

算法设计与分析(2026年春季学期)

Graph Basics

授课老师: 栗师

南京大学计算机学院

Outline

- 1 Graphs
- 2 Connectivity and Graph Traversal
 - Testing Bipartiteness
- 3 Topological Ordering
- 4 Bridges and 2-Edge-Connected Components
 - $O(n + m)$ -Time Algorithm to Find Bridges
 - Related Concept: Cut Vertices
- 5 Strong Connectivity in Directed Graphs
 - Tarjan's $O(n + m)$ -Time Algorithm for Finding SCCs

Examples of Graphs



Figure: Road Networks



Figure: Internet



Figure: Social Networks

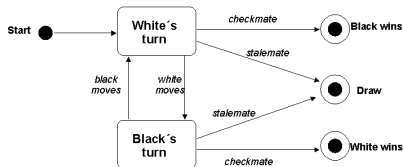
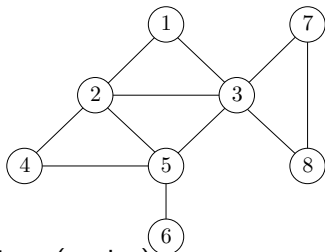


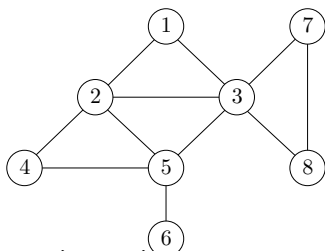
Figure: Transition Graphs

(Undirected) Graph $G = (V, E)$



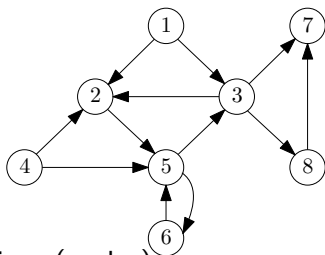
- V : set of vertices (nodes);
- E : pairwise relationships among V ;
 - (undirected) graphs: relationship is symmetric, E contains subsets of size 2

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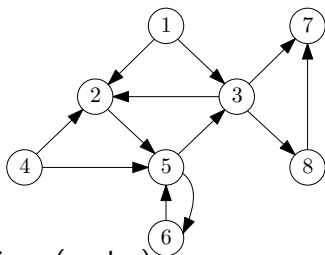
- V : set of vertices (nodes);
 - $V = \{1, 2, 3, 4, 5, 6, 7, 8\}$
- E : pairwise relationships among V ;
 - (undirected) graphs: relationship is symmetric, E contains subsets of size 2
 - $E = \{\{1, 2\}, \{1, 3\}, \{2, 3\}, \{2, 4\}, \{2, 5\}, \{3, 5\}, \{3, 7\}, \{3, 8\}, \{4, 5\}, \{5, 6\}, \{7, 8\}\}$

Directed Graph $G = (V, E)$



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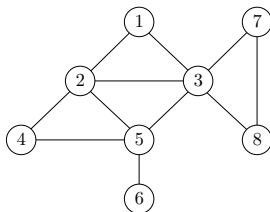
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Abuse of Notations

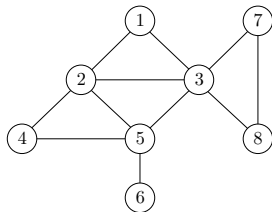
- For (undirected) graphs, we often use (i, j) to denote the set $\{i, j\}$.
- We call (i, j) an unordered pair; in this case $(i, j) = (j, i)$.



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- Social Network : Undirected
- Transition Graph : Directed
- Road Network : Directed or Undirected
- Internet : Directed or Undirected

Representation of Graphs

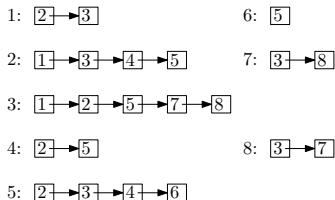
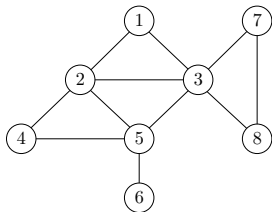


	1	2	3	4	5	6	7	8
1	0	1	1	0	0	0	0	0
2	1	0	1	1	1	0	0	0
3	1	1	0	0	1	0	1	1
4	0	1	0	0	1	0	0	0
5	0	1	1	1	0	1	0	0
6	0	0	0	0	1	0	0	0
7	0	0	1	0	0	0	0	1
8	0	0	1	0	0	0	1	0

- Adjacency matrix

- $n \times n$ matrix, $A[u, v] = 1$ if $(u, v) \in E$ and $A[u, v] = 0$ otherwise
- A is symmetric if graph is undirected

Representation of Graphs



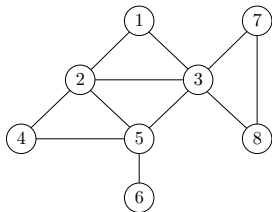
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- Linked lists

- For every vertex v , there is a linked list containing all **neighbours** of v .

Representation of Graphs



1: [2 3]

6: [5]

2: [1 3 4 5]

7: [3 8]

3: [1 2 5 7 8]

8: [3 7]

4: [2 5]

5: [2 3 4 6]

$d : (2, 4, 5, 2, 4, 1, 2, 2)$

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- Linked lists

- For every vertex v , there is a linked list containing all **neighbours** of v .
- If graph is static: store neighbors of all vertices in a length- $2m$ array, where the neighbors of any vertex are consecutive.

Comparison of Two Representations

- Assuming we are dealing with undirected graphs
- n : number of vertices
- m : number of edges, assuming $n - 1 \leq m \leq n(n - 1)/2$
- d_v : number of neighbors of v

	Matrix	Linked Lists
memory usage		
time to check $(u, v) \in E$		
time to list all neighbours of v		

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memory usage	$O(n^2)$	$O(m)$
time to check $(u, v) \in E$	$O(1)$	
time to list all neighbours of v		

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time to list all neighbours of v	$O(n)$	

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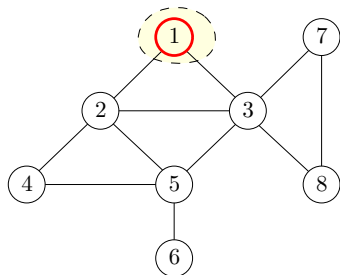
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 - Breadth-First Search (BFS)
 - Depth-First Search (DFS)

Breadth-First Search (BFS)

- Build layers $L_0, L_1, L_2, L_3, \dots$
- $L_0 = \{s\}$
- L_{j+1} contains all nodes that are not in $L_0 \cup L_1 \cup \dots \cup L_j$ and have an edge to a vertex in L_j

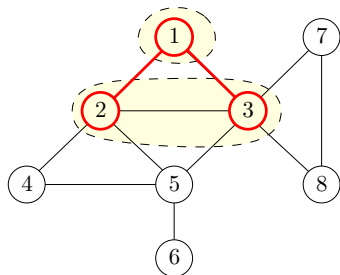
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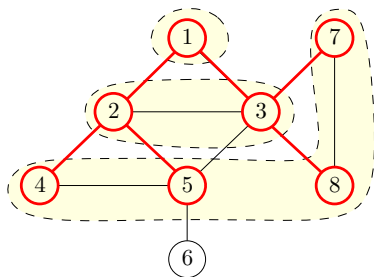
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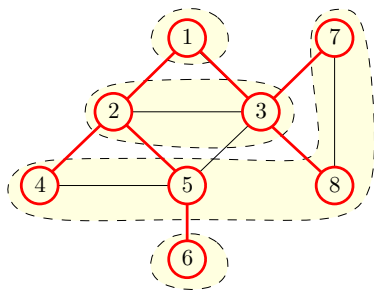
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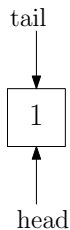
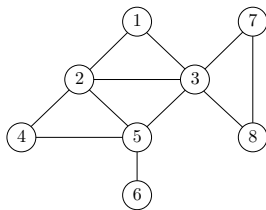
Implementing BFS using a Queue

BFS(s)

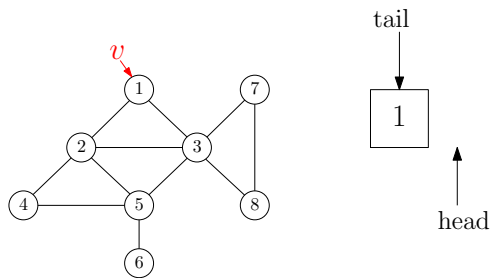
```
1:  $head \leftarrow 1, tail \leftarrow 1, queue[1] \leftarrow s$   
2: mark  $s$  as “visited” and all other vertices as “unvisited”  
3: while  $head \leq tail$  do  
4:    $v \leftarrow queue[head], head \leftarrow head + 1$   
5:   for all neighbours  $u$  of  $v$  do  
6:     if  $u$  is “unvisited” then  
7:        $tail \leftarrow tail + 1, queue[tail] = u$   
8:       mark  $u$  as “visited”
```

- Running time: $O(n + m)$.

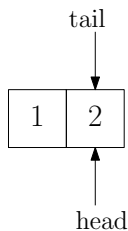
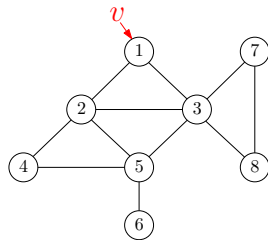
Example of BFS via Queue



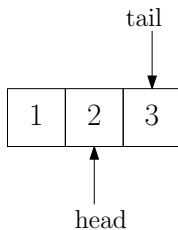
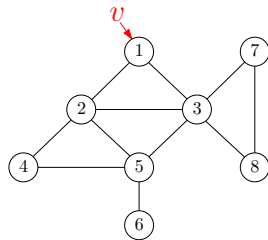
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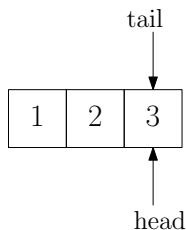
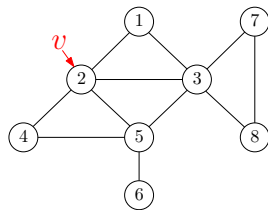
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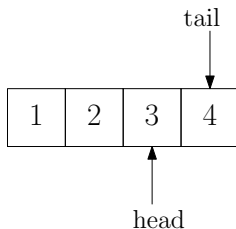
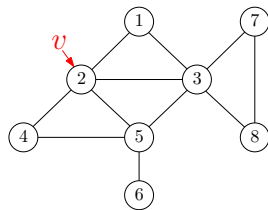
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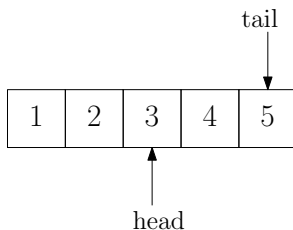
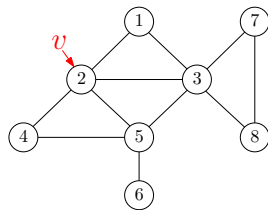
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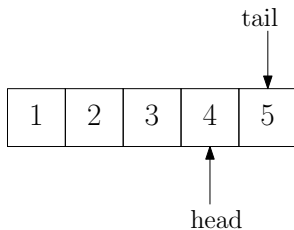
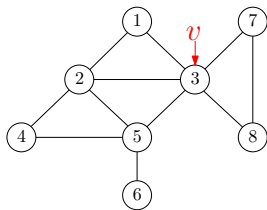
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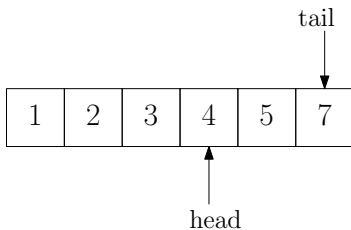
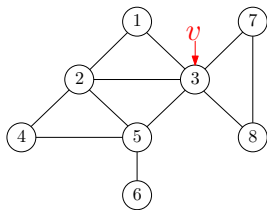
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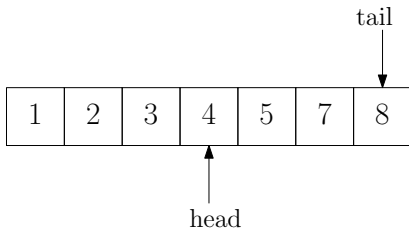
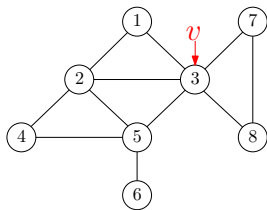
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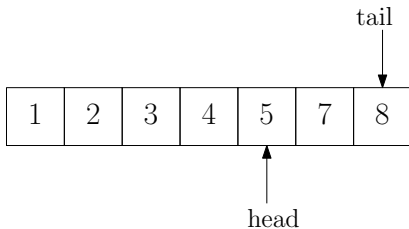
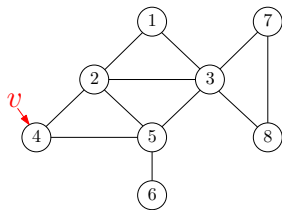
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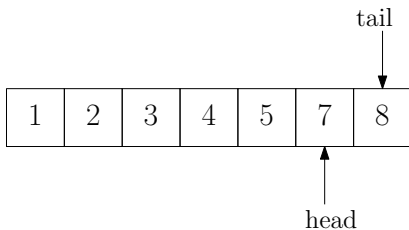
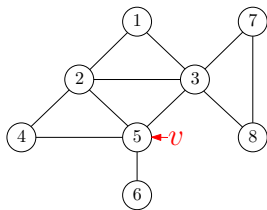
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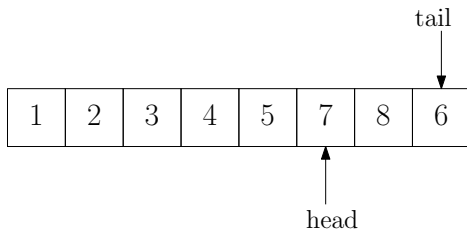
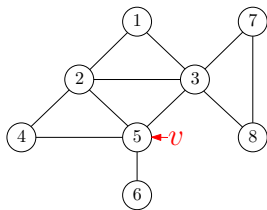
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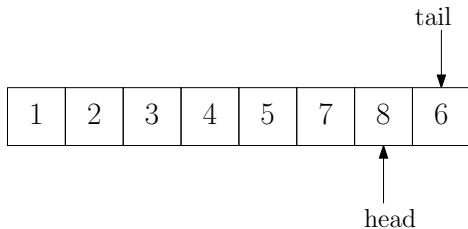
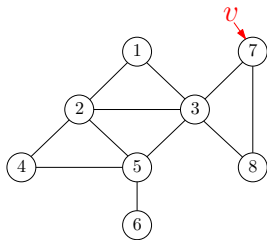
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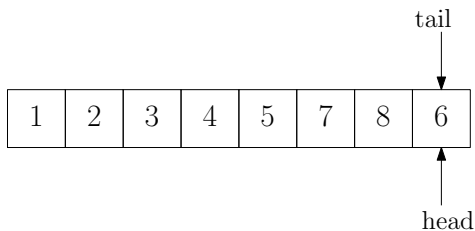
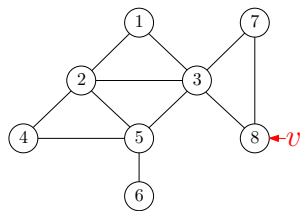
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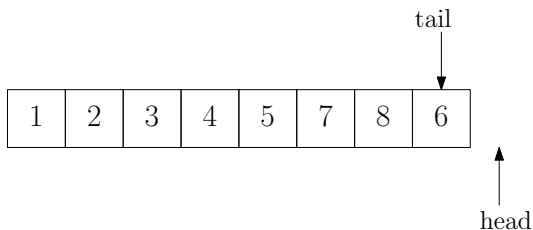
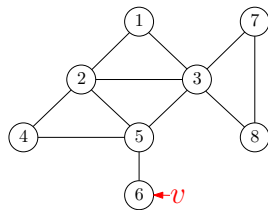
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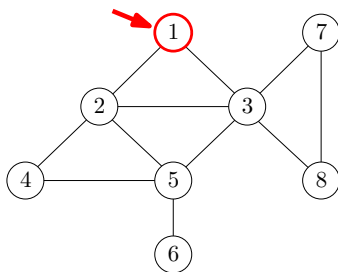


