

Running time of Kruskal's algorithm (I)

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

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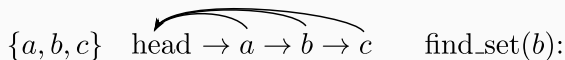
Using linked list:



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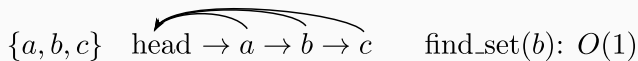
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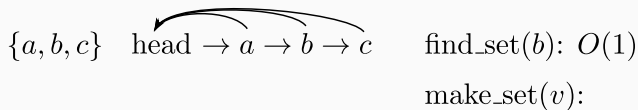
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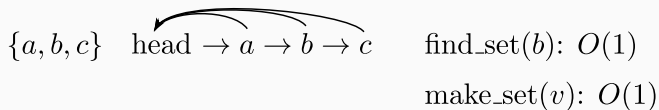
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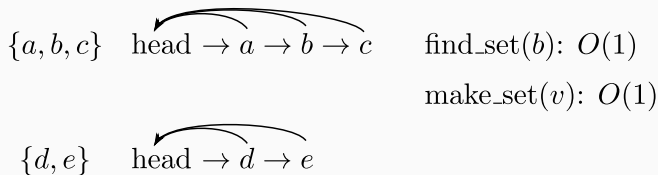
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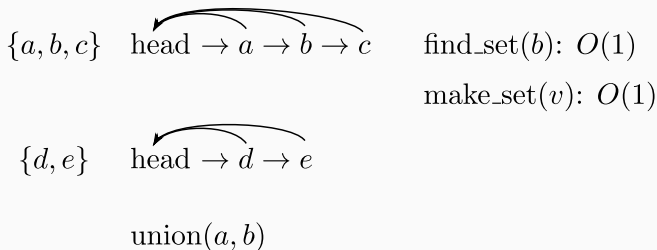
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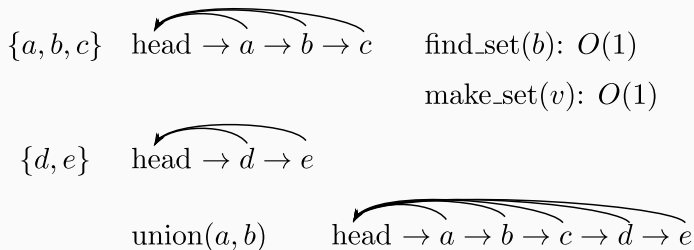
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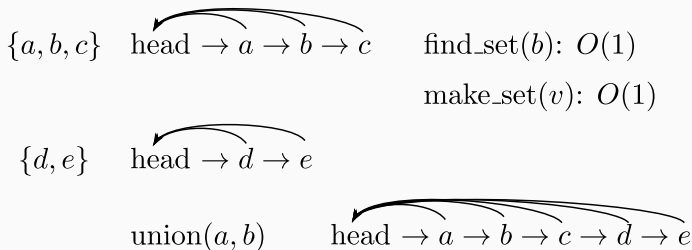
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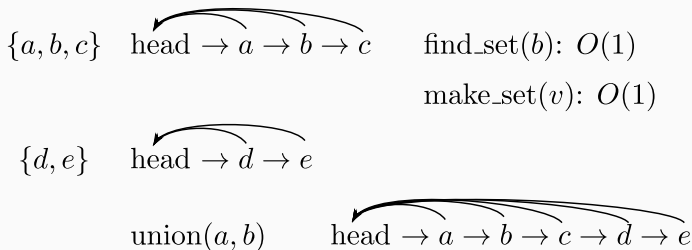


Cost of union:

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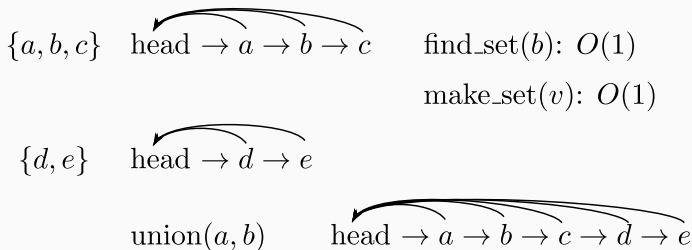


