

CMPSC 465

Data Structures and Algorithms

Spring 2022

Instructor: Chunhao Wang

Greedy algorithms

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Minimum Spanning Tree

Running time of Kruskal's algorithm (I)

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

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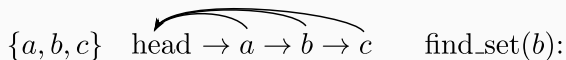
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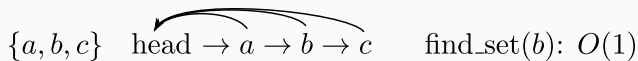
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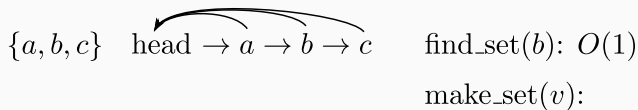
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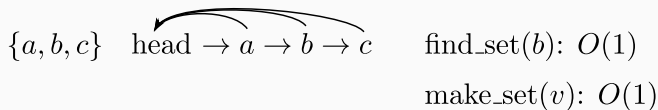
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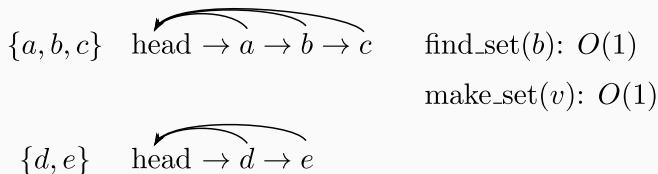
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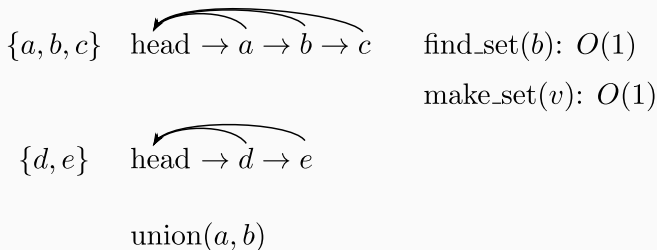
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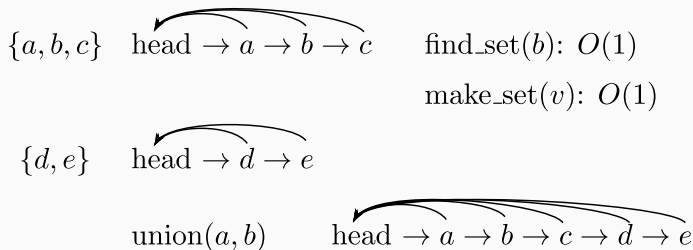
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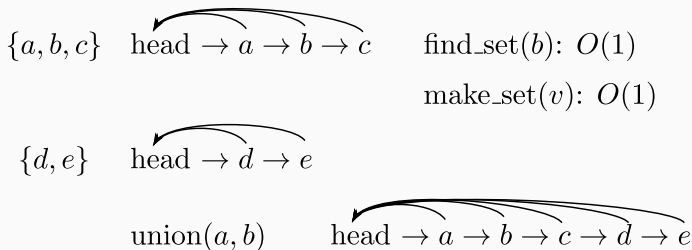
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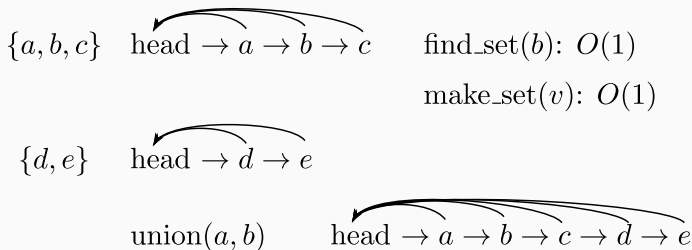


Cost of union:

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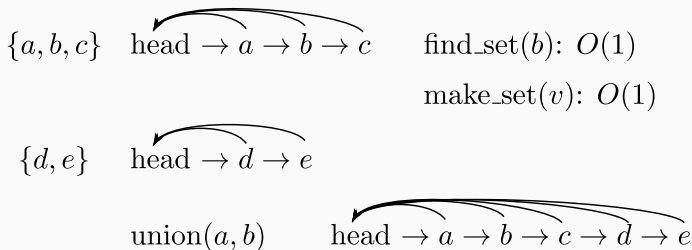