Depth-First Search (DFS)

- Starting from s
- Travel through the first edge leading out of the current vertex
- When reach an already-visited vertex (“dead-end”), go back
- Travel through the next edge
- If tried all edges leading out of the current vertex, go back

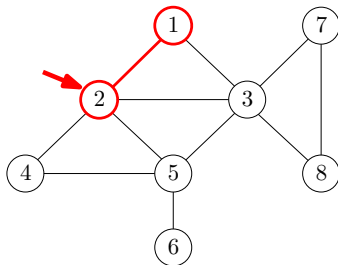
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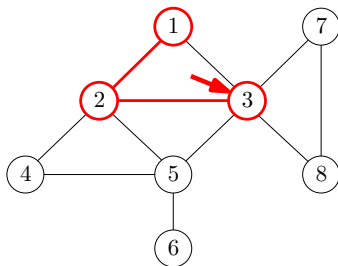
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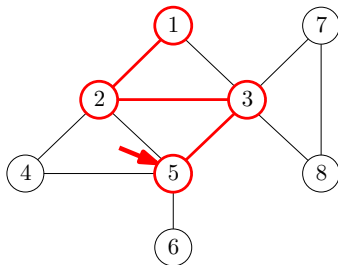
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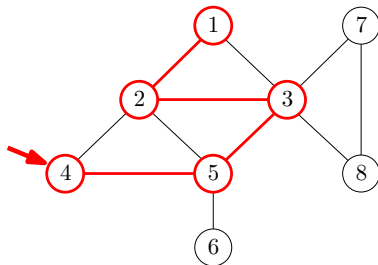
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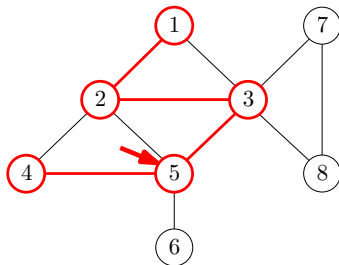
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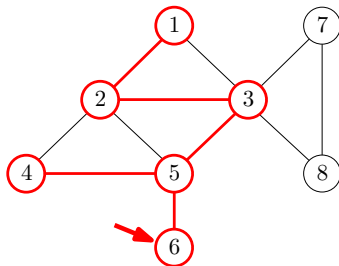
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- If tried all edges leading out of the current vertex, go back



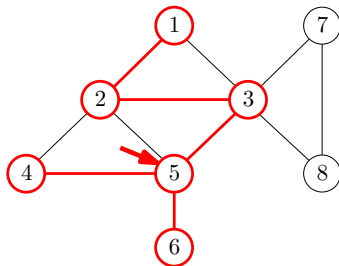
Depth-First Search (DFS)

- Starting from s
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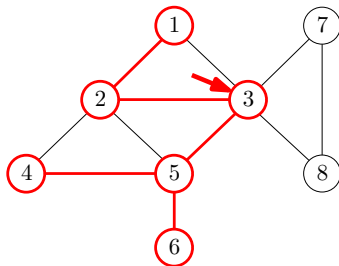
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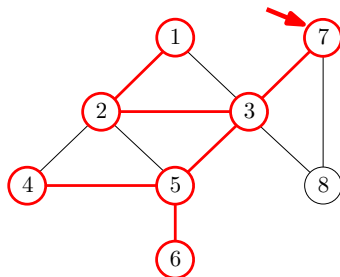
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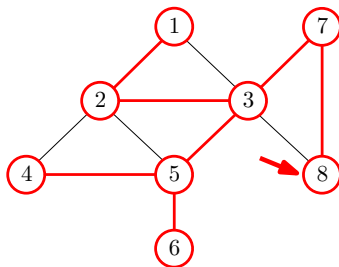
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Implementing DFS using Recursion

DFS(s)

- 1: mark all vertices as “unvisited”
- 2: recursive-DFS(s)

recursive-DFS(v)

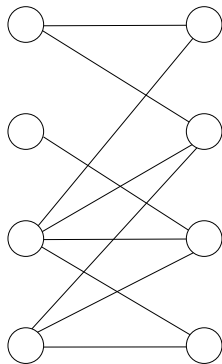
- 1: mark v as “visited”
- 2: **for** all neighbours u of v **do**
- 3: **if** u is unvisited **then** recursive-DFS(u)

Outline

- 1 Graphs
- 2 Connectivity and Graph Traversal
 - Testing Bipartiteness
- 3 Topological Ordering
- 4 Bridges and 2-Edge-Connected Components
 - $O(n + m)$ -Time Algorithm to Find Bridges
 - Related Concept: Cut Vertices
- 5 Strong Connectivity in Directed Graphs
 - Tarjan's $O(n + m)$ -Time Algorithm for Finding SCCs

Testing Bipartiteness: Applications of BFS

Def. A graph $G = (V, E)$ is a **bipartite graph** if there is a partition of V into two sets L and R such that for every edge $(u, v) \in E$, we have either $u \in L, v \in R$ or $v \in L, u \in R$.



Testing Bipartiteness

- Taking an arbitrary vertex $s \in V$

Testing Bipartiteness

- Taking an arbitrary vertex $s \in V$
- Assuming $s \in L$ w.l.o.g

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- ...

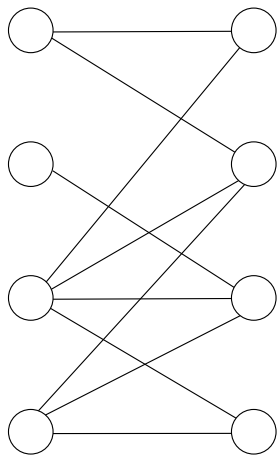
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- Assuming $s \in L$ w.l.o.g
- Neighbors of s must be in R
- Neighbors of neighbors of s must be in L
- ...
- Report “not a bipartite graph” if contradiction was found

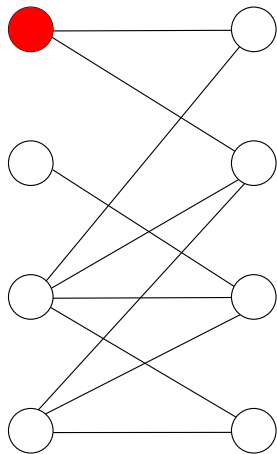
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- Assuming $s \in L$ w.l.o.g
- Neighbors of s must be in R
- Neighbors of neighbors of s must be in L
- ...
- Report “not a bipartite graph” if contradiction was found
- If G contains multiple connected components, repeat above algorithm for each component

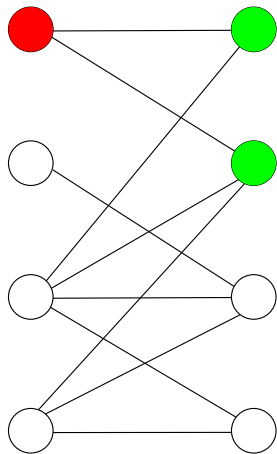
Test Bipartiteness



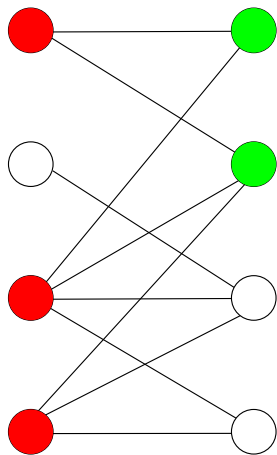
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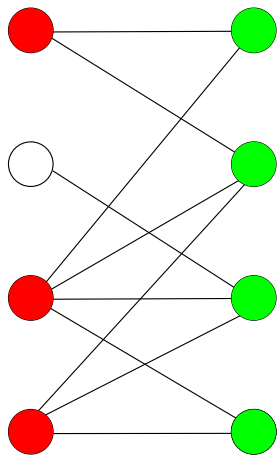
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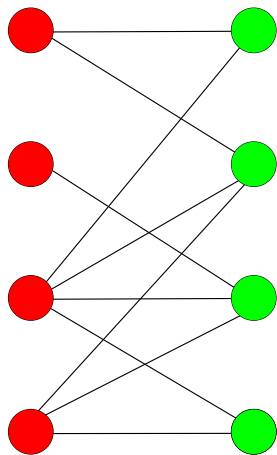
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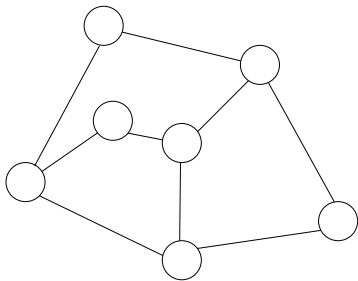
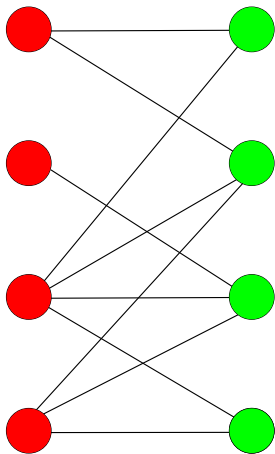
Test Bipartiteness



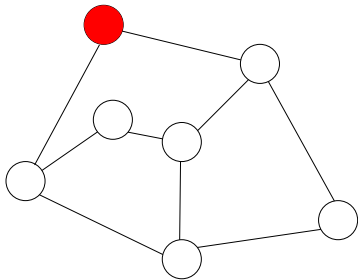
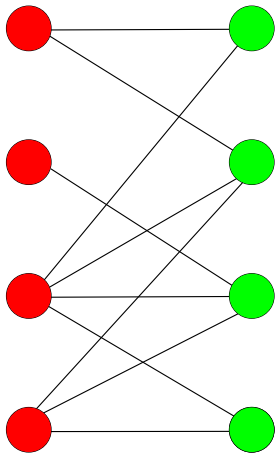
Test Bipartiteness



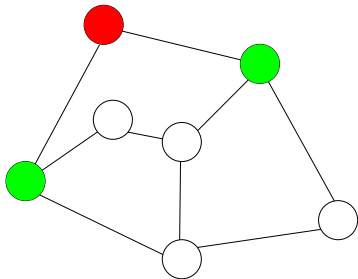
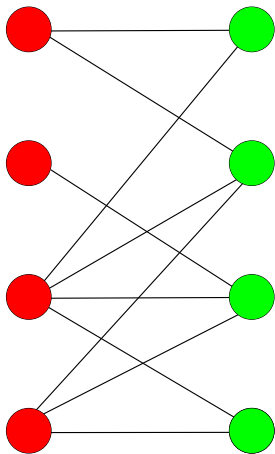
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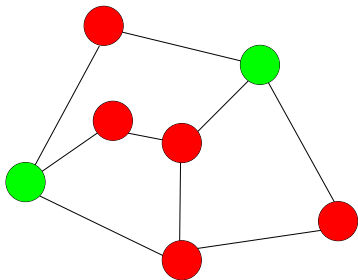
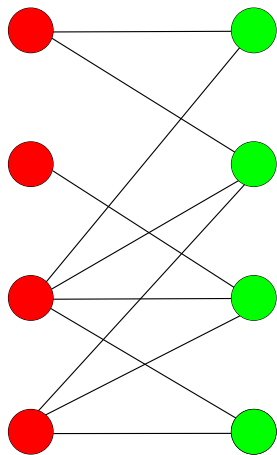
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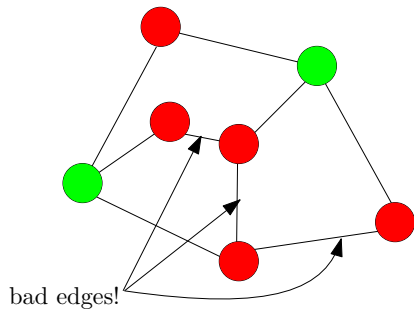
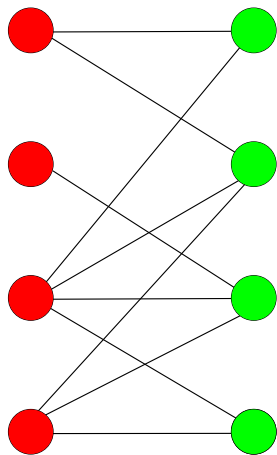
Test Bipartiteness



Test Bipartiteness



Test Bipartiteness



Testing Bipartiteness using BFS

BFS(s)

```
1:  $head \leftarrow 1, tail \leftarrow 1, queue[1] \leftarrow s$ 
2: mark  $s$  as “visited” and all other vertices as “unvisited”
3: while  $head \leq tail$  do
4:    $v \leftarrow queue[head], head \leftarrow head + 1$ 
5:   for all neighbours  $u$  of  $v$  do
6:     if  $u$  is “unvisited” then
7:        $tail \leftarrow tail + 1, queue[tail] = u$ 
8:       mark  $u$  as “visited”
```

Testing Bipartiteness using BFS

test-bipartiteness(s)

```
1:  $head \leftarrow 1, tail \leftarrow 1, queue[1] \leftarrow s$ 
2: mark  $s$  as “visited” and all other vertices as “unvisited”
3:  $color[s] \leftarrow 0$ 
4: while  $head \leq tail$  do
5:    $v \leftarrow queue[head], head \leftarrow head + 1$ 
6:   for all neighbours  $u$  of  $v$  do
7:     if  $u$  is “unvisited” then
8:        $tail \leftarrow tail + 1, queue[tail] = u$ 
9:       mark  $u$  as “visited”
10:       $color[u] \leftarrow 1 - color[v]$ 
11:    else if  $color[u] = color[v]$  then
12:      print(“ $G$  is not bipartite”) and exit
```

Testing Bipartiteness using BFS

```
1: mark all vertices as "unvisited"
2: for each vertex  $v \in V$  do
3:   if  $v$  is "unvisited" then
4:     test-bipartiteness( $v$ )
5: print("G is bipartite")
```

