

Chapter 22 Solution

<https://github.com/frc123/CLRS>

12/1/2021

22.1

22.1-1

out-degree: $\Theta(V + E)$ by simply counting the size of each adjacency list.

```
1  std::vector<int> OutDegree(const Graph& graph)
2  {
3      size_t v_size, i;
4      v_size = graph.adj.size();
5      std::vector<int> degree(v_size);
6      for (i = 0; i < v_size; ++i)
7      {
8          // assume graph.adj[i].size() takes O(n)
9          // where n is size of graph.adj[i]
10         degree[i] = graph.adj[i].size();
11     }
12     return degree;
13 }
```

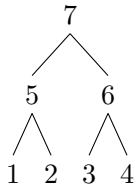
in-degree: $\Theta(V + E)$ by maintaining a counting table: each entry of the table is the counter for in-degree of the specific vertex.

```
1  std::vector<int> InDegree(const Graph& graph)
2  {
3      size_t v_size, i;
4      v_size = graph.adj.size();
5      std::vector<int> degree(v_size);
6      for (i = 0; i < v_size; ++i)
7      {
8          for (int v : graph.adj[i])
9              {
```

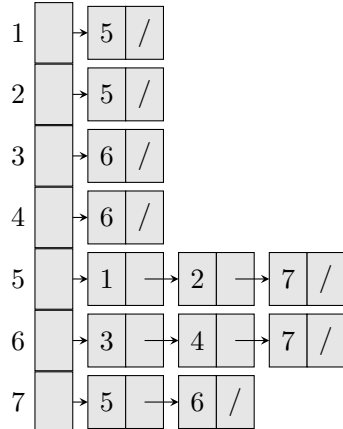
```
10         ++degree[v];
11     }
12 }
13 return degree;
14 }
```

22.1-2

Consider the following binary tree:



We have the following adjacency-list representation:



We have the following adjacency-matrix representation:

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

22.1-3

We can compute G^T from G for the adjacency-list representation in $\Theta(V + E)$ by the following algorithm:

```
1 AdjListGraph Transpose(const AdjListGraph& graph)
2 {
```

```
3     size_t size, u;
4     size = graph.adj.size();
5     AdjListGraph target(size);
6     for (u = 0; u < size; ++u)
7     {
8         for (int v : graph.adj[u])
9         {
10             target.adj[v].push_back(u);
11         }
12     }
13     return target;
14 }
```

We can compute G^T from G for the adjacency-matrix representation in $\Theta(V^2)$ by the following algorithm:

```
1 AdjMatrixGraph Transpose(const AdjMatrixGraph& graph)
2 {
3     size_t size, u, v;
4     size = graph.adj.size();
5     AdjMatrixGraph target(size);
6     for (u = 0; u < size; ++u)
7     {
8         for (v = 0; v < size; ++v)
9         {
10             target.adj[v][u] = graph.adj[u][v];
11         }
12     }
13     return target;
14 }
```

22.1-4

```
1 AdjListGraph Equivalent(const AdjListGraph& graph)
2 {
3     size_t size, u;
4     size = graph.adj.size();
5     AdjListGraph target(size);
6     std::vector<bool> edge_usage;
7     for (u = 0; u < size; ++u)
8     {
```

```
9      edge_usage = std::vector<bool>(size, false);
10     edge_usage[u] = true;
11     for (int v : graph.adj[u])
12     {
13         if (edge_usage[v] == false)
14         {
15             target.adj[u].push_back(v);
16             edge_usage[v] = true;
17         }
18     }
19 }
20 return target;
21 }
```

22.1-5

We can compute G^2 from G for the adjacency-list representation in $O(VE)$ by the following algorithm:

```
1  AdjListGraph Square(const AdjListGraph& graph)
2  {
3      size_t size, u;
4      size = graph.adj.size();
5      AdjListGraph result(size);
6      for (u = 0; u < size; ++u)
7      {
8          for (int v : graph.adj[u])
9          {
10             result.adj[u].push_back(v);
11             for (int w : graph.adj[v])
12             {
13                 result.adj[u].push_back(w);
14             }
15         }
16     }
17     return result;
18 }
```

We can compute G^2 from G for the adjacency-matrix representation in $\Theta(V^3)$ by the following algorithm (note that G^2 might be a not simple graph):

```
1  AdjMatrixGraph Square(const AdjMatrixGraph& graph)
```

```
2  {
3      size_t size, u, v, w;
4      size = graph.Rows();
5      AdjMatrixGraph result(size, size);
6      for (u = 0; u < size; ++u)
7      {
8          for (v = 0; v < size; ++v)
9          {
10             if (graph[u][v])
11             {
12                 result[u][v] = true;
13                 for (w = 0; w < size; ++w)
14                 {
15                     if (graph[v][w])
16                     {
17                         result[u][w] = true;
18                     }
19                 }
20             }
21         }
22     }
23     return result;
24 }
```

We also can optimize computation of G^2 from G by using Strassen algorithm.

Lemma 1. Let $A = (a_{ij})$ be a $n \times n$ nonnegative matrix and $B = A^2 = (b_{ij})$. Then $b_{uw} > 0$ if and only if there exists some integer $v \in [1, n]$ such that $a_{uv} > 0$ and $a_{vw} > 0$.

Proof. Contrapositive: $b_{uw} = 0 \iff (\forall v \in \mathbb{Z}_{[1,n]}, a_{uv} = 0 \vee a_{vw} = 0)$

According to equation (4.8) on page 75, we have

$$b_{uw} = \sum_{v=1}^n a_{uv} \cdot a_{vw}.$$

Note A is nonnegative matrix. Clearly, $b_{uw} = 0$ if and only if $a_{uv} = 0$ or $a_{vw} = 0$ for all integer $v \in [1, n]$ □

Claim 2. Let $A = (a_{ij})$ be the adjacency-matrix representations of graph $G = (V, E)$. Let $B = A^2 = (b_{ij})$. Then $b_{uw} > 0$ if and only if there exists a path with exactly two edges between u and w .

Proof. To prove the claim, we just need to show that “there exists a path with exactly two edges between u and w ” is equivalent to “there exists some integer $v \in [1, n]$ such that $a_{uv} > 0$ and $a_{vw} > 0$ ” so we can utilize lemma 1. Let $v \in V$. We have $(u, v) \in E$ if and only if $a_{uv} > 0$. Similarly,

$(v, w) \in E$ if and only if $a_{vw} > 0$. Also, $(u, v) \in E$ and $(v, w) \in E$ means there exists a path: $u \rightarrow v \rightarrow w$. □

Therefore, we have the following algorithm run in $\Theta(|V|^{\lg 7})$:

```
1 AdjMatrixGraph SquareByStrassen(const AdjMatrixGraph& graph)
2 {
3     size_t size, u, v, w;
4     size = graph.Rows();
5     AdjMatrixGraph result = StrassenMultiplication(graph, graph);
6     for (u = 0; u < size; ++u)
7     {
8         for (v = 0; v < size; ++v)
9         {
10             if (graph[u][v])
11             {
12                 result[u][v] = 1;
13             }
14         }
15     }
16     return result;
17 }
```

22.1-6

Notice that we can check whether a vertex is a universal sink in $\Theta(|V|)$. However, it will take $O(|V|^2)$ to check all vertex precisely. So, we want to constraint to a unique possible vertex and check that unique possible vertex.

