CS/ECE 374: Algorithms & Models of Computation

Algorithms for Minimum Spanning Trees

Lecture 19 April 4, 2023

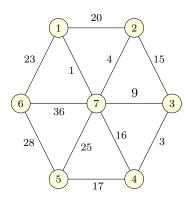
Part I

Algorithms for Minimum Spanning Tree

Minimum Spanning Tree

Input Connected graph G = (V, E) with edge costs **Goal** Find $T \subseteq E$ such that (V, T) is connected and total cost of all edges in T is smallest

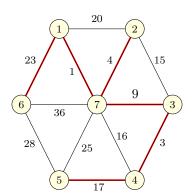
1 T is the minimum spanning tree (MST) of G



Minimum Spanning Tree

Input Connected graph G = (V, E) with edge costs **Goal** Find $T \subseteq E$ such that (V, T) is connected and total cost of all edges in T is smallest

1 T is the minimum spanning tree (MST) of G



Applications

- Network Design
 - Designing networks with minimum cost but maximum connectivity
- 2 Approximation algorithms
 - Can be used to bound the optimality of algorithms to approximate Traveling Salesman Problem, Steiner Trees, etc.
- Cluster Analysis

Some basic properties of (Spanning) Trees

- A tree T = (V, E) is a connected acyclic undirected graph
- A tree on n nodes has exactly n-1 edges
- Every tree on n > 1 nodes has at least two leaves
- A tree **T** in a graph **G** is spanning if **T** includes every node of **G**
- G is connected implies that there is a path from u to v for all nodes u, v
- A graph G is connected iff it has a spanning tree (constructive proof via Basic Search)

A useful lemma

Lemma

Let $T = (V, E_T)$ be a spanning tree of G = (V, E). For every non-tree edge $e \in E \setminus E_T$ there is a unique cycle C in T + e. For every edge $f \in C - \{e\}$, T - f + e is another spanning tree of G.

Greedy Template

```
Initially E is the set of all edges in G
T is empty (* T will store edges of a MST *)

while E is not empty do

choose e \in E

if (e \text{ satisfies condition})

add e to T

return the set T
```

Main Task: In what order should edges be processed? When should we add edge to spanning tree?







Process edges in the order of their costs (starting from the least) and add edges to T as long as they don't form a cycle.

