

# Computational Geometry · Lecture

## Range Searching

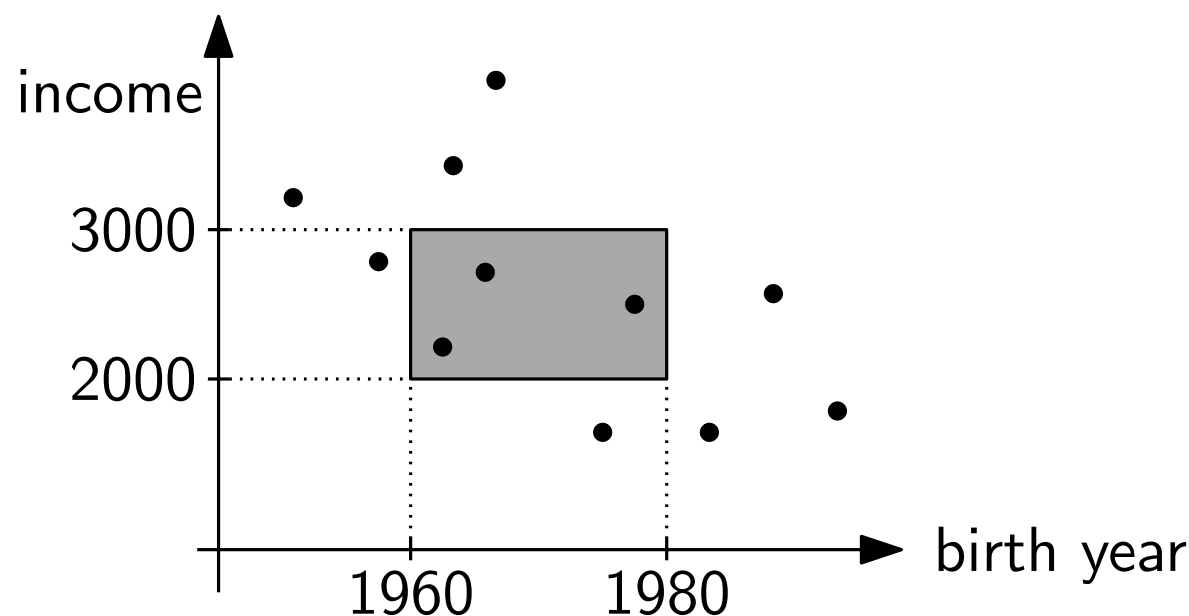
INSTITUT FÜR THEORETISCHE INFORMATIK · FAKULTÄT FÜR INFORMATIK

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# Geometry in Databases

In a personnel database, the employees of a company are anonymized and their monthly income and birth year are saved. We now want to perform a search: which employees have an income between 2,000 and 3,000 Euro and were born between 1960 and 1980?

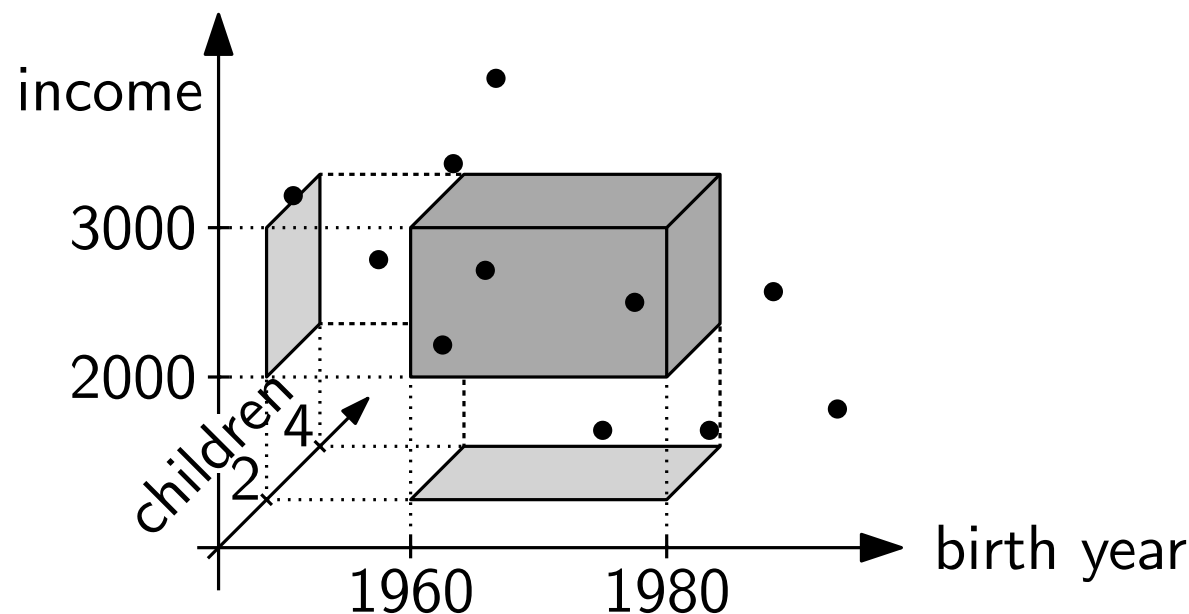


## Geometric Interpretation:

Entries are points: (birth year, income level) and the query is an axis-parallel rectangle

# Geometry in Databases

In a personnel database, the employees of a company are anonymized and their monthly income and birth year are saved. We now want to perform a search: which employees have an income between 2,000 and 3,000 Euro and were born between 1960 and 1980?



