

Announcements

- HW1 grading will be out around this weekend
 - Will also post the reference solution

CS6161: Design and Analysis of Algorithms (Fall 2020)

Minimum Spanning Trees

Instructor: Haifeng Xu

Outline

- MSTs and Accelerating Kruskal's Algorithm
- Prim's Algorithm and Its Acceleration

Minimum-Spanning Trees (MSTs)

Spanning Trees

A **spanning tree** of an undirected connected graph $G = (V, E)$ is a set of edges $T \subseteq E$ such that

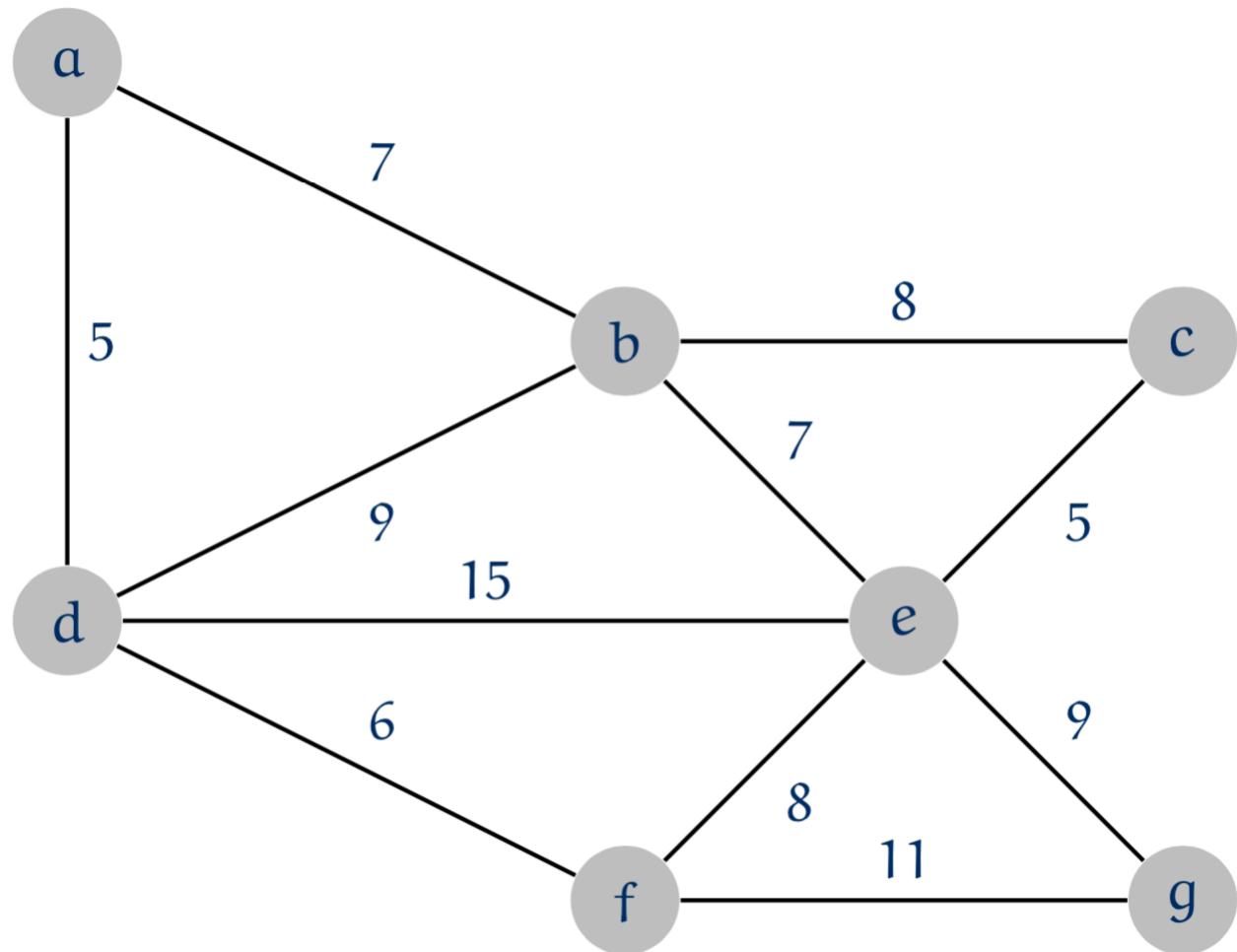
- T forms a tree, and
- (V, T) is connected.

In a weighted graph G , a **minimum spanning tree** is a spanning tree with smallest sum of edge weights

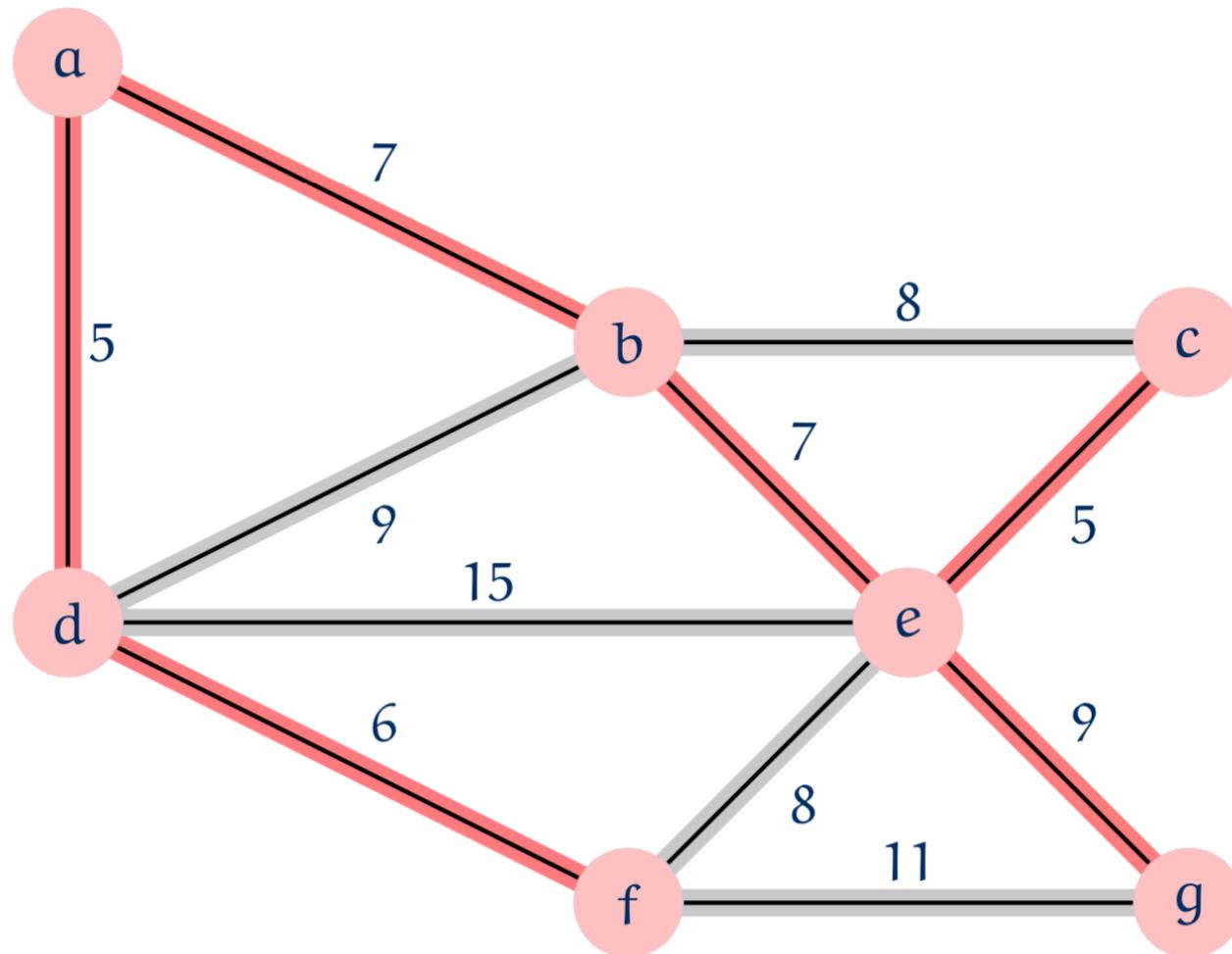
The MST problem

Find a minimum spanning tree of a weighted graph G

MST:An Example



MST:An Example



Kruskal's MST Algorithm

- Pick edges in increasing order of their costs, and add edges to T as long as they do not form a cycle

```
KruskalMST( $G$ )
```