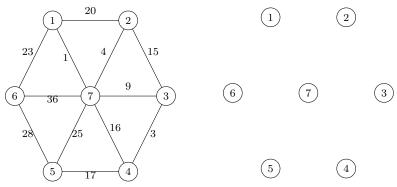


Figure: Graph G

Figure: MST of G

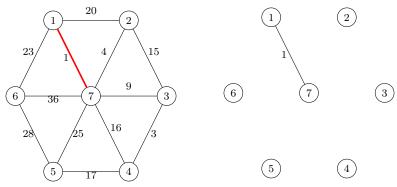


Figure: Graph G

Figure: MST of G

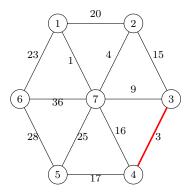


Figure: Graph G

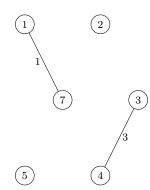


Figure: MST of G

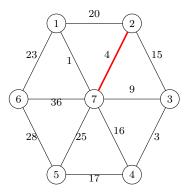


Figure: Graph G

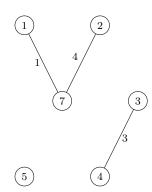


Figure: MST of G

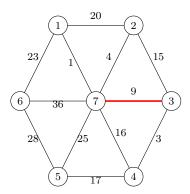


Figure: Graph G

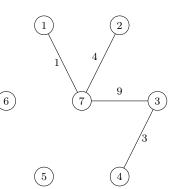


Figure: MST of G

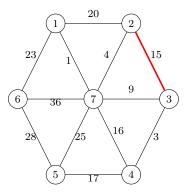


Figure: Graph G

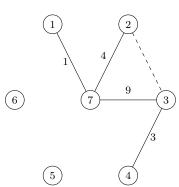


Figure: MST of G

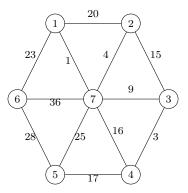


Figure: Graph G

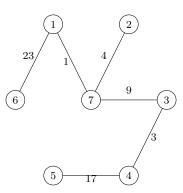


Figure: MST of G

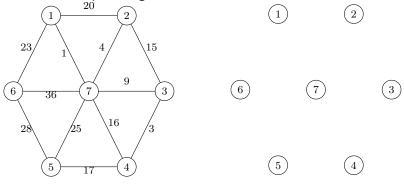


Figure: Graph G

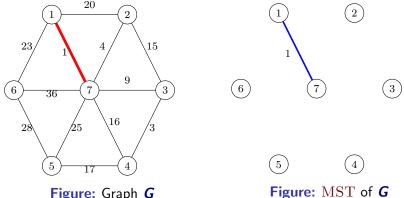


Figure: Graph G

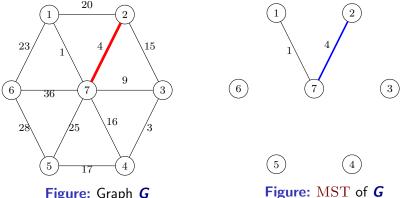


Figure: Graph G

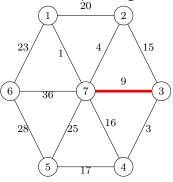


Figure: Graph G

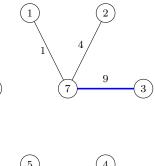


Figure: MST of G

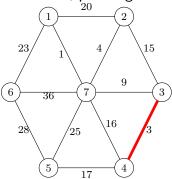


Figure: Graph G

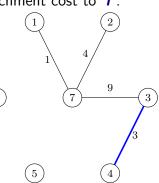


Figure: MST of G

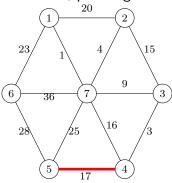


Figure: Graph G

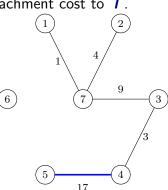


Figure: MST of G



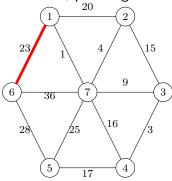


Figure: Graph G

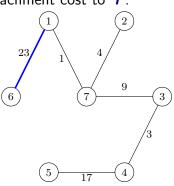


Figure: MST of G

Reverse Delete Algorithm

```
Initially E is the set of all edges in G
T is E (* T will store edges of a MST *)
while E is not empty do
    choose e ∈ E of largest cost
    if removing e does not disconnect T then
        remove e from T
return the set T
```

Returns a minimum spanning tree.

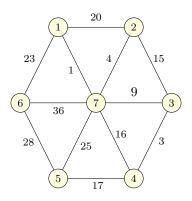


Borůvka's Algorithm

Simplest to implement. See notes. Assume G is a connected graph.

```
T is \emptyset (* T will store edges of a MST *) while T is not spanning do
X \leftarrow \emptyset
for each connected component S of T do
add to X the cheapest edge between S and V \setminus S
Add edges in X to T
```

Borůvka's Algorithm



Correctness of MST Algorithms

- Many different MST algorithms
- ② All of them rely on some basic properties of MSTs, in particular the **Cut Property** to be seen shortly.

Assumption

And for now ...

Assumption

Edge costs are distinct, that is no two edge costs are equal.

Cuts

Definition

Given a graph G = (V, E), a **cut** is a partition of the vertices of the graph into two sets $(S, V \setminus S)$.

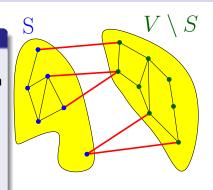
Cuts

Definition

Given a graph G = (V, E), a **cut** is a partition of the vertices of the graph into two sets $(S, V \setminus S)$.

Edges having an endpoint on both sides are the edges of the cut.

A cut edge is **crossing** the cut.



Safe and Unsafe Edges

Definition

An edge e = (u, v) is a safe edge if there is some partition of V into S and $V \setminus S$ and e is the unique minimum cost edge crossing S (one end in S and the other in $V \setminus S$).

Safe and Unsafe Edges

Definition

An edge e = (u, v) is a safe edge if there is some partition of V into S and $V \setminus S$ and e is the unique minimum cost edge crossing S (one end in S and the other in $V \setminus S$).

Definition

An edge e = (u, v) is an unsafe edge if there is some cycle C such that e is the unique maximum cost edge in C.

Safe and Unsafe Edges

Definition

An edge e = (u, v) is a safe edge if there is some partition of V into S and $V \setminus S$ and e is the unique minimum cost edge crossing S (one end in S and the other in $V \setminus S$).

Definition

An edge e = (u, v) is an unsafe edge if there is some cycle C such that e is the unique maximum cost edge in C.

Proposition

If edge costs are distinct then every edge is either safe or unsafe.

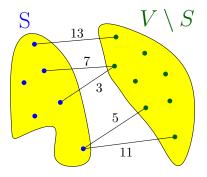
Proof.

Exercise.

Safe edge

Example...

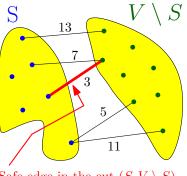
Every cut identifies one safe edge...



Safe edge

Example...

Every cut identifies one safe edge...



Safe edge in the cut $(S, V \setminus S)$

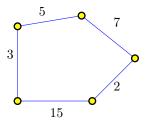
...the cheapest edge in the cut.

Note: An edge e may be a safe edge for many cuts!

Unsafe edge

Example...

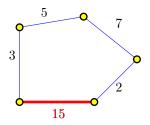
Every cycle identifies one unsafe edge...



Unsafe edge

Example...

Every cycle identifies one **unsafe** edge...



