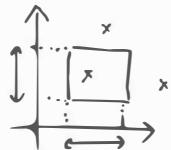


# Orthogonal Range Searching.

Prob Given  $n$  points  $p_1, \dots, p_n$ , and a rectangular range query, find all the points inside the given range.

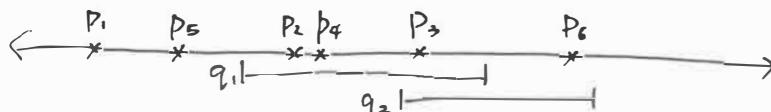


query that can be represented  
as the Cartesian product of ranges.

Naively, we can iterate through all points...  $O(n)$  alg.

$\Rightarrow$  Consider the setting where the points are fixed & range queries change. We want a sublinear query time.

## II. The 1-dimensional case.



~~Consider the query~~ Consider the query  $q = [x_1, x_2]$ .

Approach #1. Arrays.

Alg Prep 1

Sort the points.

Alg Query 1

- Do binary search to find the smallest point  $p_L$  s.t.  $p_L \leq x_1$ .
- Scan the array while outputting all points  $\leq x_2$ .

... easy to see that approach #1 uses  $O(n \log n)$  prep +  $O(\log n + k)$  query time, where  $k = \#$  reported points.

$\gg$  Easy, but hard to generalize to higher dims.

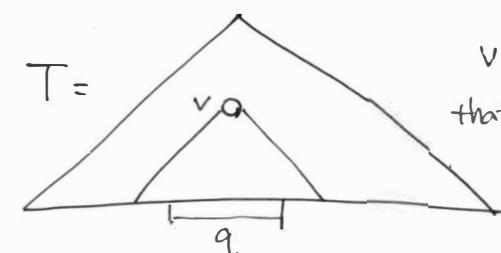
Approach #2. BST.

Alg Prep 2

$T \in \text{BST}$  with points sorted w/ coords. ~~leaf~~  
 $\Rightarrow$  leaf only at leaves.

Alg Query 2

$v \in \text{Find Split Node}(T, q)$   
report  $(v, q)$



$v$  is the lowest node  
that includes all nodes to report  
in the subtree rooted @ itself

Alg Find Split Node ( $T, q$ )

$v \leftarrow T.\text{root}$ .

while  $v$  is not a leaf and  $(x_1 \leq v.x \text{ or } x_2 > v.x)$

```

if  $x_2 \leq v.x$ 
|    $v \leftarrow v.\text{left}$ 
else  $v \leftarrow v.\text{right}$ 
return  $v$ 
```

Alg report( $v, q$ )

if  $v$  is a leaf, report  $v$  if  $v \in q$

else

$\text{cur} \leftarrow v.\text{left}$

  while  $\text{cur}$  is not a leaf

    if  $x_i \leq \text{cur}.x_l$

      report all leaves in the subtree rooted at  $\text{cur}.\text{right}$

$\text{cur} \leftarrow \text{cur}.\text{left}$

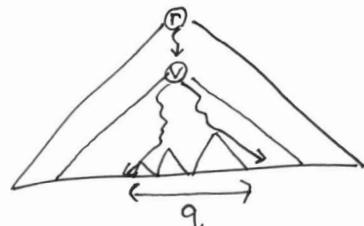
    else  $\text{cur} \leftarrow \text{cur}.\text{right}$

  if  $\text{cur}$  is a leaf, report  $\text{cur}$  if  $\text{cur} \in q$ .

  repeat once for the right symmetrically.

Analysis

Query 2 consists of two parts:



i) going down the BST

ii) reporting leaves in subtrees.

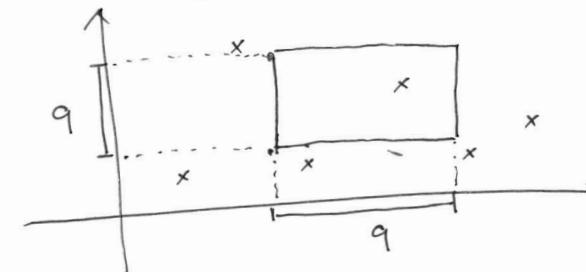
Assuming a balanced BST, i) takes  $O(\log n)$  time

Since # internal nodes < # leaves, ii) takes  $O(k)$  time

Overall, Query 2 runs in  $O(\log n + k)$  time.

