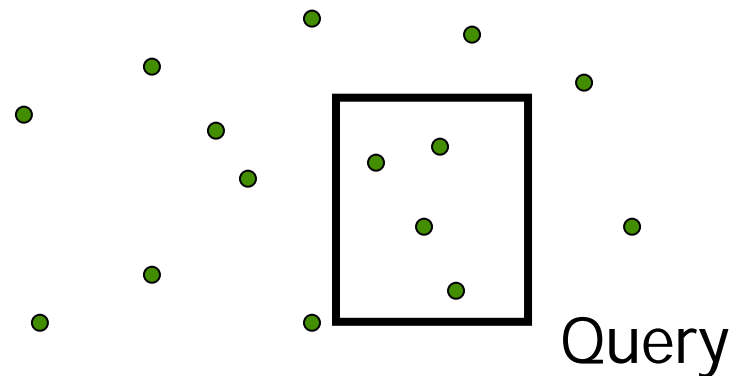


Query Processing and Advanced Queries

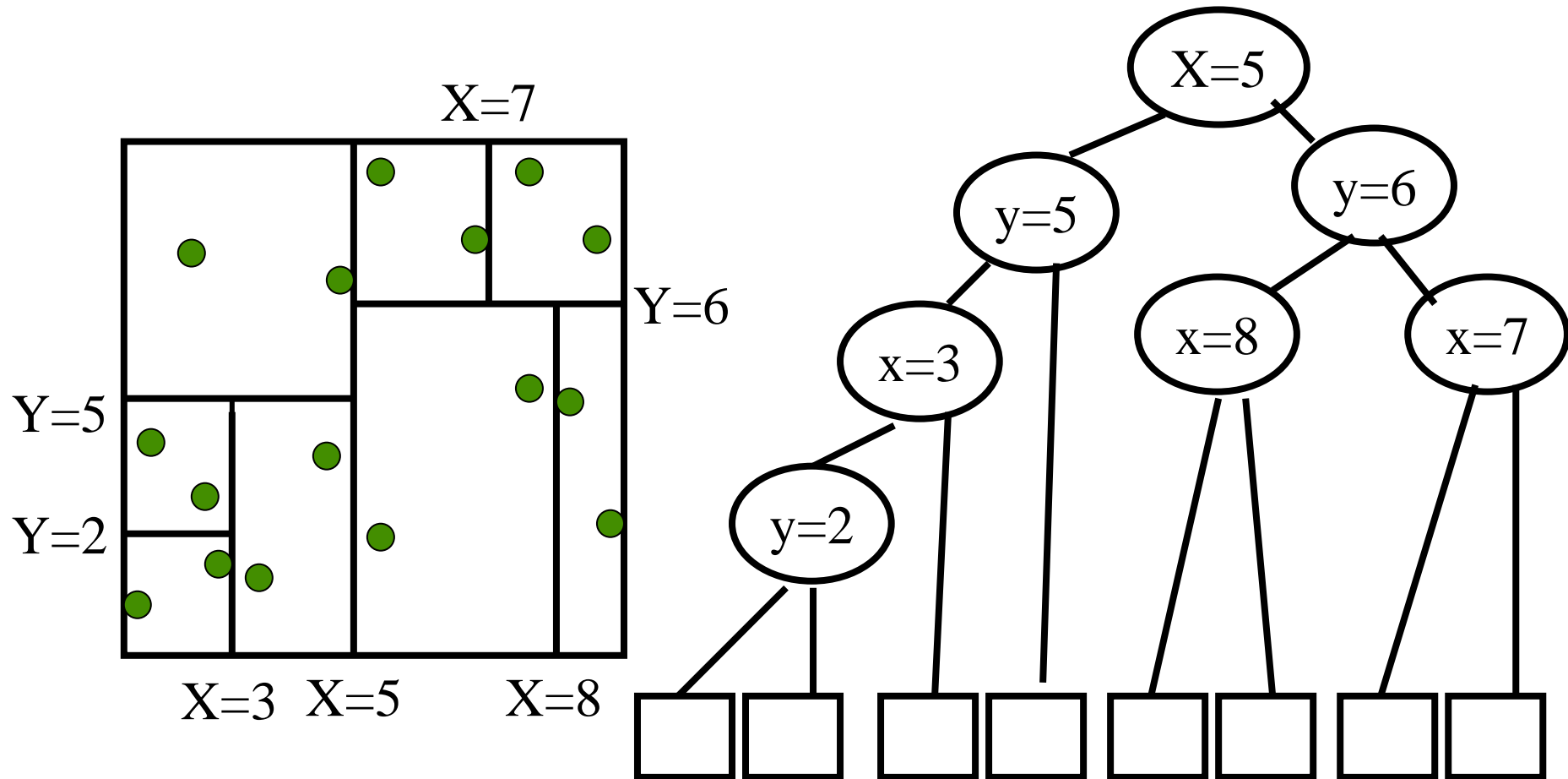
Advanced Queries (2): R-Tree

Review: PAM

- Given a point set and a rectangular query, find the points enclosed in the query
- We allow insertions/deletions online

