

How to Teach Yourself Programming

<https://abstrusegoose.com/249>

Days 1 - 10

Teach yourself variables, constants, arrays, strings, expressions, statements, functions,....



Days 11 - 21

Teach yourself program flow, pointers, references, classes, objects, inheritance, polymorphism,



Days 22 - 697

Do a lot of recreational programming. Have fun hacking but remember to learn from your mistakes.



Days 698 - 3648

Interact with other programmers. Work on programming projects together. Learn from them.



Days 3649 - 7781

Teach yourself advanced theoretical physics and formulate a consistent theory of quantum gravity.



Days 7782 - 14611

Teach yourself biochemistry, molecular biology, genetics,...



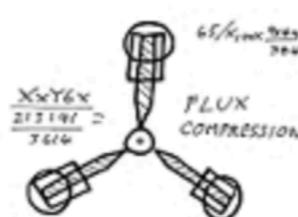
Day 14611

Use knowledge of biology to make an age-reversing potion.



Day 14611

Use knowledge of physics to build flux capacitor and go back in time to day 21.



Day 21

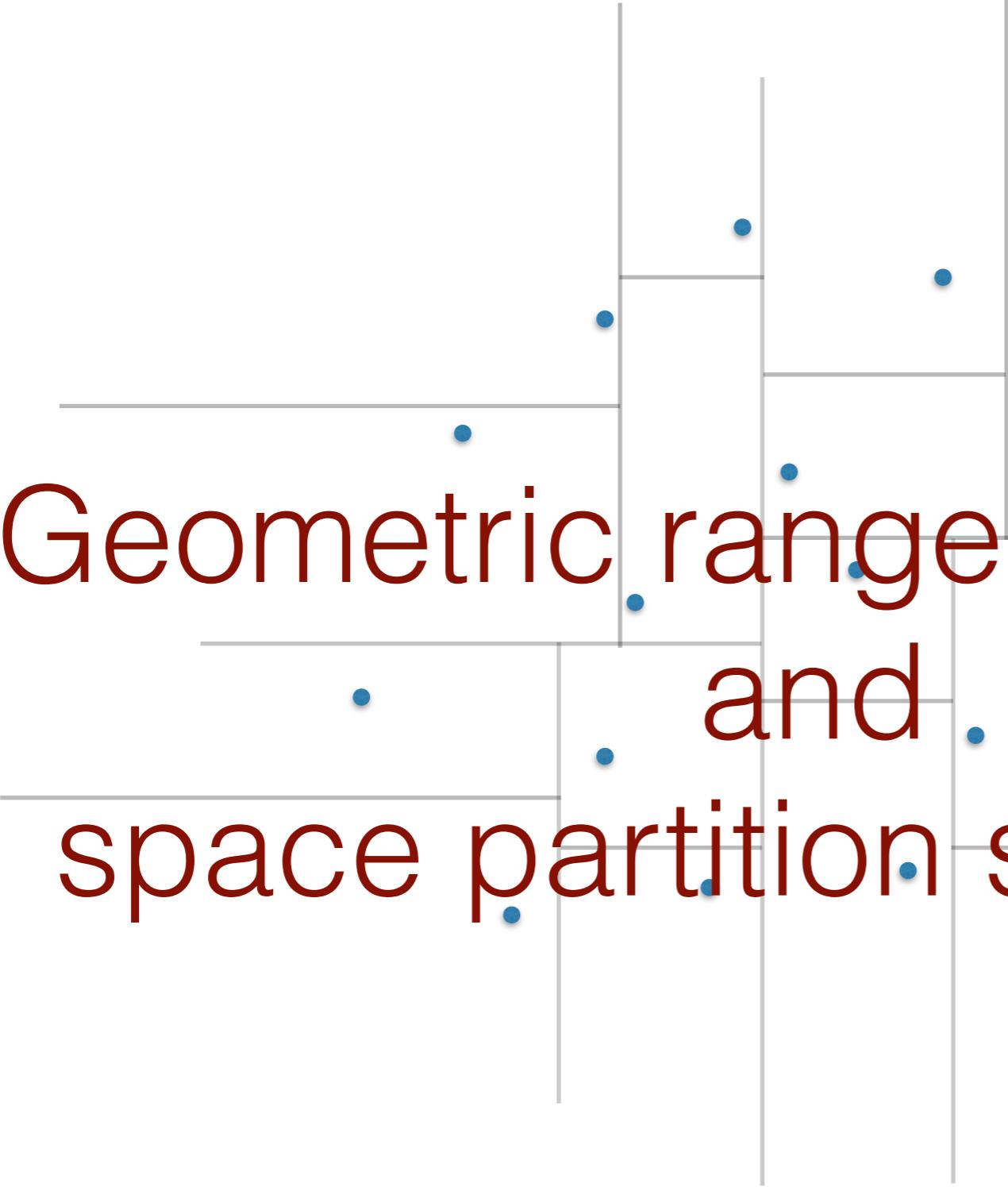
Replace younger self.



As far as I know, this is the easiest way to

"Teach Yourself C++ in 21 Days".

Seriously, why is everyone in **such a rush?**



Geometric range searching and space partition structures

Where we are

“Global” problems

- closest pair
- convex hull
- intersections
- ...

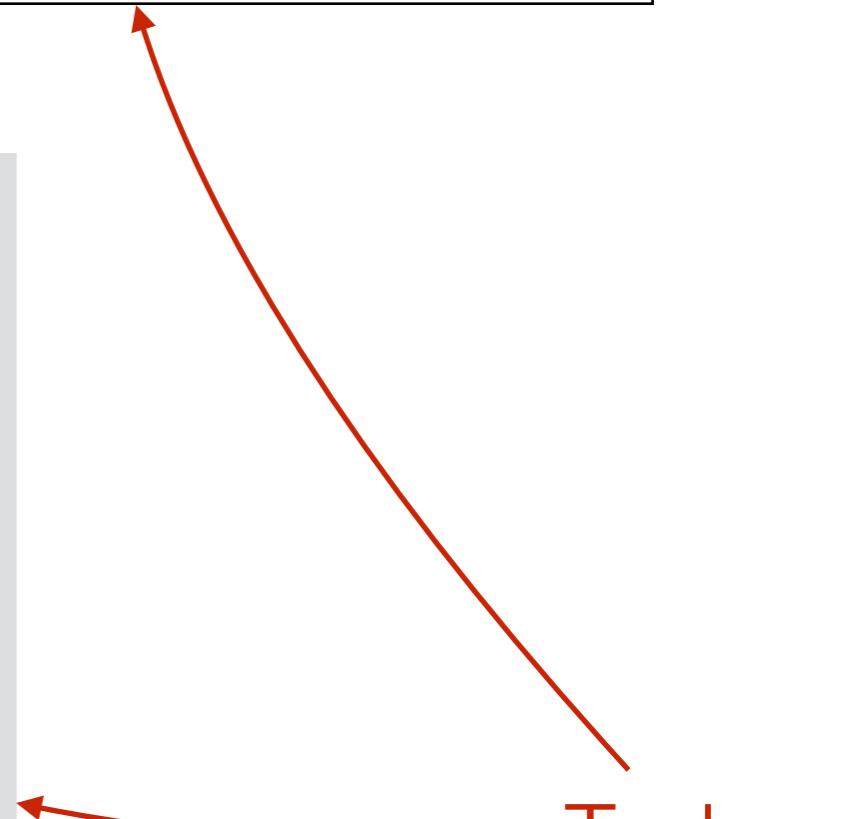
Geometric search problems

- range searching
- nearest neighbor
- k-nearest neighbor
- find all roads within 1km of current location
- ...

Techniques

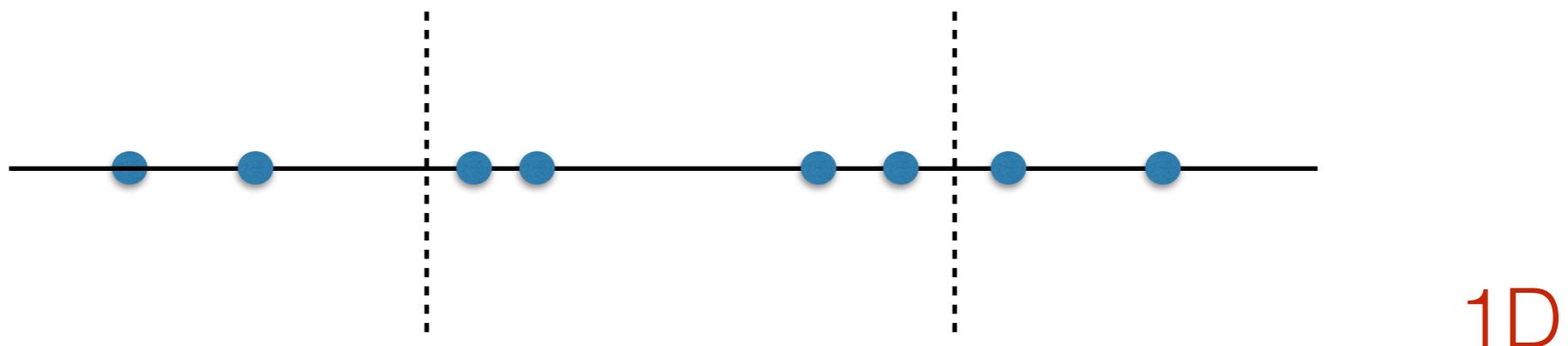
- divide-and-conquer
- incremental
- plane sweep
- space decomposition
- ...

Today



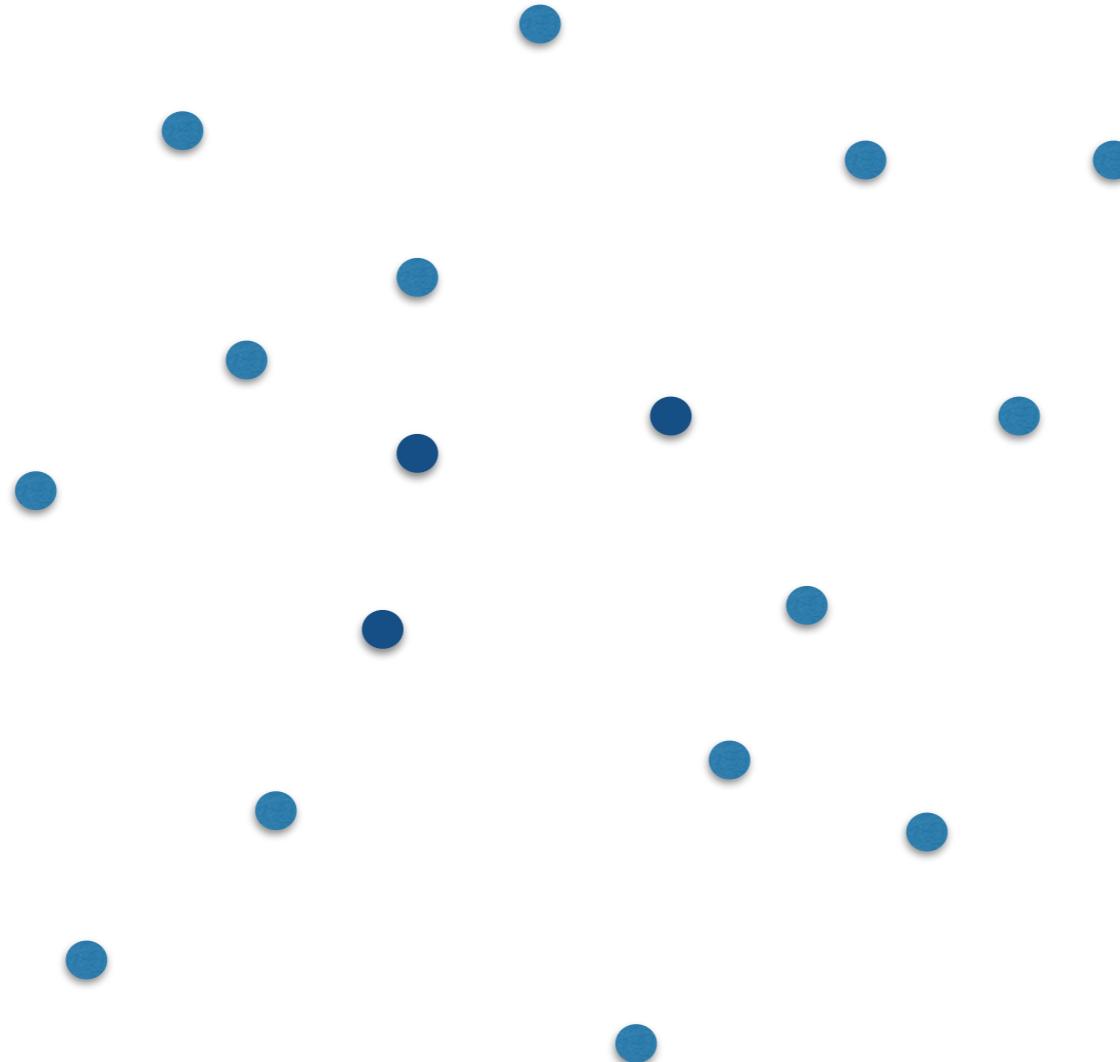
1D Range searching

Given a set of n points on the real line, preprocess them into a data structure to support fast range queries.



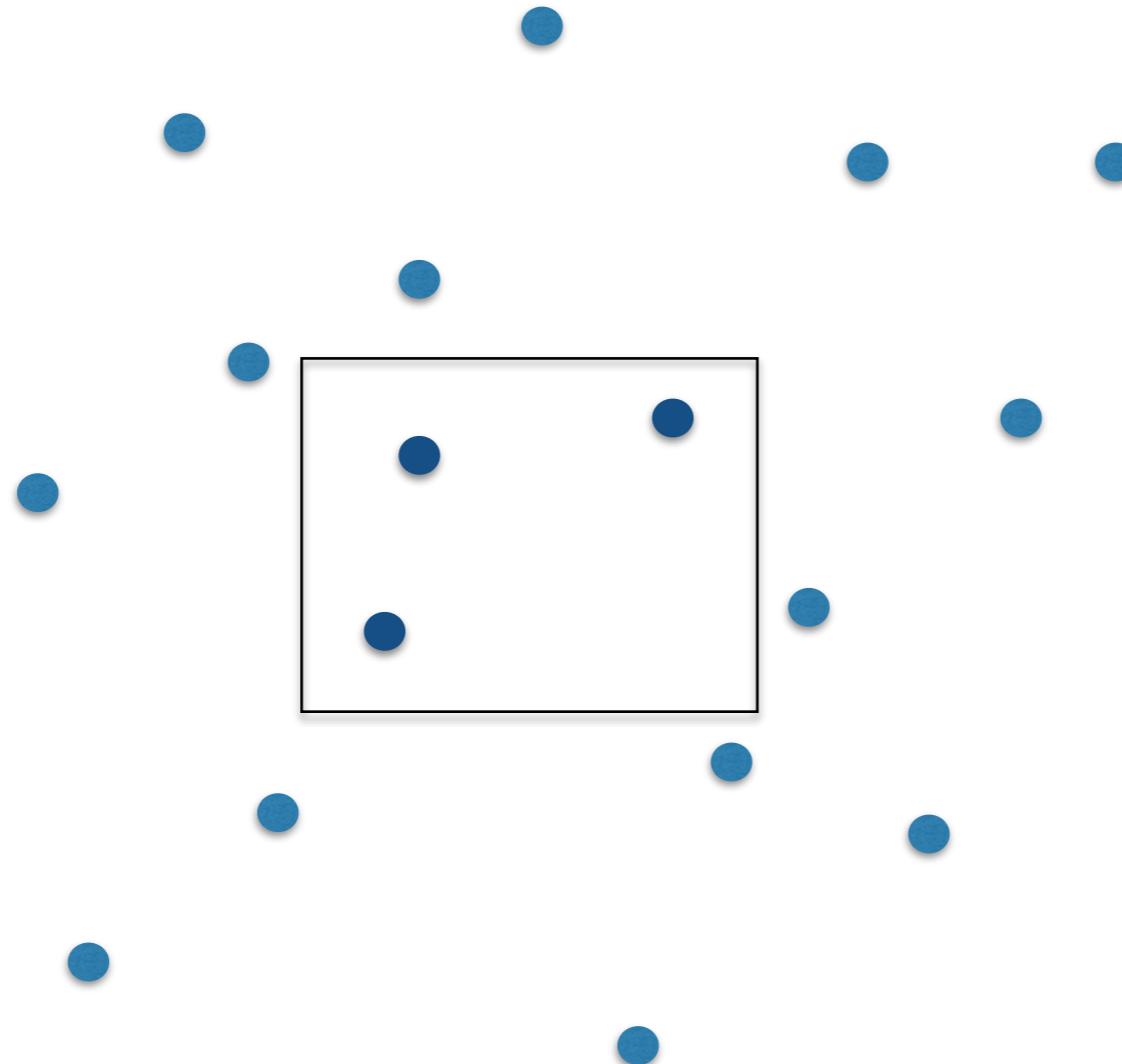
2D Range searching

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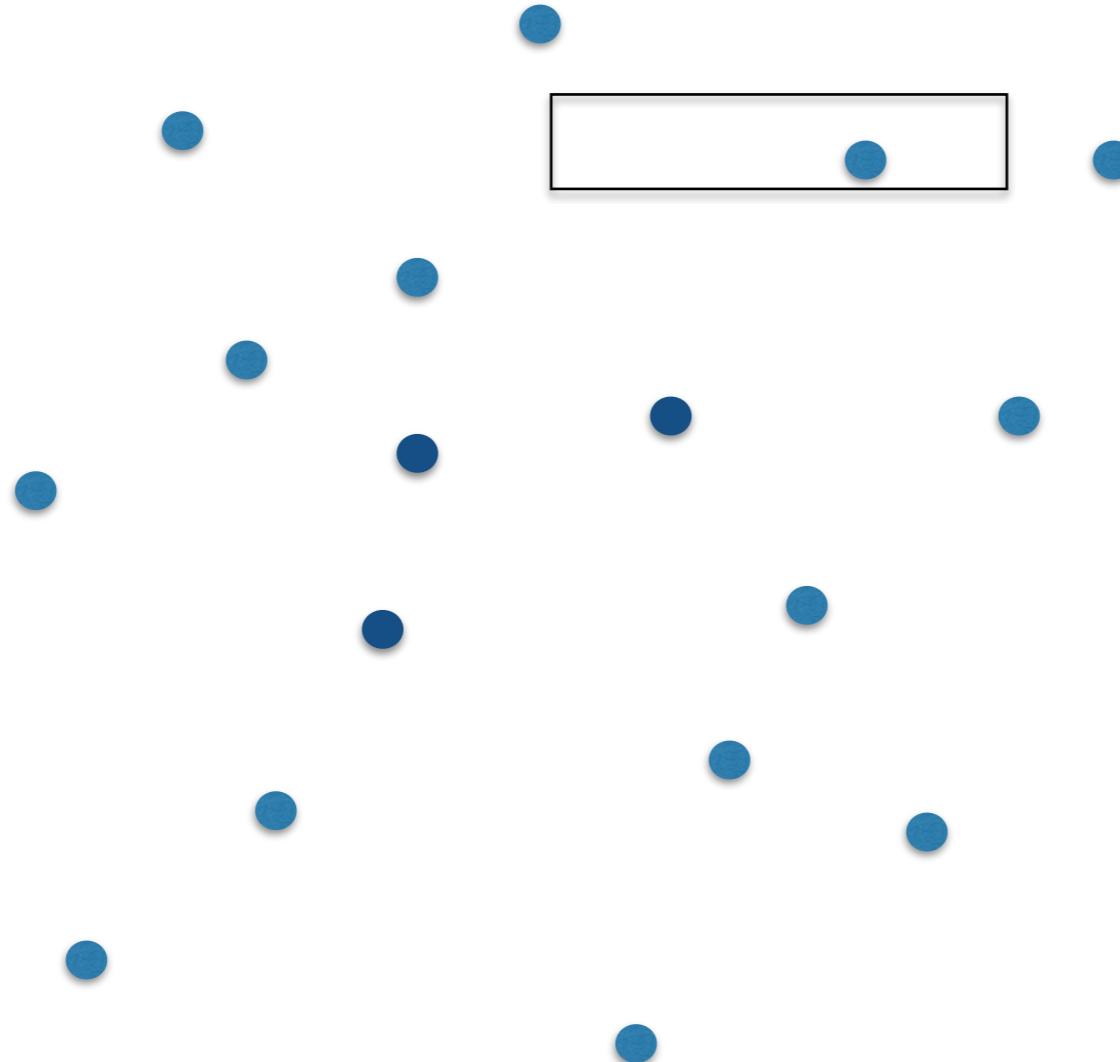
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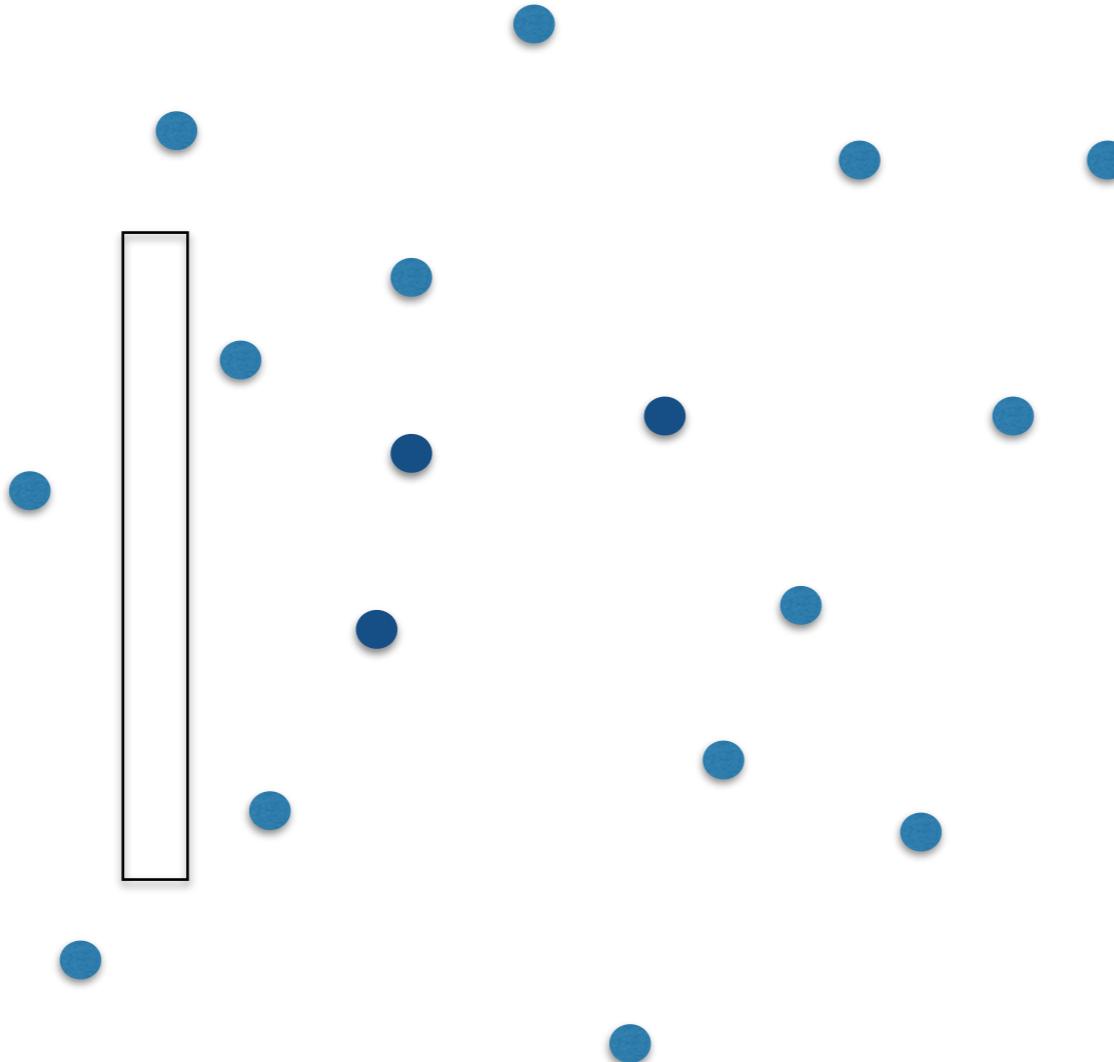
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2D Range searching

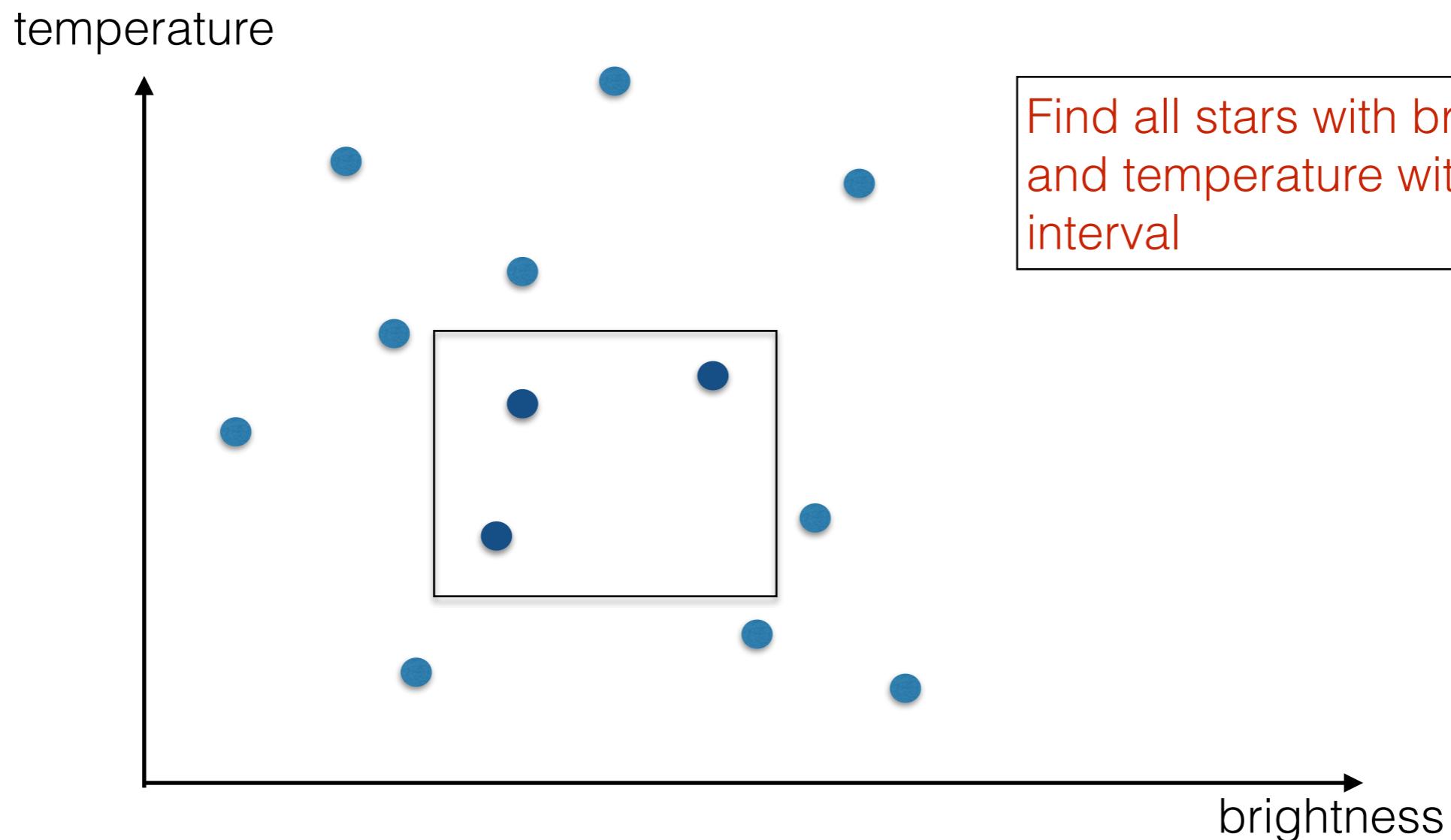
Given a set of points, preprocess them into a data structure to support fast range queries.



Why?

Arise in settings that are not geometrical.

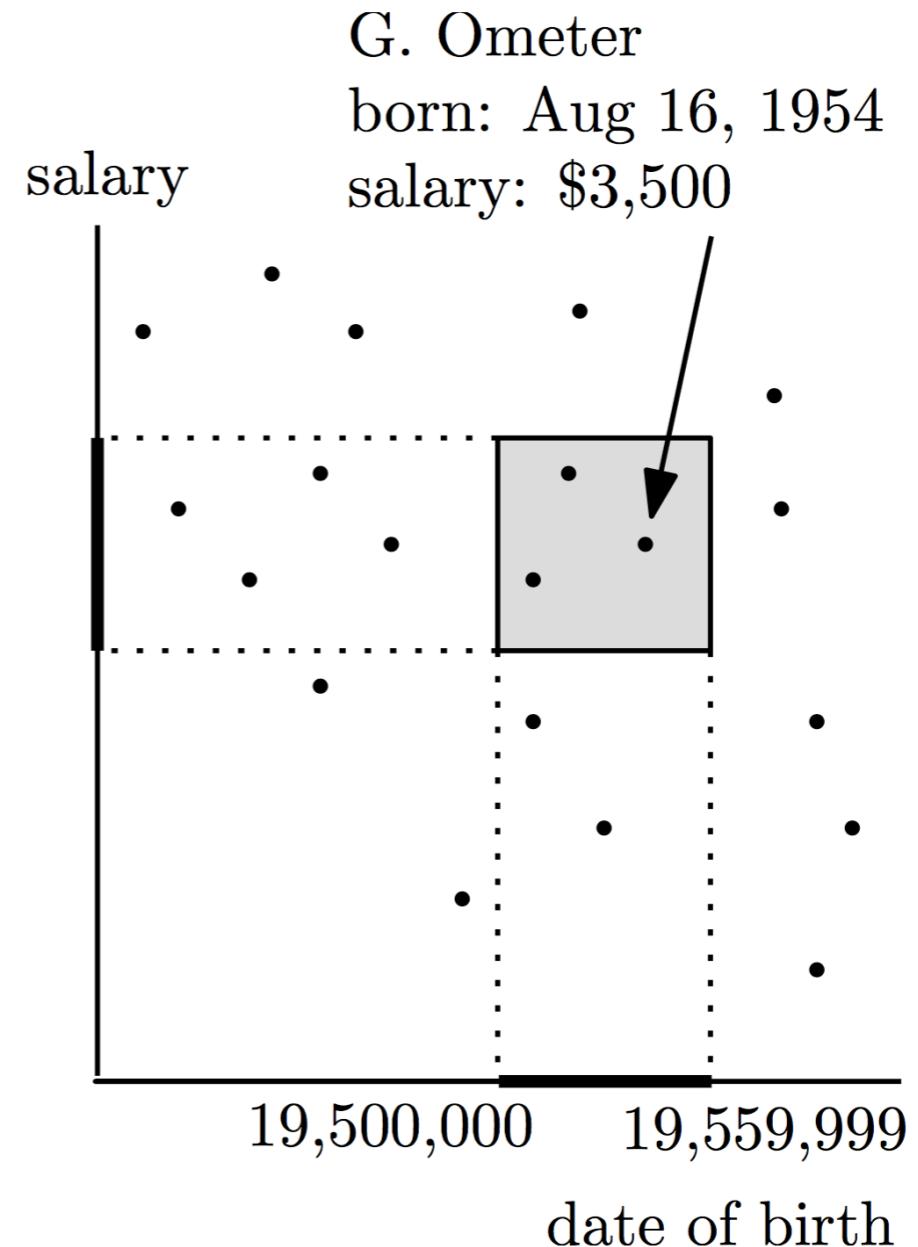
Database of stars. A star = (brightness, temperature,.....)



Why?

Database of employees. An employee = (age, salary,.....)

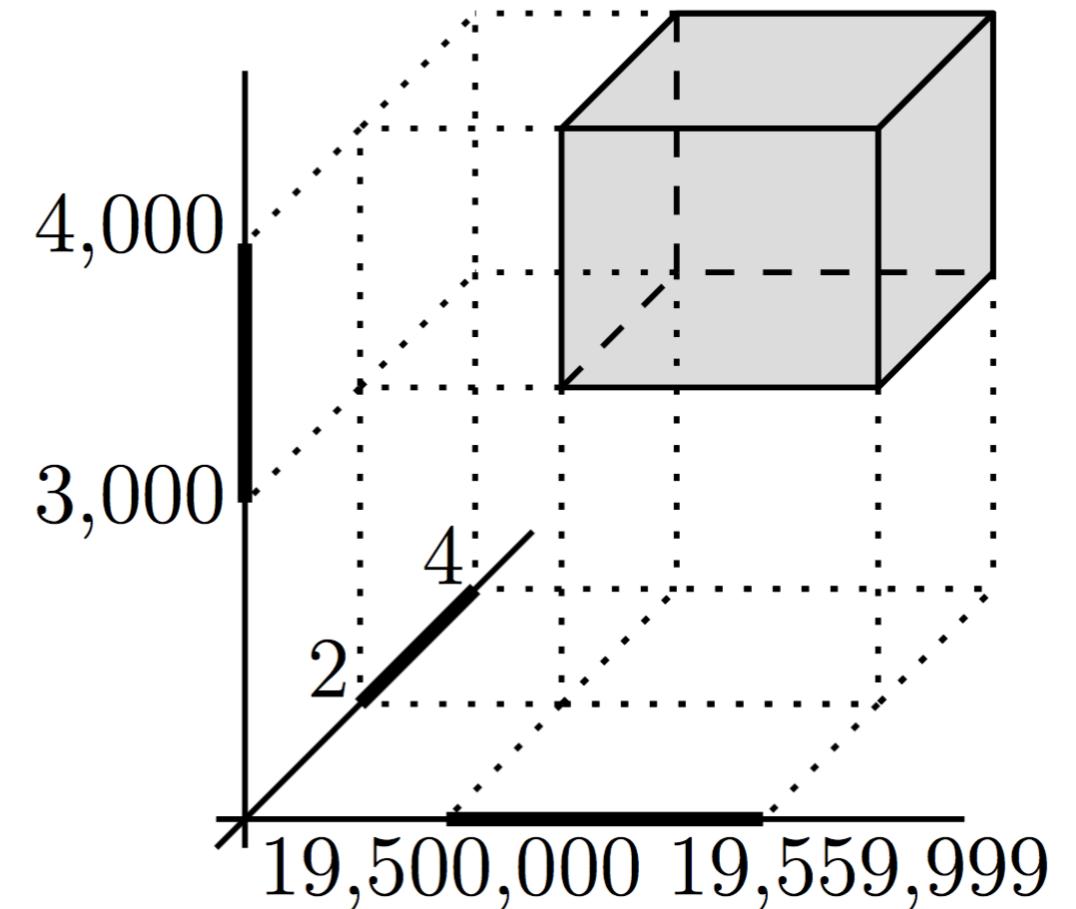
A database query may ask for all employees with age between a_1 and a_2 , and salary between s_1 and s_2



screenshot from Mark van Kreveld slides at <http://www.cs.uu.nl/docs/vakken/ga/slides5a.pdf>

Why?

Example of a 3-dimensional
(orthogonal) range query:
children in $[2, 4]$, salary in
 $[3000, 4000]$, date of birth in
 $[19,500,000, 19,559,999]$



screenshot from Mark van Kreveld slides at <http://www.cs.uu.nl/docs/vakken/ga/slides5a.pdf>

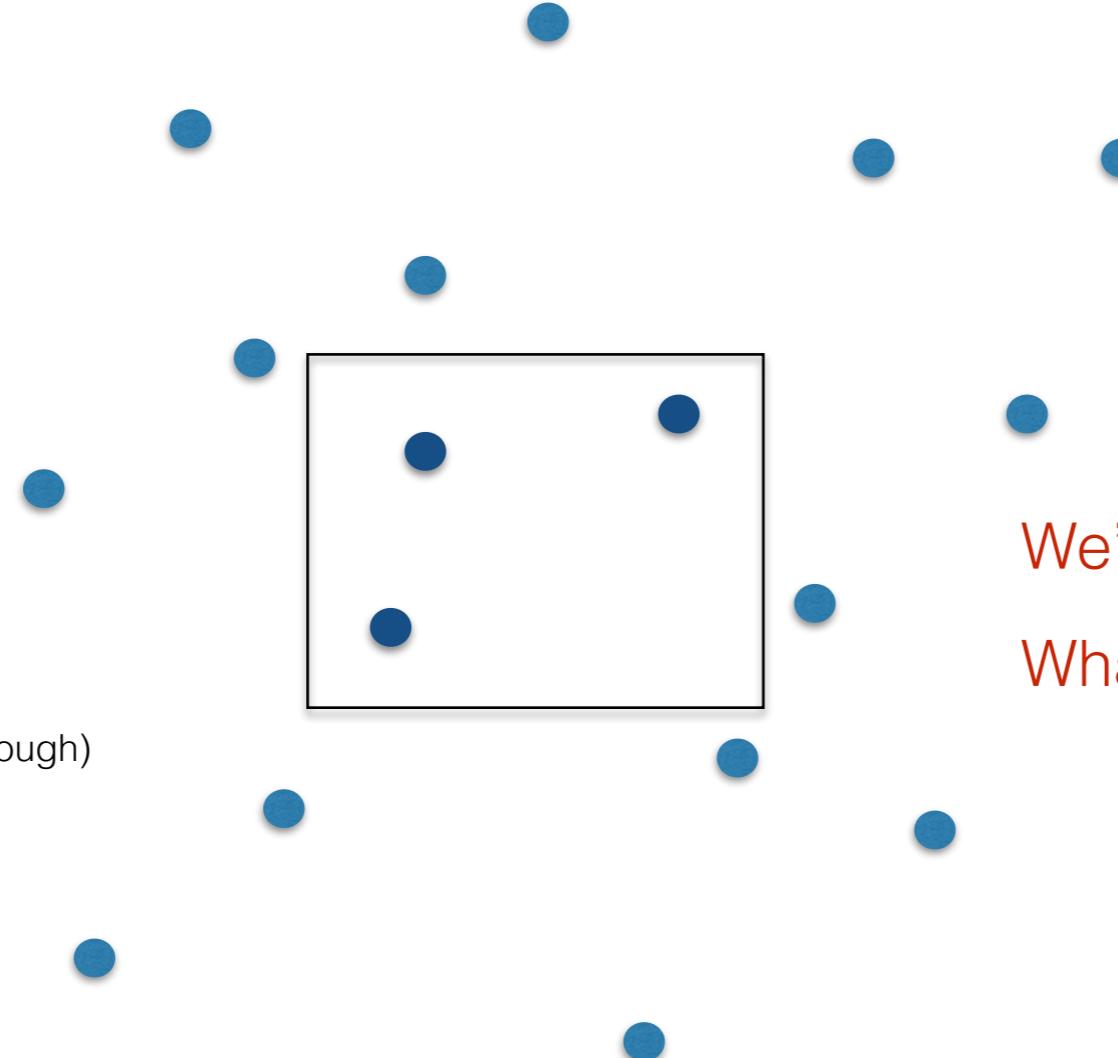
Range searching in 2D

- n: size of the input (number of segments)
- k: size of output (number of points inside range)

How?

The naive approach:

- No data structure: traverse and check in $O(n)$
- Note: good when k is large



Points are static or dynamic?

We'll assume static (it's hard enough)

We'd like to do better.

What sort of bounds can we expect?

Range searching in 2D: How?

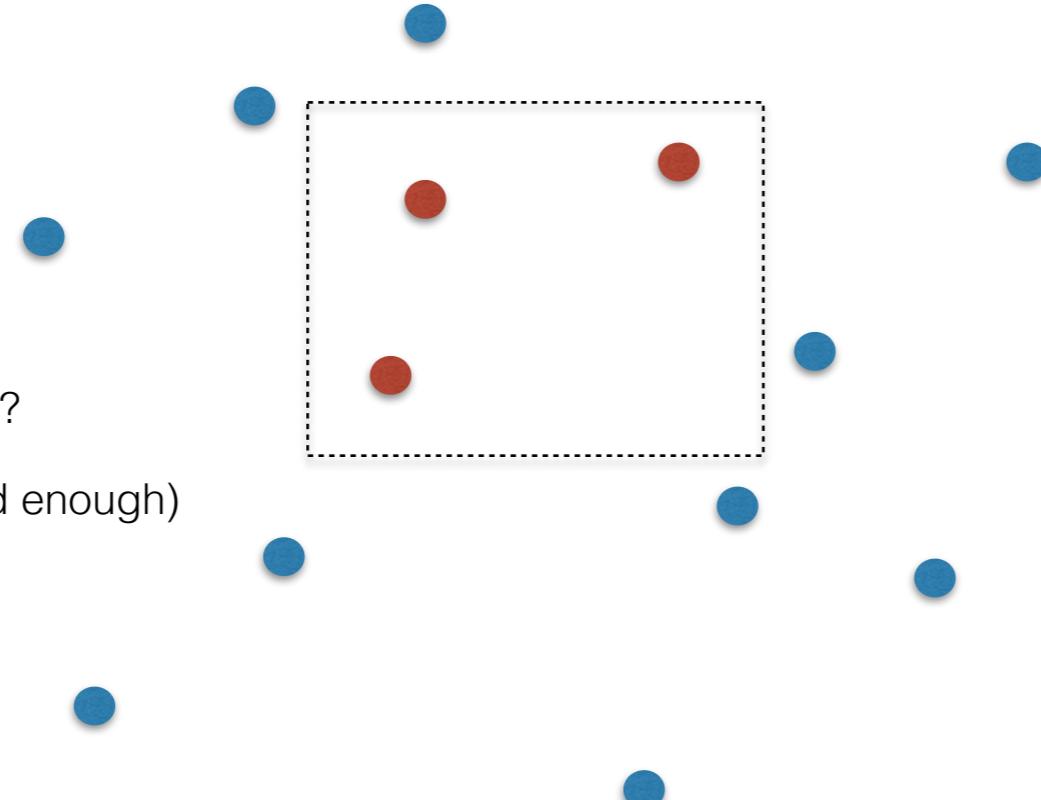
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- No data structure: traverse and check in $O(n)$
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 - Note: good when k is large

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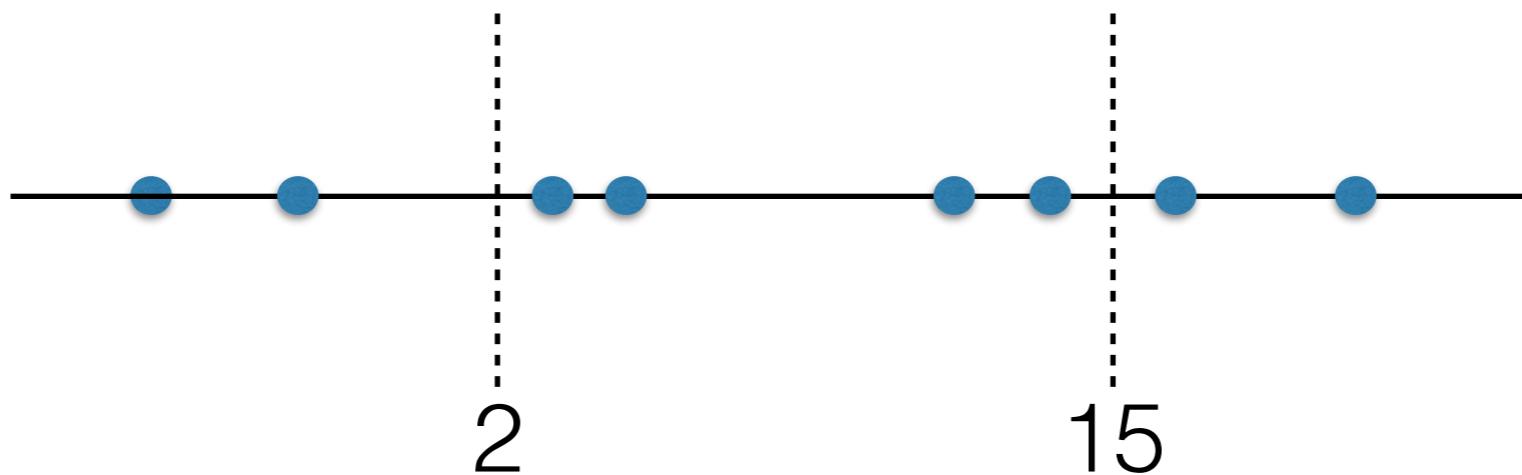
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1D range searching

1D Range searching

Given a set of n points on the real line, preprocess them into a data structure to support fast range queries.



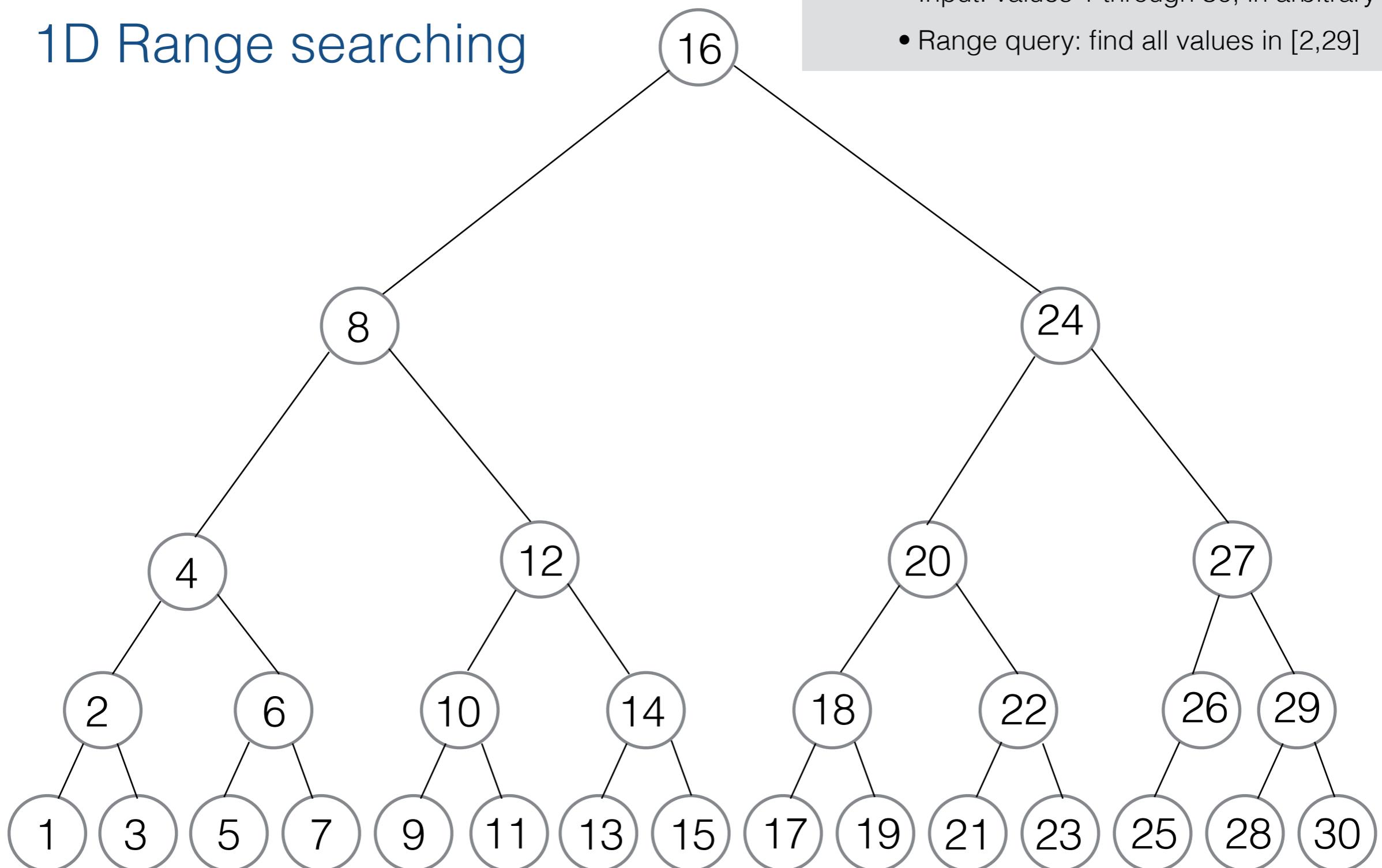
find all values in $[2, 15]$

How do we solve this and how fast?

1D Range searching

Example

- Input: values 1 through 30, in arbitrary order
- Range query: find all values in [2,29]

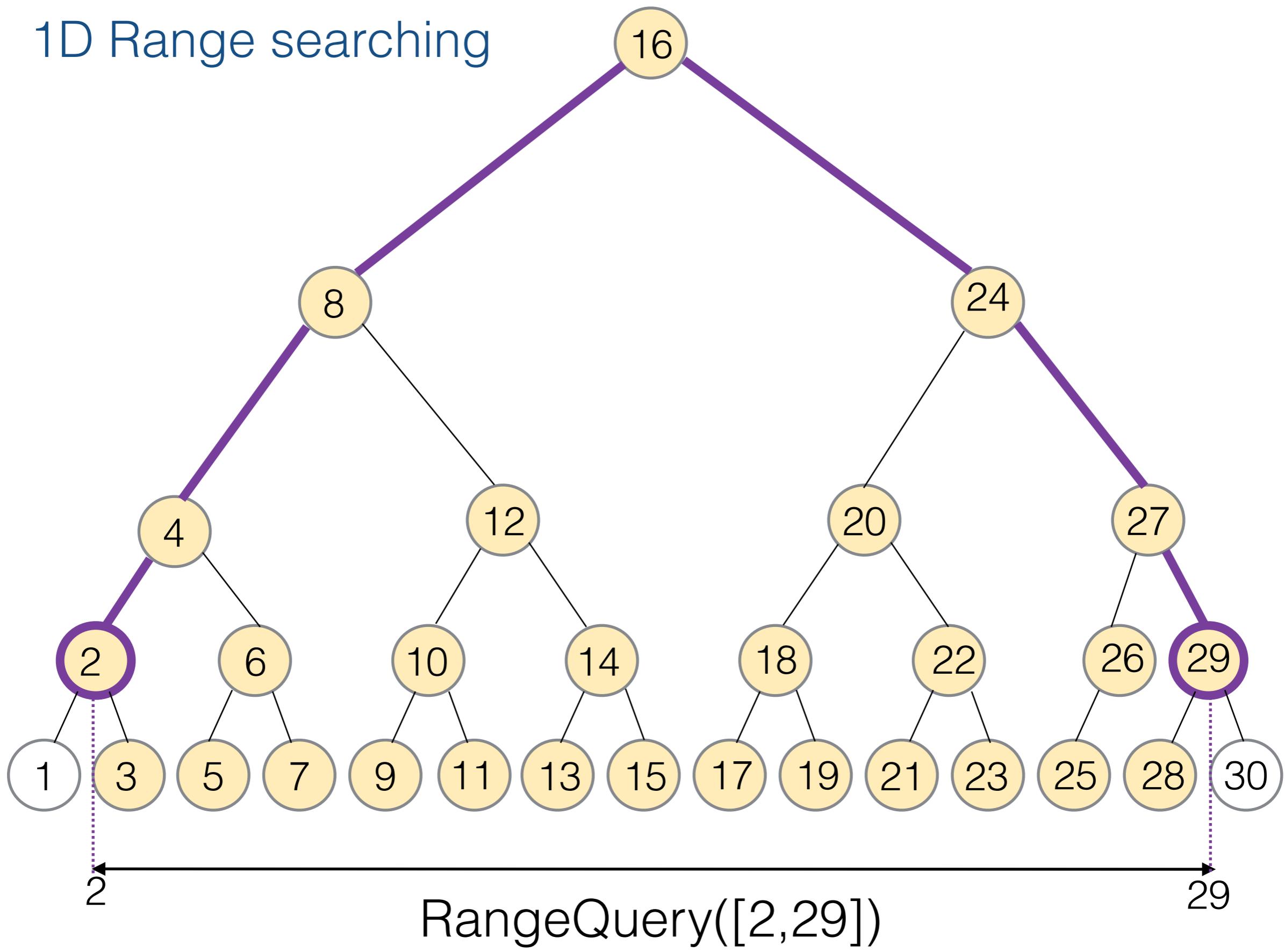


2

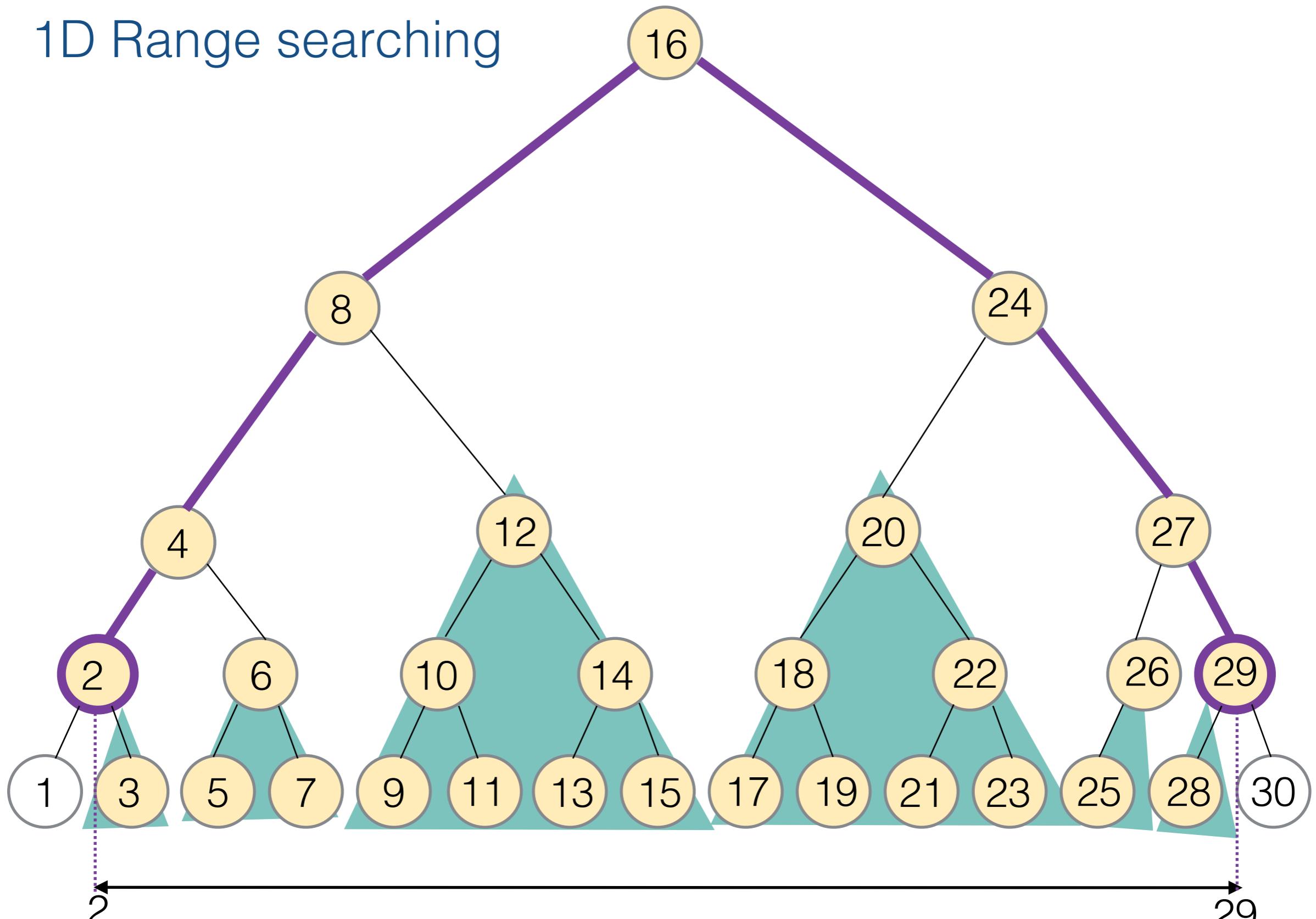
29

RangeQuery([2,29])

1D Range searching



1D Range searching



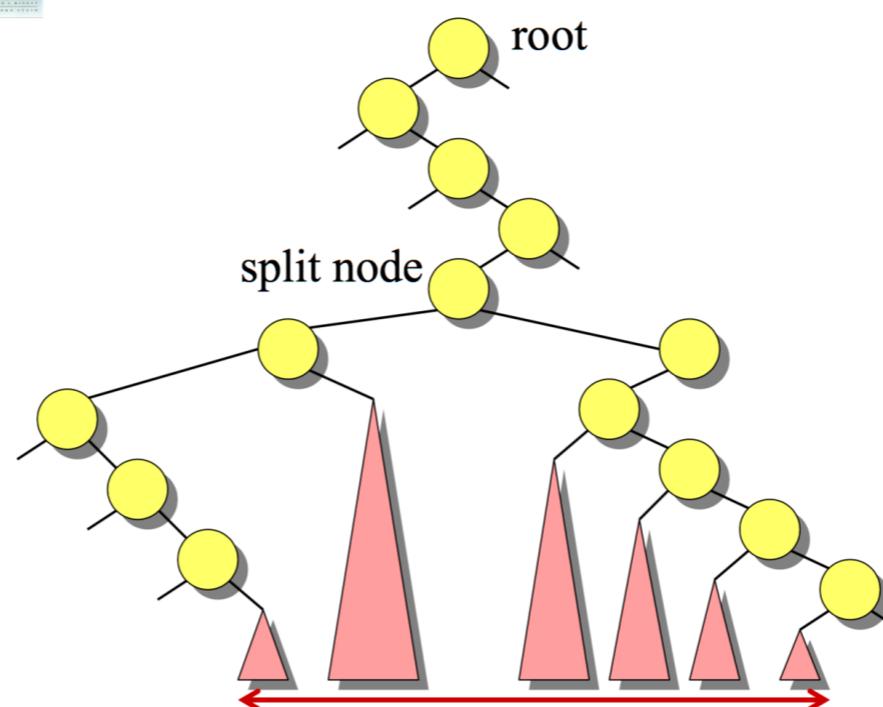
The k points in the range are in $O(\lg n)$ subtrees

1D Results

- A set of n points (1D) can be pre-processed into a BBST such that:
 - Build: $O(n \lg n)$
 - Space: $O(n)$
 - Range queries: $O(\lg n + k)$
 - Dynamic: points can be inserted/deleted in $O(\lg n)$



General 1D range query



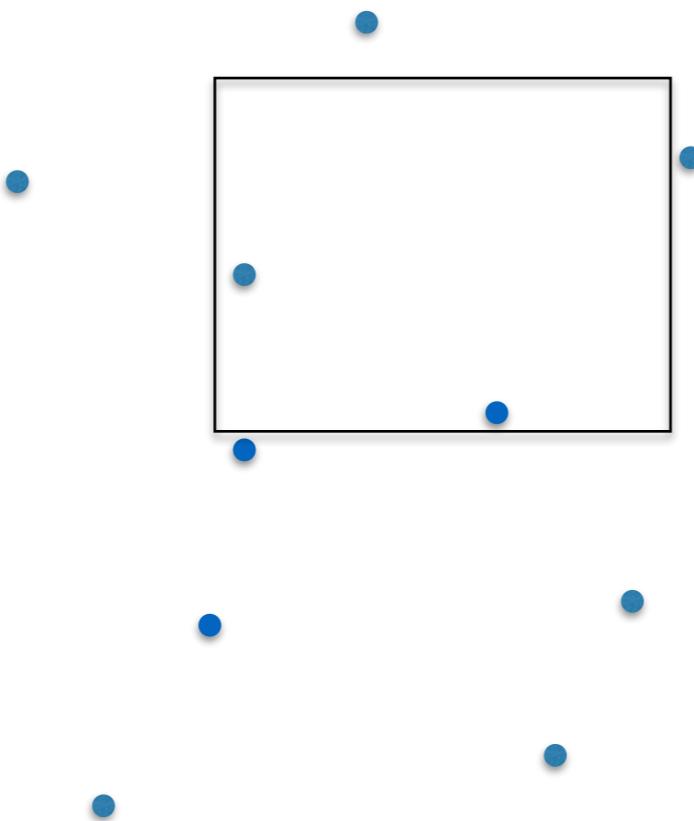
1D Results

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 - Build: $O(n \lg n)$
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 - Range queries: $O(\lg n + k)$
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2D

- A set of n 2d-points can be pre-processed into a structure such that:
 - Build: $O(n \lg n)$
 - Space: $O(n)$
 - Range queries: $O(\lg^2 n + k)$
- These bounds would be nice
But how?