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$; // T will store edges of MST
- 3 **while** T is not a spanning tree of G
- 4
- 5
- 6
- 7
- 8 **return** T

Kruskal's MST Algorithm

- Pick edges in increasing order of their costs, and add edges to T as long as they do not form a cycle

```
KruskalMST( $G$ )
```

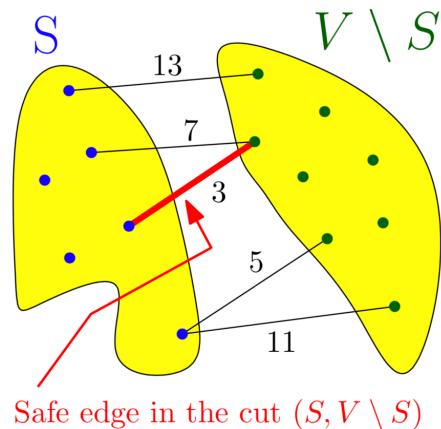
- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$; // T will store edges of MST
- 3 **while** T is not a spanning tree of G
 - 4 choose $e \in E$ of minimum cost;
 - 5 remove e from E ;
 - 6 if $T \cup \{e\}$ does not contain cycles **then**
 - 7 add e to T ;
- 8 **return** T

Correctness

- Relies on a key property of MSTs

Theorem: Given an undirected **connected** graph G with **distinct** edge costs, the set of its safe edges form the **unique** MST of G .

Recall: an edge e is a **safe** edge if there **exists** a cut $(S, V \setminus S)$ such that e is the **unique minimum cost edge** crossing this cut.

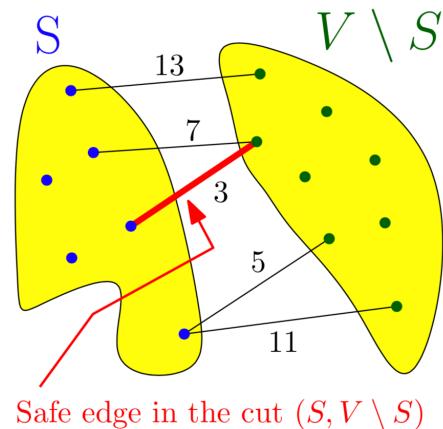


Correctness

- Relies on a key property of MSTs

Theorem: Given an undirected **connected** graph G with **distinct** edge costs, the set of its safe edges form the **unique** MST of G .

Proof idea: all edges added by Kruskal's algorithm are safe



Running Time

1. Sorting takes $O(m \log m)$, where $m = |E|$ times (note $m > n$)
2. While loop:
 - Executed for $m = |E|$ times
 - Step 6 takes $n = |V|$ timeIn total $O(mn)$ time

KruskalMST(G)

```
1  Sort edges in increasing order of costs
2  Initialize  $T = \emptyset$ ; //  $T$  will store edges of MST
3  while  $T$  is not a spanning tree of  $G$ 
    4      choose  $e \in E$  of minimum cost;
    5      remove  $e$  from  $E$ ;
    6      if  $T \cup \{e\}$  does not contain cycles then
        7          add  $e$  to  $T$ ;
8  return  $T$ 
```

Q: Can we do better, i.e., be faster?

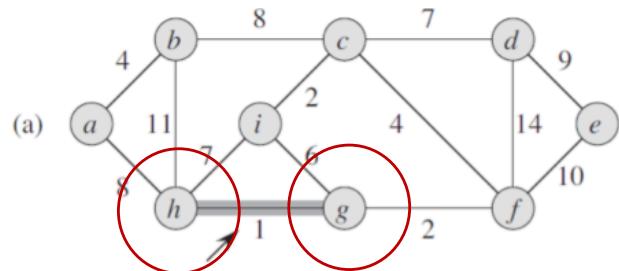
Yes, the same algorithm but more efficient implementation!

KruskalMST(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$; // T will store edges of MST
- 3 **while** T is not a spanning tree of G
 - 4 choose $e \in E$ of minimum cost;
 - 5 remove e from E ;
 - 6 if $T \cup \{e\}$ does not contain cycles **then**
 - 7 add e to T ;
 - 8 **return** T

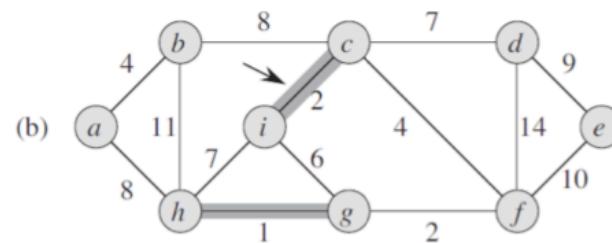
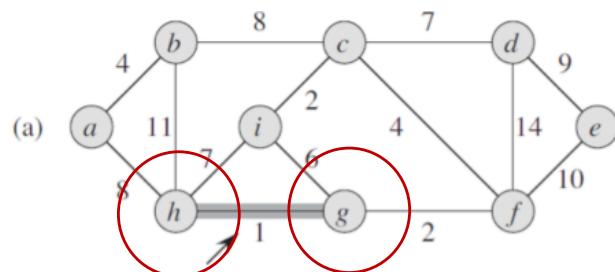
An Observation about Kruskal's Algorithm

- Each added edge “joins” two components



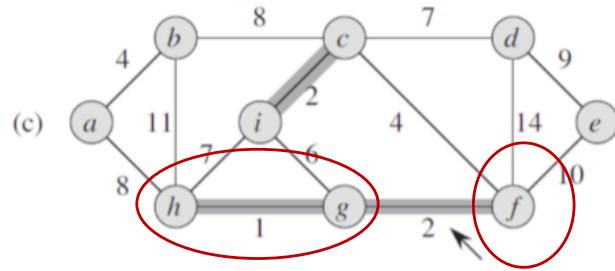
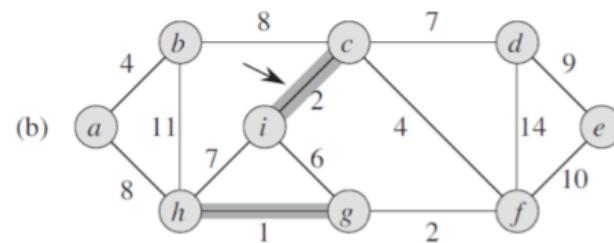
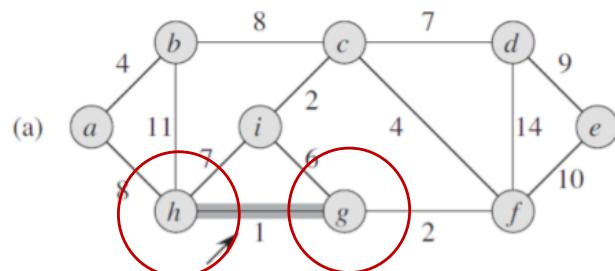
An Observation about Kruskal's Algorithm

- Each added edge “joins” two components



An Observation about Kruskal's Algorithm

- Each added edge “joins” two components



More Efficient Implementation

```
SmarterKruskal( $G$ )
```

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$; // T will store edges of MST
- 3
- 4
- 5
- 6
- 7
- 8 **return** T

More Efficient Implementation

```
SmarterKruskal( $G$ )
```

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$; *// T will store edges of MST*
- 3 Put each vertex $u \in V$ into a set by itself
- 4
- 5
- 6
- 7
- 8 **return** T

