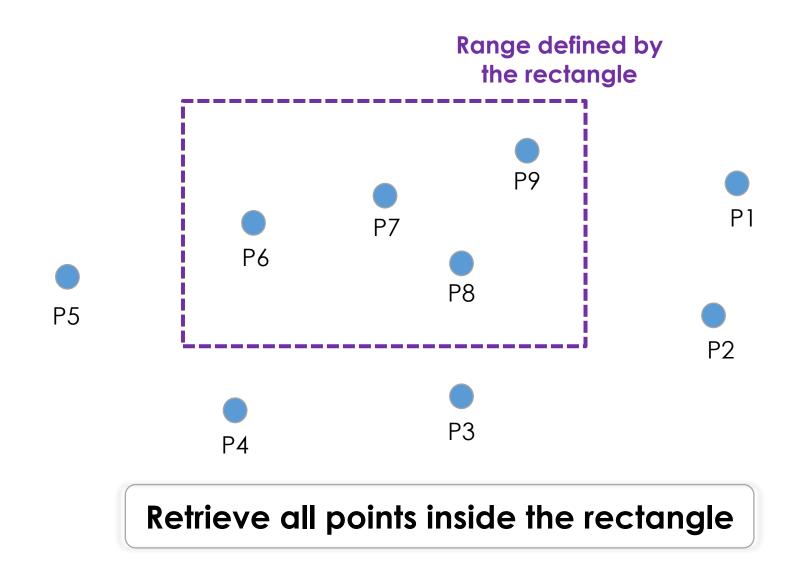
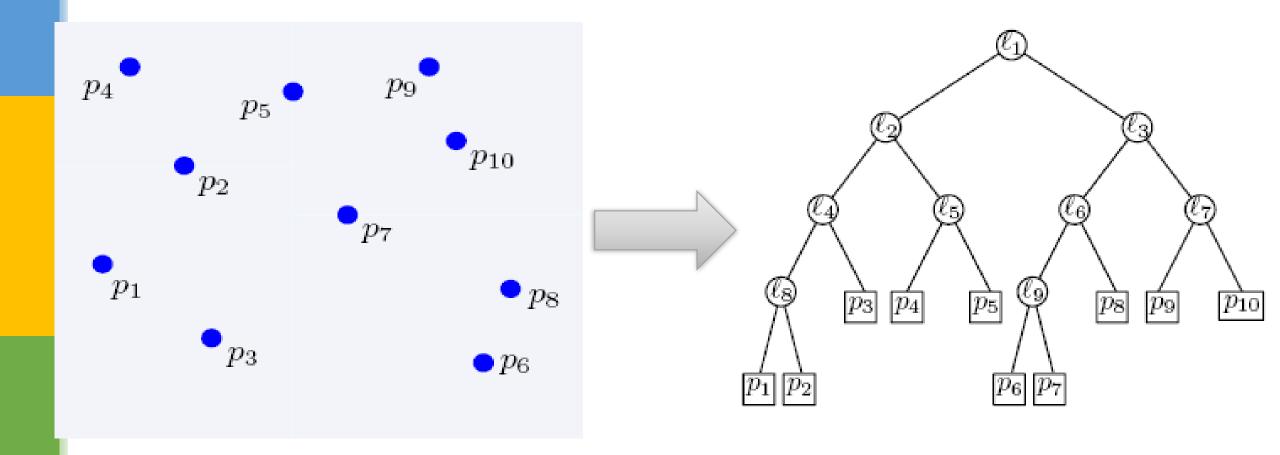
# Geometric Data Structures

# Range Queries

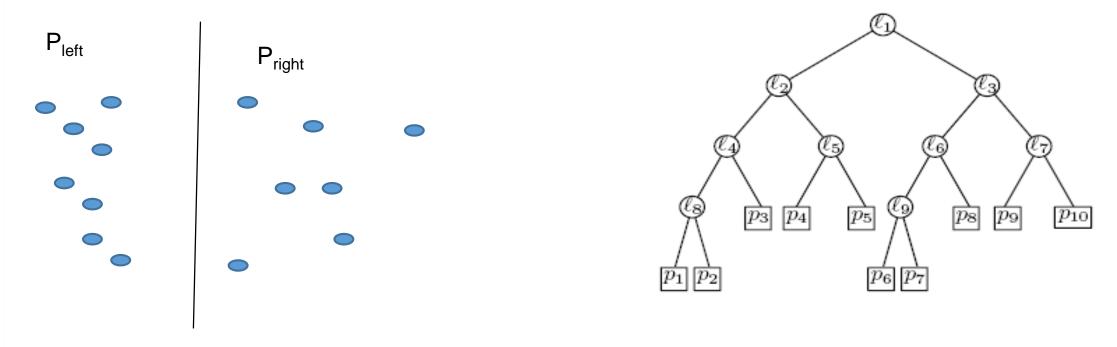


# KD Tree



#### **KD** Tree

- ☐ Every node (except leaves) represents a hyperplane that divides the space into two parts.
- ☐ Points to the left (right) of this hyperplane represent the left (right) sub-tree of that node.



#### KD Tree

As we move down the tree, we divide the space along alternating (but not always) axis-aligned hyperplanes:

Split by x-coordinate: split by a vertical line that has (ideally) half the points left or on, and half right.

Split by y-coordinate: split by a horizontal line that has (ideally) half the points below or on and half above.

Split by x-coordinate: split by a vertical line that has approximately half the

points left or on, and half right. Points on the line go to the left subtree  $ullet p_{10}$  $\ell_2$  $p_2$  $p_8$  $p_{10}$ 

Split by y-coordinate: split by a horizontal line that has approximately half the

points left or on, and half right. Points on the line go to the left subtree  $ullet p_{10}$  $p_2$  $\ell_2$  $p_8$  $p_{10}$ 

Split by x-coordinate: split by a vertical line that has approximately half the

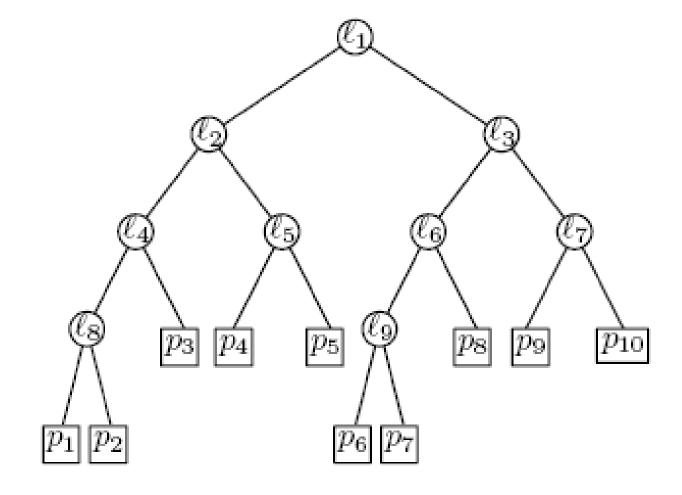
points left or on, and half right. Points on the line go to the left subtree  $lue{p_{10}}$  $\ell_2$  $p_8$  $p_5$  $p_{10}$ 

Split by y-coordinate: split by a horizontal line that has approximately half the

points left or on, and half right. Points on the line go to the left subtree  $lue{p_{10}}$  $\ell_2$  $p_8$  $p_{10}$ 

#### KD Tree Node Structure

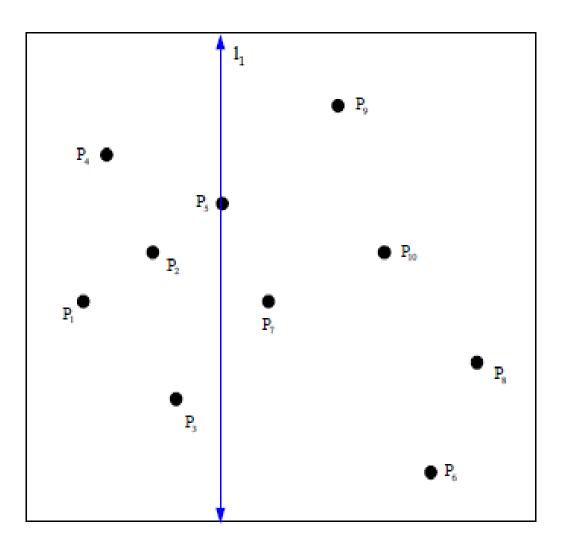
- A KD-tree node has 5 fields
  - Splitting axis
  - Splitting value
  - Data
  - Left pointer
  - Right pointer

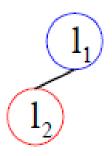


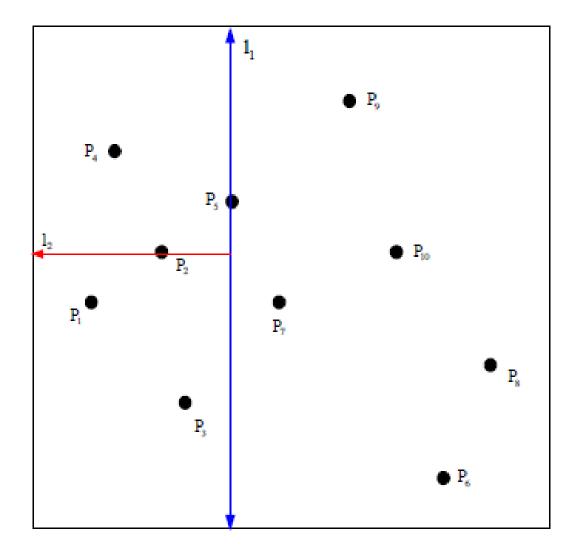
# KD Tree Splitting Strategies

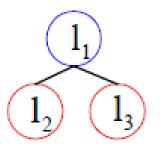
- Divide by finding median
  - Assumes all the points are available ahead of time.
- Divide perpendicular to the axis with widest spread
  - Split axes might not alternate
- And many more....

