

# Topic 13

## Heap, Set and Map

資料結構與程式設計  
Data Structure and Programming

11/25/2015

### Linear Data Types

- ◆ In previous topic and Homework #5, we have learned linear data types like list and array
  - Tradeoffs between insert/delete/find operators
  - Memory overhead
  - ➔ Constant time for “push\_back()” or “push\_front()” operation
- ◆ The best way to use linear data types is ---
  - Data are recorded in a linear sequence (i.e. only push\_back or push\_front is needed)
  - Linearly traverse each element (i.e. for(...; li++))
  - No “find”, “insert any”, nor “delete any”

## Consider the Scenario...

- ◆ Suppose we are assigning jobs sequentially to several machines ---
  - One job to one machine and we record the accumulated runtime for each machine.
  - Our machine selection criteria is to “even out” the runtime of the machines.
  - In other words, we would like to pick the machine with least accumulated runtime for the next job
- ➔ Do we need to sort ALL the elements?
- ➔ Need a priority queue

## Priority Queue

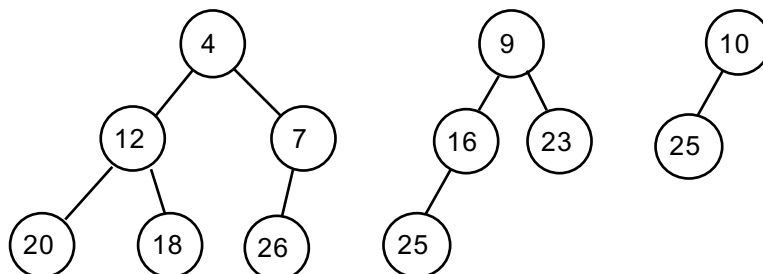
- ◆ An ADT that supports 2 operations
  - Insert
  - Delete min(or max)
- ◆ An element with arbitrary priority can be inserted to the queue
- ◆ At any time, it should take constant time to find the element with min(or max) priority and remove it from the list
  - Need to figure out which is the one with next lowest(highest) priority efficiently

## Using List or Array?

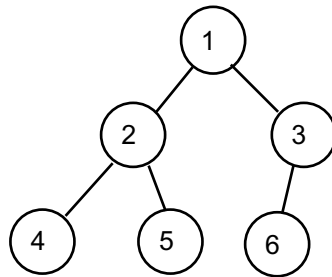
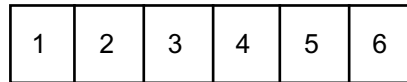
- ◆ Use linear ADT with an extra field to record the element with min(max) priority
  - Insert:  $O(1)$
  - Delete min(max):  $O(n)$   
(why?)
  
- ◆ As we learn before,  $O(n)$  is not good. We would prefer an ADT with  $O(\log n)$  for both operations

## Min (Max) Heap

- ◆ A complete binary tree in which the key value in each node is no larger (smaller) than its children



**Remember that we can use array to implement a complete binary tree...**

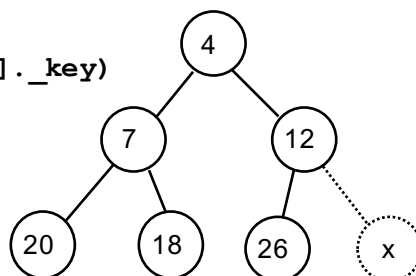


◆ Parent  
= child / 2

◆ Child  
= Parent \* 2  
or Parent \* 2 + 1

## MinHeap Insertion

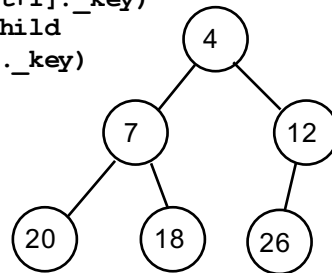
```
// Let n be the index of the last element
void MinHeap::insert(const T& x)
{
    int t = ++n; // next to the last
    while (t > 1) {
        int p = t / 2;
        if (x._key >= _heap[p]._key)
            break;
        _heap[t] = _heap[p];
        t = p;
    }
    _heap[t] = x;
}
```