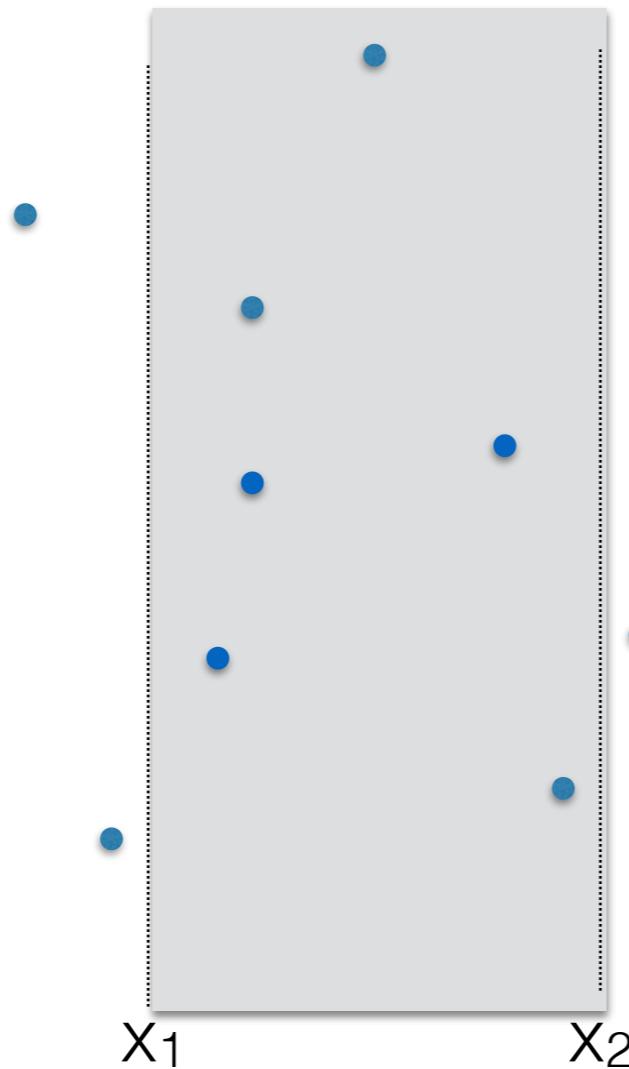
How could we use 1D structure for 2D?



Denote query $[x_1, x_2] \times [y_1, y_2]$

Could it be as simple as ..

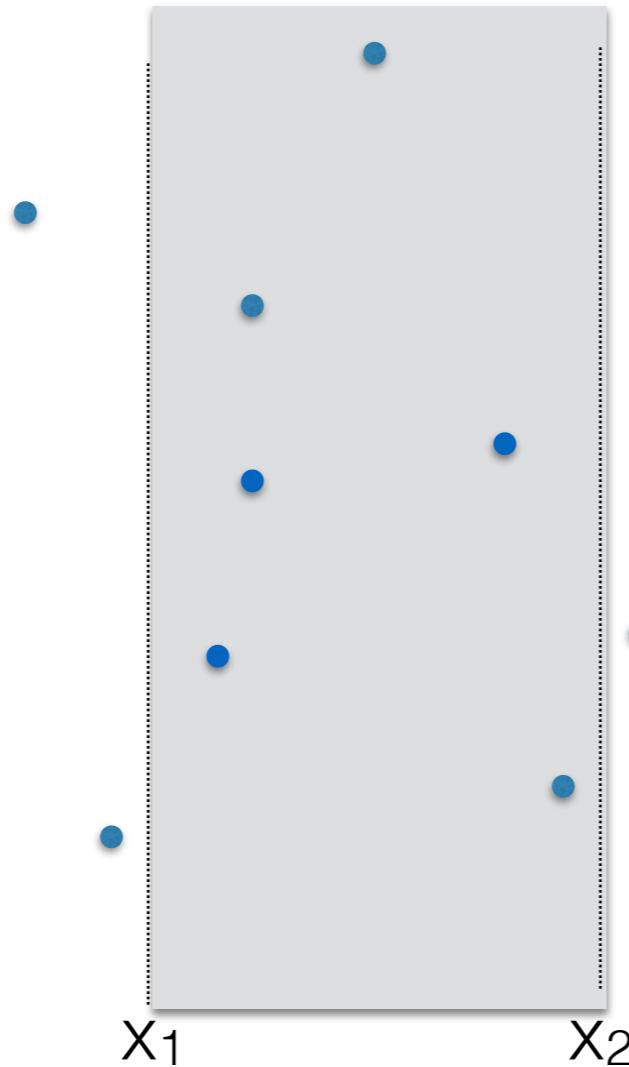
- Find all points with the x-coordinates in the correct range $[x_1, x_2]$



Denote query $[x_1, x_2] \times [y_1, y_2]$

Could it be as simple as ..

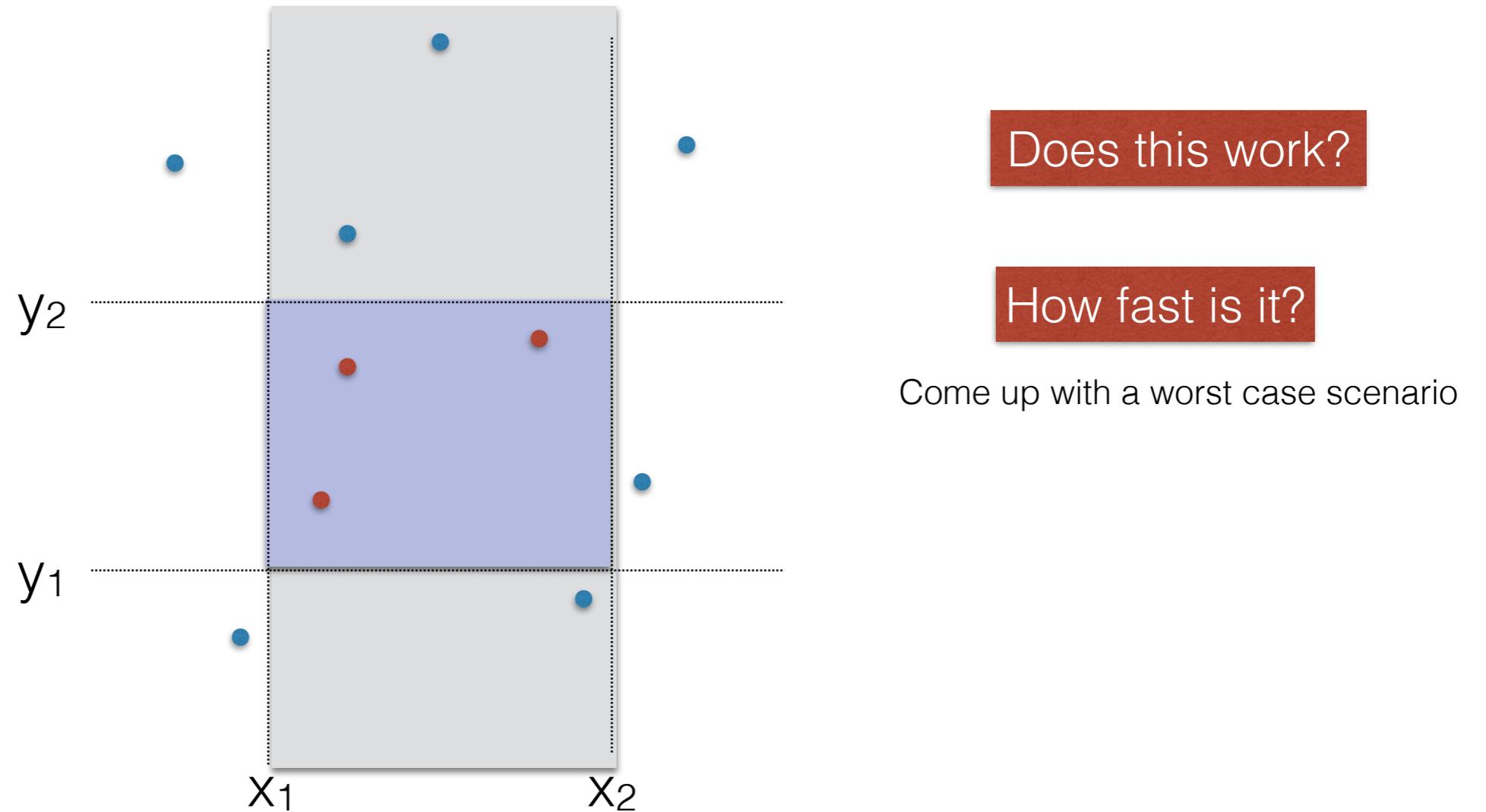
- Find all points with the x-coordinates in the correct range $[x_1, x_2]$
- Out of these points, find all points with the y-coord in the correct range $[y_1, y_2]$

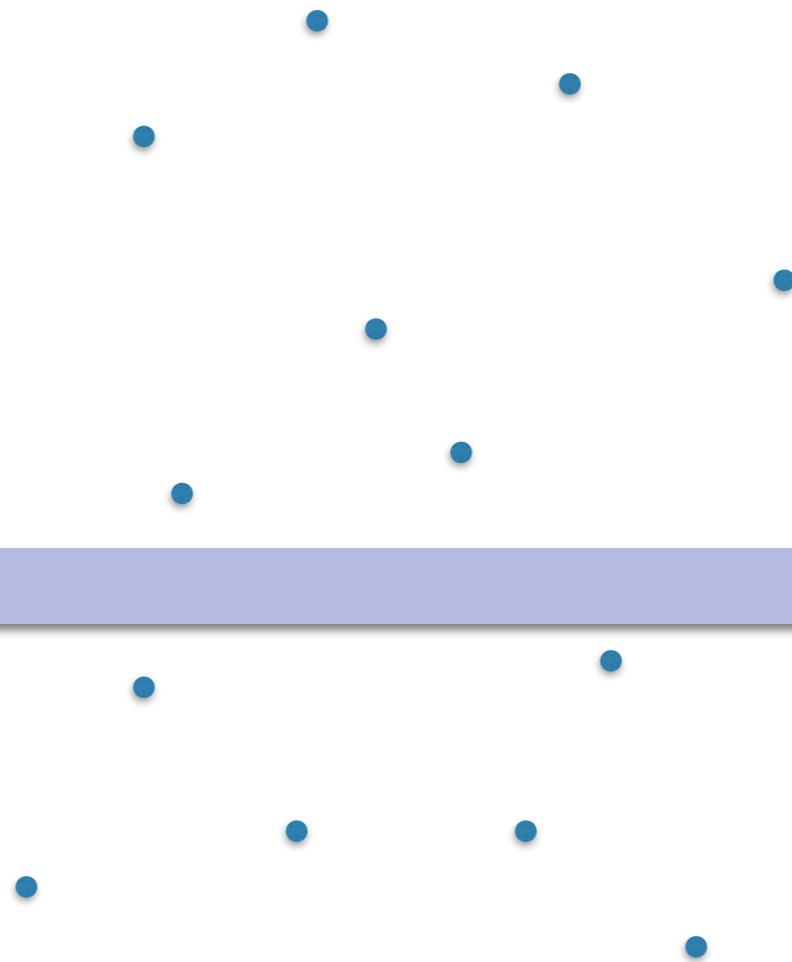


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Could it be as simple as ..

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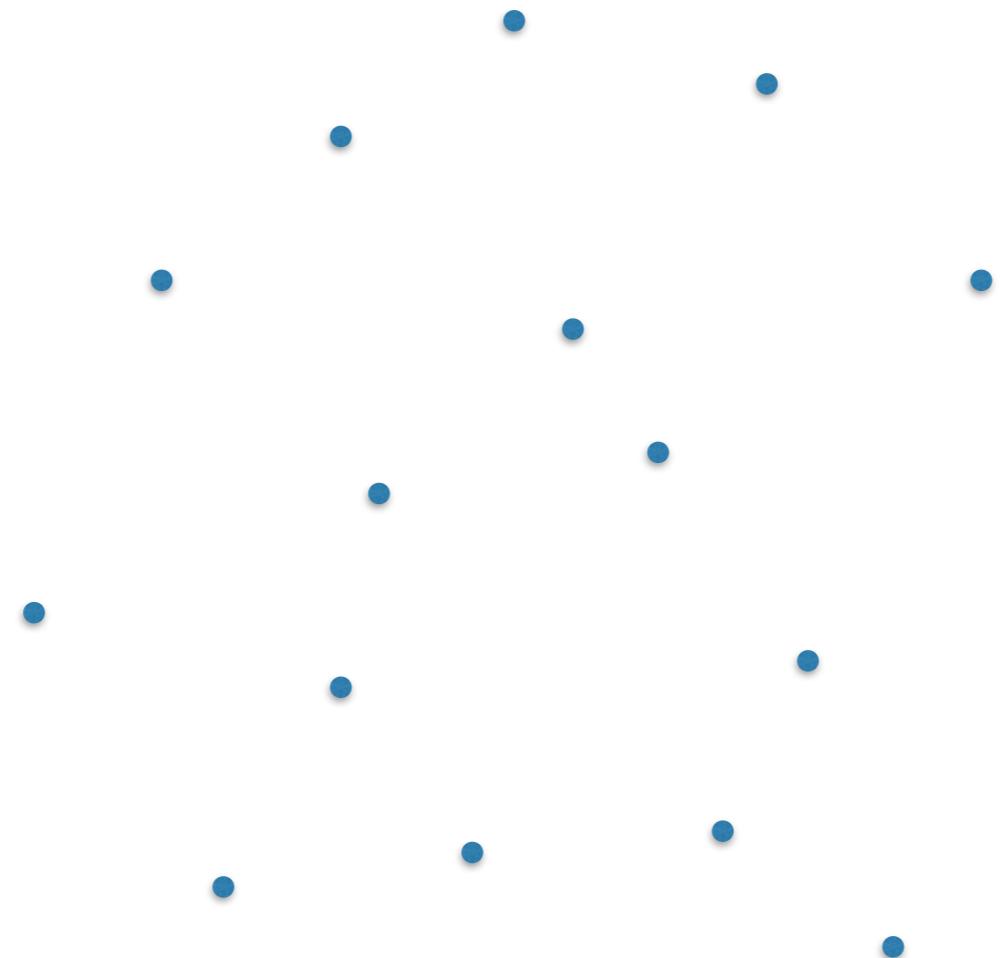




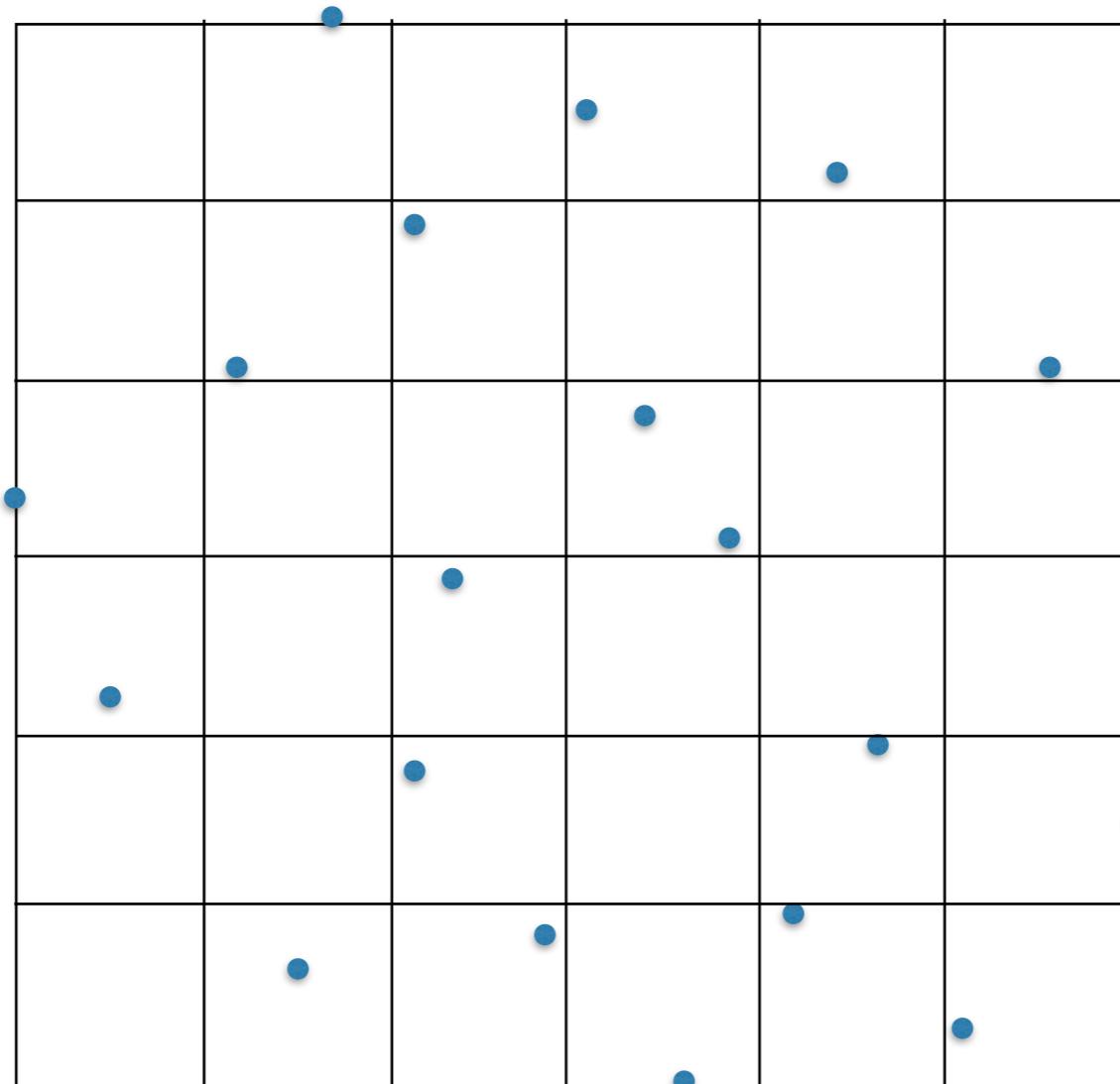
We'll partition the space, store it in a data structure and use it to speed up searching

Space decomposition methods

The simplest space decomposition is a grid



The grid method



For example...

```
class Grid {  
    double x1, x2, y1, y2;           // the bounding box of the grid  
    int m;                          // number of cells in the grid (m-by-m)  
    double cellsize_x, cellsize_y;   // size of a grid cell  
    List<point2D*> ***g;          //2D array of list*; g[i][j] contains the pointer  
                                  //to the list of points that lie in cell [i][j]  
  
    Grid (Point p[], int n, int m);  
    List<Point2D*>* rangeQuery(double x1, x2, y1, y2);  
    ....  
};
```

The grid method

- Creating a grid of m-by-m cells from a set of points P
 1. Figure out a rectangle that contains P (e.g. $x_{\min}, x_{\max}, y_{\min}, y_{\max}$)
 2. Allocate a 2d array of lists, all initially empty
 3. For each point p in P: figure out which cell i, j contains p, and insert p in the list corresponding to $g[i][j]$

The grid method

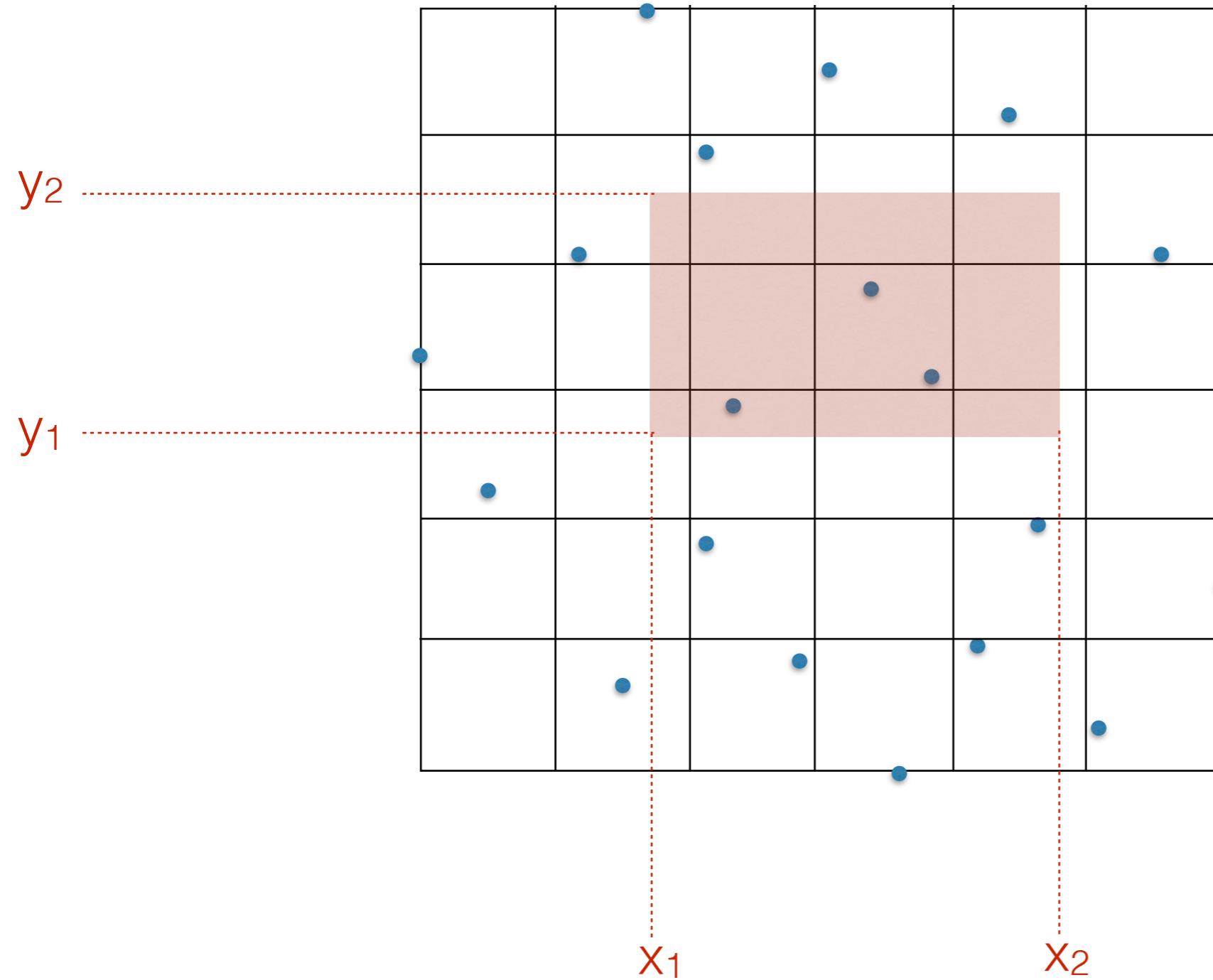
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```
g = new (List<point2D*>**)[m];
for (int i=0; i<m; i++) {
    g[i] = new (List<point2D*>*) [m];
    for (int j=0; j<m; j++) {
        g[i][j] = new List<point2D*>;
    }
}
```

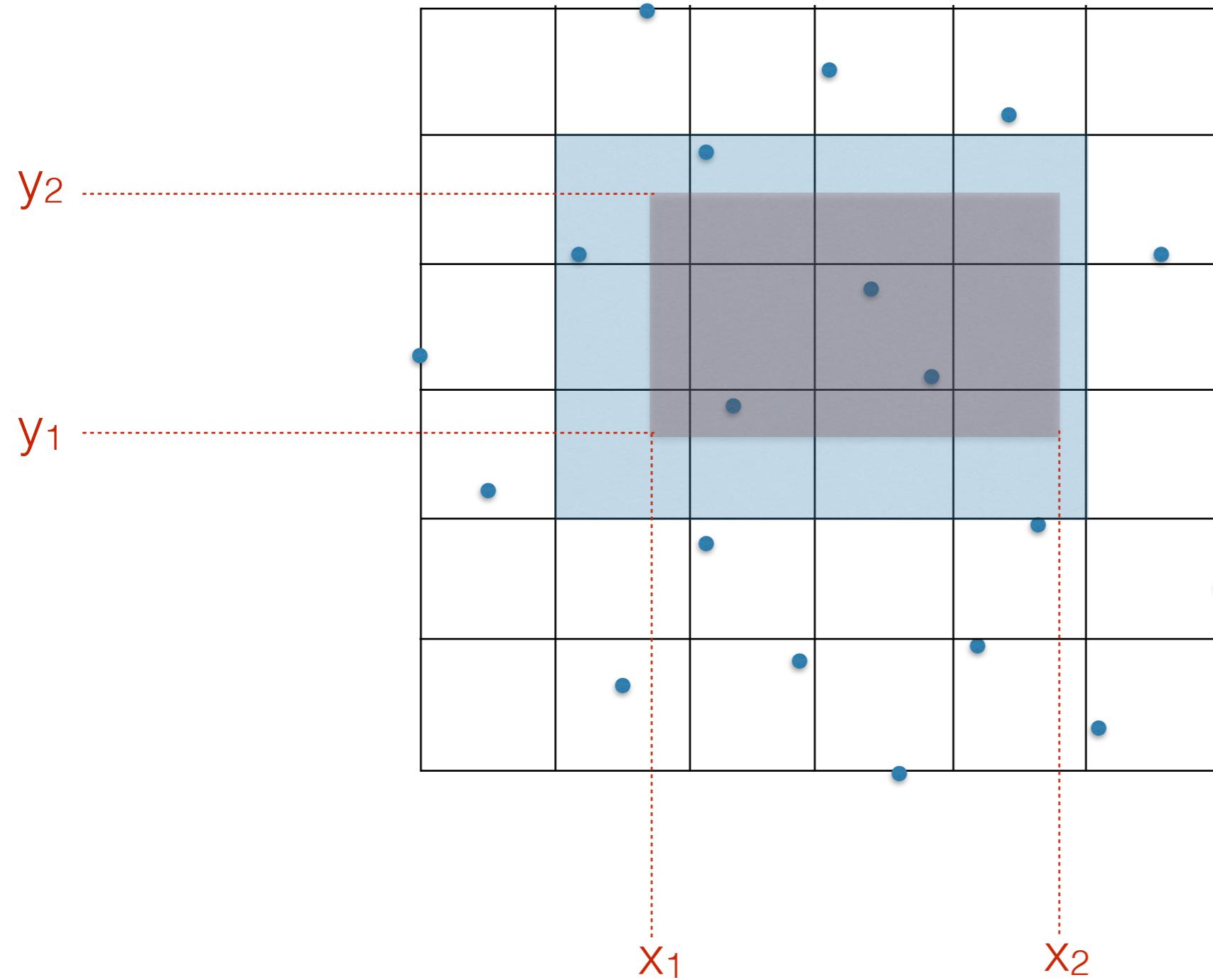
for each point p

