

CS 2009

Design and Analysis of Algorithms

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Week 10:

Minimum Spanning Trees

Examples

- Greedy algorithms! Three examples:
 - Activity selection (greedy choice: pick activity with earliest finish time)
 - Coin Change (greedy choice: take the largest possible bill or coin that does not overshoot)
 - Fractional Knapsack (greedy choice: select item with highest value/weight value until bag is full)

THE GREEDY PARADIGM

**Commit to choices one-at-a-time,
never look back,
and hope for the best.**

Greedy doesn't always work.

WHAT WE'LL COVER TODAY

- Applications of the greedy algorithm design paradigm to **Minimum Spanning Trees**
 - Prim's algorithm
 - Kruskal's algorithm

MINIMUM SPANNING TREES

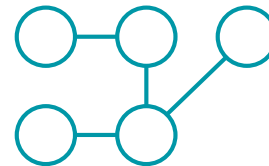
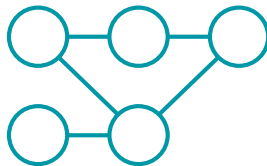
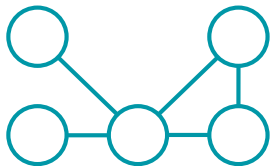
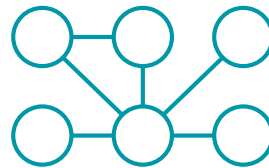
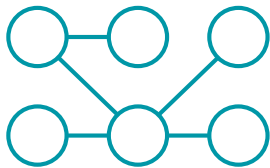
What are minimum spanning trees (MSTs)?

TREES IN GRAPHS

Let's go over some terminology that we'll be using today.

A tree is an undirected, *acyclic*, connected graph.

Which of these graphs are trees?

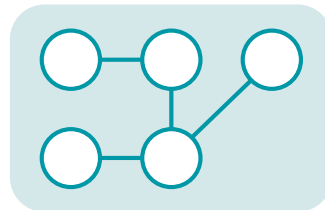
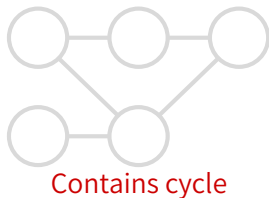
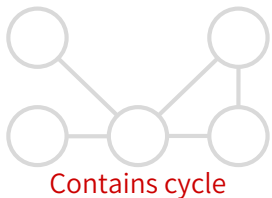
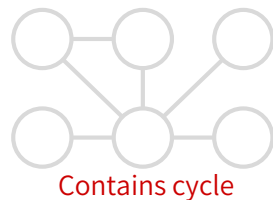
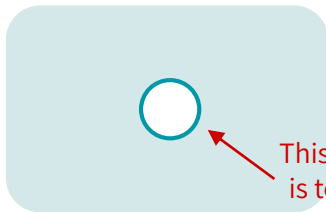
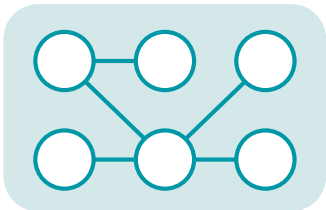


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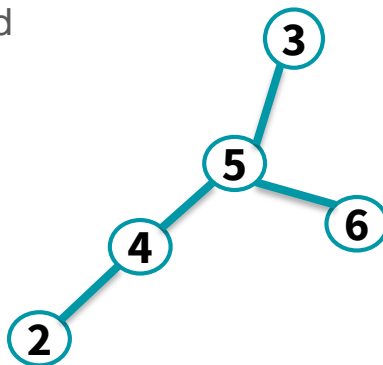
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TREES IN UNIDIRECTED GRAPHS?

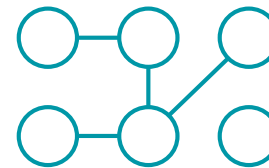
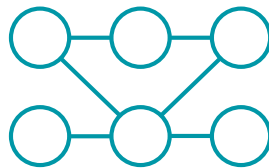
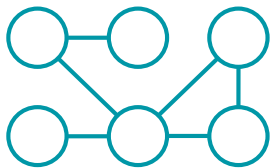
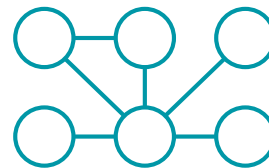
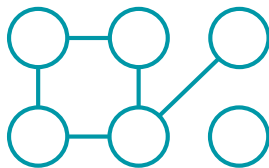
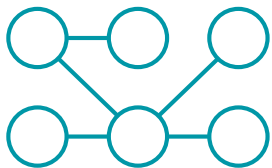
- However, in undirected graphs, there is another definition of trees
- Tree
 - A undirected graph (V, E) , where E is the set of undirected edges
 - All vertices are connected
 - $|E|=|V|-1$



SPANNING TREES

A spanning tree is a tree that connects all of the vertices in the graph

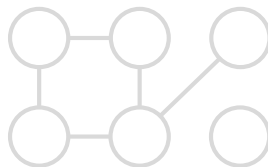
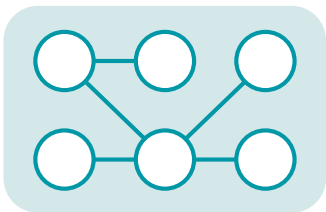
Which of these are spanning trees?



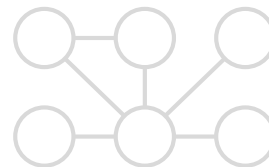
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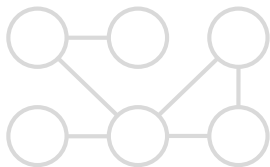
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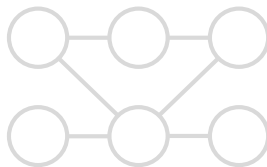
Doesn't connect all vertices



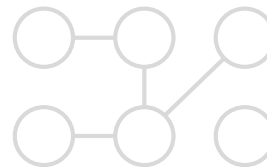
Not a tree



Not a tree



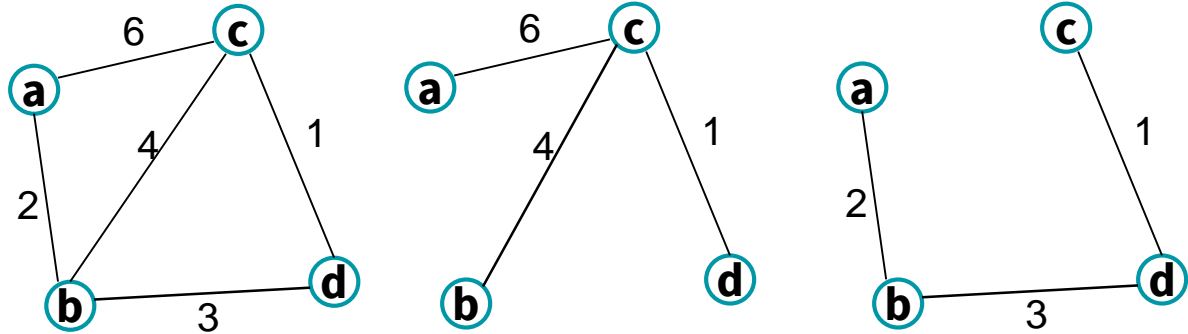
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Doesn't connect all vertices

Examples of MST

Example:



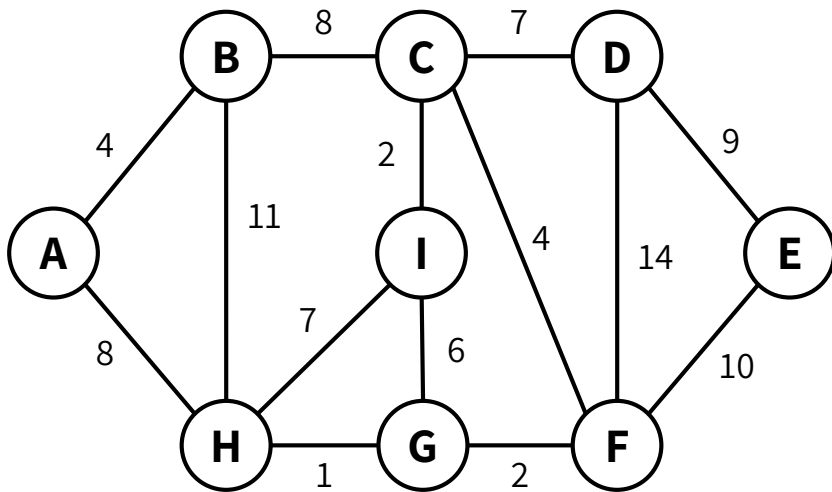
MINIMUM SPANNING TREES (MSTs)

we're going to work with **undirected, weighted, connected graphs**.

The **cost of a spanning tree** is the **sum of the weights on the edges**.

An **MST** of a graph is a spanning tree of the graph with minimum cost.

Note: A graph may have multiple spanning trees. It may also have multiple MSTs (if 2 different spanning trees have the same exact cost)



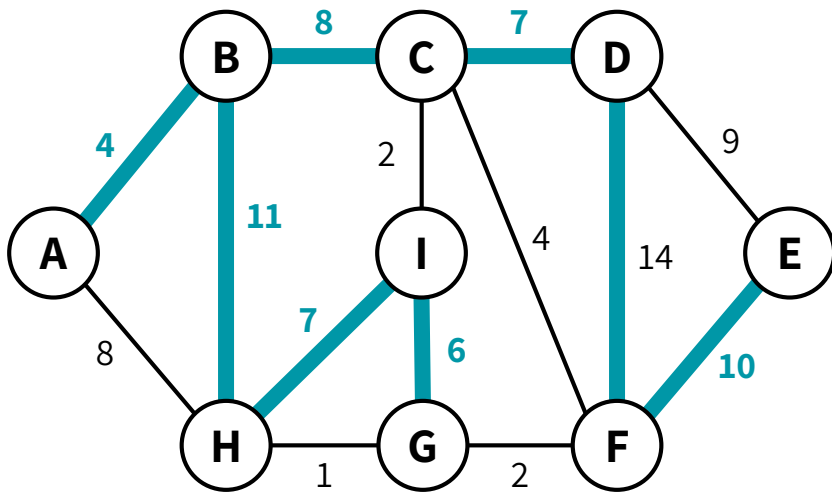
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This spanning tree has a cost of **67**.

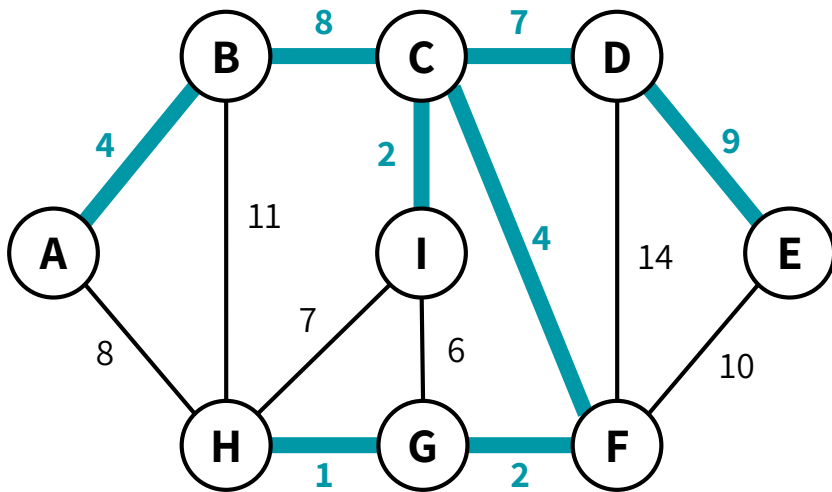
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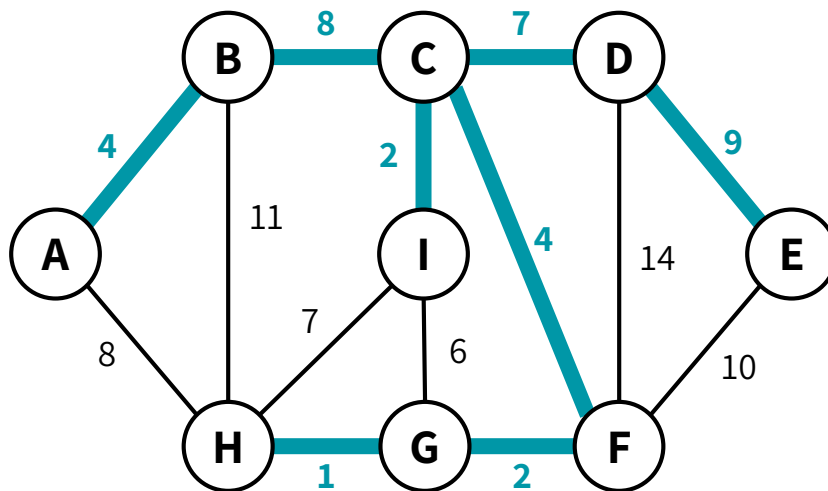
This spanning tree has a cost of **37**.

This is an MST of this graph, since there is no other spanning tree with smaller cost.

MINIMUM SPANNING TREES (MSTs)

The task for today:

Given an undirected, weighted, and connected graph G , find the minimum spanning tree (as a subset of the G 's edges)



We would return this MST.
Sometimes, there may be more than one MST as well, so return any MST of G .

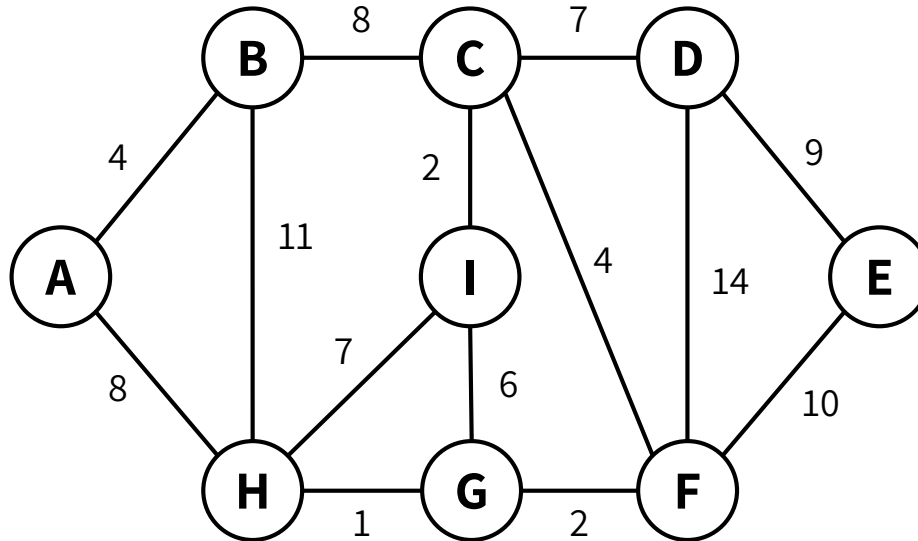
PRIM'S ALGORITHM

Greedyly add the closest vertex!