Testing Bipartiteness using BFS

```
1: mark all vertices as "unvisited"
2: for each vertex  $v \in V$  do
3:   if  $v$  is "unvisited" then
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Obs. Running time of algorithm = $O(n + m)$

Outline

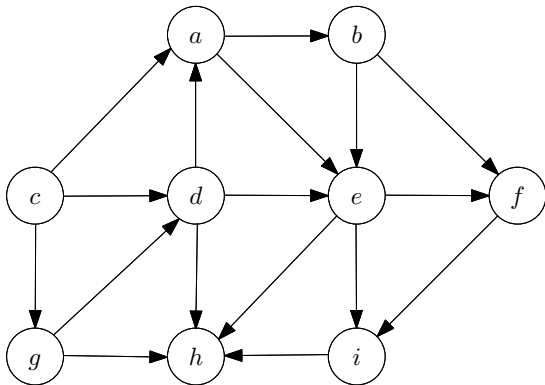
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Topological Ordering Problem

Input: a directed acyclic graph (DAG) $G = (V, E)$

Output: 1-to-1 function $\pi : V \rightarrow \{1, 2, 3 \dots, n\}$, so that

- if $(u, v) \in E$ then $\pi(u) < \pi(v)$

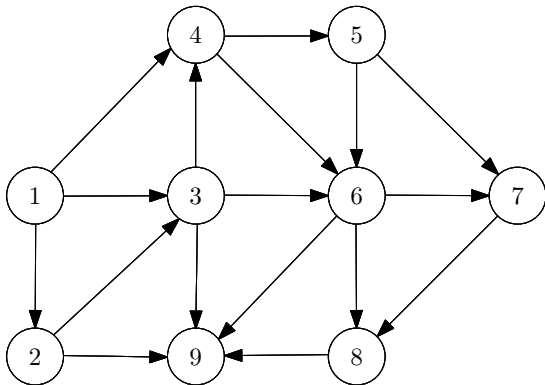


Topological Ordering Problem

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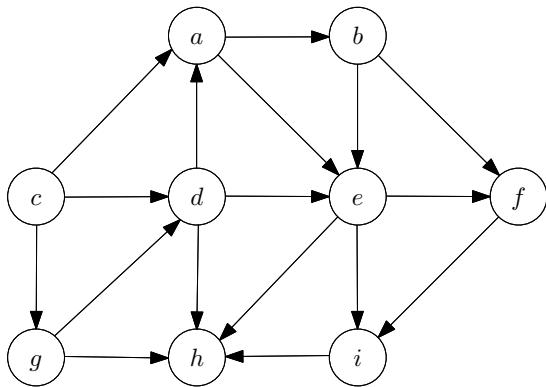
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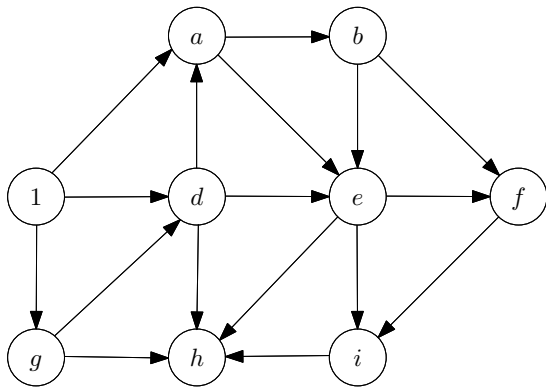
Topological Ordering

- Algorithm: each time take a vertex without incoming edges, then remove the vertex and all its outgoing edges.



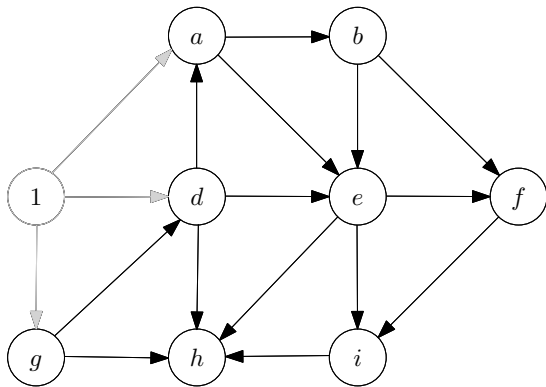
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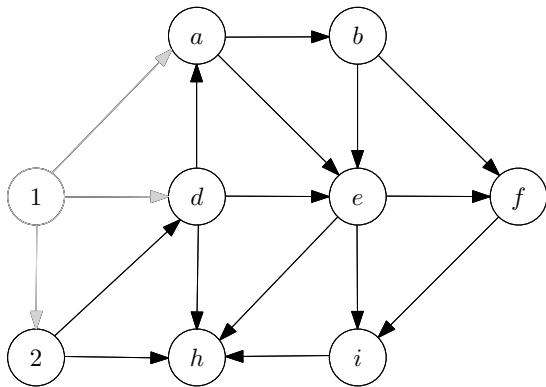
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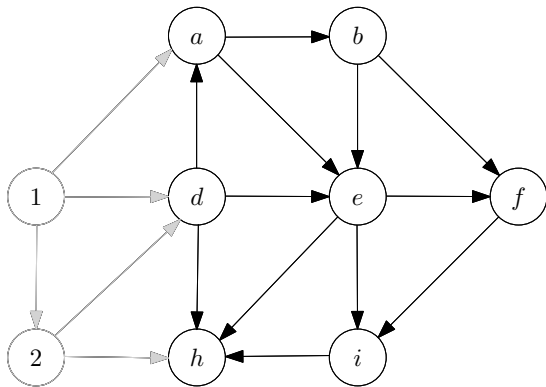
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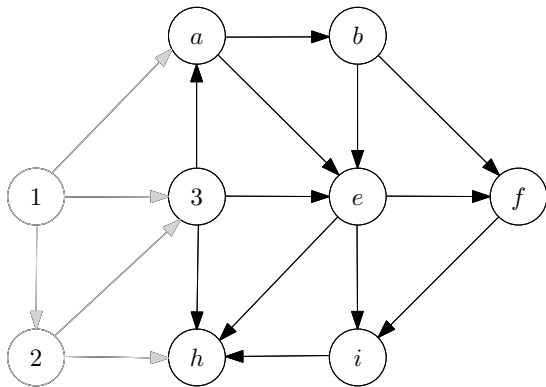
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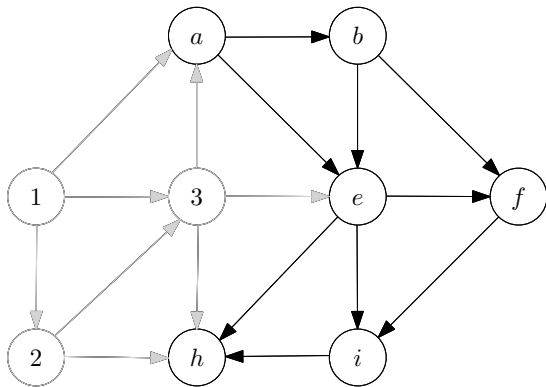
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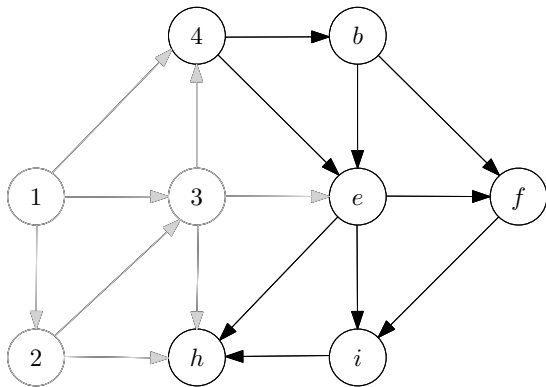
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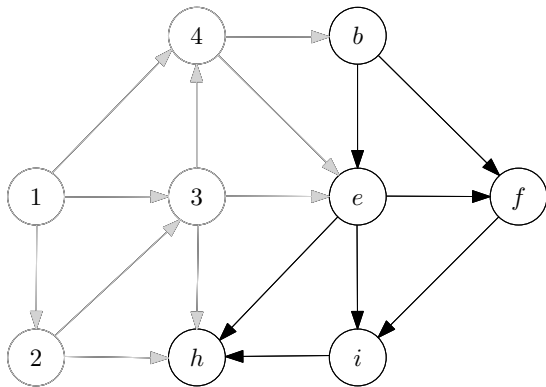
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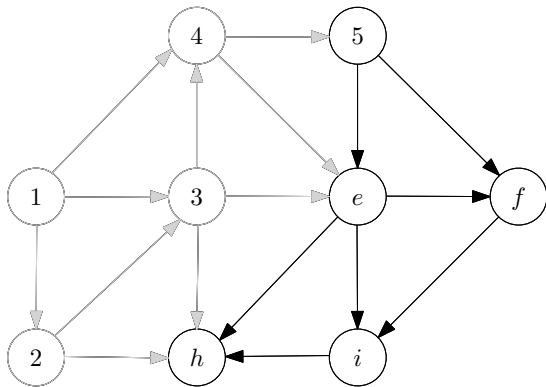
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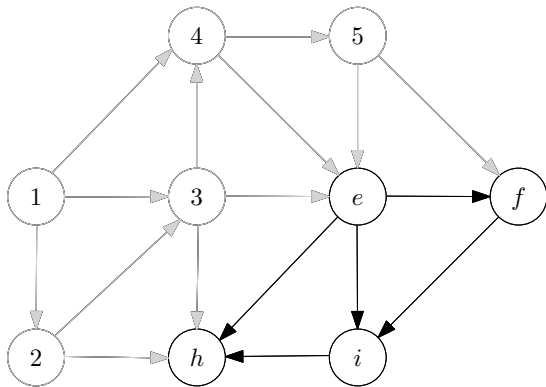
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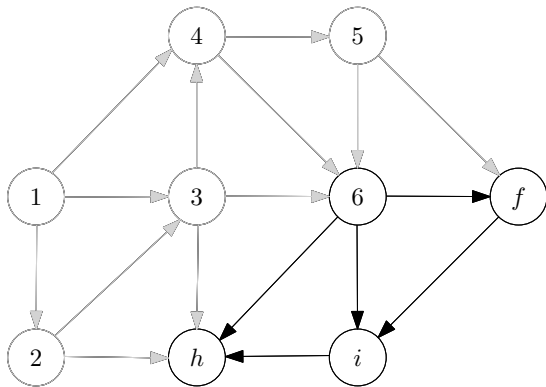
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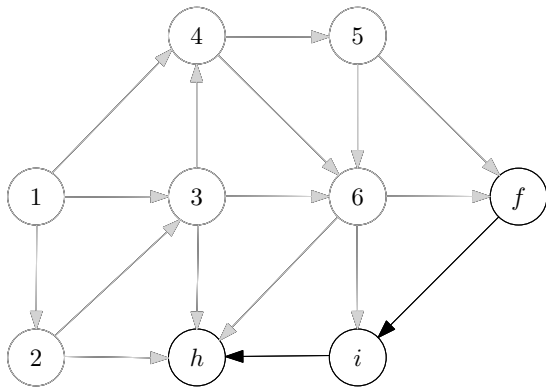
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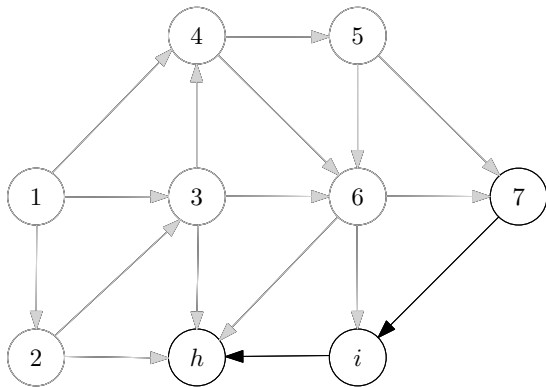
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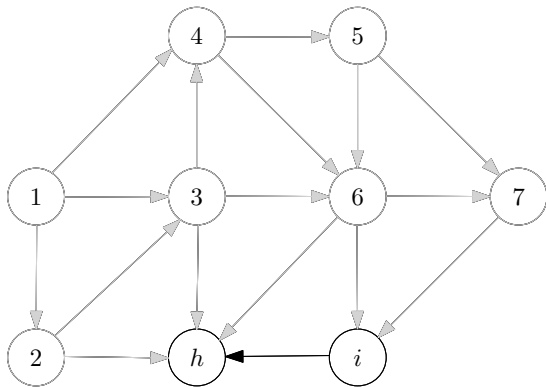
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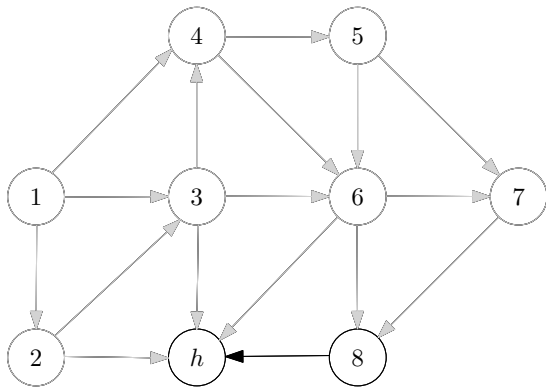
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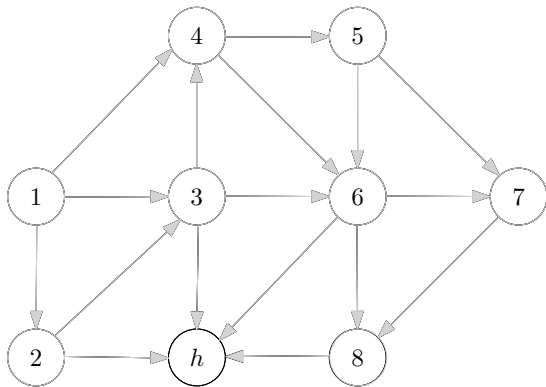
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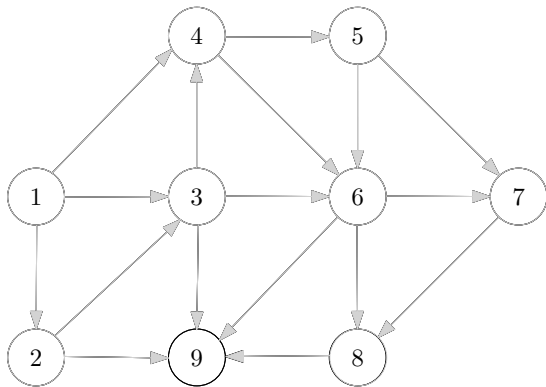
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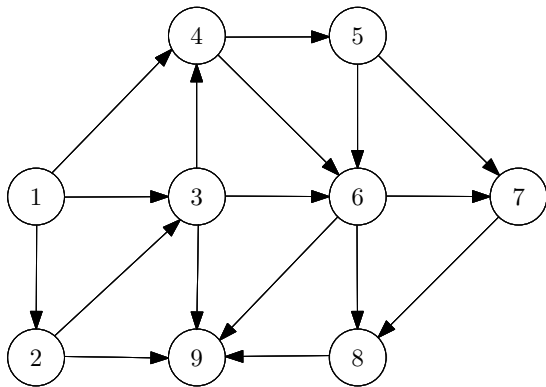
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Q: How to make the algorithm as efficient as possible?

Topological Ordering

- Algorithm: each time take a vertex without incoming edges, then remove the vertex and all its outgoing edges.

Q: How to make the algorithm as efficient as possible?

A:

- Use linked-lists of outgoing edges
- Maintain the in-degree d_v of vertices
- Maintain a queue (or stack) of vertices v with $d_v = 0$

topological-sort(G)

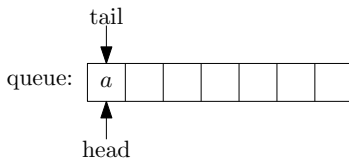
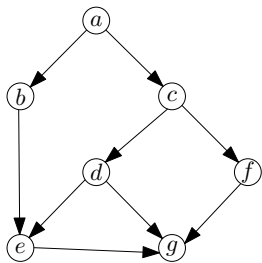
```
1: let  $d_v \leftarrow 0$  for every  $v \in V$ 
2: for every  $v \in V$  do
3:   for every  $u$  such that  $(v, u) \in E$  do
4:      $d_u \leftarrow d_u + 1$ 
5:  $S \leftarrow \{v : d_v = 0\}, i \leftarrow 0$ 
6: while  $S \neq \emptyset$  do
7:    $v \leftarrow$  arbitrary vertex in  $S, S \leftarrow S \setminus \{v\}$ 
8:    $i \leftarrow i + 1, \pi(v) \leftarrow i$ 
9:   for every  $u$  such that  $(v, u) \in E$  do
10:     $d_u \leftarrow d_u - 1$ 
11:    if  $d_u = 0$  then add  $u$  to  $S$ 
12: if  $i < n$  then output “not a DAG”
```

- S can be represented using a queue or a stack
- Running time = $O(n + m)$

S as a Queue or a Stack

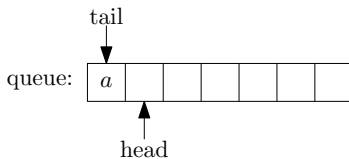
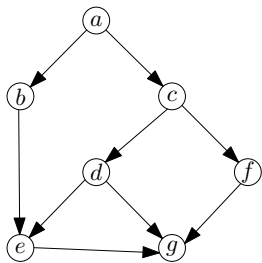
DS	Queue	Stack
Initialization	$head \leftarrow 1, tail \leftarrow 0$	$top \leftarrow 0$
Non-Empty?	$head \leq tail$	$top > 0$
Add(v)	$tail \leftarrow tail + 1$ $S[tail] \leftarrow v$	$top \leftarrow top + 1$ $S[top] \leftarrow v$
Retrieve v	$v \leftarrow S[head]$ $head \leftarrow head + 1$	$v \leftarrow S[top]$ $top \leftarrow top - 1$

Example



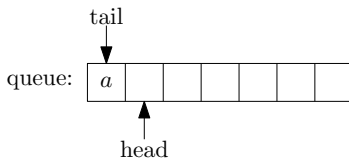
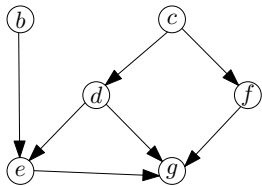
	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	1	1	1	2	1	3

Example



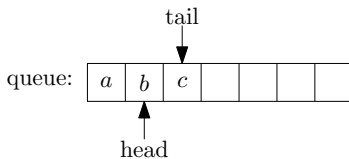
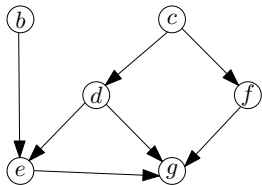
	a	b	c	d	e	f	g
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Example



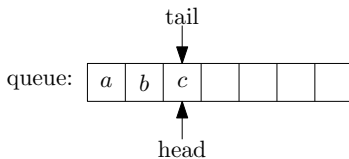
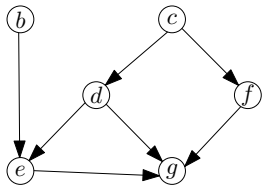
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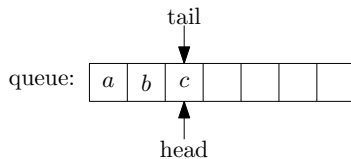
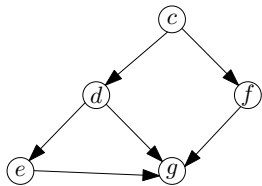
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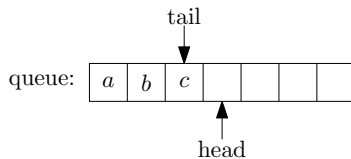
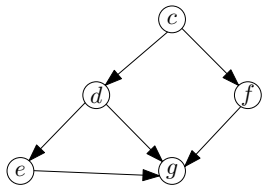
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degree	0	0	0	1	2	1	3

