Advanced algorithms and data structures

Lecture 6: van Emde Boas Trees

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Today's Lecture

van Emde Boas Trees

Predecessor search/ordered sets

Naive

Twolevel

Recursive

vEB: worst case $\mathcal{O}(\log \log |U|)$ time

RS-vEB: expected $\mathcal{O}(\log \log |U|)$ time, $\mathcal{O}(n \log \log |U|)$ space

R²S-vEB: expected $\mathcal{O}(\log \log |U|)$ time, $\mathcal{O}(n)$ space

Bonus: vEB is optimal for $w = \Theta(\log n)$

Bonus: Integer sorting in expected $\mathcal{O}(n \log \log |U|)$ time.

Predecessor search/ordered sets

Problem:

```
Given a universe U = [u] where u = 2^w,
maintain subset S \subseteq U, |S| = n under:
member(x, S): Return [x \in S].
insert(x, S): Add x to S (assumes x \notin S).
delete(x, S): Remove x from S (assumes x \in S).
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member(x, S): Return [x \in S].
insert(x, S): Add x to S (assumes x \notin S).
delete(x, S): Remove x from S (assumes x \in S).
empty(S): Return [S = \emptyset].
min(S):
               Return min S (assumes S \neq \emptyset).
               Return max S (assumes S \neq \emptyset).
\max(S):
```

Predecessor search/ordered sets

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maintain subset S \subseteq U, |S| = n under:
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member(x, S): Return [x \in S].
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$$insert(x, S)$$
: Add x to S (assumes $x \notin S$).

$$delete(x, S)$$
: Remove x from S (assumes $x \in S$).

$$empty(S)$$
: Return $[S = \emptyset]$.

$$\min(S)$$
: Return $\min S$ (assumes $S \neq \emptyset$).

$$\max(S)$$
: Return $\max S$ (assumes $S \neq \emptyset$).

predecessor(
$$x$$
, S): Return max{ $y \in S \mid y < x$ }

(assumes
$$\{y \in S \mid y < x\}$$
 is nonempty, i.e. $S \neq \emptyset$ and $x > \min(S)$).

successor(
$$x, S$$
): Return min{ $y \in S \mid y > x$ }

(assumes
$$\{y \in S \mid y > x\}$$
 is nonempty, i.e. $S \neq \emptyset$ and $x < \max(S)$).

i.e.
$$S \neq \emptyset$$
 and $x < \max(S)$

Idea: If we are willing to spend $\mathcal{O}(|U|)$ space. . .



