Lecture 7 Graphs III – Trees

Karl Bringmann, Dominic Zimmer

Saarland University

June, 2019





Midterm-Contest

- Date: 20.06.2020 at 10:00
 Meet in the Lecture Zoom at 10:00 sharp!
- ► Install & test the VM as soon as possible (≤ Thursday)!
 - Installation instructions: https://cms.sic.saarland/cp20/2/VM
 - ► Test if you can boot & connect to the VPN
 - Test if you can submit a solution for a problem
 - ► Last chance to contact us in case of problems: Office Hour at Friday, 10:00
 - Before the contest: Delete all files you created (except your cheatsheet)
- Submit your Cheat-Sheet until Thursday 23:59
 - ▶ 100.000 characters, .txt file
 - Submission on your personal status page





Overview

- ► Graphs I: Traversals and Shortest Paths
- ► Graphs II: DFS Applications and Friends
- ► Graphs III: Trees
- ► Graphs IV: Flow Problems





City Planning

After constructing a city where every citizen could visit everyone else, you do some statistics. Your traffic authorities count that there are precisely V-1 roads in your city.





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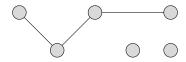






City Planning

After constructing a city where every citizen could visit everyone else, you do some statistics. Your traffic authorities count that there are precisely V-1 roads in your city. How many unique paths are there?

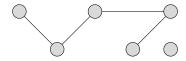






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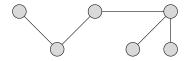






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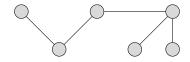




City Planning

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How many unique paths are there?



Insight

Between two vertices v and w in this graph, there exists exactly one unique path.

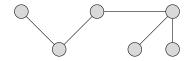




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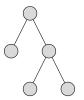
In total, there are $\binom{V}{2} = \frac{V \cdot (V-1)}{2}$ unique paths

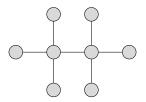




Definition: Tree

A Tree is an acyclic, connected, undirected graph.







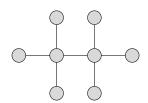


Definition: Tree

Let G be undirected. The following statements are equivalent

- ► G is called a tree
- ► G is acyclic and connected
- ▶ Between any two vertices of G, there is exactly one path
- ▶ G is acyclic and E = V 1
- ▶ G is connected and E = V 1
- ► G is minimally connected
- G is maximally acyclic





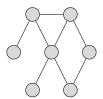


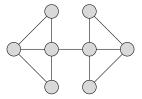


Spanning Trees

Definition: Spanning Tree

A subgraph S of G is called a $Spanning\ Tree$, if S is a tree and it contains every vertex of G.





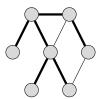


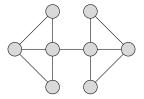


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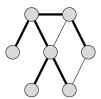


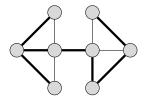


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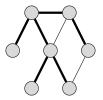


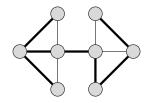


Spanning Trees

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So far, we have seen

- ► DFS/BFS Spanning Tree
- ► Shortest Path Spanning Tree





Problem: Lights Out

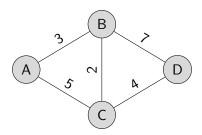
Lights Out





Problem: Lights Out

Lights Out

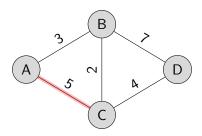






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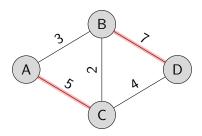






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Lights Out

The town needs to save electricity and wants to turn off some street lights. However, every citizen should still be able to drive between any two intersections along an illuminated path. Given the electricity cost per hour for each road, which roads should be left dark to save the most money?

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▶ We are given an undirected, weighted graph G





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- ► A spanning tree of G uses the minimum number of edges to connect G





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Insight

- We are given an undirected, weighted graph G
- ► A spanning tree of G uses the minimum number of edges to connect G
- Among all spanning trees, find one of minimal total edge weight





Minimum Spanning Trees

A *Minimum Spanning Tree* (MST) is a spanning tree of a weighted graph, such that the total sum of its edge weights is minimal.





Minimum Spanning Trees

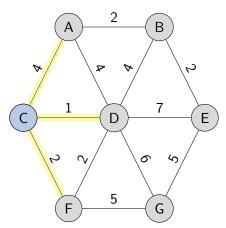
A *Minimum Spanning Tree* (MST) is a spanning tree of a weighted graph, such that the total sum of its edge weights is minimal.

- Start exploring at any vertex
- Maintain a sorted list of explored edges, the frontier
 - Take the edge of minimum weight
 - If the edge is a back edge, discard it
 - Otherwise, add it to the MST





