

COMP261

Algorithms and Data Structures

2024 Tri 1

Jyoti Sahni

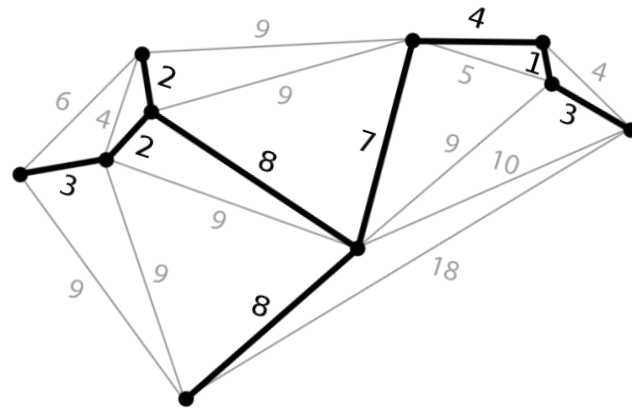
jyoti.sahni@ecs.vuw.ac.nz

Office Hours (COMP261): AM414, Thursday 10:00 – 12:00

Recap: Spanning Trees

Given a **connected, undirected**, weighted graph, a spanning tree is a subgraph that contains **all the nodes** but has no cycles (**is a tree**)

A spanning tree is defined only for a connected graph, because a tree is always connected, and in a disconnected graph of n vertices we cannot find a connected subgraph with n vertices



Recap: Spanning trees in Weighted graphs

The spanning-tree problem

- Add nodes to partial tree
- Add acyclic edges

Minimum-cost-spanning-tree problem

- Given a **connected, weighted, undirected** graph, find a spanning tree of minimum weight
- The above approaches suffice with minor changes:
 - Add nodes to partial tree approach: **Prim's Algorithm**
 - Add acyclic edges approach : **Kruskal's algorithm**

Prim's Algorithm

Given: a connected undirected weight graph

Initialize fringe to have a root node with costToTree = 0

all nodes are unvisited;

Repeat until all nodes are visited {

 Choose from fringe the unvisited node (n^*) with minimum costToTree;

 Add the corresponding edge to the spanning tree, set n^* as visited

 for each (edge (n^* , n') with one end-node n^*) {

 if (n' is not visited) then add $\langle n', (n^*, n'), \text{cost}(n^*, n') \rangle$ into the fringe;

 }

}

Kruskal's Algorithm

Given: a connected undirected weight graph (N nodes, M edges)

Initialize an empty edge set T .

Sort all graph edges by the **ascending order** of their weight values.

For each edge in the sorted edge list

 Check whether it will create a **cycle** with the edges inside T .

 If the edge doesn't introduce any cycles, add it into T .

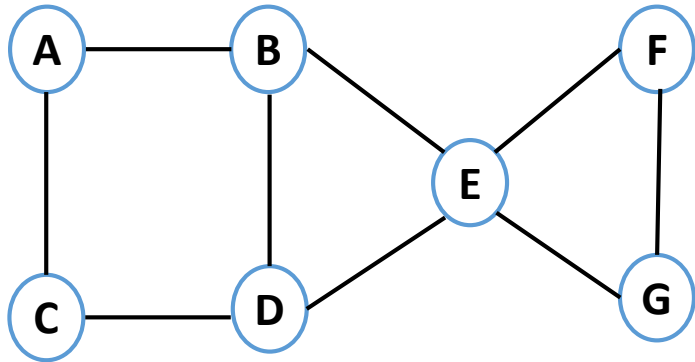
 If T has $(V-1)$ edges, exit the loop.

return T

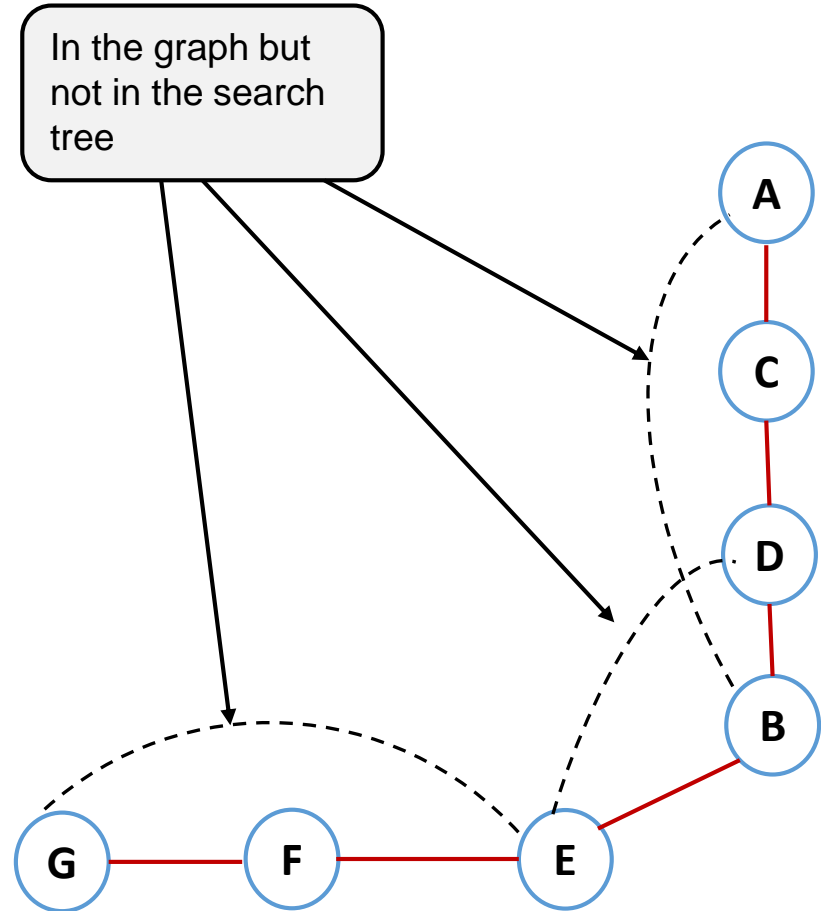
Complexity: Weighted Spanning Trees

- Depends on data structure
 - Naïve approach, if using adjacency list with **linear search**
 - Prim's : $O(|V|^2)$
 - Kruskal's: $O(\text{sorting of edges} + |V||V + E|) \sim O(|V|^2)$
// $|E| \approx$ some multiple of $|V|$
 - Priority queue
 - Prim's algorithm becomes similar to Dijkstra's
 - Complexity: $O(|E|\text{Log}|V|)$
- Can we do better in Kruskal's algorithm?
 - A new data structure: **disjoint sets**

Recap: Cycle Detection using DFS

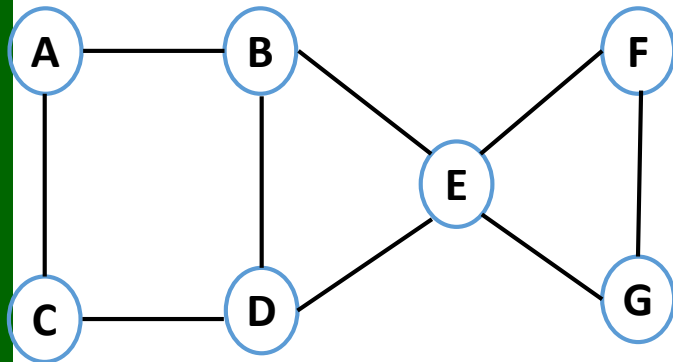


Back edge: Edge which is missing in the DFS tree but present in the graph



All the back edges which DFS skips over are part of cycles

Recap: Cycle Detection using DFS



- Steps:
 - Start DFS traversal
 - Keep track of parent of the node being visited
 - If you find a node that has already been visited but is not the parent of the current node being visited – there is a cycle

