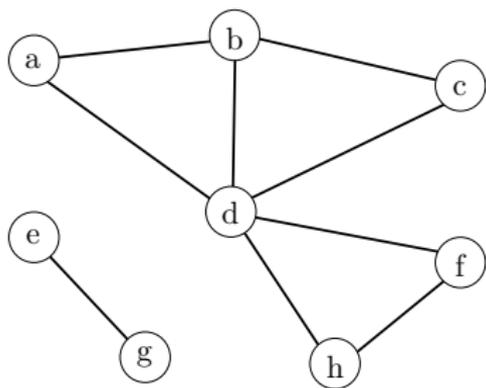
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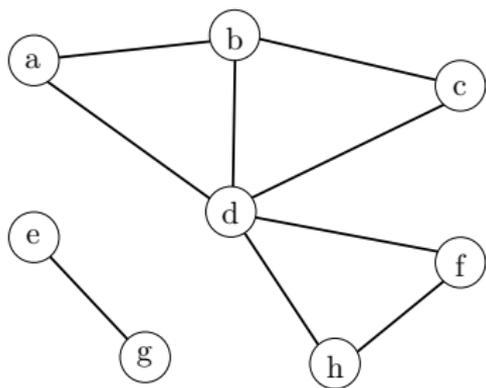
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Digraph $G = (V, E)$: Distances and Diameter

The **distance**, $d(u, v)$, from a node u to a node v in G is the *minimum* length of a path from u to v .

- If no path exists, the distance is undefined or $+\infty$.
- For graphs, $d(u, v) = d(v, u)$ for all vertices u and v .

The **diameter** of G is the *maximum* distance $\max_{u, v \in V} [d(u, v)]$ between any two vertices.

The **radius** of G is $\min_{u \in V} \max_{v \in V} [d(u, v)]$.

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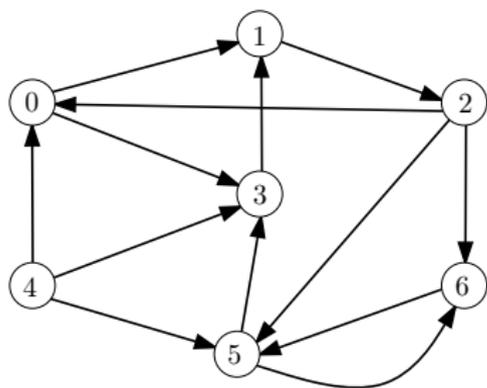
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Path Distances in Digraphs: Examples

$$\begin{aligned}
 d(0, 3) &= \min\{\text{length}_{\text{of } 0,3}; \text{length}_{\text{of } 0,1,2,6,5,3}; \text{length}_{\text{of } 0,1,2,5,3}\} \\
 &= \min\{1; 5; 4\} = 1
 \end{aligned}$$



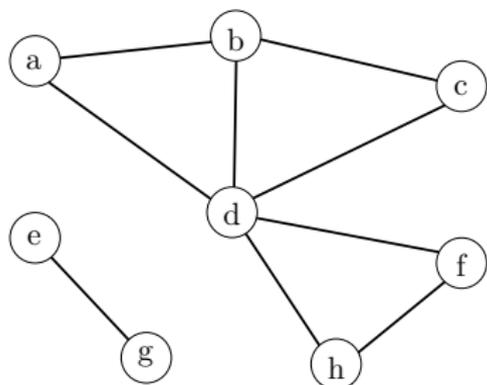
	v						
	0	1	2	3	4	5	6
$u=0$	—	1	2	1	∞	3	3
$u=1$	2	—	1	3	∞	2	2
$u=2$	1	3	—	2	∞	1	1
$u=3$	3	1	2	—	∞	3	3
$u=4$	1	2	3	1	—	1	2
$u=5$	4	2	3	1	∞	—	1
$u=6$	5	3	4	2	∞	1	—

$$\begin{aligned}
 d(0, 1) = 1, \quad d(0, 2) = 2, \quad d(0, 5) = 3, \quad d(0, 4) = \infty, \quad d(5, 5) = 0, \quad d(5, 2) = 3, \\
 d(5, 0) = 4, \quad d(4, 6) = 2, \quad d(4, 1) = 2, \quad d(4, 2) = 3
 \end{aligned}$$

Diameter: $\max\{1, 2, 1, \infty, 3, \dots, 4, \dots, 5, \dots, 1\} = \infty$

Raduis: $\min\{\infty, \infty, \dots, 3, \infty, \infty\} = 3$

Path Distances in Graphs: Examples



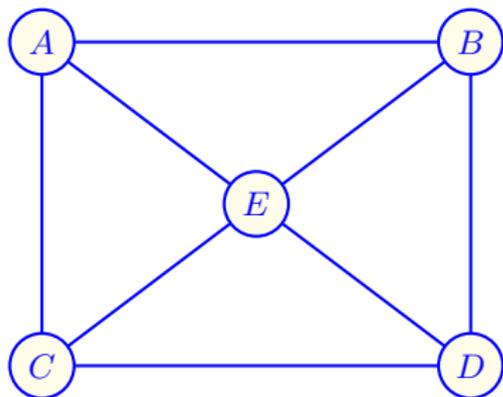
	v							
	a	b	c	d	e	f	g	h
$u=a$	0	1	2	1	∞	2	∞	2
$u=b$	1	0	1	1	∞	2	∞	2
$u=c$	2	1	0	1	∞	2	∞	2
$u=d$	1	1	1	0	∞	1	∞	1
$u=e$	∞	∞	∞	∞	0	∞	1	∞
$u=f$	2	2	2	1	∞	0	∞	1
$u=g$	∞	∞	∞	∞	1	∞	0	∞
$u=h$	2	2	2	1	∞	1	∞	0

$d(a, b) = d(b, a) = 1$, $d(a, c) = d(c, a) = 2$, $d(a, f) = d(f, a) = 2$,
 $d(a, e) = d(e, a) = \infty$, $d(e, e) = 0$, $d(e, g) = d(g, e) = 1$, $d(h, f) = d(f, h) = 1$,
 $d(d, h) = d(h, d) = 1$

Diameter: $\max\{0, 1, 2, 1, \infty, 2, \dots, 2, \dots, 2, \dots, 0\} = \infty$

Radius: $\min\{\infty, \dots, \infty\} = \infty$

Diameter / Radius of an Unweighted Graph



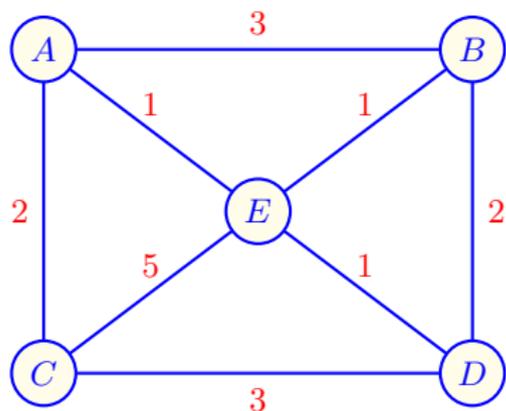
	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	$\max_v d(u,v)$
<i>A</i>	0	1	1	2	1	2
<i>B</i>	1	0	2	1	1	2
<i>C</i>	1	2	0	1	1	2
<i>D</i>	2	1	1	0	1	2
<i>E</i>	1	1	1	1	0	1

$$\begin{aligned}
 d(C, E) &= d(E, C) \\
 &= \min\{1, 1 + 1, 1 + 1, 1 + 1 + 1, 1 + 1 + 1\} = 1
 \end{aligned}$$

$$\begin{aligned}
 d(B, C) &= d(C, B) \\
 &= \min\{1 + 1, 1 + 1 + 1, 1 + 1, 1 + 1 + 1, 1 + 1, 1 + 1 + 1\} = 2
 \end{aligned}$$

Radius = 1; diameter = 2.

Diameter / Radius of a Weighted Graph



	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	$\max_v d(u,v)$
<i>A</i>	0	2	2	2	1	2
<i>B</i>	2	0	4	2	1	4
<i>C</i>	2	4	0	3	3	4
<i>D</i>	2	2	3	0	1	3
<i>E</i>	1	1	3	1	0	3

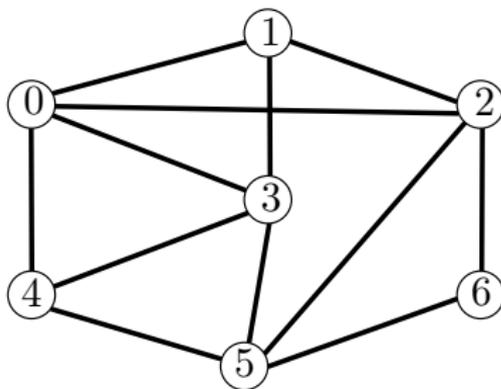
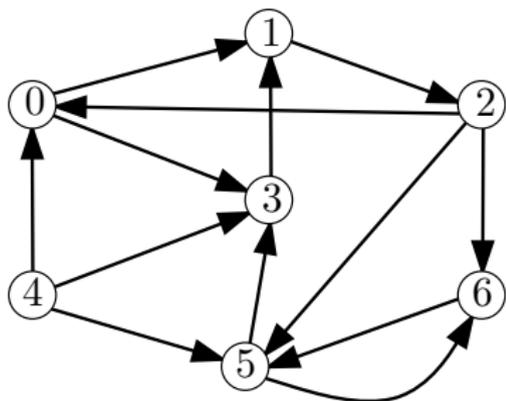
$$\begin{aligned}
 d(C, E) &= d(E, C) \\
 &= \min\{5, 2 + 1, 3 + 1, 2 + 3 + 1, 3 + 2 + 1\} = 3
 \end{aligned}$$

$$\begin{aligned}
 d(B, C) &= d(C, B) \\
 &= \min\{3 + 2, 1 + 1 + 2, 1 + 5, 1 + 1 + 3, 2 + 3, 2 + 1 + 5\} = 4
 \end{aligned}$$

Radius = 2; diameter = 4.

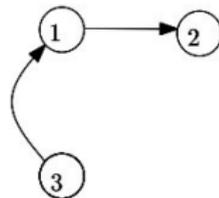
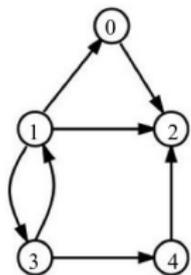
Underlying Graph of a Digraph

The **underlying graph** of a digraph $G = (V, E)$ is the graph $G' = (V, E')$ where $E' = \{\{u, v\} \mid (u, v) \in E\}$.



Sub(di)graphs

A **subdigraph** of a digraph $G = (V, E)$ is a digraph $G' = (V', E')$ where $V' \subseteq V$ and $E' \subseteq E$.