...the most expensive edge in the cycle.

Example

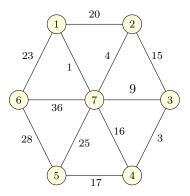


Figure: Graph with unique edge costs. Safe edges are red, rest are unsafe.

Example

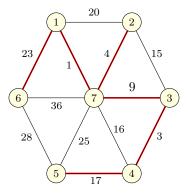


Figure: Graph with unique edge costs. Safe edges are red, rest are unsafe.

Example

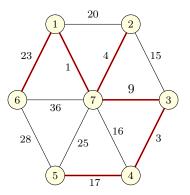


Figure: Graph with unique edge costs. Safe edges are red, rest are unsafe.

And all safe edges are in the MST in this case...

Key Observation: Cut Property

Lemma

If e is a safe edge then every minimum spanning tree contains e.

Key Observation: Cut Property

Lemma

If e is a safe edge then every minimum spanning tree contains e.

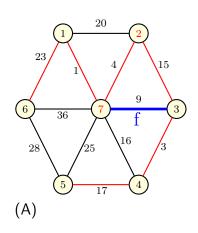
- Suppose (for contradiction) e is not in MST T.
- ② Since e is safe there is an $S \subset V$ such that e is the unique min cost edge crossing S.
- ullet Since T is connected, there must be some edge f with one end in S and the other in $V \setminus S$
- Since $c_f > c_e$, $T' = (T \setminus \{f\}) \cup \{e\}$ is a spanning tree of lower cost!

Key Observation: Cut Property

Lemma

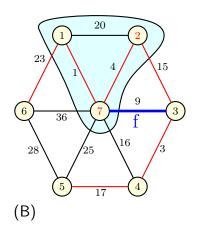
If e is a safe edge then every minimum spanning tree contains e.

- Suppose (for contradiction) e is not in MST T.
- ② Since e is safe there is an $S \subset V$ such that e is the unique min cost edge crossing S.
- 3 Since T is connected, there must be some edge f with one end in S and the other in $V \setminus S$
- Since $c_f > c_e$, $T' = (T \setminus \{f\}) \cup \{e\}$ is a spanning tree of lower cost! Error: T' may not be a spanning tree!!



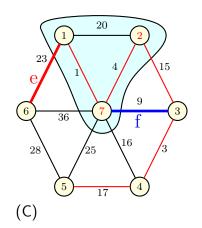
 \bullet (A) Consider adding the edge f.

Problematic example. $S = \{1, 2, 7\}$, e = (7, 3), f = (1, 6). T - f + e is not a spanning tree.



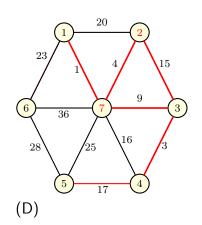
- \bullet (A) Consider adding the edge f.
- (B) It is safe because it is the cheapest edge in the cut.

Problematic example. $S = \{1, 2, 7\}$, e = (7, 3), f = (1, 6). T - f + e is not a spanning tree.



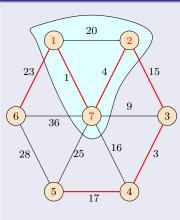
- **1** (A) Consider adding the edge f.
- (B) It is safe because it is the cheapest edge in the cut.
- (C) Lets throw out the edge e currently in the spanning tree which is more expensive than f and is in the same cut. Put it f instead...

Problematic example. $S = \{1, 2, 7\}$, e = (7, 3), f = (1, 6). T - f + e is not a spanning tree.

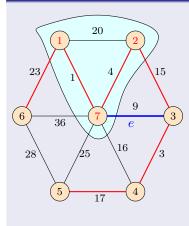


- \bullet (A) Consider adding the edge f.
- (B) It is safe because it is the cheapest edge in the cut.
- (C) Lets throw out the edge e currently in the spanning tree which is more expensive than f and is in the same cut. Put it f instead...
- (D) New graph of selected edges is not a tree anymore. BUG.

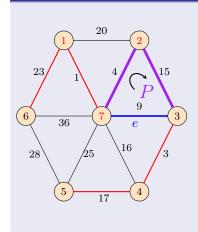




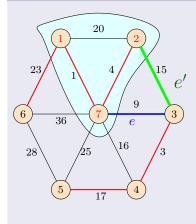
① Suppose e = (v, w) is not in MST T and e is min weight edge in cut $(S, V \setminus S)$. Assume $v \in S$.



