#### **Computational Geometry**



#### **Orthogonal Range Searching** Michael Goodrich

with slides from Carola Wenk

## **Orthogonal range searching**

**Input:** *n* points in *d* dimensions

• E.g., representing a database of *n* records each with *d* numeric fields

Query: Axis-aligned *box* (in 2D, a rectangle)

- Report on the points inside the box:
  - Are there any points?
  - How many are there?
  - List the points.



## **Orthogonal range searching**

**Input:** *n* points in *d* dimensions

- Query: Axis-aligned *box* (in 2D, a rectangle)
  - Report on the points inside the box
- **Goal:** Preprocess points into a data structure to support fast queries
  - Primary goal: *Static data structure*
  - In 1D, we will also obtain a dynamic data structure supporting insert and delete



#### **1D range searching**

In 1D, the query is an interval:



First solution:

- Sort the points and store them in an array
  - Solve query by binary search on endpoints.
  - Obtain a static structure that can list k answers in a query in  $O(k + \log n)$  time.

**Goal:** Obtain a dynamic structure that can list k answers in a query in  $O(k + \log n)$  time.

#### **1D range searching**

In 1D, the query is an interval:

New solution that extends to higher dimensions:

- Balanced binary search tree
  - New organization principle: Store points in the *leaves* of the tree.
  - Internal nodes store copies of the leaves to satisfy binary search property:
    - Node *x* stores in *key*[*x*] the maximum key of any leaf in the left subtree of *x*.



key[x] is the maximum key of any leaf in the left subtree of x.

#### **Example of a 1D range tree Note:** # internal nodes $\boldsymbol{\chi}$ = #leaves -1 = n - 1So, O(n) complexity.

key[x] is the maximum key of any leaf in the left subtree of x.

35 41





#### **Pseudocode, part 1:** Find the split node

1D-RANGE-QUERY(T,  $[x_1, x_2]$ )  $w \leftarrow T.root$ while w is not a leaf and  $(x_2 \le w.key \text{ or } w.key \le x_1)$ **do if**  $x_2 \leq w.key$ then  $w \leftarrow w.left$ else  $w \leftarrow w.right$ // w is now the split node [traverse left and right from w and report relevant subtrees]



# **Pseudocode, part 2: Traverse left and right from split node**

```
1D-RANGE-QUERY(T, [x_1, x_2])
    [find the split node]
    // w is now the split node
    if w is a leaf
    then output the leaf w if x_1 \le w.key \le x_2
    else v \leftarrow w.left
                                                        // Left traversal
          while v is not a leaf
             do if x_1 \leq w.key
                 then output all leaves in the subtree rooted at v.right
                        v \leftarrow v.left
                 else v \leftarrow v.right
          output the leaf v if x_1 \le v.key \le x_2
           [symmetrically for right traversal]
```

## **Analysis of 1D-RANGE-QUERY**

Query time: Answer to range query represented by  $O(\log n)$  subtrees found in  $O(\log n)$  time. Thus:

- Can test for points in interval in O(log *n*) time.
- Can report all k points in interval in O(k + log n) time.
- Can count points in interval in O(log n) time
- Space: O(n)
  Preprocessing time: O(n log n)



#### **2D range trees**



#### **2D** range trees

Store a *primary* 1D range tree for all the points based on *x*-coordinate.

Thus in  $O(\log n)$  time we can find  $O(\log n)$  subtrees representing the points with proper *x*-coordinate. How to restrict to points with proper *y*-coordinate?



## **2D range trees**



**Idea:** In primary 1D range tree of *x*-coordinate, <u>every</u> node stores a *secondary* 1D range tree based on *y*-coordinate for all points in the subtree of the node. Recursively search within each.



## **2D** range tree example Secondary trees 7/2 (1/1)5/8 Primary tree

#### Analysis of 2D range trees

**Query time:** In  $O(\log^2 n) = O((\log n)^2)$  time, we can represent answer to range query by  $O(\log^2 n)$  subtrees. Total cost for reporting *k* points:  $O(k + (\log n)^2)$ .

**Space:** The secondary trees at each level of the primary tree together store a copy of the points. Also, each point is present in each secondary tree along the path from the leaf to the root. Either way, we obtain that the space is  $O(n \log n)$ .

**Preprocessing time:** O(n log n)

#### *d*-dimensional range trees Each node of the secondary *y*-structure stores a tertiary *z*-structure representing the points in the subtree rooted at the node, etc. Save one log factor using fractional cascading Query time: $O(k + \log^d n)$ to report k points. **Space:** $O(n \log^{d-1} n)$

**Preprocessing time:**  $O(n \log^{d-1} n)$ 

#### **Search in Subsets**

- **Given:** Two sorted arrays  $A_1$  and A, with  $A_1 \subseteq A$ A query interval [l,r]
- **Task:** Report all elements e in  $A_1$  and A with  $l \le e \le r$
- Idea: Add pointers from *A* to  $A_1$ :  $\rightarrow$  For each  $a \in A$  add a pointer to the smallest element  $b \in A_1$  with  $b \ge a$

**Query:** Find  $l \in A$ , follow pointer to  $A_1$ . Both in A and  $A_1$  sequentially output all elements in [l,r].



**Runtime:**  $O((\log n + k) + (1 + k)) = O(\log n + k)$ 

#### Search in Subsets (cont.)

**Given:** Three sorted arrays  $A_{1,}A_{2}$ , and A, with  $A_{1} \subseteq A$  and  $A_{2} \subseteq A$ 



**Runtime:**  $O((\log n + k) + (1+k) + (1+k)) = O(\log n + k))$ 

 $Y_1 \cup Y_2$ 

Range trees:

X

#### **Fractional Cascading:** Layered Range Tree

17

15

5 /7 8 12 15

(5,80) (8,37) (15,99)

17

52

52,23)

21 38 41 52

(33, 30)

(19) (7,10) (12,3) (17,52) (21,49) (41,95) (58,59) (93,30)

58

67

(67, 89)

67

93

58

Replace 2D range tree with a layered range tree, using sorted arrays and pointers instead of the secondary range trees.

Preprocessing:  $O(n \log n)$ Query:  $O(\log n + k)$ 

#### **Fractional Cascading:** Layered Range Tree

Replace 2D range tree with a layered range tree, using sorted arrays and pointers instead of the secondary range trees.

Preprocessing:  $O(n \log n)$ Query:  $O(\log n + k)$ 



#### **d**-dimensional range trees

Query time:  $O(k + \log^{d-1} n)$  to report k points, uses fractional cascading in the last dimension Space:  $O(n \log^{d-1} n)$ Preprocessing time:  $O(n \log^{d-1} n)$ 

Best data structure to date: Query time:  $O(k + \log^{d-1} n)$  to report k points. Space:  $O(n (\log n / \log \log n)^{d-1})$ Preprocessing time:  $O(n \log^{d-1} n)$