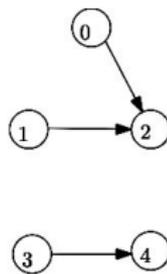
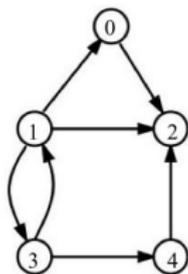


$$G = \left(\begin{array}{l} V = \{0, 1, 2, 3, 4\}, \\ E = \left\{ \begin{array}{l} (0, 2), (1, 0), (1, 2), \\ (1, 3), (3, 1), (4, 2), \\ (3, 4) \end{array} \right\} \end{array} \right)$$

$$G' = \left(\begin{array}{l} V' = \{1, 2, 3\}, \\ E' = \{(1, 2), (3, 1)\} \end{array} \right)$$

Spanning Sub(di)graphs

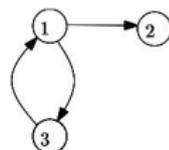
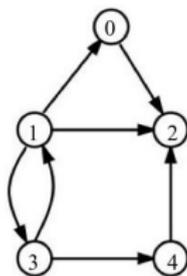
A **spanning** subdigraph contains all nodes, that is, $V' = V$.



$$G = \left(\begin{array}{l} V = \{0, 1, 2, 3, 4\}, \\ E = \left\{ \begin{array}{l} (0, 2), (1, 0), (1, 2), \\ (1, 3), (3, 1), (4, 2), \\ (3, 4) \end{array} \right\} \end{array} \right) \quad G' = \left(\begin{array}{l} V' = \{0, 1, 2, 3, 4\}, \\ E' = \left\{ \begin{array}{l} (0, 2), (1, 2), \\ (3, 4) \end{array} \right\} \end{array} \right)$$

Induced Sub(di)graphs

The subdigraph **induced** by a subset V' of V is the digraph $G' = (V', E')$ where $E' = \{(u, v) \in E \mid u \in V' \text{ and } v \in V'\}$.



$$G = \left(\begin{array}{l} V = \{0, 1, 2, 3, 4\}, \\ E = \left\{ \begin{array}{l} (0, 2), (1, 0), (1, 2), \\ (1, 3), (3, 1), (4, 2), \\ (3, 4) \end{array} \right\} \end{array} \right) \quad G' = \left(\begin{array}{l} V' = \{1, 2, 3\}, \\ E' = \left\{ \begin{array}{l} (1, 2), (1, 3), \\ (3, 1) \end{array} \right\} \end{array} \right)$$

Digraphs: Computer Representation

For a digraph G of order n with the vertices, V , labelled $0, 1, \dots, n - 1$:

The **adjacency matrix** of G :

The $n \times n$ boolean matrix (often encoded with 0's and 1's) such that its entry (i, j) is true if and only if there is an arc (i, j) from the node i to node j .

An **adjacency list** of G :

A sequence of n sequences, L_0, \dots, L_{n-1} , such that the sequence L_i contains all nodes of G that are adjacent to the node i .

Each sequence L_i may not be sorted! But we usually sort them.

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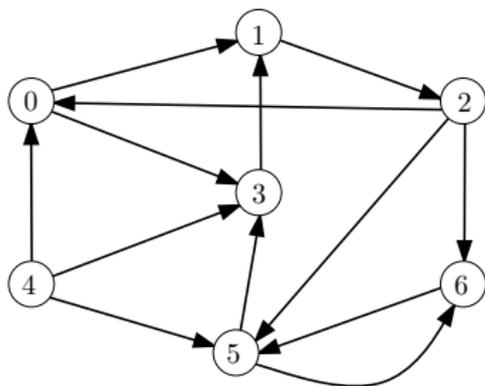
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Each sequence L_i may not be sorted! **But we usually sort them.**

Adjacency Matrix of a Digraph



Digraph $G = (V, E)$

	0	1	2	3	4	5	6
0	0	1	0	1	0	0	0
1	0	0	1	0	0	0	0
2	1	0	0	0	0	1	1
3	0	1	0	0	0	0	0
4	1	0	0	1	0	1	0
5	0	0	0	1	0	0	1
6	0	0	0	0	0	1	0

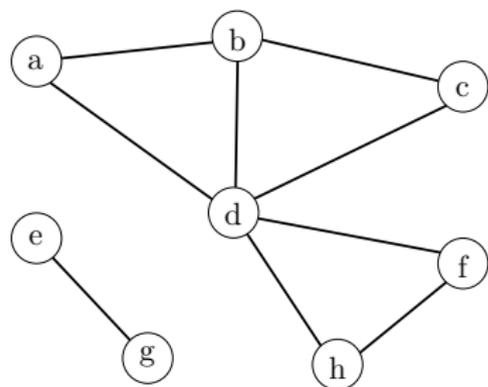
Adjacency matrix of G :

0 – a non-adjacent pair of vertices:
 $(i, j) \notin E$

1 – an adjacent pair of vertices:
 $(i, j) \in E$

The number of 1's in a row (column) is the out-(in-) degree of the related node.

Adjacency Lists of a Graph



Graph $G = (V, E)$

symbolic

0 = a: b d
 1 = b: a c d
 2 = c: b d
 3 = d: a b c f h
 4 = e: g
 5 = f: d h
 6 = g: e
 7 = h: d f

numeric

8
 1 3
 0 2 3
 1 3
 0 1 2 5 7
 6
 3 7
 4
 3 5

Special cases can be stored more efficiently:

- A complete binary tree or a heap: in an array.
- A general rooted tree: in an array `pred` of size n ;
 - `pred[i]` – a pointer to the parent of node i .

Digraph Operations w.r.t. Data Structures

Operation	Adjacency Matrix	Adjacency Lists
arc (i, j) exists?	is entry (i, j) 0 or 1	find j in list i
out-degree of i	scan row and sum 1's	size of list i
in-degree of i	scan column and sum 1's	for $j \neq i$, find i in list j
add arc (i, j)	change entry (i, j)	insert j in list i
delete arc (i, j)	change entry (i, j)	delete j from list i
add node	create new row/column	add new list at end
delete node i	delete row/column i and shuffle other entries	delete list i and for $j \neq i$, delete i from list j

Adjacency Lists / Matrices: Comparative Performance

$$G = (V, E) \quad \rightarrow \quad n = |V|; \quad m = |E|$$

Operation	array/array	list/list
arc (i, j) exists?	$\Theta(1)$	$\Theta(\alpha)^{\circ}$
out-degree of i	$\Theta(n)$	$\Theta(1)$
in-degree of i	$\Theta(n)$	$\Theta(n + m)$
add arc (i, j)	$\Theta(1)$	$\Theta(1)$
delete arc (i, j)	$\Theta(1)$	$\Theta(\alpha)$
add node	$\Theta(n)$	$\Theta(1)$
delete node i	$\Theta(n^2)$	$\Theta(n + m)$

^oHere, α denotes size of the adjacency list for vertex i .

General Graph Traversal Algorithm

(Part 1)

algorithm traverse

Input: digraph $G = (V, E)$

begin

array $colour[n]$, $pred[n]$

for $u \in V(G)$ **do**

$colour[u] \leftarrow$ WHITE

end for

for $s \in V(G)$ **do**

if $colour[s] =$ WHITE **then**

$visit(s)$

end if

end for

return $pred$

end

Three types of nodes each stage:

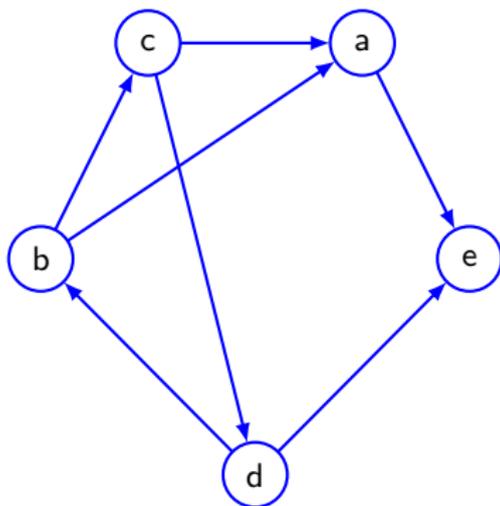
- WHITE – unvisited yet.
- GREY – visited, but some adjacent nodes are WHITE.
- BLACK – visited; only GREY adjacent nodes

General Graph Traversal Algorithm

(Part 2)

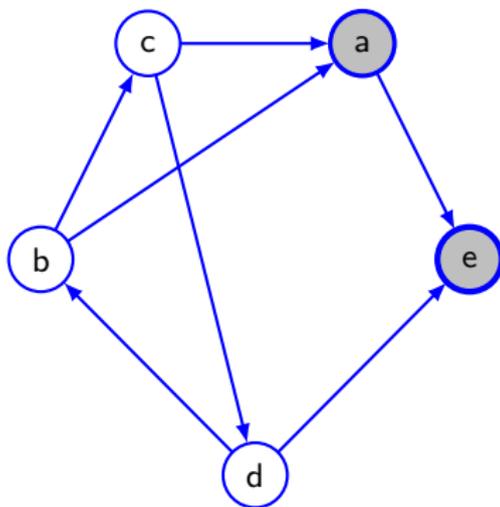
algorithm visit**Input:** node s of digraph G **begin** $colour[s] \leftarrow \text{GREY}; pred[s] \leftarrow \text{NULL}$ **while** there is a grey node **do** choose a grey node u **if** there is a white neighbour of u choose such a neighbour v $colour[v] \leftarrow \text{GREY}; pred[v] \leftarrow u$ **else** $colour[u] \leftarrow \text{BLACK}$ **end if** **end while****end**

Illustrating the General Traversal Algorithm



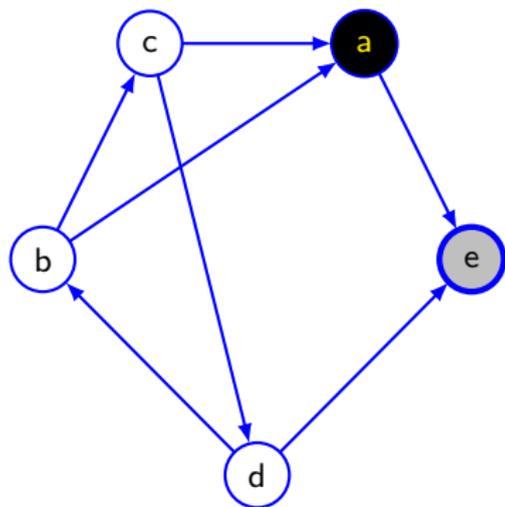
initialising all nodes WHITE

Illustrating the General Traversal Algorithm



`visit(a); colour[a] ← GREY`
e is WHITE neighbour of a:
`colour[e] ← GREY; pred[e] ← a`

