

# Computational Geometry

## Chapter 5

### Range Searching

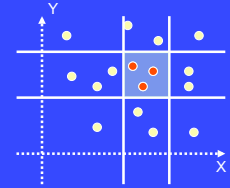
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# Orthogonal Range Searching

❑ **Problem:** Given a set of  $n$  points in  $\mathbb{R}^d$ , preprocess them such that reporting or counting the  $k$  points inside a  $d$ -dimensional axis-parallel box will be most efficient.

❑ Desired *output-sensitive* query time complexity –  $O(k+f(n))$  for reporting and  $O(f(n))$  for counting, where  $f(n)=o(n)$ , e.g.  $f(n)=\log n$ .

❑ **Sample application:** Report all cities within 100 mile radius of Boston.

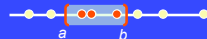


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# Range Searching – 1D

❑ In a one-dimensional world, points are real numbers and the query is two numbers  $(a, b)$ .

❑ Simple  $O(\log n)$  algorithm:  
 ■ Preprocessing: Sort points in  $O(n \log n)$ .  
 ■ Query: (Binary) search for  $a$  and  $b$  in list in  $O(\log n)$ .  
 List all values inbetween.



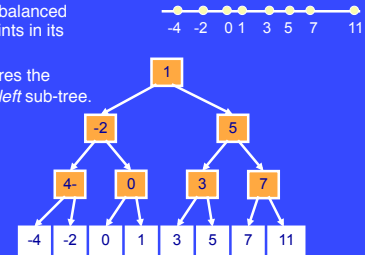
❑ Cannot be easily generalized to higher dimensions (**why not?**).

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# Range Searching – 1D Tree

❑ Range tree solution:

- Sort points.
- Construct a binary balanced tree, storing the points in its leaves.
- Each tree node stores the largest value of its *left* sub-tree.



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# Range Searching in 1D Tree

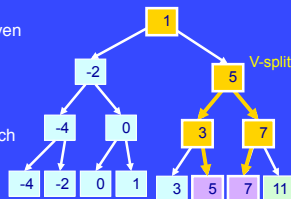
**Input Range: 3.5-8.2**

❑ Required time for finding a leaf:  $O(\log n)$ .

❑ Find the two boundaries of the given range in the leaves  $u$  and  $v$ .

❑ Report all the leaves in *maximal subtrees* between  $u$  and  $v$ .

❑ Mark the vertex at which the search paths diverge as V-split.



❑ Continue to find the two boundaries, reporting values in the subtrees:  
 When going left (right), report the entire right (left) subtree.

❑ When reaching a leaf, check it exhaustively.

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# General Idea

❑ Build a data structure storing a “small” number of canonical subsets, such that:

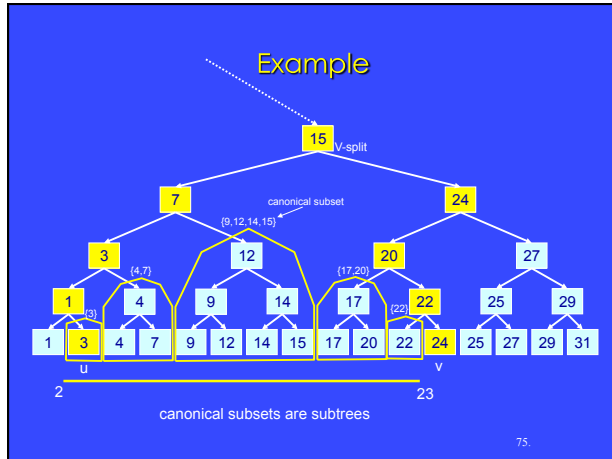
- The c.s. may overlap.
- Every query may be answered as the union of a “small” number of c.s.

❑ The geometry of the space enables this.