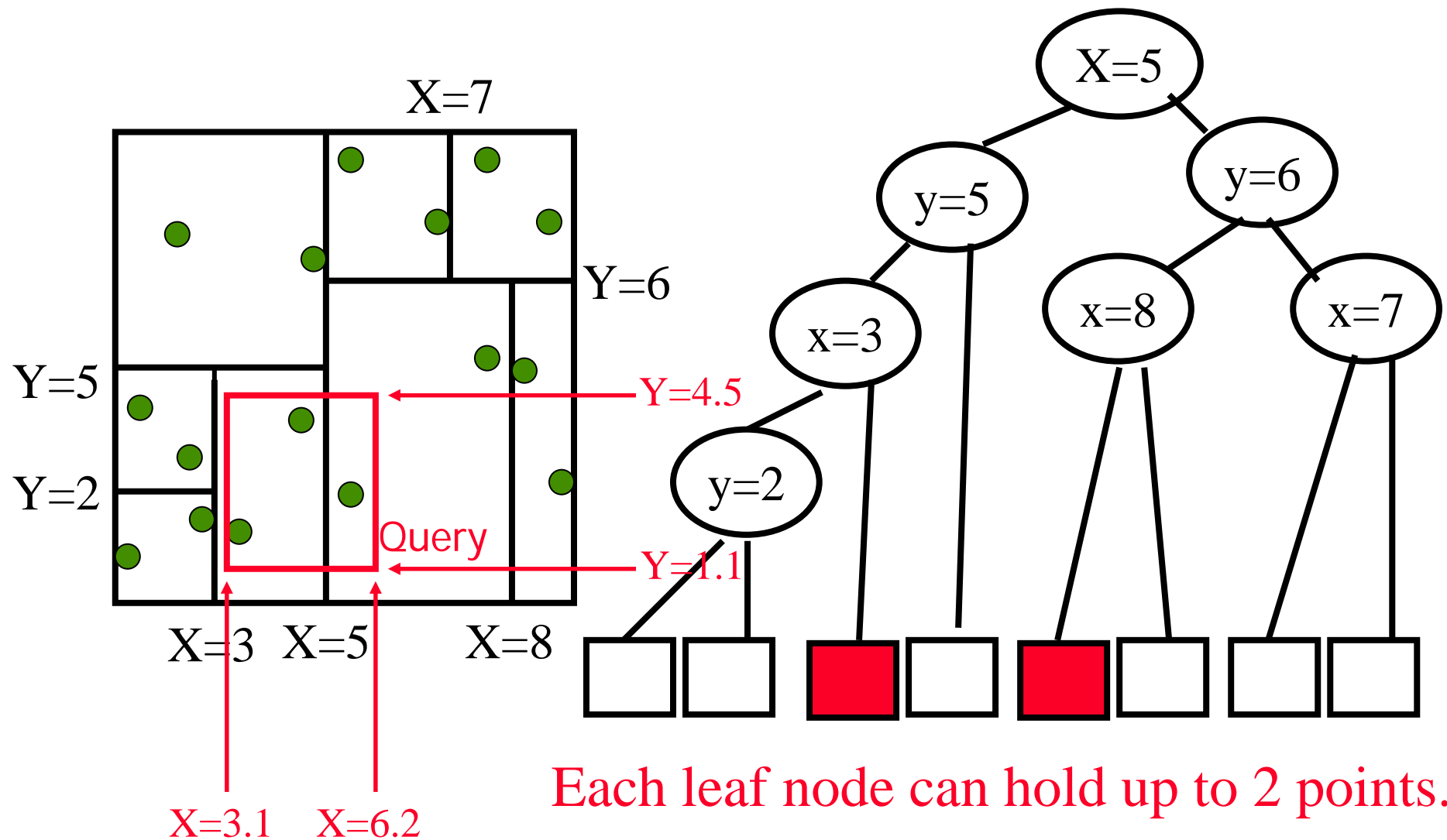


Review: kd-Tree



Each leaf node can hold up to 2 points.

Query Answering using kd-Tree



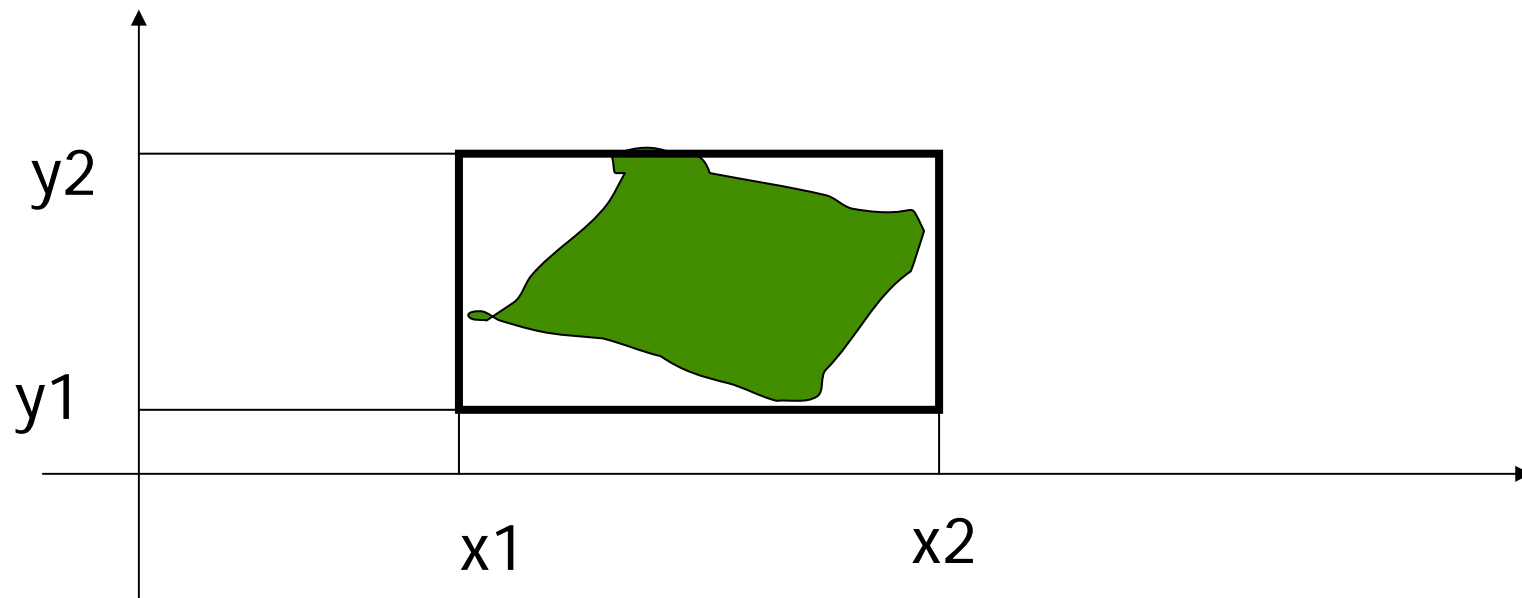
Each leaf node can hold up to 2 points.