Cost of union: $O(\text{length of the shorter list})$

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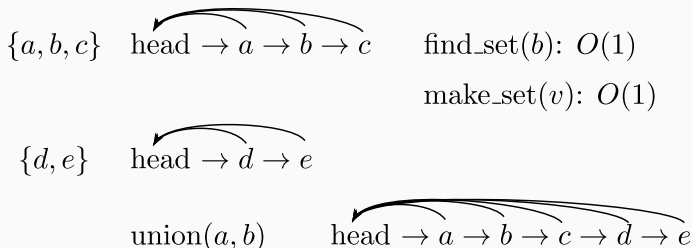
Using an array to implement it:

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

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Running time of Kruskal's algorithm (II)

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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
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8              $A := A \cup \{(u, v)\}$ ;
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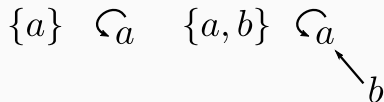
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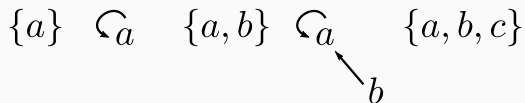
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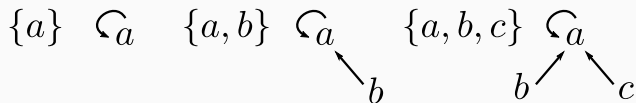
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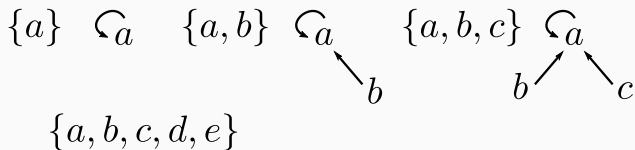
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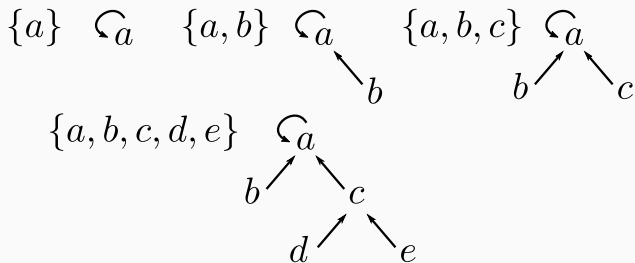
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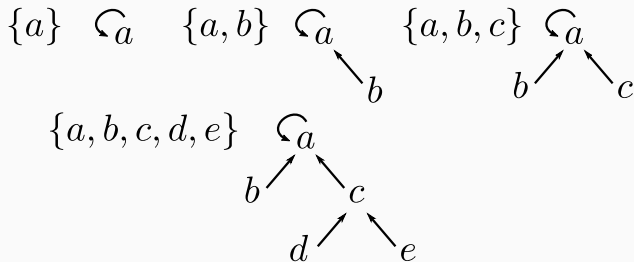
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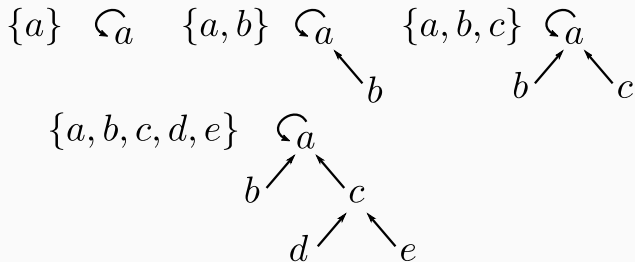
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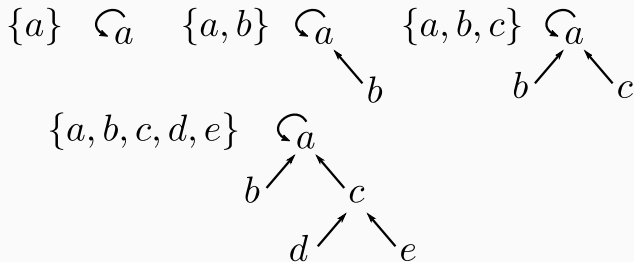
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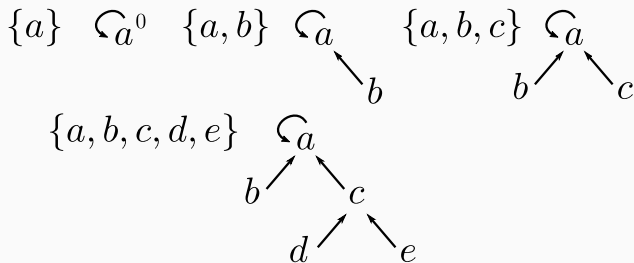
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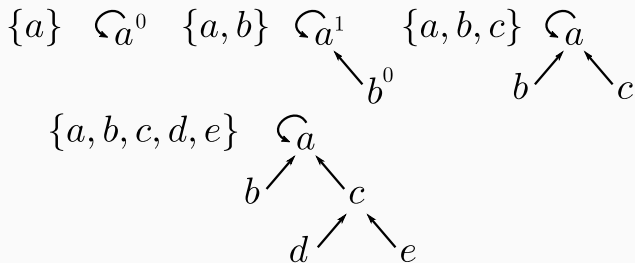
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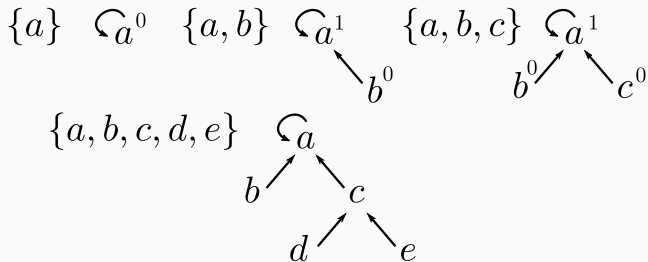
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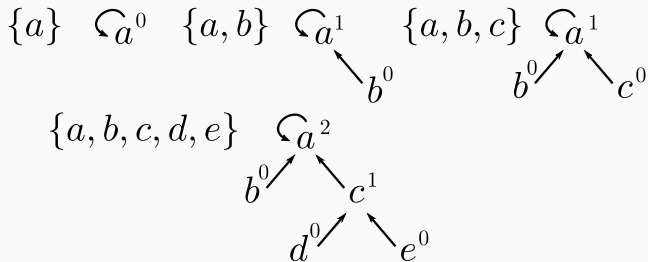
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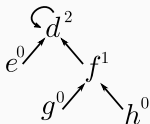
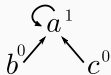
- what about union?