This problem can easily be generalized to  $d$  dimensions.

# Orthogonal Range Queries

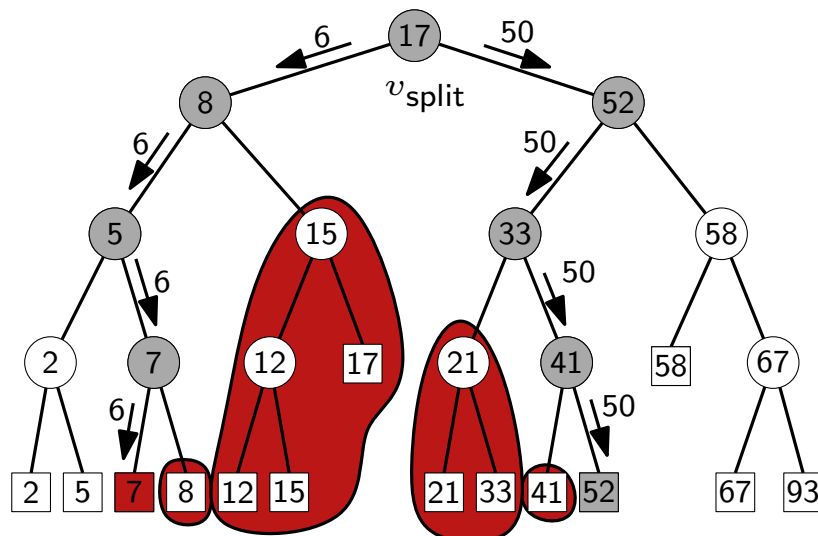
**Given:**  $n$  points in  $\mathbb{R}^d$

**Output:** A data structure that efficiently answers queries of the form  $[a_1, b_1] \times \cdots \times [a_d, b_d]$

**Problem:** Design a data structure for the case  $d = 1$ .

**Solution:** Balanced binary search tree:

- Stores points in the leaves
- Internal node  $v$  stores pivot value  $x_v$



**Example:**

Search for all points in  $[6, 50]$

**Answer:**

Points in the leaves between the search paths, (i.e.,  $\{7, 8, 12, 15, 17, 21, 33, 41\}$ )

# 1dRangeQuery

**FindSplitNode** $(T, x, x')$

$v \leftarrow \text{root}(T)$

**while**  $v$  not a leaf and  $(x' \leq x_v$  or  $x > x_v)$  **do**

**if**  $x' \leq x_v$  **then**  $v \leftarrow \text{lc}(v)$  **else**  $v \leftarrow \text{rc}(v)$

**return**  $v$

**1dRangeQuery** $(T, x, x')$

$v_{\text{split}} \leftarrow \text{FindSplitNode}(T, x, x')$

**if**  $v_{\text{split}}$  is leaf **then** report  $v_{\text{split}}$

**else**

$v \leftarrow \text{lc}(v_{\text{split}})$

**while**  $v$  not a leaf **do**

**if**  $x \leq x_v$  **then**

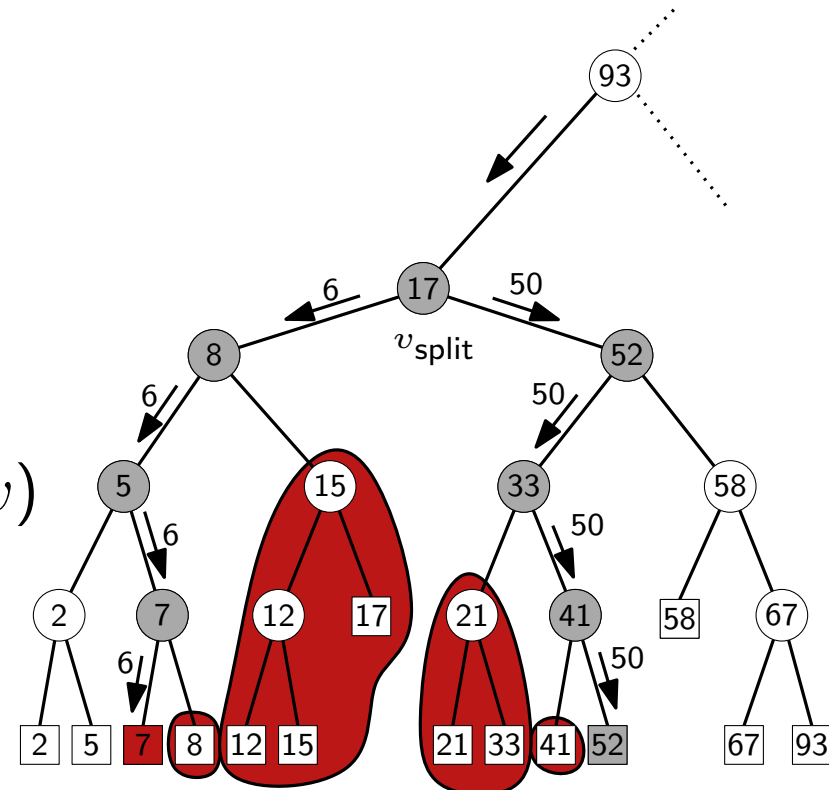
            | ReportSubtree(rc( $v$ ));  $v \leftarrow \text{lc}(v)$

