

CSCI 136

Data Structures & Advanced Programming

Graph Applications:
Minimum Cost Spanning Trees

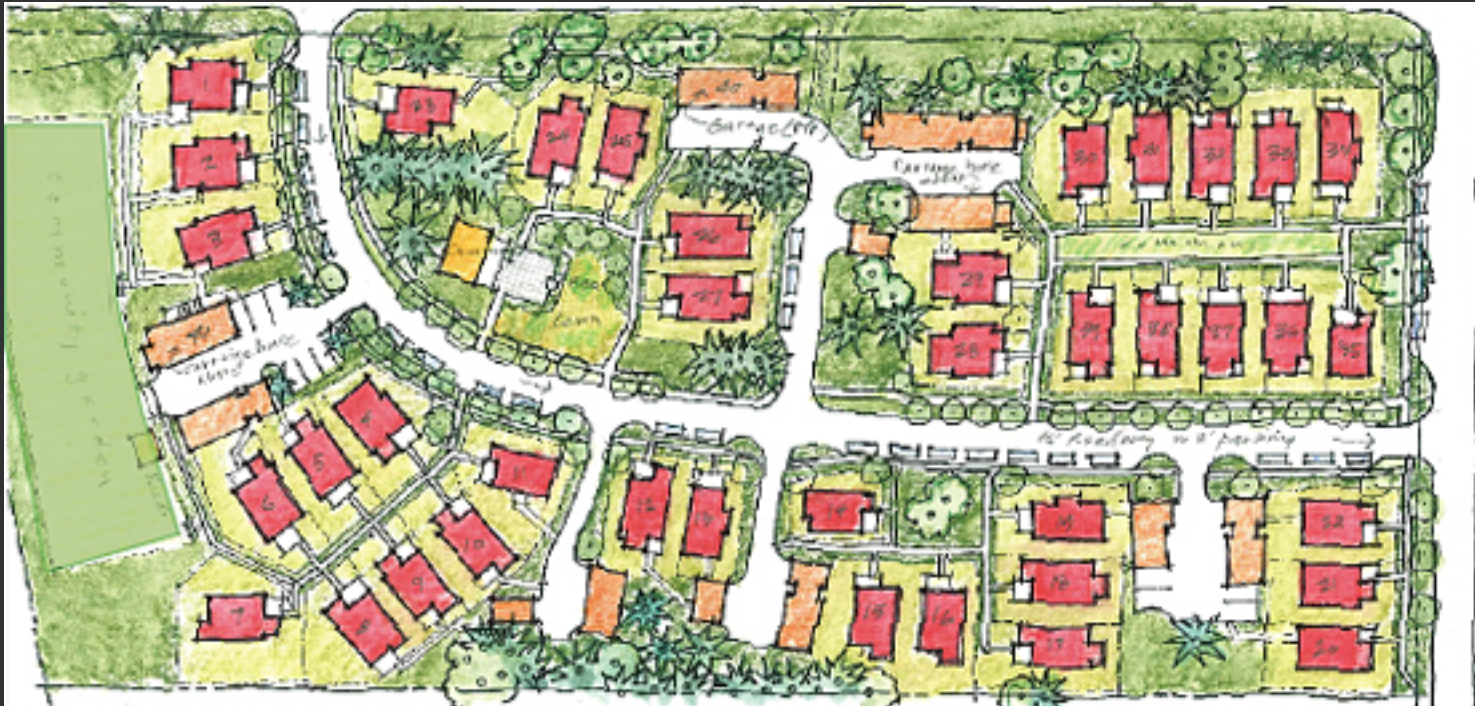
Video Outline

- Spanning subgraphs
- Spanning trees
- Prim's algorithm to calculate spanning trees with the minimum cost
 - Description
 - Proof
 - Pseudocode
 - Implementation in structure5

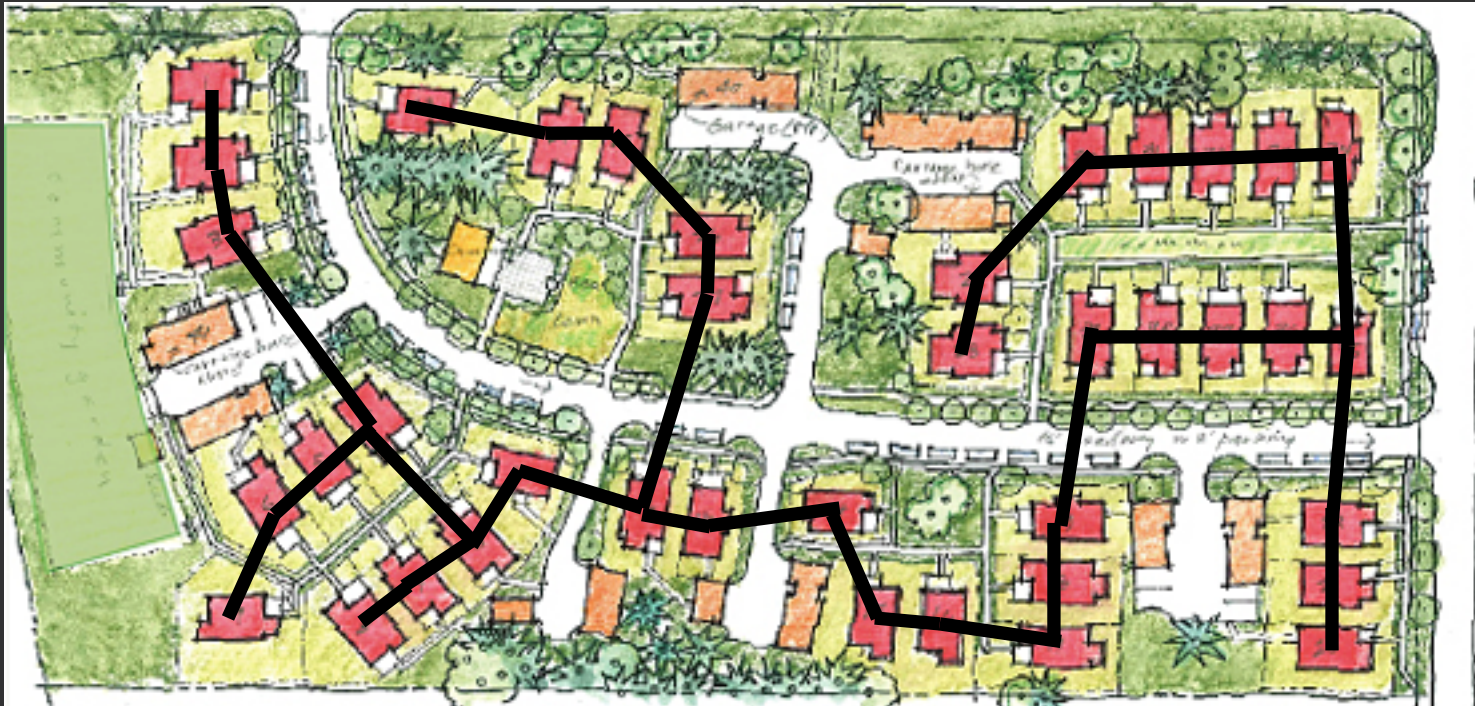
Motivation

- Let's say we have a neighborhood of houses
- Want to create an electrical grid
- Goal: each house needs to be connected to a single network
- (In other words, there is a path along the electrical wires between any two houses)
- Also works for creating ethernet networks, etc.

Motivation



Motivation



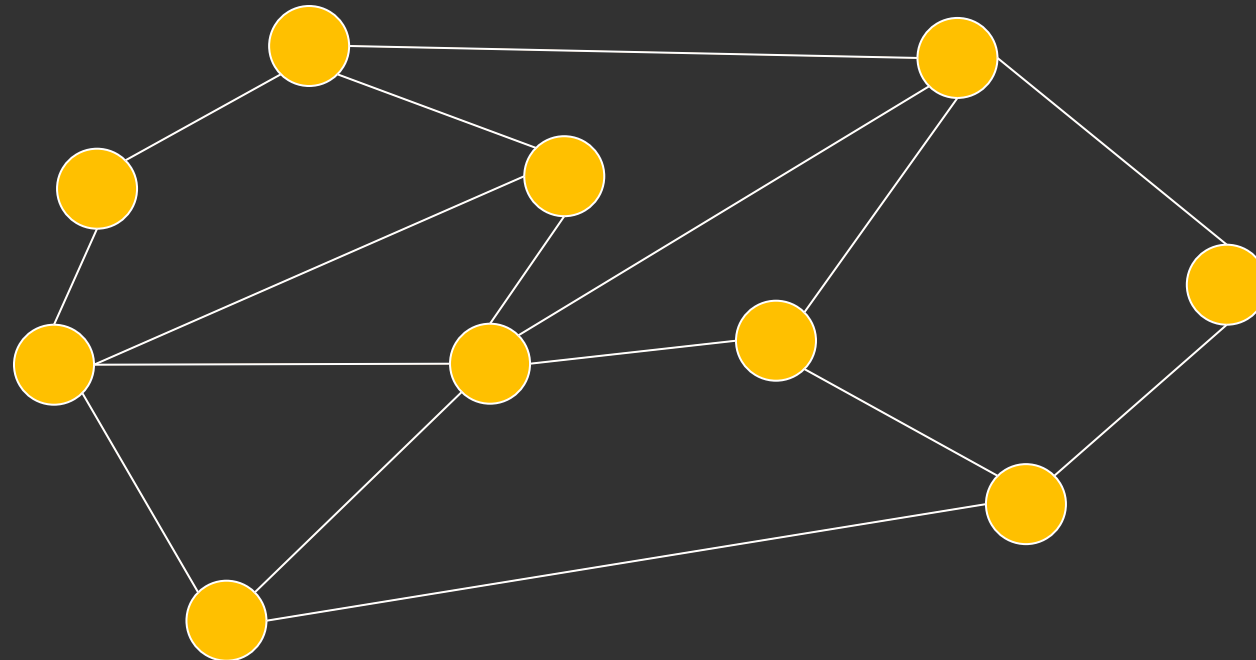
Goal

- Graph problem!
- Select edges to connect all vertices using a single tree
- Can only select edges *in the original graph*
- Want to select the minimum cost:
 - Minimum *total* of edge weights in the tree

