

Lecture 9: Integers

- word RAM model (we want to look into the bits of the representation)

predecessor problem

- maintain a set S of n words out of a universe U of size 2^w where
- insert ($x \in U$) w is the word length (e.g. $w = 32$ for 32bit int)
- delete ($x \in S$)
- predecessor ($x \in U$) = $\max\{y \in S, y \leq x\}$ assume $w \geq \log n$ which allows to store n in one word

(Note: in comparison model we know that $\Theta(\log n)$ is the answer, LB: sort, UB: BBST)

Technique	Operation	Space	
Van Emde Boas tree (1975)	$O(\log w)$	$O(U)$	} good if w is small } today
+ hashing	$O(\log w)$ w.h.p.	$\Theta(n)$	
γ -fast trees	$O(\log w)$ w.h.p.	$\Theta(n)$	
fusion trees { static dynamic	$O(\frac{\log n}{\log w})$ w.h.p.	$\Theta(n)$	} good if w is large } next week

↳ all combined: $\min(\log w, \frac{\log n}{\log w}) \leq \sqrt{\log n}$ so provably faster than Bal Bin Search Trees

Van Emde Boas Trees

How to get $\log \log U$? Binary search on the word length?

Main idea is to target this recurrence:

$$T(U) = T(\sqrt{U}) + O(1) \text{ or binary search on } w \text{ by splitting } U \text{ into } \sqrt{U} \text{ clusters of size } \sqrt{U}$$

word $x = \langle c, i \rangle$

$$c = \lfloor x / \sqrt{U} \rfloor$$

$$i = x \% \sqrt{U}$$

$$x = c \cdot \sqrt{U} + i$$

hard to do on CPU

$$c = x \gg (w/2)$$

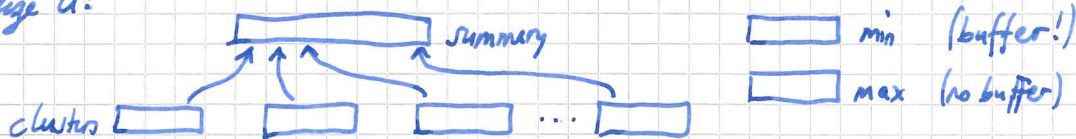
$$i = x \& ((1 \ll (w/2)) - 1)$$

$$x = (c \ll (w/2)) | i$$

easier by just halving word length



VEB of size U :



- V . cluster $[i] = \text{VEB of size } \sqrt{U}, 0 \leq i \leq \sqrt{U}$

- V . summary = VEB of size \sqrt{U} containing the indices of non-empty clusters

- V . min = minimum element in V (not stored recursively!)

- V . max = maximum element in V (also stored recursively, just auxiliary information)

successor($v, x = \langle c, i \rangle$) ← if $x < v.min$: return $v.min$
 if $i < v.cluster[c].max$:
 return $\langle c, successor(v.cluster[c], i) \rangle$ } search in lower bits
 else:
 $c' = successor(v.summary, c)$
 return $\langle c', v.cluster[c'].min \rangle$ } search in higher bits

insert($v, x = \langle c, i \rangle$):

if $v.min = None$: $v.min = v.max = x$; return (x) } update min/max
 if $x < v.min$: swap $x \leftrightarrow v.min$
 if $x > v.max$: $v.max = x$
 if $v.cluster[c].min = None$ } recursive insertion
 insert($v.summary, c$)
 insert($v.cluster[c], i$) } if both recursive calls are made, then the second is in $O(1)$, by $(x) \rightarrow$ so at most one non-trivial recursive call!

delete($v, x = \langle c, i \rangle$)

if $x = v.min$:
 if $x = v.max$:
 $v.max = v.min = None$; return (x) } update min
 $c = v.summary.min$
 $i = v.cluster[c].min$
 $v.min = \langle c, i \rangle$
 delete($v.cluster[c], i$) } recursive deletion
 if $v.cluster[c].min = None$:
 delete($v.summary, c$) } (again: if two calls are made, one is trivial by (x))
 if $v.summary.min = None$:
 $v.max = v.min$
 else:
 $c' = v.summary.max$
 $v.max = \langle c', v.cluster[c'].max \rangle$ } update max

→ recurrence $T(2^w) = T(2^{w/2}) + O(1) \rightarrow T(2^w) = \log(w)$

Saving Space: Expanded tree view:

first idea:

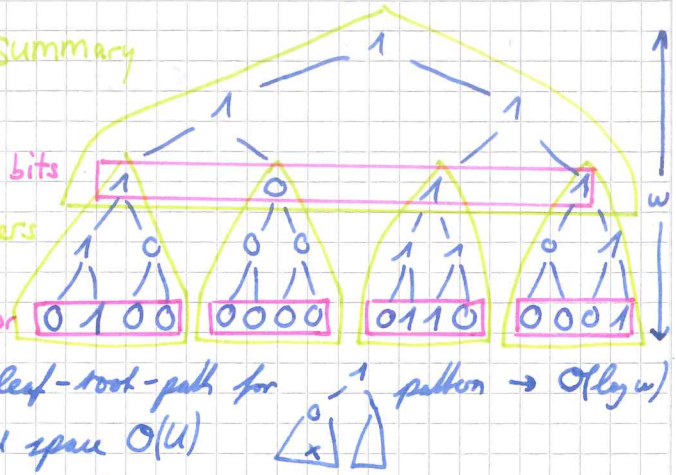
90's

- node = OR of its children
- tree stores min/max per subtree
- linked list of 1-bits
- update: $O(w)$

Summary

summary bits
clusters

bit vector

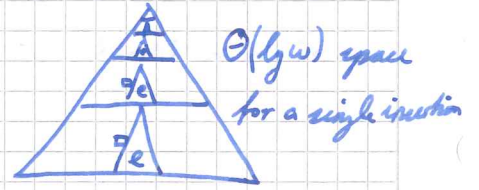


- query: bin search on the leaf-root-path for '1' pattern $\rightarrow O(\log w)$
 \rightarrow problem: $O(w)$ update and space $O(u)$

second idea:

2011

- don't store empty clusters in VEB
 - v. clusters = hash table $O(1)$ w.h.p. (FKS perfect hashing)
 - space = $O(\# \text{non-empty clusters} + 1)$
 - change each table entry to min in its child
- \rightarrow improves space to $O(n \cdot \log w)$



third idea:

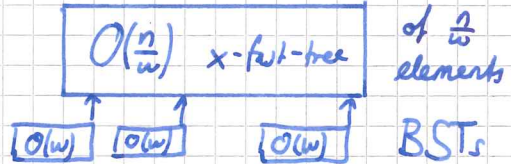
x-fast trees
1977

- take tree view idea from idea 1
 - store hash table of each level i 's position
- \rightarrow gives $O(\log w)$ query, $O(w)$ update in $O(n \cdot w)$ space (w.h.p.)

finally: y-fast trees: x-fast trees + indirection

- $O(\log w)$ query
- $O(\log w)$ amortized update w.h.p.
- $O(n)$ space

Indirection



query: top: $O(\log w)$, bottom $O(\log w)$
 update: $O(\log w)$ bottom, $O(w)$ amortized to $O(w)$ changes
 space: $O(\frac{n}{w} \cdot w + n) = O(n)$

same indirection idea works on the second idea to get to $O(n)$ space