Illustrating the General Traversal Algorithm



`visit(a); colour[a] ← GREY`

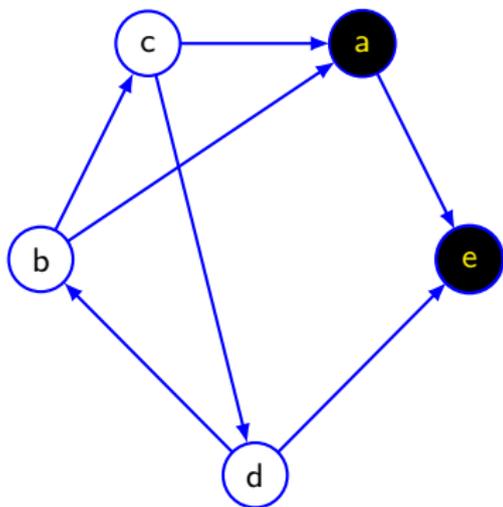
`e is WHITE neighbour of a`

`colour[e] ← GREY; pred[e] ← a`

`choose GREY a: no WHITE neighbour:`

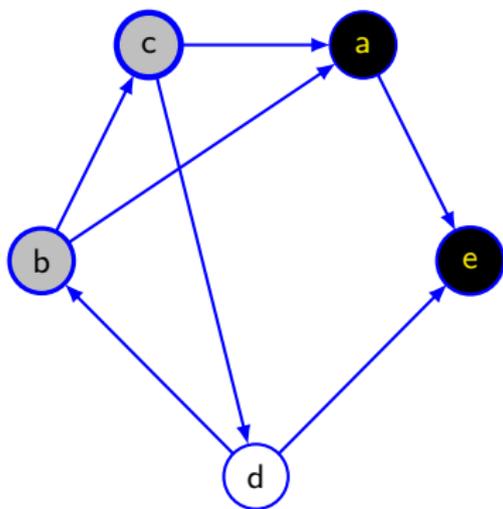
`colour[a] ← BLACK`

Illustrating the General Traversal Algorithm



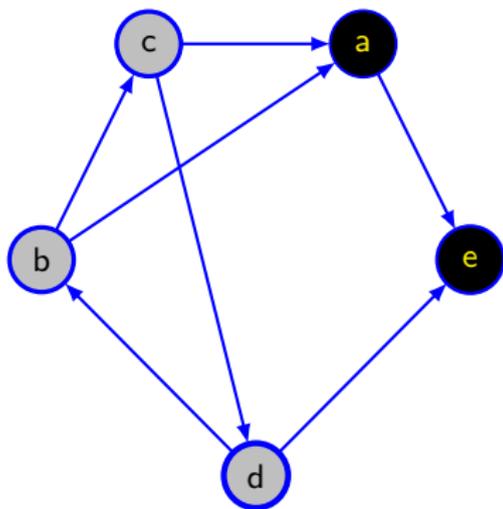
```
visit(a); colour[a] ← GREY
e is WHITE neighbour of a
  colour[e] ← GREY; pred[e] ← a
choose GREY a: no WHITE neighbour:
  colour[a] ← BLACK
choose GREY e: no WHITE neighbour:
  colour[e] ← BLACK
```

Illustrating the General Traversal Algorithm



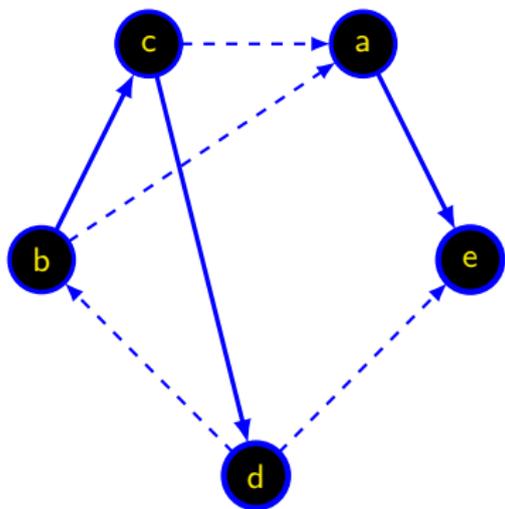
`visit(b); colour[b] ← GREY`
c is WHITE neighbour of b
`colour[c] ← GREY; pred[c] ← b`

Illustrating the General Traversal Algorithm



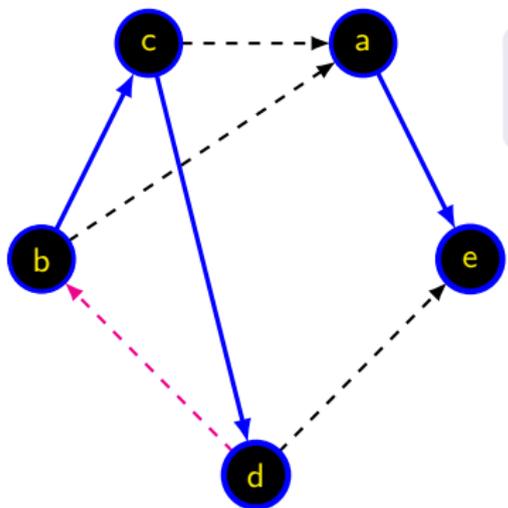
```
visit(b); colour[b] ← GREY  
c is WHITE neighbour of b  
    colour[c] ← GREY; pred[c] ← b  
d is WHITE neighbour of c  
    colour[d] ← GREY; pred[d] ← c
```

Illustrating the General Traversal Algorithm



```
visit(b); colour[b] ← GREY  
c is WHITE neighbour of b  
    colour[c] ← GREY; pred[c] ← b  
d is WHITE neighbour of c  
    colour[d] ← GREY; pred[d] ← c  
no more WHITE nodes:  
    colour[d] ← BLACK  
    colour[c] ← BLACK  
    colour[b] ← BLACK
```

Classes of Traversal Arcs



Search forest F : a set of disjoint trees spanning a digraph G after its traversal.

An arc $(u, v) \in E(G)$ is called a **tree arc** if it belongs to one of the trees of F

The arc (u, v) , which is not a tree arc, is called:

- a **forward arc** if u is an ancestor of v in F ;
- a **back arc** if u is a descendant of v in F , and
- a **cross arc** if neither u nor v is an ancestor of the other in F .

Basic Facts about Traversal Trees (for further analyses)

Theorem 5.2: Suppose we have run `traverse` on G , resulting in a search forest F .

- 1 If T_1 and T_2 are different trees in F and T_1 was explored before T_2 , then there are no arcs from T_1 to T_2 .
- 2 If G is a graph, then there can be no edges joining different trees of F .
- 3 If $v, w \in V(G)$; v is visited before w , and w is reachable from v in G , then v and w belong to the same tree of F .
- 4 If $v, w \in V(G)$ and v and w belong to the same tree T in F , then any path from v to w in G must have all nodes in T .

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Run-time Analysis of Algorithm traverse

In the **while-loop** of subroutine `visit` let:

- a (A) be lower (upper) time bound to choose a GREY node.
- b (B) be lower (upper) time bound to choose a WHITE neighbour.

Given a (di)graph $G = (V, E)$ of order $n = |V|$ and size $m = |E|$, the running time of `traverse` is:

- $O(An + Bm)$ and $\Omega(an + bm)$ with adjacency lists, and
- $O(An + Bn^2)$ and $\Omega(an + bn^2)$ with an adjacency matrix.

Time to find a GREY node: $O(An)$ and $\Omega(an)$

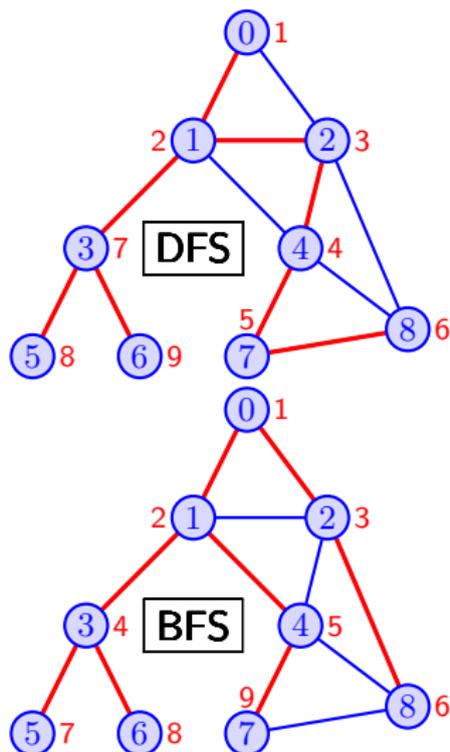
Time to find a WHITE neighbour: $O(Bm)$ and $\Omega(bm)$ (adjacency lists)

$O(Bn^2)$ and $\Omega(bn^2)$ (an adjacency matrix)

- Generally, A, B, a, b may depend on n .
- A more detailed analysis depends on the rules used.

Main Rules for Choosing Next Nodes

- Depth-first search (DFS):
 - Starting at a node v .
 - Searching as far away from v as possible via neighbours.
 - Continue from the next neighbour until no more new nodes.
- Breadth-first search (BFS):
 - Starting at a node v .
 - Searching through all its neighbours, then through all their neighbours, etc.
 - Continue until no more new nodes.
- More complicated priority-first search (PFS).