PRIM'S ALGORITHM: THE IDEA

Greedy choice:

Grow a single tree, & greedily add the shortest edge that could grow our tree

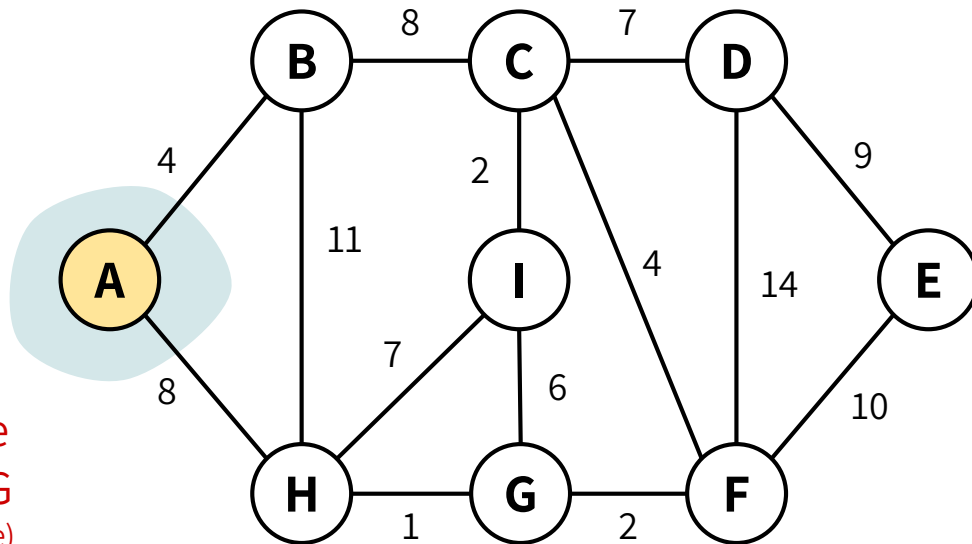


PRIM'S ALGORITHM: THE IDEA

Greedy choice:

Grow a single tree, & greedily add the shortest edge that could grow our tree

First, we can
initialize our tree
to contain a single
arbitrary node in G
(doesn't matter which node)

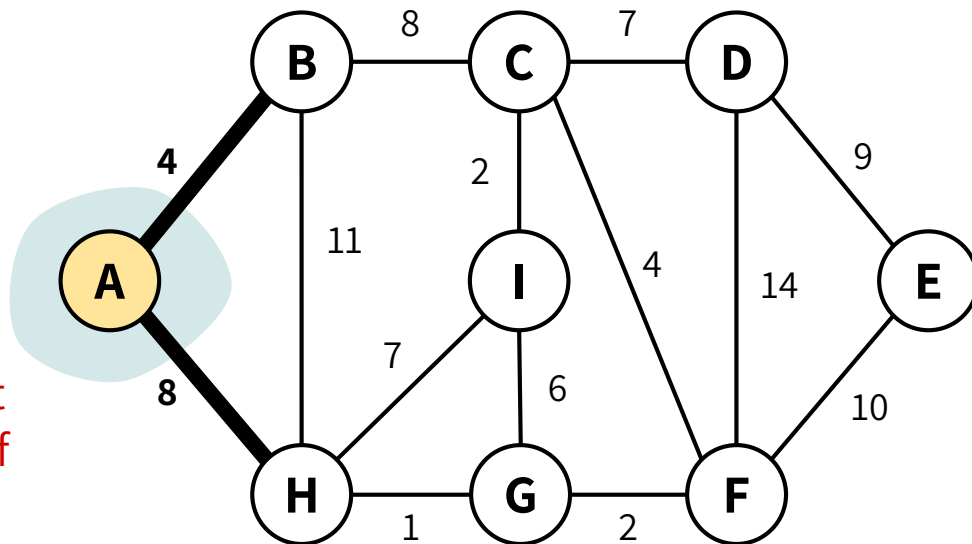


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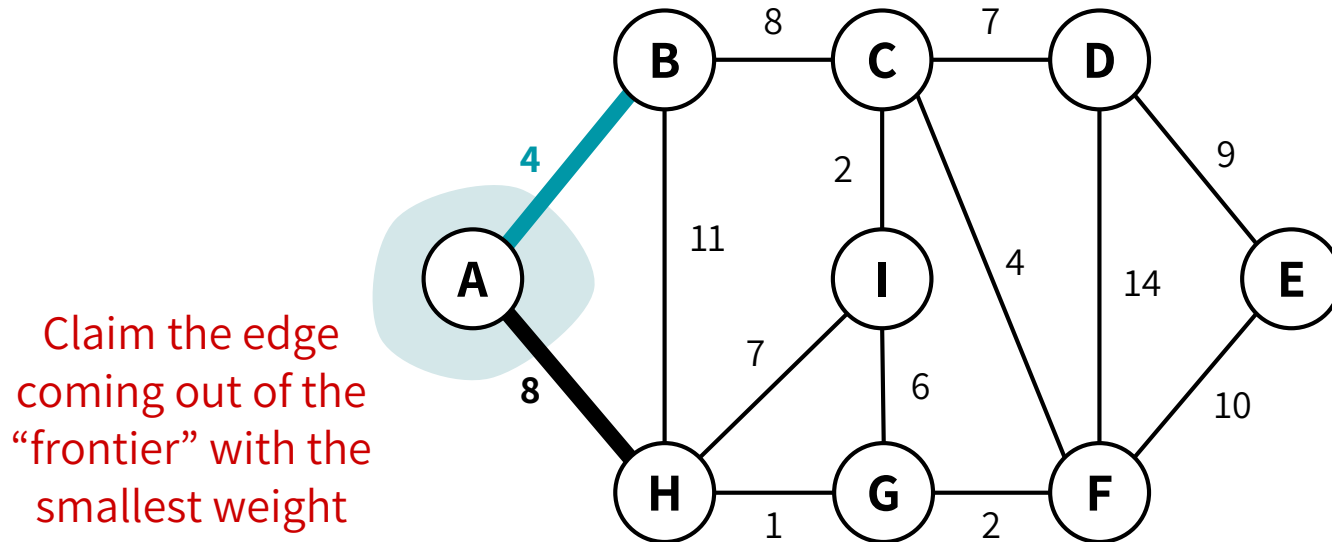
Consider the edges coming out of the “frontier” of our growing tree.



PRIM'S ALGORITHM: THE IDEA

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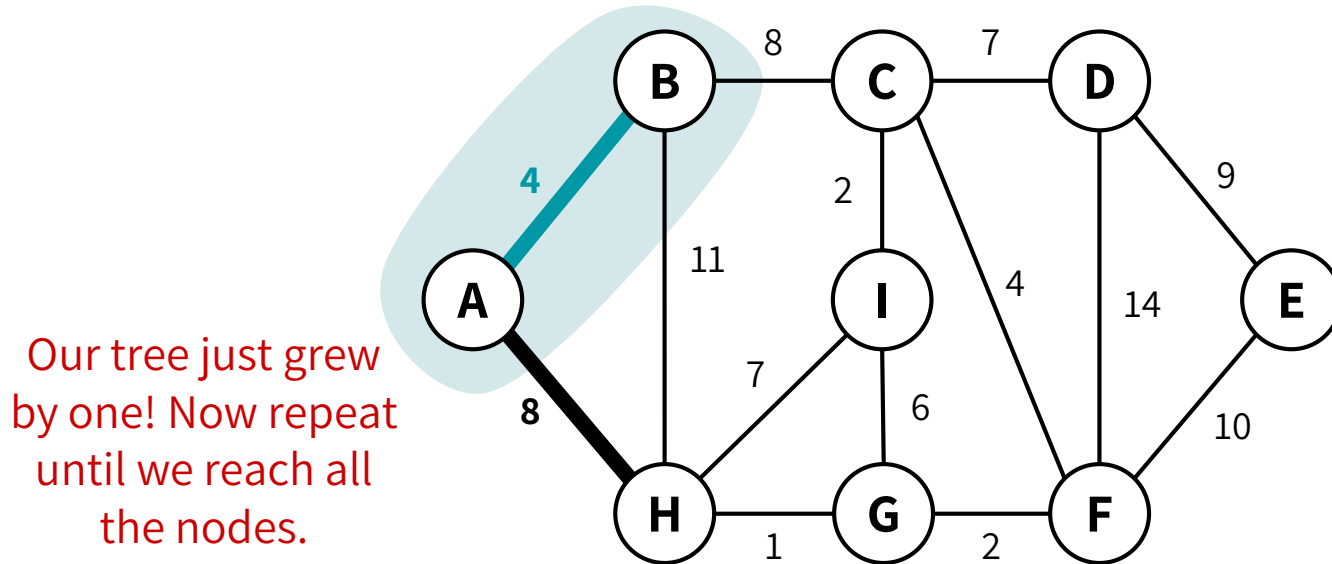
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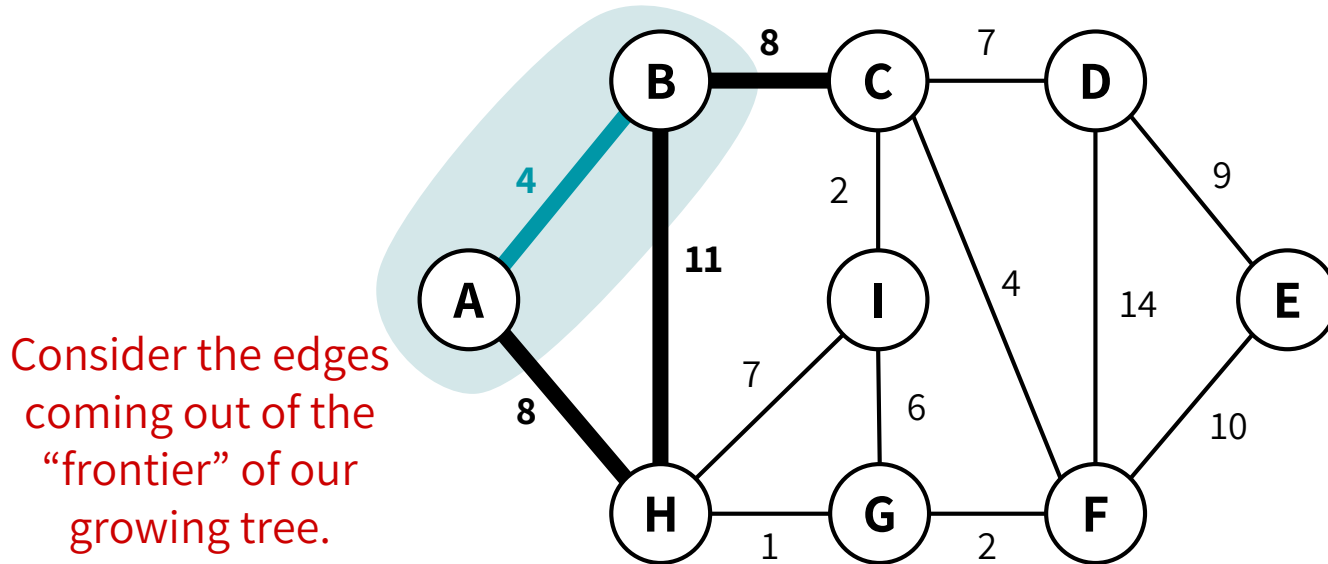
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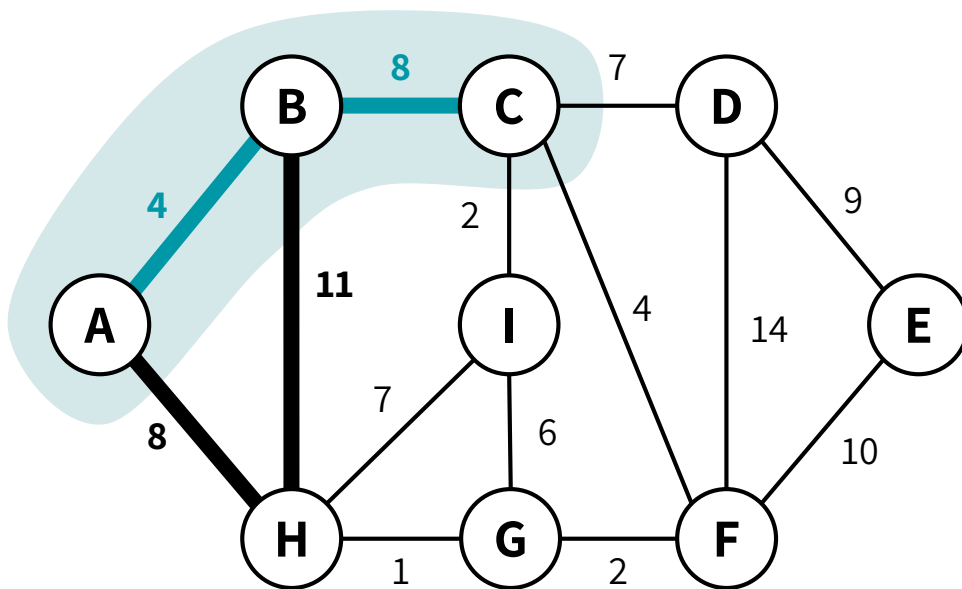


PRIM'S ALGORITHM: THE IDEA

Greedy choice:

Grow a single tree, & greedily add the shortest edge that could grow our tree

Claim the edge coming out of the “frontier” with the smallest weight (if there’s a tie, choose any)

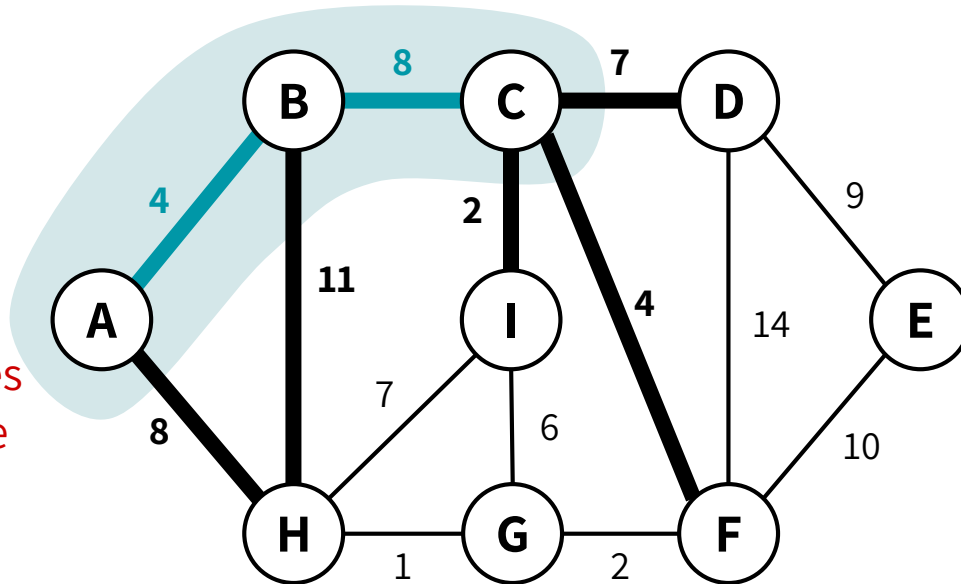


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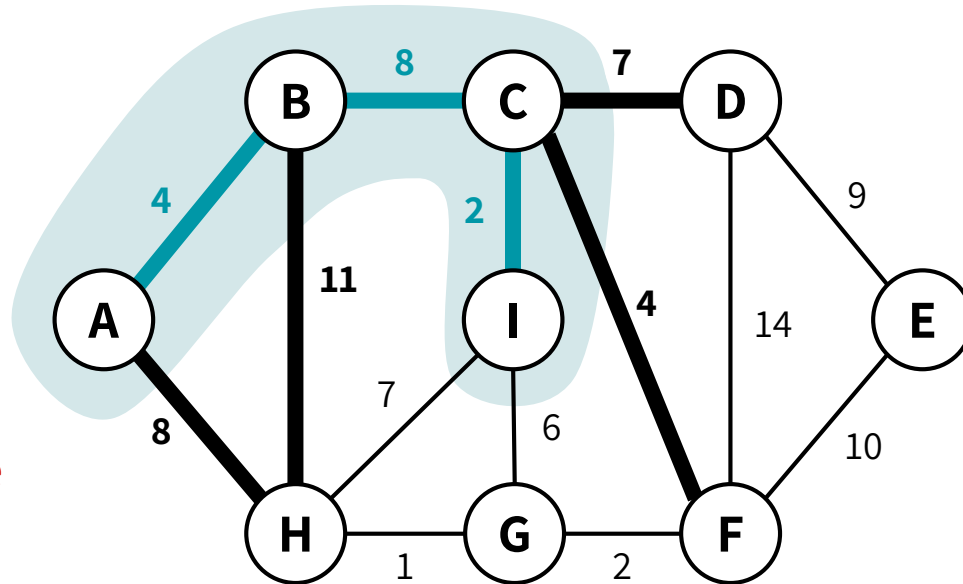