**else**  $v \leftarrow \text{rc}(v)$

    report  $v$

    // analog. for  $x'$  and rc( $v_{\text{split}}$ )

Can find *canonical subset* in linear time



# Analysis of 1dRangeQuery

**1dRangeQuery**( $T, x, x'$ )

$v_{\text{split}} \leftarrow \text{FindSplitNode}(T, x, x')$

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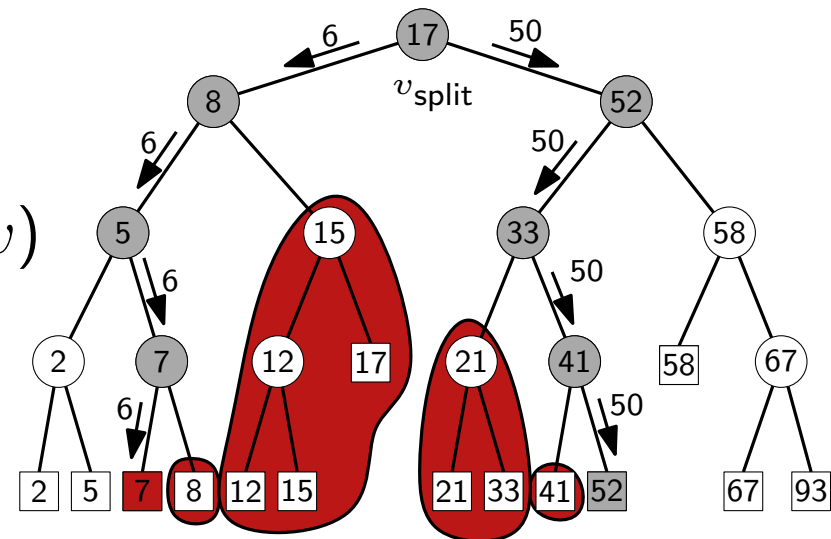
**if**  $x \leq x_v$  **then**

| ReportSubtree(rc( $v$ ));  $v \leftarrow \text{lc}(v)$

**else**  $v \leftarrow \text{rc}(v)$

report  $v$

// analog. for  $x'$  and rc( $v_{\text{split}}$ )



**Theorem 1:** A set of  $n$  points in  $\mathbb{R}$  can be preprocessed in  $O(n \log n)$  time and stored in  $O(n)$  space so that we can answer range queries in  $O(k + \log n)$  time, where  $k$  is the number of reported points.