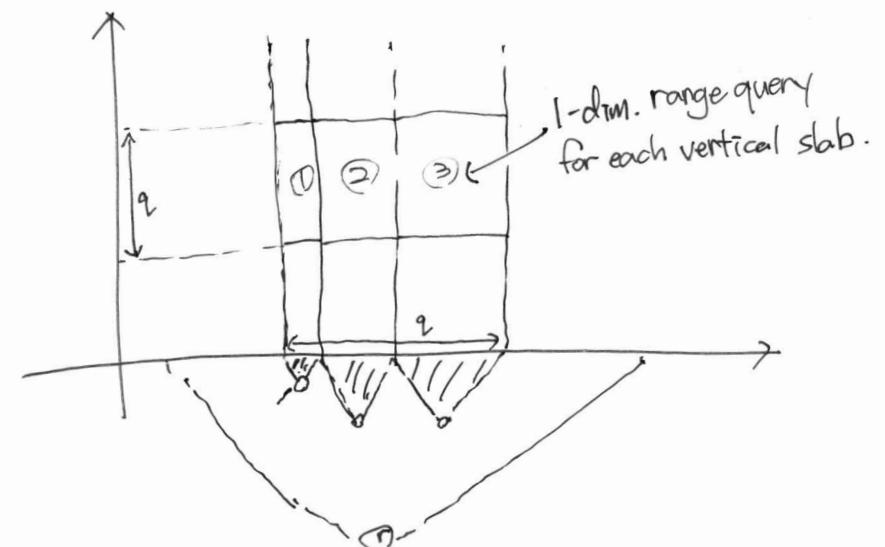
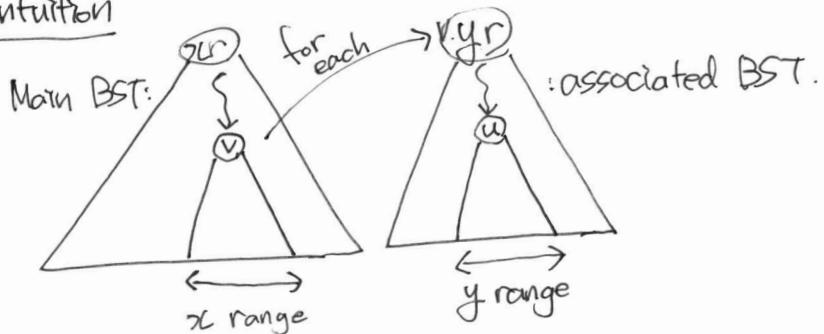
### III. The 2-dimensional case.

Approach #1. Range trees.



~~We extend the BST method to higher dims.~~

Intuition



## Alg Construct Range Tree (P)

$P_x \leftarrow$  Sort points in  $P$  w/  $x$ -coords

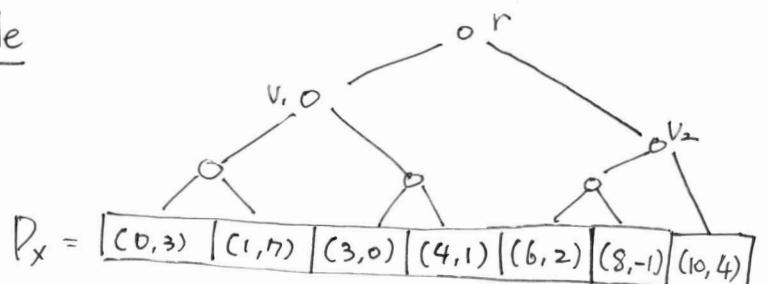
~~$P_{x,y} \leftarrow$  Sort points in  $P$  w/  $(x, y)$  coords~~

Structure the main BST using  $P_x$  so that it is balanced & full  
For each node  $v$  in the main BST in bottom-up order do:

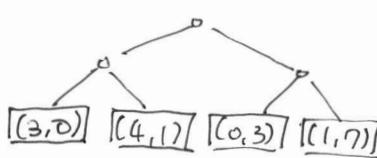
if  $v$  is a leaf, then  $v.\text{assoc}$  is a singleton BST  
w/ a single point.

o/w. merge  $v.\text{left.assoc}$  &  $v.\text{right.assoc}$  to construct  
 $v.\text{assoc}$ .