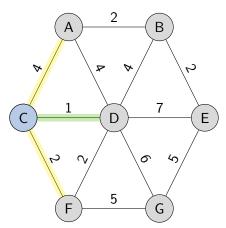
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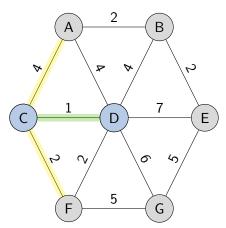
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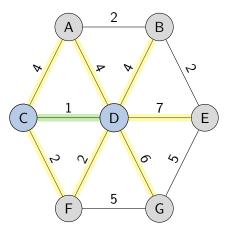
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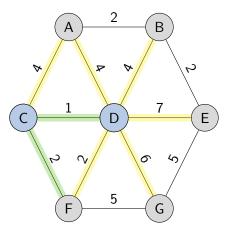
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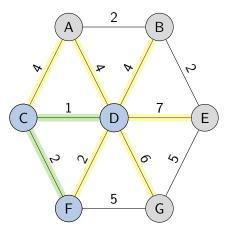
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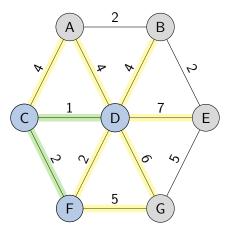
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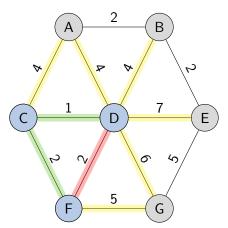
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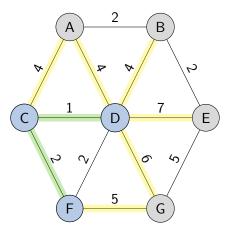
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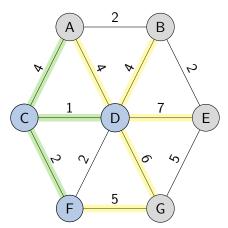
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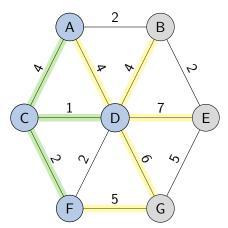
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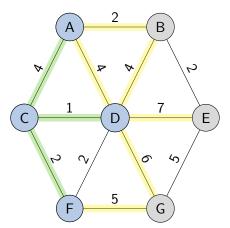
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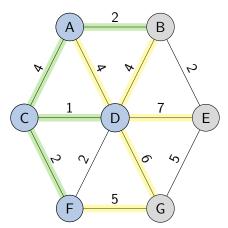
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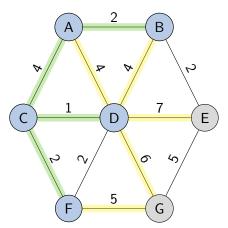
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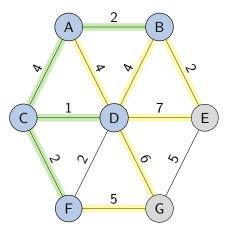
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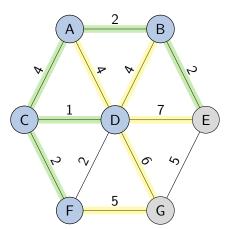
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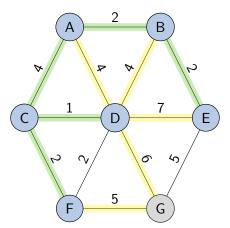
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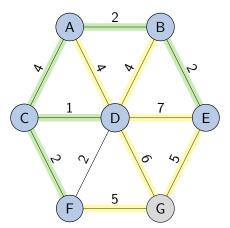
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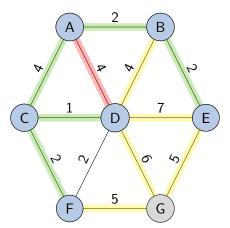
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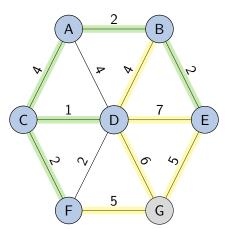
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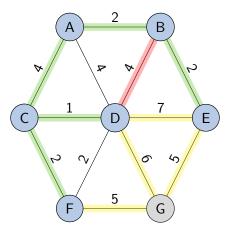
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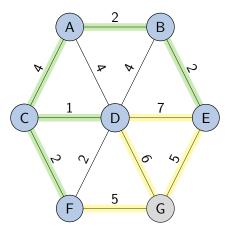
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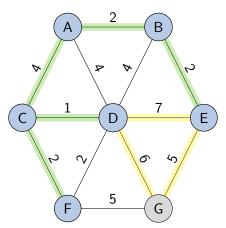
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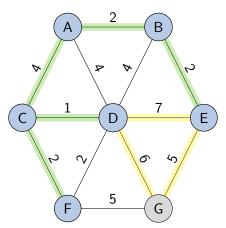
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Prim's Algorithm

Algorithm Idea (Prim's)

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Sounds Familiar?

We used the same insights in Dijkstra's SSSP algorithm





Prim's Algorithm

```
vector<bool> visited(V, false);
   vector < vector < pair < int , int >>> adj(V); // < to , weight >>
   // <weight, from, to>
   priority_queue < tuple < int , int , int >,
     vector<tuple<int , int , int >>,
     greater<tuple<int , int , int >>> PQ;
   void visit(int v) {
      visited[v] = true;
     for (auto p: adj[v]) {
     int u = p.first;
10
11
       int w = p.second;
        if (!visited[u]) PQ.push({w, v, u});
12
13
14
15
   visit(0); // arbitrary start vertex
   while (!PQ.empty()) {
16
     auto front = PQ.top(); PQ.pop();
17
     int w, from, to;
18
     tie(w, from, to) = front; // unpack tuple
19
     if (!visited[to])
20
       cout << "Add " << from << "-" << to << " to MST\n";</pre>
21
       visit (to);
22
23
```

Problem: Lights Out

Lights Out

The town needs to save electricity and wants to turn off some street lights. However, every citizen should still be able to drive between any two intersections along an illuminated path. Given the electricity cost per hour for each road, which roads should be left dark to save the most money?

