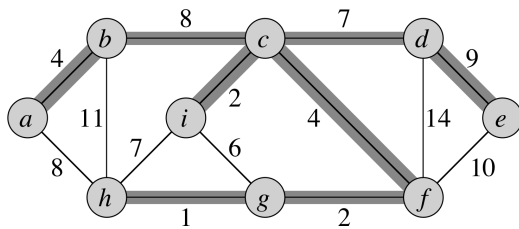


Computer Science & Engineering 423/823  
 Design and Analysis of Algorithms  
 Lecture 06 — Minimum-Weight Spanning Trees (Chapter 23)

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## MST Example



## MST-Kruskal( $G, w$ )

```

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

## Introduction

- ▶ Given a connected, undirected graph  $G = (V, E)$ , a **spanning tree** is an acyclic subset  $T \subseteq E$  that connects all vertices in  $V$ 
  - ▶  $T$  acyclic  $\Rightarrow$  a tree
  - ▶  $T$  connects all vertices  $\Rightarrow$  **spans**  $G$
- ▶ If  $G$  is weighted, then  $T$ 's weight is  $w(T) = \sum_{(u,v) \in T} w(u, v)$
- ▶ A **minimum weight spanning tree** (or **minimum spanning tree**, or **MST**) is a spanning tree of minimum weight
  - ▶ Not necessarily unique
- ▶ Applications: anything where one needs to connect all nodes with minimum cost, e.g., wires on a circuit board or fiber cable in a network

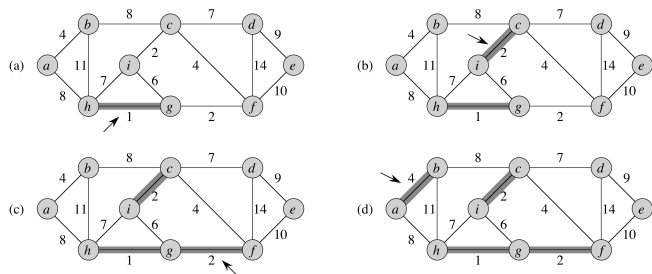
## Kruskal's Algorithm

- ▶ Greedy algorithm: Make the locally best choice at each step
- ▶ Starts by declaring each vertex to be its own tree (so all nodes together make a forest)
- ▶ Iteratively identify the minimum-weight edge  $(u, v)$  that connects two distinct trees, and add it to the MST  $T$ , merging  $u$ 's tree with  $v$ 's tree

## More on Kruskal's Algorithm

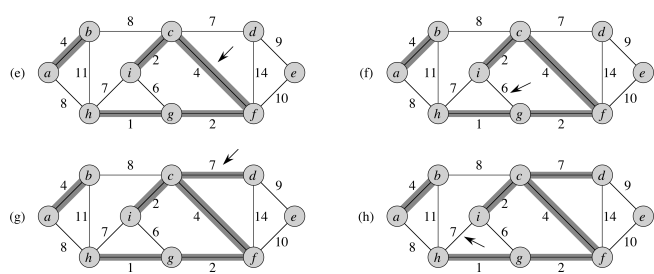
- ▶ FIND-SET( $u$ ) returns a representative element from the set (tree) that contains  $u$
- ▶ UNION( $u, v$ ) combines  $u$ 's tree to  $v$ 's tree
- ▶ These functions are based on the **disjoint-set data structure**
- ▶ More on this later

## Example (1)



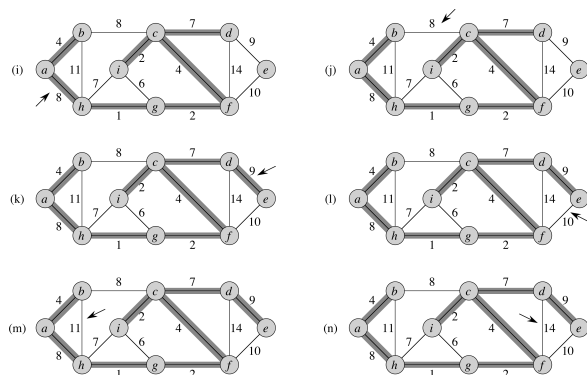
7/21

## Example (2)



8/21

## Example (3)



9/21

## Disjoint-Set Data Structure

- ▶ Given a **universe**  $U = \{x_1, \dots, x_n\}$  of elements (e.g., the vertices in a graph  $G$ ), a DSDS maintains a collection  $\mathcal{S} = \{S_1, \dots, S_k\}$  of disjoint sets of elements such that
