

ORTHOGONAL RANGE SEARCHING

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نجمه نوری

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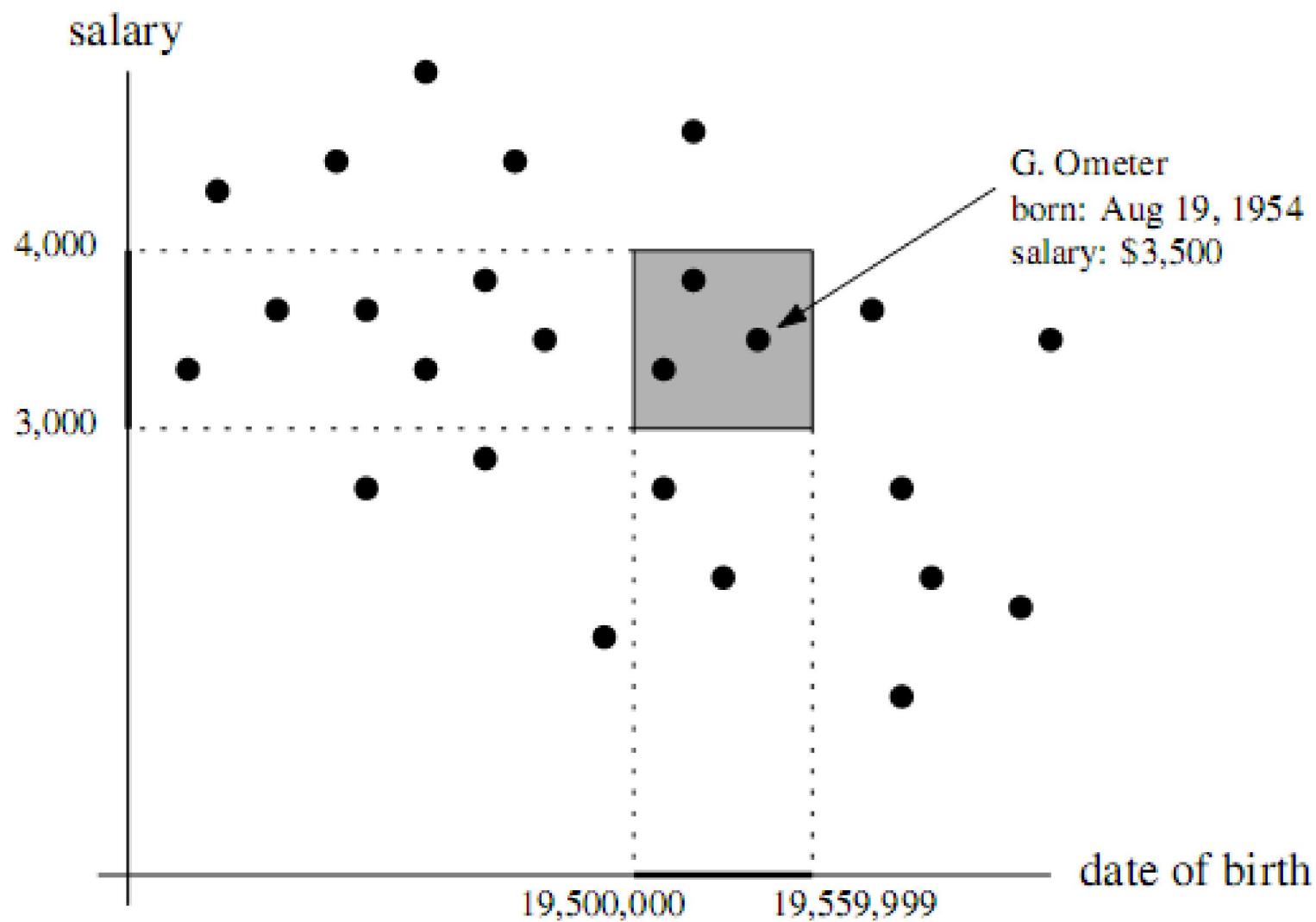
Motivation

- Many types of questions - **queries** - about data in a database can be interpreted geometrically
- To this end we transform records in a database into points in a multi-dimensional space, and we transform the queries about the records into queries on this set of points

Example

- Database for personnel administration
- In such database the name, address, date of birth, salary, and so on, of each employee are stored
- A typical query one may want to perform is to report all employees born between 1950 and 1955 who earn between \$3,000 and \$4,000 a month
- To formulate this as a geometric problem we represent each employee by a point in the plane
- The first coordinate of the point is the date of birth, represented by the integer
$$10000 \times year + 1000 \times month + day$$
- the second coordinate is the monthly salary
- With the point we store the other information