Review: Spatial Indexing

- Point Access Methods can index only points. What about regions?
 - Use the transformation technique and a PAM
 - New methods: Spatial Access Methods SAMs
 - R-tree and variations

Minimum Bounding Rectangle

- Approximate each region with a simple shape: usually Minimum Bounding Rectangle (MBR) = $[(x1, x2), (y1, y2)]$



Transformation Technique

- Map an d-dim MBR into a point: ex.

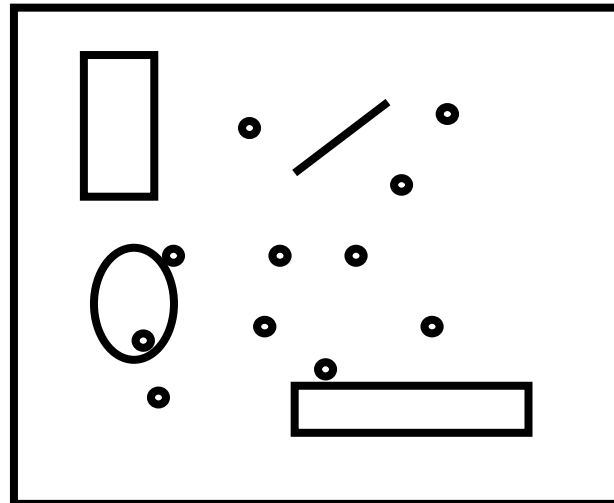
$[(x_{\min}, x_{\max}) (y_{\min}, y_{\max})] \Rightarrow$

$(x_{\min}, x_{\max}, y_{\min}, y_{\max})$

- Use a PAM to index the 2d points
- Given a range query, map the query into the 2d space and use the PAM to answer it

SAM: The Problem

- Given a collection of geometric objects (points, lines, polygons, ...)
- Organize them on disk, to answer spatial queries (e.g., range query, NN query, etc)



Indexing using SAMs

- Two steps:
 - Filtering step: Find all the MBRs (using the SAM) that satisfy the query
 - Refinement step: For each qualified MBR, check the original object against the query

R-Trees

- [Guttman 84] Main idea: allow parents to overlap!
 - => guaranteed 50% utilization
 - => easier insertion/split algorithms.
 - (only deal with Minimum Bounding Rectangles - **MBRs**)

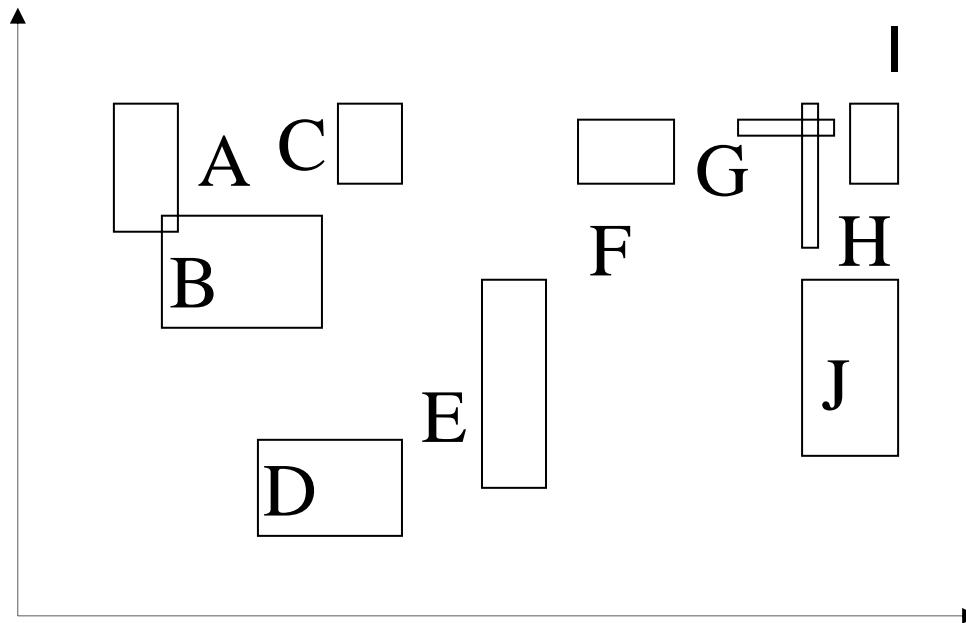


R-trees

- A multi-way external memory tree
- Index nodes and data (leaf) nodes
- All leaf nodes appear on the same level
- Every node contains between m and M entries
- The root node has at least 2 entries (children)