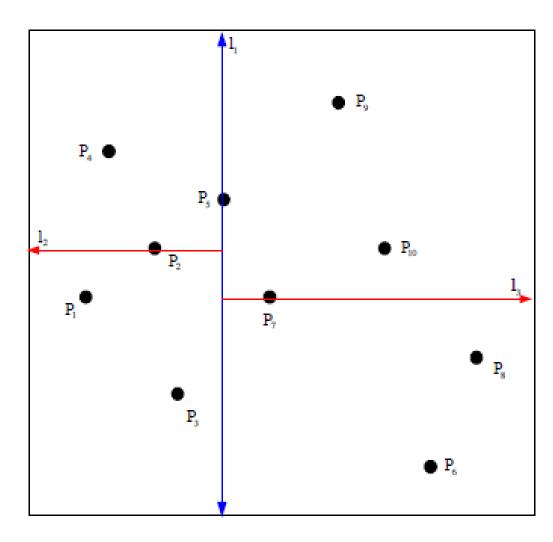


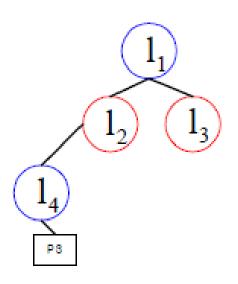


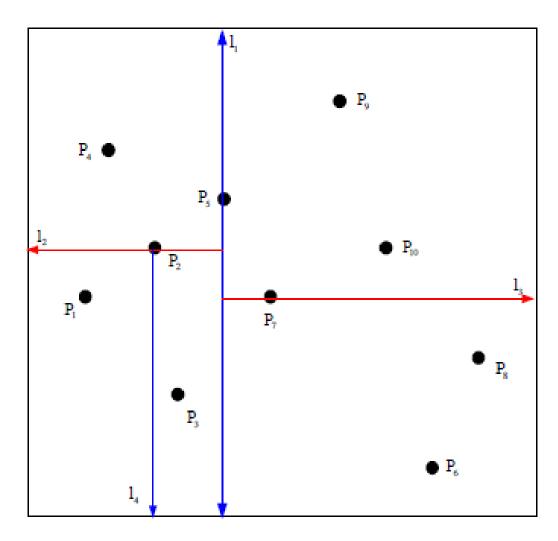


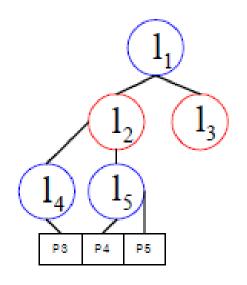


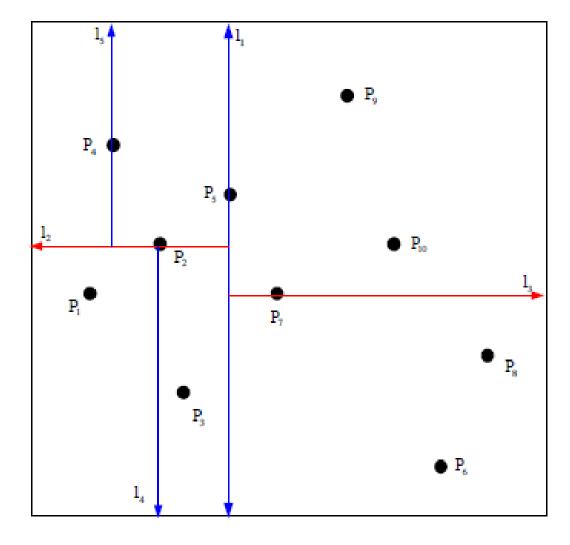


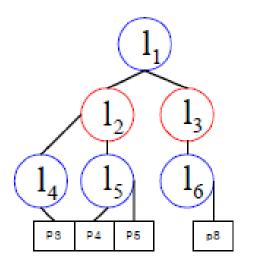


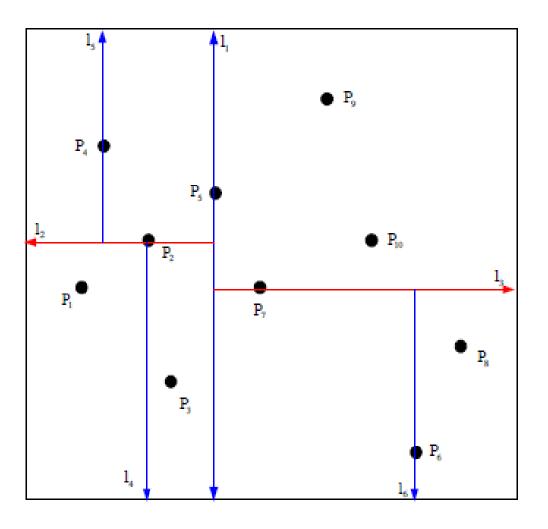


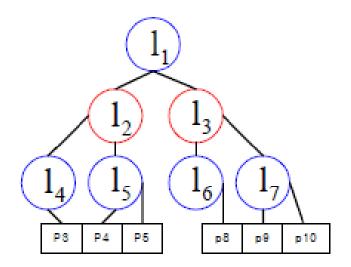


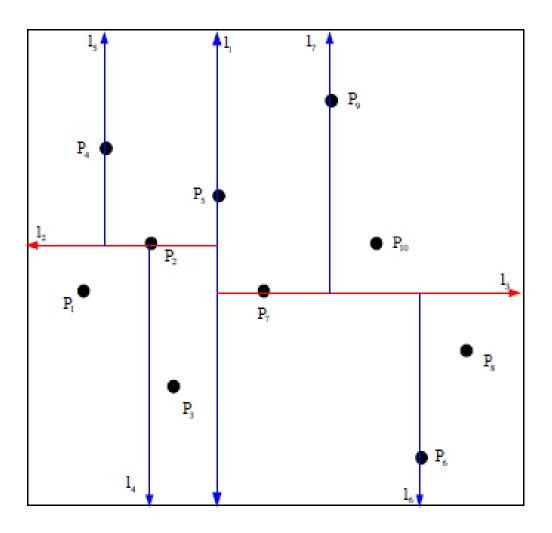


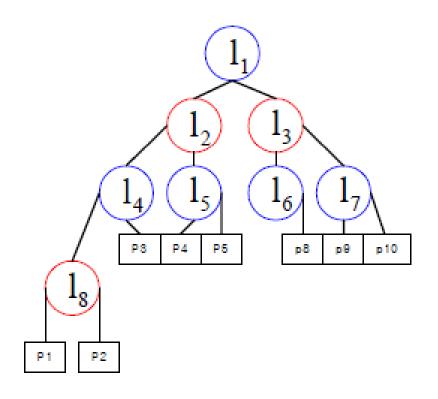


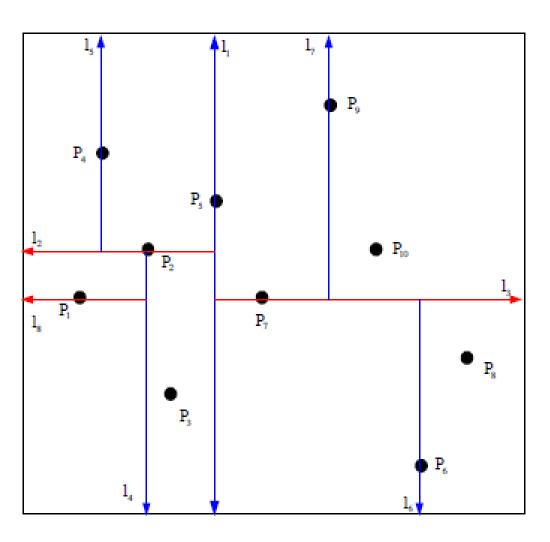


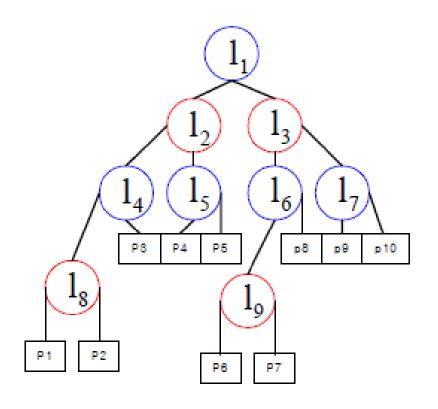


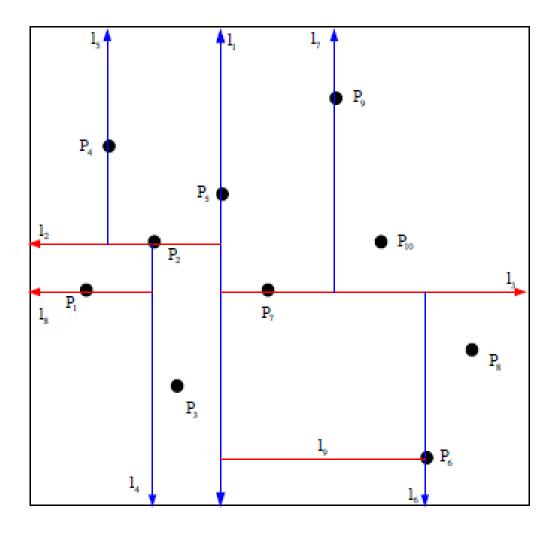




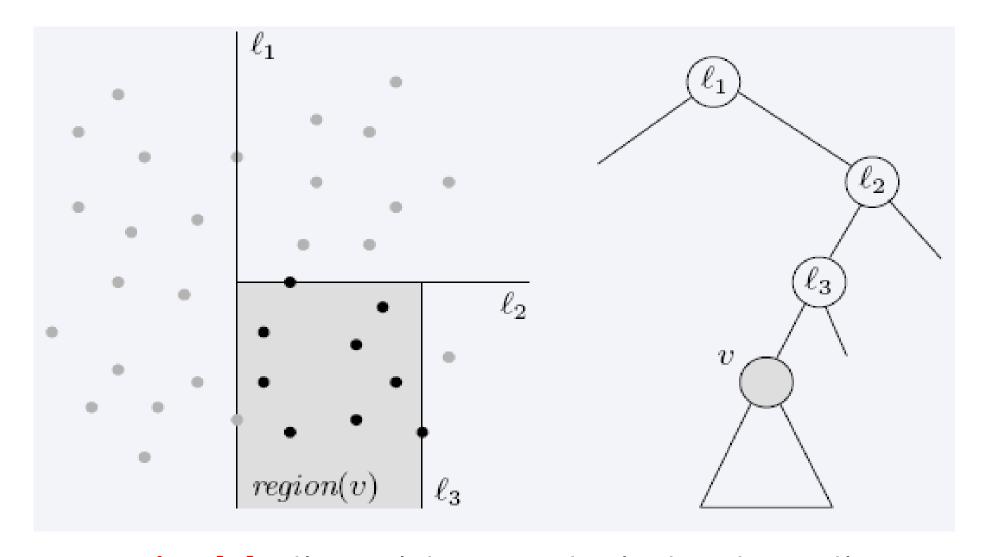








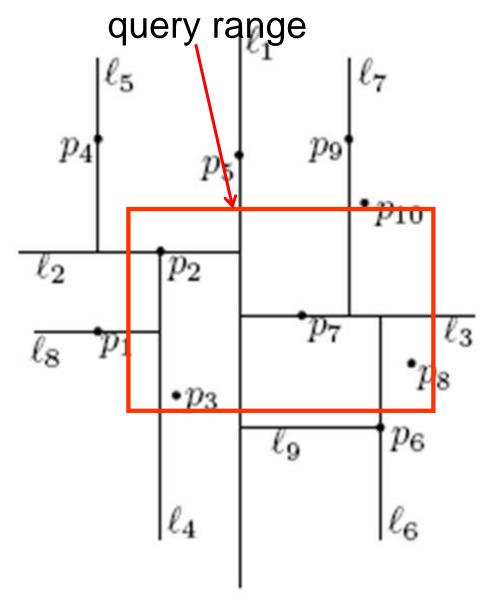
# Region of node v



Region(v): the subtree rooted at v stores the points in black dots

# KD Trees – Range Search

- Need only search nodes whose region intersects query region.
  - Report all points in subtrees whose regions are entirely contained in query range.
  - If a region is partially contained in the query range check points.

