

Data Structures - Week 3

Tuesday, March 28, 2023

10:23 AM

Priority queues: Introduction

Queue

A queue is an abstract data type supporting the following main operations:

- PushBack(e) adds an element to the back of the queue
- PopFront() extracts an element from the front of the queue

Priority Queue

A priority queue is a generalization of a queue where each element is assigned a priority and elements come out in order by priority.

Typical use cases

Scheduling jobs:

- Want to process jobs one by one in order of decreasing priority. While the current job is processed, new job may arrive.
- To add a job to the set of scheduled jobs, call Insert(job)
- To process a job with the highest priority, get it by calling ExtractMax()

Priority Queue (formally)

Priority queue is an abstract data type supporting the following main operations:

- Insert(p) adds a new element with priority p
- ExtractMax() extracts an element with maximum priority

Additional Operations

- Remove(it) removes an element pointed by iterator it
- GetMax() returns an element with maximum priority (without changing the set of elements)
- ChangePriority(it, p) changes the priority of an element pointed by it to p

Algorithms that use priority queues

- Dijkstra's algorithm: finding a shortest path in a graph
- Prim's algorithm: constructing an optimum prefix-free encoding of a string
- Heap sort: sorting a given sequence

Priority queues: Binary Heaps

Binary max-heap:

- A binary tree (each node has zero, one, or two children) where the value of each node is at least the values of its children
- For each edge of the tree, the value of the parent is at least the value of the child

GetMax

- Return the root value
- Running time $O(1)$

Insert

- Attach a new node to any leaf
- This may violate the heap property
- To fix this we let the new node sift up

SiftUp

- For this, we swap the problematic node with its parent until the property is satisfied
- Invariant: heap property is violated on at most one edge
- The edge gets closer to the root while sifting up
- Running time $O(\text{tree height})$

ExtractMax

- Replace the root with any leaf
- Again this may violate the heap property
- To fix it, we let the problematic node sift down

SITDOWN

- For this we swap the problematic node with larger child until the heap property is satisfied
- We swap with the larger child which automatically fixes one of the two bad edges
- Running time: $O(\text{tree height})$

ChangePriority

- Change the priority and let the changed element sift up or down depending on whether its priority decreased or increased
- Running time: $O(\text{tree height})$

Remove

- Change the priority of the element to infinite, let it sift up and then extract the maximum
- Now call `ExtractMax()`
- Running time: $O(\text{tree height})$

How to keep a tree shallow?

- A binary tree is complete if all its levels are filled except possibly the last one which is filled from left to right

First Advantage: Low Height

- A complete binary tree with n nodes has a height at most $O(\log n)$
- What do we pay for these advantages?
- We need to keep the tree complete
- Which binary heap operations modify the shape of the tree?
- Only Insert and ExtractMax (remove changes the shape by calling ExtractMax)

Keeping the tree complete

- To insert an element, insert it as a leaf in the leftmost vacant position in the last level and let it sift up
- To extract the maximum value replace the root by the last leaf and let it sift down

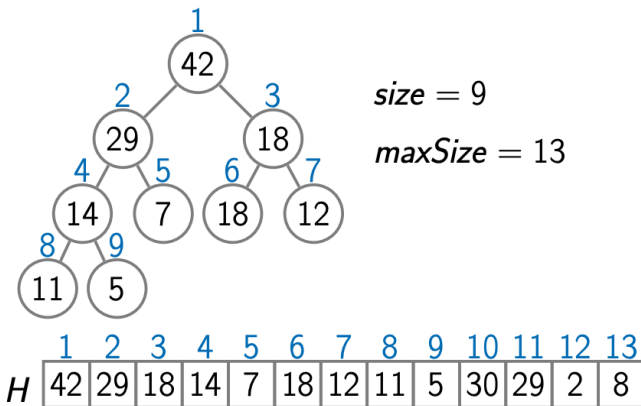
General Setting

- $maxSize$ is the maximum number of elements in the heap
- $size$ is the size of the heap
- $H[1...maxSize]$ is an array of length $maxSize$ where the heap occupies the first $size$ elements

SiftDown(i)

```
 $maxIndex \leftarrow i$   
 $\ell \leftarrow \text{LeftChild}(i)$   
if  $\ell \leq size$  and  $H[\ell] > H[maxIndex]$ :  
     $maxIndex \leftarrow \ell$   
 $r \leftarrow \text{RightChild}(i)$   
if  $r \leq size$  and  $H[r] > H[maxIndex]$ :  
     $maxIndex \leftarrow r$   
if  $i \neq maxIndex$ :  
    swap  $H[i]$  and  $H[maxIndex]$   
    SiftDown( $maxIndex$ )
```

Example



ExtractMax()

```
 $result \leftarrow H[1]$   
 $H[1] \leftarrow H[size]$   
 $size \leftarrow size - 1$   
SiftDown(1)  
return  $result$ 
```

Insert(p)

```
if  $size = maxSize$ :  
    return ERROR  
 $size \leftarrow size + 1$   
 $H[size] \leftarrow p$   
SiftUp( $size$ )
```

Remove(i)

```
 $H[i] \leftarrow \infty$   
SiftUp( $i$ )  
ExtractMax()
```

SiftUp(i)

```
while  $i > 1$  and  $H[\text{Parent}(i)] < H[i]$ :  
    swap  $H[\text{Parent}(i)]$  and  $H[i]$   
     $i \leftarrow \text{Parent}(i)$ 
```

ChangePriority(i, p)

```
 $oldp \leftarrow H[i]$   
 $H[i] \leftarrow p$   
if  $p > oldp$ :  
    SiftUp( $i$ )  
else:  
    SiftDown( $i$ )
```