Claim 3. $v \in V$ is a universal sink if and only if $(\forall w \in V, a_{vw} = 0)$ and $(\forall u \in V \setminus \{v\}, a_{uv} = 1)$.

Then we have

$$\begin{cases} a_{uv} = 1 & \text{implies } u \text{ is not a universal sink,} \\ a_{uv} = 0 \wedge u \neq v & \text{implies } v \text{ is not a universal sink.} \end{cases}$$

Thus we can eliminate a candidate vertex either u or v in $\Theta(1)$ by access a_{uv} if $u \neq v$.

Therefore, we have the following algorithm run in $\Theta(|V|)$:

```
1 // graph must be a square matrix
2 // return vertex of universal sink
3 // return -1 if universal sink not exist
4 int UniversalSink(const Matrix& graph)
5 {
```

```
6     size_t size, u, v;
7     size = graph.size();
8     // eliminate candidates
9     u = 0;
10    v = 1;
11    while (v < size)
12    {
13        if (graph[u][v])
14        {
15            ++u;
16            if (u == v)
17            {
18                ++v;
19            }
20        }
21        else
22        {
23            ++v;
24        }
25    }
26    // test the possible vertex u by claim 3
27    for (v = 0; v < size; ++v)
28    {
29        if (graph[u][v])
30            return -1;
31    }
32    for (v = 0; v < size; ++v)
33    {
34        if (graph[v][u] == false && u != v)
35            return -1;
36    }
37    return u;
38 }
```

The following algorithm runs in $\Theta(|V|)$ also:

```
1 int UniversalSinkAnother(const Matrix& graph)
2 {
3     size_t size, u, v;
4     size = graph.size();
5     u = 0;
```

```
6     v = 0;
7     while (u < size && v < size)
8     {
9         if (graph[u][v])
10        {
11            ++u;
12        }
13        else
14        {
15            ++v;
16        }
17    }
18    if (u >= size)
19        return -1;
20    for (v = 0; v < size; ++v)
21    {
22        if (graph[u][v])
23            return -1;
24    }
25    for (v = 0; v < size; ++v)
26    {
27        if (graph[v][u] == false && u != v)
28            return -1;
29    }
30    return u;
31 }
```

22.1-7

Let matrix $C = B^T = (c_{ij})$. This says C is a $|E| \times |V|$ matrix, and $c_{ij} = b_{ji}$. Let $D = BB^T = (d_{ij})$. Hence we have

$$d_{ij} = \sum_{k \in E} b_{ik} c_{kj} = \sum_{k \in E} b_{ik} b_{jk}$$

In conclusion, the meaning of d_{ij} depends on whether $i = j$.

Case 1 $i = j$

$b_{ik} b_{jk} = b_{ik} = 1 = 1 \cdot 1 = -1 \cdot -1$ implies edge k enters or leaves vertex i .

$b_{ik} b_{jk} = b_{ik} = 0$ implies edge k does not connect to vertex i .

$b_{ik} b_{jk} = b_{ik} = -1$ is impossible since $b_{ik} = b_{jk}$.

Hence d_{ij} means the total degree (in-degree + out-degree) of vertex i .

Case 2 $i \neq j$

$b_{ik}b_{jk} = 1 = 1 \cdot 1 = -1 \cdot -1$ is impossible since edge k cannot enter i and j simultaneously, and edge k cannot leave i and j simultaneously.

$b_{ik}b_{jk} = 0$ implies edge k does not connect to vertex i and j .

$b_{ik}b_{jk} = -1$ implies edge k leaves vertex i and enters j , or edge k leaves vertex j and enters i .

Hence $-d_{ij}$ means the number of edges connect to vertex i and j simultaneously.

22.1-8

Expected time to determine whether an edge is in the graph: $\Theta(1)$.

Disadvantage to use hash table: 1. we are not able to handle graphs that are not simple; 2. the worst case take $\Theta(|V|)$ time.

Suggest: utilize red-black trees containing keys v such that $(u, v) \in E$; add a counter (counter for unweighted graph; list for weighted graph) to the attributes of each node in the red-black tree to handle graphs that are not simple.

Disadvantage compared to the hash table: expect time of red-black tree is $\Theta(\lg n)$ where n is the size of elements in the red-black tree.

22.2

22.2-1

vertex 1: $d = \infty, \pi = NIL$.

vertex 2: $d = 3, \pi = 4$.

vertex 3: $d = 0, \pi = NIL$.

vertex 4: $d = 2, \pi = 5$.

vertex 5: $d = 1, \pi = 3$.

vertex 6: $d = 1, \pi = 3$.

22.2-2

vertex r: $d = 4, \pi = s$.

vertex s: $d = 3, \pi = w$.

vertex t: $d = 1, \pi = u$.

vertex u: $d = 0, \pi = NIL$.

vertex v: $d = 5, \pi = r$.

vertex w: $d = 2, \pi = x$.

vertex x: $d = 1, \pi = u$.

vertex y: $d = 1, \pi = u$.

22.2-3

We can replace all "GRAY"s with "BLACK"s. Notice line 13 is the only place to check color, and it only checks whether or not the vertex is "WHITE". Hence, as long as we can make sure all non-white vertex are in non-"WHITE" color, our algorithm works.

```
1  std::list<int> BFS(Graph& graph, int src_idx)
2  {
3      int u_idx, v_idx;
4      std::list<int> discover_result;
5      std::queue<int> q;
6      for (Vertex& u : graph.vertices)
7      {
8          u.color = Color::kWhite;
9          u.distance = INT_MAX;
10         u.prev = -1;
11     }
12     graph.vertices[src_idx].color = Color::kNonWhite;
13     graph.vertices[src_idx].distance = 0;
14     graph.vertices[src_idx].prev = -1;
15     q.push(src_idx);
16     while (q.empty() == false)
17     {
18         u_idx = q.front();
19         q.pop();
20         discover_result.push_back(u_idx);
21         Vertex& u = graph.vertices[u_idx];
22         for (int v_idx : graph.adj[u_idx])
23         {
24             Vertex& v = graph.vertices[v_idx];
25             if (v.color == Color::kWhite)
26             {
27                 v.color = Color::kNonWhite;
28                 v.distance = u.distance + 1;
29                 v.prev = u_idx;
30                 q.push(v_idx);
31             }
32         }
33     }
34     return discover_result;
35 }
```

22.2-4

$O(V^2)$ since we need to iterate all entries in the matrix.

```
1  std::list<int> BFS(Graph& graph, int src_idx)
2  {
3      int u_idx, v_idx;
4      std::list<int> discover_result;
5      std::queue<int> q;
6      for (Vertex& u : graph.vertices)
7      {
8          u.color = Color::kWhite;
9          u.distance = INT_MAX;
10         u.prev = -1;
11     }
12     graph.vertices[src_idx].color = Color::kGray;
13     graph.vertices[src_idx].distance = 0;
14     graph.vertices[src_idx].prev = -1;
15     q.push(src_idx);
16     while (q.empty() == false)
17     {
18         u_idx = q.front();
19         q.pop();
20         discover_result.push_back(u_idx);
21         Vertex& u = graph.vertices[u_idx];
22         for (v_idx = 0; v_idx < graph.size; ++v_idx)
23         {
24             if (graph.adj[u_idx][v_idx])
25             {
26                 Vertex& v = graph.vertices[v_idx];
27                 if (v.color == Color::kWhite)
28                 {
29                     v.color = Color::kGray;
30                     v.distance = u.distance + 1;
31                     v.prev = u_idx;
32                     q.push(v_idx);
33                 }
34             }
35         }
36         u.color = Color::kBlack;
37     }
```

```
38     return discover_result;  
39 }
```

