

		update time	extreme pt query	CH area
<u>History</u>	C'99	$O(\log^{1+\epsilon} n)$	$O(\log n)$	Out
	Brodal-Jacob '00	$O(\log n \log \log n)$	$O(\log n)$	Still best
	Brodal-Jacob '02	$O(\log n)$	$O(\log n)$	

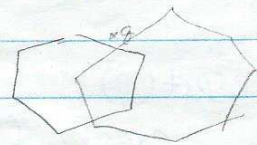
• 100 pages!

## General Dynamization Techniques

Def A query problem is decomposable if answer for input set  $S_1 \cup S_2$  can be computed from answer for  $S_1$  & answer for  $S_2$  in  $O(1)$  time

Ex range counting ✓  
 range emptiness ✓  
 nearest neighbor search ✓  
 extreme pt in CH ✓

membership in CH ✗  
 intersect CH with line ✗



## Technique 1: The "Logarithmic Method" (Bentley-Saxe '80)

If there is a static DS  $\mathcal{S}$  with

$P(n)$  preproc time,  $S(n)$  space,  $Q(n)$  query time

& problem is decomposable,

then there is a semidynamic DS  $\mathcal{S}'$  with

$O(S(n))$  space,  $O\left(\frac{P(n)}{n} \log n\right)$  insert time amortized

$O(Q(n) \cdot \log n)$  query time

i.e. total time for  $n$  inserts is  $O(P(n) \log n)$

Ex triangle range counting in 2D

$P(n) = O(n \log n)$ ,  $S(n) = O(n)$ ,  $Q(n) = O(n^{1/2+\epsilon})$   
 $\Rightarrow O(n)$  space,  $O(\log^2 n)$  insert (amort),  $(D(n) = O(\log n))$  by  
 $O(n^{1/2+\epsilon} \log n)$  query time  
 $(O(\log n)$  delete)

CH extreme pt query in 2D

( $D(n) = O(\log n)$ )  
by Chazelle

$P(n) = O(n \log n)$ ,  $S(n) = O(n)$ ,  $Q(n) = O(\log n)$   
 $\Rightarrow O(n)$  space, insert  $O(\log^2 n)$ , query  $O(\log^2 n)$   
worse than Preparata '80!  $O(\log n)$  (delete  $O(\log n)$ )

nearest neighbor search in 2D

$P(n) = O(n \log n)$ ,  $S(n) = O(n)$ ,  $Q(n) = O(\log n)$  ( $M(n) = O(n)$ )  
by Kirkpatrick

Voronoi diagram

$\Rightarrow O(n)$  space, insert  $O(\log^2 n)$ , query  $O(\log^2 n)$   
 $O(\log n)$

Lemma:  $\dots$

Proof: divide input set  $S$  into  $O(\log n)$  subsets

invariant: for each  $i = 0, \dots, \log n$ ,  
at most 1 subset at level  $i$  of size  $2^i$

insert( $p$ ):

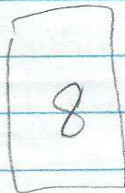
create  $\{p\}$  at level 0

whenever  $\exists$  2 subsets  $S_1, S_2$  at same level  $i$

destroy  $S_1, S_2$

build  $S_1 \cup S_2$  at level  $i+1$

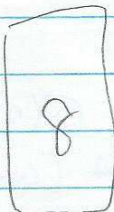
e.g.



$n = 11$

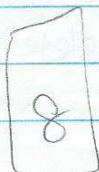
↓ insert

$\Rightarrow$

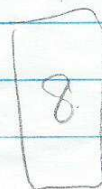


$n = 12$

$\Rightarrow$



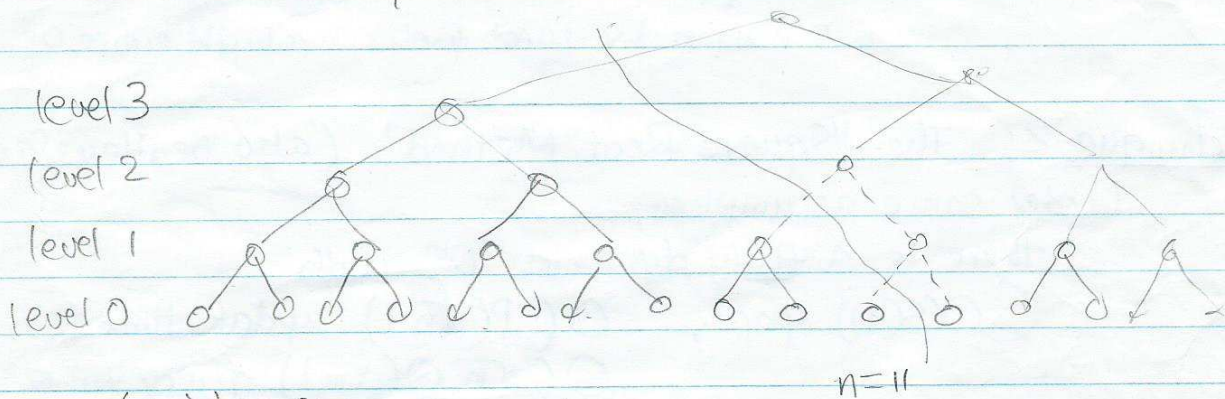
$\Rightarrow$



(Similar to binary counter)