0	A	→ C, B
1	B	→ A, D, E
2	C	→ A, D
3	D	→ C, B, E
4	E	→ B, D, F, G
5	F	→ E, G
6	G	→ E, F

DFS (A, -1), visited(A) = true

DFS (C, A), visited(C) = true

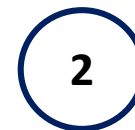
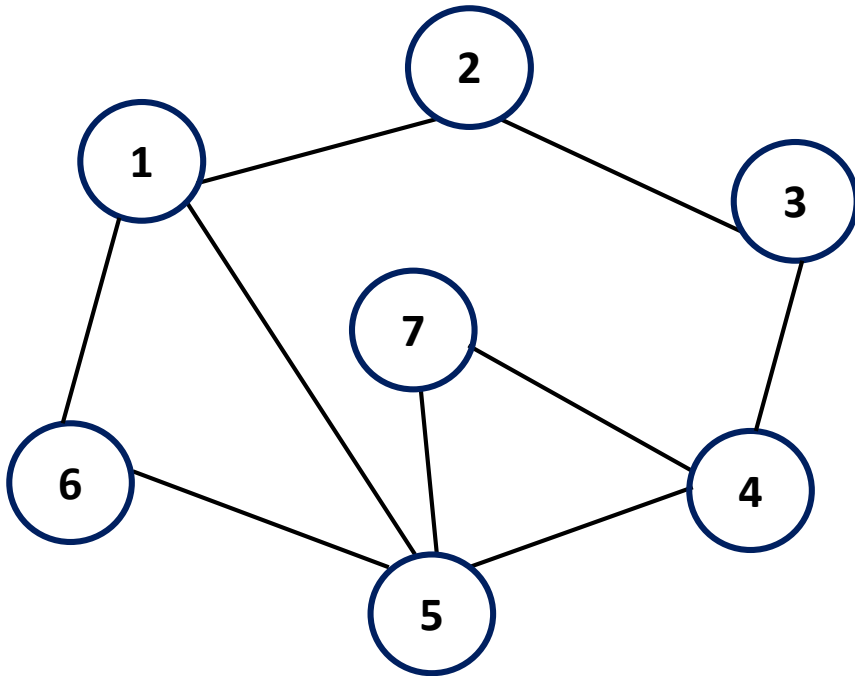
DFS (D, C), visited(D) = true

DFS (B, D), visited(B) = true

A has already been visited and $A \neq \text{parent}(B)$. Cycle found

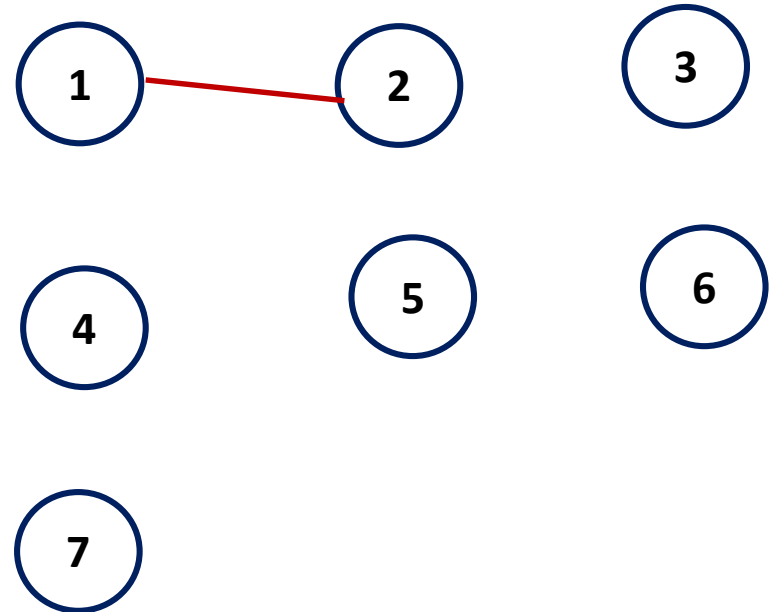
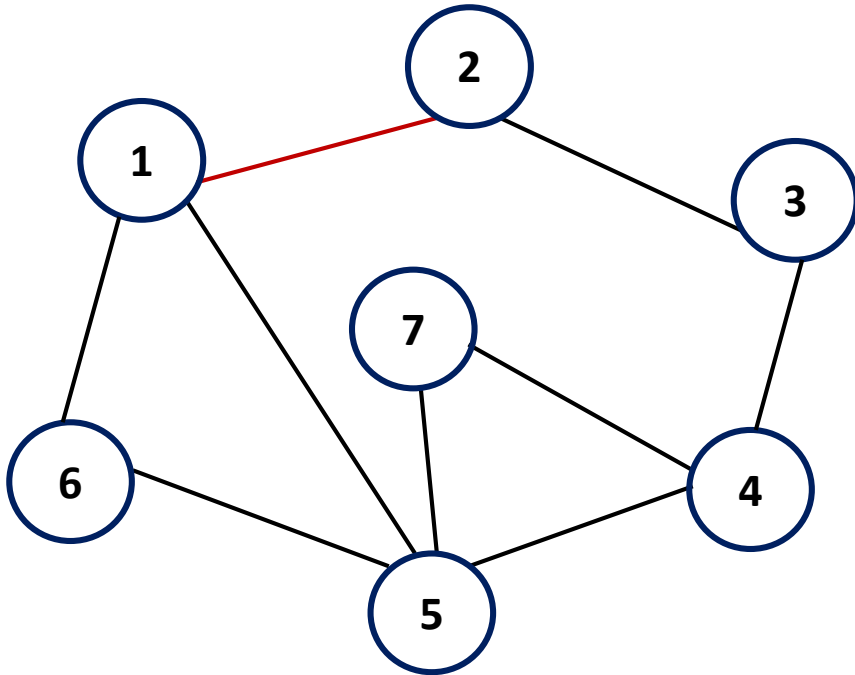
Finding a cycle – another approach

Order of edges considered: (1,2), (3,4), (5,6), (5,7), (1,5),
(1,6), (2,3), (4,7), (4,5)



