ICSI 403: DESIGN AND ANALYSIS OF ALGORITHMS

<u>Spanning Trees – Chapter 23</u>

Background - Greedy Algorithms

- "Greedy" algorithms tend to be very simple.
- They make *locally-optimized* decisions
 - i.e., they make the most advantageous choice at any given point, without regard for any possible past or future knowledge about the problem.
- As such, the most appealing choice at the moment *may* turn out, in the long run, to be a poor choice
- Greedy algorithms never back up and "change their mind" based on what they later learn

Greedy Algorithms - Example

- The Huffman Tree-building algorithm is a greedy algorithm
 - It always picks the two remaining items with the lowest weights to select next for merging into a new subtree
 - In the case of Huffman, the greedy approach does happen to build the optimal encoding tree (as Huffman proved)

Greedy Algorithms – Another Example

- Counting change using standard US coins, with the constraint of using as few coins as possible
 - Coins: half-dollar, quarter, dime, nickel, penny
 - Being greedy, and wanting to use as few coins as possible, we'll always try to use the largest coin we can.

Greedy Algorithms - Example

- Example: Count out 36 cents in change
 - Step 1: The largest coin is half-dollar; too big
 - Step 2: Next-largest coin is quarter; USE ONE
 - Step 3: We still have 11 cents to count out. The half-dollar and quarter are too big; USE A DIME
 - Step 4: The half-dollar, quarter, dime, and nickel are too big; USE A PENNY.
 - Total change = 36 cents; we're done with 3 coins
 - This happens to be the optimal solution.

Greedy Algorithms – Another Example

- Example: Count out 16 cents worth of change, but assume there's a 12-cent coin in addition to the standard denominations:
 - Step 1: The half-dollar and quarter are too big
 - Step 2: Next-largest coin is the 12-cent; USE ONE
 - Step 3: We still have 4 cents to count out. We wind up using FOUR pennies
 - Total change = 16 cents; we're done with 5 coins
 - This is NOT the optimal solution (dime, nickel, penny takes two fewer coins).
 - The decision to use the 12-cent coin looked good at the time but backed us into a non-optimal solution.

Yet Another Greedy Algorithm

- The Discrete (0-1) Knapsack problem:
 - A thief with a knapsack is breaking into a jewelry store
 - If the weight in the knapsack exceeds some threshold, the sack will break, and the thief gets nothing
 - The thief wants to maximize the value of his loot
 - If the jewelry store contains n items from which the thief can choose, there are 2^n combinations to consider: the thief may grow old while planning the heist!
 - The thief can't take part of an item; it's all-or-nothing

The Discrete Knapsack Problem (2)

- One obvious strategy: Take the item with the largest value first
 - The problem: The item with the largest value might also have a correspondingly high weight
 - Suppose we have three items, valued at \$10, \$9, and \$9, with corresponding weights of 25, 10, and 10 pounds. Further, suppose the knapsack's weight limit is 30 pounds
 - If we start with the most valuable item (\$10 and 25 pounds), we can't take another item, so we wind up with \$10 in loot.
 - The optimal solution would be to take the two \$9 (10-pound) items, for a total of \$18 (and 20 lbs).

The Discrete Knapsack Problem (3)

- Another obvious strategy: Take the item with the smallest weight first
 - The problem: If the items that are light in weight have small values compared with their weights

The Discrete Knapsack Problem (4)

- A more sophisticated strategy: Take the item with the highest value/weight ratio first.
 - Suppose we have three items:
 - Item 1: \$50 / 5 lbs = \$10/lb
 - Item 2: \$60 / 10 lbs = \$6/lb
 - Item 3: \$140 / 20 lbs = \$7/lb
 - If we take the items in order of value per weight, we would take items 1 & 3 (\$50 & \$140, with weights of 5 & 20 lbs), for a total of \$190 / 25 lbs
 - But the optimal solution (which maximizes the total value) is to take items 2 & 3 (\$60 & \$140, with weights of 10 & 20 lbs), for a total of \$200 / 30 lbs.

The Discrete Knapsack Problem (5)

The greedy approach does not work (i.e., give us an optimal solution) on the discrete knapsack problem!

The Fractional Knapsack Problem

- A variant of the discrete knapsack problem is the Fractional Knapsack Problem:
 - The thief can take part of an item.
 - Think of the Discrete problem as gold/silver bars
 - Think of the Fractional problem as gold/silver dust
 - By being able to take part of an item, there is never any wasted space in the knapsack

The Fractional Knapsack Problem

- Consider the original three items:
 - Item 1: \$50 / 5 lbs = \$10/lb
 - Item 2: \$60 / 10 lbs = \$6/lb
 - Item 3: \$140 / 20 lbs = \$7/lb
- The thief would take all of item 1 (\$50/5 lb), all of item 3 (\$140/20 lb), and have 5 lbs of knapsack space left, with which he could take 5 lbs of item #2 (\$30)
- \bullet Total haul: \$50 + 140 + 30 = \$220
- Total haul: 5 + 20 + 5 = 30 lbs

The Fractional Knapsack Problem

- Although the greedy approach does not necessarily provide an optimal solution to the discrete knapsack problem, it does provide an optimal solution to the fractional knapsack problem.
- Now, on to spanning trees...