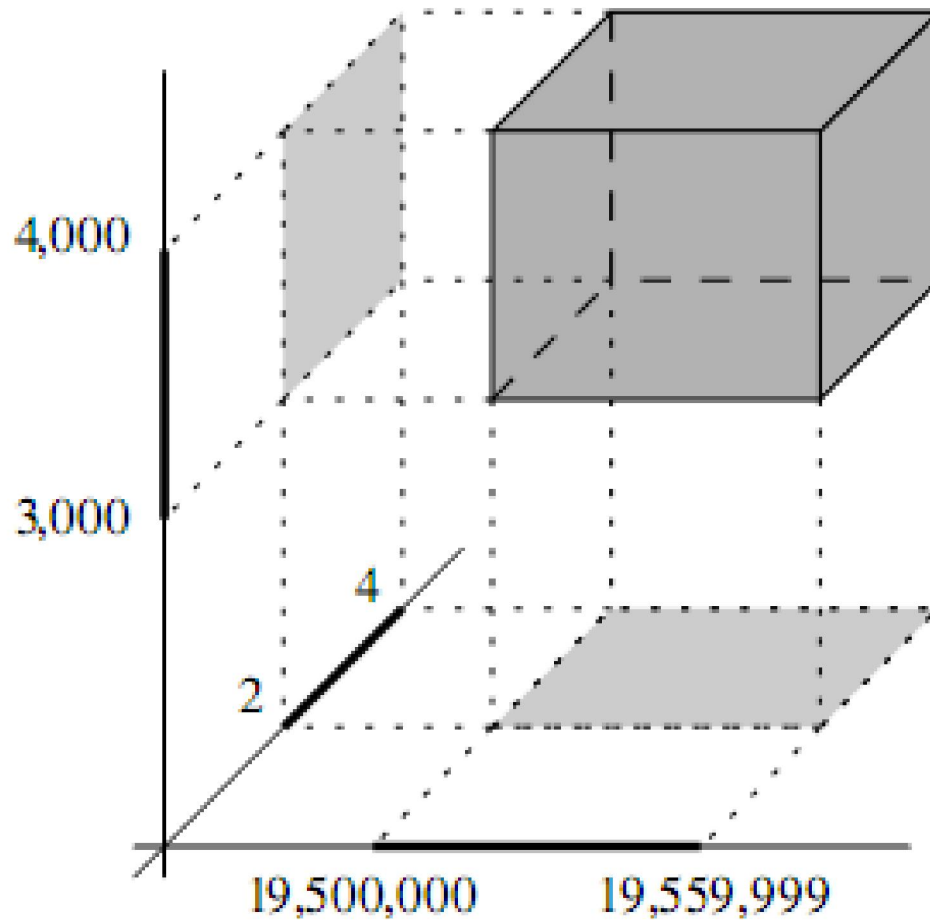
Interpreting a database query geometrically



Other Query

- If we have information about the number of children, to ask queries like “report all employees born between 1950 and 1955 who earn between \$3,000 and \$4,000 a month and have between two and four children “
- Represent each employee by a point in 3-dimensional space. The first coordinate represents the date of birth, the second coordinate the salary, the third coordinate the number of children

Interpreting of query



Such a query is called a **rectangular range query** or an **orthogonal range query**

1-Dimensional Range Searching

- $P := \{p_1, p_2, \dots, p_n\}$ is set of point on the rial line
- Data structure : a balanced binary search tree T
- Leaves store a point
- internal nodes store splitting values to guide the search
- Denote the splitting value stored at a node v by