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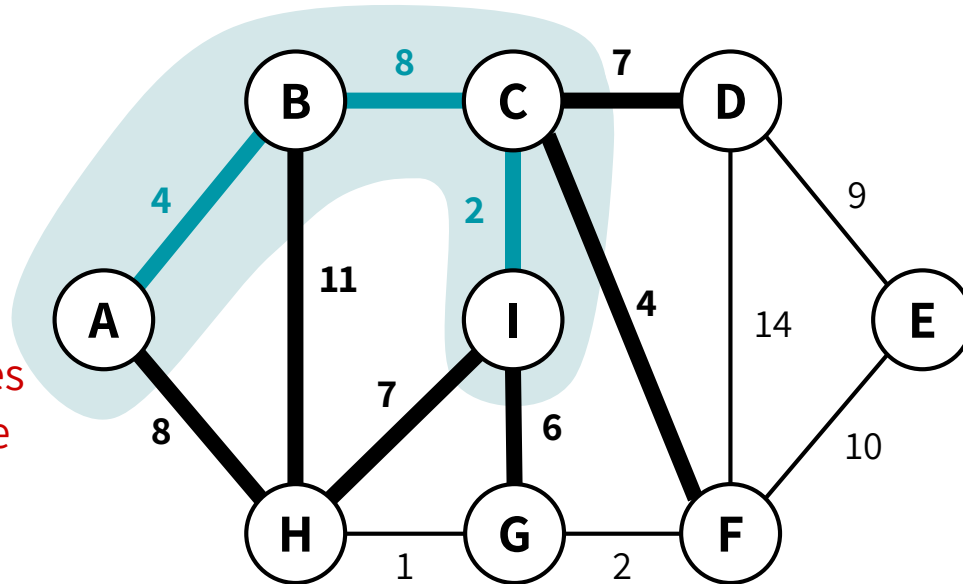


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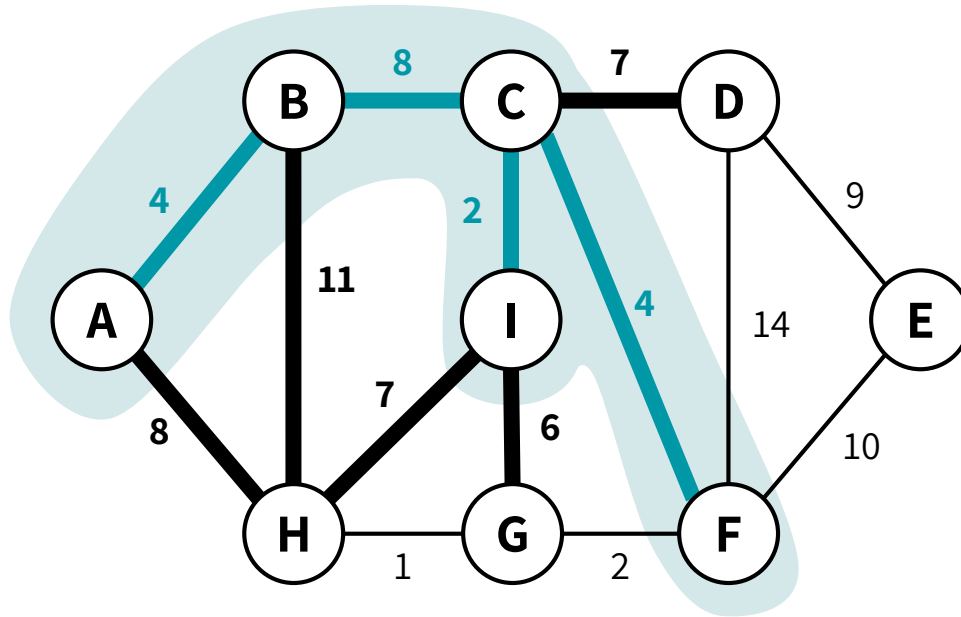


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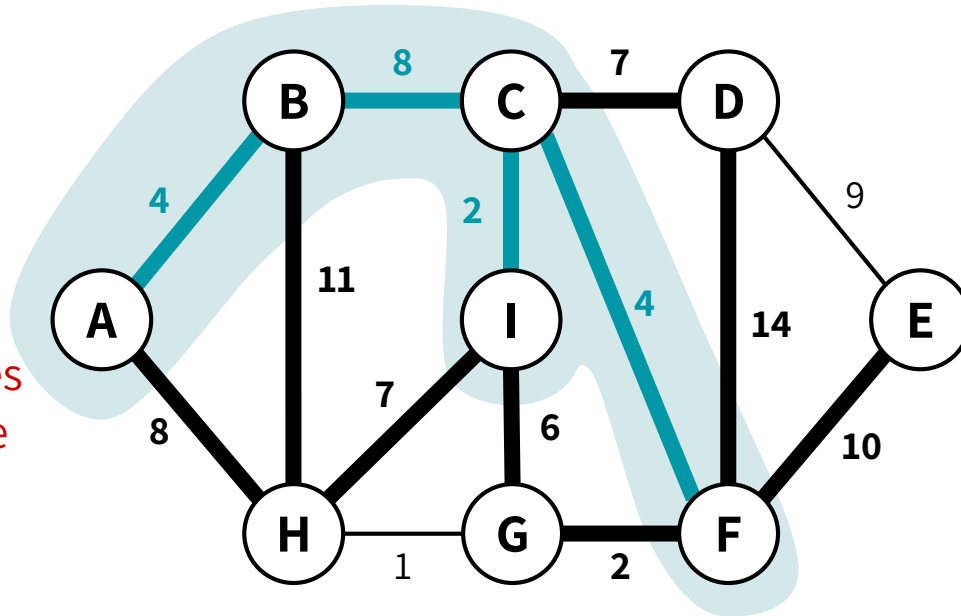


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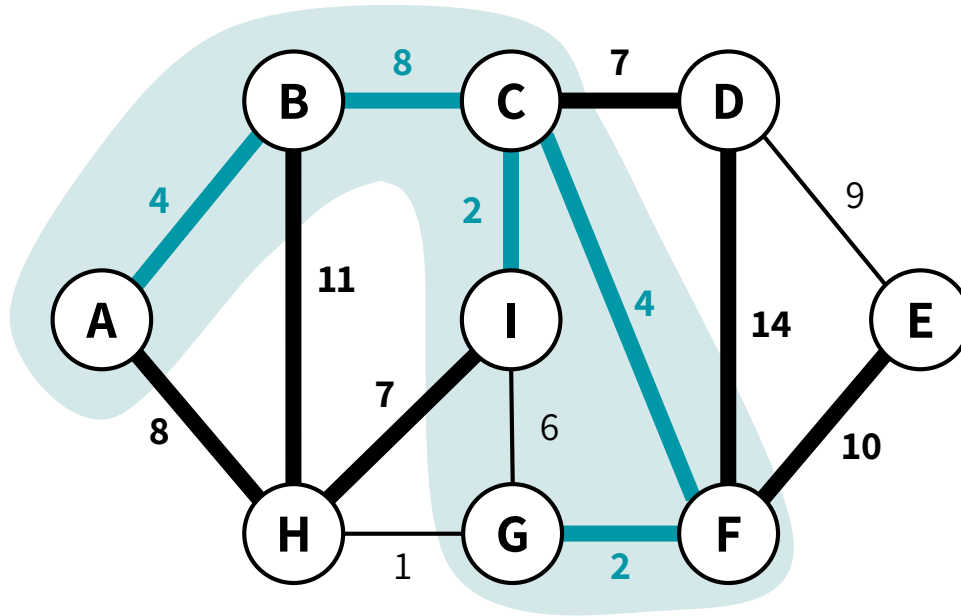


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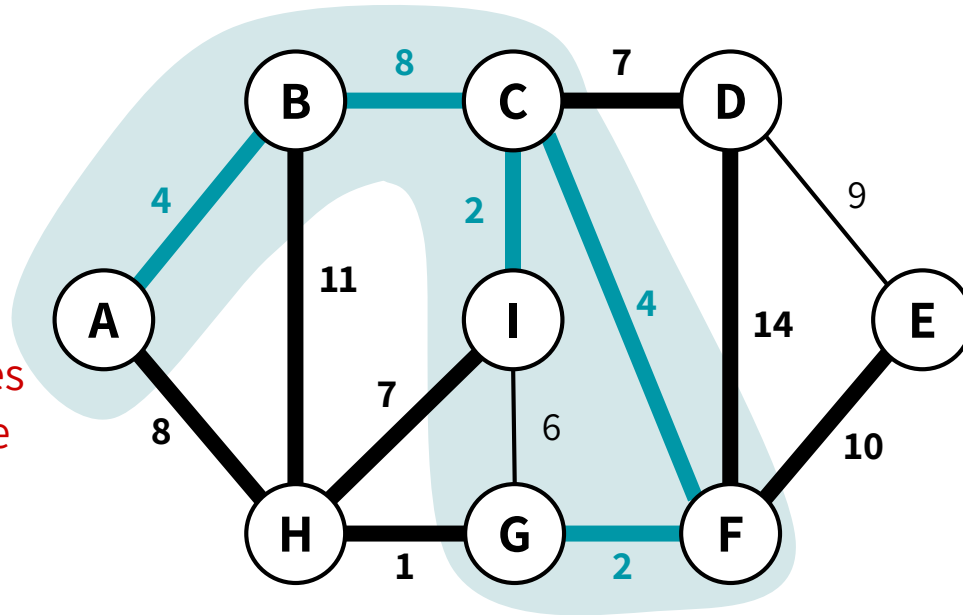


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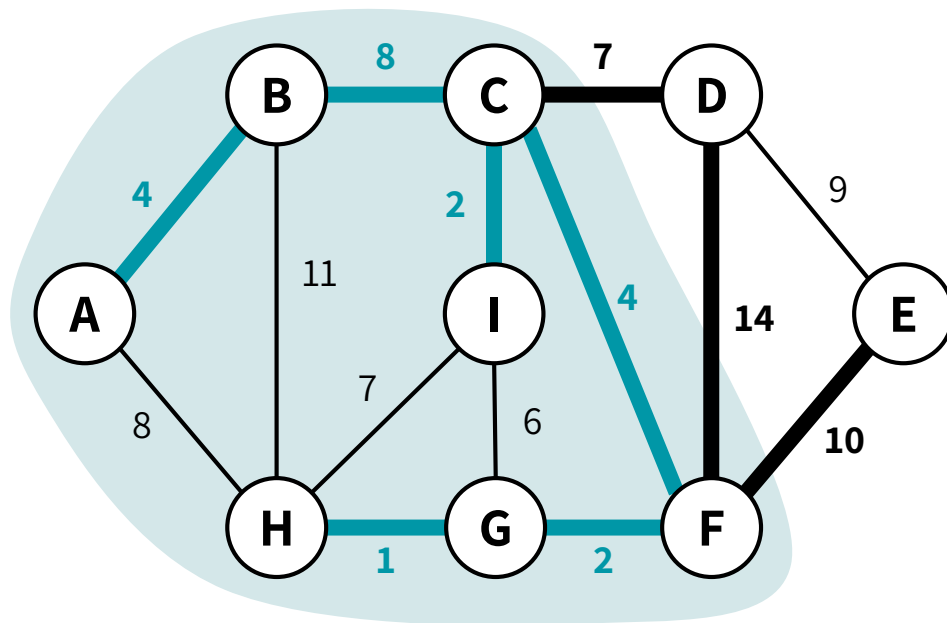


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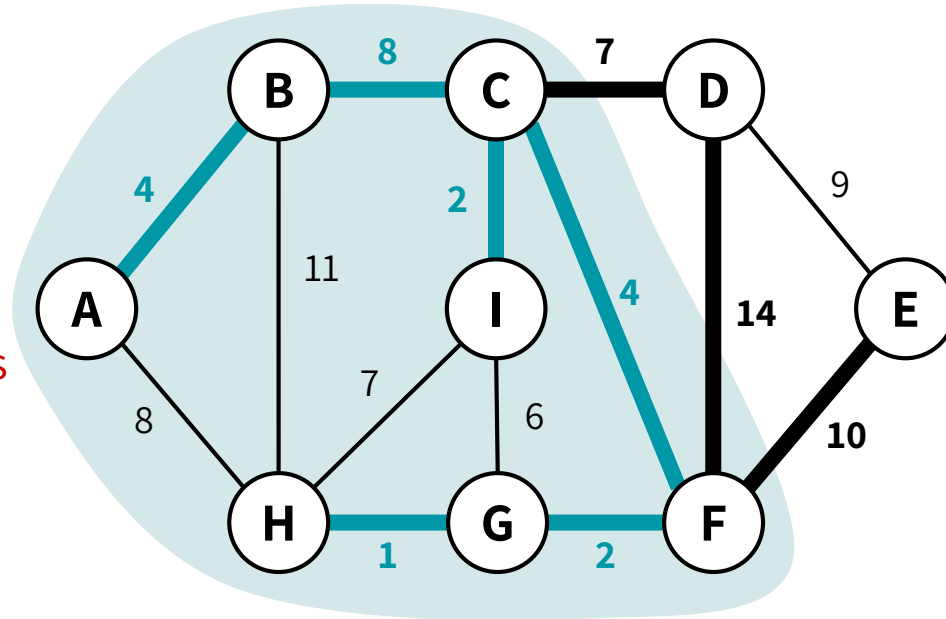