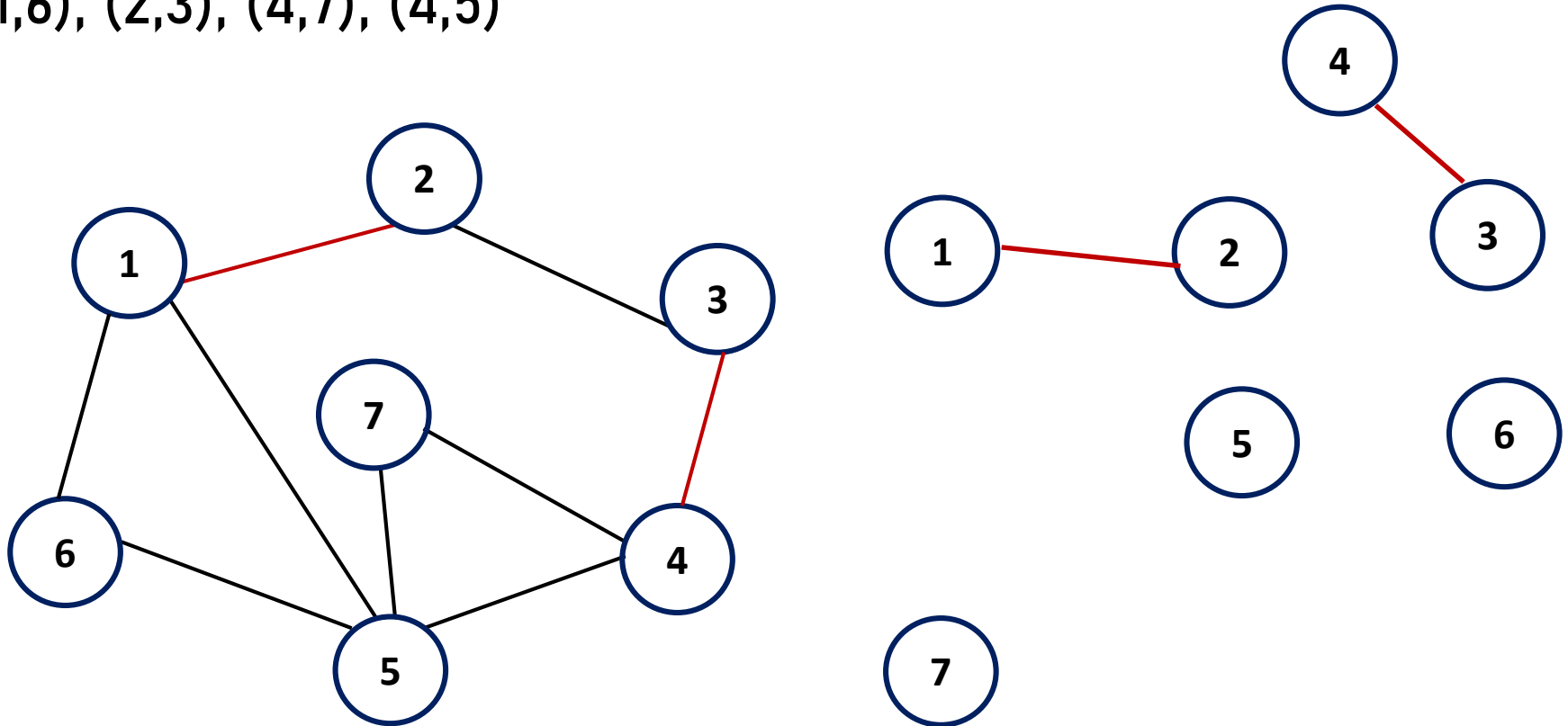
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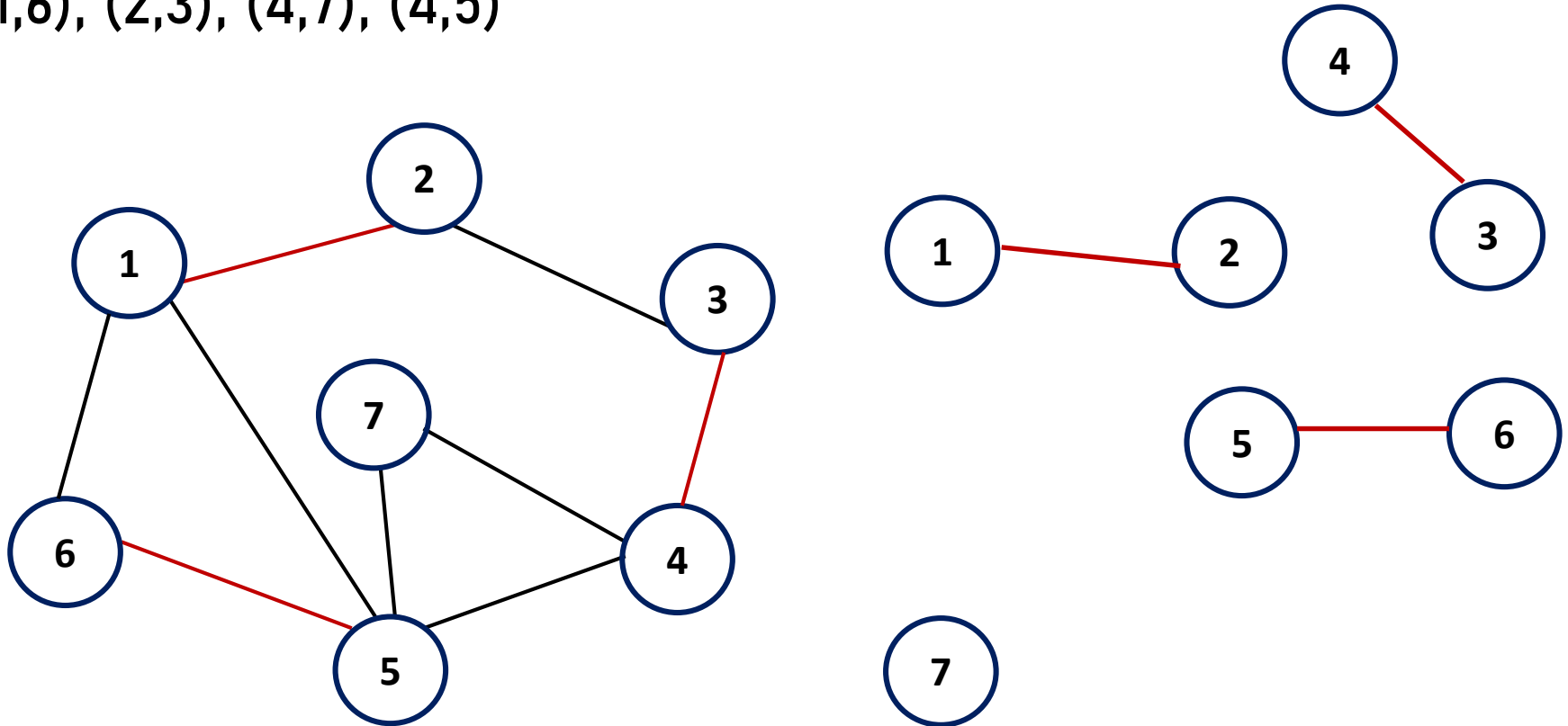
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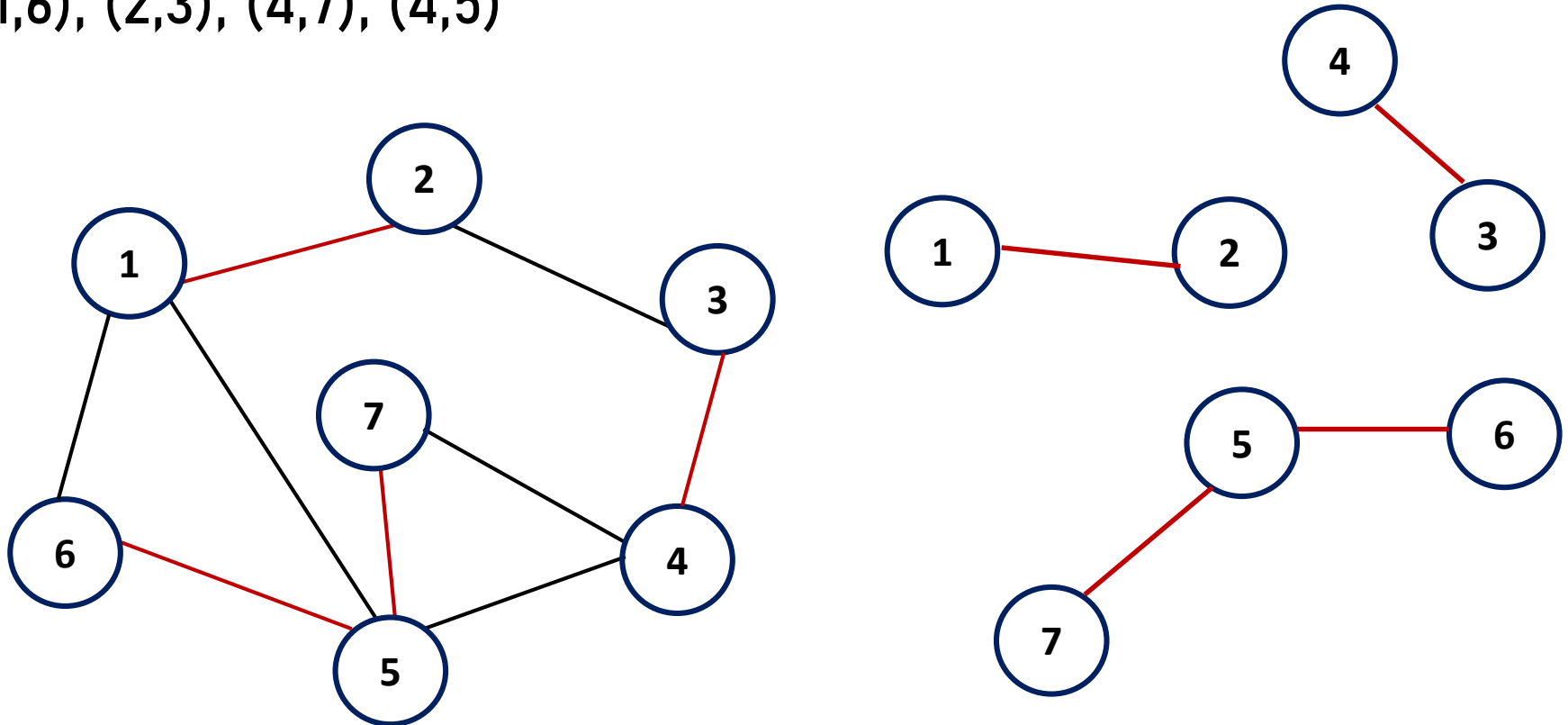
Finding a cycle – another approach

