

# Advanced Data Structures

## Lecture 10: Orthogonal Range Searching

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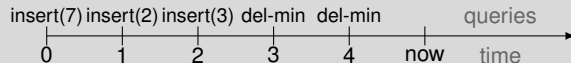
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# Recap: Retroactive Data Structures

## Operations

- $\text{INSERT}(t, \text{operation})$ : insert operation at time  $t$
- $\text{DELETE}(t)$ : delete operation at time  $t$
- $\text{QUERY}(t, \text{query})$ : ask  $\text{query}$  at time  $t$

- for a priority queue updates are
  - insert
  - delete-min
- time is integer ⓘ for simplicity otherwise use order-maintenance data structure



## Definition: Partial Retroactivity

QUERY is only allowed for  $t = \infty$  ⓘ now

## Definition: Full Retroactivity

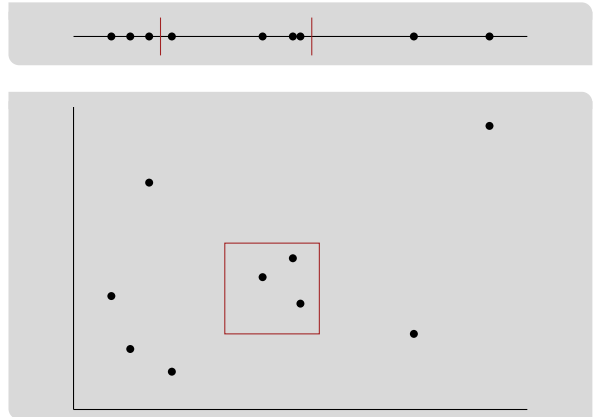
QUERY is allowed at any time  $t$

## Definition: Nonoblivious Retroactivity

INSERT, DELETE, and QUERY at any time  $t$  but also identify changed QUERY results

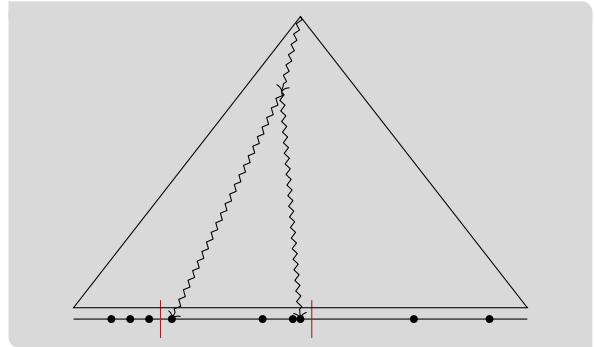
# Motivation: Query Set of Points

- given set of points  $P = \{p_1, \dots, p_n\}$  with  $p_i = (x_i, y_i)$
  - find all points in  $[x, y] \times [x', y']$
  - higher dimensions are possible
- 
- think about database queries
  - each dimension is a property
  - searching for objects fulfilling all properties of range




# 1-Dimensional Range Searching (1/2)

- consider 1-dimensional problem
  - range is  $[x..x']$
  - points  $P = \{x_1, \dots, x_n\}$  are just numbers
- build BBST where each leaf contains a point
  - inner node  $v$  store splitting value  $x_v$
- query for both  $x$  and  $x'$
  - find leaves  $b$  and  $e$  for  $x$  and  $x'$
  - let node  $v$  be node where paths to leaves split
  - report all leaves between  $b$  and  $e$



# 1-Dimensional Range Searching (2/2)

- how long does it take to report all children of a subtree with  $k$  leaves in a BBST?  **PINGO**

## Lemma: 1-Dimensional Range Searching

Let  $P$  be a set of  $n$  1-dimensional points.  $P$  can be stored in a BBST that requires  $O(n)$  words space, can be constructed in  $O(n \log n)$  time, and can answer range searching queries in  $O(\log n + occ)$  time