Cost of union: $O(\text{length of the shorter list})$

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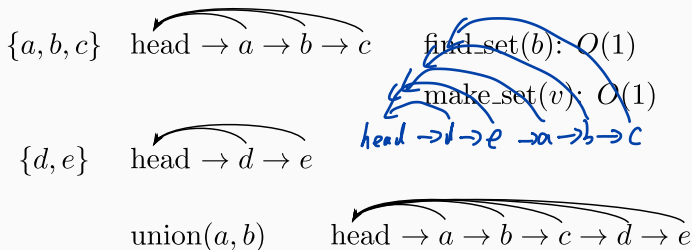
Using an array to implement it:

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

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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	$\xrightarrow{\text{union}}$	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              $A := A \cup \{(u, v)\}$ ;
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10     $\lfloor$ 
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```

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

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

$\log |V|^2 = 2 \log |V|$

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Worst-case cost for union:

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$\rightarrow O(|V|)$

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

\downarrow
 $O(|V| \log |V|)$

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Total cost of the algorithm:

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7         if find_set( $u$ )  $\neq$  find_set( $v$ ): worst case  $O(|E| \ln)$   
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Total cost of the algorithm: $O(|E| \log |V|)$

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Directed tree disjoint set:

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Directed tree disjoint set:

$$\{a\}$$

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

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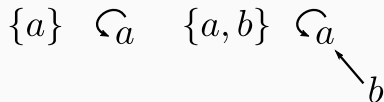
Directed tree disjoint set:

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

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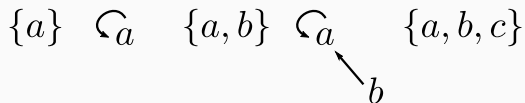
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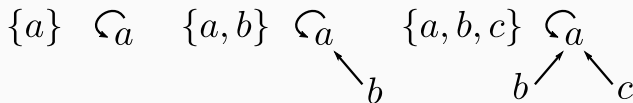
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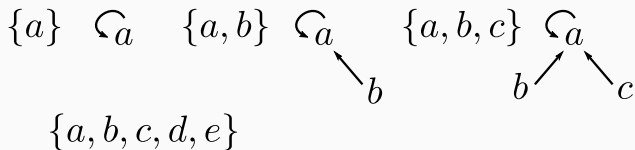
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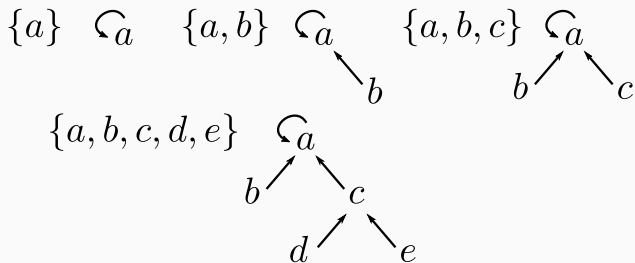
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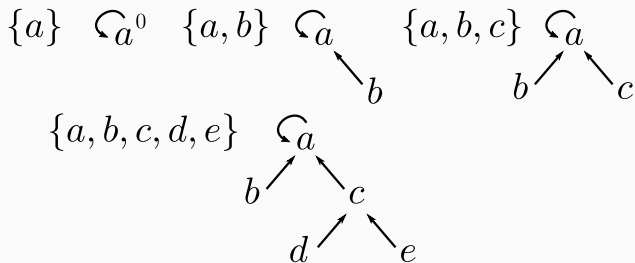
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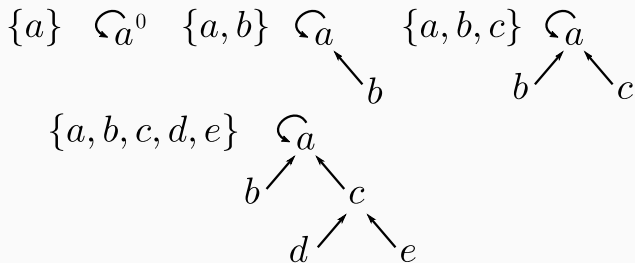
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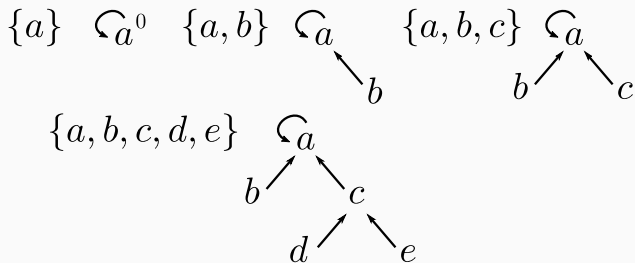
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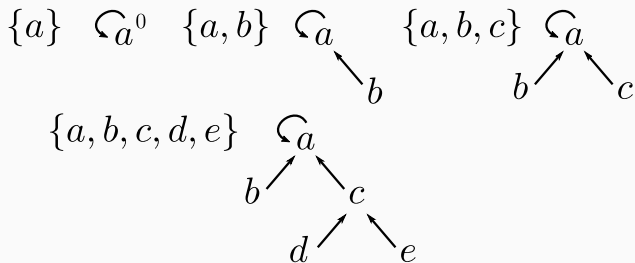
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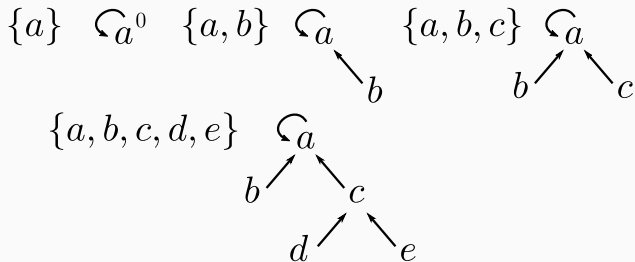
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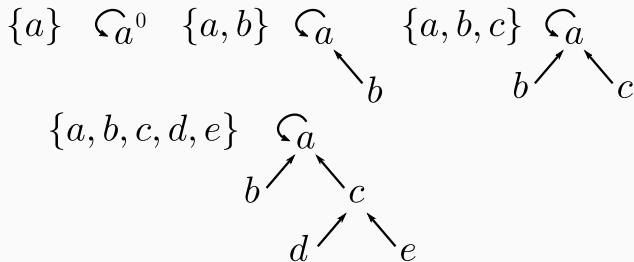
Definition

