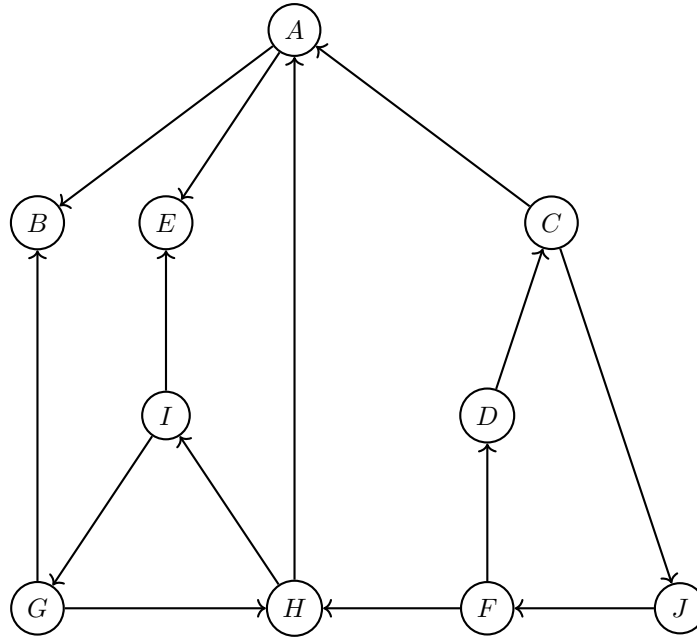


*Note:* Your TA probably will not cover all the problems. This is totally fine, the discussion worksheets are deliberately made long so they can serve as a resource you can use to practice, reinforce, and build upon concepts discussed in lecture, readings, and the homework.

## 1 Graph Traversal



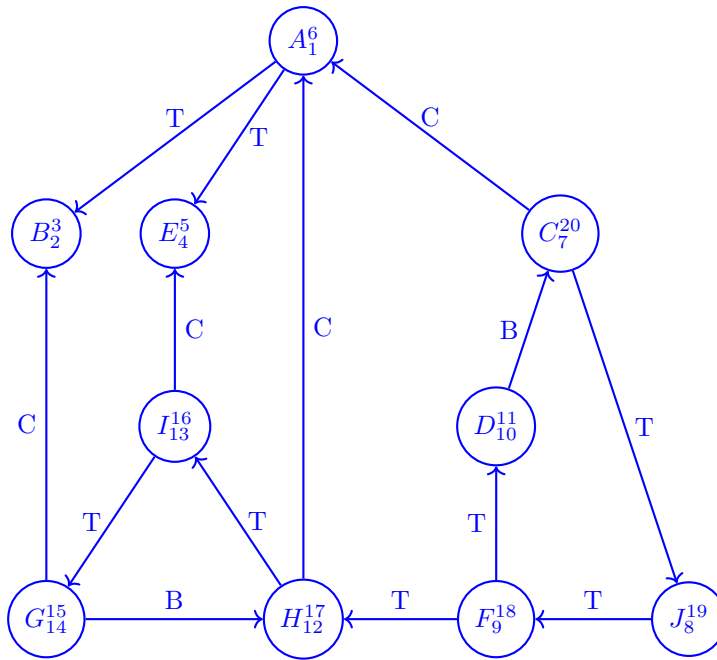
- (a) Recall that given a DFS tree, we can classify edges into one of four types:
- Tree edges are edges in the DFS tree,
  - Back edges are edges  $(u, v)$  not in the DFS tree where  $v$  is the ancestor of  $u$  in the DFS tree
  - Forward edges are edges  $(u, v)$  not in the DFS tree where  $u$  is the ancestor of  $v$  in the DFS tree
  - Cross edges are edges  $(u, v)$  not in the DFS tree where  $u$  is not the ancestor of  $v$ , nor is  $v$  the ancestor of  $u$ .

For the directed graph above, perform DFS starting from vertex A, breaking ties alphabetically. As you go, label each node with its pre- and post-number, and mark each edge as **T**ree, **B**ack, **F**orward or **C**ross.

- (b) A strongly connected component (SCC) is defined as a subset of vertices in which there exists a path from each vertex to each other vertex. What are the SCCs of the above graph?
- (c) Collapse each SCC you found in part (b) into a meta-node, so that you end up with a graph of the SCC meta-nodes. Draw this graph below, and describe its structure.