**What's the time complexity?**

## Delete Min Element

```
T& MinHeap::deleteMin()
{
    T ret = _heap[1];
    int p = 1, t = 2 * p;
    while (t <= n) {
        if (t < n) // has right child
            if (_heap[t]._key > _heap[t+1]._key)
                ++t; // to the smaller child
        if (_heap[n]._key < _heap[t]._key)
            break;
        _heap[p] = _heap[t];
        p = t;
        t = 2 * p;
    }
    _heap[p] = _heap[n--];
    return ret;
}
```



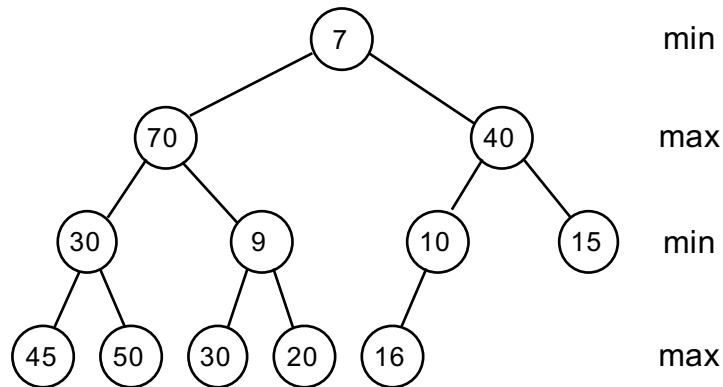
What's the time complexity?

## Min(Max) Heap

- ◆ Simple implementation (just an array)
- ◆ Good insertion and deleteMin complexity
  - $O(\log n)$  vs.  $O(n)$

What if you want to  
delete min AND delete max?

## Min-Max Heap



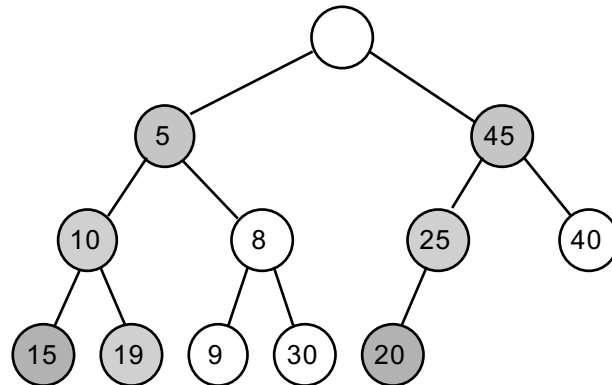
- Insert, delete min, delete max: all  $O(\log n)$  (why?)

## Deap

### ◆ Double-ended heap

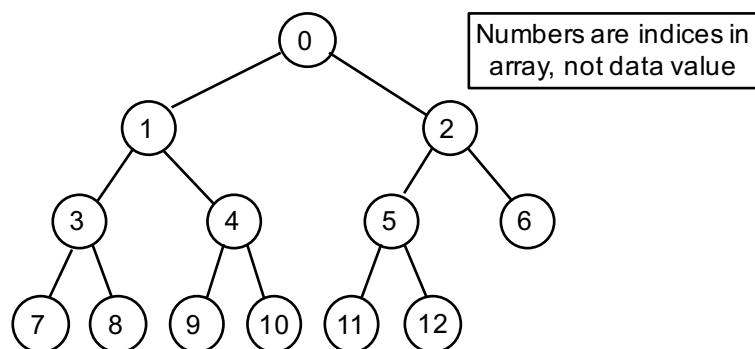
1. The root contains no element
2. The left subtree is a min heap
3. The right subtree is a max heap
4. Let  $i$  be any node in the left subtree. Let  $j$  be the corresponding node in the right subtree. If such a  $j$  node does not exist, then let  $j$  be the corresponding parent of  $i$ .  
→ The key in node  $i$  is less than or equal to that in  $j$ .