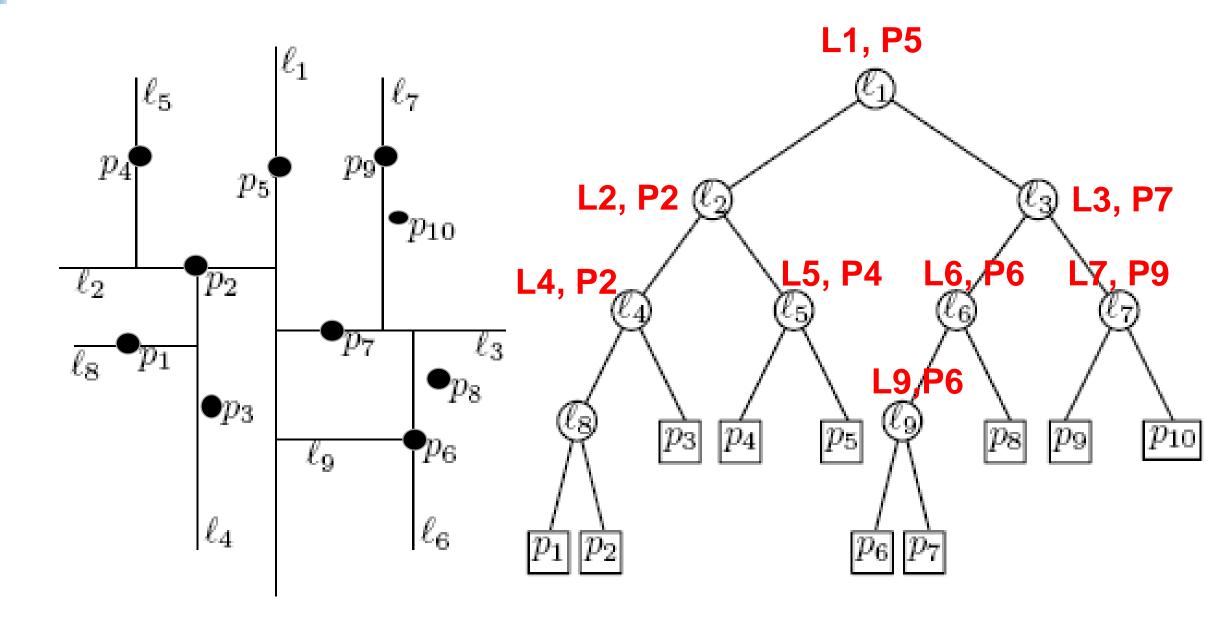


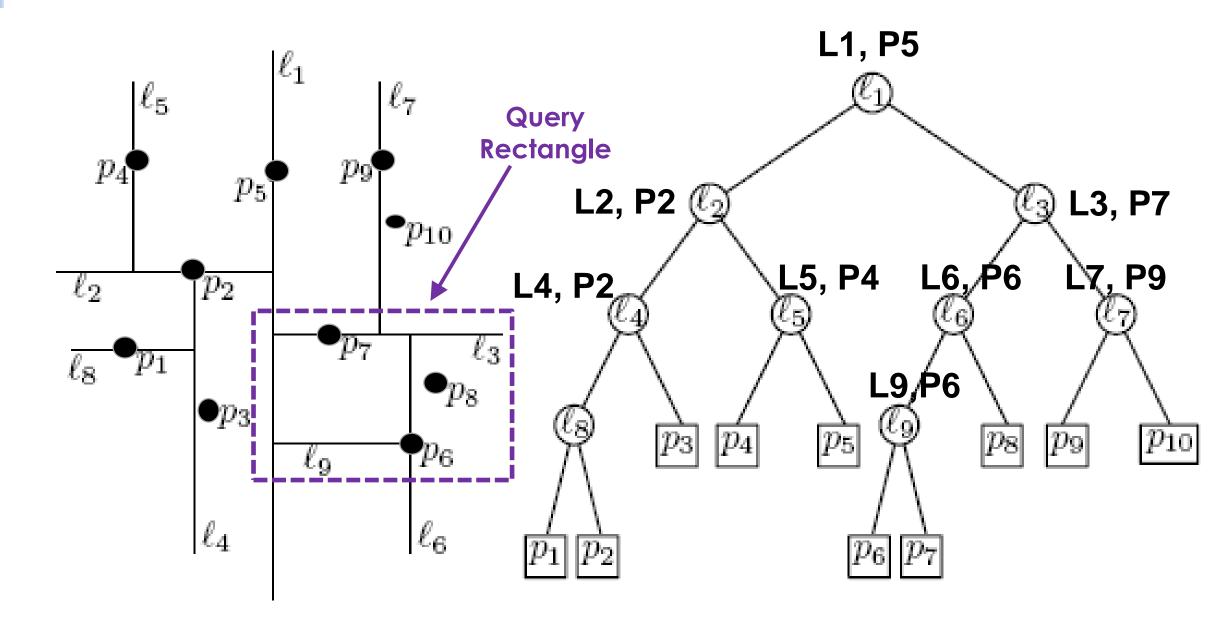
# KD-tree: range queries

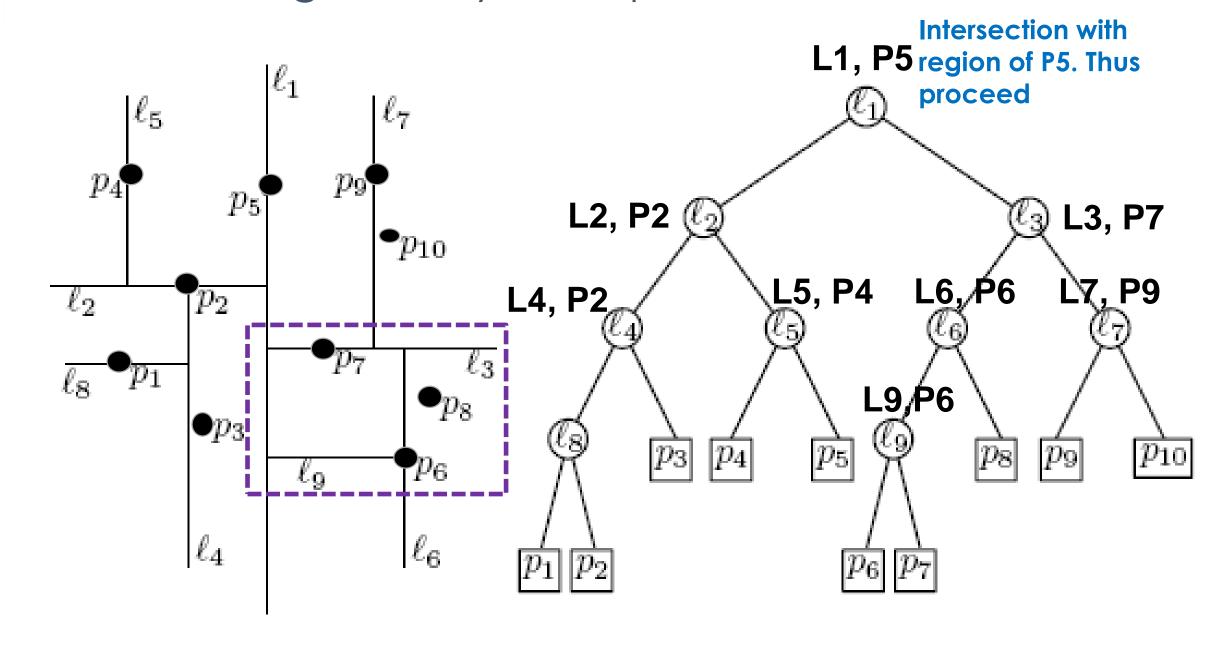
Recursive procedure starting from v = root

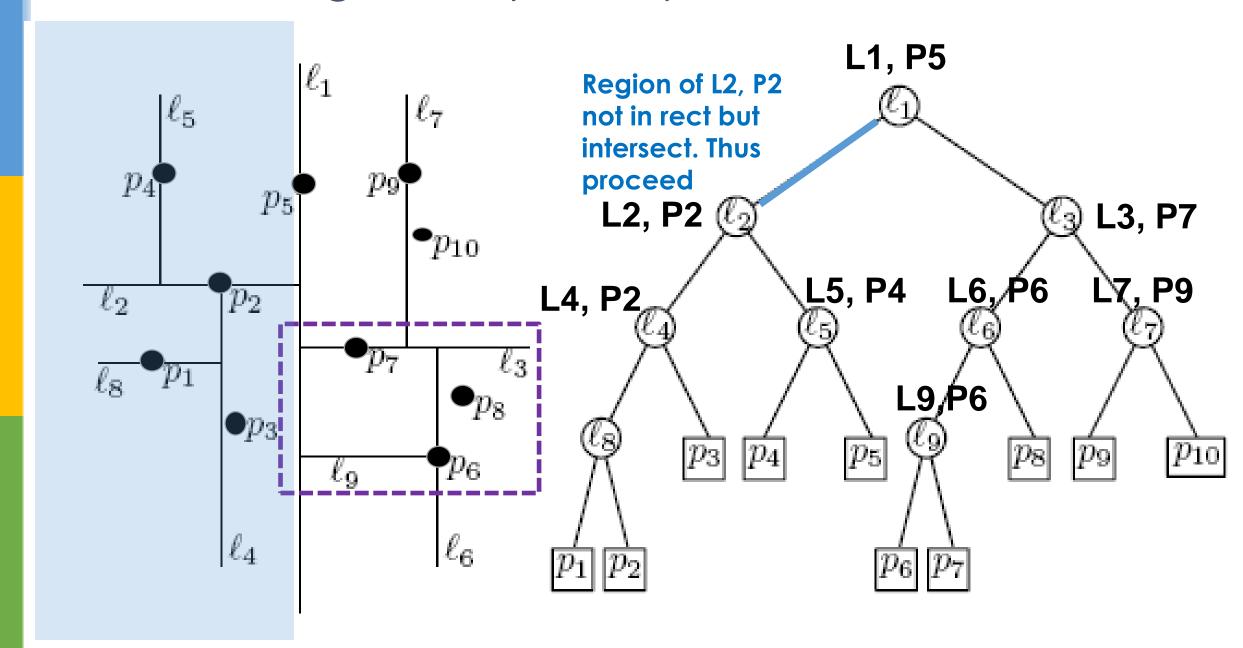
#### SearchKDtree (v, query rectangle R)