Solution Idea

- Compute a MST of the input graph
- Sum the weights of edges not included in MST





Union-Find Disjoint Set

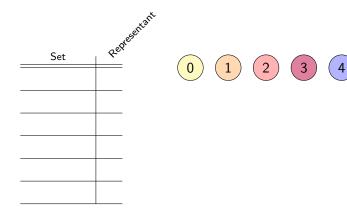
Union-Find

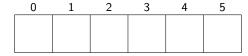
Union-Find Dijsoint Set (short: Union-Find) is a datastructure to manage disjoint sets. Every disjoint set is represented by one of its elements. A Union-Find supports the operations

- unionSet(i,j) unify sets i and j
- findSet(i) find the representant of i













Set	\ \delta es
{0}	0
{1}	1
{2}	2
{3}	3
{4}	4
{5}	5











0	1	2	3	4	5





Set	L Ses
{0}	0
{1}	1
{2}	2
{3}	3
{4}	4
{5}	5











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





epresentant

unionSet(1,	5)
-------------	----

Set	Res
{0}	0
{1}	1
{2}	2
{3}	3
{4}	4
	5

0	(1





5

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





Set Qente (1) (1) (2) (2) (3) (4) (4)

{5}

unionSet(1, 5)

0

2

3

4

5

parent:

5

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





 unionSet(1, 5)

0

2

3)

4) (



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





Set {0} {2} 2 {3} 3 {4} 4 $\{1,5\}$ 5 unionSet(1, 5)









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





 unionSet(1, 5)

0





1

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





 unionSet(1, 5)

0

2) (

)



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





 unionSet(4, 2)

0







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





Set Red (2) (2) (3) (4) (4) (4) (5) (5)

unionSet(4, 2)







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





Set Rept (1) (1) (1) (2) (2) (2) (3) (3) (4) (4) (4) (1,5) (5)

unionSet(4, 2)







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





 unionSet(4, 2)







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





Set Q²⁶

{0} 0

{2,4} 2

{3} 3

{1,5} 5

unionSet(4, 2)







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





Set Qest

{0} 0

{2,4} 2

{3} 3

{1,5} 5

unionSet(4, 2)







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





Set Qest

{0} 0

{2,4} 2

{3} 3

{1,5} 5

findSet(4)

0





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





findSet(4)







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





findSet(4)







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





Set Qest

{0}

(0)

(2,4)

(3)

(1,5)

5

findSet(4) = 2

0





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





Set Q²⁶

{0} 0

{2,4} 2

{3} 3

{1,5} 5

unionSet(2, 5)







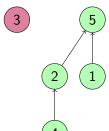
0	1	2	3	4	5
0	5	2	3	2	5





 unionSet(2, 5)

0



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0	5	2	3	2	5





Set Qest

{0} 0

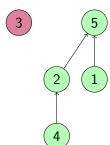
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{3} 3

{1,5} 5

unionSet(2, 5)

0



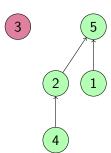
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 unionSet(2, 5)

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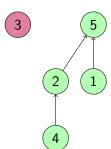
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 unionSet(2, 5)

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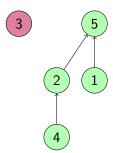
0	1	2	3	4	5
0	5	2	3	2	5





 unionSet(2, 5)

0



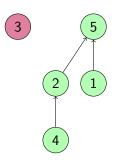
0	1	2	3	4	5
0	5	5	3	2	5





 findSet(4)

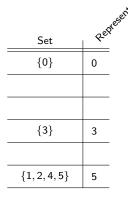
0



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

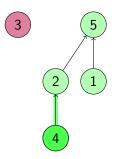






findSet(4)

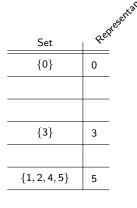




0	1	2	3	4	5
0	5	5	3	2	5

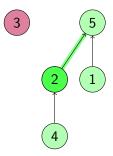






findSet(4)





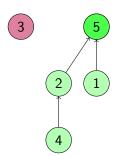
0	1	2	3	4	5
0	5	5	3	2	5





Set {0} {3} 3 $\{1, 2, 4, 5\}$ 5 findSet(4)

0



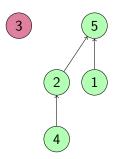
0	1	2	3	4	5
0	5	5	3	2	5





findSet(4)	= 5
------------	-----

0



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

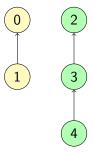




Example Code

```
class UnionFind {
   private:
3
    vector<int> parent;
   public:
      UnionFind(int N) {
        parent.resize(N);
6
       for (int i = 0; i < N; i++) parent[i] = i;
8
     int findSet(int i) {
9
        if (parent[i] == i)
10
11
          return i:
        else
12
13
          return findSet(parent[i]);
14
     bool isSameSet(int i, int j) {
15
        return findSet(i) == findSet(j);
16
17
     void unionSet(int i, int j) {
18
        i = findSet(i), j = findSet(j);
19
        if (!isSameSet(i, j)) // or (i != j)
20
          parent[i] = j;
21
22
                                  14
```

Union-by-Rank

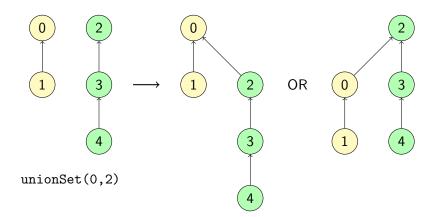


unionSet(0,2)





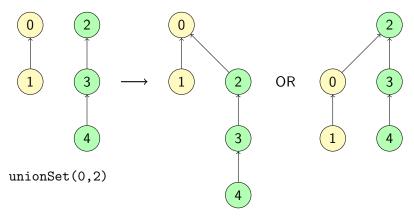
Union-by-Rank







Union-by-Rank

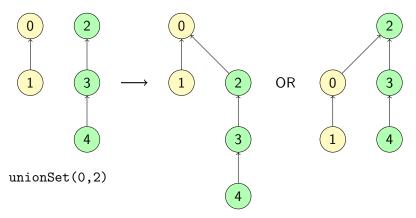


▶ findSet's running time depends on the depth of the tree





Union-by-Rank

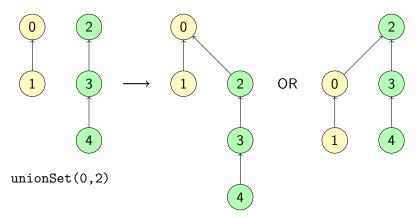


- findSet's running time depends on the depth of the tree
- Store the approximate depth of each vertex as rank