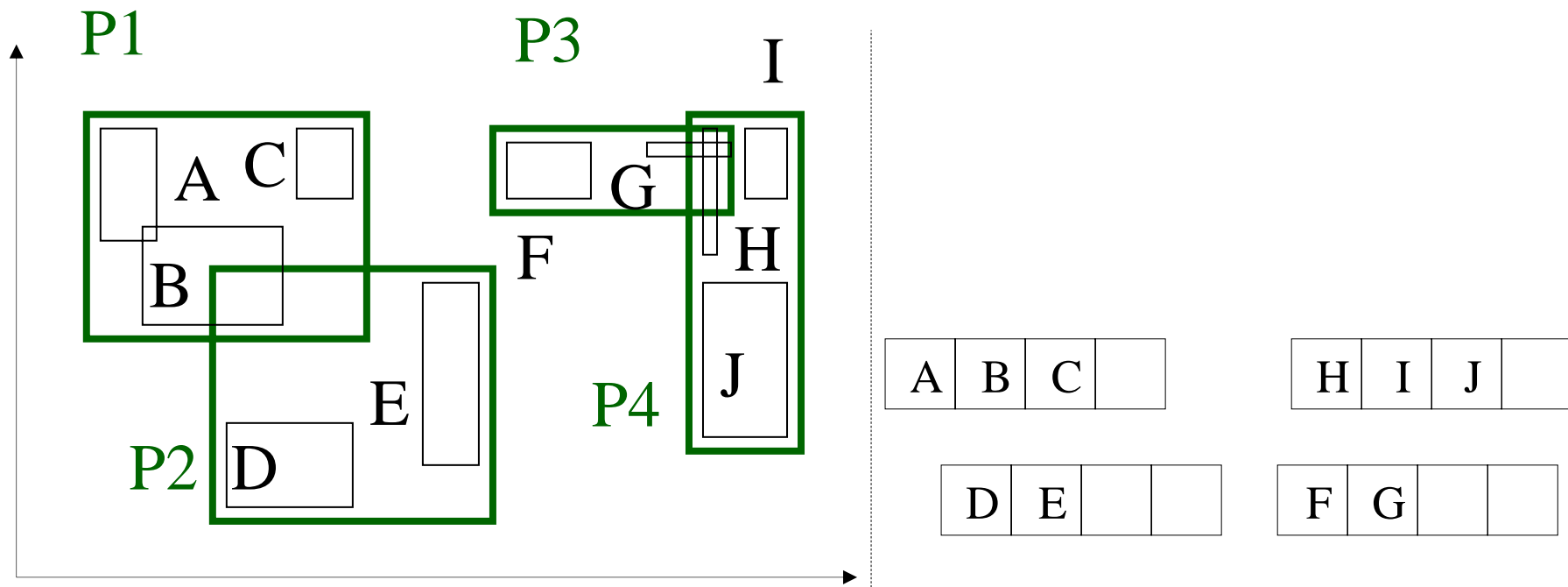
Example

- Fan-out = 4: group nearby rectangles to parent MBRs; each group -> disk page



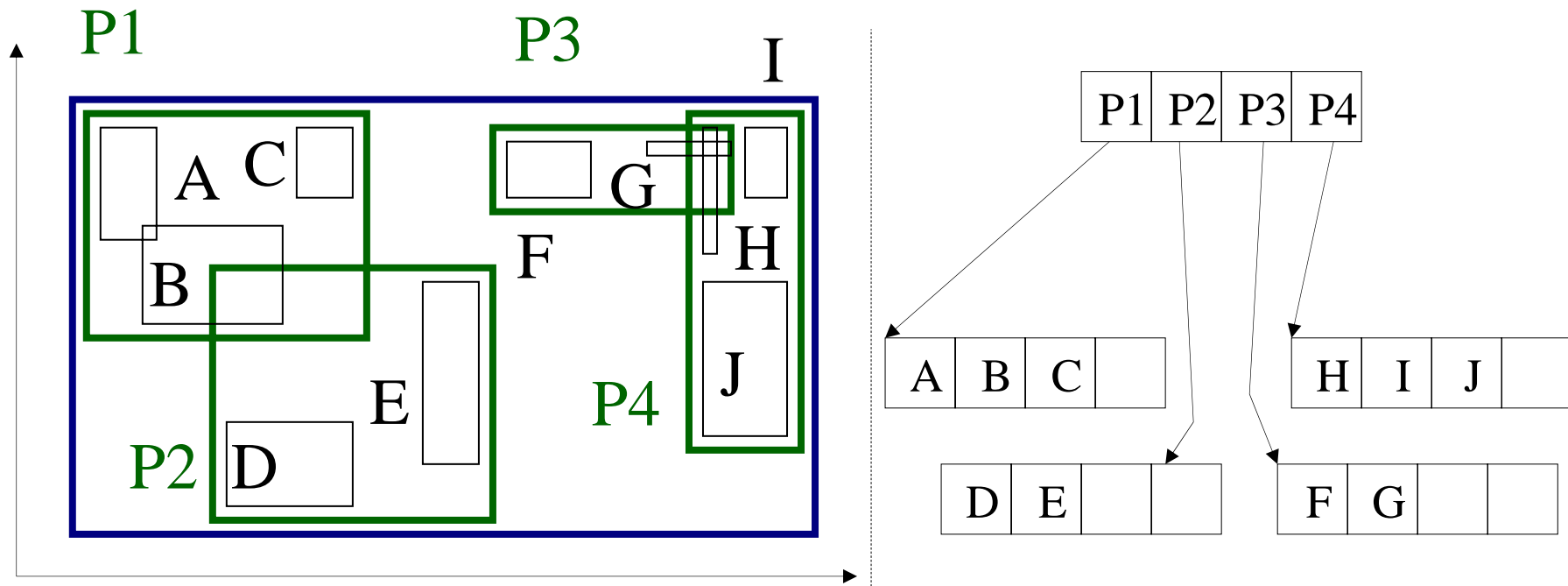
Example

■ F=4