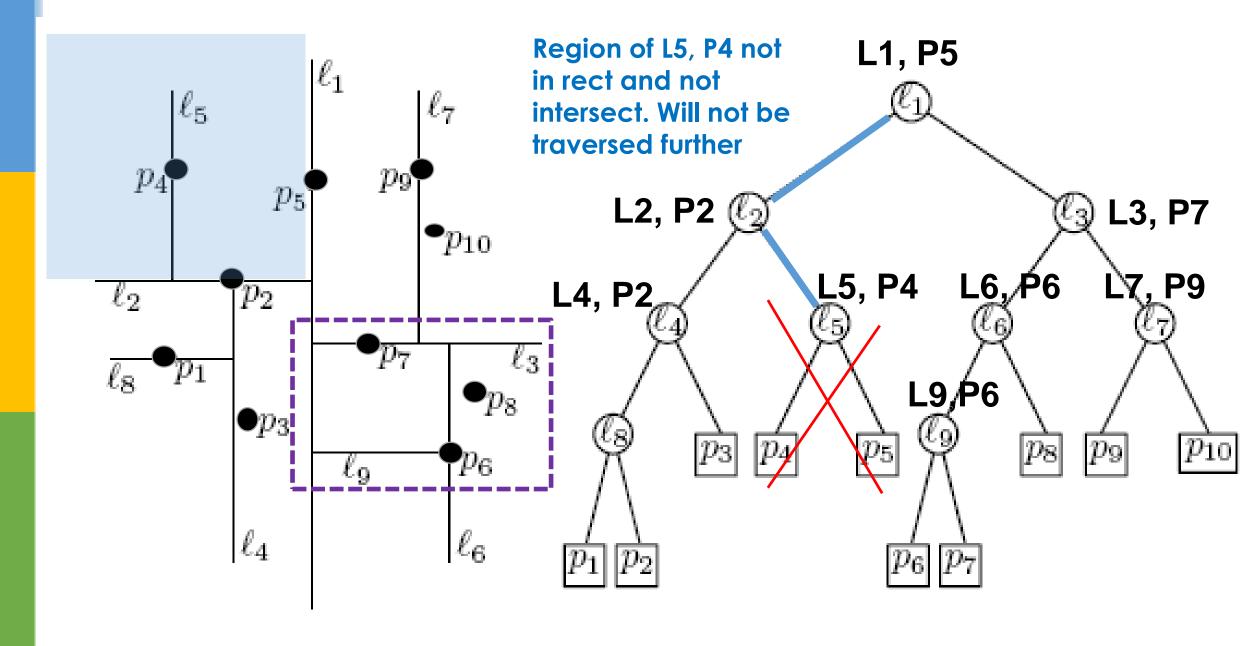
- 1. If v is a leaf, then report the points stored in v if it lies in R
- 2. Else
  - 2.1 if Region(lc(v)) is contained in R, then report all points in the subtree(v).
  - 2.2 Else If Region(Ic(v)) intersects with R, then SearchKDtree(Ic(v), R)
  - 2.3 If Region(rc(v)) is contained in R, then report all points in the subtree(v). intersects R
  - 2. 4 Else if Region(rc(v)) intersects R, then SearchKDtree(rc(v), R)

