

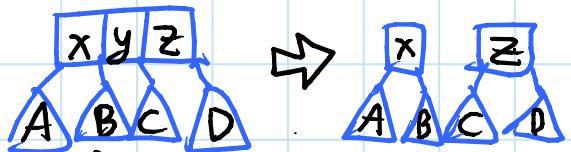
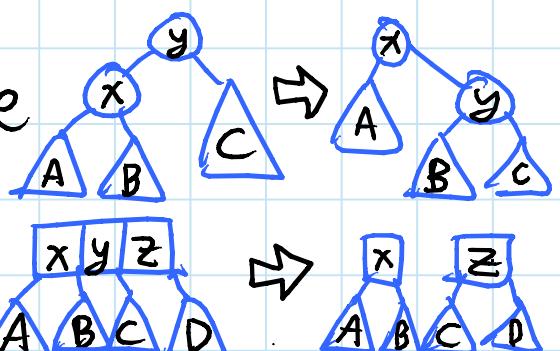
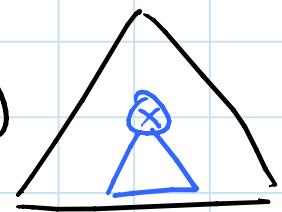
TODAY: Augmentation

- easy tree augmentation
- order-statistic trees
- finger search trees
- range trees

Idea: modify "off-the-shelf" data structure to store additional information  $\hookrightarrow$  updates

Easy tree augmentation:

- goal: store  $f(\text{subtree rooted at } x)$  at each node  $x$  in  $x.f$
- suppose  $x.f$  can be computed  $\rightarrow$  UPDATED in  $O(1)$  time from  $x$ , children, &  $\text{children}.f$
- if modify set  $S$  of nodes (data, children) then costs  $O(\# \text{ancestors of nodes in } S)$  to update  $f(x)$ 's (walk up from  $S$ )
- so  $O(\lg n)$  updates in
  - AVL trees: e.g. rotate  $\Rightarrow$  update  $y$  then  $x$
  - 2-3 trees: e.g. split  $\Rightarrow$  update  $x$  &  $z$  ( $\leftarrow$  also update up the tree)



# Order-statistic trees: (from 6.006)

- ADT/interface: (Abstract Data Type)
  - insert(x) / delete(x) / successor(x)
  - rank(x): find x's index in sorted order  
(= # elements < x if all distinct)
  - select(i): find element of rank i
- idea: use easy tree augmentation to store subtree size:  $f(\text{subtree}) = \# \text{ nodes in it}$   
 $\Rightarrow x.\text{size} = 1 + \sum(c.\text{size} \text{ for } c \text{ in } x.\text{children})$
- say, AVL trees  $\Rightarrow$  binary (2-3 trees also work)
- rank(x):
  - rank =  $x.\text{left.size} + 1^*$
  - walk up to root from x
    - when go left ( $x \rightarrow x'$ ):  
rank +=  $x'.\text{left.size} + 1$
- select(i):
  - $x = \text{root}$
  - - rank =  $x.\text{left.size} + 1^*$
    - if  $i = \text{rank}$ : return  $x$
    - if  $i < \text{rank}$ :  $x = x.\text{left}$
    - if  $i > \text{rank}$ :  $x = x.\text{right}$   
 $i -= \text{rank}$
  - repeat