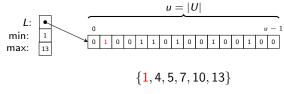
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How fast is:
empty(S), min(S), and max(S)?
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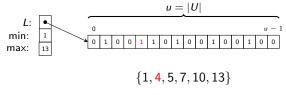
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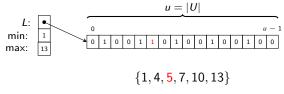
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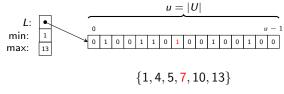
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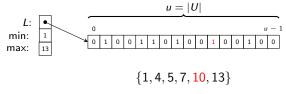
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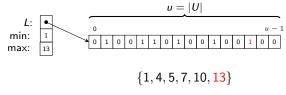
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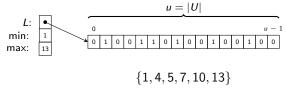
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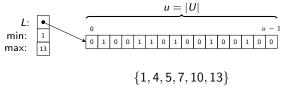
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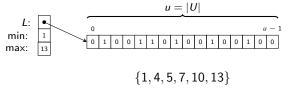
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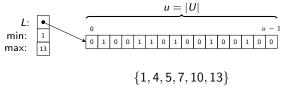
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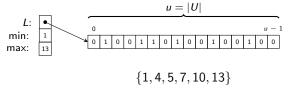
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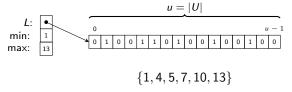
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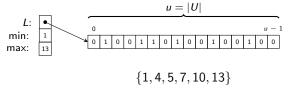
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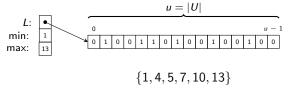
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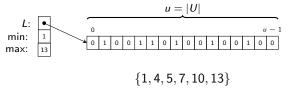
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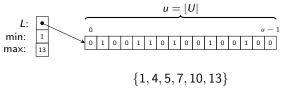
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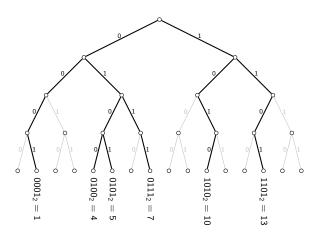
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```
How fast is:  \begin{split} & \operatorname{empty}(S), \ \min(S), \ \operatorname{and} \ \max(S)? \ \operatorname{worst} \ \operatorname{case} \ \mathcal{O}(1). \\ & \operatorname{member}(x,S)? \ \operatorname{worst} \ \operatorname{case} \ \mathcal{O}(1). \\ & \operatorname{predecessor}(x,S) \ \operatorname{and} \ \operatorname{successor}(x,S)? \ \operatorname{worst} \ \operatorname{case} \ \Theta(|U|). \\ & \operatorname{delete}(x,S)? \ \operatorname{worst} \ \operatorname{case} \ \mathcal{O}(1). \end{split}
```

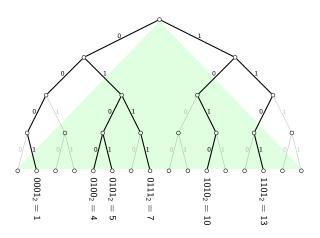
Bit-Trie

Idea: Think of each key as a *w*-bit string describing a path in a binary *trie* (= a special kind of tree). The naive structure ignores all intermediate branches and jump directly to the leaves. What if we split each key into a high and a low part?



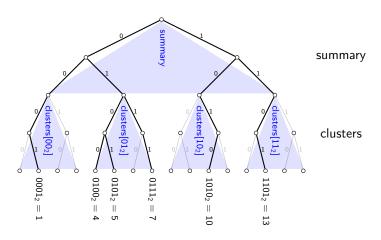
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Idea: Split each key into *high* and *low* parts. Use naive for each.

Recall that $U = [2^w]$, and define:

$$hi_w(x) := \left\lfloor \frac{x}{2^{\lceil w/2 \rceil}} \right\rfloor$$

$$lo_w(x) := x \mod 2^{\lceil w/2 \rceil}$$

$$x_w(h, \ell) := h \cdot 2^{\lceil w/2 \rceil} + \ell$$

$$w = \log_2|U| \text{ bits}$$

$$\text{hi}_w(x) \qquad \log_w(x)$$

 $\operatorname{Hote that} X = \operatorname{Hidex}_{W}(\operatorname{Hi}_{W}(X), \operatorname{IO}_{W}(X)).$

Now let the structure directly store the values:

$$\mathsf{min} := \mathsf{min}(S) \text{ if } S \neq \emptyset, \text{ else } 1$$
 $\mathsf{max} := \mathsf{max}(S) \text{ if } S \neq \emptyset, \text{ else } 0$

and use the naive structure to store

$$\begin{split} \text{summary} &:= \{ \mathsf{hi}_w(x) \mid w \in S \} \\ & \mathsf{clusters}[h] := \{ \ell \in [2^{\lceil w/2 \rceil}] \mid \mathsf{index}_w(h,\ell) \in S \} \qquad \forall h \in [2^{\lfloor w/2 \rfloor}] \\ \mathsf{note that} \ S &= \bigcup_{h \in [2^{\lfloor w/2 \rfloor}]} \{ \mathsf{index}_w(h,\ell) \mid \ell \in \mathsf{clusters}[h] \}. \end{split}$$

Idea: Split each key into *high* and *low* parts. Use naive for each.

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$$\begin{aligned} \operatorname{hi}_w(x) &:= \left\lfloor \frac{x}{2^{\lceil w/2 \rceil}} \right\rfloor \\ \operatorname{lo}_w(x) &:= x \bmod 2^{\lceil w/2 \rceil} \\ \operatorname{index}_w(h,\ell) &:= h \cdot 2^{\lceil w/2 \rceil} + \ell \\ \operatorname{note that} x &= \operatorname{index}_w(\operatorname{hi}_w(x), \operatorname{lo}_w(x)). \end{aligned}$$

$$x: \begin{array}{c} w = \log_2 |U| \text{ bits} \\ \\ hi_w(x) & lo_w(x) \end{array}$$

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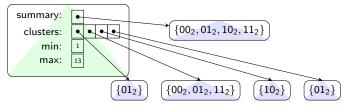
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note that $x = index_w(hi_w(x), lo_w(x))$.

$$\begin{split} \operatorname{summary} &:= \{\operatorname{hi}_w(x) \mid w \in S\} \\ \operatorname{clusters}[h] &:= \{\ell \in [2^{\lceil w/2 \rceil}] \mid \operatorname{index}_w(h,\ell) \in S\} \qquad \forall h \in [2^{\lfloor w/2 \rfloor}] \\ \operatorname{note that} S &= \bigcup_{h \in [2^{\lfloor w/2 \rfloor}]} \{\operatorname{index}_w(h,\ell) \mid \ell \in \operatorname{clusters}[h]\}. \end{split}$$

We can draw the structure for the set $S = \{1,4,5,7,10,13\} = \{0001_2,0100_2,0101_2,0111_2,1010_2,1101_2\} \subseteq [2^4]$ as:



```
How fast (worst case) is:

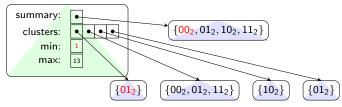
empty(S), min(S), max(S)?

member(x, S)?

predecessor(x, S), successor(x, S)?

delete(x, S)?
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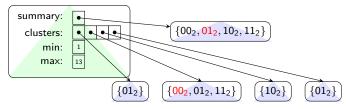
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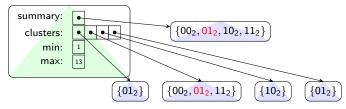
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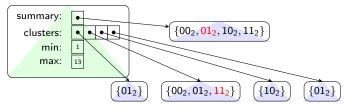
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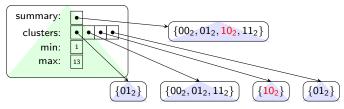


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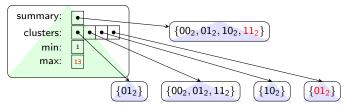


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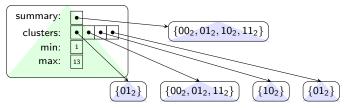
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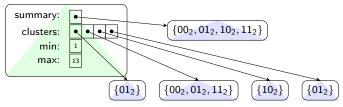
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empty(S), min(S), max(S)?

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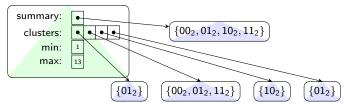
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How fast (worst case) is:
empty(S), min(S), max(S)?
\mathcal{O}(1)
member(x, S)?
predecessor(x, S), successor(x, S)
```

```
function EMPTY(S)
return S.min > S.max
function MIN(S)
\Rightarrow Assumes S \neq \emptyset
return S.min
function MAX(S)
\Rightarrow Assumes S \neq \emptyset
return S.max
```

delete(x, S)?

insert(x, S)

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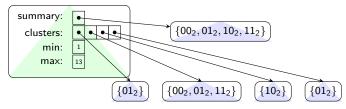
```
empty(S), min(S), max(S)? O(1)
member(x, S)?
```

```
predecessor(x, S), successor(x, S)
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delete(x, S)

insert(x, S)?

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```
How fast (worst case) is:

empty(S), min(S), max(S)?

\mathcal{O}(1)

member(x, S)?