More Efficient Implementation

```
SmarterKruskal( $G$ )
```

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$; *// T will store edges of MST*
- 3 Put each vertex $u \in V$ into a set by itself
- 4 **foreach** $e = (u, v) \in E$ in sorted order **do**
- 5
- 6
- 7
- 8 **return** T

More Efficient Implementation

```
SmarterKruskal( $G$ )
```

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$; // T will store edges of MST
- 3 Put each vertex $u \in V$ into a set by itself
- 4 **foreach** $e = (u, v) \in E$ in sorted order **do**
- 5 **if** u and v belong to different sets **then**
- 6 add e to T ;
- 7 merge the two sets containing u and v ;
- 8 **return** T

- Need an efficient data structure to
 - Check if two elements belong to same set
 - Merge two sets

Data Structure: Union-Find

Union-Find

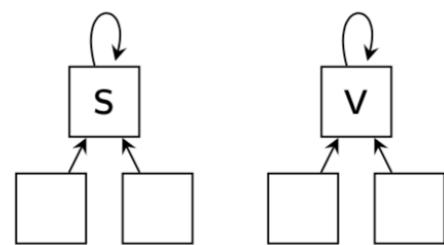
Store a set of disjoint sets with the following operations:

- $\text{Make-Set}(V)$: generate a set $\{\nu\}$ for each $\nu \in V$. Name of set $\{\nu\}$ is ν
- $\text{Find}(u)$: find the name of the set containing vertex u
- $\text{Union}(u, v)$: merge the sets named u and v . Name of the new set is either u or v .

➤ The running time of Kruskal algorithm will depend on the implementation of this data structure.

Union-Find: Implementation

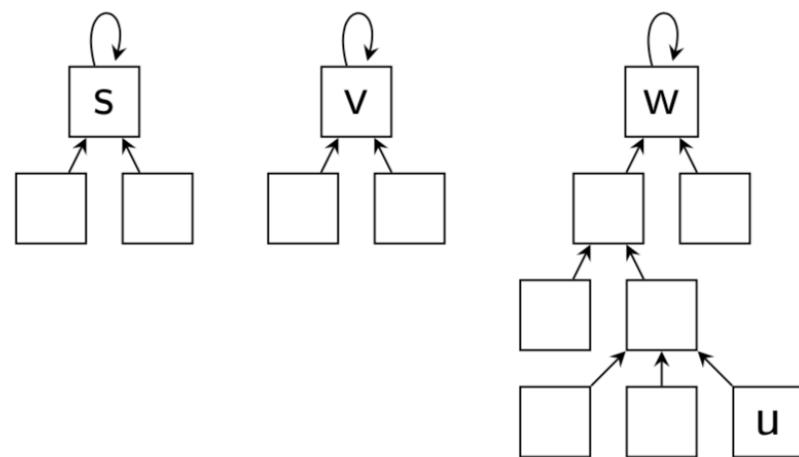
- Sets are represented as trees, by pointers towards roots. All elements in one tree belong to a set with root's name



Figure

Union-Find: Implementation

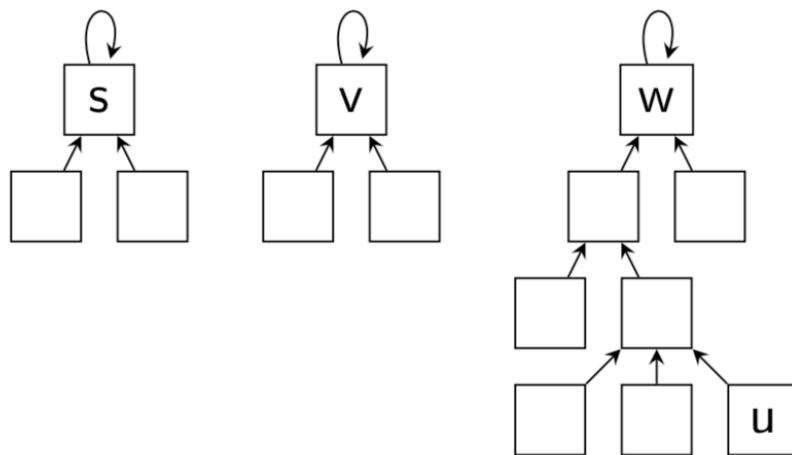
- Sets are represented as trees, by pointers towards roots. All elements in one tree belong to a set with root's name.
 - $\text{Find}(u)$: Traverse from u to the root



Figure

Union-Find: Implementation

- Sets are represented as trees, by pointers towards roots. All elements in one tree belong to a set with root's name.
 - $\text{Find}(u)$: Traverse from u to the root
 - $\text{Union}(u, v)$: Make root of u (smaller set) point to root of v . Takes $O(1)$ time



Figure

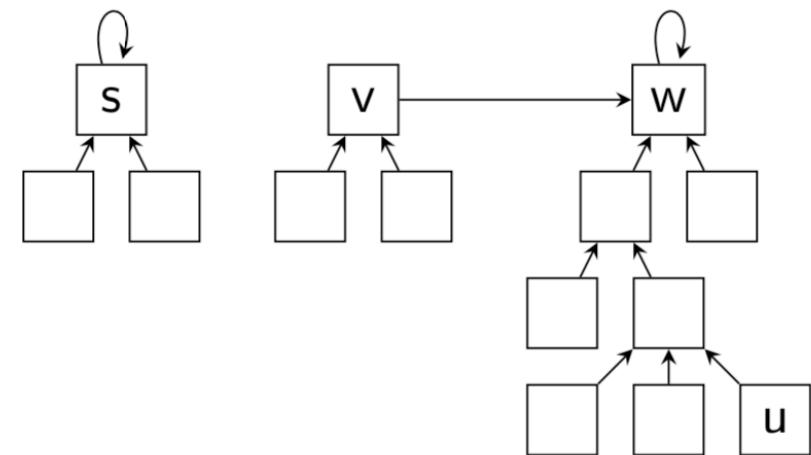
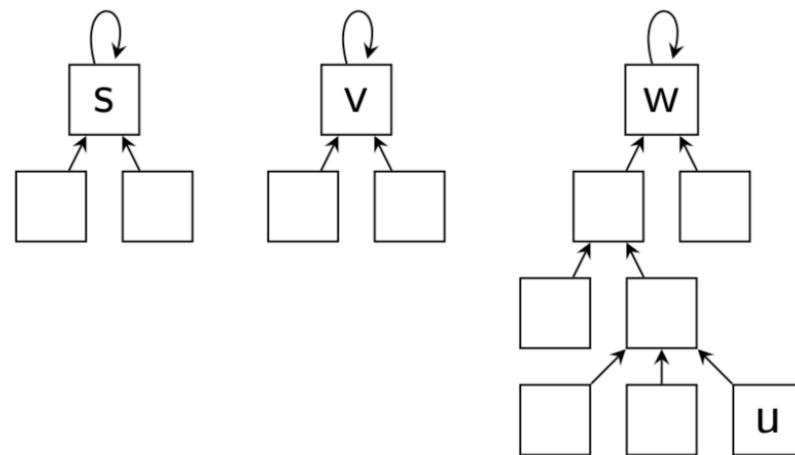


Figure: $\text{Union}(\text{Find}(v), \text{Find}(u))$

Union-Find: Implementation

- Sets are represented as trees, by pointers towards roots. All elements in one tree belong to a set with root's name.
 - $\text{Find}(u)$: Traverse from u to the root
 - $\text{Union}(u, v)$: Make root of u (smaller set) point to root of v . Takes $O(1)$ time.
- Each vertex u has a pointer $\text{parent}(u)$ to its parent