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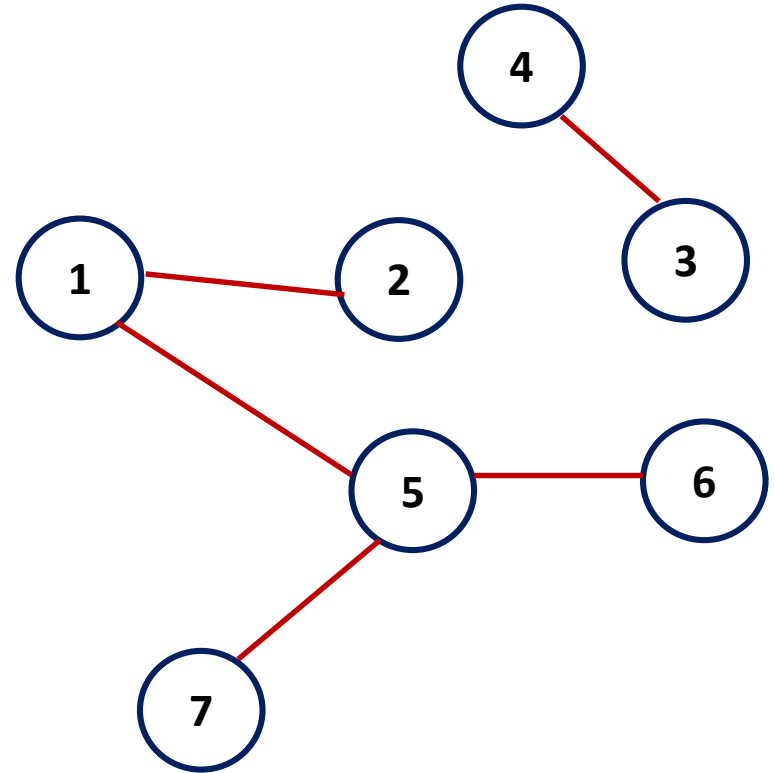
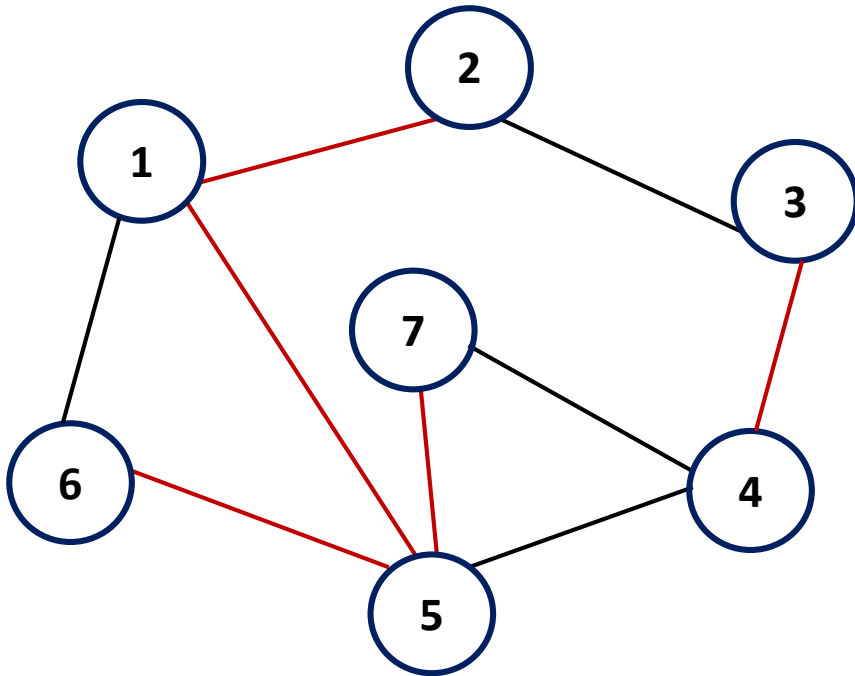
Finding a cycle – another approach

Order of edges considered: (1,2), (3,4), (5,6), (5,7), (1,5),
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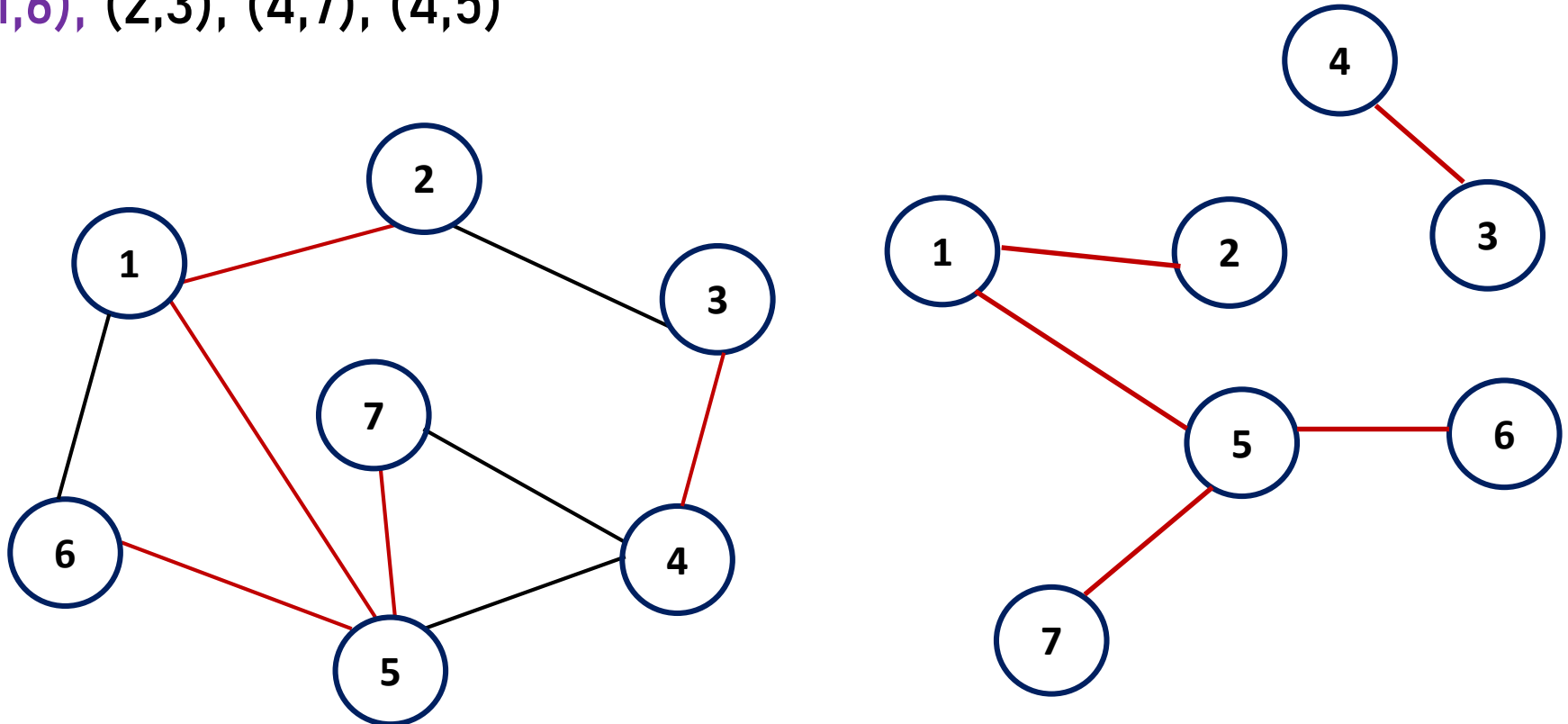
Finding a cycle – another approach