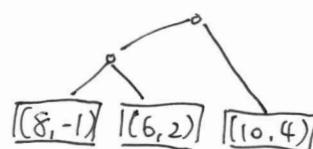
## Example



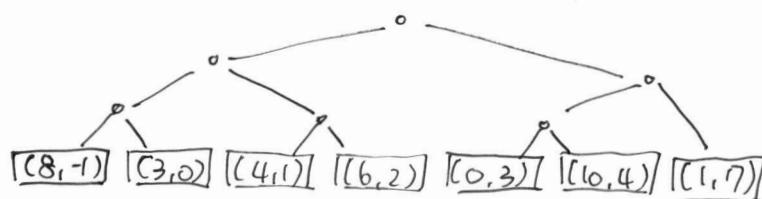
$v_1.\text{assoc}$ :



$v_2.\text{assoc}$ :



$r.\text{assoc}$ :



i.e., Merge like in mergesort.

Analysis The range tree of  $P$  can be constructed in  $O(n \log n)$  time  
and uses  $O(n \log n)$  space.

pf) 1. space. Each point  $p_x$  is stored only in the assoc.

$\text{BST}_1$  from a leaf to the root.  $\therefore$  a point appears exactly  
once in ~~the~~ the assoc. BST's of a given depth.

$D$   $\Rightarrow$  the overall storage of all the assoc. BSTs  
at level  $D$  is  $O(n)$ . Since the tree is balanced,  
the range tree uses  $O(n \log n)$  storage.

2. time. Structuring the main BST takes  $O(n)$  time.

Sorting  $P$  takes  $O(n \log n)$  time. Merging takes  
overall  $O(n \log n)$  time using a similar analysis w/ merge sort.

» We can answer axis-parallel rectangular range queries  
in  $O(\log^2 n + k)$  time if we perform 1-dim. range  
queries for each BST.

Remark. For points in  $d \geq 2$ -dimensional space,

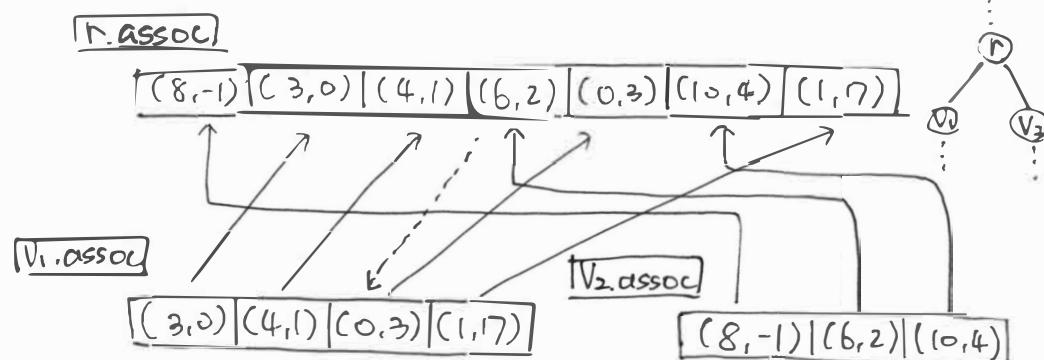
we can construct a range tree in  $O(n \log^{d-1} n)$  time & space  
which can answer axis-parallel hyperrectangular queries  
in  $O(\log^d n + k)$  time.

Approach #1-1. Range trees w/ fractional cascading.

~~Observation~~ We do not need to maintain a BST for the associated structure at the final level.

Question How can we avoid searching at every associated structure at the final level?

» Return to the mergesort intuition.



Originally, we would have to search in both  $V_1.assoc$  &  $V_2.assoc$ . resulting in a  $O(\log n + k)$  RT. However, note that we can maintain pointers so that we only perform binary search @ the root & report consecutive points when necessary.

Example consider range query  $[-1:7] \times [2:6]$ .

We first find (6,2) using binary search on  $r.assoc$ .

Since the x-range of  $V_1$  is in  $[-1:7]$ , we report on  $V_1$ .