$\pi(x)$: parent of x

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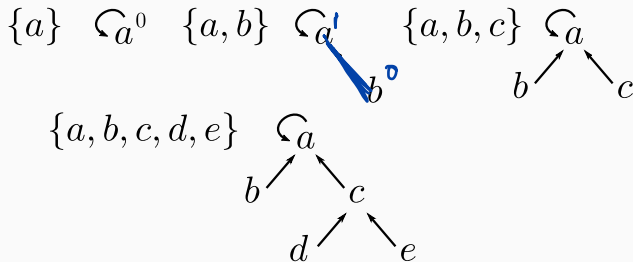
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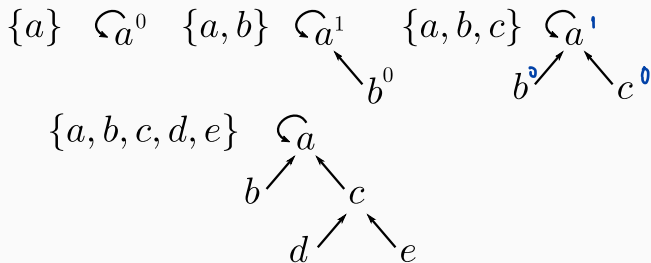
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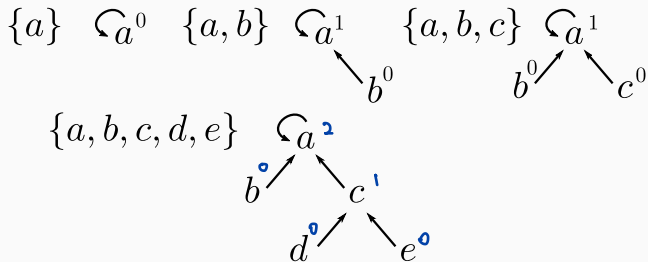
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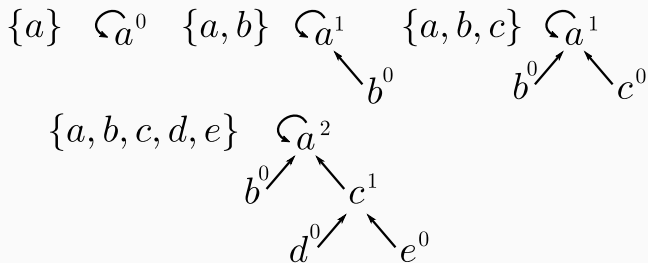
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Operations of direct tree disjoint set (I)