Figure

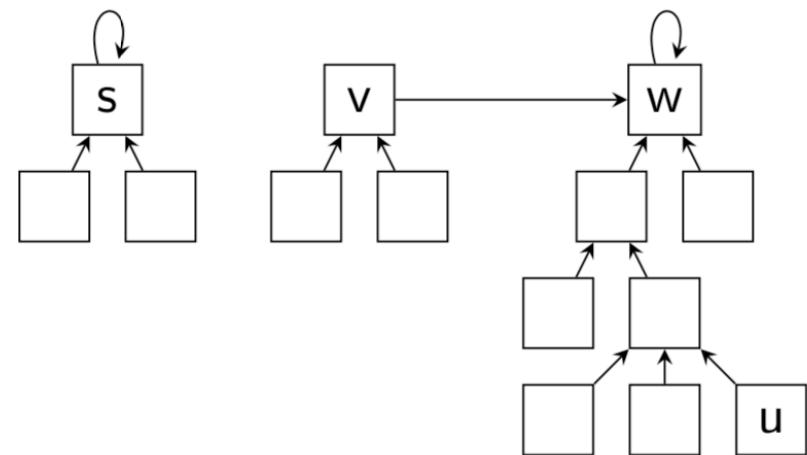


Figure: $\text{Union}(\text{Find}(v), \text{Find}(u))$

Union-Find: Implementation (FYI)

Algorithm: Make-Set(G):

```
foreach  $u \in V$  do  
    parent( $u$ ) =  $u$ ;
```

Algorithm: Find(u):

```
while parent( $u$ )  $\neq u$  do  
     $u$  = parent( $u$ );  
return  $u$ 
```

Algorithm: Union(u, v):

```
(* parent( $u$ ) =  $u$  & parent( $v$ ) =  $v$  *)  
if |component( $u$ )|  $\leq$  |component( $v$ )| then  
    parent( $u$ ) =  $v$   
else  
    parent( $v$ ) =  $u$   
set new component size to |component( $u$ ) + component( $v$ )|.
```

Union-Find: Time Analysis

- $\text{Make-Set}(V)$: $O(n)$
- $\text{Union}(u, v)$: $O(1)$ time
- $\text{Find}(u)$: in $O(\text{depth of the tree})$ time

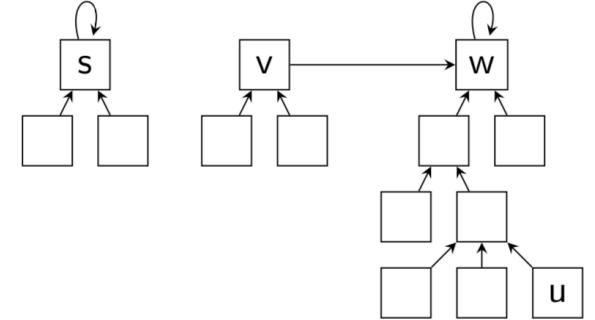


Figure: $\text{Union}(\text{Find}(v), \text{Find}(u))$

Fact: The maximum depth of trees in union-find is $O(\log n)$.

Proof:

- $\text{Depth}(v)$ increase by at most 1 only when the set containing v changes its name
- If $\text{Depth}(v)$ increases, then size of the set containing v at least doubles
- Maximum set size is n ; so depth of any node is at most $O(\log n)$.

Union-Find: Time Analysis

- $\text{Make-Set}(V)$: $O(n)$
- $\text{Union}(u, v)$: $O(1)$ time
- $\text{Find}(u)$: in $O(\log n)$ time

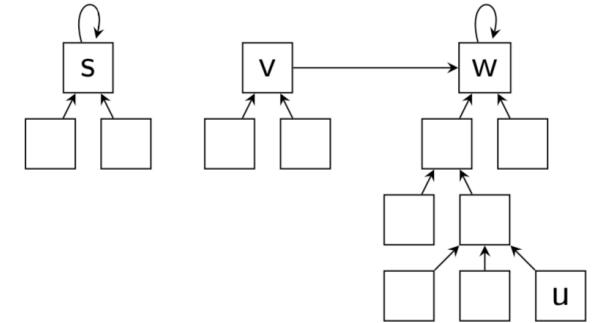


Figure: Union(Find(v), Find(u))

Fact: The maximum depth of trees in union-find is $O(\log n)$.

Proof:

- $\text{Depth}(v)$ increase by at most 1 only when the set containing v changes its name
- If $\text{Depth}(v)$ increases, then size of the set containing v at least doubles
- Maximum set size is n ; so depth of any tree is at most $O(\log n)$.

Running Time Analysis: SmarterKruskal

SmarterKruskal(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Put each vertex $u \in V$ into a set by itself \longrightarrow Make-Set(V): $O(n)$
- 4 **foreach** $e = (u, v) \in E$ in sorted order **do** \longrightarrow repeat m times
- 5 **if** u and v belong to different sets **then**
- 6 add e to T ;
- 7 merge the two sets containing u and v ; \longrightarrow Union(u, v) $O(1)$ time
- 8 **return** T

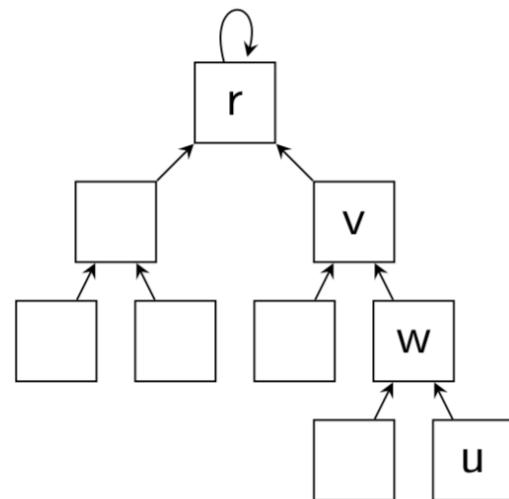
$O(m \log m)$

Find(u), Find(v) and then compare: $O(\log n)$

In total: $O(m \log m) + m \times O(\log n) = O(m \log m)$

Can Even Further Speed up!

- $\text{Find}(u)$ can be done in almost $O(1)$ time
- Main idea: point directly to the root
- Slightly better, but cannot lead to order-wise improvement since sorting is the bottleneck, and takes $O(m \log m)$.



Figure

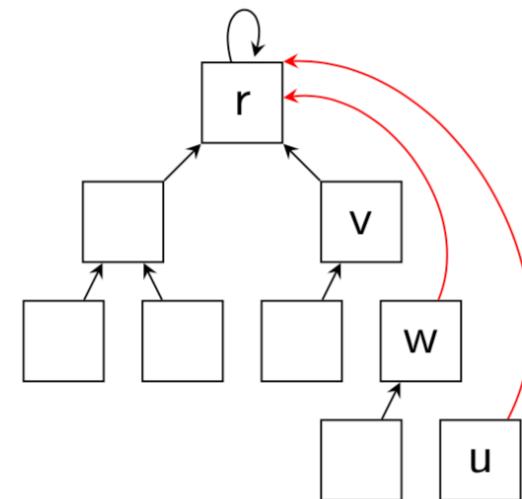


Figure: After $\text{Find}(u)$

Outline

- MSTs and Accelerating Kruskal's Algorithm
- Prim's Algorithm and Its Acceleration

Recall Our Greedy Framework

```
SomeGreedyMSTAlgo( $G$ )
```

- 1 Initialize $T = \emptyset$; // T will store edges of MST
- 2 **while** T is not a spanning tree of G
- 3 choose $e \in E$ that “improves” solution;
- 4 add e to T
- 5 **return** T

Prim's Algorithm uses a different Greedy Procedure to add edges

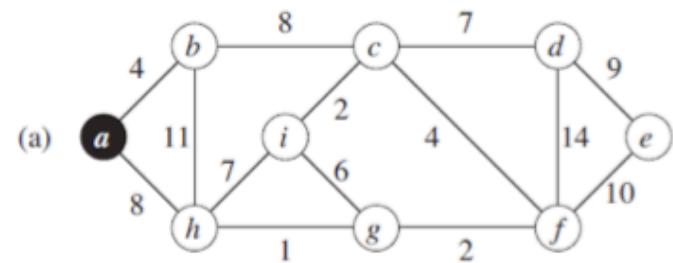
- T maintained by the algorithm will remain a tree
- In each iteration, pick edges with least attachment cost to T .