Spanning Trees

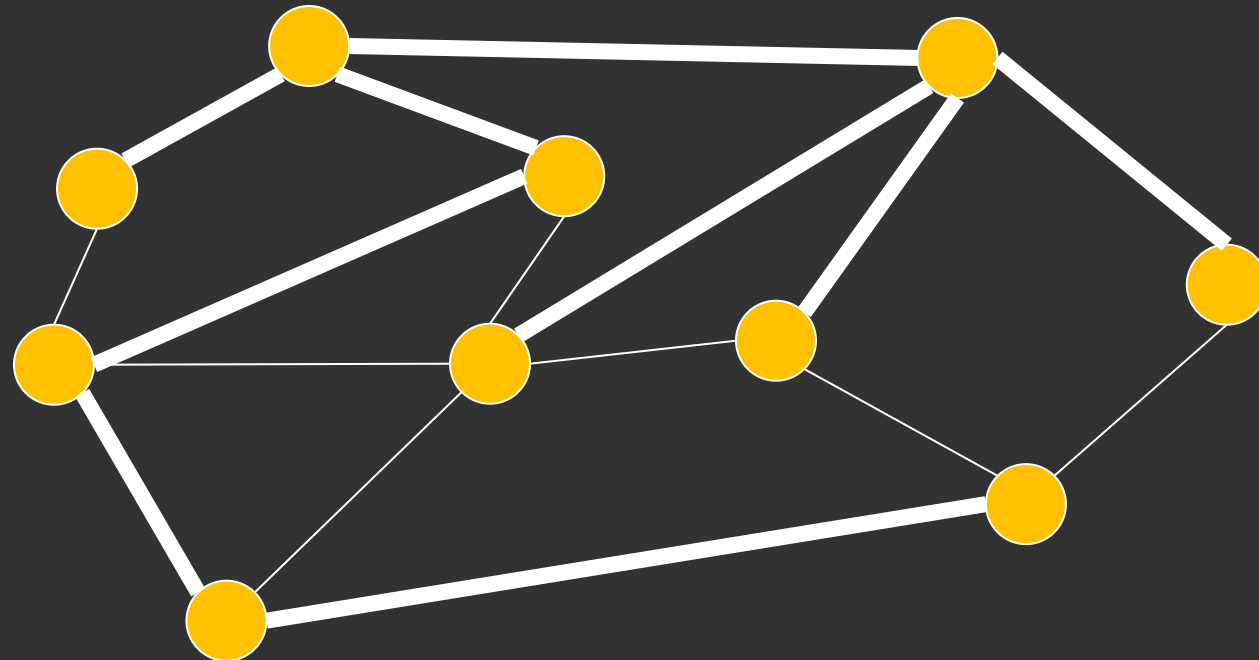
Collection of vertices and edges from the original graph

- A **spanning tree** is a subgraph that **covers** all the vertices using the *minimum number of edges*



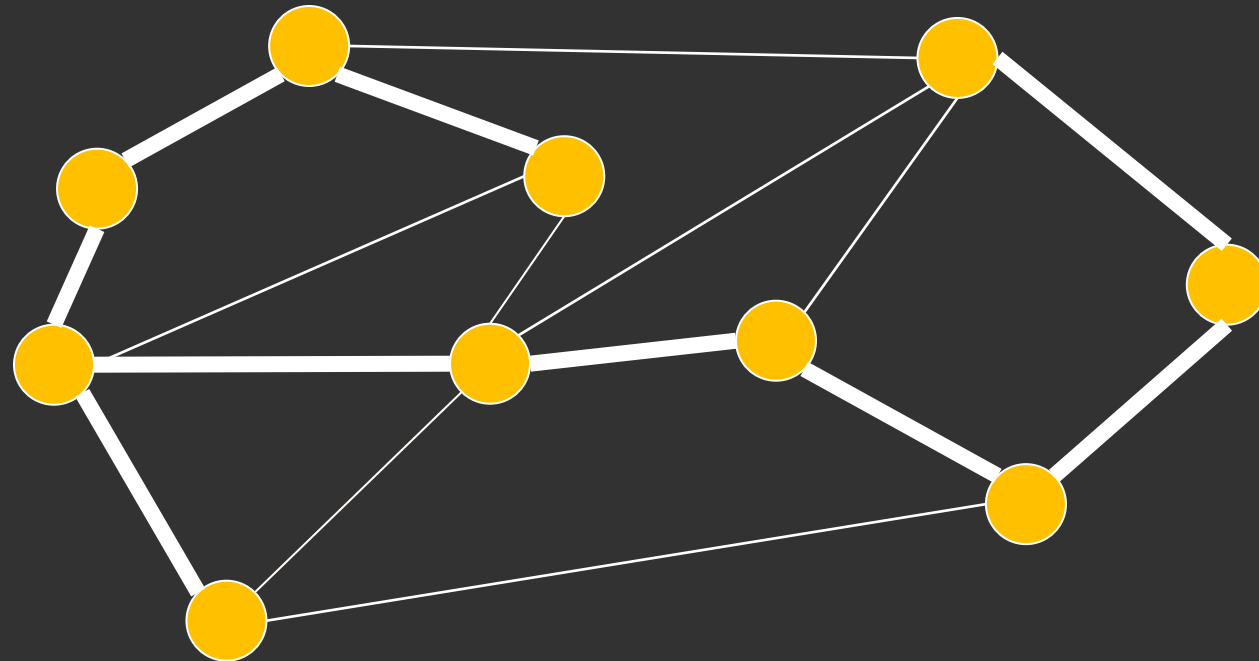
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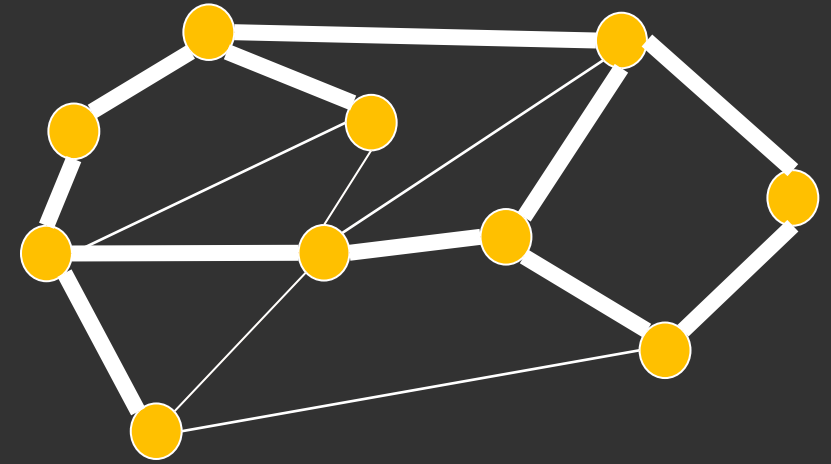


Why Trees?

Theorem: The minimum collection of edges that connects all vertices is a tree

Proof idea:

- If it's not a tree, then it contains some cycle C
- If we remove an edge from C , the resulting collection of edges still connects all vertices in the tree
- Repeat this process of removing edges until no more cycles remain
- Now we are left with a tree

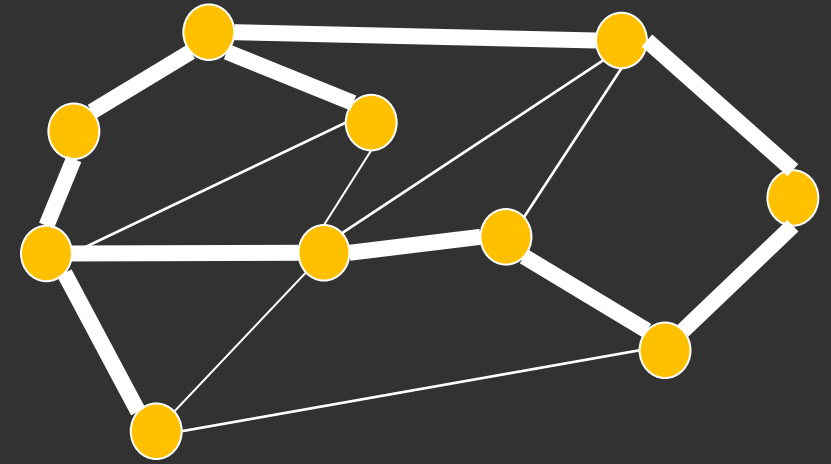


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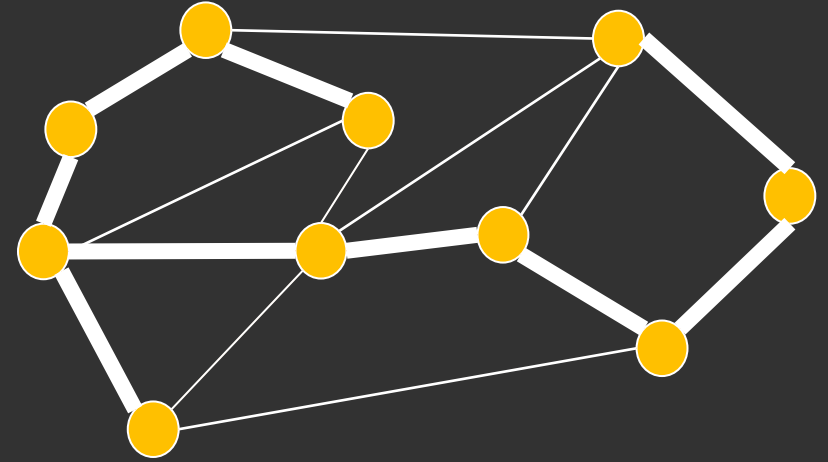