\mathcal{O}(1) + 1 \times \text{naive} = \mathcal{O}(1)

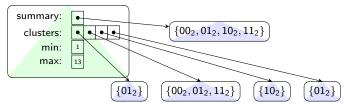
predecessor(x, S), successor(x, S)?
```

```
 \begin{array}{c} \textbf{function} \ \ {\tt MEMBER}_w(x,S) \\ \textbf{return} \ \ {\tt MEMBER}_{\lceil w/2 \rceil} \big( {\sf lo}_w(x), S. {\sf clusters}[{\sf hi}_w(x)] \big) \end{array}
```

delete(x, S)?

insert(x, S)?

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\mathcal{O}(1)

member(x, S)?

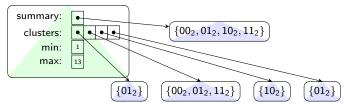
\mathcal{O}(1) + 1 \times \text{naive} = \mathcal{O}(1)

predecessor(x, S), successor(x, S)?
```

delete(x, S)

insert(x, S)?

We can draw the structure for the set $S=\{1,4,5,7,10,13\}=\{0001_2,0100_2,0101_2,0111_2,1010_2,1101_2\}\subseteq [2^4]$ as:



```
How fast (worst case) is: empty(S), \min(S), \max(S)? \mathcal{O}(1) member(x, S)? \mathcal{O}(1) + 1 \times \text{naive} = \mathcal{O}(1) predecessor(x, S), successor(x, S)? \mathcal{O}(1) + 1 \times \text{naive} = \Theta(2^{\lceil w/2 \rceil}) = \Theta(\sqrt{|U|})
```

```
function PREDECESSOR<sub>w</sub>(x, S)

▷ Assumes S \neq \emptyset and min(S) < x

if x > S.max then

return S.max

p \leftarrow \text{hi}_w(x), s \leftarrow \text{lo}_w(x), C \leftarrow S.clusters[p]

if not EMPTY(C) and C.min < s then

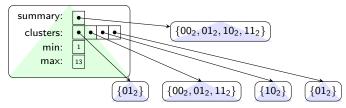
return index<sub>w</sub>(p, PREDECESSOR<sub>[w/2]</sub>(s, C))

p \leftarrow \text{PREDECESSOR}_{\lfloor w/2 \rfloor}(p, S.summary)

return index<sub>w</sub>(p, p.clusters[p].max)
```

insert(x, S)?

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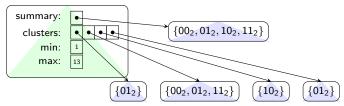