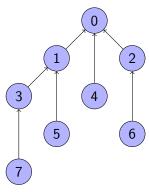
Union-by-Rank



- findSet's running time depends on the depth of the tree
- Store the approximate depth of each vertex as rank
- Use rank to keep the total depth small during unionSet



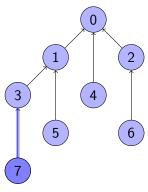




findSet(7)



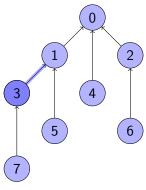




findSet(7)



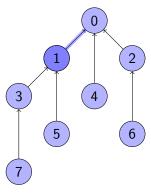




findSet(7)





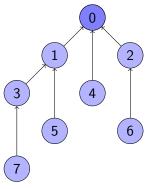


findSet(7)



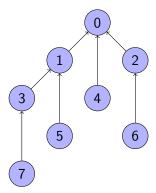


Path Compression



findSet(7)

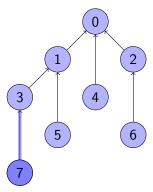




$$findSet(7) = 0$$



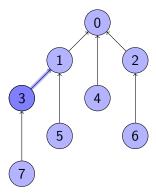




$$findSet(7) = 0$$



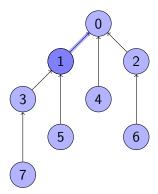




$$findSet(7) = 0$$



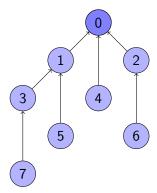




$$findSet(7) = 0$$



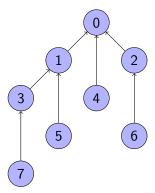




$$findSet(7) = 0$$



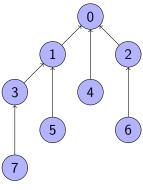




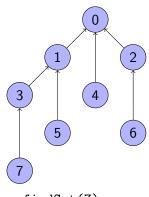
$$findSet(7) = 0$$





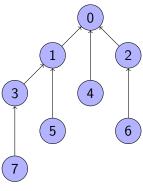


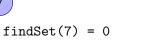
$$findSet(7) = 0$$

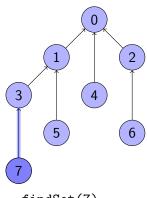


findSet(7)



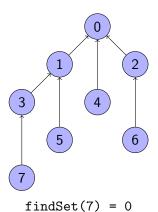


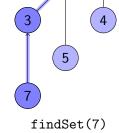




findSet(7)

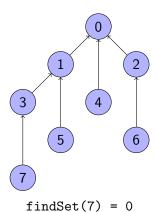


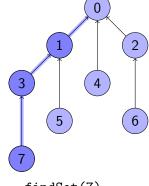






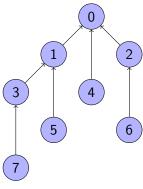
Path Compression

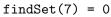


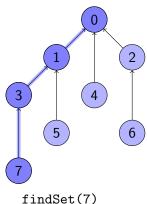


findSet(7)

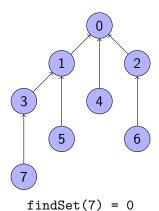


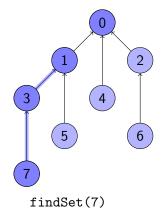




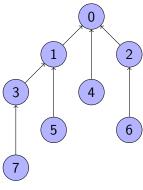




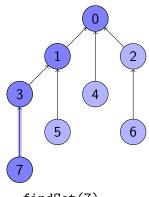








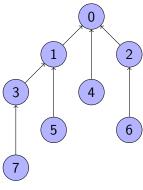
$$findSet(7) = 0$$



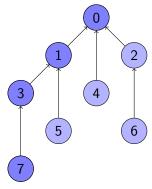
findSet(7)







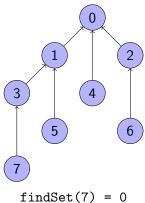
$$findSet(7) = 0$$



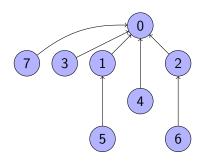
$$findSet(7) = 0$$







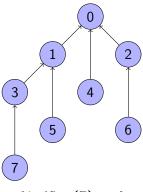


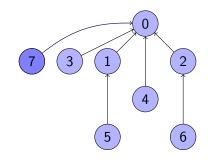


$$findSet(7) = 0$$







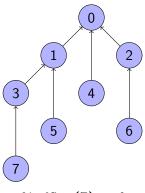


$$findSet(7) = 0$$

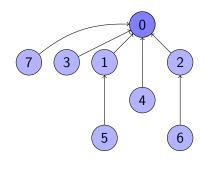
$$findSet(7) = 0$$





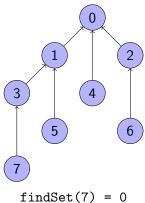


$$findSet(7) = 0$$

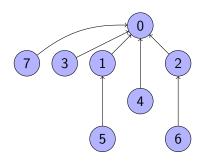


$$findSet(7) = 0$$







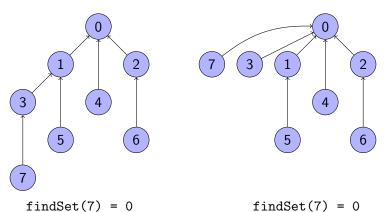


$$findSet(7) = 0$$



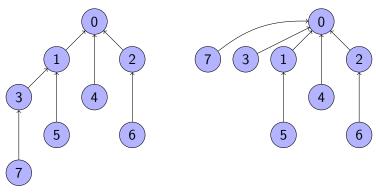


Path Compression



► The first findSet query updates all parents on the path to the root





$$findSet(7) = 0$$

$$findSet(7) = 0$$

- ► The first findSet query updates all parents on the path to the root
- Subsequent findSet queries benefit from shallower tree