```
j = (p.x - x_min)/cellsize_x;
i = (y_max - p.y/cellsize_y;
g[i][j]->insert(&p);
```

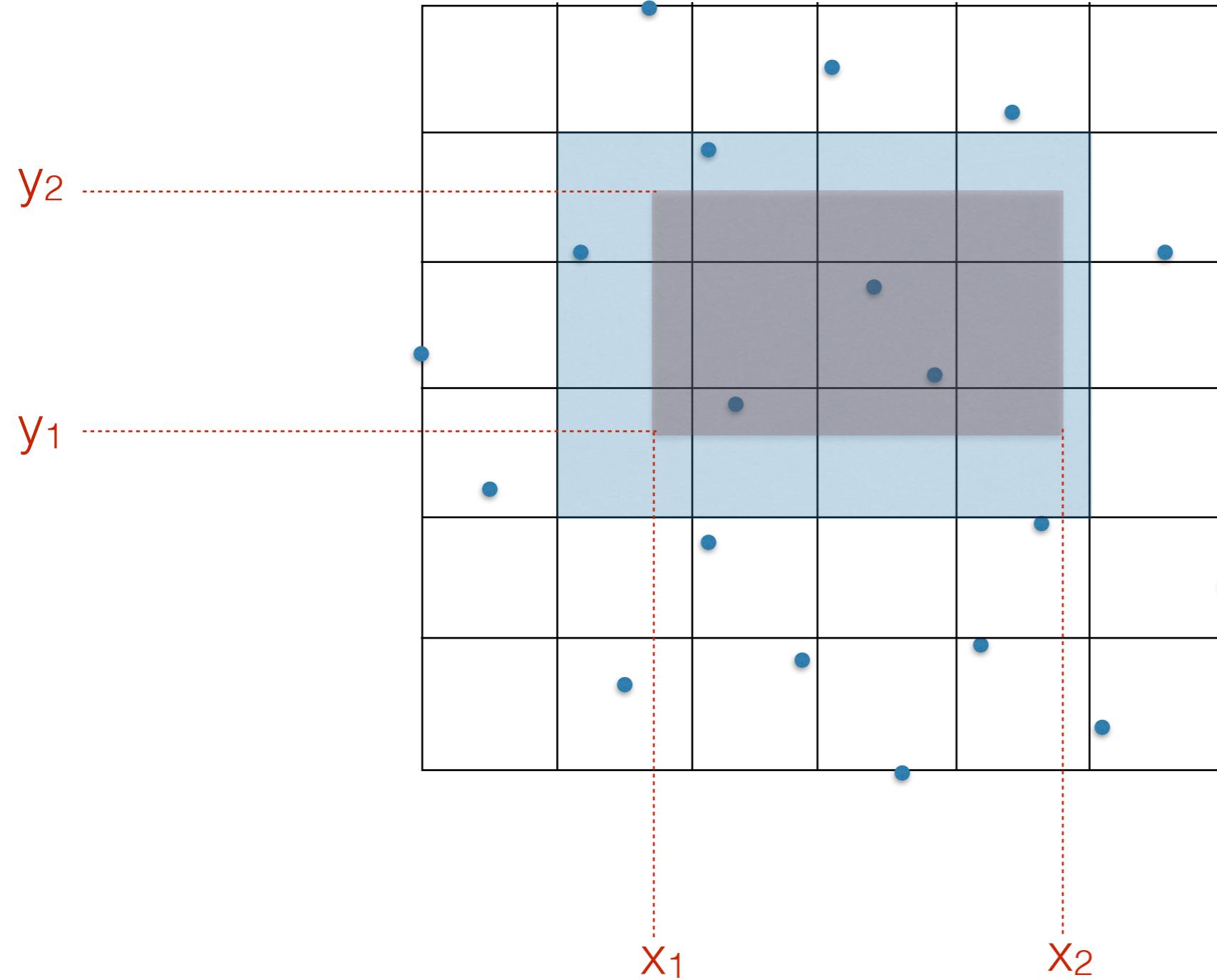
How do we answer range searches with a grid?



How do we answer range searches with a grid?



How do we answer range searches with a grid?



Analysis

- How long does a range query take ?
- How many points in a cell?
- How do the points look like for the worst-case to be good?
- How to chose m?

The grid method

- + Grids perform well if points are uniformly distributed
- + Grids can be used as heuristic for many other problems besides range searching (e.g. closest pair, neighbor queries)
- - No worst case guarantees
- + simple to implement

We've seen this!

2d search trees

3d search trees

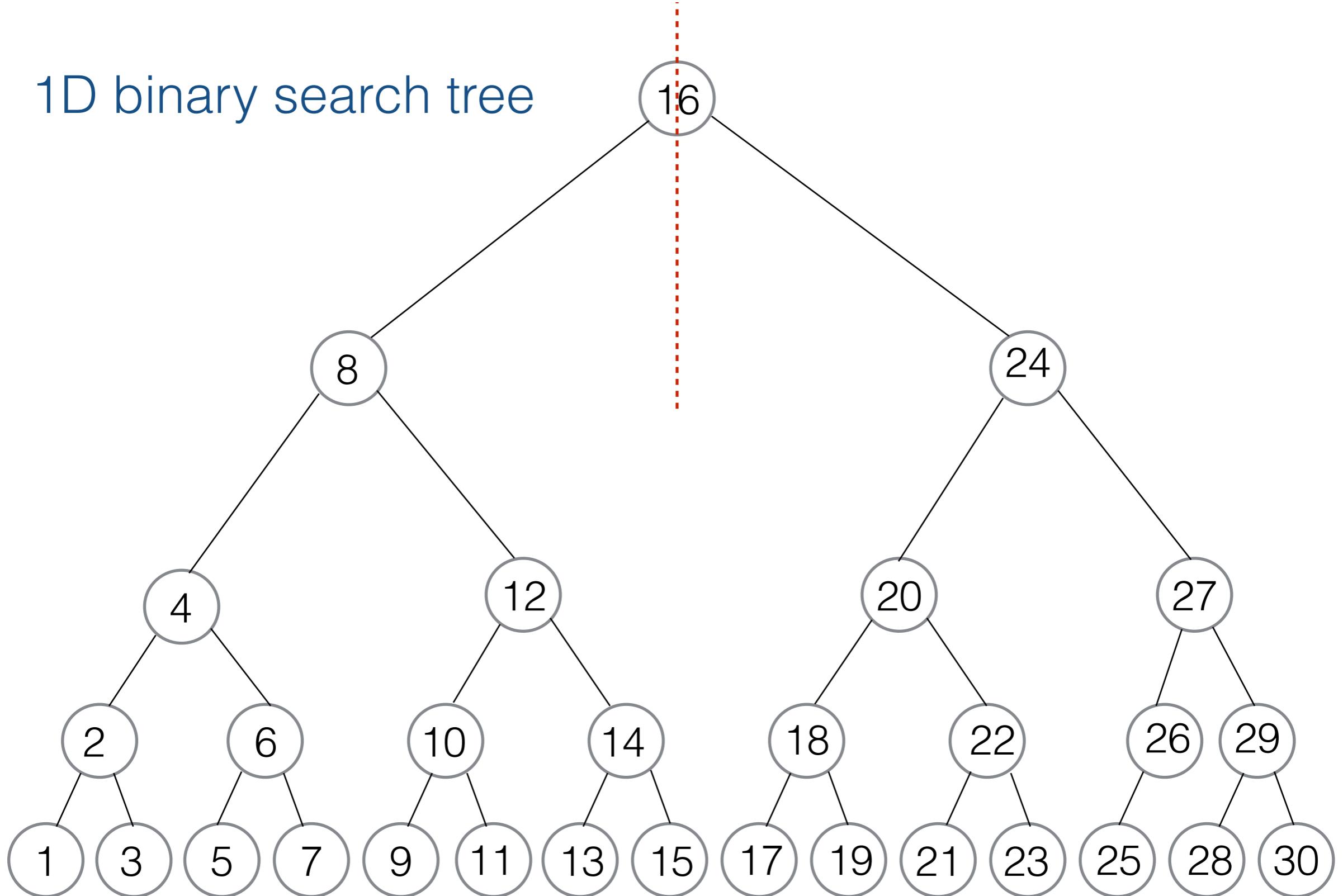
4d search trees

...

k-dimensional search trees

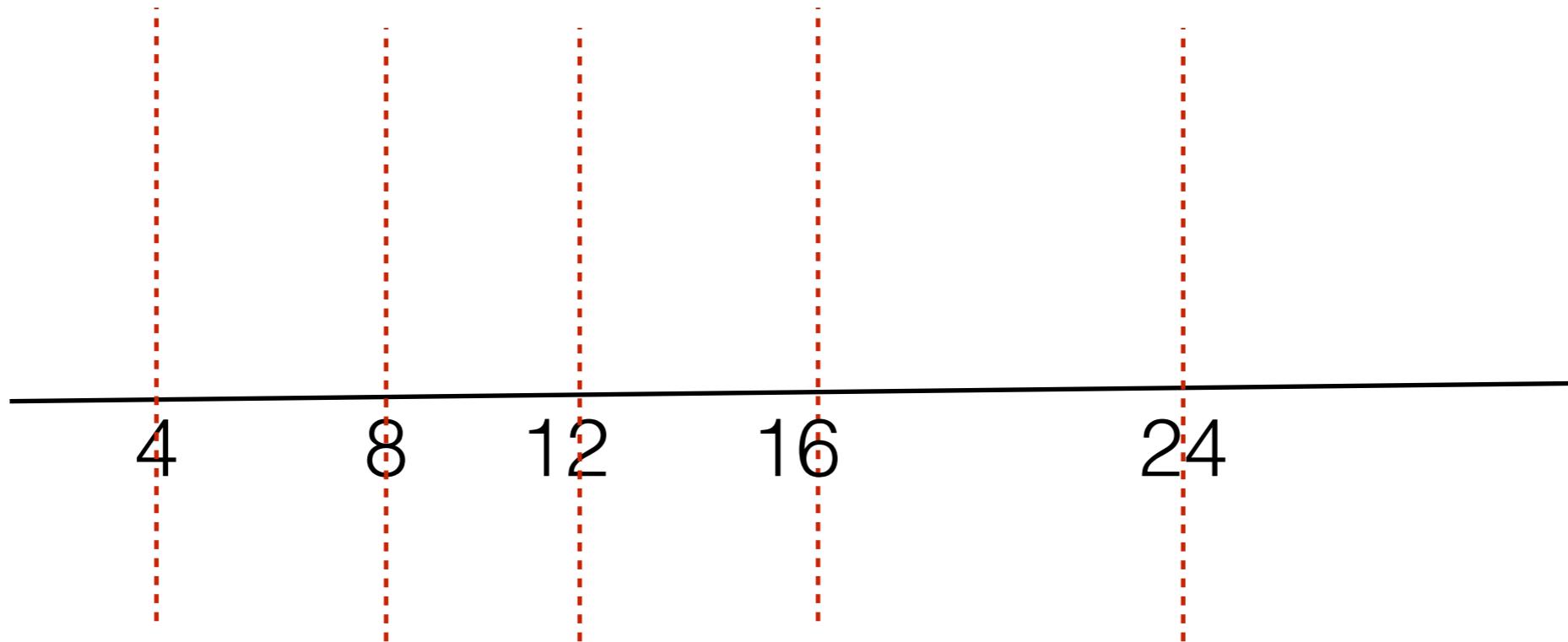
kd trees

1D binary search tree



a space decomposition: left tree represents all values ≤ 16 ; right tree represents all values > 16 ; and so on

1D binary search tree

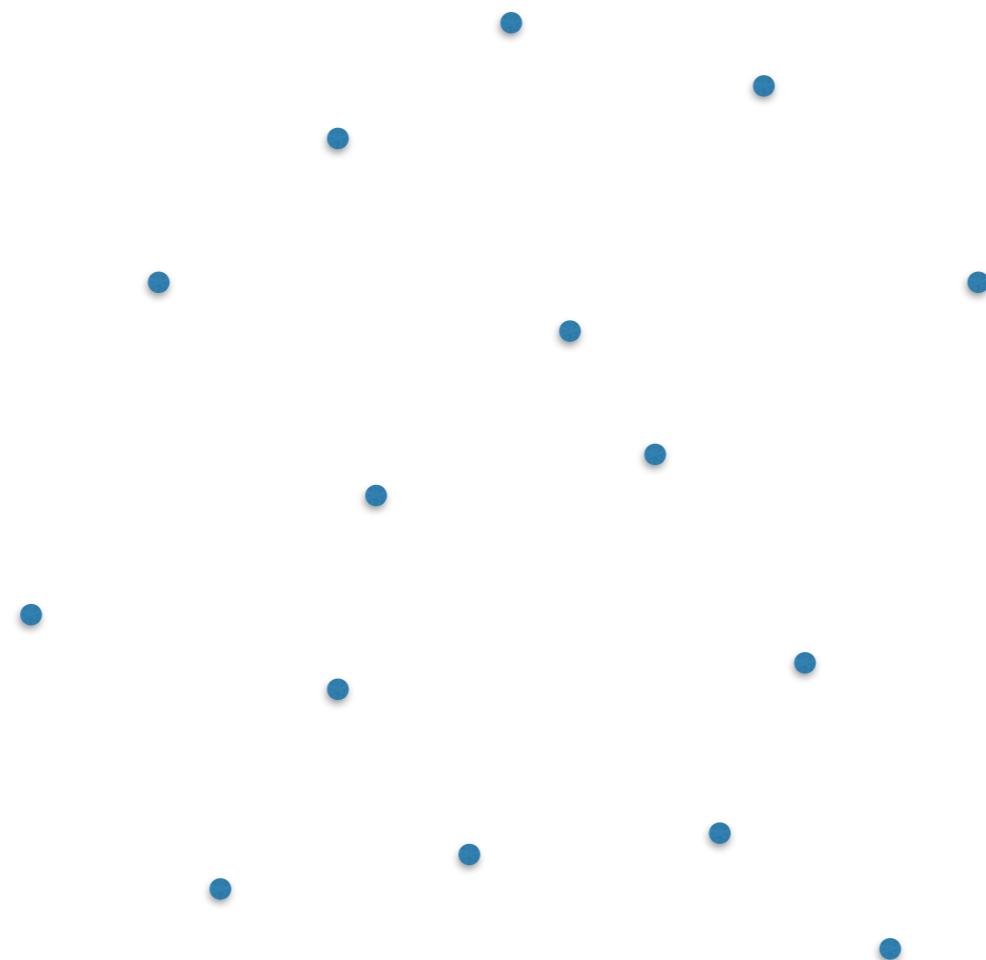


- to search for a value, find the region of space where it would be if it were in the input

2d binary search trees

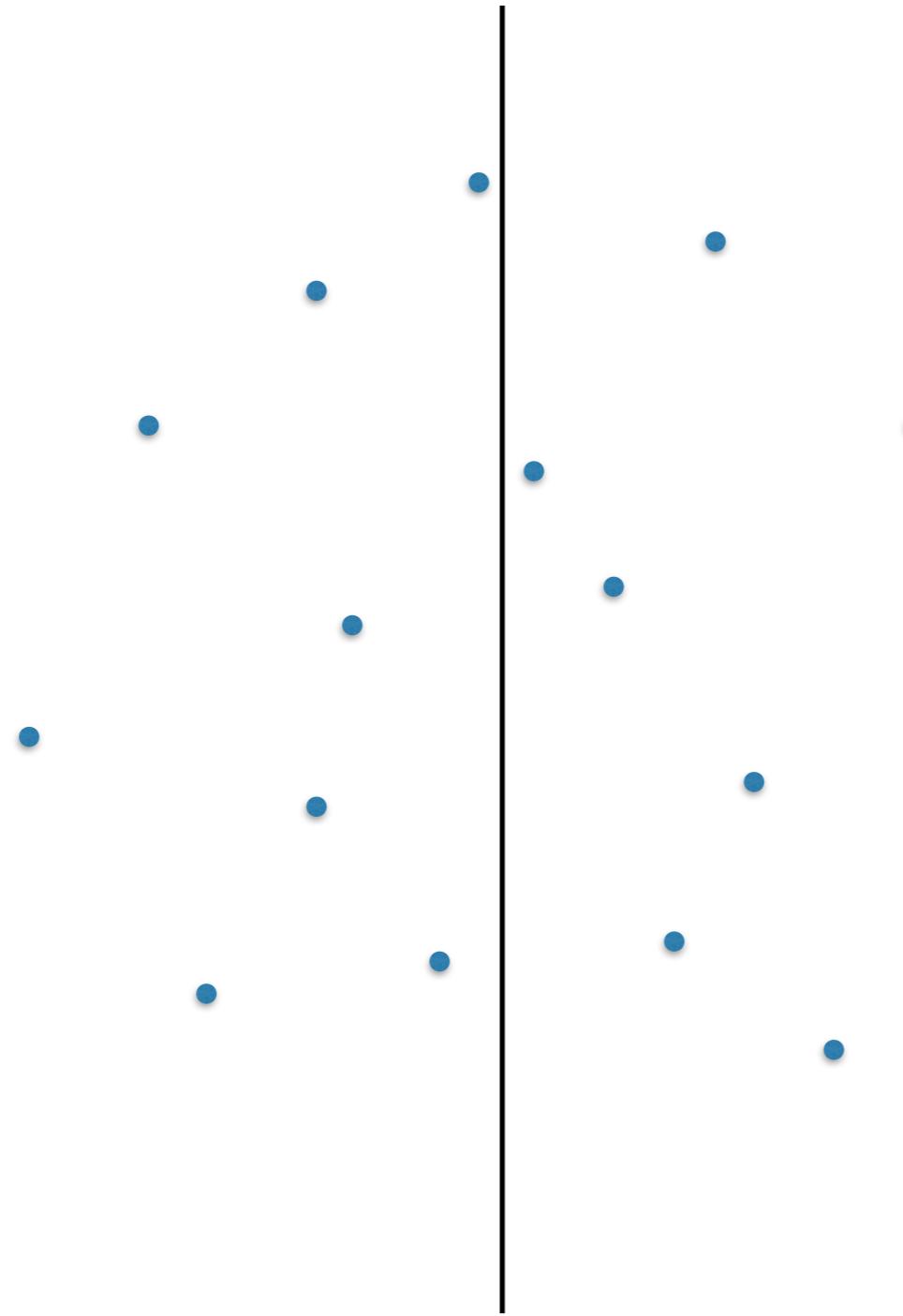
- **The idea:** A binary tree which recursively subdivides the plane by vertical and horizontal cut lines
- Vertical and horizontal lines alternate
- Cut lines are chosen to split the points in two (=> logarithmic height)

2d binary search trees



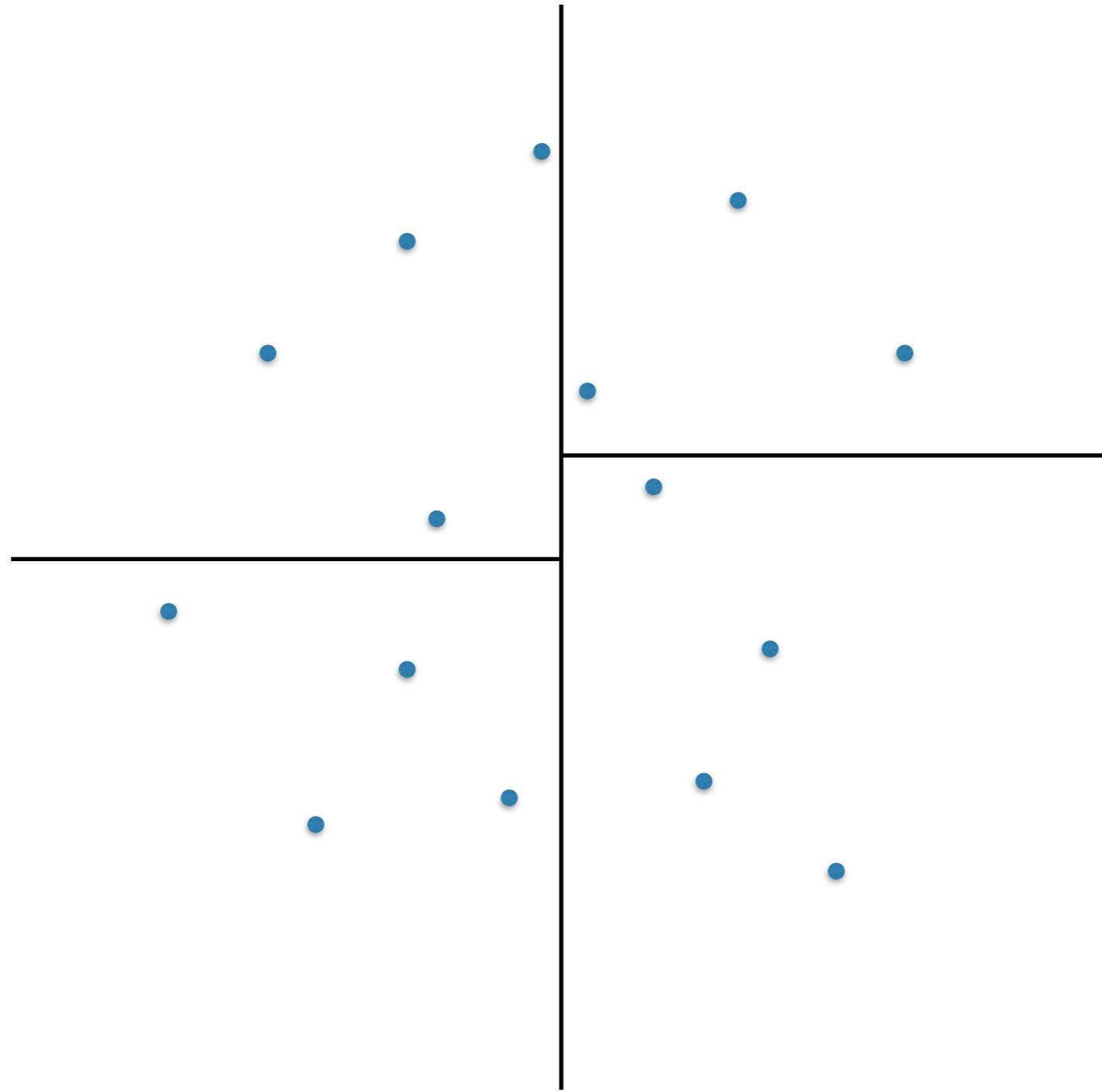
2d binary search trees

split points in two halves with a vertical line

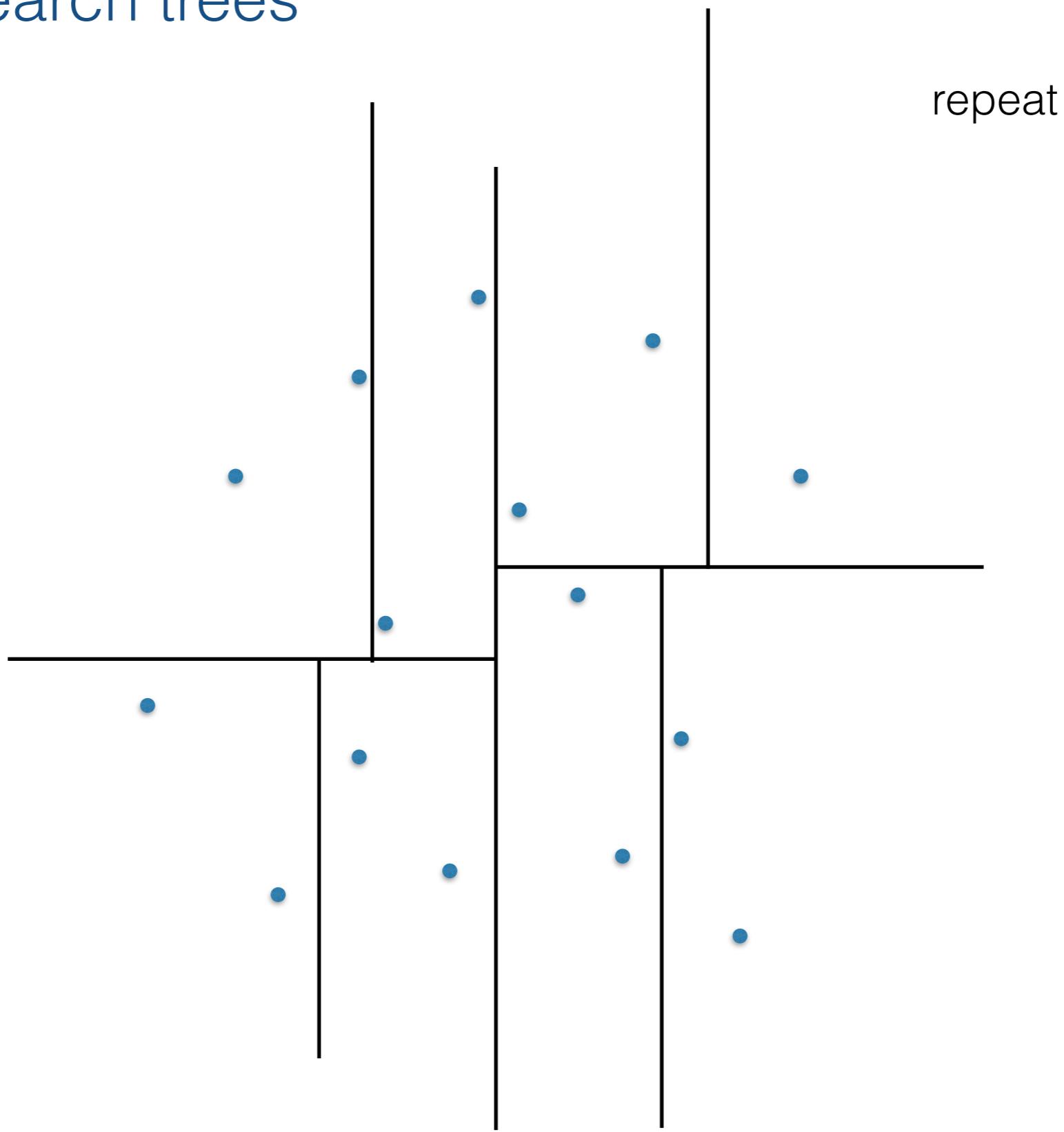


2d binary search trees

split each side into half with a horizontal line

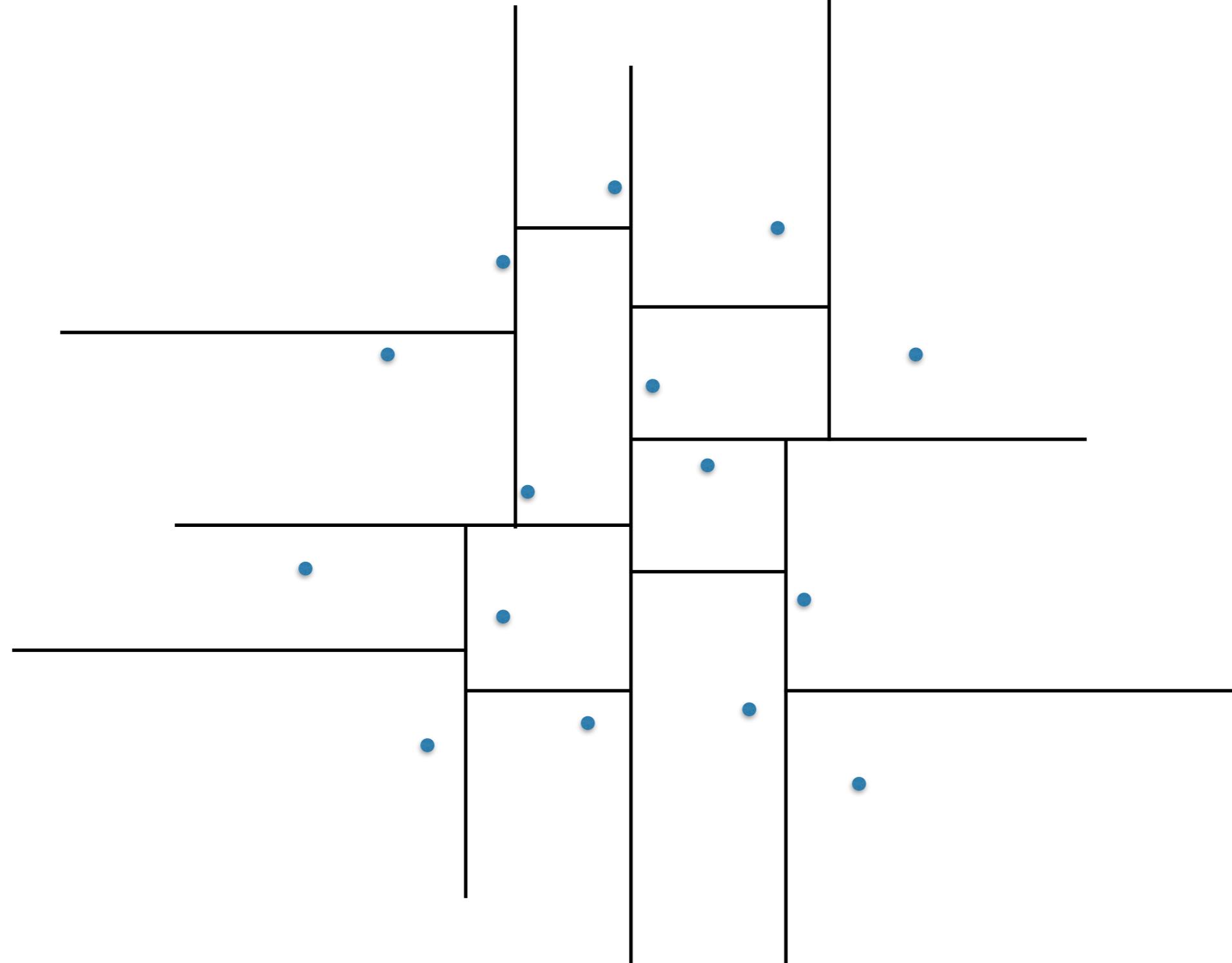


2d binary search trees



2d binary search trees

How to find a line that splits the points in half?



2d binary search trees

How to find a line that splits the points in half?

Variants:

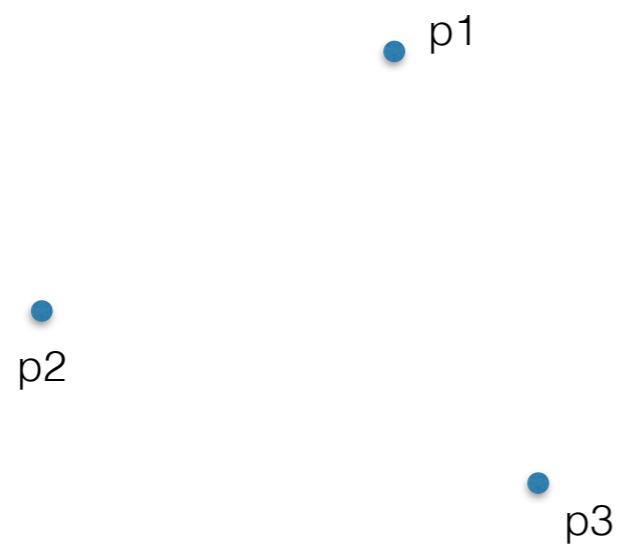
- Choose the cut line so that it goes through the median point, and store the median in the internal node.
- Choose the cut line so that it falls in between the points. Internal nodes store lines, and points are only in leaves.
- Choose the cut line so it goes through the median point. Internal nodes store lines, and points are only in leaves.
 - if n is even, assign the median to the e.g. smaller (left/below) one, consistently



This is standard and simplifies the details

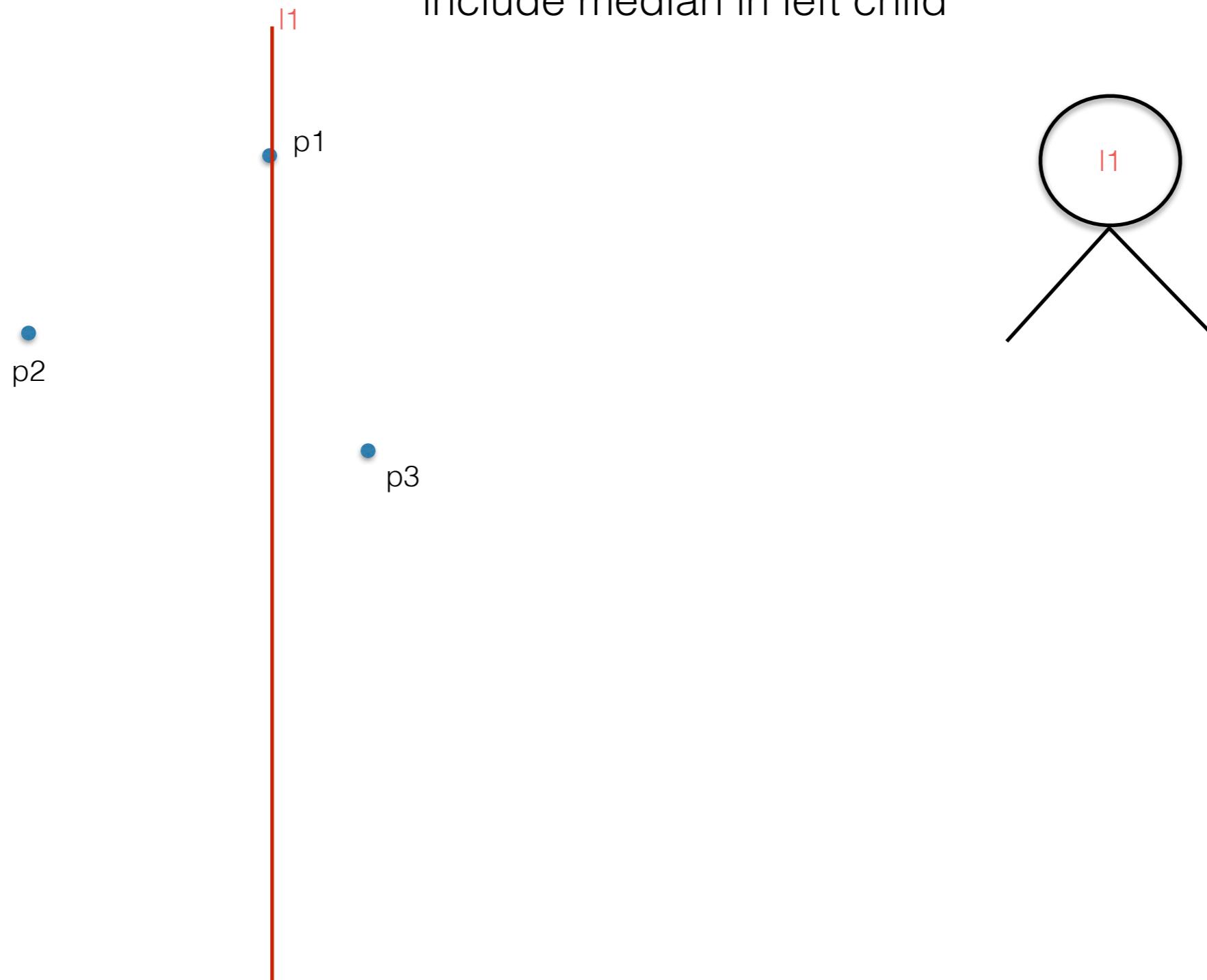
Let's see what this means

2d binary search trees

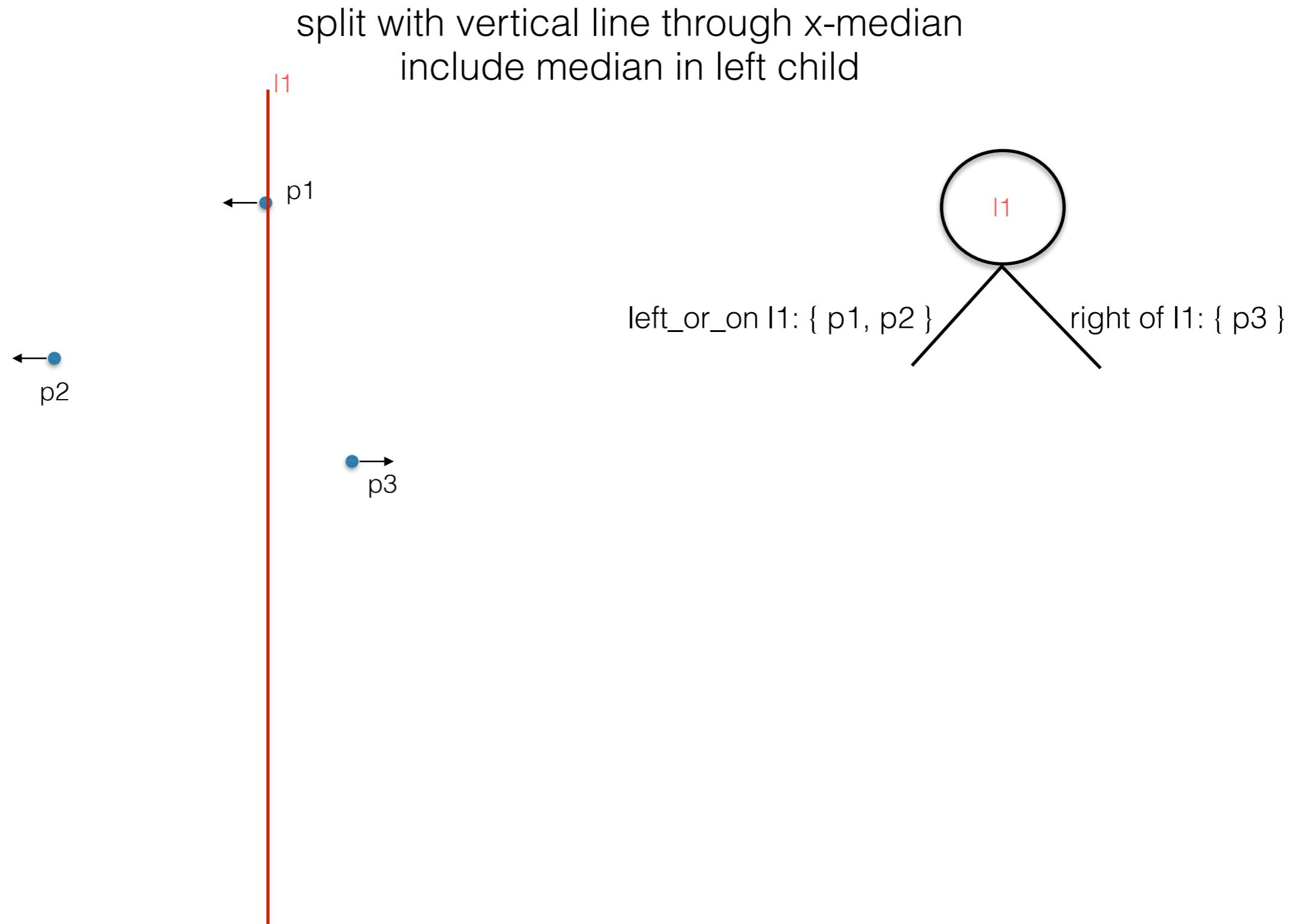


2d binary search trees

split with vertical line through x-median
include median in left child

