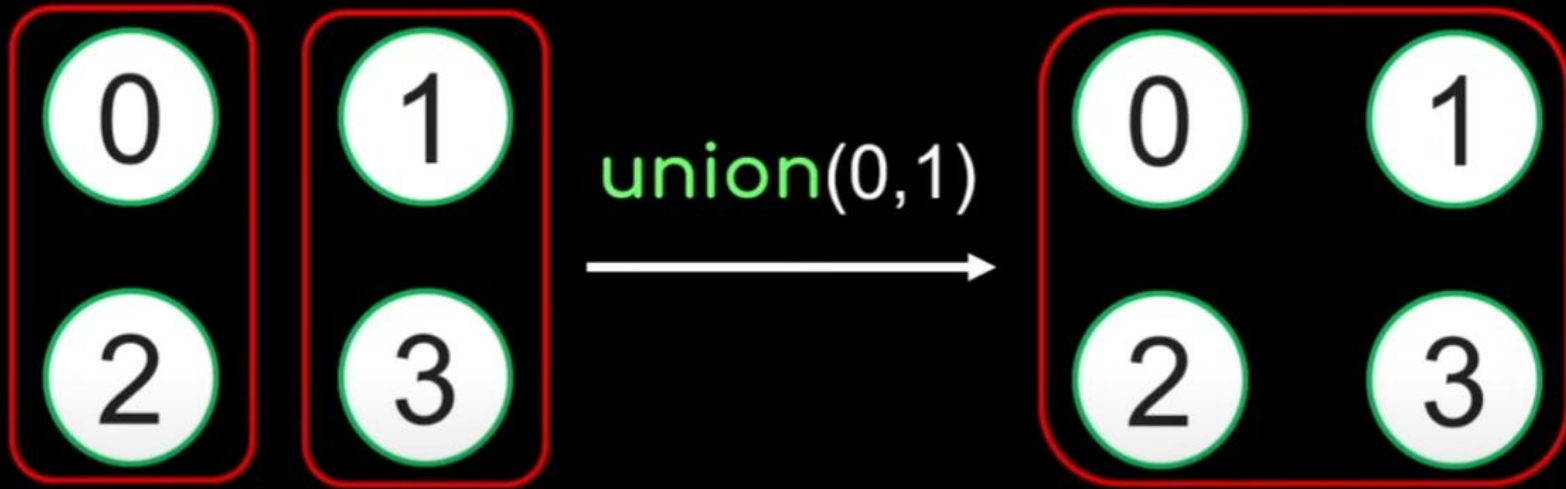


Minimal Spanning Tree and applications in clustering

**Mustafa Hajij**

## A quick introduction to Union-Find

1. **union**(x, y) Unions the groups containing x and y

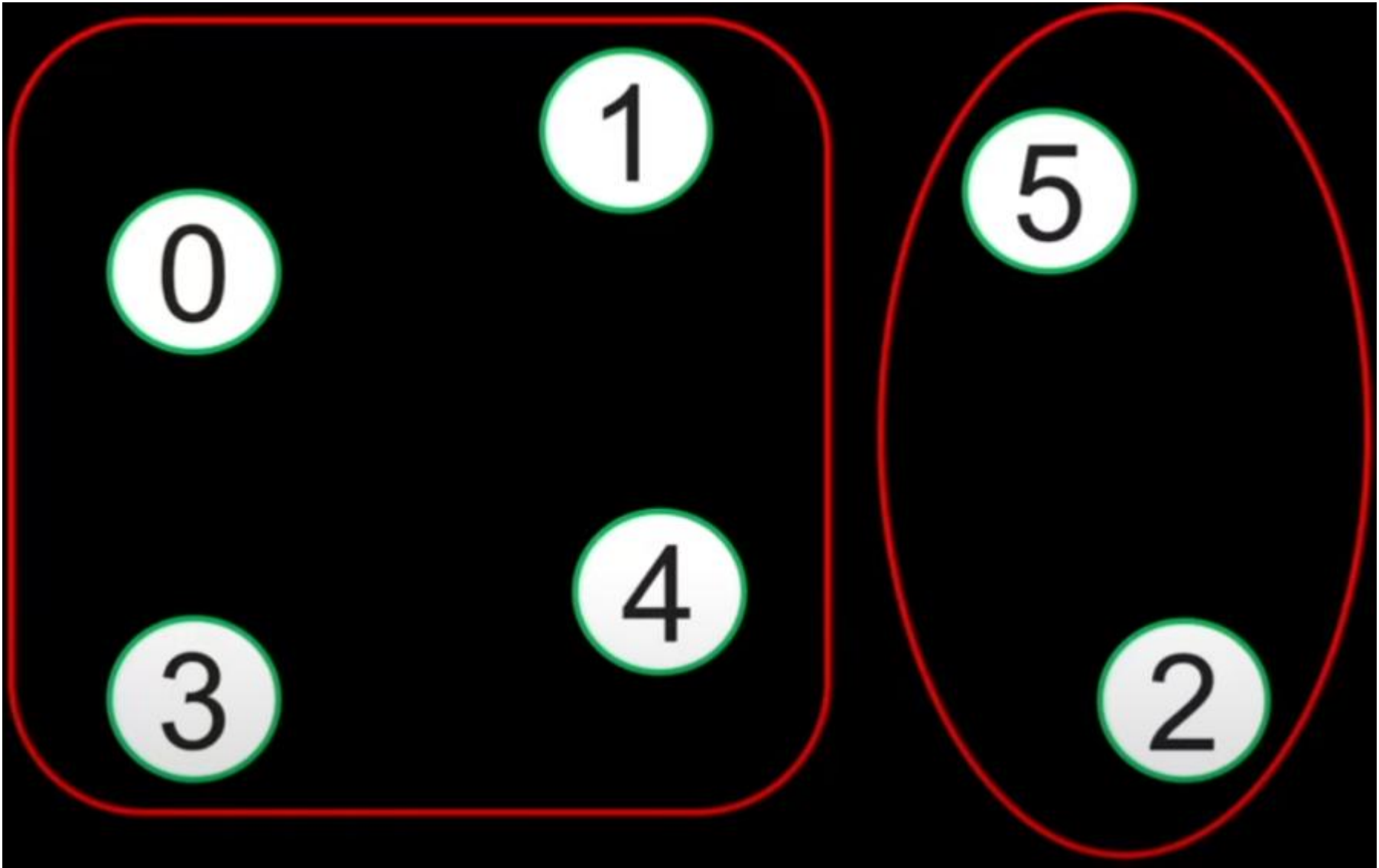


2. **find**(x) Find the group x belongs to

