

Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

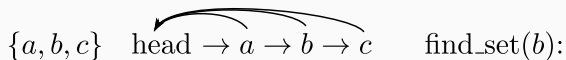
Using linked list:



Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

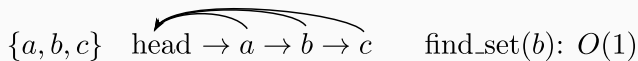
Using linked list:



Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

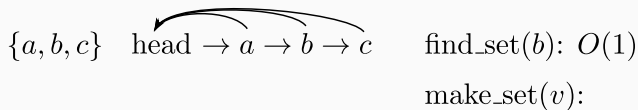
Using linked list:



Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

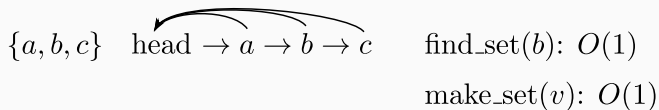
Using linked list:



Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

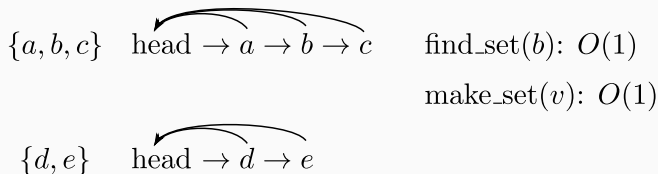
Using linked list:



Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

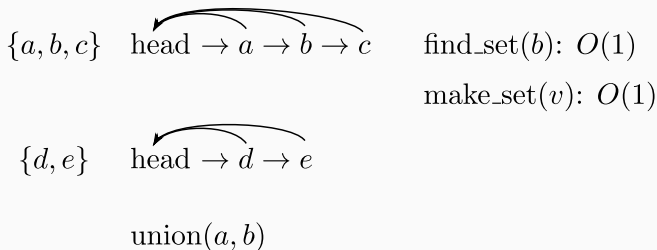
Using linked list:



Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

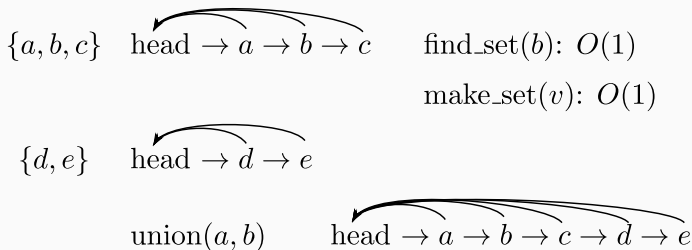
Using linked list:



Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

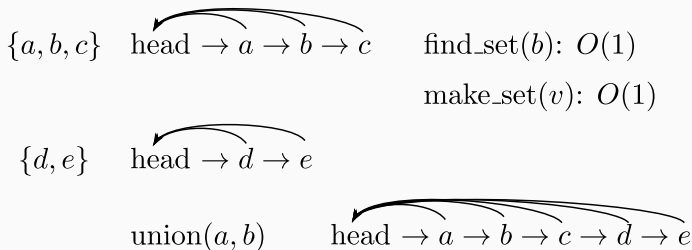
Using linked list:



Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

Using linked list:

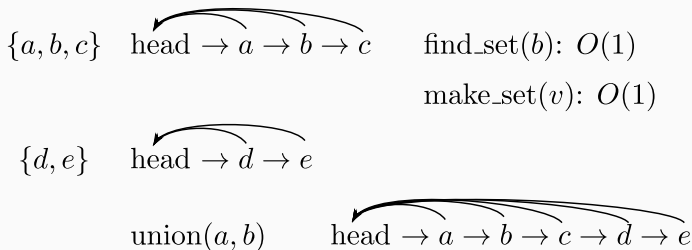


Cost of union:

Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

Using linked list:

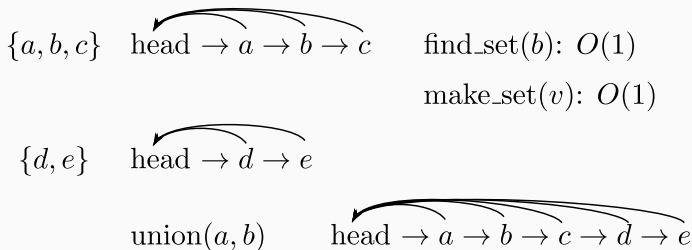


Cost of union: $O(\text{length of the shorter list})$

Running time of Kruskal's algorithm (I)

Depends on how we implement `make_set`, `find_set`, and `union`

Using linked list:



Cost of union: $O(\text{length of the shorter list})$

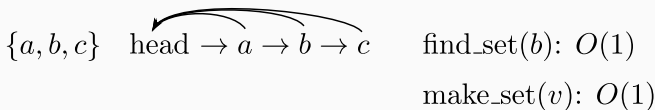
Using an array to implement it:

vertex	1	2	3	4	5	$\xrightarrow{\text{union}}$
head	1	1	1	4	4	

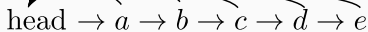
Running time of Kruskal's algorithm (I)

Depends on how we implement ~~make_set~~, ~~find_set~~, and ~~union~~

Using linked list:



$\text{union}(a, b)$



Cost of union: $O(\text{length of the shorter list})$

Using an array to implement it:

vertex	1	2	3	4	5	$\xrightarrow{\text{union}}$	1	2	3	4	5
head	1	1	1	4	4		1	1	1	1	1

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):
2     Set  $A := \{\}$ ;
3     for  $v \in V$ :
4          $\lfloor$  make_set( $v$ )
5     Sort  $E$  in increasing order of edge weights
6     for  $(u, v) \in E$ :
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):
8              $A := A \cup \{(u, v)\}$ ;
9              $\lfloor$  union( $u, v$ );
10     $\lfloor$ 
```

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):
2     Set  $A := \{\}$ ;
3     for  $v \in V$ :
4         make_set( $v$ );
5     Sort  $E$  in increasing order of edge weights
6     for  $(u, v) \in E$ :
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):
8              $A := A \cup \{(u, v)\}$ ;
9             union( $u, v$ );
```

$|E| = O(|V|^2)$

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

$O(|V|)$ worst-case } worse case $O(|E| \cdot |V|)$

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):
2     Set  $A := \{\}$ ;
3     for  $v \in V$ :
4          $\lfloor$  make_set( $v$ ) ; //  $O(|V|)$ 
5     Sort  $E$  in increasing order of edge weights ; //  $O(|E| \log |V|)$ 
6     for  $(u, v) \in E$ :
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):
8              $A := A \cup \{(u, v)\}$ ;
9              $\lfloor$  union( $u, v$ );
```


Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4          $\lfloor$  make_set( $v$ ) ; //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ; //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9              $\lfloor$  union( $u, v$ );  
         $\lfloor$ 
```

Worst-case cost for union:

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4          $\lfloor$  make_set( $v$ ) ; //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ; //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9              $\lfloor$  union( $u, v$ );  
         $\lfloor$ 
```

Worst-case cost for union: $O(|V|)$.

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4          $\lfloor$  make_set( $v$ ) ; //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ; //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9              $\lfloor$  union( $u, v$ );
```

Worst-case cost for union: $O(|V|)$. What about the cost for lines 6-9?