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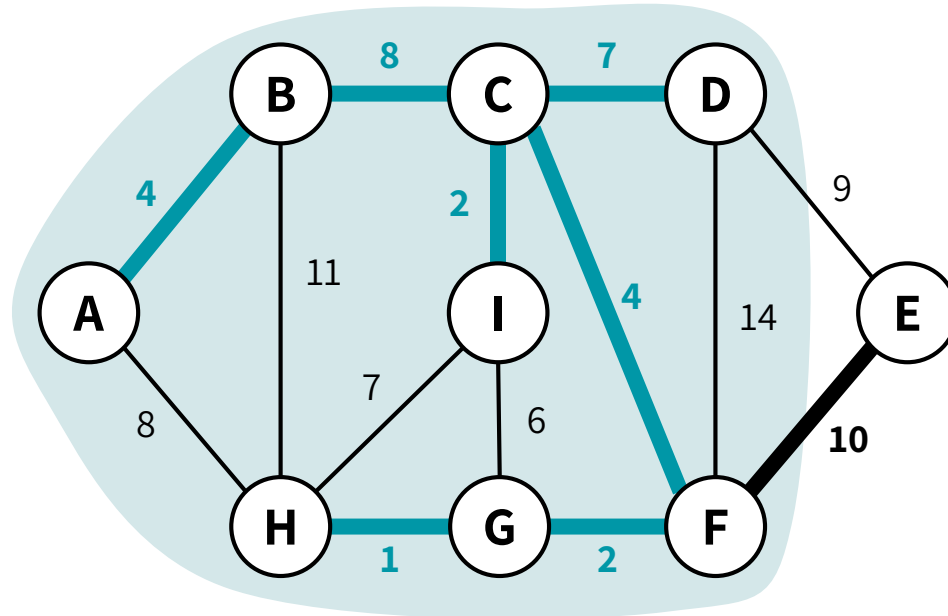


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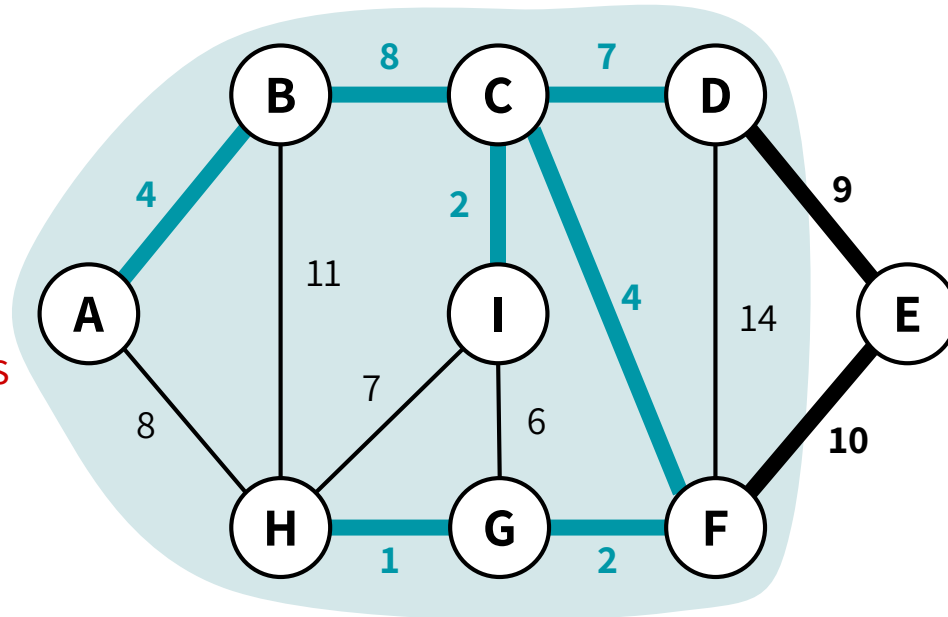


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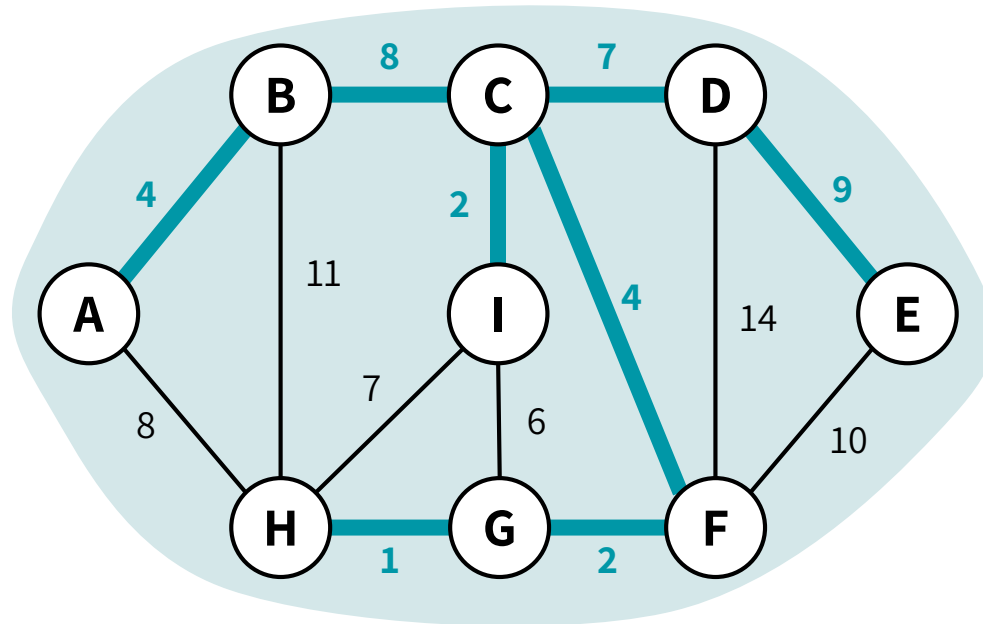


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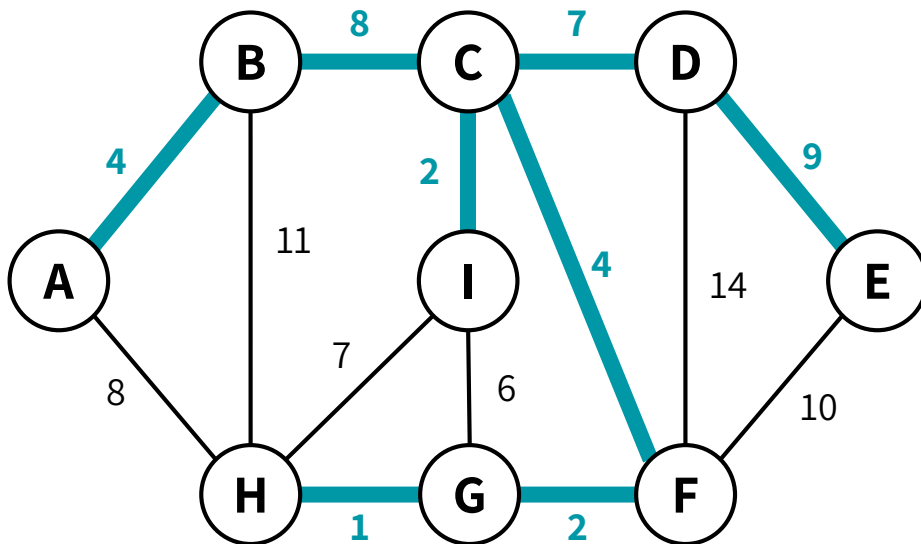
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PRIM'S ALGORITHM: THE IDEA

Greedy choice:

Grow a single tree, & greedily add the shortest edge that could grow our tree



And we're done!
This is our MST.
(with weight 37)

PRIM'S ALGORITHM: SLOW VERSION

NAIVE-PRIM($G = (V, E)$, s):

MST = {}

visited = {s}

while len(visited) < n:

 find the lightest edge (x, v) in E s.t.

- x in visited
- v not in visited

 MST.add((x, v))

 visited.add(v)

return MST

If we manually find the lightest edge each iteration, it could be $O(E)$ time per iteration..

(Naive) Runtime: $O(V \cdot E)$

(We'll speed this up by using smart data structures...)

PRIM'S ALGORITHM: SLOW VERSION

NAIVE-PRIM($G = (V, E)$, s):

$MST \leftarrow \emptyset$

How should we actually implement this?

Each vertex that's not yet reached by the growing tree keeps track of:

- 1) the **distance** from itself to the growing spanning tree using *one edge*
- 2) **how to get to there** (the closest neighbor that's reached by the tree already)

return MST

(Naive) Runtime: $O(V \cdot E)$

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HOW DO WE IMPLEMENT THIS?

Each vertex that's not yet reached by the growing tree keeps track of:

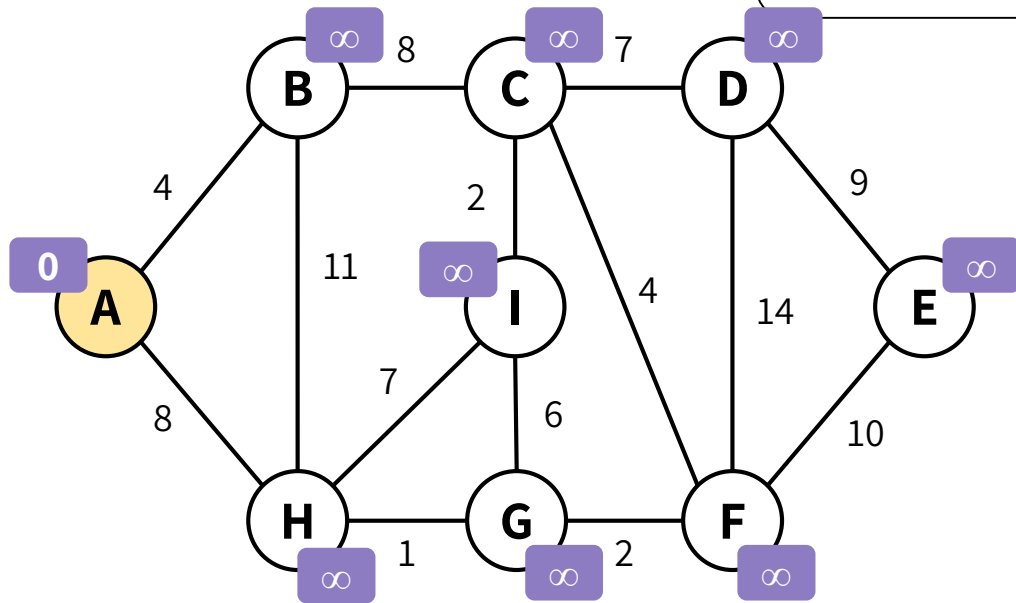
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PRIM($G = (V, E), s$):

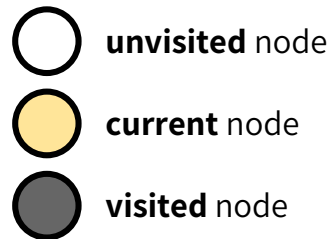
MST = {}

visited = {s}

for all v besides s: $d[v] = \infty$ and $k[v] = \text{NULL}$



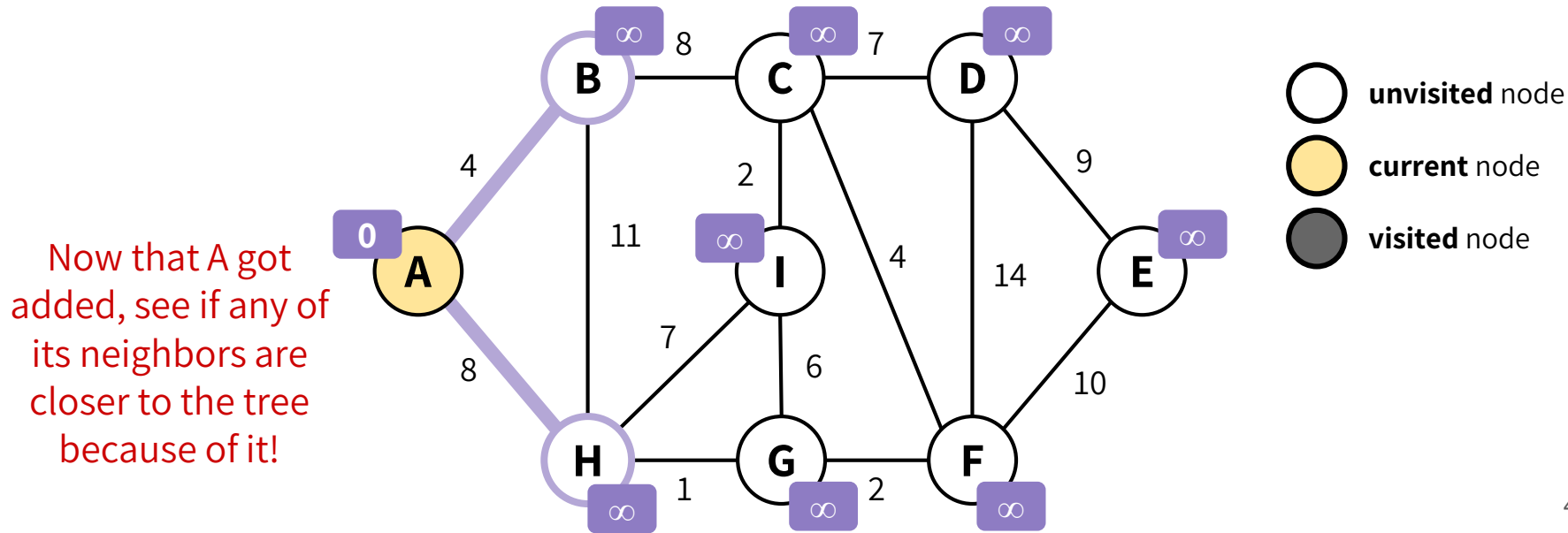
A is part of the growing tree first



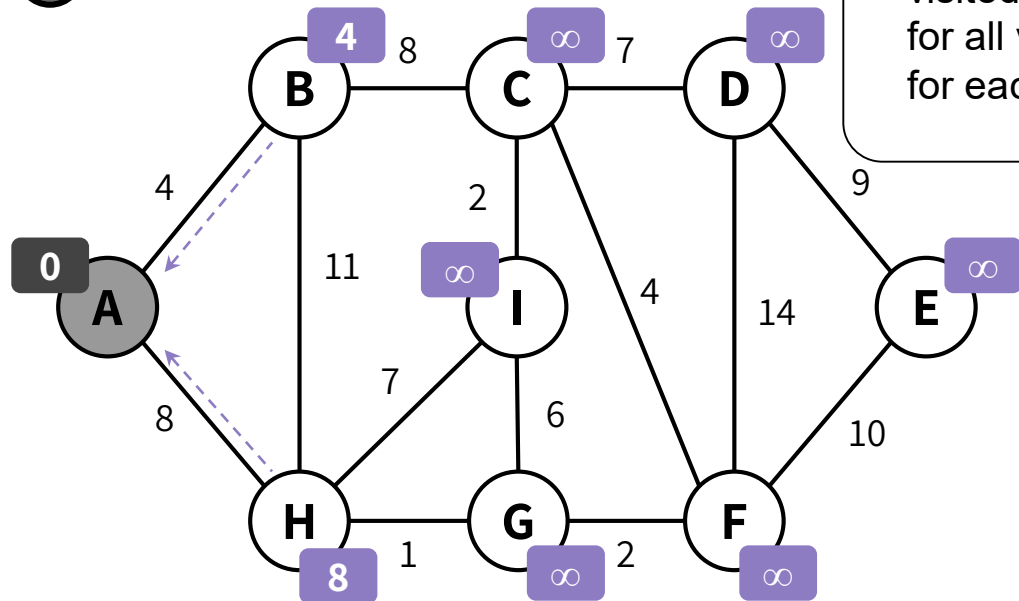
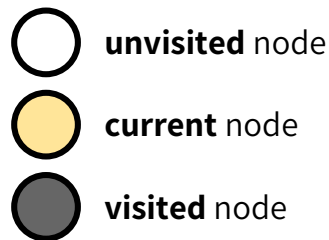
HOW DO WE IMPLEMENT THIS?