We want a data structure that can do the above two operations easily and quickly (Union and Find).

Ref : <https://www.youtube.com/watch?v=ayW5B2W9hfo>

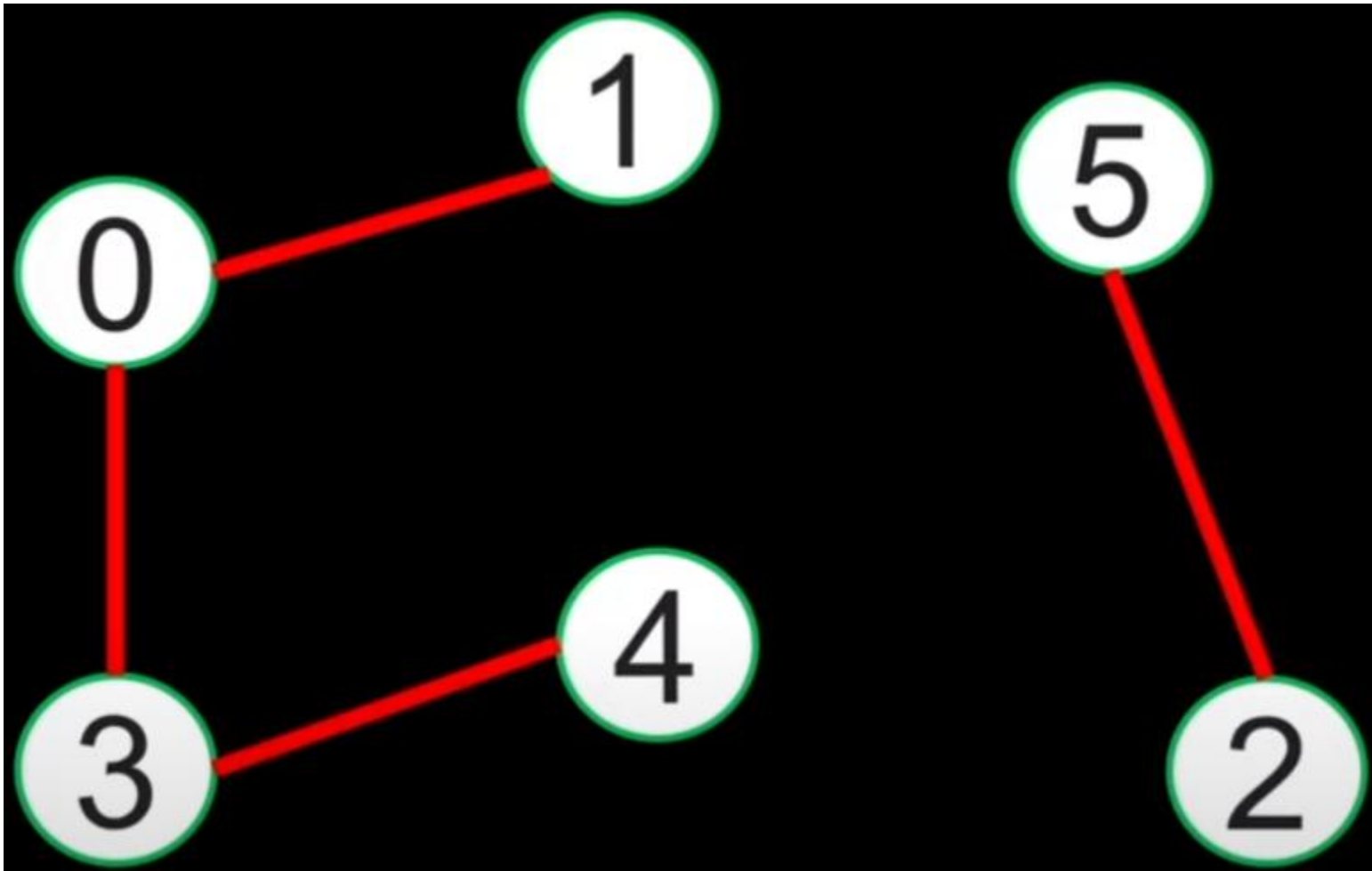
## A quick introduction to Union-Find



Lets discuss find first.