Improved Example Code

```
class UnionFind {
   private:
3
       vector<int> parent , rank; // add rank vector
   public:
     UnionFind(int N) {
        rank.assign(N, 0); // initialize
6
        parent.resize(N);
        for (int i = 0; i < N; i++) parent[i] = i;
8
9
     void unionSet(int i, int j) {
10
11
        i = findSet(i), j = findSet(j);
        if (!isSameSet(i, j)) { // or (i != j)
12
          if (rank[i] > rank[j]) { // union by rank
13
            parent[i] = i;
14
          } else {
15
            parent[i] = j;
16
            if (rank[i] == rank[j]) rank[j]++;
17
18
19
20
```





Improved Example Code





Improved Example Code

▶ Initially, we had $\mathcal{O}(N)$ worst case running time per query





Improved Example Code

```
int findSet(int i) {
21
        if (parent[i] == i)
22
          return i:
23
            // path compression
24
          return parent[i] = findSet(parent[i]);
25
26
     bool isSameSet(int i, int j) {
27
        return findSet(i) == findSet(j);
28
29
```

- ▶ Initially, we had O(N) worst case running time per query
- ► With Path Compression and Union by Rank, we get *almost* constant* running time per query





Conclusion

The Union-Find data structure is a useful tool for managing (potentially) large disjoint sets. It makes testing whether two elements belong to the same set easy.





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In particular, Union-Find is used in

Kruskal's Algorithm for MSTs





Conclusion

The Union-Find data structure is a useful tool for managing (potentially) large disjoint sets. It makes testing whether two elements belong to the same set easy.

In particular, Union-Find is used in

- Kruskal's Algorithm for MSTs
- ► Tarjan's Algorithm for LCA





Kruskal's Algorithm

Algorithm Idea (Prim's)

- ► Start exploring at any vertex
- Maintain a sorted list of explored edges, the frontier
 - ► Take an edge of minimum weight
 - If the edge is a back edge, discard it
 - Otherwise, add it to the MST





Kruskal's Algorithm

Algorithm Idea (Prim's)

- Start exploring at any vertex
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 - ► Take an edge of minimum weight
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- ► Sort the edges by weight
- Greedily take an edge of minimum cost
 - If adding the edge forms a cycle, discard it
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Kruskal's Algorithm

Algorithm Idea (Prim's)

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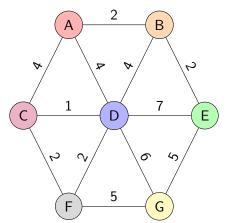
- Sort the edges by weight
- Greedily take an edge of minimum cost
 - If adding the edge forms a cycle, discard it
 - Union-Find
 - Otherwise, add it to the MST





Kruskal's Algorithm

- Sort the edges by weight
- ► Greedily take an edge of minimum cost, avoid cycles

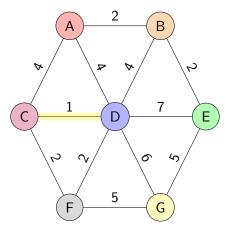






Kruskal's Algorithm

- ► Sort the edges by weight
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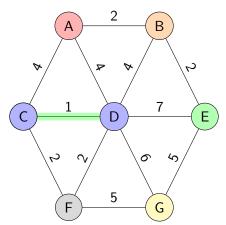






Kruskal's Algorithm

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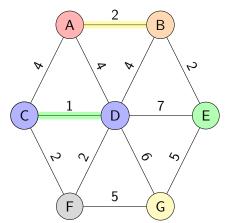






Kruskal's Algorithm

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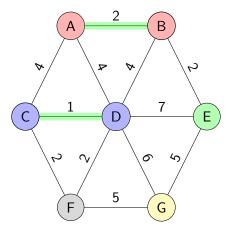






Kruskal's Algorithm

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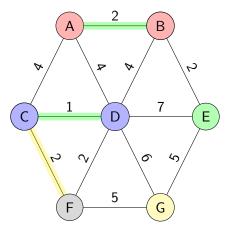






Kruskal's Algorithm

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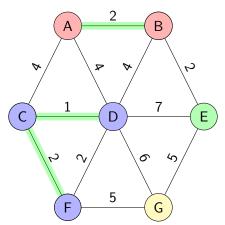






Kruskal's Algorithm

- Sort the edges by weight
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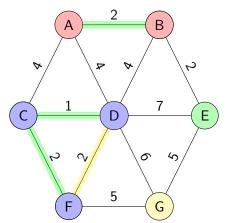






Kruskal's Algorithm

- ► Sort the edges by weight
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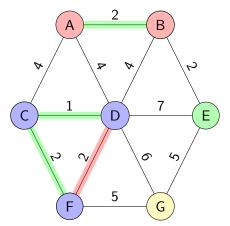






Kruskal's Algorithm

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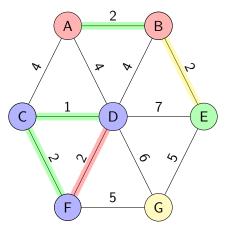






Kruskal's Algorithm

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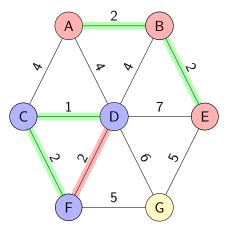






Kruskal's Algorithm

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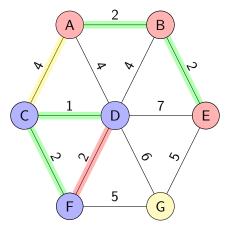






Kruskal's Algorithm

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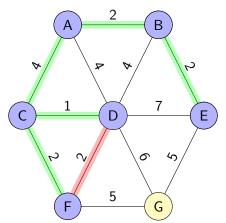






