

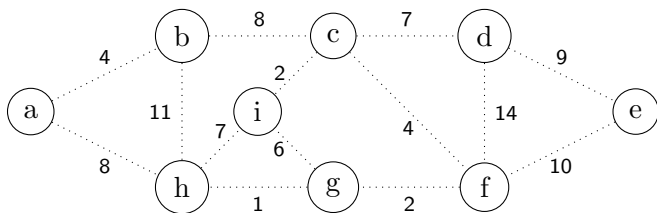
# CSCB63 – Design and Analysis of Data Structures

Akshay Arun Bapat<sup>1</sup>

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<sup>1</sup>based on notes by Anya Tafliovich, Anna Bretscher and Albert Lai

## introduction



An (edge-)weighted graph

Applications?

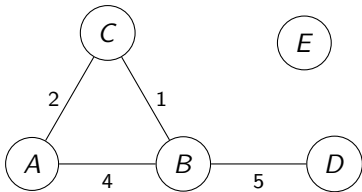


## weighted graph

A weighted (edge-weighted) graph consists of:

- a set of vertices  $V$
- a set of edges  $E$
- weights: a map  $w : E \rightarrow \mathbb{R}$  (usually  $\geq 0$ )
  - if undirected graph:  $(u, v)$  and  $(v, u)$  have the same weight
  - if directed graph:  $(u, v)$  and  $(v, u)$  may have different weights

## storing a weighted graph



Adjacency matrix:

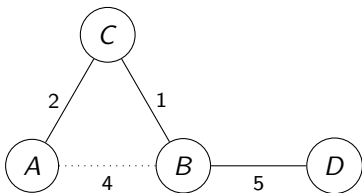
	A	B	C	D	E
A	0	4	2	$\infty$	$\infty$
B	4	0	1	5	$\infty$
C	2	1	0	$\infty$	$\infty$
D	$\infty$	5	$\infty$	0	$\infty$
E	$\infty$	$\infty$	$\infty$	$\infty$	0

Adjacency lists:

	adjacency list
A	(B,4), (C,2)
B	(A,4), (C,1), (D,5)
C	(A,2), (B,1)
D	(B,5)
E	

## minimum spanning tree

- common task #1 on weighted graphs
- find a spanning tree
  - a tree that covers all vertices
  - a tree  $T$  such that every vertex  $v \in V$  is an endpoint of at least one edge in  $T$
- minimise the sum of the weights of the edges used
  - $weight(T) = \sum_{(u,v) \in T} weight(u,v)$
  - want tree  $T$  with minimum  $weight(T)$



Usually just for undirected, connected graphs.

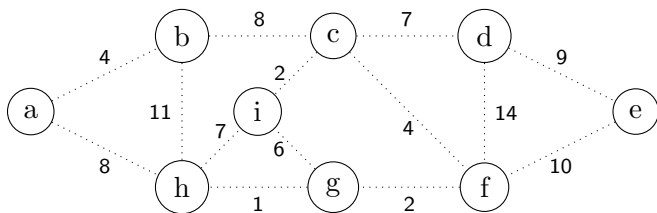
## Kruskal's algorithm: idea

Kruskal's algorithm finds a MST by successive mergers.

1. At first, each vertex is its own small cluster/tree/set.
2. Find an edge of minimum weight, use it to merge two clusters/trees/sets into one.
  - Do not create cycles!
3. Do it again. . .
4. In general, find an edge of minimum weight that crosses two clusters; merge them into one.

Correctness idea: at each iteration find the cheapest way to merge two trees.

## Kruskal's algorithm: example

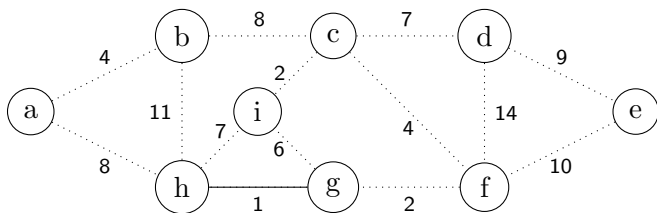


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
(g,i,6), (c,d,7), (h,i,7), (a,h,8), (b,c,8),  
(d,e,9), (e,f,10), (b,h,11), (d,f,14)]