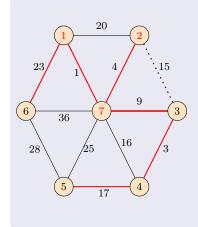
- Suppose e = (v, w) is not in MST T and e is min weight edge in cut $(S, V \setminus S)$. Assume $v \in S$.
- T is spanning tree: there is a unique path P from v to w in T



- ① Suppose e = (v, w) is not in MST T and e is min weight edge in cut $(S, V \setminus S)$. Assume $v \in S$.
- 2 T is spanning tree: there is a unique path P from v to w in T



- Suppose e = (v, w) is not in MST T and e is min weight edge in cut $(S, V \setminus S)$. Assume $v \in S$.
- T is spanning tree: there is a unique path P from v to w in T
- **1** Let w' be the first vertex in P belonging to $V \setminus S$; let v' be the vertex just before it on P, and let e' = (v', w')



- ① Suppose e = (v, w) is not in MST T and e is min weight edge in cut $(S, V \setminus S)$. Assume $v \in S$.
- 2 T is spanning tree: there is a unique path P from v to w in T
- 3 Let w' be the first vertex in P belonging to $V \setminus S$; let v' be the vertex just before it on P, and let e' = (v', w')
- $T' = (T \setminus \{e'\}) \cup \{e\}$ is spanning tree of lower cost. (Why?)

Proof of Cut Property (contd)

Observation

 $T' = (T \setminus \{e'\}) \cup \{e\}$ is a spanning tree.

Proof.

T' is connected.

T' is a tree

23

Proof of Cut Property (contd)

Observation

 $T' = (T \setminus \{e'\}) \cup \{e\}$ is a spanning tree.

Proof.

T' is connected.

Removed e' = (v', w') from T but v' and w' are connected by the path P - f + e in T'. Hence T' is connected if T is.

T' is a tree

Proof of Cut Property (contd)

Observation

 $T' = (T \setminus \{e'\}) \cup \{e\}$ is a spanning tree.

Proof.

T' is connected.

Removed e' = (v', w') from T but v' and w' are connected by the path P - f + e in T'. Hence T' is connected if T is.

T' is a tree

T' is connected and has n-1 edges (since T had n-1 edges) and hence T' is a tree



Safe Edges form a Tree

Lemma

Let G be a connected graph with distinct edge costs, then the set of safe edges form a connected graph.

- Suppose not. Let S be a connected component in the graph induced by the safe edges.
- $oldsymbol{\circ}$ Consider the edges crossing $oldsymbol{S}$, there must be a safe edge among them since edge costs are distinct and so we must have picked it.



Safe Edges form an MST

Corollary

Let G be a connected graph with distinct edge costs, then set of safe edges form the unique $\overline{\mathrm{MST}}$ of G.

25

Safe Edges form an MST

Corollary

Let G be a connected graph with distinct edge costs, then set of safe edges form the unique $\overline{\mathrm{MST}}$ of G.

Consequence: Every correct \overline{MST} algorithm when G has unique edge costs includes exactly the safe edges.

Cycle Property

Lemma

If e is an unsafe edge then no \overline{MST} of G contains e.

Proof.

Exercise.

Note: Cut and Cycle properties hold even when edge costs are not distinct. Safe and unsafe definitions do not rely on distinct cost assumption.

Prim's Algorithm

Pick edge with minimum attachment cost to current tree, and add to current tree.

Proof of correctness.

① If e is added to tree, then e is safe and belongs to every MST.

2 Set of edges output is a spanning tree

27

Prim's Algorithm

Pick edge with minimum attachment cost to current tree, and add to current tree.

Proof of correctness.

- lacktriangle If e is added to tree, then e is safe and belongs to every MST.
 - $oldsymbol{0}$ Let $oldsymbol{S}$ be the vertices connected by edges in $oldsymbol{T}$ when $oldsymbol{e}$ is added.

2 Set of edges output is a spanning tree

Prim's Algorithm

Pick edge with minimum attachment cost to current tree, and add to current tree.

Proof of correctness.

- If e is added to tree, then e is safe and belongs to every MST.
 - $oldsymbol{0}$ Let $oldsymbol{S}$ be the vertices connected by edges in $oldsymbol{T}$ when $oldsymbol{e}$ is added.
 - e is edge of lowest cost with one end in S and the other in
 V \ S and hence e is safe.

27

2 Set of edges output is a spanning tree

Prim's Algorithm

Pick edge with minimum attachment cost to current tree, and add to current tree.

Proof of correctness.

- If e is added to tree, then e is safe and belongs to every MST.
 - $oldsymbol{0}$ Let $oldsymbol{S}$ be the vertices connected by edges in $oldsymbol{\mathcal{T}}$ when $oldsymbol{e}$ is added.
 - e is edge of lowest cost with one end in S and the other in
 V \ S and hence e is safe.
- Set of edges output is a spanning tree
 - **①** Set of edges output forms a connected graph: by induction, \boldsymbol{S} is connected in each iteration and eventually $\boldsymbol{S} = \boldsymbol{V}$.

27

Prim's Algorithm

Pick edge with minimum attachment cost to current tree, and add to current tree.

Proof of correctness.

- If e is added to tree, then e is safe and belongs to every MST.
 - $oldsymbol{0}$ Let $oldsymbol{S}$ be the vertices connected by edges in $oldsymbol{\mathcal{T}}$ when $oldsymbol{e}$ is added.
 - e is edge of lowest cost with one end in S and the other in
 V \ S and hence e is safe.
- Set of edges output is a spanning tree
 - Set of edges output forms a connected graph: by induction, S is connected in each iteration and eventually S = V.
 - ② Only safe edges added and they do not have a cycle

Kruskal's Algorithm

Pick edge of lowest cost and add if it does not form a cycle with existing edges.