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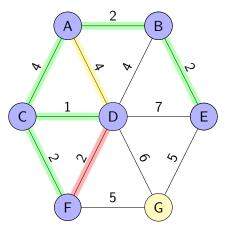






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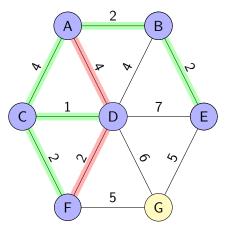






Kruskal's Algorithm

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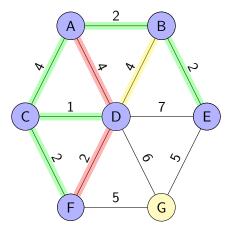






Kruskal's Algorithm

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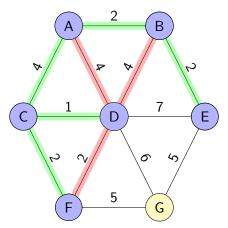






Kruskal's Algorithm

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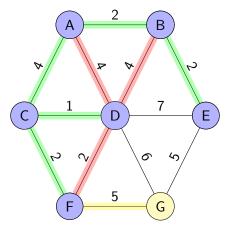






Kruskal's Algorithm

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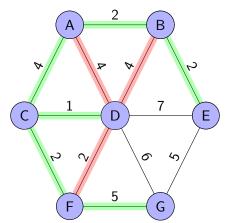






Kruskal's Algorithm

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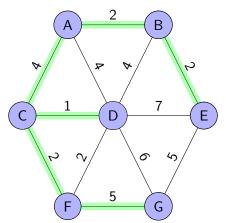






Kruskal's Algorithm

- ► Sort the edges by weight
- ► Greedily take an edge of minimum cost, avoid cycles







Example Code: Kruskal's Algorithm

```
auto UF = UnionFind(V);
   // <weight, u, v>
   vector<tuple<int , int , int >> edgeList;
   // sort increasing by weight
   sort(edgeList.begin(), edgeList.end());
8
   for (auto edge: edgeList) {
     int w. u. v:
     tie(w, u, v) = edge; // unpack tuple
10
      if (!UF.isSameSet(u, v)) {
11
       cout << "Add " << u << "-" << v << " to MST" << endl;</pre>
12
       UF.unionSet(u, v);
13
14
15
```





Overview

Prim's Algorithm

- Uses PriorityQueue
- Greedily picks edges from the frontier
- Avoids cycles using visited flags
- ▶ Runs in O(E log E)

Kruskal's Algorithm

- Requires Union-Find
 - But results in shorter code
- Greedily picks edges from sorted list
- Avoids cycles using Union-Find
- Runs in $\mathcal{O}(E \log E)$





Overview

Prim's Algorithm

- Uses PriorityQueue
- Greedily picks edges from the frontier
- Avoids cycles using visited flags
- Runs in $\mathcal{O}(E \log E)$

Kruskal's Algorithm

- Requires Union-Find
 - But results in shorter code
- Greedily picks edges from sorted list
- Avoids cycles using Union-Find
- Runs in $\mathcal{O}(E \log E)$
- ▶ Both algorithms can early exit once (V-1) edges have been added to the MST





Union-Find

Conclusion

The Union-Find data structure is a useful tool for managing (potentially) large disjoint sets. It makes testing whether two elements belong to the same set easy.

In particular, Union-Find is used in

- Kruskal's Algorithm for MSTs
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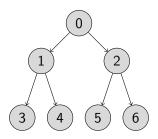
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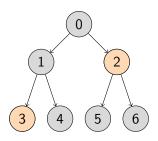
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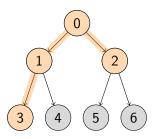
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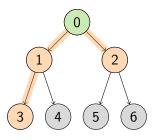
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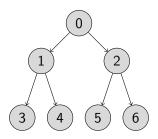
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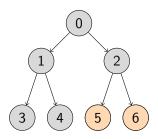
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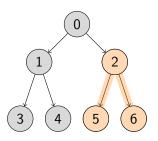
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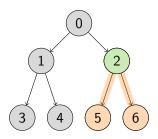
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Paths $\langle 5, 2, 0 \rangle$ and $\langle 6, 2, 0 \rangle$ share the suffix $\langle 2, 0 \rangle$.

Thus, LCA(5, 6) = 2.





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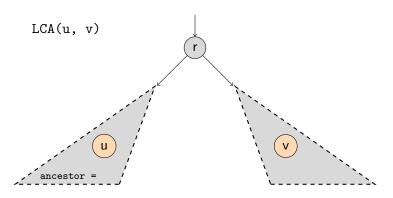
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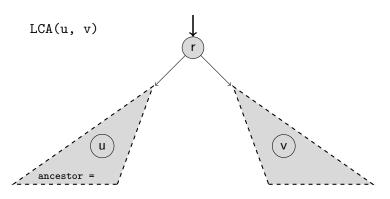




- Traverse the Tree via DFS
- After exploring v, if u was finished before, ancestor[u] is the LCA
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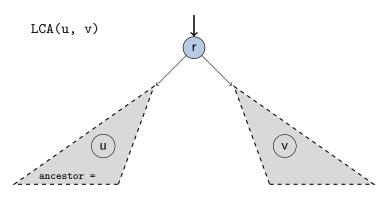




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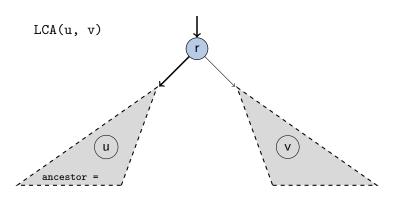