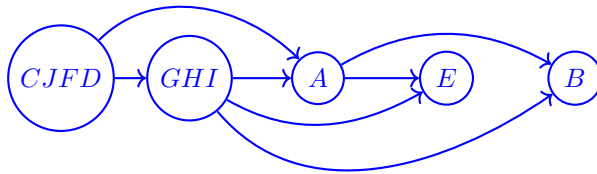
**Solution:**

(a)



(b)  $\{A\}, \{B\}, \{E\}, \{G, H, I\}, \{C, J, F, D\}$

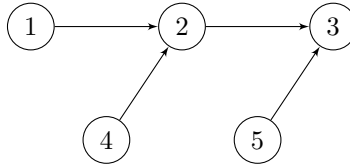
(c)



## 2 Finding Clusters

We are given a directed graph  $G = (V, E)$ , where  $V = \{1, \dots, n\}$ , i.e. the vertices are integers in the range 1 to  $n$ . For every vertex  $i$  we would like to compute the value  $m(i)$  defined as follows:  $m(i)$  is the smallest  $j$  such from which you can reach vertex  $i$ . (As a convention, we assume that  $i$  is reachable from  $i$ .)

**Example:** Consider the following directed graph with 5 vertices:



The  $m(i)$  values are:

- $m(1) = 1$ : Only vertex 1 can reach vertex 1 (itself).
- $m(2) = 1$ : Vertices 1, 2, and 4 can reach vertex 2. The smallest is 1.
- $m(3) = 1$ : Vertices 1, 2, 3, 4, and 5 can all reach vertex 3. The smallest is 1.
- $m(4) = 4$ : Only vertex 4 can reach vertex 4 (itself).
- $m(5) = 5$ : Only vertex 5 can reach vertex 5 (itself).

(a) Show that the values  $m(1), \dots, m(n)$  can be computed in  $O(|V| + |E|)$  time.

**Solution:** The algorithm is as follows.

**procedure** DFS-CLUSTERS( $G$ )

**while** there are unvisited nodes in  $G$  **do**

    Run DFS on  $G$  starting from the numerically-first unvisited node  $j$

**for**  $i$  visited by this DFS **do**  $m(i) := j$

To see that this algorithm is correct, note that if a vertex  $i$  is assigned a value then that value is the smallest of the nodes that can reach it in  $G$ , and every node is assigned a value because the loop does not terminate until this happens.

The running time is  $O(|V| + |E|)$  since the algorithm is just a modification of DFS.

(b) Suppose we instead define  $m(i)$  to be the smallest  $j$  that can be reached from  $i$ , instead of the smallest  $j$  from which you can reach  $i$ . Can we use the same DFS approach from part (a) If not, what goes wrong, and how can we fix it?

**Solution:** We might try a forward DFS approach: run DFS and track the minimum vertex reachable from each vertex. When we finish exploring a vertex  $v$ , we set  $m(v) = \min(v, \min_{(v,u) \in E} m(u))$ .

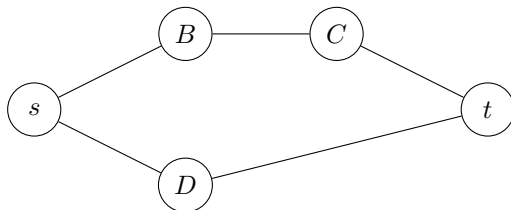
The problem is **back edges**. A back edge goes from a vertex to one of its ancestors in the DFS tree. When we're at vertex  $v$  and see a back edge to ancestor  $u$ , we know  $v$  can reach  $u$ . But  $u$  is still on the DFS stack, we haven't finished computing  $m(u)$  yet! And  $m(u)$  depends on  $m(v)$  (since  $u$  can reach  $v$  via tree edges). This circular dependency means we can't resolve the minimum in a single forward pass.

**The fix:** Reverse all edges in  $G$ , i.e., replace each edge  $(u, v)$  with the edge  $(v, u)$ , and then use our algorithm from part (a). This works because in the reversed graph  $G^R$ , we can reach  $j$  from  $i$  if and only if we can reach  $i$  from  $j$  in the original graph  $G$ . So finding the smallest vertex that can reach  $i$  in  $G^R$  is exactly finding the smallest vertex reachable from  $i$  in  $G$ .

### 3 Odd Shortest Path

Given an undirected graph  $G = (V, E)$  and two vertices  $s, t \in V$ , find the shortest path from  $s$  to  $t$  that uses an **odd number of edges**, or report that no such path exists. Your algorithm should run in  $O(|V| + |E|)$  time.

**Example:** Consider the following undirected graph:



The shortest path from  $s$  to  $t$  overall is  $s \rightarrow D \rightarrow t$  with length 2 (even). However, this path has an **even** number of edges, so it doesn't count!

The shortest **odd-length** path is  $s \rightarrow B \rightarrow C \rightarrow t$  with length 3.