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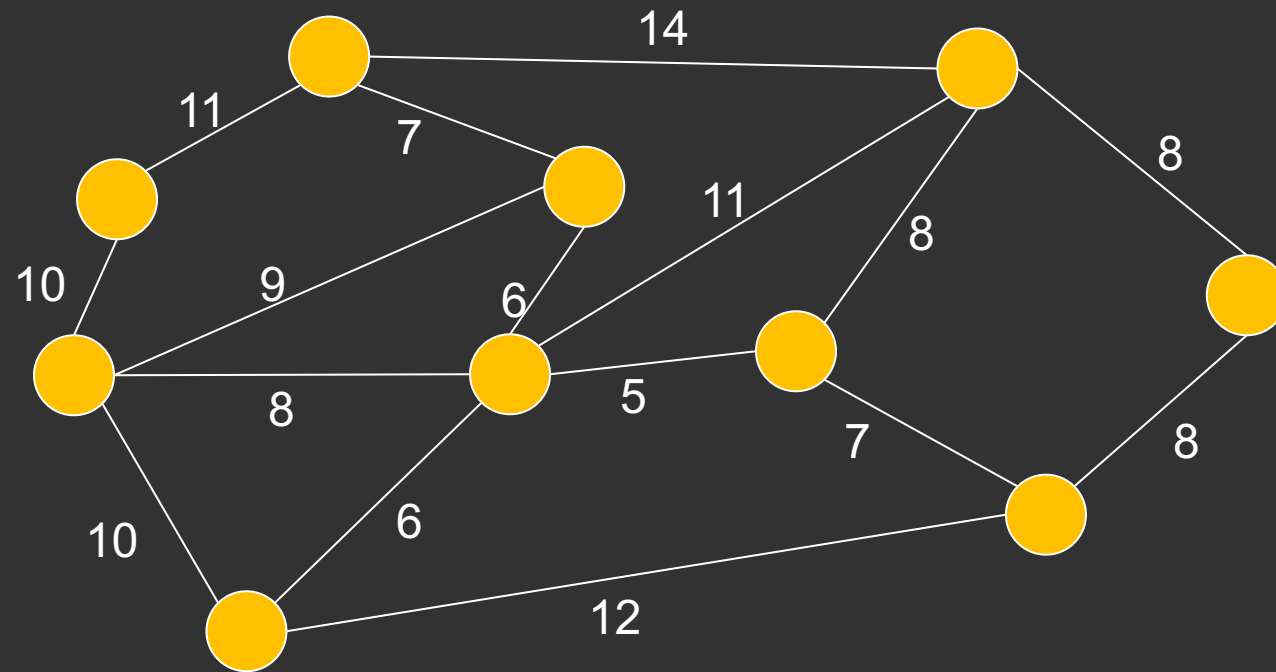
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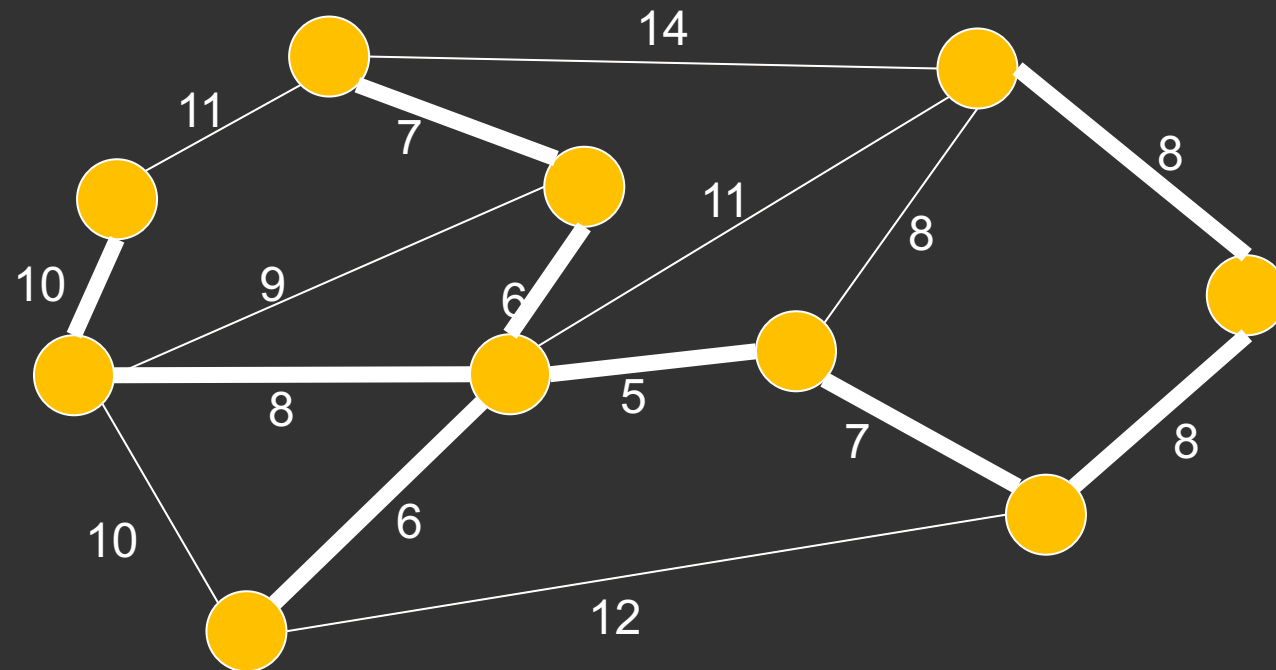
Minimum-Cost Spanning Trees

- Suppose we're given a graph that is:
 - connected, and
 - has weighted edges (integer, float, double, etc.)
- A **minimum cost spanning tree** is a spanning tree where the *sum of all the edge weights* is the smallest possible

Minimum-Cost Spanning Trees



Minimum-Cost Spanning Trees

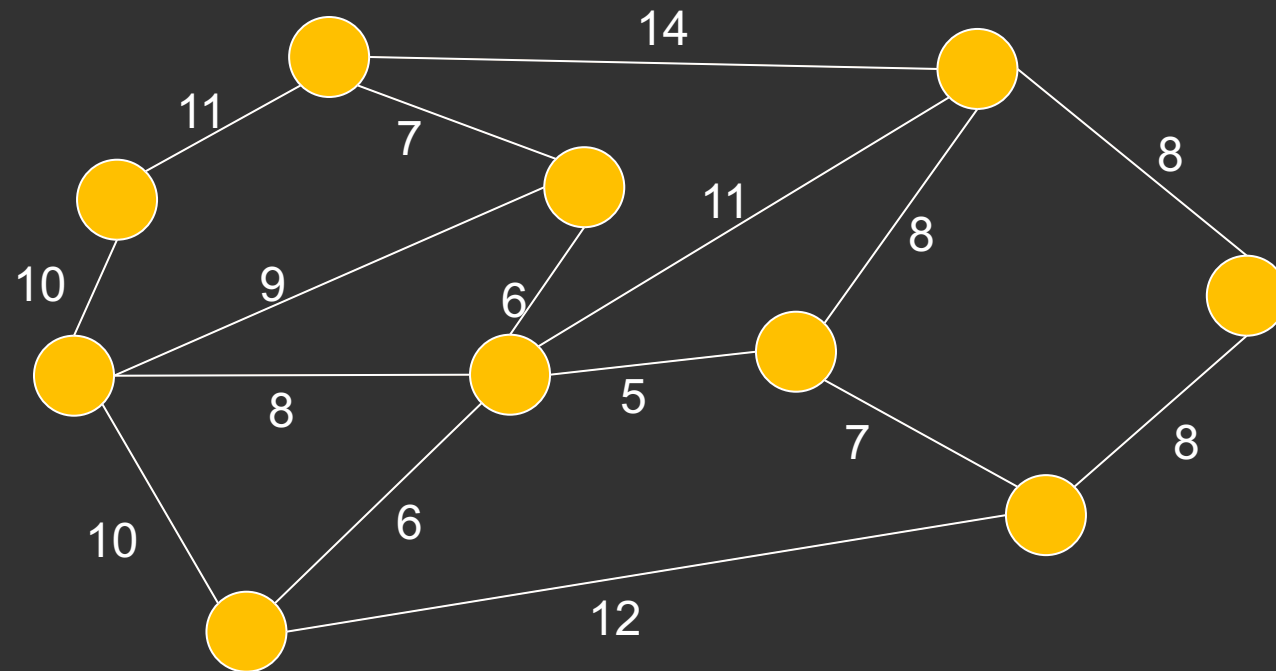


How can we find an MCST?

- Let's start with some node in our tree
- We need to hook it up to the network...how?
- Let's use the cheapest edge
- Now we have a two-node network. Need to expand this network again...how?
- Take the cheapest edge connecting either of the two nodes currently in the network to an outside node
- Repeat until n nodes in the tree!