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4         make_set( $v$ ) ;                               //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ;    //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9             union( $u, v$ );
```

Worst-case cost for union: $O(|V|)$. What about the cost for lines 6-9?

Consider a single $v \in V$. Once it's touched in some union operation, the size of the set at least doubles.

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4         make_set( $v$ ) ;                                //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ;      //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9             union( $u, v$ );
```

Worst-case cost for union: $O(|V|)$. What about the cost for lines 6-9?

Consider a single $v \in V$. Once it's touched in some union operation, the size of the set at least doubles. Since the maximum size of a set can be $|V|$, each v is touched at most $O(\log |V|)$ times

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4         make_set( $v$ ) ;                                //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ;      //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9             union( $u, v$ );
```

Worst-case cost for union: $O(|V|)$. What about the cost for lines 6-9?

Consider a single $v \in V$. Once it's touched in some union operation, the size of the set at least doubles. Since the maximum size of a set can be $|V|$, each v is touched at most $O(\log |V|)$ times

At most $|V|$ vertices are involved in union operations, so the total cost of lines 6-9:

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4         make_set( $v$ ) ;                               //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ;      //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9             union( $u, v$ );
```

Worst-case cost for union: $O(|V|)$. What about the cost for lines 6-9?

Consider a single $v \in V$. Once it's touched in some union operation, the size of the set at least doubles. Since the maximum size of a set can be $|V|$, each v is touched at most $O(\log |V|)$ times

At most $|V|$ vertices are involved in union operations, so the total cost of lines 6-9: $O(|V| \log |V|)$

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4         make_set( $v$ ) ; //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ; //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9             union( $u, v$ );
```

Worst-case cost for union: $O(|V|)$. What about the cost for lines 6-9?

Consider a single $v \in V$. Once it's touched in some union operation, the size of the set at least doubles. Since the maximum size of a set can be $|V|$, each v is touched at most $O(\log |V|)$ times

At most $|V|$ vertices are involved in union operations, so the total cost of lines 6-9: $O(|V| \log |V|)$

Total cost of the algorithm: $O(|V|) + O(|E| \log |V|) + O(|V| \log |V|)$

Running time of Kruskal's algorithm (II)

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2     Set  $A := \{\}$ ;  
3     for  $v \in V$ :  
4         make_set( $v$ ) ;                               //  $O(|V|)$   
5     Sort  $E$  in increasing order of edge weights ;    //  $O(|E| \log |V|)$   
6     for  $(u, v) \in E$ :  
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8              $A := A \cup \{(u, v)\}$ ;  
9             union( $u, v$ );
```

Worst-case cost for union: $O(|V|)$. What about the cost for lines 6-9?

Consider a single $v \in V$. Once it's touched in some union operation, the size of the set at least doubles. Since the maximum size of a set can be $|V|$, each v is touched at most $O(\log |V|)$ times

At most $|V|$ vertices are involved in union operations, so the total cost of lines 6-9: $O(|V| \log |V|)$

Total cost of the algorithm: $O(|E| \log |V|)$

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:

$$\{a\}$$

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:

$$\{a\} \hookrightarrow a$$

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

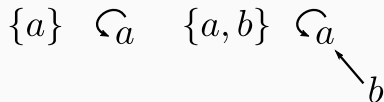
Directed tree disjoint set:

$$\{a\} \hookrightarrow_a \{a, b\}$$

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

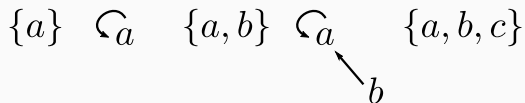
Directed tree disjoint set:



Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

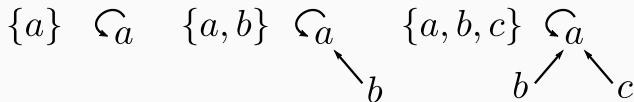
Directed tree disjoint set:



Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

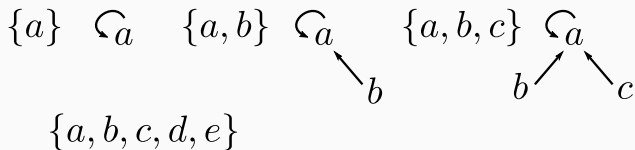
Directed tree disjoint set:



Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

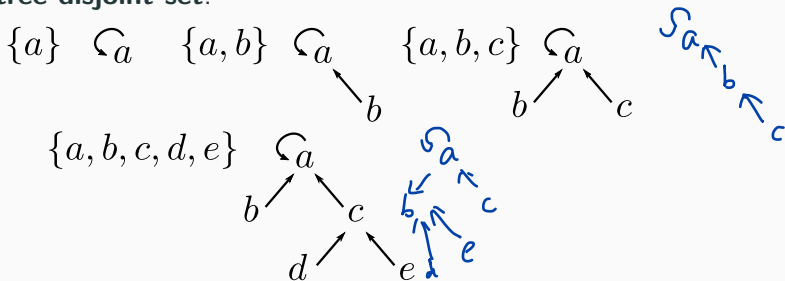
Directed tree disjoint set:



Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

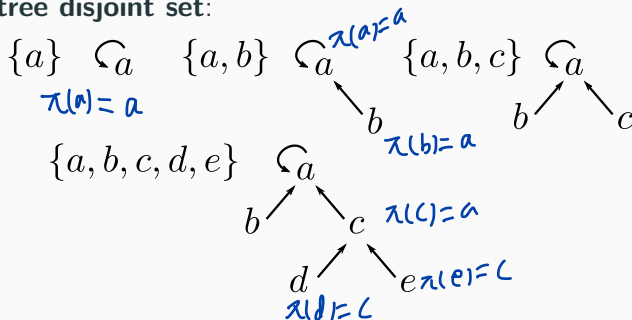
Directed tree disjoint set:



Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:



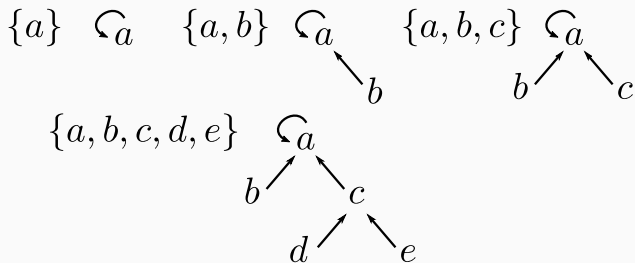
Definition

$\pi(x)$: parent of x

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:



Definition

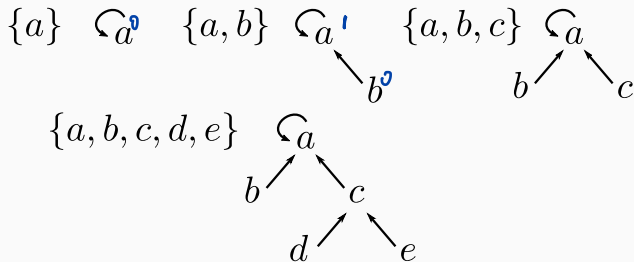
$\pi(x)$: parent of x

root node: x s.t. $\pi(x) = x$

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:



Definition

$\pi(x)$: parent of x

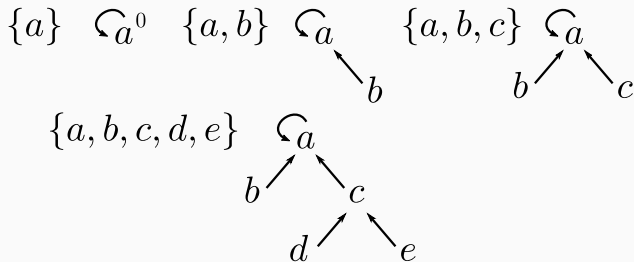
root node: x s.t. $\pi(x) = x$

$\text{rank}(x)$: number of the edges in the longest simple path from x to a leaf

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:



Definition

$\pi(x)$: parent of x

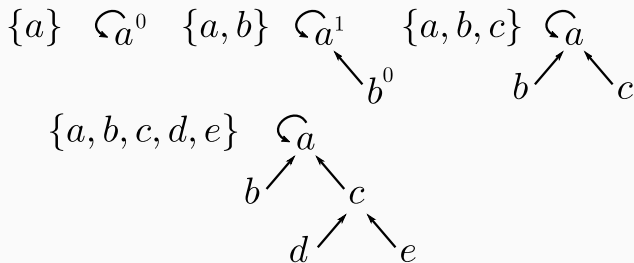
root node: x s.t. $\pi(x) = x$

$\text{rank}(x)$: number of the edges in the longest simple path from x to a leaf

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:



Definition

$\pi(x)$: parent of x

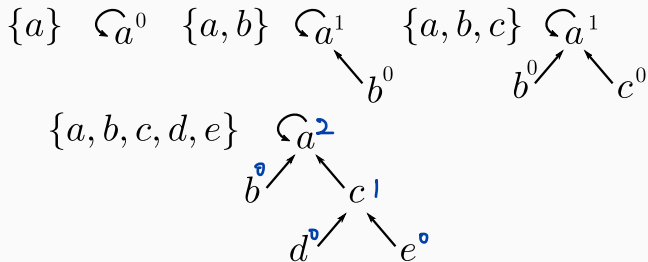
root node: x s.t. $\pi(x) = x$

$\text{rank}(x)$: number of the edges in the longest simple path from x to a leaf

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:



Definition

$\pi(x)$: parent of x

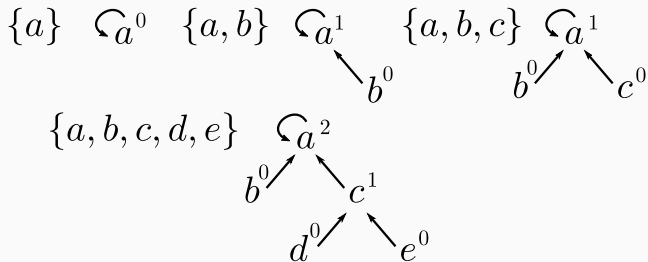
root node: x s.t. $\pi(x) = x$

$\text{rank}(x)$: number of the edges in the longest simple path from x to a leaf

Alternative data structure

The linked-list implementation is good enough, but there exist better data structures to improve the **worst-case** cost for union

Directed tree disjoint set:



Definition

$\pi(x)$: parent of x

root node: x s.t. $\pi(x) = x$

$\text{rank}(x)$: number of the edges in the longest simple path from x to a leaf

Operations of direct tree disjoint set (I)

- `make_set(v)`

Operations of direct tree disjoint set (I)

- `make_set(v)`

def `make_set(v):`

$\pi(v) := v;$

$\text{rank}(v) = 0;$

$\hookrightarrow v^0$

Operations of direct tree disjoint set (I)

- `make_set(v)`

def `make_set(v):`

```
     $\pi(v) := v;$   
     $\text{rank}(v) = 0;$ 
```

Cost: $O(1)$

Operations of direct tree disjoint set (I)

- `make_set(v)`

def `make_set(v):`

```
     $\pi(v) := v;$   
     $\text{rank}(v) = 0;$ 
```

Cost: $O(1)$

- `find_set(v)`

Operations of direct tree disjoint set (I)

- `make_set(v)`

```
def make_set( $v$ ):
```

```
     $\pi(v) := v$ ;
```

```
    rank( $v$ ) = 0;
```

Cost: $O(1)$

- `find_set(v)`

```
def find_set( $v$ ):
```

```
    while  $v \neq \pi(v)$ :
```

```
         $v := \pi(v)$ ;
```

```
    return  $v$ ;
```

Operations of direct tree disjoint set (I)

- `make_set(v)`

```
def make_set( $v$ ):
```

```
     $\pi(v) := v$ ;
```

```
    rank( $v$ ) = 0;
```

Cost: $O(1)$

- `find_set(v)`

```
def find_set( $v$ ):
```

```
    while  $v \neq \pi(v)$ :
```

```
         $v := \pi(v)$ ;
```

```
    return  $v$ ;
```

Cost: $O(\text{depth of the node in the tree})$

Operations of direct tree disjoint set (I)

- `make_set(v)`

```
def make_set( $v$ ):
```

```
     $\pi(v) := v$ ;
```

```
    rank( $v$ ) = 0;
```

Cost: $O(1)$

- `find_set(v)`

```
def find_set( $v$ ):
```

```
    while  $v \neq \pi(v)$ :
```

```
         $v := \pi(v)$ ;
```

```
    return  $v$ ;
```

Cost: $O(\text{depth of the node in the tree})$

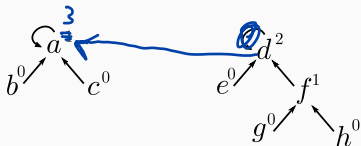
- what about union?