**Solution:** The key idea is to transform the problem using **state duplication**. We create a new graph  $G' = (V', E')$  that tracks the parity (odd/even) of the number of edges traversed so far.

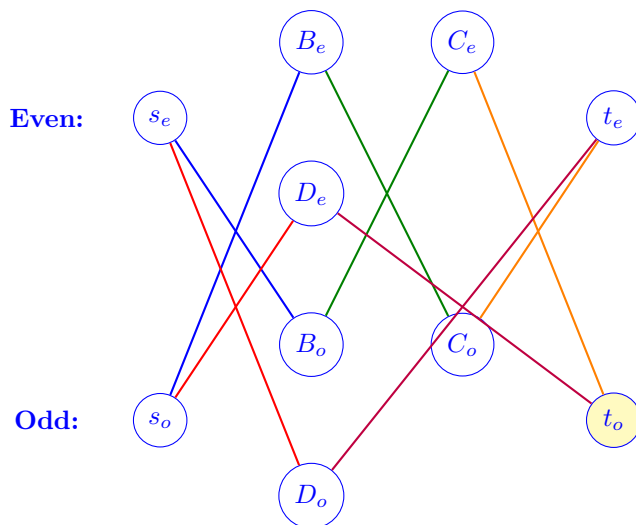
**Construction of  $G'$ :**

- For each vertex  $v \in V$ , create two copies:  $v_{\text{even}}$  and  $v_{\text{odd}}$ .
- For each edge  $(u, v) \in E$ , add edges:
  - $(u_{\text{even}}, v_{\text{odd}})$  crossing an edge from even parity goes to odd parity
  - $(u_{\text{odd}}, v_{\text{even}})$  crossing an edge from odd parity goes to even parity

**Algorithm:** Run BFS from  $s_{\text{even}}$  in  $G'$ . The answer is the distance to  $t_{\text{odd}}$ .

(Note: We don't technically need  $s_{\text{odd}}$  or  $t_{\text{even}}$  since we always start at  $s_{\text{even}}$  and only care about reaching  $t_{\text{odd}}$ , but including them keeps the construction simple.)

**Example:**



**Runtime:**  $|V'| = 2|V|$  and  $|E'| = 2|E|$ , so BFS runs in  $O(|V| + |E|)$  time.

## 4 Bottleneck Spanning Tree

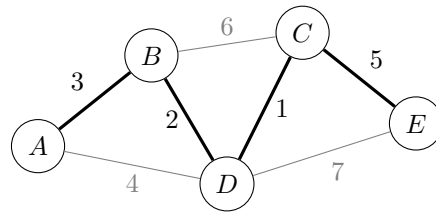
Recall that a spanning tree of a connected, undirected graph  $G = (V, E)$  is a subgraph  $T = (V, E_T)$  that:

- Contains all vertices of  $G$
- Is a tree (connected and acyclic)
- Uses only edges from  $E$

A **Minimum Spanning Tree (MST)** is a spanning tree that minimizes the *total* weight of all edges:

$$\text{MST minimizes } \sum_{e \in T} w(e)$$

**Example:** Consider the following weighted graph and its MST (bold edges):



A **Bottleneck Spanning Tree (BST)** is a spanning tree that minimizes the weight of the *heaviest* edge:

$$\text{BST minimizes } \max_{e \in T} w(e)$$

- (a) Is every Bottleneck Spanning Tree also a Minimum Spanning Tree? Prove or give a counterexample.

**Solution: No.** Consider the example graph above. The MST is  $\{A-B, B-D, D-C, C-E\}$  with weights  $\{3, 2, 1, 5\}$ , total weight 11, and max edge 5.

Now consider the alternative spanning tree  $\{A-D, B-D, D-C, C-E\}$  with weights  $\{4, 2, 1, 5\}$ . This tree has total weight 12 and max edge 5.

Both trees have the same maximum edge weight (5), so both are Bottleneck Spanning Trees. However, the second tree has a larger total weight ( $12 > 11$ ), so it is **not** a Minimum Spanning Tree.

- (b) Is every Minimum Spanning Tree also a Bottleneck Spanning Tree? Prove or give a counterexample.

**Solution:** Suppose  $T$  is an MST but not a Bottleneck Tree. Then there exists some other tree  $T_{\text{bottle}}$  with a smaller max-weight edge. Let  $e$  be the max-weight edge in  $T$ . Removing  $e$  splits  $T$  into two components.  $T_{\text{bottle}}$  must have an edge  $e'$  connecting these two components. Since  $T_{\text{bottle}}$  has a smaller max weight,  $w(e') < w(e)$ . But then we could swap  $e$  for  $e'$  to get a tree with total weight less than  $T$ , contradicting that  $T$  was an MST.