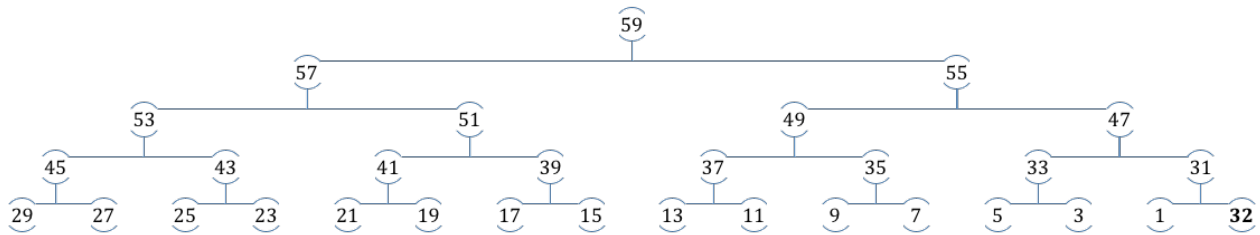


Data Structures and Algorithms, Spring 2015
Homework #5
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5.1 (1) R-8.24

Draw an example of a heap whose keys are all the odd numbers from 1 to 59 (with no repeats), such that the insertion of an entry with key 32 would cause up-heap bubbling to proceed all the way up to a child of the root (replacing that child's key with 32).



[Fig. 1]

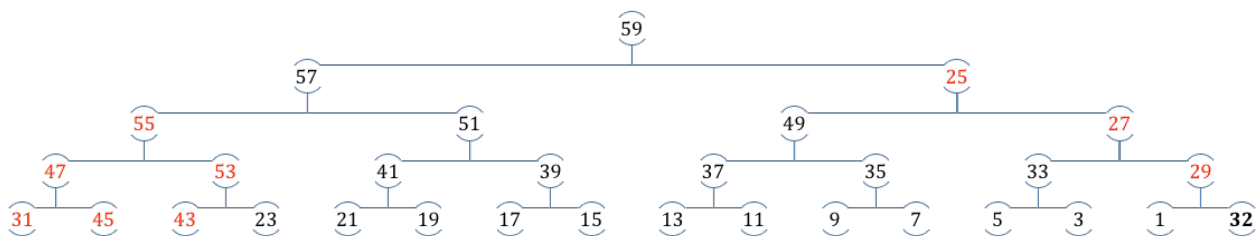
The original (trivial) case of such heap is in [Fig. 1].

Modification on [Fig.1] :

(1) pick up 31, 47, 55

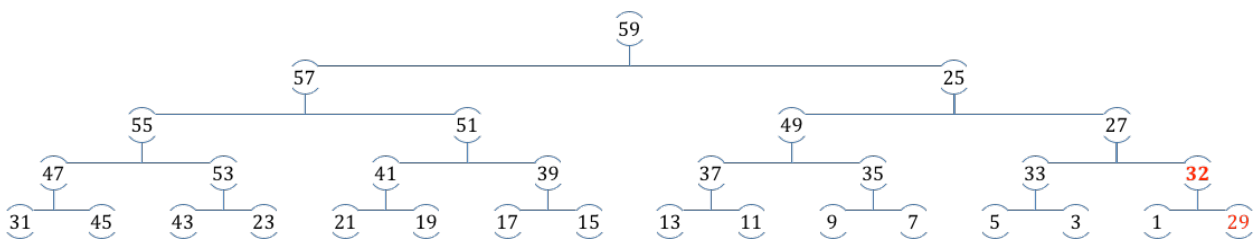
(2) replaced with 29, 27, 25

(3) maintain the heap, and then we can get the desired heap [Fig. 2]