Operations of direct tree disjoint set (II)

- union:

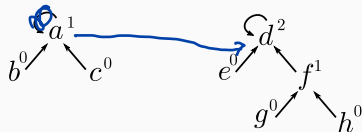
Operations of direct tree disjoint set (II)

- union:

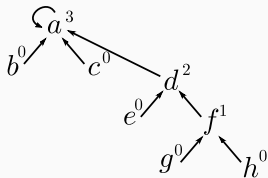


Operations of direct tree disjoint set (II)

- union:

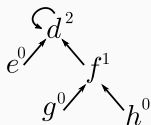
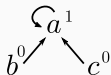


Option 1

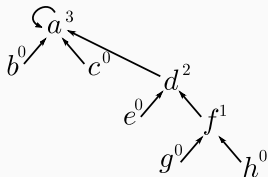


Operations of direct tree disjoint set (II)

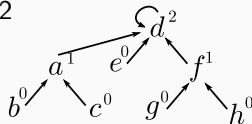
- union:



Option 1

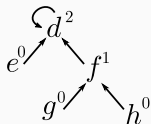
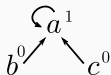


Option 2

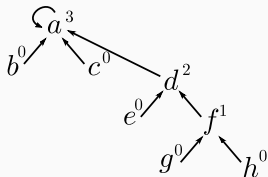


Operations of direct tree disjoint set (II)

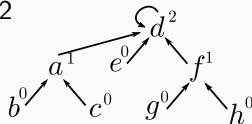
- union:



Option 1



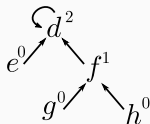
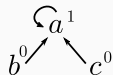
Option 2



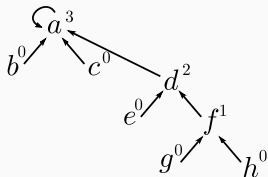
better!

Operations of direct tree disjoint set (II)

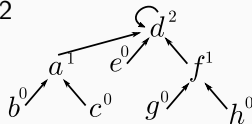
- union:



Option 1



Option 2



better!

Basic idea: attach the smaller ranked tree to a larger one

Operations of direct tree disjoint set (II)

```
def union( $x, y$ ):
```


Operations of direct tree disjoint set (II)

def union(x, y):

$r_x := \text{find_set}(x)$, $r_y := \text{find_set}(y)$;

Operations of direct tree disjoint set (II)

```
def union( $x, y$ ):  
     $r_x := \text{find\_set}(x), r_y := \text{find\_set}(y);$   
    if rank( $r_x$ ) > rank( $r_y$ ):  
        |
```

Operations of direct tree disjoint set (II)

```
def union( $x, y$ ):  
     $r_x := \text{find\_set}(x), r_y := \text{find\_set}(y);$   
    if rank( $r_x$ ) > rank( $r_y$ ):  
         $\pi(r_y) := r_x;$ 
```

Operations of direct tree disjoint set (II)

```
def union( $x, y$ ):  
     $r_x := \text{find\_set}(x), r_y := \text{find\_set}(y);$   
    if rank( $r_x$ ) > rank( $r_y$ ):  
         $\pi(r_y) := r_x;$   
    else:  
        |
```

Operations of direct tree disjoint set (II)

```
def union( $x, y$ ):  
     $r_x := \text{find\_set}(x), r_y := \text{find\_set}(y);$   
    if rank( $r_x$ ) > rank( $r_y$ ):  
         $\pi(r_y) := r_x;$   
    else:  
         $\pi(r_x) := r_y;$ 
```

Operations of direct tree disjoint set (II)

def union(x, y):

$r_x := \text{find_set}(x), r_y := \text{find_set}(y);$

if rank(r_x) > rank(r_y):

$\pi(r_y) := r_x;$

else:

$\pi(r_x) := r_y;$

if rank(r_x) == rank(r_y):

Operations of direct tree disjoint set (II)

```
def union( $x, y$ ):  
     $r_x := \text{find\_set}(x), r_y := \text{find\_set}(y);$   
    if rank( $r_x$ ) > rank( $r_y$ ):  
         $\pi(r_y) := r_x;$   
    else:  
         $\pi(r_x) := r_y;$   
        if rank( $r_x$ ) == rank( $r_y$ ):  
            rank( $r_y$ ) := rank( $r_y$ ) + 1;
```

Operations of direct tree disjoint set (II)

def union(x, y):

$r_x := \text{find_set}(x)$, $r_y := \text{find_set}(y)$;

if rank(r_x) > rank(r_y):

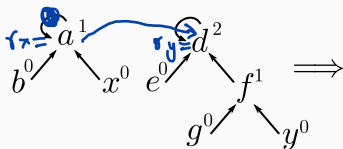
$\pi(r_y) := r_x$;

else:

$\pi(r_x) := r_y$;

if rank(r_x) == rank(r_y):

$\text{rank}(r_y) := \text{rank}(r_y) + 1$;



Operations of direct tree disjoint set (II)

def union(x, y):

$r_x := \text{find_set}(x)$, $r_y := \text{find_set}(y)$;

if rank(r_x) > rank(r_y):

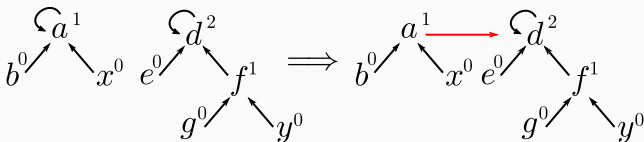
$\pi(r_y) := r_x$;

else:

$\pi(r_x) := r_y$;

if rank(r_x) == rank(r_y):

$\text{rank}(r_y) := \text{rank}(r_y) + 1$;



Operations of direct tree disjoint set (II)

def union(x, y):

$r_x := \text{find_set}(x), r_y := \text{find_set}(y);$

if rank(r_x) > rank(r_y):

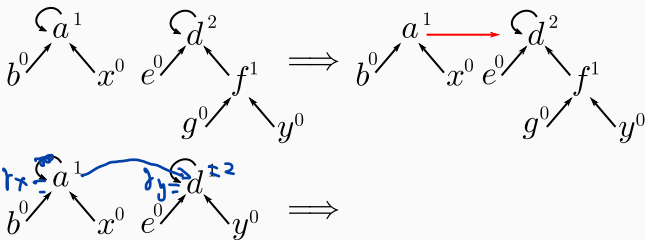
$\pi(r_y) := r_x;$

else:

$\pi(r_x) := r_y;$

if rank(r_x) == rank(r_y):