Prim's Algorithm

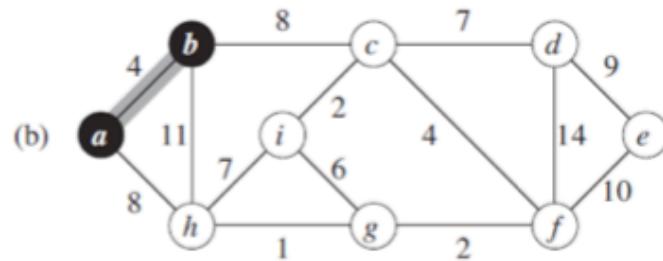
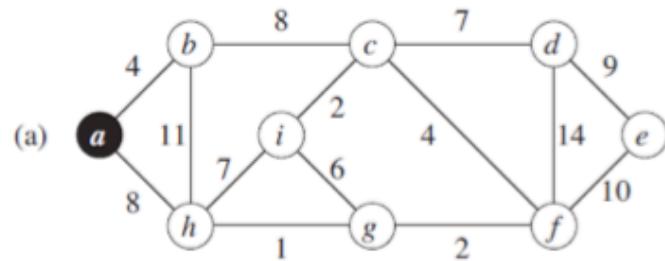
PrimMST(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$; // T stores edges of MST
- 3 Initialize $S = \{1\}$; // S stores nodes of MST
- 4 **while** $|S| < n$ // i.e., T is not a ST
- 5 choose $e = (u, v) \in E$ of minimum cost, such
 that $u \in S$ and $v \in V \setminus S$;
- 6 $T = T \cup \{e\}$;
- 7 $S = S \cup \{v\}$;
- 8 **return** T

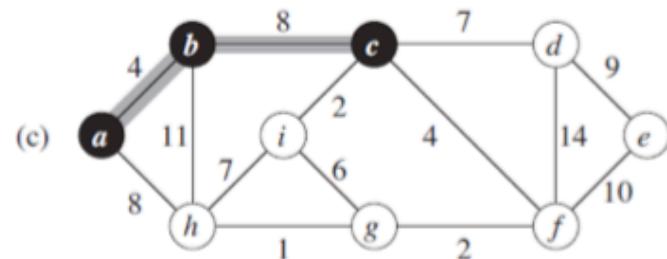
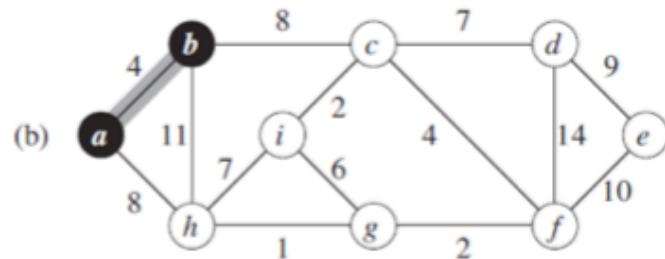
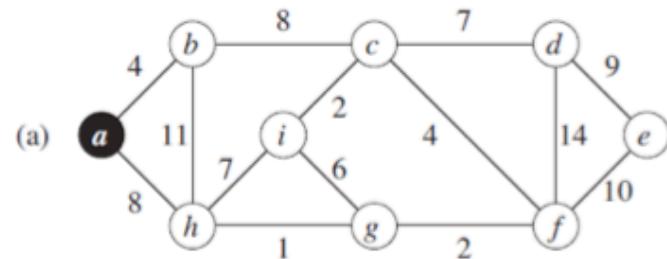
Prim's Algorithm: Example Execution



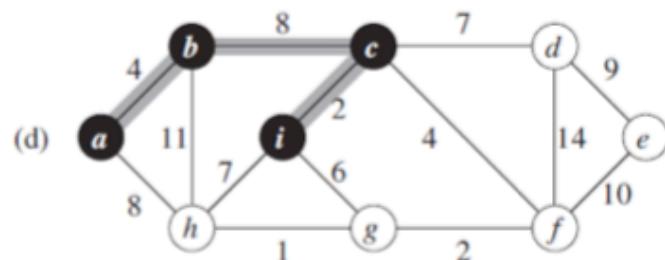
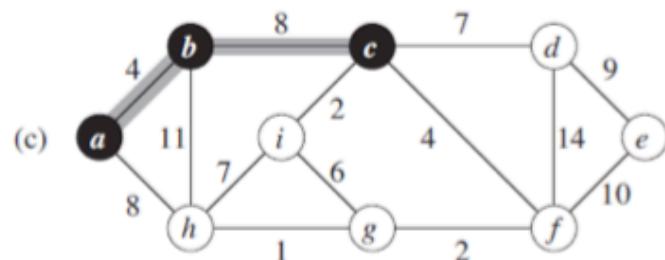
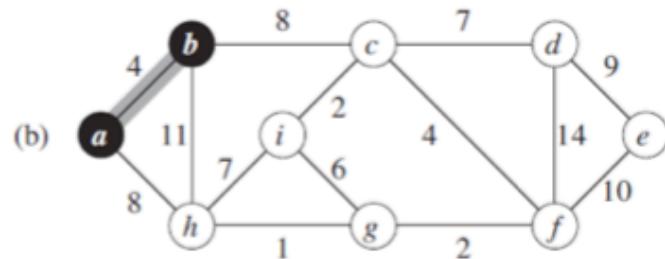
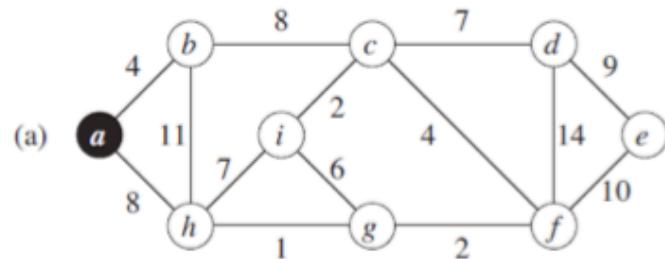
Prim's Algorithm: Example Execution



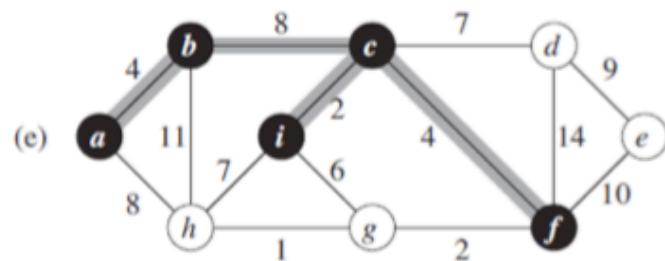
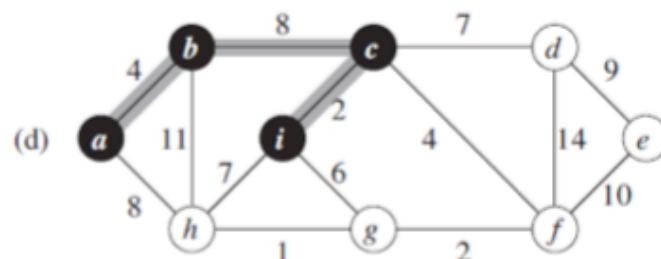
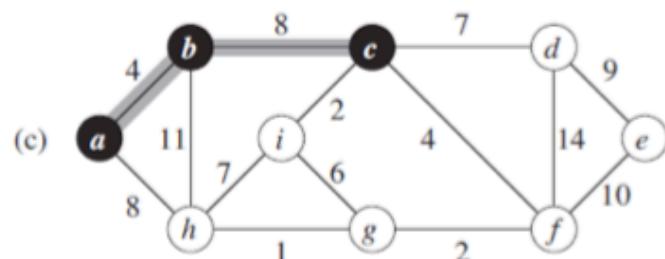
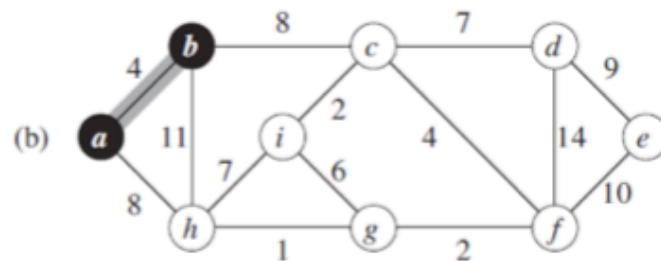
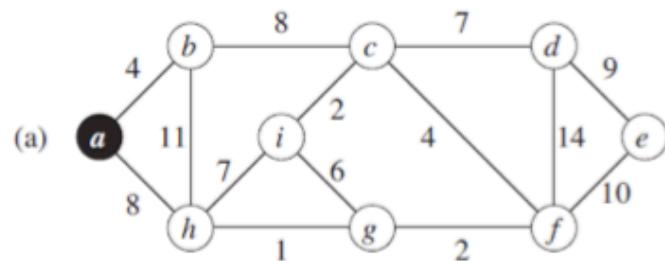
Prim's Algorithm: Example Execution