Motivation

- Given: a number of towns in a remote area, where roads are very expensive (per mile) to build
- Wanted: a way of getting to/from all towns in the region (not necessarily directly)
 - Going from town A to town B by way of some other town X is perfectly acceptable – there just has to be SOME path from A to B.
- Constraint: Given the cost of road building, minimize the total length of all roads required

Motivation – Another Example

- Given: an electric circuit, in which a number of pins need to be connected together (a common point on the schematic)
- Wanted: a way of getting all such pins connected (not necessarily directly to each other), so that they're electrically equivalent
- Constraint: Minimize the total length of all connecting wire required

Spanning Trees

- We can model these problems using <u>spanning trees</u>.
- Given a connected, undirected graph G = (V, E), and a weight w(u, v) for each edge $(u, v) \in E$, ...
- We want to find a subset of E, which we'll call T (i.e., $T \subset E$), such that

$$w(t) = \sum_{(u,v)\in T} w(u,v)$$

is minimized

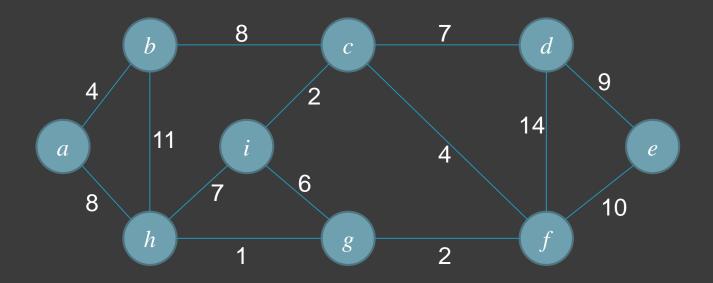
Spanning Trees

- Since T is acyclic, and it connects all of the vertices, it must form a (single) tree, which "spans" the graph, making it a <u>spanning tree</u>.
 - It contains all nodes but has no cycles.
- The problem of finding the tree T that minimizes the weights is called the minimum spanning tree problem.
 - Note that what we're minimizing is the total weight of all of the edges in the tree; not the size of the tree itself (the number of vertices or edges it contains). It's more accurately called the "minimum-weight spanning tree" problem.

Minimum Spanning Trees (MST)

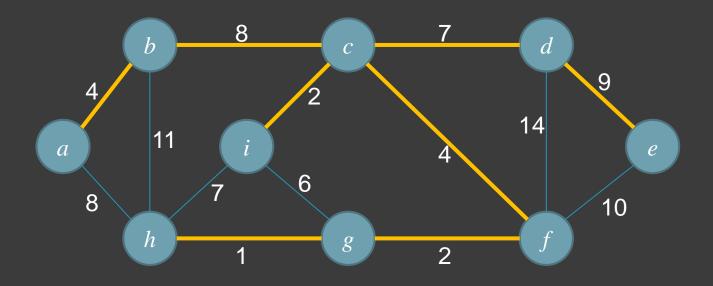
- The two algorithms we consider today, Prim's and Kruskal's, both find minimum spanning trees by using a greedy algorithm.
- We've already seen that greedy algorithms are not always guaranteed to find the optimal solutions.
- For the minimum spanning tree problem, however, just like the fractional knapsack problem, they DO always find an optimal solution!

MST Example



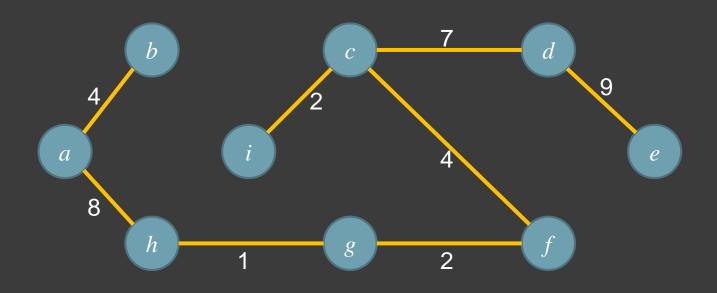
 Given this graph, with vertices a, b, ..., i, and edges as shown, with labeled weights, find the (a) minimum spanning tree.

MST Example



 You can get from any vertex to any other vertex using this tree. The sum of the weights is 37.

MST Example



• If we drop (b, c), which had a weight of 8, and replace it with (a, h), which also had a weight of 8, we still have a spanning tree with a weight of 37. "The" MST may not be unique!

MST Properties

- Some properties of an MST:
 - It has |V| 1 edges.
 - It has no cycles.
 - It might not be unique

How To Grow an MST?

- Building up the solution:
 - We will build a set A of edges.
 - Initially, A has no edges.
 - As we add edges to A, maintain a loop invariant:
 - A is a subset of some MST.
 - Add only edges that maintain the invariant. If A is a subset of some MST, an edge (u, v) is safe for A if and only if A U {(u, v)} is also a subset of some MST. So we will add only safe edges.

The Generic MST Algorithm

GENERIC-MST(G, w)

- $1 A = \emptyset$
- 2 **while** A does not form a spanning tree
- 3 find and edge (u, v) that is safe for A
- $\overline{A} = A \cup \{(u, v)\}$
- 5 return A

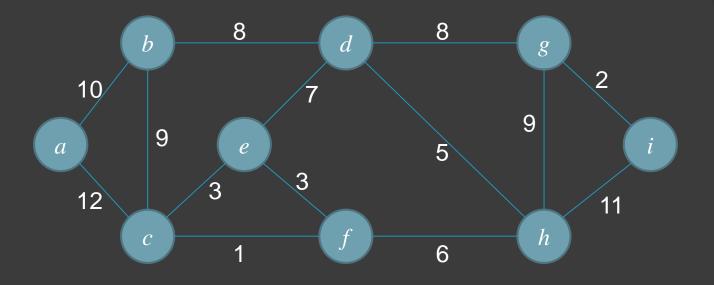
Loop Invariant and Safe Edges

- We can use the loop invariant (A is a subset of some MST) to show that this generic algorithm works.
- Initialization: The empty set trivially satisfies the loop invariant.
- Maintenance: Since we add only safe edges,
 A remains a subset of some MST.
- Termination: All edges added to A are in an MST, so when we stop, A is a spanning tree that is also an MST.

Loop Invariant and Safe Edges

- Obviously, the key to making this work lies in the details of "find an edge...that is safe for A"
- O HOW?
- An example:

Loop Invariant and Safe Edges

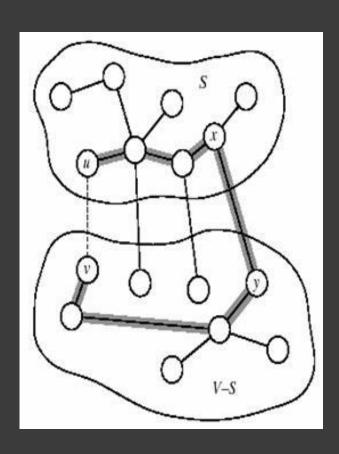


• Edge (c, f) has the lowest weight of any edge in the graph. Is this safe for $A = \emptyset$?