Clusters: {a}, {b}, {c}, {d}, {e}, {f}, {g}, {h}, {i}

MST: { }

## Kruskal's algorithm: example



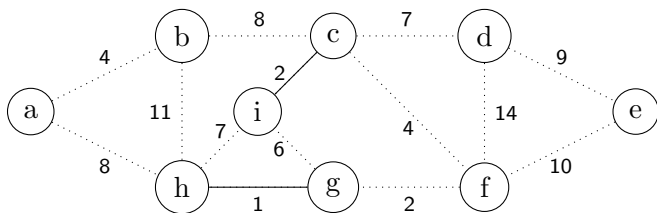
L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
(g,i,6), (c,d,7), (h,i,7), (a,h,8), (b,c,8),  
(d,e,9), (e,f,10), (b,h,11), (d,f,14)]

Clusters: {a}, {b}, {c}, {d}, {e}, {f}, {g,h}, {i}

MST: { (g,h), }



## Kruskal's algorithm: example

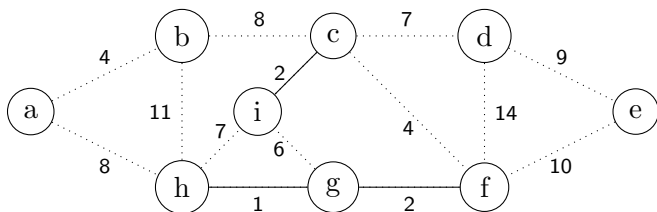


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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(d,e,9), (e,f,10), (b,h,11), (d,f,14)]

Clusters: {a}, {b}, {c,i}, {d}, {e}, {f}, {g,h}

MST: { (g,h), (c,i), }

## Kruskal's algorithm: example

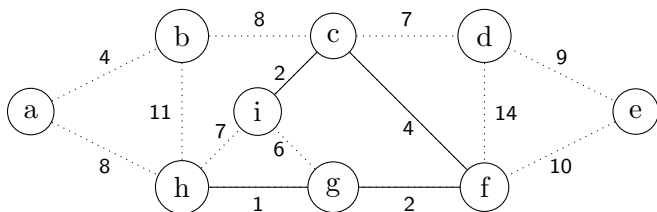


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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(d,e,9), (e,f,10), (b,h,11), (d,f,14)]

Clusters: {a}, {b}, {c,i}, {d}, {e}, {f,g,h}

MST: { (g,h), (c,i), (f,g), }

## Kruskal's algorithm: example

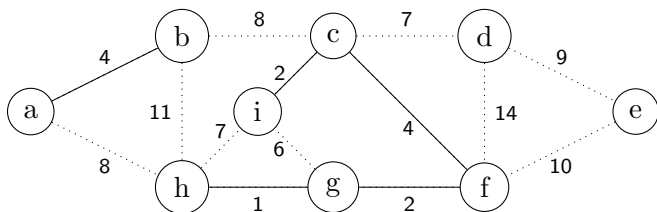


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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(d,e,9), (e,f,10), (b,h,11), (d,f,14)]

Clusters: {a}, {b}, {d}, {e}, {c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), }

## Kruskal's algorithm: example

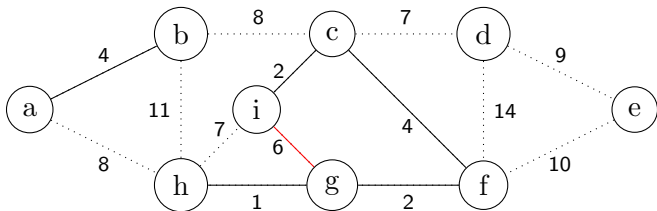


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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Clusters: {a,b}, {d}, {e}, {c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), (a,b), }

## Kruskal's algorithm: example

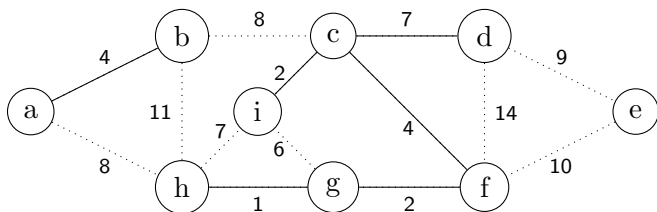


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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Clusters: {a,b}, {d}, {e}, {c,i,f,g,h}