Each vertex that's not yet reached by the growing tree keeps track of:

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- 2) **how to get to there** (the closest neighbor that's reached by the tree already)



HOW DO WE IMPLEMENT THIS?



PRIM($G = (V, E), s$):

MST = {}

visited = {s}

for all v besides s: $d[v] = \infty$ and $k[v] = \text{NULL}$

for each neighbor v of s: $d[v] = w(s, v)$ and $k[v] = s$

Update their estimates, and now A is officially done.

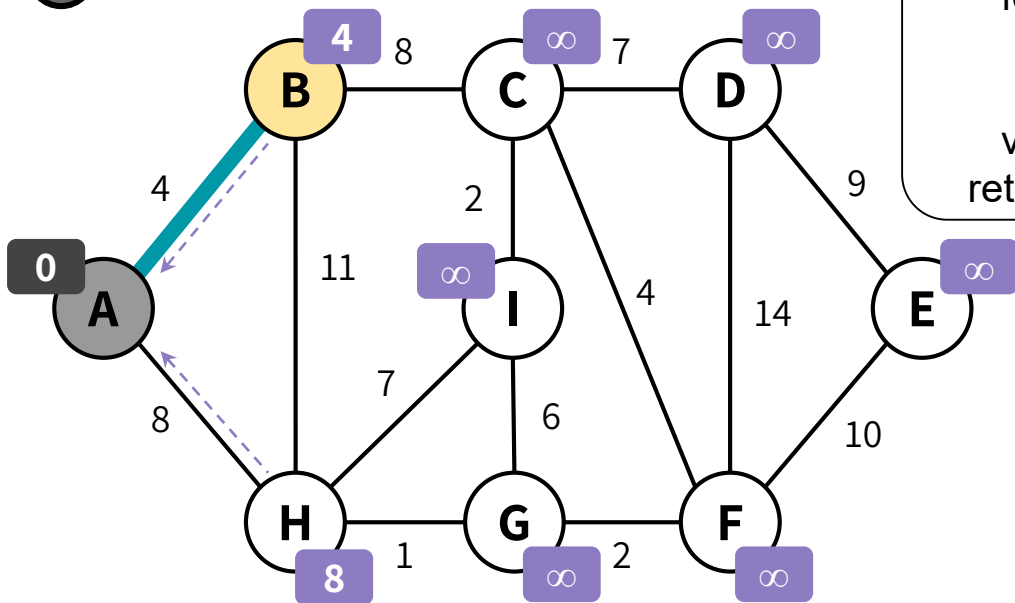
Time to choose the lightest edge on the frontier (i.e. the edge whose endpoint has the lowest distance stored)

HOW DO WE IMPLEMENT THIS?

○ **unvisited** node

● **current** node

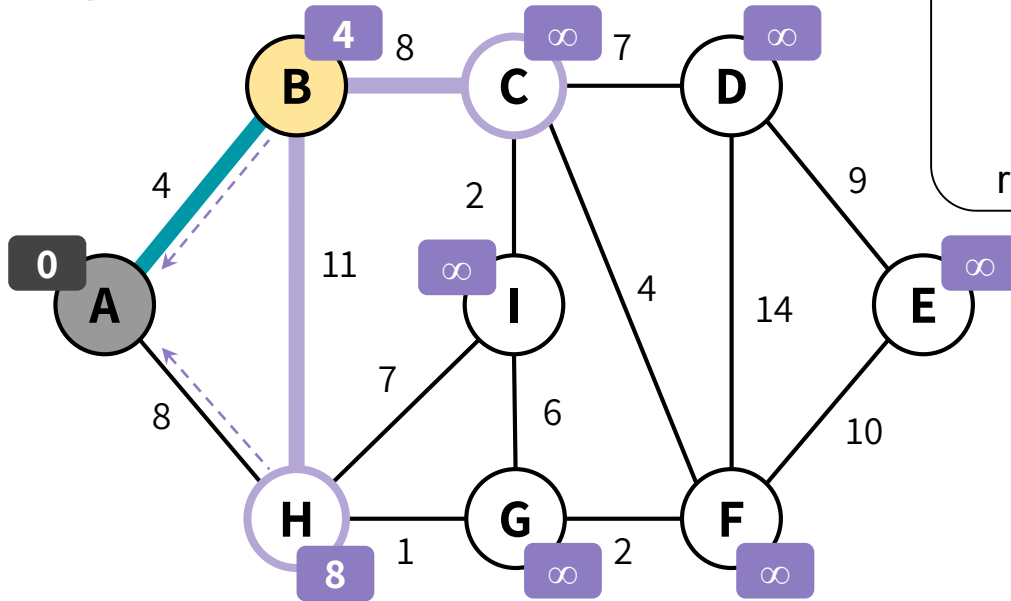
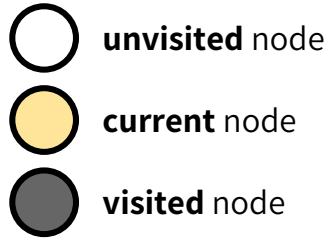
● **visited** node



```
while len(visited) < n:  
    x = unvisited vertex v with smallest d[v] value  
    MST.add((K[x], x))  
    for each unreached neighbor v of x:  
        d[v] = min(w(x,v), d[v])  
        if d[v] was updated: k[v] = x  
    visited.add(x)  
return MST
```

B is the closest node to the growing tree.

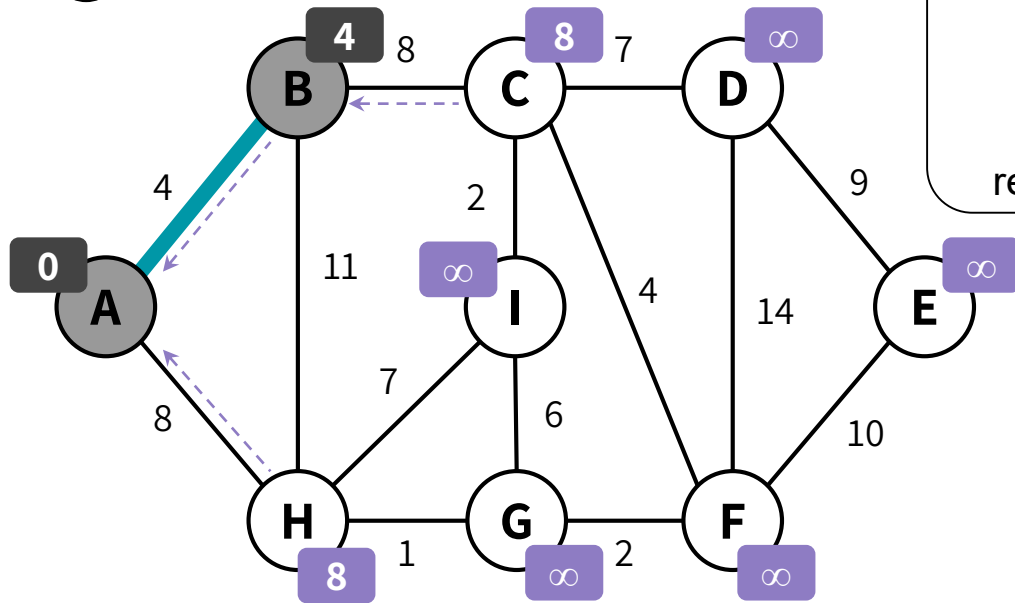
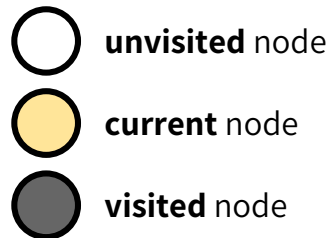
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        if d[v] was updated: k[v] = x  
    visited.add(x)  
return MST
```

Now that B is reached by the tree, see if any of its neighbors are closer to the tree because of it!

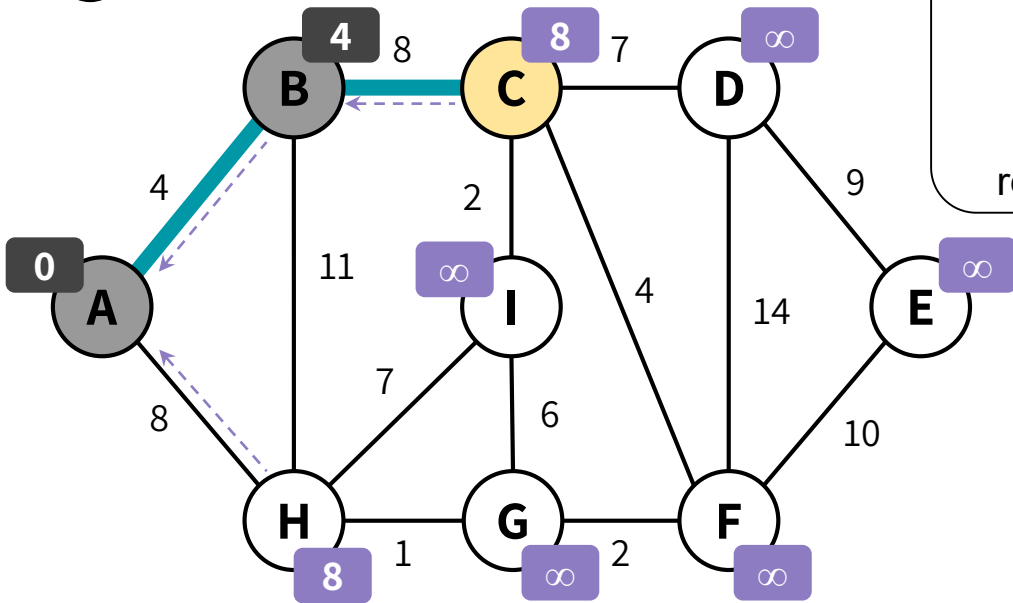
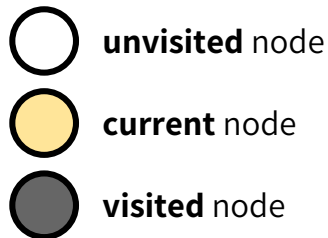
HOW DO WE IMPLEMENT THIS?



```
while len(visited) < n:  
    x = unvisited vertex v with smallest d[v] value  
    MST.add((K[x], x))  
    for each unreached neighbor v of x:  
        d[v] = min(w(x,v), d[v])  
        if d[v] was updated: k[v] = x  
    visited.add(x)  
return MST
```

Update their estimates, and now B is officially done.

HOW DO WE IMPLEMENT THIS?