Proof of correctness.

 $\textbf{0} \ \ \mathsf{lf} \ \boldsymbol{e} = (\boldsymbol{u}, \boldsymbol{v}) \ \mathsf{is} \ \mathsf{added} \ \mathsf{to} \ \mathsf{tree}, \ \mathsf{then} \ \boldsymbol{e} \ \mathsf{is} \ \mathsf{safe}$

Set of edges output is a spanning tree : exercise

Kruskal's Algorithm

Pick edge of lowest cost and add if it does not form a cycle with existing edges.

Proof of correctness.

- If e = (u, v) is added to tree, then e is safe
 - When algorithm adds e let S and S' be the connected components containing u and v respectively

Set of edges output is a spanning tree : exercise



Kruskal's Algorithm

Pick edge of lowest cost and add if it does not form a cycle with existing edges.

Proof of correctness.

- If e = (u, v) is added to tree, then e is safe
 - When algorithm adds e let S and S' be the connected components containing u and v respectively
 - ② e is the lowest cost edge crossing S (and also S').

Set of edges output is a spanning tree : exercise



Kruskal's Algorithm

Pick edge of lowest cost and add if it does not form a cycle with existing edges.

Proof of correctness.

- 1 If e = (u, v) is added to tree, then e is safe
 - When algorithm adds **e** let **S** and **S**' be the connected components containing **u** and **v** respectively
 - \mathbf{Q} \mathbf{e} is the lowest cost edge crossing \mathbf{S} (and also \mathbf{S}).
 - If there is an edge e' crossing S and has lower cost than e, then e' would come before e in the sorted order and would be added by the algorithm to T
- 2 Set of edges output is a spanning tree : exercise



Correctness of Borůvka's Algorithm

Proof of correctness.

Argue that only safe edges are added.

29 / 44

Correctness of Reverse Delete Algorithm

Reverse Delete Algorithm

Consider edges in decreasing cost and remove an edge if it does not disconnect the graph

Proof of correctness.

Argue that only unsafe edges are removed.

30

When edge costs are not distinct

Heuristic argument: Make edge costs distinct by adding a small tiny and different cost to each edge

31

When edge costs are not distinct

Heuristic argument: Make edge costs distinct by adding a small tiny and different cost to each edge

Formal argument: Order edges lexicographically to break ties

- $lackbox{0}$ $e_i \prec e_j$ if either $c(e_i) < c(e_j)$ or $(c(e_i) = c(e_j)$ and i < j)
- 2 Lexicographic ordering extends to sets of edges. If $A, B \subseteq E$, $A \neq B$ then $A \prec B$ if either c(A) < c(B) or (c(A) = c(B)) and $A \setminus B$ has a lower indexed edge than $B \setminus A$
- **3** Can order all spanning trees according to lexicographic order of their edge sets. Hence there is a unique MST.

When edge costs are not distinct

Heuristic argument: Make edge costs distinct by adding a small tiny and different cost to each edge

Formal argument: Order edges lexicographically to break ties

- $lackbox{0}$ $e_i \prec e_j$ if either $c(e_i) < c(e_j)$ or $(c(e_i) = c(e_j)$ and i < j)
- 2 Lexicographic ordering extends to sets of edges. If $A, B \subseteq E$, $A \neq B$ then $A \prec B$ if either c(A) < c(B) or (c(A) = c(B)) and $A \setminus B$ has a lower indexed edge than $B \setminus A$
- **3** Can order all spanning trees according to lexicographic order of their edge sets. Hence there is a unique MST.

When edge costs are not distinct

Heuristic argument: Make edge costs distinct by adding a small tiny and different cost to each edge

Formal argument: Order edges lexicographically to break ties

- $lackbox{0}$ $e_i \prec e_j$ if either $c(e_i) < c(e_j)$ or $(c(e_i) = c(e_j)$ and i < j)
- 2 Lexicographic ordering extends to sets of edges. If $A, B \subseteq E$, $A \neq B$ then $A \prec B$ if either c(A) < c(B) or (c(A) = c(B)) and $A \setminus B$ has a lower indexed edge than $B \setminus A$
- 3 Can order all spanning trees according to lexicographic order of their edge sets. Hence there is a unique MST.

Prim's, Kruskal, and Reverse Delete Algorithms are optimal with respect to lexicographic ordering.

Edge Costs: Positive and Negative

- Algorithms and proofs don't assume that edge costs are non-negative! MST algorithms work for arbitrary edge costs.
- Another way to see this: make edge costs non-negative by adding to each edge a large enough positive number. Why does this work for MSTs but not for shortest paths?
- Can compute maximum weight spanning tree by negating edge costs and then computing an MST.