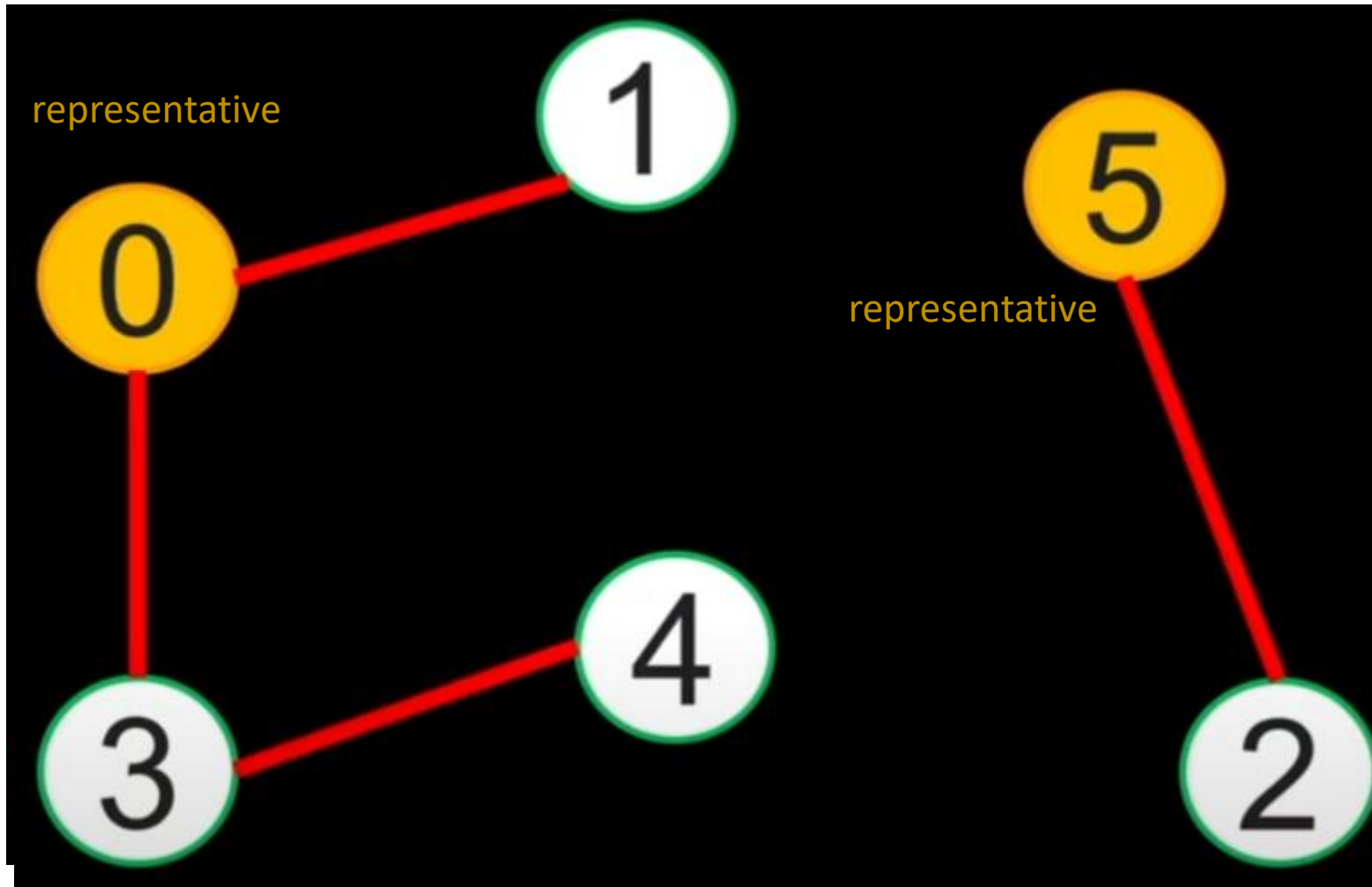
Ref :<https://www.youtube.com/watch?v=ayW5B2W9hfo>