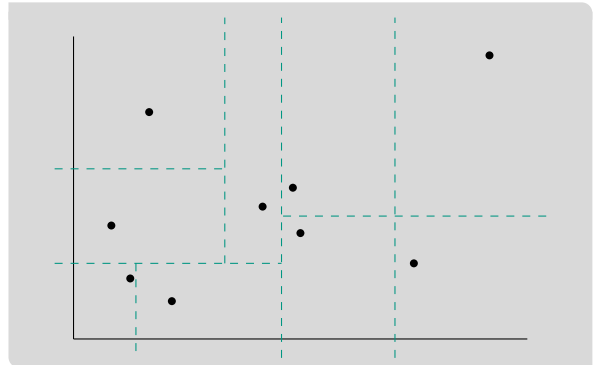
## Proof (Sketch Time)

- reporting all children in a subtree requires  $O(occ)$  time
- BBST has depth  $O(\log n)$
- search paths starting at  $v$  have length  $O(\log n)$
- report all subtrees to the right of the left path
- report all subtrees to the left of the right path

# 2-Dimensional Rectangular Range Searching

## Important

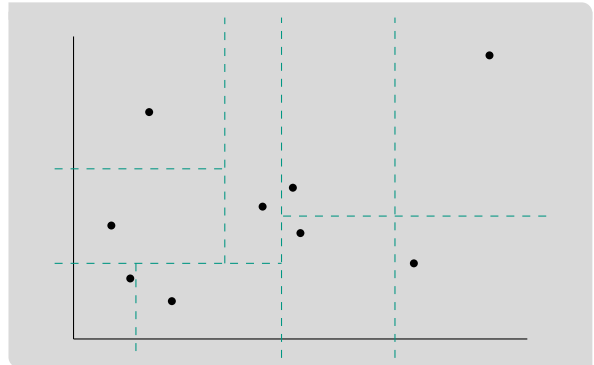
- for now: assume now two points have the same  $x$ - or  $y$ -coordinate
- generalize 1-dimensional idea
- 1-dimensional
  - split number of points in half at each node
  - points consist of one value
- 2-dimensional
  - points consist of two values
  - split number of points in half w.r.t. one value
  - switch between values depending on depth



# Kd-Trees (1/4)

- considering the 2-dimensional case
- each inner node at an even depth
  - splits the leaves in its subtree in half
  - using the  $x$ -coordinate
- each inner node at an odd depth
  - splits the leaves in its subtree in half
  - using the  $y$ -coordinate
- until each region contains a single point
- each leaf represents a point

- splitting in linear time is complicated
- better presort based on  $x$ - and  $y$ -coordinate
- inner nodes store splitter (line)





## Kd-Trees (2/4)

### Lemma: Kd-Tree Construction

A kd-tree for a set of  $n$  points requires  $O(n)$  words space and can be constructed in  $O(n \log n)$  time

### Proof (Sketch: Space)

- there are  $O(n)$  leaves
- there are  $O(n)$  inner nodes
- a binary tree requires  $O(1)$  words per node
- $O(n)$  words total space


### Proof (Sketch: Time)

- finding the splitter is easy due to presorted points
- splitting requires  $T(n)$  time with

$$T(n) = \begin{cases} O(1) & n = 1 \\ O(n) + 2T(\lceil n/2 \rceil) & n > 1 \end{cases}$$

- results in  $O(n \log n)$  running time
- dominates presorting

## Kd-Trees (3/4)

- use splitter depending on depth to identify paths through tree
  - if a region is fully contained in query: report region
  - if a region is intersected by query: check if point has to be reported
- 
- precomputation of requires note necessary
  - current region can be computed during query
  - using splitters
- 
- example on the board 

# Kd-Trees (4/4)


## Lemma: Kd-Tree Query

A query with an axis-parallel rectangle in a Kd-tree storing  $n$  points in the plane can be performed in  $O(\sqrt{n} + occ)$  time