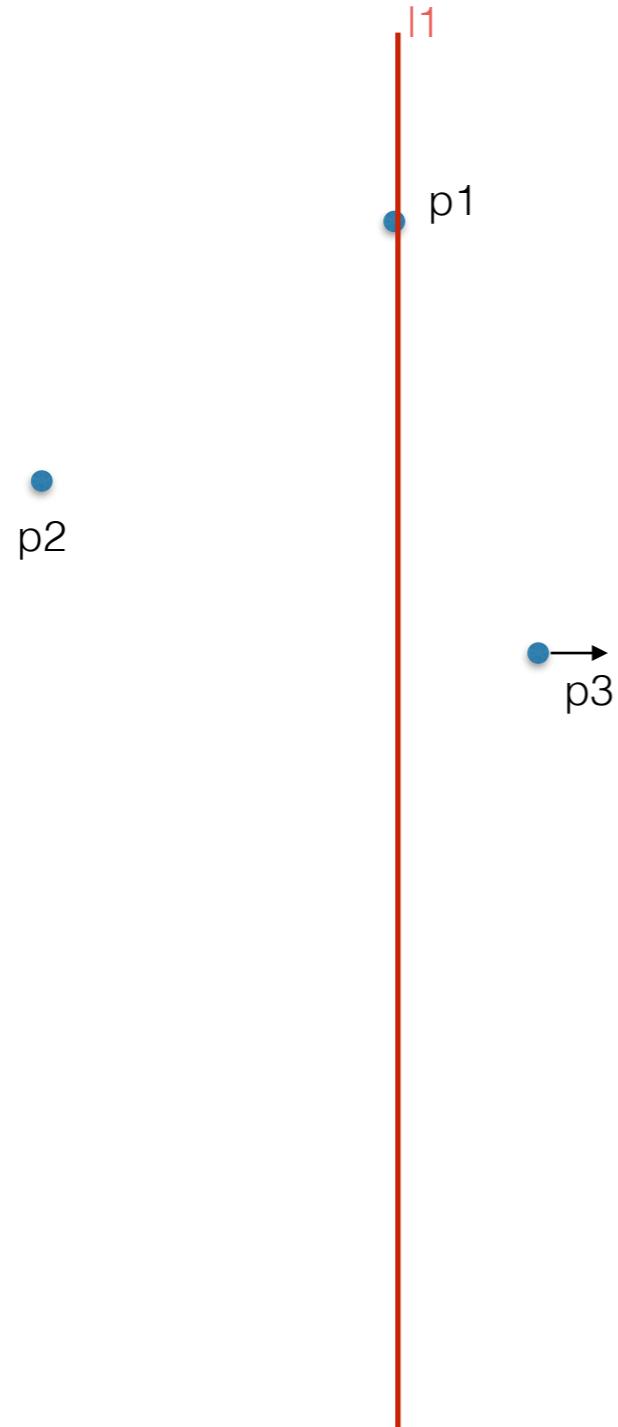


2d binary search trees

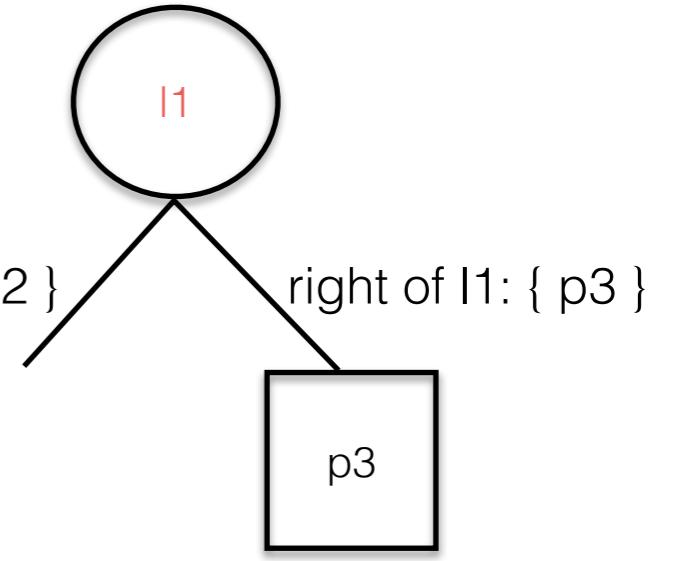


2d binary search trees

right of l1: p3 => leaf

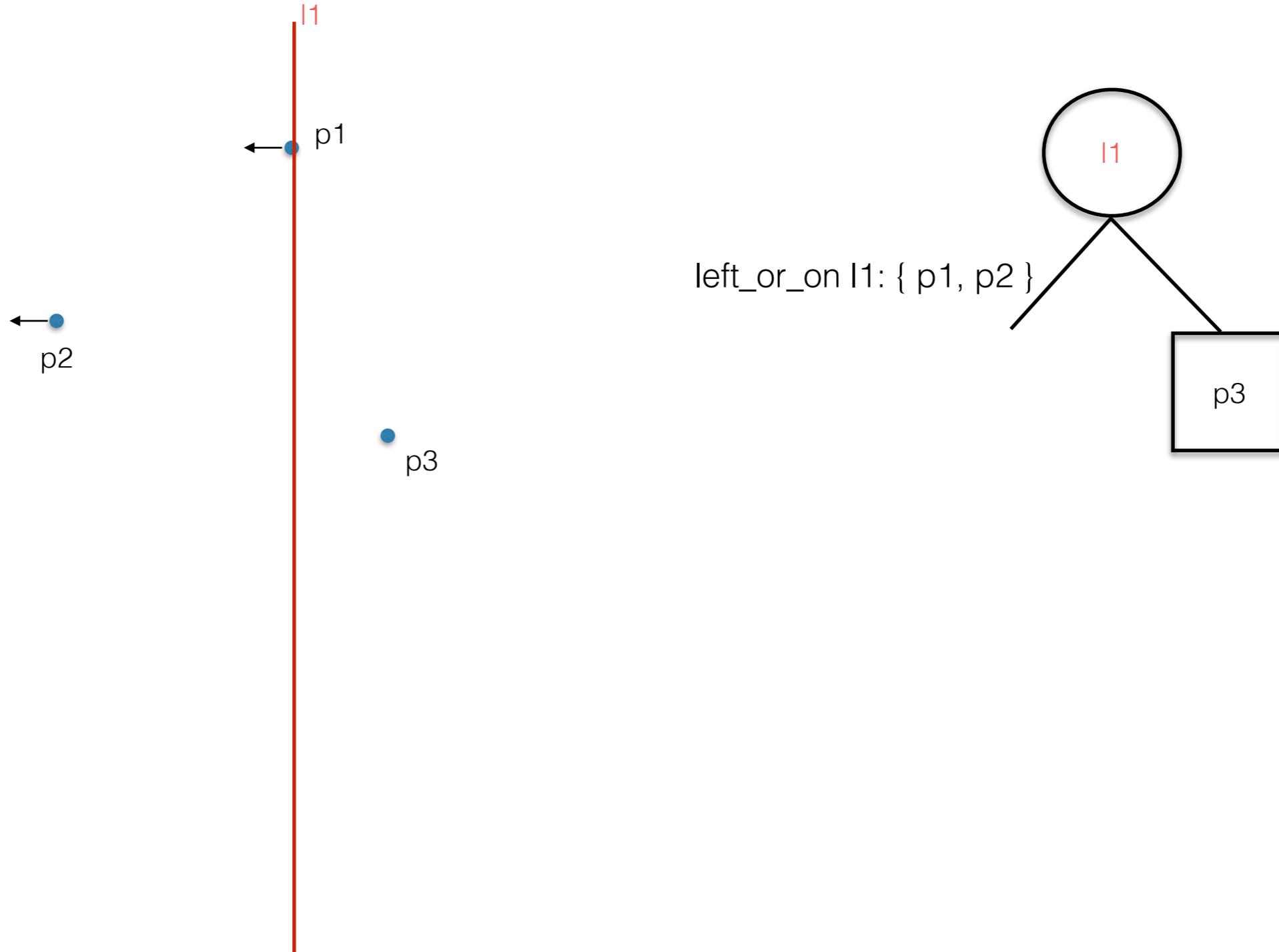


left_or_on l1: { p1, p2 }



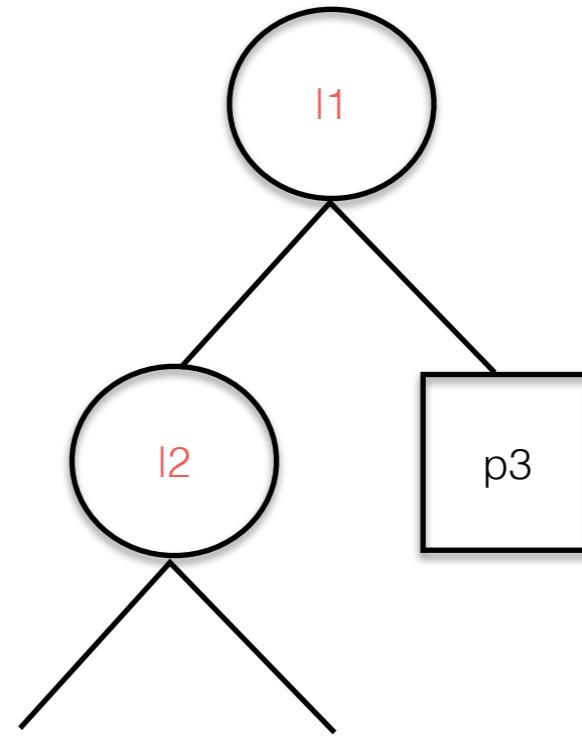
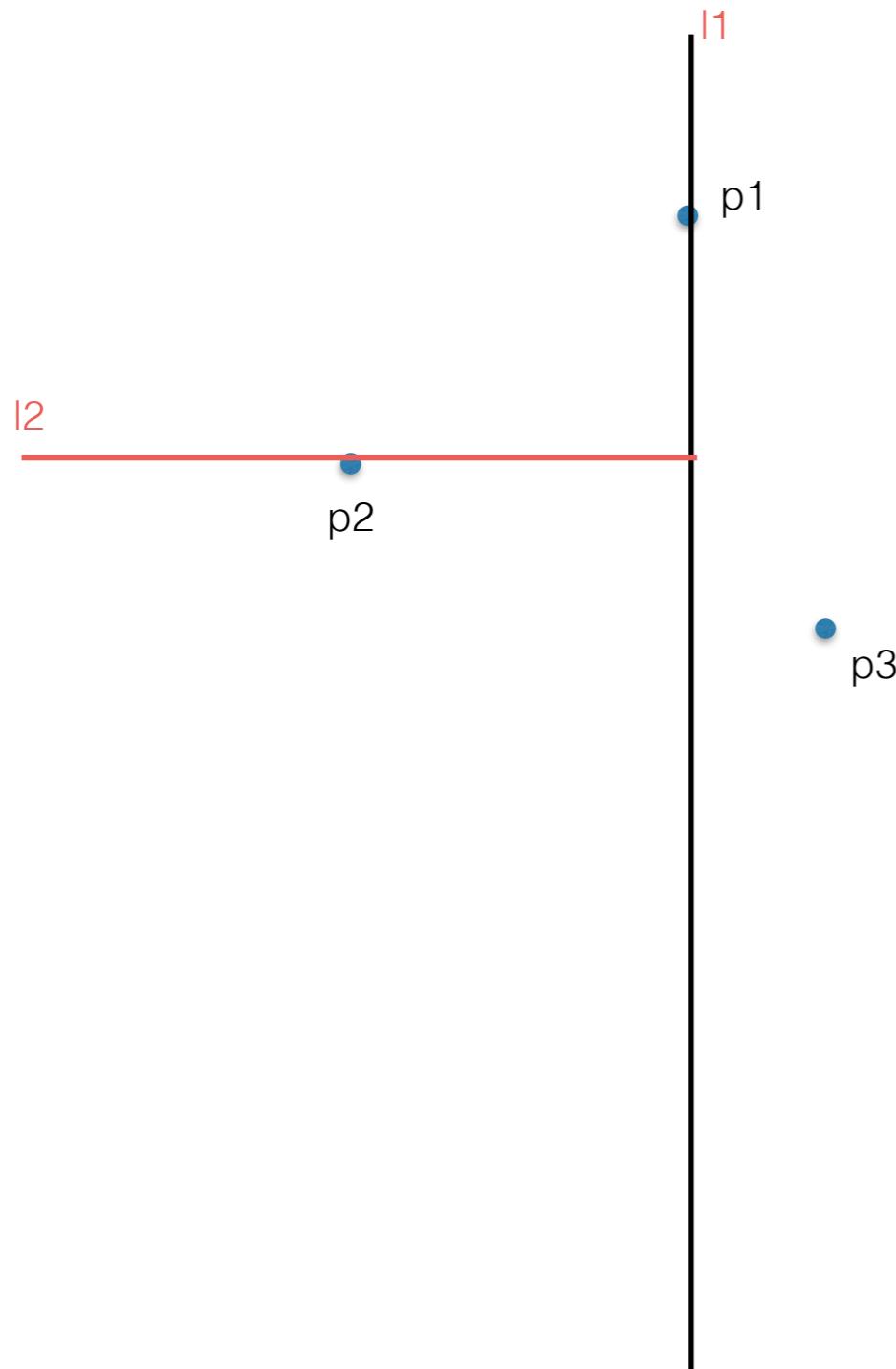
2d binary search trees

left_on l1: p1,p2 => recurse

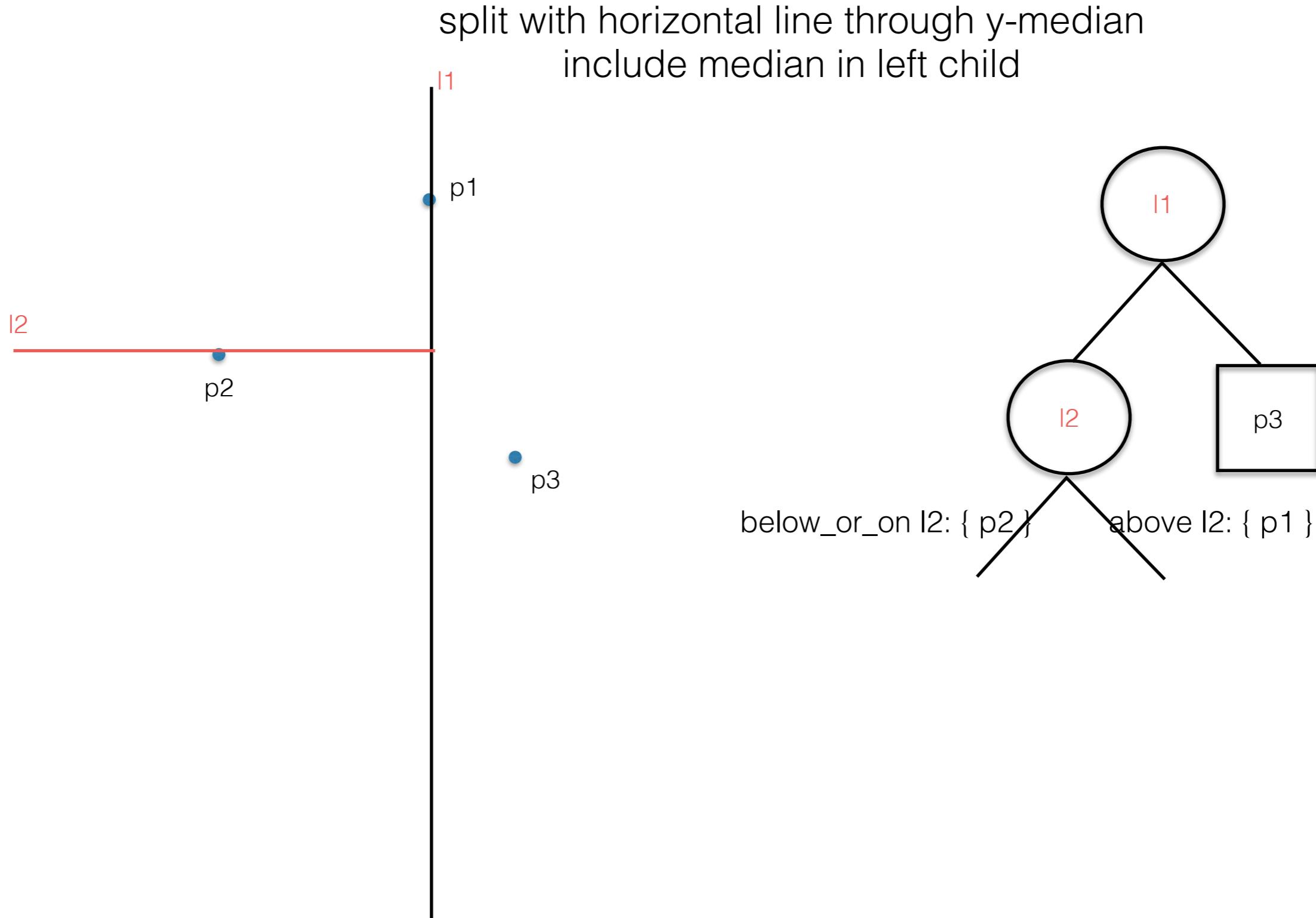


2d binary search trees

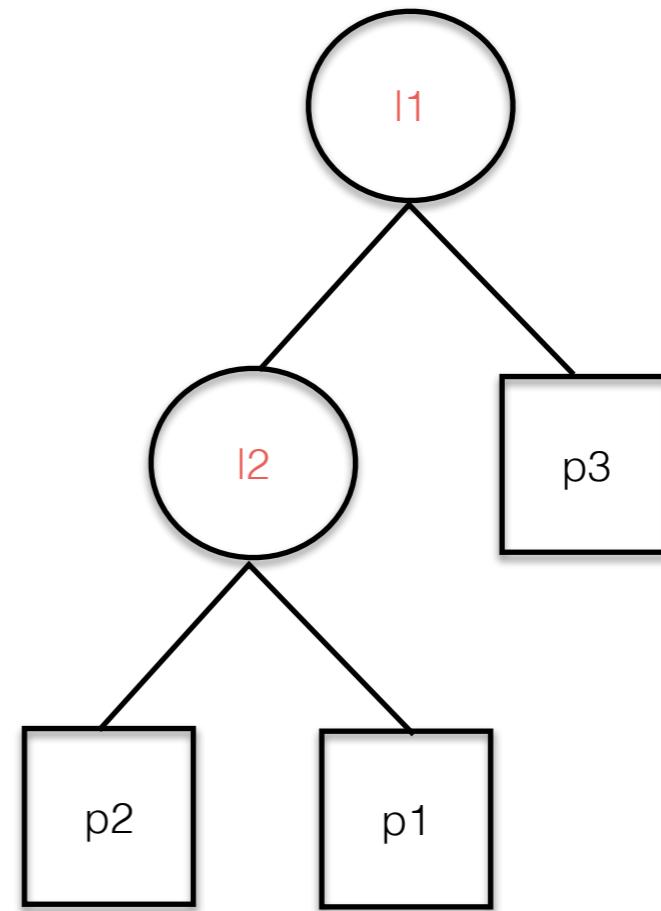
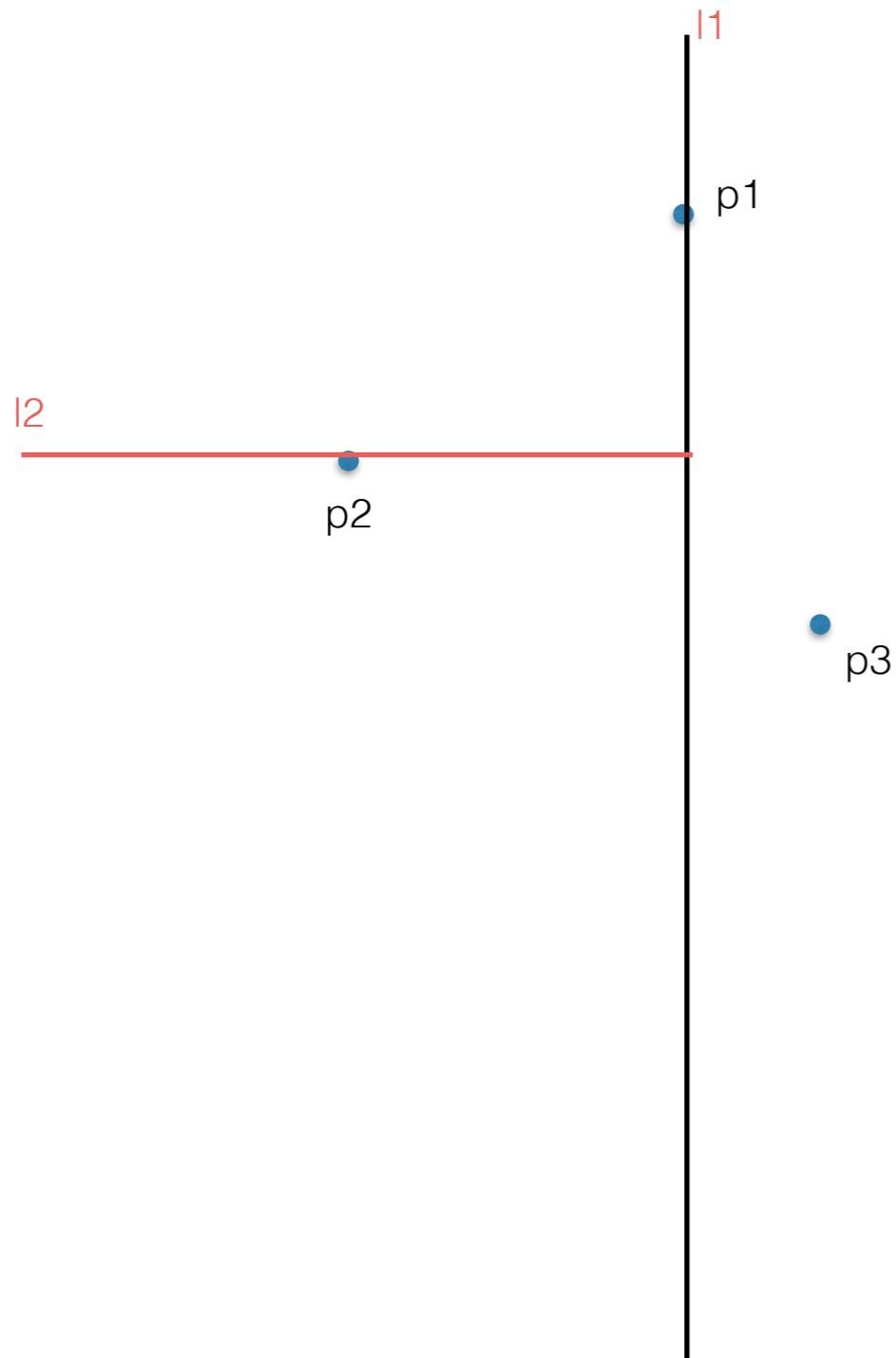
split with horizontal line through y-median
include median in left child



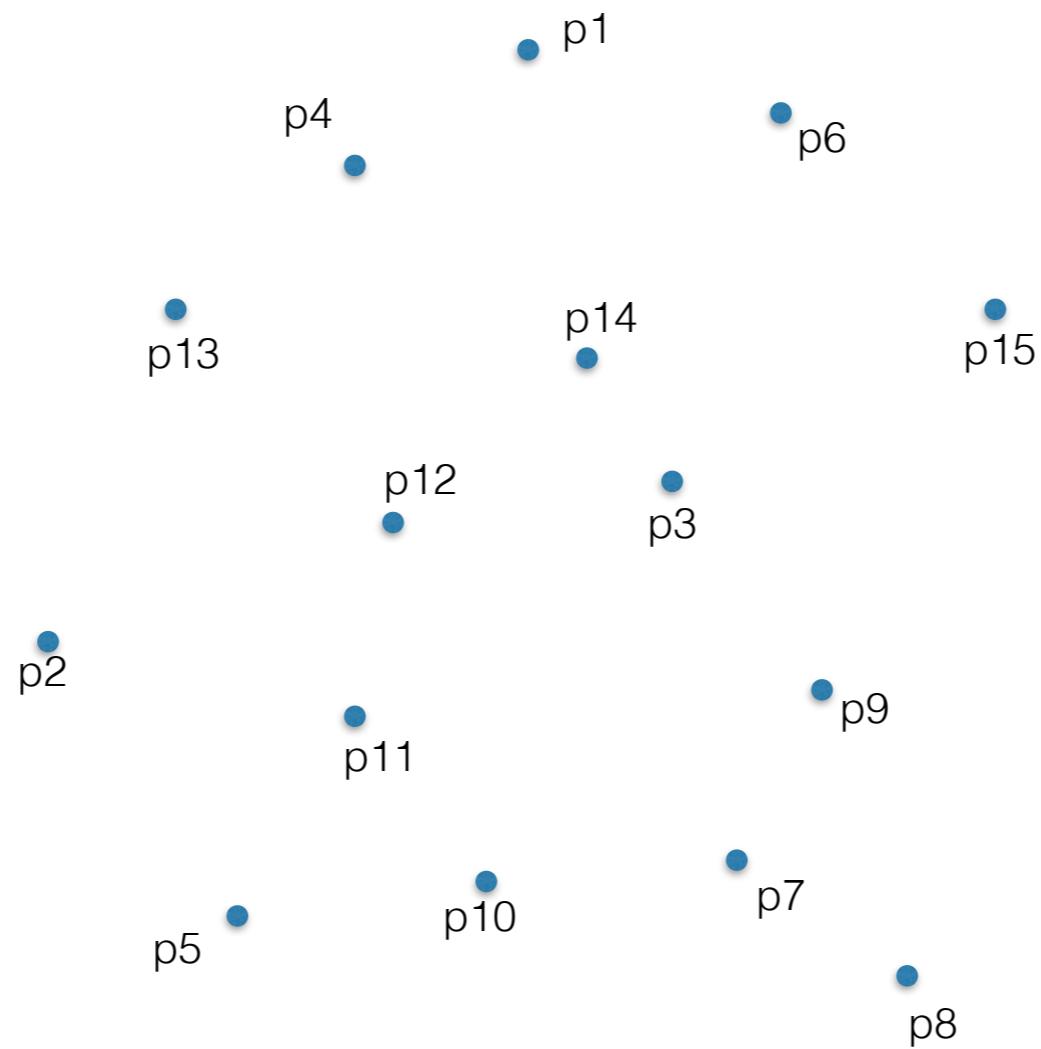
2d binary search trees



2d binary search trees

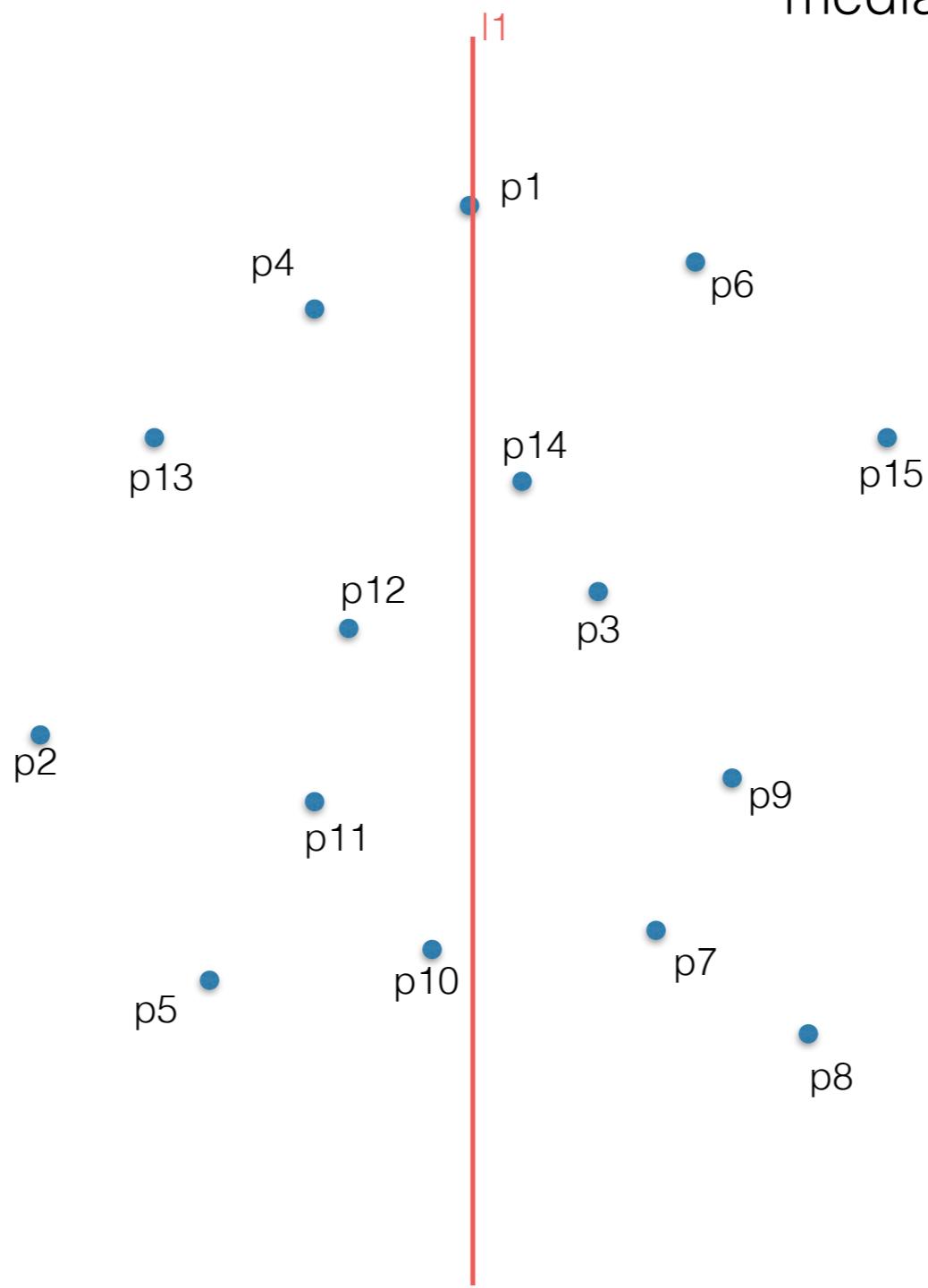


A bigger example



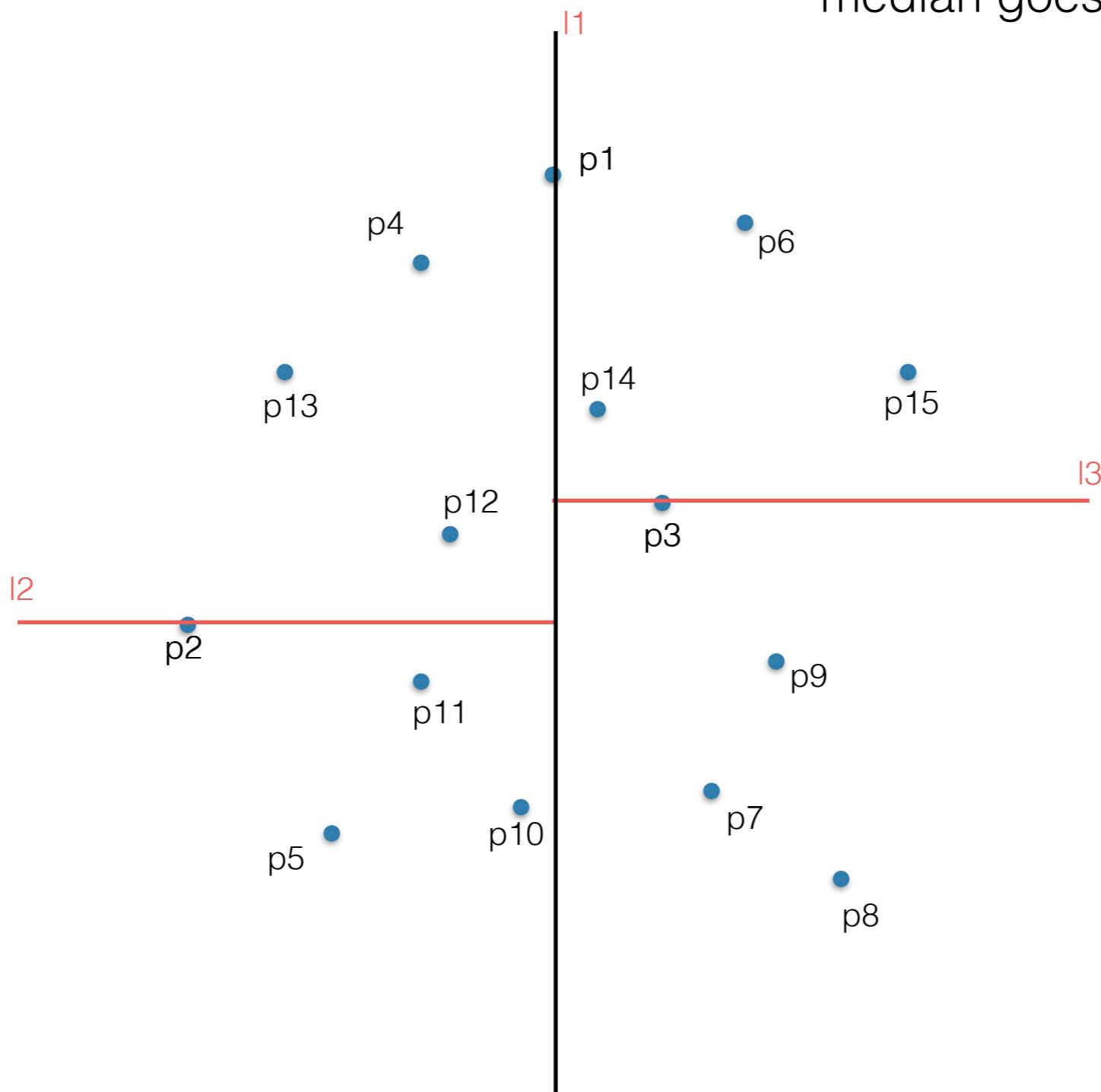
A bigger example

split with vertical line through x-median
median goes to the left side

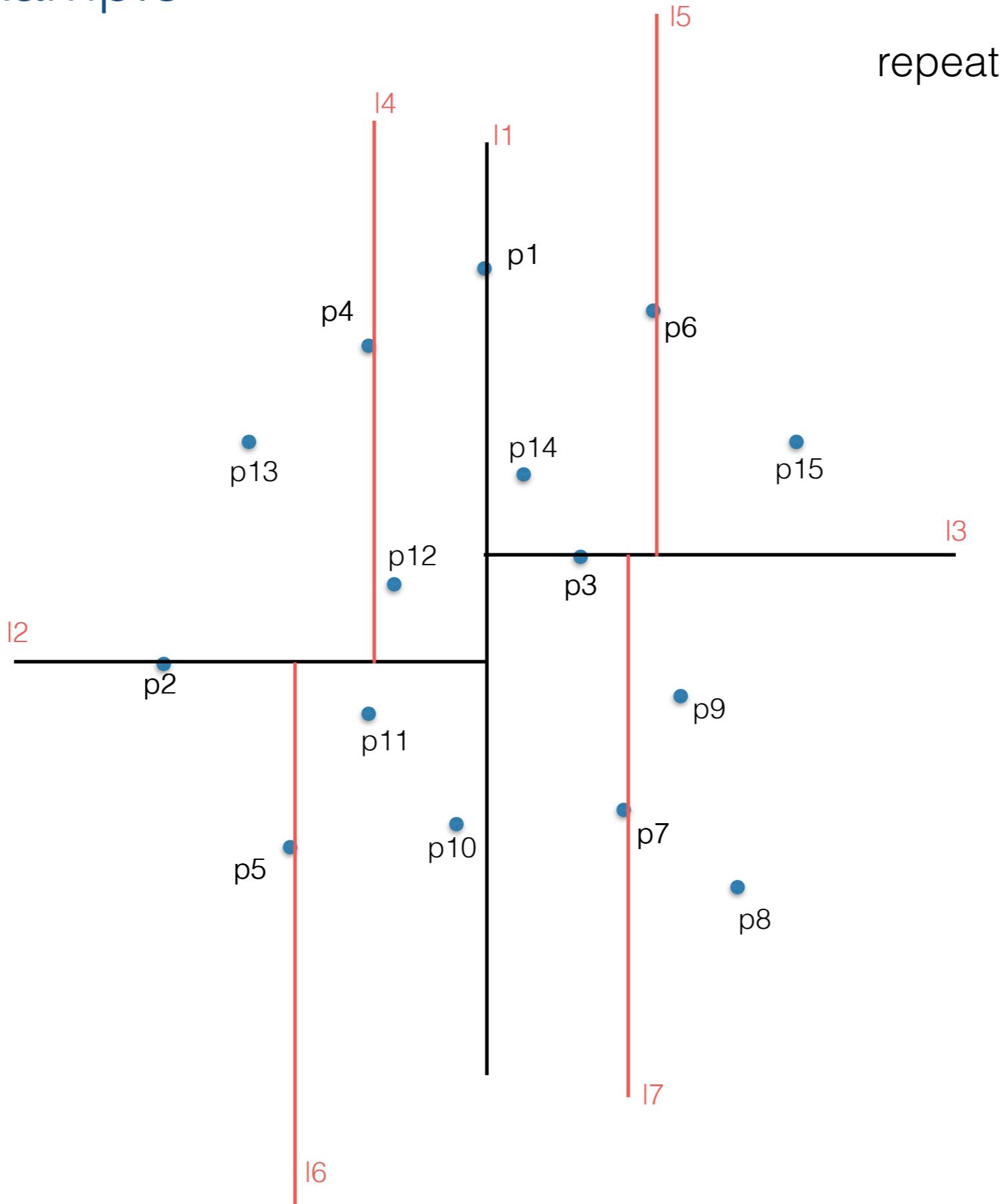


A bigger example

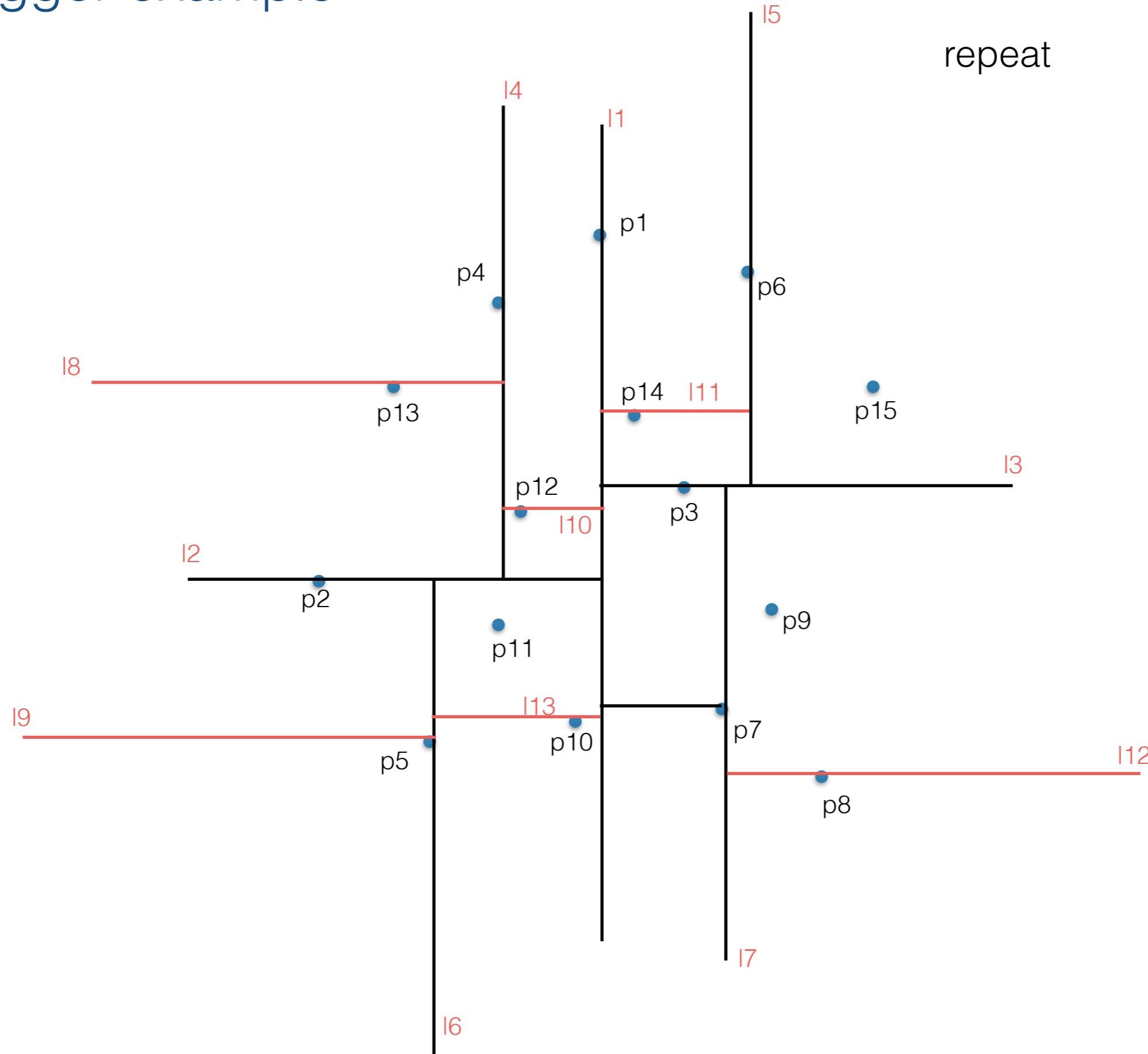
split each side with horizontal line through y-median
median goes to the left side



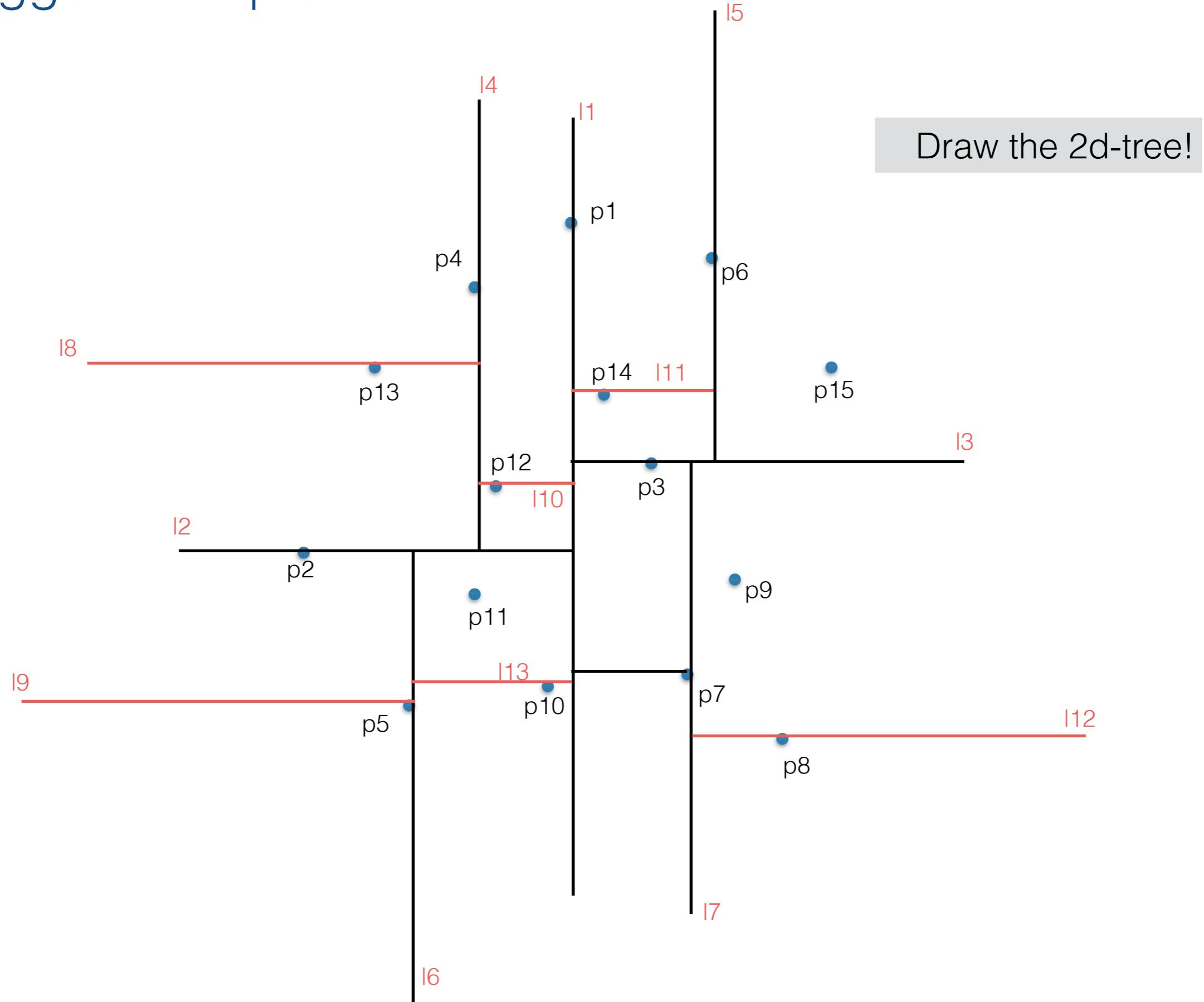
A bigger example



A bigger example



A bigger example



2d binary search trees

Analysis

1. How to build it and how fast?
2. How much space does it take?
3. How to answer range queries and how fast?

2d binary search trees construction

Algorithm BUILDKDTREE($P, depth$)

1. **if** P contains only one point
2. **then return** a leaf storing this point
3. **else if** $depth$ is even
4. **then** Split P with a vertical line ℓ through the median x -coordinate into P_1 (left of or on ℓ) and P_2 (right of ℓ)
5. **else** Split P with a horizontal line ℓ through the median y -coordinate into P_1 (below or on ℓ) and P_2 (above ℓ)
6. $v_{\text{left}} \leftarrow \text{BUILDKDTREE}(P_1, depth + 1)$
7. $v_{\text{right}} \leftarrow \text{BUILDKDTREE}(P_2, depth + 1)$
8. Create a node v storing ℓ , make v_{left} the left child of v , and make v_{right} the right child of v .
9. **return** v

2d binary search trees construction

Analysis?

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9. **return** v

2d binary search trees construction

1. How to build it and how fast?

- Let $T(n)$ be the time needed to build a 2d tree of n points
- Then

$$T(n) = 2T(n/2) + O(n)$$

- This solves to **$O(n \lg n)$**
- Practical notes
 - The $O(n)$ median finding algorithm is not practical. Either do a randomized median finding (QuickSelect); or
 - Better: pre-sort P on x- and y-coord and pass them along as argument, and maintain the sorted sets through recursion

P_1 -sorted-by-x, P_1 -sorted-by-y

P_2 -sorted-by-x, P_2 -sorted-by-y

2d binary search trees