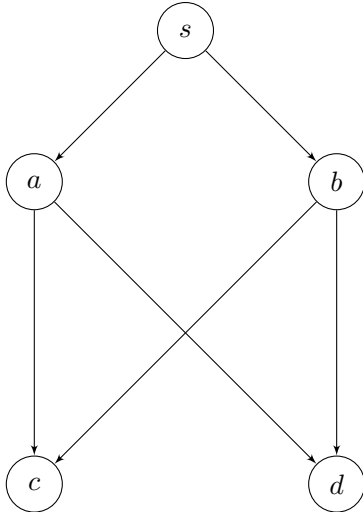
22.2-5

For all reachable vertices $v \in V$, we have $v.d = \delta(s, v)$ by Theorem 22.5. For unreachable vertices $v \in V$, we have $v.d = \infty = \delta(s, v)$. Since $\delta(s, v)$ is independent of the order in which the vertices appear in each adjacency list, so does $v.d$ for all $v \in V$.

If t is in front of x in the adjacency list of w , then t will be enqueued before x enqueue, and $u.\pi = t$, so $(t, u) \in E_\pi$. If t is in after x in the adjacency list of w , then t will be enqueued after x enqueue, and $u.\pi = x$, so $(x, u) \in E_\pi$.

22.2-6

Consider the following directed graph $G = (V, E)$:



Consider $E_\pi = \{(s, a), (s, b), (a, c), (b, d)\}$. This E_π cannot be produced by running BFS on G . The only two possible predecessor subgraph is

$$(V, \{(s, a), (s, b), (a, c), (a, d)\})$$

and

$$(V, \{(s, b), (s, a), (b, c), (b, d)\}).$$

22.2-7

Let $G = (V, E)$ be an undirected graph.

Let each vertex $v \in V$ represent a professional wrestler, and let each edge $e \in E$ represent a rivalry.

```
1  bool DesignateWrestlerType(Graph& graph)
2  {
3      int u_idx, v_idx;
4      std::queue<int> q;
5      for (Vertex& u : graph.vertices)
6      {
7          u.color = Color::kWhite;
8      }
9      graph.vertices[0].color = Color::kNonWhite;
10     graph.vertices[0].type = WrestlerType::kBabyfaces;
11     q.push(0);
12     while (q.empty() == false)
13     {
14         u_idx = q.front();
15         q.pop();
16         Vertex& u = graph.vertices[u_idx];
17         for (int v_idx : graph.adj[u_idx])
18         {
19             Vertex& v = graph.vertices[v_idx];
20             if (v.color == Color::kWhite)
21             {
22                 v.color = Color::kNonWhite;
23                 graph.vertices[v_idx].type = (u.type == WrestlerType::kBabyfaces) ?
24                     WrestlerType::kHeels : WrestlerType::kBabyfaces;
25                 q.push(v_idx);
26             }
27             else
28             {
29                 if (u.type == v.type)
30                     return false;
31             }
32         }
33     }
34     return true;
35 }
```

22.2-8

According to Theorem B.2 (Properties of free trees), We have T is connected, acyclic, and $|E| = |V| - 1$. This says we can do BFS in $O(|V| + |E|) = O(|V| + |V| - 1) = O(|V|)$ time. If we

can do constant times BFS, we will be able to find diameter in $O(|V|)$ time.

Let arbitrary node $s \in V$ be the source node of BFS. Since T is connected, all nodes in V are reachable. After performing BFS on T from the source node s , we have $v.d = \delta(s, v)$ for each $v \in V$. Let s be the root node of the tree T . Then $v.d$ is the depth of node v in the tree since any two vertices in a tree are connected by a unique simple path (Theorem B.2 (2)) and the length of the path from s to v is $\delta(s, v)$, which is depth of node v . We can recursively traverse the tree to get the height of each node based on depth, and we can get the longest path contains the root of the subtree by height of children of the root.

Since BFS takes $O(|V|)$ and traverses tree takes $O(|V|)$, our algorithm runs in $O(|V|)$.

```
1  int DiameterAux(Graph& graph, int subtree_root_idx)
2  {
3      int first_largest_h, second_largest_h, longest_path_length;
4      first_largest_h = -1;
5      second_largest_h = -1;
6      longest_path_length = INT_MIN;
7      Vertex& subtree_root = graph.vertices[subtree_root_idx];
8      for (int child_idx : graph.adj[subtree_root_idx])
9      {
10         if (child_idx != subtree_root.prev)
11         {
12             Vertex& child = graph.vertices[child_idx];
13             longest_path_length = std::max(longest_path_length,
14                 DiameterAux(graph, child_idx));
15             if (child.height > first_largest_h)
16             {
17                 second_largest_h = first_largest_h;
18                 first_largest_h = child.height;
19             }
20             else if (child.height > second_largest_h)
21             {
22                 second_largest_h = child.height;
23             }
24         }
25     }
26     subtree_root.height = first_largest_h + 1;
27     longest_path_length = std::max(longest_path_length,
28         first_largest_h + second_largest_h + 2);
29     return longest_path_length;
30 }
```

```
31
32 int Diameter(Graph& graph)
33 {
34     int u_idx, v_idx;
35     std::queue<int> q;
36     // let vertex 0 be the root node
37     graph.vertices[0].depth = 0;
38     graph.vertices[0].prev = -1;
39     q.push(0);
40     while (q.empty() == false)
41     {
42         u_idx = q.front();
43         q.pop();
44         Vertex& u = graph.vertices[u_idx];
45         for (int v_idx : graph.adj[u_idx])
46         {
47             if (v_idx != u.prev)
48             {
49                 Vertex& v = graph.vertices[v_idx];
50                 v.depth = u.depth + 1;
51                 v.prev = u_idx;
52                 q.push(v_idx);
53             }
54         }
55     }
56     return DiameterAux(graph, 0);
57 }
```

22.2-9

```
1 std::list< std::pair<int, int> > TraverseEdge(Graph& graph, int src_idx)
2 {
3     int u_idx, v_idx;
4     std::list< std::pair<int, int> > traverse_result;
5     std::list< std::pair<int, int> >::iterator result_it;
6     std::queue<int> q;
7     for (Vertex& u : graph.vertices)
8     {
9         u.color = Color::kWhite;
10    }
```

```
11     graph.vertices[src_idx].color = Color::kGray;
12     graph.vertices[src_idx].result_pos = traverse_result.end();
13     q.push(src_idx);
14     while (q.empty() == false)
15     {
16         u_idx = q.front();
17         q.pop();
18         Vertex& u = graph.vertices[u_idx];
19         result_it = u.result_pos;
20         for (int v_idx : graph.adj[u_idx])
21         {
22             Vertex& v = graph.vertices[v_idx];
23             if (v.color == Color::kWhite)
24             {
25                 v.color = Color::kGray;
26                 traverse_result.emplace(result_it, u_idx, v_idx);
27                 v.result_pos =
28                     traverse_result.emplace(result_it, v_idx, u_idx);
29                 q.push(v_idx);
30             }
31             else if (v.color == Color::kGray)
32             {
33                 traverse_result.emplace(result_it, u_idx, v_idx);
34                 traverse_result.emplace(result_it, v_idx, u_idx);
35             }
36         }
37         u.color = Color::kBlack;
38     }
39     return traverse_result;
40 }
```

Let each spot of the maze be a vertex in V . If there is no wall between two spots, we add an edge to connect two vertices. Then we traverse each edge in E by using the above algorithm to find the way out of the maze.