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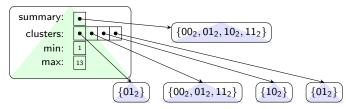
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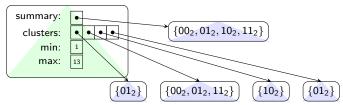
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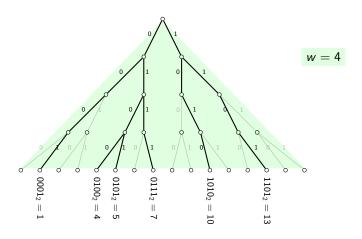
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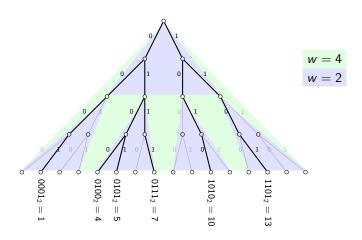


```
function INSERT<sub>w</sub>(x, S)
How fast (worst case) is:
                                                                    \triangleright Assumes x \notin S
empty(S), min(S), max(S)?
                                                                    if EMPTY(S) then
\mathcal{O}(1)
                                                                         S.\min \leftarrow x. S.\max \leftarrow x
member(x, S)?
                                                                    if x < S.min then S.min \leftarrow x
\mathcal{O}(1) + 1 \times \mathsf{naive} = \mathcal{O}(1)
                                                                    if x > S.max then S.max \leftarrow x
predecessor(x, S), successor(x, S)?
                                                                    p \leftarrow \text{hi}_w(x), s \leftarrow \text{lo}_w(x)
\mathcal{O}(1) + 1 \times \text{naive}
                                                                    if EMPTY(S.clusters[p]) then
=\Theta(2^{\lceil w/2 \rceil})=\Theta(\sqrt{|U|})
                                                                         INSERT_{|w/2|}(p, S.summary)
delete(x, S)?
                                                                    INSERT \lceil w/2 \rceil (s, S.clusters[p])
\mathcal{O}(1) + 2 \times \text{naive} = \Theta(\sqrt{|U|})
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\mathcal{O}(1) + 2 \times \text{naive} = \mathcal{O}(1)
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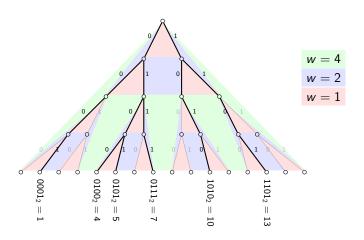
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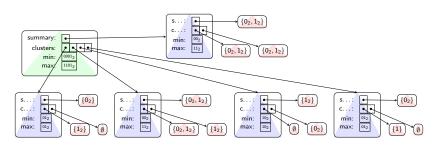


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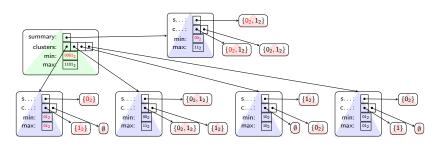
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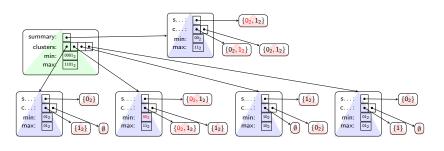
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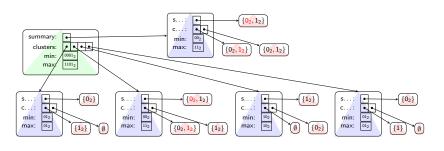
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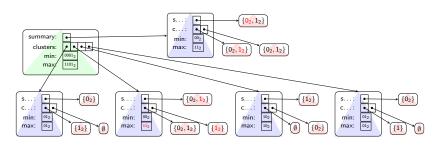
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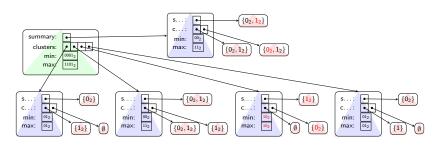
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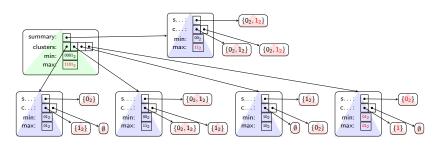
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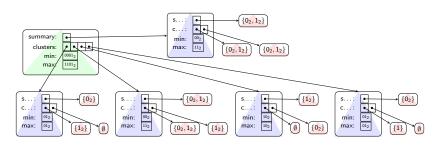
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Theorem

The recursion depth of this structure, when used on the universe $U = [2^w]$ is $\lceil \log_2 w \rceil = \mathcal{O}(\log \log |U|)$.

Proof.

Let d(w) be the recursion depth when working with a universe of size 2^w . We will prove by induction that $d(w) = \lceil \log_2 w \rceil$.

The base case is w=1 where there is no recursion, so

$$d(1) = 0 = \lceil \log_2(1) \rceil.$$

For the induction case, suppose w>1 and that $d(w')=|\log_2(w')|$ for all $w'\in [w]_+$. Now let $w'=\lceil w/2\rceil\in [w]_+$, then the largest universe size used in the recursion is $2^{w'}$, and

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insert(x, S) and delete(x, S)?

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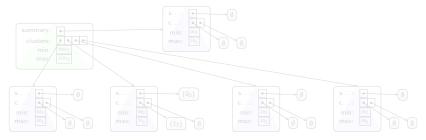
```
Using the recursive structure, how fast is: empty(S), \min(S), and \max(S)? worst case \mathcal{O}(1). member(x, S)? \mathcal{O}(1) + 1 \times \text{recursion} = \mathcal{O}(d(w)) = \mathcal{O}(\log \log |U|). predecessor(x, S) and \operatorname{successor}(x, S)? O(1) + 1 \times \operatorname{recursion} = \mathcal{O}(d(w)) = \mathcal{O}(\log \log |U|). insert(x, S) and \operatorname{delete}(x, S)? O(1) + 2 \times \operatorname{recursion} = \Theta(2^{d(w)}) = \Theta(w) = \Theta(\log |U|).
```

How can we improve the update time? Somehow make insert and delete recurse in only one substructure.

Idea: Exclude min(S) and/or max(S) from the set of keys stored in summary and clusters. (CLRS excludes only min(S), I exclude both).

Specifically, redefine summary and clusters to recursively store the sets:

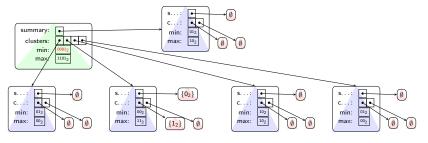
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\begin{split} \text{summary} &:= \{ \mathsf{hi}_w(x) \mid w \in S \setminus \{ \mathsf{min}, \mathsf{max} \} \} \\ \mathsf{clusters}[h] &:= \{ \ell \in [2^{\lceil w/2 \rceil}] \mid \mathsf{index}_w(h,\ell) \in S \setminus \{ \mathsf{min}, \mathsf{max} \} \} \quad \forall h \in [2^{\lfloor w/2 \rfloor}] \end{split}
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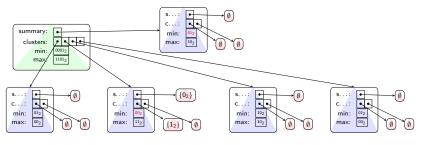
```
\begin{split} & \mathsf{summary} := \{\mathsf{hi}_w(x) \mid w \in \mathcal{S} \setminus \{\mathsf{min}, \mathsf{max}\} \} \\ & \mathsf{clusters}[h] := \{\ell \in [2^{\lceil w/2 \rceil}] \mid \mathsf{index}_w(h,\ell) \in \mathcal{S} \setminus \{\mathsf{min}, \mathsf{max}\} \} \quad \forall h \in [2^{\lfloor w/2 \rfloor}] \end{split}
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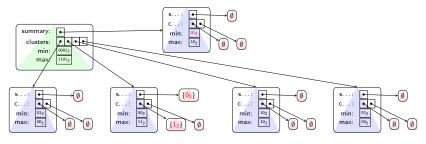
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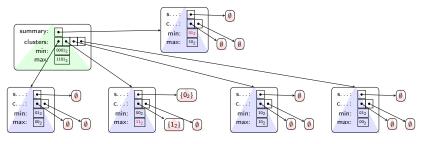
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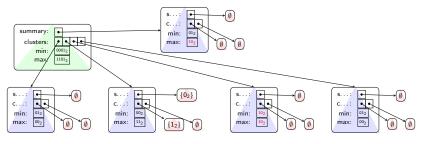