Edge Costs: Positive and Negative

- Algorithms and proofs don't assume that edge costs are non-negative! MST algorithms work for arbitrary edge costs.
- Another way to see this: make edge costs non-negative by adding to each edge a large enough positive number. Why does this work for MSTs but not for shortest paths?
- Can compute maximum weight spanning tree by negating edge costs and then computing an MST.
 - Question: Why does this not work for shortest paths?

Part II

Data Structures for MST: Priority Queues and Union-Find

33

No complex data structure needed.

```
T is ∅ (* T will store edges of a MST *)
while T is not spanning do
    X ← ∅
    for each connected component S of T do
        add to X the cheapest edge between S and V \ S
    Add edges in X to T
return the set T
```

• $O(\log n)$ iterations of while loop. Why?

No complex data structure needed.

```
T is ∅ (* T will store edges of a MST *)
while T is not spanning do
    X ← ∅
    for each connected component S of T do
        add to X the cheapest edge between S and V \ S
    Add edges in X to T
return the set T
```

- $O(\log n)$ iterations of while loop. Why? Number of connected components shrink by at least half since each component merges with one or more other components.
- Each iteration can be implemented in O(m) time.

No complex data structure needed.

```
T is ∅ (* T will store edges of a MST *)
while T is not spanning do
    X ← ∅
    for each connected component S of T do
        add to X the cheapest edge between S and V \ S
    Add edges in X to T
return the set T
```

- O(log n) iterations of while loop. Why? Number of connected components shrink by at least half since each component merges with one or more other components.
- Each iteration can be implemented in O(m) time.

Running time: $O(m \log n)$ time.

No complex data structure needed.

```
T is ∅ (* T will store edges of a MST *)
while T is not spanning do
    X ← ∅
    for each connected component S of T do
        add to X the cheapest edge between S and V \ S
    Add edges in X to T
return the set T
```

- $O(\log n)$ iterations of while loop. Why? Number of connected components shrink by at least half since each component merges with one or more other components.
- Each iteration can be implemented in O(m) time.

Running time: $O(m \log n)$ time.

Advantages of the algorithm: no data structures and easy to implement in parallel setting

34

Prim's Algorithm

T maintained by algorithm will be a tree. Start with a node in T. In each iteration, pick edge with least attachment cost to T.

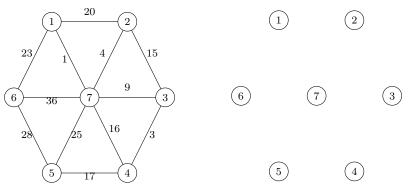


Figure: Graph G

Implementing Prim's Algorithm

```
Prim_ComputeMST
    E is the set of all edges in G
    S = \{1\}
    T is empty (* T will store edges of a MST *)
    while S \neq V do
        pick e = (v, w) \in E such that
             v \in S and w \in V - S
             e has minimum cost
        T = T \cup e
        S = S \cup w
    return the set T
```

Analysis

Implementing Prim's Algorithm

```
Prim_ComputeMST
    E is the set of all edges in G
    S = \{1\}
    T is empty (* T will store edges of a MST *)
    while S \neq V do
        pick e = (v, w) \in E such that
             v \in S and w \in V - S
             e has minimum cost
        T = T \cup e
        S = S \cup w
    return the set T
```

Analysis

1 Number of iterations = O(n), where n is number of vertices

Implementing Prim's Algorithm

```
Prim_ComputeMST
    E is the set of all edges in G
    S = \{1\}
    T is empty (* T will store edges of a MST *)
    while S \neq V do
        pick e = (v, w) \in E such that
             v \in S and w \in V - S
             e has minimum cost
        T = T \cup e
        S = S \cup w
    return the set T
```

Analysis

- 1 Number of iterations = O(n), where n is number of vertices
- 2 Picking e is O(m) where m is the number of edges

Implementing Prim's Algorithm

```
Prim_ComputeMST
    E is the set of all edges in G
    S = \{1\}
    T is empty (* T will store edges of a MST *)
    while S \neq V do
        pick e = (v, w) \in E such that
             v \in S and w \in V - S
             e has minimum cost
        T = T \cup e
        S = S \cup w
    return the set T
```

Analysis

- 1 Number of iterations = O(n), where n is number of vertices
- 2 Picking e is O(m) where m is the number of edges
- **3** Total time *O*(*nm*)

More Efficient Implementation

```
Prim_ComputeMST
      E is the set of all edges in G
      S = \{1\}
      T is empty (* T will store edges of a MST *)
      for \mathbf{v} \notin \mathbf{S}, \mathbf{a}(\mathbf{v}) = \min_{\mathbf{w} \in \mathbf{S}} \mathbf{c}(\mathbf{w}, \mathbf{v})
      for v \not\in S, e(v) = w such that w \in S and c(w, v) is minimum
      while S \neq V do
           pick \mathbf{v} with minimum \mathbf{a}(\mathbf{v})
            T = T \cup \{(e(v), v)\}
           S = S \cup \{v\}
           update arrays a and e
     return the set T
```