Safe Edges

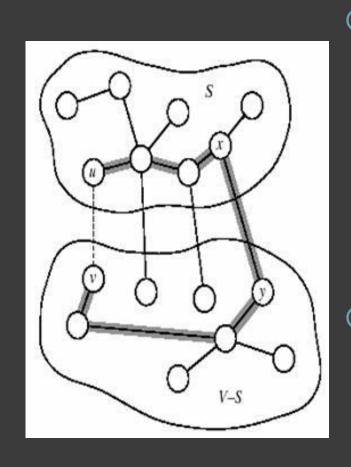
- Intuitively: Let $S \subseteq V$ be any set of vertices that includes c but not f (so that f is in V S).
- In any MST, there has to be at least one edge that connects S with V-S.
- Why not choose the edge with minimum weight (which would be (c, f) in this case)?
- Some definitions:

Safe Edges – Definitions (1)



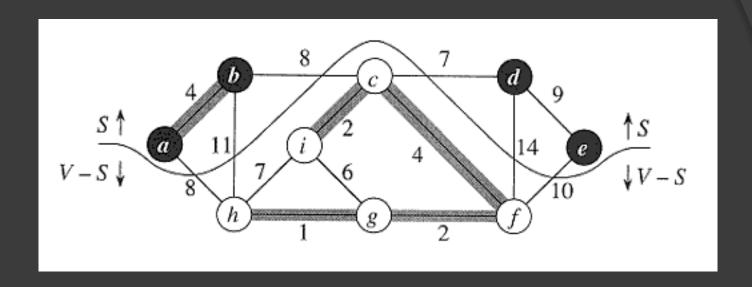
- A <u>cut</u> (S, V S) is a partition of vertices into disjoint sets V and S V.
- Edge $(u, v) \in E$ <u>crosses</u> the cut (S, V S) if one endpoint is in S and the other is in V S.
- A cut <u>respects</u> A if and only if no edge in A crosses the cut.

Safe Edges – Definitions (2)



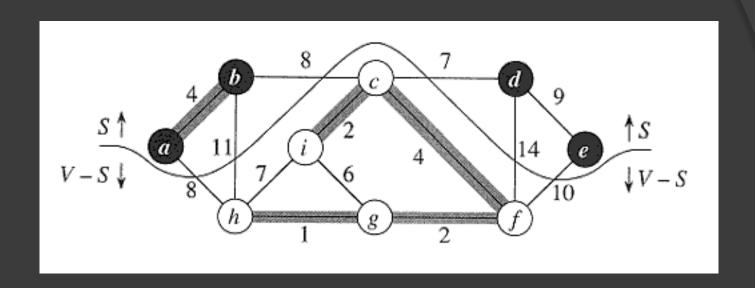
- An edge is a <u>light edge</u>
 crossing a cut if and only if its
 weight is minimum over all
 edges crossing the cut. For a
 given cut, there can be > 1
 light edge crossing it.
- If A is shown as shaded edges, the cut does not respect A; it cuts the edge (x, y)

Safe Edges – Definitions (3)



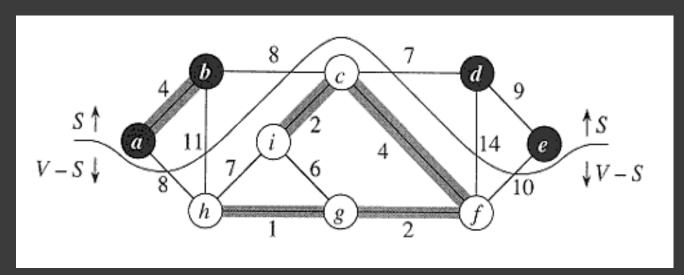
- Vertices in S are shown in black, vertices in V-S are shown in white.
- Edges crossing the cut connect a black vertex to a white one.

Safe Edges – Definitions (4)

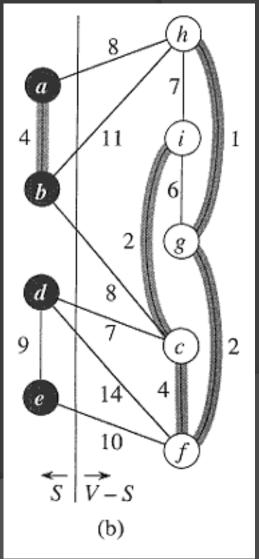


- A subset A of the edges is shaded.
- ullet Note that the cut (S, V S) respects A, since no edge of A crosses the cut

Safe Edges – Definitions (5)



- Another way of viewing this cut.
- An edge crosses the cut if it connects a vertex on the left (in S) with a vertex on the right (in V – S)



Theorem 23.1

• Let A be a subset of some MST, (S, V - S) be a cut that respects A, and (u, v) be a light edge crossing (S, V - S). Then (u, v) is safe for A.

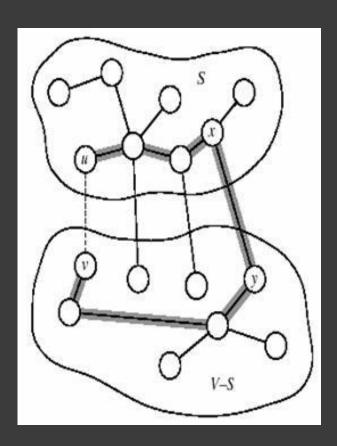
Proof

- Let T be an MST that includes A.
- If *T* contains (*u*, *v*), done.
- Assume T does not contain (u, v). We'll construct a different MST T that includes $A \cup \{(u, v)\}$.