22.3

22.3-1

Directed:

$\begin{array}{c} \backslash \\ i \end{array} \begin{array}{c} j \\ \end{array}$	WHITE	GRAY	BLACK
W	tree/back/forward/cross	back/cross	cross
G	tree/forward	tree/back/forward	tree/forward/cross
B	N/A	back	tree/back/forward/cross
Undirected:			
$\begin{array}{c} \backslash \\ i \end{array} \begin{array}{c} j \\ \end{array}$	WHITE	GRAY	BLACK
W	tree/back	tree/back	N/A
G	tree/back	tree/back	tree/back
B	N/A	tree/back	tree/back

22.3-2

Discovery and finish time:

q 1/16

s 2/7

v 3/6

w 4/5

t 8/15

x 9/12

z 10/11

y 13/14

r 17/20

u 18/19

Classification of each edge:

(q,s) tree

(s,v) tree

(v,w) tree

(w,s) back

(q,t) tree

(t,x) tree

(x,z) tree
(z,x) back
(t,y) tree
(y,q) back
(q,w) forward
(r,u) tree
(u,y) cross
(r,y) cross

22.3-3

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

22.3-4

The algorithm only check if the color is white or non white, so we can use only two colors.

```
1 void DFSVisit(Graph& graph, int u_idx, int& time)
2 {
3     Vertex& u = graph.vertices[u_idx];
4     ++time;
5     u.discovery_time = time;
6     u.color = Color::kNonWhite;
7     for (int v_idx : graph.adj[u_idx])
8     {
9         Vertex& v = graph.vertices[v_idx];
10        if (v.color == Color::kWhite)
11        {
12            v.prev = u_idx;
13            DFSVisit(graph, v_idx, time);
14        }
15    }
16    ++time;
17    u.finishing_time = time;
18 }
19
20 void DFS(Graph& graph)
21 {
```

```
22     int time, u_idx;
23     for (Vertex& u : graph.vertices)
24     {
25         u.color = Color::kWhite;
26     }
27     time = 0;
28     for (u_idx = 0; u_idx < graph.size; ++u_idx)
29     {
30         Vertex& u = graph.vertices[u_idx];
31         if (u.color == Color::kWhite)
32         {
33             u.prev = -1;
34             DFSVisit(graph, u_idx, time);
35         }
36     }
37 }
```

22.3-5

(a)

Proof. Edge (u, v) is a tree edge or forward edge if and only if v is a proper descendant of u . By Corollary 22.8 (Nesting of descendants' intervals), we have $u.d < v.d < v.f < u.f$ if and only if v is a proper descendant of u . \square

(b)

Proof. Edge (u, v) is a back edge if and only if u is a descendant of v (include self-loop). By Theorem 22.7 (Parenthesis theorem), we have $v.d \leq u.d < u.f \leq v.f$ if and only if u is a descendant of v immediately. \square

(c)

Proof. Edge (u, v) is a cross edge if and only if u is neither descendant of v nor ancestor of v . By Theorem 22.7 (Parenthesis theorem), we have $v.d < v.f < u.d < u.f$ or $u.d < u.f < v.d < v.f$ if and only if u is neither descendant of v nor ancestor of v . Thus, we have $v.d < v.f < u.d < u.f \implies (u, v)$ is a cross edge.

We claim (u, v) is a cross edge $\implies v.d < v.f < u.d < u.f$. Assume $u.d < u.f < v.d < v.f$, for the purpose of contradiction. This says u is black and v is white at some point during the DFS. However, there is an edge from u to v , which is a contradiction. \square

22.3-6

Proof. By definition of tree edge, edge (u, v) is classified as a tree edge if v is first discovered by exploring edge (u, v) , which is equivalent to (u, v) is encountered first. By Theorem 22.10, every edge in an undirected graph is either a tree edge or a back edge. Thus, if (u, v) is not classified as a tree edge, (u, v) is classified as a back edge, which is equivalent to (v, u) is encountered first. \square

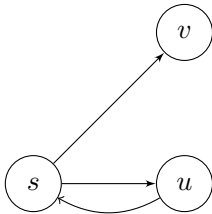
22.3-7

```
1  struct StackElement
2  {
3      int vertex_idx;
4      AdjList::iterator adj_it_curr;
5      AdjList::iterator adj_it_end;
6  };
7
8  void DFS(Graph& graph)
9  {
10     int time, u_idx, v_idx;
11     std::stack< StackElement > path;
12     AdjList vertices_list;
13     for (u_idx = 0; u_idx < graph.size; ++u_idx)
14     {
15         Vertex& u = graph.vertices[u_idx];
16         u.color = Color::kWhite;
17         vertices_list.push_back(u_idx);
18     }
19     time = 0;
20     path.push({ -1, vertices_list.begin(), vertices_list.end() });
21     while (true)
22     {
23         u_idx = path.top().vertex_idx;
24         AdjList::iterator& v_it = path.top().adj_it_curr;
25         if (v_it == path.top().adj_it_end)
26         {
27             path.pop();
28             if (path.empty())
29                 break;
30             Vertex& u = graph.vertices[u_idx];
31             u.color = Color::kBlack;
32             ++time;
```

```
33         u.finishing_time = time;
34     }
35     else
36     {
37         v_idx = *v_it;
38         Vertex& v = graph.vertices[v_idx];
39         if (v.color == Color::kWhite)
40         {
41             ++time;
42             v.discovery_time = time;
43             v.color = Color::kGray;
44             v.prev = u_idx;
45             path.push({ v_idx, graph.adj[v_idx].begin(), graph.adj[v_idx].end() });
46         }
47         ++v_it;
48     }
49 }
50 }
```

22.3-8

Consider the following directed graph:



Let s be the source vertex, and let u be in front of v in the adjacency list of s . Thus, we have the following result of DFS:

vertex	d	f
s	0	5
u	1	2
v	3	4

There exist a path from u to v : $u \rightarrow s \rightarrow v$.

22.3-9

Consider the same case as the solution of 22.3-8.