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## Kruskal's algorithm: example

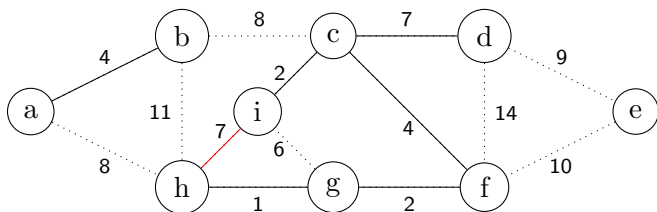


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Clusters: {a,b}, {e}, {d,c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), (a,b), (c,d), }

## Kruskal's algorithm: example

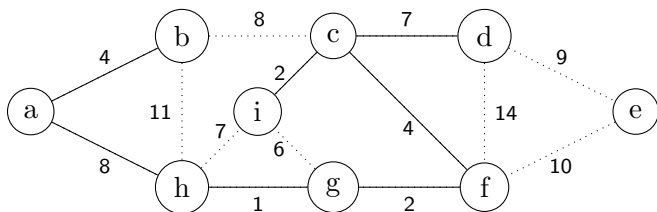


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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Clusters: {a,b}, {e}, {d,c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), (a,b), (c,d), }

## Kruskal's algorithm: example



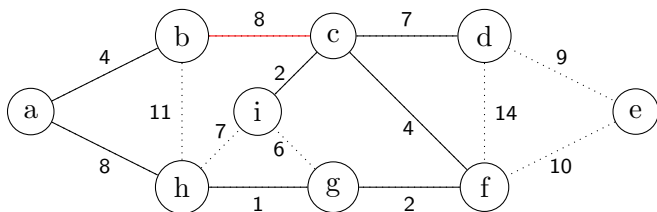
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(d,e,9), (e,f,10), (b,h,11), (d,f,14)]

Clusters: {e}, {a,b,d,c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), (a,b), (c,d), (a,h), }



## Kruskal's algorithm: example

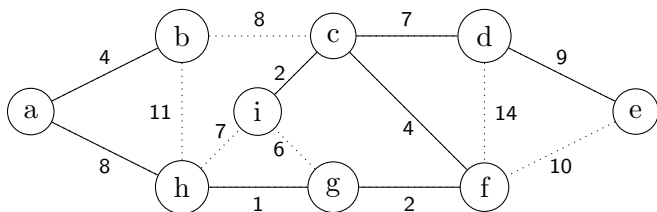


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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Clusters: {e}, {a,b,d,c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), (a,b), (c,d), (a,h), }

## Kruskal's algorithm: example

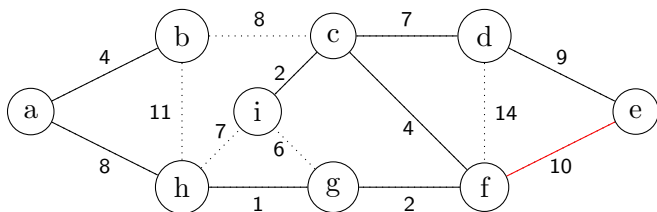


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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Clusters: {e,a,b,d,c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), (a,b), (c,d), (a,h), (d,e) }

## Kruskal's algorithm: example

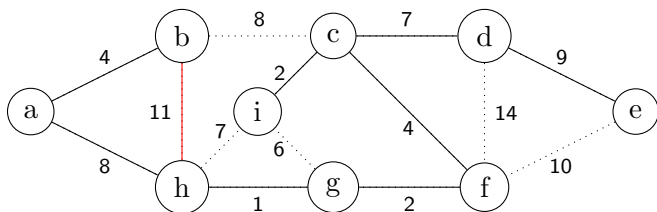


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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## Kruskal's algorithm: example

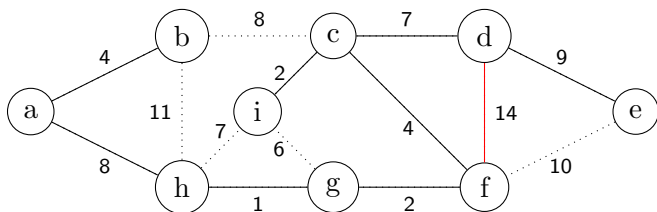


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Clusters: {e,a,b,d,c,i,f,g,h}