2. How much space does it take?

2d binary search trees

2. How much space does it take?

$O(n)$

2d binary search trees

3. How to answer range queries?

Let's work through an example to get the intuition.

Range queries

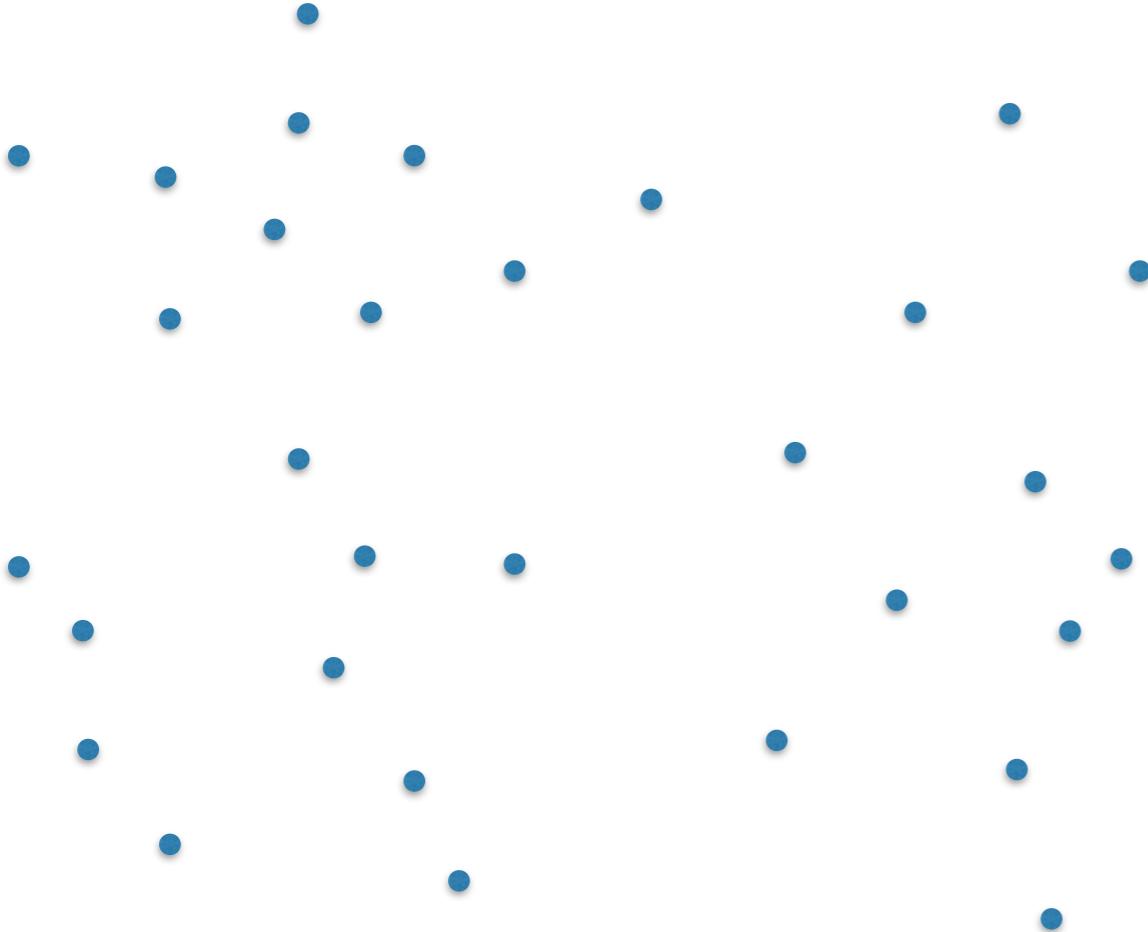
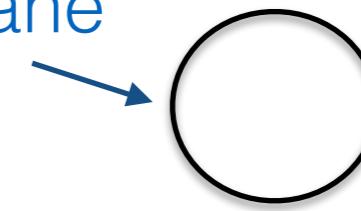
We'll use that:

A 2d-tree defines a **hierarchical** partition of the space, where each node in the tree represents a region of space.

Let's see what this means..

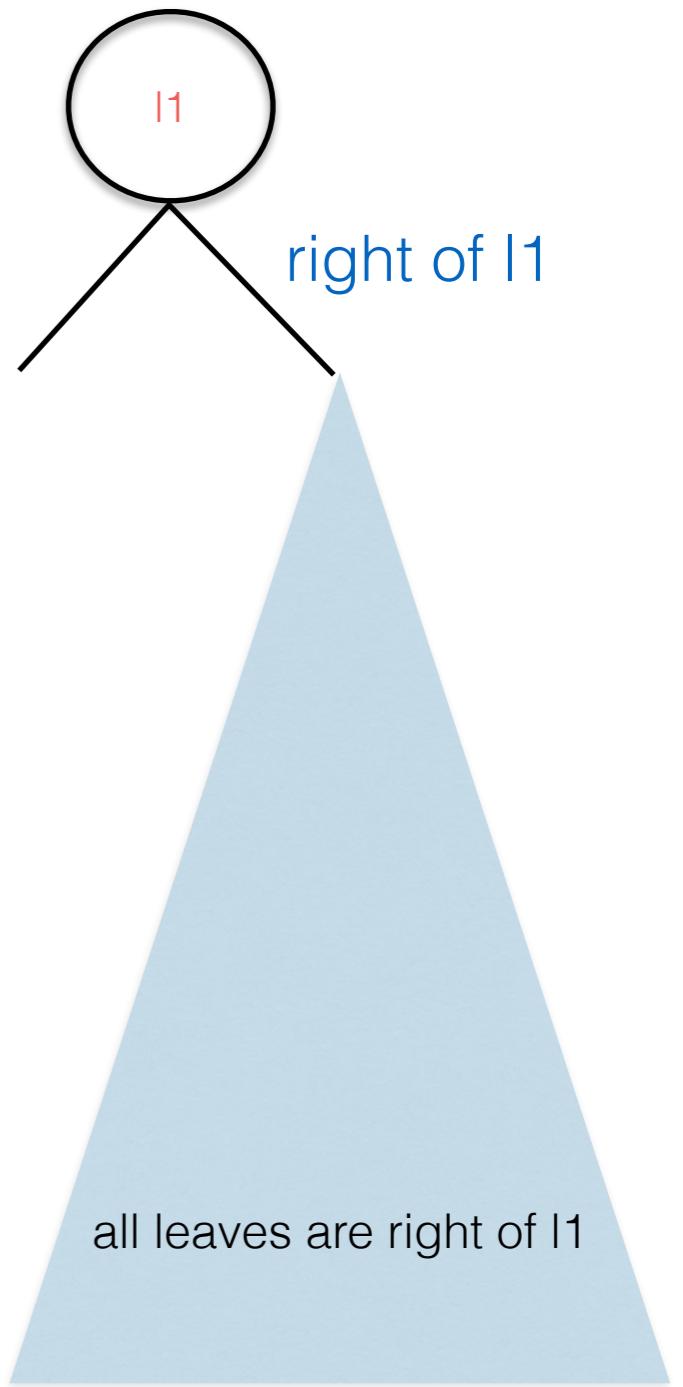
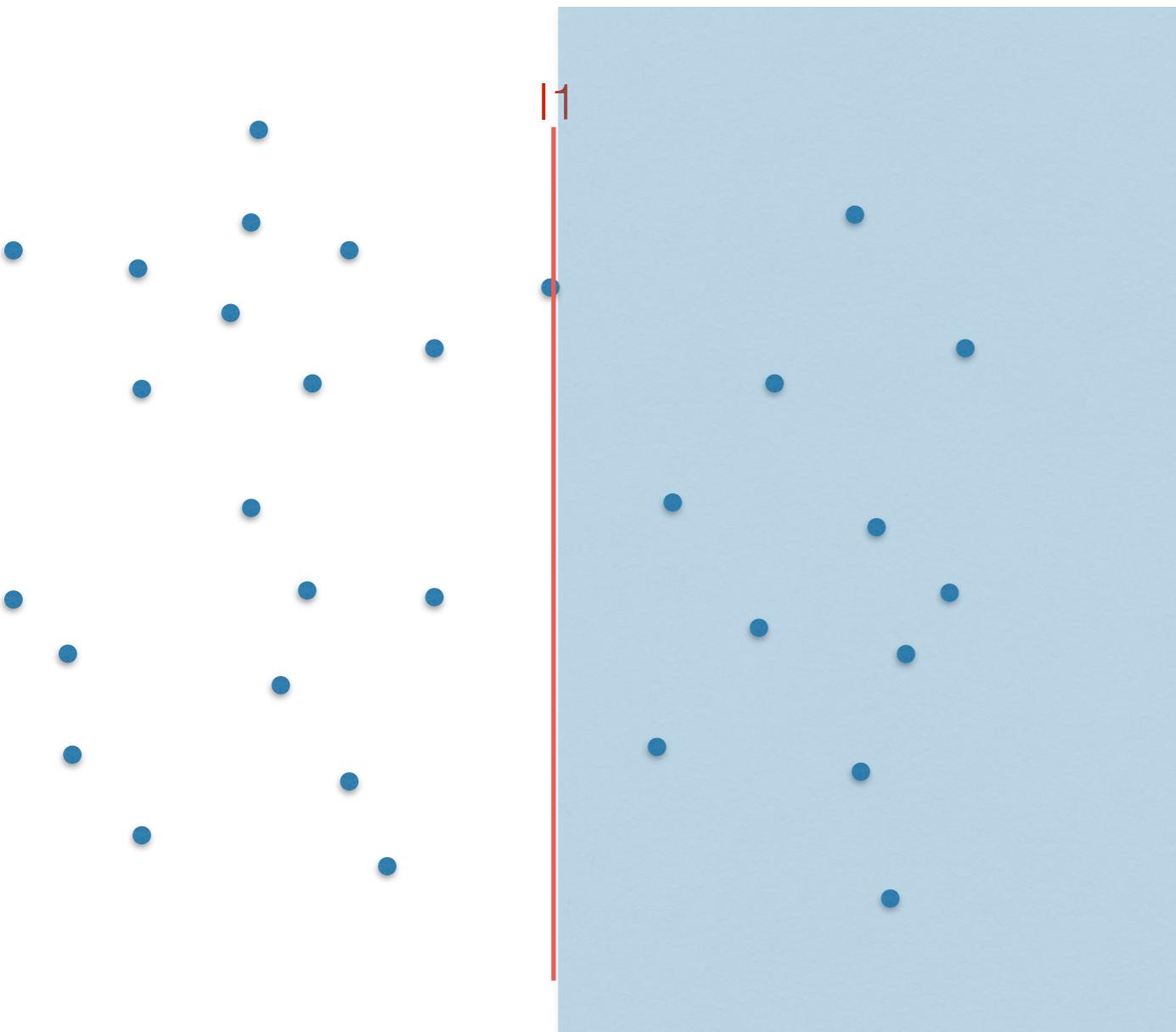
The **region** of a node

whole plane



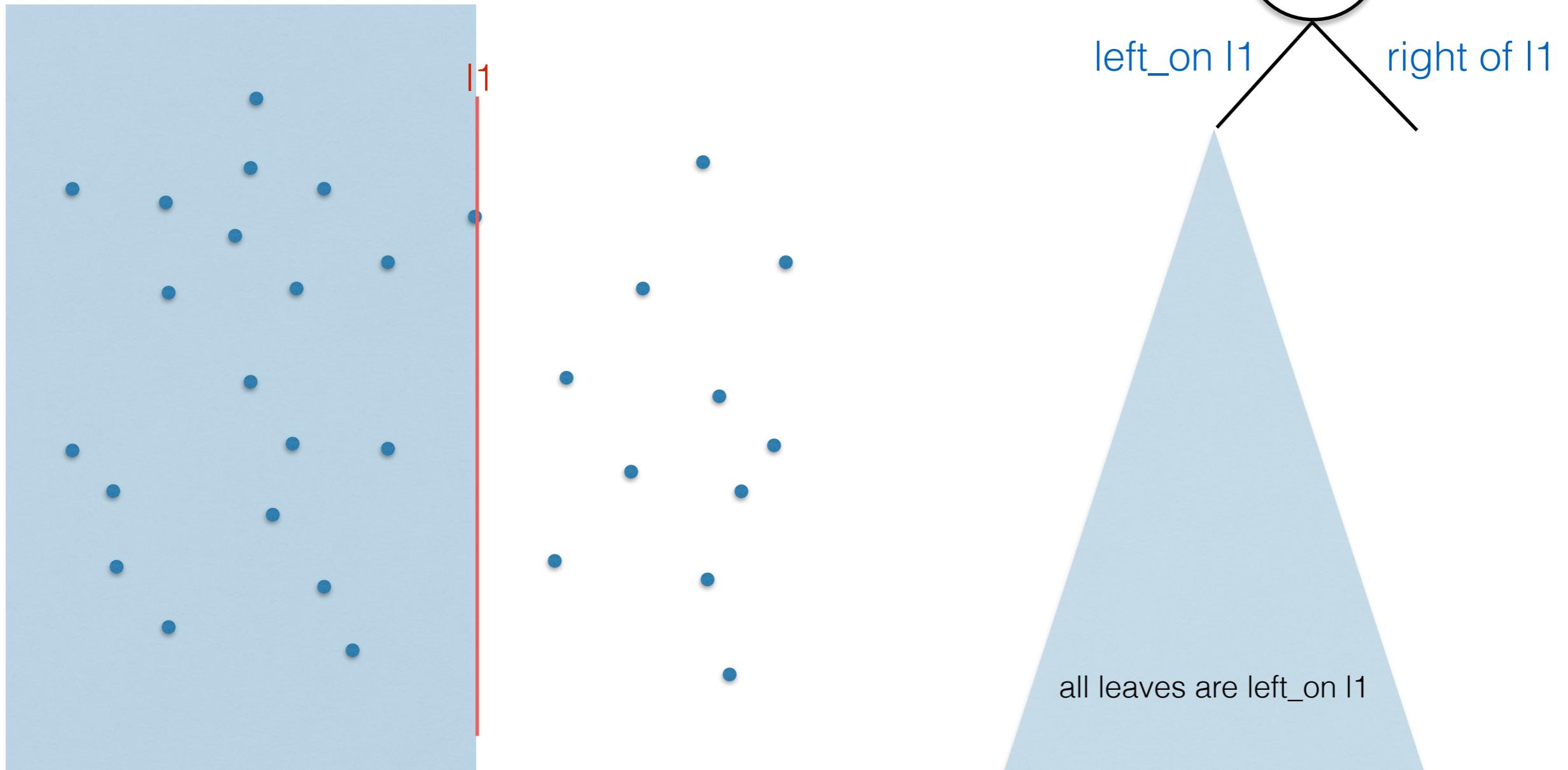
Each node in the tree corresponds to a region in the plane.

The **region** of a node



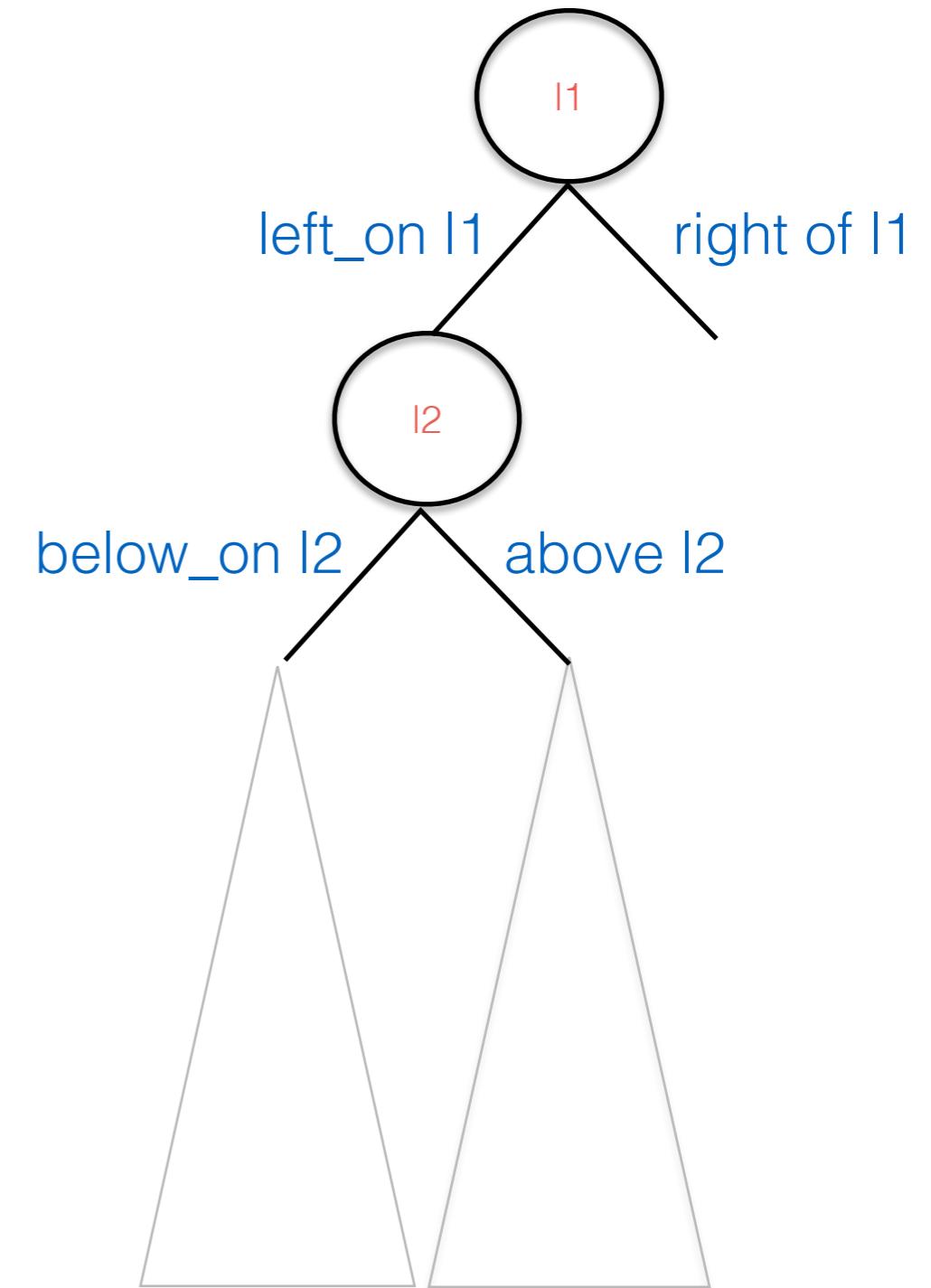
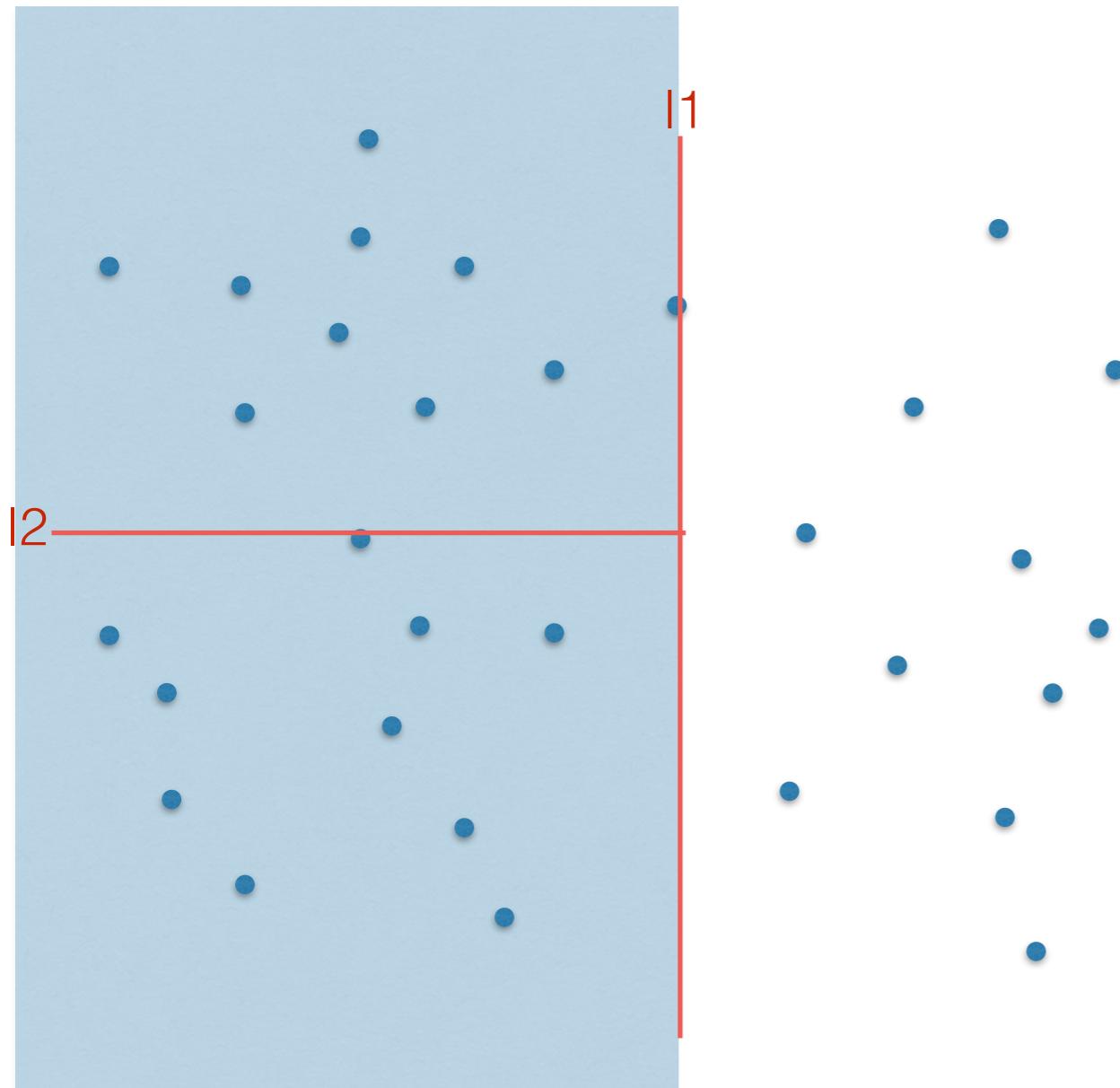
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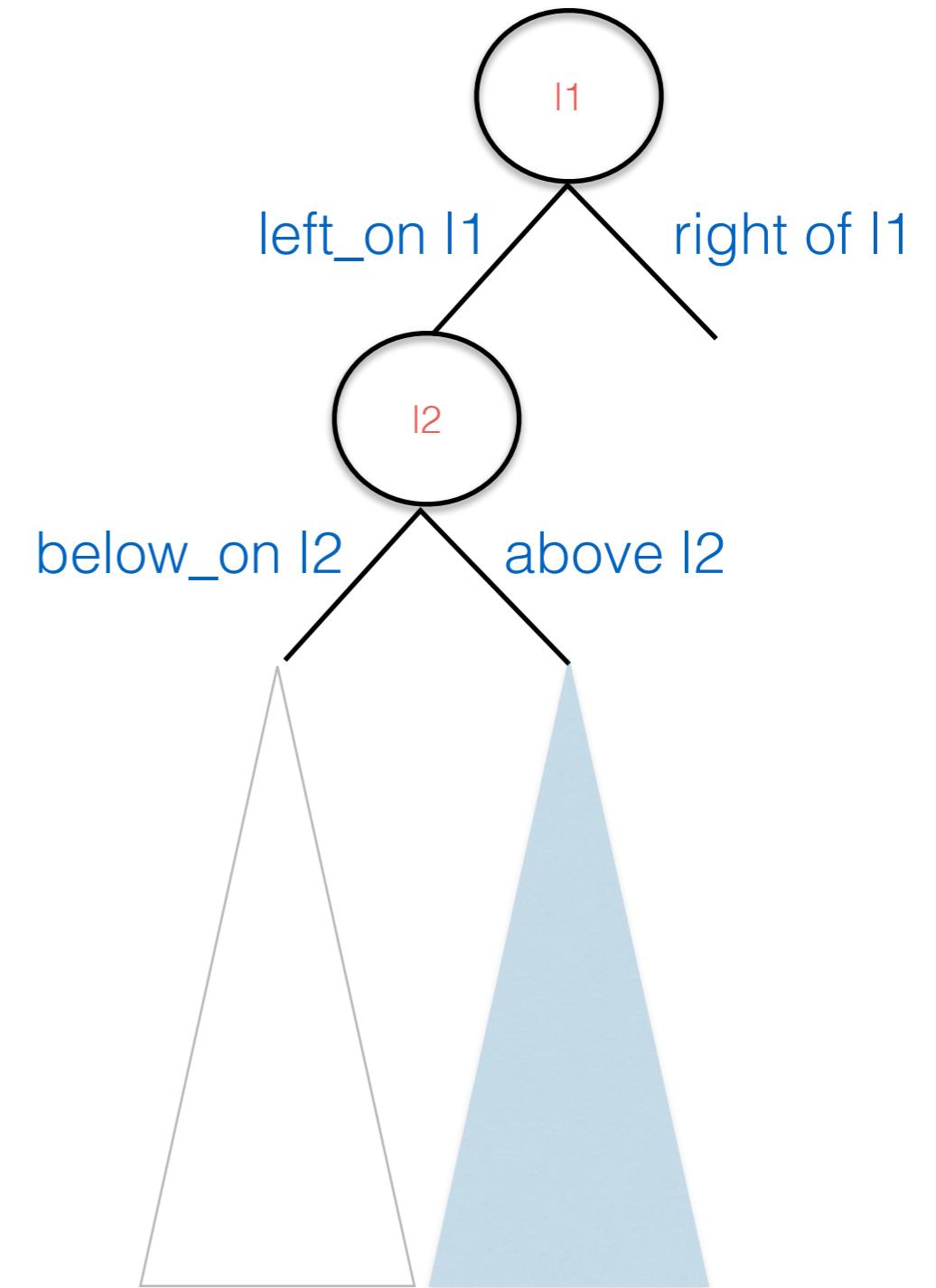
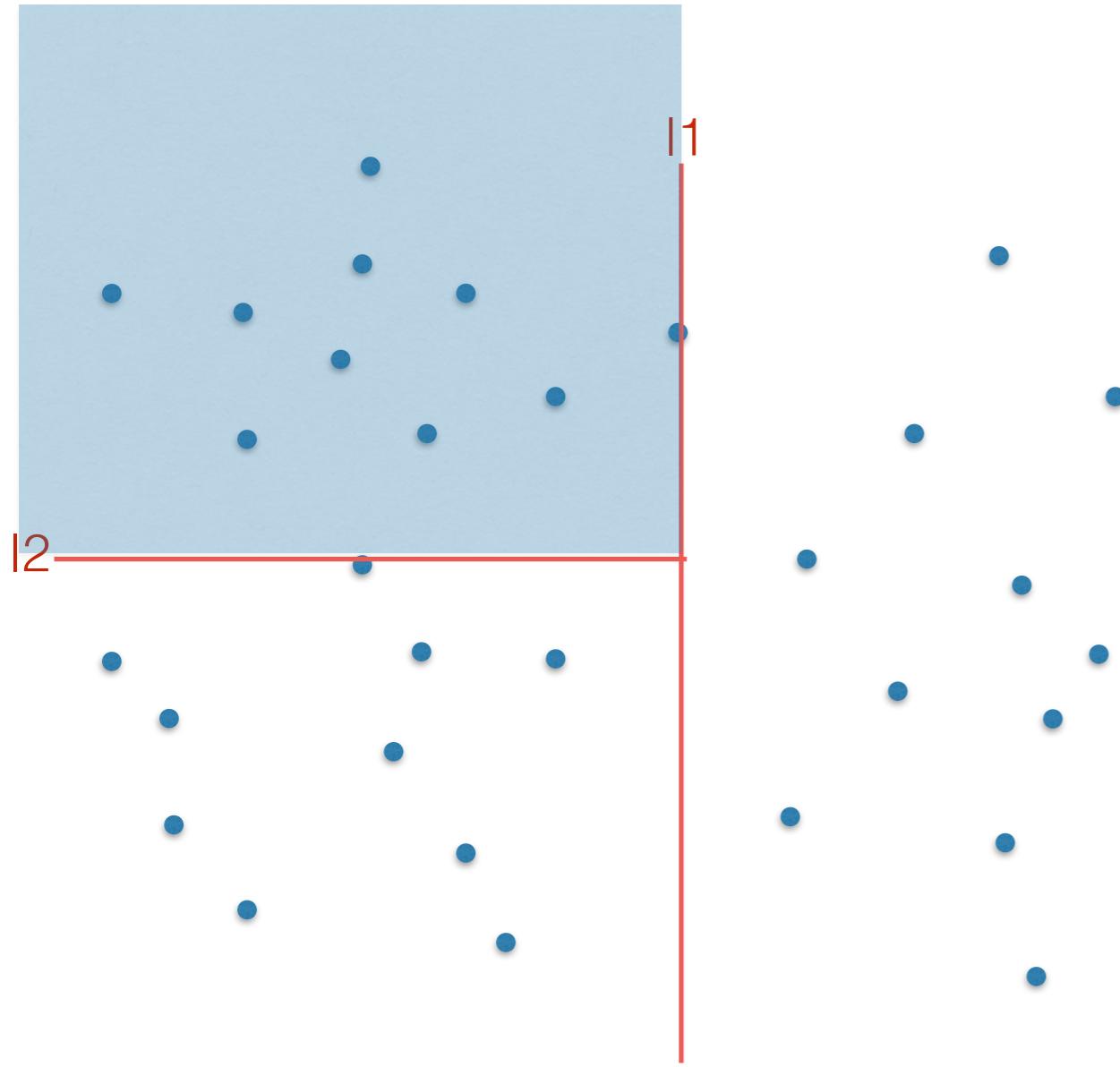
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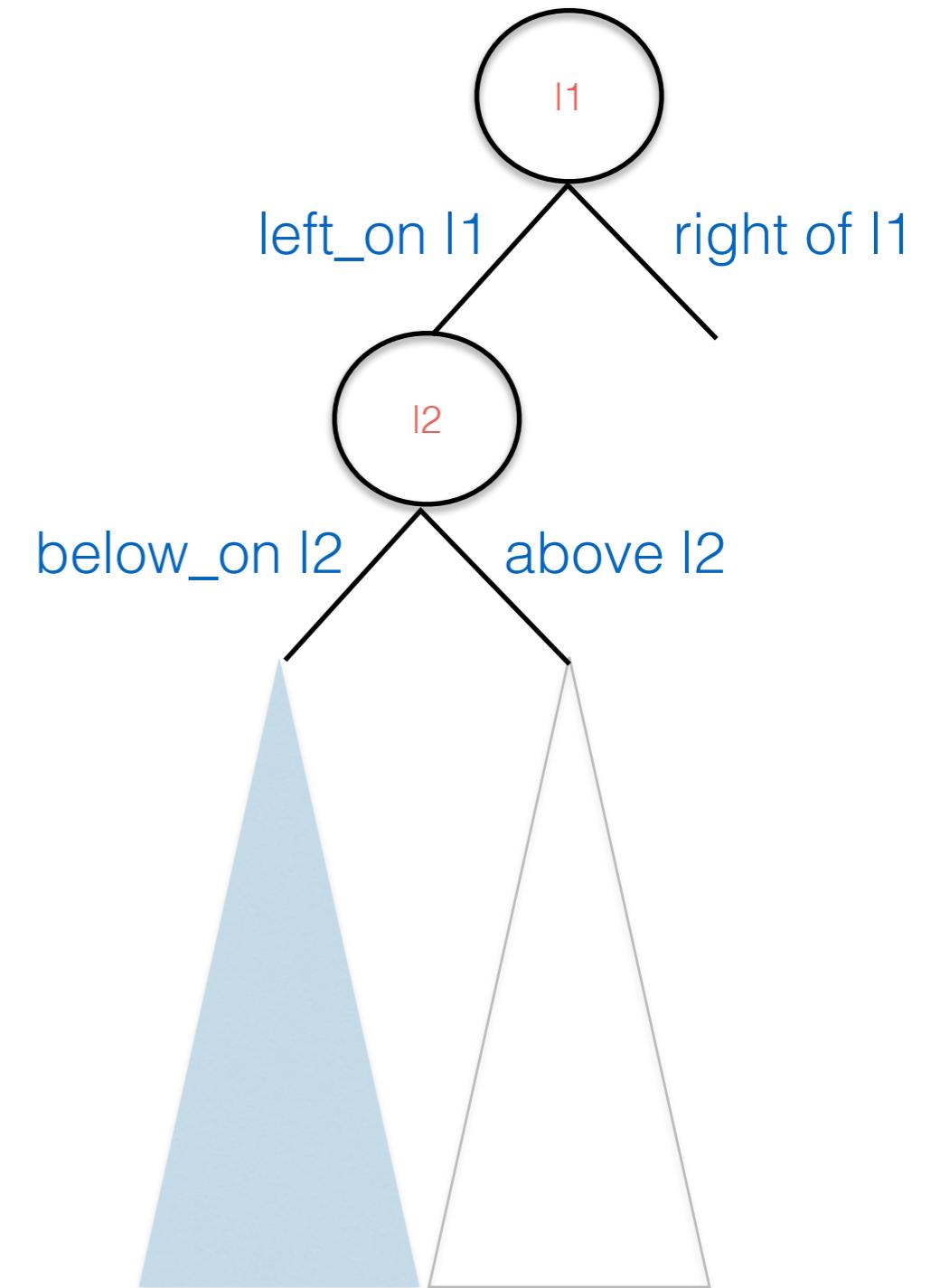
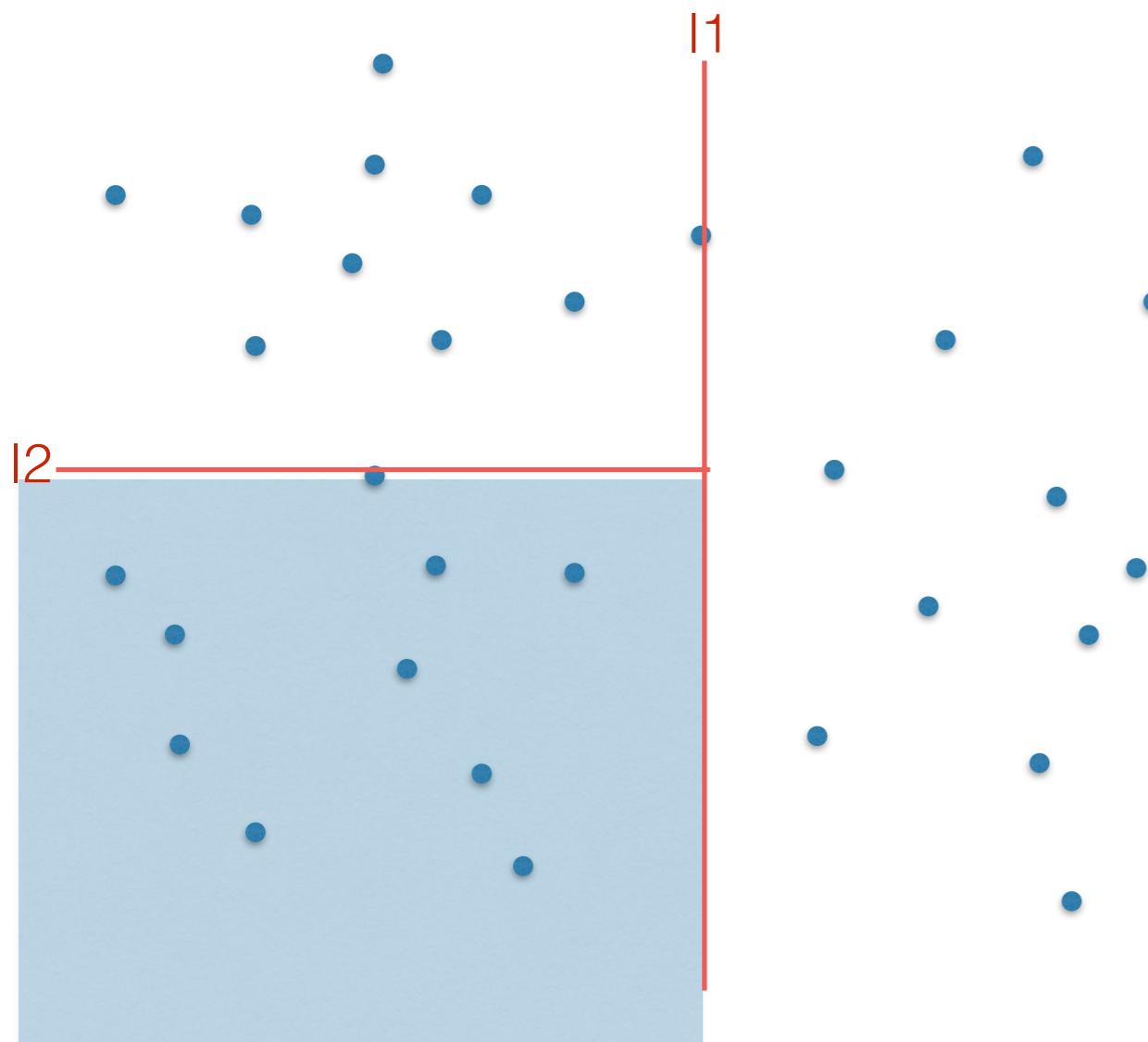
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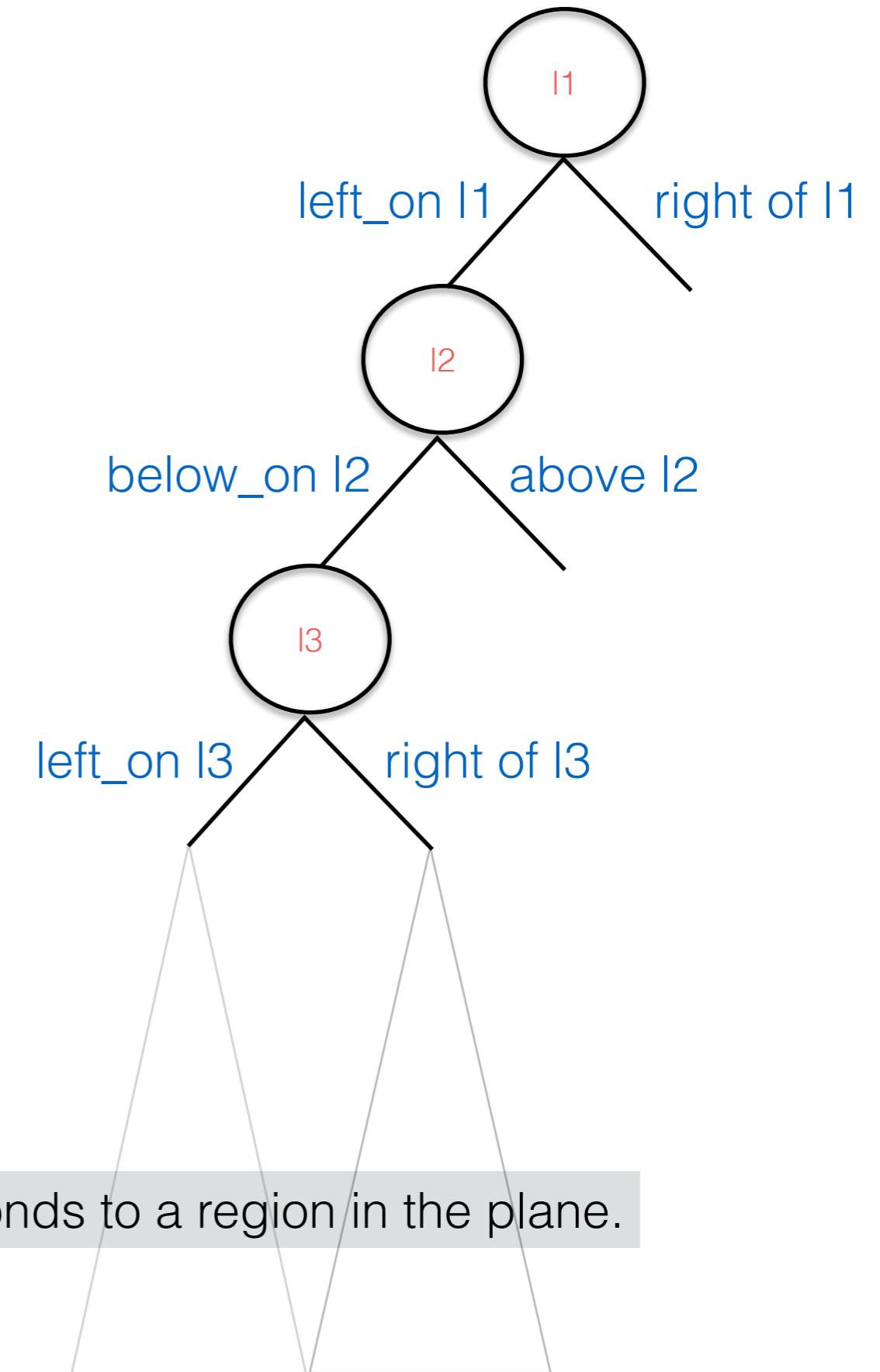
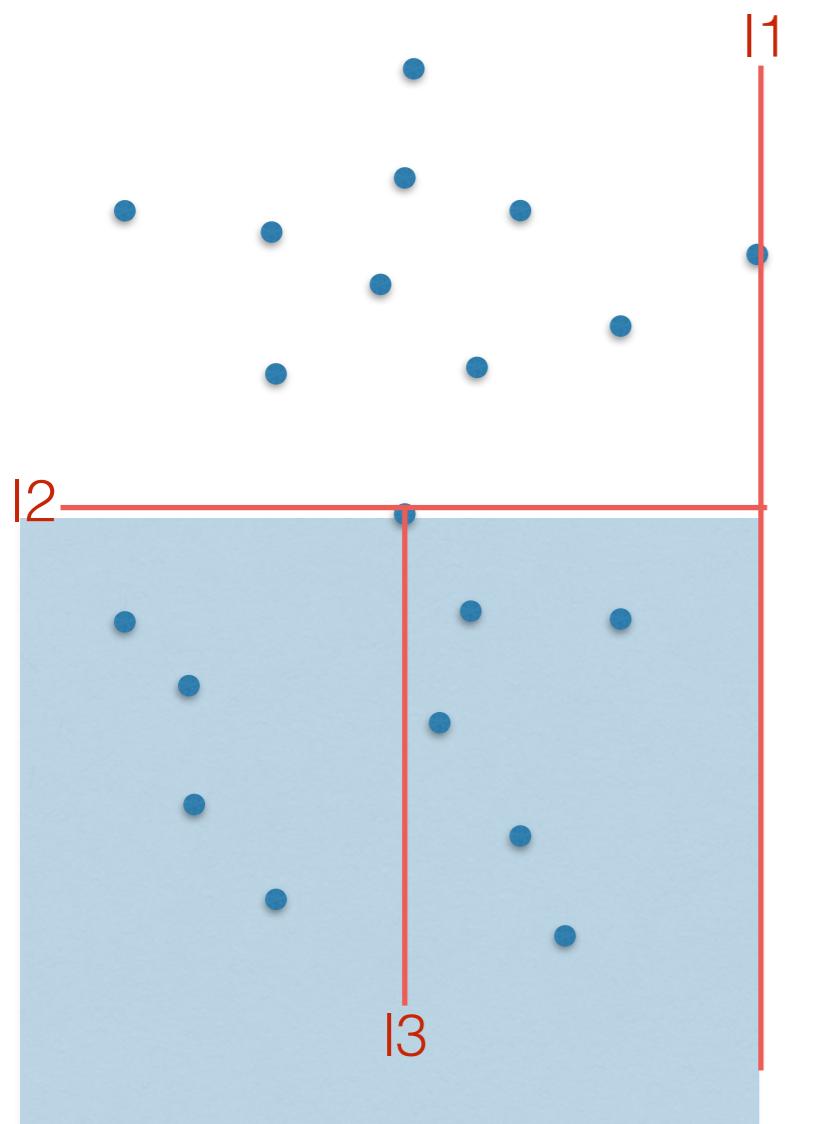
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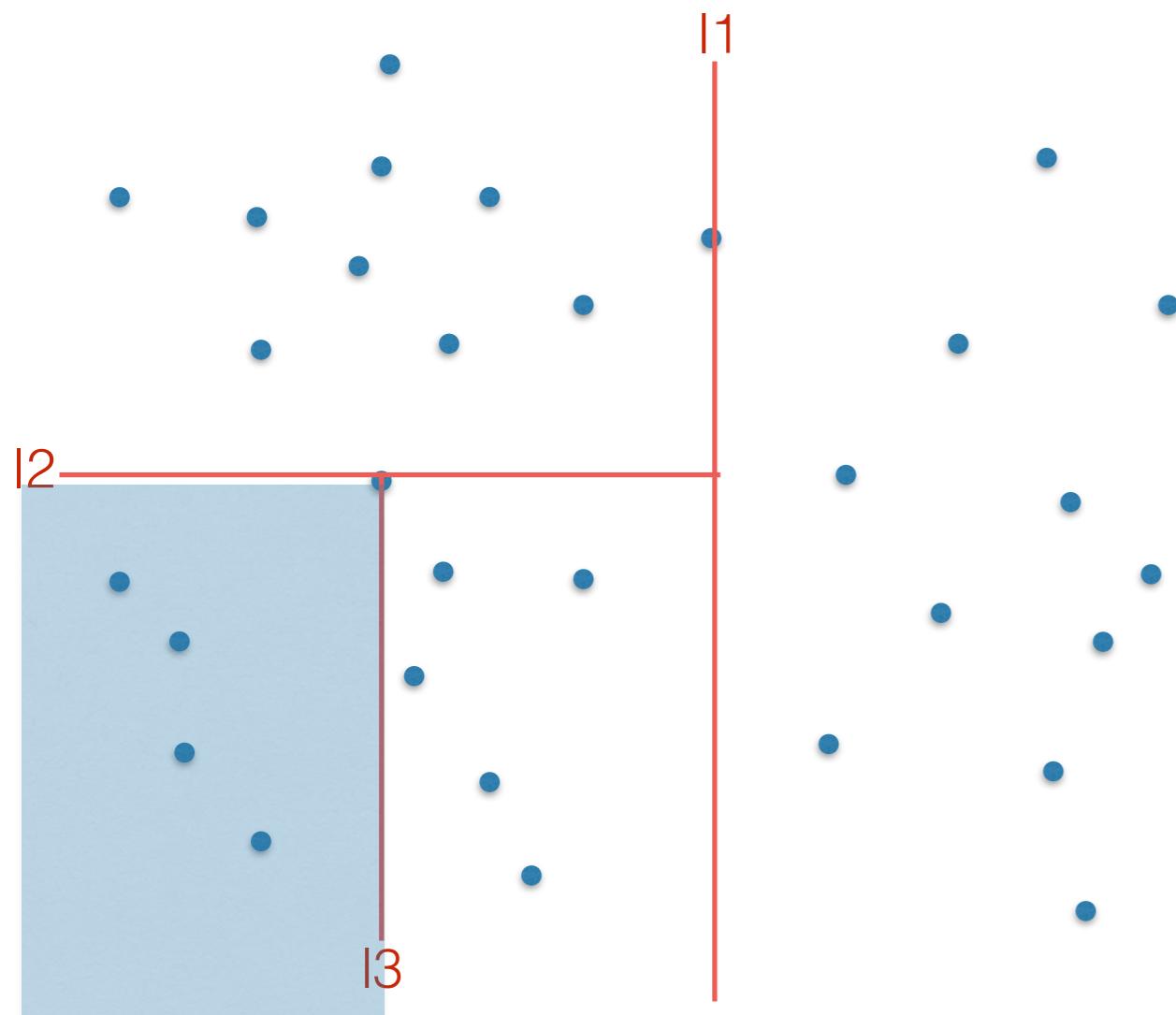
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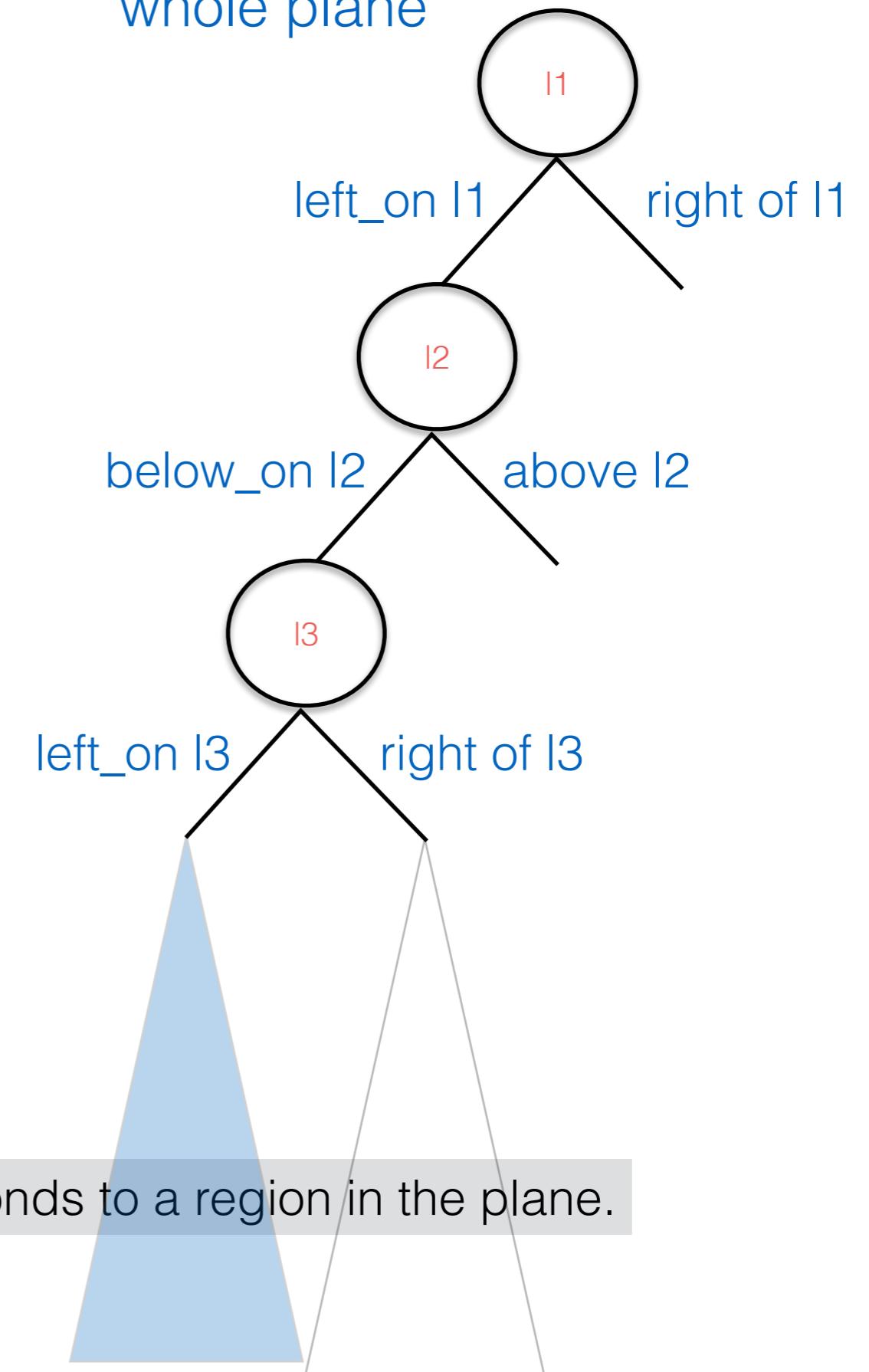


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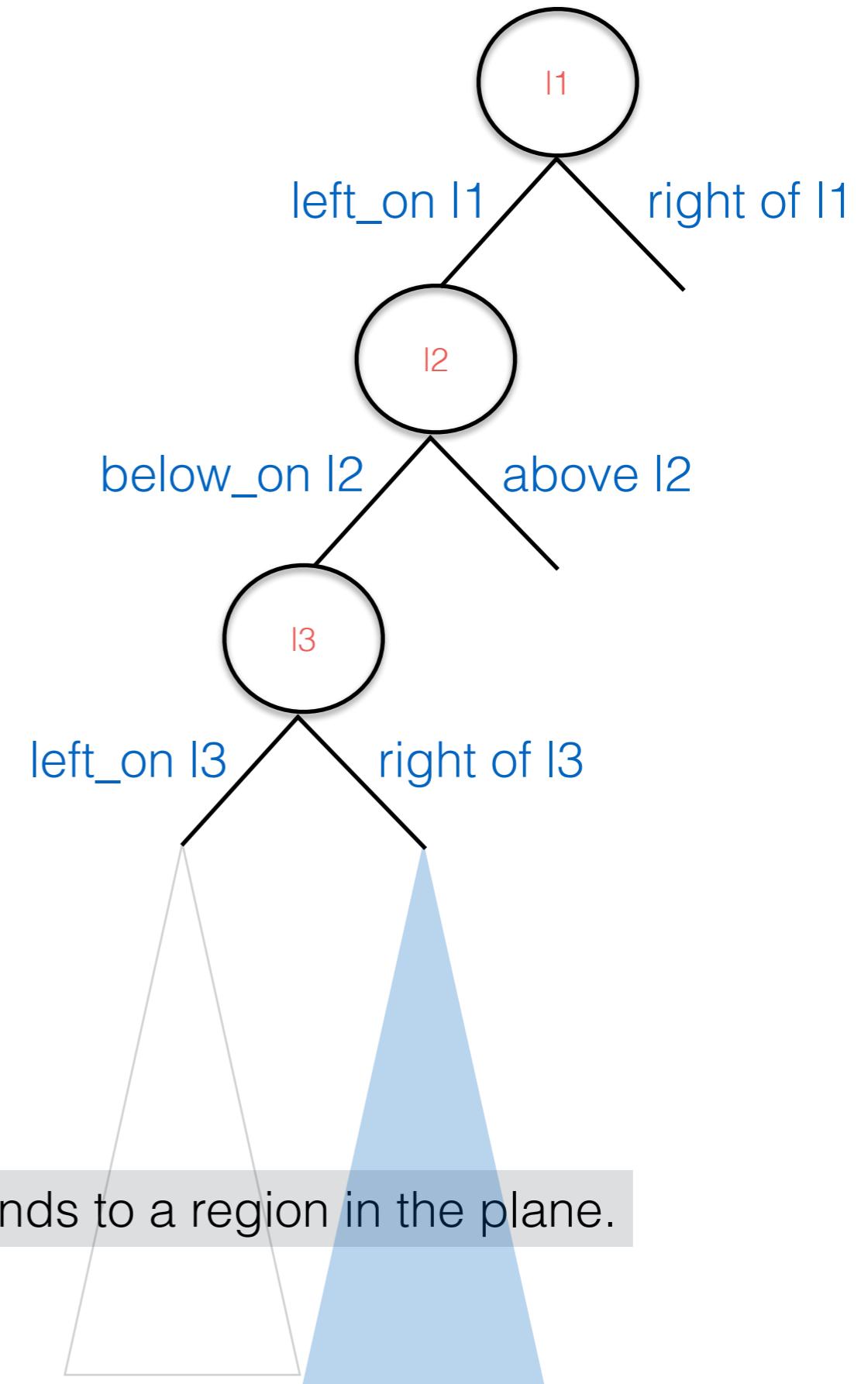
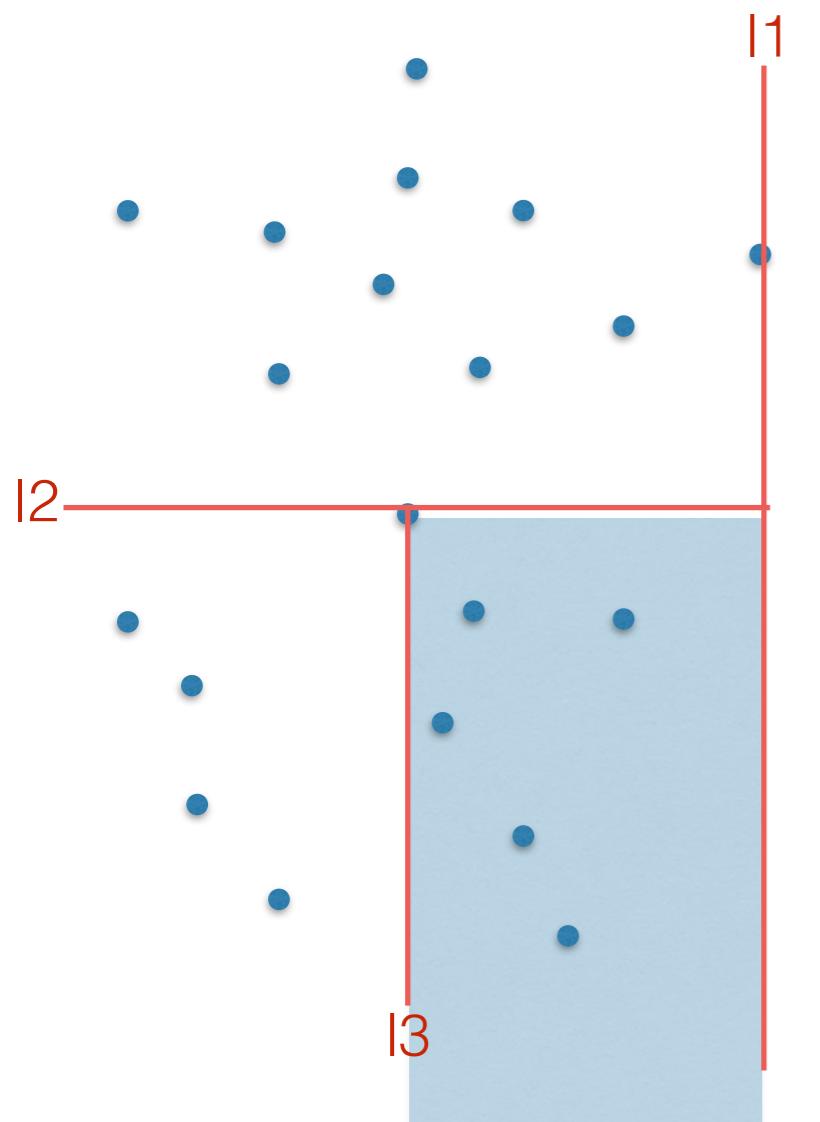


whole plane



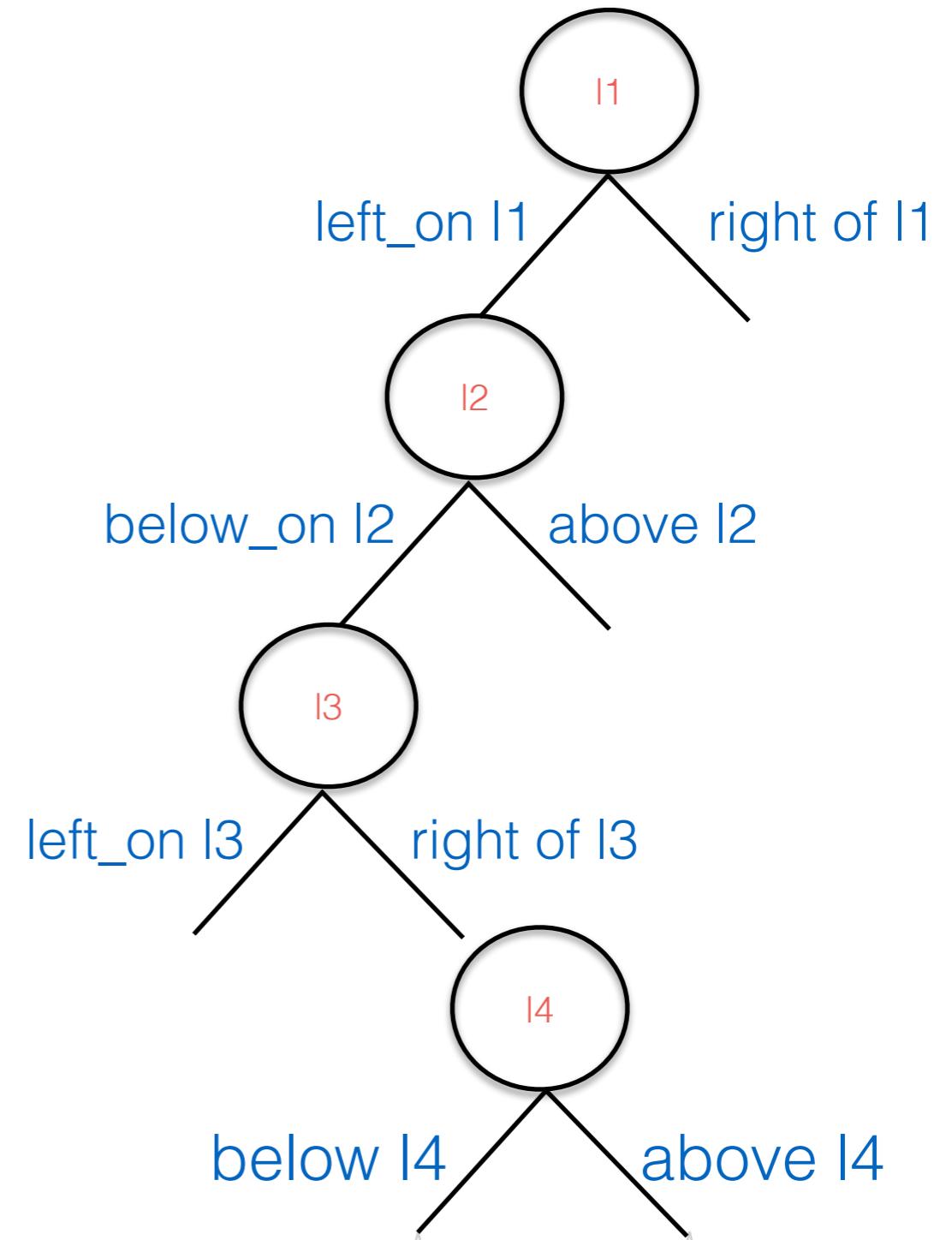
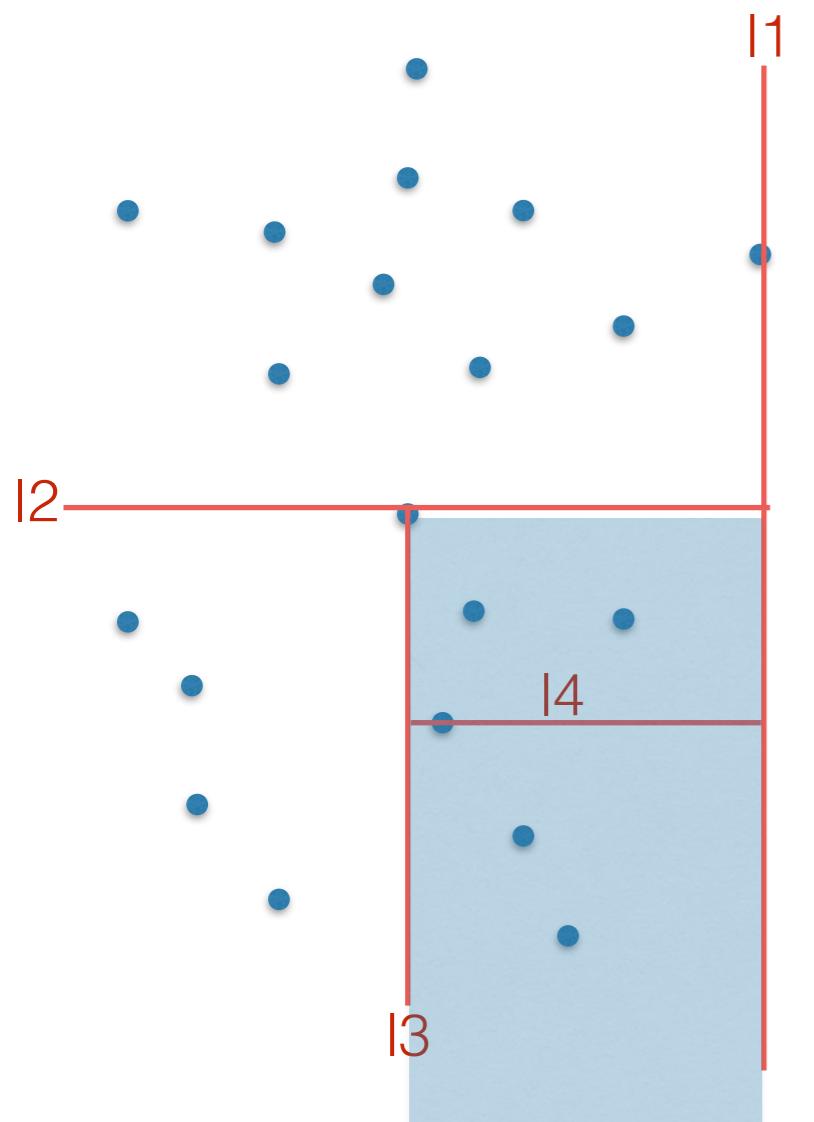
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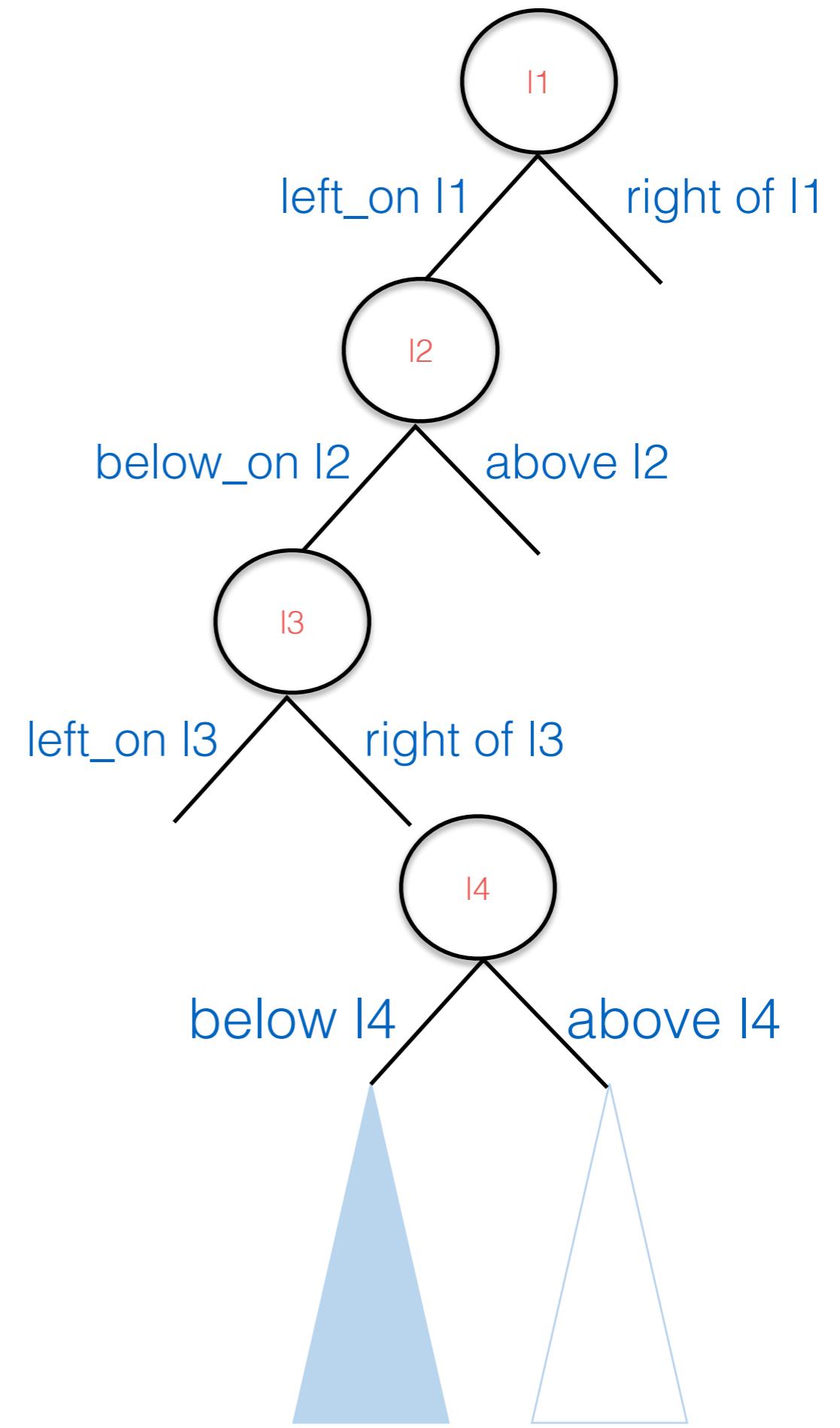
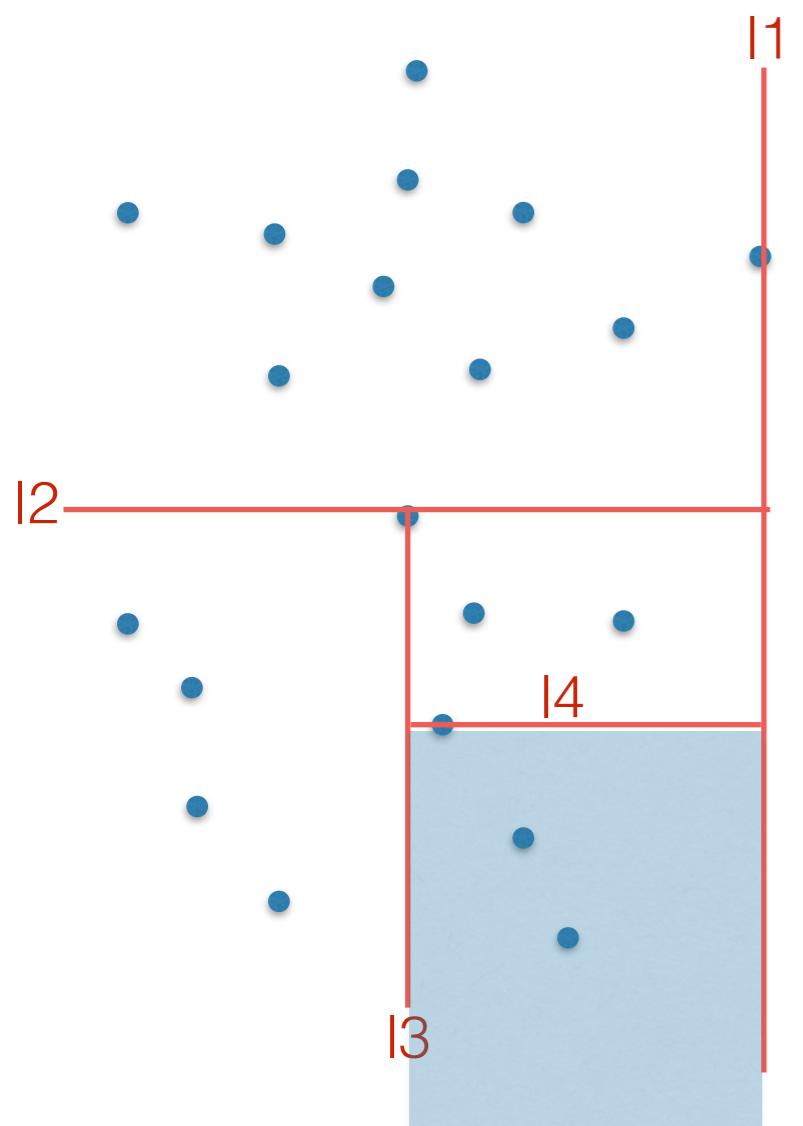
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The **region** of a node

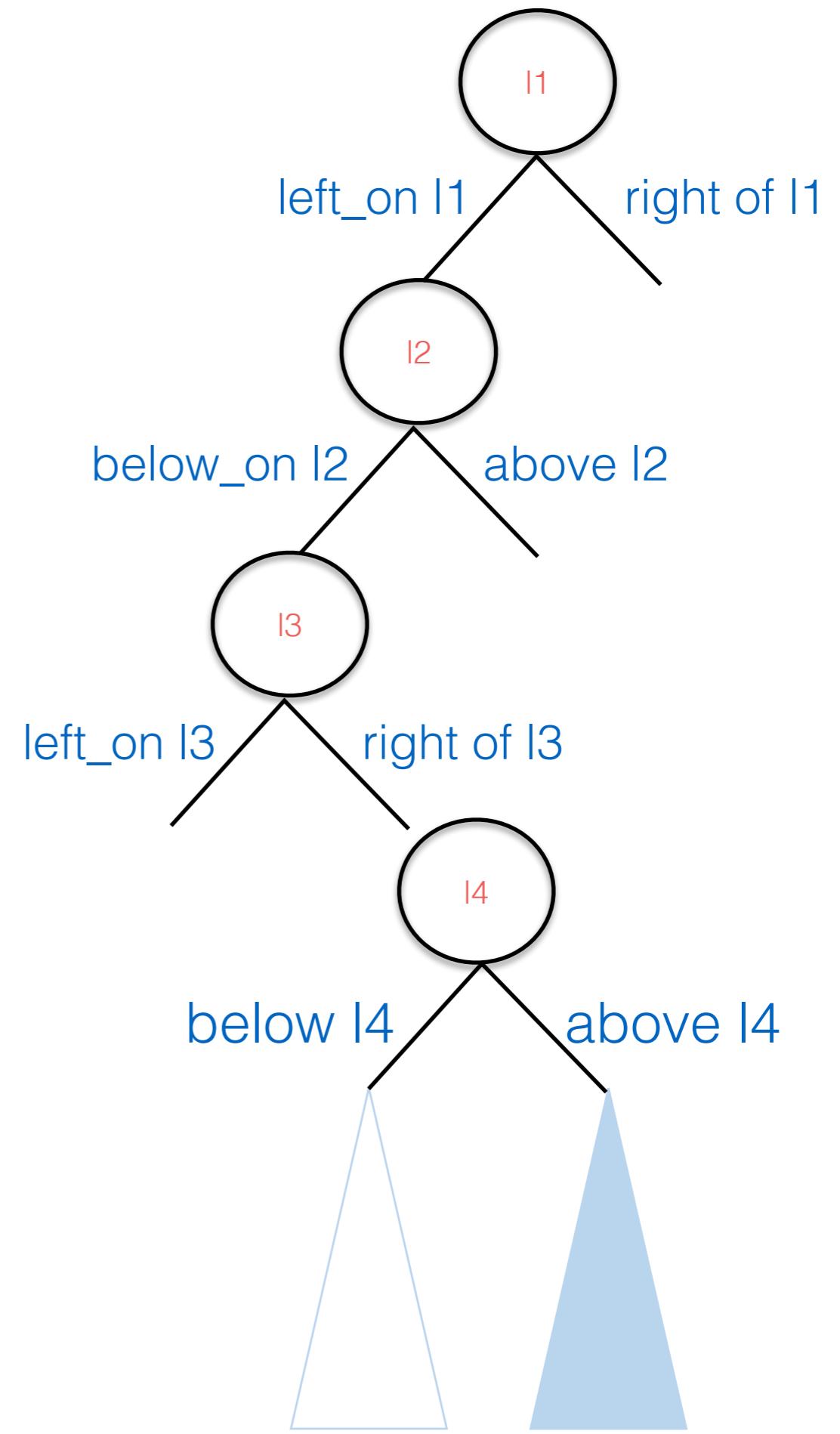
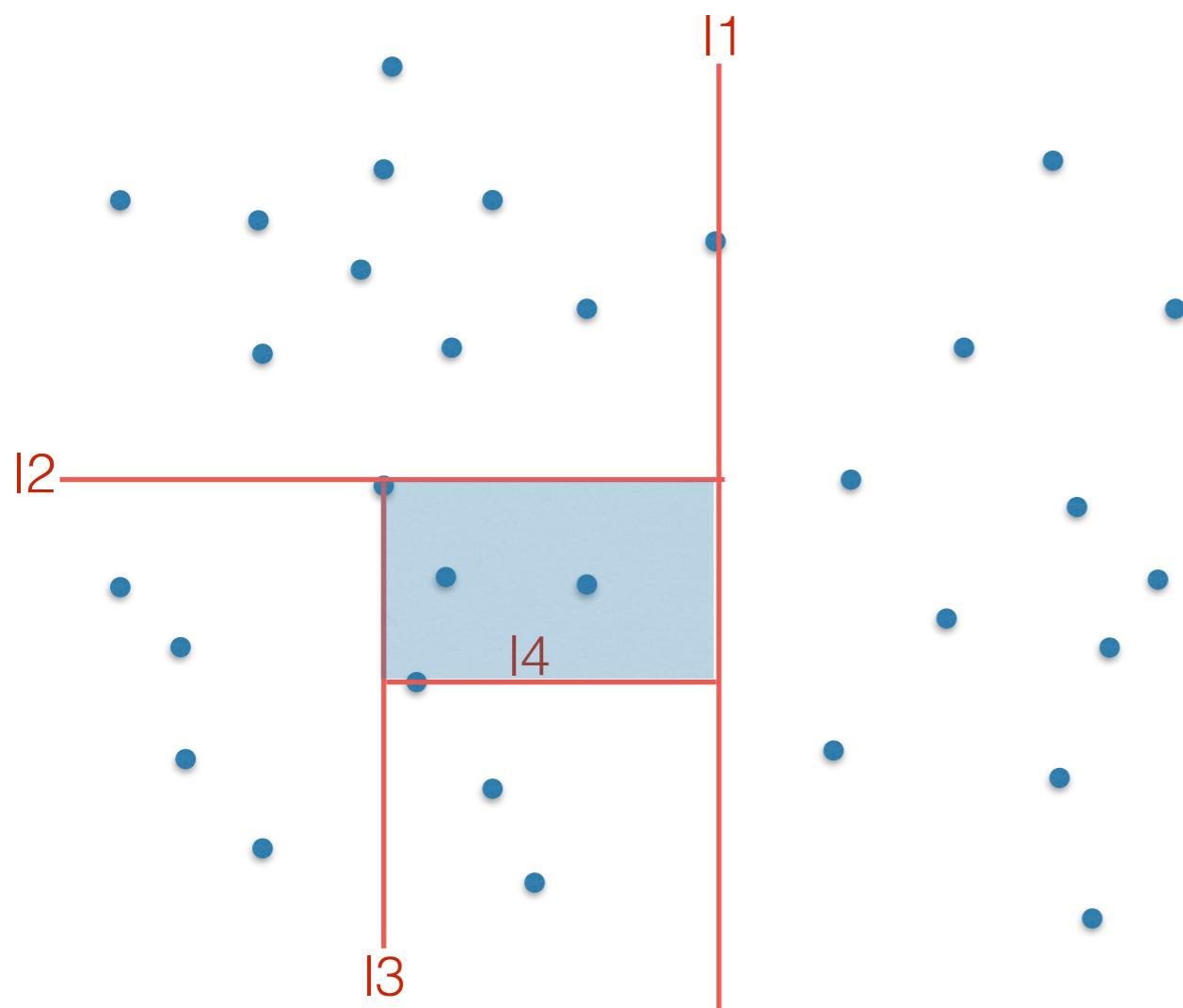


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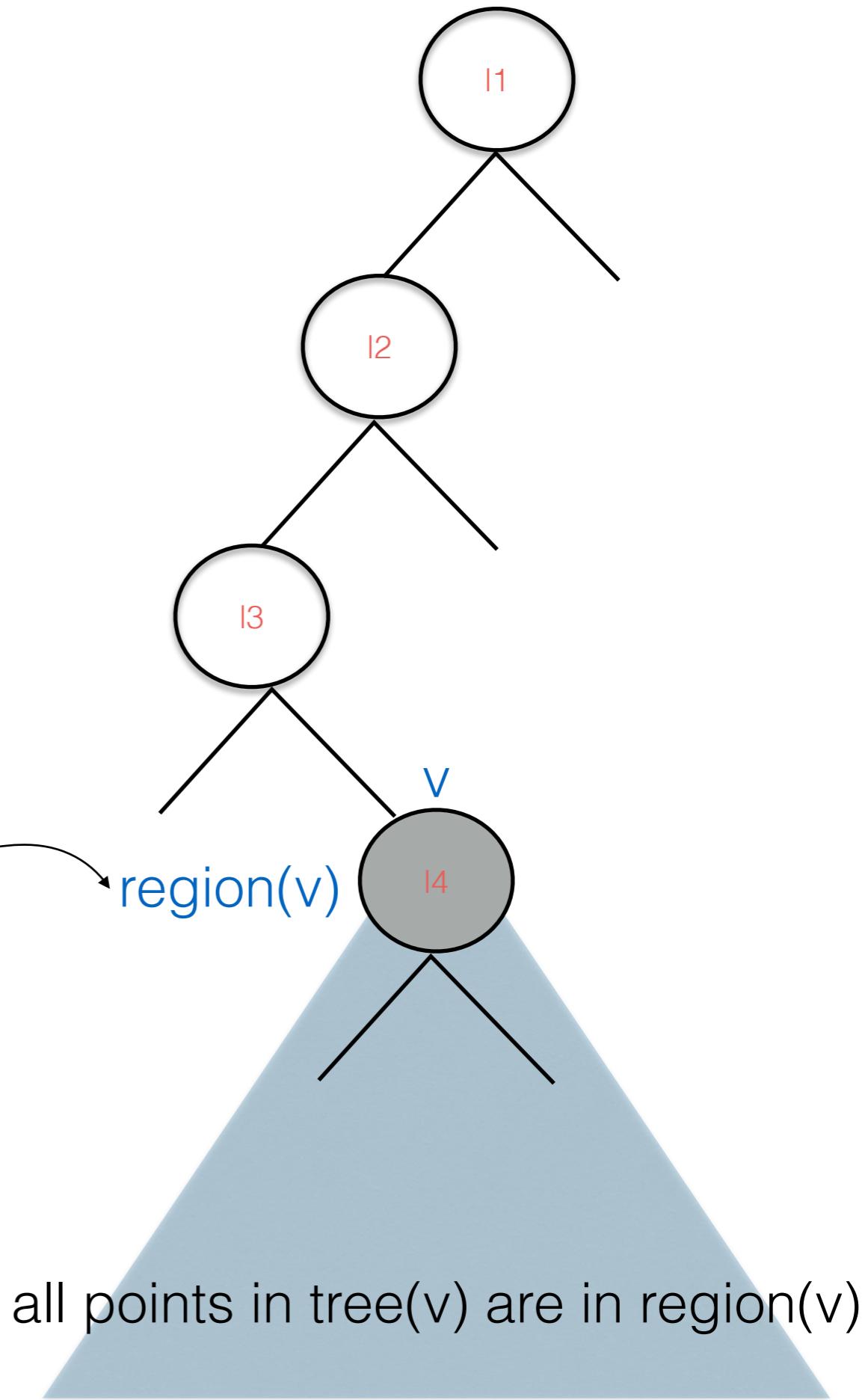
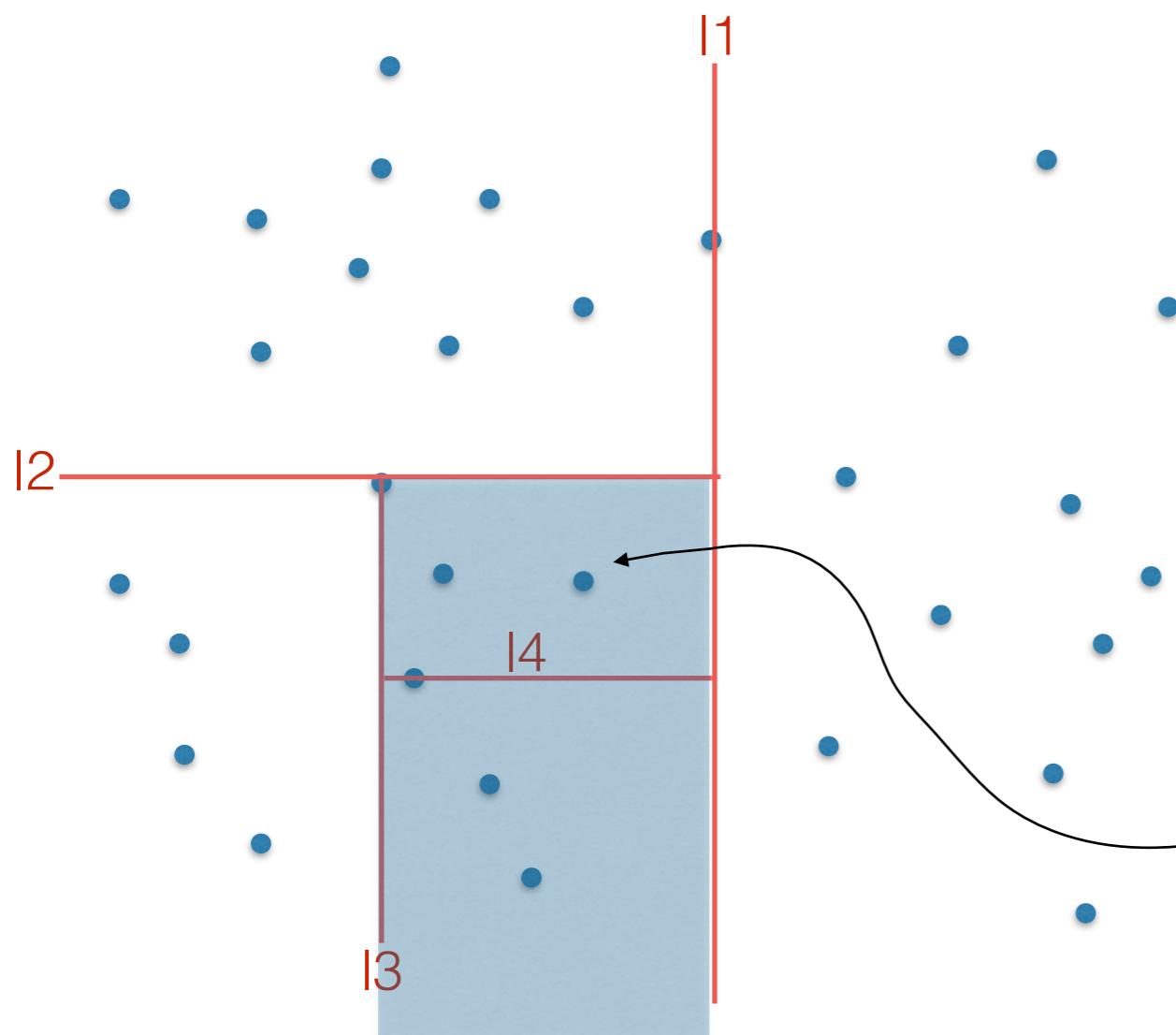
The **region** of a node



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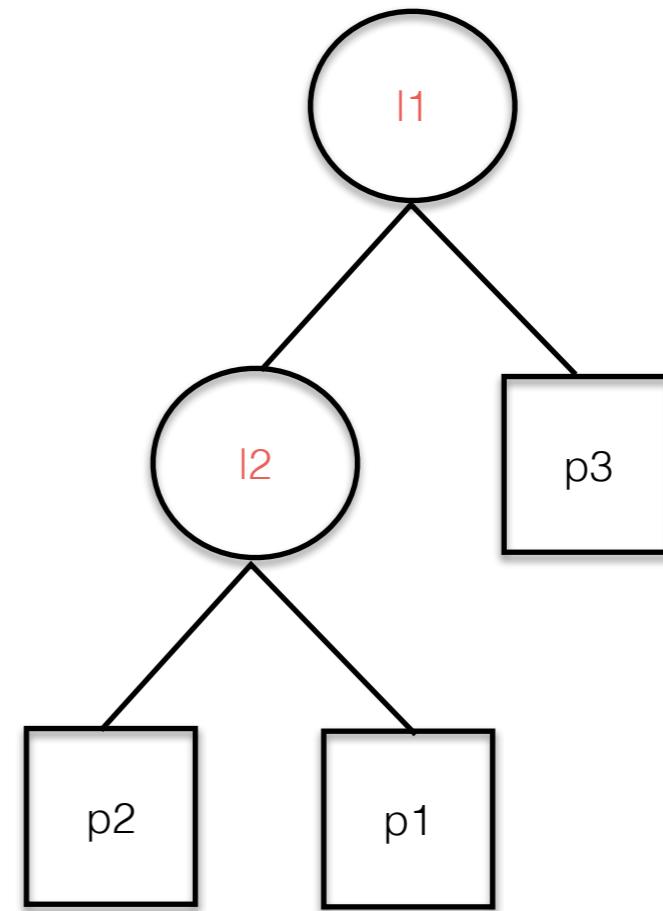
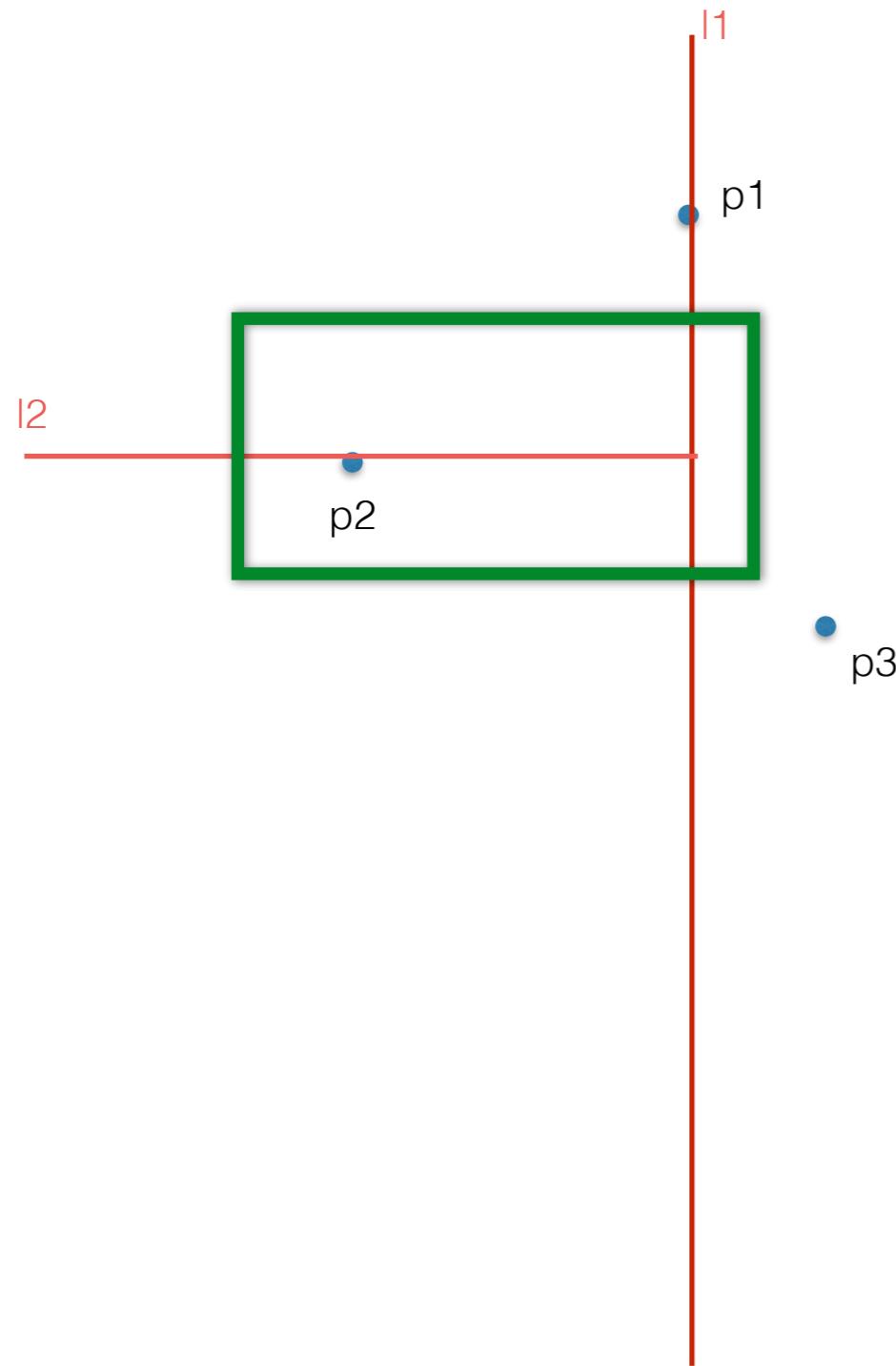


Each node in the tree corresponds to a region in the plane.

all points in $\text{tree}(v)$ are in $\text{region}(v)$

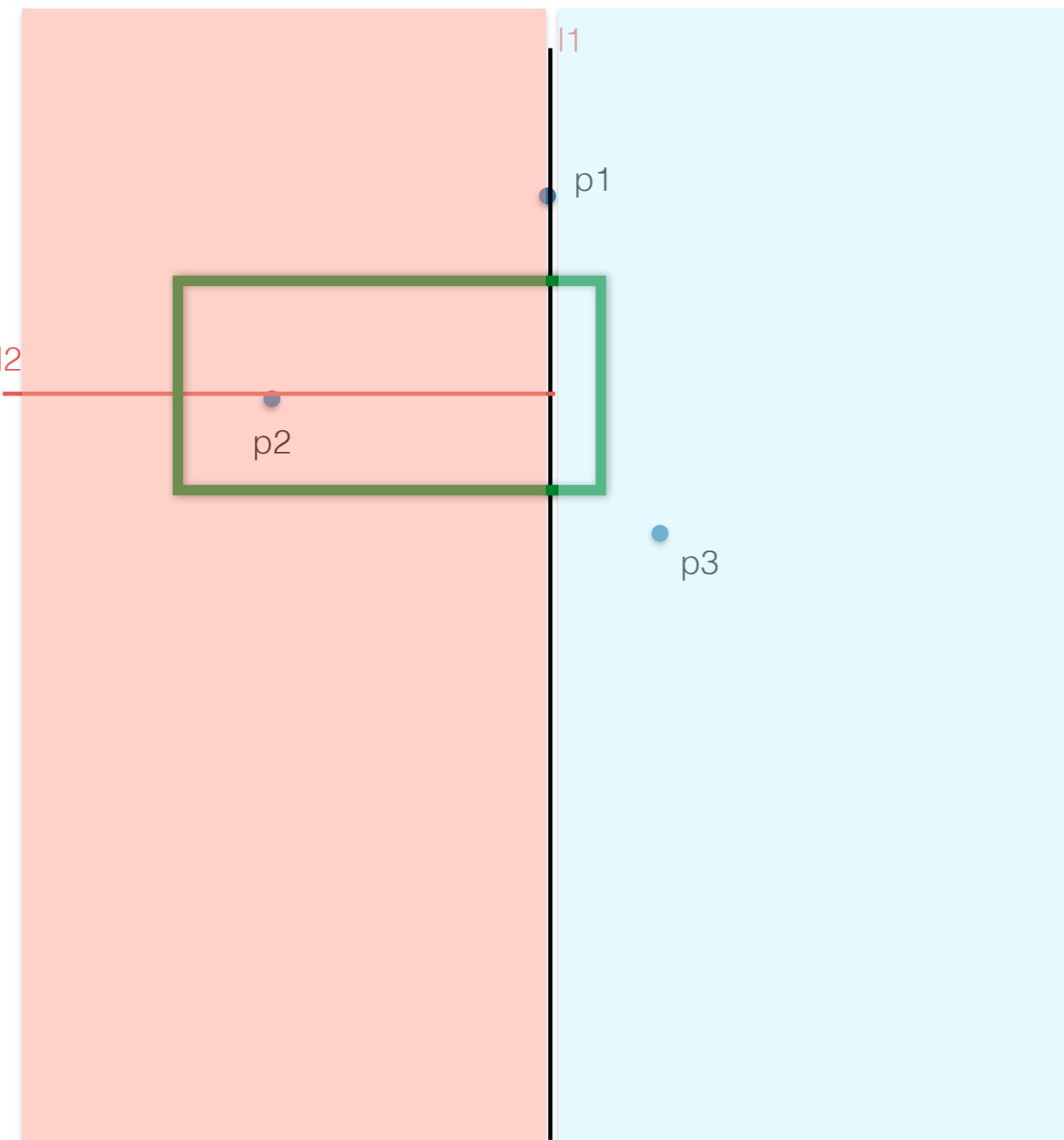
We'll use this insight to answer range queries

Range queries on 2d-binary-search-trees



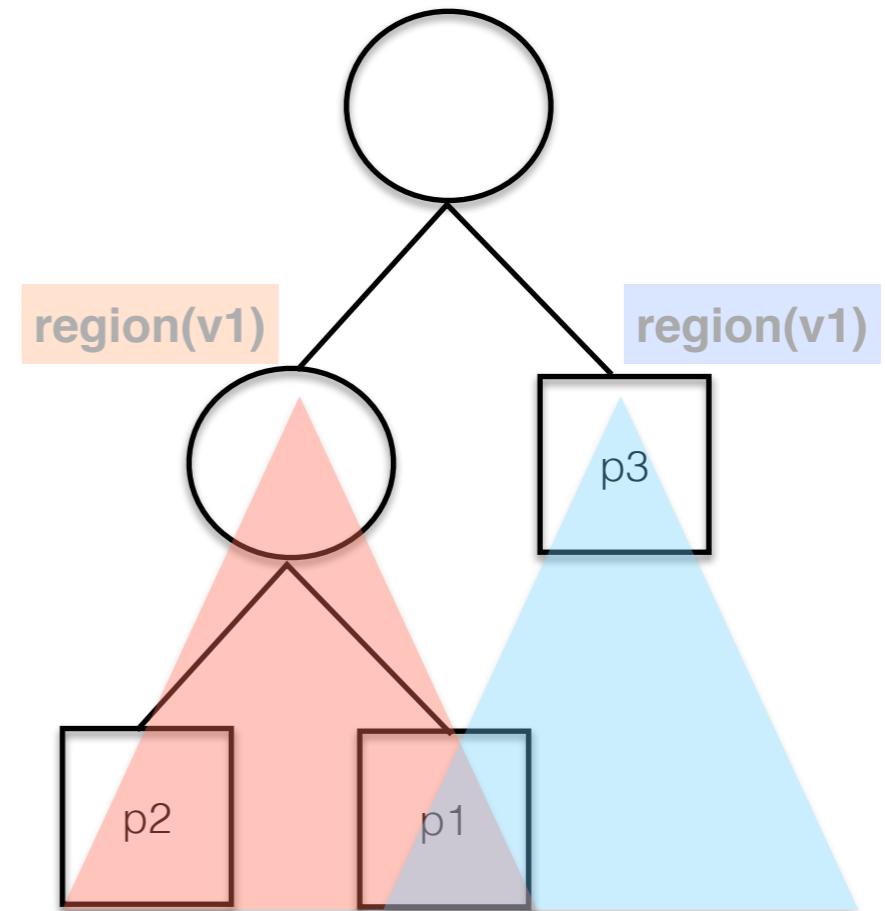
Let's bring in the space partition defined by the tree

We are at the root node, looking at the two children. To which child should send the query to?

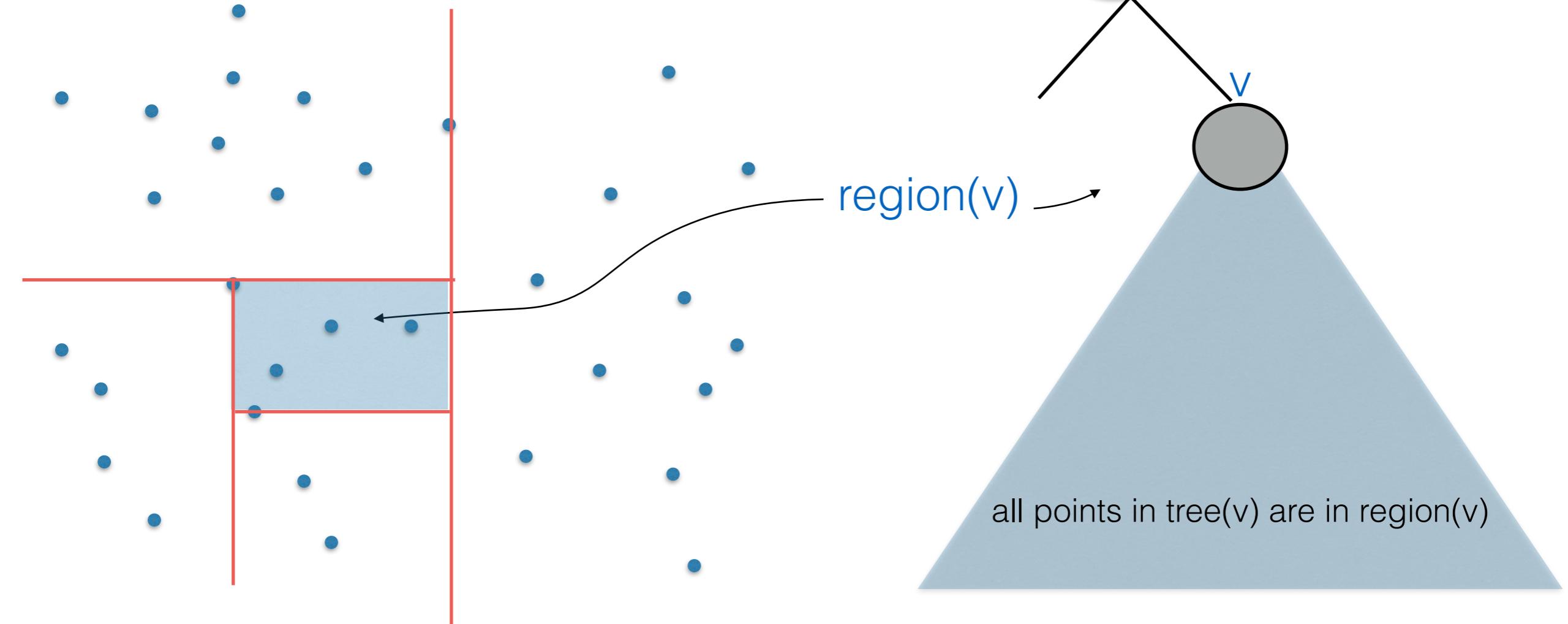