22.3-10

Directed graph:

```
1  struct Edge
2  {
3      int vertex_from;
4      int vertex_to;
5      enum Type { kTree, kBack, kForward, kCross } type;
6
7      Edge(int vertex_from, int vertex_to, Type type)
8          : vertex_from(vertex_from), vertex_to(vertex_to), type(type) {}
9  };
10
11 void DFSVisit(Graph& graph, int u_idx, int& time, std::list<Edge>& result)
12 {
13     Vertex& u = graph.vertices[u_idx];
14     ++time;
15     u.discovery_time = time;
16     u.color = Color::kGray;
17     for (int v_idx : graph.adj[u_idx])
18     {
19         Vertex& v = graph.vertices[v_idx];
20         if (v.color == Color::kWhite)
21         {
22             result.emplace_back(u_idx, v_idx, Edge::kTree);
23             v.prev = u_idx;
24             DFSVisit(graph, v_idx, time, result);
25         }
26         else if (v.color == Color::kGray)
27         {
28             result.emplace_back(u_idx, v_idx, Edge::kBack);
29         }
30         else if (u.discovery_time < v.discovery_time)
31         {
32             result.emplace_back(u_idx, v_idx, Edge::kForward);
33         }
34         else
35         {
36             result.emplace_back(u_idx, v_idx, Edge::kCross);
37         }
38     }
39     u.color = Color::kBlack;
40     ++time;
```

```
41     u.finishing_time = time;
42 }
43
44 std::list<Edge> DFS(Graph& graph)
45 {
46     int time, u_idx;
47     std::list<Edge> result;
48     for (Vertex& u : graph.vertices)
49     {
50         u.color = Color::kWhite;
51         u.prev = -1;
52     }
53     time = 0;
54     for (u_idx = 0; u_idx < graph.size; ++u_idx)
55     {
56         if (graph.vertices[u_idx].color == Color::kWhite)
57         {
58             DFSVisit(graph, u_idx, time, result);
59         }
60     }
61     return result;
62 }
```

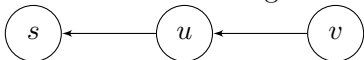
Undirected graph:

```
1 void DFSVisitUndirected(Graph& graph, int u_idx, int& time, std::list<Edge>& result)
2 {
3     Vertex& u = graph.vertices[u_idx];
4     ++time;
5     u.discovery_time = time;
6     u.color = Color::kGray;
7     for (int v_idx : graph.adj[u_idx])
8     {
9         Vertex& v = graph.vertices[v_idx];
10        if (u.prev != v_idx)
11        {
12            if (v.color == Color::kWhite)
13            {
14                result.emplace_back(u_idx, v_idx, Edge::kTree);
15                v.prev = u_idx;
16                DFSVisitUndirected(graph, v_idx, time, result);
17            }
18        }
19    }
20 }
```

```
17         }
18         else if (v.color == Color::kGray)
19         {
20             result.emplace_back(u_idx, v_idx, Edge::kBack);
21         }
22     }
23 }
24 u.color = Color::kBlack;
25 ++time;
26 u.finishing_time = time;
27 }
28
29 std::list<Edge> DFSUndirected(Graph& graph)
30 {
31     int time, u_idx;
32     std::list<Edge> result;
33     for (Vertex& u : graph.vertices)
34     {
35         u.color = Color::kWhite;
36         u.prev = -1;
37     }
38     time = 0;
39     for (u_idx = 0; u_idx < graph.size; ++u_idx)
40     {
41         if (graph.vertices[u_idx].color == Color::kWhite)
42         {
43             DFSVisitUndirected(graph, u_idx, time, result);
44         }
45     }
46     return result;
47 }
```

22.3-11

Consider the following directed graph:



Let s be the first source vertex, let u be the second, and let v be the third. Thus, we have the following result of DFS:

vetex	d	f
s	1	2
u	3	4
v	5	6

There exist a depth-first tree containing only u .

22.3-12

```
1 void DFSVisitUndirected(Graph& graph, int u_idx, int& time, int cc)
2 {
3     Vertex& u = graph.vertices[u_idx];
4     ++time;
5     u.discovery_time = time;
6     u.color = Color::kGray;
7     u.connecting_component = cc;
8     for (int v_idx : graph.adj[u_idx])
9     {
10         Vertex& v = graph.vertices[v_idx];
11         if (u.prev != v_idx)
12         {
13             if (v.color == Color::kWhite)
14             {
15                 v.prev = u_idx;
16                 DFSVisitUndirected(graph, v_idx, time, cc);
17             }
18         }
19     }
20     u.color = Color::kBlack;
21     ++time;
22     u.finishing_time = time;
23 }
24
25 void DFSUndirected(Graph& graph)
26 {
27     int time, u_idx, cc;
28     for (Vertex& u : graph.vertices)
29     {
30         u.color = Color::kWhite;
31         u.prev = -1;
32     }
```

```
33     time = 0;
34     cc = 0;
35     for (u_idx = 0; u_idx < graph.size; ++u_idx)
36     {
37         if (graph.vertices[u_idx].color == Color::kWhite)
38         {
39             ++cc;
40             DFSVisitUndirected(graph, u_idx, time, cc);
41         }
42     }
43 }
```

22.3-13

We can run DFS multiple times with source on each vertex on the graph. A directed graph is not signly connected if there exist some forward edges or cross edges that connect vertices in the same depth-first tree (i.e. in the same component). Hence, we have the following algorithm runs in $O(|V|(|V| + |E|))$:

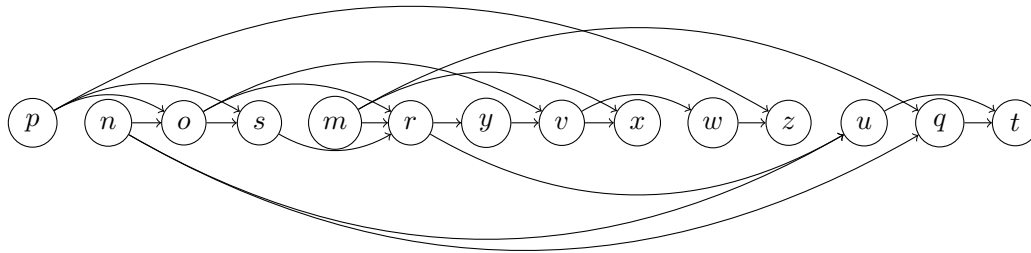
```
1  bool DFSVisit(Graph& graph, int u_idx, int& time)
2  {
3      Vertex& u = graph.vertices[u_idx];
4      ++time;
5      u.discovery_time = time;
6      u.color = Color::kGray;
7      for (int v_idx : graph.adj[u_idx])
8      {
9          Vertex& v = graph.vertices[v_idx];
10         if (v.color == Color::kWhite)
11         {
12             v.prev = u_idx;
13             DFSVisit(graph, v_idx, time);
14         }
15         else if (v.color == Color::kBlack)
16         {
17             return false;
18         }
19     }
20     u.color = Color::kBlack;
21     ++time;
22     u.finishing_time = time;
```

```
23     return true;
24 }
25
26 bool DFSWithSrc(Graph& graph, int src_vertex)
27 {
28     int time, u_idx;
29     for (Vertex& u : graph.vertices)
30     {
31         u.color = Color::kWhite;
32         u.prev = -1;
33     }
34     time = 0;
35     return DFSVisit(graph, src_vertex, time);
36 }
37
38 bool CheckSinglyConnected(Graph& graph)
39 {
40     int u_idx;
41     for (u_idx = 0; u_idx < graph.size; ++u_idx)
42     {
43         if (DFSWithSrc(graph, u_idx) == false)
44             return false;
45     }
46     return true;
47 }
```

22.4

22.4-1

vetex	d	f
m	1	20
n	21	26
o	22	25
p	27	28
q	2	5
r	6	19
s	23	24
t	3	4
u	7	8
v	10	17
w	11	14
x	15	16
y	9	18
z	12	13



22.4-2

We can perform topological sort on G and utilize dynamic programming to count simple path in linear time.