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## Kruskal's algorithm: example

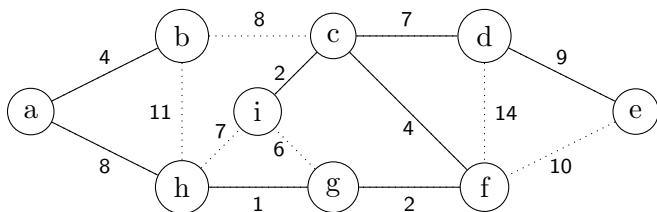


L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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Clusters: {e,a,b,d,c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), (a,b), (c,d), (a,h), (d,e) }

## Kruskal's algorithm: example



L: [(g,h,1), (c,i,2), (f,g,2), (c,f,4), (a,b,4),  
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Clusters: {e,a,b,d,c,i,f,g,h}

MST: { (g,h), (c,i), (f,g), (c,f), (a,b), (c,d), (a,h), (d,e) }

## Kruskal's algorithm

0.  $T :=$  new container for edges
1.  $L :=$  edges sorted in non-decreasing order by weight
2. for each vertex  $v$ :
3.     $v.\text{cluster} := \text{make-cluster}(v)$
4. for each  $(u, v)$  in  $L$ :
5.    if  $u.\text{cluster} \neq v.\text{cluster}$ :
6.      $T.\text{add}((u,v))$
7.     merge  $u.\text{cluster}$  and  $v.\text{cluster}$
8. return  $T$

## storing clusters

An easy way for now:

- each cluster is a linked list
- $v.\text{cluster}$  is pointer to  $v$ 's owning linked list
- $u.\text{cluster} \neq v.\text{cluster}$  is: pointer equality,  $\Theta(1)$  time
- merging two clusters is merging two linked lists:
  - a lot of vertices may need their  $v.\text{cluster}$ 's updated!



## storing clusters

An easy way for now, continued...

Choose to always move the smaller list to the larger one:

- in the best case: smaller list has one node: 1 update
- in the worst case: smaller list has (almost) as many nodes as larger list
- in the worst case: the size of cluster roughly doubles as a result
- then how many such merges can we do?
- each v.cluster is updated at most:  $\log |V|$  times

A much better way will appear later in this course.

## Kruskal's algorithm: time

Let  $n = |V|$  and  $m = |E|$ . Then:

- Collecting and sorting edges:  $\Theta(m \log m)$ .
- v.cluster updates:  $\mathcal{O}(\log n)$  per vertex, so  $\mathcal{O}(n \log n)$  total
- the rest is  $\Theta(1)$  per vertex or edge

Total:  $\mathcal{O}(n \log n + m \log m)$  time.

But lets look at  $n$  and  $m$ :

- maximum number of edges in a graph with  $n$  vertices:  
 $n(n-1)/2$
- then

$$\begin{aligned} m &\leq n(n-1)/2 \leq n^2 \\ \therefore \log m &\leq \log(n^2) = 2 \log n \\ \therefore \log m &\in \mathcal{O}(\log n) \end{aligned}$$

Then total time is  $\mathcal{O}((n+m) \log n)$ .

## Prim's algorithm: idea

Prim's algorithm finds a MST by a BFS with a twist:

- the queue is replaced with a minimum priority queue
- with an additional operation `decrease-priority(vertex, new-priority)`
  - **Exercise:** show that `decrease-priority` is  $\mathcal{O}(\log n)$  where  $n$  is the size of the priority queue

Keep unvisited vertices in the priority queue:

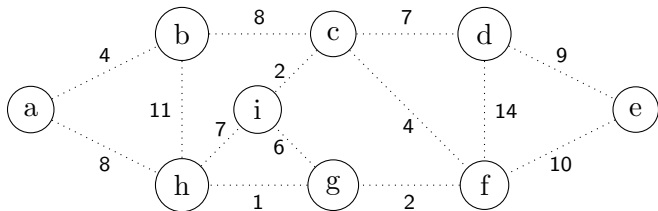
$priority(v)$  = minimum weight of any edge between  $v$  and tree

$priority(v) = \infty$  if no such edge

The algorithm grows a tree by one edge at a time.

Correctness idea: every time we `extract-min`, we get the cheapest edge to add to the tree.