# Orthogonal Range Queries for $d = 2$

**Given:** Set  $P$  of  $n$  points in  $\mathbb{R}^2$

**Goal:** A data structure to efficiently answer range queries of the form  $R = [x, x'] \times [y, y']$

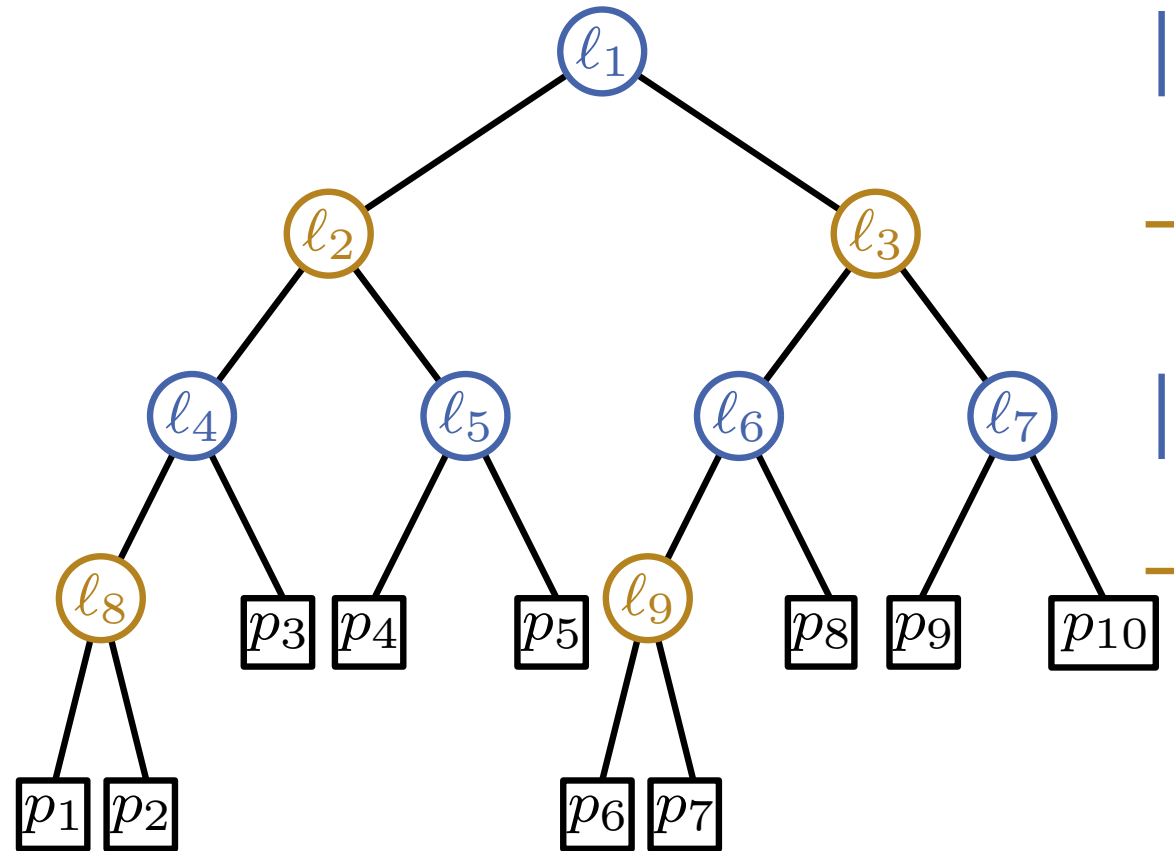
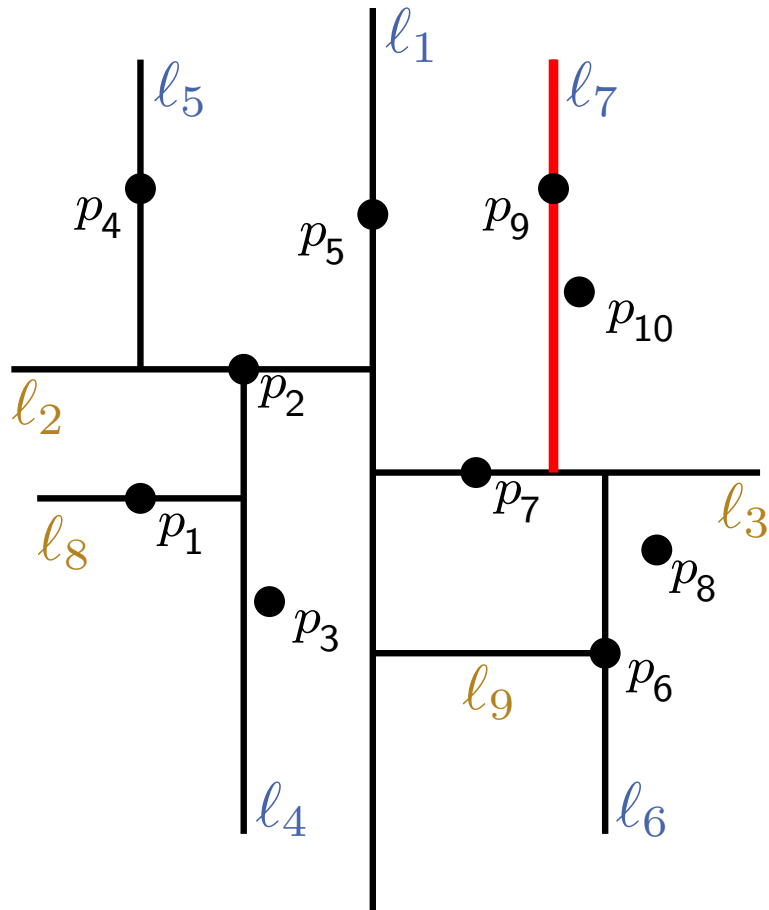
## Ideas for generalizing the 1d case?

### Solutions:

- *one* search tree, alternate search for  $x$  and  $y$  coordinates  
→ ***kd-Tree***
- *primary* search tree on  $x$ -coordinates,  
several *secondary* search trees on  $y$ -coordinates  
→ **Range Tree**

**Temporary assumption:** general position, that is no two points have the same  $x$ - or  $y$ -coordinates

# *kd*-Trees: Example





# $kd$ -Trees: Construction

BuildKdTree( $P$ ,  $depth$ )