Denote $c[u]$ as the number of simple paths from u to t in G . We have the following recursive solution:

$$c[u] = \begin{cases} 1 & \text{if } u = t, \\ 0 & \text{if } u.f < t.f, \\ \sum_{(u,v) \in E} c[v] & \text{if } u.f > t.f. \end{cases}$$

The following bottom-up algorithm runs in $O(V + E)$:

```
1  // graph must be DAG
```

```
2  // s and t must be in the graph
3  int CountSimplePaths(Graph& graph, int s, int t)
4  {
5      int u;
6      std::vector<int> dp_count(graph.size, 0);
7      std::list<int> topo_sort = TopologicalSort(graph);
8      // find t
9      std::list<int>::reverse_iterator rit = topo_sort.rbegin();
10     while (rit != topo_sort.rend())
11     {
12         if (*rit == s)
13         {
14             return 0;
15         }
16         else if (*rit == t)
17         {
18             ++rit;
19             break;
20         }
21         ++rit;
22     }
23     // perform dp
24     dp_count[t] = 1;
25     while (rit != topo_sort.rend())
26     {
27         u = *rit;
28         for (int v : graph.adj[u])
29         {
30             dp_count[u] += dp_count[v];
31         }
32         if (u == s)
33             break;
34         ++rit;
35     }
36     return dp_count[s];
37 }
```

22.4-3

A forest contains multiple free trees (components). By Theorem B.2, $|E| = |V| - 1$ in a free tree. Hence, we can perform DFS on the graph. For each component, if the number of edges is more than the number of vertices, we can determine the graph contains a cycle. Hence, we can bound the running time in $O(|V| + |E|) = O(|V|)$.

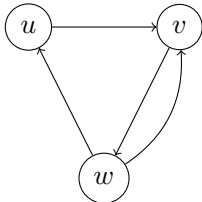
```
1  bool DFSVisit(Graph& graph, int u_idx, int& time)
2  {
3      Vertex& u = graph.vertices[u_idx];
4      ++time;
5      u.discovery_time = time;
6      u.color = Color::kGray;
7      for (int v_idx : graph.adj[u_idx])
8      {
9          Vertex& v = graph.vertices[v_idx];
10         if (u.prev != v_idx)
11         {
12             if (v.color == Color::kWhite)
13             {
14                 v.prev = u_idx;
15                 DFSVisit(graph, v_idx, time);
16             }
17             else if (v.color == Color::kGray)
18             {
19                 return false;
20             }
21         }
22     }
23     u.color = Color::kBlack;
24     ++time;
25     u.finishing_time = time;
26     return true;
27 }
28
29 bool IsAcyclic(Graph& graph)
30 {
31     int time, u_idx;
32     for (Vertex& u : graph.vertices)
33     {
34         u.color = Color::kWhite;
```

```

35     u.prev = -1;
36 }
37 time = 0;
38 for (u_idx = 0; u_idx < graph.size; ++u_idx)
39 {
40     if (graph.vertices[u_idx].color == Color::kWhite)
41     {
42         if (DFSVisit(graph, u_idx, time) == false)
43         {
44             return false;
45         }
46     }
47 }
48 return true;
49 }
```

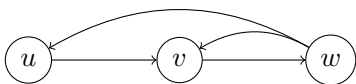
22.4-4

The claim is incorrect. Observed “bad” edges produced by topological sort are exactly back edges classified during DFS. However, DFS might not minimize back edges. Consider the following graph:



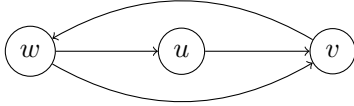
Let u be our source edge of DFS. Then we have the following DFS result and topological sort result which only has two “bad” edges: (w, u) and (w, v) .

vertex	d	f
u	1	6
v	2	5
w	3	4



However, if we let w be our source edge of DFS and let u be in front of v in the adjacency list of w , we have the following DFS result and topological sort result which has only one “bad” edge: (v, w) .

vetex	d	f
u	2	5
v	3	4
w	1	6



22.4-5

```
1  std::list<int> TopologicalSort(Graph& graph)
2  {
3      int u;
4      std::list<int> topo_sort;
5      std::vector<int> in_degree_counter(graph.size, 0);
6      std::queue<int> zero_in_degree;
7      // compute in-degree
8      for (AdjList& adj_list : graph.adj)
9      {
10         for (int v : adj_list)
11         {
12             ++in_degree_counter[v];
13         }
14     }
15     // init queue contains vertices with zero in-degree
16     for (u = 0; u < graph.size; ++u)
17     {
18         if (in_degree_counter[u] == 0)
19         {
20             zero_in_degree.push(u);
21         }
22     }
23     // perform topological sort
24     while (zero_in_degree.empty() == false)
25     {
26         u = zero_in_degree.front();
27         zero_in_degree.pop();
28         topo_sort.push_back(u);
29         for (int v : graph.adj[u])
30         {
```



```
31         --in_degree_counter[v];
32         if (in_degree_counter[v] == 0)
33         {
34             zero_in_degree.push(v);
35         }
36     }
37 }
38 return topo_sort;
39 }
```

If the graph has cycles, the array (std::vector) “in_degree_counter” contains non-zero values at the end of the procedure, so there exist some vertices not contained in the result of topological sort.

22.5

22.5-1

The number of strongly connected components of a graph will stay same or decrease if a new edge is added.

22.5-2

By Exercise 22.3-2, the result of DFS(G) (the first DFS) is the following:

vetex	d	f
q	1	16
s	2	7
v	3	6
w	4	5
t	8	15
x	9	12
z	10	11
y	13	14
r	17	20
u	18	19

Hence DFS(G^T) (the second DFS), in the main loop of DFS, consider the vetices in the following order: r u q t y x z s v w.

Therefore, the strongly connected components of the graph is the following:

r

u

q y t

x z

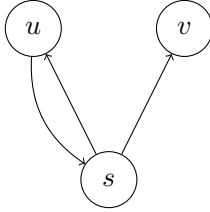
s w v

22.5-3

The claim is incorrect.

Construct the sequence $A = \langle a_1, a_2, \dots, a_{|V|} \rangle$ where $a_i \in V$ for $i \in [1, |V|]$. WLOG, let $a_i.f$ to be strictly increase for $i \in [1, |V|]$. Let C be arbitrary strongly connected component. Let $B = \langle b_1, b_2, \dots, b_{|C|} \rangle$ be the subsequence of A such that elements in B are vertex of C . Note that $f(C) = \max_{u \in C} \{u.f\}$. This says $f(C) = b_{|C|}.f$. In the main loop of the second DFS, the procedure consider the vetices in the order of A . This says, in the case of $|C| > 1$, the procedure will consider b_1 before consider $b_{|C|}$, unlike the process in STRONGLY-CONNECTED-COMPONENTS. However, we have $b_1.f = \min_{u \in C} \{u.f\}$. Let C' be a strongly connected component that is distinct from C . Hence, by lemma 22.14, there might be a edge (b_1, v) where $v \in C'$ and $b_1.f < v.f \leq f(C') < f(C)$ (especially $b_1.f < f(C')$), which means depth-first forest produced by the second DFS might not output strongly connected components correctly.

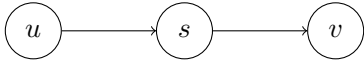
Consider the following counterexample:



Let s be the source vertex in the first DFS, and let u be in front of v in the adjacency list of s . The result of the first DFS is the following:

vetex	d	f
s	1	6
u	2	3
v	4	5

Then the second DFS, in the main loop, will consider the vetices in the following order: u v s. Hence the second DFS will produce the following depth-first forest:



The depth-first forest implies u-s-v are in a single strongly connected component, which is incorrect.

