

## Lecture 4: Divide and Conquer: van Emde Boas Trees

- Series of Improved Data Structures
- Insert, Successor
- Delete
- Space

This lecture is based on personal communication with Michael Bender, 2001.

### Goal

We want to maintain  $n$  elements in the range  $\{0, 1, 2, \dots, u - 1\}$  and perform Insert, Delete and Successor operations in  $\mathcal{O}(\log \log u)$  time.

- If  $n = n^{\mathcal{O}(1)}$  or  $n^{(\log n)^{\mathcal{O}(1)}}$ , then we have  $\mathcal{O}(\log \log n)$  time operations
  - Exponentially faster than Balanced Binary Search Trees
  - Cooler queries than hashing
- Application: Network Routing Tables
  - $u$  = Range of IP Addresses  $\rightarrow$  port to send ( $u = 2^{32}$  in IPv4)

**Where might the  $\mathcal{O}(\log \log u)$  bound arise ?**

- Binary search over  $\mathcal{O}(\log u)$  elements
- Recurrences
  - $T(\log u) = T\left(\frac{\log u}{2}\right) + \mathcal{O}(1)$
  - $T(u) = T(\sqrt{u}) + \mathcal{O}(1)$

### Improvements

We will develop the van Emde Boas data structure by a series of improvements on a very simple data structure.

## Bit Vector

We maintain a vector  $V$  of size  $u$  such that  $V[x] = 1$  if and only if  $x$  is in the set. Now, inserts and deletes can be performed by just flipping the corresponding bit in the vector. However, successor/predecessor requires us to traverse through the vector to find the next 1-bit.

- Insert/Delete:  $\mathcal{O}(1)$
- Successor/Predecessor:  $\mathcal{O}(u)$

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	1	0	0	0	0	0	0	0	1	1	0	0	0	0	1

Figure 1: Bit vector for  $u = 16$ . The current set is  $\{1, 9, 10, 15\}$ .

## Split Universe into Clusters

We can improve performance by splitting up the range  $\{0, 1, 2, \dots, u-1\}$  into  $\sqrt{u}$  clusters of size  $\sqrt{u}$ . If  $x = i\sqrt{u} + j$ , then  $V[x] = V.Cluster[i][j]$ .

$$low(x) = x \bmod \sqrt{u} = j$$

$$high(x) = \left\lfloor \frac{x}{\sqrt{u}} \right\rfloor = i$$

$$index(i, j) = i\sqrt{u} + j$$

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	1	0	0	0	0	0	0	0	1	1	0	0	0	0	1
$V.Cluster[0]$				$V.Cluster[1]$				$V.Cluster[2]$				$V.Cluster[3]$			

Figure 2: Bit vector ( $u = 16$ ) split into  $\sqrt{16} = 4$  clusters of size 4.

- Insert:
  - Set  $V.cluster[high(x)][low(x)] = 1$   $\mathcal{O}(1)$

- Mark cluster  $high(x)$  as non-empty  $\mathcal{O}(1)$
- Successor:
  - Look within cluster  $high(x)$   $\mathcal{O}(\sqrt{u})$
  - Else, find next non-empty cluster  $i$   $\mathcal{O}(\sqrt{u})$
  - Find minimum entry  $j$  in that cluster  $\mathcal{O}(\sqrt{u})$
  - Return  $index(i, j)$  Total =  $\mathcal{O}(\sqrt{u})$

## Recurse

The three operations in Successor are also Successor calls to vectors of size  $\sqrt{u}$ . We can use recursion to speed things up.

- $V.cluster[i]$  is a size- $\sqrt{u}$  van Emde Boas structure ( $\forall 0 \leq i < \sqrt{u}$ )
- $V.summary$  is a size- $\sqrt{u}$  van Emde Boas structure
- $V.summary[i]$  indicates whether  $V.cluster[i]$  is nonempty

INSERT( $V, x$ )

- 1  $Insert(V.cluster[high(x)], low[x])$
- 2  $Insert(V.summary, high[x])$

So, we get the recurrence:

$$\begin{aligned}
 T(u) &= 2T(\sqrt{u}) + \mathcal{O}(1) \\
 T'(\log u) &= 2T'\left(\frac{\log u}{2}\right) + \mathcal{O}(1) \\
 \implies T(u) &= T'(\log u) = \mathcal{O}(\log u)
 \end{aligned}$$

SUCCESSOR( $V, x$ )

- 1  $i = high(x)$
- 2  $j = Successor(V.cluster[i], j)$
- 3 **if**  $j == \infty$
- 4      $i = Successor(V.summary, i)$
- 5      $j = Successor(V.cluster[i], -\infty)$
- 6 **return**  $index(i, j)$

$$\begin{aligned}
 T(u) &= 3T(\sqrt{u}) + \mathcal{O}(1) \\
 T'(\log u) &= 3T'\left(\frac{\log u}{2}\right) + \mathcal{O}(1) \\
 \implies T(u) &= T'(\log u) = \mathcal{O}((\log u)^{\log 3}) \approx \mathcal{O}((\log u)^{1.585})
 \end{aligned}$$

To obtain the  $\mathcal{O}(\log \log u)$  running time, we need to reduce the number of recursions to one.