## A quick introduction to Union-Find



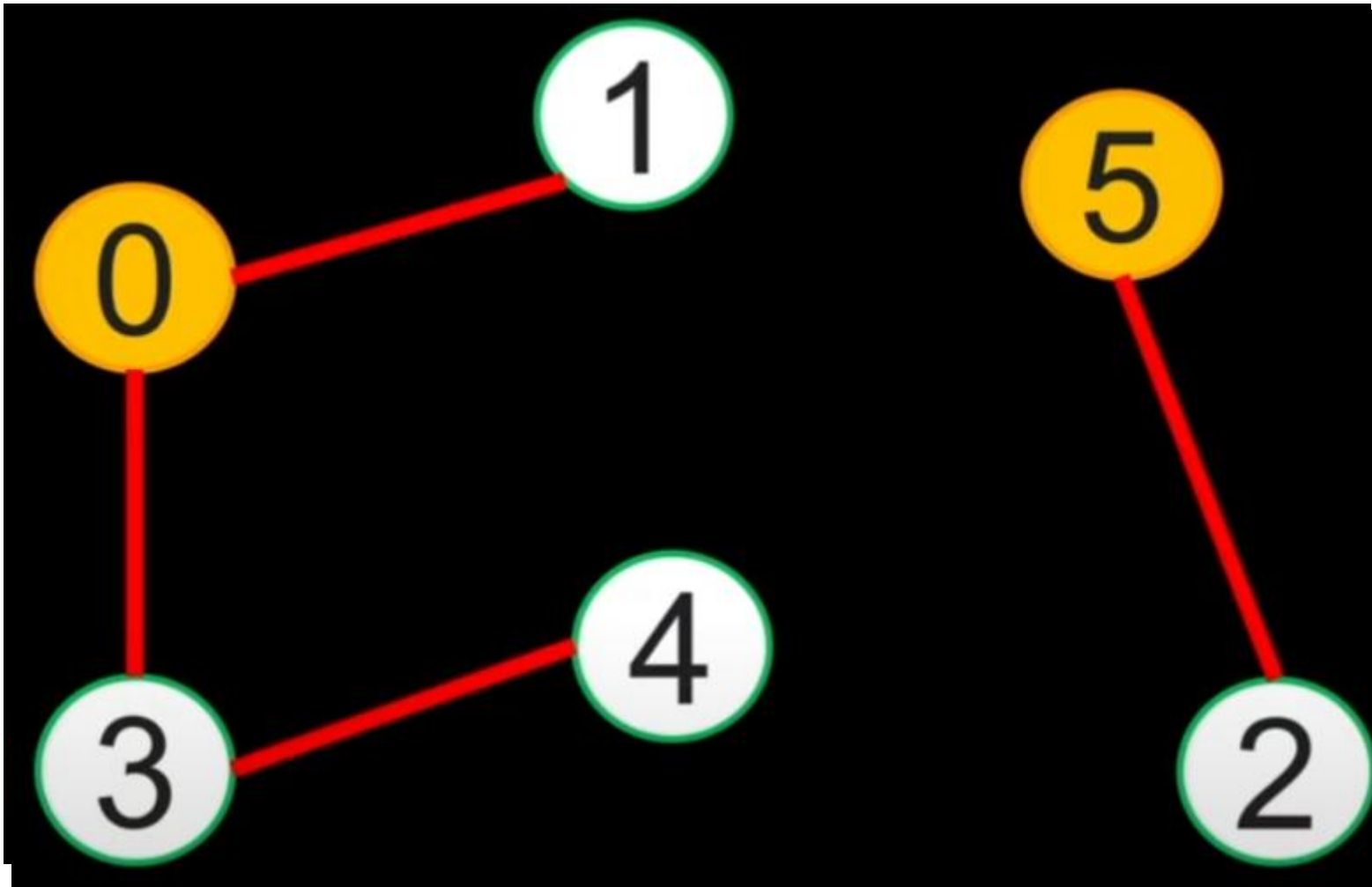
Connect all elements that belong to the same group by edges

## A quick introduction to Union-Find



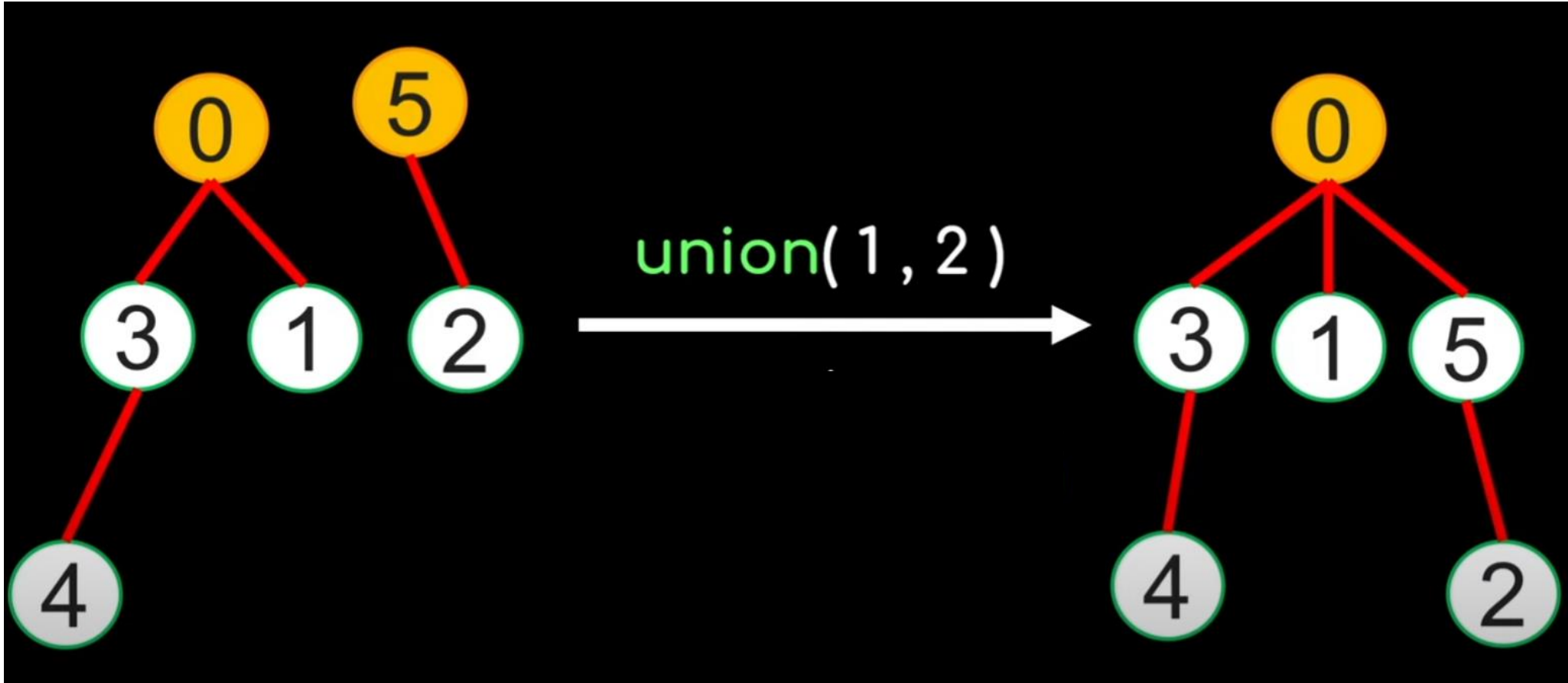
- Next we set a **representative** to each group.
- We design the data structure such that **find(4)=0** and **find(2)=5** Ref :<https://www.youtube.com/watch?v=ayW5B2W9hfo>