22.5-4

Proof. Let $G^{SCC} = (V^{SCC}, E^{SCC})$, Observed that G and G^T have exactly the same strongly connected components (page 616), we have $(G^T)^{SCC} = (V^{SCC}, (E^T)^{SCC})$. To show $G^{SCC} = ((G^T)^{SCC})^T$, we claim $E^{SCC} = ((E^T)^{SCC})^T$.

(\subseteq) Suppose that $(u, v) \in E^{SCC}$. Then $u, v \in V^{SCC}$ and $(u, v) \in E$ where u and v are in distinct components. By definition of E^T , $(v, u) \in E^T$. Since u and v are in distinct components and $u, v \in V^{SCC}$, we have $(v, u) \in (E^T)^{SCC}$. Hence $(u, v) \in ((E^T)^{SCC})^T$.

(\supseteq) Similar to the proof of (\subseteq). □

22.5-5

```
1  void DFSVisitComputeComponentGraph(Graph& graph, int u_idx,
2      std::list<int>& adj_list)
3  {
4      Vertex& u = graph.vertices[u_idx];
5      for (int v_idx : graph.adj[u_idx])
6      {
7          Vertex& v = graph.vertices[v_idx];
8          if (v.component < 0)
9          {
10             v.component = u.component;
11             DFSVisitComputeComponentGraph(graph, v_idx, adj_list);
12         }
13         else if (v.component != u.component)
14         {
15             adj_list.push_back(v.component);
16         }
17     }
18 }
19
20 std::list< std::list<int> > DFSComputeComponentGraphAdjList(Graph& graph,
21     const std::list<int>& order)
22 {
23     int curr_component, v_idx;
24     std::list< std::list<int> > adj_lists;
25     for (Vertex& u : graph.vertices)
26     {
27         u.component = INT_MIN;
28     }
29     curr_component = 0; // index of component graph start from 0
30     for (int u_idx : order)
31     {
32         Vertex& u = graph.vertices[u_idx];
33         if (u.component < 0)
```

```
34     {
35         // DFS find edges (in adj_list) connected with curr_component
36         u.component = curr_component;
37         adj_lists.emplace_back(std::list<int>());
38         DFSVisitComputeComponentGraph(graph, u_idx, adj_lists.back());
39         ++curr_component;
40     }
41 }
42 return adj_lists;
43 }
44
45 Graph ComponentGraphAdjLists(Graph& graph)
46 {
47     int i;
48     std::list<int> topo_sort = TopologicalSort(graph);
49     Graph tranpose = Transpose(graph);
50     std::list< std::list<int> > tranpose_component_graph_adj_lists =
51         DFSComputeComponentGraphAdjList(tranpose, topo_sort);
52     // build ComponentGraph
53     Graph component_graph(tranpose_component_graph_adj_lists.size());
54     std::list< std::list<int> >::iterator adj_lists_it =
55         tranpose_component_graph_adj_lists.begin();
56     for (i = 0; i < component_graph.size; ++i)
57     {
58         // adjacency list might exist repeat elements
59         // we delete repeat elements by utilizing counter
60         // Note that the adjacency list is for  $(G^T)^{SCC}$ 
61         // By Exercise 22.5-4, we showed that  $G^{SCC} = ((G^T)^{SCC})^T$ 
62         std::vector<bool> counter(i + 1, false);
63         for (std::list<int>::iterator it = adj_lists_it->begin();
64             it != adj_lists_it->end(); ++it)
65         {
66             if (counter[*it] == false)
67             {
68                 counter[*it] = true;
69                 component_graph.adj[*it].push_back(i);
70             }
71         }
72         ++adj_lists_it;
73     }
```

```
74     for (i = 0; i < graph.size; ++i)
75     {
76         component_graph.vertices_lists_from_original_graph
77             [transpose.vertices[i].component].push_back(i);
78     }
79     return component_graph;
80 }
```

22.5-6

Notice that a simple directed cycle constructed by vertices of the component minimized edges in each component. We can utilize the procedure in 25.5-5 to get component graph in $O(V + E)$. Therefore, we have the following algorithm which takes $O(V + E)$ time:

```
1  Graph CreateGraphExercise22_5_6(Graph& graph)
2  {
3      int component_idx, u;
4      Graph component_graph = ComponentGraphAdjLists(graph);
5      Graph result_graph(graph.size);
6      for (component_idx = 0; component_idx < component_graph.size; ++component_idx)
7      {
8          // build cycle
9          std::list<int>& vertices =
10             component_graph.vertices_lists_from_original_graph[component_idx];
11         if (vertices.front() != vertices.back())
12         {
13             u = vertices.back();
14             for (int v : vertices)
15             {
16                 result_graph.adj[u].push_back(v);
17                 u = v;
18             }
19         }
20         // build edges among components
21         u = vertices.front();
22         std::list<int>& adj = component_graph.adj[component_idx];
23         for (int c : adj)
24         {
25             int v = component_graph.vertices_lists_from_original_graph[c].front();
26             result_graph.adj[u].push_back(v);
27         }
```

```

28     }
29     return result_graph;
30 }
```

22.5-7

Let $G = (V, E)$ and $G^{SCC} = (V^{SCC}, E^{SCC})$. Suppose that G has strongly connected components C_1, C_2, \dots, C_k . The vertex set V^{SCC} is $\{s_1, s_2, \dots, s_k\}$, and it contains a vertex s_i for each strongly connected component C_i of G .

Lemma 4. Let $x \in C_i$ and $y \in C_j$. $x \rightsquigarrow y \iff s_i \rightsquigarrow s_j$

Proof. (\implies) Suppose that $x \rightsquigarrow y$. Then $\exists w_1, w_2, \dots, w_p \in V$ such that $(x, w_1), (w_1, w_2), \dots, (w_p, y) \in E$. Suppose that c_1, c_2, \dots, c_p are the indexes of components contain w_1, w_2, \dots, w_p ; i.e. $w_l \in C_{c_l}$ for all $l \in [1, p]$. By definition of E^{SCC} , we have $(s_i, s_{c_1}), (s_{c_1}, s_{c_2}), \dots, (s_{c_p}, s_j) \in E^{SCC}$. Thus, $s_i \rightsquigarrow s_j$.

(\impliedby) Suppose that $s_i \rightsquigarrow s_j$. Then $\exists w_1, w_2, \dots, w_p \in V^{SCC}$ such that $(s_i, w_1), (w_1, w_2), \dots, (w_p, s_j) \in E^{SCC}$. Suppose that c_1, c_2, \dots, c_p are the indexes of components w_1, w_2, \dots, w_p ; i.e. $w_l = s_{c_l}$ for all $l \in [1, p]$. Pick arbitrary $u_l \in C_{c_l}$ for all $l \in [1, p]$. By definition of strongly connected components and E^{SCC} , we have $x \rightsquigarrow u_1 \rightsquigarrow u_2 \dots u_p \rightsquigarrow y$. Thus, $x \rightsquigarrow y$. \square

Lemma 5. G is semiconnected if and only if G^{SCC} is semiconnected.