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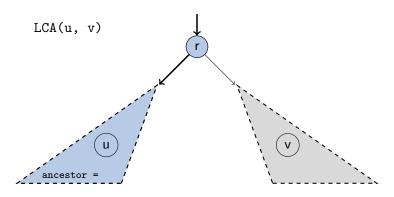




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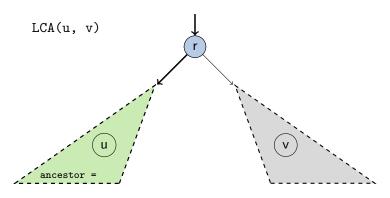




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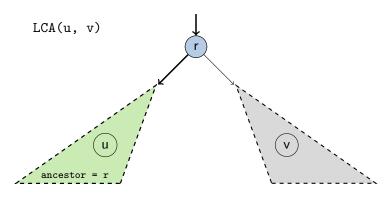




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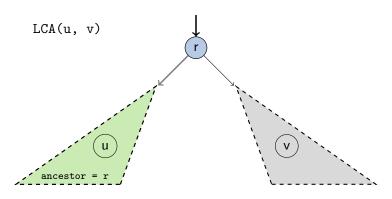




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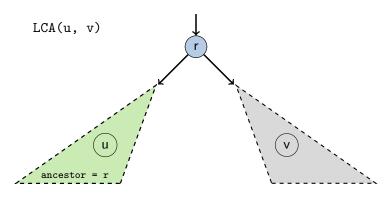




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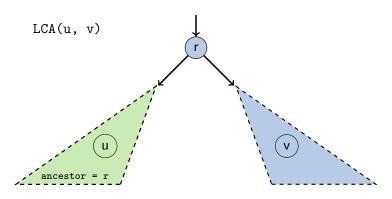




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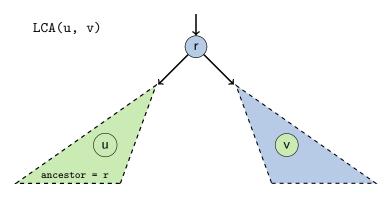




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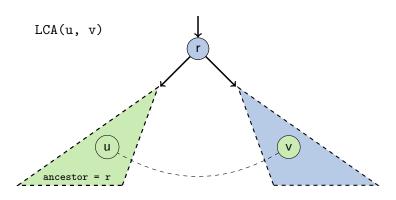




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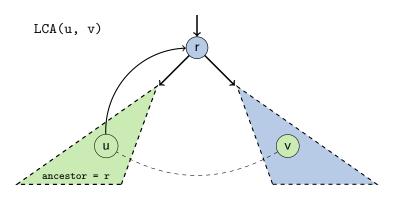




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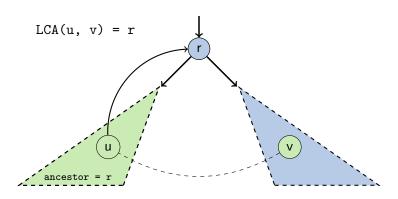




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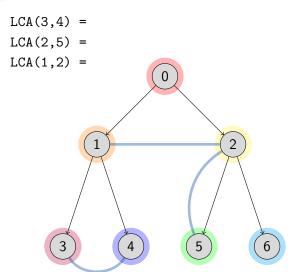




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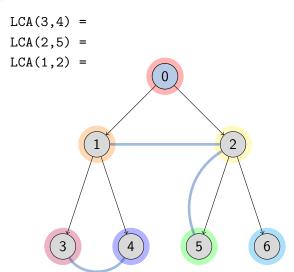