## Proof (Sketch)

- $O(occ)$  time necessary to report points
- look at number of regions intersected by any vertical line
- upper bound for the regions intersected by query (for left and right edge of rectangle)
- upper bound for top and bottom edges are the same

## Proof (Sketch, cnt.)

- for vertical lines consider every inner node at odd depth
- starting at root's children
- can intersect two regions
- number of nodes is  $\lceil n/4 \rceil$   halved at each level
- number of intersected regions is  $Q(n)$  with

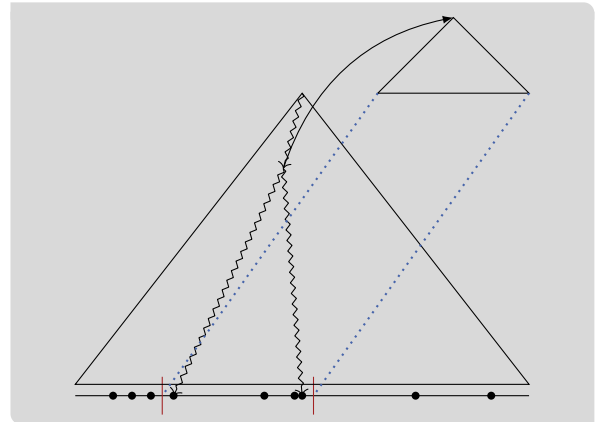
$$Q(n) = \begin{cases} O(1) & n = 1 \\ 2 + 2Q(\lceil n/4 \rceil) & n > 1 \end{cases}$$

- results in  $Q(n) = O(\sqrt{n})$
- $O(\sqrt{n} + k)$  total running time


# Range Trees (1/4)

- one BBST build on the  $x$ -coordinates
  - same as for 1-dimensional queries
- each inner node is associated with a set of points
- build a BBST for the  $y$ -coordinates of associated points for each inner node
  - store points in leaves not just  $y$ -coordinates
  - this BBST is used for reporting

- space-query-time trade-off
- faster queries but larger



## Range Trees (2/4)

- the BBST for the  $x$ -coordinates requires  $O(n)$  words of space
- how much space do the associated BBSTs require in total?  **PINGO**

### Lemma: Space Range Tree

A range tree on a set of  $n$  points in the plane requires  $O(n \log n)$  words space

### Proof (Sketch)

- BBST for  $x$ -coordinates has depth  $O(\log n)$
- all points are represented on **each** depth **exactly** once

### Proof (Sketch, cnt.)

- all associated BBSTs on each depth contain every point exactly once
  - total size of all BBSTs on each depth is  $O(n)$
  - total space  $O(n \log n)$  words
- how much faster is the range tree?

## Range Trees (3/4)

- 2-dimensional rectangular range search reduced to two 1-dimensional range searches
- look in BBST for  $x$ -coordinates  $\textcircled{i}$  same as 1-dimensional case
- instead of reporting subtrees to the right/left of paths search associated BBSTs
- report results in leaves of associated BBSTs

### Lemma: Range Tree Query Time

A query with an axis-parallel rectangle in a range tree storing  $n$  points requires  $O(\log^2 n + occ)$  time

### Proof (Sketch)

- each search in an associated BBST  $t$  requires  $O(\log n + occ_t)$  time
- $O(\log n)$  associated BSSTs  $T$  are searched  $\textcircled{i}$  as seen in 1-dimensional case
- total query time  $\sum_{t \in T} O(\log n + occ_t)$
- $\sum_{t \in T} O(occ_t) = O(occ)$
- $\sum_{t \in T} O(\log n) = O(\log^2 n)$
- total time:  $O(\log^2 n + occ)$

# Range Trees (4/4)

- range trees can be generalized to higher dimensions
- for each dimension add an additional associated BBST
- reporting in final BBST
- $d$ -dimensional queries are  $d$  1-dimensional queries