**if**  $|P| = 1$  **then**

**return** leaf with a point in  $P$

**else**

**if**  $depth$  even **then**

        divide  $P$  vertically at

$\ell : x = x_{\text{median}(P)}$  in

$P_1$  (Points left of or on  $\ell$ ) and

$P_2 = P \setminus P_1$

**else**

        divide  $P$  horizontal at

$\ell : y = y_{\text{median}(P)}$  in

$P_1$  (points above or on  $\ell$ ) and

$P_2 = P \setminus P_1$

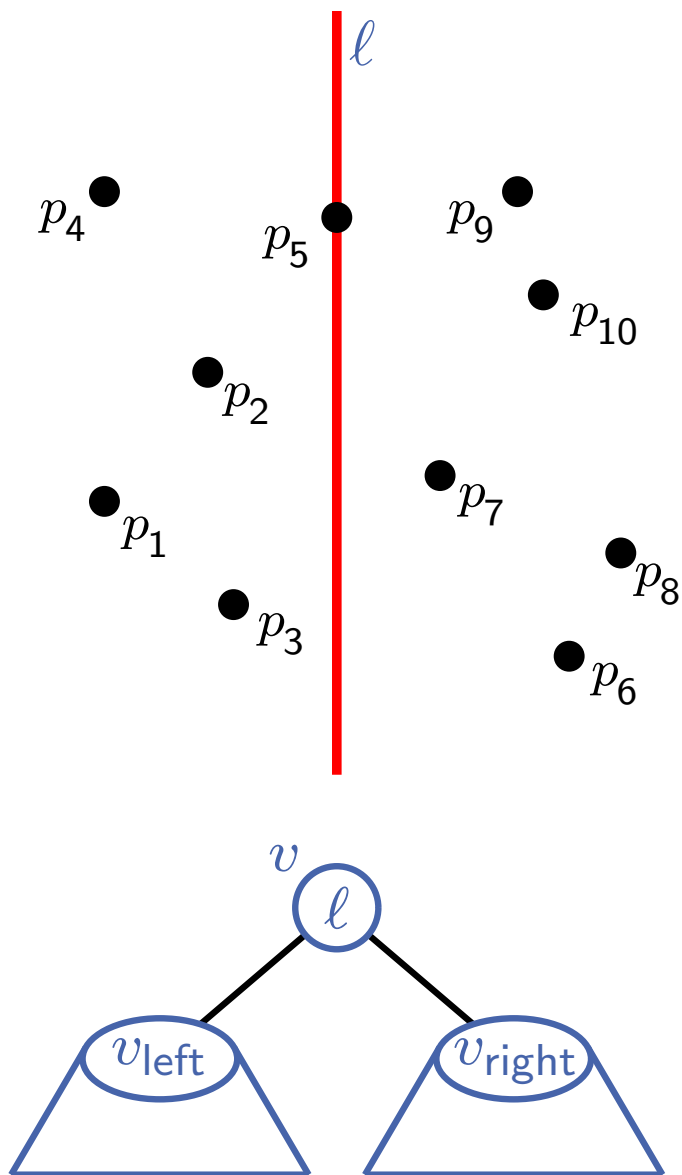
$v_{\text{left}} \leftarrow \text{BuildKdTree}(P_1, depth + 1)$

$v_{\text{right}} \leftarrow \text{BuildKdTree}(P_2, depth + 1)$

    Create node  $v$ , which stores  $\ell$

    make  $v_{\text{left}}$  and  $v_{\text{right}}$  children of  $v$

**return**  $v$



# Analysis of $kd$ -Tree Construction

**Lemma 1:** A  $kd$ -tree for  $n$  points in  $\mathbb{R}^2$  can be constructed in  $O(n \log n)$  time, using  $O(n)$  space.

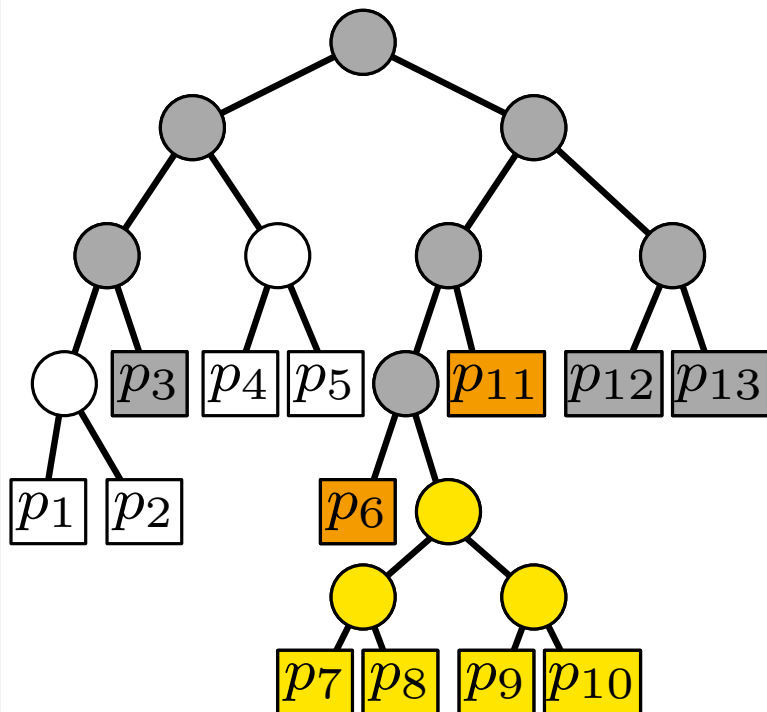
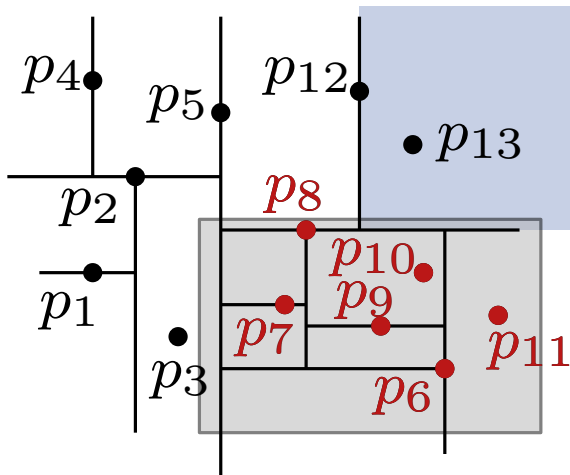
## Proof sketch:

- Determine median:
  - make two lists sorted on  $x$ - and  $y$ -coordinates
  - at each step, determine median and divide the lists
- We get the following recurrence:

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

- Solves to  $T(n) = O(n \log n)$  (analogous to MergeSort)
- Linear space, since we are using a binary tree with  $n$  leaves.