Theorem 23.1 (2)

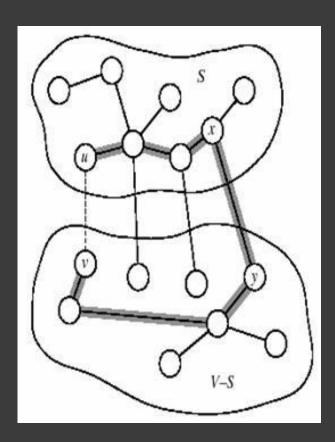
- Recall: a tree has unique path between each pair of vertices. Since T is an MST, it contains a unique path p between u and v. Path p must cross the cut (S, V S) at least once. Let (x, y) be an edge of p that crosses the cut. From how we chose (u, v), must have $w(u, v) \le w(x, y)$.
- Since the cut respects A, edge (x, y) is not in A in the graph.

Theorem 23.1 (3)



- \bullet To form T' from T:
- Remove (x, y), which breaks T into two components.
- Add (u, v), reconnecting the trees.
- So $T' = T \{(x, y)\} \cup \{(u, v)\},$ and T' is a spanning tree.

Theorem 23.1 (4)



- w(T') = w(T) w(x, y) + w(u, v) $\leq w(T)$, since $w(u, v) \leq w(x, y)$.
- Since T' is a spanning tree, $w(T') \le w(T)$, and T is an MST, then T' must be an MST.
- Need to show that (u, v) is safe for A:
- $A \subseteq T$ and (x, y) not in A; thus, $A \subseteq T$, and $A \cup \{(u, v)\} \subseteq T$
- Since T is an MST, (u, v) is safe for A

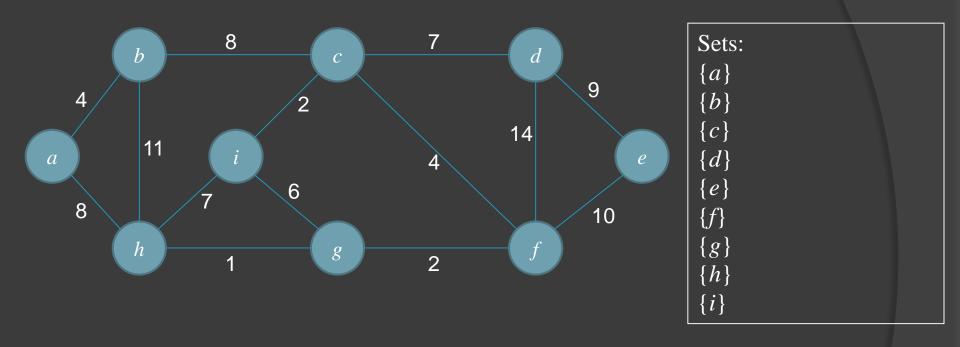
Kruskal's & Prim's Algorithms

- Both are elaborations of GENERIC-MST.
- They differ in how they select a safe edge in line 3 of GENERIC-MST.
- In Kruskal's algorithm, the set A is a forest (multiple trees). The safe edge we add always connects two distinct trees
- In Prim's algorithm, the set A is a single tree. The safe edge we add is always a least-weight edge connecting the tree to a vertex not yet in the tree.

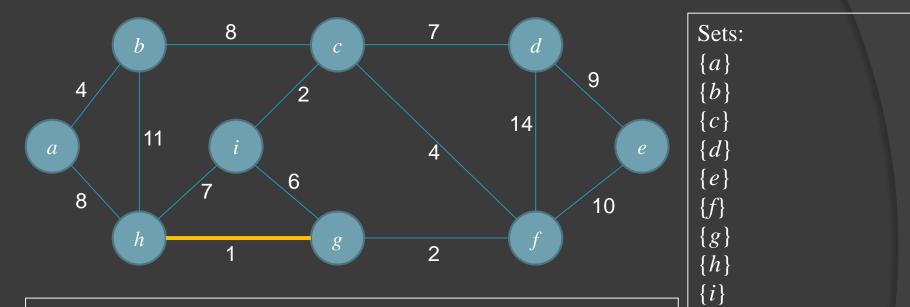
- Kruskal's algorithm depends on the existence of few set-based functions:
- Make-Set(x) creates a set containing the single item x.
- FIND-SET(x) looks through the sets it is maintaining, and determines which set x belongs to
- UNION(u, v) merges two sets (one containing u and the other containing v) into one set (u)

MST-KRUSKAL Algorithm

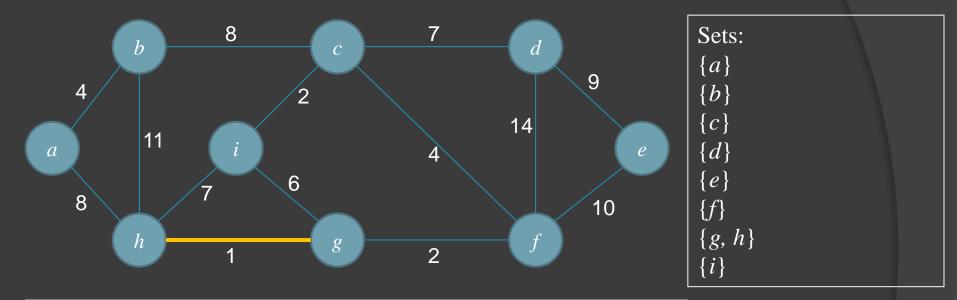
```
MST-KRUSKAL(G, w)
1 A = \emptyset
2 for each vertex v \in G.V
      MAKE-SET(v)
4 sort the edges of E by increasing w (weight) value
5 for each edge (u, v) \in E (taken in weight order)
      if FIND-SET(u) \neq FIND-SET(v)
          A = A \cup \{(u, v)\}
          UNION(u, v)
 return A
```



- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets

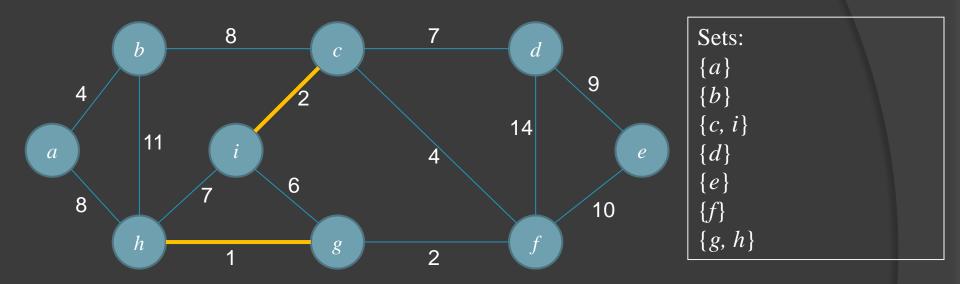


- The lowest-weight edge is (h, g), and h and g are in different sets, so include the edge and merge the two sets
 - Start with every vertex in its own set
 - Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
 - If we include the edge, merge the two sets