## Deap Example



- Insert, delete min, delete max: all  $O(\log n)$  (why?)
  - But faster than min-max heap by a constant factor
  - Algorithm is simpler

## Deap Implementation

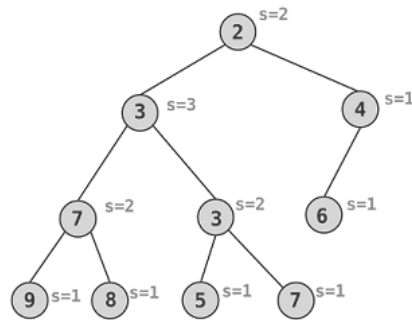


- Given a node 'i', how to find the "corresponding parent" or "corresponding child"?
- When insertion or deletion, what should we do when the node value is greater/smaller than its corresponding parent/child?

## More Varieties of Heaps: Leftist Heap

◆ In contrast to a *binary heap*, a leftist heap attempts to be very unbalanced.

- $s\text{-value}(v)$ :  
the distance to the nearest leaf.
- In addition to the heap property, the right child of each node has the lower  $s$ -value.



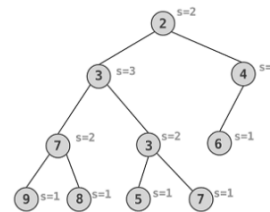
◆ Support  
“combine(heap1, heap2)” in  $O(\log n)$

## Leftist Heap: Huh?

◆ Remember: “combine(heap1, heap2)” in  $O(\log n)$

- Both “insert” and “deleteMin” operations can be realized by “combine”. (How?)

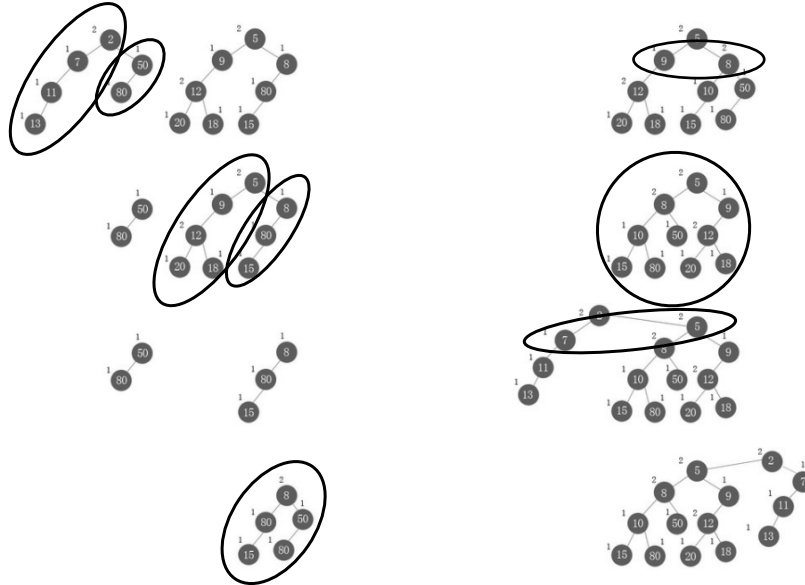
```
combine(h1, h2) {
    compare(min(h1), min(h2));
    // let min(hi) < min(hj)
    if (right(hi) == NULL)
        right(hi) = hj;
    else
        combine(right(hi), hj);
    // hj is now the combined heap
    if (s(right(hj)) > s(left(hj)))
        swap(right(hj), left(hj));
}
```





## Leftist Tree: Combine

[src] <http://blog.yam.com/rockmanray/article/44962825>



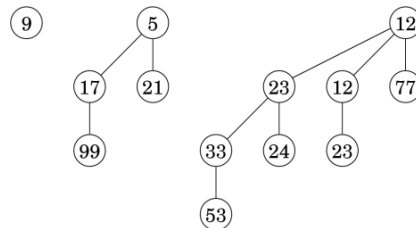
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## More Varieties of Heaps: Binomial heap

- ◆ Binomial tree of order  $k$ 
  - Binomial tree of order 0 is a single node
  - The root of a binomial tree of order  $k$  has  $k$  children, who are roots of binomial trees of order  $k-1, k-2, \dots, 0$
  - Has exactly  $2^k$  nodes; height =  $k$
- ◆ Binomial heap
  - A collection of Binomial trees
  - Most operations have the complexity  $O(\log n)$
  - But the amortized complexity is either  $O(1)$  or  $O(\log n)$



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## Binomial Heap: Properties

- ◆ Given a binomial heap with  $n$  nodes:
  - The node containing the min element is a root of  $B_0, B_1, \dots$ , or  $B_k$ .
  - It contains the binomial tree  $B_i$  iff  $b_i = 1$ , where  $b_k \cdot b_{k-1} \cdot b_{k-2} \cdot \dots \cdot b_1 \cdot b_0$  is binary representation of  $n$ .
  - It has  $\leq \lfloor \log_2 n \rfloor + 1$  binomial trees.
  - Its height  $\leq \lfloor \log_2 n \rfloor$ .