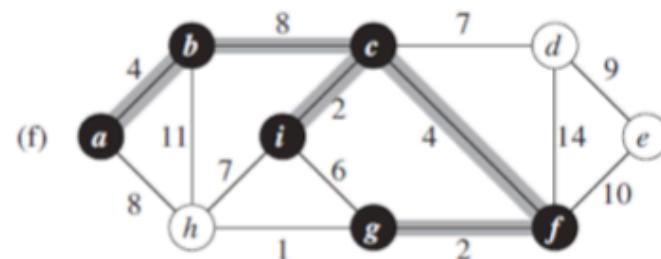
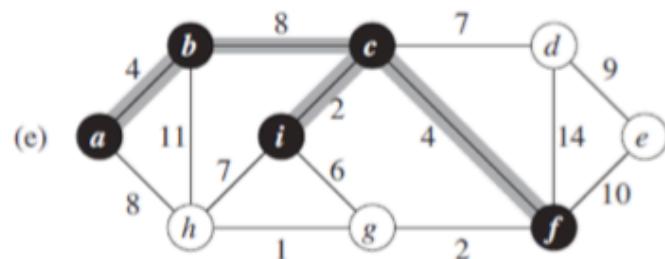
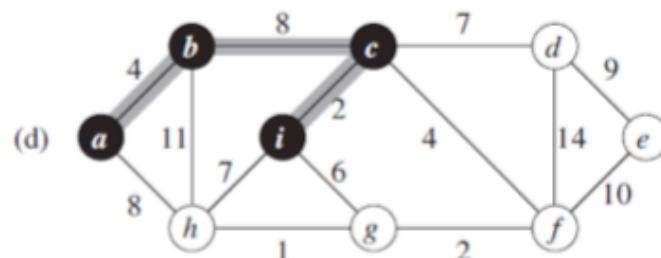
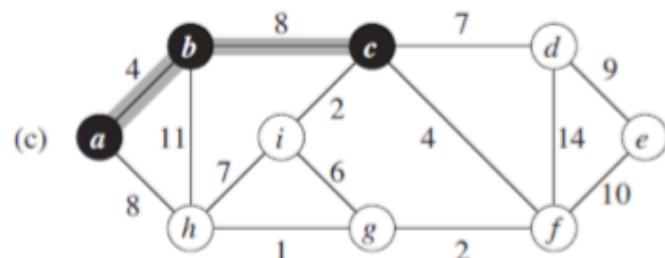
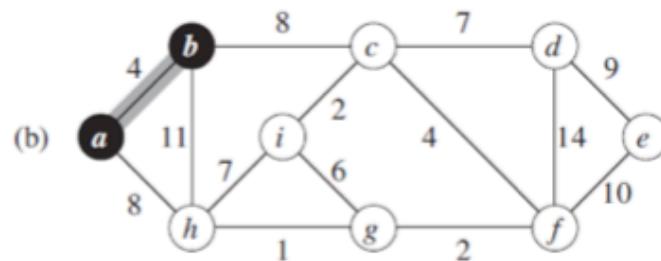
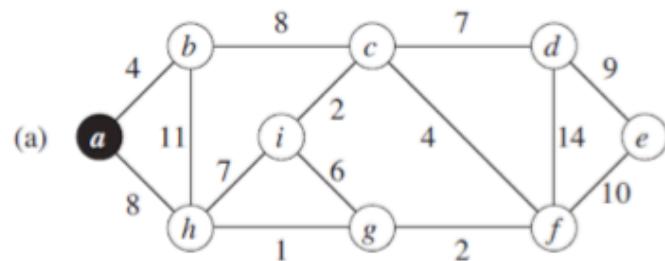
Prim's Algorithm: Example Execution



Prim's Algorithm: Example Execution

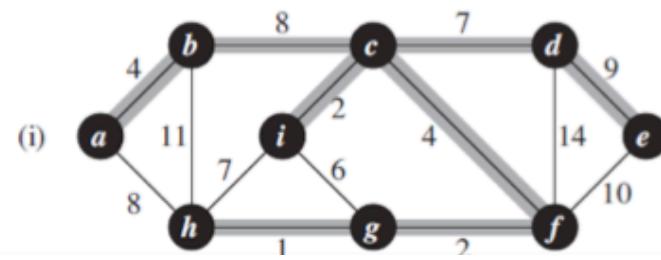
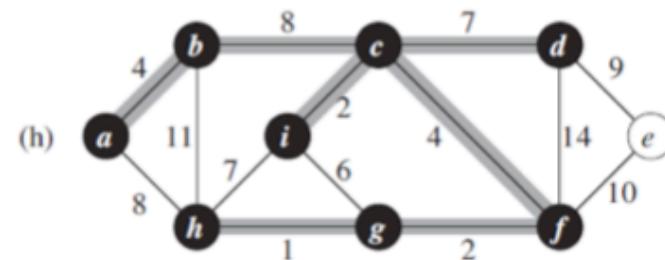
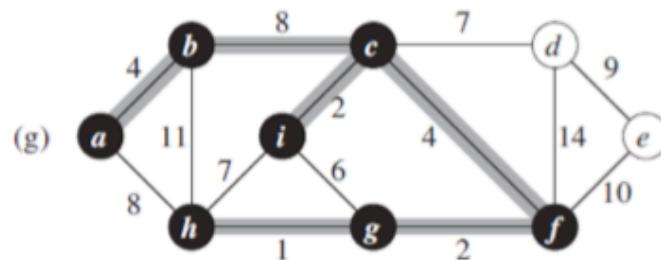


Prim's Algorithm: Example Execution

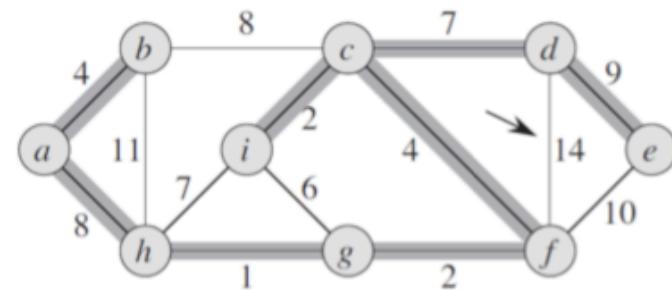


Prim's Algorithm: Example Execution

Similar remaining steps...



Compare with the MST by Kruskal's Algorithm



Correctness of Prim's Algorithm

- Want to show that all added edges are safe edges
- Again, assume edges all have different costs (not essential)

Proof of Correctness

- If e is added to the tree, then e is safe
 - Let S be the vertices connected by edges in T when e is added
 - e is the minimum cost edge crossing cut $(S, V \setminus S)$.
- S is connected in each iteration and eventually $S = V$.

Prim's Algorithm: Time Complexity

PrimMST(G)

$O(m \log m)$

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **while** $|S| < n$ Repeat n times
 - 5 choose $e = (u, v) \in E$ of minimum cost, such that $u \in S$ and $v \in V \setminus S$;
 - 6 $T = T \cup \{e\}$;
 - 7 $S = S \cup \{v\}$;
- 8 **return** T

In total: $O(m \log m) + n \times O(m) = O(mn)$

Prim's Algorithm: Time Complexity

PrimMST(G)

$O(m \log m)$

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **while** $|S| < n$ Repeat n times
 - 5 choose $e = (u, v) \in E$ of minimum cost, such that $u \in S$ and $v \in V \setminus S$;
 - 6 $T = T \cup \{e\}$;
 - 7 $S = S \cup \{v\}$;
- 8 **return** T

$O(m)$ times

Q: Can we do better?