Order of edges considered: (1,2), (3,4), (5,6), (5,7), (1,5),
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Finding a cycle – another approach

Order of edges considered: (1,2), (3,4), (5,6), (5,7), (1,5),
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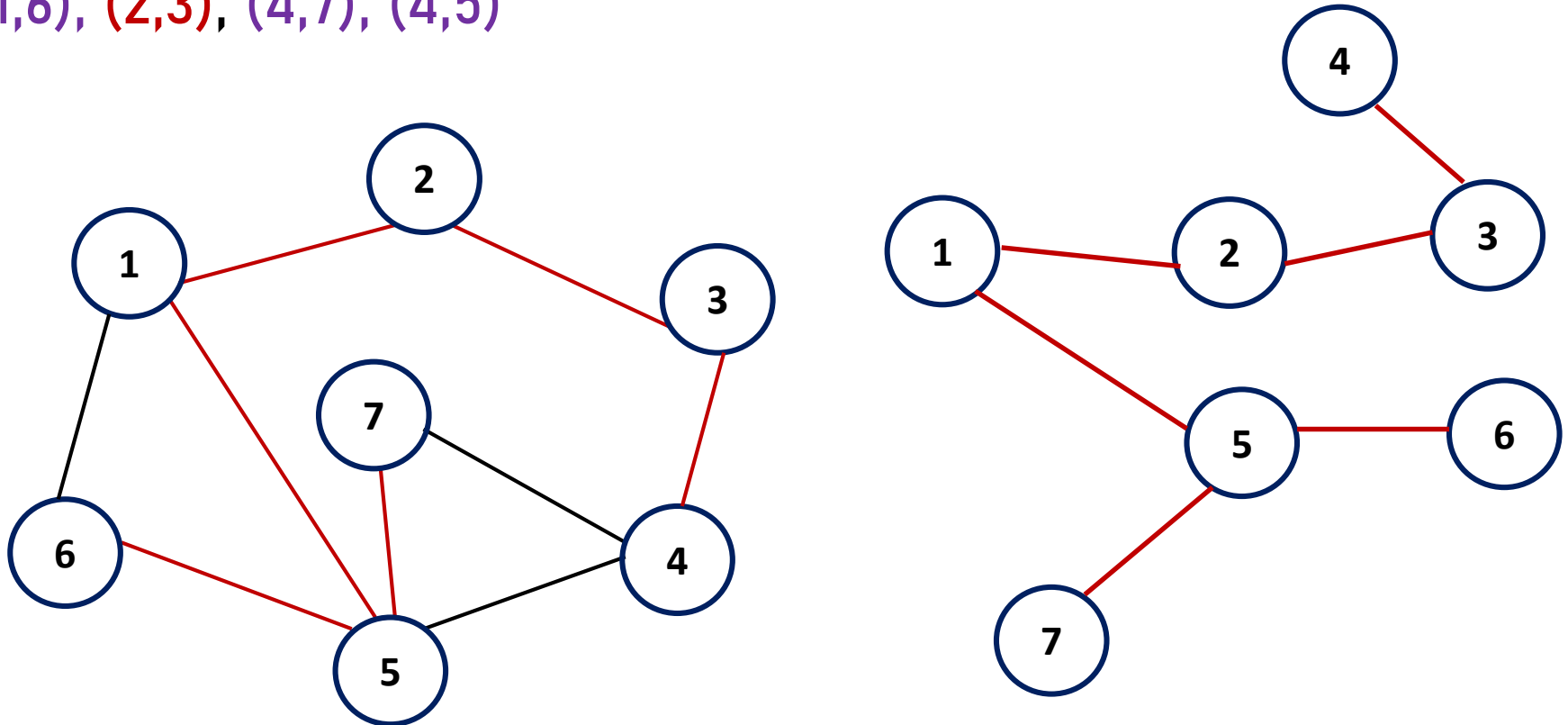


1 and 6 are in the same tree – ignore

Can Kruskal's use this approach?

Kruskal's with the new approach

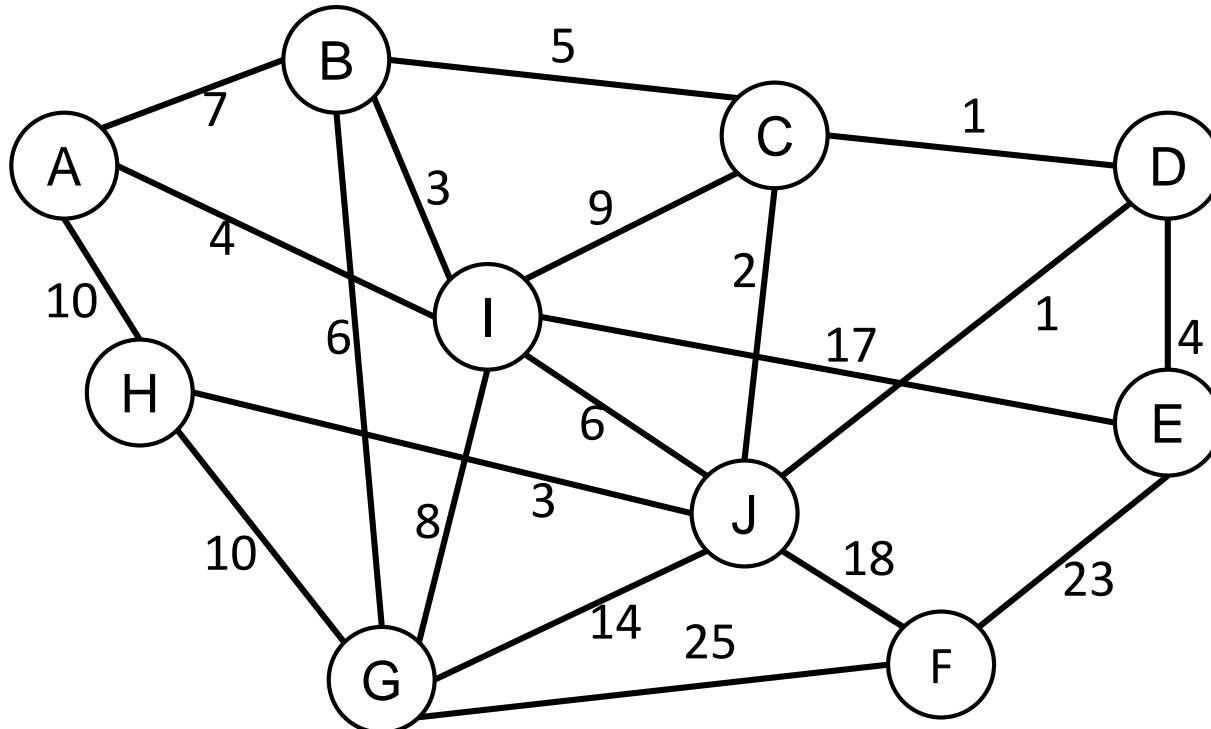
Order of edges considered: (1,2), (3,4), (5,6), (5,7), (1,5),
(1,6), (2,3), (4,7), (4,5)



Stop when $N-1$ edges have been found

Kruskal's Algorithm

- Merge trees
 - Initially, each node is a **single-node tree**
 - At each step, **merge two trees into one**
 - The merge cost is the **minimum** (min-cost edge)



Kruskal's Algorithm

Given: a connected undirected weight graph (N nodes, M edges)

Set **forest** as N node sets, each containing a node;

Set **fringe** as a priority queue of all the edges $\langle n1, n2, length \rangle$;

Set **tree** as an empty set of edges;

Repeat until **forest** contains only one tree or edges is empty {

 Get and remove $\langle n1^*, n2^*, length^* \rangle$ as the edge with **minimum length** from fringe;

 If ($n1^*$ and $n2^*$ are in different sets in forest) {

 Merge the two sets in **forest**;

 Add the edge to **tree**;

 }

}

return **tree**;