[src] <http://www.cs.princeton.edu/~wayne/leinberg-tardos/pdf/BinomialHeaps.pdf>

## Binomial Heap: Operations

- ◆ Similar to Leftist Heap, the operations of Binomial Heap can be realized by the “compose” (aka. “meld”) operation.
- ◆ Compose operation:
  - Binary addition
  - Given two binomial heaps
 
$$H_1 := \{ (B_3, B_2, B_1, B_0) = (1, 1, 0, 1) \}, \text{ and}$$

$$H_2 := \{ (B_4, B_3, B_2, B_1, B_0) = (1, 0, 1, 0, 1) \}.$$
 The composed binomial heap
 
$$H_m := \{ (B_5, B_4, B_3, B_2, B_1, B_0) = (1, 0, 0, 0, 1, 0) \}.$$

## Binomial Heap: Compose Operation

- ◆ Atomic operation:
  - Given two binomial trees  $B_i, B_j$ , with the same order  $k$ , then  $\text{compose}(B_i, B_j)$ :
    1. Connect the roots  $r_i, r_j$  of  $B_i, B_j$ .
    2. Choose  $\min(r_i, r_j)$  as the root of the composed tree
    3. The composed tree is of order  $k+1$

→ What if we have three binomial trees with the same order?
- ◆ The compose operation of two binomial heaps:
  1. Align the binomial trees of both heaps
  2. From the trees with the least order, perform tree composition
  3. Propagate to the next order of tree if necessary
- ◆ What's the time complexity?  $O(\log n)$

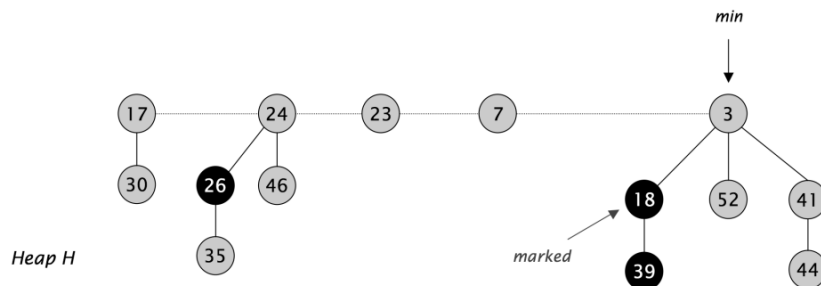
## Binomial Heap: Other Operations

- ◆ FindMin
  - // remember: It has  $\leq \lfloor \log_2 n \rfloor + 1$  binomial trees
  - $O(\log n)$
- ◆ DeleteMin
  - Note: after the "min" is removed, the corresponding binomial tree (of order  $k$ ) is broken and becomes  $k$  binomial trees
  - It just becomes "compose" operations of some binomial trees // How many?
  - $O(\log n)$
- ◆ DeleteNode(iterator pos)
  - $O(\log n)$
- ◆ Insert( $x$ )
  - $O(\log n)$

## More Varieties of Heaps: Fibonacci heap

### ◆ Fibonacci heap

- Especially useful when `deleteMin()` & `delete(n)` are rarely called → amortized  $O(\log n)$
- All other operations are  $O(1)$



## Fibonacci Heap

### ◆ Basic idea

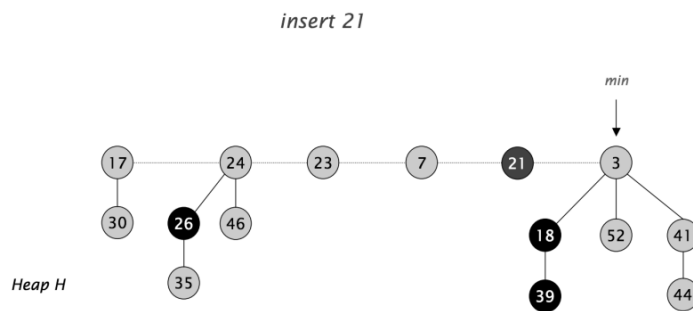
- Similar to binomial heaps, but less rigid structure
- Binomial heap: eagerly consolidate trees after each insert (maintain binomial structure)
- Fibonacci heap: lazily defer consolidation until next **delete-min**