Disjoint Sets: Naïve Implementations

A disjoint-set data structure supports the following operations:

- $\text{MakeSet}(x)$ creates a singleton set $\{x\}$
- $\text{Find}(x)$ returns ID of the set containing x :
 - If x and y lie in the same set the $\text{Find}(x) = \text{Find}(y)$
 - Otherwise, $\text{Find}(x) \neq \text{Find}(y)$
- $\text{Union}(x, y)$ merge two sets containing x and y

Preprocess(*maze*)

```
for each cell  $c$  in maze:  
    MakeSet( $c$ )  
for each cell  $c$  in maze:  
    for each neighbor  $n$  of  $c$ :  
        Union( $c, n$ )
```

IsReachable(A, B)

```
return Find( $A$ ) = Find( $B$ )
```

Naïve Implementations

For simplicity, we assume that our n objects are just integers $1, 2, \dots, n$

Using the smallest element as ID

- Use the smallest element of the set as its ID
- Use array `smallest[1...n]`: `smallest[i]` stores the smallest element in the set i belongs to

belongs to

MakeSet(*i*)

`smallest[i] ← i`

Find(*i*)

`return smallest[i]`

Running time: $O(1)$

Union(*i, j*)

`i_id ← Find(i)`

`j_id ← Find(j)`

`if i_id = j_id:`

`return`

`m ← min(i_id, j_id)`

`for k from 1 to n:`

`if smallest[k] in {i_id, j_id}:
 smallest[k] ← m`

Running time: $O(n)$

- Current bottleneck: Union
- What basic data structure allows for efficient merging?
- Linked list!
- Idea: represent a set as a linked list, use the list tail as ID of the set
- Pros:
 - Running time of Union is $O(1)$
 - Well defined ID
- Cons:
 - Running time of Find is $O(n)$ as we need to traverse the list to find its tail
 - Union(*x, y*) works in time $O(1)$ only if we can get the tail of the list of *x* and the head of the list of *y* in constant time!

Disjoint Sets: efficient implementations

- Represent each set as a rooted tree
- ID of a set is the root of the tree
- Use array parent[1...*n*]: parent[*i*] is the parent of *i*, or *i* if it is the root

MakeSet(*i*)

MakeSet(*i*)

$\text{parent}[i] \leftarrow i$

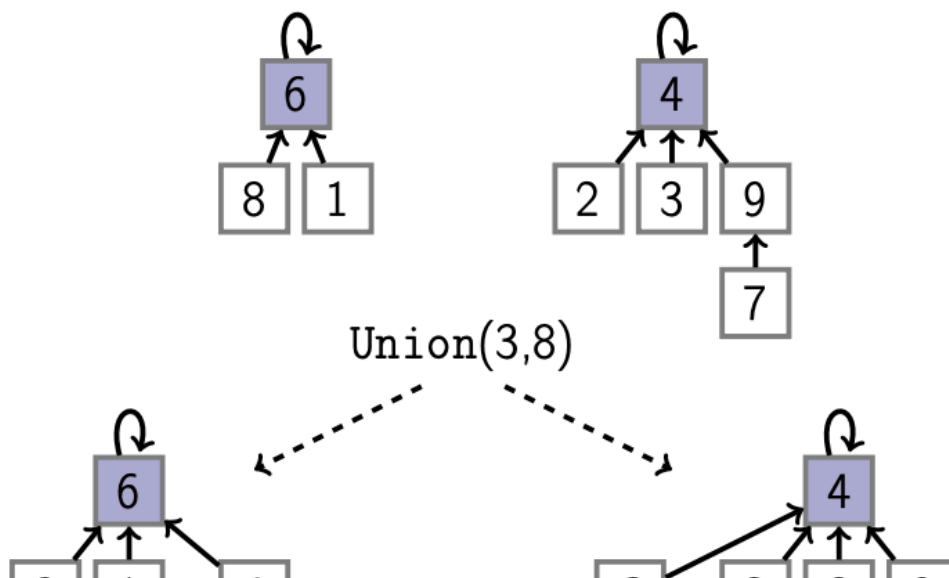
Running time: $O(1)$

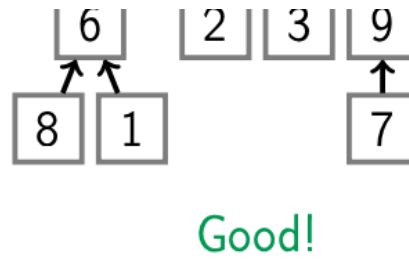
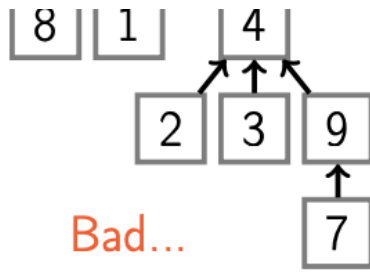
Find(*i*)

```
while  $i \neq \text{parent}[i]$ :  
     $i \leftarrow \text{parent}[i]$   
return  $i$ 
```

Running time: $O(\text{tree height})$

- How to merge two trees?
 - Hang one of the trees under the root of the other one
- Which one to hang?
 - A shorter one, since we would like to keep the trees shallow





- When merging two trees we hang a shorter one under the root of a taller one
- To quickly find a height of a tree, we will keep the height of each subtree in an array `rank[1...n]`: `rank[i]` is the height of the subtree whose root is `i`
- (The reason we call it rank, but not height will become clear later)
- Hanging a shorter tree under a taller one is called a union by rank heuristic