More Efficient Implementation

`SmarterPrim(G)`

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;

More Efficient Implementation

SmarterPrim(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **for** $u \notin S$, initialize $a(u) = \arg \min_{e=(u,v), v \in S} c_e$
// $a(u)$ stores the cheapest edge from S to u

About notation “ $\arg \min$ ”

If $x = 4$ minimizes $(x - 2)^2$ subject to $x \geq 4$, we say $4 = \arg \min_{x \geq 4} (x - 2)^2$

More Efficient Implementation

SmarterPrim(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **for** $u \notin S$, initialize $a(u) = \arg \min_{e=(u,v), v \in S} c_e$
// $a(u)$ stores the cheapest edge from S to u
- 5 **while** $|S| < n$
- 6
- 7
- 8
- 9
- 10

More Efficient Implementation

SmarterPrim(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **for** $u \notin S$, initialize $a(u) = \arg \min_{e=(u,v), v \in S} c_e$
// $a(u)$ stores the cheapest edge from S to u
- 5 **while** $|S| < n$
 - 6 pick *minimum* $a(u) = (u, v)$;
 - 7 $T = T \cup \{a(u)\}$;
 - 8 $S = S \cup \{u\}$;
 - 9
- 10

More Efficient Implementation

SmarterPrim(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **for** $u \notin S$, initialize $a(u) = \arg \min_{e=(u,v), v \in S} c_e$
// $a(u)$ stores the cheapest edge from S to u
- 5 **while** $|S| < n$
 - 6 pick *minimum* $a(u) = (u, v)$;
 - 7 $T = T \cup \{a(u)\}$;
 - 8 $S = S \cup \{u\}$;
 - 9 update array a ;
- 10 **return** T

More Efficient Implementation

SmarterPrim(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **for** $u \notin S$, initialize $a(u) = \arg \min_{e=(u,v), v \in S} c_e$
// $a(u)$ stores the cheapest edge from S to u
- 5 **while** $|S| < n$
 - 6 pick *minimum* $a(u) = (u, v)$;
 - 7 $T = T \cup \{a(u)\}$;
 - 8 $S = S \cup \{u\}$;
 - 9 update array a ;
- 10 **return** T

Running time depends on implementation of highlighted steps

More Efficient Implementation

SmarterPrim(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **for** $u \notin S$, initialize $a(u) = \arg \min_{e=(u,v), v \in S} c_e$
- 5 **while** $|S| < n$
 - 6 pick *minimum* $a(u) = (u, v)$;
 - 7 $T = T \cup \{a(u)\}$;
 - 8 $S = S \cup \{u\}$;
 - 9 update array a ;
- 10 **return** T

➤ Implementation of Step 4

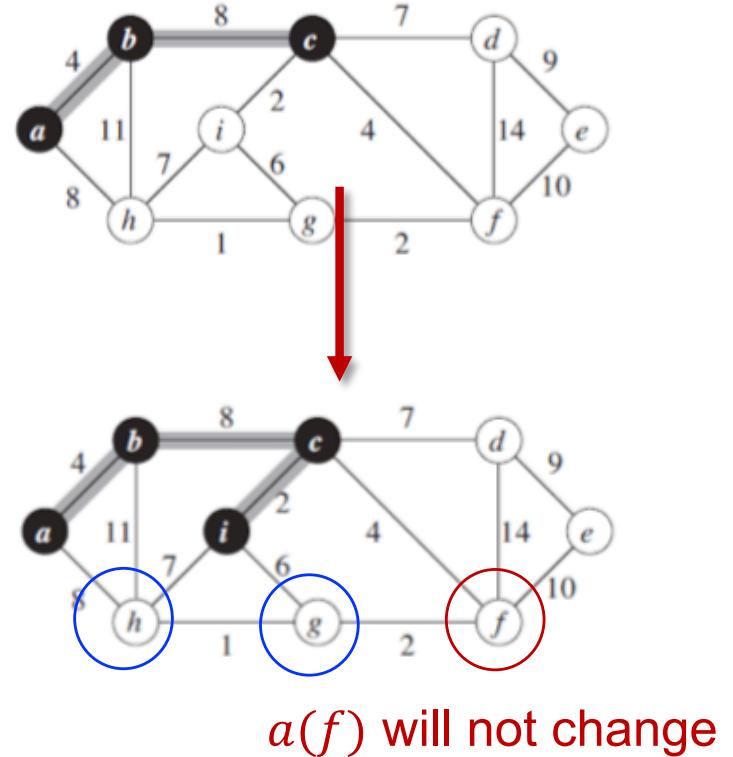
- For each u , check all edges that connect u to S
- Each edge is checked for at most twice
- In total, $O(m)$ time

More Efficient Implementation

SmarterPrim(G)

```

1 Sort edges in increasing order of costs
2 Initialize  $T = \emptyset$ ;
3 Initialize  $S = \{1\}$ ;
4 for  $u \notin S$ , initialize  $a(u) = \arg \min_{e=(u,v), v \in S} c_e$ 
5 while  $|S| < n$ 
6     pick minimum  $a(u) = (u, v)$ ;
7      $T = T \cup \{a(u)\}$ ;
8      $S = S \cup \{u\}$ ;
9     update array  $a$ ;
10    return  $T$ 
```



➤ Implementation of Step 9: how to update $a(v)$ for any $v \notin S$?

- First of all, only need to update the v that connects to u
- For these v 's, simply compare $c_{a(v)}$ and c_{uv}
- During entire “while loop”, each edge is considered at most once, so update $a(u)$ for at most m times

More Efficient Implementation

SmarterPrim(G)

- 1 Sort edges in increasing order of costs
- 2 Initialize $T = \emptyset$;
- 3 Initialize $S = \{1\}$;
- 4 **for** $u \notin S$, initialize $a(u) = \arg \min_{e=(u,v), v \in S} c_e$
- 5 **while** $|S| < n$
 - 6 **pick minimum** $a(u) = (u, v)$;
 - 7 $T = T \cup \{a(u)\}$;
 - 8 $S = S \cup \{u\}$;
 - 9 **update array** a ;
- 10 **return** T

- Implementation of Step 6: which data structure is efficient for repeatedly retrieving min?
 - Priority queue!

Priority Queues

Store a set S of n elements, where each element $v \in S$ has an associated key value $k(v)$, with the following operations:

- **Make-Queue**: create an empty queue
- **Find-Min**: find the minimum key in S
- **Extract-Min**: remove $v \in S$ with minimum key and return it
- **Decrease-Key($v, k'(v)$)**: decrease key of v from $k(v)$ to $k'(v)$
- **Add($v, k(v)$)**: add new element v with key $k(v)$ to S

➤ Prim requires

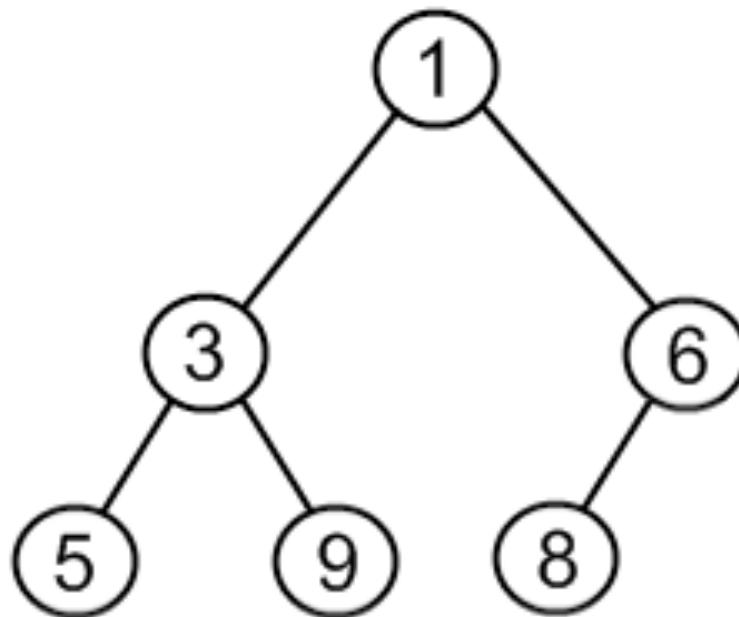
- $O(n)$ Extract-Min for **pick minimum $a(u)$**
- $O(m)$ Decrease-Key for **update array a** (whenever u added to S , only update u 's neighbors' key value)