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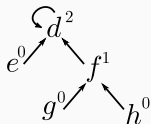
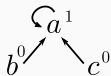
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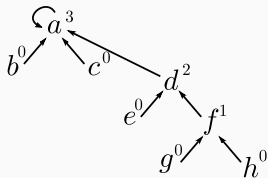


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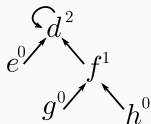
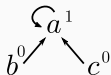


Option 1

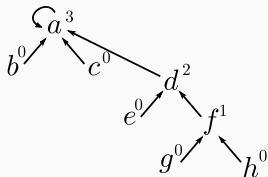


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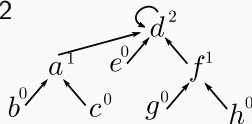
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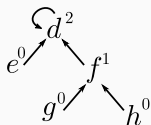
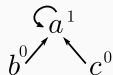


Option 2

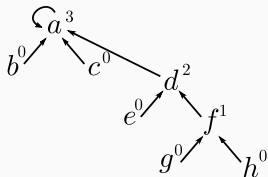


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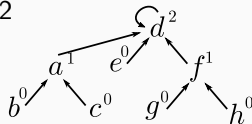
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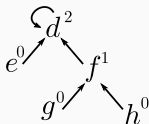
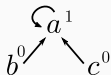
Option 2



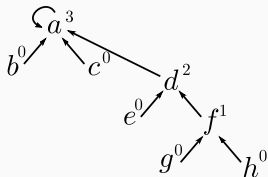
better!