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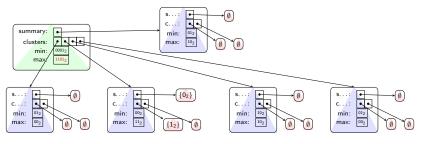
We can draw the van Emde Boas tree for the set $S=\{1,4,5,7,\textcolor{red}{10},13\}=\{0001_2,0100_2,0101_2,0111_2,\textcolor{red}{1010_2},1101_2\}\subseteq [2^4]$ as:



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```



```
\triangleright Assumes S \neq \emptyset and x > S.min
 1: function PREDECESSOR<sub>w</sub>(x, S)
         if x > S.max then
             return S.max
 3.
        if w = 1 then
 4.
             return S.min
 5.
         p \leftarrow \text{hi}_w(x), s \leftarrow \text{lo}_w(x), C \leftarrow S.\text{clusters}[p]
 6:
         if not EMPTY(C) and C.min < s then
 7:
             return index<sub>w</sub>(p, PREDECESSOR_{\lceil w/2 \rceil}(s, C))
 8:
         if EMPTY(S.summary) or p \leq S.summary.min then
9:
             return S.min
10:
         p \leftarrow \text{PREDECESSOR}_{|w/2|}(p, S.\text{summary})
11:
         return index<sub>w</sub>(p, S.clusters[p].max)
12:
```

Theorem

PREDECESSOR_w(x, S) takes worst case $\mathcal{O}(d(w)) = \mathcal{O}(\log \log |U|)$ time.