## A quick introduction to Union-Find



- Two elements  $u$  and  $v$  belong to the same group iff  $\text{find}(u) = \text{find}(v)$

## A quick introduction to Union-Find



- Union (say `union(2,1)` ) is a matter of merging the trees together!

## A quick introduction to Union-Find

In short a union find data structure allows for three operations :

- (1) MAKE-SET( $v$ ) : Make a connected component from the node.  $O(1)$
- (2) FIND-SET( $u$ ) : given a node  $u$ , returns a pointer to the connected component it belongs to.  $O(\log(n))^*$
- (3) UNION( $u, v$ ): Given two nodes  $u, v$  that may belong to two separate connected components, merge these two separate connected components into a single one.  $O(\log(n))^*$

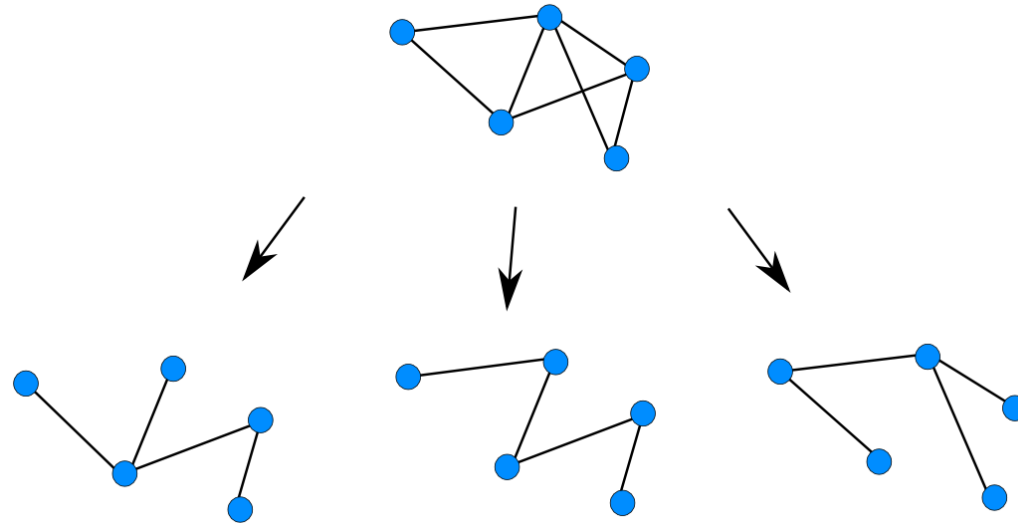
\* The optimal complexity is in fact  $O(\alpha(n))$ ,  $\alpha(n)$  is the extremely slow-growing inverse [Ackermann function](#).



# Spanning Tree

Let  $G = (V, E)$  be a connected weighted graph. A **spanning tree** for  $G$  is a subgraph of  $G$  which includes all of the vertices of  $G$  and is a tree.

A graph might have more than one spanning tree

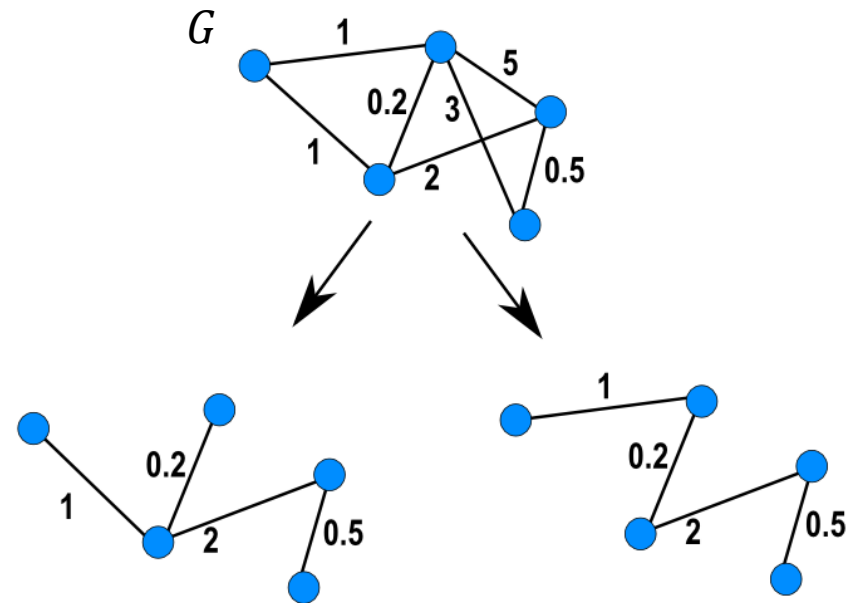


Spanning trees for  $G$

# Minimal Spanning Tree

Let  $G = (V, E, w)$  be a connected weighted graph. A **minimal spanning tree** for  $G$  is a spanning tree whose sum of edge weights is as small as possible.