- `make_set(v)`

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def `make_set(v):`

$\pi(v) := v;$

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Cost: $O(1)$

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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 makefind_set( $v$ ):
```

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

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

```
    return  $v$ 
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Operations of direct tree disjoint set (I)

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```

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

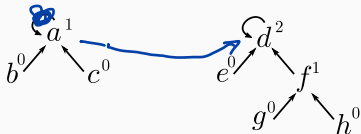
- what about union?

Operations of direct tree disjoint set (II)

- union:

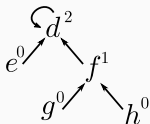
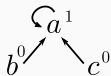
Operations of direct tree disjoint set (II)

■ union:

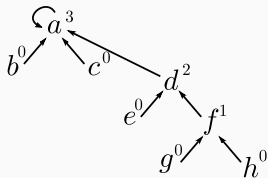


Operations of direct tree disjoint set (II)

- union:

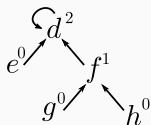
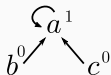


Option 1

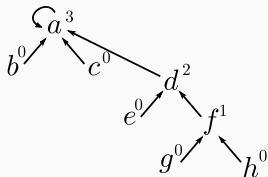


Operations of direct tree disjoint set (II)

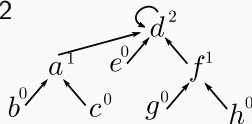
- union:



Option 1

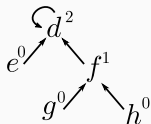
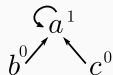


Option 2

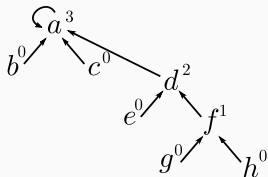


Operations of direct tree disjoint set (II)

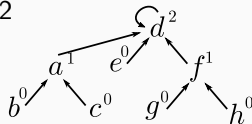
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Option 1



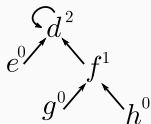
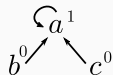
Option 2



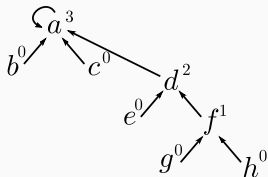
better!

Operations of direct tree disjoint set (II)

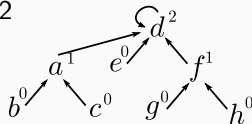
- union:



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)

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    else:  
        |
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Operations of direct tree disjoint set (II)

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$\pi(r_y) := r_x$;

else:

$\pi(r_x) := r_y$;

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

 |



Operations of direct tree disjoint set (II)

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    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)

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

if rank(r_x) > rank(r_y):