Can left child contain points in the range?

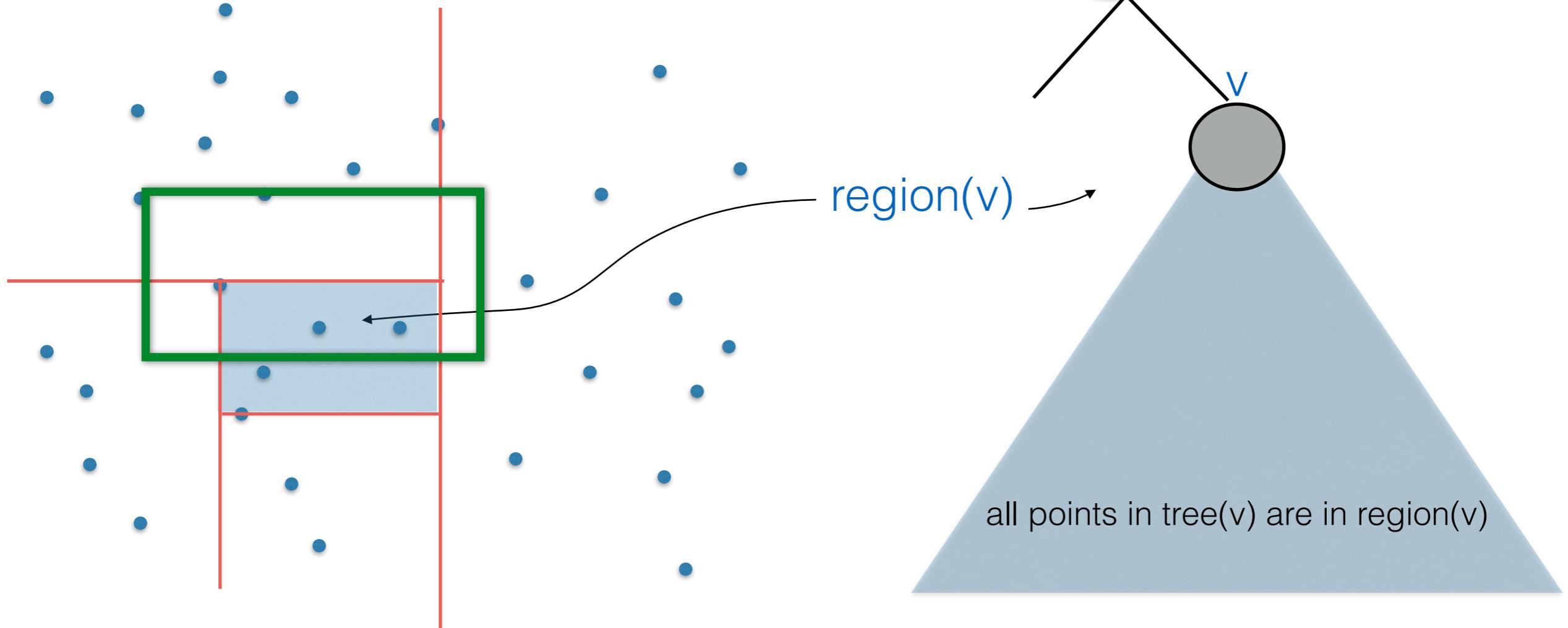
Can right child contain points in the range?

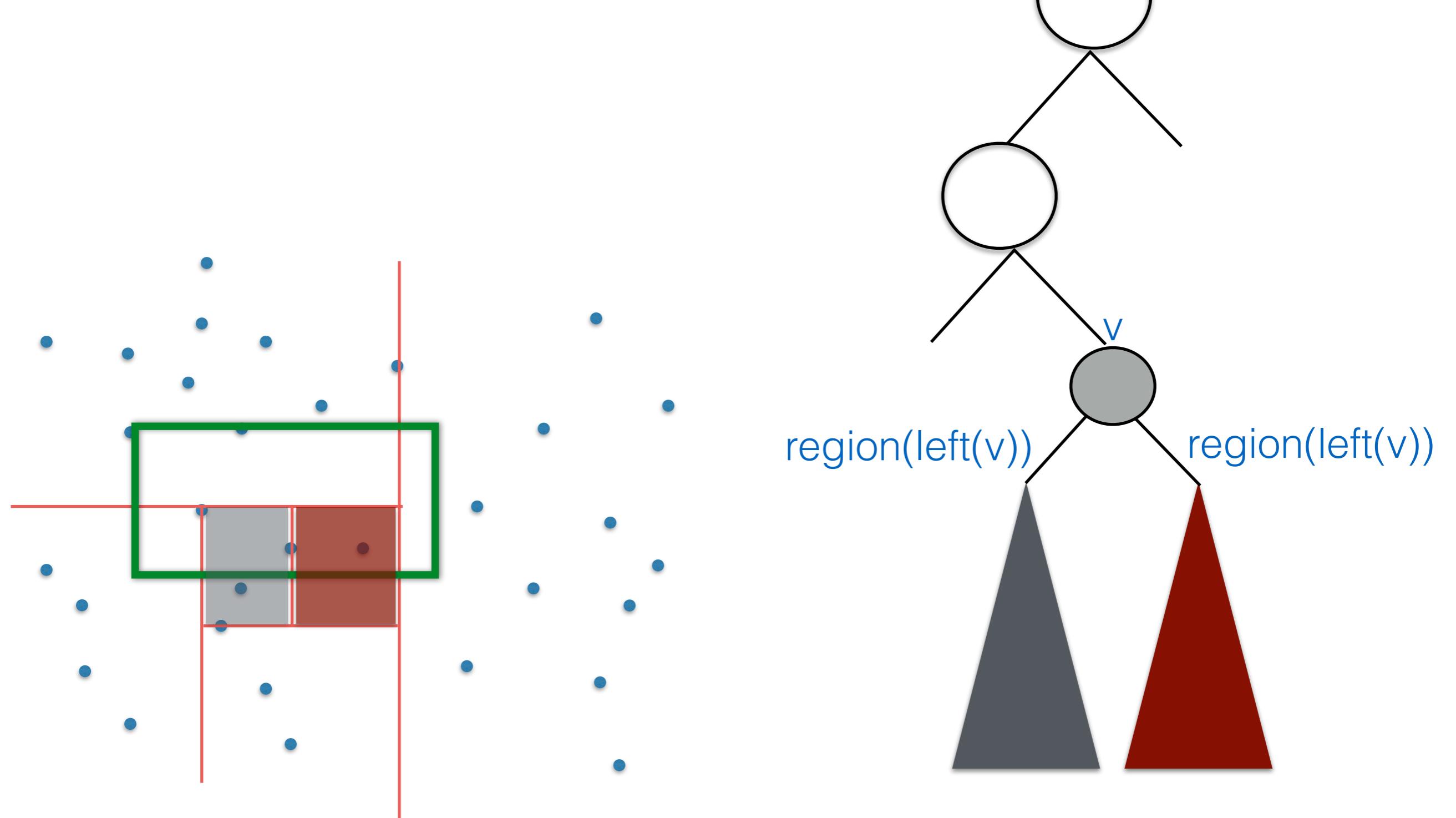


Range queries: general idea



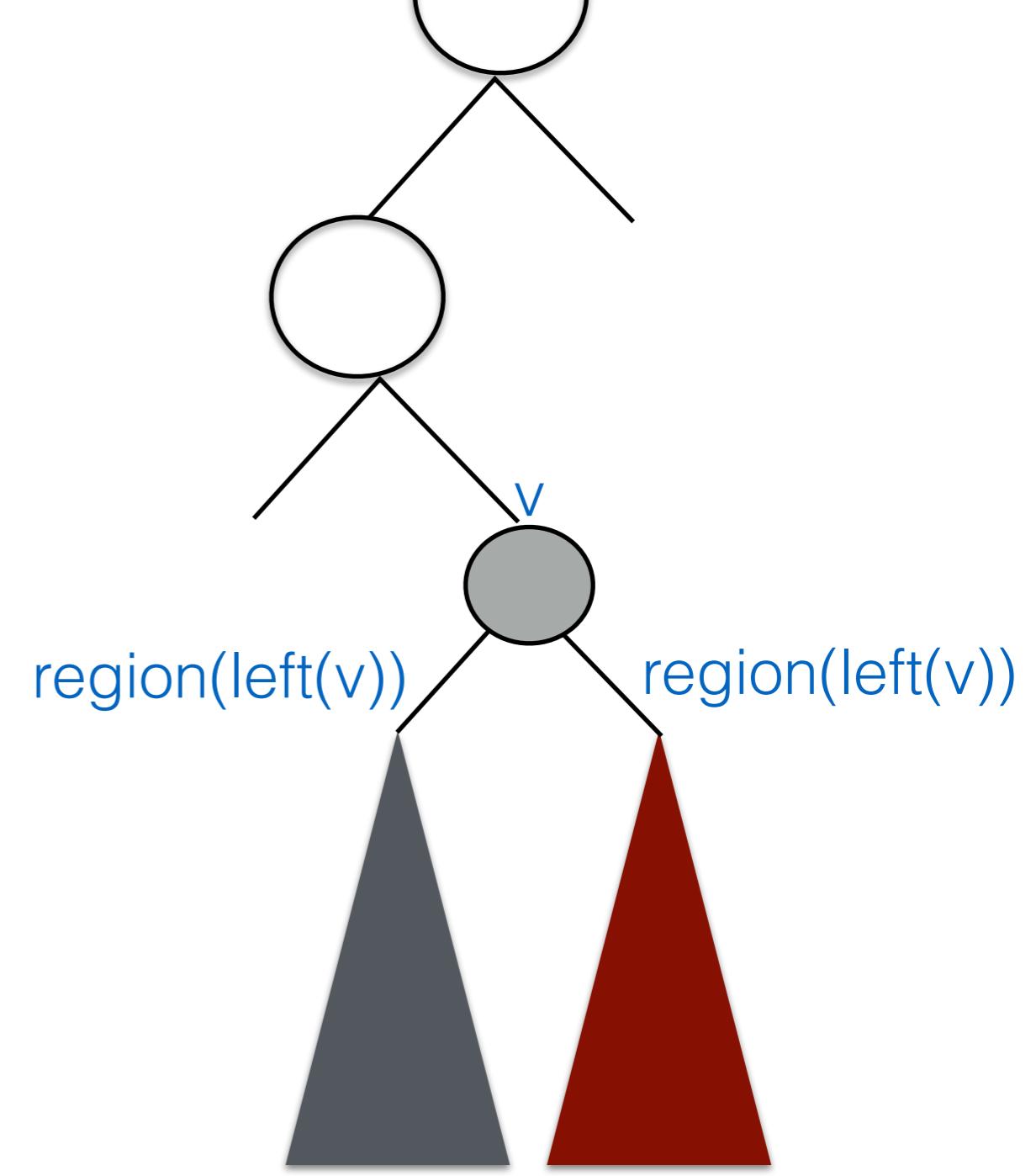
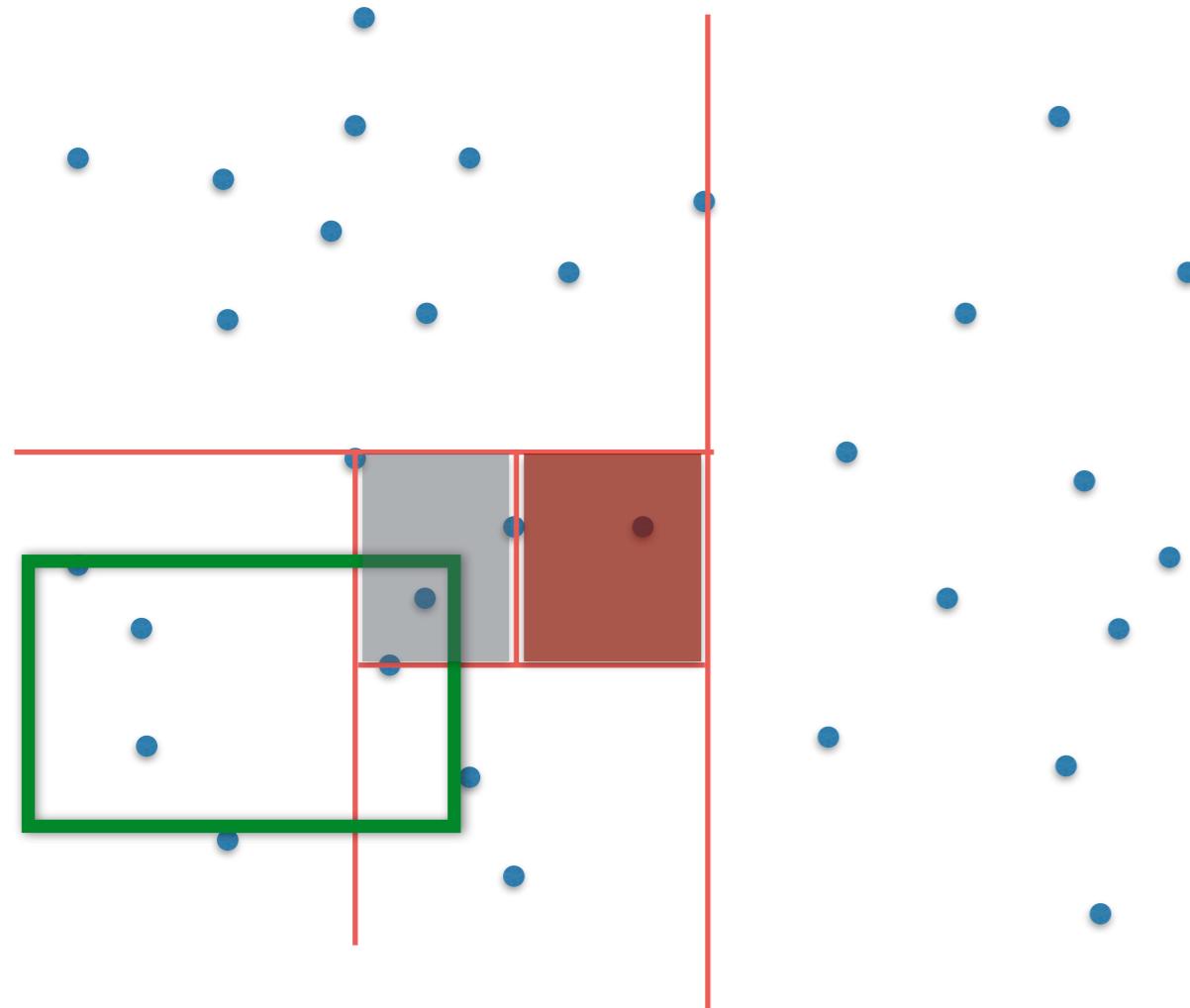
Range queries: general idea





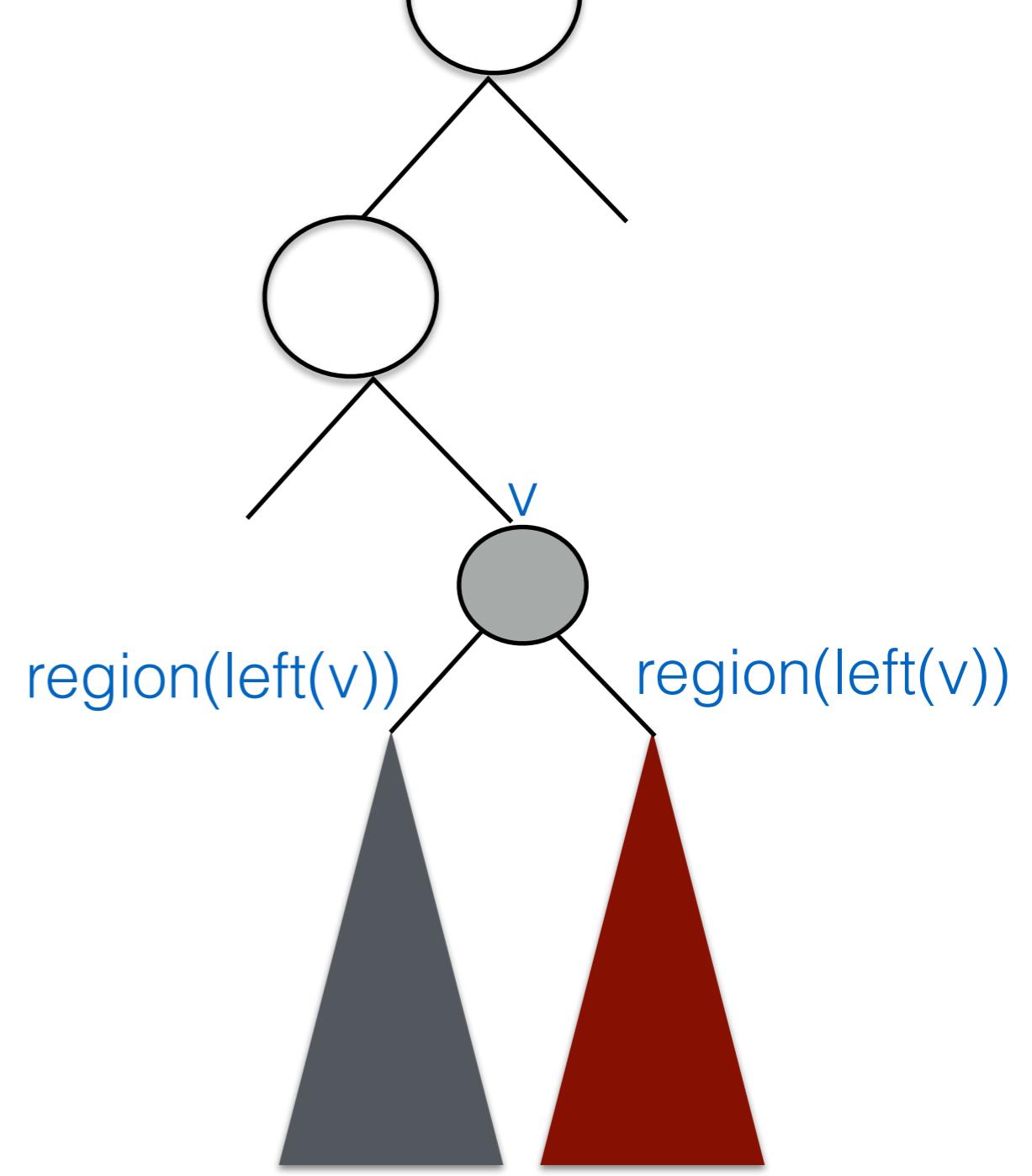
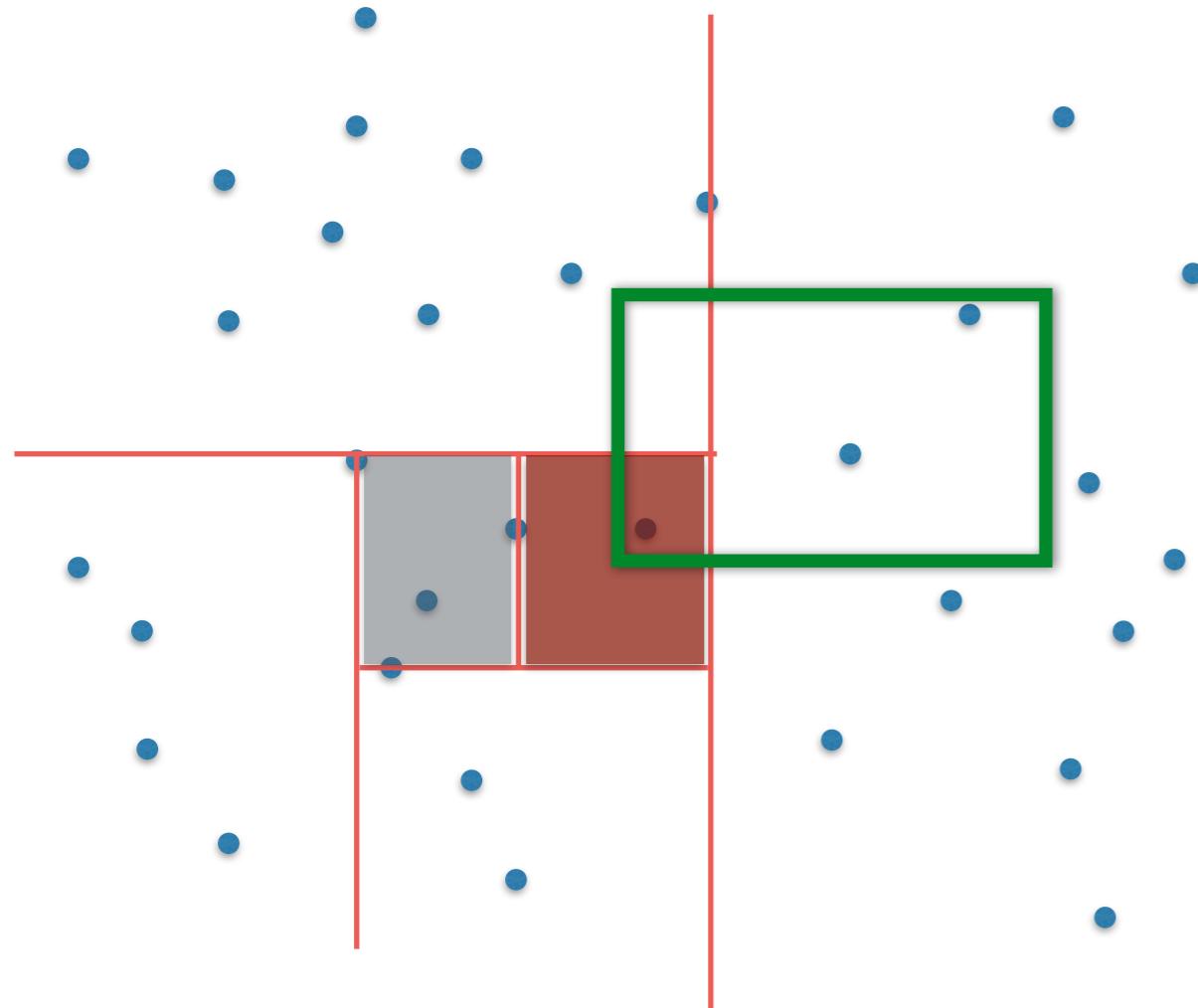
Case 1:range intersects both children

Case 2



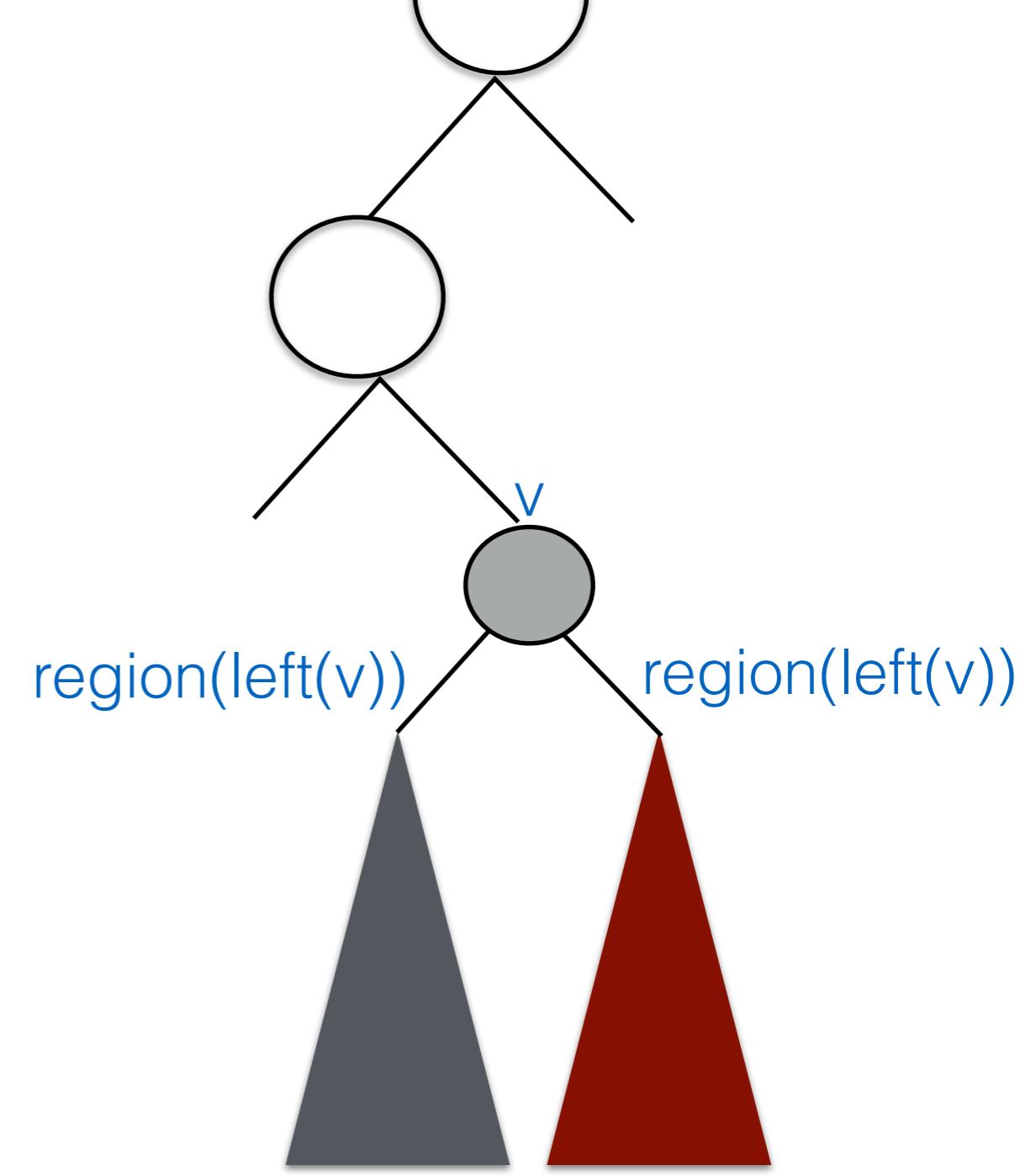
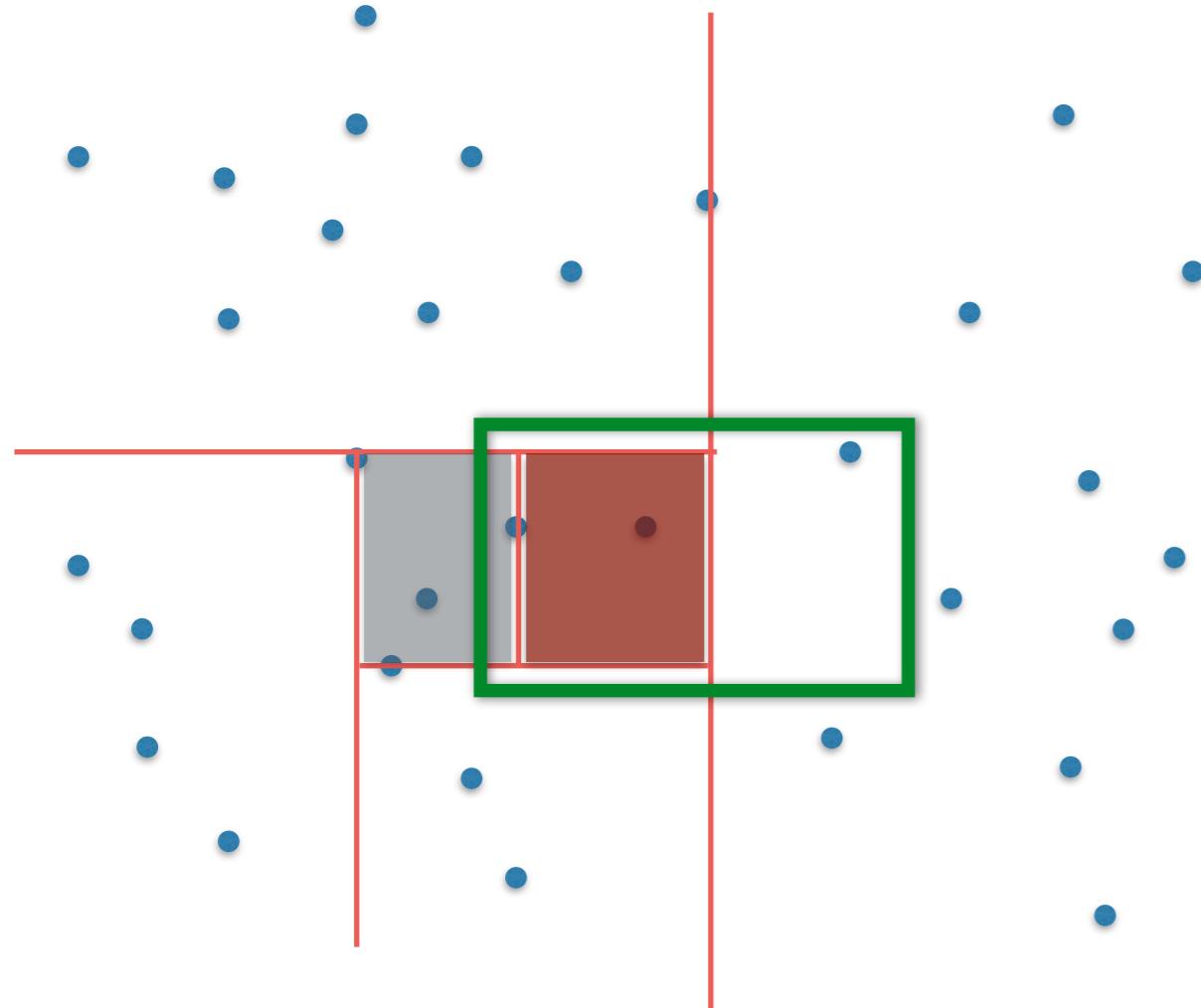
Case 2: range intersects only one child

Range queries: general idea



Case 2: range intersects only one child

Range queries: general idea



Case 3: child completely contained in range

Algorithm SEARCHKDTREE(v, R)

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

Output. All points 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** SEARCHKDTREE($lc(v), R$)
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** SEARCHKDTREE($rc(v), R$)

How long does a range query take?

To analyze the time to answer a range query we'll look at the nodes visited in the tree

Here a standard analysis does not work..

If at any node we would visit **one** child => $O(\lg n)$

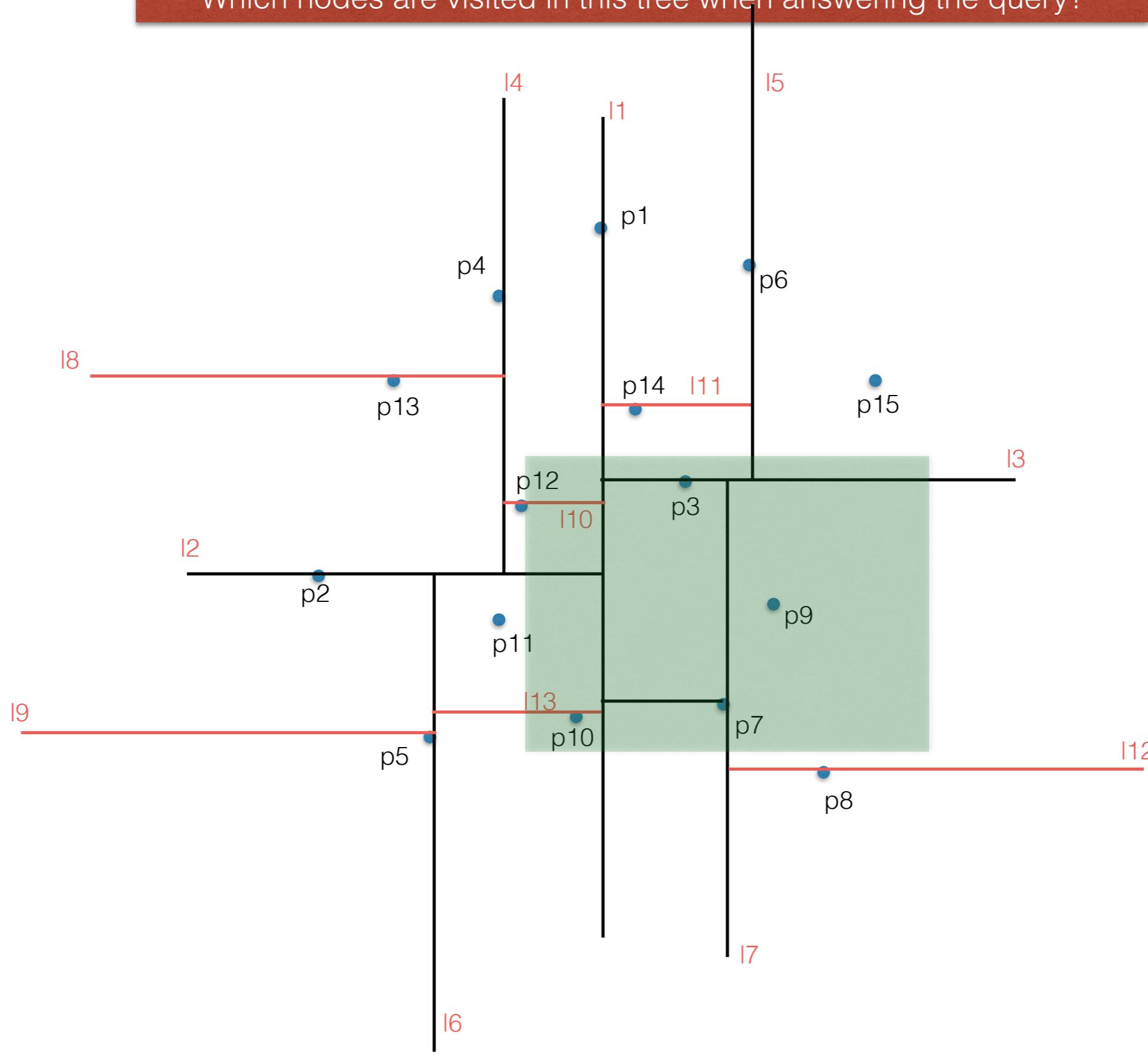
If at any node we would visit **both** children => $O(n)$

Here we are in between

We visit the children intersected by the query range, which can be one or both

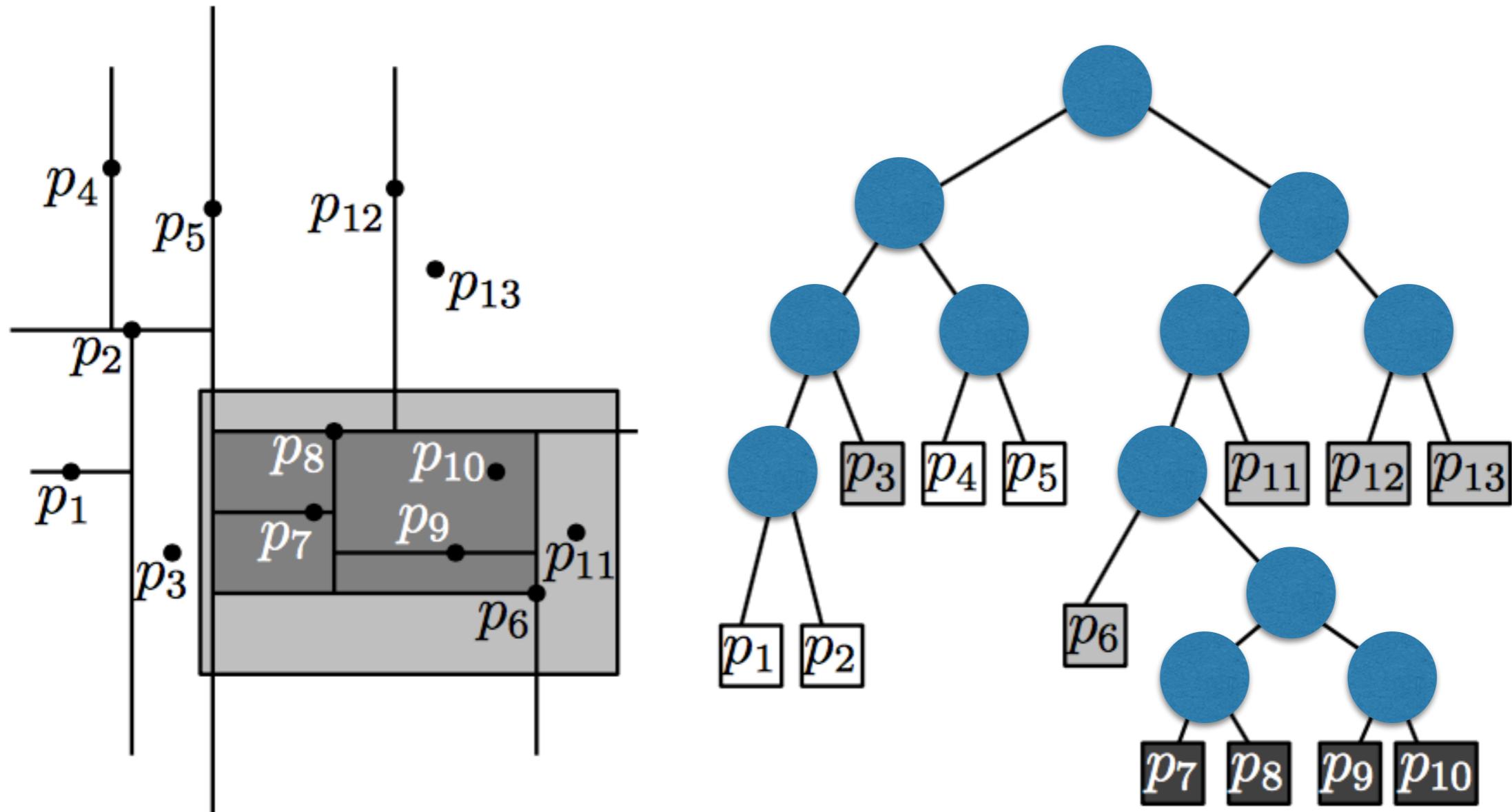
To analyze the time to answer a range query we'll look at the nodes visited in the tree when answering a query

Which nodes are visited in this tree when answering the query?



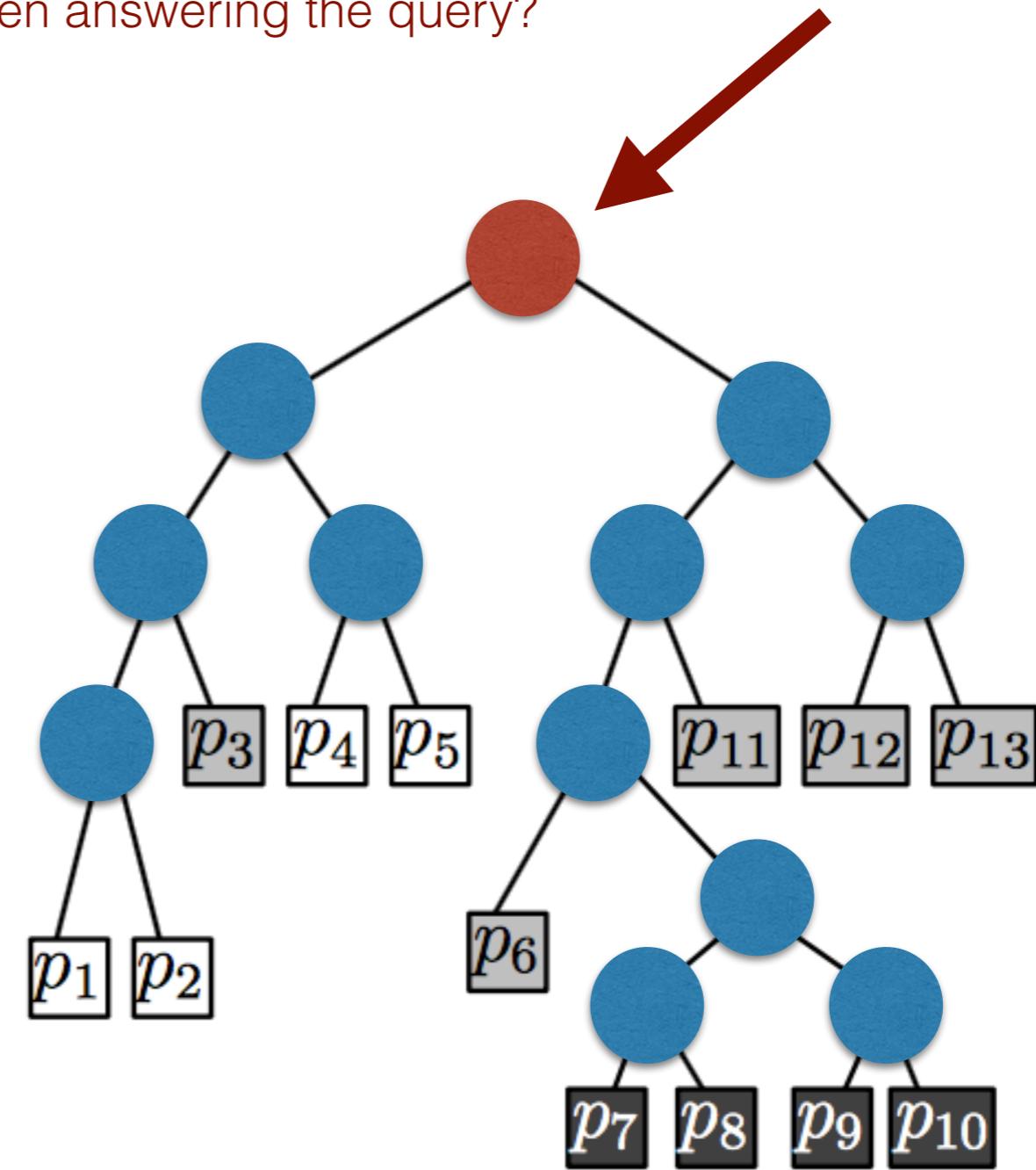
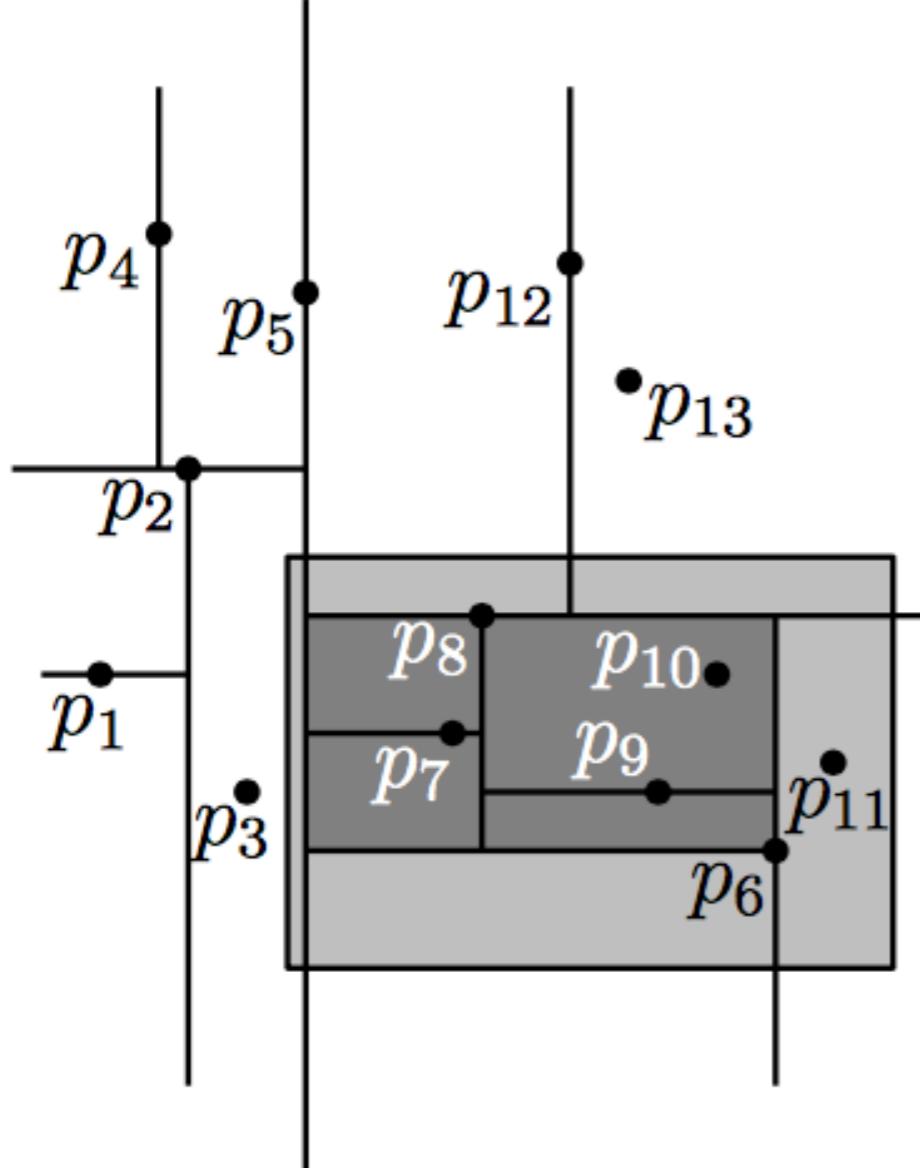
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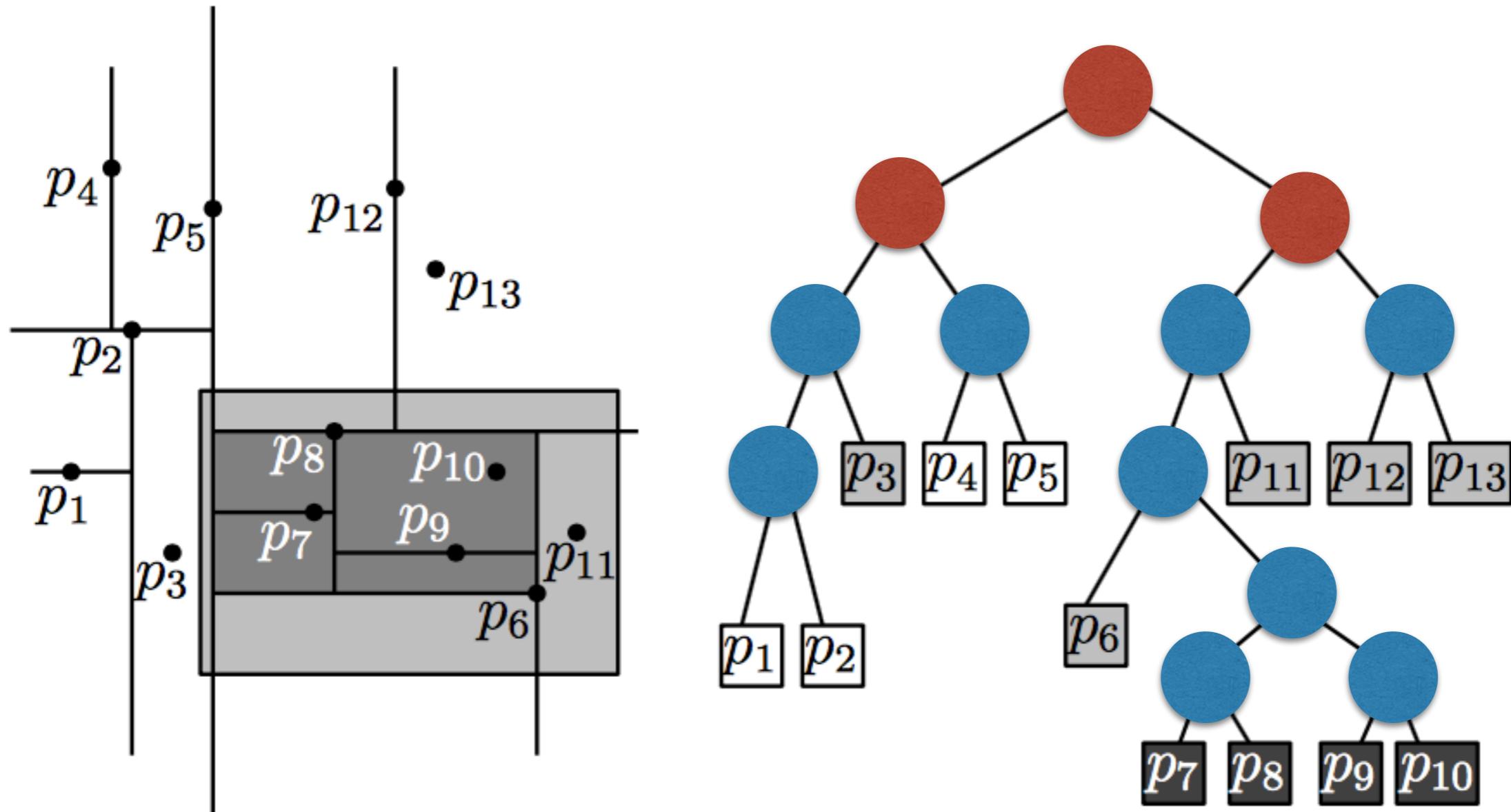
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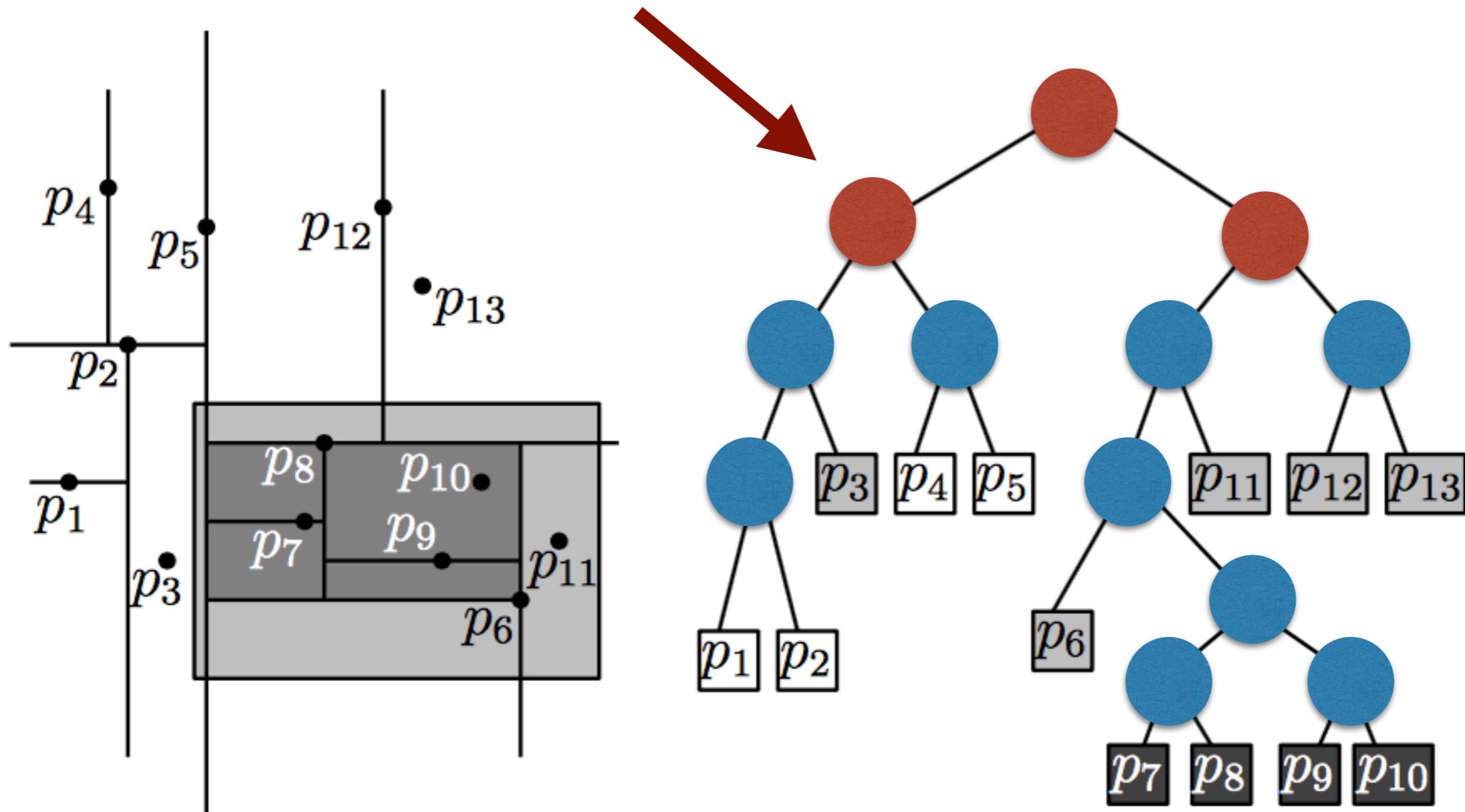
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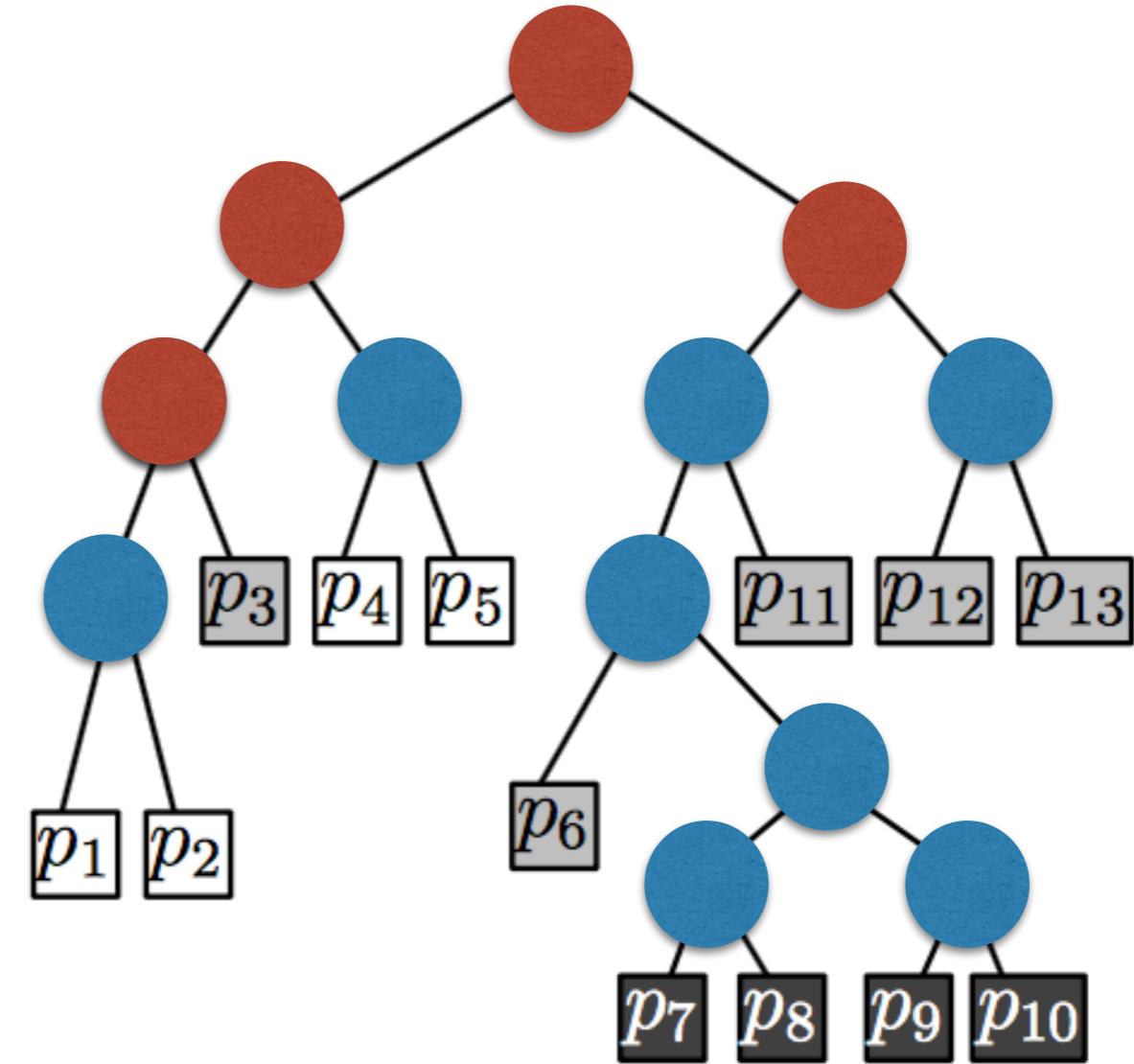
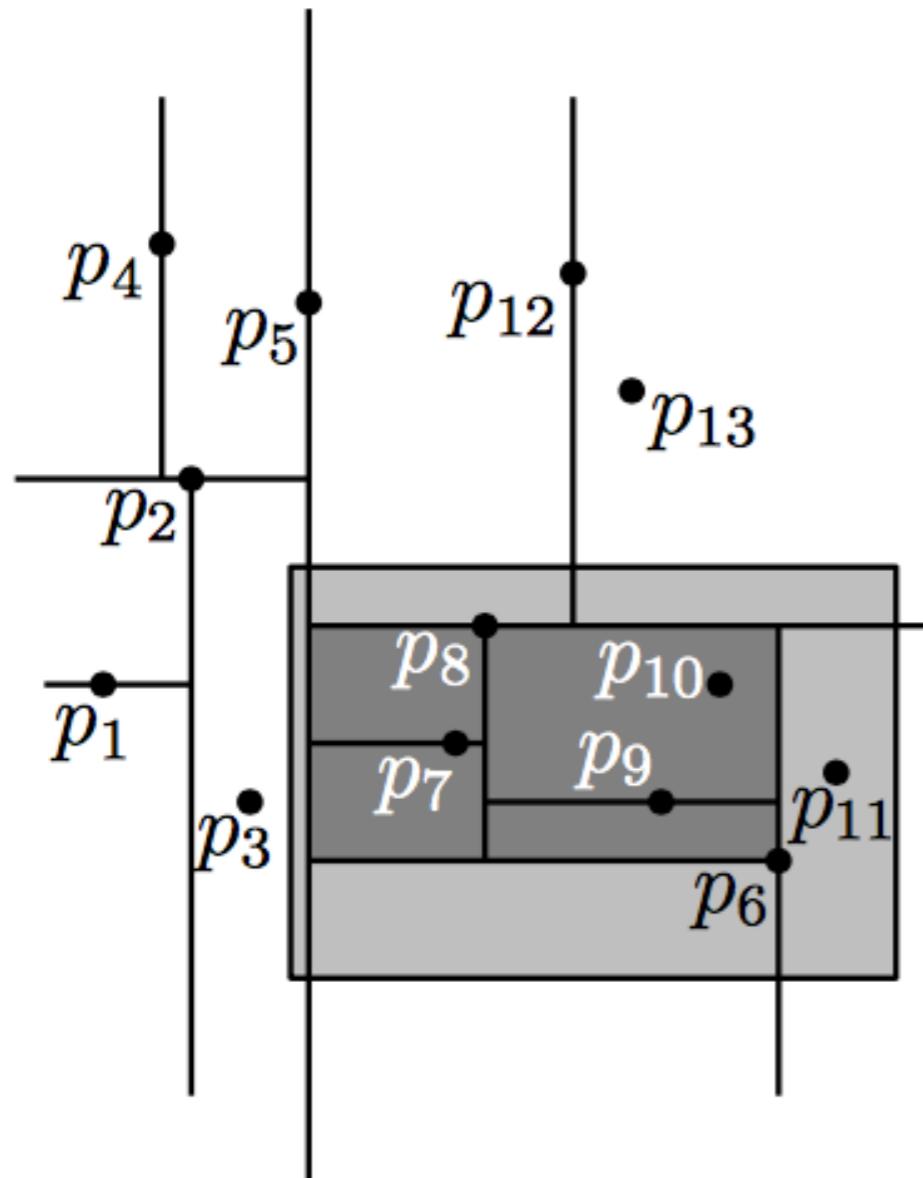
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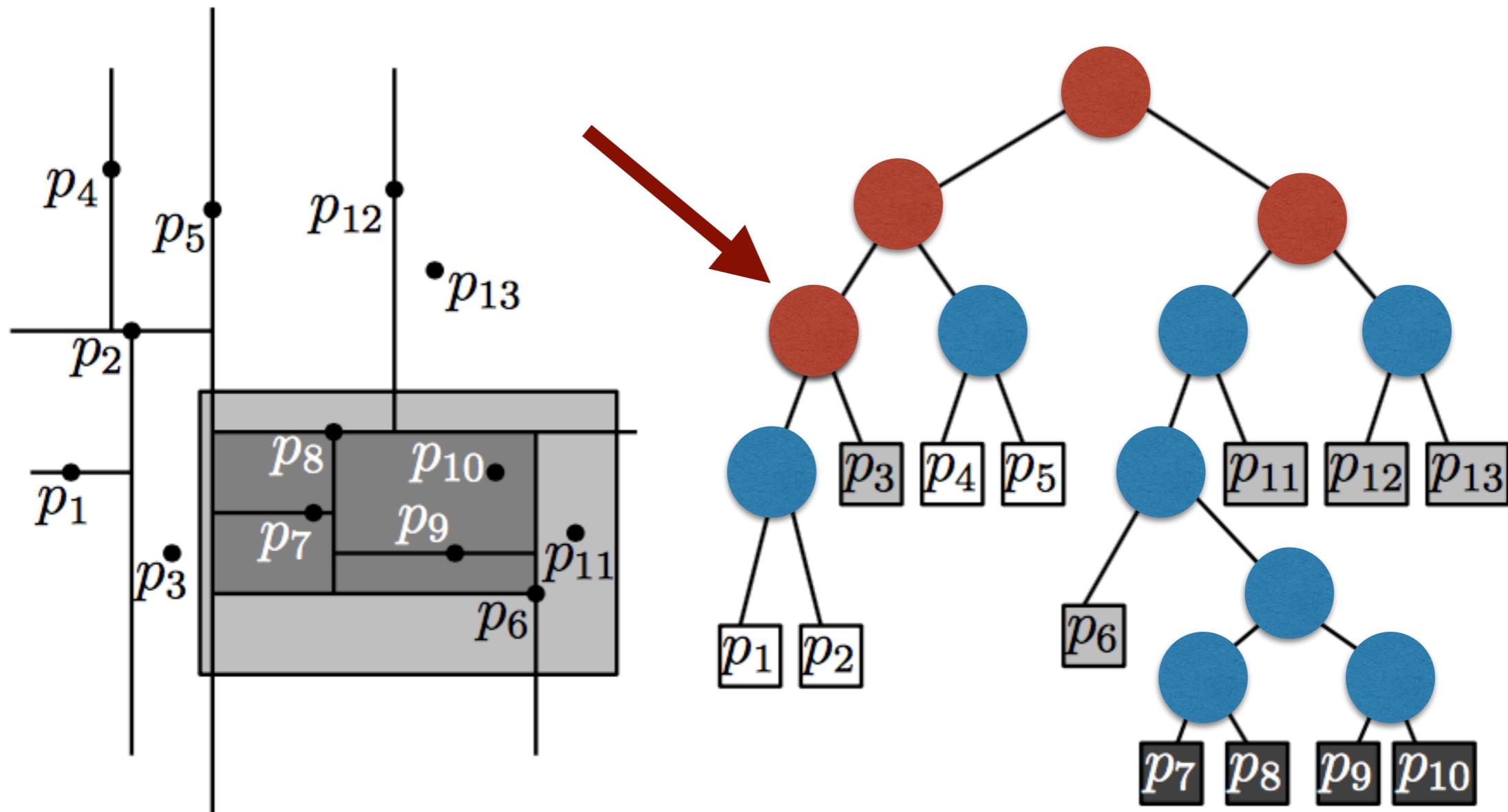
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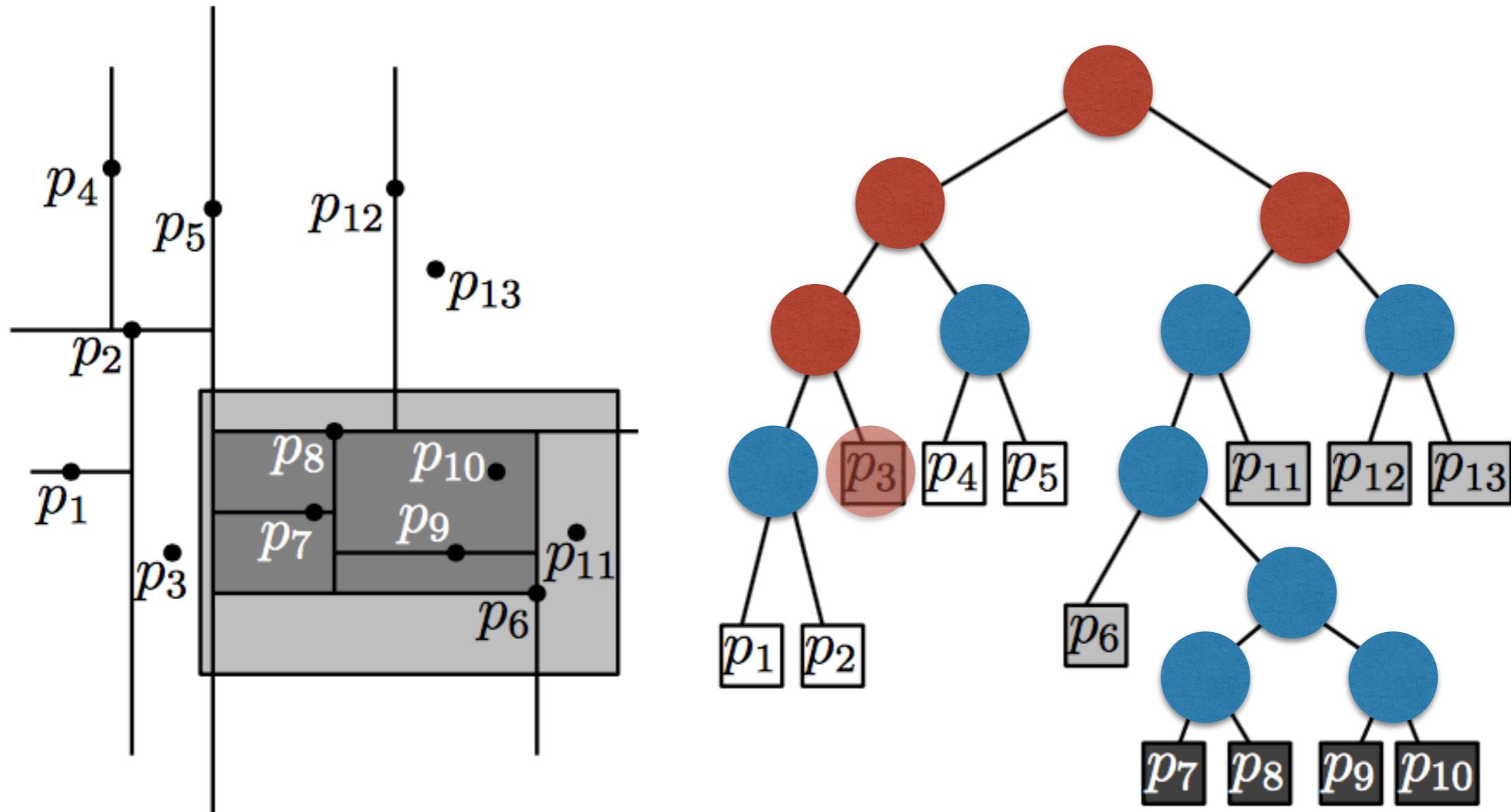
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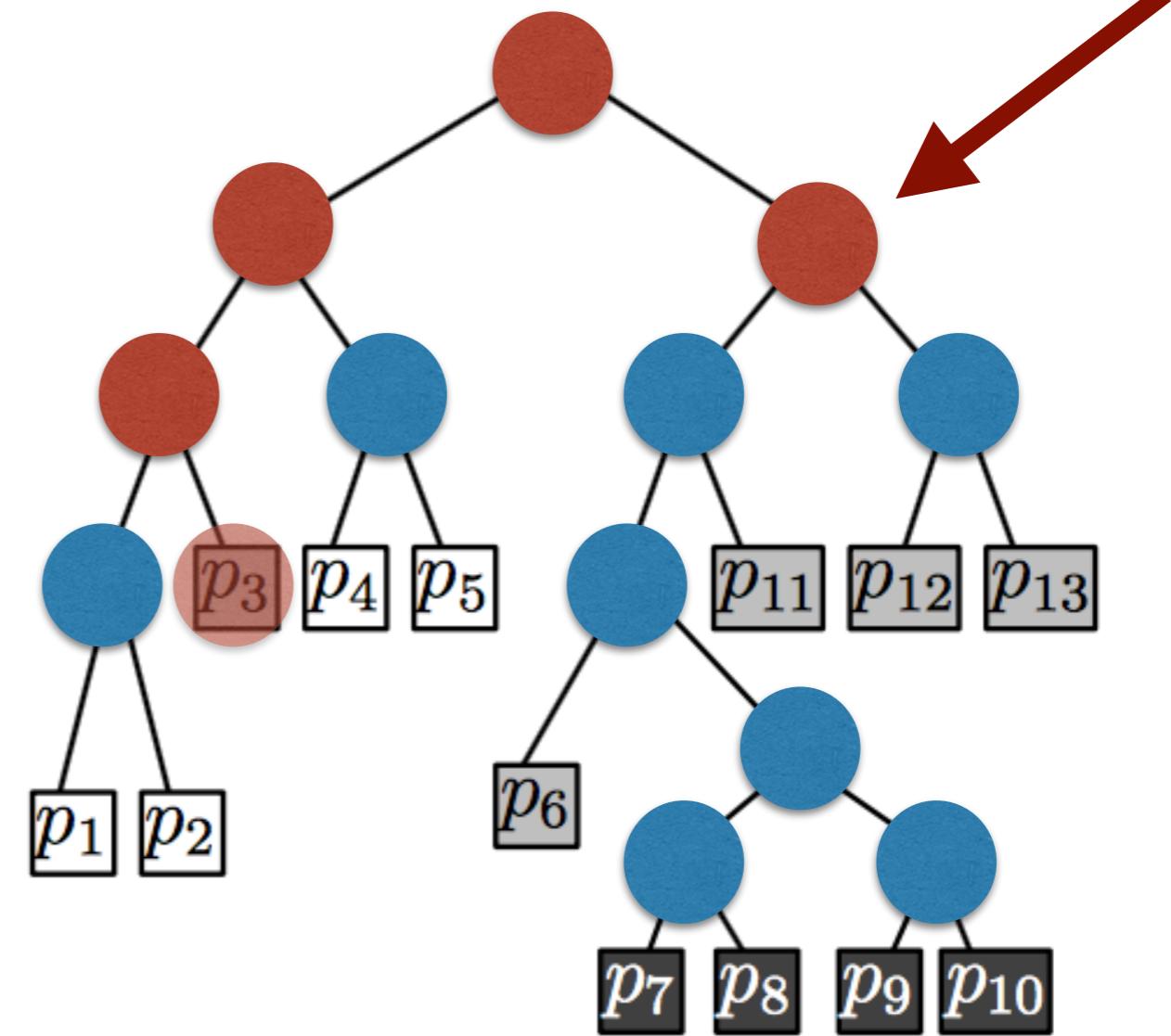
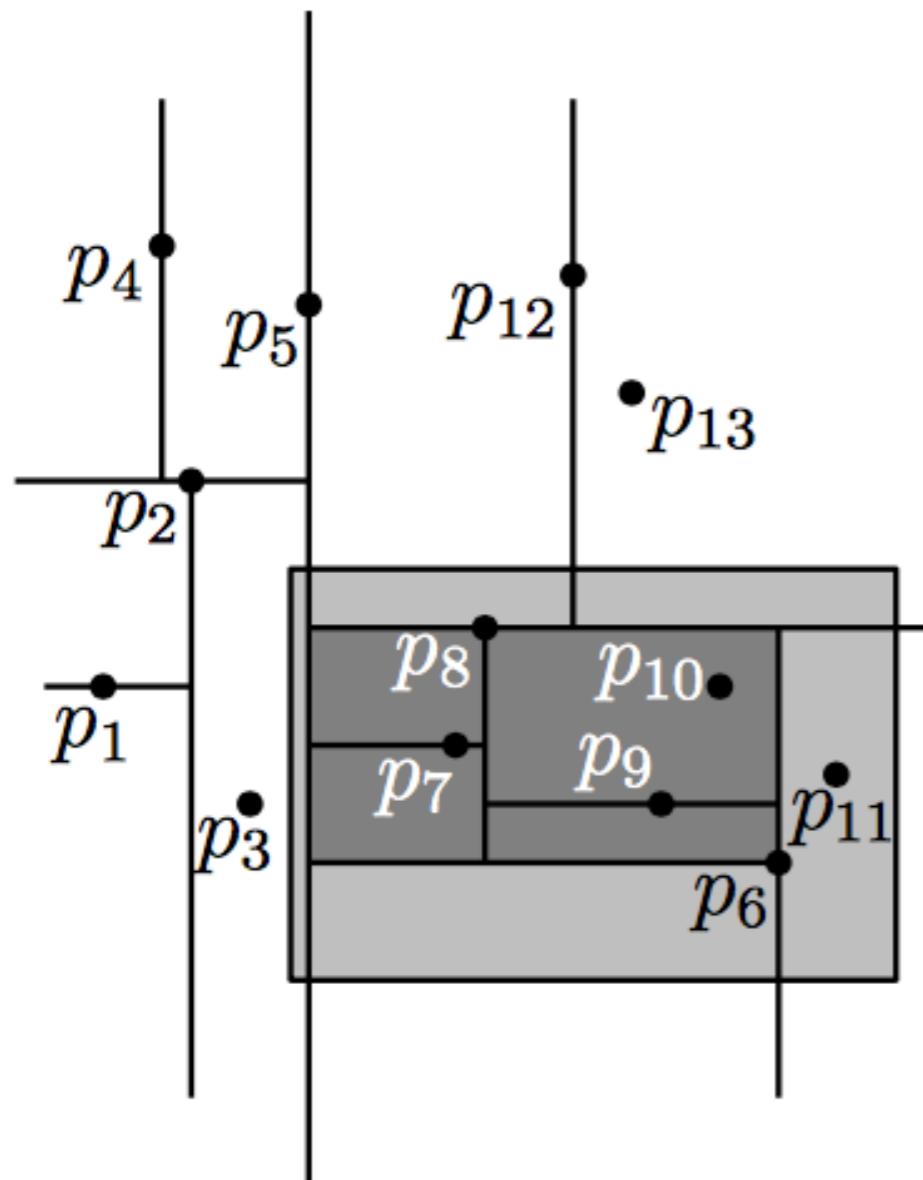
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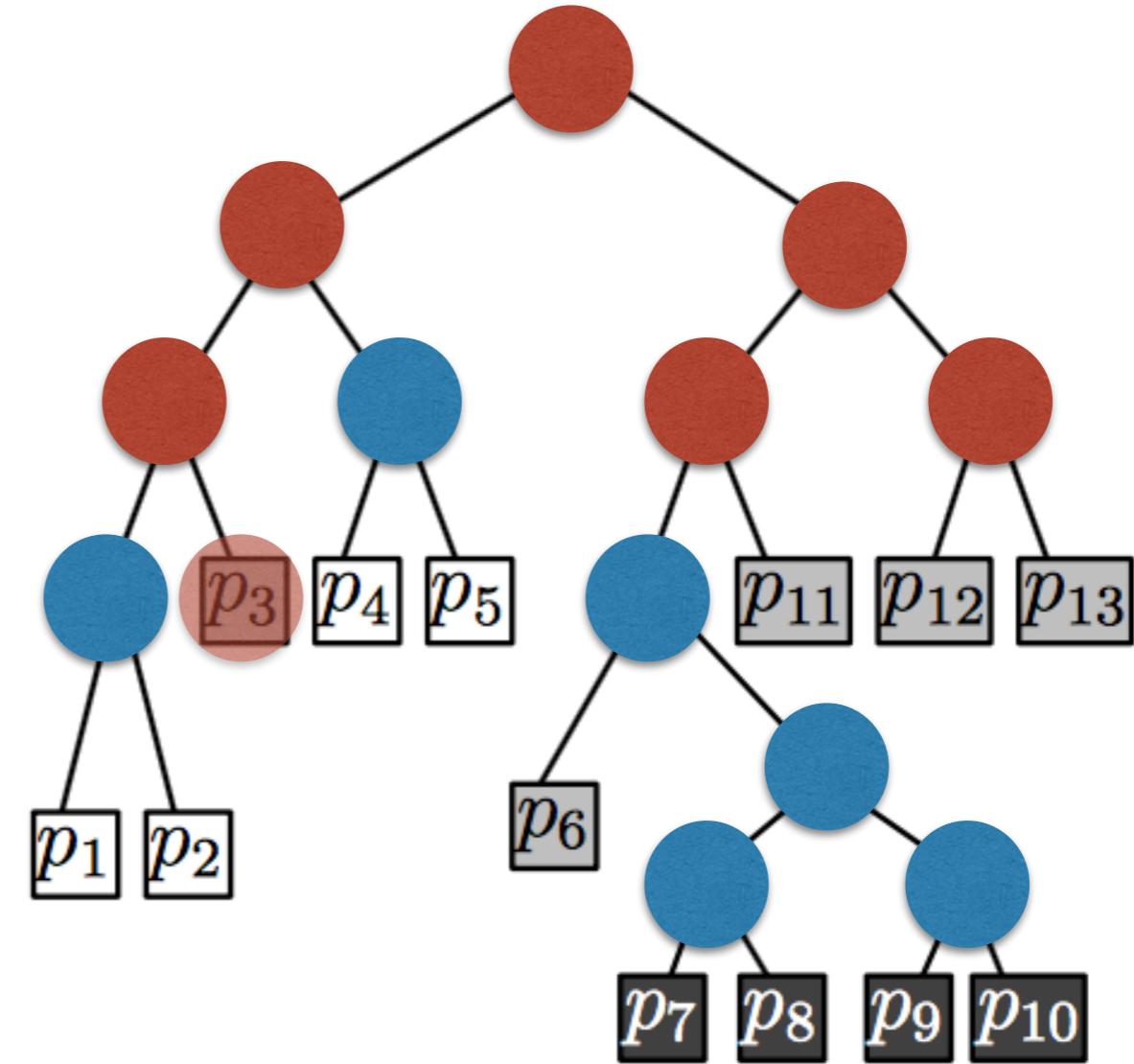
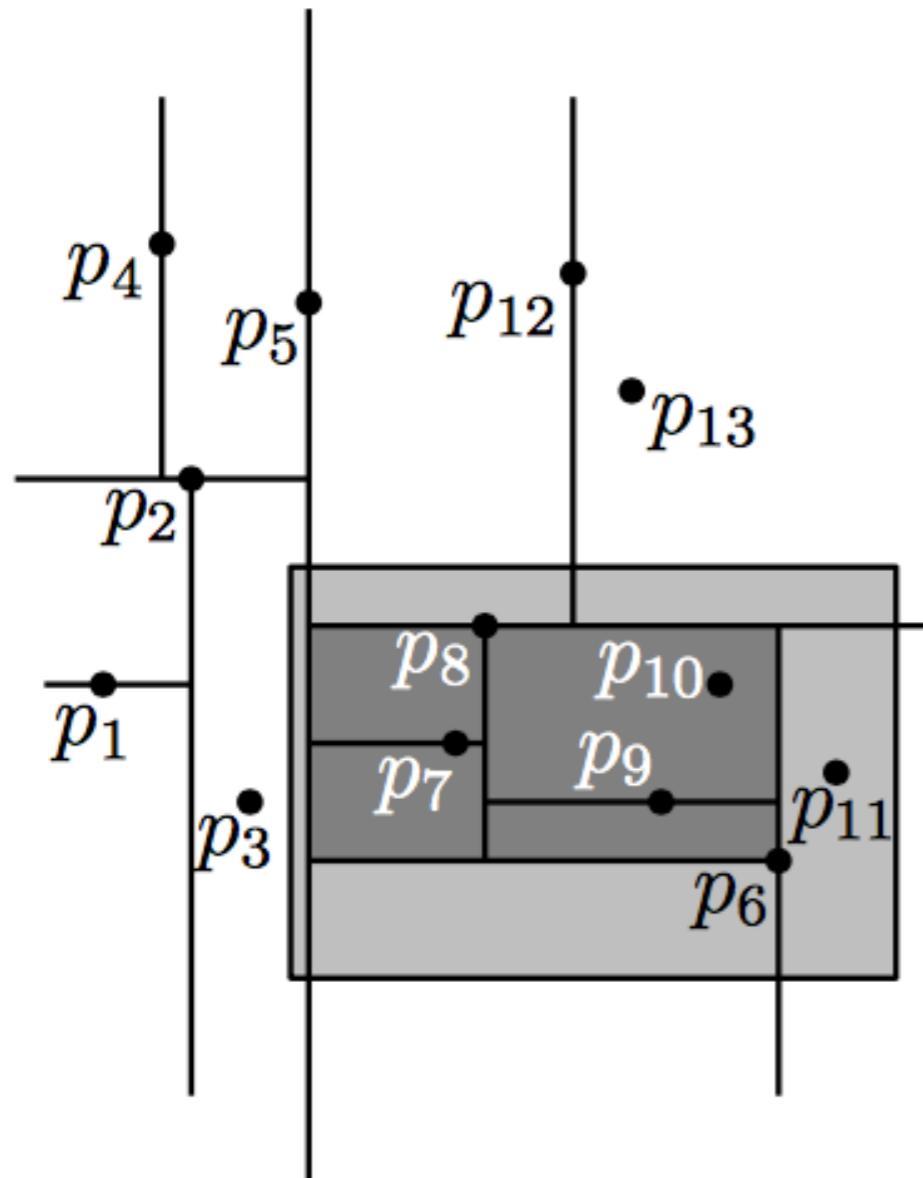
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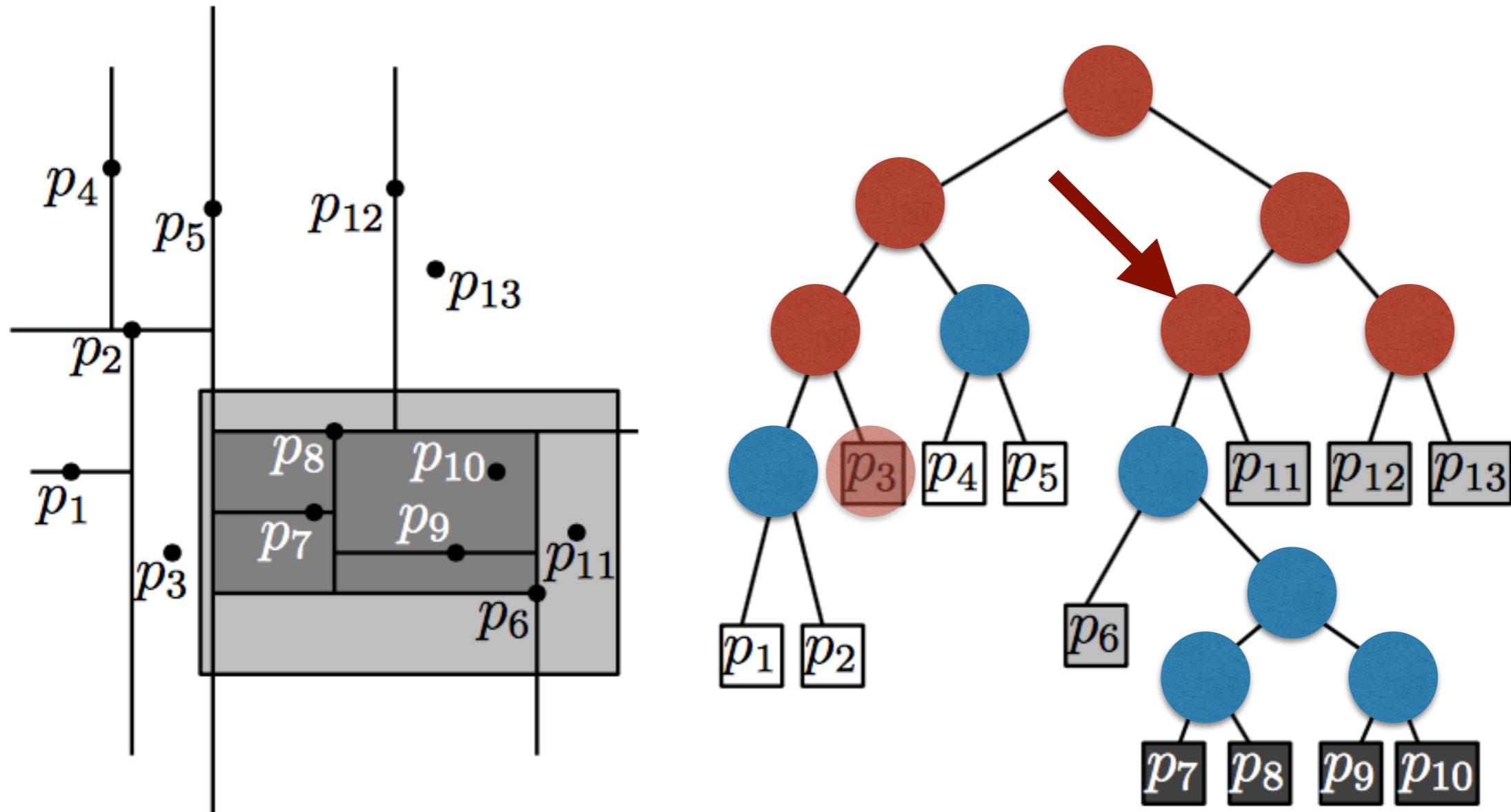
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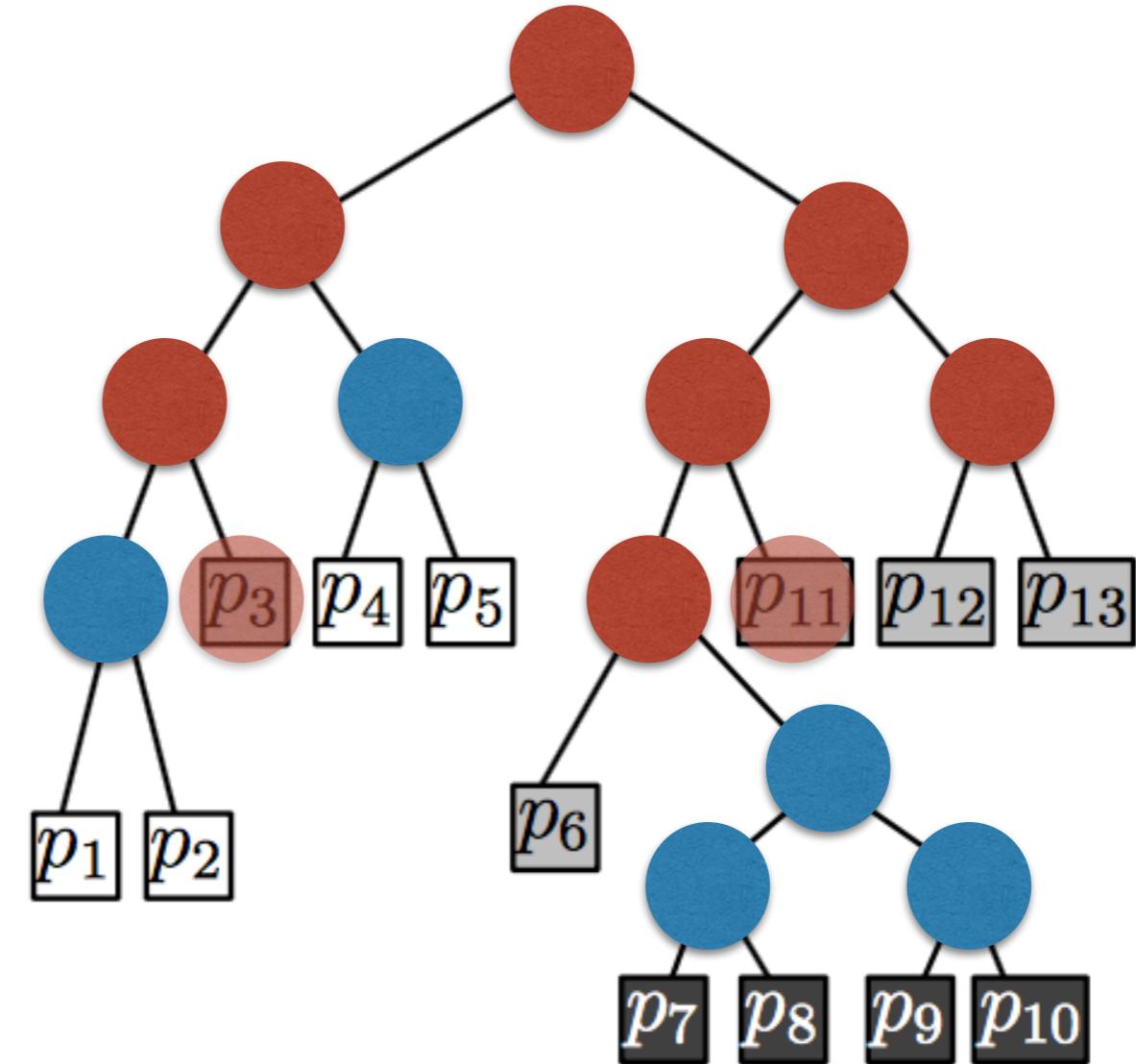