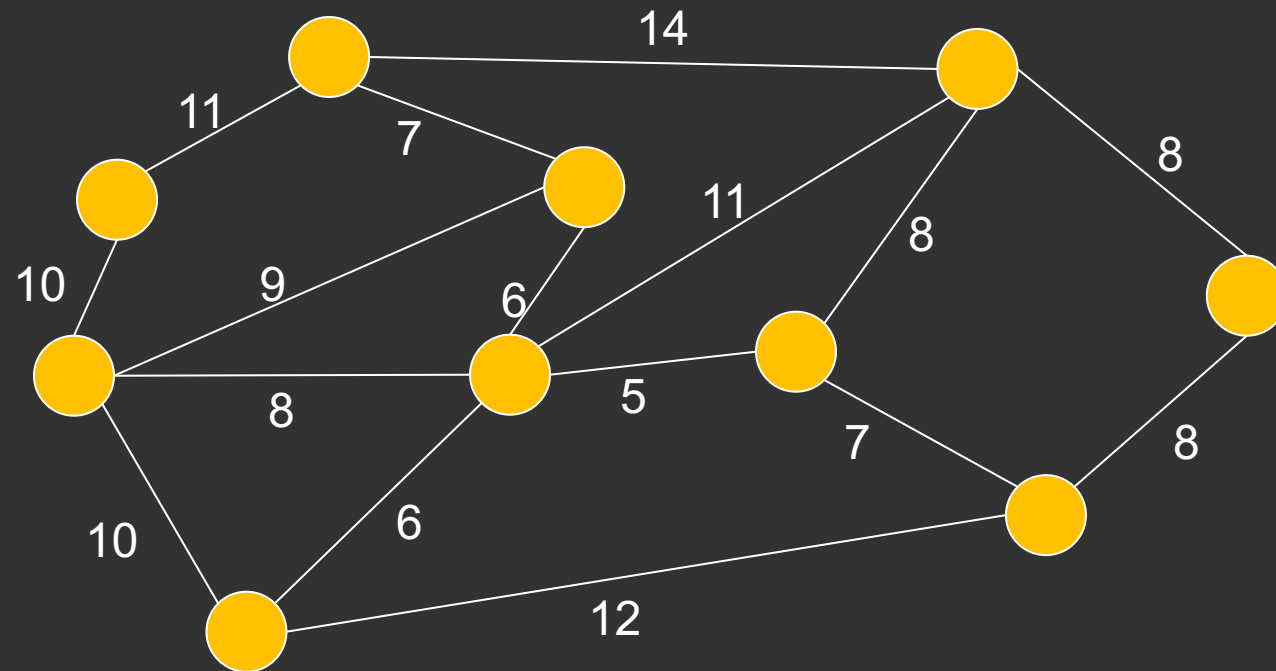
Prim's Algorithm for MCSTs

- Let's walk through an example to solidify the algorithm. In this example, not all edge weights are unique.



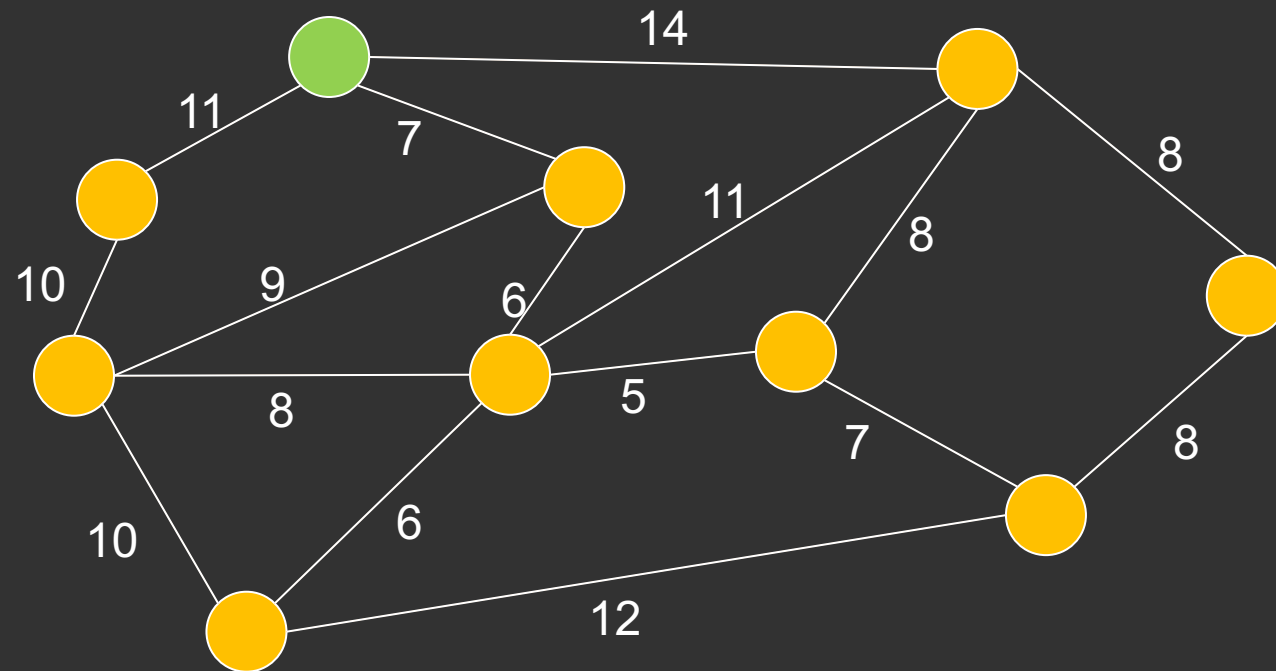
Prim's Algorithm for MCSTs

- Start by picking some vertex



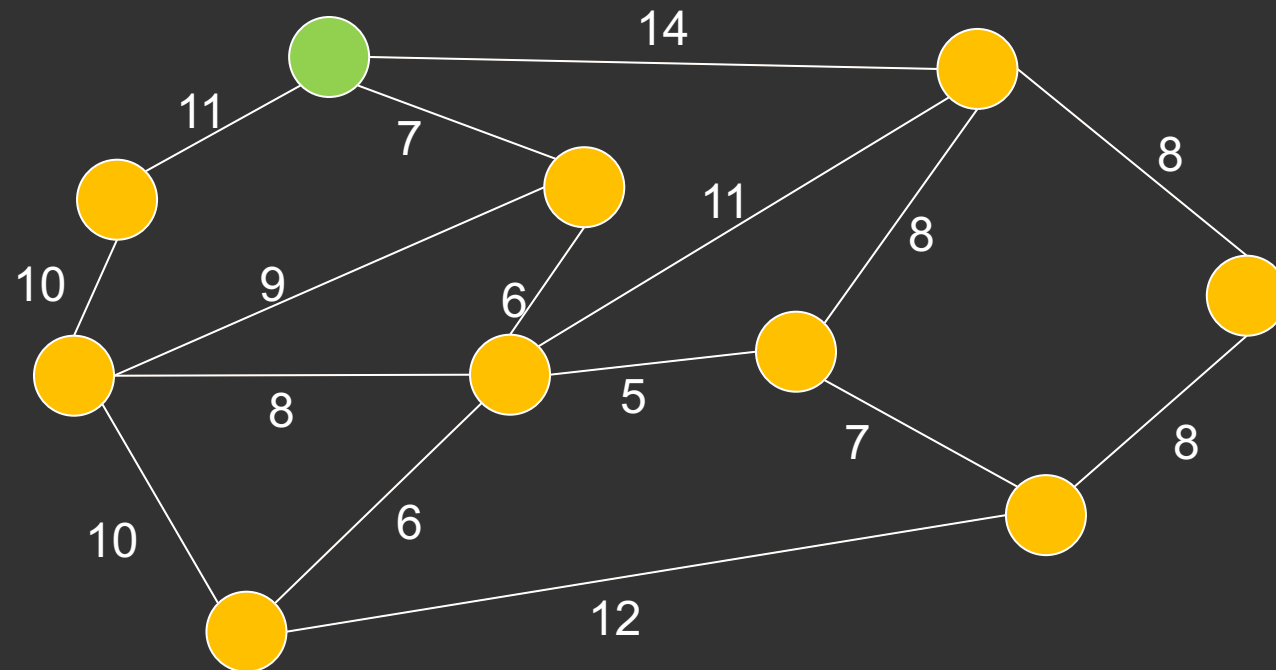
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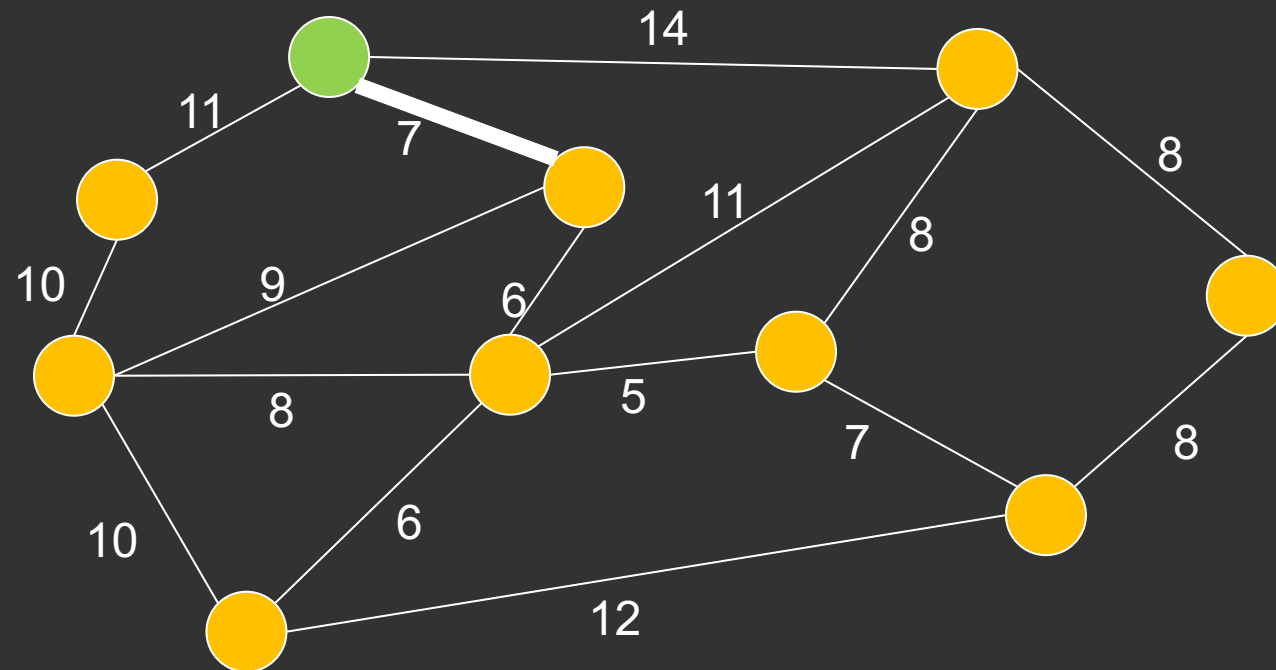
Prim's Algorithm for MCSTs

- We'll note our current tree in green, and other vertices in orange. Select an edge with the cheapest cost that connects a green vertex to an orange vertex.