## Prim's algorithm: example

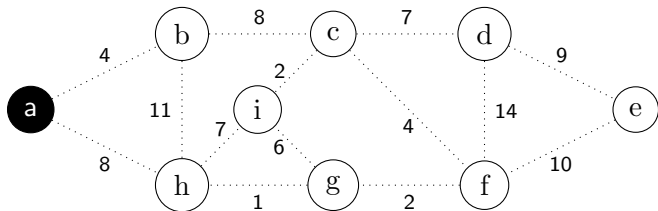


Priority queue contains vertices *not* in tree:

vertex	a	b	c	d	e	f	g	h	i
priority	0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
pred									

MST: {            }

## Prim's algorithm: example

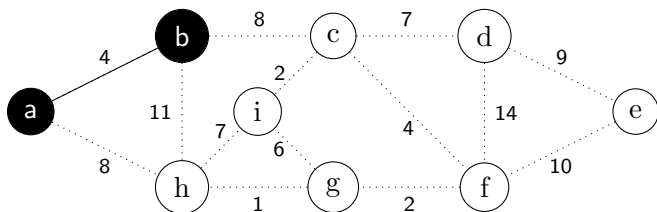


Priority queue contains vertices *not* in tree:

vertex	b	h	c	d	e	f	g	i
priority	4	8	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
pred	a	a						

MST: {            }

## Prim's algorithm: example

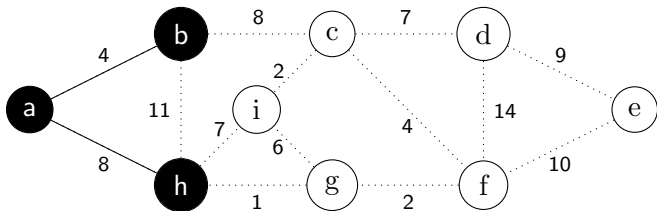


Priority queue contains vertices *not* in tree:

vertex	h	c	d	e	f	g	i
priority	8	8	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
pred	a	b					

MST: { (a,b),        }

## Prim's algorithm: example

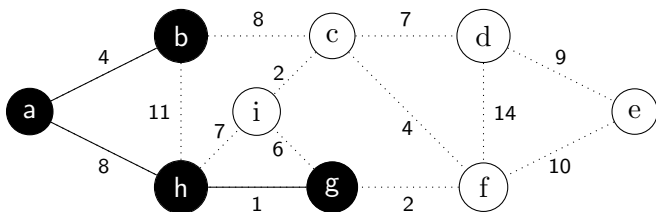


Priority queue contains vertices *not* in tree:

vertex	g	i	c	d	e	f
priority	1	7	8	$\infty$	$\infty$	$\infty$
pred	h	h	b			

MST: { (a,b), (a,h), }

## Prim's algorithm: example



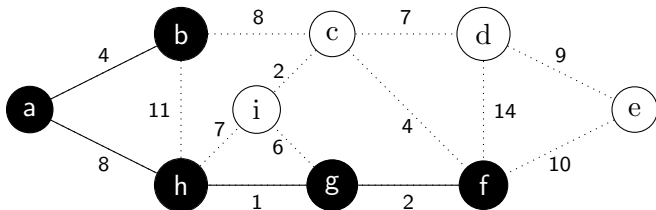
Priority queue contains vertices *not* in tree:

vertex	f	i	c	d	e
priority	2	6	8	$\infty$	$\infty$
pred	g	g	b		

MST: { (a,b), (a,h), (h,g),      }



## Prim's algorithm: example

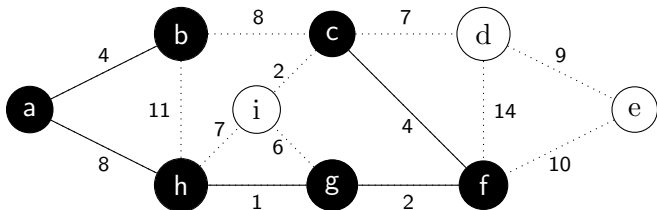


Priority queue contains vertices *not* in tree:

vertex	c	i	e	d
priority	4	6	10	14
pred	f	g	f	f

MST: { (a,b), (a,h), (h,g), (g,f), }

## Prim's algorithm: example

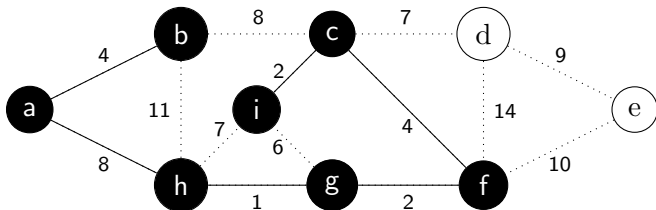


Priority queue contains vertices *not* in tree:

vertex	i	d	e
priority	2	7	10
pred	c	c	f

MST: { (a,b), (a,h), (h,g), (g,f), (c,f), }

## Prim's algorithm: example

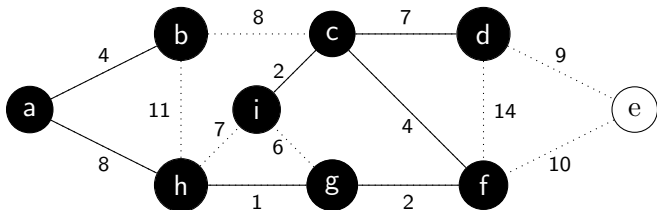


Priority queue contains vertices *not* in tree:

vertex	d	e
priority	7	10
pred	c	f

MST: { (a,b), (a,h), (h,g), (g,f), (c,f), (c,i), }

## Prim's algorithm: example

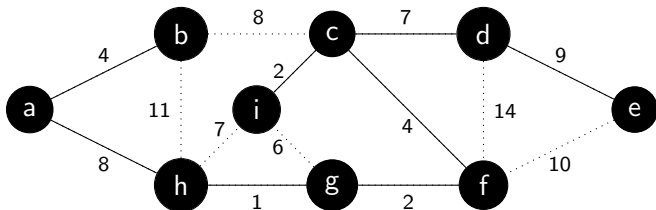


Priority queue contains vertices *not* in tree:

vertex	e
priority	9
pred	d

MST:  $\{ (a,b), (a,h), (h,g), (g,f), (c,f), (c,i), (c,d), \}$

## Prim's algorithm: example



Priority queue contains vertices *not* in tree:

vertex	
priority	
pred	

MST: { (a,b), (a,h), (h,g), (g,f), (c,f), (c,i), (c,d), (d,e) }

## Prim's algorithm

```
0. T := new container for edges
1. PQ := new min-heap()
2. start := pick a vertex
3. PQ.insert(0, start)
4. for each vertex v != start: PQ.insert(inf, v)
5. while not PQ.is-empty():
6.     u := PQ.extract-min()
7.     T.add((u.pred, u))
8.     for each v in u's adjacency list:
9.         if v in PQ and w(u, v) < priority(v):
10.            PQ.decrease-priority(v, w(u,v))
11.            v.pred := u
12. return T
```

## Prim's algorithm: time

Let  $n = |V|$  and  $m = |E|$ . Then:

- every vertex enters and leaves min-heap once
  - enters in the beginning only; continue until heap is empty
  - $\mathcal{O}(\log n)$  each, for a total of  $\mathcal{O}(n \log n)$
- with every edge may call decrease-priority
  - $\mathcal{O}(\log n)$  each, for a total of  $\mathcal{O}(m \log n)$
- the rest can be done in  $\Theta(1)$  per vertex or per edge

Total time worst case:  $\mathcal{O}((n + m) \log n)$

## Kruskal's algorithm

0.  $T :=$  new container for edges
1.  $L :=$  edges sorted in non-decreasing order by weight
2. for each vertex  $v$ :
3.     $v.\text{cluster} := \text{make-cluster}(v)$
4. for each  $(u, v)$  in  $L$ :
5.    if  $u.\text{cluster} \neq v.\text{cluster}$ :
6.      $T.\text{add}((u,v))$
7.     merge  $u.\text{cluster}$  and  $v.\text{cluster}$
8. return  $T$



## Kruskal's algorithm: correctness

Kruskal's algorithm maintains the loop invariants:

1. each cluster is a tree
2.  $T \subseteq T_{min}$  for some MST  $T_{min}$

Initially  $T$  is empty and clusters are single vertices, so trivially true.