$$x_v$$

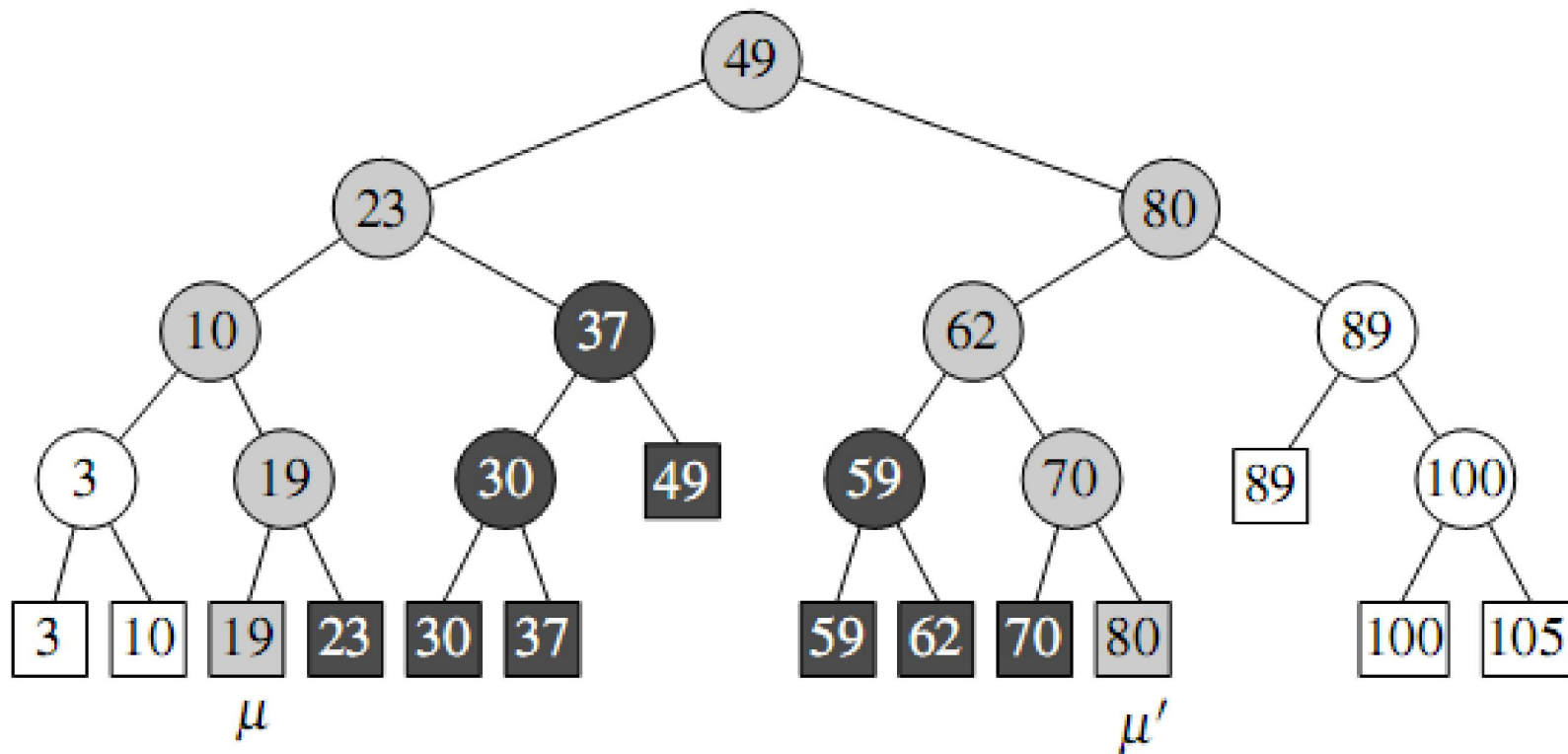
1-Dimensional Range Searching

To report the points in a query range $[x, x']$:

- Search with x and x' in T
- μ and μ' be the two leaves where the search end , respectively
- point in the interval $[x, x']$ are the ones stored in the leaves in between μ and μ' plus, possibly the point store at μ and μ'

A 1-dimensional range query in a binary search tree

search with the interval $[18, 77]$



1-Dimensional Range Searching

How can we find the leaves in between μ and μ' ?

- They are the leaves of certain subtrees in between the search paths to μ and μ'
- The subtrees that we select are rooted at nodes v in between the two search paths whose parents are on the search path

To find these nodes :