Example



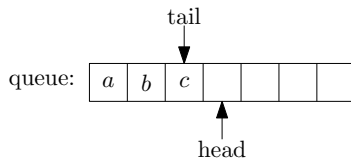
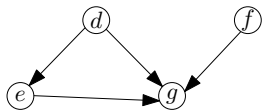
	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
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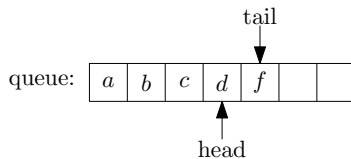
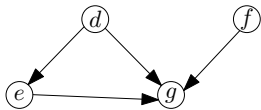
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Example



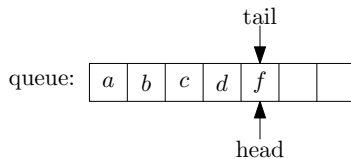
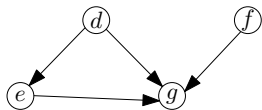
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degree	0	0	0	0	1	0	3

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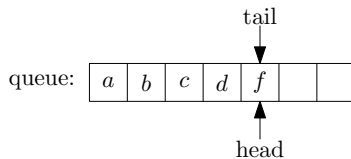
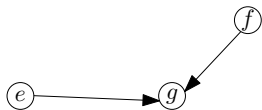
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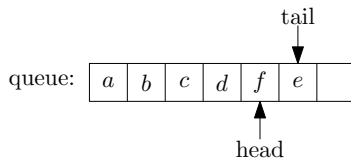
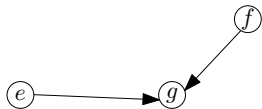
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Example



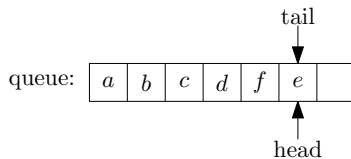
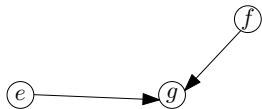
	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	2

Example



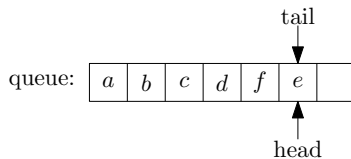
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Example



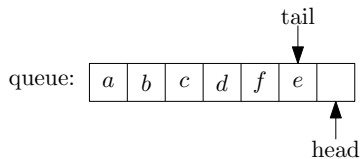
	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	2

Example



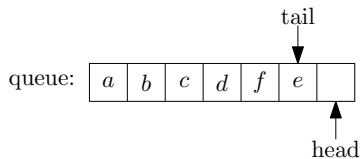
	a	b	c	d	e	f	g
degree	0	0	0	0	0	0	1

Example



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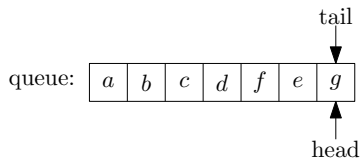
Example



⑨

	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	0

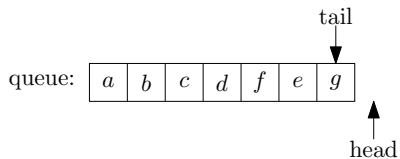
Example



⑨

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degree	0	0	0	0	0	0	0

Example



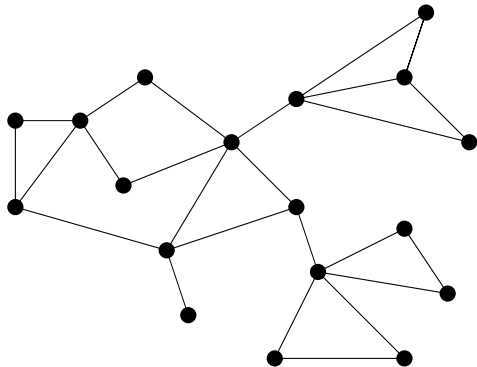
⑨

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Outline

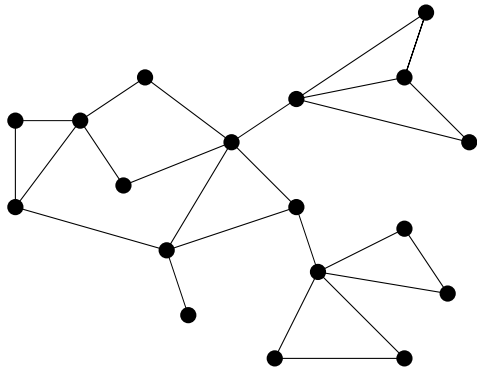
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Def. Given $G = (V, E)$, $e \in E$ is called a **bridge** if the removal of e from G will increase its number of connected components.



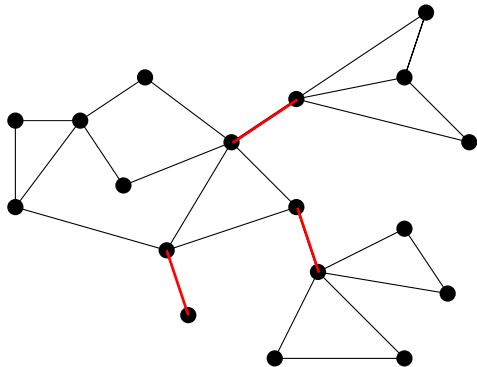
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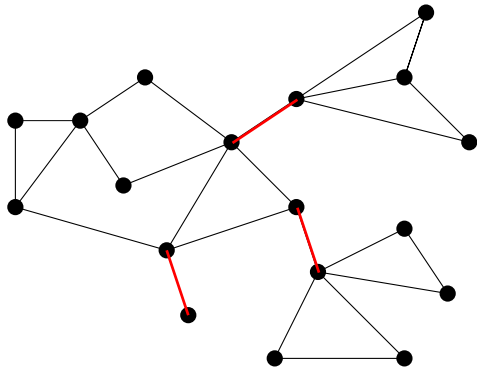
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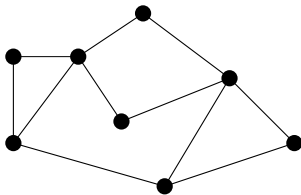
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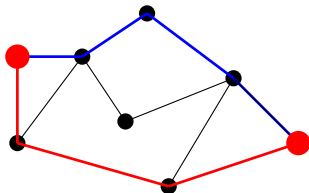
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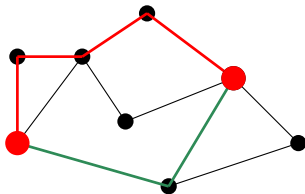
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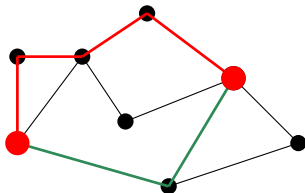
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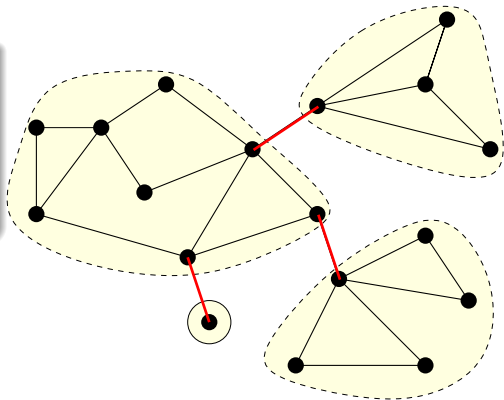


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Lemma Let B be the set of bridges in a graph $G = (V, E)$. Then, every connected component in $(V, E \setminus B)$ is 2-edge-connected. Every such component is called a **2-edge-connected component** of G .

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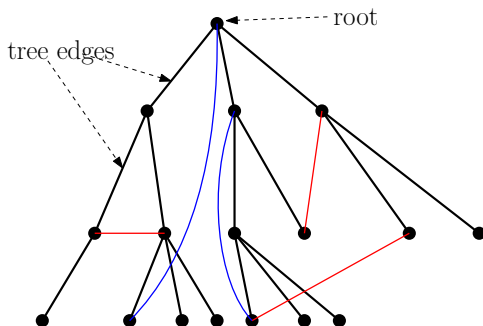
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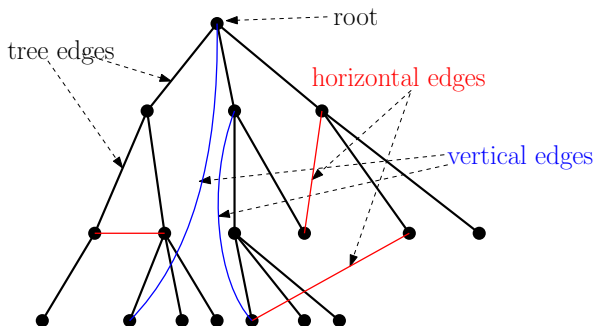
Vertical and Horizontal Edges

- $G = (V, E)$: connected graph
- $T = (V, E_T)$: rooted spanning tree of G



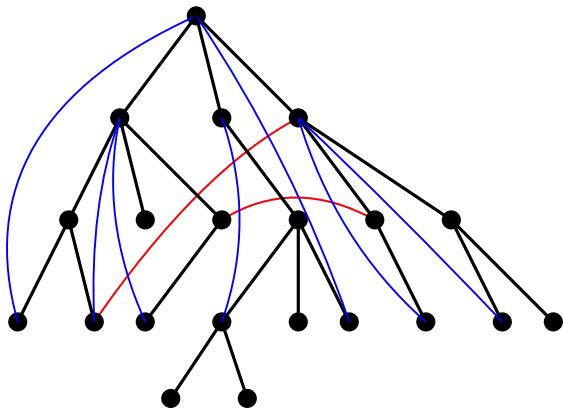
Vertical and Horizontal Edges

- $G = (V, E)$: connected graph
- $T = (V, E_T)$: rooted spanning tree of G
- $(u, v) \in E \setminus E_T$ is
 - **vertical** if one of u and v is an ancestor of the other in T ,
 - **horizontal** otherwise.



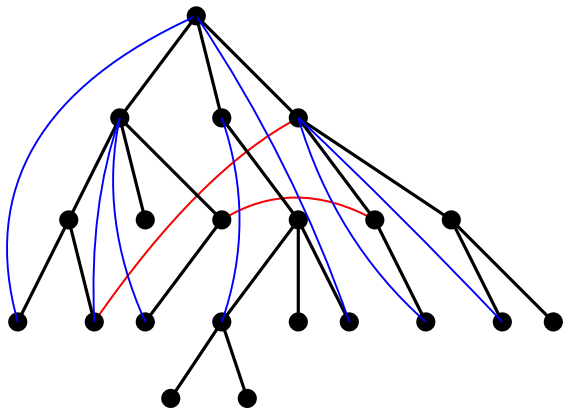
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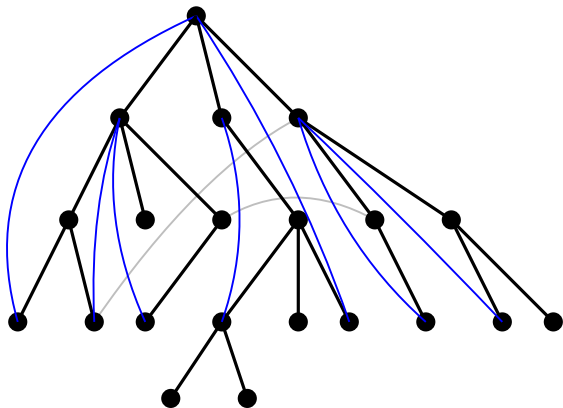
T : a DFS tree for G



Q: Can there be a horizontal edges (u, v) w.r.t T ?

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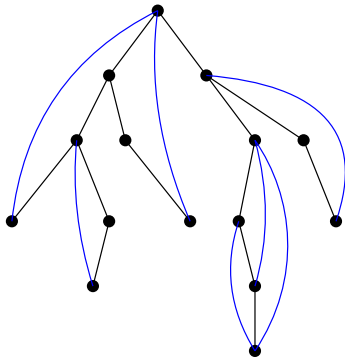
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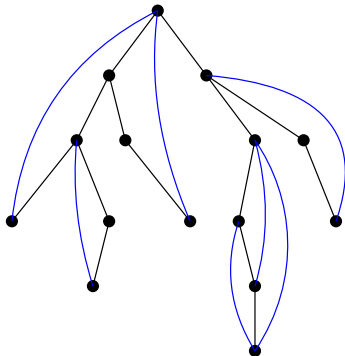
Q: Can there be a horizontal edges (u, v) w.r.t T ?

A: No!

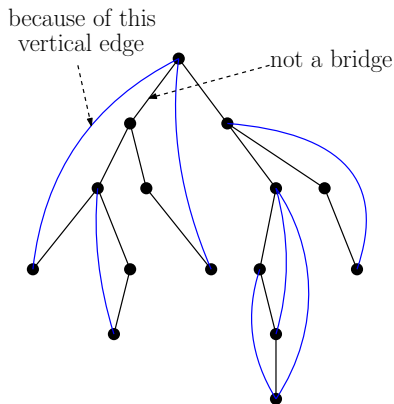
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- T : a DFS tree for G
- G contains only tree and vertical edges



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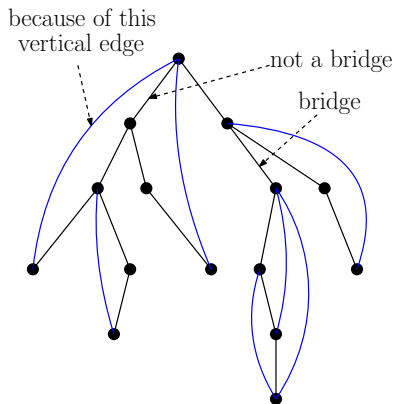
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Lemma

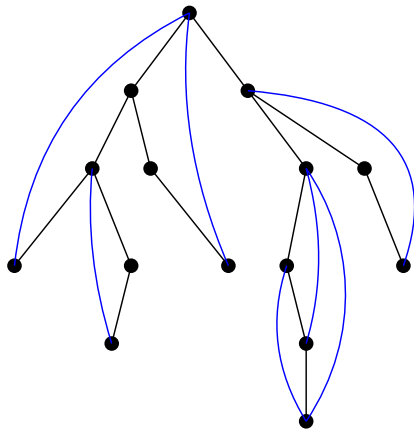
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- (u, v) is not a bridge $\iff \exists$ vertical edge connecting an (inclusive) descendant of v and an (inclusive) ancestor of u

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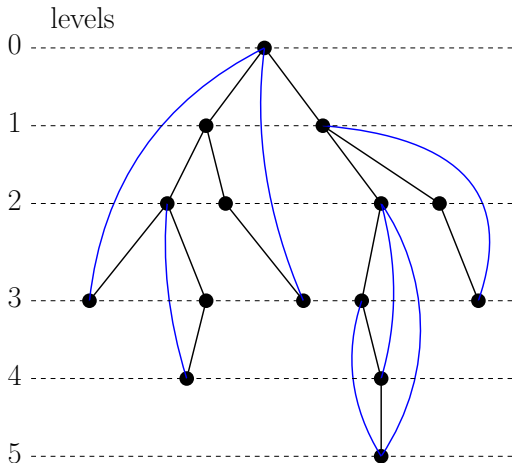


Lemma

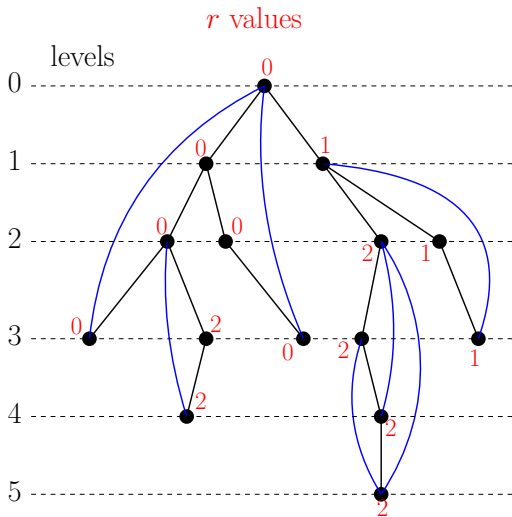
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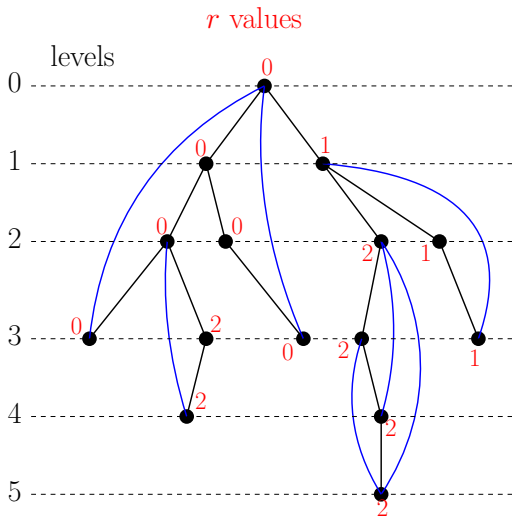
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- T_v : subtree rooted at v
- $v.r$: the smallest level that can be reached by a vertical edge from T_v
- $(parent(u), u)$ is a bridge if and only if $u.r \geq u.l$.



recursive-DFS(v)