A graph might have more than one minimal spanning tree. However, if all edges in the graph have unique weights then the minimal spanning tree is unique.



Minimal spanning trees for  $G$

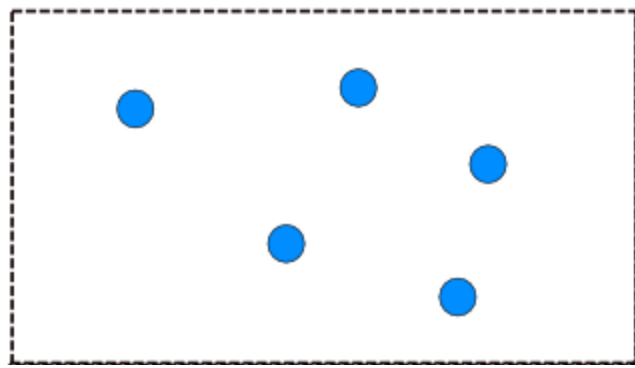
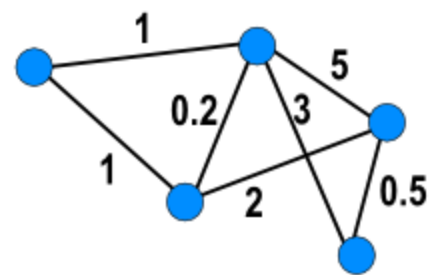
# Kruskal's Algorithm

Let  $G = (V, E, w)$  be a connected weighted graph. The Kruskal's algorithm is a greedy algorithm.

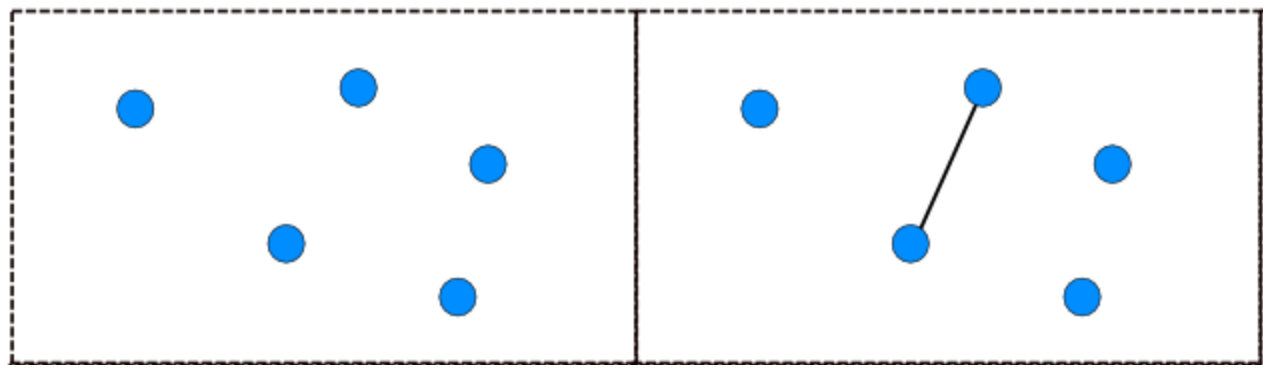
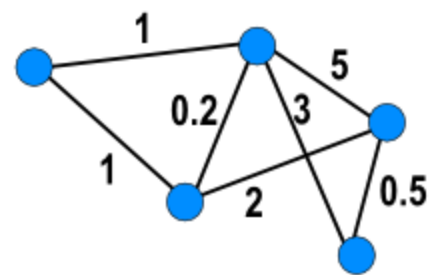
Informally, the algorithm can be given by the following three steps :

1. Set  $V_T$  to be  $V$ , Set  $E_T = \{\}$ . Let  $S = E$
2. While  $S$  is not empty and  $T$  is not a spanning tree
  1. Select an edge  $e$  from  $S$  with the minimum weight and delete  $e$  from  $S$ .
  2. If  $e$  connects two separate trees of  $T$  then add  $e$  to  $E_T$