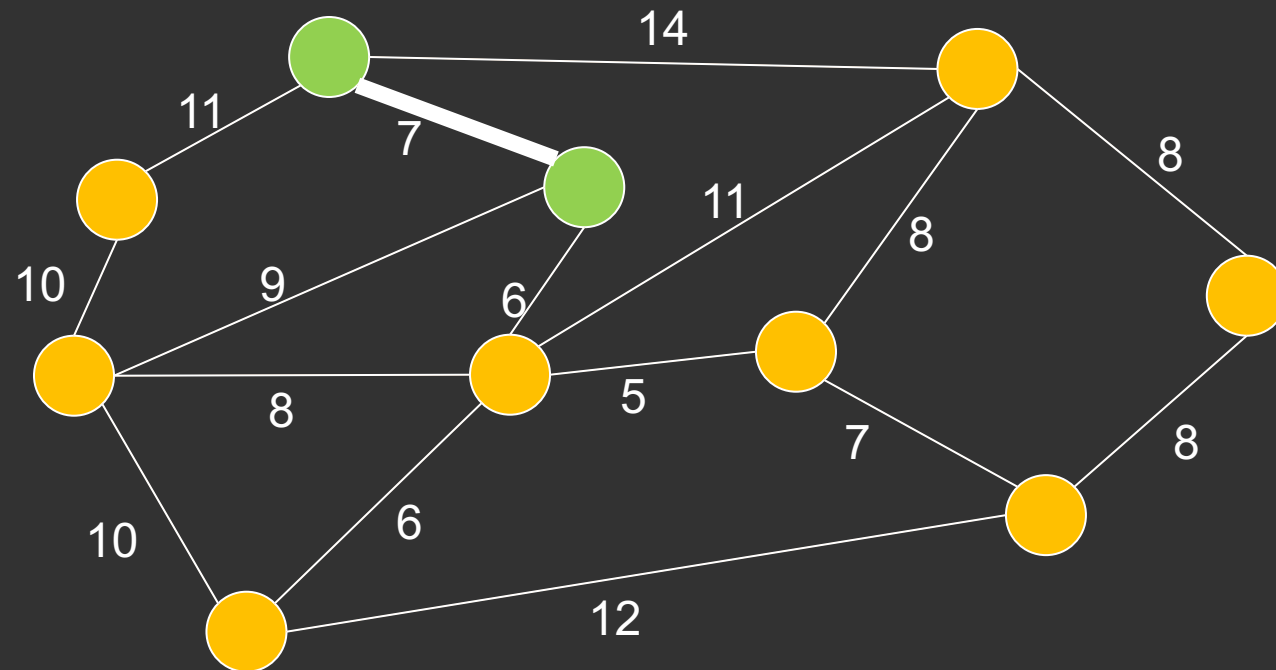
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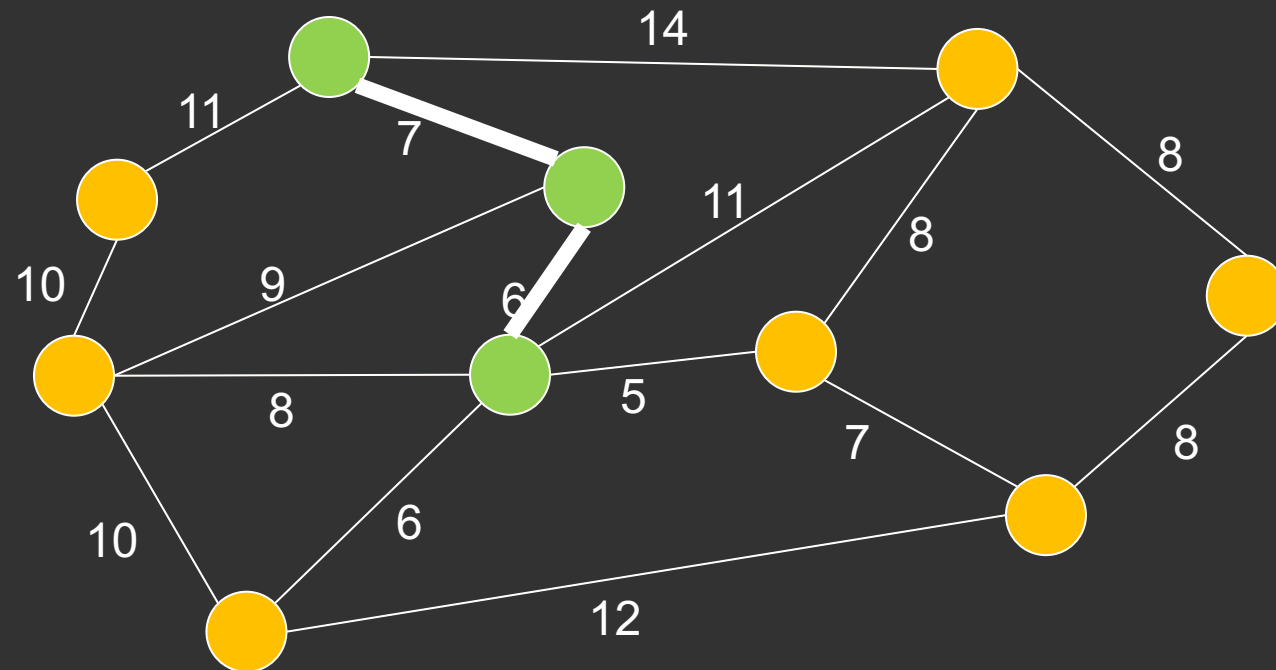
Prim's Algorithm for MCSTs

- Continue this process of adding an edge with the cheapest cost connecting a green vertex to an orange vertex until all vertices are green.