Kruskal's Algorithm

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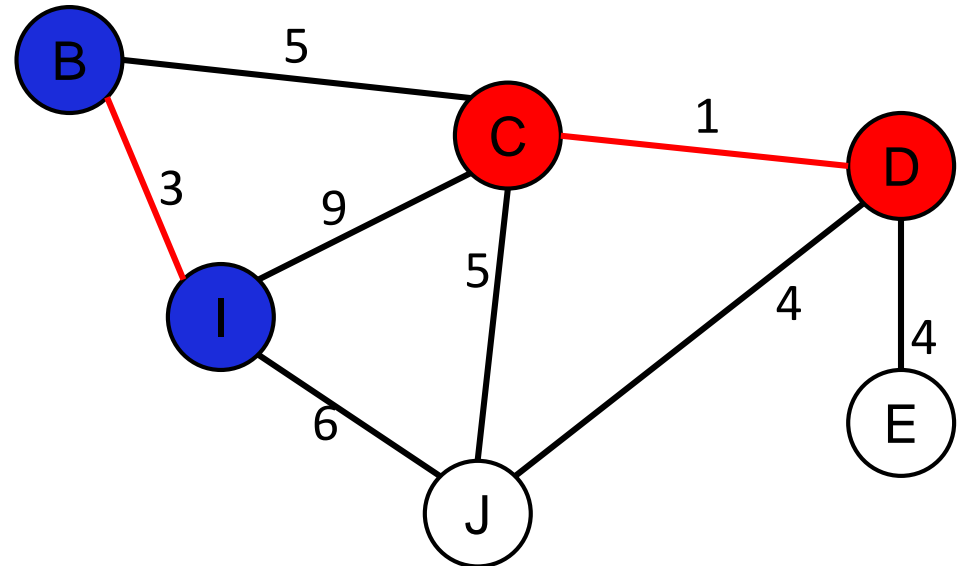
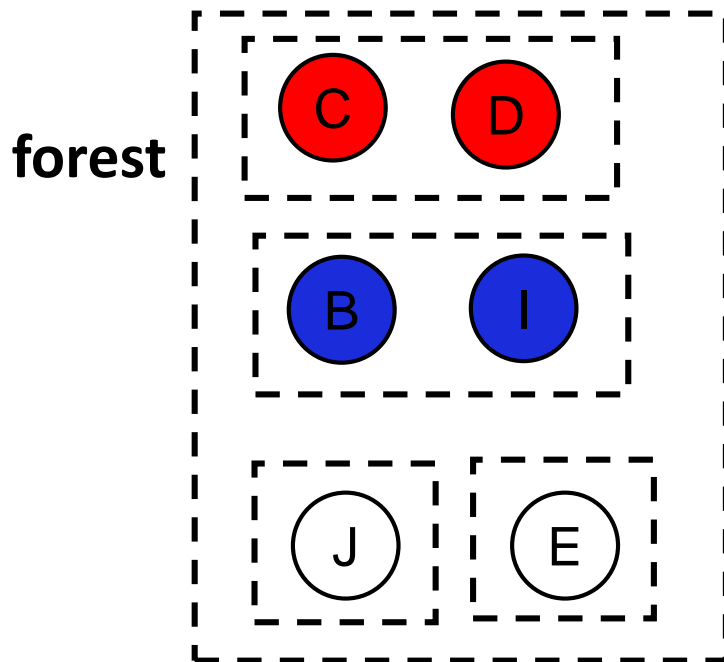
Need a way to –

- Efficiently **find** if two nodes are in sets in a forest
- Efficiently merge (**Union**) two trees

A new data structure: disjoint sets

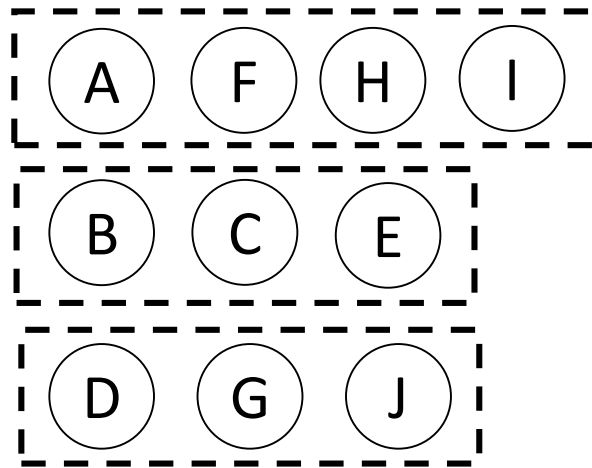
Find and Union Operation

- **Find**: Determine whether two elements belong to the same set
- **Union**: merge two sets into one
- The cost of Find and Union depends on the data structure of the forest: **set of sets**



Set of Sets: Data Structures

- Option 1: set of sets (e.g. `HashSet<HashSet<Node>>`)
 - Cost of find: iterate over all sets
 - Cost of union: add all the elements from one set to another

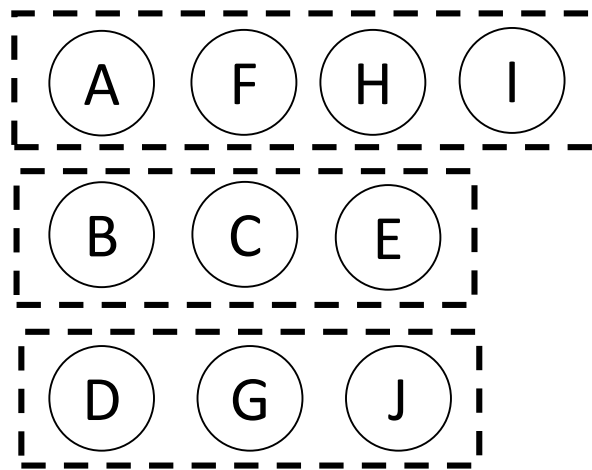


Set of Sets: Data Structures

- Option 1: set of sets (e.g.

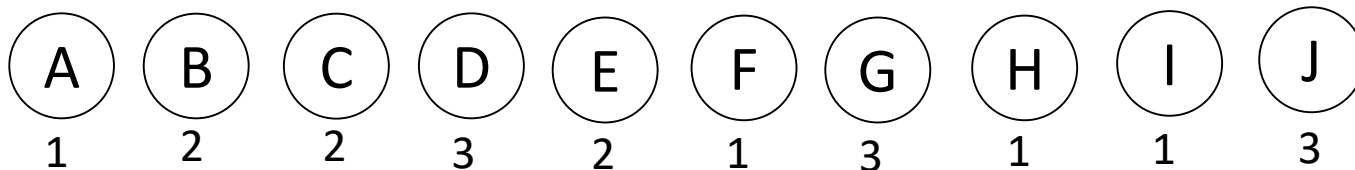
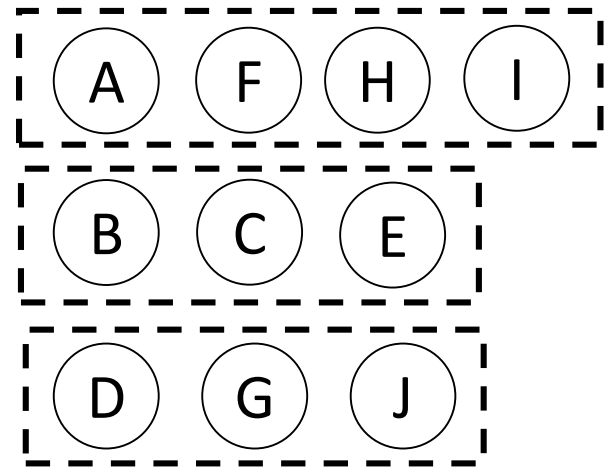
`HashSet<HashSet<Node>>`)

- Cost of find: iterate over all sets, $O(n)$
- Cost of union: add all the elements from one set to another, $O(n)$