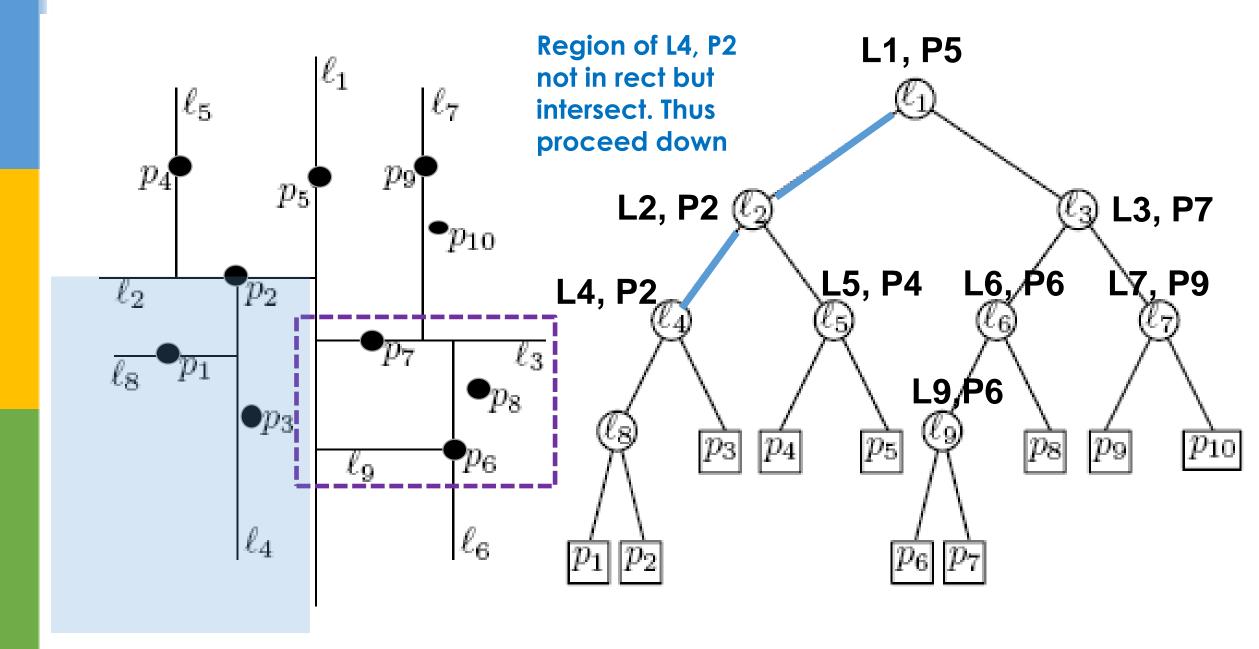


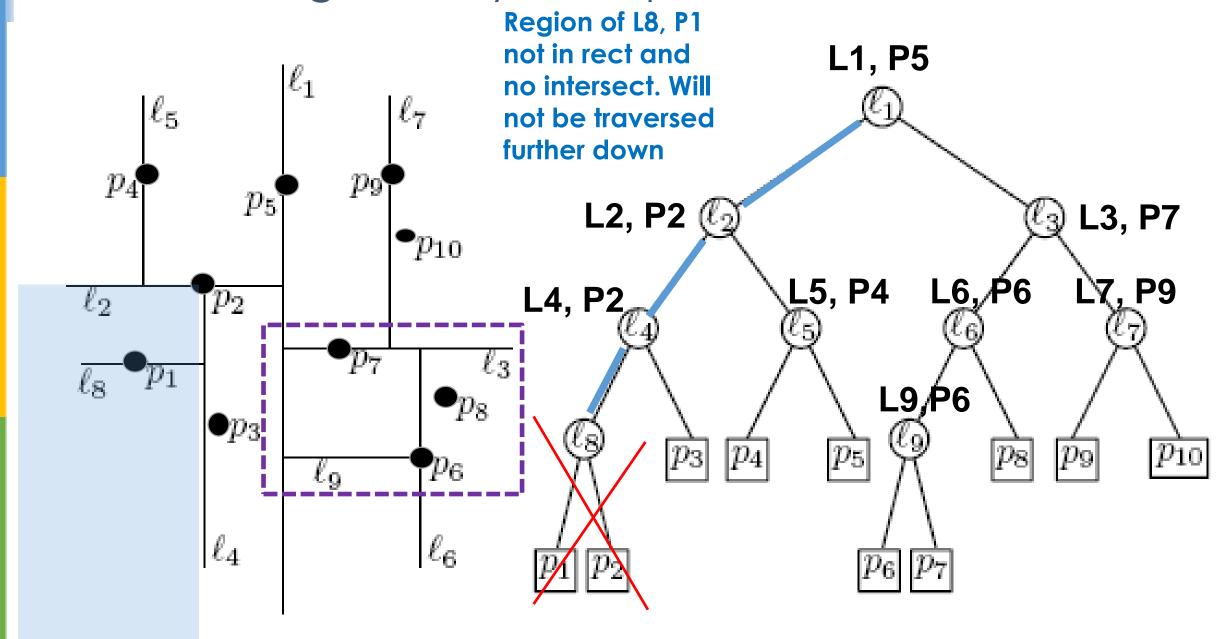


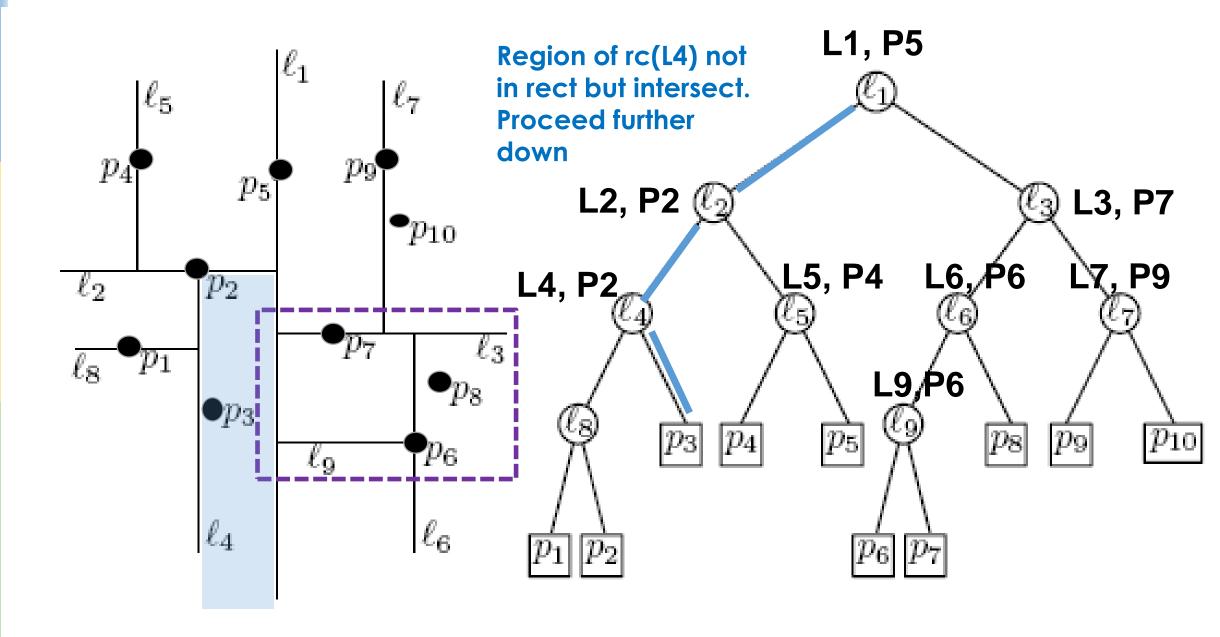


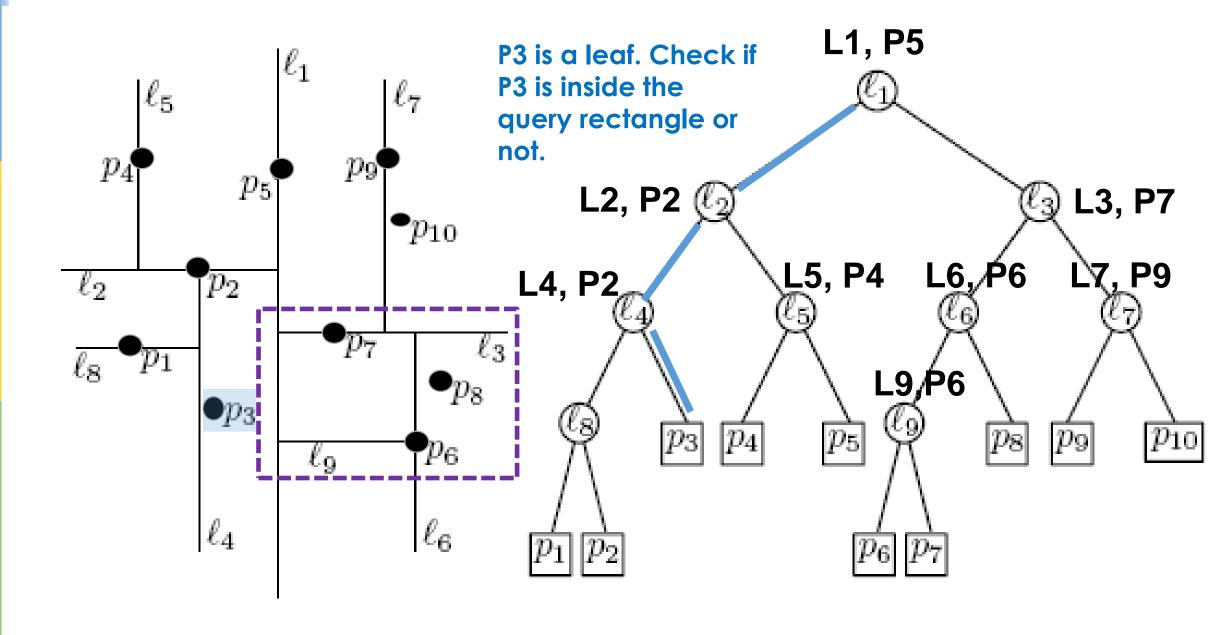


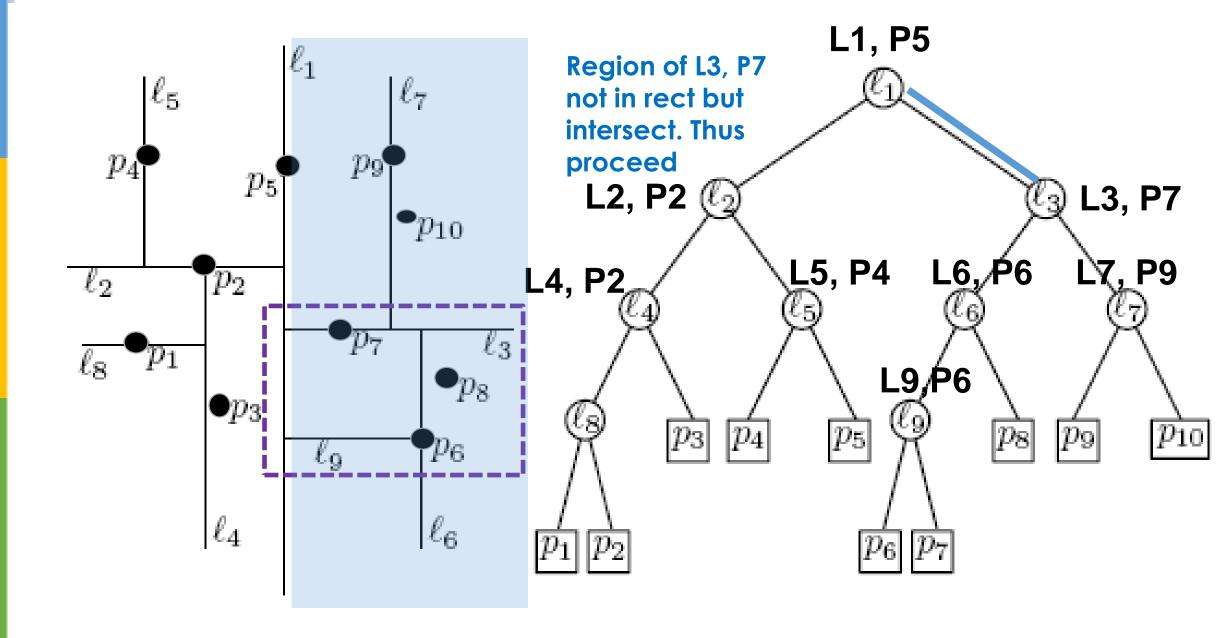


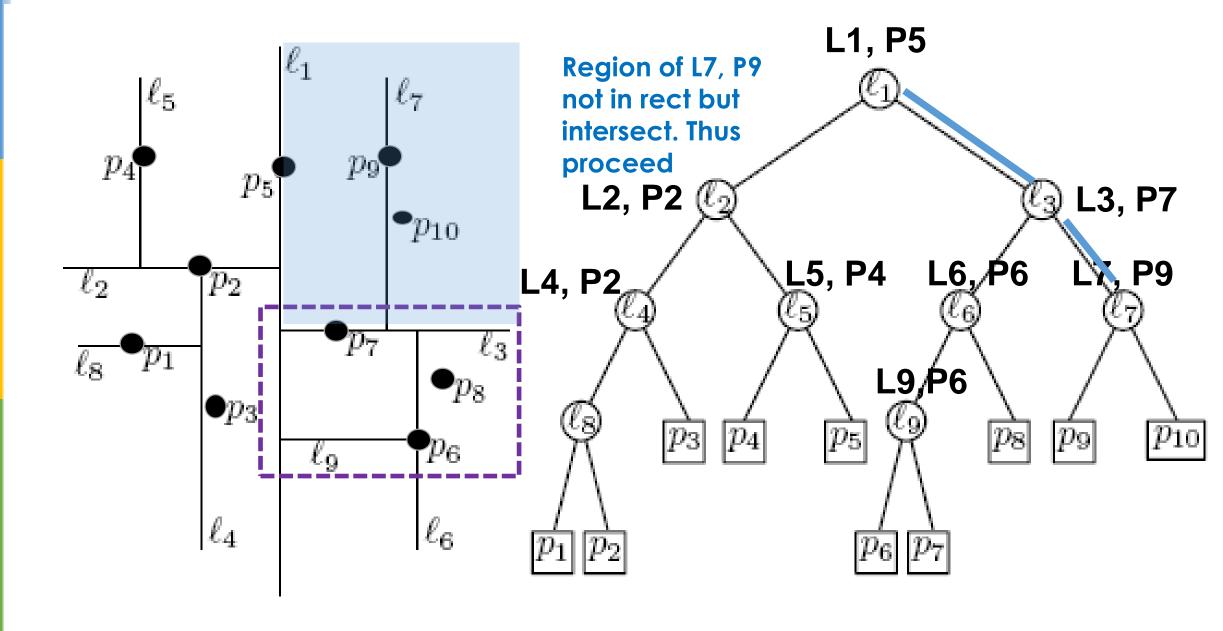


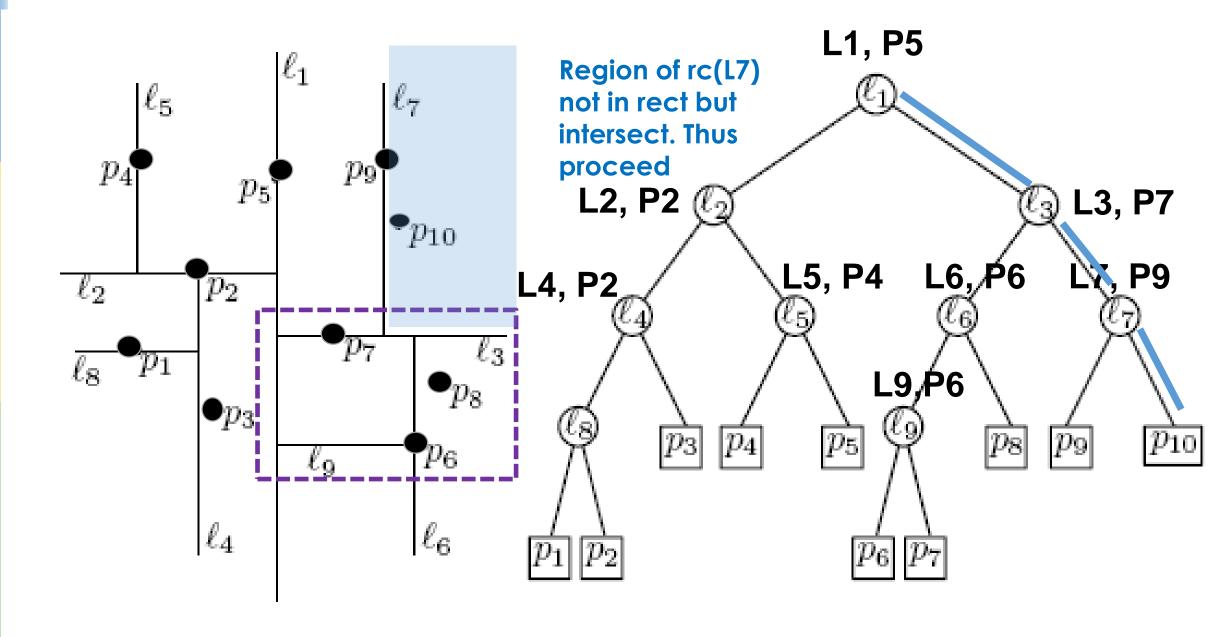


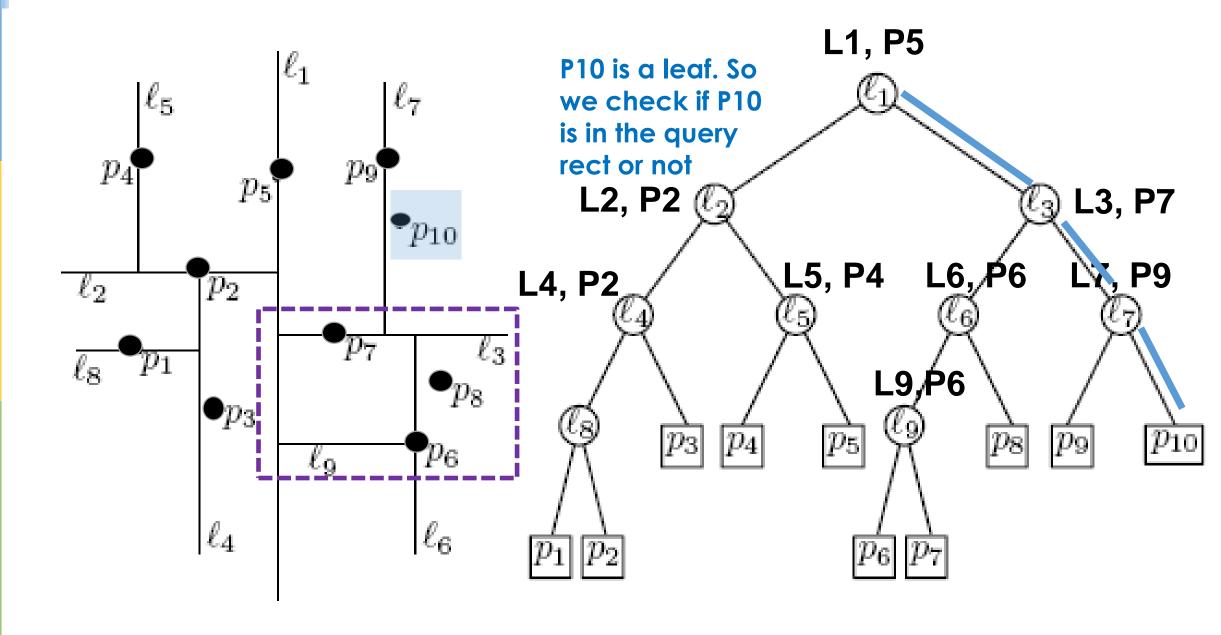


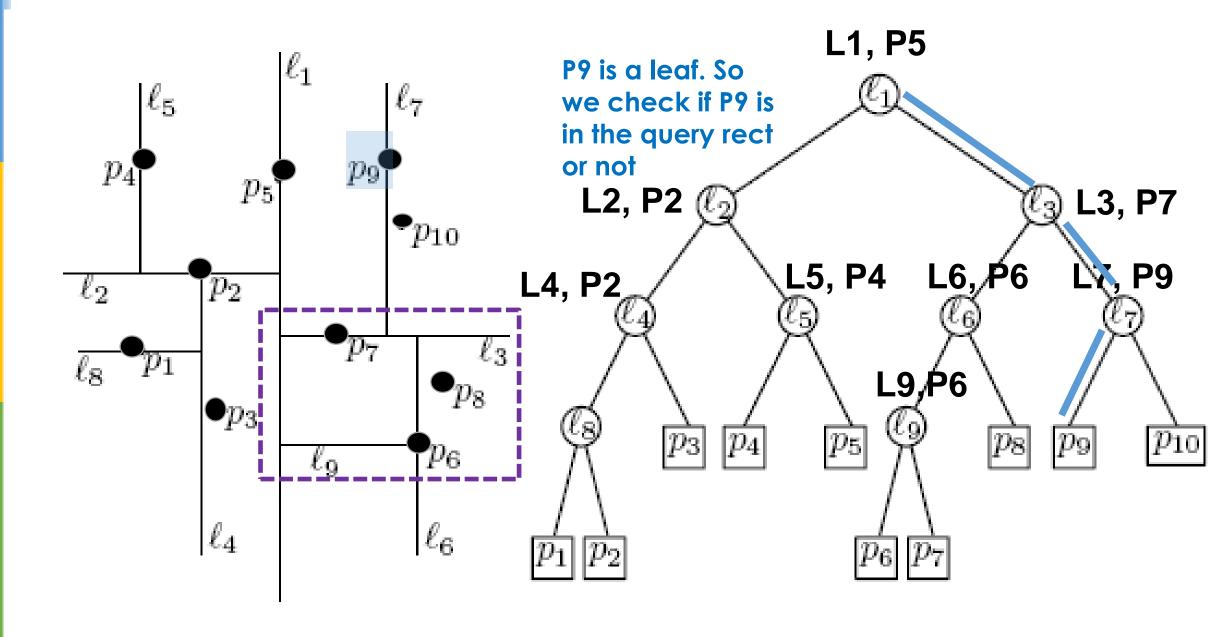


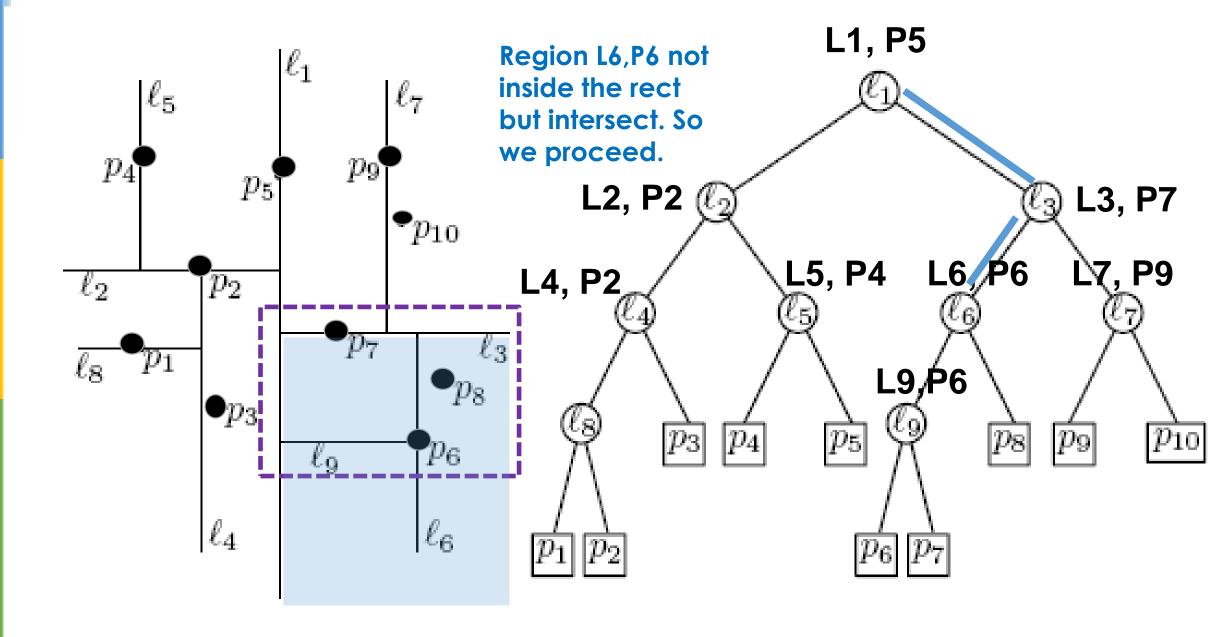


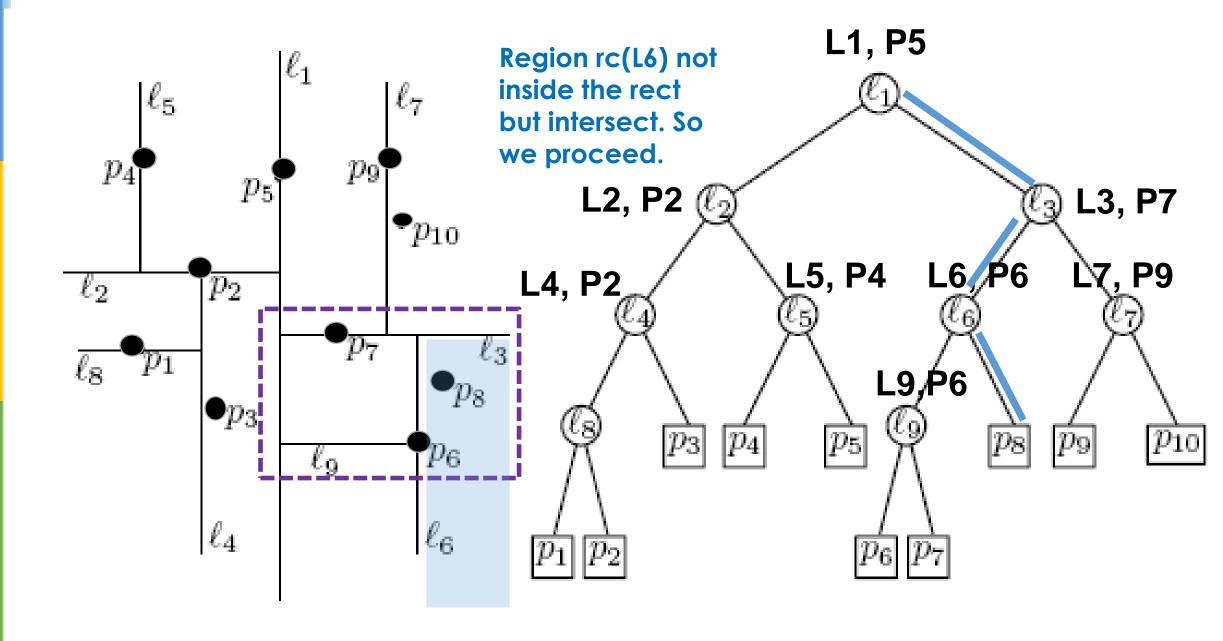


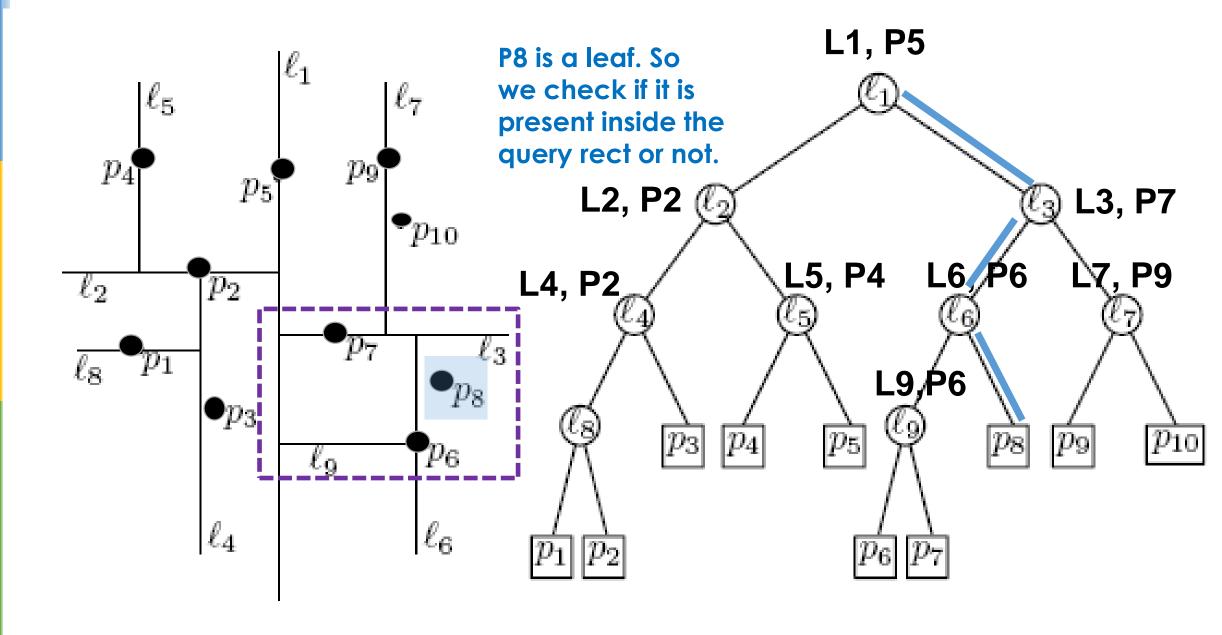


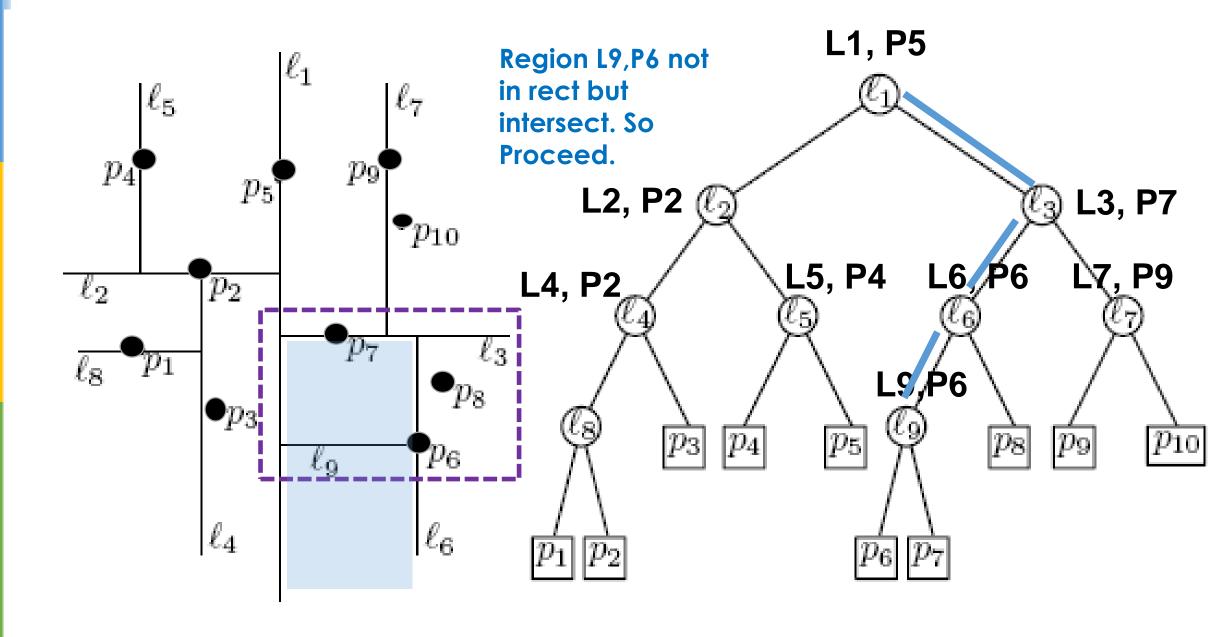


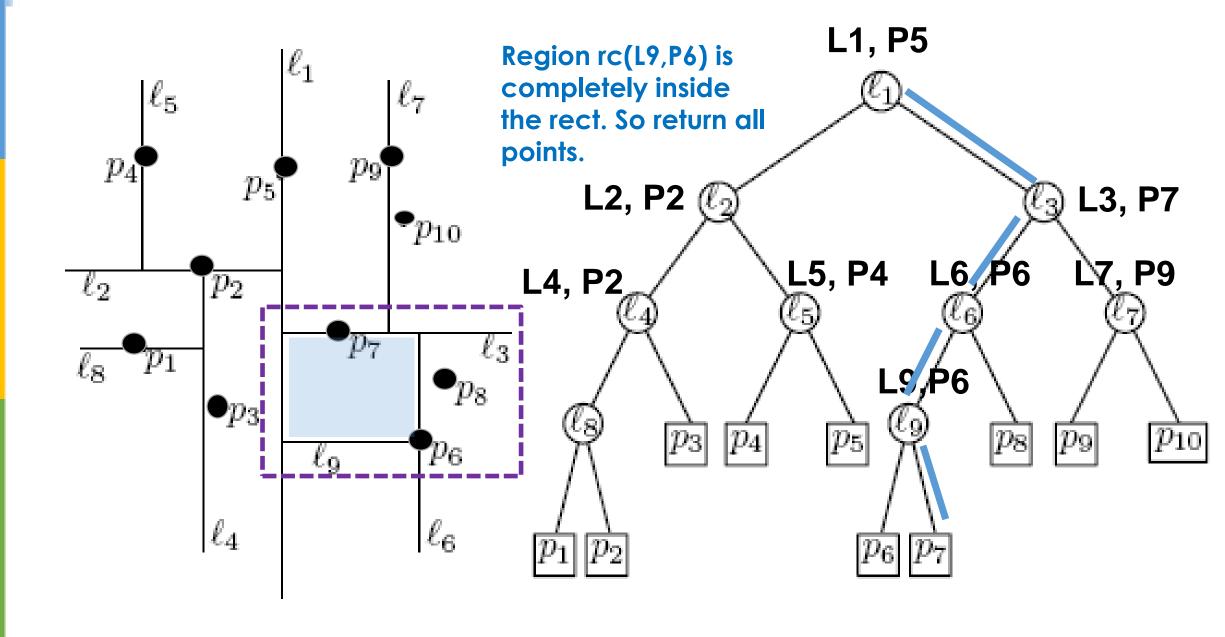