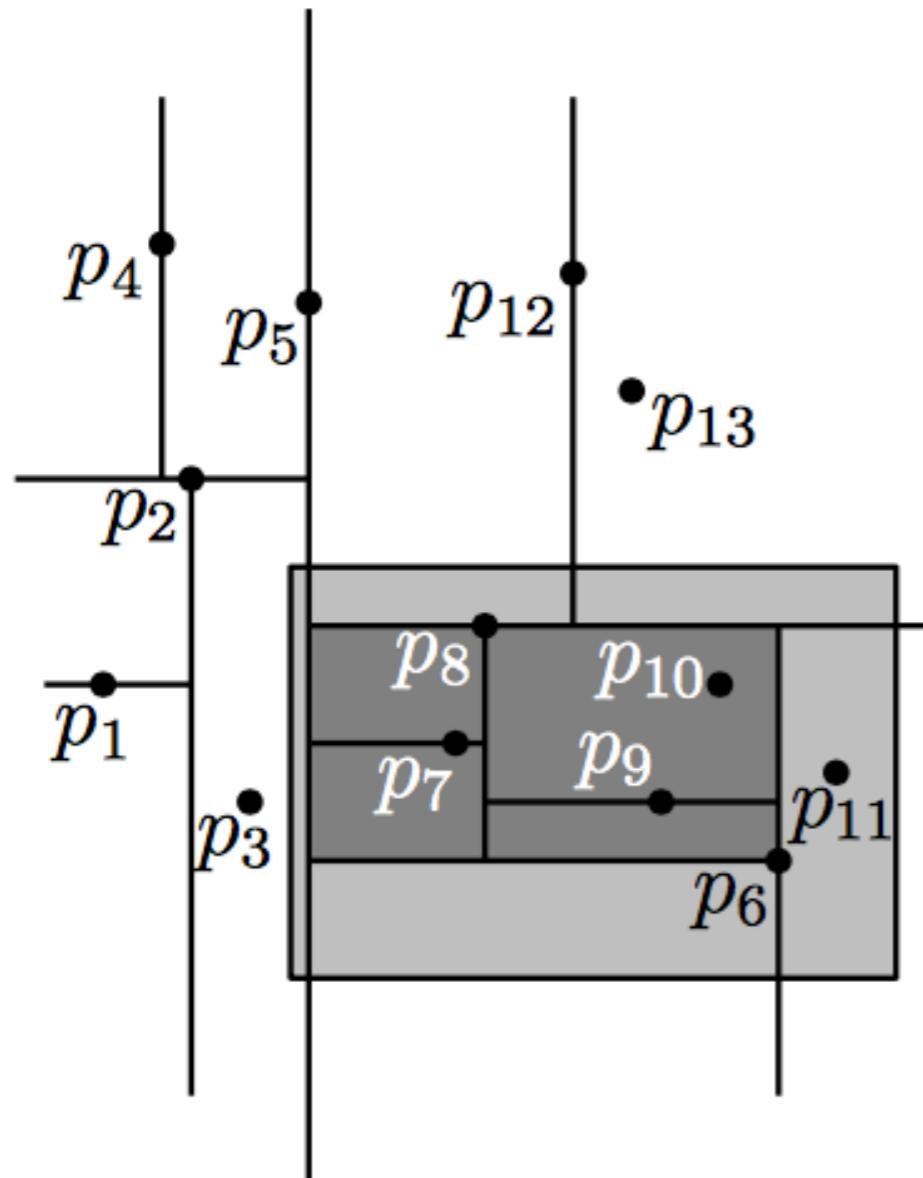
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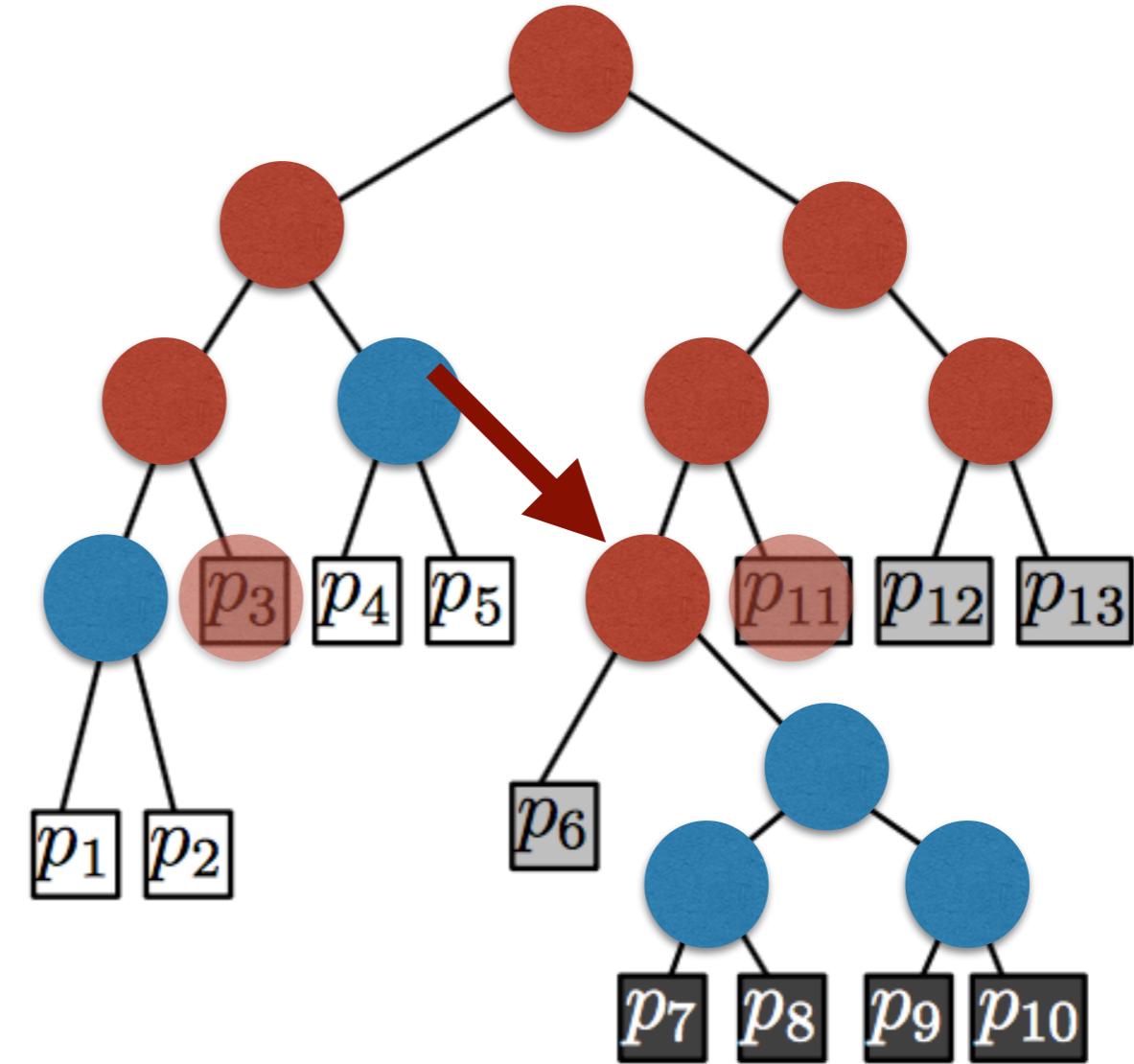
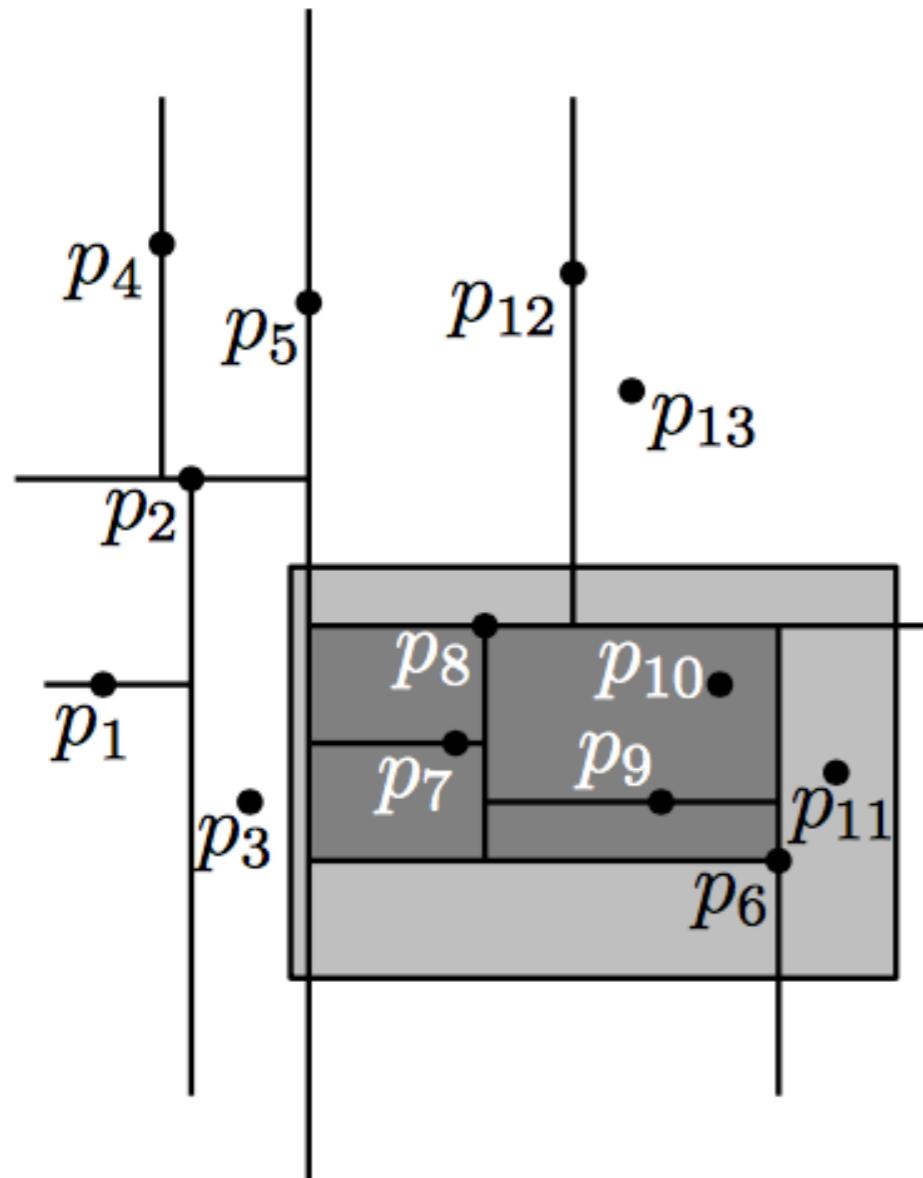
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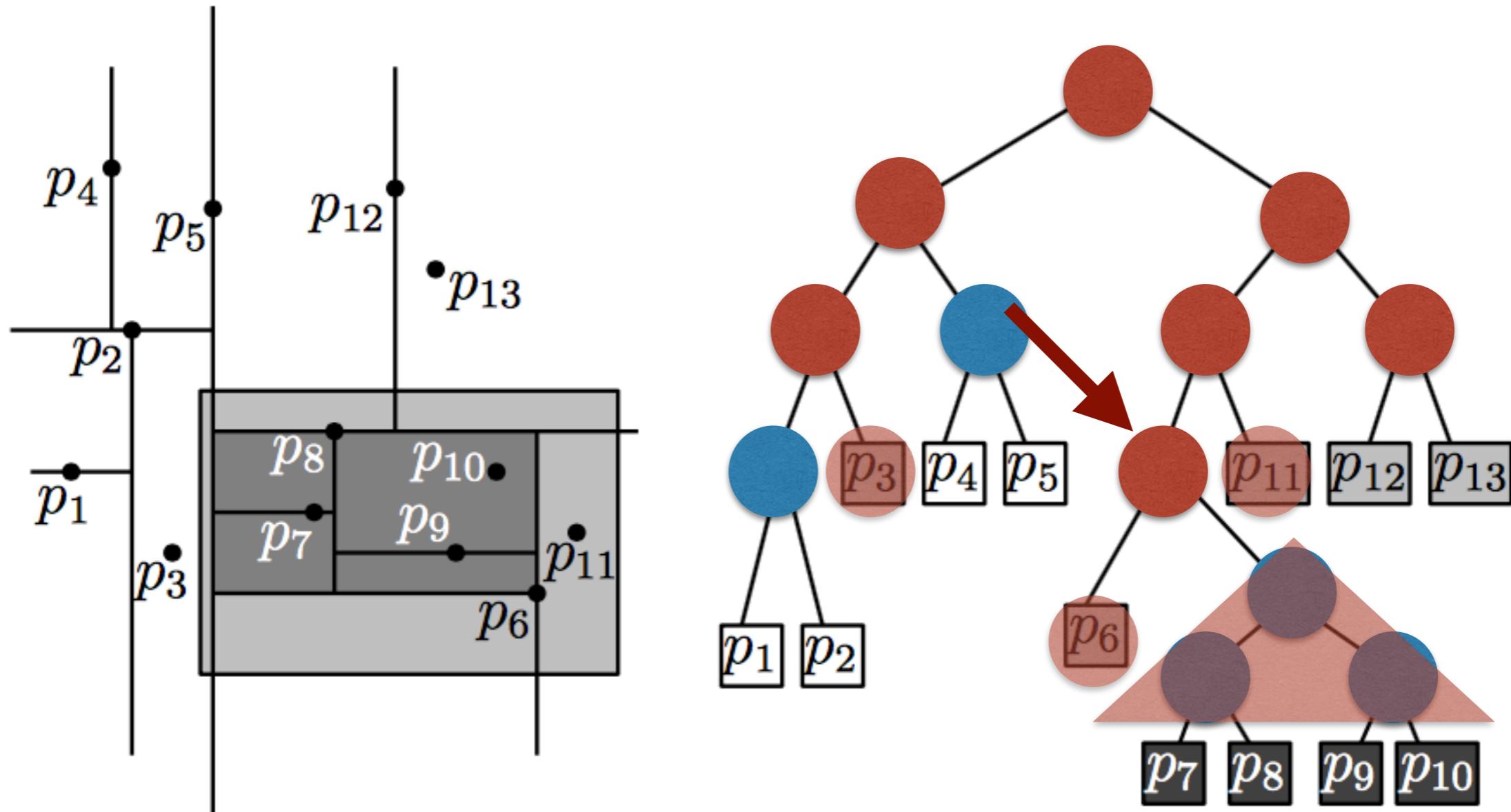
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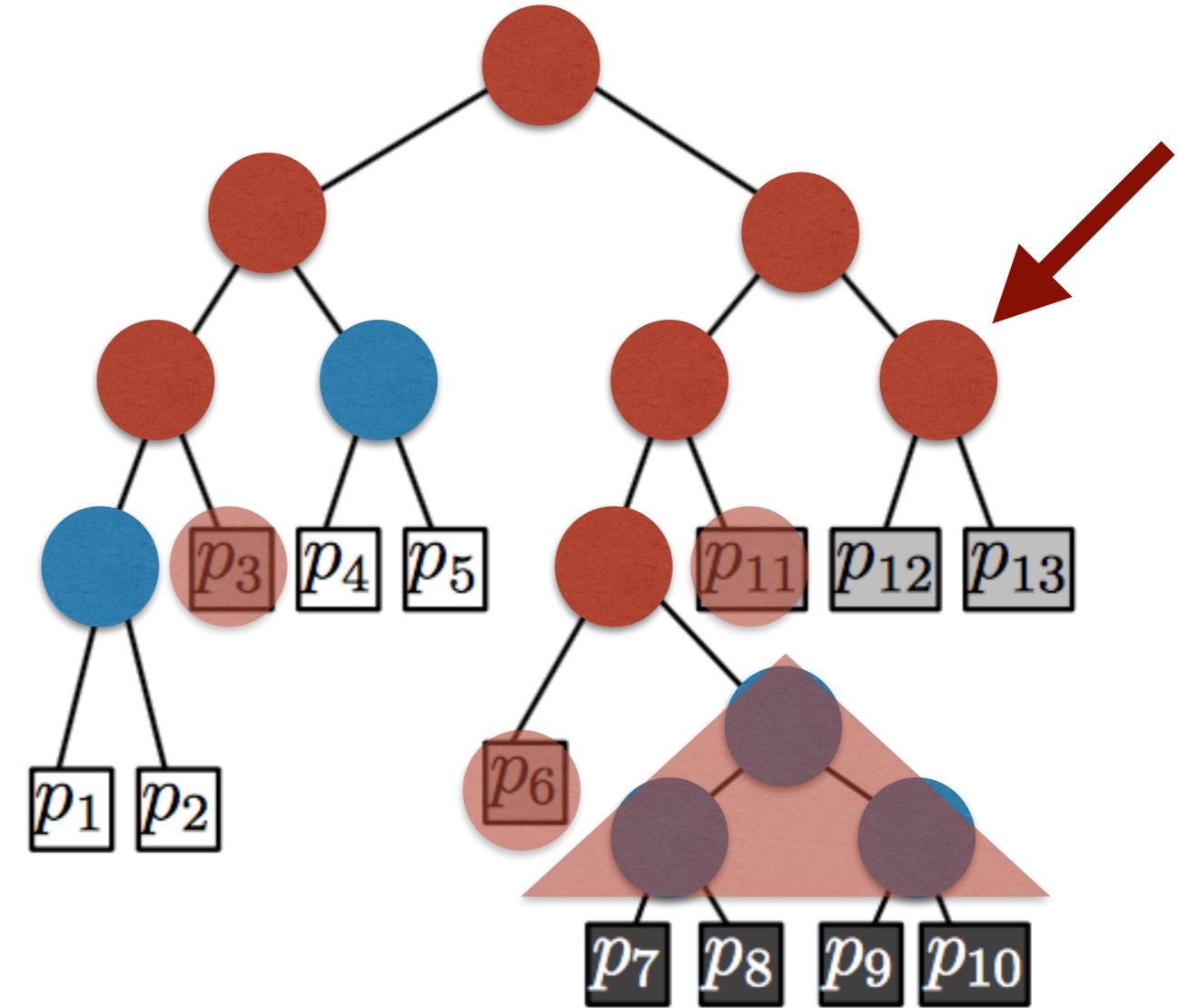
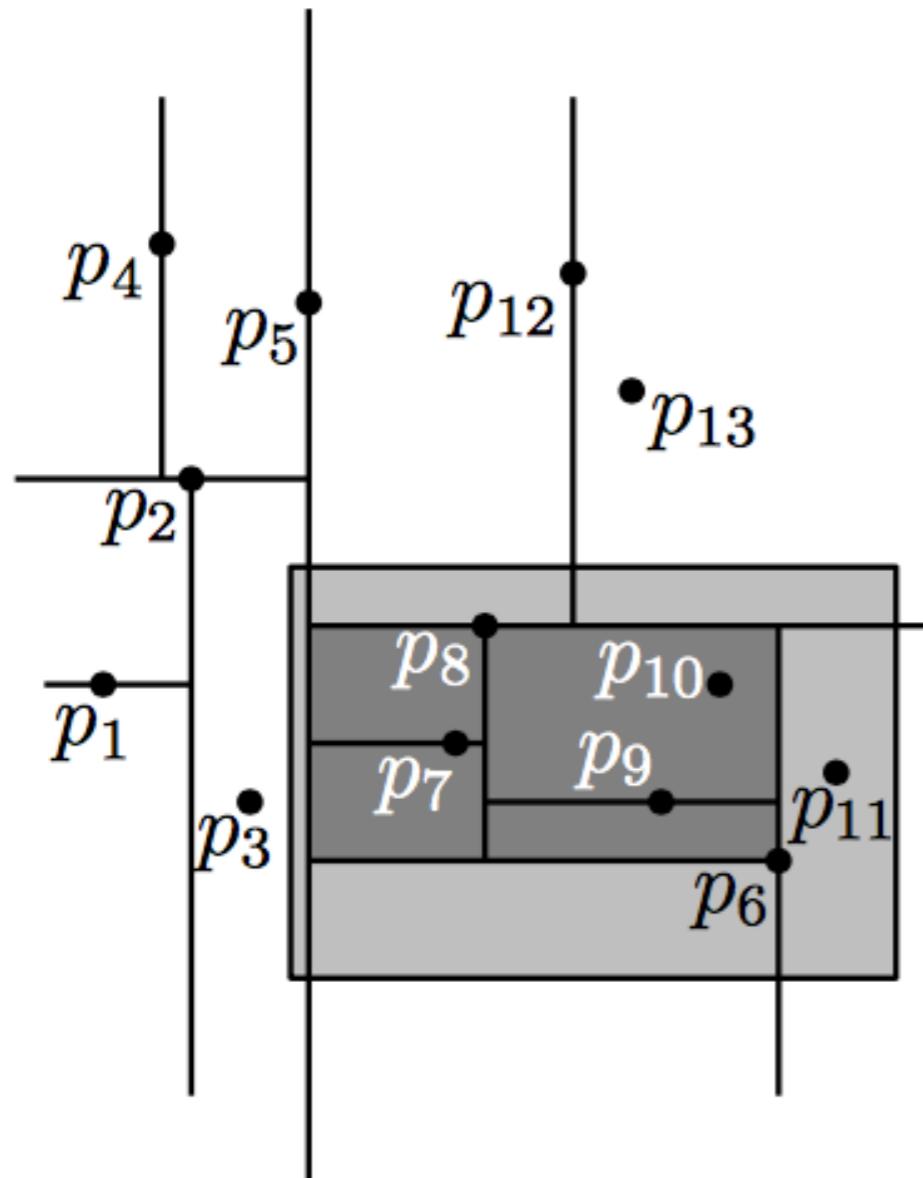
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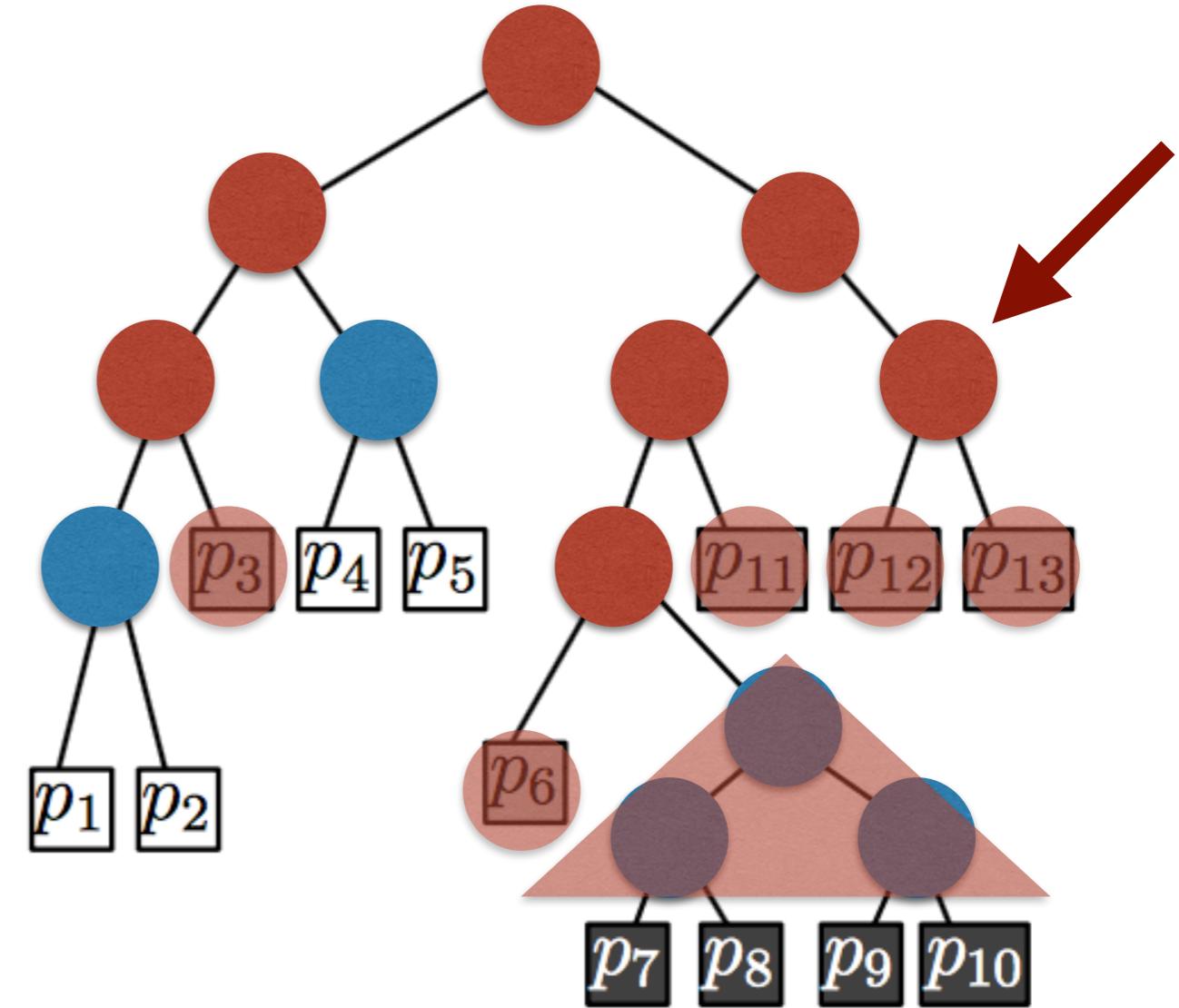
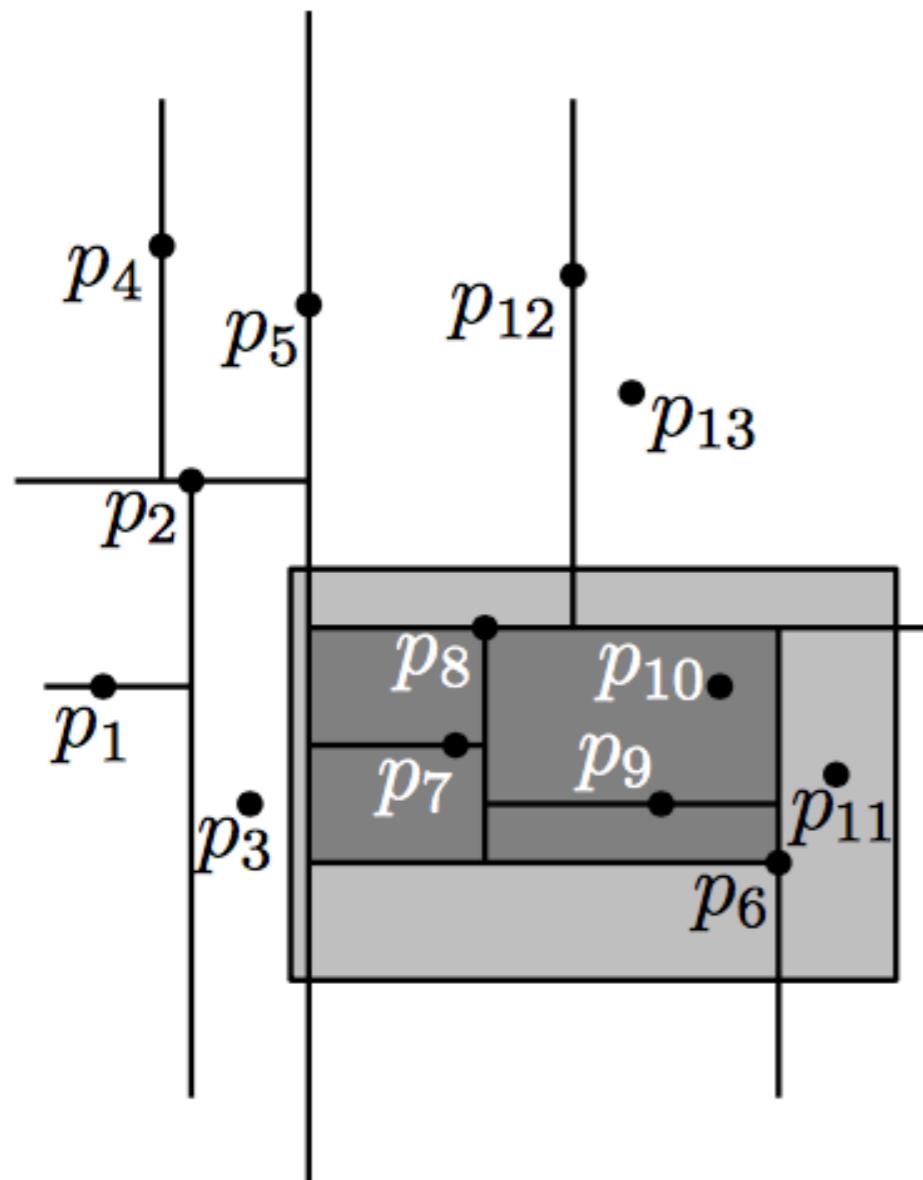
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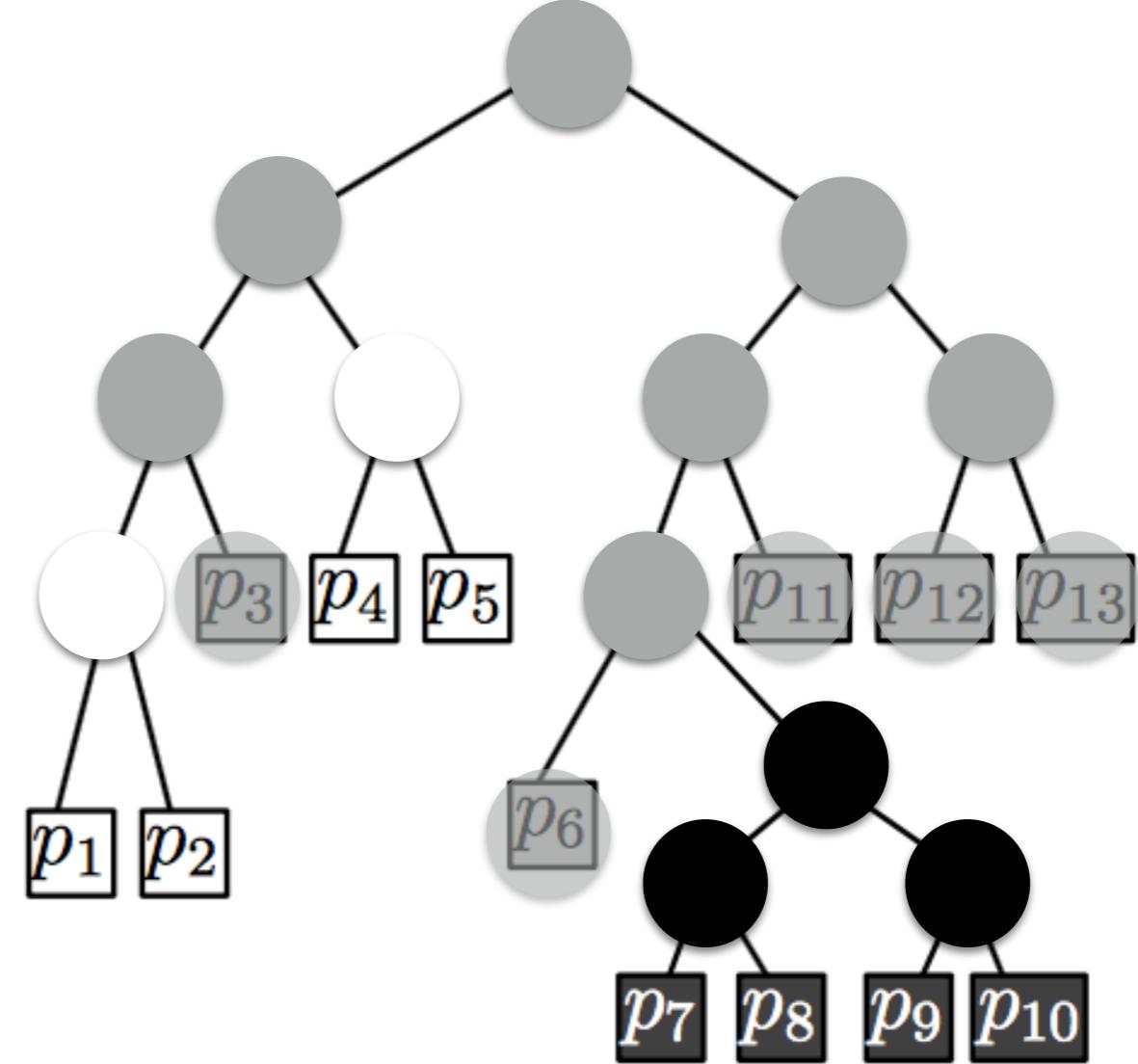
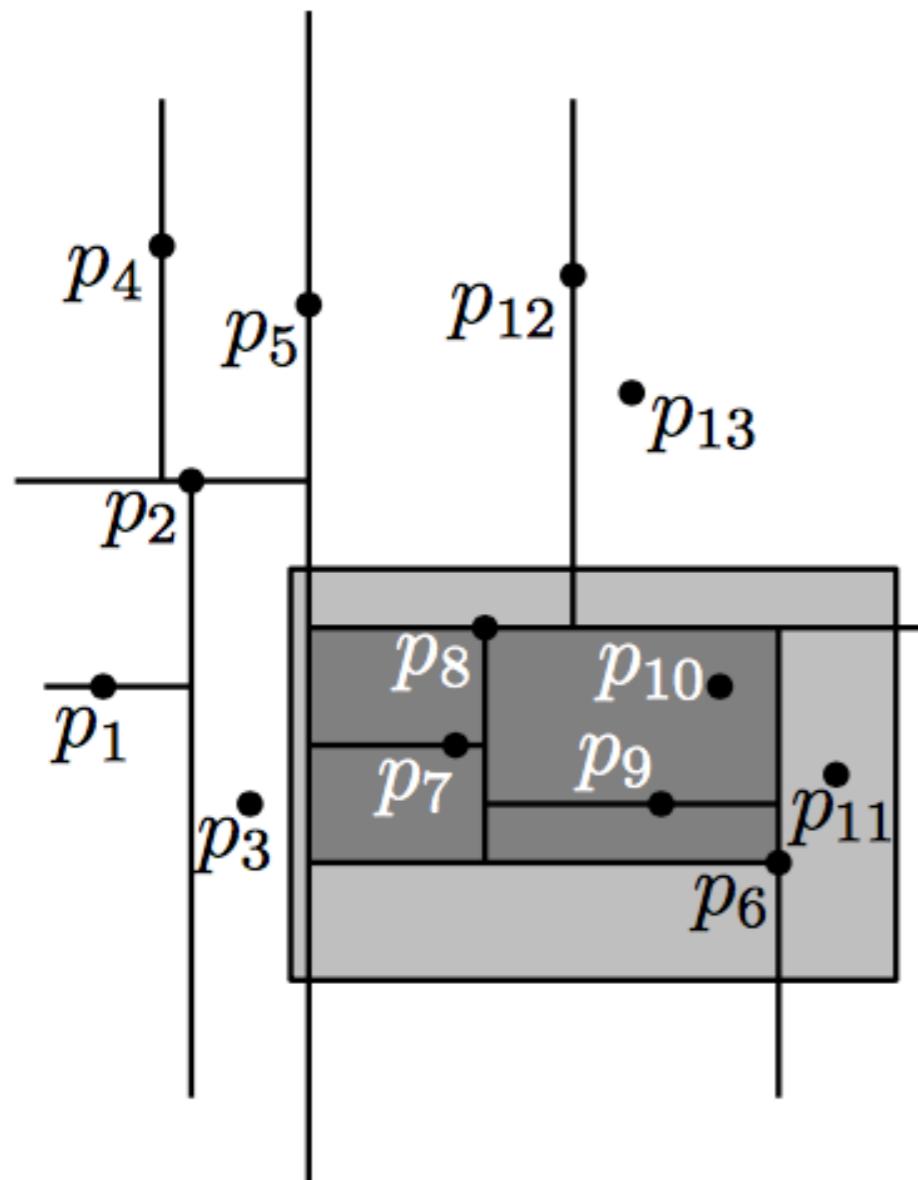


To analyze the time to answer a range query we'll look at the nodes visited in the tree

Which nodes are visited in this tree when answering the query?



- nodes never visited by the query
- visited by the query, but unclear if they lead to output
- visited by the query, whole subtree is output





- nodes never visited by the query
- visited by the query, but unclear if they lead to output
- visited by the query, whole subtree is output

Consider $\text{region}(\text{node})$ and how it intersects range R



nodes never visited by the query

visited by the query, but unclear if they lead to output

visited by the query, whole subtree is output

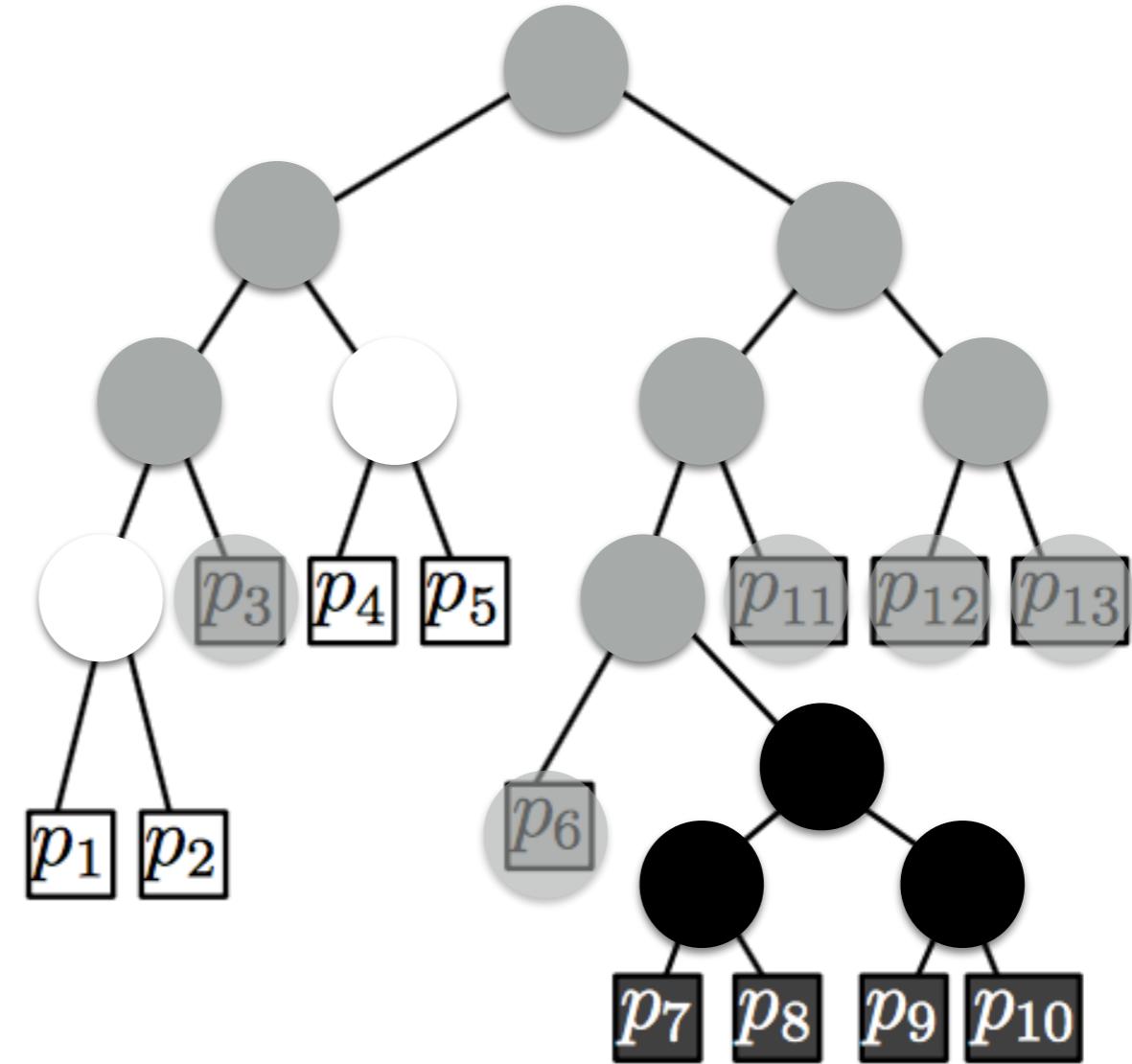
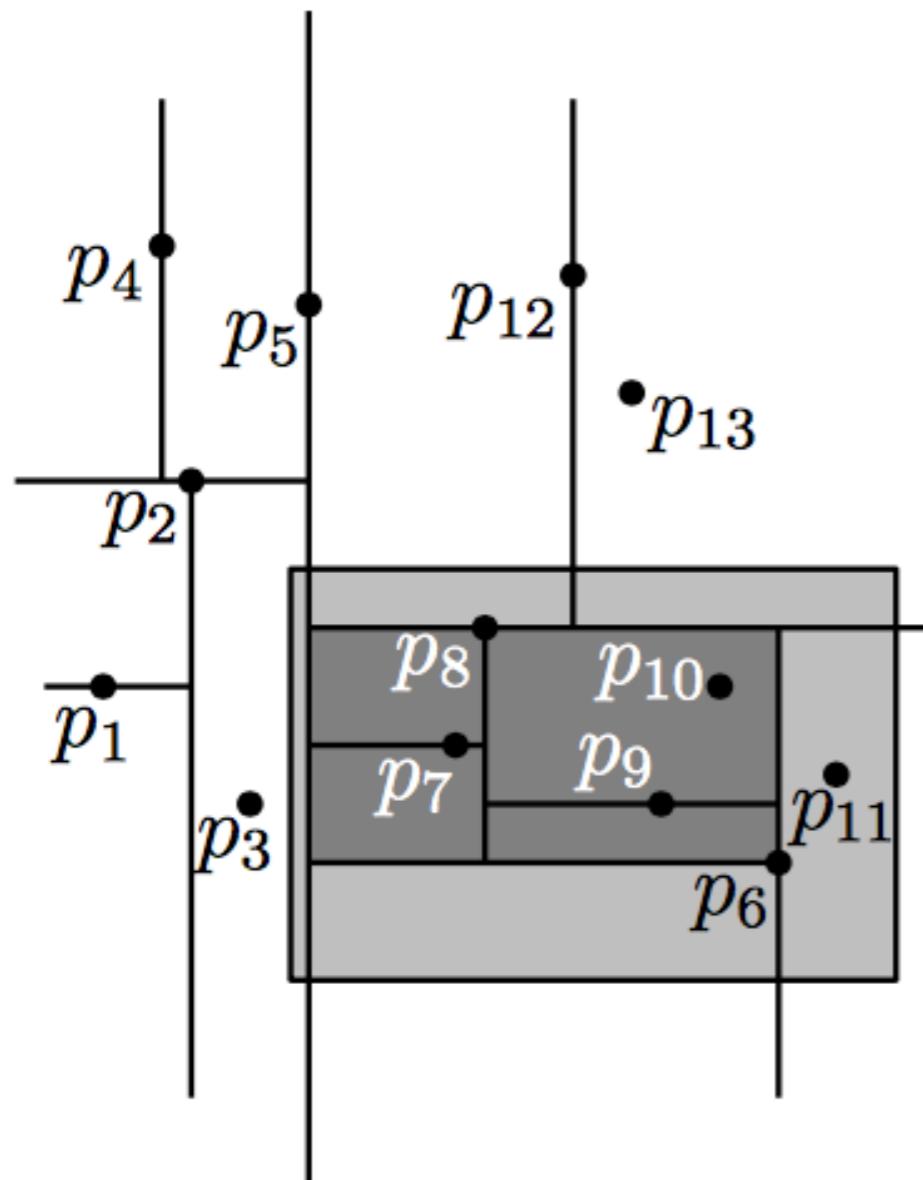
R does not intersect region(v)

Consider $\text{region}(\text{node})$ and how it intersects range R

region(v) is contained in R

R intersects region(v), but region(v) not contained in R

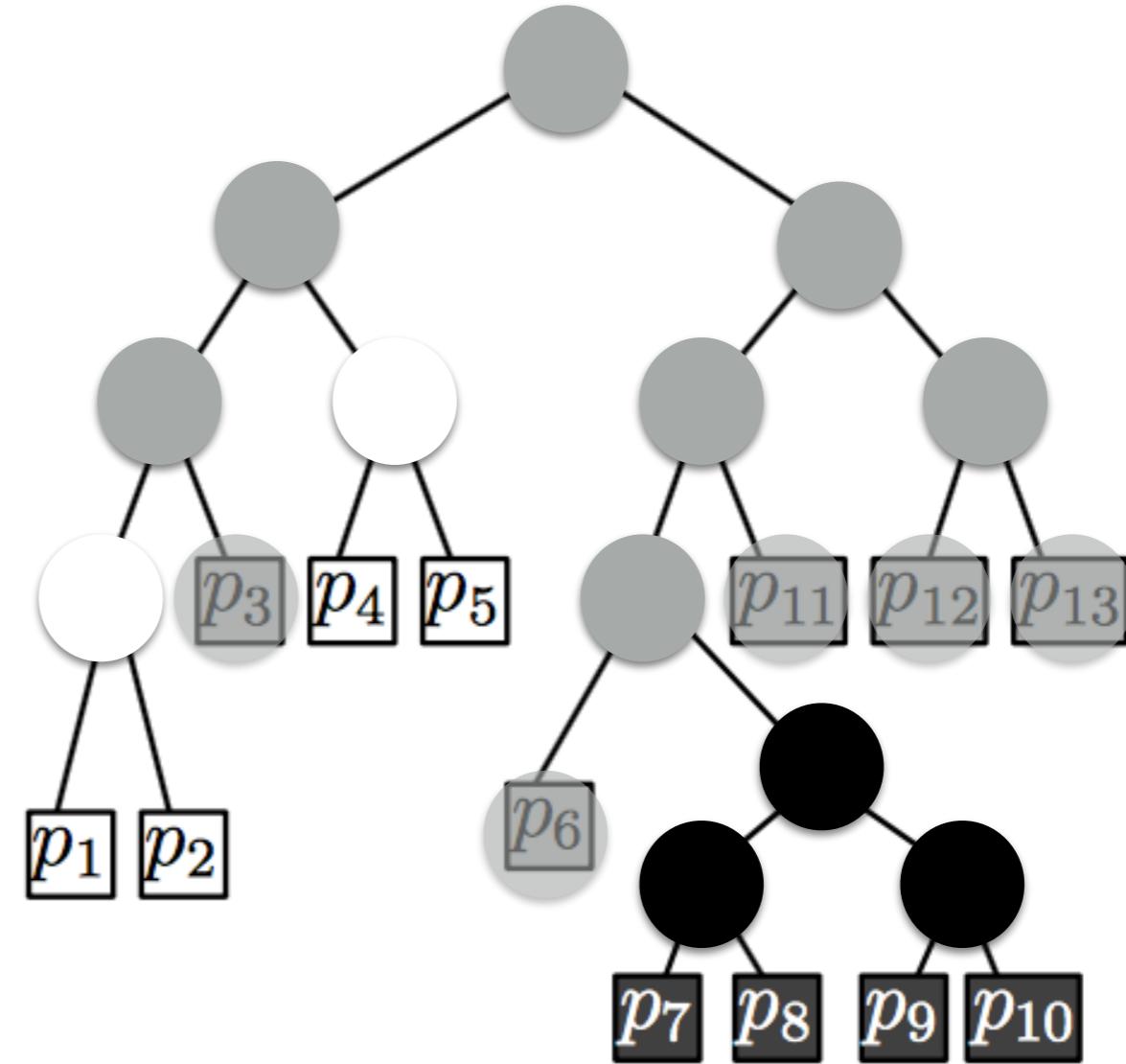
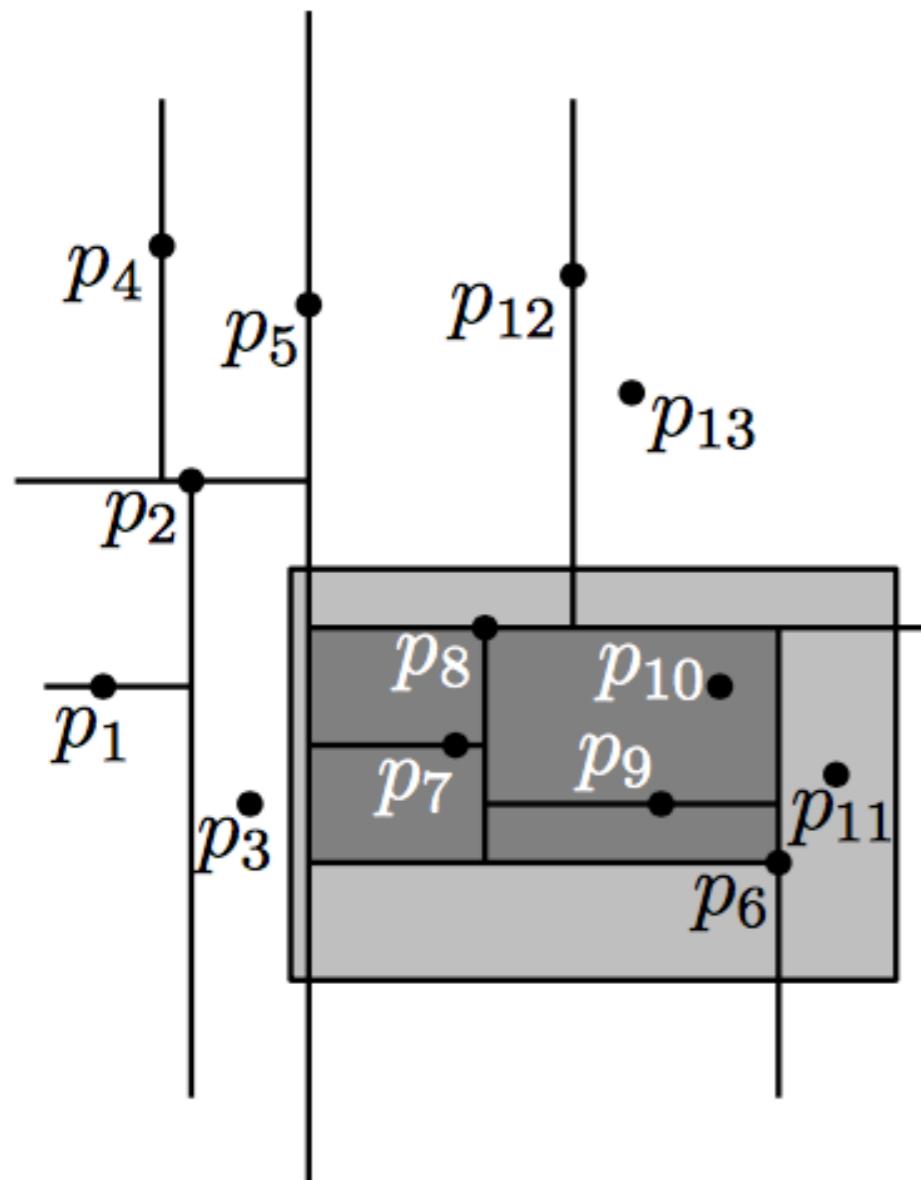
Time to answer range query = $O(\text{nb.black} + \text{nb.grey nodes})$



How many black nodes?

Observation: Each black leaf contain a point that's reported => k leaves

Can be shown that the nb. of internal black nodes is $k-1$

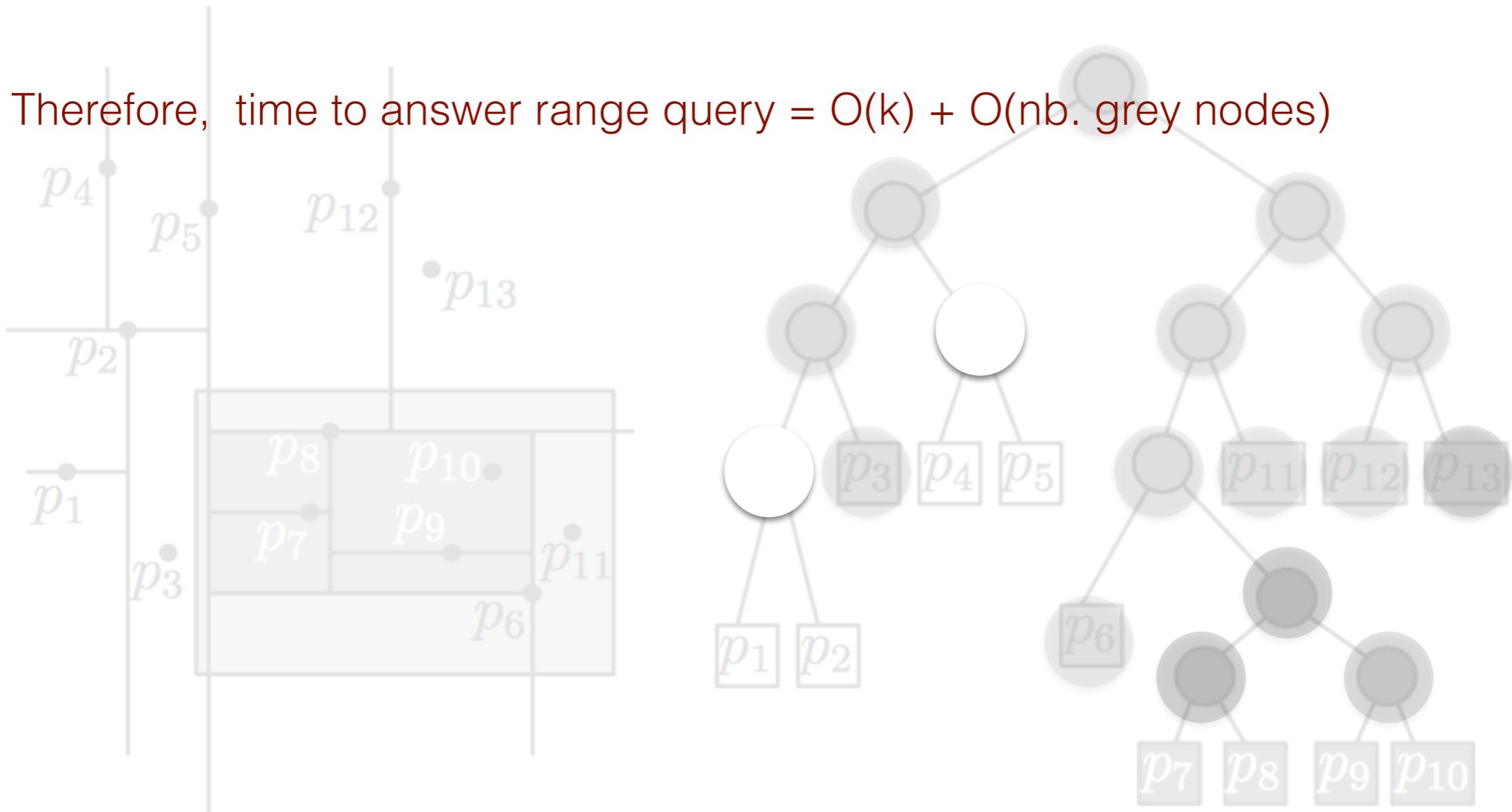


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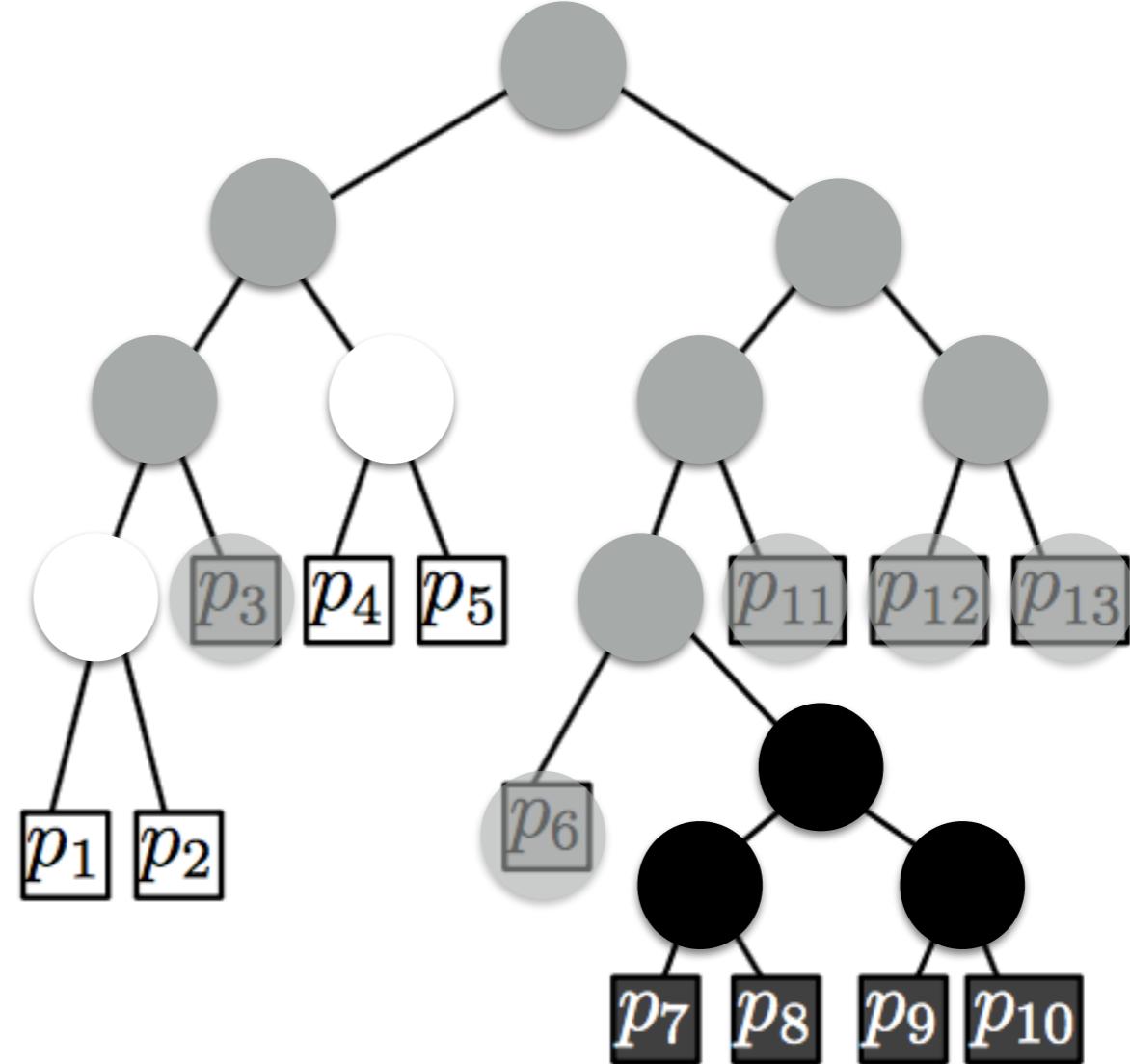
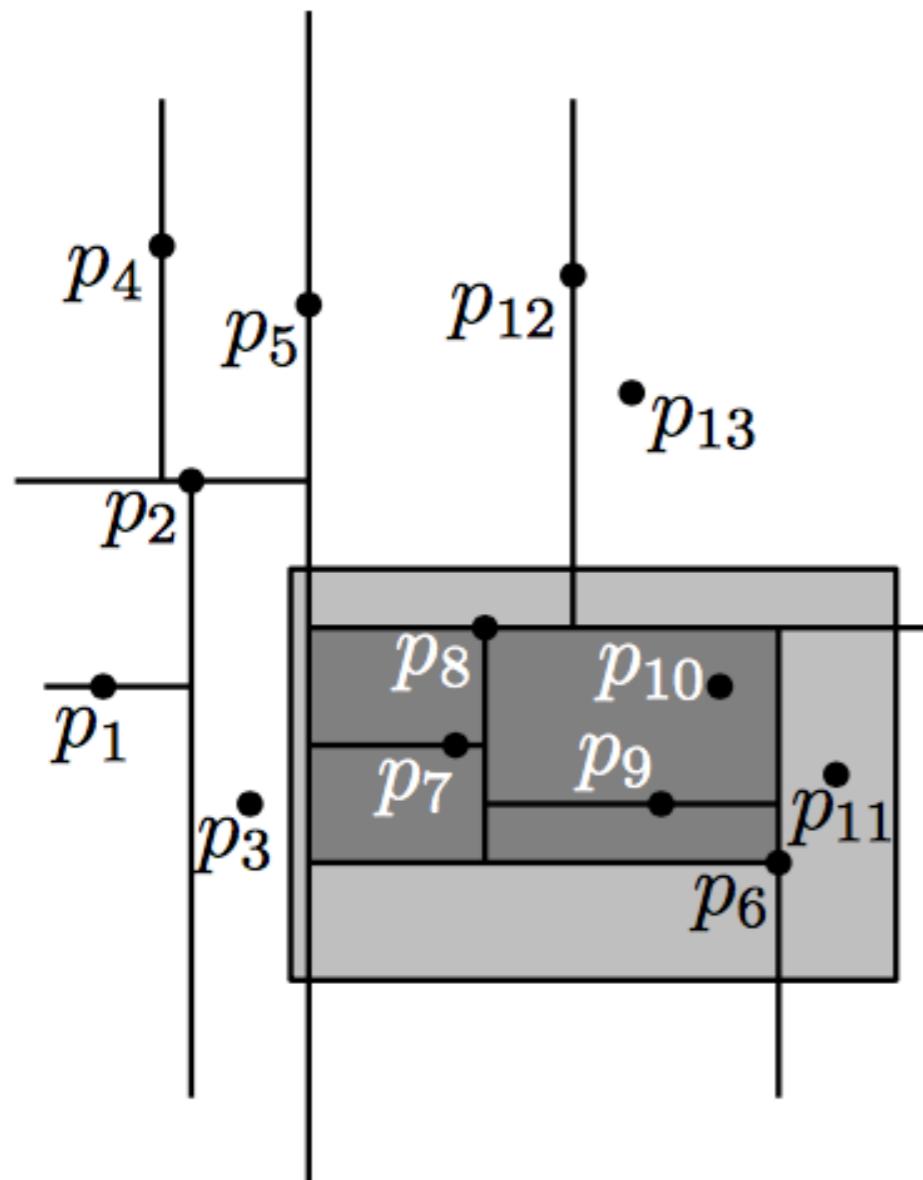
Therefore, time to answer range query = $O(k) + O(\text{nb. grey nodes})$



How many grey nodes?



region(v) intersects R, but region(v) not contained in R



How many grey nodes?



region(v) intersects R , but region(v) not contained in R

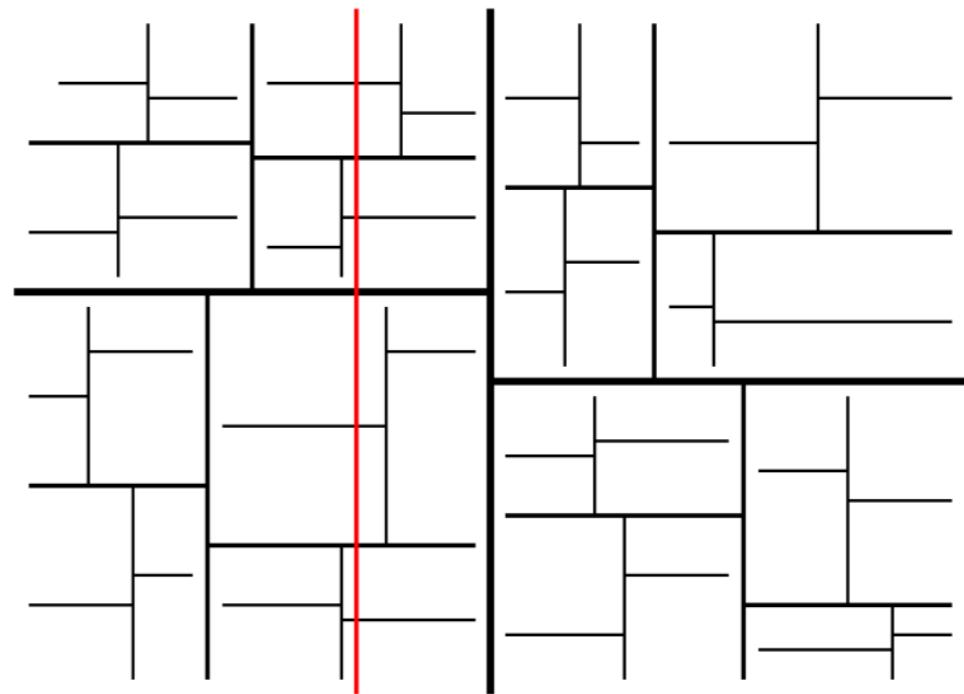
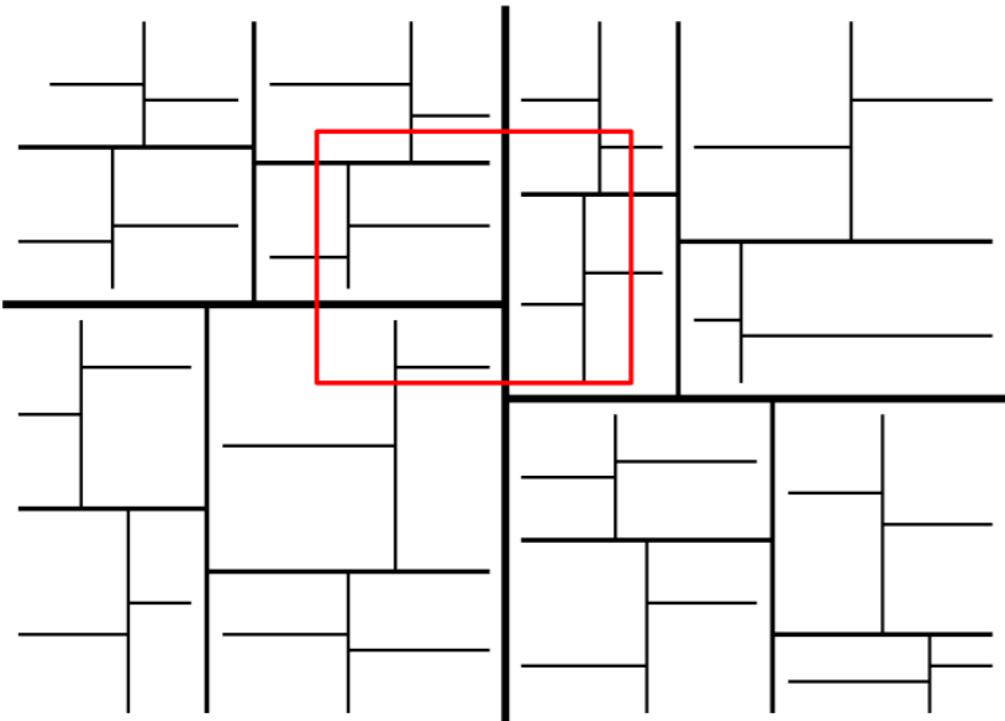
→ How many nodes are such that the boundary of their region intersects the boundary of the range?

How many grey nodes?



region(v) intersects R , but region(v) not contained in R

→ How many nodes are such that the boundary of their region intersects the boundary of the range?