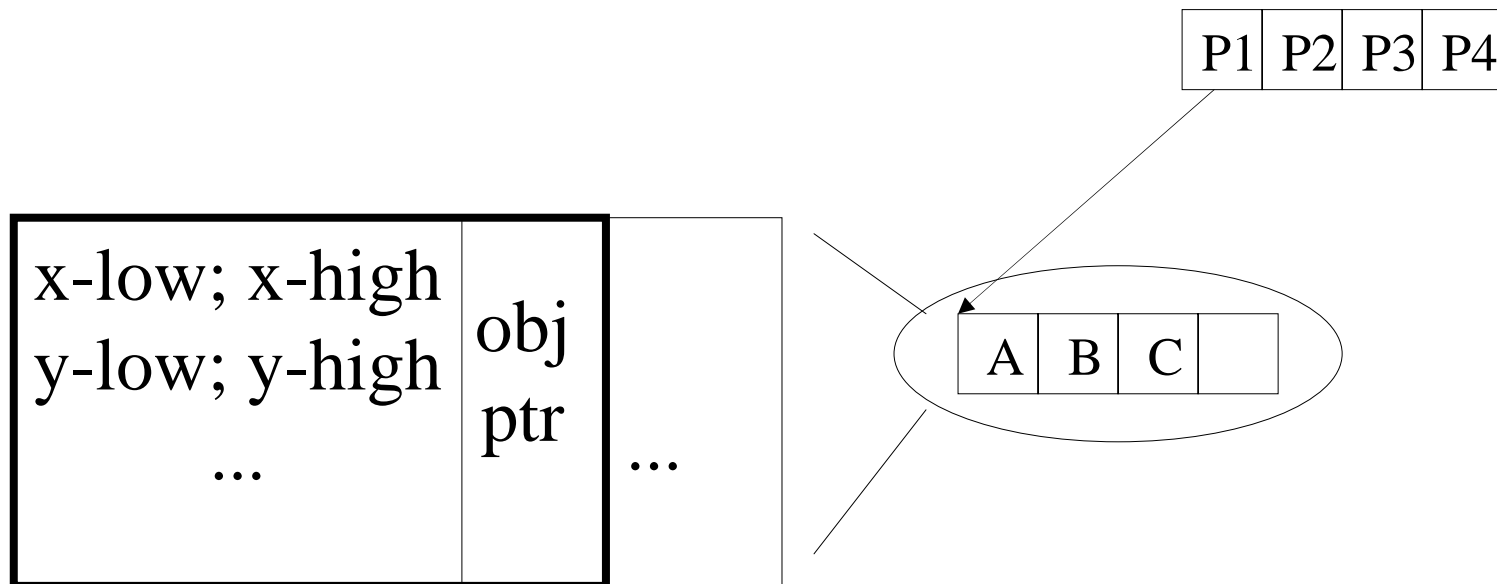
Example

■ F=4



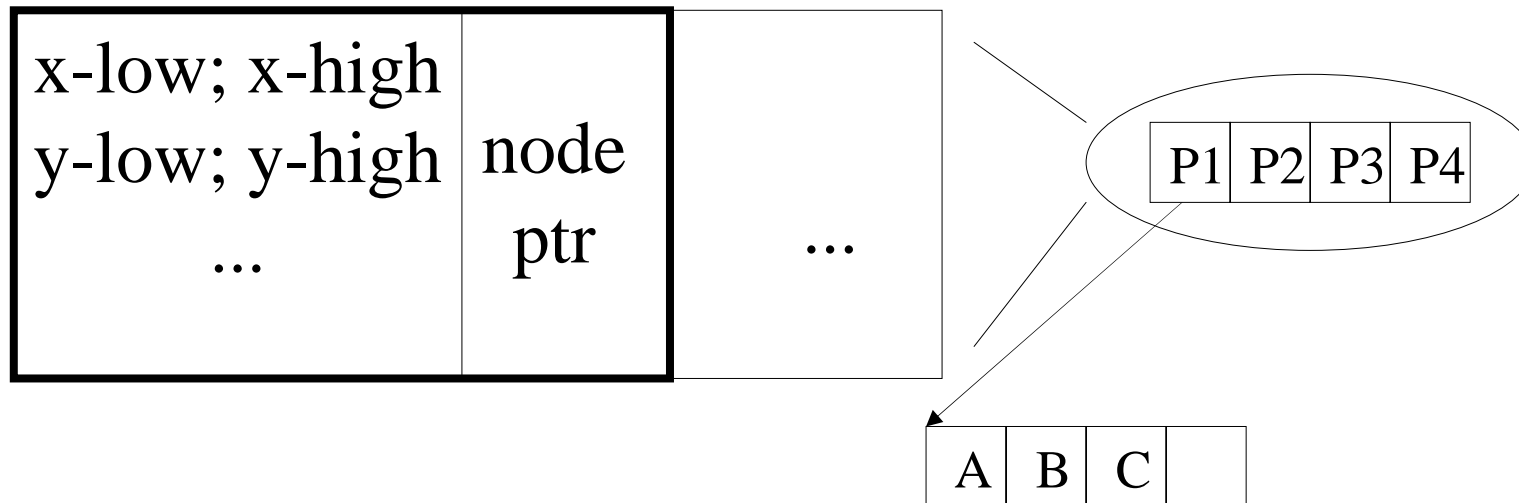
R-Trees - Format of Nodes

- {(MBR; obj_ptr)} for leaf nodes

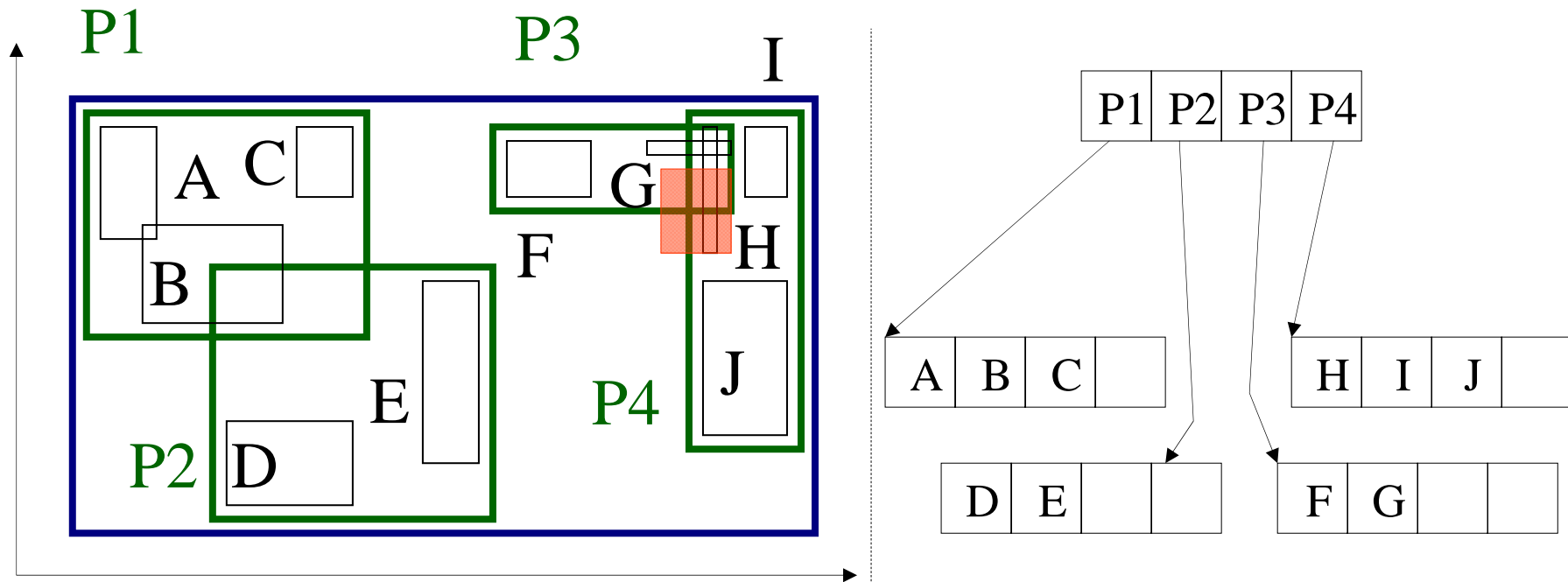


R-Trees - Format of Nodes