Depth-first Search (DFS) Algorithm

(Part 1)

algorithm dfs

Input: digraph $G = (V(G), E(G))$

begin

stack S ; array $colour[n], pred[n], seen[n], done[n]$

for $u \in V(G)$ **do**

$colour[u] \leftarrow \text{WHITE}; pred[u] \leftarrow \text{NULL}$

end for

$time \leftarrow 0$

for $s \in V(G)$ **do**

if $colour[s] = \text{WHITE}$ **then**

$dfsvisit(s)$

end if

end for

return $pred, seen, done$

end

Depth-first Search (DFS) Algorithm

(Part 2)

algorithm dfsvisit

Input: node s

begin

$colour[s] \leftarrow \text{GREY}; seen[s] \leftarrow time + +;$

$S.push_top(s)$

while not $S.isempty()$ **do**

$u \leftarrow S.get_top()$

if there is a v adjacent to u **and** $colour[v] = \text{WHITE}$ **then**

$colour[v] \leftarrow \text{GREY}; pred[v] \leftarrow u$

$seen[v] \leftarrow time + +; S.push_top(v)$

else $S.del_top();$

$colour[u] \leftarrow \text{BLACK}; done[u] \leftarrow time + +;$

end if

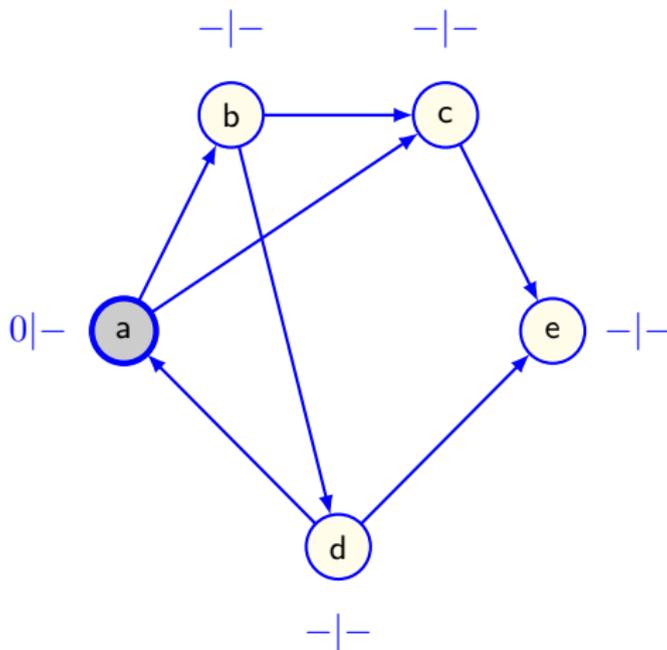
end while

end

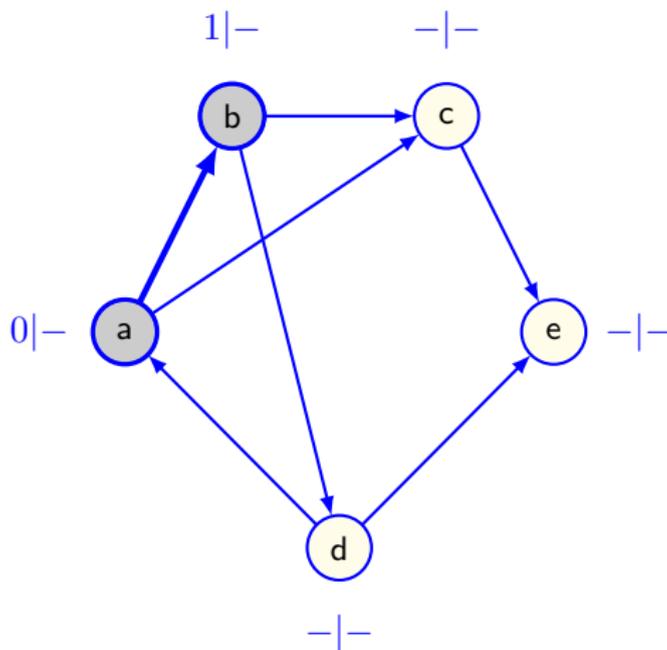
Recursive View of DFS Algorithm

```
algorithm rec_dfs_visit
  Input: node  $s$ 
begin
   $colour[s] \leftarrow \text{GREY}$ 
   $seen[s] \leftarrow time ++$ 
  for each  $v$  adjacent to  $s$  do
    if  $colour[v] = \text{WHITE}$  then
       $pred[v] \leftarrow s$ 
      rec_dfs_visit( $v$ )
    end if
  end for
   $colour[s] \leftarrow \text{BLACK}$ 
   $done[s] \leftarrow time ++$ 
end
```

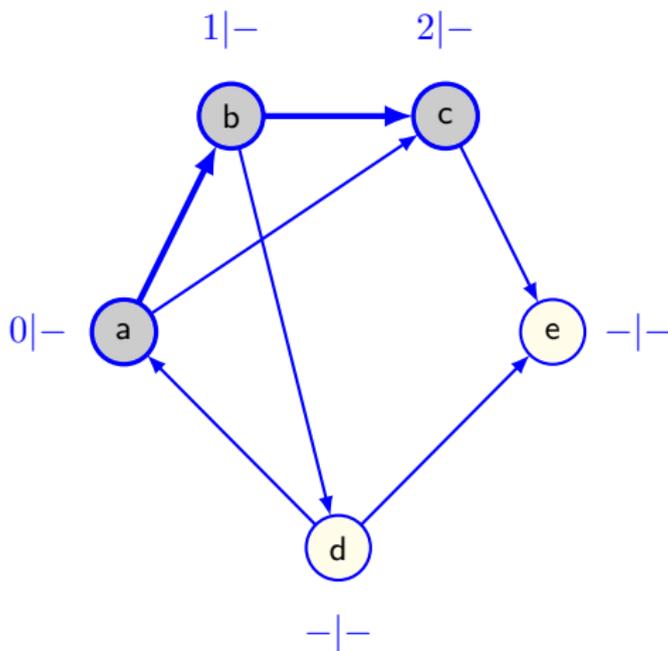
DFS: An Example ($seen[v] \mid done[v]$): $time = 0; 1$



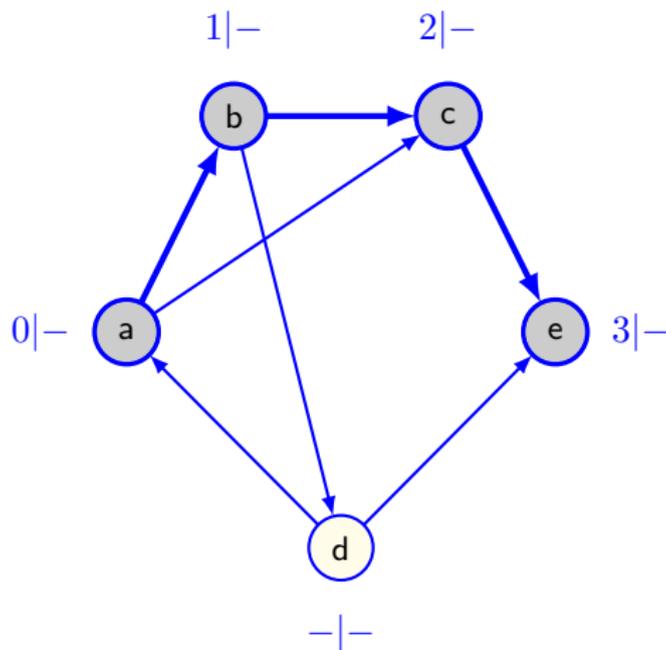
DFS: An Example ($seen[v] \mid done[v]$): $time = 1; 2$



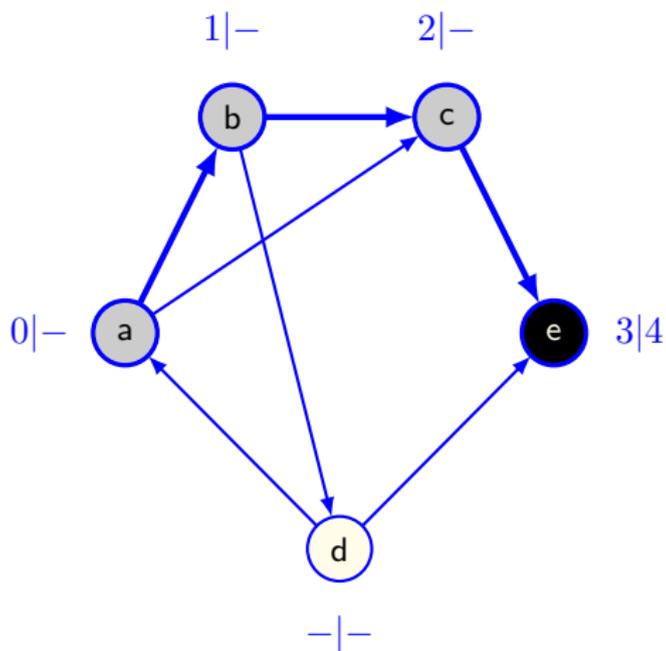
DFS: An Example ($seen[v] \mid done[v]$): $time = 2, 3$



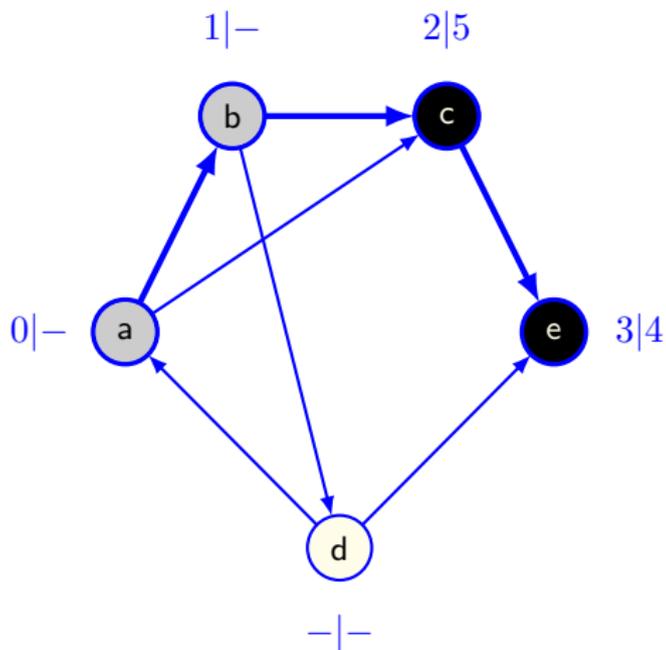
DFS: An Example ($seen[v] \mid done[v]$): $time = 3; 4$



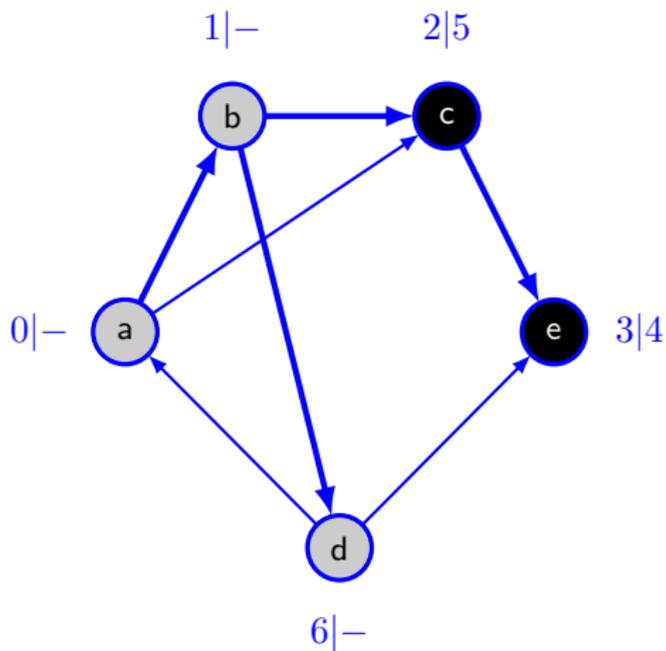
DFS: An Example ($seen[v] \mid done[v]$): $time = 4; 5$