$\text{rank}(r_y) := \text{rank}(r_y) + 1;$



Operations of direct tree disjoint set (II)

def union(x, y):

$r_x := \text{find_set}(x)$, $r_y := \text{find_set}(y)$;

if rank(r_x) > rank(r_y):

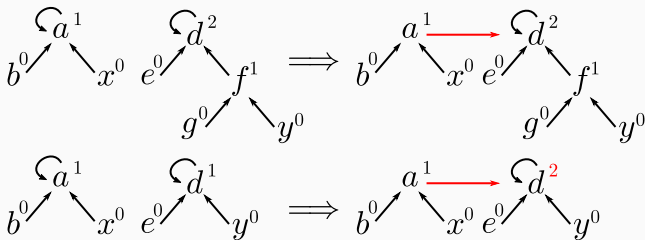
└ $\pi(r_y) := r_x$;

else:

└ $\pi(r_x) := r_y$;

└ **if** rank(r_x) == rank(r_y):

└└ rank(r_y) := rank(r_y) + 1;



Operations of direct tree disjoint set (II)

def union(x, y):

$r_x := \text{find_set}(x), r_y := \text{find_set}(y);$

if rank(r_x) > rank(r_y):

└ $\pi(r_y) := r_x;$

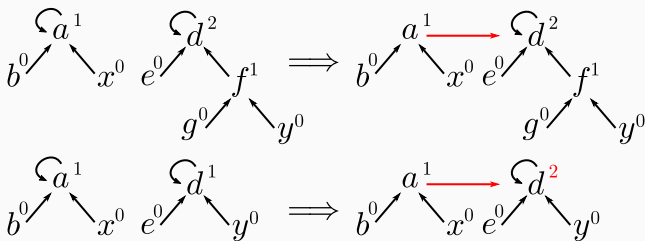
else:

└ $\pi(r_x) := r_y;$

└ **if** rank(r_x) == rank(r_y):

└ └ rank(r_y) := rank(r_y) + 1;

Cost: dominated by find_set



Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

Any root node of rank k has at least 2^k nodes in it

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

Any root node of rank k has at least 2^k nodes in it

Proof.

By induction: base case has $k = 0$ and $2^0 = 1$.

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

Any root node of rank k has at least 2^k nodes in it

Proof.

By induction: base case has $k = 0$ and $2^0 = 1$.

Assume the statement is true for $k - 1$.

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

$\nearrow^{\log |v|}$ $\llcorner^{|v|}$

Any root node of rank k has at least 2^k nodes in it

Proof.

By induction: base case has $k = 0$ and $2^0 = 1$.

Assume the statement is true for $k - 1$. By observation: after merging, the number of nodes is $\geq 2^{k-1} + 2^{k-1} = 2^k$ □

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

Any root node of rank k has at least 2^k nodes in it

Proof.

By induction: base case has $k = 0$ and $2^0 = 1$.

Assume the statement is true for $k - 1$. By observation: after merging, the number of nodes is $\geq 2^{k-1} + 2^{k-1} = 2^k$ □

By the lemma, if we have $|V|$ nodes, the maximum rank is $\log |V|$. So

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

Any root node of rank k has at least 2^k nodes in it

Proof.

By induction: base case has $k = 0$ and $2^0 = 1$.

Assume the statement is true for $k - 1$. By observation: after merging, the number of nodes is $\geq 2^{k-1} + 2^{k-1} = 2^k$ □

By the lemma, if we have $|V|$ nodes, the maximum rank is $\log |V|$. So

- the cost of find_set:

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

Any root node of rank k has at least 2^k nodes in it

Proof.

By induction: base case has $k = 0$ and $2^0 = 1$.

Assume the statement is true for $k - 1$. By observation: after merging, the number of nodes is $\geq 2^{k-1} + 2^{k-1} = 2^k$ □

By the lemma, if we have $|V|$ nodes, the maximum rank is $\log |V|$. So

- the cost of find_set: $O(\log |V|)$

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

Any root node of rank k has at least 2^k nodes in it

Proof.

By induction: base case has $k = 0$ and $2^0 = 1$.

Assume the statement is true for $k - 1$. By observation: after merging, the number of nodes is $\geq 2^{k-1} + 2^{k-1} = 2^k$ □

By the lemma, if we have $|V|$ nodes, the maximum rank is $\log |V|$. So

- the cost of find_set: $O(\log |V|)$
- the cost of union:

Cost of find_set using directed tree disjoint set

Observation

Root node with rank k is formed by the merge of two rank $k - 1$ trees

Lemma

Any root node of rank k has at least 2^k nodes in it

Proof.

By induction: base case has $k = 0$ and $2^0 = 1$.

Assume the statement is true for $k - 1$. By observation: after merging, the number of nodes is $\geq 2^{k-1} + 2^{k-1} = 2^k$ □

By the lemma, if we have $|V|$ nodes, the maximum rank is $\log |V|$. So

- the cost of find_set: $O(\log |V|)$
- the cost of union: $O(\log |V|)$

Total running time of Kruskal using directed tree disjoint set

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):
2   Set  $A := \{\}$ ;
3   for  $v \in V$ :
4     make_set( $v$ ) ; //  $O(|V|)$ 
5   Sort  $E$  in increasing order of edge weights ; //  $O(|E| \log |V|)$ 
6   for  $(u, v) \in E$ :
7     if find_set( $u$ )  $\neq$  find_set( $v$ ):
8        $A := A \cup \{(u, v)\}$ ;
9       union( $u, v$ );  $\rightarrow \log |v|$ 
```

$\left. \begin{array}{l} \rightarrow \log |v| \\ \log |v| \end{array} \right\} O(E / \log |V|)$

Total running time of Kruskal using directed tree disjoint set

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):
2   Set  $A := \{\}$ ;
3   for  $v \in V$ :
4     | make_set( $v$ ) ;                                //  $O(|V|)$ 
5   Sort  $E$  in increasing order of edge weights ;      //  $O(|E| \log |V|)$ 
6   for  $(u, v) \in E$ :
7     | if find_set( $u$ )  $\neq$  find_set( $v$ ):
8       | |  $A := A \cup \{(u, v)\}$ ;
9       | | union( $u, v$ );
```

Lines 6-9:

Total running time of Kruskal using directed tree disjoint set

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2   Set  $A := \{\}$ ;  
3   for  $v \in V$ :  
4     | make_set( $v$ ) ;                                //  $O(|V|)$   
5   Sort  $E$  in increasing order of edge weights ;      //  $O(|E| \log |V|)$   
6   for  $(u, v) \in E$ :  
7     | if find_set( $u$ )  $\neq$  find_set( $v$ ):  
8       |    $A := A \cup \{(u, v)\}$ ;  
9       |   union( $u, v$ );
```

Lines 6-9: $O(|E| \log |V|)$

Total running time of Kruskal using directed tree disjoint set

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):
2     Set  $A := \{\}$ ;
3     for  $v \in V$ :
4          $\lfloor$  make_set( $v$ ) ; //  $O(|V|)$ 
5     Sort  $E$  in increasing order of edge weights ; //  $O(|E| \log |V|)$ 
6     for  $(u, v) \in E$ :
7         if find_set( $u$ )  $\neq$  find_set( $v$ ):
8              $A := A \cup \{(u, v)\}$ ;
9              $\lfloor$  union( $u, v$ );
```

Lines 6-9: $O(|E| \log |V|)$

Total cost:

Total running time of Kruskal using directed tree disjoint set

```
1 def KRUSKAL_MST(undirected  $G = (V, E)$ , weights  $w = (w_e)_{e \in E}$ ):  
2   Set  $A := \{\}$ ;  
3   for  $v \in V$ :  
4     make_set( $v$ ) ; //  $O(|V|)$   
5   Sort  $E$  in increasing order of edge weights ; //  $O(|E| \log |V|)$ 
```

```
6   for  $(u, v) \in E$ :
```

```
7     if find_set( $u$ )  $\neq$  find_set( $v$ ):
```

```
8        $A := A \cup \{(u, v)\}$ ;
```

```
9       union( $u, v$ );
```

Lines 6-9: $O(|E| \log |V|)$

Total cost: $O(|E| \log |V|)$

	linked list	directed tree disjoint set
make_set	$O(1)$	$O(1)$
find_set	$O(1)$	$O(\log V)$
union	$O(V)$	$O(\log V)$