Suppose (1) and (2) are true before line 4.

- on line 5, if  $u.cluster \neq v.cluster$ , then
- since  $u$ 's cluster is a tree and  $v$ 's cluster is a different tree,
- then the merged cluster (line 7) is a tree

## Kruskal's algorithm: correctness

Suppose (1) and (2) are true before line 4.

- if  $(u, v) \in T_{min}$ , then choose  $T'_{min} = T_{min}$  and done
- if  $(u, v) \notin T_{min}$ , then partition  $V$  into  $S$  and  $V - S$  such that  $u$ 's cluster  $\subseteq S$ ,  $v$ 's cluster  $\subseteq V - S$ , and no  $T$  edge between  $S$  and  $V - S$
- in  $T_{min}$  there is a unique simple path connecting  $u$  and  $v$
- in  $T_{min}$  there is some edge  $(u', v')$  connecting  $S$  and  $V - S$
- without  $(u', v')$ ,  $T_{min}$  disconnected;  $(u, v)$  would reconnect
- $(u, v)$  is the minimum-weight edge in  $L$  connecting two clusters
- $\therefore weight(u, v) \leq weight(u', v')$
- then choose  $T'_{min} = T_{min} - \{(u', v')\} + \{(u, v)\}$  is an MST

## Prim's algorithm

```
0. T := new container for edges
1. PQ := new min-heap()
2. start := pick a vertex
3. PQ.insert(0, start)
4. for each vertex v != start: PQ.insert(inf, v)
5. while not PQ.is-empty():
6.   u := PQ.extract-min()
7.   T.add((u.pred, u))
8.   for each v in u's adjacency list:
9.     if v in PQ and w(u, v) < priority(v):
10.      PQ.decrease-priority(v, w(u,v))
11.      v.pred := u
12. return T
```

## Prim's algorithm: correctness

Prim's algorithm maintains the loop invariants:

1.  $T$  contains vertices in  $V - PQ$
2. for each  $v$  in  $PQ$ ,  $priority(v)$  = minimum weight of any edge between  $v$  and  $T$
3.  $T \subseteq T_{min}$  for some MST  $T_{min}$

Initially  $T$  is empty,  $PQ$  contains all of  $V$ , and all priorities are  $\infty$ , so trivially true.

Suppose (1), (2), and (3) are true before line 5.

- line 6 extracts  $u$  from  $PQ$ , line 7 adds edge  $(u.pred, u)$  to  $T$ , so (1)
- lines 8-11 update priorities of vertices adjacent to  $u$ , so (2)

## Prim's algorithm: correctness

Suppose (1), (2), and (3) are true before line 5. Let  $p = u.pred$ .

- if  $(p, u) \in T_{min}$ , then choose  $T'_{min} = T_{min}$  and done
- if  $(p, u) \notin T_{min}$ , then in  $T_{min}$  there is a unique simple path connecting  $p$  and  $u$
- in  $T_{min}$  there is some edge  $(x, y)$  where  $x$  no longer in  $PQ$  and  $y$  in  $PQ$  on a path from  $p$  to  $u$
- without  $(x, y)$ ,  $T_{min}$  disconnected;  $(p, u)$  would reconnect
- $u$  was just extracted from  $PQ$ , so  
 $weight(p, u) = priority(u) \leq priority(y) = weight(x, y)$
- then choose  $T'_{min} = T_{min} - \{(x, y)\} + \{(p, u)\}$  is an MST

## General Theorem

Suppose

- $T \subseteq T_{min}$
- can partition  $V$  into  $S$  and  $V - S$  (cut), such that
  - no  $T$  edge between  $S$  and  $V - S$
  - $(u, v)$  is the cheapest edge (light edge) connecting  $V$  and  $V - S$  (crosses the cut)

Then  $T + \{(u, v)\} \subseteq T'_{min}$

- if  $(u, v) \notin T_{min}$
- $T_{min}$  has a unique simple path from  $u$  to  $v$ , via some edge  $(u', v')$  with  $u' \in S$  and  $v' \in V - S$
- $T_{min}$  without  $(u', v')$  disconnected;  $(u, v)$  would reconnect
- $weight(u, v) \leq weight(u', v')$
- Choose  $T'_{min} = T_{min} - \{(u', v')\} + \{(u, v)\}$