## Lemma: Higher Dimensions Range Tree

A  $d$ -dimensional range tree (for  $d \geq 2$ ) storing  $n$  points in the plane requires  $O(n \log^{d-1} n)$  words space and can answer queries in  $O(\log^d n + occ)$  time

## Proof (Sketch Query Time)


- recursive query time  $Q_d(n)$  with  $Q_2(n) = O(\log^2 n)$
- $Q_d(n) = O(\log n) + O(\log n) \cdot Q_{d-1}(n)$
- solves to  $Q_d(n) = O(\log^d n)$
- $O(occ)$  time for reporting

## Proof (Sketch Construction Space)

- recursive space  $S_d(n)$  with  $S_2(n) = O(n \log n)$  words
- $T_d(n) = O(n \log n) + O(\log n) \cdot T_{d-1}(n)$
- solves to  $S_d(n) = O(n \log^{d-1} n)$

## Fractional Cascading (1/2)

- sorted sets  $S_1, \dots, S_m$
- $|S_1| = n$  and  $S_{i+1} \subseteq S_i$
- report elements in range  $[x..x']$  in  $S_1, \dots, S_m$

- how much time does a naive algorithm with binary search require?  **PINGO**
- $O(m \log n + occ)$  time
- $O(m + \log n + occ)$  time possible with fractional cascading

- in addition to  $S_i$  store pointers to  $S_{i+1}$
- for each element in  $S_i$  store pointer to successor in  $S_{i+1}$
- possible because  $S_{i+1} \subseteq S_i$



# Fractional Cascading (2/2)

## Lemma:

Given sets  $S_1, \dots, S_m$  with  $|S_1| = n$  and  $S_{i+1} \subseteq S_i$ , find a range in all  $S_i$ 's using fractional cascading requires  $O(m + \log n + occ)$  time

## Proof (Sketch)

- binary search on  $S_1$  requires  $O(\log n)$  time
- following pointer to  $S_2$  requires  $O(1)$  time
- scanning  $S_2$  requires  $O(occ)$  time
- following pointer to  $S_3$  requires  $O(1)$  time
- repeat  $m$  times
- total:  $O(m + \log n + occ)$  time

- how to apply to range trees?
- instead of associated BBSTs store leaf data in arrays for all nodes but root
- each node has associated data
- store **two** successor pointers to the associated data in the left and right child
- two pointers to cover all possible paths
- this is a **layered range tree**

# Query Layered Range Trees

- search in BBST for  $x$ -coordinates remains the same
- to search  $y$ -coordinates first search associated BBST of root
- same as initial binary search for fractional cascading
- continue to follow pointers in associated data and scan to report queries

## Lemma: Query time Layered Range Tree

A query with an axis-parallel rectangle in a layered range tree storing  $n$  points in the plane can be performed in  $O(\log n + occ)$  time

## Proof (Sketch)

- the initial search requires  $O(\log n)$  time
- the search in the associated BBST of the root requires  $O(\log n)$  time
- remaining searches in associated data  $a$  requires  $O(1 + occ_a)$  time
- each point is reported once
- total time:  $O(\log n + occ)$

# General Sets of Points

- all solutions requires unique  $x$  and  $y$ -coordinate combination
- big limitation for applications
- remember database motivation

- store  $(x|k)$  as coordinate with  $x$  being the  $x$ -coordinate and  $k$  a unique key
- same for  $y$ -coordinates
- compare points using  $(x|k) < (x'|k') \iff x < x' \text{ or } (x = x' \text{ and } k < k')$

- range queries  $[x..x'] \times [y..y']$  become

$$[(x|-\infty)..(x'|\infty)] \times (y|-\infty)..[(y'|\infty)]$$

# Conclusion and Outlook

## This Lecture

- orthogonal range searching

## Next Lecture

- geometric data structures
- Q&A
- results of evaluation

## Advanced Data Structures

