

Minimum Spanning Tree

54
L16

Lemma: (Recall problem and lemma from last lecture)

Prim's Algorithm

```

 $T \leftarrow \{v\}$  (arbitrary)
while  $|T| < n$  do
    add the shortest
    edge  $(a, b)$  with
     $a \in T$  and  $b \notin T$  to  $T$ 
return  $T$ 

```

Kruskal's Algorithm

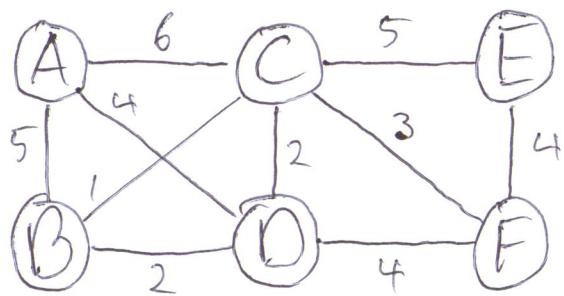
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 $E \leftarrow$  Sort all edges by length
 $T \leftarrow \emptyset$ 
for  $i \leftarrow 1$  to  $|E|$  do
    add  $E[i]$  to  $T$  if it
    does not create a cycle
return  $T$ 

```

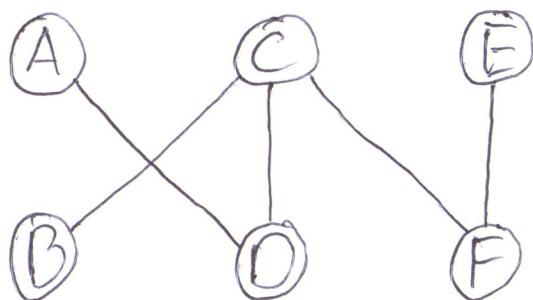
I'll analyze Kruskal's algorithm - Prim's is in A4!

Example:



Edges:
 $BC, CD, BD,$
 $CF, DF, EF,$
 AD, AB, CE, AC .

Result:



- Correct? There are two parts to this: (55)

Claim: ~~All~~ The graph returned by Kruskal is connected.

Proof: Suppose it is not. Then there are at least two connected components C_1 and C_2 . ~~Each edge joining C_1 and C_2~~ But the shortest edge joining C_1 and C_2 would have been added to T , so this is impossible. \square

Claim: Before and after each iteration of the loop, there is an MST containing all edges of T .

Proof: The claim holds initially, as any MST contains the empty set.

Suppose that the claim is true before the loop. Let ~~M~~ be than MST containing all edges of T , and let e be the next edge considered. If e creates a cycle in T , or $e \in M$ then we are done. If e is added to T , but $e \notin M$, then e creates a cycle in M that has an edge $f \notin T$. Then $T' = T - f + e$ is also a spanning tree, and it has the same weight as T , since the algorithm took e before f , so $w(e) \leq w(f)$. \square

- Terminates? Yes.

- Efficient?

Sorting the edges takes $O(|E| \log |E|) = O(|E| \log |V|)$ since ~~$|E|=O(|V|^2)$~~ . $|E| \leq |V|^2$.

Creating T and adding edges to it can be done in $O(|V|)$ time.

If we can check for a cycle in $O(s)$ time, the for-loop takes $O(|E|s)$ time.

$$\Rightarrow \text{Total: } O(|E|(\log |V| + s))$$

So if we can check for cycles in $O(\log |V|)$ time, this is $O(|E| \log |V|)$ total.

We use a union-find data structure.

This ~~can also~~ supports two operations:

- Union(x, y): merge the sets containing x and y
- Find(x): return the unique identifier of the set containing x .

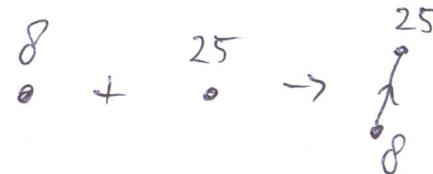
"does (u, v) create a cycle" \Rightarrow "Find(u) \neq Find(v)?"

It works as follows:

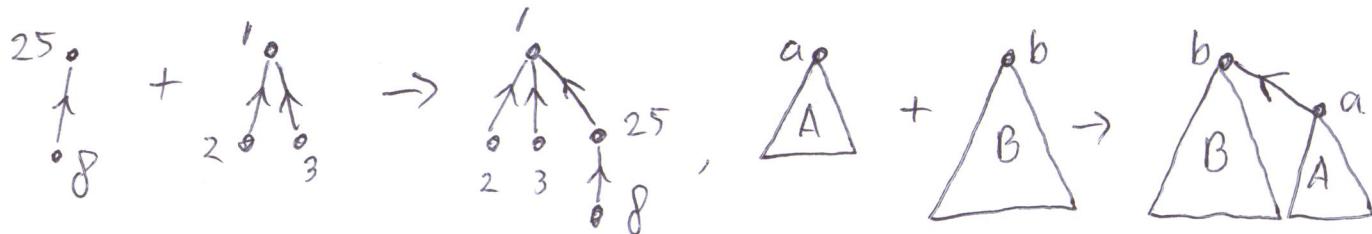
- Start by making each vertex its own set:

1	2	3	4	...	n
0	0	0	0	...	0

- When 2 sets merge, we create a rooted tree: (57)

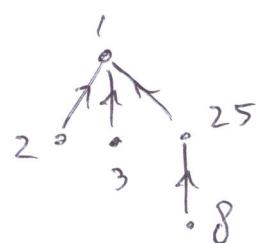


- When 2 trees merge, we make the smaller a child of the larger one with more: tree with fewer levels



If each root stores the number of levels, ~~mergethion(x,y)~~ takes $O(1)$ time.
merging

- We implement $\text{Find}(x)$ by walking to the root of x 's tree:



$$\text{Find}(8) = \text{Find}(25) = \text{Find}(1) = 1$$

$$\text{Find}(2) = \text{Find}(1) = 1$$

This takes $O(\# \text{levels})$ time.

Claim: A tree with k levels has at least 2^{k-1} nodes.

Proof: This is true initially, since $2^0 = 2^0 = 1$.

If the level doesn't increase with a merge, it stays true. If the level does increase, both trees had k levels, so the new tree has at least $2^{k-1} + 2^{k-1} = 2 \cdot 2^{k-1} = 2^k = 2^{(k+1)-1}$ nodes. \square

This means that the maximum level with n vertices is $O(\log n)$.

Thus, both Union and Find take $O(\log n)$ time.

With a small improvement, called path compression, this can be brought down to $O(\alpha(n))$. Here $\alpha(n)$ is an extremely slowly growing function called the inverse Ackermann function. For any realistic input size, $\alpha(n) < 5$.