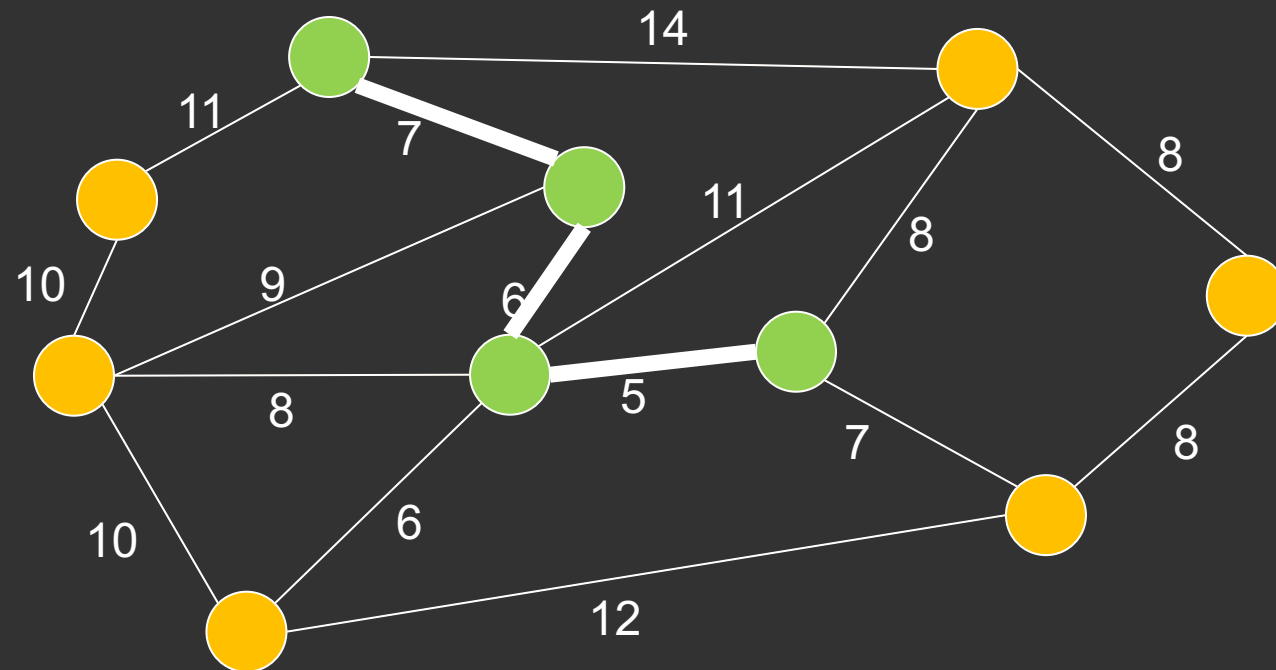
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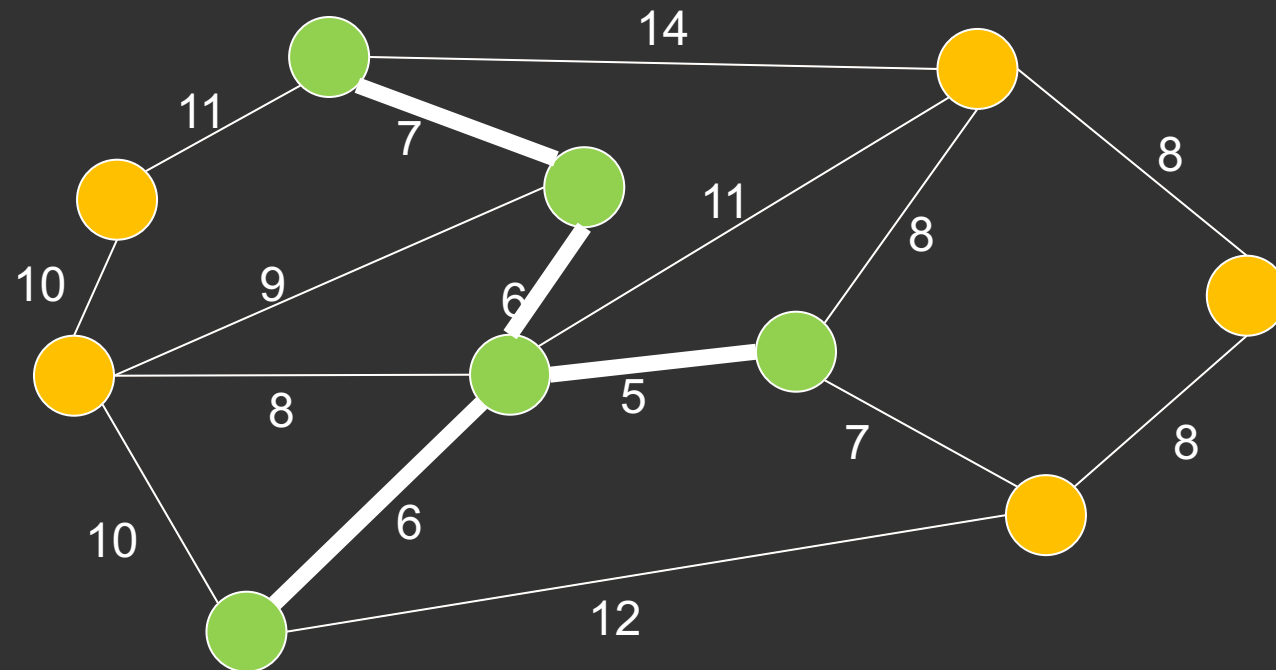
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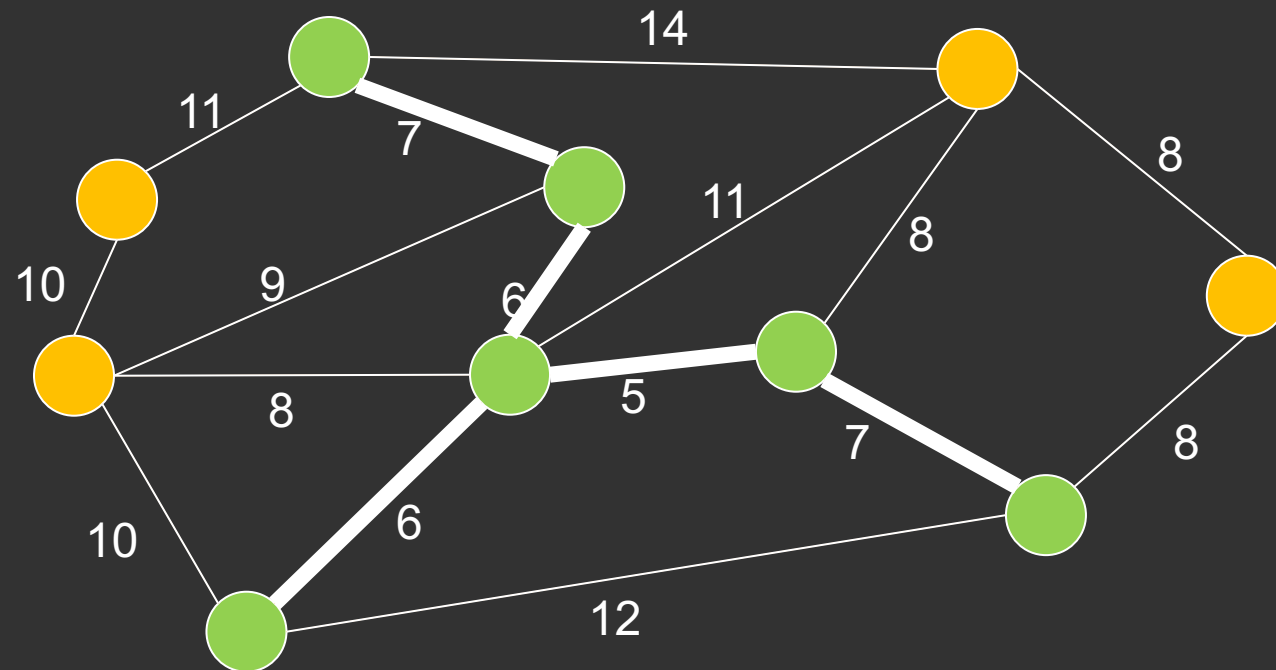
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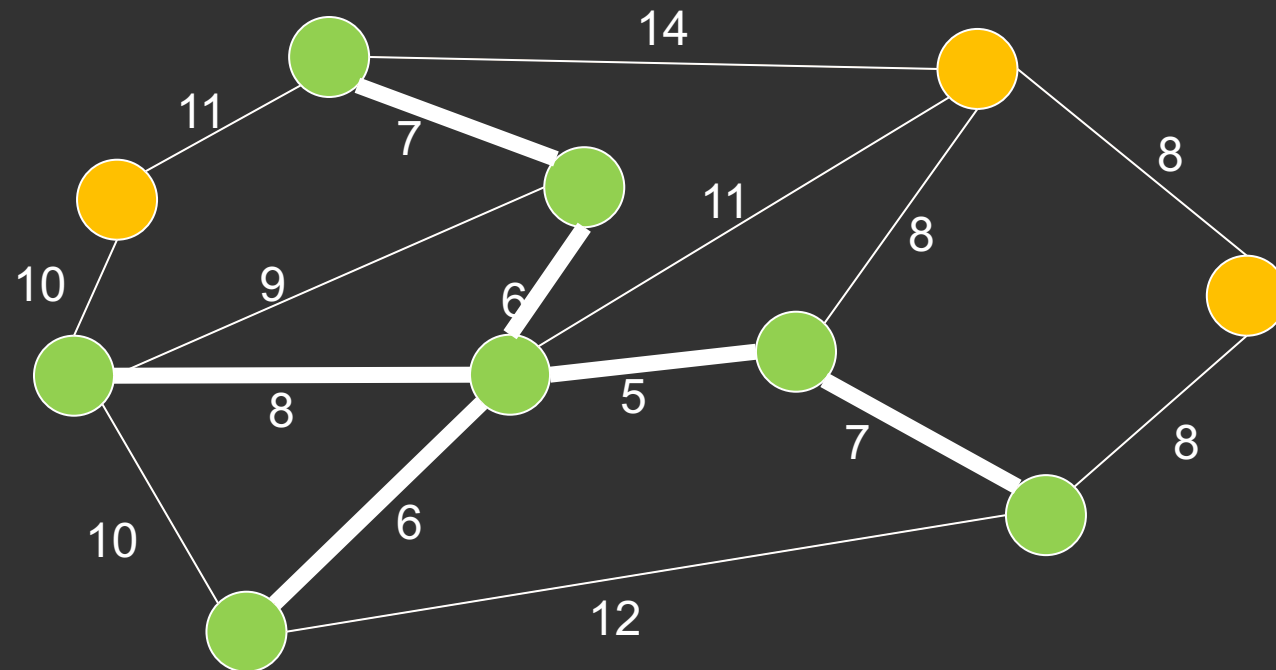
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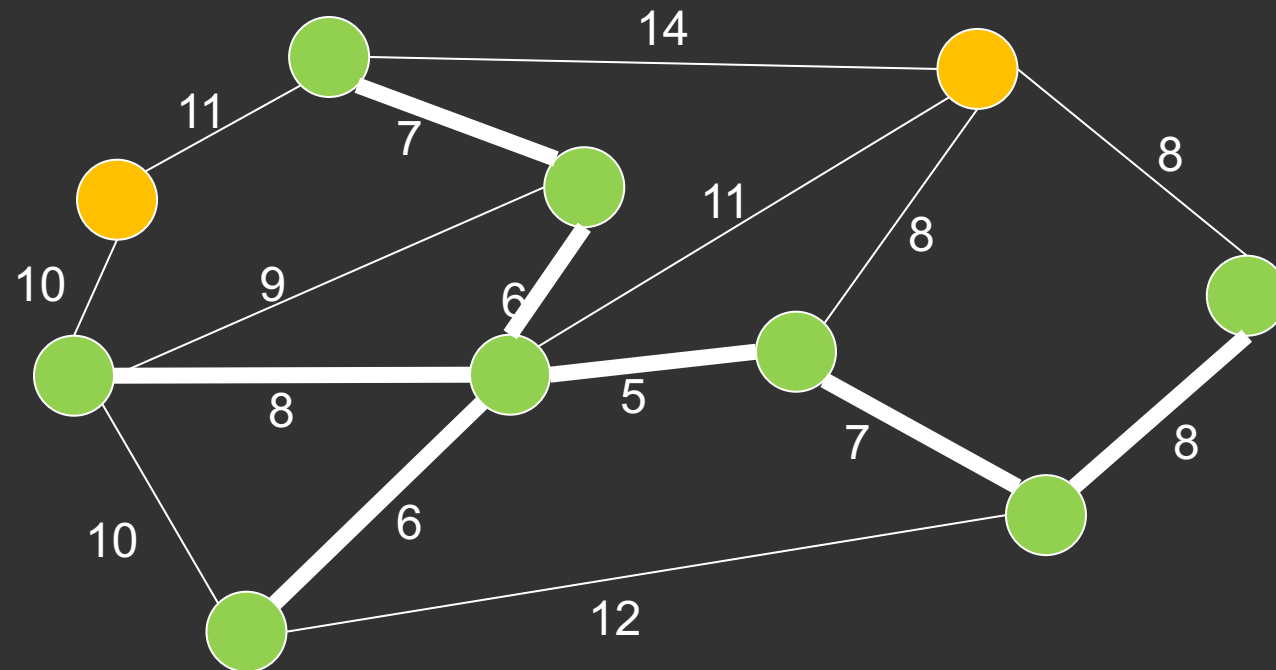
Prim's Algorithm for MCSTs

- What if we have multiple cheapest edges? Ties can be broken arbitrarily. There may be multiple valid minimum cost spanning trees!



