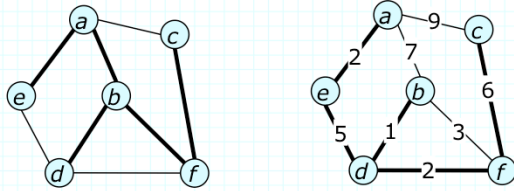


# Minimum Spanning Trees and $d$ -Heaps

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# Minimum Spanning Trees



- A *spanning tree* of an undirected graph  $G=(V,E)$  is a tree  $T=(V,E')$ , for which  $E' \subseteq E$
- In a graph in which edges have *costs* the *minimum spanning tree* problem is to find a spanning tree  $T=(V,E')$  for which  $\sum_{e \in E'} \text{cost}(e)$  is as small as possible
- Variety of direct applications; often appears as a sub-problem in other optimization problems

## The Greedy Method

- A *cut* in  $G=(V,E)$  is a division of  $V$  into two parts  $X, X'$ 
  - » an edge *crosses* cut if one endpoint is in  $X$  and other is in  $X'$
- The *greedy method* for solving the minimum spanning tree problem is a general algorithmic pattern
  - » at each step it colors an edge either *blue* (accepted) or *red* (rejected); when all edges are colored, the blue edges form a minimum spanning tree
  - » coloring rules
    - *Blue rule*. Select a cut with no blue edges, but at least one uncolored edge; select a minimum cost uncolored edges crossing the cut and color it blue
    - *Red rule*. Select a simple cycle with no red edges and at least one uncolored edge; select a maximum cost uncolored edge on the cycle and color it red

## Correctness of Greedy Method

- Greedy method maintains *color invariant*
    - » there is an MST containing all the blue edges and no red ones
  - **Theorem 6.1 (Tarjan)**. Greedy method colors all edges of a connected graph and maintains color invariant
- Proof.* Suppose invariant is true before a “blue step”
- » let  $e = \{x, y\}$  be selected edge, let  $T = (V, F)$  be an MST containing all blue edges (and no red ones) before the step
  - » if  $e \in F$ ,  $T$  contains all blue edges (and no red ones) after the step
  - » if  $e \notin F$ , there is some other edge  $e'$  on simple path from  $x$  to  $y$  in  $T$  that is also in the cut selected by the blue rule ( $e'$  is not blue)
    - $T' = (V, F \cup \{e\} - \{e'\})$  is a spanning tree
    - since  $e'$  is not blue and  $\text{cost}(e) \leq \text{cost}(e')$ ,  $T'$  is an MST and  $T'$  contains all the blue edges (and no red ones) after the step

- Suppose invariant is true before a “red step”
  - » let  $e = \{x, y\}$  be selected edge and let  $T = (V, F)$  be an MST that contains no red edges (and all blue edges) before the step
  - » if  $e \notin F$  then  $T$  contains no red edges (and all blue) after step
  - » if  $e \in F$ , then removing  $e$  from  $T$  splits  $T$  into subtrees  $T_1$  and  $T_2$ 
    - there is some edge  $e'$  that is not in  $T$ , on the cycle selected by the red rule that joins a vertex in  $T_1$  to a vertex in  $T_2$  ( $e'$  is not red)
    - $T' = (V, F \cup \{e'\} - \{e\})$  is a spanning tree.
    - Since  $cost(e) \geq cost(e')$ ,  $T'$  is an MST.  $T'$  contains no red edges (and all blue) after the step
- To see that all edges are colored, suppose that at some point  $e = \{u, v\}$  remains uncolored.
  - » if  $u$  and  $v$  are connected by a blue path then that path plus  $e$  forms a cycle that the red rule can be applied to
  - » if  $u$  and  $v$  are not connected by a blue path, then there is a cut crossed by  $e$  that the blue rule can be applied to ■

# Prim's Algorithm

- Build single blue tree, from an arbitrary starting vertex by repeating following step  $n-1$  times
  - » select a minimum cost edge incident to the blue tree containing the starting vertex and color it blue
- The algorithm can also be expressed as follows.

```

procedure minspantree(graph  $G$ , set  $Tedges$ );
  vertex  $w, u, v$ ; set  $S$ ,  $Tvertices$ ;
   $Tvertices := \{1\}$ ;  $Tedges := \{\}$ ;  $S := neighbors(1)$ ;
  do  $S \neq \{\} \Rightarrow$ 
    select min cost edge  $\{w, u\}$  from  $w \in Tvertices$  to  $u \in S$ 
     $Tvertices := Tvertices \cup \{u\}$ ;  $Tedges := Tedges \cup \{w, u\}$ ;
     $S := S - \{u\}$ ;
    for  $\{u, v\} \in edges(u) \Rightarrow$  if  $v \notin Tvertices \Rightarrow S := S \cup \{v\}$  fi rof;
  od; // edges not added to tree_edges are implicitly colored red
end;