```
1: mark  $v$  as "visited"
2:  $v.r \leftarrow \infty$ 
3: for all neighbours  $u$  of  $v$  do
4:   if  $u$  is unvisited then  $\triangleright u$  is a child of  $v$ 
5:      $u.l \leftarrow v.l + 1$ 
6:     recursive-DFS( $u$ )
7:     if  $u.r \geq u.l$  then claim  $(v, u)$  is a bridge
8:     if  $u.r < v.r$  then  $v.r \leftarrow u.r$ 
9:   else if  $u.l < v.l - 1$  then  $\triangleright u$  is ancestor but not parent
10:    if  $u.l < v.r$  then  $v.r \leftarrow u.l$ 
```

finding-bridges

```
1: mark all vertices as "unvisited"
2: for every  $v \in V$  do
3:   if  $v$  is unvisited then
4:      $v.l \leftarrow 0$ 
5:     recursive-DFS( $v$ )
```

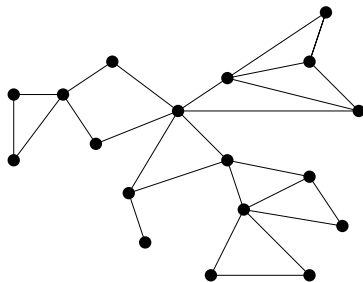
- Running time: $O(n + m)$

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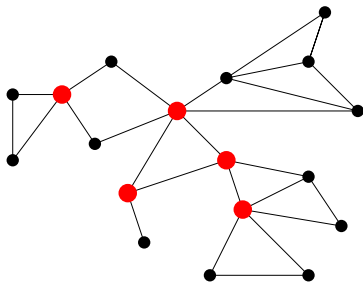
Cut vertices

Def. A vertex is a **cut vertex** of $G = (V, E)$ if its removal will increase the number of connected components of G .



Cut vertices

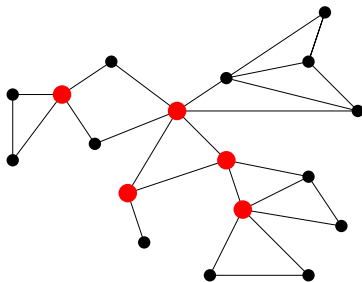
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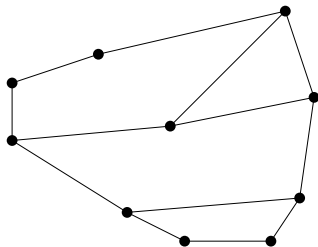
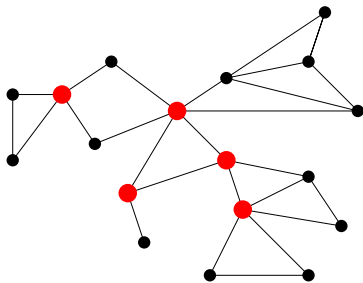
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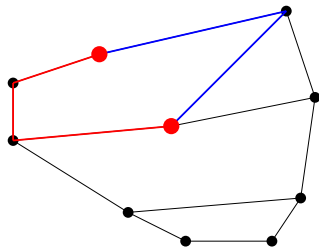
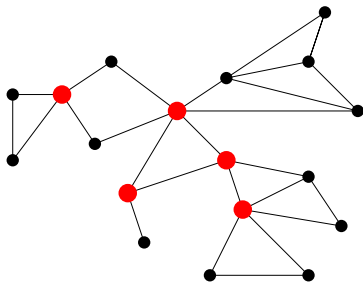
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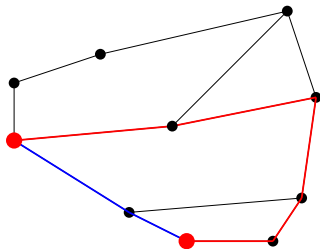
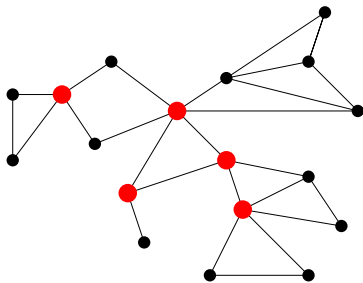
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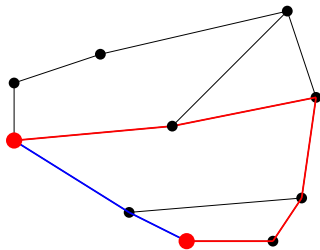
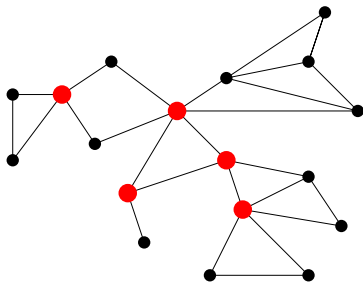


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Lemma A graph $G = (V, E)$ with $|V| \geq 3$ does not contain a cut vertex, if and only if it is biconnected.



Q: How can we find the cut vertices?

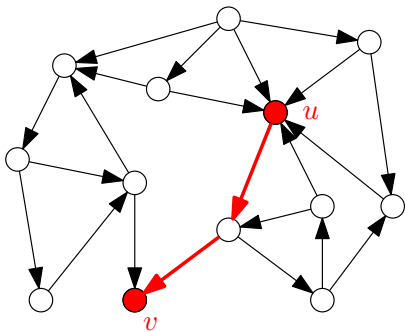
Q: How can we find the cut vertices?

A: With a small modification to the algorithm for finding bridges.

Outline

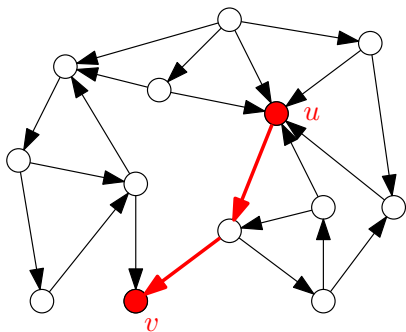
- 1 Graphs
- 2 Connectivity and Graph Traversal
 - Testing Bipartiteness
- 3 Topological Ordering
- 4 Bridges and 2-Edge-Connected Components
 - $O(n + m)$ -Time Algorithm to Find Bridges
 - Related Concept: Cut Vertices
- 5 Strong Connectivity in Directed Graphs
 - Tarjan's $O(n + m)$ -Time Algorithm for Finding SCCs

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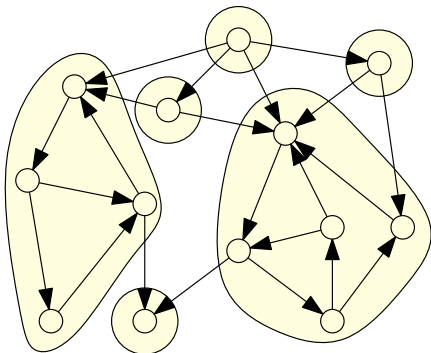
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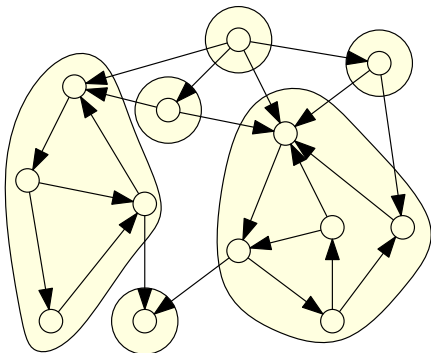
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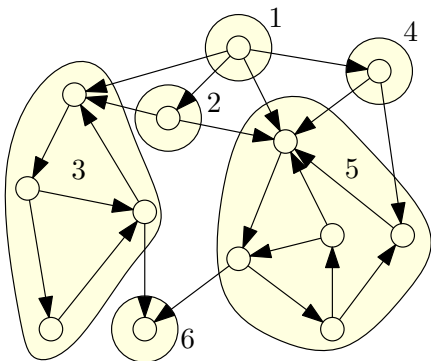


