

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

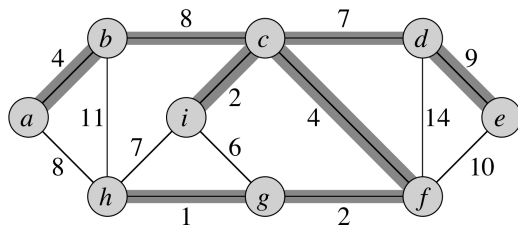
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Introduction

- ▶ 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

MST Example



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

MST-Kruskal(G, w)

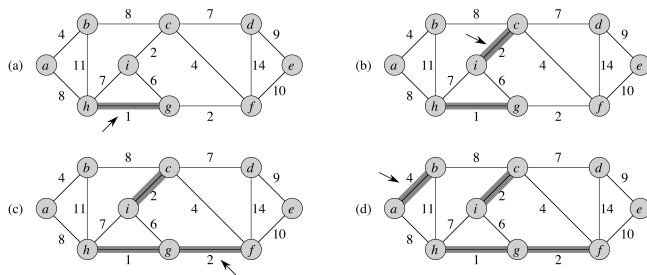
```

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    |     |  $A = A \cup \{(u, v)\}$ 
9    |     | UNION( $u, v$ )
10   | end
11 return  $A$ 
    
```

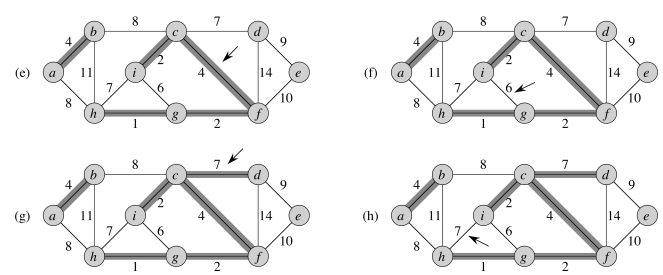
Notes and Questions

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

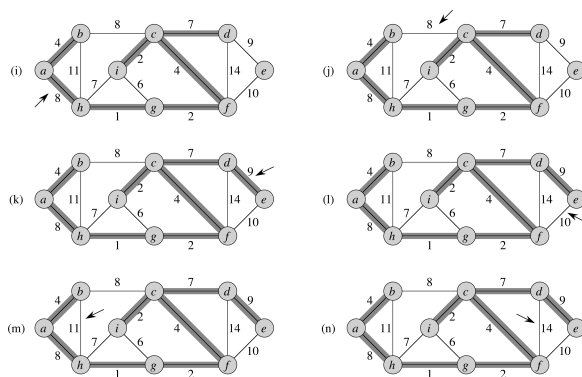
Example (1)



Example (2)



Example (3)



Notes and Questions

- ▶ 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

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 **inverse Ackerman's** $\alpha(e) = o(\lg^* e) = o(\log \log e)$ is **very** slowly growing:

$$\alpha(e) = \begin{cases} 0 & \text{if } 0 \leq e \leq 2 \\ 1 & \text{if } e = 3 \\ 2 & \text{if } 4 \leq e \leq 7 \\ 3 & \text{if } 8 \leq e \leq 2047 \\ 4 & \text{if } 2048 \leq e \leq 2^{2048} (\gg 10^{600}) \end{cases} \quad \lg^*(e) = \begin{cases} 0 & \text{if } e \leq 1 \\ 1 & \text{if } 1 < e \leq 2 \\ 2 & \text{if } 2 < e \leq 4 \\ 3 & \text{if } 4 < e \leq 16 \\ 4 & \text{if } 16 < e \leq 65536 \\ 5 & \text{if } 65536 < e \leq 2^{65536} \end{cases}$$

Analysis of Kruskal's Algorithm

Notes and Questions

- ▶ 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 $\log |E| = O(\log |V|)$

Prim's Algorithm

Notes and Questions

- ▶ 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)

Notes and Questions

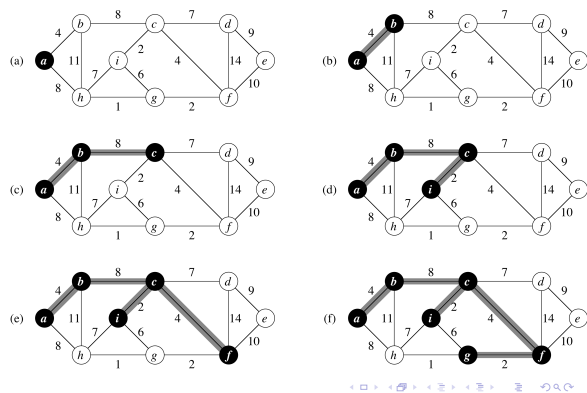
```
1  A = {}
2  for each vertex v in V do
3    key[v] = ∞
4    π[v] = NIL
5  end
6  key[r] = 0
7  Q = V
8  while Q ≠ {} do
9    u = EXTRACT-MIN(Q)
10   for each v in Adj[u] do
11     if v in Q and w(u, v) < key[v] then
12       π[v] = u
13       key[v] = w(u, v)
14     end
15   end
16 end
```

More on Prim's Algorithm

Notes and Questions

- ▶ $key[v]$ 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)



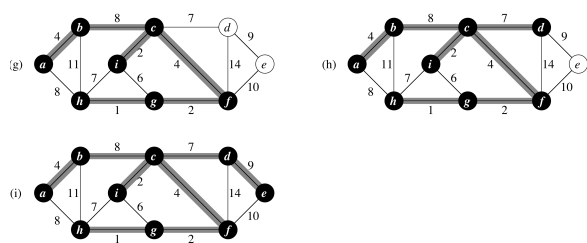
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Notes and Questions

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Example (2)



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Analysis of Prim's Algorithm

- **Invariant:** Prior to each iteration of the while loop:
 1. Nodes already in MST are exactly those in $V \setminus Q$
 2. For all vertices $v \in Q$, if $\pi[v] \neq \text{NIL}$, then $\text{key}[v] < \infty$ and $\text{key}[v]$ 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

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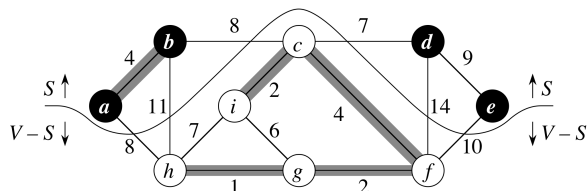
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Proof of Correctness of Both Algorithms

- 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)



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Notes and Questions

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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 $A \cup \{(u, v)\} \subseteq T'$

□

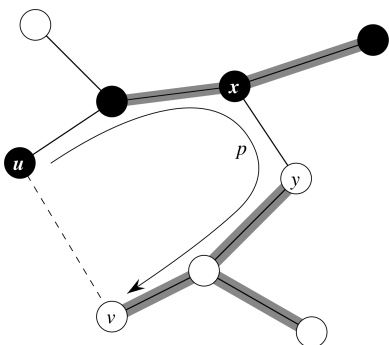
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Proof of Correctness of Both Algorithms (3)



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Notes and Questions

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