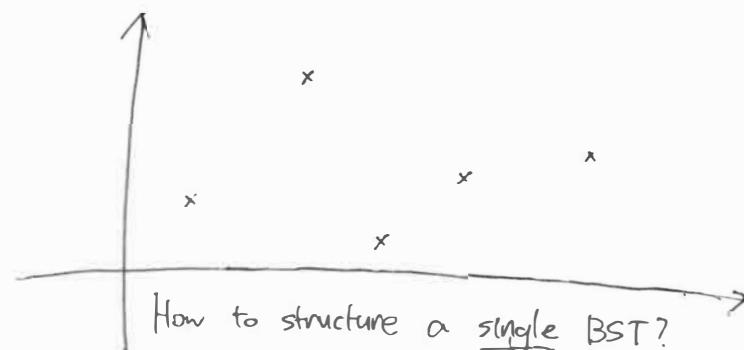
The least elt. in  $V_1.assoc$  no smaller than (6,2) is (0,3), which is in the rect. However the next elt. (1,7) is out-of-bounds,

Thm Let  $P$  be a set of  $n$  points in a  $d$ -dim. space, where  $d \geq 1$ . A (layered) range tree for  $P$  uses  $O(n \log^{d-1} n)$  storage and it can be constructed in  $O(n \log^{d-1} n)$  time. With the range tree, we can report the points in  $P$  that lie in a rectangular query range in  $O(\log^{d-1} n + k)$  time, where  $k = \#$  reported pts.

Approach #2. kd-trees.

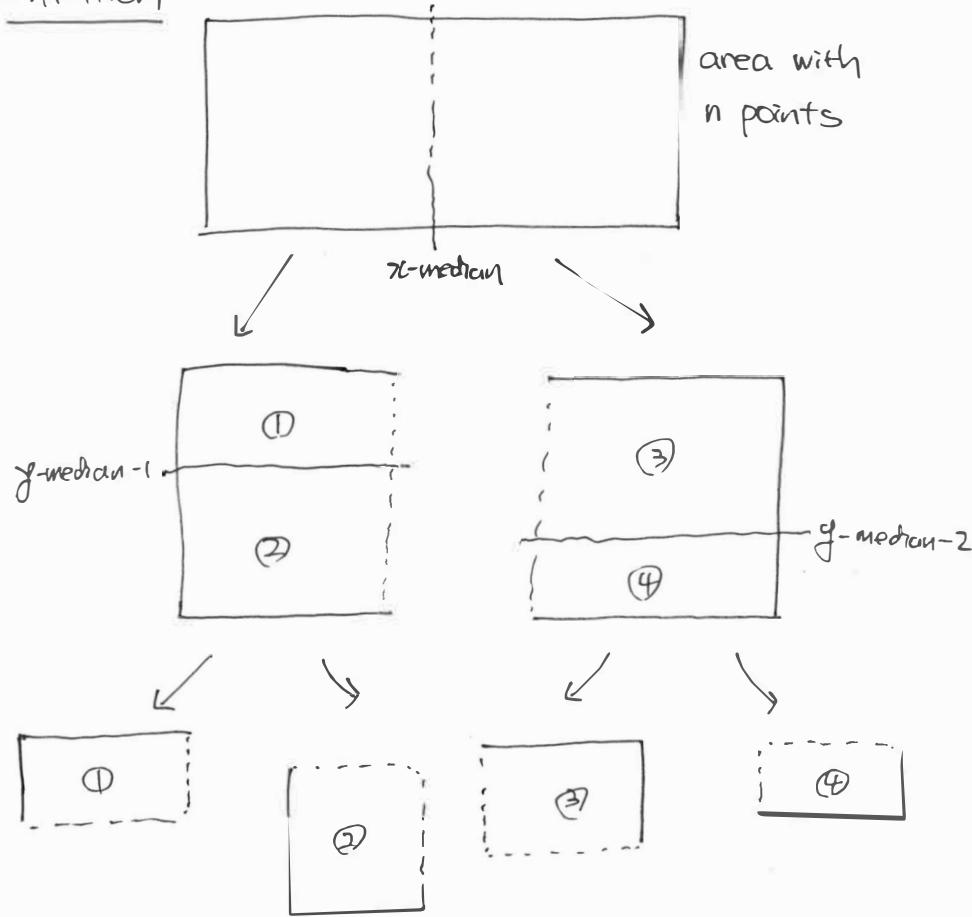
> We want a data structure w/ linear storage at the cost of querying time.

=> We can no longer "layer" structures for each dim.



Answer Alternate b/w splitting with X and splitting with Y.

## Intuition



We report points in regions that are fully contained by the query rectangle.

+ We can extend to  $d > 2$  dimensions by using ~~medians along~~ the median along the  $(D \bmod d)^{\text{th}}$  axis, where  $D$  is the depth in the kd-tree.