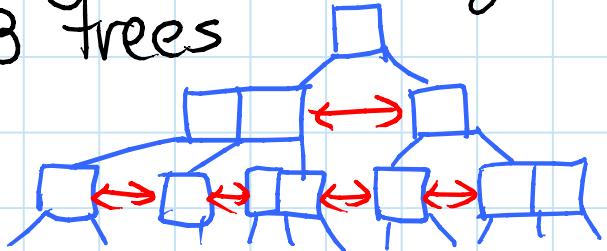


\* omit for indices starting at  $\emptyset$

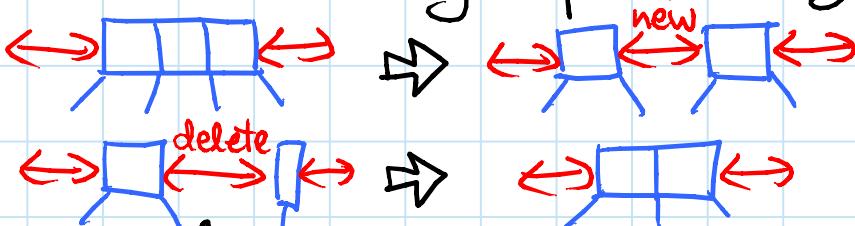
- e.g. can't maintain rank of each node:  
insert(-∞) would change all ranks

## Finger search trees: [Brown & Tarjan 1980]

- goal: if already found  $y$ , search( $x$  from  $y$ )  
should only take  $O(\lg \text{rank}(x) - \text{rank}(y))$
- idea: level-linked 2-3 trees
  - each node points to next & previous on same level



- maintain during split/merge:



- store all keys in the leaves:

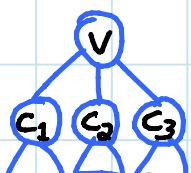


- nonleaf nodes don't store keys



- maintain min & max of each subtree  
(via easy tree augmentation)

⇒ can still do (top-down) search( $x$ ):



- say at vertex  $v$  with children  $c_1, c_2, c_3$
- look at min & max of each child  $c_i$
- if  $c_i.\text{min} \leq x \leq c_i.\text{max}$ : go down to  $c_i$
- if  $c_i.\text{max} < x < c_{i+1}.\text{min}$ :  
return  $c_i.\text{max}$  (predecessor)  
or  $c_{i+1}.\text{min}$  (successor)

- search( $x$  from  $y$ ):

-  $v = \text{leaf containing } y$  (given)

→ if  $v.\min \leq x \leq v.\max$ :

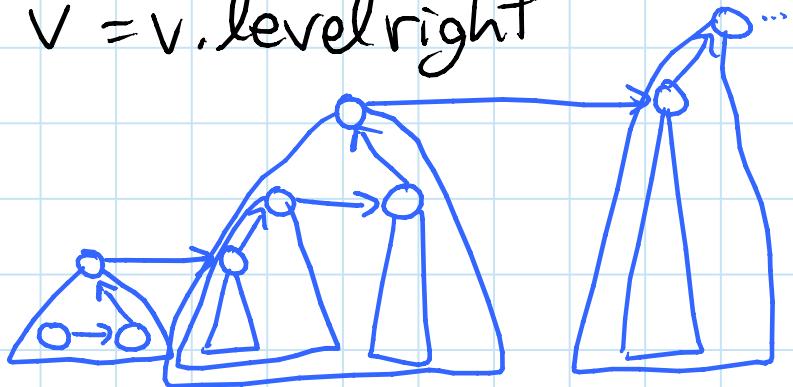
do top-down search for  $x$  from  $v$   
(i.e. within rooted subtree at  $v$ )

- if  $x < v.\min$ :  $v = v.\text{level left}$

- elif  $x > v.\max$ :  $v = v.\text{level right}$

-  $v = v.\text{parent}$

repeat



### Analysis:

- start at leaf level (height  $\emptyset$ )

- each round, go up 1 level

⇒ at step  $i$ , level link (height  $i$ ) skips  
 $\approx c^i$  keys/ranks, where  $c \in [2, 3]$

⇒ if  $|\text{rank}(x) - \text{rank}(y)| = k$

then reach  $x$  in  $\Theta(\lg k)$  steps

(and downward search also  $\Theta(\lg k)$ )