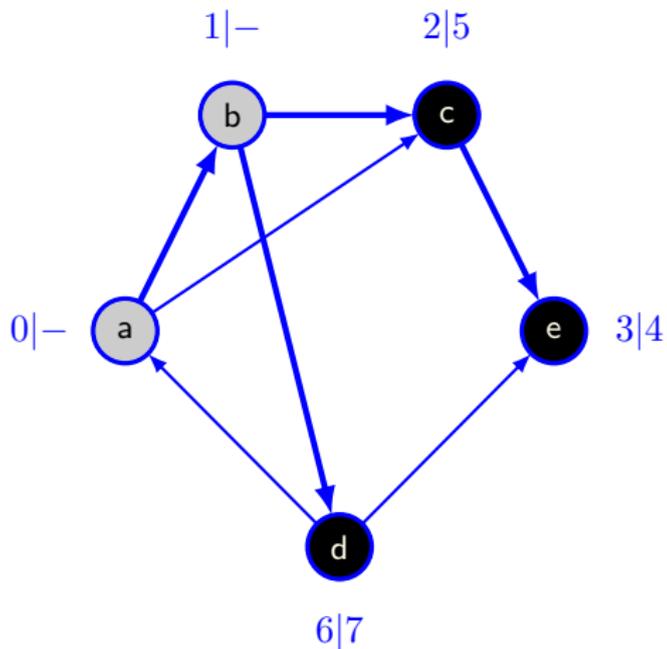
DFS: An Example ($seen[v] \mid done[v]: time = 5, 6$)



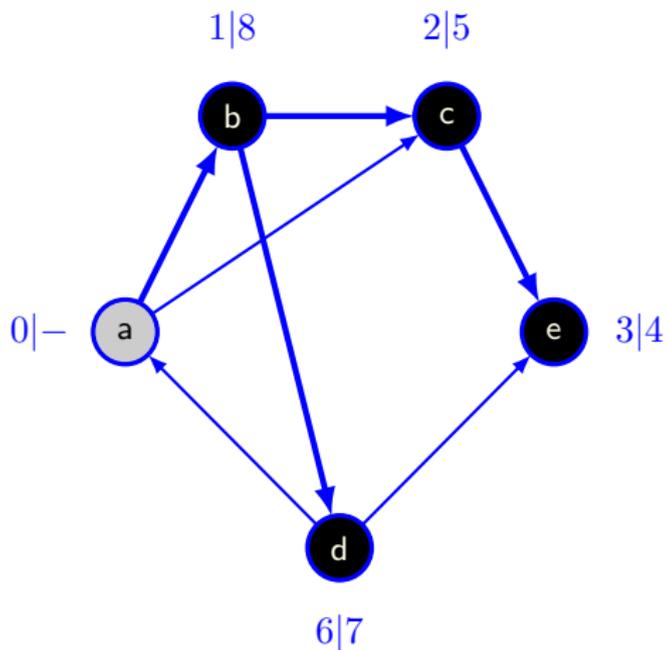
DFS: An Example ($seen[v] \mid done[v]$): $time = 6, 7$



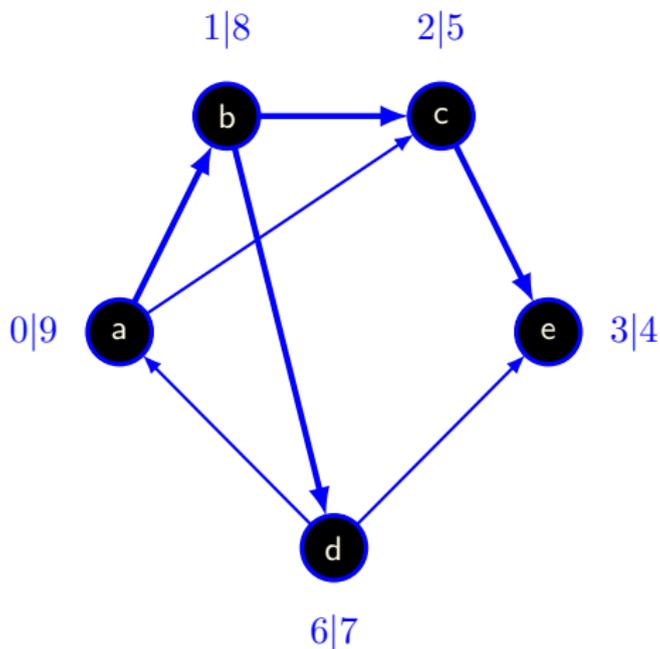
DFS: An Example ($seen[v] \mid done[v]$): $time = 7, 8$



DFS: An Example ($seen[v] \mid done[v]$): $time = 8, 9$



DFS: An Example ($seen[v] \mid done[v]$): $time = 9, 10$



Basic Properties of Depth-first Search

Next GREY node chosen \leftarrow the last one coloured GREY thus far.

- Data structure for this “last in, first out” order – a **stack**.

Each call to `dfs_visit(v)` terminates only when all nodes reachable from v via a path of WHITE nodes have been seen.

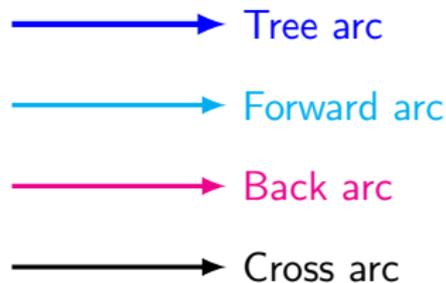
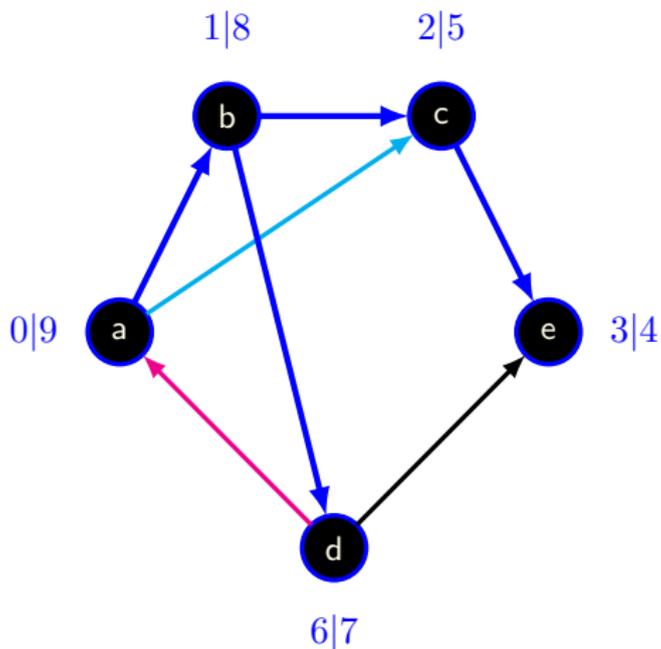
If (v, w) is an arc, then for a

- **tree** or **forward arc**: $seen[v] < seen[w] < done[w] < done[v]$
 - Example in Slide 52: $(a, b) : 0 < 1 < 8 < 9$; $(b, c) : 1 < 2 < 5 < 8$;
 $(a, c) : 0 < 2 < 5 < 9$;
- **back arc**: $seen[w] < seen[v] < done[v] < done[w]$:
 - Example in Slide 52: $(d, a) : 0 < 6 < 7 < 9$;
- **cross arc**: $seen[w] < done[w] < seen[v] < done[v]$.
 - Example in Slide 52: $(d, e) : 3 < 4 < 6 < 7$;

Hence, there are no cross edges on a graph.

Tree, Forward, Back, and Cross Arcs

(Example in Slide 52)



Using DFS to Determine Ancestors of a Tree

Theorem 5.5

Suppose that DFS on a digraph G results in a search forest F . Let $v, w \in V(G)$ and $seen[v] < seen[w]$.

- 1 If v is an ancestor of w in F , then

$$seen[v] < seen[w] < done[w] < done[v].$$

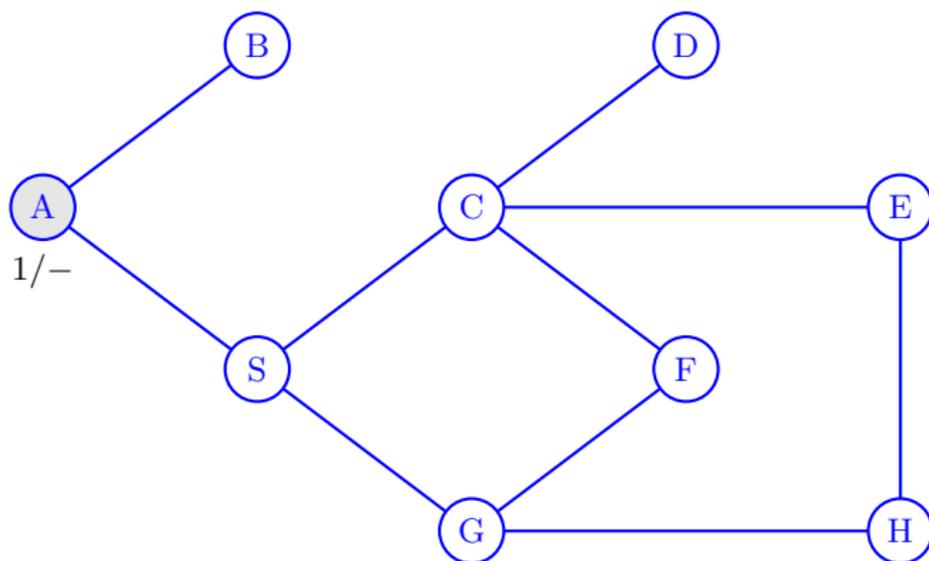
- 2 If v is not an ancestor of w in F , then

$$seen[v] < done[v] < seen[w] < done[w].$$

Proof.

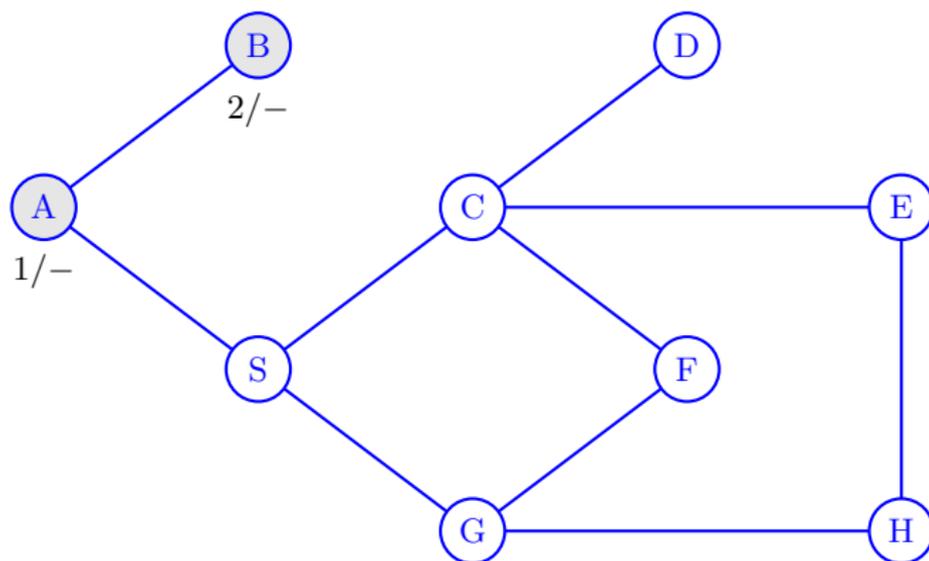
- 1 This part follows from the recursive nature of DFS.
- 2 If v is not an ancestor of w in F , then w is also not an ancestor v .
 - Thus v is in a subtree, which was completely explored before the subtree of w .