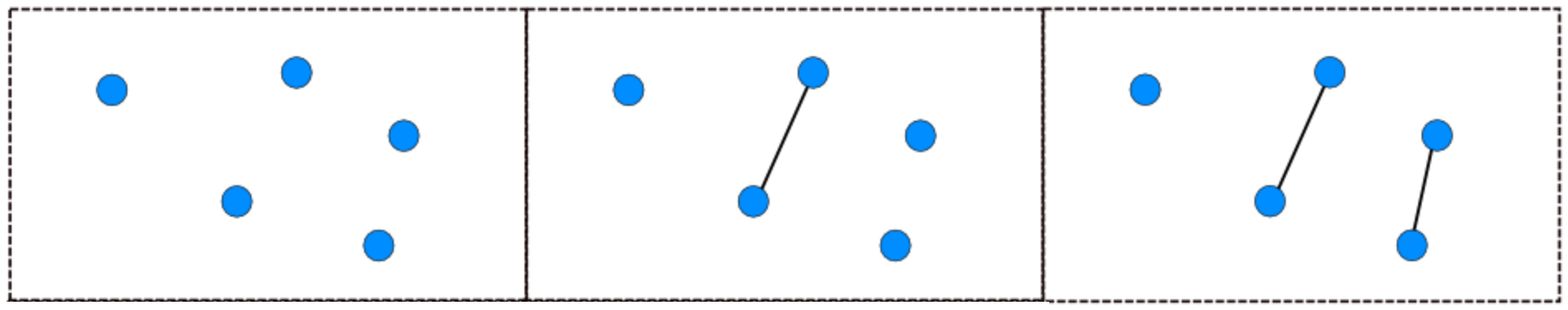
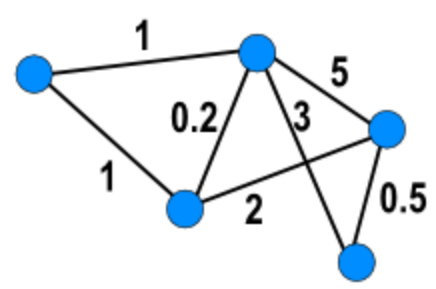
# Kruskal's Algorithm Example



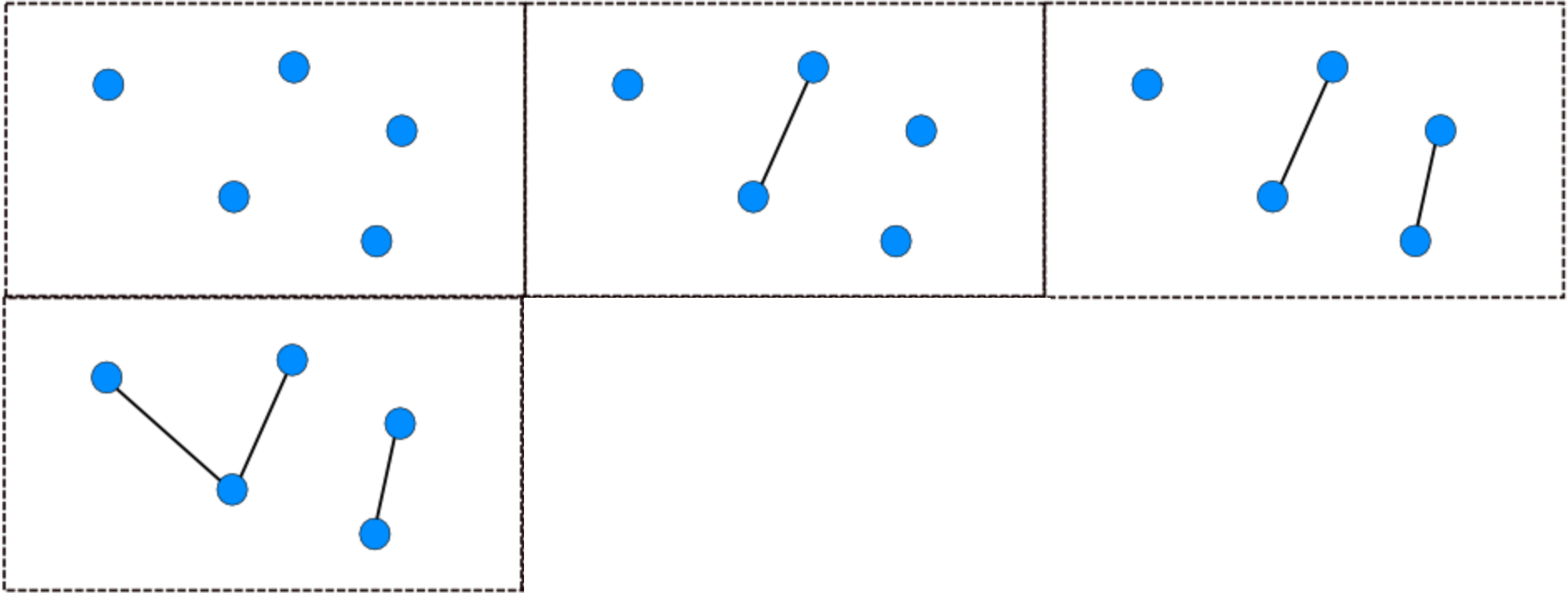
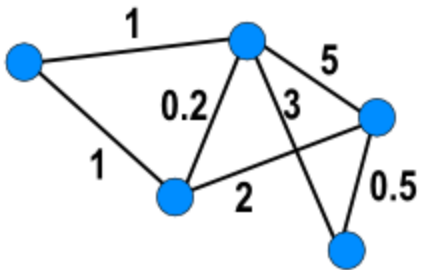
# Kruskal's Algorithm Example