  - ▶ Each element  $x_i$  is in exactly one set  $S_j$
  - ▶ No set  $S_j$  is empty
- ▶ Membership in sets is dynamic (changes as program progresses)
- ▶ Each set  $S \in \mathcal{S}$  has a **representative element**  $x \in S$
- ▶ Chapter 21

10/21

## Disjoint-Set Data Structure (2)

- ▶ DSDS implementations support the following functions:
  - ▶ **MAKE-SET**( $x$ ) takes element  $x$  and creates new set  $\{x\}$ ; returns pointer to  $x$  as set's representative
  - ▶ **UNION**( $x, y$ ) takes  $x$ 's set ( $S_x$ ) and  $y$ 's set ( $S_y$ , assumed disjoint from  $S_x$ ), merges them, destroys  $S_x$  and  $S_y$ , and returns representative for new set from  $S_x \cup S_y$
  - ▶ **FIND-SET**( $x$ ) returns a pointer to the representative of the unique set that contains  $x$
- ▶ Section 21.3: can perform  $d$  D-S operations on  $e$  elements in time  $O(d \alpha(e))$ , where  $\alpha(e) = o(\lg^* e) = o(\log e)$  is very slowly growing:

$$\alpha(e) = \begin{cases} 0 & \text{if } 0 \leq e \leq 2 \\ 1 & \text{if } e = 3 \\ 2 & \text{if } 4 \leq e \leq 7 \\ 3 & \text{if } 8 \leq e \leq 2047 \\ 4 & \text{if } 2048 \leq e \leq 2^{2048} (\gg 10^{600}) \end{cases}$$

11/21

## Analysis of Kruskal's Algorithm

- ▶ Sorting edges takes time  $O(|E| \log |E|)$
- ▶ Number of disjoint-set operations is  $O(|V| + |E|)$  on  $O(|V|)$  elements, which can be done in time  $O((|V| + |E|) \alpha(|V|)) = O(|E| \alpha(|V|))$  since  $|E| \geq |V| - 1$
- ▶ Since  $\alpha(|V|) = o(\log |V|) = O(\log |E|)$ , we get total time of  $O(|E| \log |E|) = O(|E| \log |V|)$  since  $\log |E| = O(\log |V|)$

12/21

## Prim's Algorithm

- Greedy algorithm, like Kruskal's
- In contrast to Kruskal's, Prim's algorithm maintains a single tree rather than a forest
- Starts with an arbitrary tree root  $r$
- Repeatedly finds a minimum-weight edge that is incident to a node not yet in tree

## MST-Prim( $G, w, r$ )