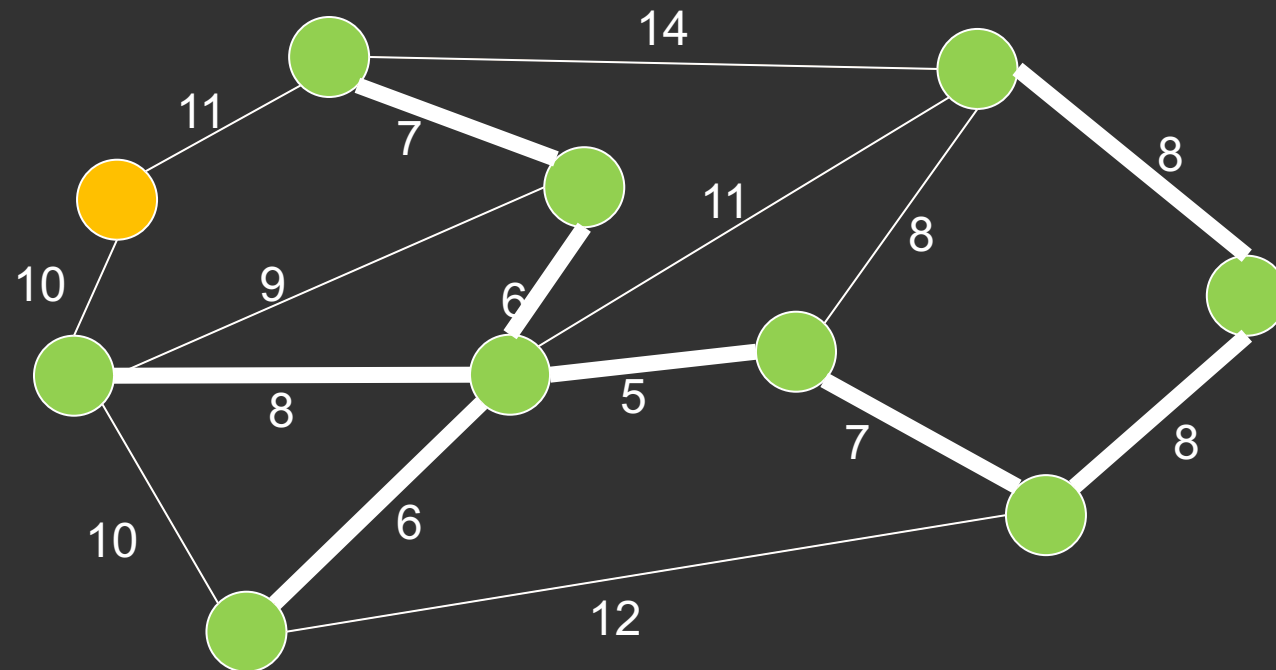
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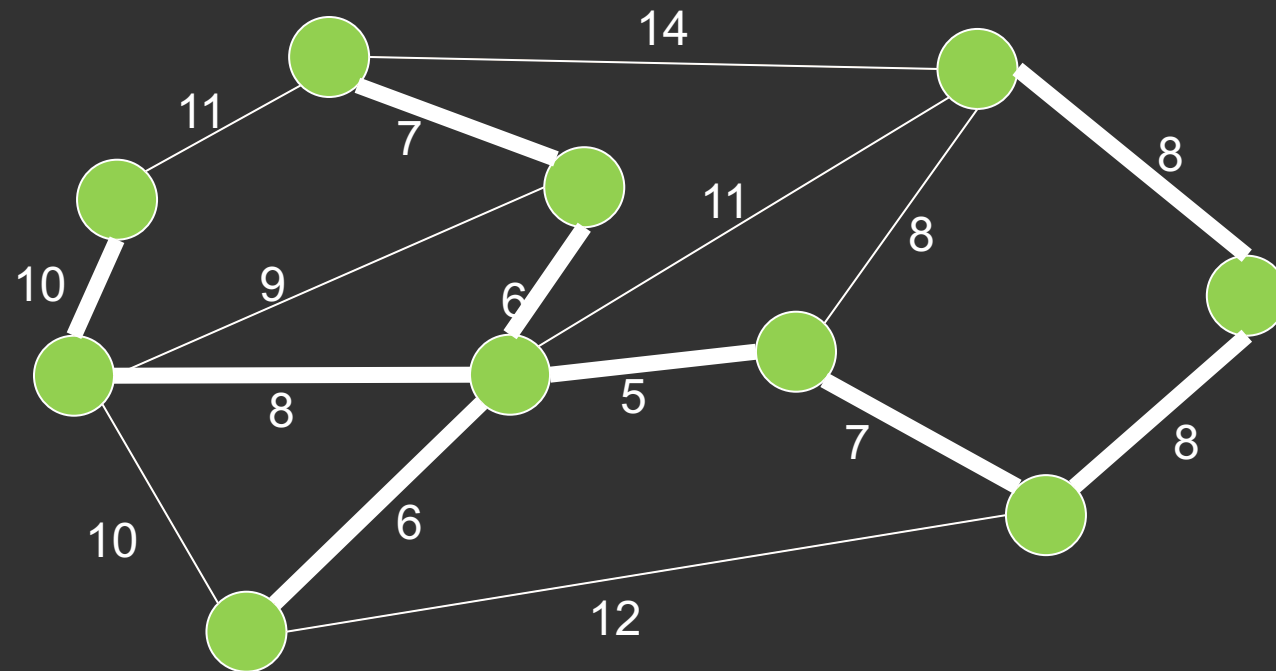
Prim's Algorithm for MCSTs

- What if we have multiple cheapest edges? Ties can be broken arbitrarily. There may be multiple valid minimum cost spanning trees!



Prim's Algorithm for MCSTs

- Once all vertices are green, we have constructed a minimum cost spanning tree.



Prim's Algorithm

- The greedy algorithm we just described is called **Prim's algorithm**
 - It always find a minimum-cost spanning tree for any connected graph (even if the weights are negative)!
- How can we argue that Prim's algorithm is optimal?
- Why is it always a good idea to connect the current network to the cheapest node outside of the network?

The Key to Prim's Algorithm: "Cut Property"

Def: Sets V_1 and V_2 form a *partition* of a set V if

$$V_1 \cup V_2 = V \text{ and } V_1 \cap V_2 = \emptyset$$

- In other words, V_1 and V_2 together contain all of the vertices in V , but no vertex is in both V_1 and V_2 .

Lemma: Let $G=(V,E)$ be a connected graph and let V_1 and V_2 be a partition of V . *Every* MCST of G contains *a* cheapest edge between V_1 and V_2

Proof Sketch: Cut Property

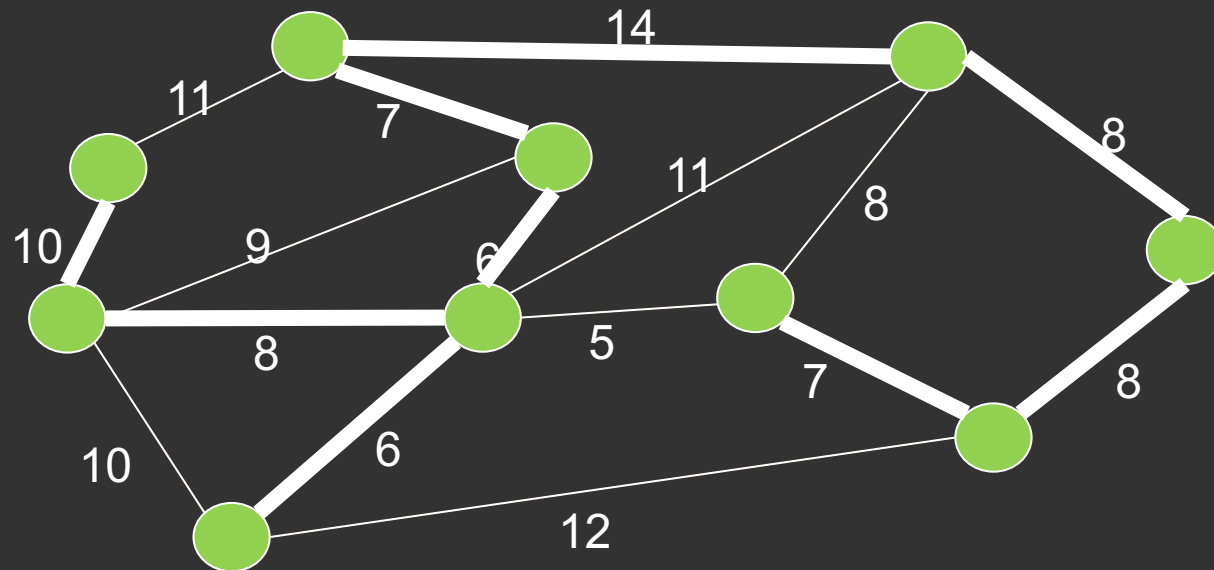
If there's one cheapest edge, it's in the MCST. If there's a tie, one of them is in the MCST

Lemma: Let $G=(V,E)$ be a connected graph and let V_1 and V_2 be a partition of V . Every MCST of G contains a cheapest edge between V_1 and V_2

- Let e be a cheapest edge between V_1 and V_2
- Let T be a MCST of G .
 - If $e \notin T$, then $T \cup \{e\}$ contains a cycle C and e is an edge of C
 - Some other edge e' of C must also be between V_1 and V_2 ; since e is a cheapest edge, so $w(e') = w(e)$
 - (If it weren't, we could replace e with e' and T 's cost would be cheaper, but that's impossible because T was a MCST.)