- search for the node v_{split} where the paths to x and x' split

$lc(v)$ and $rc(v)$ denote the left and right child of

Find split node

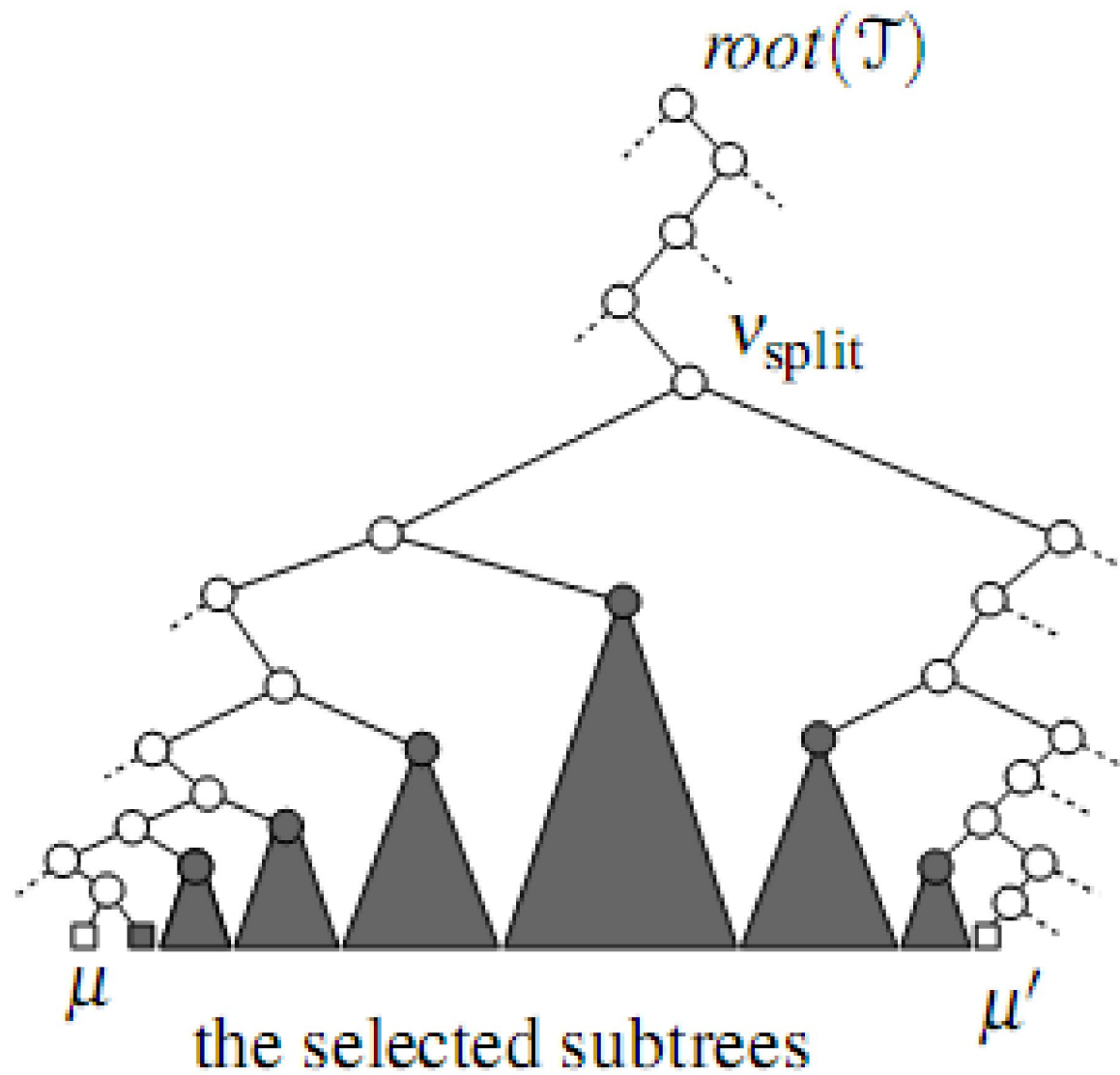
FINDSPLITNODE(T, x, x')

Input. A tree and two values x and x' with $x \leq x'$.

Output. The node v where the paths to x and x' split,
or the leaf where both paths end.

1. $v \leftarrow \text{root}(T)$
2. While v is not a leaf and $(x' \leq x_v \text{ or } x \geq x_v)$
3. do if $x' \leq x_v$
4. then $v \leftarrow \text{lc}(v)$
5. else $v \leftarrow \text{rc}(v)$
6. return v

The selected subtree



The selected subtree

Algorithm 1DRANGEQUERY($T, [x, x']$)

Input. A binary search tree T and a range $[x, x']$.

Output. All points stored in T that lie in the range.

1. $v_{split} \leftarrow FINDSPLITNODE(T, x, x')$
2. if v_{split} is a leaf
3. then Check if the point stored at v_{split} must be reported.
4. else (*Follow the path to x and report the points in subtrees right of the path.*)
5. $v \leftarrow lc(v_{split})$
6. while v is not a leaf
7. do if $x \leq x_v$
8. then REPORTSUBTREE($rc(v)$)
9. $v \leftarrow lc(v)$
10. else $v \leftarrow rc(v)$
11. Check if the point stored at the leaf v must be reported.
12. Similarly, follow the path to x' , report the points in subtree left of the path ,
and check if the poin stored at the leaf where the path ends must be reported.