```

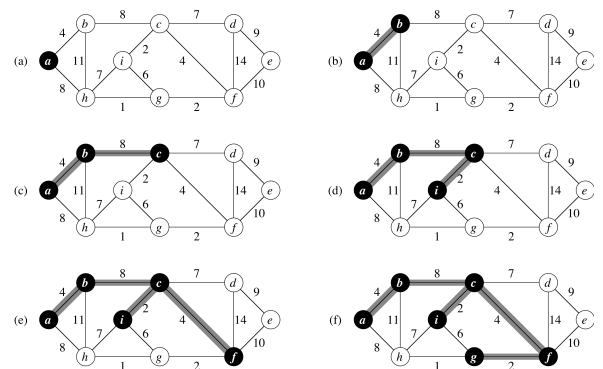
1   $A = \emptyset$ 
2  for each vertex  $v \in V$  do
3     $key[v] = \infty$ 
4     $\pi[v] = \text{NIL}$ 
5  end
6   $key[r] = 0$ 
7   $Q = V$ 
8  while  $Q \neq \emptyset$  do
9     $u = \text{EXTRACT-MIN}(Q)$ 
10   for each  $v \in \text{Adj}[u]$  do
11     if  $v \in Q$  and  $w(u, v) < key[v]$  then
12        $\pi[v] = u$ 
13        $key[v] = w(u, v)$ 
14     end
15   end
16 end

```

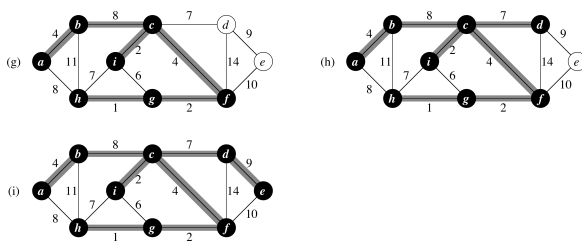
## More on Prim's Algorithm

- $key[v]$  is the weight of the minimum weight edge from  $v$  to any node already in MST
- EXTRACT-MIN uses a **minimum heap** (minimum priority queue) data structure
  - Binary tree where the key at each node is  $\leq$  keys of its children
  - Thus minimum value always at top
  - Any subtree is also a heap
  - Height of tree is  $\Theta(\log n)$
  - Can build heap on  $n$  elements in  $O(n)$  time
  - After returning the minimum, can filter new minimum to top in time  $O(\log n)$
  - Based on Chapter 6

## Example (1)



## Example (2)

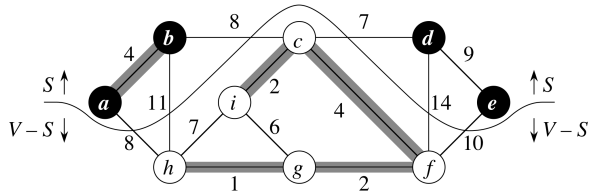


## Analysis of Prim's Algorithm

- **Invariant:** Prior to each iteration of the while loop:
  - Nodes already in MST are exactly those in  $V \setminus Q$
  - For all vertices  $v \in Q$ , if  $\pi[v] \neq \text{NIL}$ , then  $key[v] < \infty$  and  $key[v]$  is the weight of the lightest edge that connects  $v$  to a node already in the tree
- Time complexity:
  - Building heap takes time  $O(|V|)$
  - Make  $|V|$  calls to EXTRACT-MIN, each taking time  $O(\log |V|)$
  - For loop iterates  $O(|E|)$  times
    - In for loop, need constant time to check for queue membership and  $O(\log |V|)$  time for decreasing  $v$ 's key and updating heap
  - Yields total time of  $O(|V| \log |V| + |E| \log |V|) = O(|E| \log |V|)$
  - Can decrease total time to  $O(|E| + |V| \log |V|)$  using Fibonacci heaps

## Proof of Correctness of Both Algorithms

- Both algorithms use greedy approach for optimality
- Maintain **invariant** that at any time, set of edges  $A$  selected so far is subset of some MST
  - $\Rightarrow$  Optimal substructure property
- Each iteration of each algorithm looks for a **safe edge**  $e$  such that  $A \cup \{e\}$  is also a subset of an MST
  - $\Rightarrow$  Greedy choice
- Prove invariant via use of **cut**  $(S, V - S)$  that **respects**  $A$  (no edges span cut)



19/21

## Proof of Correctness of Both Algorithms (2)

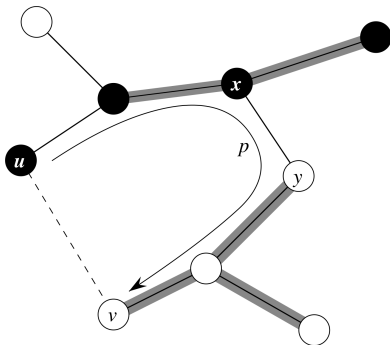
- Theorem:** Let  $A \subseteq E$  be included in some MST of  $G$ ,  $(S, V - S)$  be a cut respecting  $A$ , and  $(u, v) \in E$  be a minimum-weight edge crossing cut. Then  $(u, v)$  is a safe edge for  $A$ .
- Proof:**
  - Let  $T$  be an MST including  $A$  and not including  $(u, v)$
  - Let  $p$  be path from  $u$  to  $v$  in  $T$ , and  $(x, y)$  be edge from  $p$  crossing cut ( $\Rightarrow$  not in  $A$ )
  - Since  $T$  is a spanning tree, so is  $T' = T - \{(x, y)\} \cup \{(u, v)\}$
  - Both  $(u, v)$  and  $(x, y)$  cross cut, so  $w(u, v) \leq w(x, y)$
  - So,  $w(T') = w(T) - w(x, y) + w(u, v) \leq w(T)$
  - $\Rightarrow T'$  is MST
  - $\Rightarrow (u, v)$  safe for  $A$  since  $A \cup \{(u, v)\} \subseteq T'$

□

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20/21

## Proof of Correctness of Both Algorithms (3)



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21/21