Operations of direct tree disjoint set (II)

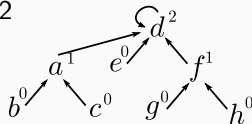
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Option 1



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):

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

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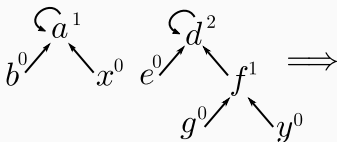
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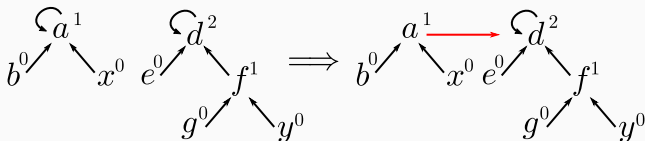
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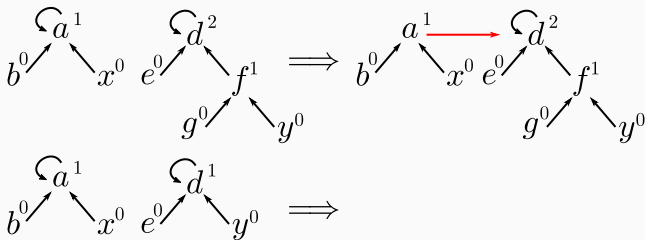
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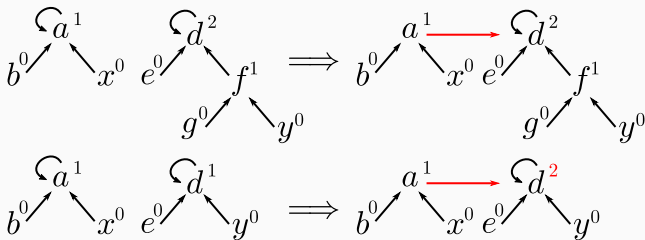
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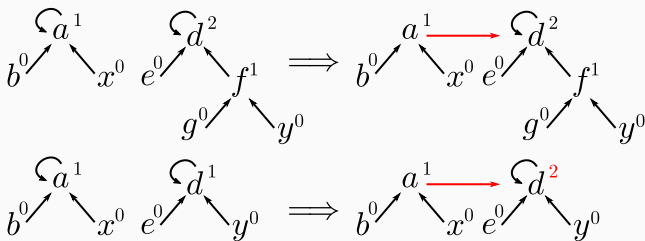
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Cost: dominated by find_set



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

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

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Total running time of Kruskal using directed tree disjoint set

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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( $v$ ):
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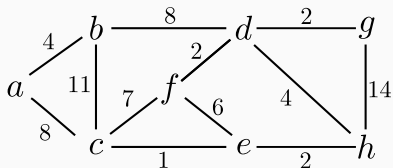
Prim's algorithm

Intuition: iteratively grows the tree

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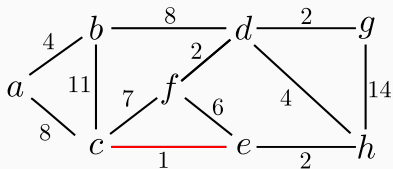
Kruskal



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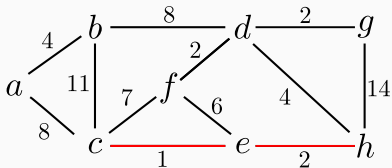
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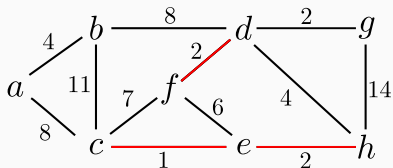
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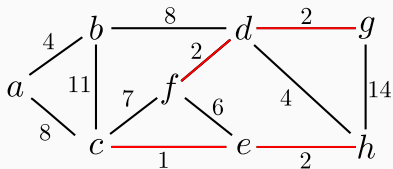
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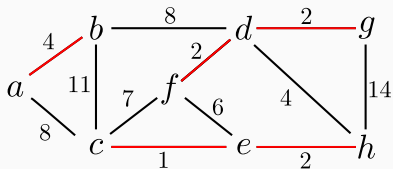
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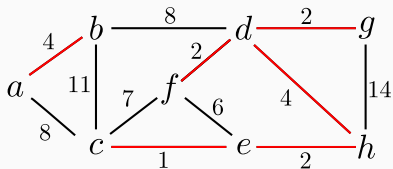
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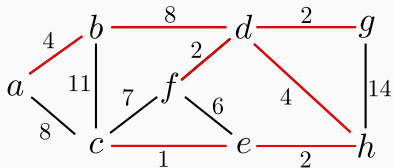
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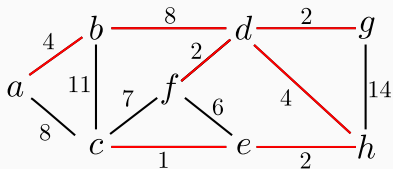
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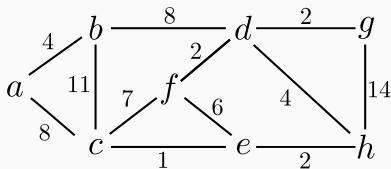
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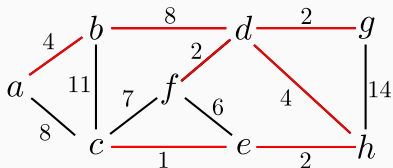
starting with c