- Define equivalence relation: u and v are related if they are **reachable from each other**
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- After contracting each SCC, G becomes a **directed-acyclic (multi-)graph (DAG)**.

Q: How can we check if a directed graph $G = (V, E)$ is strongly-connected?

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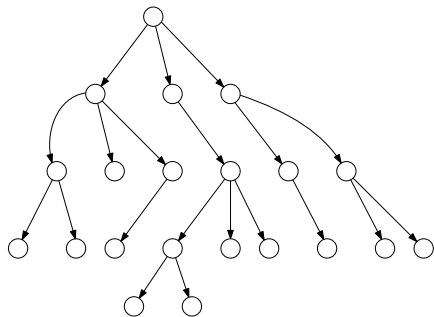
A: A much harder problem. Tarjan's $O(n + m)$ -time algorithm.

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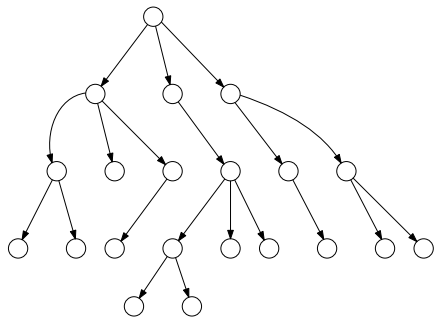
Type of Edges w.r.t a Directed DFS Tree

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- assuming every vertex is reachable from the root of T



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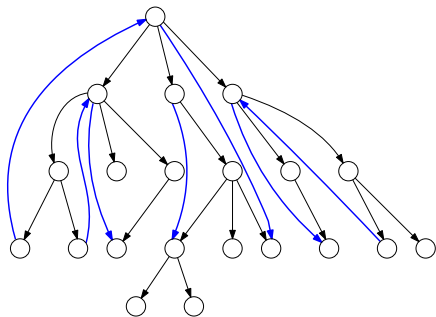


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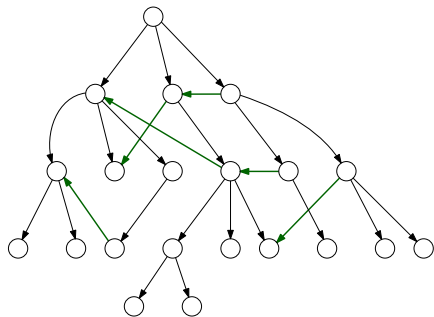


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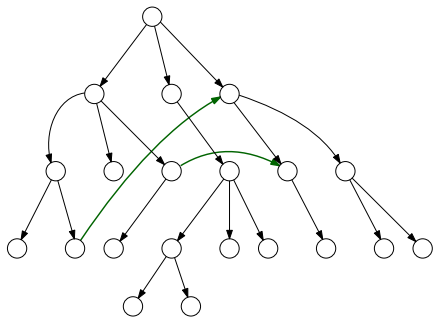


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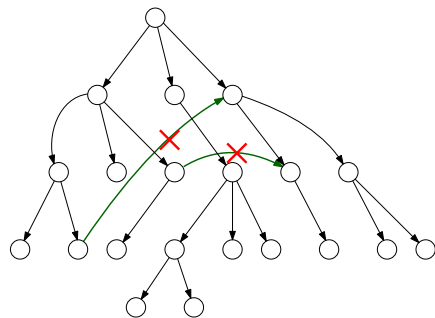
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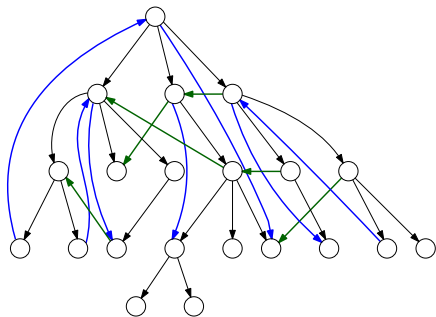
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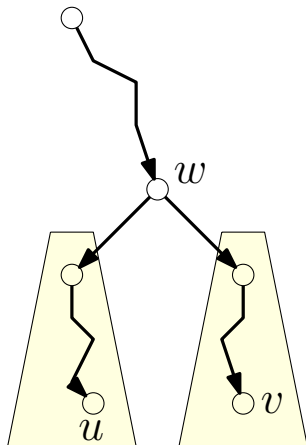
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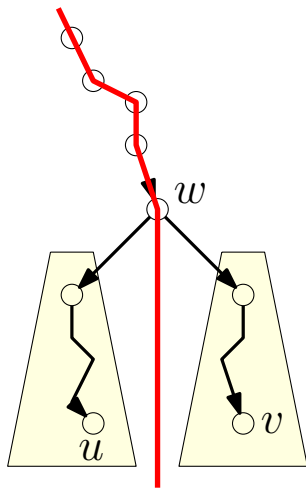
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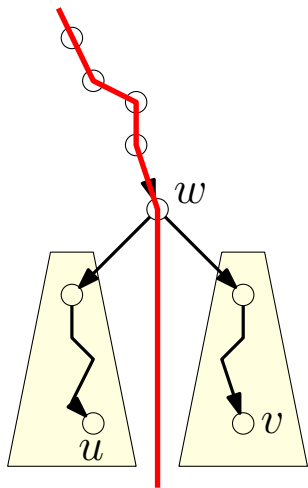
- Idea: using leftward, upwards and tree edges, u can not reach v without touching w or its ancestors. □



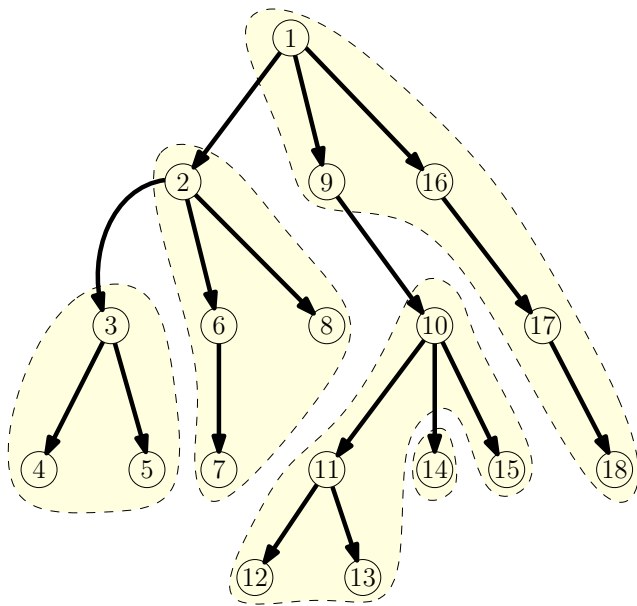
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Lemma The vertices in every SCC of G induce a sub-tree in T .



An Intermediate Algorithm to Keep in Mind

- 1: build the DFS tree T
- 2: **while** T is not empty **do**
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Proof.

- from v , we can reach any vertex in T_v (using tree edges, easy)
- from any vertex in T_v , we can reach v (harder)
- no edges go out of T_v (by our choice, easy)



Illustration of Intermediate Algorithm

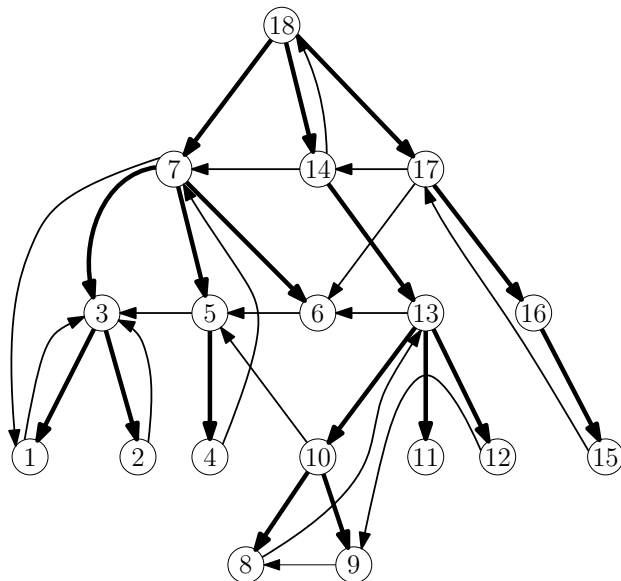


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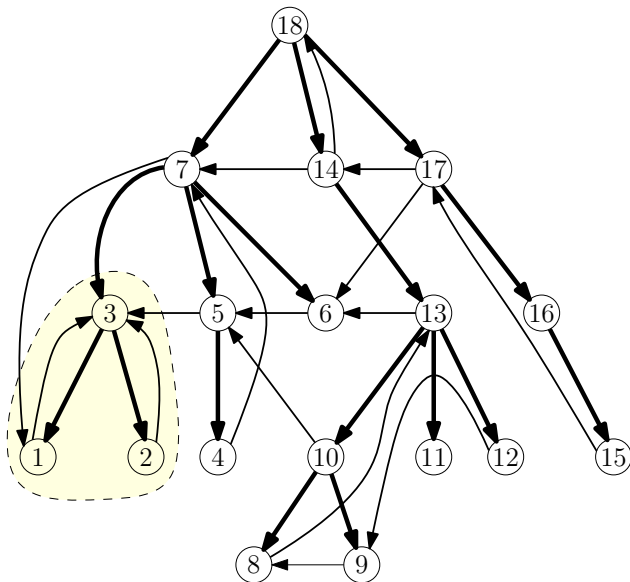


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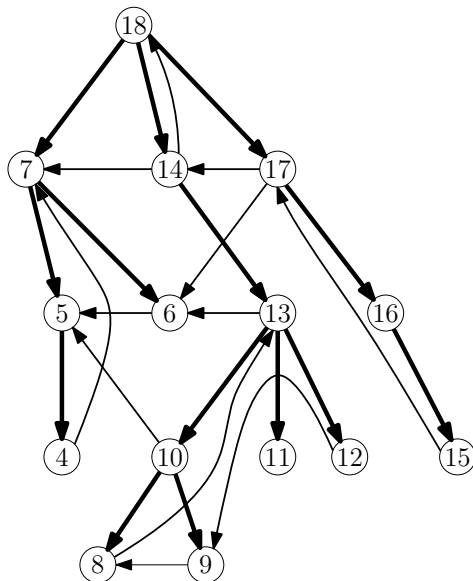


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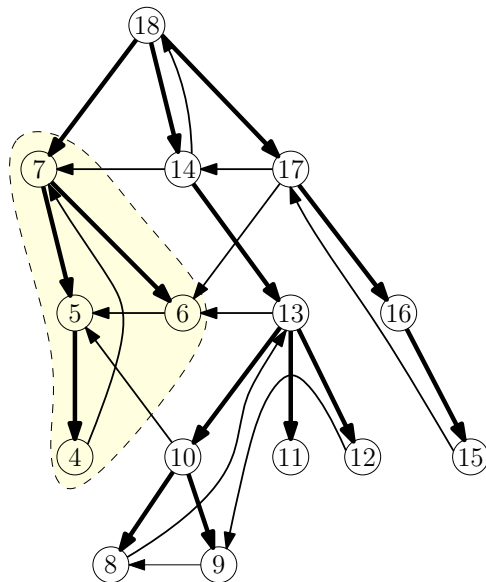


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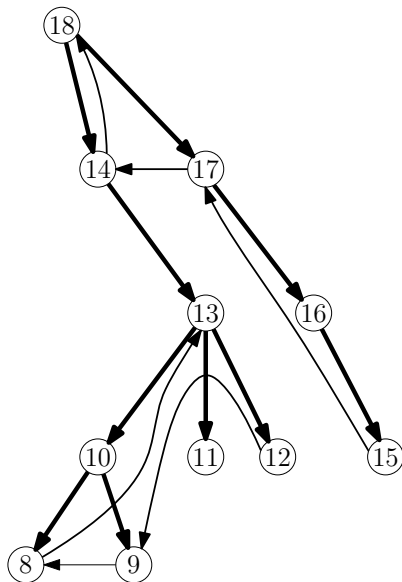