(1011  $\rightarrow$  1100)

Viewed as a tree, over time:



(similar to merge sort)

Space  $\sum_{i=0}^{\log n} S(2^i) = O(S(n))$  [assume  $\frac{S(n)}{n}$  is nondecreas.]

query  $\leq \sum_{i=0}^{\log n} Q(2^i)$  by decomposability  
 $\leq O(Q(n) \log n)$

total time for  $n$  inserts

$\leq \sum_{i=0}^{\log n} P(2^i) \frac{n}{2^i}$  (# builds at level  $i$  is  $\frac{n}{2^i}$ )

$\leq \sum_{i=0}^{\log n} P(n) \frac{n}{n}$  [assume  $\frac{P(n)}{n}$  is nondecreas.]

$= O(P(n) \log n)$   $\square$

Remarks: (i) amortized  $\rightarrow$  worst case

(idea - spread rebuild over multiple updates messy!)

(ii) insert time improves to  $O(\frac{M(n)}{n} \log n)$

where  $M(n) =$  merging time

(iii) tradeoffs: base-b version ...

(iv) if  $\mathcal{S}$  supports delete in  $D(n)$  time,

$\mathcal{S}'$  also supports delete in  $O(D(n))$  time

(if  $n$  drops by const factor, rebuild entire DS...)

Technique 2: The "Square Root Method" (also Bentley-Saxe)

Under same assumption,

there is a fully dynamic DS with

$O(S(n))$  space,  $O(P(\sqrt{n}))$  update time,

$O(\sqrt{n} Q(\sqrt{n}))$  query time.

Proof: divide  $S$  into  $\sqrt{n}$  subsets of size  $\sqrt{n}$

insert( $p$ ):

let  $S_i =$  smallest subset

destroy  $S_i$

build  $S_i \cup \{p\}$

delete( $p$ ):

let  $S_i =$  subset containing  $p$

destroy  $S_i$

build  $S_i - \{p\}$   $\square$

Ex nearest neighbor in 2D

update  $O(\sqrt{n} \log n)$

query  $O(\sqrt{n} \log n)$

[C'19:  $O(\log^4 n)$  update]  
 $O(\log^2 n)$  query

via the logarithmic method  
+ shallow cuttings

Technique 3:

Under same assumption,

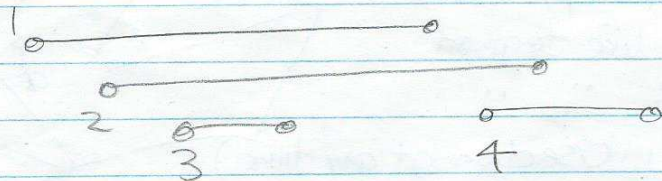
there is an offline dynamic DS  $\mathcal{S}'$  with

$\uparrow$   
entire update sequence is given in advance

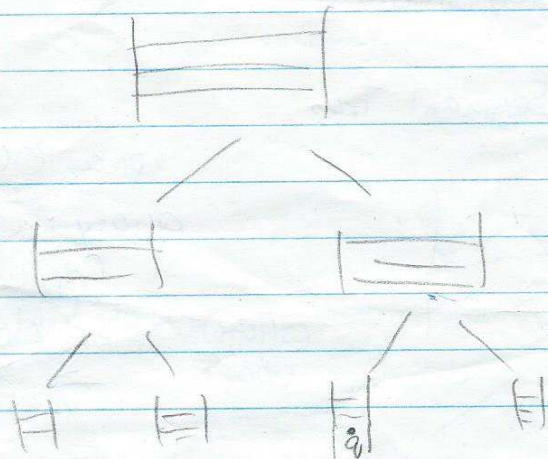
$O(S(n) \log n)$  space,  $O(\frac{P(n)}{n} \log n)$  amort. update time,

$O(Q(n) \log n)$  query time.

Proof: for each element, consider its lifespan / time interval



build segment tree for these  $n$  intervals!



for each node, store all its long segs in a DS  $\mathcal{S}$

Space  $O(S(n) \log n)$ , total update time  $O(P(n) \log n)$ ,  
 query  $O(Q(n) \log n)$   $\square$

Rmks - generalizes to semi-online

(insert is online; when element is inserted, told its delete time)

- don't need decomposability

if  $\mathcal{S}$  supports insert in worst-case  $I(n)$  time:  
 then  $\mathcal{S}'$  has  $O(I(n) \log n)$  amort. update time  
 $O(Q(n))$  query time