# Orthogonal range searching:

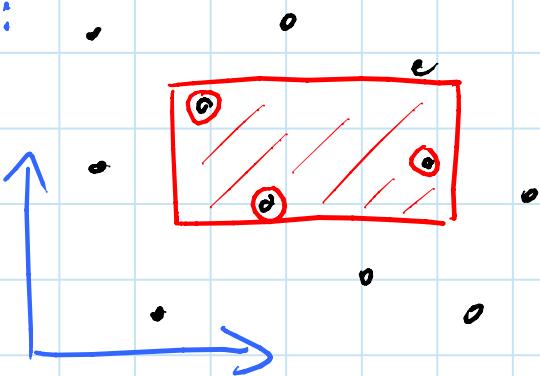
preprocess  $n$  points in  $d$  dimensions  
into a (static) data structure

Supporting range query:

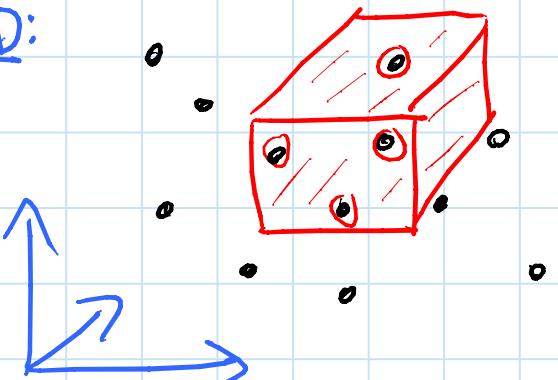
find  $k$  points in given axis-aligned box  
(rectangle in 2D)

OR count # points

2D:



3D:



1D:



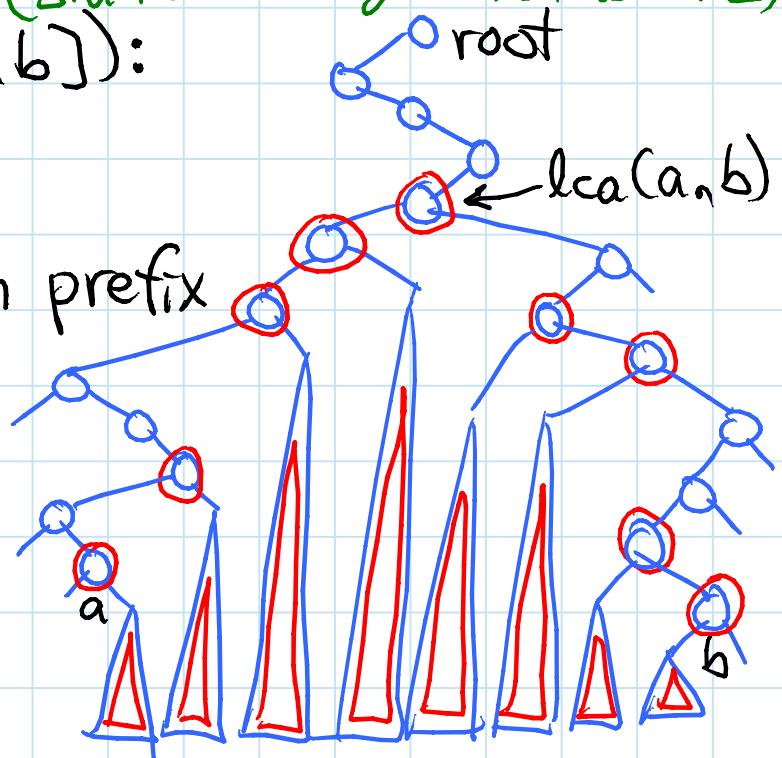
query = interval

- sorted array: binary search, walk right  
 $\Rightarrow O(\lg n + k)$  to report  $k$   
 (count in  $O(\lg n)$  via 2 binary searches  
 $+ \text{subtract}$ )

- finger search tree: (dynamic)  
 search, finger search right by 1, ...  
 $\Rightarrow O(\lg n + k)$  also  
 (counting harder...)