# Range Queries in a $kd$ -Tree



SearchKdTree( $v, R$ )

**if**  $v$  leaf **then**

    report point  $p$  in  $v$  when  $p \in R$

**else**

**if** region( $lc(v)$ )  $\subseteq R$  **then**

        ReportSubtree( $lc(v)$ )

**else**

**if** region( $lc(v)$ )  $\cap R \neq \emptyset$  **then**

            SearchKdTree( $lc(v), R$ )

**if** region( $rc(v)$ )  $\subseteq R$  **then**

        ReportSubtree( $rc(v)$ )

**else**

**if** region( $rc(v)$ )  $\cap R \neq \emptyset$  **then**

            SearchKdTree( $rc(v), R$ )

**Lemma 2:** A range query with an axis-aligned rectangle  $R$  in a  $kd$ -tree on  $n$  points may use  $O(\sqrt{n} + k)$  time, where  $k$  is the number of reported points.

## Proof sketch:

- Calls to ReportSubtree take  $O(k)$  time in total
- Still missing:  
Number of remaining nodes visited  
→ Exercise

# Orthogonal Range Queries for $d = 2$

**Given:** Set  $P$  of  $n$  points in  $\mathbb{R}^2$

**Goal:** A data structure to efficiently answer range queries of the form  $R = [x, x'] \times [y, y']$

## Ideas for generalizing the 1d case?

### Solutions:

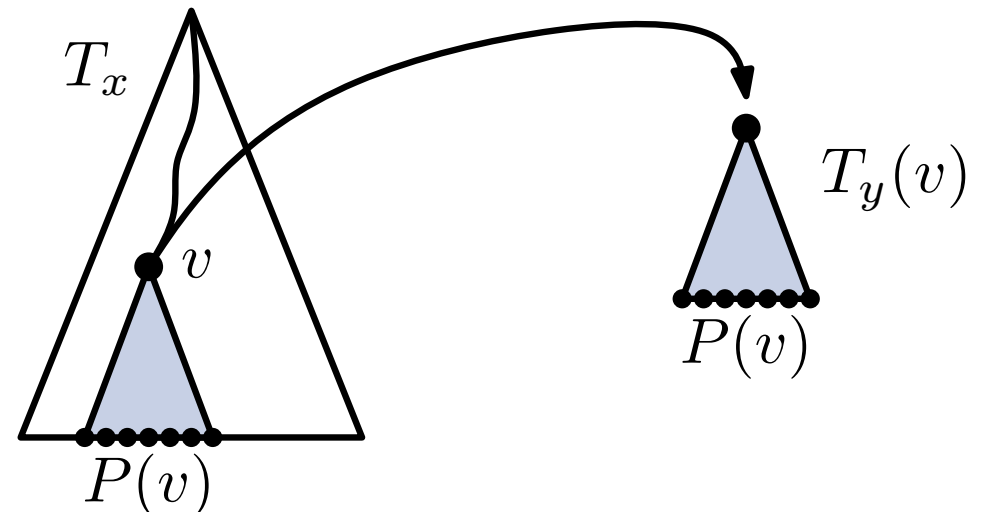
- *one* search tree, alternate search for  $x$  and  $y$  coordinates  
→ ***kd-Tree*** ✓
- *primary* search tree on  $x$ -coordinates,  
several *secondary* search trees on  $y$ -coordinates  
→ **Range Tree**

**Temporary assumption:** general position, that is no two points have the same  $x$ - or  $y$ -coordinates

# Range Trees

**Idea:** Use 1-dimensional search trees on two levels:

- a 1d search tree  $T_x$  on  $x$ -coordinates
- in each node  $v$  of  $T_x$  a 1d search tree  $T_y(v)$  stores the canonical subset  $P(v)$  on  $y$ -coordinates
- compute the points by  $x$ -query in  $T_x$  and subsequent  $y$ -queries in the auxiliary structures  $T_y$  for the subtrees in  $T_x$



# Range Trees: Construction

BuildRangeTree( $P$ )

**if**  $|P| = 1$  **then**

    Create leaf  $v$  for the point in  $P$

**else**

    Split  $P$  at  $x_{\text{median}}$  into  $P_1 = \{p \in P \mid p_x \leq x_{\text{median}}\}$ ,  $P_2 = P \setminus P_1$

$v_{\text{left}} \leftarrow \text{BuildRangeTree}(P_1)$

$v_{\text{right}} \leftarrow \text{BuildRangeTree}(P_2)$

    Create node  $v$  with pivot  $x_{\text{median}}$  and children  $v_{\text{left}}$  and  $v_{\text{right}}$

$T_y(v) \leftarrow$  binary search tree for  $P$  w.r.t  $y$ -coordinates

**return**  $v$

**Problem:** How much space and runtime does BuildRangeTree use?

**Lemma 3:** A Range Tree for  $n$  points in  $\mathbb{R}^2$  uses  $O(n \log n)$  space and can be constructed in  $O(n \log n)$  time.