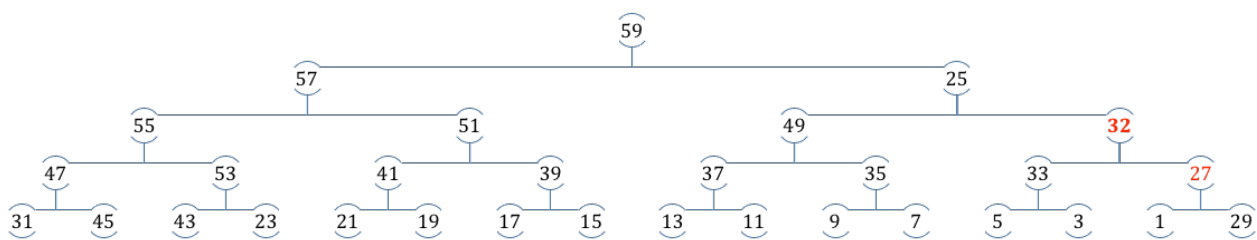


[Fig. 2] -> Answer!

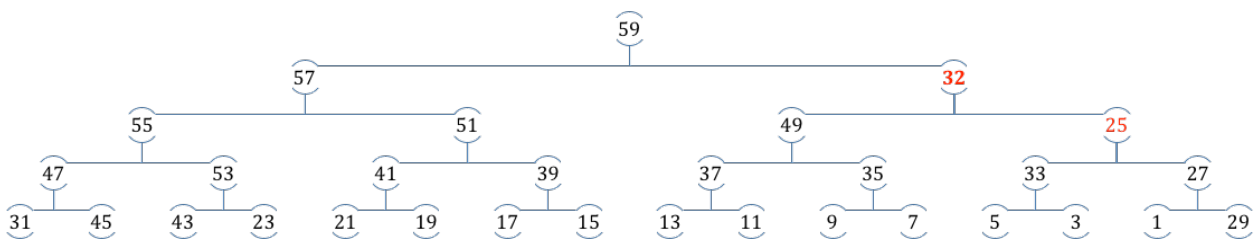
Process of up-heap bubbling:



[Fig. 3]



[Fig. 4]



[Fig. 5]

5.1 (2) C-8.4

Show how to implement the stack ADT using only a priority queue and one additional member variable.

```

1 class Stack {
2
3     class Node {
4
5         Node(int p, DATATYPE e)
6             :priority(p), element(e){};
7         int priority; //key
8         DATATYPE element; //value
9     };
10
11     priority_queue< Node > Q; // operator'<' is defined by node's priority
12     int top_priority = 0; //one additional member variable
13
14     void push(DATATYPE input)
15     {
16         Q.push( Node(top_priority++, input) );
17     }
18     DATATYPE pop()
19     {
20         top_priority--;
21         return q.pop().element;
22     }
23 };

```

[Fig. 6]

5.1 (3) C-8.14

Given a heap T and a key k , give an algorithm to compute all the entries in T with a key less than or equal to k . For example, given the heap of Figure 8.12(a) and query $k = 7$, the algorithm should report the entries with keys 2, 4, 5, 6, and 7 (but not necessarily in this order). Your algorithm should run in time proportional to the number of entries returned.

```

1 // suppose heap T is realized by "Level Numbering" method
2 // that is: (see p.295)
3 // If v is the root of T, then  $f(v) = 1$ 
4 // If v is the left child of node u, then  $f(v) = 2f(u)$ 
5 // If v is the right child of node u, then  $f(v) = 2f(u) + 1$ 
6
7 class Node {
8
9     Node(int _k, DATATYPE _v)
10         :key(_k), value(_v){};
11     int key;
12     DATATYPE value;
13 };
14 void dfs( Node heap[], int id, int k)
15 {
16     if( heap[id].key > k ) return;
17     cout << heap[id].value << " ";
18     dfs( heap, id * 2, k);
19     dfs( heap, id * 2 + 1, k);
20 }
21 void findKeyLowerThanK( Node T[], int k )
22 {
23     dfs( T, 1, k ); //1-based
24 }

```

[Fig. 7]

Ans. (see Fig.7)

In order to run in time proportional to the number of entries returned and NOT modifying the original heap, I chose depth-first search to realize the algorithm "findKeyLowerThanK".