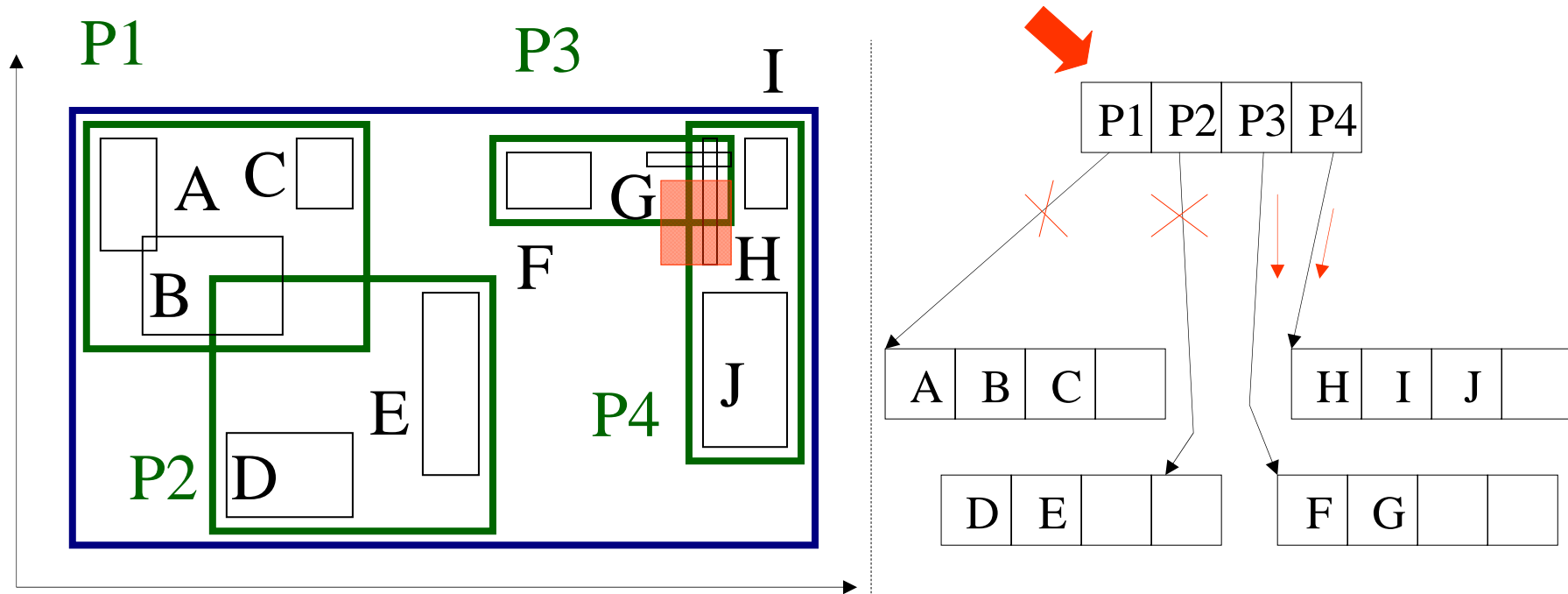
- {(MBR; node_ptr)} for non-leaf nodes



R-Trees: Search



R-Trees: Search



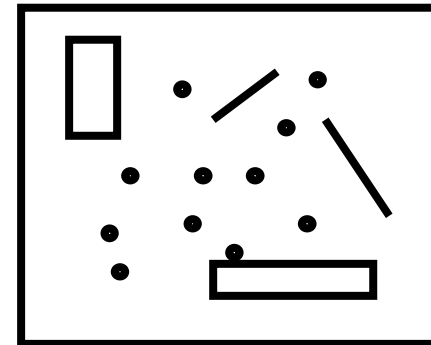
R-Trees: Search

■ Main points:

- Every parent node completely covers its 'children'
- A child MBR may be covered by more than one parent - it is stored under **ONLY ONE** of them.
- A point query may follow multiple branches.
- Everything works for any(?) dimensionality

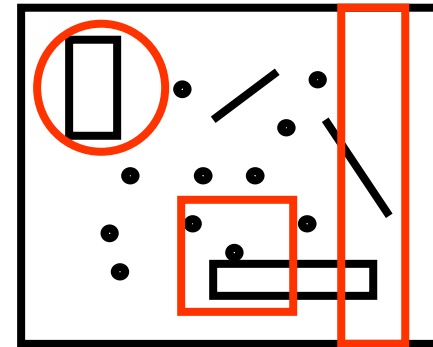
Spatial Queries

- Given a collection of geometric objects (points, lines, polygons, ...)
- organize them on disk, to answer efficiently
 - range queries
 - k-nn queries



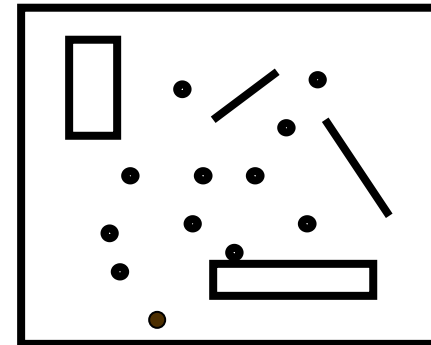
Spatial Queries

- Given a collection of geometric objects (points, lines, polygons, ...)
- organize them on disk, to answer efficiently
 - range queries
 - k-nn queries



Spatial Queries

- Given a collection of geometric objects (points, lines, polygons, ...)
- organize them on disk, to answer efficiently
 - range queries
 - **k-nn queries**



R-Trees - Range Search

pseudocode:

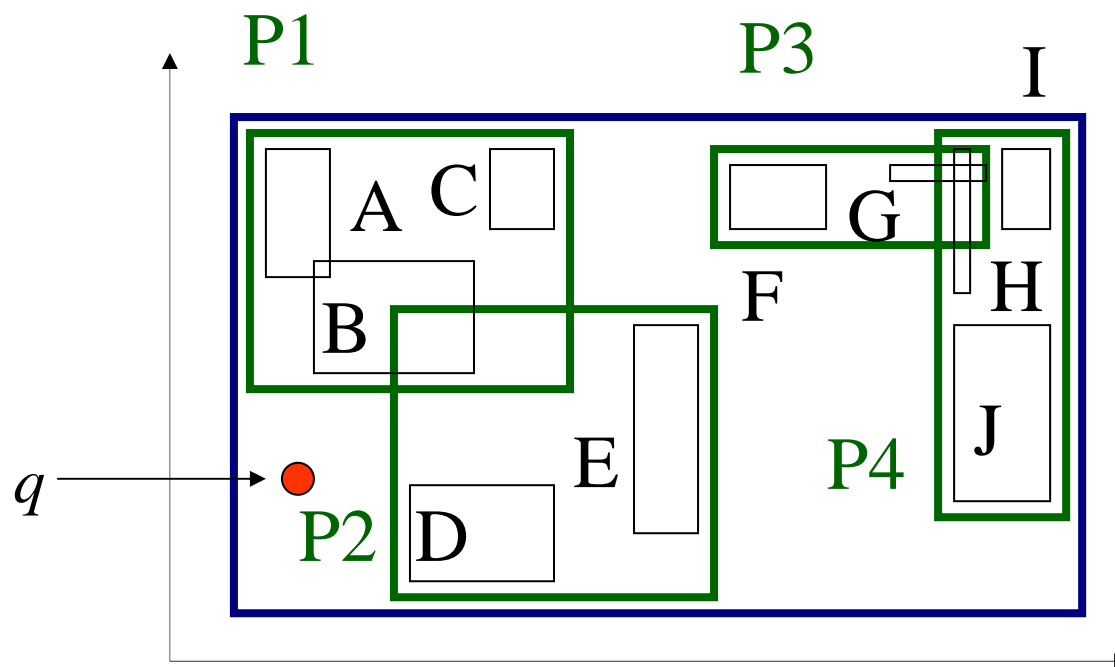
check the root

for each branch,

if its MBR intersects the query rectangle

apply range-search (or print out, if this
is a leaf)

R-Trees - NN Search



Two Metrics to Ordering the NN Search

- *MINDIST* (P, R) is the minimum distance between a point P and a rectangle R .
- If the point is inside the rectangle, $MINDIST = 0$;
- If the point is outside the rectangle, $MINDIST$ is the minimal possible distance from the point to any object in or on the perimeter of the rectangle.