More Efficient Implementation

```
Prim_ComputeMST
      E is the set of all edges in G
      S = \{1\}
      T is empty (* T will store edges of a MST *)
      for \mathbf{v} \notin \mathbf{S}, \mathbf{a}(\mathbf{v}) = \min_{\mathbf{w} \in \mathbf{S}} \mathbf{c}(\mathbf{w}, \mathbf{v})
      for v \not\in S, e(v) = w such that w \in S and c(w, v) is minimum
      while S \neq V do
           pick \mathbf{v} with minimum \mathbf{a}(\mathbf{v})
            T = T \cup \{(e(v), v)\}
           S = S \cup \{v\}
           update arrays a and e
     return the set T
```

More Efficient Implementation

```
Prim_ComputeMST
      E is the set of all edges in G
      S = \{1\}
      T is empty (* T will store edges of a MST *)
      for \mathbf{v} \notin \mathbf{S}, \mathbf{a}(\mathbf{v}) = \min_{\mathbf{w} \in \mathbf{S}} \mathbf{c}(\mathbf{w}, \mathbf{v})
      for v \notin S, e(v) = w such that w \in S and c(w, v) is minimum
      while S \neq V do
           pick \mathbf{v} with minimum \mathbf{a}(\mathbf{v})
            T = T \cup \{(e(v), v)\}
           S = S \cup \{v\}
           update arrays a and e
      return the set T
```

Maintain vertices in $V \setminus S$ in a priority queue with key a(v).

Priority Queues

Data structure to store a set S of n elements where each element $v \in S$ has an associated real/integer key k(v) such that the following operations

- makeQ: create an empty queue
- 2 findMin: find the minimum key in S
- extractMin: Remove $v \in S$ with smallest key and return it
- **add**(v, k(v)): Add new element v with key k(v) to S
- **5** Delete(v): Remove element v from S
- **6 decreaseKey** (v, k'(v)): *decrease* key of v from k(v) (current key) to k'(v) (new key). Assumption: $k'(v) \le k(v)$
- o meld: merge two separate priority queues into one

Prim's using priority queues

```
E is the set of all edges in G S = \{1\} T is empty (* T will store edges of a MST *) for v \not\in S, a(v) = \min_{w \in S} c(w, v) for v \not\in S, e(v) = w such that w \in S and c(w, v) is minimum while S \neq V do pick v with minimum a(v) T = T \cup \{(e(v), v)\} S = S \cup \{v\} update arrays a and e return the set T
```

Maintain vertices in $V \setminus S$ in a priority queue with key a(v)

Prim's using priority queues

```
\boldsymbol{E} is the set of all edges in \boldsymbol{G}
S = \{1\}
T is empty (* T will store edges of a MST *)
for \mathbf{v} \notin \mathbf{S}, \mathbf{a}(\mathbf{v}) = \min_{\mathbf{w} \in \mathbf{S}} \mathbf{c}(\mathbf{w}, \mathbf{v})
for v \not\in S, e(v) = w such that w \in S and c(w, v) is minimum
while S \neq V do
      pick \mathbf{v} with minimum \mathbf{a}(\mathbf{v})
      T = T \cup \{(e(v), v)\}
      S = S \cup \{v\}
      update arrays a and e
return the set T
```

Maintain vertices in $V \setminus S$ in a priority queue with key a(v)

• Requires O(n) extractMin operations

39

Prim's using priority queues

```
\boldsymbol{E} is the set of all edges in \boldsymbol{G}
S = \{1\}
T is empty (* T will store edges of a MST *)
for \mathbf{v} \not\in \mathbf{S}, \mathbf{a}(\mathbf{v}) = \min_{\mathbf{w} \in \mathbf{S}} \mathbf{c}(\mathbf{w}, \mathbf{v})
for v \not\in S, e(v) = w such that w \in S and c(w, v) is minimum
while S \neq V do
      pick \mathbf{v} with minimum \mathbf{a}(\mathbf{v})
      T = T \cup \{(e(v), v)\}
      S = S \cup \{v\}
      update arrays a and e
return the set T
```

Maintain vertices in $V \setminus S$ in a priority queue with key a(v)

- Requires O(n) extractMin operations
- 2 Requires O(m) decrease Key operations

Running time of Prim's Algorithm

- O(n) extractMin operations and O(m) decreaseKey operations
 - ① Using standard Heaps, extractMin and decreaseKey take $O(\log n)$ time. Total: $O((m+n)\log n)$
 - ② Using Fibonacci Heaps, $O(\log n)$ for extractMin and O(1) (amortized) for decreaseKey. Total: $O(n \log n + m)$.

Running time of Prim's Algorithm

- O(n) extractMin operations and O(m) decreaseKey operations
 - ① Using standard Heaps, extractMin and decreaseKey take $O(\log n)$ time. Total: $O((m+n)\log n)$
 - ② Using Fibonacci Heaps, $O(\log n)$ for extractMin and O(1) (amortized) for decreaseKey. Total: $O(n \log n + m)$.