## Maintain Min and Max

We store the minimum and maximum entry in each structure. This gives an  $\mathcal{O}(1)$  time overhead for each *Insert* operation.

SUCCESSOR( $V, x$ )

```

1   $i = \text{high}(x)$ 
2  if  $\text{low}(x) < V.\text{cluster}[i].\text{max}$ 
3       $j = \text{Successor}(V.\text{cluster}[i], \text{low}(x))$ 
4  else  $i = \text{Successor}(V.\text{summary}, \text{high}(x))$ 
5       $j = V.\text{cluster}[i].\text{min}$ 
6  return  $\text{index}(i, j)$ 

```

$$\begin{aligned}
 T(u) &= T(\sqrt{u}) + \mathcal{O}(1) \\
 \implies T(u) &= \mathcal{O}(\log \log u)
 \end{aligned}$$

## Don't store Min recursively

The *Successor* call now needs to check for the min separately.

**if**  $x < V.\text{min}$  : **return**  $V.\text{min}$  (1)

INSERT( $V, x$ )

```

1  if  $V.min == None$ 
2       $V.min = V.max = x$     ▷  $\mathcal{O}(1)$  time
3      return
4  if  $x < V.min$ 
5      swap( $x \leftrightarrow V.min$ )
6  if  $x > V.max$ 
7       $V.max = x$ 
8  if  $V.cluster[high(x)] == None$ 
9      Insert( $V.summary, high(x)$ )    ▷ First Call
10 Insert( $V.cluster[high(x)], low(x)$ )    ▷ Second Call

```

If the **first call** is executed, the **second call** only takes  $\mathcal{O}(1)$  time. So

$$T(u) = T(\sqrt{u}) + \mathcal{O}(1)$$

$$\implies T(u) = \mathcal{O}(\log \log u)$$

DELETE( $V, x$ )

```

1  if  $x == V.min$     ▷ Find new min
2       $i = V.summary.min$ 
3      if  $i = None$ 
4           $V.min = V.max = None$     ▷  $\mathcal{O}(1)$  time
5          return
6       $V.min = index(i, V.cluster[i].min)$     ▷ Unstore new min
7  Delete( $V.cluster[high(x)], low(x)$ )    ▷ First Call
8  if  $V.cluster[high(x)].min == None$ 
9      Delete( $V.summary, high(x)$ )    ▷ Second Call
10 ▷ Now we update  $V.max$ 
11 if  $x == V.max$ 
12 if  $V.summary.max = None$ 
13 else
14      $i = V.summary.max$ 
15      $V.max = index(i, V.cluster[i].max)$ 

```

If the **second call** is executed, the **first call** only takes  $\mathcal{O}(1)$  time. So

$$T(u) = T(\sqrt{u}) + \mathcal{O}(1)$$

$$\implies T(u) = \mathcal{O}(\log \log u)$$

## Lower Bound [Patrascu & Thorup 2007]

Even for static queries (no Insert/Delete)

- $\Omega(\log \log u)$  time per query for  $u = n^{(\log n)^{\mathcal{O}(1)}}$
- $\mathcal{O}(n \cdot \text{poly}(\log n))$  space

## Space Improvements

We can improve from  $\Theta(u)$  to  $\mathcal{O}(n \log \log u)$ .

- Only create nonempty clusters
  - If  $V.min$  becomes *None*, deallocate  $V$
- Store  $V.cluster$  as a hashtable of nonempty clusters
- Each insert may create a new structure  $\Theta(\log \log u)$  times (each empty insert)
  - Can actually happen [Vladimir Čunát]
- Charge pointer to structure (and associated hash table entry) to the structure

This gives us  $\mathcal{O}(n \log \log u)$  space (but randomized).

## Indirection

We can further reduce to  $\mathcal{O}(n)$  space.

- Store vEB structure with  $n = \mathcal{O}(\log \log u)$  using BST or even an array
  - $\implies \mathcal{O}(\log \log n)$  time once in base case
- We use  $\mathcal{O}(n / \log \log u)$  such structures (disjoint)
  - $\implies \mathcal{O}(\frac{n}{\log \log u} \cdot \log \log u) = \mathcal{O}(n)$  space for small
- Larger structures “store” pointers to them
  - $\implies \mathcal{O}(\frac{n}{\log \log u} \cdot \log \log u) = \mathcal{O}(n)$  space for large
- Details: Split/Merge small structures

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