```

need fast  
method to select  
min cost edge

# The Heap Data Structure

- A *heap* is a data structure consisting of a collection of items, each having a key; the basic operations are:
  - »  $insert(i, k, h)$  – add item  $i$  to heap  $h$  using  $k$  as the key value
  - »  $deletemin(h)$  – delete and return a minimum key item in  $h$
  - »  $changekey(i, k, h)$  – change the key of item  $i$  in heap  $h$  to  $k$
  - »  $key(i, h)$  – return the key value for item  $i$
- The  $d$ -heap is one implementation of the heap data structure that has an integer parameter  $d$ 
  - » running time of  $O(\log_d n)$  for  $insert$  and for  $changekey$  operations that decrease the key value
  - » running time of  $O(d \log_d n)$  for  $deletemin$  and for  $changekey$  operations that increase the key value
  - » can choose value of  $d$  to optimize algorithm performance

# Prim's Algorithm Using a Heap

```

procedure minspanntree(graph G, set Tedges);
  vertex u,v; set tree_vertices;
  heap S; mapping cheap: vertex → edge;
  Tvertices := {1}; tree_edges := {};
  for {1,v} ∈ edges(1) ⇒ insert(v,cost(1,v),S); cheap(v) := {1,v} rof;
  do S ≠ {} ⇒
    u := deletemin(S);
    Tvertices := Tvertices ∪ {u}; Tedges := Tedges ∪ {cheap(u)};
    for {u,v} ∈ edges(u) ⇒
      if v ∈ S and cost(u,v) < key(v) ⇒
        changekey(v,cost(u,v),S); cheap(v) := {u,v};
      | v ∉ S and v ∉ Tvertices ⇒
        insert(v,cost(u,v),S); cheap(v) := {u,v}
      fi;
    rof;
  od;
end;

```

Note that heap stores vertices, not edges

every edge examined twice

every *changkey* reduces key value

each vertex inserted once



## Analysis of Prim's Algorithm

- Assume that  $T_{vertices}$  is implemented as a bit vector and  $T_{edges}$  as a list
- Non-heap operations within main **do**-loop but outside **for**-loop use constant time per iteration
- The **do**-loop is executed exactly  $n$  times
- The **for**-loop is executed  $2m$  times
- Heap operation counts
  - » at most  $n$  *deletemins*,  $n$  *inserts*,  $m$  *changekeys*
  - » *changekey* operations all decrease the key value
- Choosing  $d = \lfloor 2 + m/n \rfloor$  gives  $O\left(m \frac{\log n}{\log(2 + m/n)}\right)$

## C++ Version

```
// Find min spanning tree of graf and return it in mst
void prim(Wgraph& graf, Wgraph& mst) {
    vertex u,v; edge e;
    edge *cheap = new edge[graf.n()+1];
    Dheap nodeHeap(graf.n(),2+graf.m()/graf.n());
    for (e = graf.firstAt(1); e != 0; e = graf.nextAt(1,e)) {
        u = graf.mate(1,e); nodeHeap.insert(u,graf.weight(e));
        cheap[u] = e;
    }
    while (!nodeHeap.empty()) {
        u = nodeHeap.deletemin();
        e = mst.join(graf.left(cheap[u]),graf.right(cheap[u]));
        mst.setWeight(e,graf.weight(cheap[u]));
        for (e = graf.firstAt(u); e != 0; e = graf.nextAt(u,e)) {
            v = graf.mate(u,e);
            if (nodeHeap.member(v) && graf.weight(e) < nodeHeap.key(v)) {
                nodeHeap.changekey(v, graf.weight(e)); cheap[v] = e;
            } else if (!nodeHeap.member(v) && mst.firstAt(v) == 0) {
                nodeHeap.insert(v, graf.w(e)); cheap[v] = e;
            } } }
    delete [] cheap;
}
```

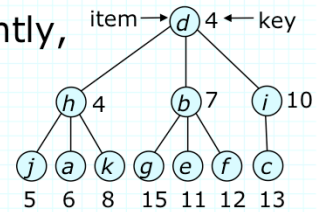
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## $d$ -Heaps

- Heaps can be implemented efficiently, using *heap-ordered tree*

- » each tree node contains one *item* with a real-valued *key*

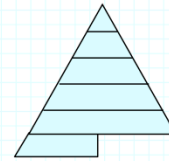
- » key of each node  $\geq$  key of its parent



- A  $d$ -heap is *heap-shaped*, heap-ordered  $d$ -ary tree

- » let  $T$  be an infinite  $d$ -ary tree, with vertices numbered in breadth-first order