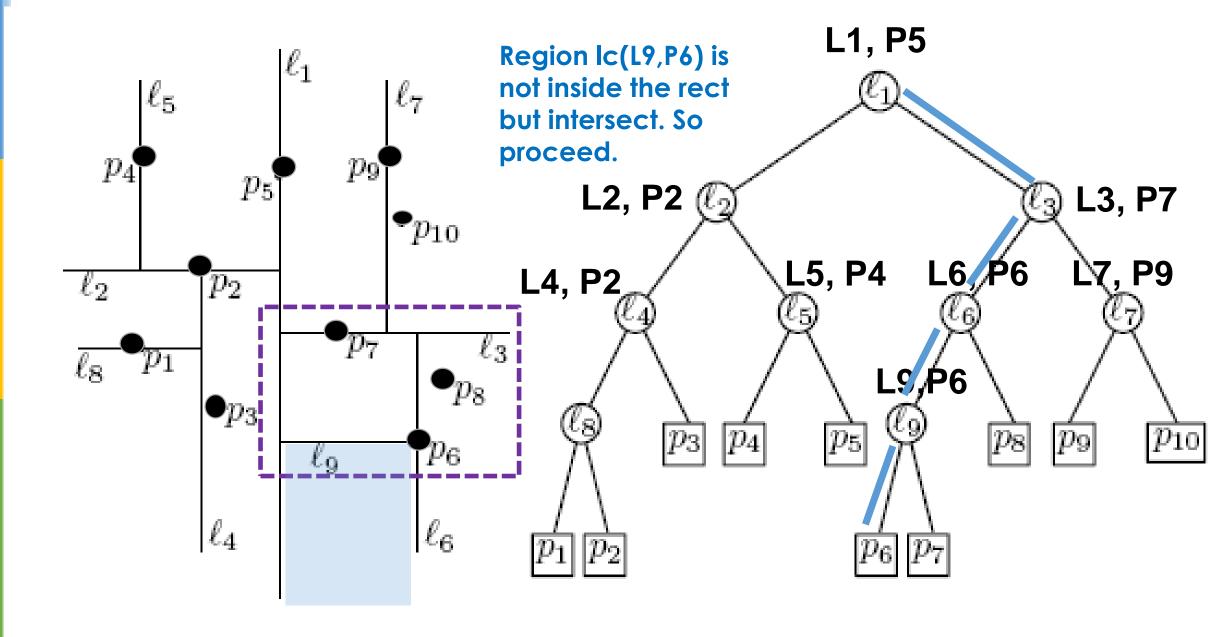


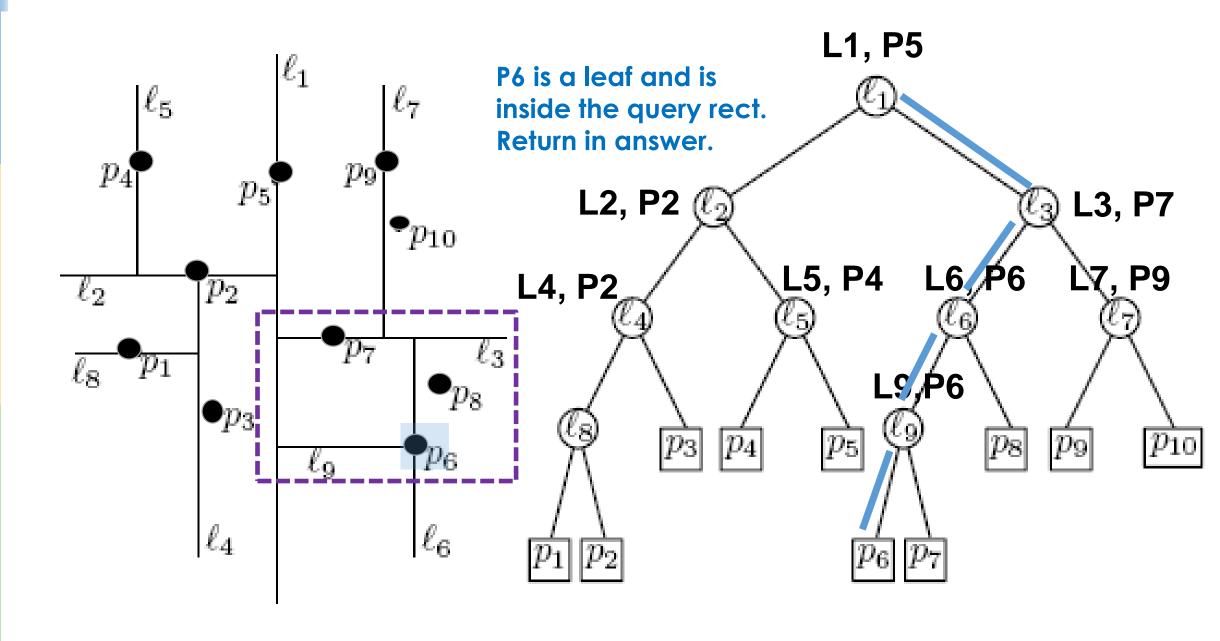




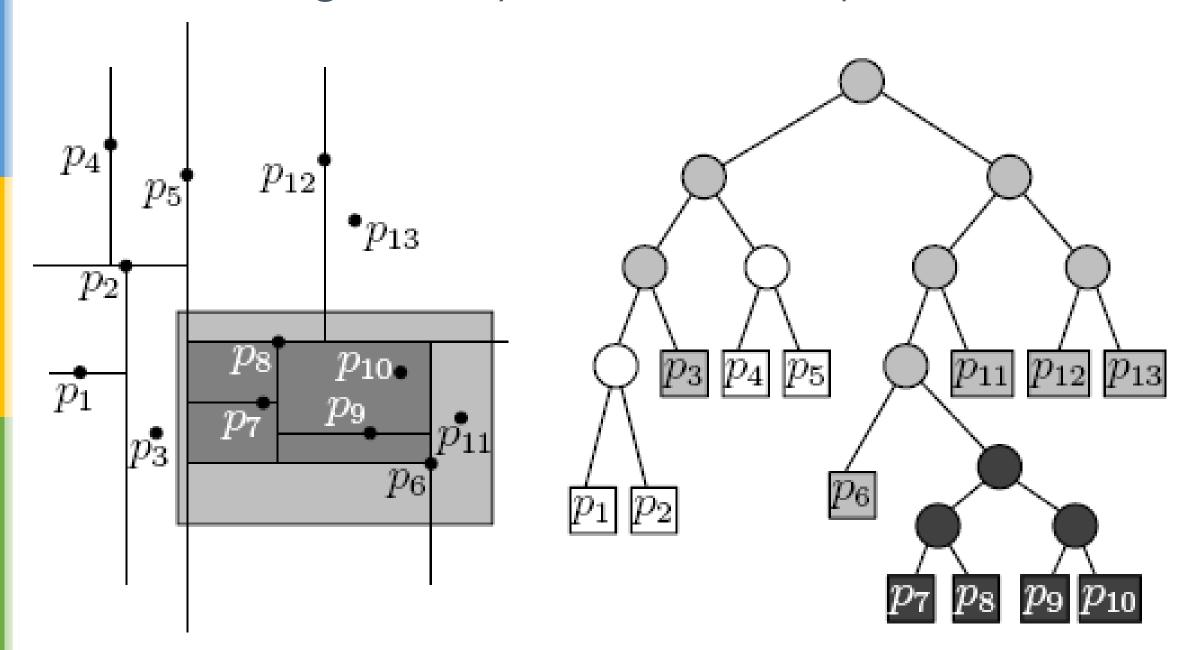






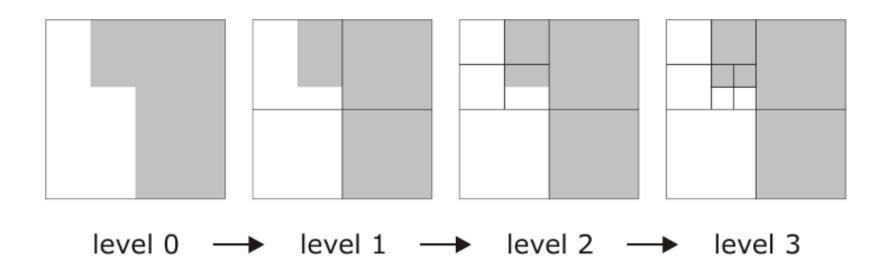


# KD-tree: Range Query Another Example



# Region Quadtrees

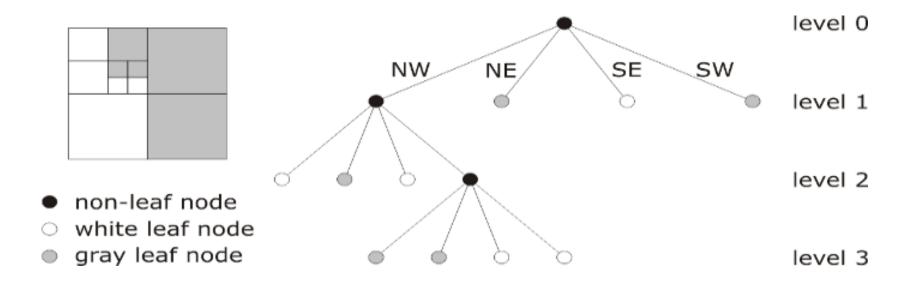
- Quadtree is a tree structure where every non-leaf node has exactly four descendents
- Region quadtrees recursively subdivide non-homogenous square arrays of cells into four equal sized quadrants
- Decomposition continues until all squares bound homogenous regions



Some slides borrowed from "GIS a computational perspective: second edition" by M. Worboys CRC press 2004.