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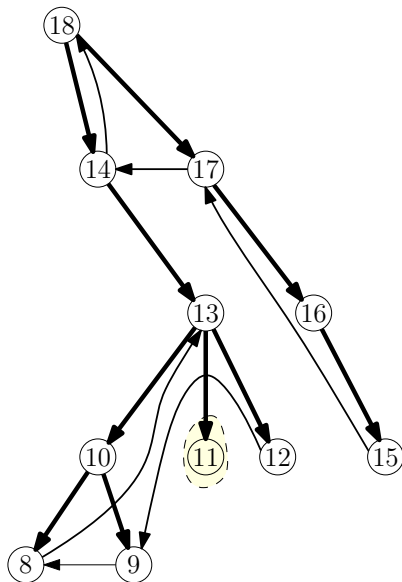


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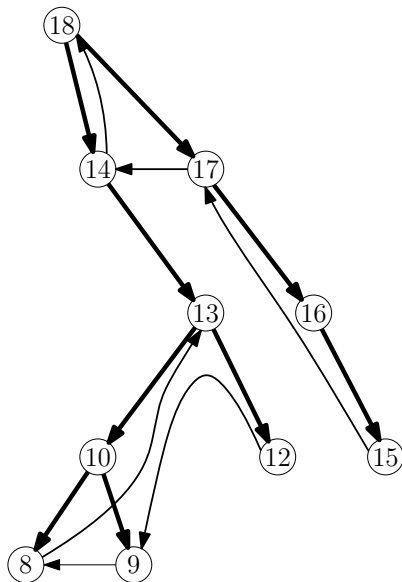


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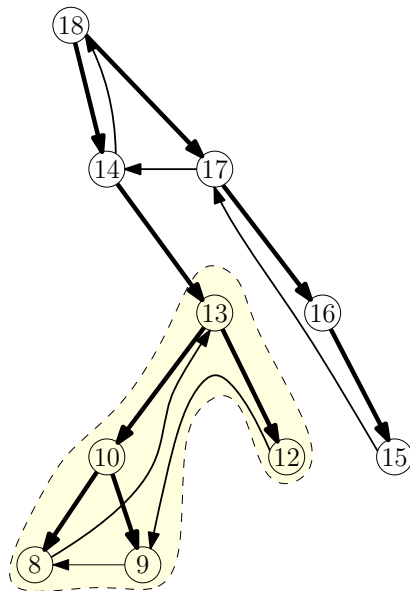


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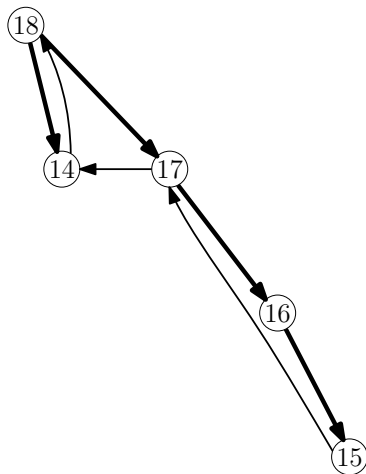


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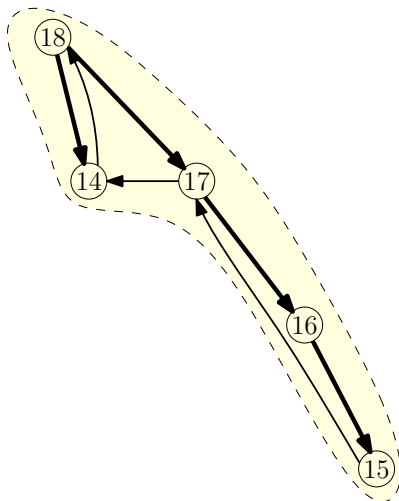


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Tarjan's $O(n + m)$ -Time Algorithm

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- *stack*: store alive vertices, in visiting order
- *onstack*[v]: whether v is in the stack (i.e, alive)
- $v.i$: the rank of v using the pre-traversal order
- $v.r$, for an **alive** v : the minimum of $u.i$, over all vertices u that can be reached from v , **using alive edges**

Illustration of Tarjan's Algorithm

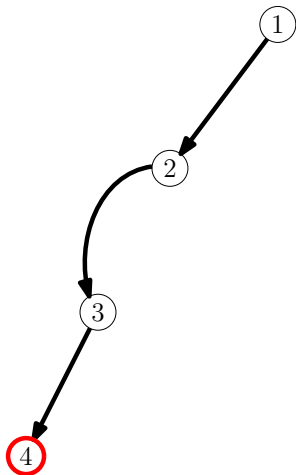


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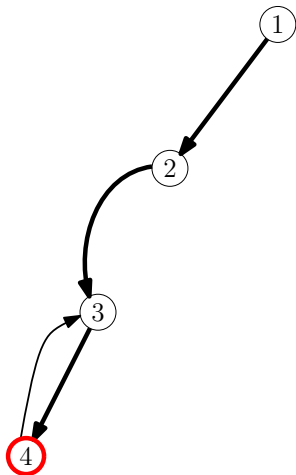


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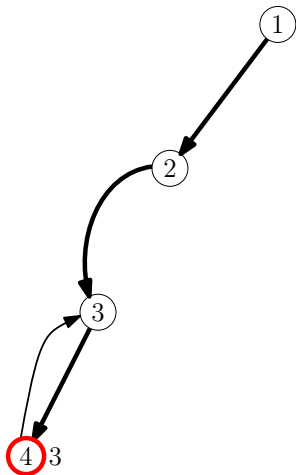


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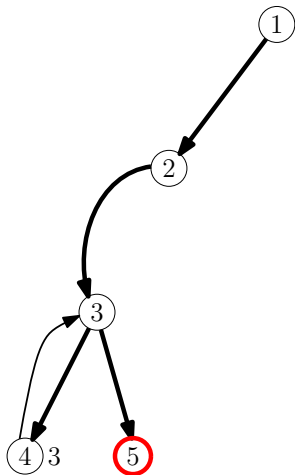


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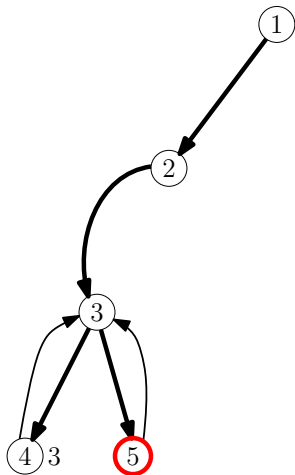


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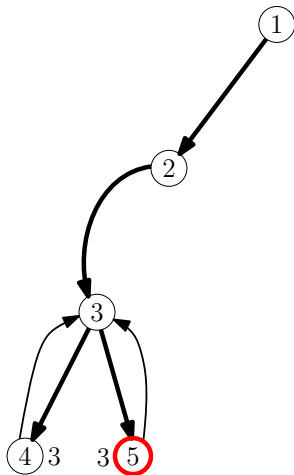


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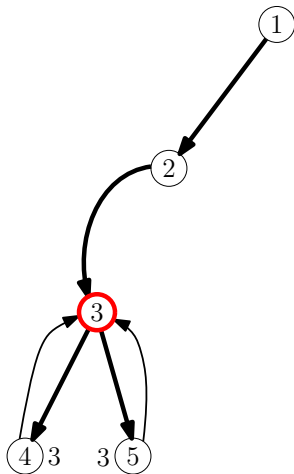


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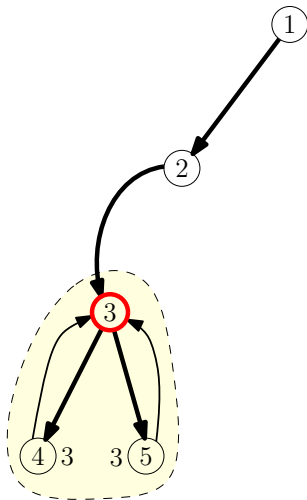


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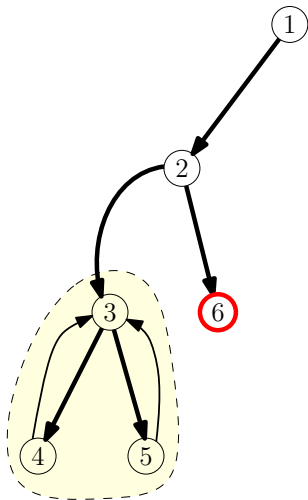


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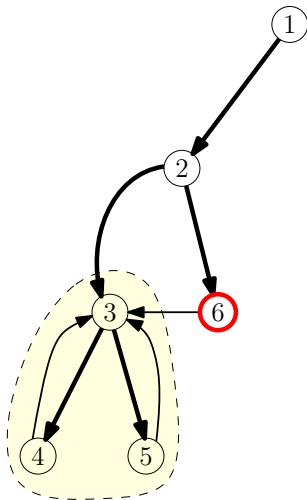


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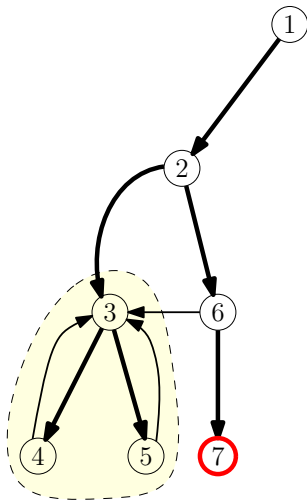


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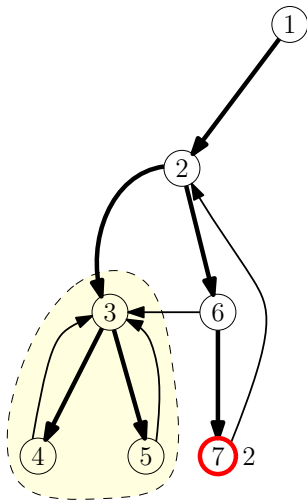


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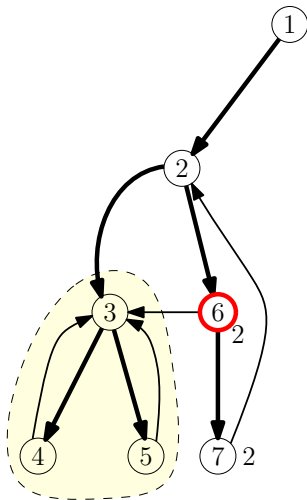


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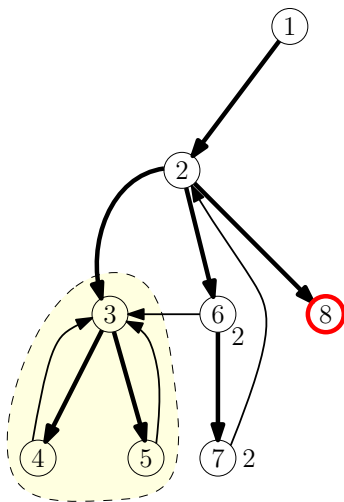


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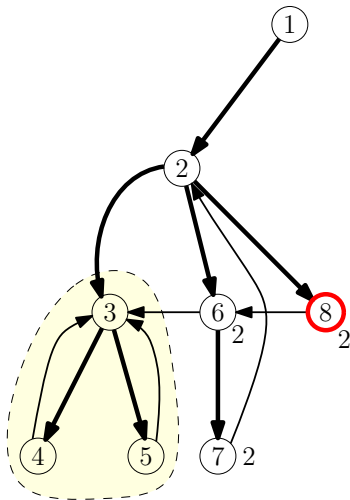


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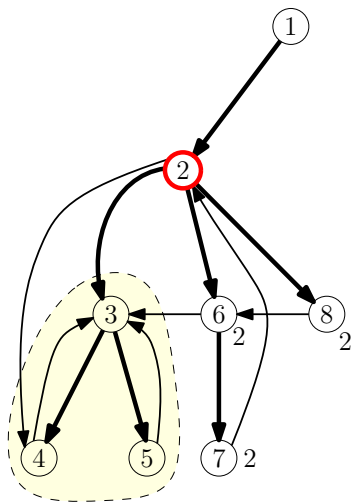


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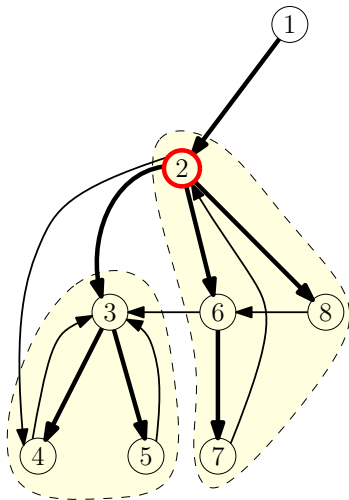