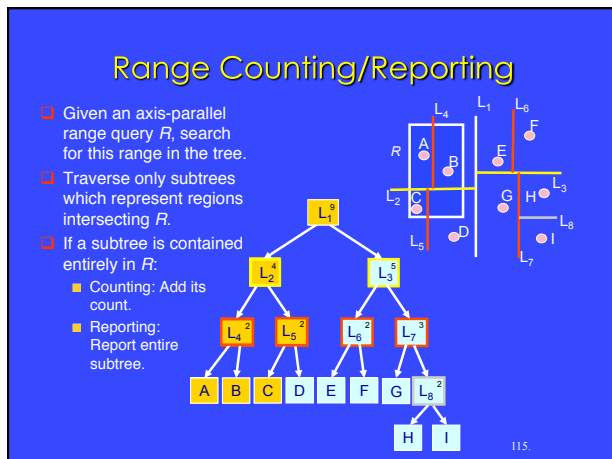
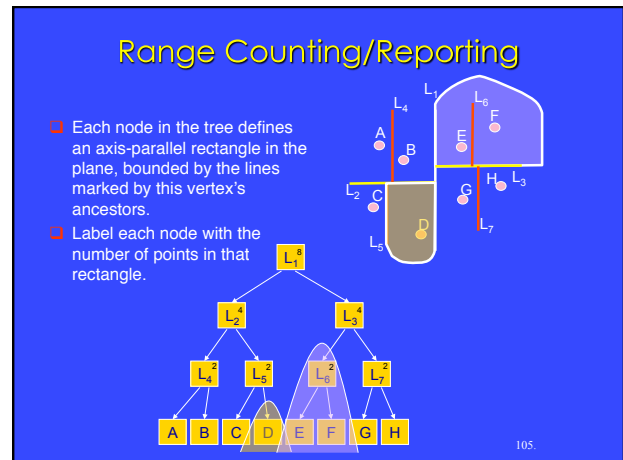
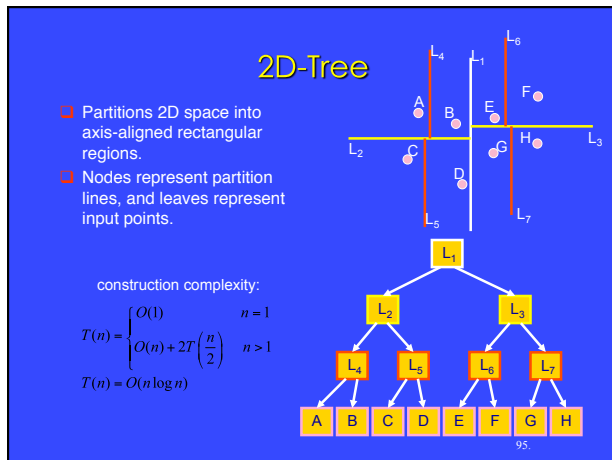
❑ Two extremes:

- Singletons –  $O(k)$  query time, even for counting.
- Power set –  $O(1)$  query time,  $O(2^n)$  storage.

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- ### 2D-Trees
- Given a set of points in 2D.
  - Bound the points by a rectangle.
  - Split the points into two (almost) equal size groups, using a horizontal or vertical line.
  - Continue recursively to partition the subsets, until they are small enough.
  - Canonical subsets are subtrees.
- 
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- ### Runtime Complexity
- $k$  nodes are reported. How much time is spent on internal nodes? The nodes visited are those that are **stabbed** by  $R$  but not contained in  $R$ . How many such nodes are there?
  - **Theorem:** Every side of  $R$  stabs  $O(\sqrt{n})$  cells of the tree.
  - **Proof:** Extend the side to a full line (WLOG - horizontal):
    - In the first level it stabs two children.
    - In the next level it stabs (only) two of the four grandchildren.
    - Thus, the recursive equation is:
- $$Q(n) = \begin{cases} 1 & n = 1 \\ 2 + 2Q\left(\frac{n}{4}\right) & \text{otherwise} \end{cases} = O(\sqrt{n})$$
- Total query time:  $O(\sqrt{n} + k)$ .
- 125.

## Kd-Trees – Higher Dimensions

- For a  $d$ -dimensional space:
  - Construction time:  $O(d \log n)$ .
  - Space Complexity:  $O(dn)$ .
  - Query time complexity:  $O(dn^{1-1/d}+k)$ .

**Question:** Are kd trees useful for non-orthogonal range queries, e.g. disks, convex polygons ?

Fact: Using *interval trees*, orthogonal range queries may be solved in  $O(\log^{d-1}n+k)$  time and  $O(n \log^{d-1}n)$  space.