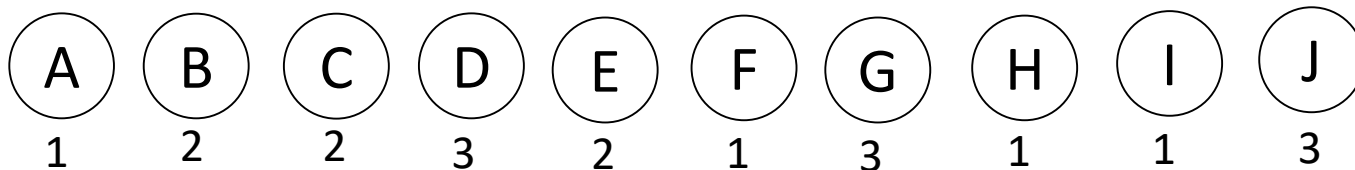
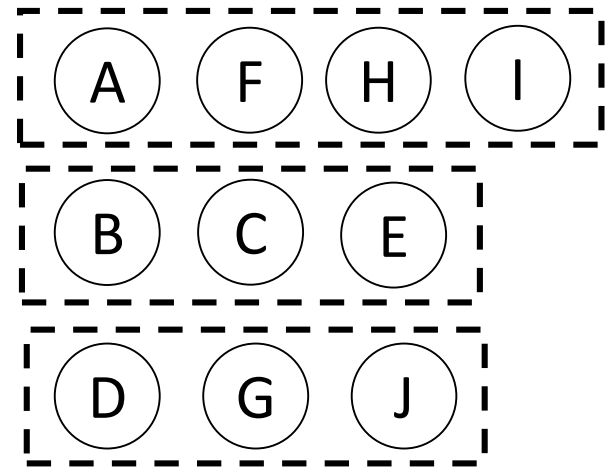
Set of Sets: Data Structures

- Option 2: mark each node with set ID
 - **Cost of find**: check whether the two elements have the same set ID
 - **Cost of union**: iterate all the nodes, change the set ID of one set to another



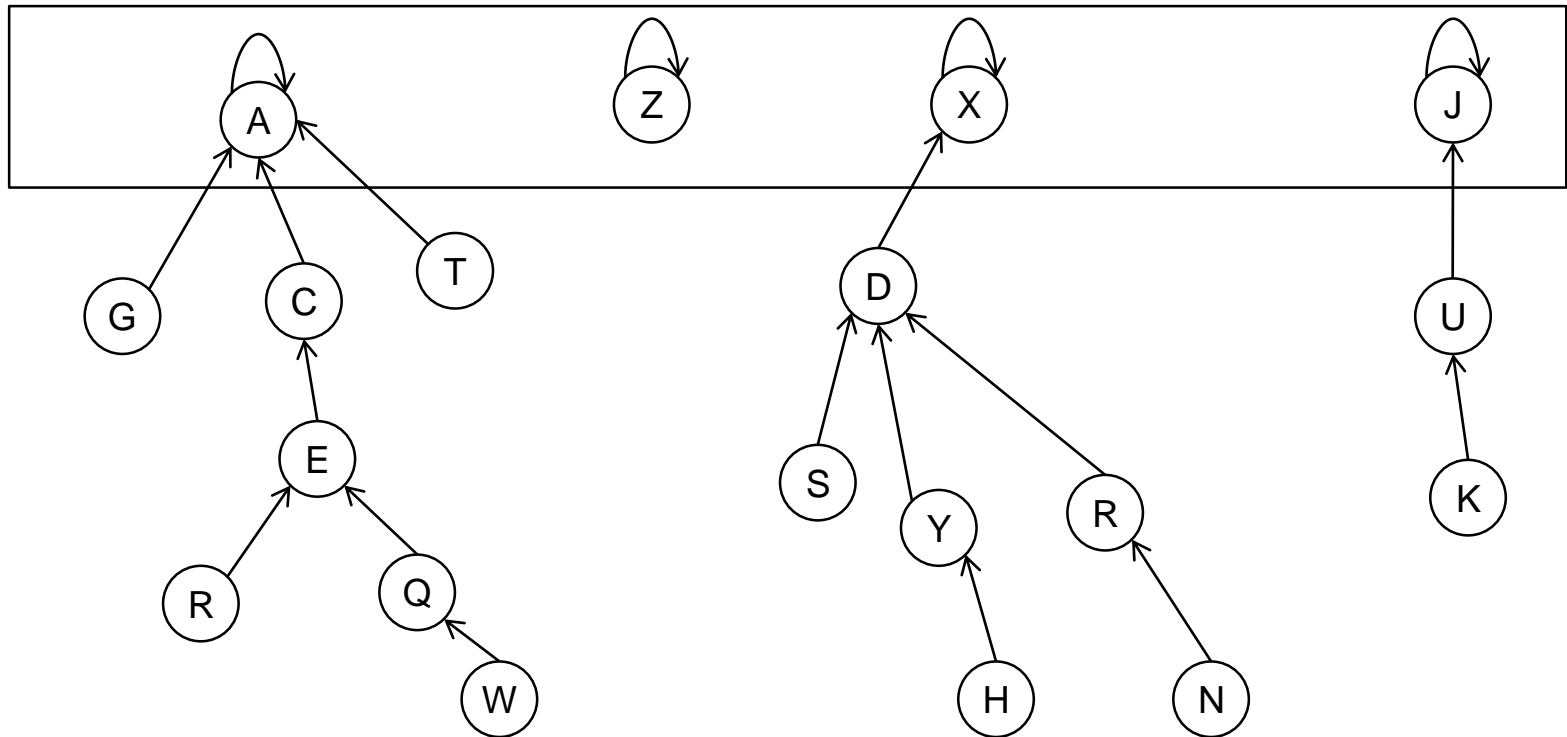
Set of Sets: Data Structures

- Option 2: mark each node with set ID
 - **Cost of find**: check whether the two elements have the same set ID $O(1)$
 - **Cost of union**: iterate all the nodes, change the set ID of one set to another $O(n)$



Set of Sets: Data Structures

- Option 3 (the best): disjoint-set (union-find) data structure
 - Set of inverted trees
 - Each set is represented by a linked tree with links pointing towards the root
 - **Forest** = set of root nodes



Disjoint Set

// make a new set with element x

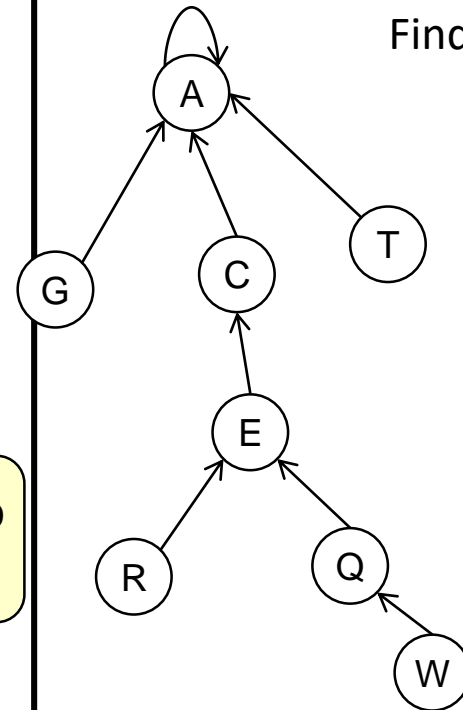
```
MakeSet(x) {  
  x.parent = x;  
  add x to forest;  
}
```



// find the root of the set that x belongs to

```
Find(x) {  
  if (x.parent == x) { // x is the root  
    return x;  
  } else {  
    root = Find(x.parent);  
    return root;  
  }  
}
```

Recursively go up to the root

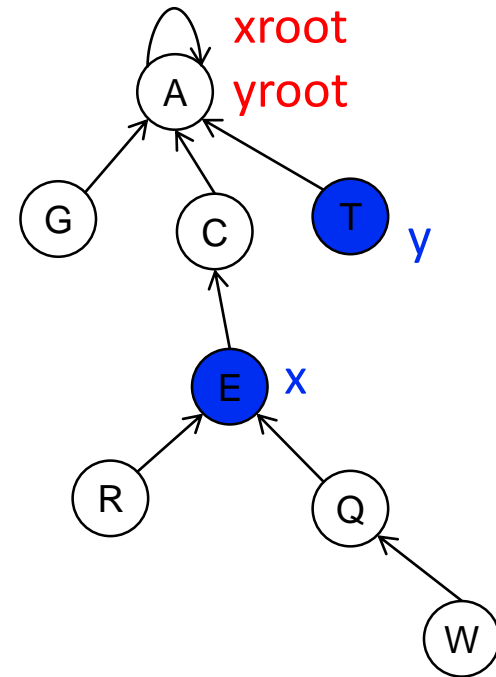


Find(A) = A
Find(G) = A
Find(E) = A
Find(W) = A
...

