

# Advanced Data Structures

## Lecture 5

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## 1 Word RAM model

Words of size  $w$ , which can store a pointer or a value.  $w > \log u$ . An instruction on a word or pair of words takes  $O(1)$ . Following a pointer takes  $O(1)$ .

## 2 Predecessor/Successor Integer Problem

Given a set  $S$  of  $n$  integers from some universe  $U = 1, \dots, u - 1$ , where  $n \ll u$

### Operations

- member
- insert
- delete
- succ
- pred
- find-min/max

### Various Implementations

f	Data structure	Time	Space
1962	Binary Trees	$O(\log n)$	$O(n)$
1975	van Emde Boas trees	$O(\log \log u)$	$O(u)$
????	x-fast trie	$O(\log \log u)$	$O(n \log u)$
1983	y-fast trie	$O(\log \log u)$	$O(n)$
1993	fusion trees	$O(\frac{\log n}{\log \log u})^1$	$O(n)$

### 2.1 van Emde Boas Trees

The intuition is that we do a binary search on the representation, which gives us  $O(\log \log n)$ .

Basically, store 0 or 1 in a giant array. Then build a tree on top of it, where each node is 1 if any of the children are 1, and 0 if all of them are 0. Note that the depth of this tree is  $O(\log u)$ .

<sup>1</sup>Note that if  $\log \log u = \omega(\sqrt{\log n})$ , then we can sort integers in  $O(n\sqrt{n})$

To reduce the size further, we store the tree as a bitfield. Note that for some word  $x$ , there are  $\frac{1}{2} \log u$  high bits, and the same number of low bits. Note that there are  $\sqrt{u}$  groups of the giant array with  $\sqrt{u}$  elements in them. As such, we pack two lower trees into each word, and use higher layers to route to the lower trees.

**Formally** Recursive data structure with  $\sqrt{u} + 1$  sub-structures, each for a range  $\sqrt{u}$ . We refer to these as  $S[0], \dots, S[\sqrt{u} - 1]$ . We refer to the summary structure as  $S.summary$ .  $S.summary[i] = 1$  iff  $S[i]$  is not empty.

### 2.1.1 Membership

```
member(x):
    S[high(x)].member(low(x))
```

**Complexity**

$$T[\log u] = T\left[\frac{\log u}{2}\right] + \Theta(1) = \Theta(\log \log u)$$

### 2.1.2 Insertion

```
insert(x):
    S[high(x)].insert(low(x))
    S.summary.insert(high(x))
```

**Complexity**

$$T[\log u] = 2T\left[\frac{\log u}{2}\right] + \Theta(1) = \Theta(\log u)$$

**A better way** Store the first element explicitly (without making the recursive calls) in  $S.min$ .

```
insert(x):
    if S.min is nil:
        S.max := x
        S.min := x and return
    if x < S.min:
        swap x, S.min
    if x > S.max:
        S.max := x
    if S[high(x)].min is nil:
        S[high(x)].min := low(x)
        S[high(x)].max := low(x)
        S.summary.insert(high(x))
    else:
        S[high(x)].insert(low(x))
```

### 2.1.3 Successor

We also need to store the last element explicitly (without making the recursive calls) in  $S.max$ .

```

succ(x):
    if x < S.min:
        return S.min
    if low(x) < S[high(x)].max:
        return S[high(x)].succ(low(x))
    else
        succ_high = S.summary.succ(high(x))
        return S[succ_high].min | (succ_high * sqrt(u))

```

## 2.2 x-fast tries

Similar to van Emde Boas trees, but store the path as a sparse tree. We just need to find the closest node  $y$  that is 1 on the path from  $x$  to the root. If  $x$  is in the right subtree of  $y$ ,  $\text{pred}(x)$  is the maximum in the left subtree. If  $x$  is in the left subtree,  $\text{succ}(x)$  is the minimum in the right subtree. Every node in this tree we will give a name which is the binary prefix path to the node for each element  $x$  in the set  $S$ .

We store this in a dynamic hash table, which requires  $\Theta(n \cdot \log u)$ .

**Successor/Predecessor** Find  $y$  by binary search on the path from  $x$  to the root (binary prefixes of  $x$ ). This takes  $O(\log \log u)$  time.

**Insert** Insert 1 for all the binary prefixes of  $x$ . This takes  $O(\log u)$  amortized time.

**Member** Check the hash for  $x$ .  $O(1)$

## 2.3 y-fast tries

Chunk into  $n' = \frac{n}{\log u}$  groups of  $\log u$  elements, store them in BSTs. Track roots of groups in x-fast trie. This gives us  $\Theta(n' \cdot \log u) = \Theta(n)$  space, and the BSTs take  $\frac{n}{\log u} \log u = \Theta(n)$  space. Note that the operations on the BSTs take  $O(\log n)$ . We need to ensure that the size of each of these BSTs is in  $[\frac{1}{2} \log u, 2 \log u]$ .

**Member**

**Successor/Predecessor**

**Insertion**

**Indirection** Compress successive elements of  $S$  into groups of  $\Theta(\log u)$ . Store the group as a BST, where all operations take  $O(\log \log u)$ . Store a representative of the group in a x-fast trie. The representative values are not required to even be in the groups.

← There are  $O(\log n)$  variants of all operations on BSTs, including split/merge

**Queries** Given  $x$  - find the successor/predecessor  $y$  of  $x$  in the x-fast trie. Run the operation in the BST for  $y$ , or in the BST after  $y$ .

**Insertion/Deletion** Find the BST  $y$  for  $x$ , and run the operation on  $y$ . If  $y$  becomes too large, split it into  $y_1, y_2$ , and enter them into the x-fast trie. If  $y$  becomes too small, merge it with its neighbor, potentially splitting the combined tree, and update the x-fast trie.

Note that insertions/deletions on the x-fast trie happen once every  $\Theta(\log u)$  operations, and therefore take  $\Theta(\log \log u)$  amortized time.

## 3 Self-adjusting Data Structures - Competitive Analysis

### 3.1 Competitive Analysis

Competitive Analysis compares our algorithm against one that has perfect knowledge. Is used in the Advanced Algorithms course to evaluate online algorithms.

**Worst case**  $\Theta(n)$

**Average case** Access each element with uniform probability.  $\Theta(n)$

**Stochastic** Access each element  $i$  with probability  $p_i$  such that  $p_1 \geq p_2 \geq \dots \geq p_n$ .  $\sum_{i=1}^n i \cdot p_i$

**Static generic** Sequence of queries  $\sigma = x_1, \dots, x_m$ . Define  $f_i$  be the number of queries for element  $i$ . An optimal static solution is to order the elements by  $f_i$  descending.

**Dynamic generic** we are allowed to change the order of the list in runtime. The dynamic optimal solution is the most efficient solution given perfect knowledge.

**Sleator Trajan Cost Model** When we search for the element  $x$  in the place  $i$ , we pay  $i$ . To move the element  $x$  to the front is free. Switching between neighbors costs us 1.

### 3.2 Linked Lists

A list of  $n$  elements  $1, 2, \dots, n$ .

#### 3.2.1 Access

Walk the list to find the element.

**Possible strategy**

**Move to Front** Move each element to the front when accessed.

**Frequency Count** Each time an element is accessed, reorder the list by frequency.

**Transpose** Promote on each access.

## **4 next week**

splay trees