# Minimum Spanning Trees using Kruskal's Algorithm with Path Compression and Union by Rank

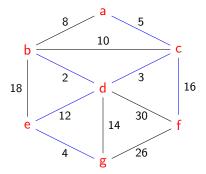
Alan R. Hahn

Clemson University

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#### Definition

Given an undirected graph G with real valued weight on each edge, a Minimum Spanning Tree T of a graph G is a spanning tree of G such that the sum of the edge weights in T is minimum with respect to all spanning trees of G.



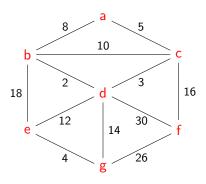
The blue subtree is the minimum spanning tree of G.

Consider an undirected graph G with real valued weight on each edge.

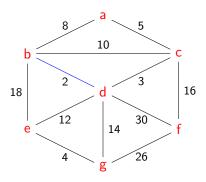
#### Kruskal's Algorithm

Apply the following step to the edges of G in non - decreasing order by edge weight:

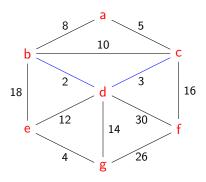
## Inclusion Step



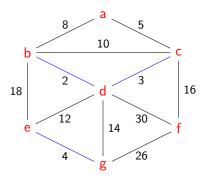
After 0 steps



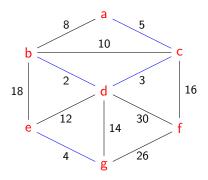
After 1 step



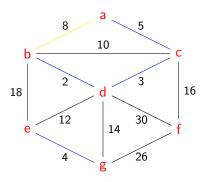
After 2 steps



After 3 steps

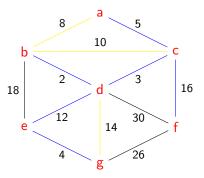


After 4 steps



After 5 steps

If the edge  $e = \{u, v\}$  is such that u and v are in the same blue subtree, leave e uncolored. Else, color e blue.



After all steps, terminate and return the blue subtree which is the minimum spanning tree of G.

#### Requirements

- Sort the list of edges *E* of *G*
- Check whether two vertices share a blue subtree

#### The Partition Class

A class to represent a Partition of a given input set S. To identify a given part of a partition, an arbitrary but unique element will be maintained within each part as a representative element.

- initialize\_partition(S): Initialize the partition as a collection of one element parts, one for each s in S
- find(s): Return the representative element of the part containing the element s
- link(x,y): Union the two parts whose representative elements are x and y, and choose a new representative element to represent the one part

$$S = \{a, b, c, d, e, f, g\}$$

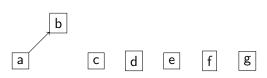
 $initialize\_partition(S)$ 

a b c d e f g

$$S = \{a, b, c, d, e, f, g\}$$

$$link(a, b):$$

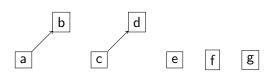
[a:b, b:b, c:c, d:d, e:e, f:f, g:g]



$$S = \{a, b, c, d, e, f, g\}$$

$$link(c, d):$$

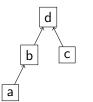
[a:b, b:b, c:d, d:d, e:e, f:f, g:g]



$$S = \{a, b, c, d, e, f, g\}$$

$$link(b, d):$$

[a:b, b:d, c:d, d:d, e:e, f:f, g:g]



е

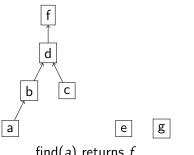
f

g

$$S = \{a, b, c, d, e, f, g\}$$

$$link(d, f)$$

[a:b, b:d, c:d, d:f, e:e, f:f, g:g]

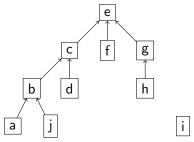


find(a) returns f

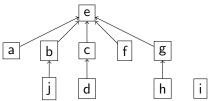
#### Two Heuristics to improve runtime

- Path Compression: For an input vertex a and representative element s, find(a) returns s and updates to s the parent of every node on the path from a to s
- Union by Rank: Changes the map and the link method so that along with the parent node, the map of a given vertex s returns a non-negative integer function called the rank of s. For each element its rank is initialized to 0. Given two representatives x and y, if without loss of generality rank(x) > rank(y), link chooses as the representative of the union of the parts which contain x and y to be x. If rank(x) = rank(y), one of rank(x), rank(y) is increased by one and the element chosen to have higher rank is then the representative of the union of the parts which contain x and y

#### Path Compression



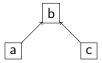
find(a) returns e and updates the parent node of every vertex along the path from a to e:



#### Union by Rank



С



#### References

- Kumar, S., Spezzano, F., Subrahmanian, V., & Faloutsos, C. (2016).
   Edge weight prediction in weighted signed networks. In *Data mining* (icdm), 2016 ieee 16th international conference on (pp. 221-230).
- Tarjan, R. E. (1983). Data structures and network algorithms.
   Philadelphia, PA, USA: Society for Industrial and Applied Mathematics.