Prove Correctness Of The Algorithm

lemma 5.1

- Algorithm 1DRANGEQUERY reports exactly those point that lie in the query range

proof

- Any reported point p lies in the query range
- Any point p in the range is reported

Summarizes the results for 1-dimensional range searching

Theorem 5.2

- Let P be a set of n points in 1-dimensional space. The set P can be stored in a balanced binary search tree, which uses $O(n)$ storage and has $O(n \log n)$ construction time, such that the points in a query range can be reported in time $O(k + \log n)$, where k is the number of reported points.

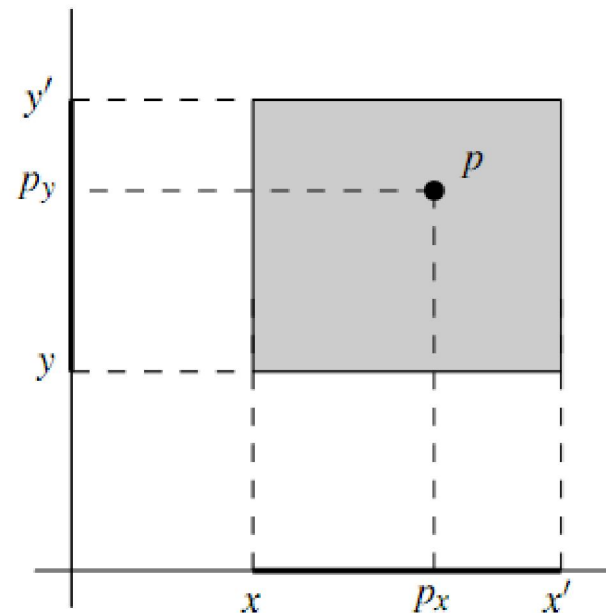
Kd-Trees

2-dimensional rectangular range searching problem

- In this section we assume that no two points in p have the same x-coordinate, and no two points have the same y-coordinate
- A 2-dimensional rectangular range query on p asks for the point from p lying inside a query rectangle $[x : x'] \times [y : y']$

- A point $p := (p_x, p_y)$ lies inside this rectangle if and only if

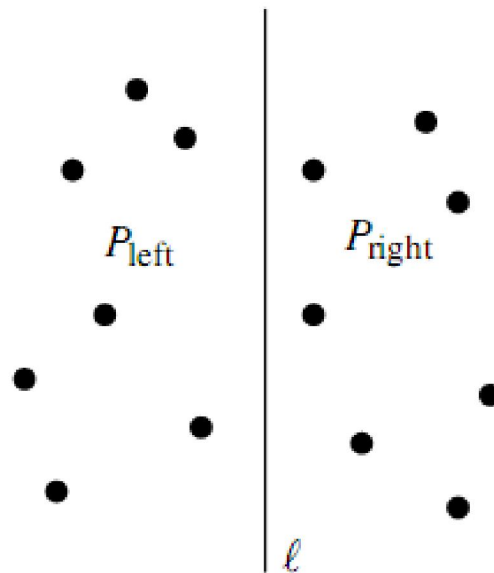
$$p_x \in [x, x'] \quad \text{and} \quad p_y \in [y, y'].$$