➤ Binary heap implementation: both operations take $O(\log n)$

➤ Total time $O((m + n) \log n) = O(m \log m)$

An Example of Priority Queue via Binary Heap

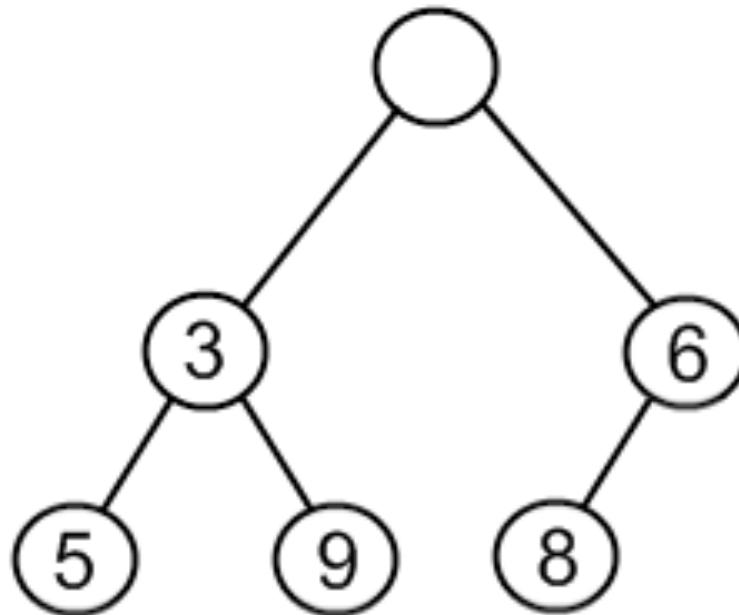
Extract-Min



Main property: any parent is smaller than both children

An Example of Priority Queue via Binary Heap

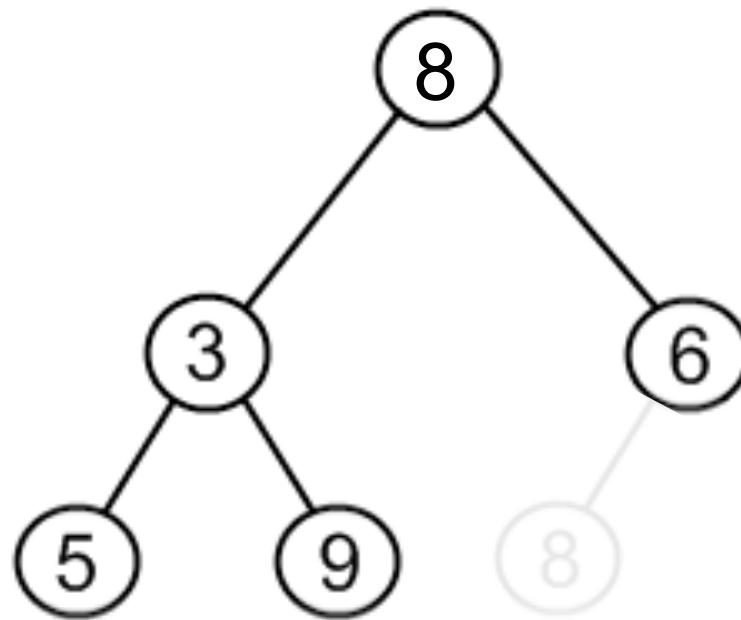
Extract-Min



Main property: any parent is smaller than both children

An Example of Priority Queue via Binary Heap

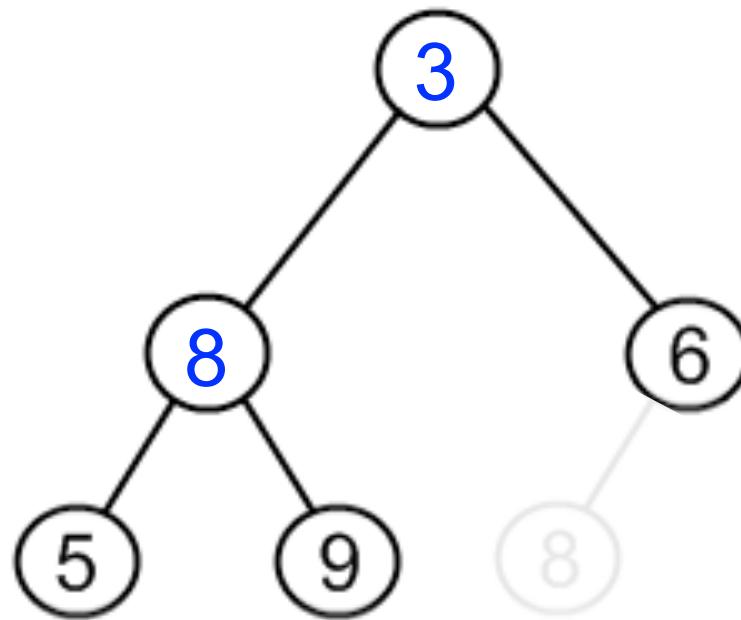
Extract-Min



Main property: any parent is smaller than both children

An Example of Priority Queue via Binary Heap

Extract-Min

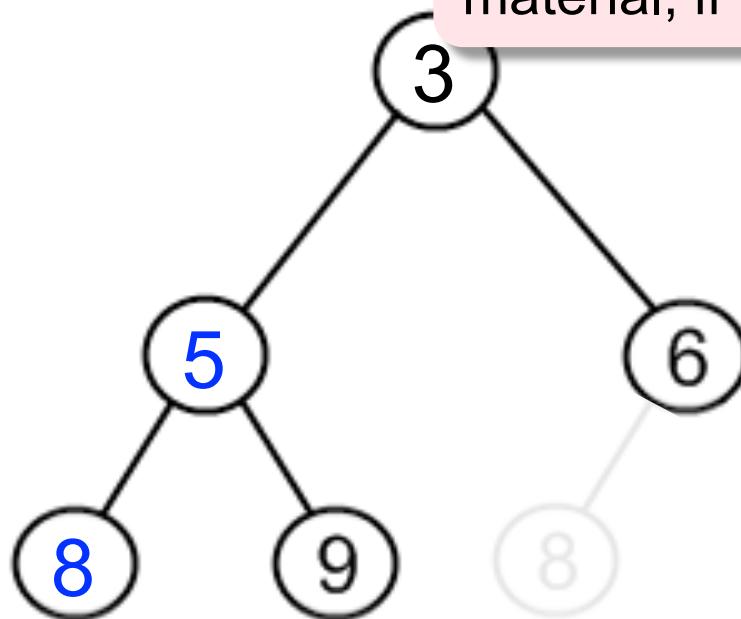


Main property: any parent is smaller than both children

An Example of Priority Queue via Binary Heap

Extract-Min

Please review data structure material, if not appear familiar to you



Main property: any parent is smaller than both children

More About MSTs

- Many different MST algorithms; all of them rely on some basic properties of MSTs
 - There is a randomized algorithm that runs in $O(m)$ expected time
- **Still open:** Is there an $O(m)$ time deterministic algorithm?

Lessons Learned

- One problem may have different greedy algorithms that all work
- The design of proper data structure may help to speed up algorithms
 - So, algorithm design is *not only* about inventing the algorithm *but also* about efficient implementation of the algorithm

Thank You

Haifeng Xu

University of Virginia

hx4ad@virginia.edu