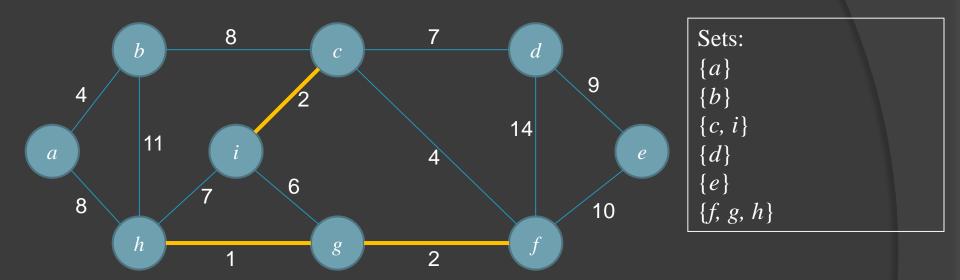
The lowest-weight edges are (c, i) and (g, f). Vertices c and i are in different sets, so include the edge and merge the sets

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



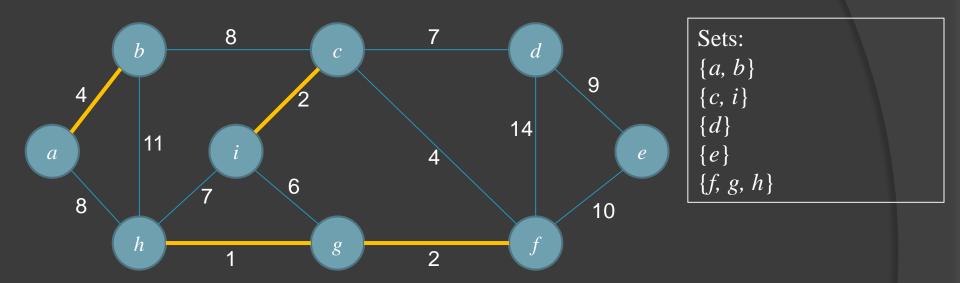
The lowest-weight remaining edge is (g, f). Vertices g and f are in different sets, so include the edge and merge the sets

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



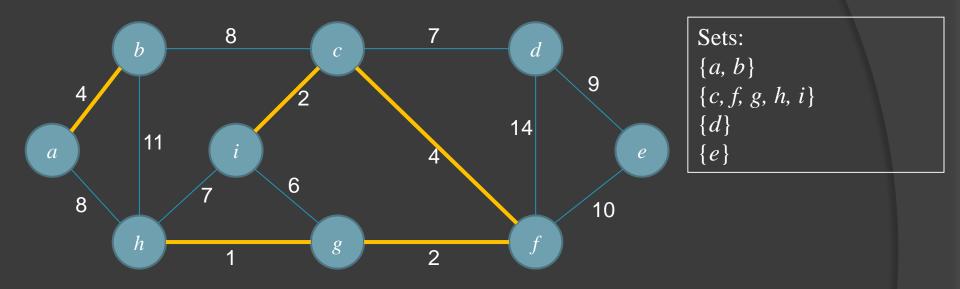
The lowest-weight remaining edges are (a, b) and (c, f) Vertices a and b are in different sets, so include the edge and merge the sets

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



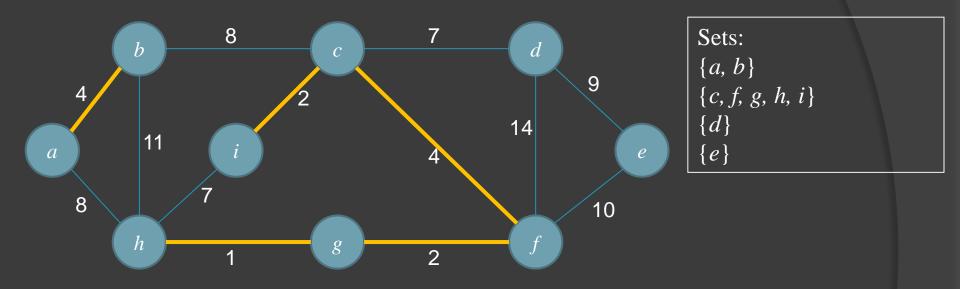
The lowest-weight remaining edge is (c, f) Vertices c and f are in different sets, so include the edge and merge the sets

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



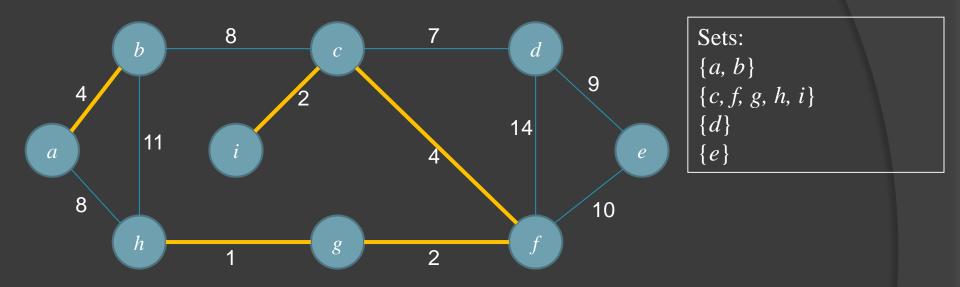
The lowest-weight remaining edge is (g, i) Vertices g and i are in the same set, so we drop this edge from consideration, and keep going