```
while len(visited) < n:  
    x = unvisited vertex v with smallest d[v] value  
    MST.add((K[x], x))  
    for each unreached neighbor v of x:  
         $d[v] = \min(w(x,v), d[v])$   
        if d[v] was updated:  $k[v] = x$   
    visited.add(x)  
return MST
```

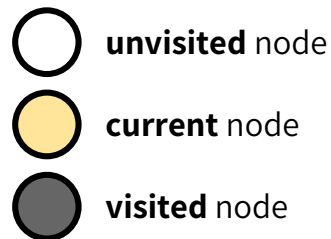
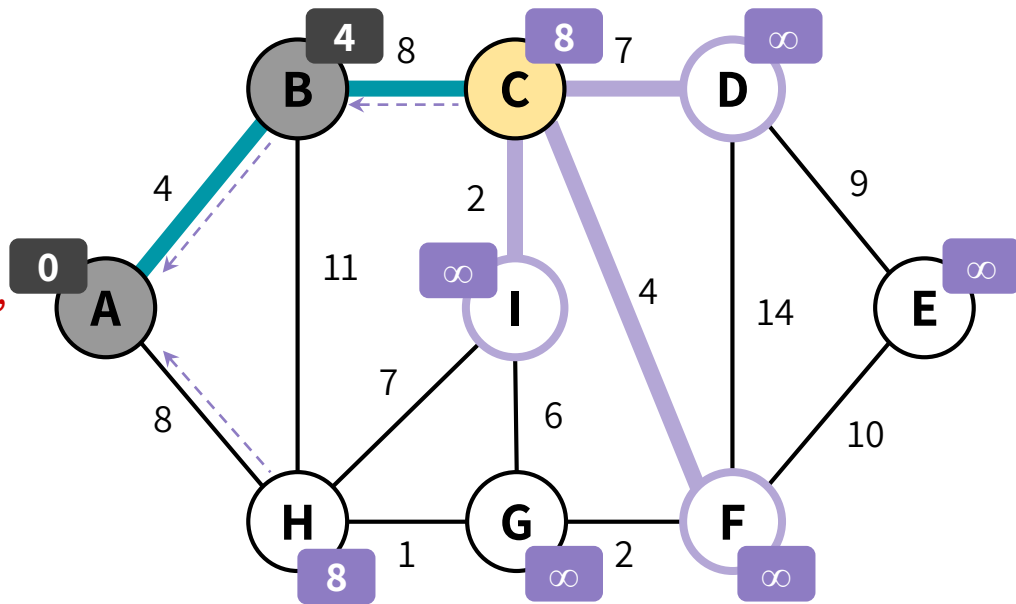
C is the closest node to the growing tree.

HOW DO WE IMPLEMENT THIS?

Each vertex that's not yet reached by the growing tree keeps track of:

- 1) the **distance** from itself to the growing spanning tree using *one edge*
- 2) **how to get to there** (the closest neighbor that's reached by the tree already)

Now that C is reached by the tree, see if any of its neighbors are closer to the tree because of it!



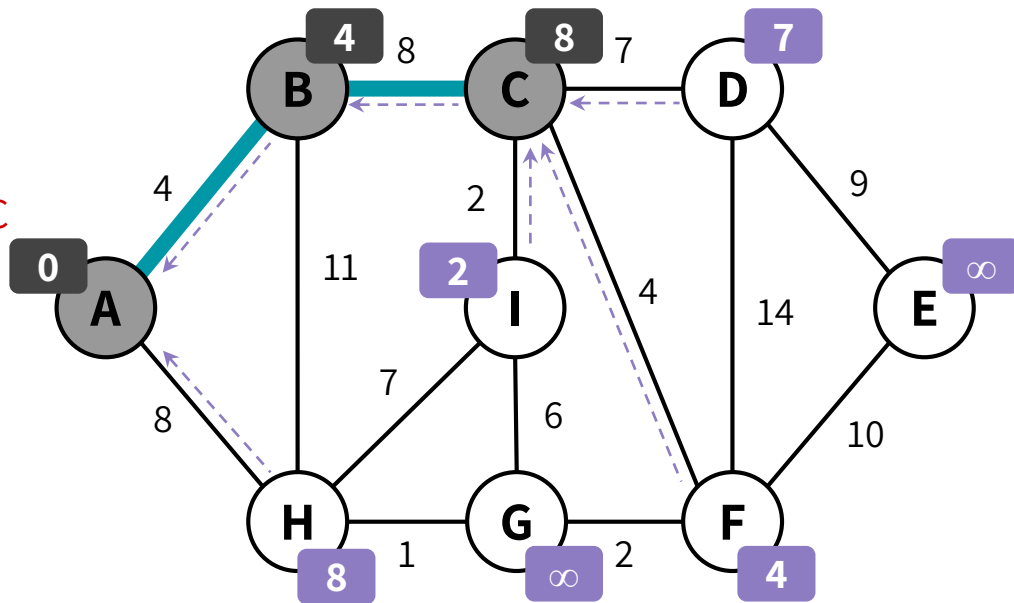
HOW DO WE IMPLEMENT THIS?

Each vertex that's not yet reached by the growing tree keeps track of:

- 1) the **distance** from itself to the growing spanning tree using *one edge*
- 2) **how to get to there** (the closest neighbor that's reached by the tree already)

Update their estimates, and now C is officially done.

Time to choose the lightest edge on the frontier (i.e. the edge whose endpoint has the lowest distance stored)



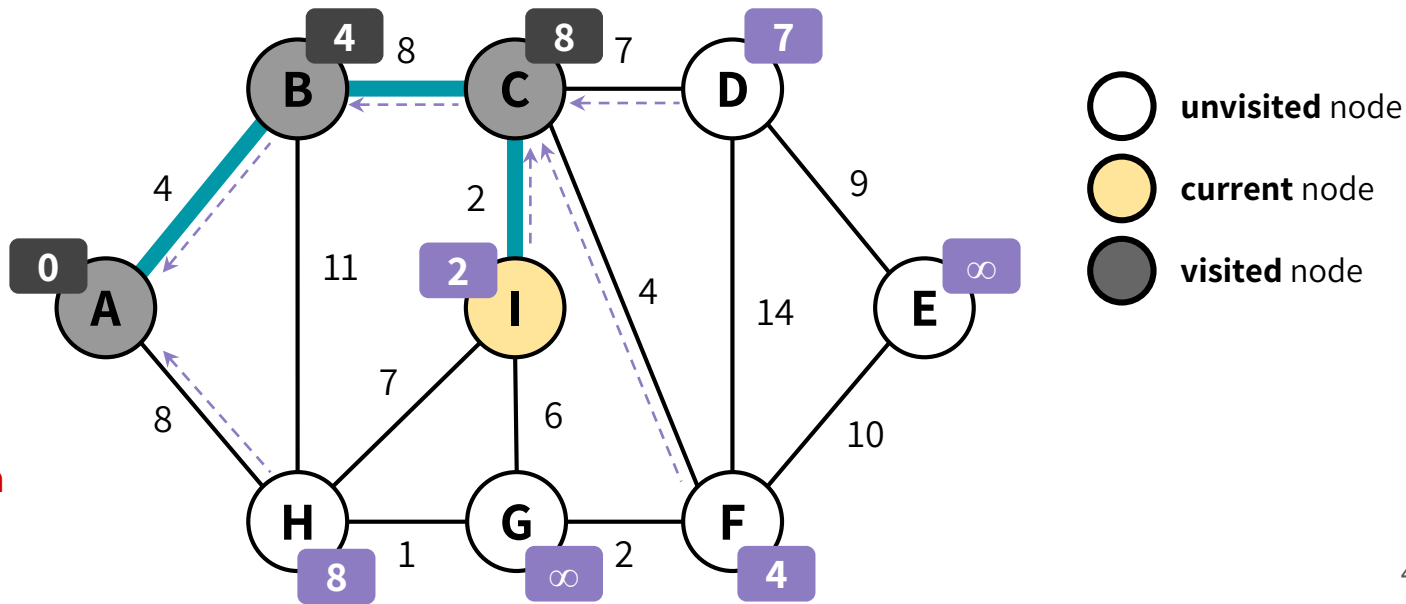
HOW DO WE IMPLEMENT THIS?

Each vertex that's not yet reached by the growing tree keeps track of:

- 1) the **distance** from itself to the growing spanning tree using *one edge*
- 2) **how to get to there** (the closest neighbor that's reached by the tree already)

I is the closest node to the growing tree.

Since we recorded how to get to the tree from I, we know which edge to add.

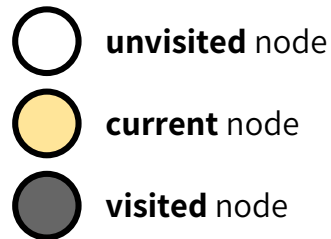
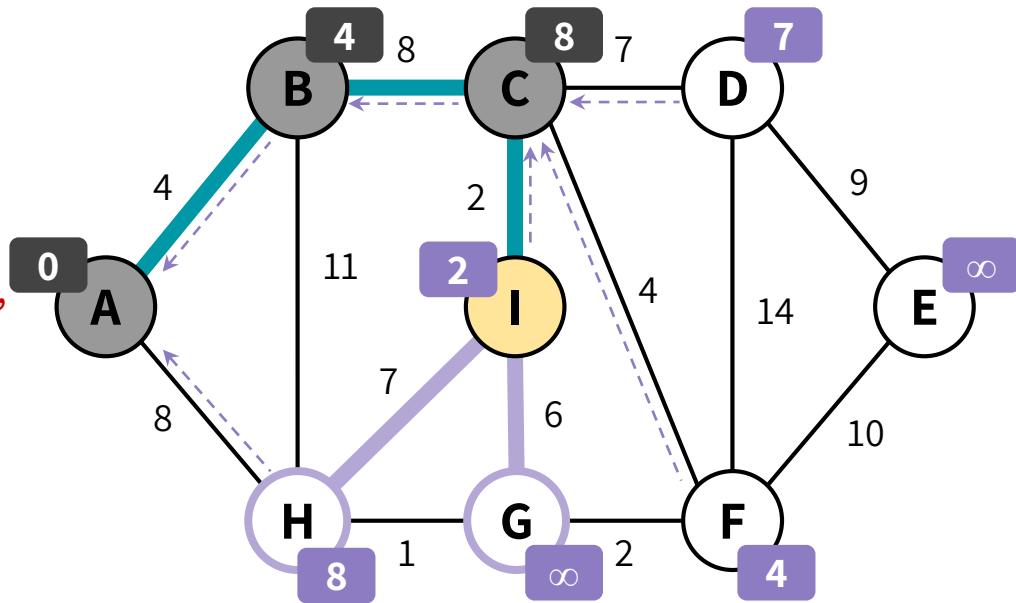


HOW DO WE IMPLEMENT THIS?

Each vertex that's not yet reached by the growing tree keeps track of:

- 1) the **distance** from itself to the growing spanning tree using *one edge*
- 2) **how to get to there** (the closest neighbor that's reached by the tree already)

Now that I is reached by the tree, see if any of its neighbors are closer to the tree because of it!



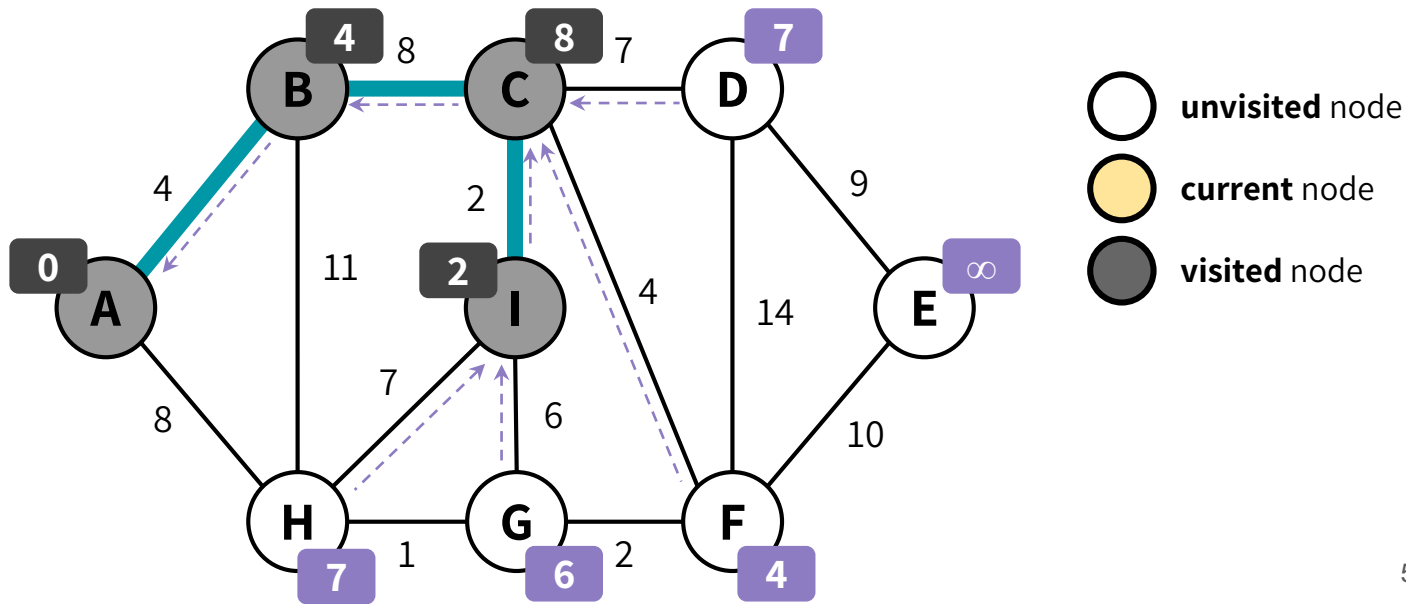
HOW DO WE IMPLEMENT THIS?

Each vertex that's not yet reached by the growing tree keeps track of:

- 1) the **distance** from itself to the growing spanning tree using *one edge*
- 2) **how to get to there** (the closest neighbor that's reached by the tree already)

Update their estimates, and now I is officially done.

Time to choose the lightest edge on the frontier (i.e. the edge whose endpoint has the lowest distance stored)



PRIM'S ALGORITHM: PSEUDOCODE

PRIM($G = (V, E)$, s):


MST = {}

visited = { s }

for all v besides s : $d[v] = \infty$ and $k[v] = \text{NULL}$

for each neighbor v of s : $d[v] = w(s, v)$ and $k[v] = s$

$k[v]$ stores the node in the growing tree that is closest to v (using one edge)



while $\text{len}(\text{visited}) < n$:

x = unvisited vertex v with smallest $d[v]$ value

MST.add(($K[x]$, x))

for each unreached neighbor v of x :

$d[v] = \min(w(x, v), d[v])$

if $d[v]$ was updated: $k[v] = x$

visited.add(x)

return MST

Runtime (using Min-heap): $O(E \log V)$

CLRS textbook version PSEUDOCODE For PRIM'S ALGORITHM