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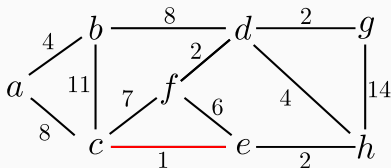
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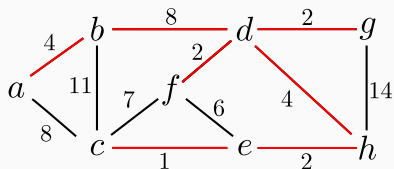
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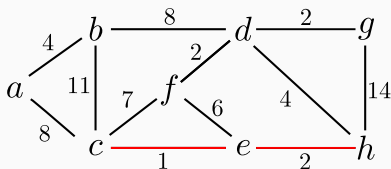
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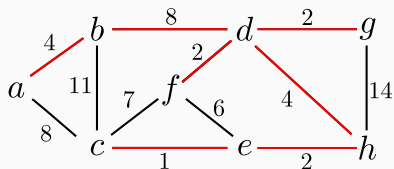
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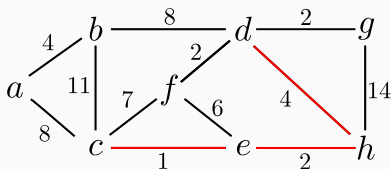
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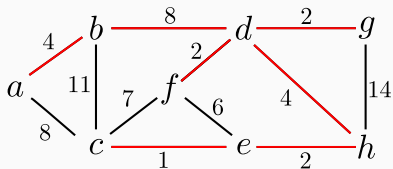
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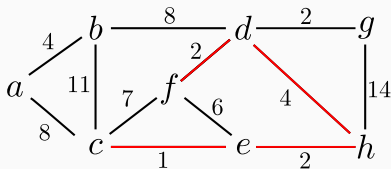
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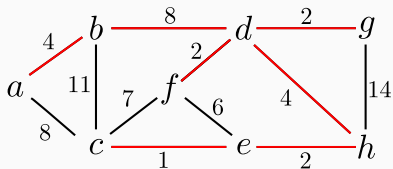
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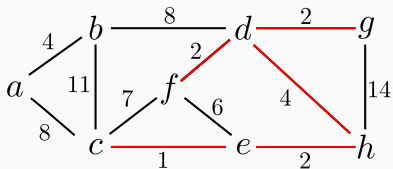
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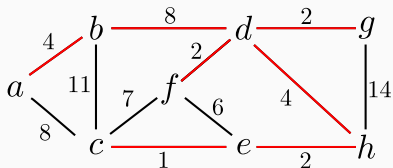
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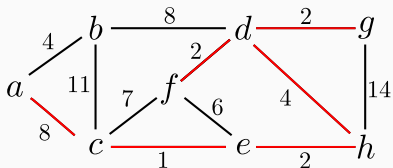
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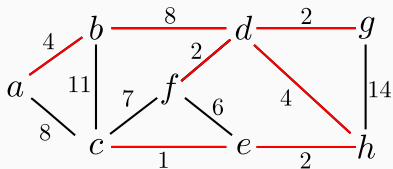
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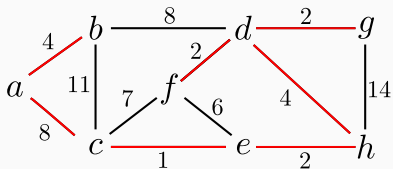
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Prim's algorithm: pseudocode