## 1D range tree:

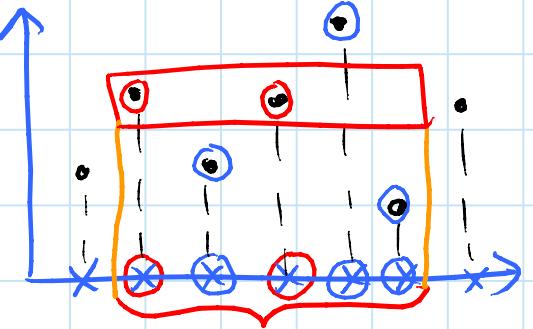
- complete BST (static ~ for dynamic, use AVL)
- range-query ([a, b]):
  - Search(a)
  - Search(b)
  - trim common prefix
  - return  $O(\lg n)$  nodes & subtrees "in between"



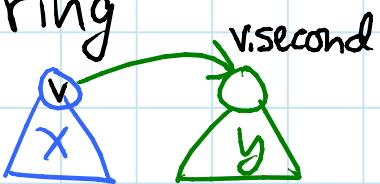
- $O(\lg n)$  to implicitly represent answer
- $O(\lg n + k)$  to traverse k outputs
- $O(\lg n)$  count via subtree size augmentation

## 2D range tree:

- primary 1D range tree keyed on x coordinate storing all points



- every node  $v$  in primary x-tree stores secondary 1D range tree, keyed on y coordinate, storing all points in  $v$ 's subtree



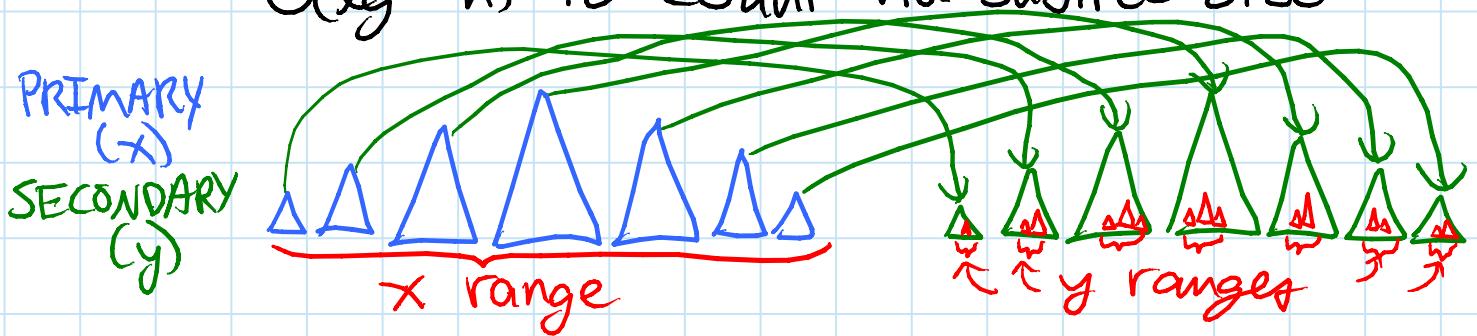
- range-search:

- use primary x-tree to find points in correct x range (implicitly)
  - $O(\lg n)$  points: check manually
  - $O(\lg n)$  subtrees: for each  $v$ , use  $v$ 's secondary y-tree to find points in correct y range (implicitly)

$\Rightarrow$  implicit representation as  $O(\lg^2 n)$  nodes & subtrees (of secondary trees)

$\Rightarrow O(\lg^2 n + k)$  to report  $k$  answers

-  $O(\lg^2 n)$  to count via subtree size



Space:  $O(n \lg n)$

- $O(n)$  for primary tree
- each point appears in  $O(\lg n)$  secondary trees (one per ancestor)

OR: each level of primary tree stores all points in secondary trees

d-D range trees:

- recurse from primary  $\rightarrow$  secondary  $\rightarrow \dots$
- query:  $O(\lg^d n + k)$
- space:  $O(n \lg^{d-1} n)$

Chazelle's improvement:

$$O(\lg^{d-1} n + k)$$

$$O(n \left(\frac{\lg n}{\lg \lg n}\right)^{d-1})$$

(see 6.851)

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6.046J / 18.410J Design and Analysis of Algorithms  
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