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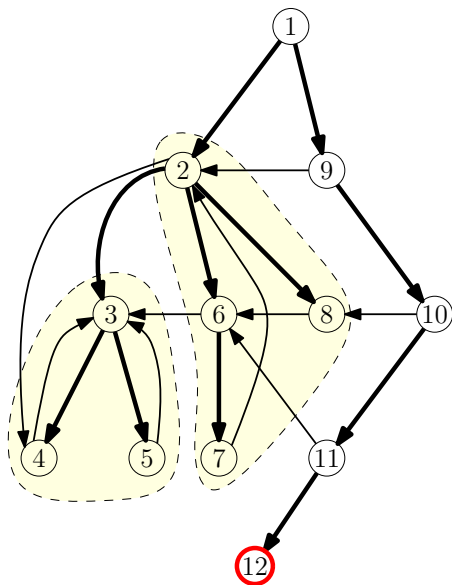


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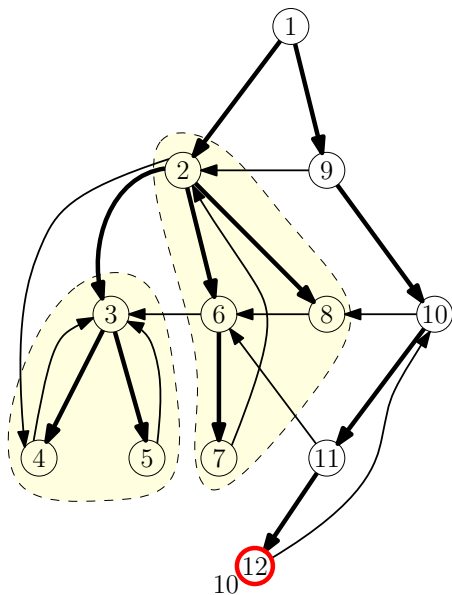


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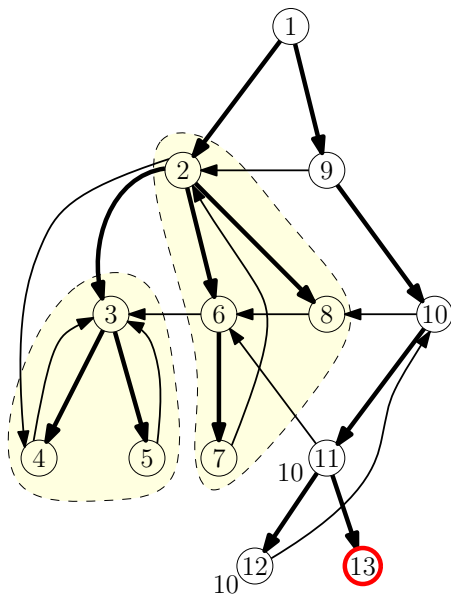


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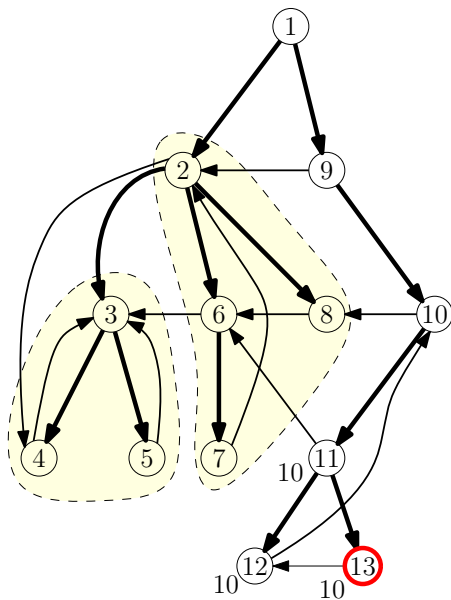


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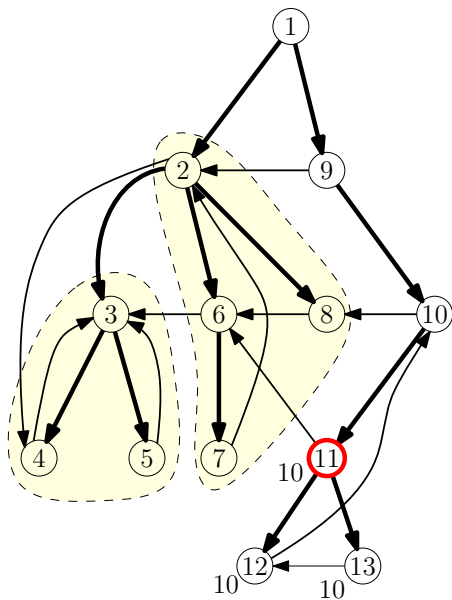


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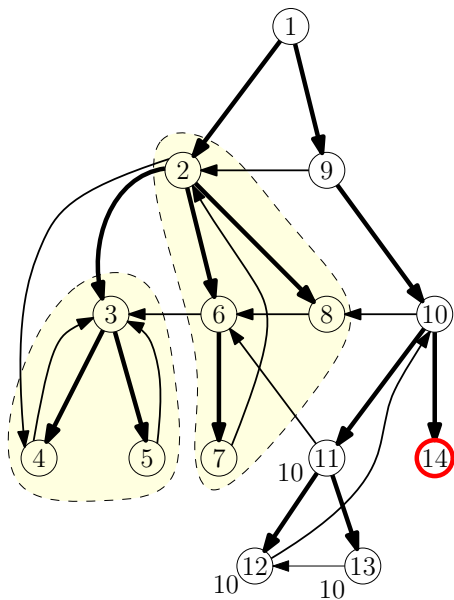


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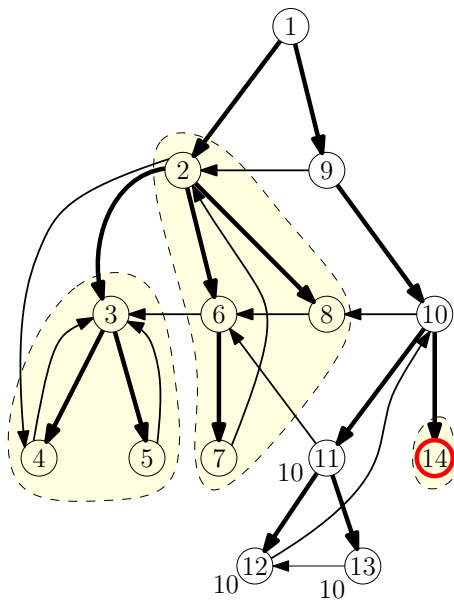


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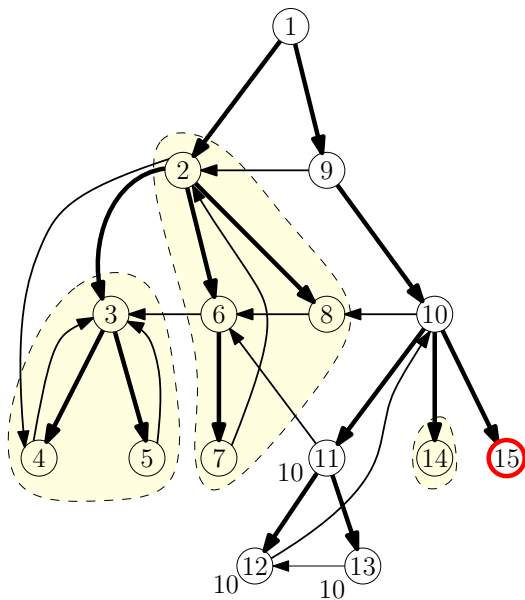


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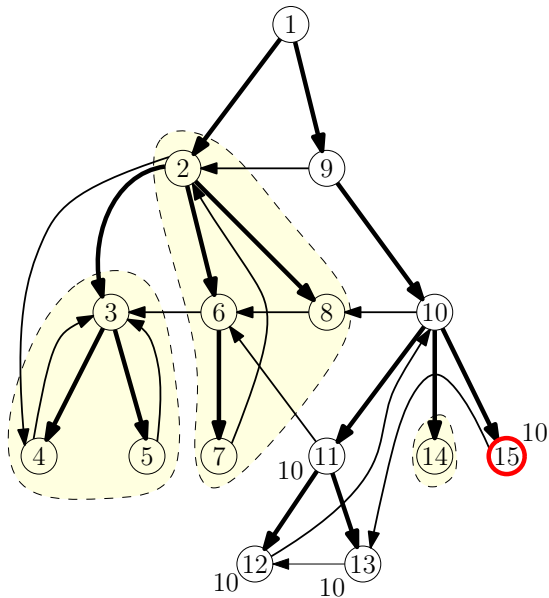


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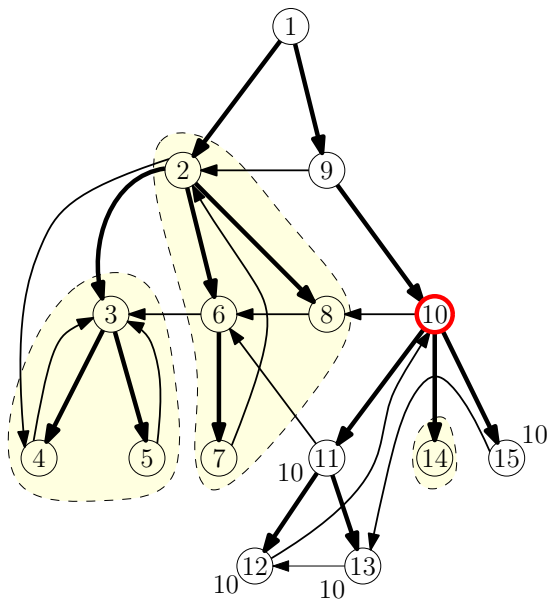


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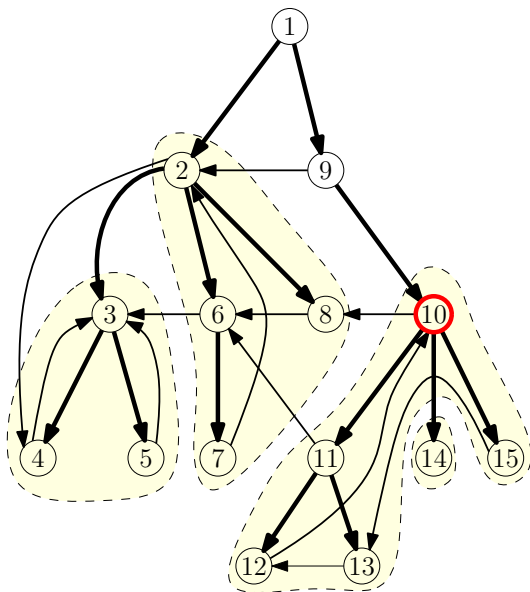


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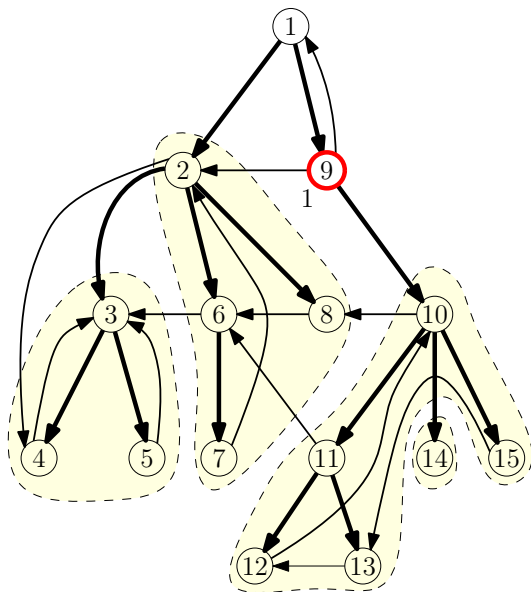


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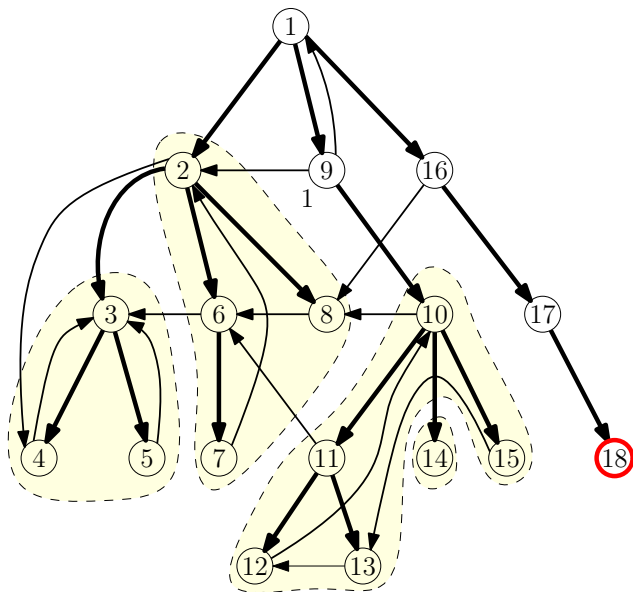


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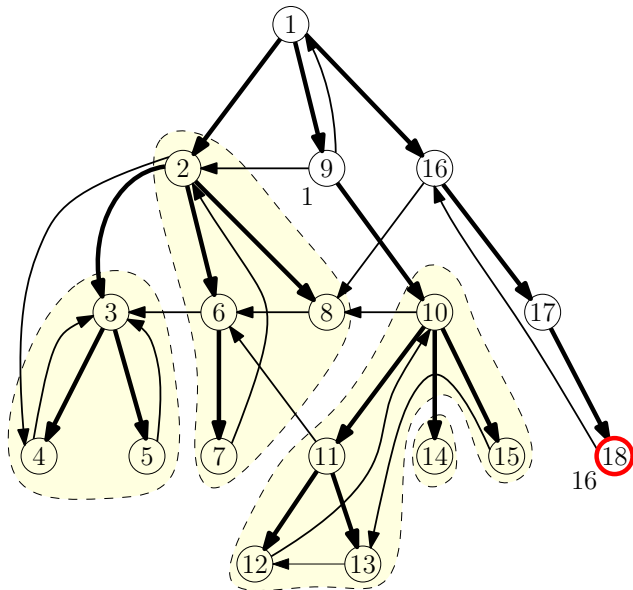


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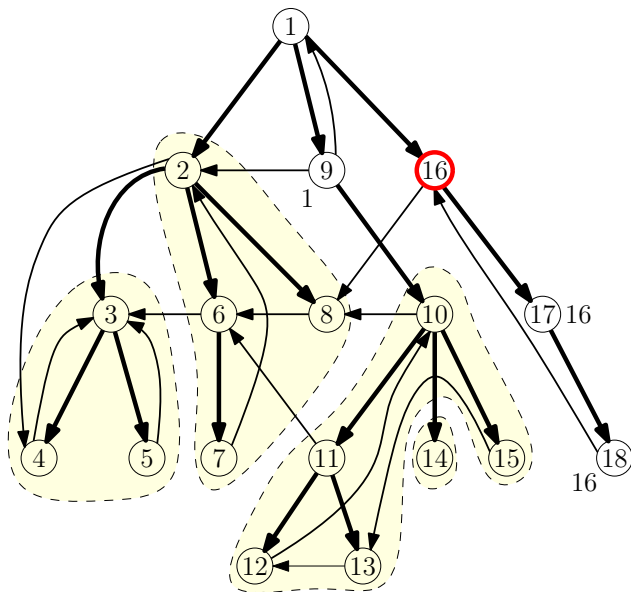


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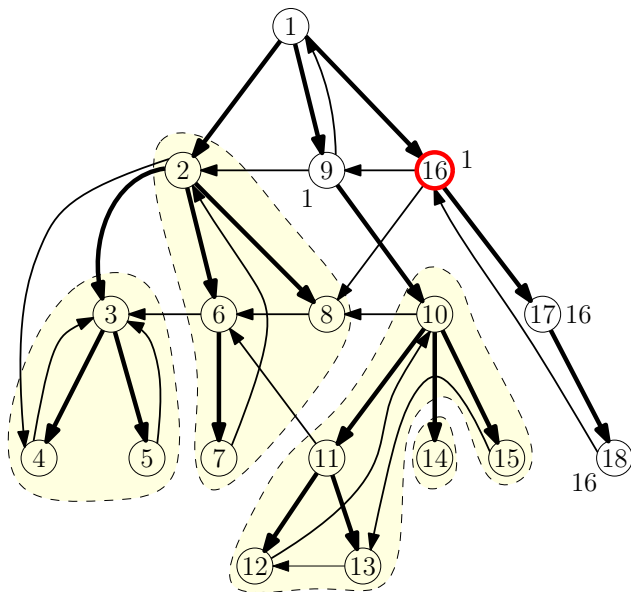


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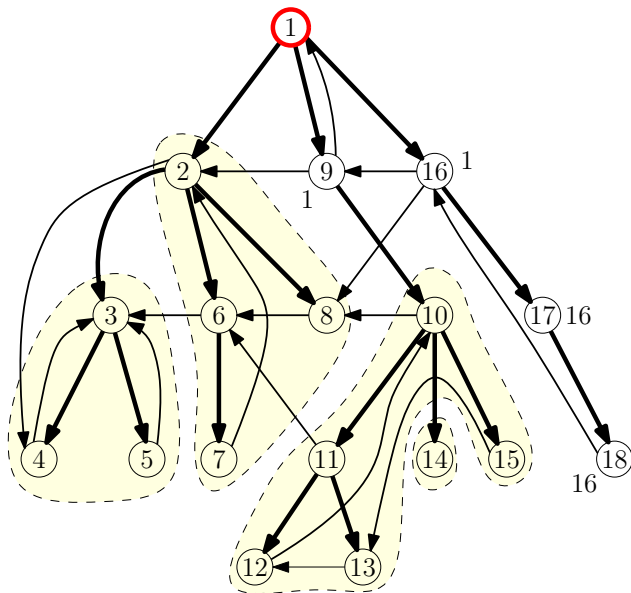
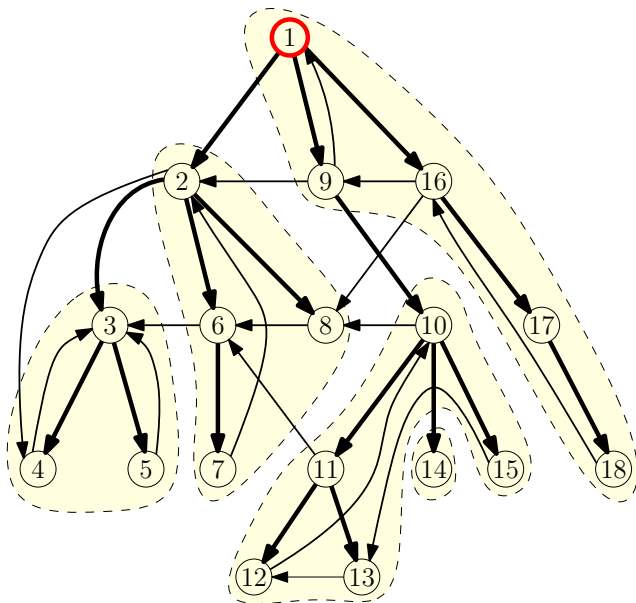


Illustration of Tarjan's Algorithm



finding strongly connected components

- 1: $stack \leftarrow \text{empty stack}, i \leftarrow 0$
- 2: **for** every $v \in V$ **do**: $v.i \leftarrow \perp, onstack[i] \leftarrow \text{false}$
- 3: **for** every $v \in V$ **do**
- 4: **if** $v.i = \perp$ **then** recursive-DFS(v)

recursive-DFS(v)

- 1: $i \leftarrow i + 1, v.i \leftarrow i, v.r \leftarrow i$
- 2: $stack.push(v), onstack[v] \leftarrow \text{true}$
- 3: **for** every outgoing edge (v, u) of v **do**
- 4: **if** $u.i = \perp$ **then** recursive-DFS(u)
- 5: **if** $onstack[u]$ and $u.r < v.r$ **then** $v.r \leftarrow u.r$
- 6: **if** $v.r = v.i$ **then**
- 7: pop all vertices in $stack$ after v , including v itself
- 8: set $onstack$ of these vertices to be **false**
- 9: declare that these vertices form an SCC

Running time of the algorithm is $O(n + m)$.