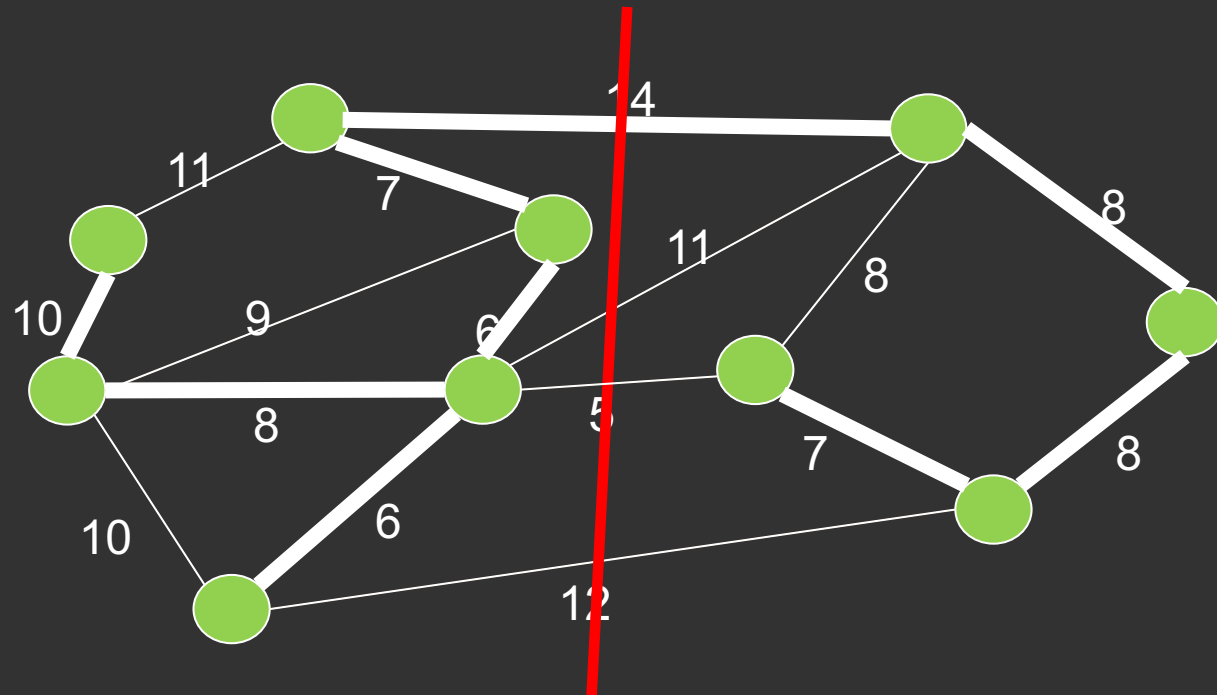
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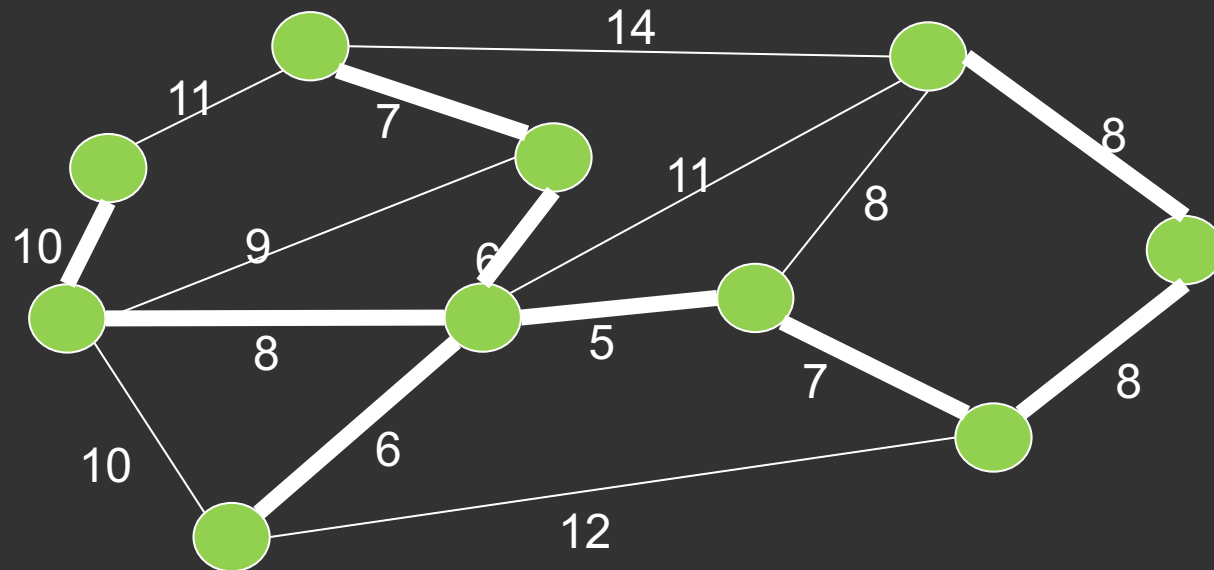
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Using The Cut Property to Prove Prim

We'll assume all edge costs are distinct

(Not necessary but otherwise proof is slightly less elegant)

Let T be a tree produced by the greedy algorithm, and suppose T^* is a MCST for G .

Claim: $T = T^*$

Idea of Proof: Show that every edge added to the tree T by the greedy algorithm is in T^*

Clearly the first edge added to T is in T^*

Why? Use the cut property!

Using The Cut Property to Prove Prim

Now use induction!

- Suppose that, for some $k \geq 1$, the first k edges added to T are in T^* . These form a tree T_k
- Let V_1 be the vertices of T_k and let $V_2 = V - V_1$
- Now, the greedy algorithm will add to T the cheapest edge e between V_1 and V_2
- But any MCST contains the (only!) cheapest edge between V_1 and V_2 , so e is in T^*
- Thus the first $k+1$ edges of T are in T^*

Where we are

- Prim's works!
 - Sometimes greedy choices don't work well, but we proved that they are always optimal in this case.
- How can we implement Prim's?
- First: write pseudocode
 - What methods do we need our data structures to support? Which ones must be fast?
- Then: decide on specifics

Prim's Algorithm

```
let v be a vertex of G;
set  $V_1 \leftarrow \{v\}$ , and  $V_2 \leftarrow V - \{v\}$ 
let  $A \leftarrow \emptyset$  // A will contain ALL edges between  $V_1$  and  $V_2$ 
while ( $|V_1| < |V|$ ) :
    add to A all edges incident to v
    // note: A now may have edges with both ends in  $V_1$ 
    repeat :
        remove cheapest edge e from A
    until e is an edge between  $V_1$  and  $V_2$ 
    add e to MCST
    let v  $\leftarrow$  the vertex of e that is in  $V_2$ 
    move v from  $V_2$  to  $V_1$ 
```

Implementing Prim's Algorithm

- We'll "build" the MCST by marking its edges as "visited"
- We'll "build" V_1 by marking its vertices visited
- **Question:** How should we represent A ?
 - What operations are important to A ?
 - Add all edges that are incident to some vertex
 - Remove a cheapest edge
 - We'll use a priority queue!
- When we remove an edge from A , we must verify it has one end in each of V_1 and V_2

ComparableEdge Class

- Values in a `PriorityQueue` need to implement `Comparable`
- We wrap edges of the PQ in a class called `ComparableEdge`
 - It requires the label used by graph edges to be of a `Comparable` type (e.g., `Integer`)

MCST: The Code

```
PriorityQueue<ComparableEdge<String,Integer>> q =  
    new VectorHeap<ComparableEdge<String,Integer>>();  
  
String v;           // current vertex  
Edge<String,Integer> e; // current edge  
boolean searching;  // still building tree?  
  
g.reset();          // clear visited flags  
  
// select a node from the graph, if any  
Iterator<String> vi = g.iterator();  
if (!vi.hasNext())  
    return; // graph is empty!  
v = vi.next();
```

MCST: The Code

```
do {  
    // Add vertex to MCST and add all outgoing edges  
    // to the priority queue  
  
    g.visit(v); // all  $V_1$  are visited  
  
    for (String neighbor : g.neighbors(v)) {  
        // turn it into outgoing edge  
        e = g.getEdge(v, neighbor);  
        // add the edge to the priority queue  
        q.add(new ComparableEdge<String,Integer>(e));  
    }  
  
    ...  
}
```

MCST: The Code

```
...
searching = true; // looking for an edge btwn  $V_1$  &  $V_2$ 
while (searching && !q.isEmpty()) {
    // grab next shortest edge
    e = q.remove();
    // Is e between  $V_1$  and  $V_2$ ?
    v = e.there();
    if (g.isVisited(v)) v = e.here();
    if (!g.isVisited(v)) {
        searching = false;
        g.visitEdge(g.getEdge(e.here(),
                               e.there()));
    }
}
} while (!searching);
```


Prim : Space Complexity

- Graph: $O(|V| + |E|)$
 - Each vertex and edge uses a constant amount of space
- Priority Queue $O(|E|)$
 - Each edge takes up constant amount of space
- Every other object (including the neighbor iterator) uses a constant amount of space
- Result: $O(|V| + |E|)$
 - Optimal in Big-O sense!

Prim : Time Complexity

Assume Map ops are $O(1)$ time

For each iteration of do ... while loop

- Add neighbors to queue: $O(\text{deg}(v) \log |E|)$
 - Iterator operations are $O(1)$ [Why?]
 - Adding an edge to the queue is $O(\log |E|)$
- Find next edge: $O(\# \text{ edges checked} * \log |E|)$
 - Removing an edge from queue is $O(\log |E|)$ time
 - All other operations are $O(1)$ time

Prim : Time Complexity

Over *all* iterations of do ... while loop

Step I: Add neighbors to queue:

- For each vertex, it's $O(\text{deg}(v) \log |E|)$ time
- Adding over all vertices gives

$$\sum_{v \in V} \text{deg}(v) \log |E| = \log |E| \sum_{v \in V} \text{deg}(v) = \log |E| * 2 |E|$$

Prim : Time Complexity

Over *all* iterations of do ... while loop

Step 2: Find next edge: $O(\# \text{ edges checked} * \log |E|)$

- Each edge is checked at most once
- Adding over all edges gives $O(|E| \log |E|)$ again

Thus, overall time complexity (worst case) of Prim's Algorithm is $O(|E| \log |E|)$

Summary

- Prim's algorithm finds a MCST for a single connected component of any graph $G=(V,E)$
- It is a greedy algorithm, but
- it finds a globally optimal solution!
- Careful analysis of the required operations helps us choose the best data structures to maximize performance.