Prim's algorithm and Dijkstra's algorithms are similar. Where is the difference?

Process edges in the order of their costs (starting from the least) and add edges to T as long as they don't form a cycle.

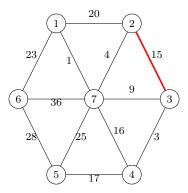


Figure: Graph G

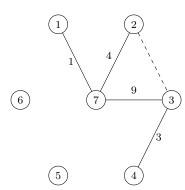


Figure: MST of G

$Kruskal_ComputeMST$

```
Initially E is the set of all edges in G
T is empty (* T will store edges of a MST *)

while E is not empty do

choose e \in E of minimum cost

if (T \cup \{e\}) does not have cycles)

add e to T

return the set T
```

Kruskal_ComputeMST

```
Initially E is the set of all edges in G
T is empty (* T will store edges of a MST *)

while E is not empty do

choose e \in E of minimum cost

if (T \cup \{e\}) does not have cycles)

add e to T

return the set T
```

```
Kruskal_ComputeMST

Initially E is the set of all edges in G
T is empty (* T will store edges of a MST *)

while E is not empty do

choose e \in E of minimum cost

if (T \cup \{e\}) does not have cycles)

add e to T

return the set T
```

• Presort edges based on cost. Choosing minimum can be done in O(1) time

```
Kruskal_ComputeMST

Initially E is the set of all edges in G
T is empty (* T will store edges of a MST *)

while E is not empty do

choose e \in E of minimum cost

if (T \cup \{e\} \text{ does not have cycles})

add e to T

return the set T
```

• Presort edges based on cost. Choosing minimum can be done in O(1) time

```
Kruskal_ComputeMST

Initially E is the set of all edges in G
T is empty (* T will store edges of a MST *)

while E is not empty do

choose e \in E of minimum cost

if (T \cup \{e\} \text{ does not have cycles})

add e to T

return the set T
```

- Presort edges based on cost. Choosing minimum can be done in O(1) time
- ② Do BFS/DFS on $T \cup \{e\}$. Takes O(n) time

```
Kruskal_ComputeMST

Initially E is the set of all edges in G
T is empty (* T will store edges of a MST *)

while E is not empty do

choose e \in E of minimum cost

if (T \cup \{e\}) does not have cycles)

add e to T

return the set T
```

- Presort edges based on cost. Choosing minimum can be done in O(1) time
- **2** Do **BFS/DFS** on $T \cup \{e\}$. Takes O(n) time
- 3 Total time $O(m \log m) + O(mn) = O(mn)$

Implementing Kruskal's Algorithm Efficiently

Kruskal_ComputeMST Sort edges in E based on cost T is empty (* T will store edges of a MST *) each vertex u is placed in a set by itself while E is not empty dopick $e = (u, v) \in E$ of minimum cost if u and v belong to different sets add e to Tmerge the sets containing u and vreturn the set T

Implementing Kruskal's Algorithm Efficiently

```
Kruskal_ComputeMST
   Sort edges in E based on cost
   T is empty (* T will store edges of a MST *)
   each vertex u is placed in a set by itself
   while E is not empty do
        pick e = (u, v) ∈ E of minimum cost
        if u and v belong to different sets
        add e to T
        merge the sets containing u and v
   return the set T
```

Need a data structure to check if two elements belong to same set and to merge two sets.

Implementing Kruskal's Algorithm Efficiently

```
Kruskal_ComputeMST
   Sort edges in E based on cost
   T is empty (* T will store edges of a MST *)
   each vertex u is placed in a set by itself
   while E is not empty do
        pick e = (u, v) ∈ E of minimum cost
        if u and v belong to different sets
        add e to T
        merge the sets containing u and v
   return the set T
```

Need a data structure to check if two elements belong to same set and to merge two sets.

Using Union-Find data structure can implement Kruskal's algorithm in $O((m+n)\log m)$ time.

Best Known Asymptotic Running Times for MST

Prim's algorithm using Fibonacci heaps: $O(n \log n + m)$. If m is O(n) then running time is $\Omega(n \log n)$.

Best Known Asymptotic Running Times for MST

Prim's algorithm using Fibonacci heaps: $O(n \log n + m)$. If m is O(n) then running time is $\Omega(n \log n)$.

Question

Is there a linear time (O(m + n)) time) algorithm for MST?

Best Known Asymptotic Running Times for MST

Prim's algorithm using Fibonacci heaps: $O(n \log n + m)$. If m is O(n) then running time is $\Omega(n \log n)$.

Question

Is there a linear time (O(m + n) time) algorithm for MST?

- O(m+n) time using bit operations in RAM model [Fredman, Willard 1994]
- O(m+n) expected time (randomized algorithm) [Karger, Klein, Tarjan 1995]
- **4** $O((n+m)\alpha(m,n))$ time Chazelle 2000]
- **5** Still open: Is there an O(n + m) time deterministic algorithm in the comparison model?