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



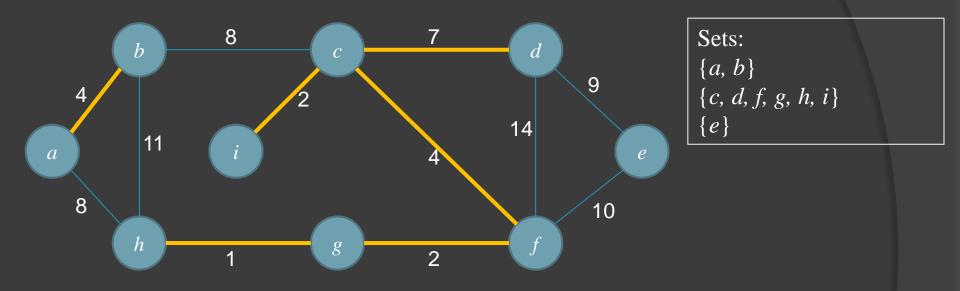
The lowest-weight remaining edges are (h, i) and (c, d). Vertices h and i are in the same set, so we drop this edge from consideration, and keep going

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



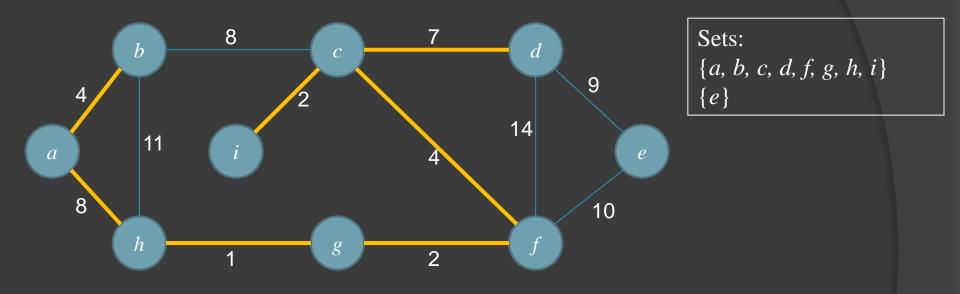
The lowest-weight remaining edge is (c, d). Vertices c and d are in different sets, so include this edge and merge the sets

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



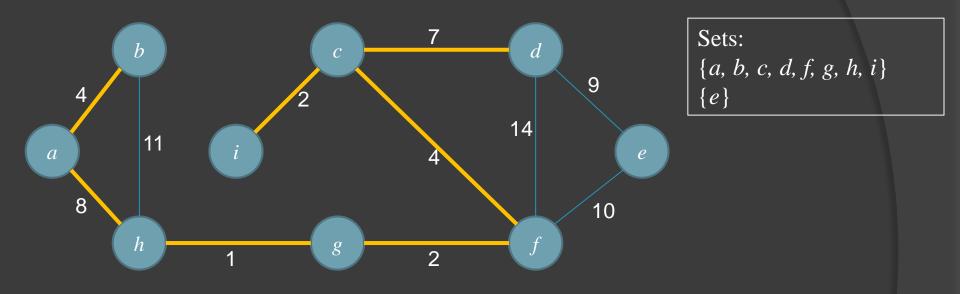
The lowest-weight remaining edges are (a, h) and (b, c). Vertices a and h are in different sets, so include this edge and merge the sets

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
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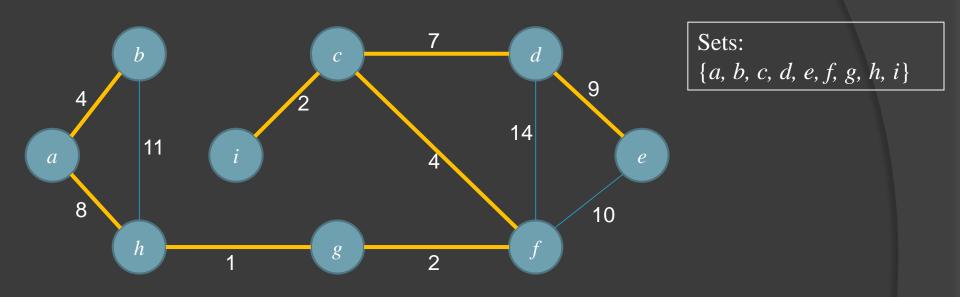
The lowest-weight remaining edge is (b, c). Vertices b and c are in the same set, so drop this edge from consideration and keep going

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



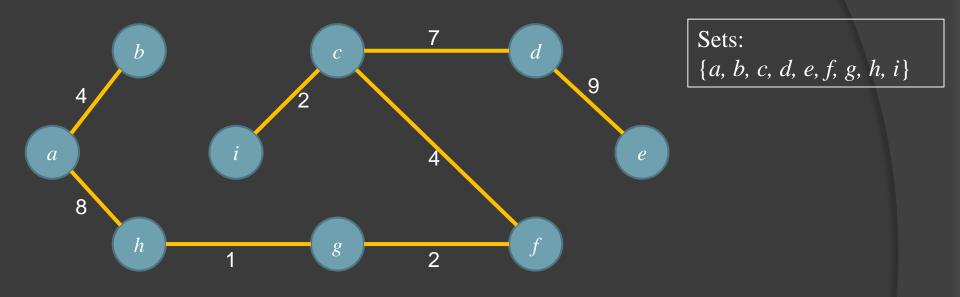
The lowest-weight remaining edge is (d, e). Vertices d and e are in different sets, so include this edge and merge the two sets

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



We will next consider the edges (f, e), (b, h), and (d, f) (in that order). Because they are all already in the same set, we will drop them all.