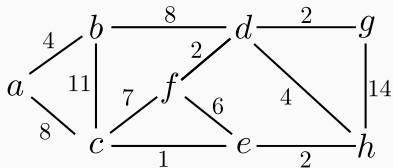
Prim's algorithm

Intuition: iteratively grows the tree

Prim's algorithm

Intuition: iteratively grows the tree

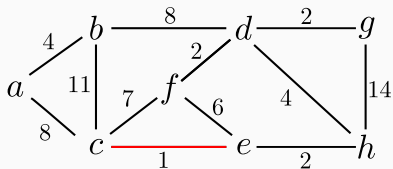
Kruskal



Prim's algorithm

Intuition: iteratively grows the tree

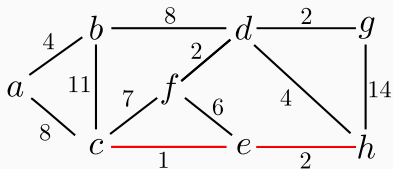
Kruskal



Prim's algorithm

Intuition: iteratively grows the tree

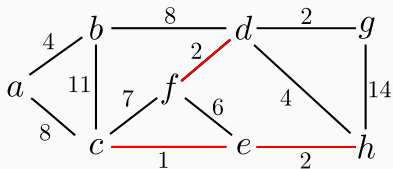
Kruskal



Prim's algorithm

Intuition: iteratively grows the tree

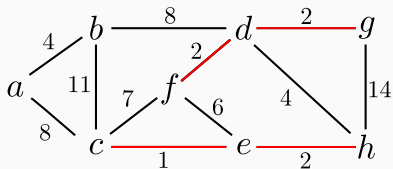
Kruskal



Prim's algorithm

Intuition: iteratively grows the tree

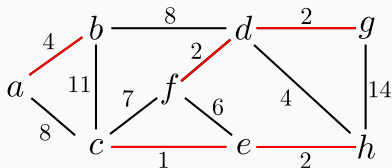
Kruskal



Prim's algorithm

Intuition: iteratively grows the tree

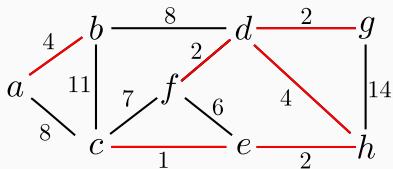
Kruskal



Prim's algorithm

Intuition: iteratively grows the tree

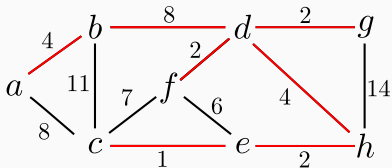
Kruskal



Prim's algorithm

Intuition: iteratively grows the tree

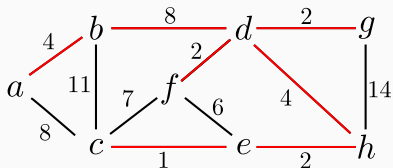
Kruskal



Prim's algorithm

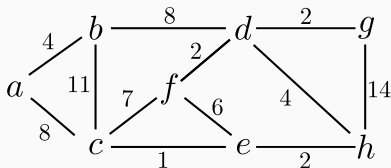
Intuition: iteratively grows the tree

Kruskal



Prim

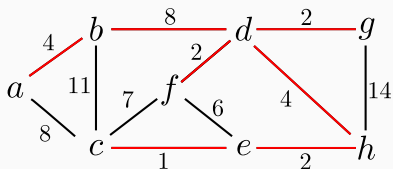
starting with c



Prim's algorithm

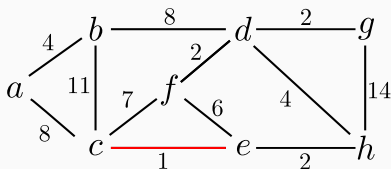
Intuition: iteratively grows the tree

Kruskal



Prim

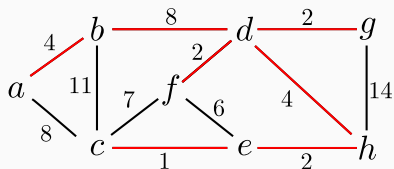
starting with c



Prim's algorithm

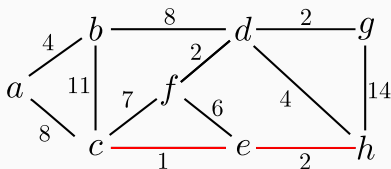
Intuition: iteratively grows the tree

Kruskal



Prim

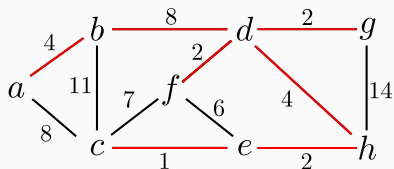
starting with c



Prim's algorithm

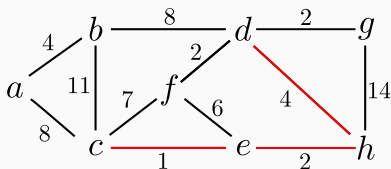
Intuition: iteratively grows the tree

Kruskal



Prim

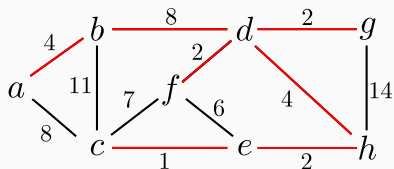
starting with c



Prim's algorithm

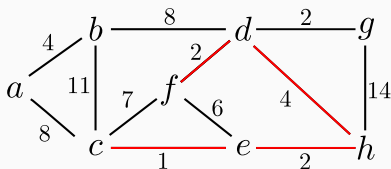
Intuition: iteratively grows the tree

Kruskal



Prim

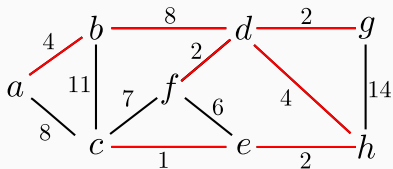
starting with c



Prim's algorithm

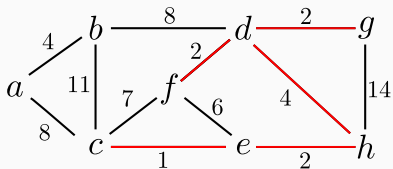
Intuition: iteratively grows the tree

Kruskal



Prim

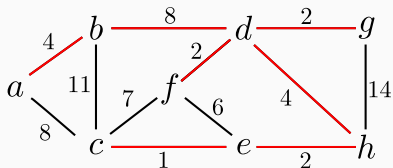
starting with c



Prim's algorithm

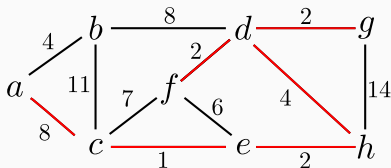
Intuition: iteratively grows the tree

Kruskal



Prim

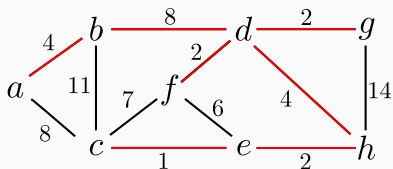
starting with c



Prim's algorithm

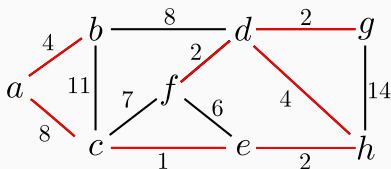
Intuition: iteratively grows the tree

Kruskal



Prim

starting with c



Prim's algorithm: pseudocode

Let S be the set included in the tree so far

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$$



Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev}(v) = \text{nil}$;

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

$\text{cost}(u_0) := 0$;

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

$\text{cost}(u_0) := 0$;

$H := \text{make_queue}(V)$;

// keys are $\text{cost}(v)$

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

$\text{cost}(u_0) := 0$;

$H := \text{make_queue}(V)$;

// keys are $\text{cost}(v)$

while H is not empty:

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

$\text{cost}(u_0) := 0$;

$H := \text{make_queue}(V)$;

// keys are $\text{cost}(v)$

while H is not empty:

$v = \text{delete_min}(H)$;

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

$\text{cost}(u_0) := 0$;

$H := \text{make_queue}(V)$;

// keys are $\text{cost}(v)$

while H is not empty:

$v = \text{delete_min}(H)$;

for $e := (v, z) \in E$:

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

$\text{cost}(u_0) := 0$;

$H := \text{make_queue}(V)$;

// keys are $\text{cost}(v)$

while H is not empty:

$v = \text{delete_min}(H)$;

for $e := (v, z) \in E$:

if $\text{cost}(z) > w_e$:

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

$\text{cost}(u_0) := 0$;

$H := \text{make_queue}(V)$;

// keys are $\text{cost}(v)$

while H is not empty:

$v = \text{delete_min}(H)$;

for $e := (v, z) \in E$:

if $\text{cost}(z) > w_e$:

$\text{cost}(z) := w_e$;

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

$\text{cost}(v) := \min_{e=(u,v) \text{ s.t. } u \in S} w_e$ and $\text{prev}(\cdot)$ is used to keep track of the tree

def PRIM_MST(*undirected* $G = (V, E)$, *weights* $w = (w_e)_{e \in E}$):

for $v \in V$:

$\text{cost}(v) := \infty$;

$\text{prev} := \text{nil}$;

 Pick any initial vertex u_0 ;

$\text{cost}(u_0) := 0$;

$H := \text{make_queue}(V)$;

// keys are $\text{cost}(v)$

while H is not empty:

$v = \text{delete_min}(H)$;

for $e := (v, z) \in E$:

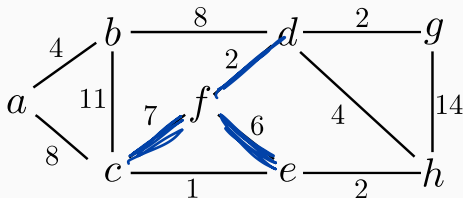
if $\text{cost}(z) > w_e$:

$\text{cost}(z) := w_e$;

$\text{prev}(z) := v$;

Prim's algorithm: a running example

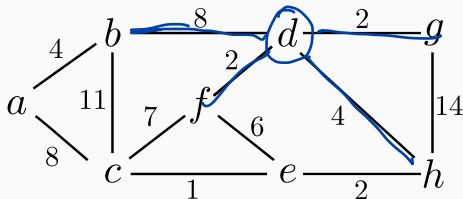
Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil 7 f	∞/nil 2 f	∞/nil 6 f	0/ nil	∞/nil	∞/nil

Prim's algorithm: a running example

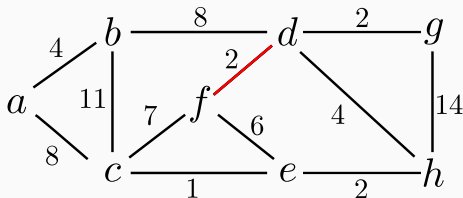
Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	0/ nil	∞/nil	∞/nil
f	∞/nil	∞/nil 8/ d	7/ f	∞/nil 2/ f	6/ f		∞/nil 2/ d	∞/nil 4/ d

Prim's algorithm: a running example

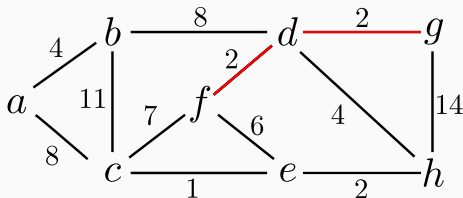
Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	0/ nil	∞/nil	∞/nil
f	∞/nil	∞/nil	7/ f	2/ f	6/ f		∞/nil	∞/nil
f, d	∞/nil	8/ d	7/ f		6/ f		2/ d	4/ d