# Range Queries in a Range Tree

Reminder:

~~1dRangeQuery( $T, x, x'$ )~~ **2dRangeQuery( $T, [x, x'] \times [y, y']$ )**

$v_{\text{split}} \leftarrow \text{FindSplitNode}(T, x, x')$

**if**  $v_{\text{split}}$  is leaf **then** report  $v_{\text{split}}$

**else**

$v \leftarrow \text{lc}(v_{\text{split}})$

**while**  $v$  not leaf **do**

**if**  $x \leq x_v$  **then**

~~ReportSubtree( $\text{rc}(v)$ )~~ **1dRangeQuery( $T_y(\text{rc}(v)), y, y'$ )**

$v \leftarrow \text{lc}(v)$

**else**  $v \leftarrow \text{rc}(v)$

report  $v$

// analogous for  $x'$  and  $\text{rc}(v_{\text{split}})$

**Lemma 4:** A range query in a Range Tree takes  $O(\log^2 n + k)$  time, where  $k$  is the number of reported points.

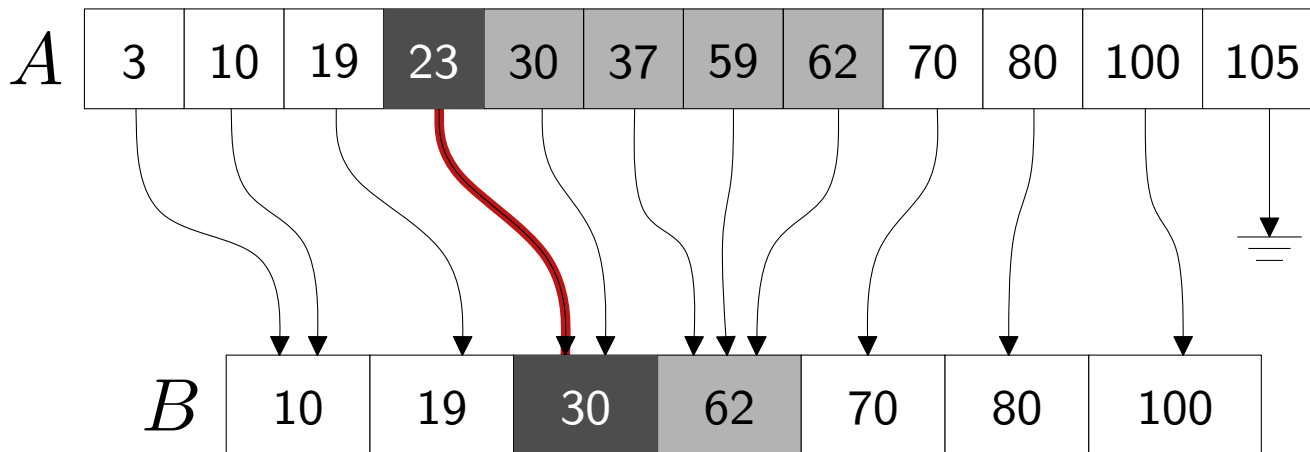


# Range Queries with Fractional Cascading

**Observation:** Range queries in a Range Tree perform  $O(\log n)$  1d queries, each taking  $O(\log n + k_v)$  time.  
The query interval  $[y, y']$  is always the same!

**Idea:** Use this property to accelerate the 1d queries to  $O(1 + k_v)$  time

**Example:** Two sets  $B \subseteq A \subseteq \mathbb{R}$  in sorted arrays

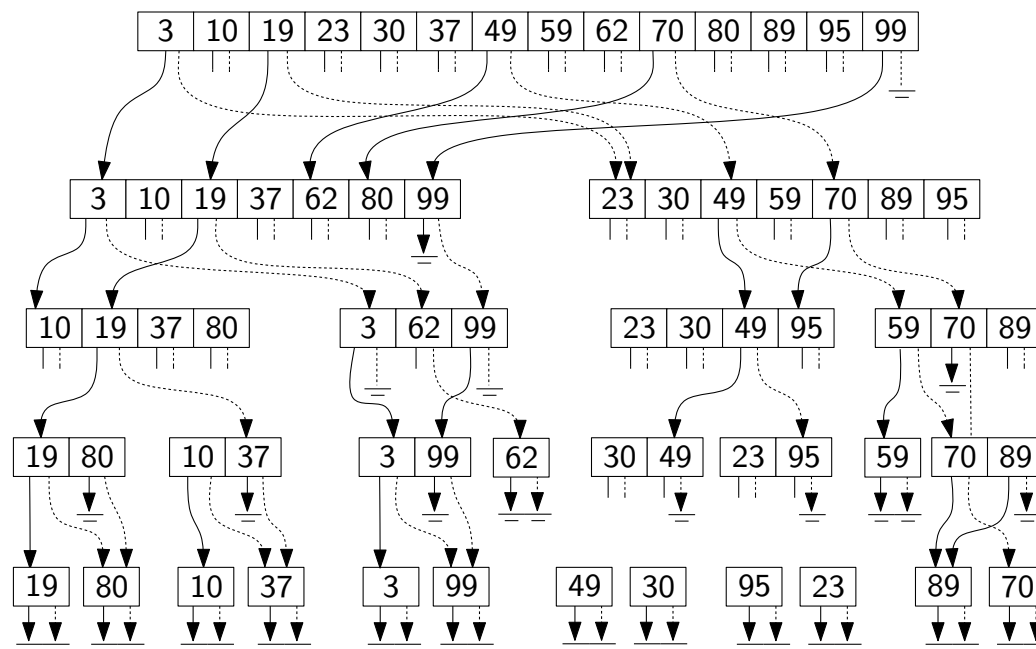
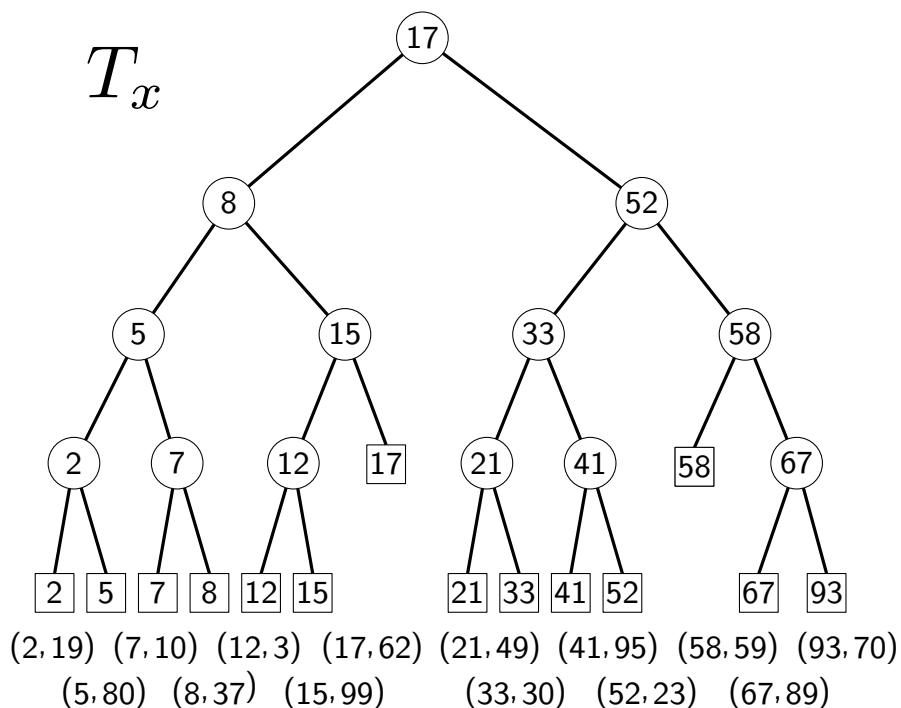