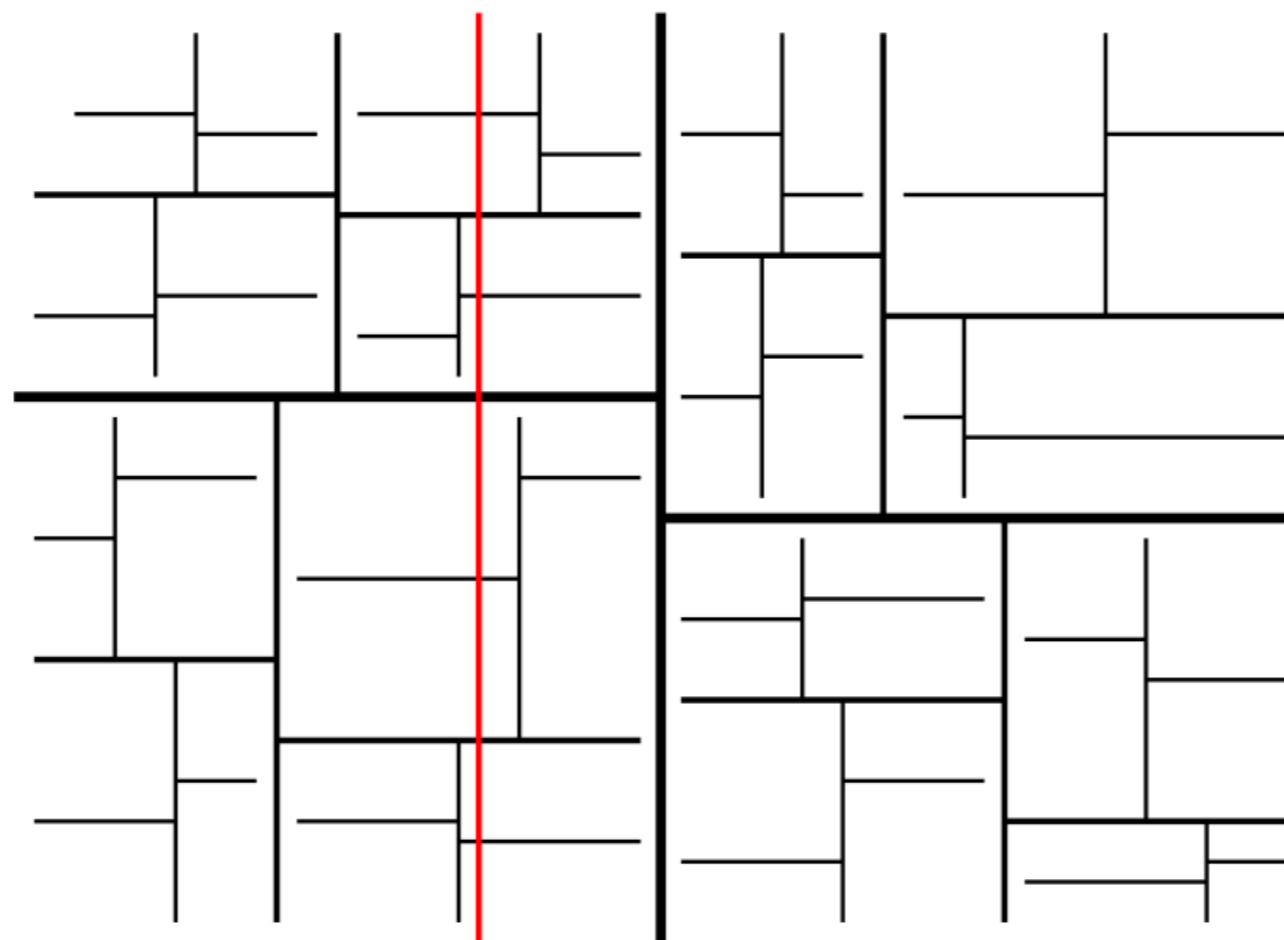


Simplified problem:

We'll count the number of nodes whose region intersects a vertical line l .

Number of nodes v such that $\text{region}(v)$ intersects a vertical line l ?

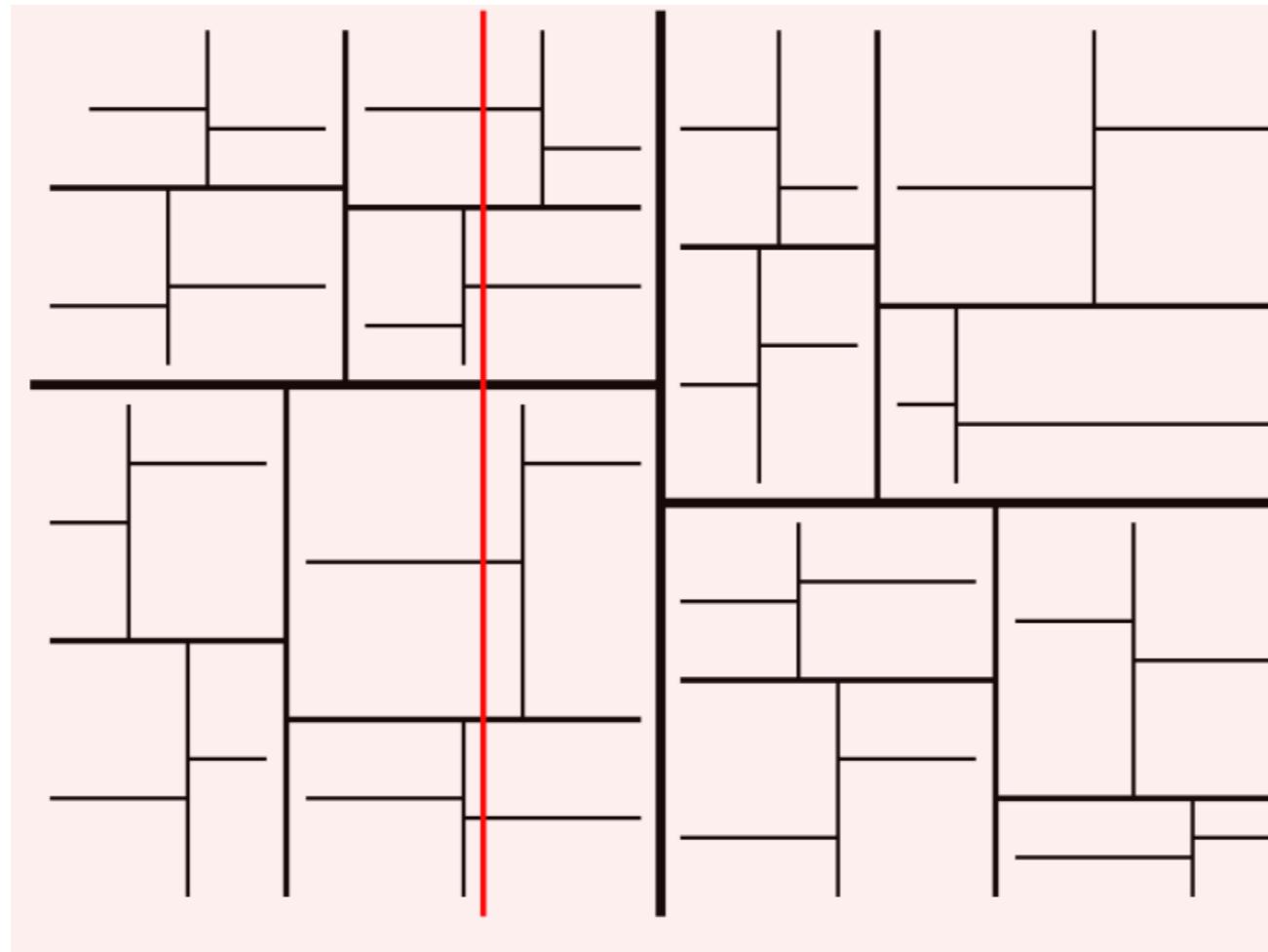
We'll think recursively, starting at the root:



Number of nodes v such that $\text{region}(v)$ intersects a vertical line l ?

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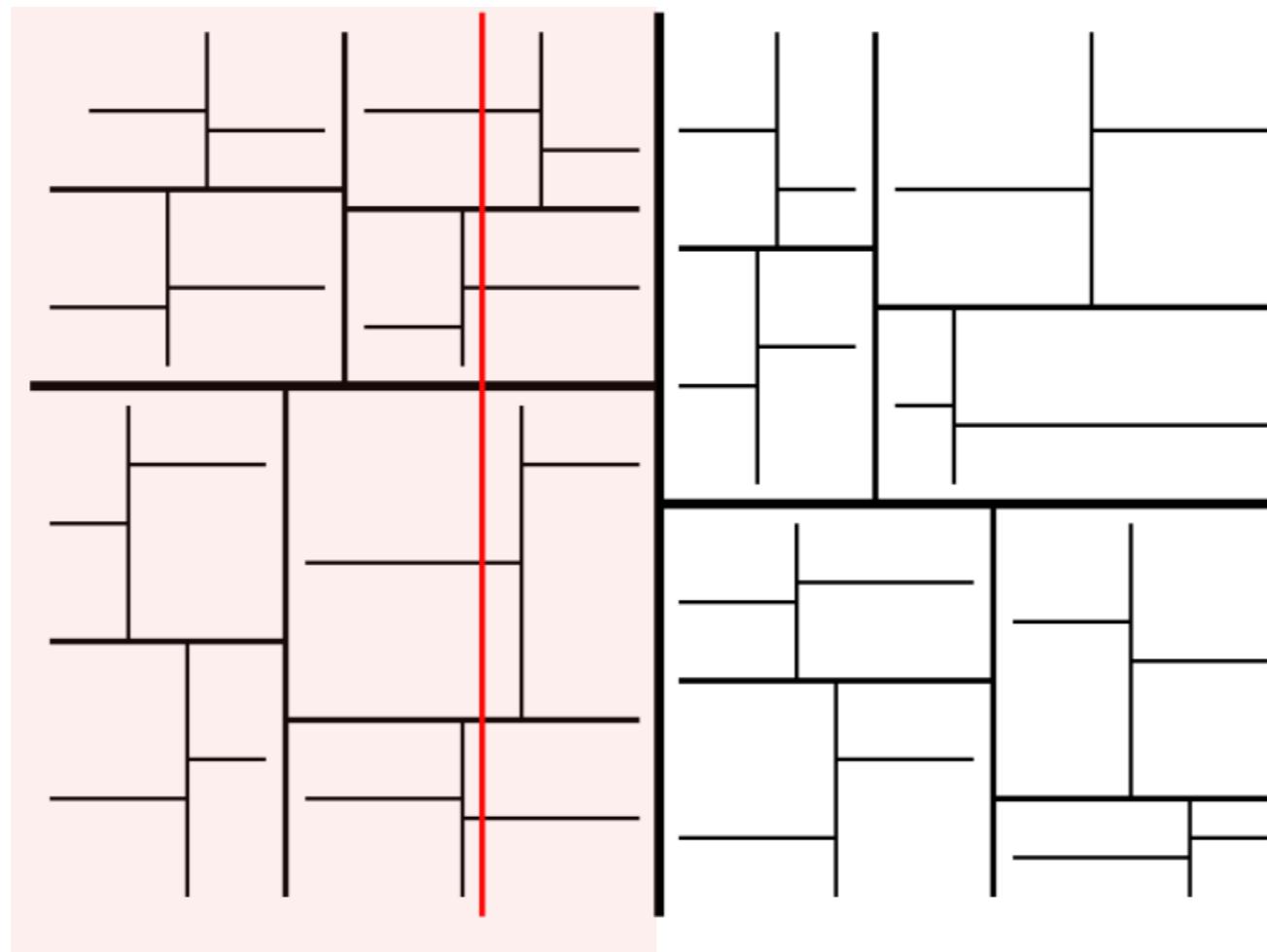
- depth=0: $\text{region}(\text{root})$ intersects l + 1



Number of nodes v such that $\text{region}(v)$ intersects a vertical line l ?

We'll think recursively, starting at the root:

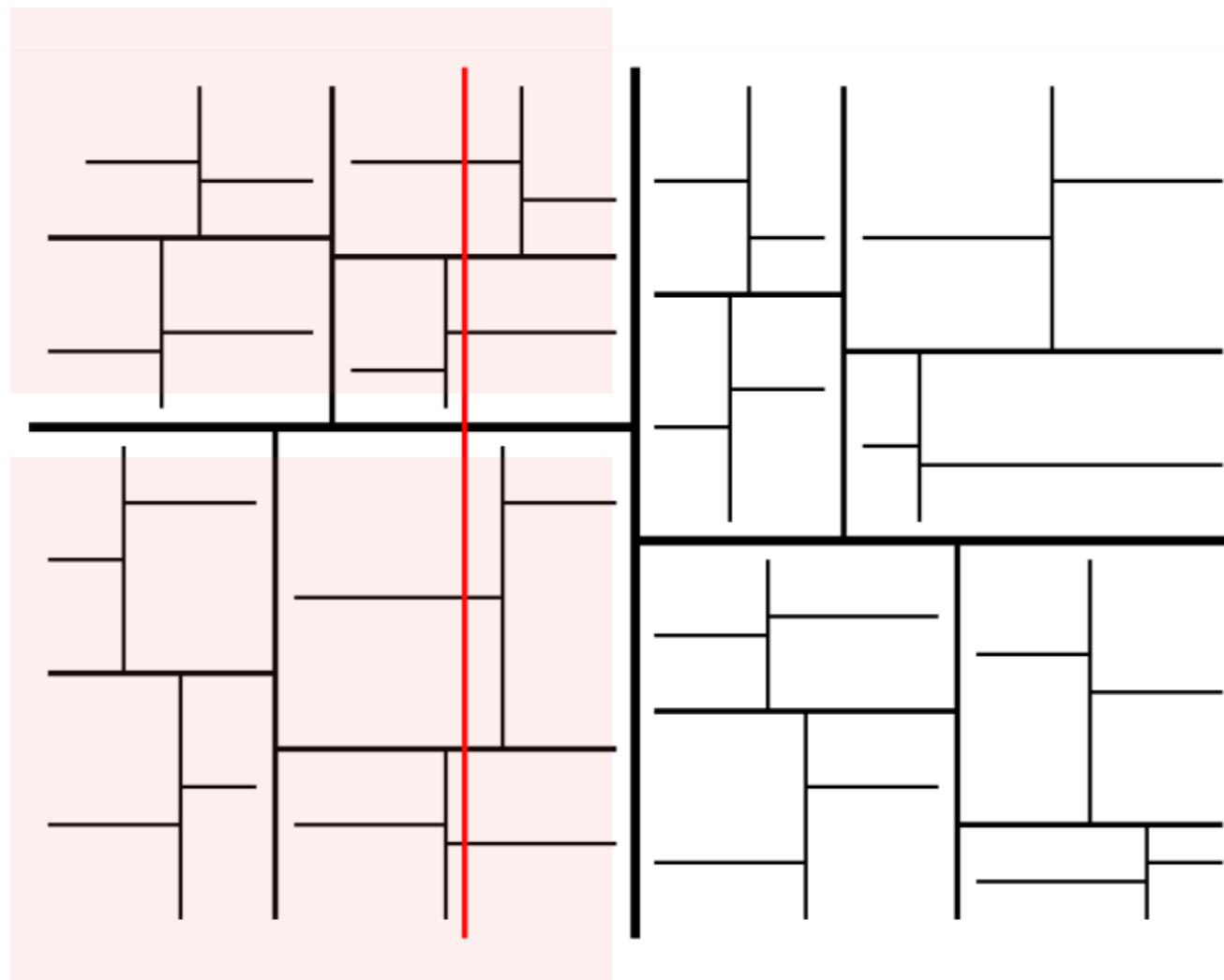
- depth=1: only one of {left, right} child intersects l + 1



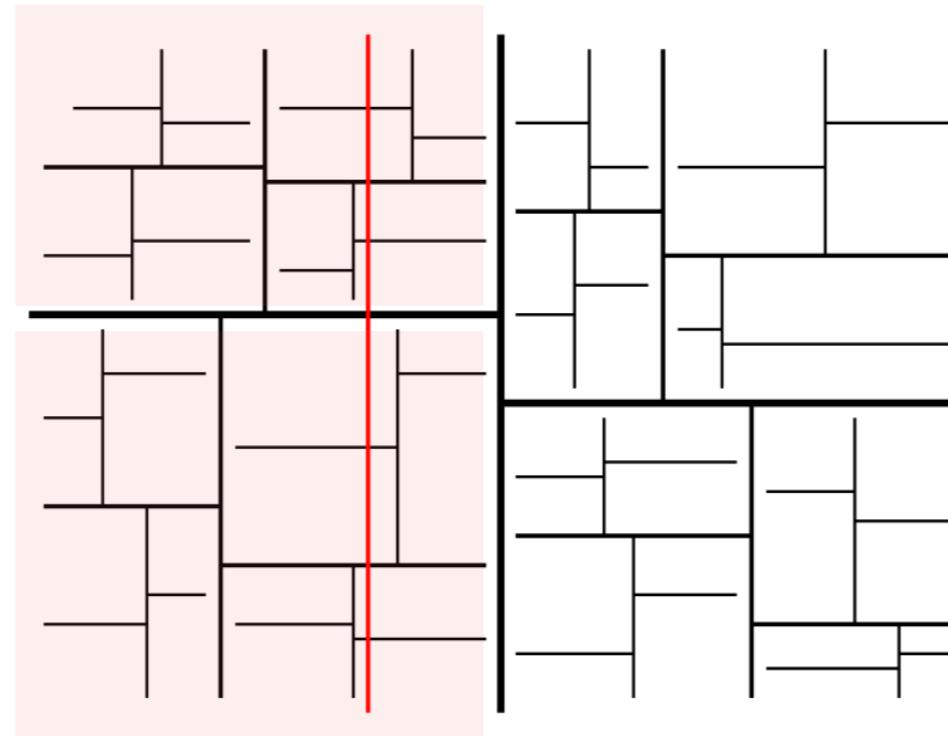
Number of nodes v such that $\text{region}(v)$ intersects a vertical line l ?

We'll think recursively, starting at the root:

- depth=2: both {left, right} child intersect l recurse



- Let $G(n)$ = nb. of nodes in a kd-tree of n points whose regions interest a vertical line l .
- Then $G(n) = 2 + 2G(n/4)$, and $G(1) = 1$



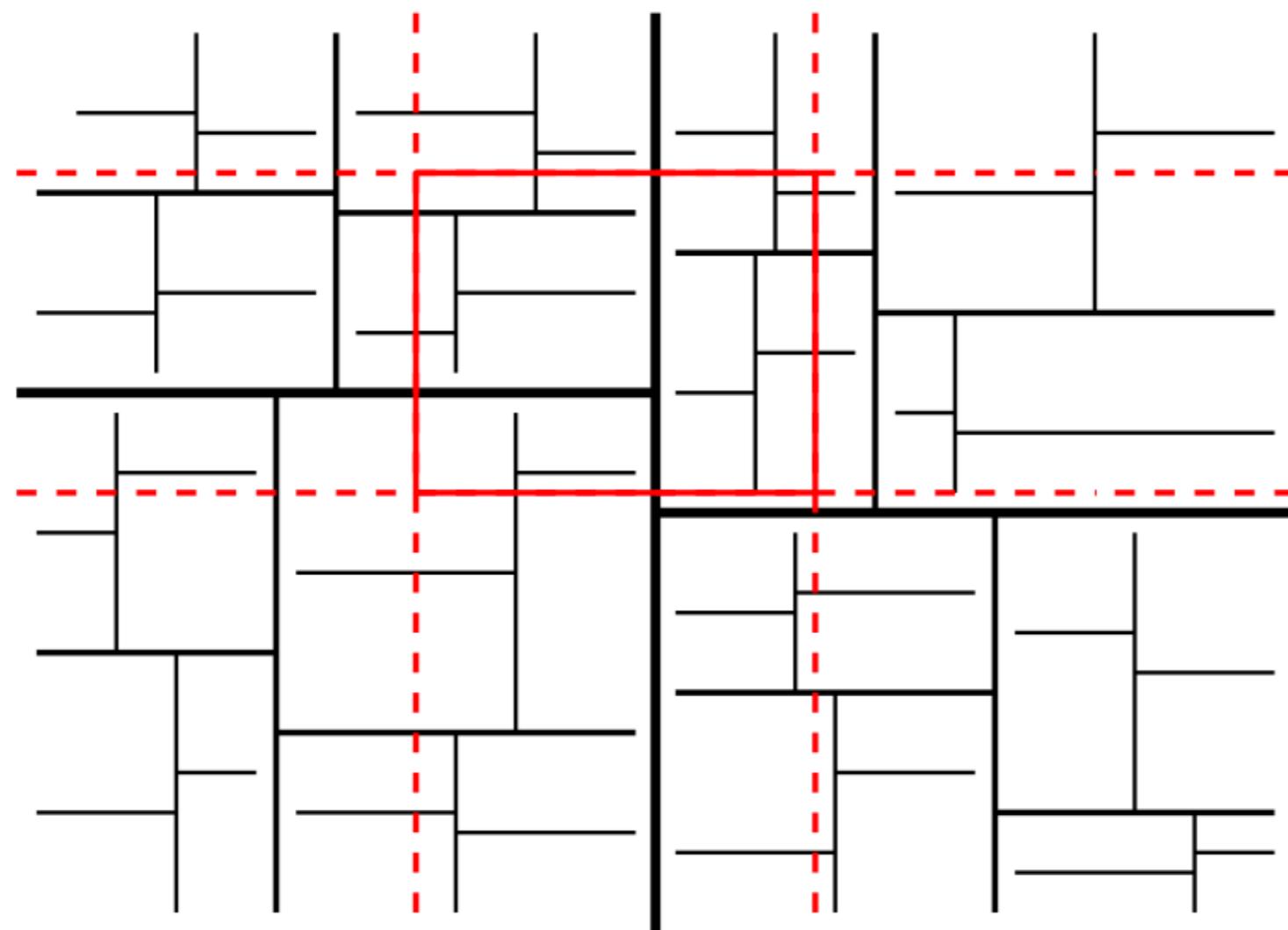
- This solves to $G(n) = O(\sqrt{n})$

Result: Any vertical or horizontal line l stabs $O(\sqrt{n})$ regions in the tree.

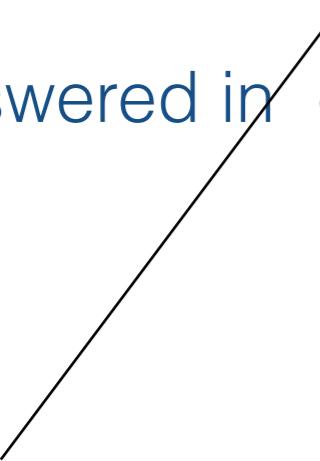
What we got so far:

- The number of grey nodes if the query were a vertical line is $O(\sqrt{n})$
- The same is true if it were a horizontal line
- How about a query rectangle?

- The nb. grey nodes for a query rectangle is at most the nb. grey nodes for two vertical and two horizontal lines, so it is at most $4 \times O(\sqrt{n}) = O(\sqrt{n})$



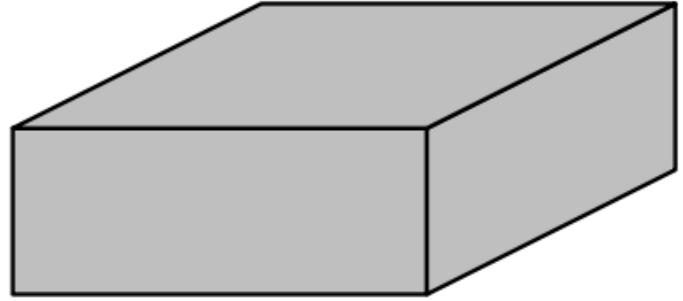
- **Theorem:** A set of n points in the plane can be preprocessed in $O(n \lg n)$ time into a data structure of $O(n)$ size so that any 2D range query can be answered in $O(\sqrt{n} + k)$ time, where k is the nb. points reported.



kd-tree (2d-tree)

kd-tree in higher dimensions

kd-tree in 3D: 3d-tree



- A 3D kd-tree alternates splits on x-, y- and z-dimensions
- A 3D range query is a cube
- Construction: The construction of a 2D kd-tree extends to 3D
- Answering range queries: Exactly the same as in 2D
- Analysis:

Let $G_3(n)$ be the number of grey nodes for a query with an axes-parallel plane in a 3D kd-tree

$$G_3(1) = 1$$

$$G_3(n) = 4 \cdot G_3(n/8) + O(1)$$

Higher dimensions

Theorem: A set of n points in d -space can be preprocessed in $O(n \log n)$ time into a data structure of $O(n)$ size so that any d -dimensional range query can be answered in $O(n^{1-1/d} + k)$ time, where k is the number of answers reported