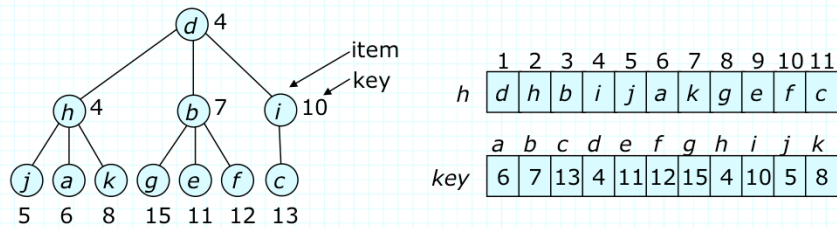
- » a subtree of  $T$  is *heap-shaped* if its vertices have consecutive numbers  $1, 2, \dots, n$



- The depth of a  $d$ -heap with  $n$  vertices is  $\leq \lceil \log_d n \rceil$

## Implementing $d$ -Heaps as Arrays

- $D$ -heap can be stored in an array in breadth-first order
  - » allows indices for parents and children to be calculated directly, eliminating the need for pointers



- If  $i$  is index of item  $x$ , then  $\lceil (i-1)/d \rceil$  is index of  $p(x)$ ; indices of children of  $x$  are in range  $[d(i-1)+2 .. di+1]$
- When key of item is decreased, restore heap-order, by repeatedly swapping the item with its parent
  - » similarly, for increasing an item's key

## $d$ -Heap Operations

```
item function findmin(heap h);  
  return if  $h = \{\}$   $\Rightarrow$  null |  $h \neq \{\}$   $\Rightarrow$   $h(1)$  fi;  
end;
```

insert  $i$  at  
position  $x$  or  
above

```
procedure siftup(item  $i$ , integer  $x$ , modifies heap  $h$ );  
  integer  $p$ ;  
   $p := \lceil (x-1)/d \rceil$ ;  
  do  $p \neq 0$  and  $key(h(p)) > key(i) \Rightarrow$   
     $h(x) := h(p)$ ;  $x := p$ ;  $p := \lceil (p-1)/d \rceil$  ;  
  od;  
   $h(x) := i$ ;  
end;
```

at most one  
iteration per  
level in heap

```
procedure insert(item  $i$ , modifies heap  $h$ );  
  siftup( $i, |h| + 1, h$ );  
end;
```

```
procedure siftdown(item  $i$ , integer  $x$ , modifies heap  $h$ );
```

```
integer  $c$ ;
```

```
 $c :=$  minchild( $x, h$ );
```

```
do  $c \neq 0$  and  $key(h(c)) < key(i) \Rightarrow$ 
```

```
     $h(x) := h(c)$ ;  $x := c$ ;  $c :=$  minchild( $x, h$ );
```

```
od;
```

```
 $h(x) := i$ ;
```

```
end;
```

insert  $i$  at position  
 $x$  or below

at most one iteration  
per level in heap

```
integer function minchild(integer  $x$ , heap  $h$ );
```

```
integer  $i$ ,  $minc$ ;
```

```
 $minc := d(x-1) + 2$ ;
```

```
if  $minc > |h| \Rightarrow$  return 0 fi;
```

```
 $i := minc + 1$ ;
```

```
do  $i \leq \min\{|h|, dx+1\} \Rightarrow$ 
```

```
    if  $key(h(i)) < key(h(minc)) \Rightarrow minc := i$  fi;
```

```
     $i := i + 1$ ;
```

```
od;
```

```
return  $minc$ ;
```

```
end;
```

at most  $d$   
iterations

```
procedure delete(item  $i$ , modifies heap  $h$ );  
  item  $j$ ;  $j := h(|h|)$ ;  $h(|h|) := \text{null}$ ;  
  if  $i \neq j$  and  $\text{key}(j) \leq \text{key}(i) \Rightarrow \text{siftup}(j, h^{-1}(i), h)$ ;  
    |  $i \neq j$  and  $\text{key}(j) > \text{key}(i) \Rightarrow \text{siftdown}(j, h^{-1}(i), h)$ ;  
  fi;  
end;
```

$h^{-1}$  implemented  
using auxiliary  
"position-of array"

```
item function deletemin(modifies heap  $h$ );  
  item  $i$ ;  
  if  $h = \{\}$   $\Rightarrow$  return null; fi;  
   $i := h(1)$ ; delete( $h(1)$ ,  $h$ );  
  return  $i$ ;  
end;
```

```
procedure changekey(item  $i$ , keytype  $k$ , modified heap  $h$ );  
  item  $ki$ ;  $ki := \text{key}(i)$ ;  $\text{key}(i) := k$ ;  
  if  $k < ki \Rightarrow \text{siftup}(i, h^{-1}(i), h)$ ;  
    |  $k > ki \Rightarrow \text{siftdown}(i, h^{-1}(i), h)$ ;  
  fi;  
end;
```