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets



All nodes are connected. Total MST weight is 4 + 8 + 1 + 2 + 4 + 2 + 7 + 9 = 37

- Start with every vertex in its own set
- Consider edges in increasing order ONLY if the edge connects vertices in two different sets.
- If we include the edge, merge the two sets

MST-Kruskal Algorithm - Analysis

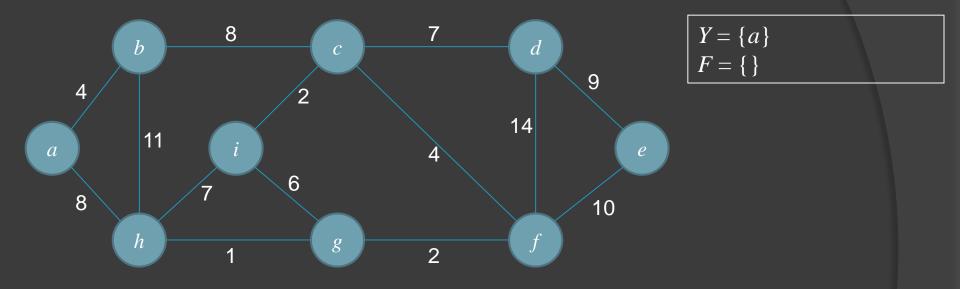
```
MST-KRUSKAL(G, w)
1 A = \emptyset
2 for each vertex v \in G.V
       MAKE-SET(v)
4 sort the edges of E by increasing w (weight) value
5 for each edge (u, v) \in E (taken in weight order)
       if FIND-\overline{SET}(u) \neq \overline{FIND}-\overline{SET}(v)
          A = A \cup \{(u, v)\}
          UNION(u, v)
9 return A
```

Total time: $O(E \lg V)$ (see p. 633)

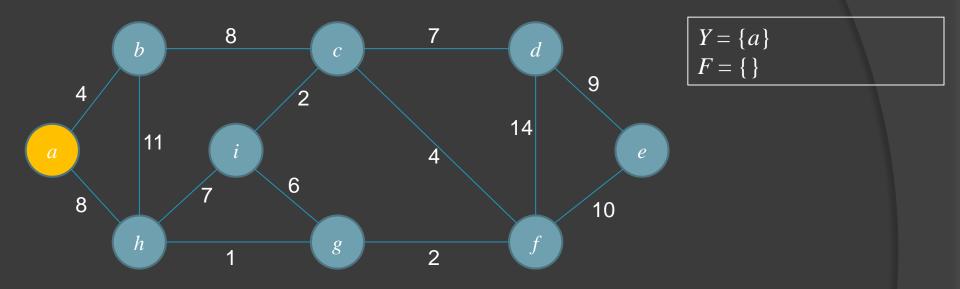
- In Prim's algorithm, the set A is one tree. The safe edge we add is always a least-weight edge connecting the tree to a vertex not yet in the tree.
- Start with an *empty* subset of the edges (F), and a subset of the vertices (Y) containing only a single (arbitrarily chosen) vertex, v_1 .
- A vertex nearest to Y (separated by the lowest weight edge) is a vertex in (V-Y) connected to a vertex in Y by an edge of minimum weight.
- The vertex nearest to Y is added to Y, and the edge is added to F. We repeat until Y = V

MST-Prim Algorithm

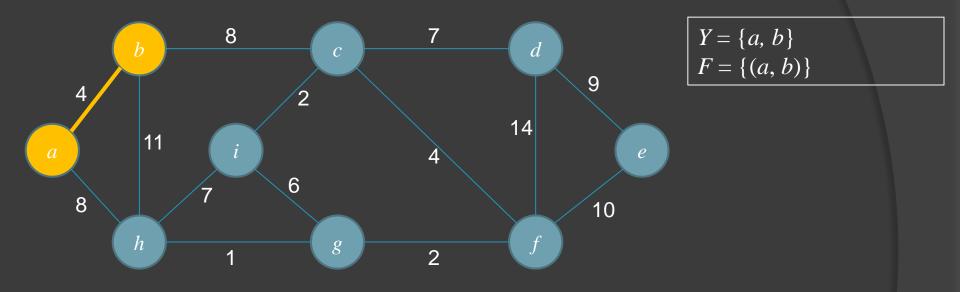
```
MST-PRIM(G, w, r) // r is an arbitrarily chosen vertex
 1 for each u \in G.V
 2 \overline{u.key} = \overline{\infty}
 3 u.\pi = NIL
 4 r.key = 0
 5 Q = G.V
                            //Q is a min-priority queue
 6 while Q \neq \emptyset
            u = \text{EXTRACT-MIN}(Q)
 8
            for each v \in Adj[u]
 9
                if v \in Q and w(u, v) < v.key
10
                    v.\pi = u
                    v.key = w(u, v)
```



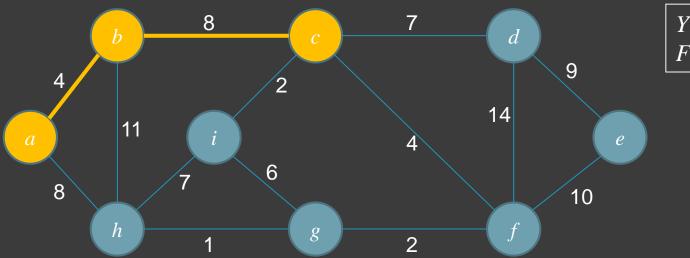
Let's start (arbitrarily) from a.



- Let's start (arbitrarily) from a.
- The lowest-weight edge connecting Y to (V-Y) is (a, b)



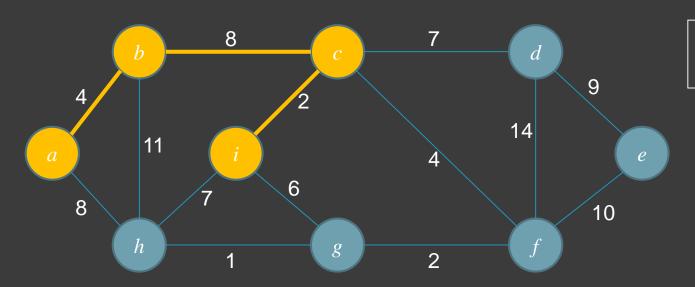
- The lowest-weight edges connecting Y to (V-Y) are (b, c) and (a, h). We will arbitrarily choose (b, c).