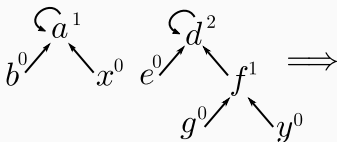
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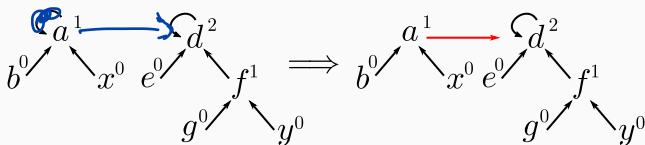
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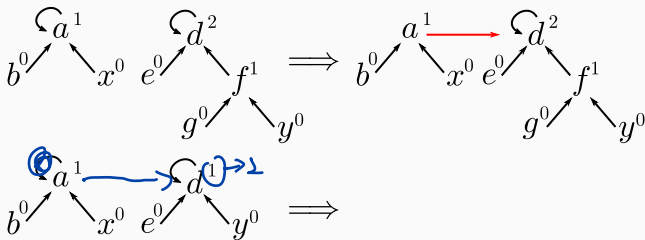
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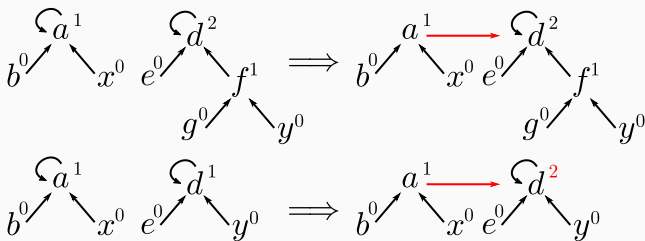
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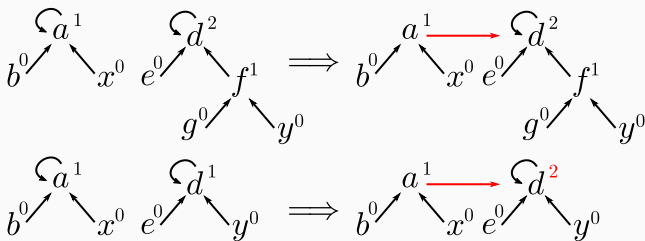
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Cost: dominated by find_set



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

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Lemma

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

Cost of find_set using directed tree disjoint set

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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

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Cost of find_set using directed tree disjoint set

Observation

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Lemma

$$\log |V| \quad \sqrt{|V|}$$

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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 $\geq 2^{k-1} + 2^{k-1} = 2^k$ □

$|V|$ nodes, max rank?

Cost of find_set using directed tree disjoint set

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

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

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- the cost of union:

Cost of find_set using directed tree disjoint set

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- 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      $\lfloor$  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       union( $u, v$ );
```

$\rightarrow \log |v|$
 $\hookrightarrow \log |v|$
 $\left. \begin{array}{l} \text{if find_set}(u) \neq \text{find_set}(v): \\ \quad A := A \cup \{(u, v)\}; \\ \quad \text{union}(u, v); \end{array} \right\} O(|E| \log |V|)$

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Lines 6-9:

Total running time of Kruskal using directed tree disjoint set

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

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

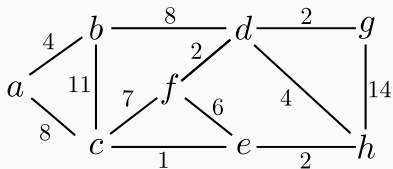
Prim's algorithm

Intuition: iteratively grows the tree

Prim's algorithm

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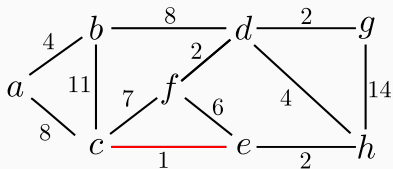
Kruskal



Prim's algorithm

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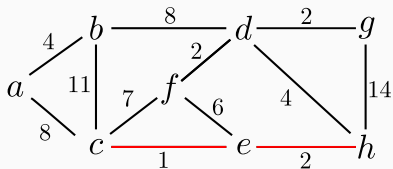
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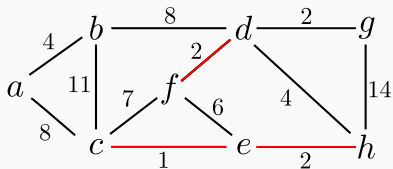
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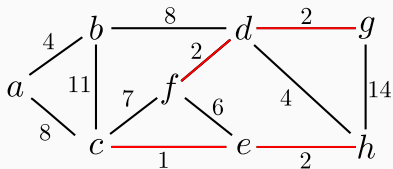
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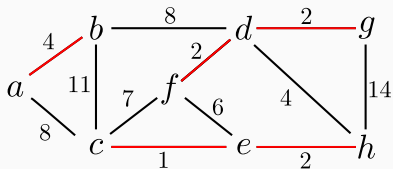
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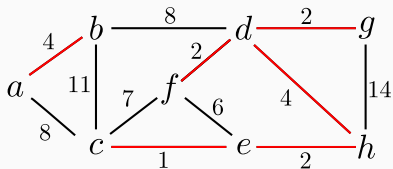
Kruskal



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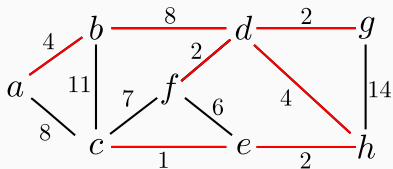
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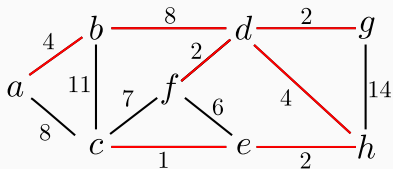
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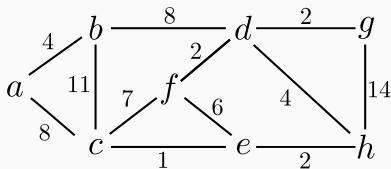
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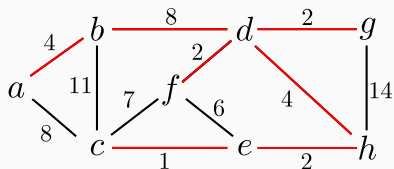
starting with c



Prim's algorithm

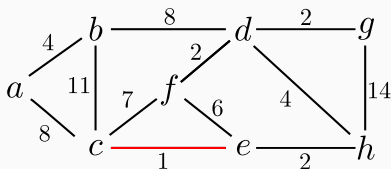
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Kruskal



Prim

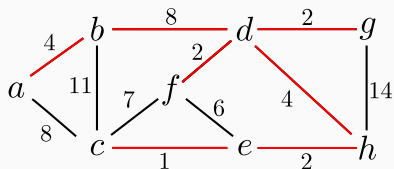
starting with c



Prim's algorithm

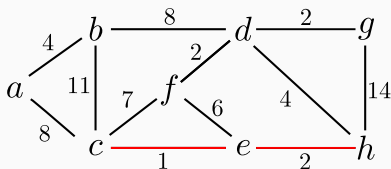
Intuition: iteratively grows the tree

Kruskal



Prim

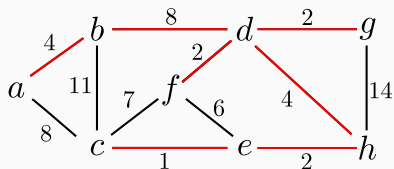
starting with c



Prim's algorithm

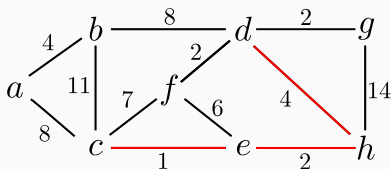
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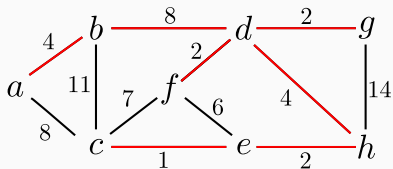
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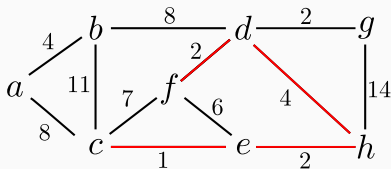
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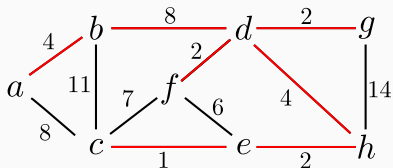
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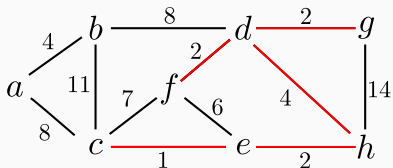
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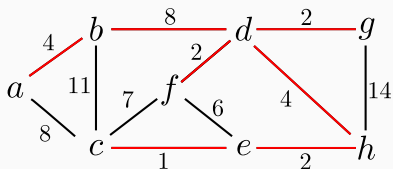
starting with c



Prim's algorithm

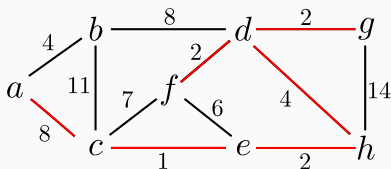