### ◆ Properties

- Set of heap-ordered trees.
- Maintain pointer to minimum element
- Set of marked nodes

## Fibonacci Heap: Insert Operation

- ◆ Create a new singleton tree.
- ◆ Add to root list; update min pointer (if necessary)  $\rightarrow O(1)$



(Ref) <https://www.cs.princeton.edu/~wayne/teaching/fibonacci-heap.pdf>

## Fibonacci Heap: DeleteMin Operation

- ◆ Let H be a Fibonacci heap and x be a node
  - Rank(x): number of children of node x
  - Rank(H): max rank of any node in heap H
  - Tree(H): number of trees in heap H
- ◆ DeleteMin
  - Delete min; meld its children into root list; update min
  - Consolidate trees so that no two roots have same rank
  - $\rightarrow$  Time complexity:  $O(\text{rank}(H)) + O(\text{trees}(H))$
  - $\rightarrow$  Amortized cost:  $O(\text{rank}(H))$

## Heap Operations Supported in STL

- ◆ STL does not have a “heap” class
  - Instead, it support several operations that can operate on “array” like data structure
- ◆ Operations
  - `void make_heap(first, last[, comp]);`
  - `void push_heap(first, last[, comp]);`
  - `void pop_heap(first, last[, comp]);`
  - `void sort_heap(first, last[, comp]);`
  - `bool is_heap(first, last[, comp]);`
  - ➔ *first, last: RandomAccessIterator*
  - ➔ *comp: StrictWeakOrdering (optional)*

## Summary: Heap Structures

- ◆ Pros:
  1. Good complexity of “insert”, “delete min(max)”, ... operations
  2. Simple data structure (low memory overhead)
  3. Simpler algorithms (than BST)
- ◆ Con
  1. Data are not sorted
    - ➔ Still have  $O(n)$  for “find” operation

## Review: Binary Search Trees

### ◆ Binary Search Trees (BSTs)

- Left subtree  $\leq$  this  $\leq$  right subtree
- Complexity depends on the height of the tree
- Worst case: can be degenerated as a tree with height  $O(n)$

### ◆ Balanced BSTs

- The heights of left subtree and right subtree are somewhat balanced
  - Height  $\sim O(\log n)$
- Examples: AVL, 2-3, 2-3-4, red-black, splay trees
- Algorithms for their operations are complicated

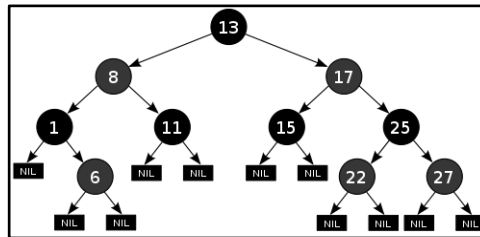
## Sorted ADT in STL

### ◆ Also classified as “Associative Containers”

1. set
  2. multiset
  3. map
  4. multimap
- ➔ Implemented in “red black tree”

## Red Black Tree

- ◆ A node is either red or **black**. The root is **black**
- ◆ All leaves are black (i.e. All leaves are same color as the root.)
- ◆ Every red node must have two **black** child nodes.
- ◆ Every path from a given node to any of its descendant leaves contains the same number of **black** nodes.
- ◆ Memory efficient
- ◆ Although balancing is NOT perfect,  $O(\log n)$  for insert, delete, and find



## class set in STL

- ◆ To store elements in a set
  - e.g. { 2, 3, 5, 7, 9 }
- ◆ `set<Key[, Compare, Alloc]>`
  - class Key: element type
  - class Compare: how the elements are compared (optional; default = `less<Key>`)
  - class Alloc: used for internal memory management (optional; default = `alloc`)