Disjoint Set

```
// union the sets of x and y  
Union(x, y) {  
    xroot = Find(x);  
    yroot = Find(y);  
    if (xroot == yroot) {  
        // x and y belong to  
        // the same set  
        return;  
    } else {  
        xroot.parent = yroot;  
        remove xroot from forest;  
    }  
}
```

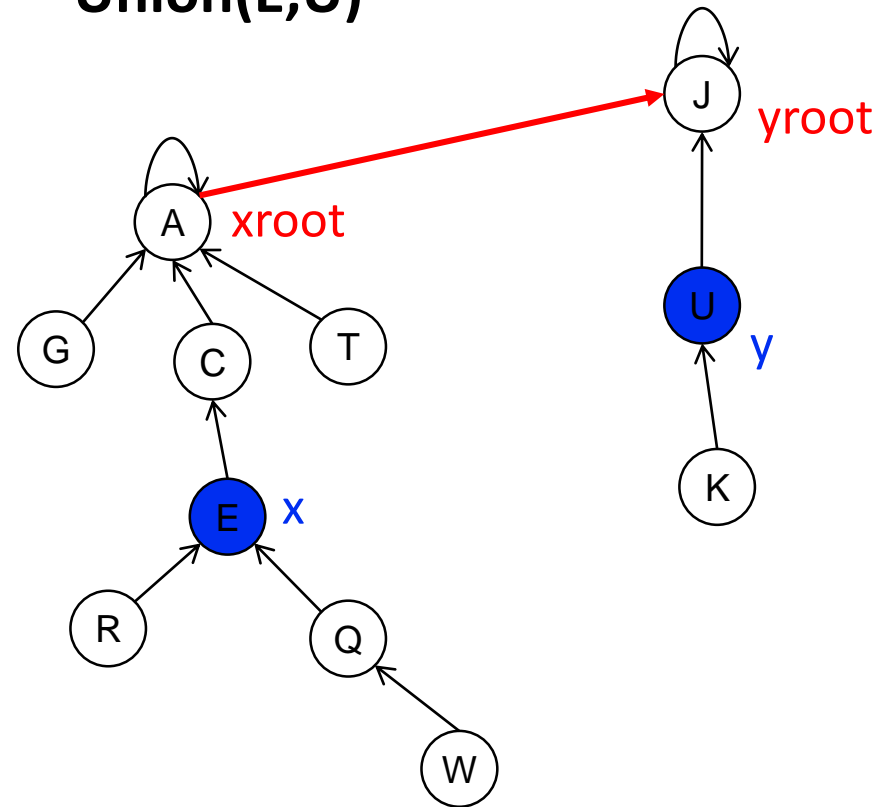
Union(E,T)



Disjoint Set

```
// union the sets of x and y  
Union(x, y) {  
  xroot = Find(x);  
  yroot = Find(y);  
  if (xroot == yroot) {  
    // x and y belong to  
    // the same set  
    return;  
  } else {  
    xroot.parent = yroot;  
    remove xroot from forest;  
  }  
}
```

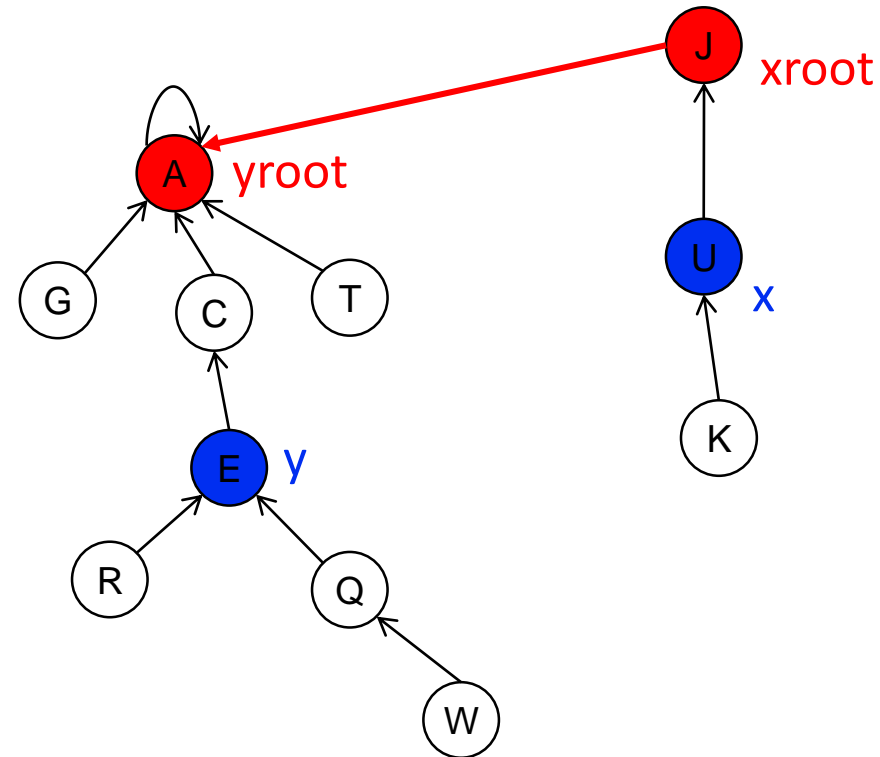
Union(E,U)



Disjoint Set

```
// union the sets of x and y  
Union(x, y) {  
  xroot = Find(x);  
  yroot = Find(y);  
  if (xroot == yroot) {  
    // x and y belong to  
    // the same set  
    return;  
  } else {  
    xroot.parent = yroot;  
    remove xroot from forest;  
  }  
}
```

Union(U,E)



Order can change the depth of the resultant tree

Disjoint Set

- To reduce complexity, **always merge shorter trees into deeper ones**

```
MakeSet(x) {  
    x.parent = x;  
    x.depth = 0;  
    add x to forest;  
}
```

```
Find(x) {  
    if (x.parent == x) {  
        return x;  
    } else {  
        root = Find(x.parent);  
        return root;  
    }  
}
```

```
Union(x, y) {  
    xroot = Find(x);  
    yroot = Find(y);  
    if (xroot == yroot) {  
        return;  
    } else {  
        if (xroot.depth < yroot.depth) {  
            xroot.parent = yroot;  
            remove xroot from forest;  
        } else {  
            yroot.parent = xroot;  
            remove yroot from forest;  
            if (xroot.depth == yroot.depth)  
                xroot.depth ++;  
        }  
    }  
}
```

Kruskal's with Disjoint Sets

- If we use Disjoint sets data structure.
 - We maintain each connected component as a disjoint set:
 - Initially each vertex is in its own disjoint set: $O(|V|)$
 - Edges are maintained in a sorted order or in a priority queue: $O(|E| \log |E|)$
 - When considering an edge (u, v) to include, we check if u and v are in the same disjoint set (Find operations):
 - If yes, they form a cycle
 - Else, we include the edge and perform a Union operation on disjoint sets containing u and v .
 - Since we always merge shorter trees into deeper ones, the worst case time complexity of Find and Union is $O(\log |V|)$
 - In worst case we iterate through all edges: $O(|E| \log |V|)$.
 - Overall complexity: $O(|V| + (|E| \log |E|) + (|E| \log |V|) = O(|E| \log |E|)$ (as $|E| \geq |V| - 1$)

Test: May 9

**Syllabus: Everything we've covered
since week 7 up to today**

Best of luck for the test!