# Region Quadtrees

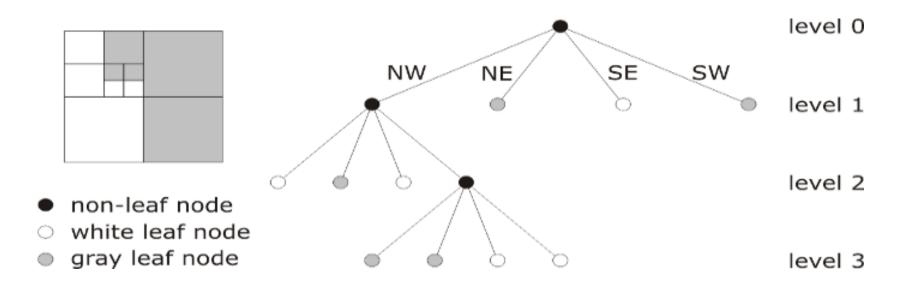
- Quadtrees take full advantage of the spatial structure, adapt to variable spatial detail
- Inefficient for highly inhomogeneous rasters
- Very sensitive to changes in the embedding space (e.g., translation, rotation)



Some slides borrowed from "GIS a computational perspective: second edition" by M. Worboys CRC press 2004.

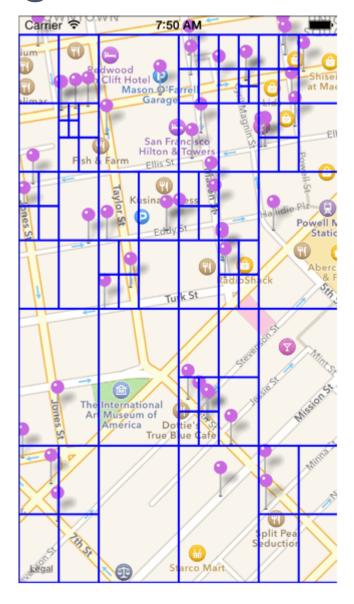
# Region Quadtrees

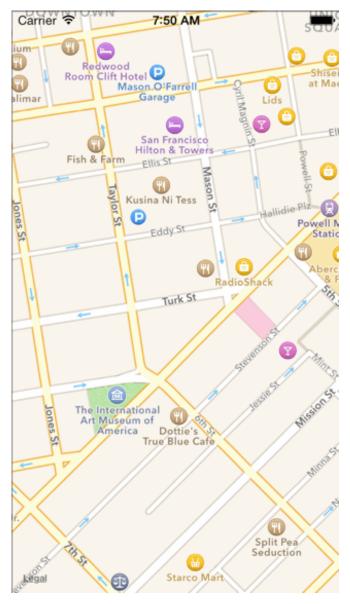
- Quadtrees take full advantage of the spatial structure, adapt to variable spatial detail
- Inefficient for highly inhomogeneous rasters
- Very sensitive to changes in the embedding space (e.g., translation, rotation)
- More useful for representing/compressing raster data.



Some slides borrowed from "GIS a computational perspective: second edition" by M. Worboys CRC press 2004.

#### Region Quadtrees for points Insertion Animation





Animation available at this link: https://robots.thoughtbot.com/how-to-handle-large-amounts-of-data-on-maps

#### Region Quadtrees for points Range Query Example



Animation available at this link: https://robots.thoughtbot.com/how-to-handle-large-amounts-of-data-on-maps