```
MST-PRIM( $G, w, r$ )
1  for each  $u \in G.V$ 
2       $u.key = \infty$ 
3       $u.\pi = \text{NIL}$ 
4   $r.key = 0$ 
5   $Q = G.V$ 
6  while  $Q \neq \emptyset$ 
7       $u = \text{EXTRACT-MIN}(Q)$ 
8      for each  $v \in G.Adj[u]$ 
9          if  $v \in Q$  and  $w(u, v) < v.key$ 
10              $v.\pi = u$ 
11              $v.key = w(u, v)$ 
```

Runtime (Build Min heap line 1-5): $O(V)$

(while loop excute $|V|$ and EXTRACT-MIN $\log V$): $O(V \log V)$

For loop line 8-11: $O(E)$

Total Prim Algo Runtime = $O(V \log V + E \log V) = O(E \log V)$???

Alg: HEAPSORT(*A*)

1. BUILD-MAX-HEAP(*A*)
2. **for** *i* \leftarrow length[*A*] **downto** 2
3. **do** exchange *A*[1] \leftrightarrow *A*[*i*]
4. MAX-HEAPIFY(*A*, 1, *i* - 1)

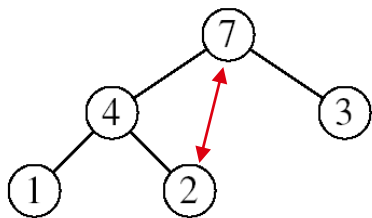
$O(\lg n)$ } $n-1$ times

- Running time: $O(n \lg n)$ --- Can be shown to be $\Theta(n \lg n)$

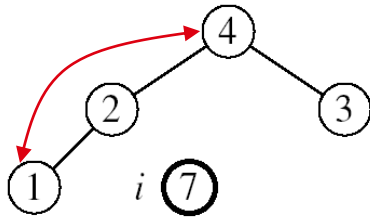
Example:

$A=[7, 4, 3, 1, 2]$

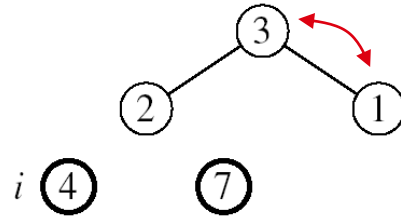
From Previous Lecture Slides



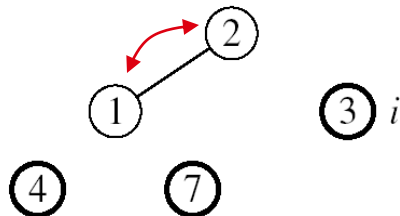
MAX-HEAPIFY(A, 1, 4)



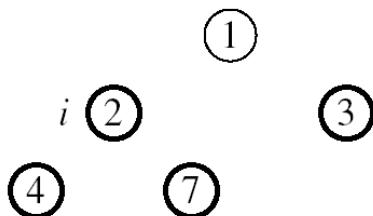
MAX-HEAPIFY(A, 1, 3)



MAX-HEAPIFY(A, 1, 2)



MAX-HEAPIFY(A, 1, 1)



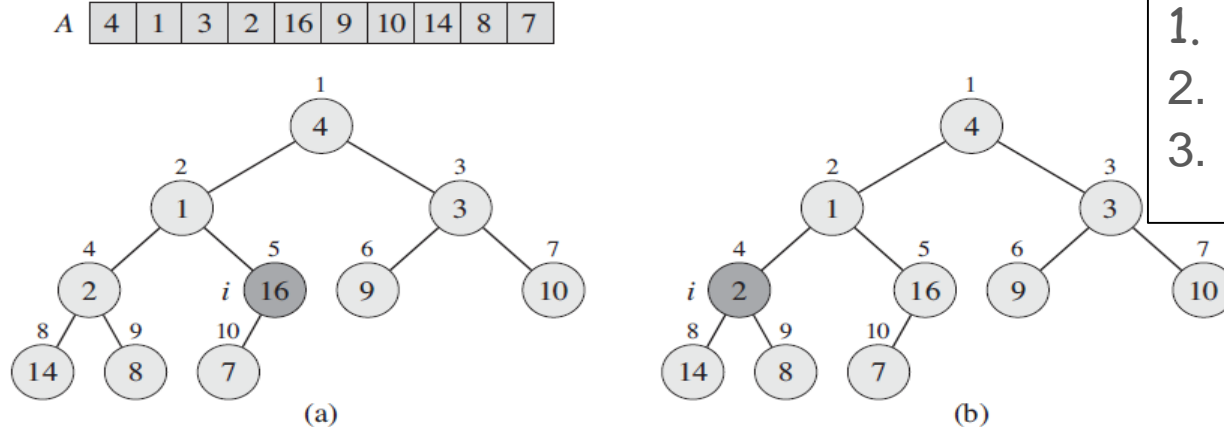
A

1	2	3	4	7
---	---	---	---	---

Build Max Heap Procedure

- Convert an array $A[1 \dots n]$ into a max-heap ($n = \text{length}[A]$)
- The elements in the subarray $A[(\lfloor n/2 \rfloor + 1) \dots n]$ are leaves
- Apply MAX-HEAPIFY on elements between 1 and $\lfloor n/2 \rfloor$

Figure 6.3 :



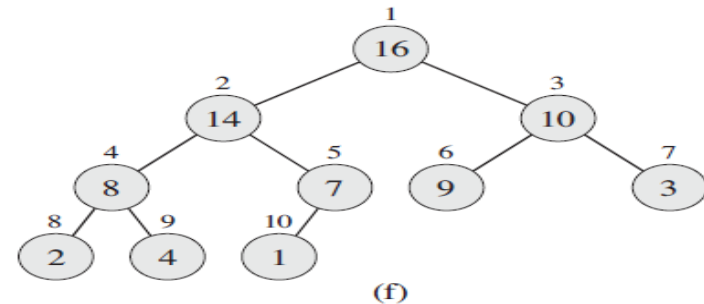
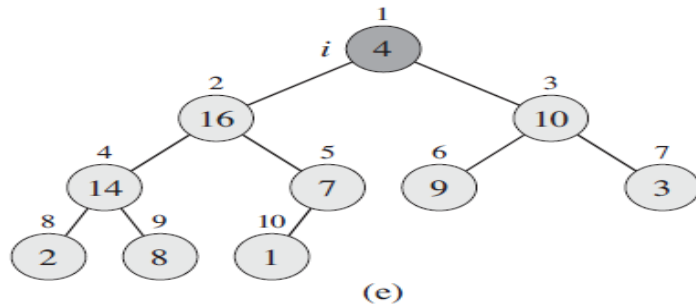
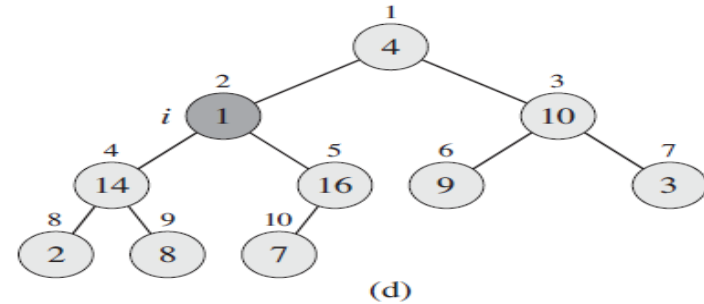
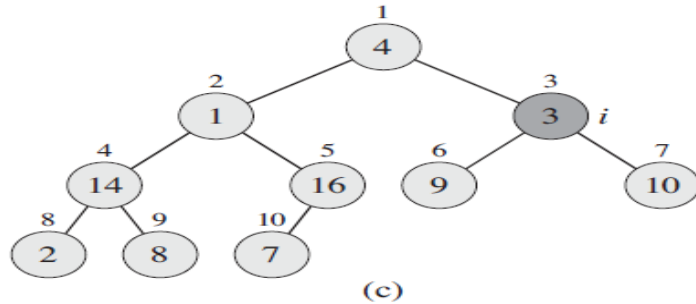
Alg: BUILD-MAX-HEAP(A)

1. $n = \text{length}[A]$
2. **for** $i \leftarrow \lfloor n/2 \rfloor$ **downto** 1
3. **do** MAX-HEAPIFY(A, i, n)

Build Max Heap Procedure

From Previous Lecture Slides

- Figure 6.3 :

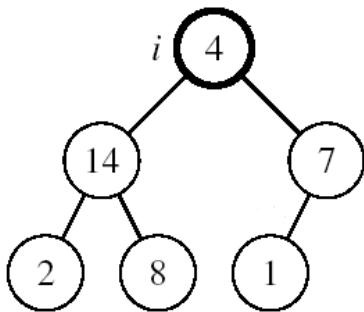


Maintaining the Heap Property

From Previous Lecture Slides

- Assumptions:

- Left and Right subtrees of i are max-heaps
- $A[i]$ may be smaller than its children



Alg: MAX-HEAPIFY(A, i, n)

1. $l \leftarrow \text{LEFT}(i)$
2. $r \leftarrow \text{RIGHT}(i)$
3. **if** $l \leq n$ and $A[l] > A[i]$
4. **then** $\text{largest} \leftarrow l$
5. **else** $\text{largest} \leftarrow i$
6. **if** $r \leq n$ and $A[r] > A[\text{largest}]$
7. **then** $\text{largest} \leftarrow r$
8. **if** $\text{largest} \neq i$
9. **then** exchange $A[i] \leftrightarrow A[\text{largest}]$
10. **MAX-HEAPIFY**($A, \text{largest}, n$)

APPLICATIONS OF MSTs

Network design

Find the most cost-effective way to connect cities with roads/water/electricity/phone

Image processing

Image segmentation, which finds connected regions in the image with minimal differences

Cluster analysis

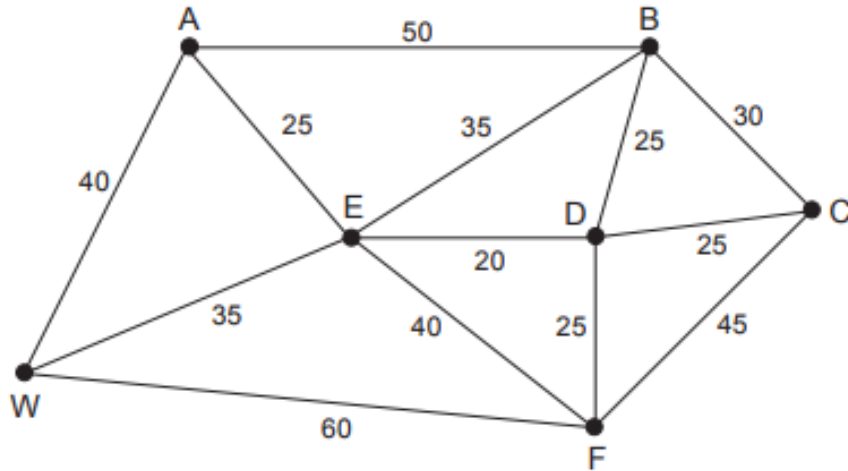
Find clusters in a dataset (one of the algorithms we'll see today can be modified slightly to basically do this)

Useful primitive

Finding an MST is often useful as a subroutine or approximation for more advanced graph algorithms

PRIM'S ALGORITHM: VERSION

Travel Agency wants to setup a public transport system between all the cities. The passenger fare in rupees between the cities are shown in the **Figure-2**. How should all cities be linked to maximize the total fare.
[Hint: Use Spanning Tree]



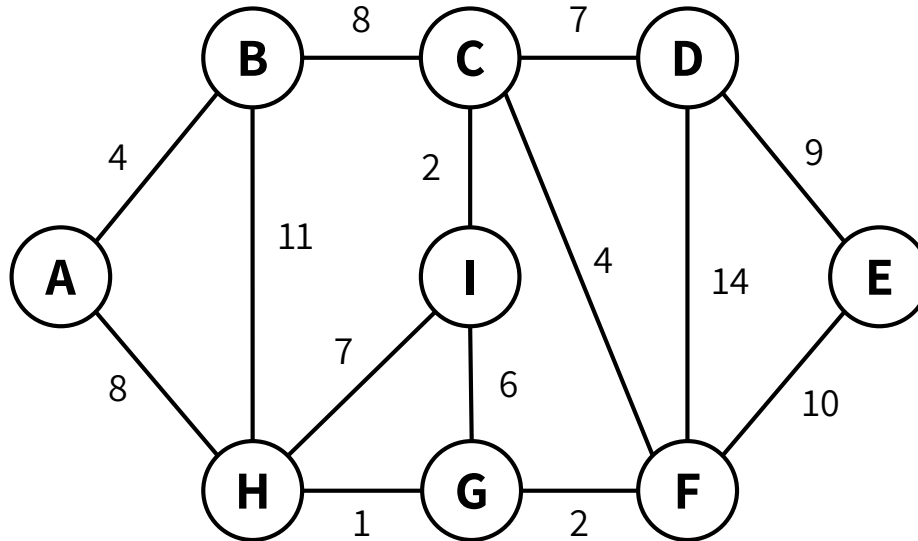
KRUSKAL'S ALGORITHM

Greedily add the cheapest edge!

KRUSKAL'S ALGORITHM: THE IDEA

Greedy choice:

Maintain a forest of trees, & greedily add the cheapest edge to combine trees