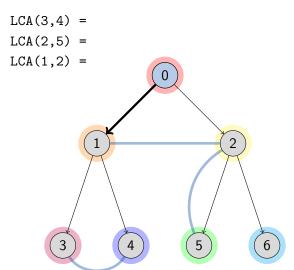






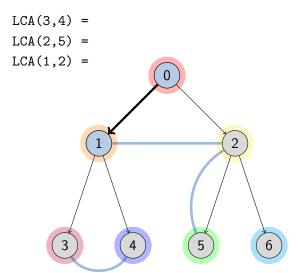






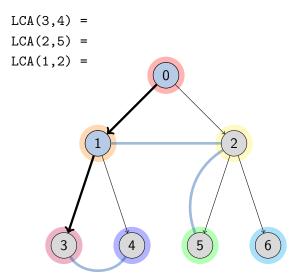






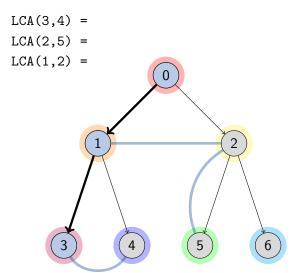






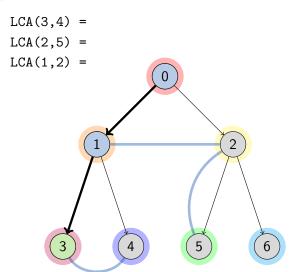






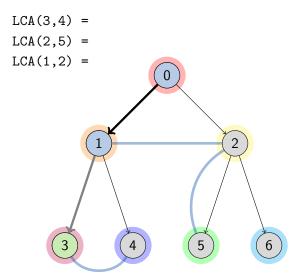






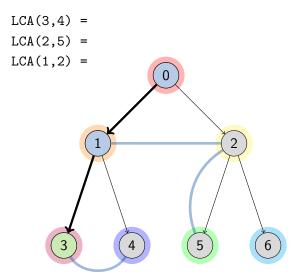






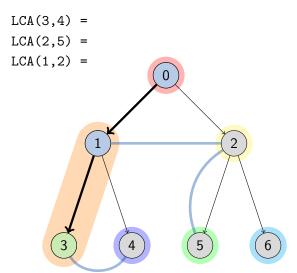






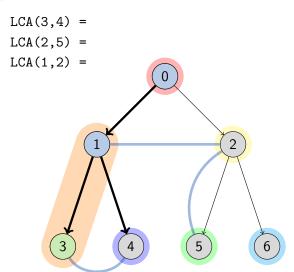






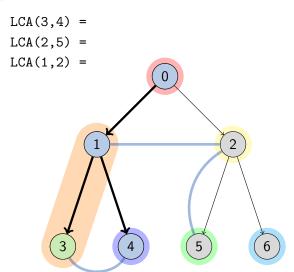






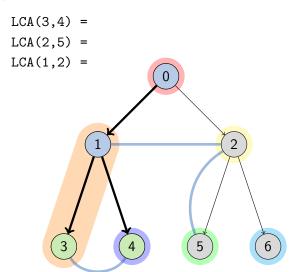






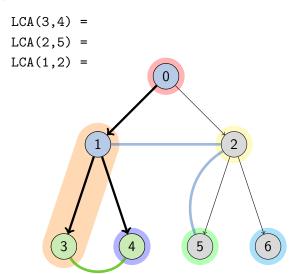






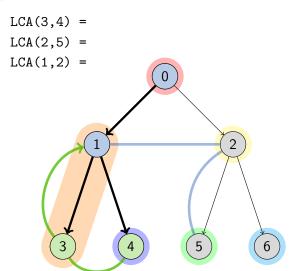














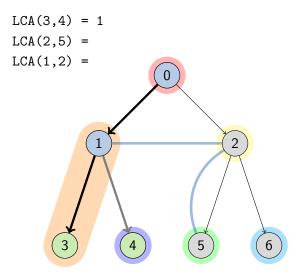










































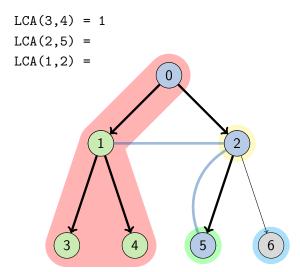










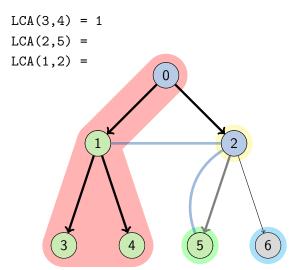


























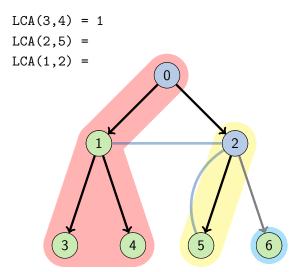














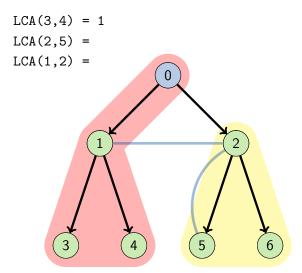


















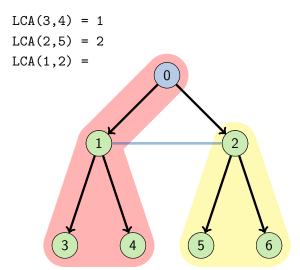






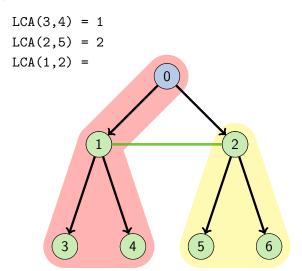














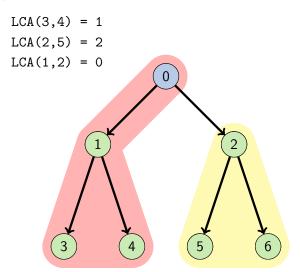






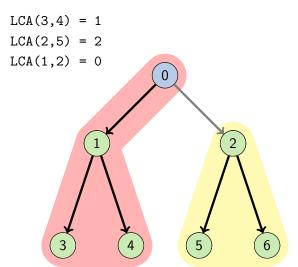






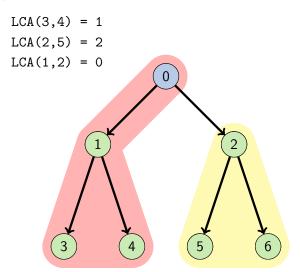






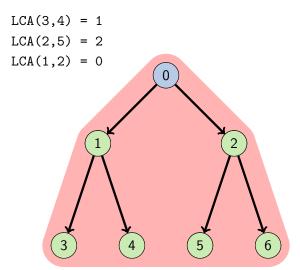






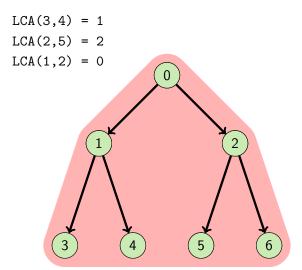
















Example Code: Tarjan LCA

```
int V, E, root, Q;
   vector<vector<int>>> adj(V); // directed, rooted tree
3
   vector<bool> visited(V, false);
   auto UF = UnionFind(V);
   vector<vector<int>>> queries(V); // each vertex ' queries
   vector<int> ancestor(V);
   for (int i = 0; i < V; i++) ancestor[i] = i;
9
   void dfs (int u) {
10
     for (auto v: adj[u]) {
11
       dfs(v);
12
13
       UF.unionSet(u, v);  // combine with children
       ancestor [UF.findSet(u)] = u; // don't lose ancestor
14
15
    visited[u] = true;
16
     for (auto v: queries[u]) // check all (u,v) query pairs
17
       if (visited[v])
18
         cout << "LCA of " << u << " and " << v <<
19
                  " is " << ancestor[UF.findSet(v)] << endl;
20
21
   dfs(root);
```



Trees

Recap

- ► Trees
- ► Prim's Algorithm
- ▶ Union-Find
- ► Kruskal's Algorithm
- ► LCA