$$\forall o \in R, MINDIST(P, R) \leq \|(P, o)\|$$

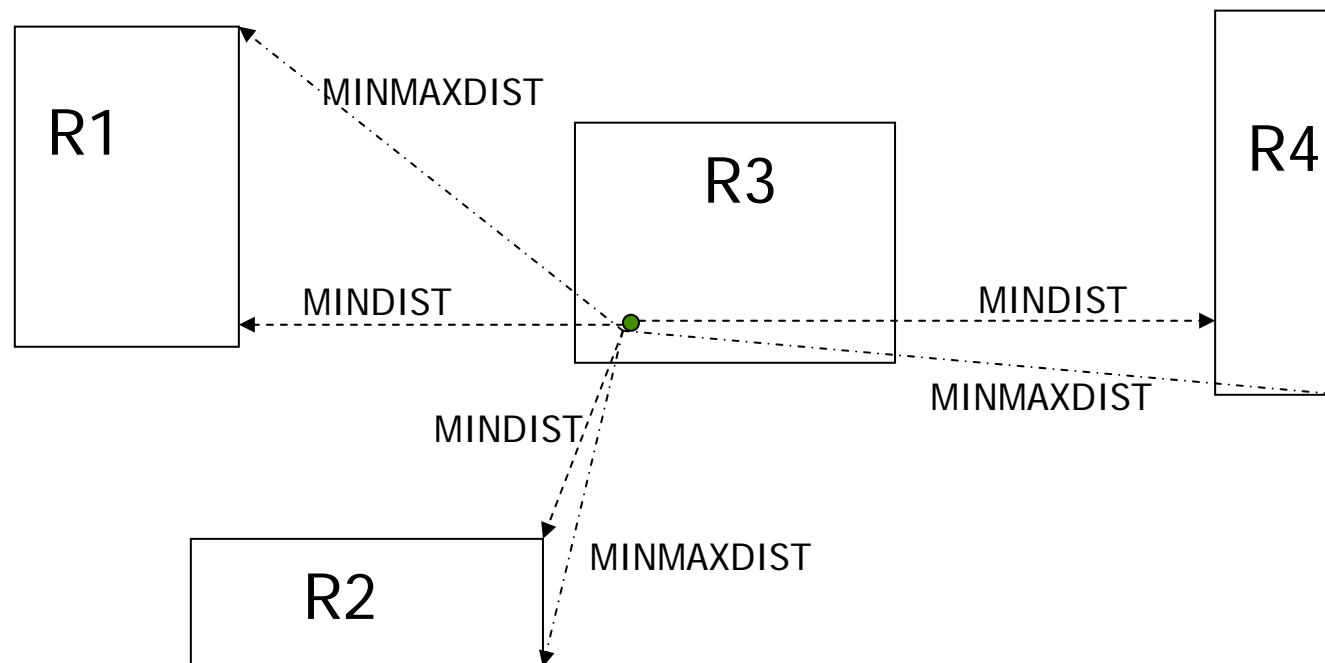
MINMAXDIST

- $\text{MINMAXDIST}(P, R)$: for each dimension, find the closest face, compute the distance to the furthest point on this face and take the minimum of all these (d) distances
- $\text{MINMAXDIST}(P, R)$ is the smallest possible upper bound of distances from P to R
- MINMAXDIST guarantees that there is at least one object in R with a distance to P smaller or equal to it.

$$\exists o \in R, \|(P, o)\| \leq \text{MINMAXDIST}(P, R)$$

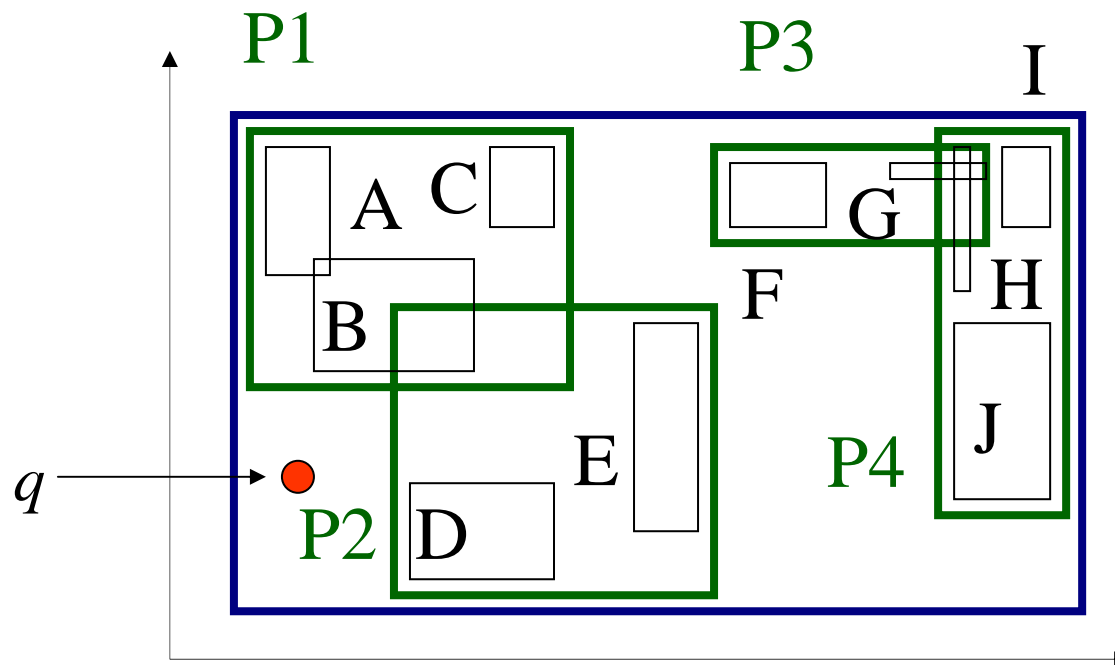
MINDIST and MINMAXDIST

- $MINDIST(P, R) \leq NN(P) \leq MINMAXDIST(P, R)$



R-Trees - NN Search

- Q: How? (find near neighbor; refine...)



R-Trees - NN Search

- Q: How? (find near neighbor; refine...)