Prim's algorithm: a running example

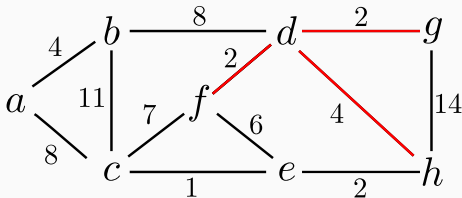
Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	0/ nil	∞/nil	∞/nil
f	∞/nil	∞/nil	7/ f	2/ f	6/ f		∞/nil	∞/nil
f, d	∞/nil	8/ d	7/ f		6/ f		2/ d	4/ d
f, d, g	∞/nil	8/ d	7/ f		6/ f			4/ d

Prim's algorithm: a running example

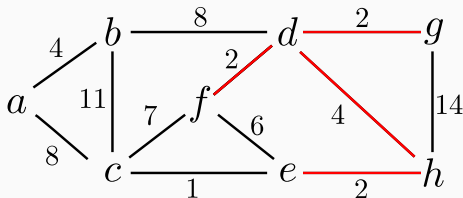
Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	0/ nil	∞/nil	∞/nil
f	∞/nil	∞/nil	7/ f	2/ f	6/ f		∞/nil	∞/nil
f, d	∞/nil	8/ d	7/ f		6/ f		2/ d	4/ d
f, d, g	∞/nil	8/ d	7/ f		6/ f			4/ d
f, d, g, h	∞/nil	8/ d	7/ f		2/ h			

Prim's algorithm: a running example

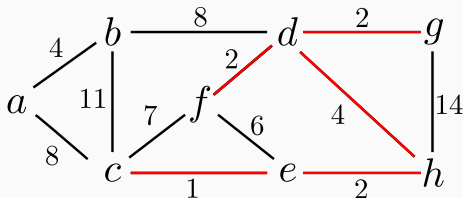
Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	0/ nil	∞/nil	∞/nil
f	∞/nil	∞/nil	7/ f	2/ f	6/ f		∞/nil	∞/nil
f, d	∞/nil	8/ d	7/ f		6/ f		2/ d	4/ d
f, d, g	∞/nil	8/ d	7/ f		6/ f			4/ d
f, d, g, h	∞/nil	8/ d	7/ f		2/ h			
f, d, g, h, e	∞/nil	8/ d	1/ e					

Prim's algorithm: a running example

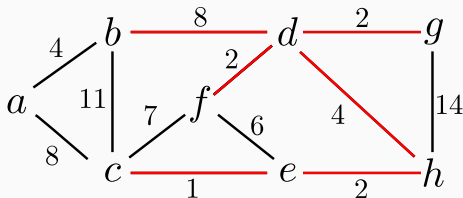
Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	0/ nil	∞/nil	∞/nil
f	∞/nil	∞/nil	7/ f	2/ f	6/ f		∞/nil	∞/nil
f, d	∞/nil	8/ d	7/ f		6/ f		2/ d	4/ d
f, d, g	∞/nil	8/ d	7/ f		6/ f			4/ d
f, d, g, h	∞/nil	8/ d	7/ f		2/ h			
f, d, g, h, e	∞/nil	8/ d	1/ e					
f, d, g, h, e, c	8/ c	8/ d						

Prim's algorithm: a running example

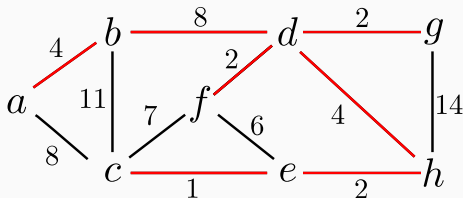
Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	0/ nil	∞/nil	∞/nil
f	∞/nil	∞/nil	7/ f	2/ f	6/ f		∞/nil	∞/nil
f, d	∞/nil	8/ d	7/ f		6/ f		2/ d	4/ d
f, d, g	∞/nil	8/ d	7/ f		6/ f			4/ d
f, d, g, h	∞/nil	8/ d	7/ f		2/ h			
f, d, g, h, e	∞/nil	8/ d	1/ e					
f, d, g, h, e, c	8/ c	8/ d						
f, d, g, h, e, c, b	4/ b							

Prim's algorithm: a running example

Starting with f



Set S	a	b	c	d	e	f	g	h
$\{\}$	∞/nil	∞/nil	∞/nil	∞/nil	∞/nil	0/ nil	∞/nil	∞/nil
f	∞/nil	∞/nil	7/ f	2/ f	6/ f		∞/nil	∞/nil
f, d	∞/nil	8/ d	7/ f		6/ f		2/ d	4/ d
f, d, g	∞/nil	8/ d	7/ f		6/ f			4/ d
f, d, g, h	∞/nil	8/ d	7/ f		2/ h			
f, d, g, h, e	∞/nil	8/ d	1/ e					
f, d, g, h, e, c	8/ c	8/ d						
f, d, g, h, e, c, b	4/ b							
f, d, g, h, e, c, b, a								

Greedy algorithms

Huffman Encoding (Textbook Section 5.2)

Huffman Encoding

An encoding scheme used in, e.g., MP3 encoding

Data: a string S of symbols over an alphabet Γ

Goal: find a binary encoding e of Γ resulting in minimum encoded length of S

Denote the encoded string by S_e

Different encodings

Consider $\Gamma = \{a, b, c\}$

Stats on S : a appears 45 times, b 16 times, and c twice

- Fixed-length encoding

$$a \rightarrow 00$$

$$e_1: b \rightarrow 01 \quad |S_{e_1}| = 45 \times 2 + 16 \times 2 + 2 \times 2 = 126$$

$$c \rightarrow 10$$

- Variable-length encoding

$$a \rightarrow 0$$

$$e_2: b \rightarrow 10 \quad |S_{e_2}| = 45 \times 1 + 16 \times 2 + 2 \times 2 = 81$$

$$c \rightarrow 11$$

$$a \rightarrow 0$$

- Be careful! $e_2: b \rightarrow 1$ Decoding will lead to ambiguity

$$c \rightarrow 01$$

Prefix-free encoding

$$a \rightarrow 0$$

Consider the bad encoding e_2 : $b \rightarrow 1$ How to decode 010110?

$$c \rightarrow 01$$

ababba?, *ccba?*, *abcba?*, or ...?

To avoid ambiguity, we need the encoding to be **prefix-free**

Definition

An encoding is **prefix-free** if no codeword is a prefix of any other codewords