Intuition: iteratively grows the tree

Kruskal



Prim

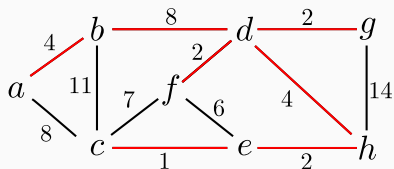
starting with c



Prim's algorithm

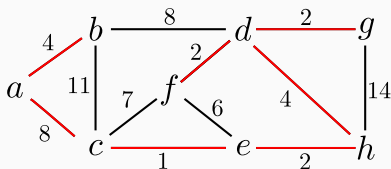
Intuition: iteratively grows the tree

Kruskal



Prim

starting with c



Prim's algorithm: pseudocode

Let S be the set included in the tree so far

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

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

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

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

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

for $v \in V$:

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

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 Pick any initial vertex u_0 ;

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

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

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while H is not empty:

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

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

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

while H is not empty:

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

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

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

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

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

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

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

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

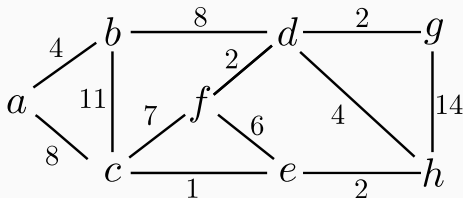
if $\text{cost}(z) > w_e$:

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

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

Prim's algorithm: a running example

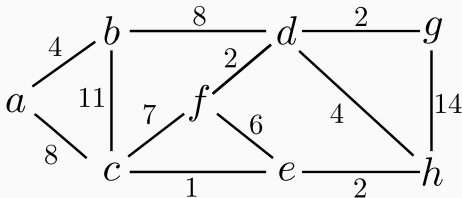
Starting with f



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

Prim's algorithm: a running example

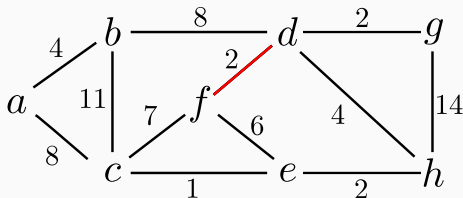
Starting with f



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

Prim's algorithm: a running example

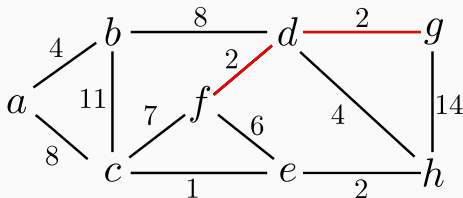
Starting with f



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

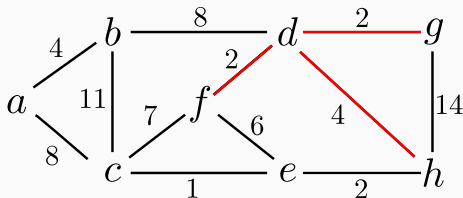
Starting with f



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

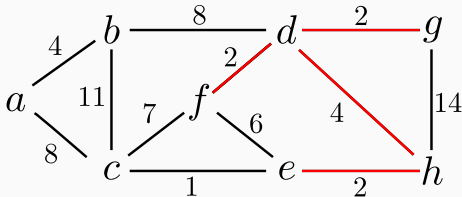
Starting with f



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

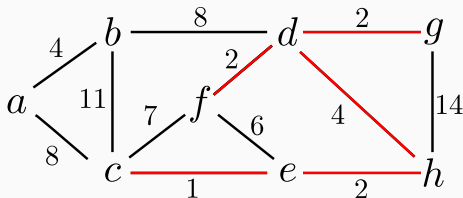
Starting with f



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

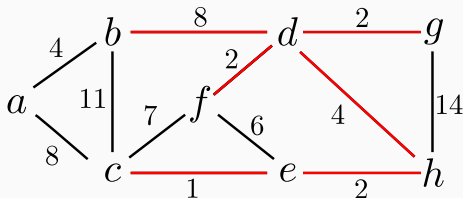
Starting with f



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

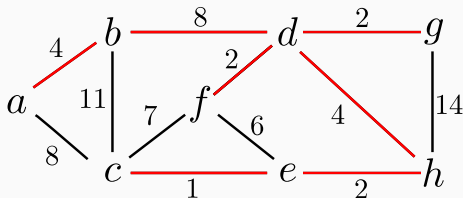
Starting with f



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)