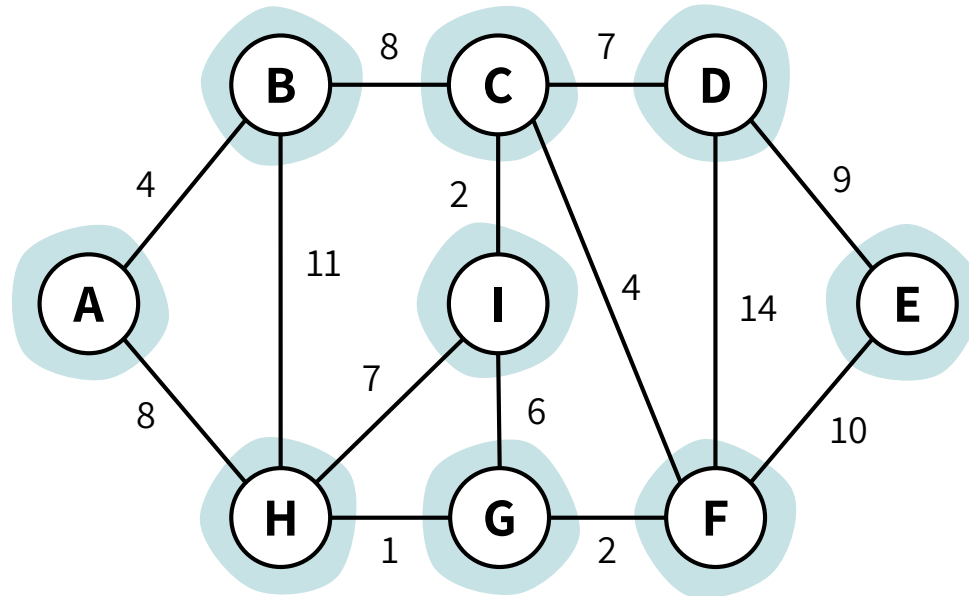


KRUSKAL'S ALGORITHM: THE IDEA

Greedy choice:

Maintain a forest of trees, & greedily add the cheapest edge to combine trees

Every node on its
own starts as an
individual tree in
this forest

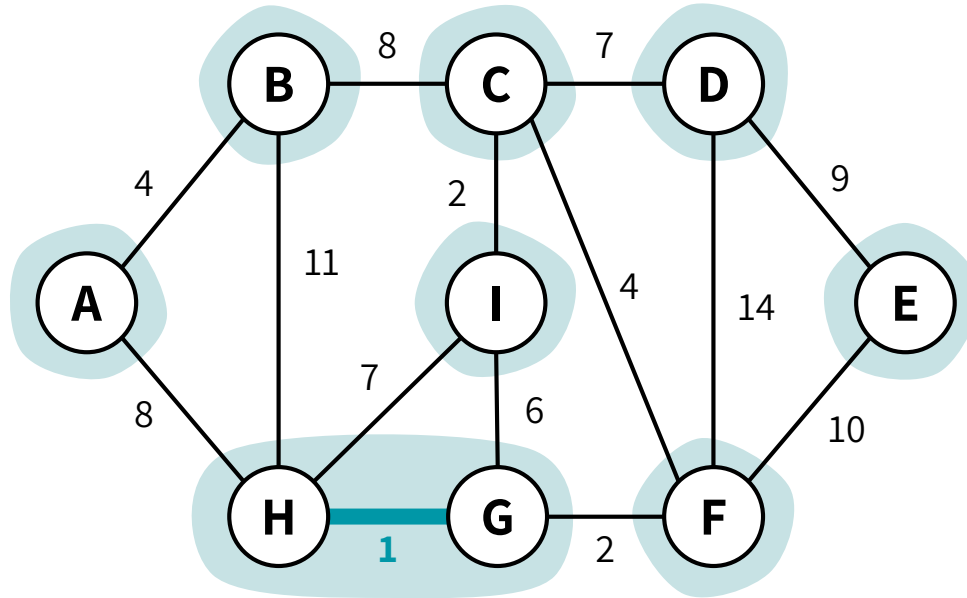


KRUSKAL'S ALGORITHM: THE IDEA

Greedy choice:

Maintain a forest of trees, & greedily add the cheapest edge to combine trees

Choose the
cheapest edge that
would combine
two trees
(i.e. that won't cause a cycle)

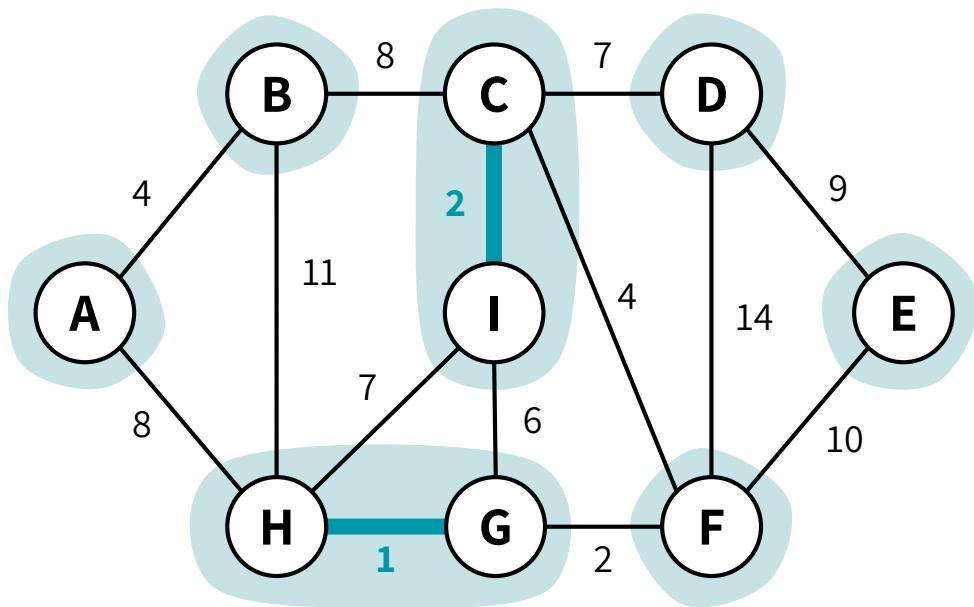


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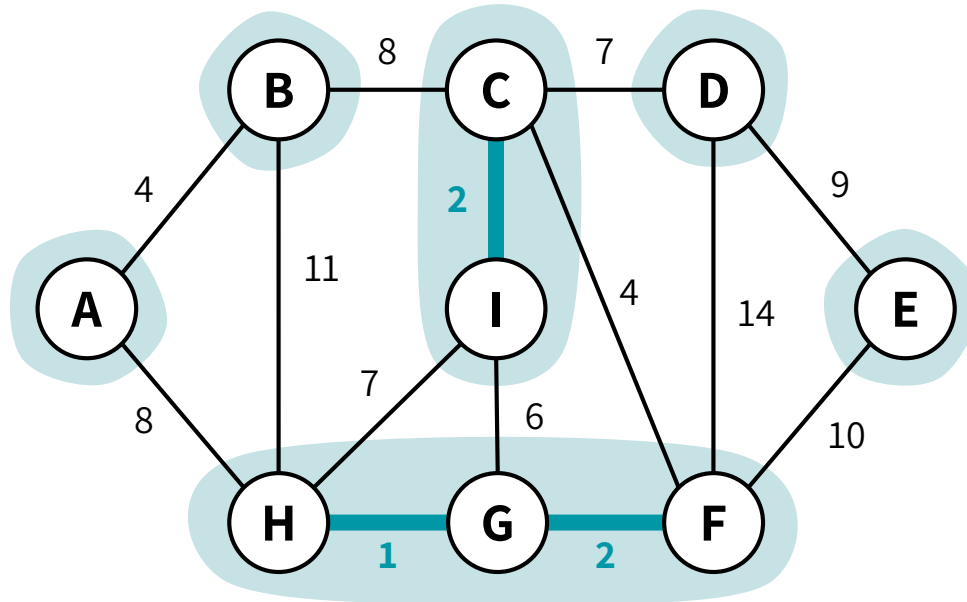
If there's a tie, choose
one of the edges

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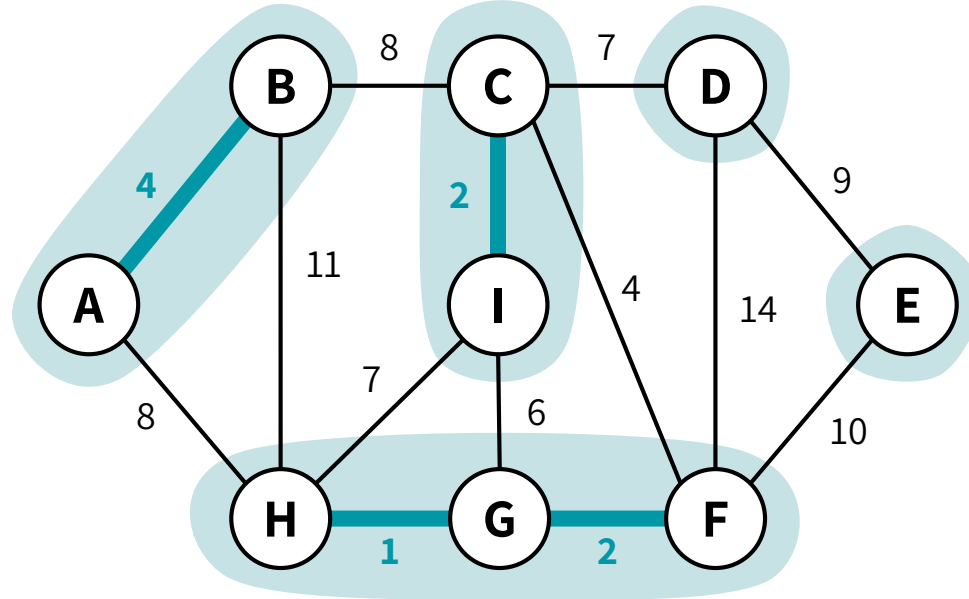


KRUSKAL'S ALGORITHM: THE IDEA

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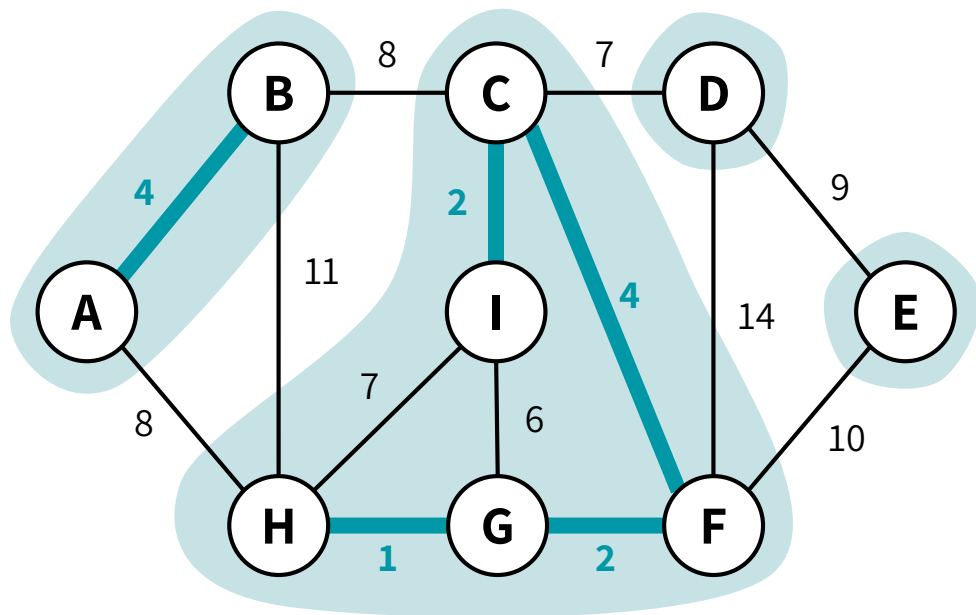


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KRUSKAL'S ALGORITHM: THE IDEA

Greedy choice:

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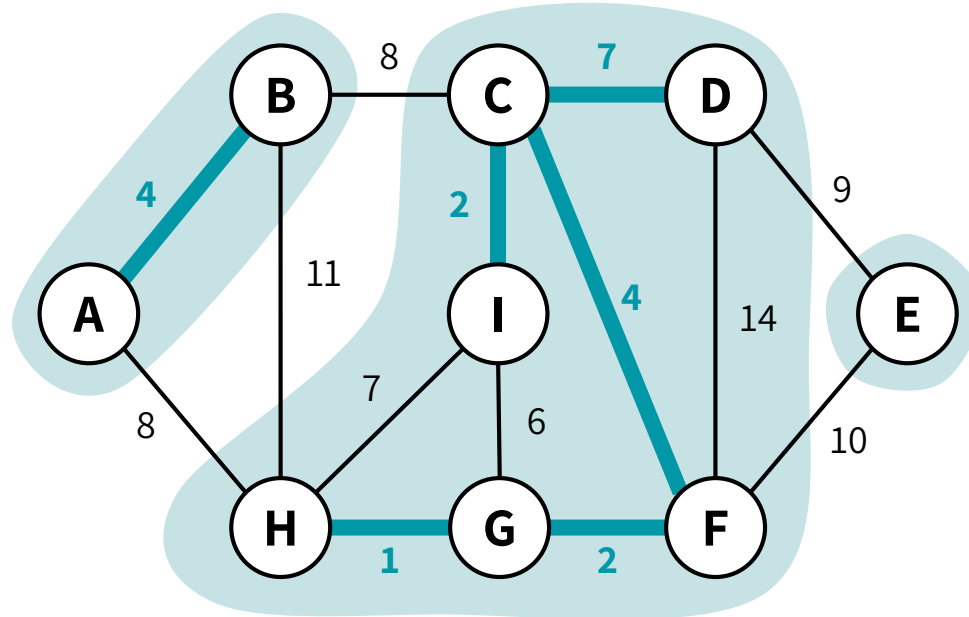
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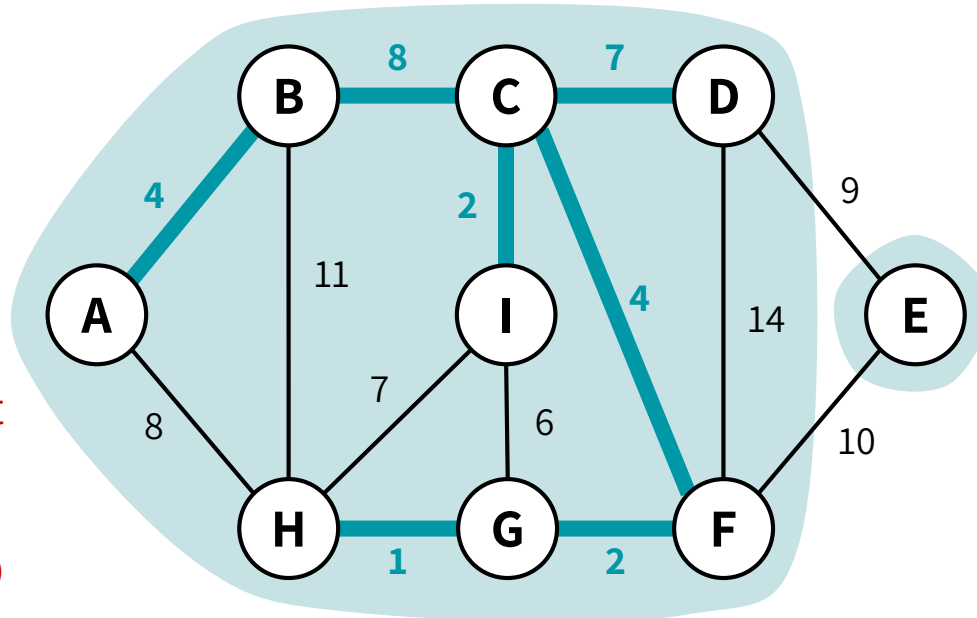


KRUSKAL'S ALGORITHM: THE IDEA

Greedy choice:

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Choose the
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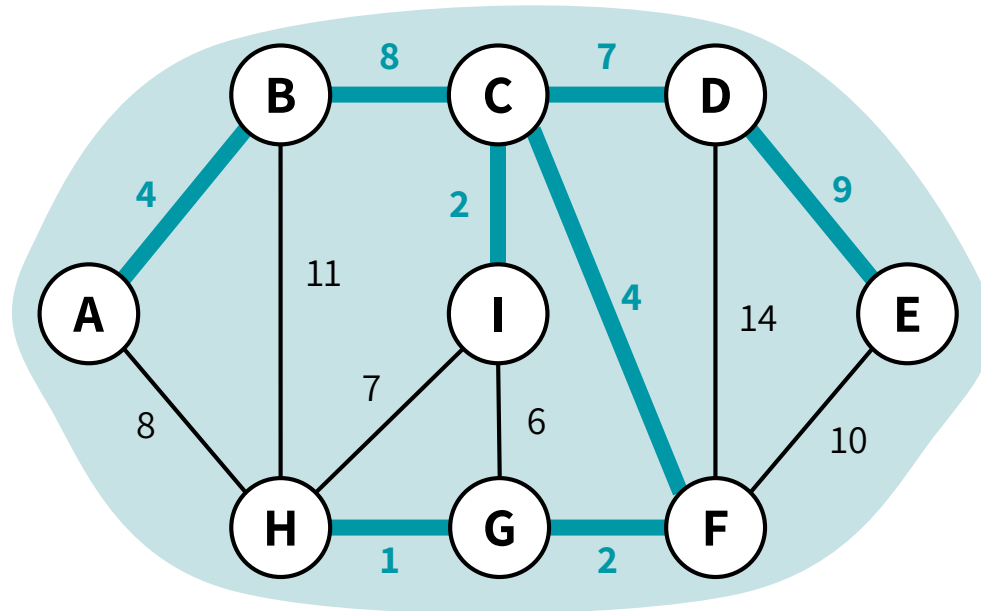


KRUSKAL'S ALGORITHM: THE IDEA

Greedy choice:

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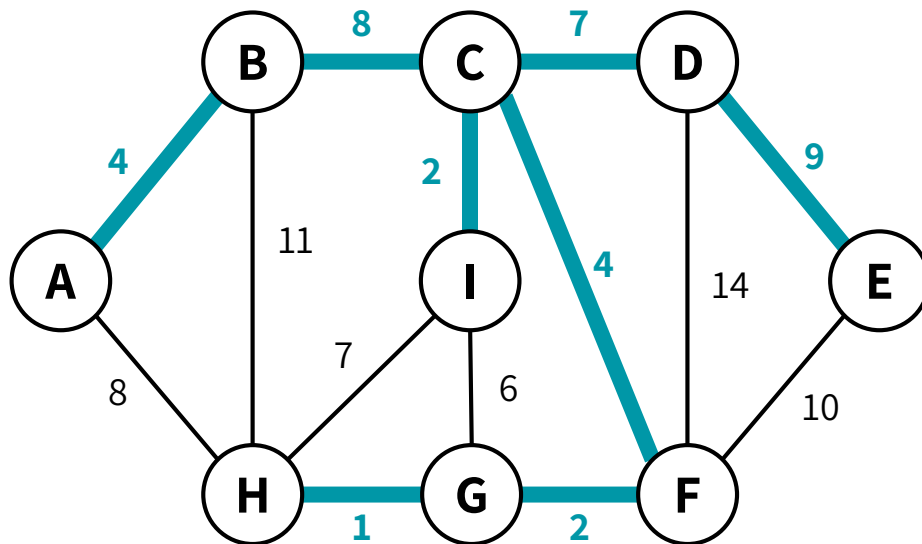
Choose the
cheapest edge that
would combine
two trees
(i.e. that won't cause a cycle)



KRUSKAL'S ALGORITHM: THE IDEA

Greedy choice:

Maintain a forest of trees, & greedily add the cheapest edge to combine trees



We're done!
This is the MST.

KRUSKAL'S ALGORITHM: PSEUDOCODE

KRUSKAL-NOT-VERY-DETAILED($G = (V, E)$):

E-SORTED = E sorted by weight in non-decreasing order

MST = {}

for v in V:

put v in its own tree

for (u,v) in E-SORTED:

if u's tree and v's tree are not the same:

MST.add((u,v))

merge u's tree with v's tree

return MST

KRUSKAL'S ALGORITHM: PSEUDOCODE

KRUSKAL-NOT-VERY-DETAILED($G = (V, E)$):

E-SORTED = E sorted by weight in non-decreasing order

MST = {}

for v in V :

put v in its own tree

for (u, v) in E-SORTED:

if u 's tree and v 's tree are not the same:

 MST.add((u, v))

merge u 's tree with v 's tree

return MST

To implement these lines, we'll use a ***Union-Find data structure***, which supports 3 operations: **MAKE-SET(x)**, **FIND(x)**, and **UNION(x,y)**

KRUSKAL'S ALGORITHM: PSEUDOCODE

KRUSKAL($G = (V, E)$):

E-SORTED = E sorted by weight in non-decreasing order

MST = {}

for v in V:

MAKE-SET(v)

for (u,v) in E-SORTED:


if FIND(u) **!= FIND**(v):

MST.add((u,v))

UNION(u,v)

return MST

Basically, the time to sort the edge weights dominates the runtime.
 $O(E \log E) = O(E \log V)$, since $E \leq V^2$



(With union-find data structure) **Runtime = $O(E \log V)$**

CLRS textbook version PSEUDOCODE For KRUSKAL'S ALGORITHM

MST-KRUSKAL(G, w)

```
1   $A = \emptyset$ 
2  for each vertex  $v \in G.V$ 
3      MAKE-SET( $v$ )
4  sort the edges of  $G.E$  into nondecreasing order by weight  $w$ 
5  for each edge  $(u, v) \in G.E$ , taken in nondecreasing order by weight
6      if FIND-SET( $u$ )  $\neq$  FIND-SET( $v$ )
7           $A = A \cup \{(u, v)\}$ 
8          UNION( $u, v$ )
9  return  $A$ 
```

since $E \leq V^2$, we have $\log E = O(\log V)$

$O(E \log E) = O(E \log V)$,

Runtime (Time to sort line 4): $O(E \log E)$ (merge sort)

(Make Set $|V|$, for loop 5-8 : $O(E)$)

Total Algo Runtime = $O(E \log E)$