Proof. (\implies) Suppose that G is semiconnected. That is $\forall x, y \in V, x \rightsquigarrow y \vee y \rightsquigarrow x$. We prove by induction on the number of strongly connected components of G . Our inductive hypothesis is $\forall u, v \in \{s_1, s_2, \dots, s_k\}, u \rightsquigarrow v \vee v \rightsquigarrow u$. The base case is when G contains no vertex, which is trivial. Suppose that $\forall u, v \in \{s_1, s_2, \dots, s_k\}, u \rightsquigarrow v \vee v \rightsquigarrow u$. Let $x \in C_t$ where $t \in [1, k]$ and $y \in C_{k+1}$ be arbitrary choices. Since $x, y \in V$, we have $x \rightsquigarrow y \vee y \rightsquigarrow x$. If $x \rightsquigarrow y$, then, by lemma 4, we have $s_t \rightsquigarrow s_{k+1}$. Similarly, if $y \rightsquigarrow x$, we have $s_{k+1} \rightsquigarrow s_t$. Since $t \in [1, k]$ is a arbitrary choice, by induction, we conclude G^{SCC} is semiconnected.

(\impliedby) Suppose that G^{SCC} is semiconnected. That is $\forall x, y \in V^{SCC}, x \rightsquigarrow y \vee y \rightsquigarrow x$. We prove by induction on the number of strongly connectly components of G . Our inductive hypothesis is $\forall u, v \in C_1 \cup C_2 \dots C_k, u \rightsquigarrow v \vee v \rightsquigarrow u$. The base case is when G contains no vertex, which is trivial. Suppose that $\forall u, v \in C_1 \cup C_2 \dots C_k, u \rightsquigarrow v \vee v \rightsquigarrow u$. Let $x \in C_t$ where $t \in [1, k]$ and $y \in C_{k+1}$ be arbitrary choices. Since $s_t, s_{k+1} \in V^{SCC}$, we have $s_t \rightsquigarrow s_{k+1} \vee s_{k+1} \rightsquigarrow s_t$. If $s_t \rightsquigarrow s_{k+1}$, then, by lemma 4, we have $x \rightsquigarrow y$. Similarly, if $s_{k+1} \rightsquigarrow s_t$, we have $y \rightsquigarrow x$. Since $t \in [1, k]$ is a arbitrary choice, by induction, we conclude G is semiconnected. \square

Lemma 6. G is semiconnected if and only if we are able to put all elements of V in a sequence $\langle v_1, v_2, \dots, v_{|V|} \rangle$ such that $v_1 \rightsquigarrow v_2 \dots v_{|V|-1} \rightsquigarrow v_{|V|}$.

Proof. (\implies) Suppose that G is semiconnected. That is $\forall u, v \in V, u \rightsquigarrow v \vee v \rightsquigarrow u$. Let $A = \langle a_1, a_2, \dots, a_k \rangle$ be a sequence. We prove by induction on k . The base case is $k = 0$, which is trivial. Suppose that $a_1 \rightsquigarrow a_2 \dots a_{k-1} \rightsquigarrow a_k$. Pick arbitrary element $u \in V$ and $u \notin A$. If $\exists i \in [1, k)$ such that $a_i \rightsquigarrow u \rightsquigarrow a_{i+1}$, then we insert u between a_i and a_{i+1} ; otherwise, insert u

before a_1 or after a_k appropriately. Re-assigning index of A , we get $A = \langle a_1, a_2, \dots, a_k, a_{k+1} \rangle$ where $a_1 \rightsquigarrow a_2 \dots a_k \rightsquigarrow a_{k+1}$.

(\Leftarrow) Immediately. □

Lemma 7. Let G be arbitrary DAG. If we are able to put all elements of V in a sequence $\langle v_1, v_2, \dots, v_{|V|} \rangle$ such that $v_1 \rightsquigarrow v_2 \dots v_{|V|-1} \rightsquigarrow v_{|V|}$, then $\{(v_1, v_2), (v_2, v_3), \dots, (v_{|V|-1}, v_{|V|})\} \subseteq E$ and the result of topological sort order of G is $v_1, v_2, \dots, v_{|V|}$.

Proof. Suppose that $v_1 \rightsquigarrow v_2 \dots v_{|V|-1} \rightsquigarrow v_{|V|}$. If $|V| = 1$, the lemma is trivial correct. Let integer $i \in [1, |V|)$. Assume $(v_i, v_{i+1}) \notin E$, for the purpose of contradiction. Then $\exists j \neq i, i+1$ such that $v_i \rightsquigarrow v_j \rightsquigarrow v_{i+1}$. If $|V| = 2$, we obviously have a contradiction. If $j < i$, we have $v_i \rightsquigarrow v_j$ and $v_j \rightsquigarrow v_i$. If $j > i+1$, we have $v_j \rightsquigarrow v_{i+1}$ and $v_{i+1} \rightsquigarrow v_j$. Contradict to G is a DAG.

In order to get other order of $\langle v_1, v_2, \dots, v_{|V|} \rangle$. we must swap some elements in the sequence, and no linear ordering is possible, which contradicts to the definition of topological sort. □

Theorem 8. (Correctness) Suppose that the result of topological sort order of G^{SCC} is $v_1, v_2, \dots, v_{|V^{SCC}|}$. G is semiconnected if and only if $\{(v_1, v_2), (v_2, v_3), \dots, (v_{|V^{SCC}|-1}, v_{|V^{SCC}|})\} \subseteq E^{SCC}$

Proof. G is semiconnected $\xLeftrightarrow{\text{lemma 5}} G^{SCC}$ is semiconnected $\xLeftrightarrow{\text{lemma 6}}$

$\exists \langle v_1, v_2, \dots, v_{|V^{SCC}|} \rangle$ where $v \in V^{SCC}$ such that $v_1 \rightsquigarrow v_2 \dots v_{|V^{SCC}|-1} \rightsquigarrow v_{|V^{SCC}|}$

(\Rightarrow) Suppose that $v_1 \rightsquigarrow v_2 \dots v_{|V^{SCC}|-1} \rightsquigarrow v_{|V^{SCC}|}$. Since G^{SCC} is a DAG, by lemma 7, we have $\{(v_1, v_2), (v_2, v_3), \dots, (v_{|V^{SCC}|-1}, v_{|V^{SCC}|})\} \subseteq E$.

(\Leftarrow) Suppose that $\{(v_1, v_2), (v_2, v_3), \dots, (v_{|V^{SCC}|-1}, v_{|V^{SCC}|})\} \subseteq E^{SCC}$. Then, obviously, we have $v_1 \rightsquigarrow v_2 \dots v_{|V^{SCC}|-1} \rightsquigarrow v_{|V^{SCC}|}$. □

Therefore, we have the following $O(V + E)$ algorithm:

```
1  bool Semiconnected(Graph& graph)
2  {
3      int u;
4      Graph component_graph = ComponentGraphAdjLists(graph);
5      std::list<int> topo_sort = TopologicalSort(component_graph);
6      std::list<int>::iterator it = topo_sort.begin();
7      std::list<int>::iterator end = topo_sort.end();
8      u = *it;
9      ++it;
10     while (it != end)
11     {
12         if (std::find(component_graph.adj[u].begin(), component_graph.adj[u].end(),
13             *it) == component_graph.adj[u].end())
14             return false;
15         u = *it;
16     }
```

```
17     return true;  
18 }
```

Updating...