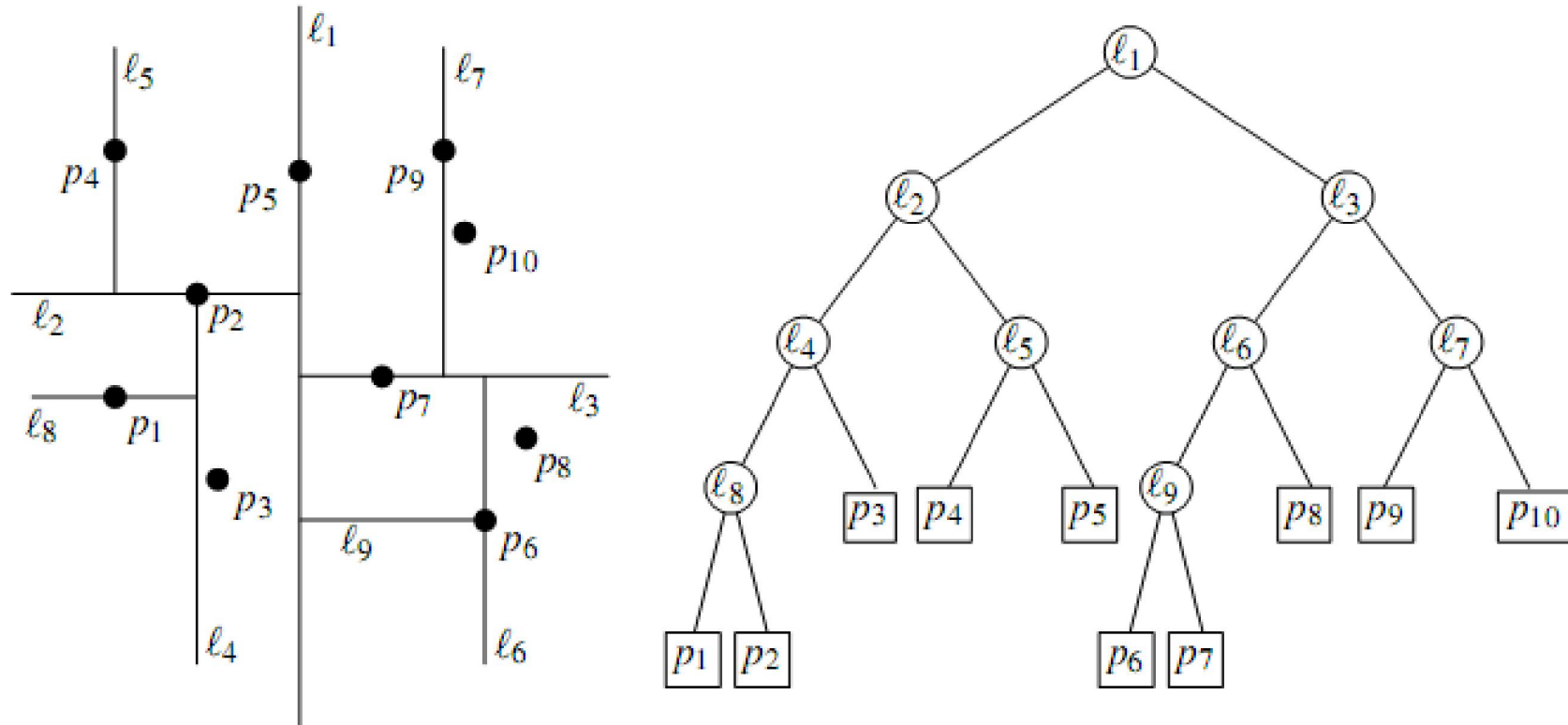
Kd-Trees

How can we generalize structure for 1-dimensional ?

- First split on x-coordinate, next on y-coordinate, then again on x-coordinate, and so on
- The splitting line is stored at the root
- P_{left} the subset of point to the left or on the splitting line, is stored in the left subtree, and P_{right} , the subset to the right of it, is stored in the right subtree



A kd-tree



- A tree like this is called a **kd -tree**

Construct a kd-tree

Algorithm BUILDKDTREE (P, depth)

Input .A set of points P and the current depth depth .

Output. The root of a kd - tree storing P .

1. if P contains only one point
2. then return a leaf storing this point
3. else if depth is even
4. the n Split P into two subsets with a vertical line l through t he median
 x - coordinate of the points in P . Let p_1 be the set of points to
 th e left of l or on l and p_2 be the set of points to the right of l .
5. else Split P into two subsets with a horizontal line l through t he median
 y - coordinate of the points in P . Let p_1 be the set of points to
 th e left of l or on l and p_2 be the set of points to the right of l .
6. $v_{\text{left}} \leftarrow \text{BUILDKDTREE } E(p_1, \text{depth} + 1)$
7. $v_{\text{right}} \leftarrow \text{BUILDKDTREE } E(p_2, \text{depth} + 1)$
8. Create a node v storing l , make v_{left} the left child of v , nd make v_{right}
 the right child of v
9. return v

Sorted list x-coordinate

1	4	2	3	5	7	9	10	6	8
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Sorted list y-coordinate

6	3	8	1	7	2	10	5	9	4
---	---	---	---	---	---	----	---	---	---

1	4	2	3	5	7	9	10	6	8
---	---	---	---	---	---	---	----	---	---

3	1	2	5	4	6	8	7	10	9
---	---	----------	---	---	---	---	----------	----	---

1	2	4	3	5	7	6	9	10	8
---	---	---	---	---	---	---	---	----	---

3	1	2	5	4	6	8	7	10	9
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Storage And Building Time A kd-tree