DFS: *seen/done*: step 1

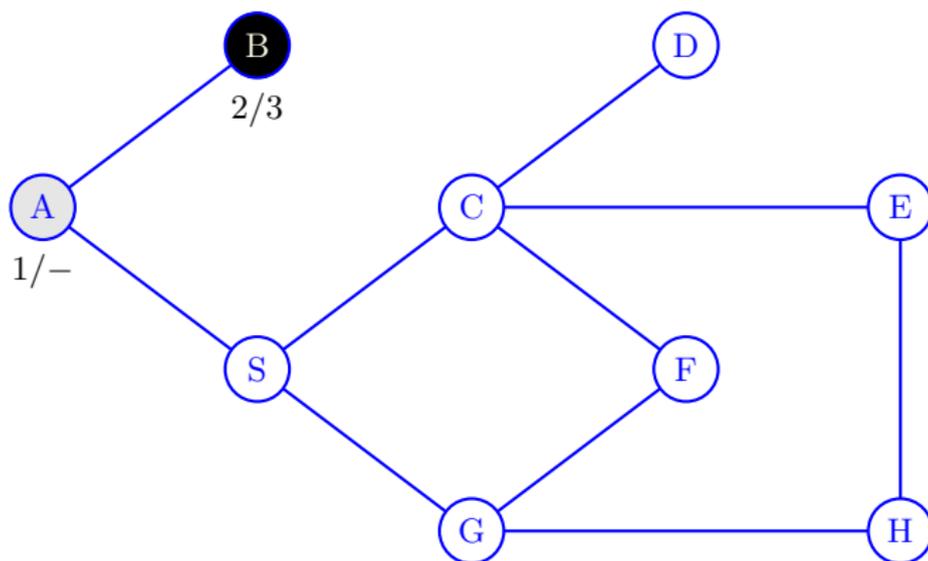
Preorder (WHITE to GREY): *seen* A
1

Postorder (GREY to BLACK) *done*

DFS: *seen/done*: step 2

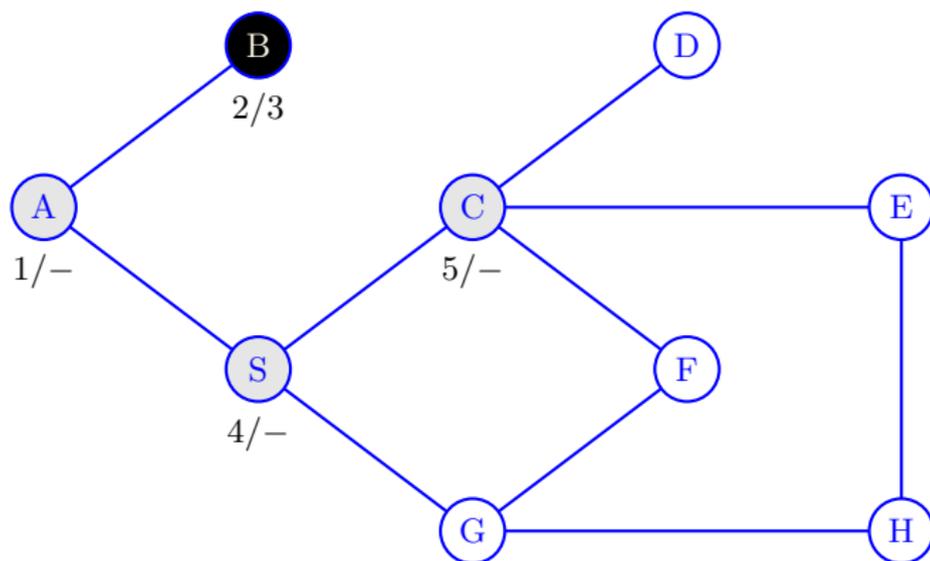
Preorder (WHITE to GREY): *seen* A B
 1 2

Postorder (GREY to BLACK) *done*

DFS: *seen/done*: step 3

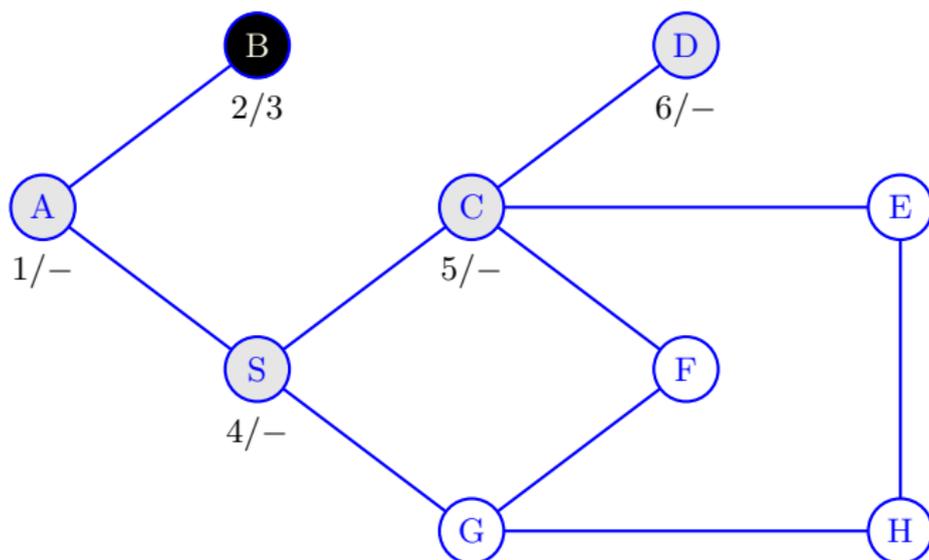
Preorder (WHITE to GREY): *seen* A B
 1 2

Postorder (GREY to BLACK) *done* B
 3

DFS: *seen/done*: step 5

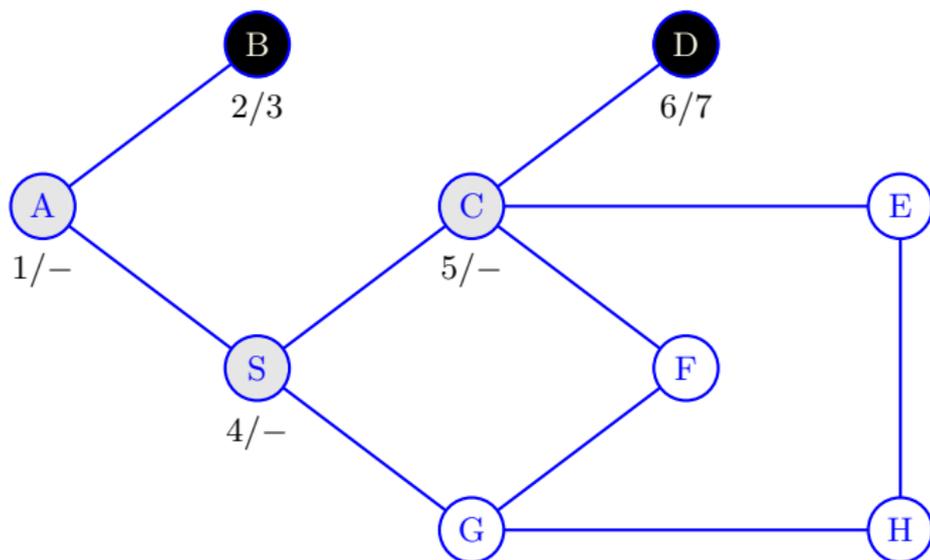
Preorder (WHITE to GREY): *seen* A B S C
 1 2 4 5

Postorder (GREY to BLACK) *done* B
 3

DFS: *seen/done*: step 6

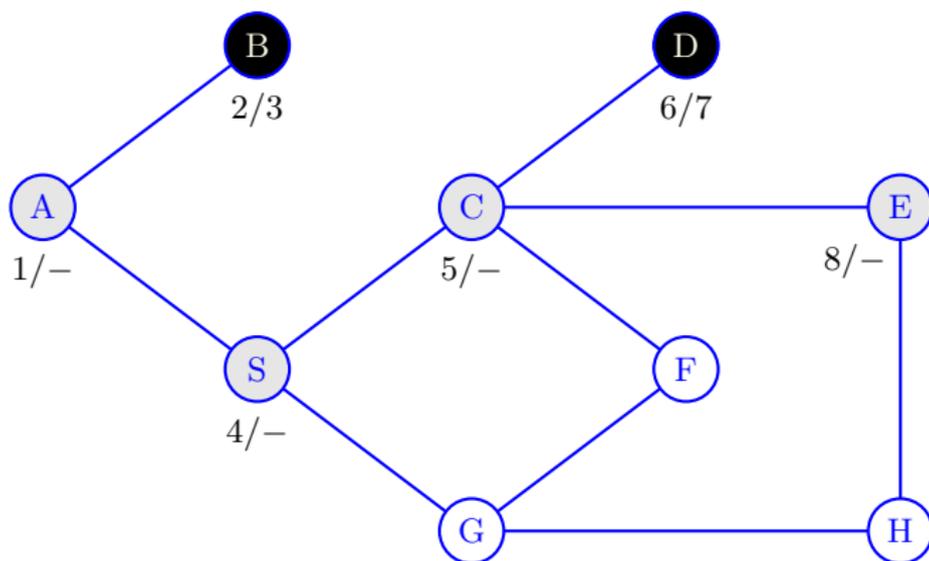
Preorder (WHITE to GREY): *seen* A B S C D
 1 2 4 5 6