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

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for $v \in V$:

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

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

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 Pick any initial vertex u_0 ;

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

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 Pick any initial vertex u_0 ;

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

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

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while H is not empty:

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

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 Pick any initial vertex u_0 ;

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

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

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

Prim's algorithm: pseudocode

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

Prim's algorithm: pseudocode

Let S be the set included in the tree so far

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for $v \in V$:

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

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

 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

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

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

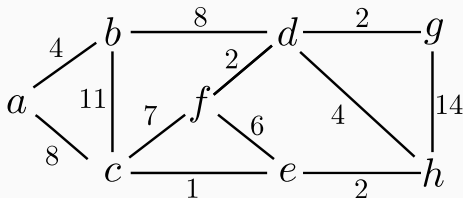
if $\text{cost}(z) > w_e$:

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

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

Prim's algorithm: a running example

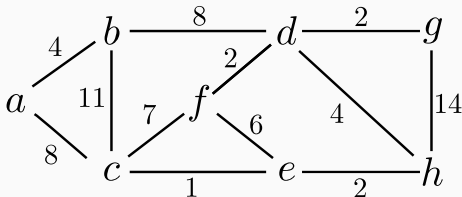
Starting with f



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

Prim's algorithm: a running example

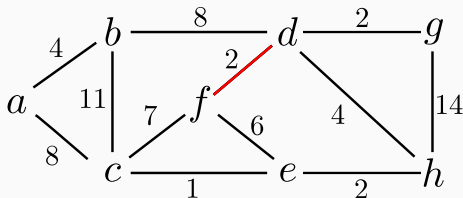
Starting with f



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

Prim's algorithm: a running example

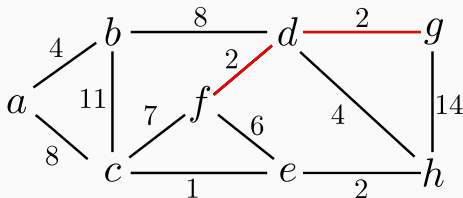
Starting with f



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

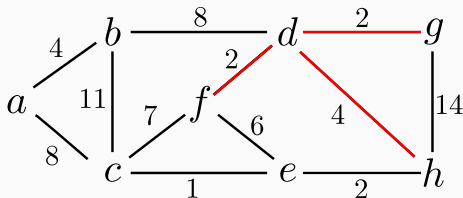
Starting with f



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

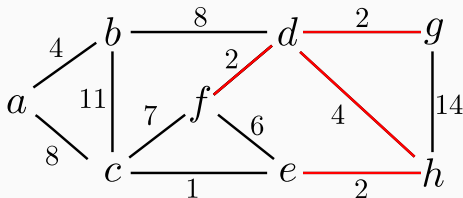
Starting with f



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

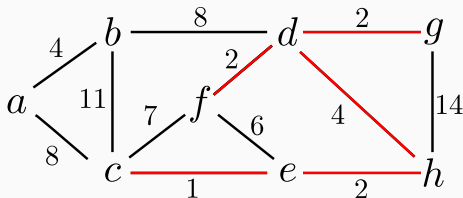
Starting with f



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

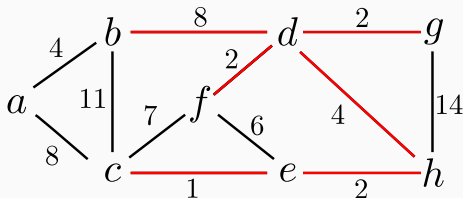
Starting with f



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

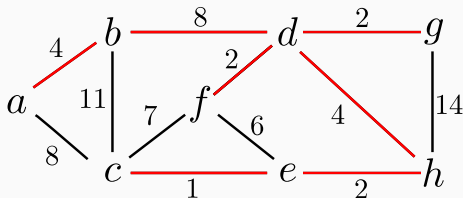
Starting with f



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								