The buiding time $T(n)$:

$$T(n) = \begin{cases} O(1), & \text{if } n = 1 \\ O(n) + 2T\left(\left\lceil \frac{n}{2} \right\rceil\right), & \text{if } n > 1 \end{cases}$$

which solves to $O(n \log n)$

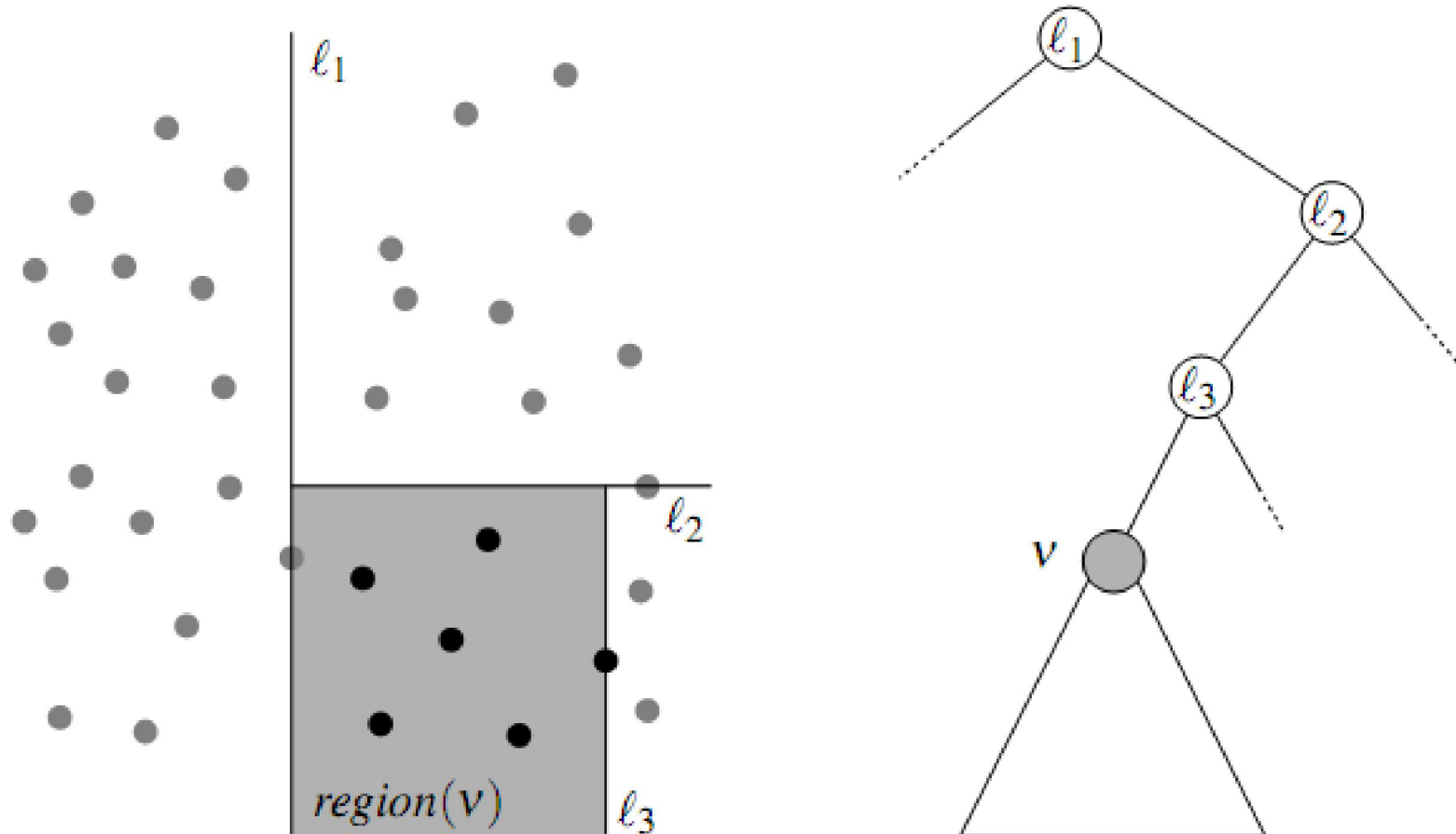
lemma 5.3

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

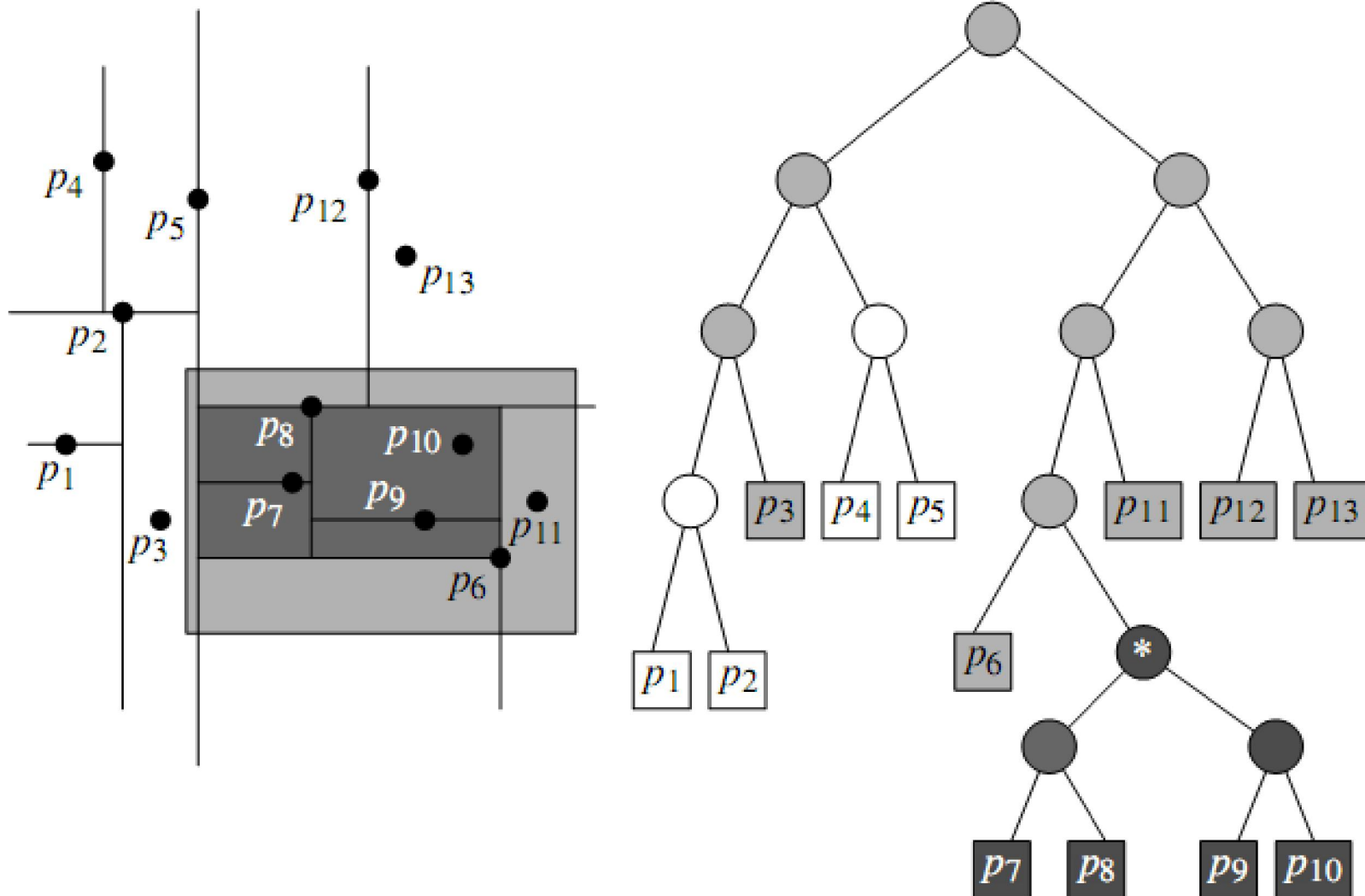
Corresponding Between Nodes In A KD-tree And Regions In The Plane

- The region corresponding to a node v is a rectangle
- It is bounded by splitting lines stored at ancestors of v which denote by $region(v)$
- We have to search the subtree rooted at v only if the query rectangle intersects $region(v)$

Corresponding Between Nodes In A KD-tree And Regions In The Plane



A query on a kd-tree



Algorithm SEARCH KDTREE(v, R)

Input. The root of (a subtree of) a kd - tree , and a range R .

Output. All point at leaves below v that lie in the range.

1. if v is a leaf
2. then Report the point stored at v if it lies in R .
3. else if $region(lc(v))$ is fully contained in R
4. then REPORTSUBTREE($lc(v)$)
5. else if $region(lc(v))$ intersects R
6. then SEARCH KDTREE($lc(v), R$)
7. if $region(rc(v))$ is fully contained in R
8. then REPORTSUBTREE($rc(v)$)
9. else if $region(rc(v))$ intersects R
10. then SEARCH KDTREE($rc(v), R$)

$$region\ lc(v) = region(v) \cap l(v)^{left}$$