Proof

It makes at most one recursive call.

```
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                                                 \triangleright Assumes S \neq \emptyset and x > S.min
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Proof.

It makes at most one recursive call.

```
1: function INSERT<sub>w</sub>(x, S)
                                                                        \triangleright Assumes x \notin S
         if EMPTY(S) then
 2:
              S.min \leftarrow x, S.max \leftarrow x, return
 3:
         if S.min = S.max then
 4.
              if x < S.min then S.min \leftarrow x
 5:
              if x > S.max then S.max \leftarrow x
 6.
 7:
              return
         if x < S.min then S.min \leftrightarrow x
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 g.
         p \leftarrow \text{hi}_w(x), s \leftarrow \text{lo}_w(x)
10:
         if EMPTY(S.clusters[p]) then
11:
              INSERT_{|w/2|}(p, S.summary)
12:
         INSERT<sub>[w/2]</sub>(s, S.clusters[p])
13:
```

Theorem

INSERT_w(x, S) takes worst case $\mathcal{O}(d(w)) = \mathcal{O}(\log \log |U|)$ time.

Proof

It makes at most one recursive call on a non-empty substructure, and inserting in an empty substructure takes constant time.

 \triangleright Assumes $x \notin S$

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Idea: Use a hash table instead of an array to store clusters $[\cdot]$, and don't store empty substructures.

Why does this change updates from worst case $\mathcal{O}(\log \log |U|)$ to expected $\mathcal{O}(\log \log |U|)$ time?

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Why does this only use $\mathcal{O}(n \cdot d(w)) = \mathcal{O}(n \log \log |U|)$ space? Because the empty structure uses $\mathcal{O}(1)$ space and INSERT_w only creates or updates $\mathcal{O}(d(w))$ substructures in the worst case. Each of these costs at most an additional $\mathcal{O}(1)$ space

Idea: Partition S into "chunks" of size $\Theta(\min\{n,\log\log|U|\})$, and store only one element representing each chunk in the RS-vEB structure. I.e. RS-vEB stores only $\mathcal{O}(\max\{1,n/\log\log|U|\})$ elements, using $\mathcal{O}(n)$ space.

We can store each chunk as a sorted linked list, and keep a hash table mapping each representative to its chunk. This also takes $\mathcal{O}(n)$ space.

PREDECESSOR and SUCCESSOR uses the RS-vEB structure to find the nearest two representatives in $\mathcal{O}(\log \log |U|)$ time, and can then spend linear time in the size of the two chunks to find the result.

INSERT may have to split a chunk that becomes too large and insert a new representative in the RS-vEB structure. Splitting the chunk can take linear time in the size of the chunk, and together with inserting the new representative into the RS-vEB structure, this still only takes expected $\mathcal{O}(\log\log|\mathcal{U}|)$ time.

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Bonus: vEB is optimal for $w = \Theta(\log n)$

In fact, $\mathcal{O}(\log w)$ is the optimal query time when $w \in \Theta(\log n)$, or equivalently when $n \in 2^{\Theta(w)}$. This was shown in:

Time-space trade-offs for predecessor search,
Mihai Pătrașcu and Mikkel Thorup,
STOC'06: Proceedings of the thirty-eighth annual ACM symposium on
Theory of Computing, pages 232–240.
https://doi.org/10.1145/1132516.1132551

Idea: Insert all elements in R²S-vEB structure, use MIN(S) then repeatedly SUCCESSOR(x, S).

This takes expected $\mathcal{O}(n \log \log |U|) = \mathcal{O}(n \log w)$ time.

Compare this to other famous sorting methods:

QUICKSORT Takes expected $O(n \log n)$ time.

COUNTINGSORT Takes $\mathcal{O}(n+|U|)$ time.

RADIXSORT Takes
$$\mathcal{O}((n+r)\log_r|U|) = \mathcal{O}((n+r)\frac{w}{\log r})$$
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- ► A slightly less naive two-level implementation.
- A recursive implementation with good query time (proto-vEB).
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