5.1 (4)

Main Idea:

A technique for quickly estimating how similar two sets are.

Application:

1. Being applied in large-scale clustering problems, such as *clustering documents by the similarity of their sets of words*.
2. In data mining, Cohen et al. (2001) use MinHash as a tool for *association rule learning*.

Techniques:

1. MinHash Signature for Sets

將這些比較長、比較大的每一個集合都以一個特殊的 *signature* 來代替，單純透過比較這些 *signatures*，就能知道其所代表的集合間的 *Jaccard similarity*；如此一來，原本計算過程中需要對上萬個元素進行比較的運算就能替換為 *signature* 長度的運算。

2. Locality-Sensitive Hashing for Documents

MinHash中，我們透過LSH將具有類似 *signature* 的集合分到相同的 *bucket*；由於較相似的集合會比不相似的集合們更有機會分配到同一個 *bucket*，我們只需要在輸出結果前檢查那些被放在同一個 *bucket* 中的集合即可。

Processes:

1. 首先須決定 *minhash signature* 的長度 n ，並對各個集合算出其 *minhash signature*。
2. 依據應用情境選擇 b 與 r 值，要記住 b 與 r 決定了集合被納入 *candidate pairs* 的機率；如果我們希望避免 *false negative*，則減少 r 值，增加 b 值；如果希望限制 *false positive*，則可增加 r ，減少 b 。
3. 以選擇的參數進行 *LSH* 運算，取得 *candidate pairs*。
4. 對各個 *candidate pair* 的 *minhash signature* 計算其 *Jaccard similarity*，確認是否相似。
5. 若前述步驟都進行完畢尚有運算資源可運用，則可實際取得集合內容，再次確認其相似性。

Jaccard similarity Def:

$$J(A, B) = \frac{|A \cap B|}{|A \cup B|}.$$

ref:

<http://web.stanford.edu/class/cs276b/handouts/minhash.pdf>

<http://en.wikipedia.org/wiki/MinHash>

<http://shihpeng.org/tag/minhash/>

5.1 (5)

```
1 typedef unsigned long long int longint;
2
3 // O(1) Time Complexity
4 longint postfixHash(string str,int k);
5
6 int binary_search(string str_1, string str_2)
7 {
8     int left = 0;
9     int right = str.size() - 1;
10    while( left <= right )
11    {
12        int mid = (left + right) / 2;
13
14        if(postfixHash(str_1, mid) != postfixHash(str_2, mid))
15        {
16            if(str_1.size() == 1)
17                return mid;
18            else
19                left = mid + 1;
20        }
21        else // (postfixHash(str_1, mid) == postfixHash(str_2, mid))
22            right = mid - 1;
23    }
24    return "KEY_NOT_FOUND";
25 }
```

[Fig. 8]

Ans. (see Fig.8)

//Discuss and justify the time complexity of your algorithm:

我的做法是，引用二元搜尋樹的概念加上postfixHash()，目標為找到在兩長度相同的字串中，找到唯一不同字元的位置(因此回傳 int)。Binary_search的時間複雜度為 $O(\log N)$ ，而postHash()以由題目訂定為 $O(1)$ ，所以總共為 $O(\log N)$ 。

5.1 (6)

```
1 typedef unsigned long long int longint;
2
3 longint hash(string s) //BONUS
4 {
5     longint out = 0;
6     for(int i = 0; i < s.size(); i++)
7     {
8         out *= 27;
9         out += s[i] - 'a' + 1; //hash("a") == 1
10    }
11    return out;
12 }
13 // consider hash("register"):
14 // 190329075127, still in the range of longint
15 // consider hash("volatile"):
16 // 236112196676, still in the range of longint
```

[Fig. 9]

Ans. (see Fig.9)

為了方便計算，我使用27進位而非26進位，讓hash("a") 為1。

如此一來，這個hash可以用linear time的時間算出一個給定string 的hash value，有效率地計算此值，且發生collision的機率極低，在給定的32 個 standard words中無碰撞發生。