Postorder (GREY to BLACK) *done* B
 3

DFS: *seen/done*: step 7

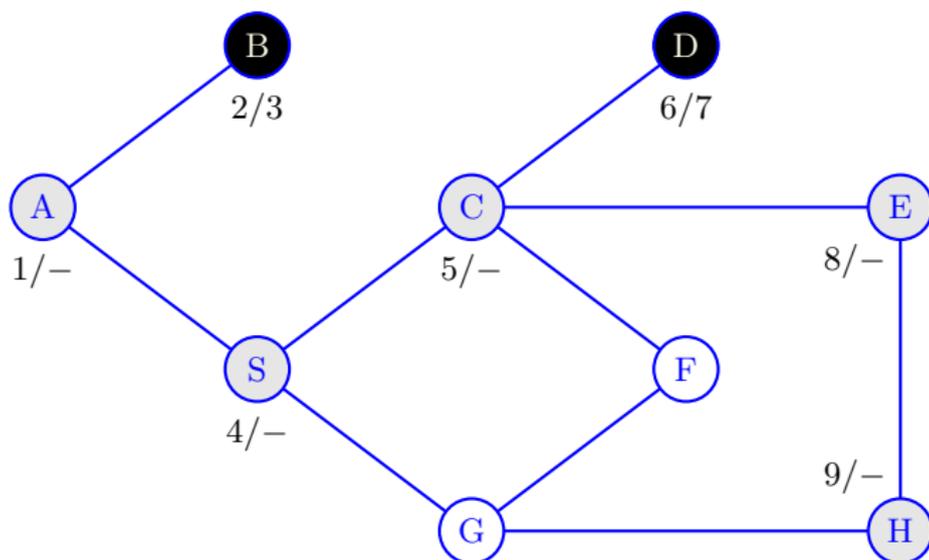
Preorder (WHITE to GREY): *seen* A B S C D
 1 2 4 5 6

Postorder (GREY to BLACK) *done* B D
 3 7

DFS: *seen/done*: step 8

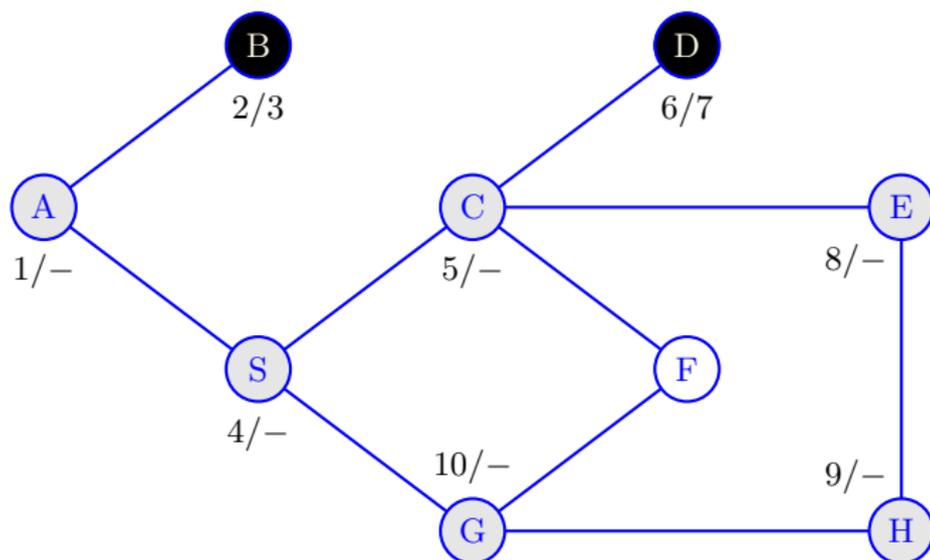
Preorder (WHITE to GREY): *seen* A B S C D E
 1 2 4 5 6 8

Postorder (GREY to BLACK) *done* B D
 3 7

DFS: *seen/done*: step 9

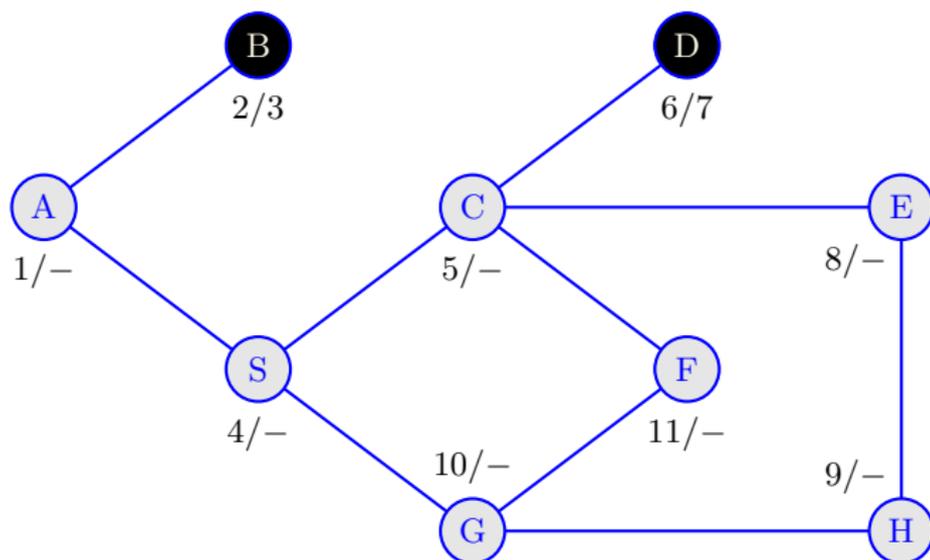
Preorder (WHITE to GREY): *seen* A B S C D E H
 1 2 4 5 6 8 9

Postorder (GREY to BLACK) *done* B D
 3 7

DFS: *seen/done*: step 10

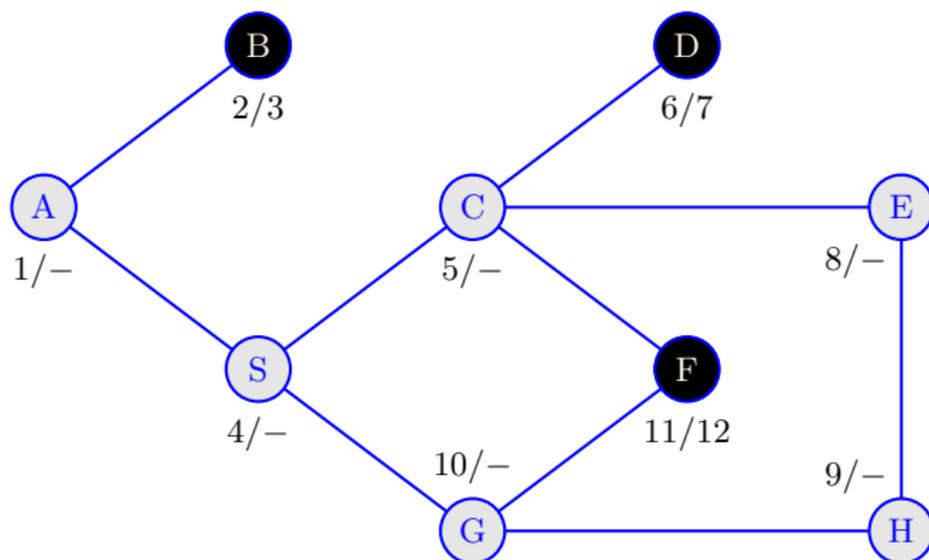
Preorder (WHITE to GREY): *seen* A B S C D E H G
 1 2 4 5 6 8 9 10

Postorder (GREY to BLACK) *done* B D
 3 7

DFS: *seen/done*: step 11

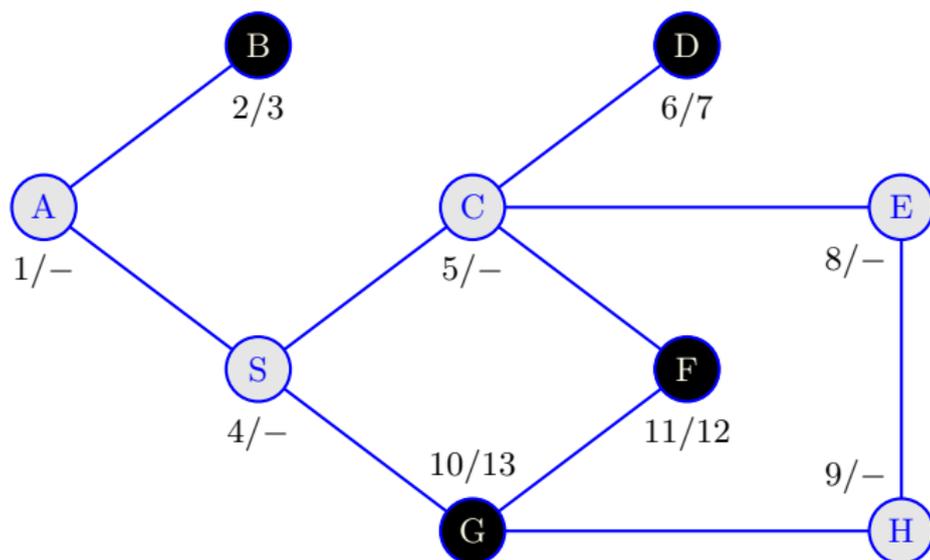
Preorder (WHITE to GREY): *seen* A B S C D E H G F
 1 2 4 5 6 8 9 10 11

Postorder (GREY to BLACK) *done* B D
 3 7

DFS: *seen/done*: step 12

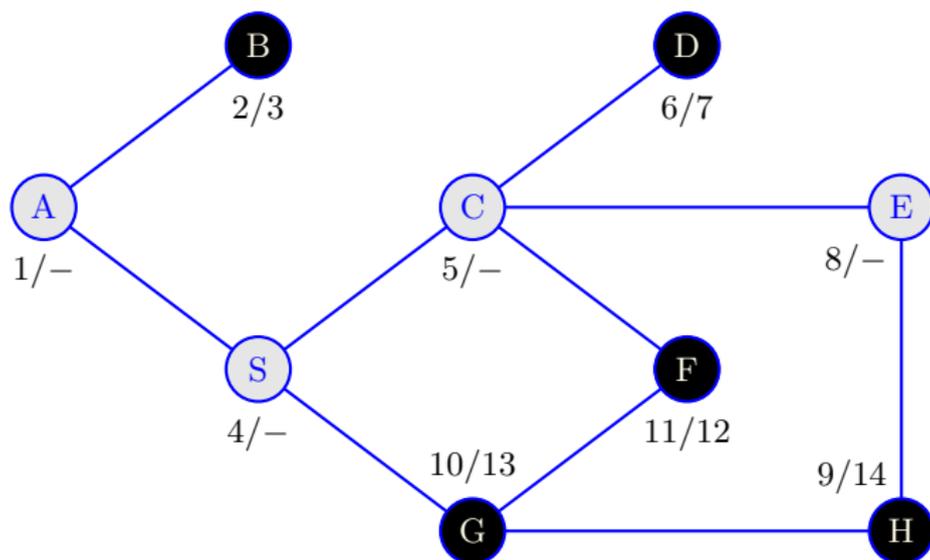
Preorder (WHITE to GREY): *seen* A B S C D E H G F
 1 2 4 5 6 8 9 10 11