## Member Functions in class set

1. `iterator begin() const;`  
`iterator end() const;`
2. `pair<iterator, bool> insert(const value_type& x);`  
`iterator insert(iterator pos, const value_type& x);`  
`void insert(InputIterator, InputIterator);`
3. `void erase(iterator pos);`  
`size_type erase(const key_type& k);`  
`void erase(iterator first, iterator last);`
4. `iterator find(const key_type& k) const;`
5. `size_type count(const key_type& k) const;`
6. `iterator lower_bound(const key_type& k) const;`  
`iterator upper_bound(const key_type& k) const;`  
`pair<iterator, iterator> equal_range(const key_type& k) const;`

## Other Functions for class set

1. `includes`
  - Check if one set is included in another
2. `set_union`
3. `set_intersection`
4. `set_difference`
5. `set_symmetric_difference`
  - $(A - B) \cup (B - A)$

## class multiset in STL

- ◆ Unlike “set”, where elements with same value are stored only once, in multiset, they can be stored repeatedly
  - e.g. { 2, 3, 5, 5, 6, 7, 7, 7 }
- ◆ multiset<Key[, Compare, Alloc]>
  - class Key: element type
  - class Compare: how the elements are compared (optional; default = less<Key>)
  - class Alloc: used for internal memory management (optional; default = alloc)

## class map in STL

- ◆ In many applications, data are associated with keys (or id's)
  - For example, (id, student record)
  - e.g. { (Mary, 90), (John, 85), (Sam, 71) ... }
- ◆ class map<Key, Data[, Compare, Alloc]>
  - class Key: compare data type
  - class Data: value type
  - class Compare: how the elements are compared (optional; default = less<Key>)
  - class Alloc: used for internal memory management (optional; default = alloc)

## Example of using class map (1)

```
map<string, unsigned> scoreMap;  
scoreMap["Mary"] = 90;  
scoreMap["John"] = 85;  
scoreMap["Sam"] = 71;  
unsigned maryScore = scoreMap["Mary"];  
cout << "Mary's score = " << maryScore << endl;  
map<string, unsigned>::iterator mi;  
mi = scoreMap.find("John");  
if (mi != scoreMap.end())  
    cout << "John's score = " << (*mi).second << endl;  
➔ How about "map<const char*, unsigned>"?
```

## Comments about map::operator []

- ◆ Since operator[] might insert a new element into the map, it can't possibly be a const member function.
- ◆ Note that the definition of operator[] is extremely simple: m[k] is equivalent to `((m.insert(value_type(k, data_type()))).first).second`.
  - value\_type = pair<Key, Data>
  - insert(value\_type) returns a pair<map::iterator, bool>
- ◆ Strictly speaking, this member function is unnecessary: it exists only for convenience.

## Bad example of using class map

```
map<const char*, unsigned> mmm;  
map<const char*, unsigned>::iterator mi;  
char buf[1024];  
cin >> buf; mmm[buf] = 10;  
cin >> buf; mmm[buf] = 20;  
cin >> buf; unsigned s1 = mmm[buf];  
cout << buf << " = " << s1 << endl;  
cin >> buf; unsigned s2 = mmm[buf];  
cout << buf << " = " << s2 << endl;
```

## Example of using class map (2)

```
string str;  
for (int i = 0; i < 5; ++i) {  
    cin >> str; mm.insert(pair<string, int>(str, i));  
}  
while (1) {  
    cin >> str;  
    map<string, int>::iterator mi = mm.find(str);  
    if (mi == mm.end()) {  
        cout << "Not found!!" << endl;  
        break;  
    }  
    cout << (*mi).first << " = " << (*mi).second << endl;  
}
```

## Conclusion: Set and Map

- ◆ “set” and “map” are useful data structures when we need to perform efficient “insert”, “erase”, and “find” operations
  - Usually implemented by balanced binary search trees
  - Implementation efforts can be high
  - Using STL may be a good choice
- ◆ Remember, unbalanced BSTs may not be a bad choice
  - Most randomly inserted BSTs are somewhat balanced
- ◆ Remember, there’s no free lunch
  - Overhead in insert (vs. push\_back)
  - If we don’t need to do “erase” or “find” during insertions... (what’s the alternative?)