Computer Science & Engineering 423/823 Design and Analysis of Algorithms

Lecture 04 — Minimum-Weight Spanning Trees (Chapter 23)

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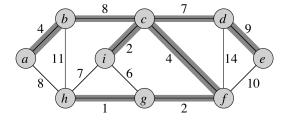
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Introduction

- Given a connected, undirected graph G = (V, E), a **spanning tree** is an acyclic subset $T \subseteq E$ that connects all vertices in V
 - ightharpoonup T acyclic \Rightarrow a tree
 - ➤ T connects all vertices ⇒ 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

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MST Example



(B) (B) (E) (E) (E) (D)

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

4 D > 4 B > 4 E > 4 E > E 9 9 C

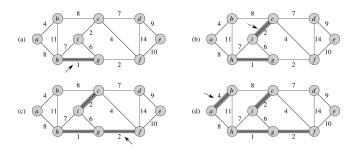
MST-Kruskal(G, w)

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 w6 for each edge $(u, v) \in E$, taken in nondecreasing order do 7 | if Find-Set(u) \neq Find-Set(v) then 8 | $A = A \cup \{(u, v)\}$ 9 | Union(u, v) 10 | 11 end 12 return A

More on Kruskal's Algorithm

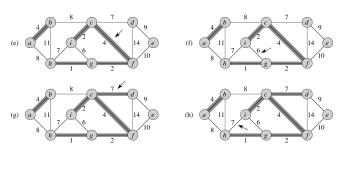
- \blacktriangleright ${\rm FIND\text{-}SET}(u)$ returns a representative element from the set (tree) that contains u
- ▶ UNION(u, v) combines u's tree to v's tree
- ▶ These functions are based on the disjoint-set data structure
- ▶ More on this later

Example (1)



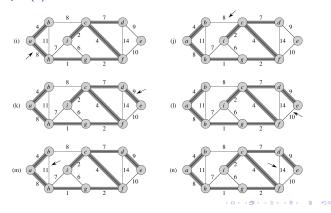
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Example (2)



4 m >

Example (3)



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 $S = \{S_1, \dots, S_k\}$ of disjoint sets of elements such that
 - ▶ Each element x_i is in exactly one set S_i
 - ▶ No set S_j is empty
- ► Membership in sets is dynamic (changes as program progresses)
- ▶ Each set $S \in S$ 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 $S_y \cup S_y$
 - ► 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:

$$\alpha(e) = \begin{cases} 0 & \text{if } 0 \le e \le 2\\ 1 & \text{if } e = 3\\ 2 & \text{if } 4 \le e \le 7\\ 3 & \text{if } 8 \le e \le 2047\\ 4 & \text{if } 2048 \le e \le 16^{512} \end{cases}$$

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

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

40 × 40 × 42 × 42 × 2 990

MST-Prim(G, w, r)

```
1 A = ∅
2 for each vertex v \in V do
         key[v] = \infty
        \pi[v] = \text{NIL}
5 end
6 key[r] = 0
7 Q = V
    while Q \neq \emptyset do
         u = \text{Extract-Min}(Q)
         for each v \in Adj[u] do

if v \in Q and w(u, v) < key[v] then

\pi[v] = u
10
11
12
13
                     key[v] = w(u, v)
14
15
16 end
```

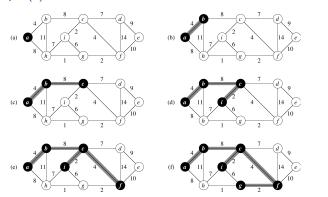
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More on Prim's Algorithm

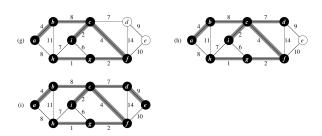
- key[v] is the weight of the minimum weight edge from v to any node already in MST
- EXTRACT-MIN uses a minimum heap (minimum priority queue) data structure
 - $\,\blacktriangleright\,$ Binary tree where the key at each node is \le keys of its children
 - Thus minimum value always at top
 - Any subtree is also a heap
 - ► Height of tree is [lg 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:
 - 1. Nodes already in MST are exactly those in $V \setminus Q$
 - 2. For all vertices $v \in Q$, if $\pi[v] \neq \mathrm{NIL}$, 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