## Analysis of $d$ -Heap Operations

```

heap function makeheap(set of item s);
  integer j; heap h;
  h := {};
  for i ∈ s ⇒ j := |h|+1; h(j) = i; rof;
  j = [(|h|-1)/d];
  do j > 0 ⇒ siftdown(h(j),j,h); j = j-1; od;
  return h;
end;

```

- Each execution of *siftup* (and hence *insert*) takes  $O(\log_d n)$  time, while each execution of *siftdown* takes  $O(d \log_d n)$  time
- Time for *changekey* depends on whether keys increase or decrease
  - » if keys always decrease, can make *changekey* faster using a large  $d$
- The running time for *makeheap* is  $O(f)$  where

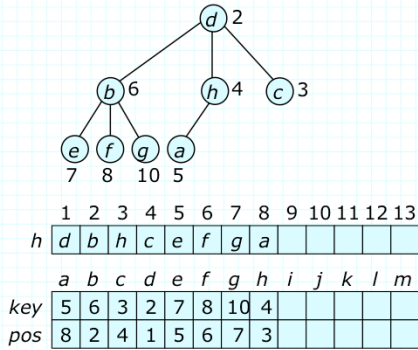
$$f(n) = \frac{n}{d}d + \frac{n}{d^2}2d + \frac{n}{d^3}3d + \dots$$

which is  $O(n)$

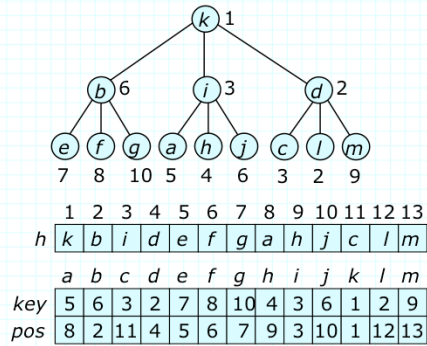


# Exercises

1. The figure below shows a 3-heap.



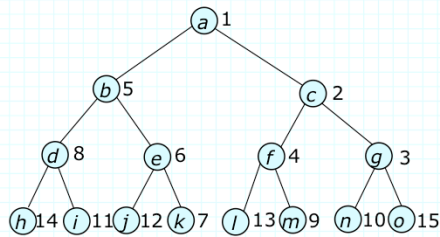
Show the heap contents after inserting new items *i*, *j*, *k*, *l* and *m* with keys 3, 6, 1, 2, 9. Show the heap state both in picture form and in array form, as above.



Delete items *d* and *k* and show the array contents after these operations.

2. Construct an example of a 2-heap on 15 items for which a *deletemin* operation requires the largest amount of time possible. Construct an example of a 2-heap on 15 items for which a sequence of 15 *deletemin* operations requires the maximum amount of time possible.

The 2-heap shown below satisfies both parts of the question.



3. The correctness of any data structure operation depends on its maintaining certain essential *invariants* of the data structure. The data portion of the class declaration for the C++ implementation of the *d*-heap data structure is shown below. What invariants must be maintained by programs that operate on it? List as many as you can think of.

```

class dheap {
int N; // max # of items in heap
int n; // # of items in heap
int d; // base of heap
item *h; // {h[i]} is set of items
int *pos; // position of item
keytyp *kvec; //key of item i
...};

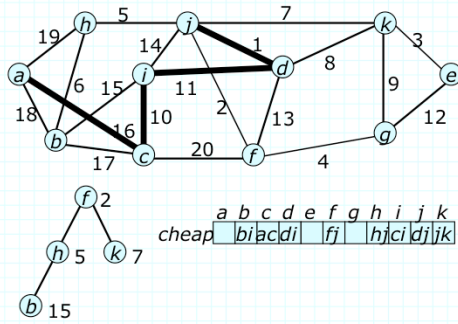
```

```

d > 1
0 ≤ n ≤ N
for 1 ≤ i ≤ n, 1 ≤ h[i] ≤ n
for 1 ≤ i ≤ n, 1 ≤ pos[i] ≤ n
for 2 ≤ i ≤ n, kvec[h[⌊i/d⌋]] = kvec[h[i]]
for 1 ≤ i ≤ n, pos[h[i]] = i
for 1 ≤ i < j ≤ n, h[i] ≠ h[j], pos[i] ≠ pos[j]

```

4. The figure below shows an intermediate state in the execution of Prim's algorithm. The heap and the *cheap* mapping are shown below ( $d=2$ ).



Show the state after four more steps have been performed.