link  $a \in A$   
with smallest  
 $b \geq a$  in  $B$

Search interval  $[20, 65]$  **Pointer yields starting point for second search in  $O(1)$  time**

# Speed-up with Fractional Cascading

- In Range Trees we have  $P(\text{lc}(v)) \subseteq P(v)$  and  $P(\text{rc}(v)) \subseteq P(v)$  as the canonical sets.
- Define for each array element  $A(v)[i]$  two pointers into the arrays  $A(\text{lc}(v))$  and  $A(\text{rc}(v))$   
 → **Layered Range Tree**
- In the split node a binary search takes  $O(\log n)$  time, then it takes  $O(1)$  time to follow the pointers in the children



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→ **Layered Range Tree**
- In the split node a binary search takes  $O(\log n)$  time, then it takes  $O(1)$  time to follow the pointers in the children

**Theorem 2:** A Layered Range Tree on  $n$  points in  $\mathbb{R}^2$  can be constructed in  $O(n \log n)$  time and space. Range queries take  $O(\log n + k)$  time, where  $k$  is the number of reported points.

# Arbitrary Point Sets

**So far:** Points in general position, where no two points have the same  $x$ - or  $y$ -coordinate

**Idea:** Instead of  $\mathbb{R}$ , use pairs of numbers  $(a|b)$  with total order  $\leftrightarrow$  lexicographic order

$$p = (p_x, p_y) \longrightarrow \hat{p} = ((p_x|p_y), (p_y|p_x)) \longrightarrow$$

unique coord.

$$\text{Rectangle } R = [x, x'] \times [y, y']$$



$$\hat{R} = [(x| - \infty), (x'| + \infty)] \times [(y| - \infty), (y'| + \infty)]$$

**Then:**  $p \in R \Leftrightarrow \hat{p} \in \hat{R}$

# Summary

**Given:** Set  $P$  of  $n$  points in  $\mathbb{R}^2$

**Construct:** Data structures with efficient range queries of the form  $R = [x, x'] \times [y, y']$

→ We have seen two alternatives

	<i>kd</i> -Tree	Range Tree
Preprocessing	$O(n \log n)$	$O(n \log n)$
Space	$O(n)$	$O(n \log n)$
Query time	$O(\sqrt{n} + k)$	$O(\log^2 n + k)$

## How can the data structures generalize to $d$ -dimensions?

- $kd$ -Trees function analogously and by dividing the points alternately on  $d$  coordinates. Space is still  $O(n)$ , construction  $O(n \log n)$  and the query time is  $O(n^{1-1/d} + k)$ .
- Range Trees can be built recursively: the auxiliary search tree on the first coordinate is a  $(d - 1)$ -dimensional Range Tree. The construction and space takes  $O(n \log^{d-1} n)$  time; a query takes  $O(\log^d n + k)$  time, and with fractional cascading,  $O(\log^{d-1} n + k)$  time.

## Is it possible to query for other objects (e.g., polygons) with these data structures?

Yes, we can transform any polygon into a point in 4d space (exercise) or we can use windowing queries (comes in a later lecture).

**Question:** Can we adapt these data structures for dynamic point sets?

- Inserting points
- Removing points

1) **Divided kd-trees** [van Kreveld, Overmars '91]  
support updates in  $O(\log n)$  time, but the query time is  $O(\sqrt{n \log n} + k)$

2) **Augmented dynamic range trees** [Mehlhorn, Näher '90]  
support updates in  $O(\log n \log \log n)$  time and queries in  $O(\log n \log \log n + k)$  time