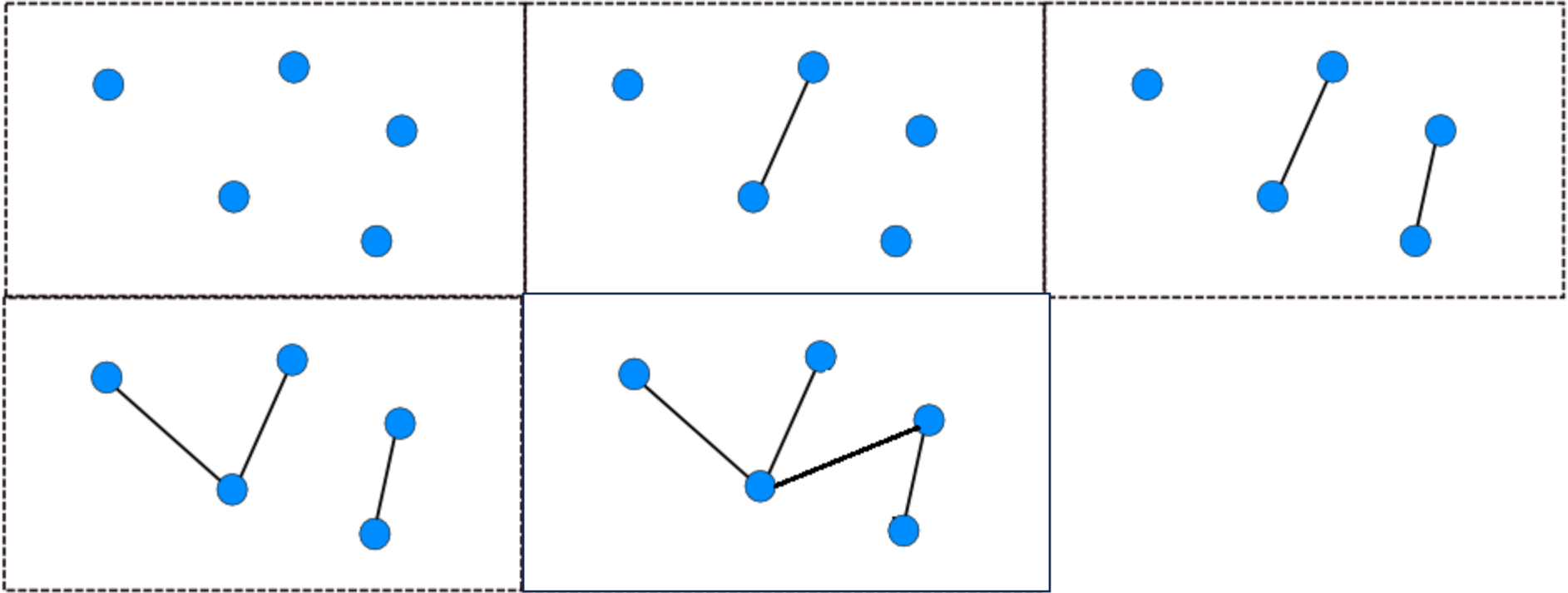
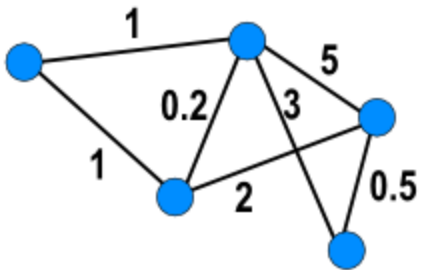
# Kruskal's Algorithm Example



# Kruskal's Algorithm Example



# Kruskal's Algorithm Example





# Kruskal's Algorithm

Let  $G = (V, E, w)$  be a connected weighted graph. The Kruskal's algorithm is a greedy algorithm

This can be implemented using [union-find](#) data-structure

```
1- A = {}
2- foreach  $v \in V$ :
3-     MAKE-SET( $v$ )
4- foreach  $(u, v)$  in  $E$  ordered by  $\text{weight}(u, v)$ , increasing:
5-     if FIND-SET( $u$ )  $\neq$  FIND-SET( $v$ ):
6-          $A = A \cup \{(u, v)\}$ 
7-         UNION( $u, v$ )
8- return A
```

## Prim's Algorithm

Let  $G = (V, E, w)$  be a connected weighted graph. The Prim's algorithm is a greedy algorithm

Informally, the algorithm can be given by the following three steps :

1. Select an arbitrary vertex  $v$  from  $V$ . Set  $V_T = \{v\}$  and  $E_T = \{ \}$
2. Grow the tree by one edge : choose an edge  $e(u,v)$  from the set  $E$  with the lowest cost such that  $u$  in  $V_T$  and  $v$  is in  $V \setminus V_T$  then add  $v$  to  $V_T$  and add  $e$  to  $E_T$
3. If  $V_T = V$  break, otherwise go to step 2.

## Application to Clustering : Zahn's algorithm

Suppose that we are given a set of a weighted graph  $G$ .

1. Construct the MST of  $G$  (using say Kruskal's algorithm).
2. Remove the inconsistent edges to obtain a collection of connected components (clusters).
3. Repeat step (2) as long as the termination condition is not satisfied.

In this case, an edge in the tree is called **inconsistent** if it has a length more than a certain given length  $L$ . Zahn's algorithm that we used to obtain a clustering algorithm on point cloud can be simply used to obtain a clustering algorithm on graphs as follows. The connected components of the remaining forest are the clusters of the graph.

Question : how can you apply this algorithm to point cloud?