$$Y = \{a, b, c\}$$

 $F = \{(a, b), (b, c)\}$

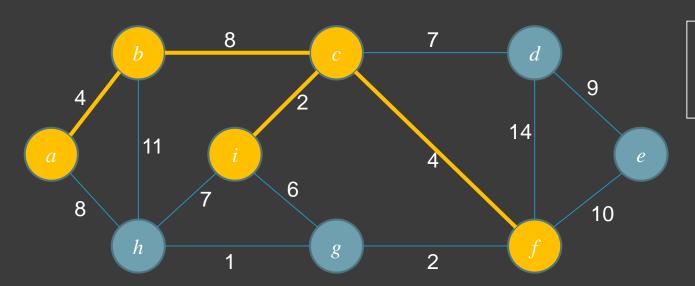
- The lowest-weight edge connecting Y to (V-Y) is (c, i)



$$Y = \{a, b, c, i\}$$

 $F = \{(a, b), (b, c), (c, i)\}$

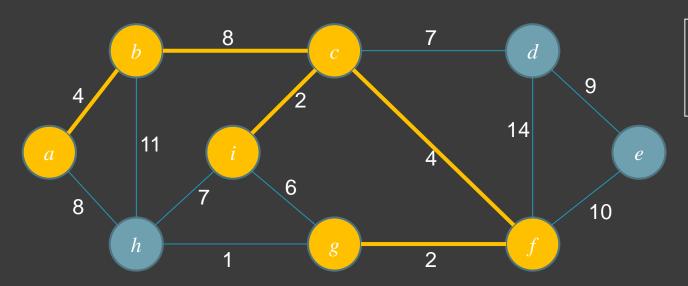
- The lowest-weight edge connecting Y to (V-Y) is (c, f)



$$Y = \{a, b, c, f, i\}$$

 $F = \{(a, b), (b, c), (c, i), (c, f)\}$

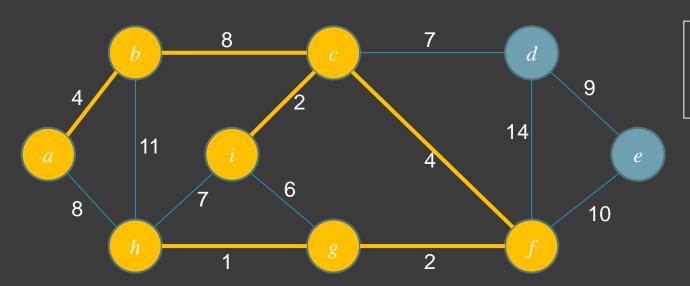
- The lowest-weight edge connecting Y to (V-Y) is (f, g)



$$Y = \{a, b, c, f, g, i\}$$

$$F = \{(a, b), (b, c), (c, i), (c, f), (f, g)\}$$

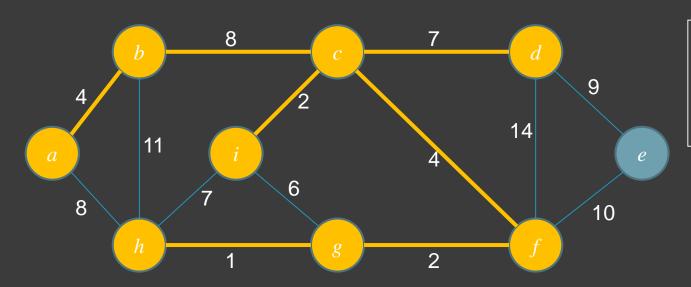
- The lowest-weight edge connecting Y to (V-Y) is (g, h)
- ullet Include this edge in F, and include h in Y.



$$Y = \{a, b, c, f, g, h, i\}$$

$$F = \{(a, b), (b, c), (c, i), (c, f), (f, g), (g, h)\}$$

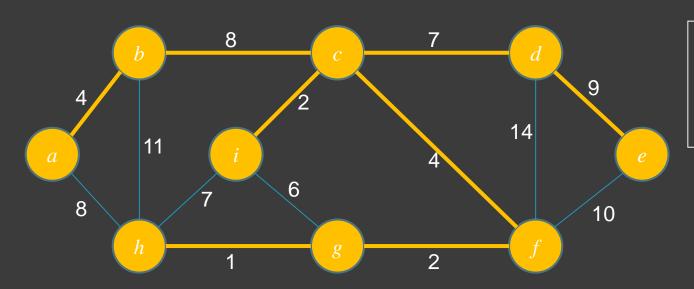
- The lowest-weight edge connecting Y to (V-Y) is (c, d)



$$Y = \{a, b, c, d, f, g, h, i\}$$

$$F = \{(a, b), (b, c), (c, i), (c, f), (f, g), (g, h), (c, d)\}$$

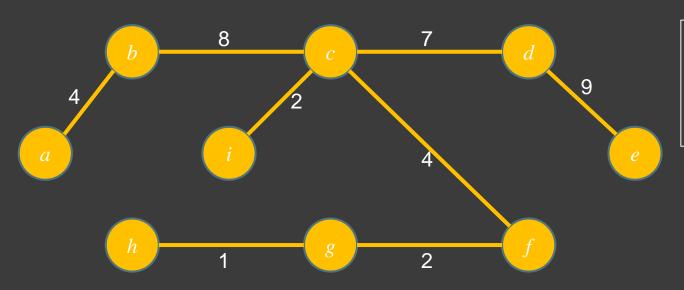
- The lowest-weight edge connecting Y to (V-Y) is (d, e)



$$Y = \{a, b, c, d, e, f, g, h, i\}$$

$$F = \{(a, b), (b, c), (c, i), (c, f), (f, g), (g, h), (c, d), (d, e)\}$$

- Now Y = V (we have added all of the vertices to Y)
- Remove all remaining edges



$$Y = \{a, b, c, d, e, f, g, h, i\}$$

$$F = \{(a, b), (b, c), (c, i), (c, f), (f, g), (g, h), (c, d), (d, e)\}$$

- Now Y = V (we have added all of the vertices to Y)
- Remove all remaining edges
- MST Weight = 4 + 8 + 7 + 9 + 2 + 4 + 1 + 2 = 37

- Depends on how the priority queue is implemented. A MIN-HEAP is a good way, but a FIBONACCI-HEAP (Chapter 19) is even better.

End of Chapter 23

? Questions?