Postorder (GREY to BLACK) *done* B D F
 3 7 12

DFS: *seen/done*: step 13

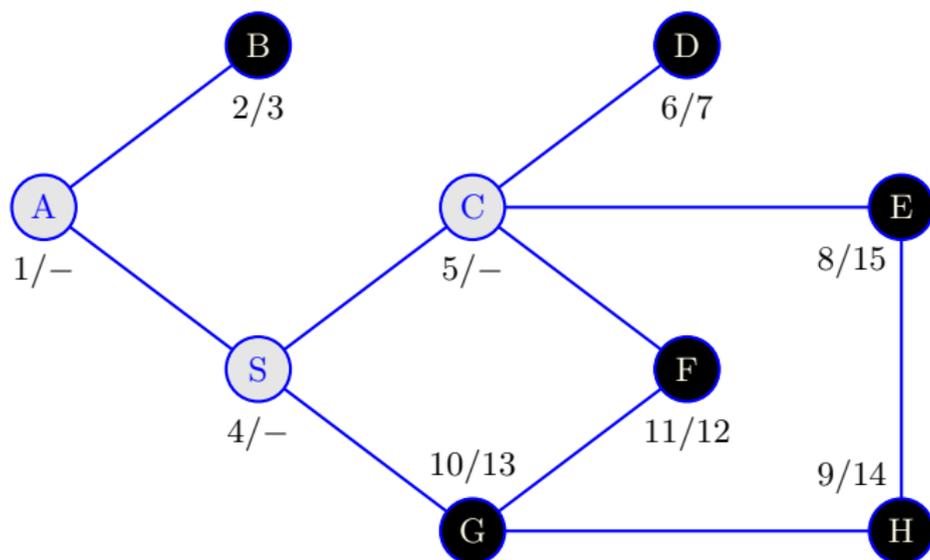
Preorder (WHITE to GREY): *seen* A B S C D E H G F
 1 2 4 5 6 8 9 10 11

Postorder (GREY to BLACK) *done* B D F G
 3 7 12 13

DFS: *seen/done*: step 14

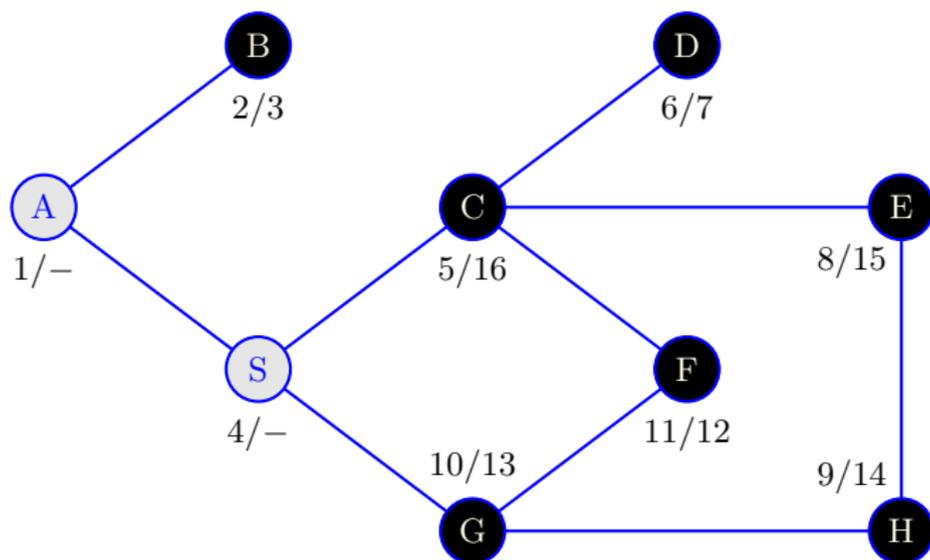
Preorder (WHITE to GREY): *seen* A B S C D E H G F
 1 2 4 5 6 8 9 10 11

Postorder (GREY to BLACK) *done* B D F G H
 3 7 12 13 14

DFS: *seen/done*: step 15

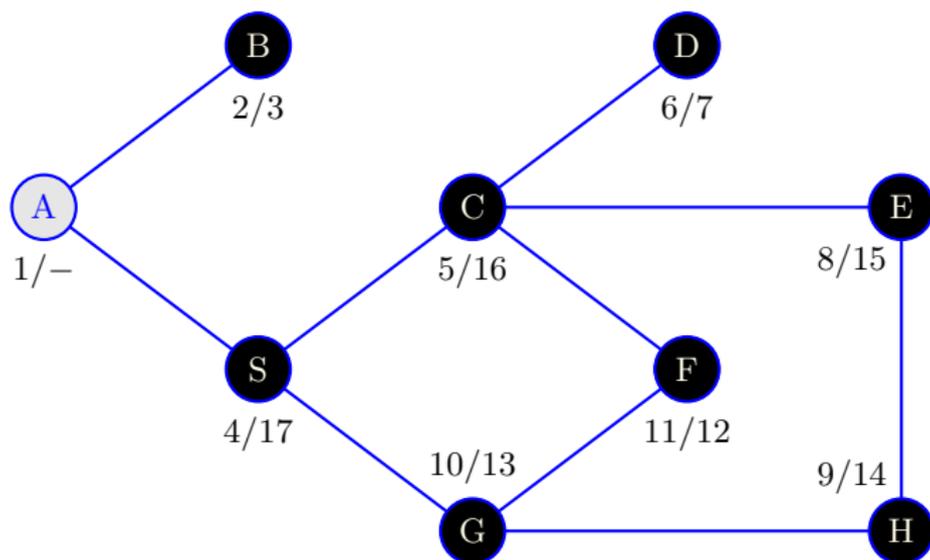
Preorder (WHITE to GREY): *seen* A B S C D E H G F
 1 2 4 5 6 8 9 10 11

Postorder (GREY to BLACK) *done* B D F G H E
 3 7 12 13 14 15

DFS: *seen/done*: step 16

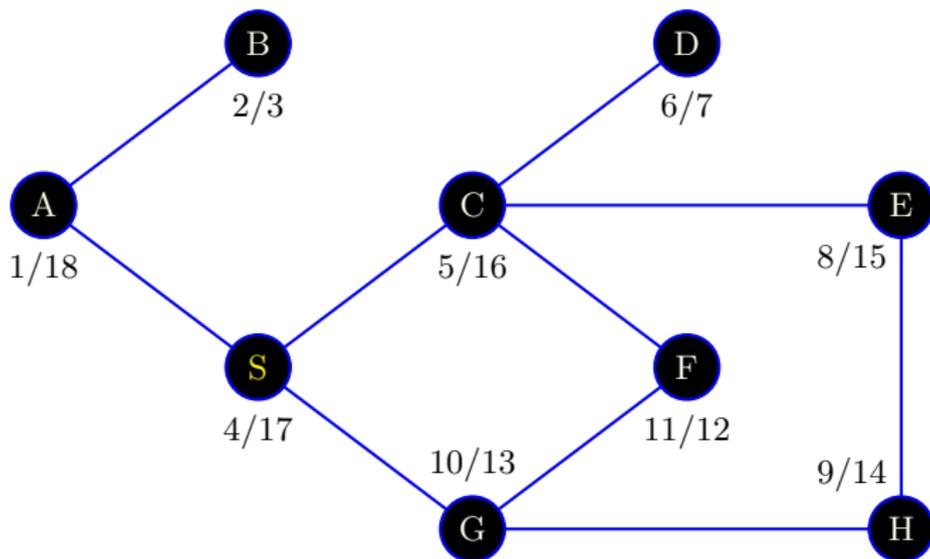
Preorder (WHITE to GREY): *seen* A B S C D E H G F
 1 2 4 5 6 8 9 10 11

Postorder (GREY to BLACK) *done* B D F G H E C
 3 7 12 13 14 15 16

DFS: *seen/done*: step 17

Preorder (WHITE to GREY): *seen* A B S C D E H G F
 1 2 4 5 6 8 9 10 11

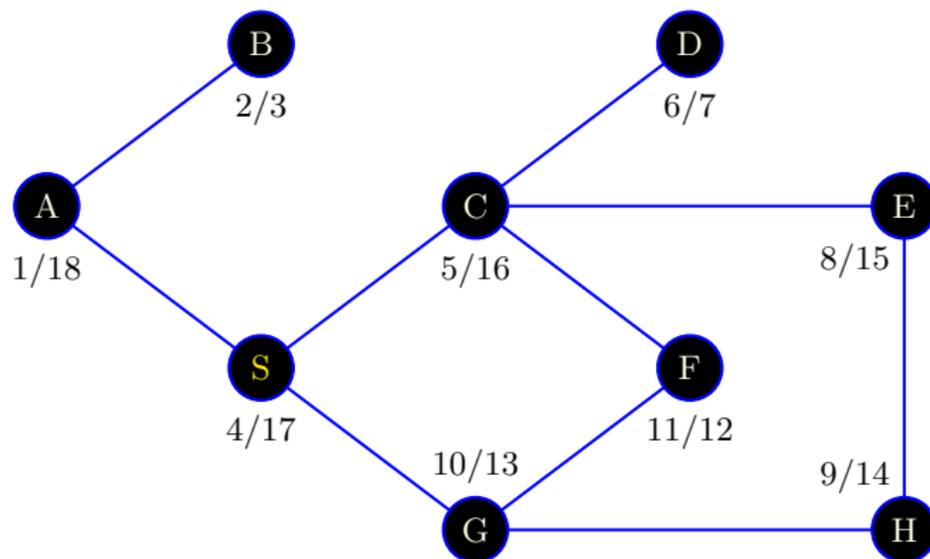
Postorder (GREY to BLACK) *done* B D F G H E C S
 3 7 12 13 14 15 16 17

DFS: *seen/done*: step 18

Preorder (WHITE to GREY): *seen* A B S C D E H G F
 1 2 4 5 6 8 9 10 11

Postorder (GREY to BLACK) *done* B D F G H E C S A
 3 7 12 13 14 15 16 17 18

Determining Ancestors of a Tree: Examples



$A \rightarrow B$: $\text{seen}[A] = 1 < \text{seen}[B] = 2 < \text{done}[B] = 3 < \text{done}[A] = 18$

$S \rightarrow H$: $\text{seen}[S] = 4 < \text{seen}[H] = 9 < \text{done}[H] = 14 < \text{done}[S] = 17$

$B \nrightarrow D$: $\text{seen}[B] = 2 < \text{done}[B] = 3 < \text{seen}[D] = 6 < \text{done}[D] = 7$

$D \nrightarrow G$: $\text{seen}[D] = 6 < \text{done}[D] = 7 < \text{seen}[G] = 10 < \text{done}[G] = 13$