

Spatial data structures in 2D

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Application areas

- ◆ **geographic information systems (GIS)**
 - area, line and point entities
 - huge databases (10^4 to 10^9 objects)
- ◆ **image analysis, recognition**
- ◆ **computer graphics, games**
 - 2D & 3D algorithm speedup (ray casting, collisions)
- ◆ **industry, CAD**
 - VLSI design, component positioning, collisions



Elementary tasks I

- ◆ **point localization** in 2D net
 - looking for an area object which contains the point
- ◆ **nearest N points** from the given center
 - global variation: looking for the closest point pair
- ◆ **curve intersections** (polylines)
 - collisions in a set of curves (polylines)
- ◆ **point objects closest to the given curve** (route)



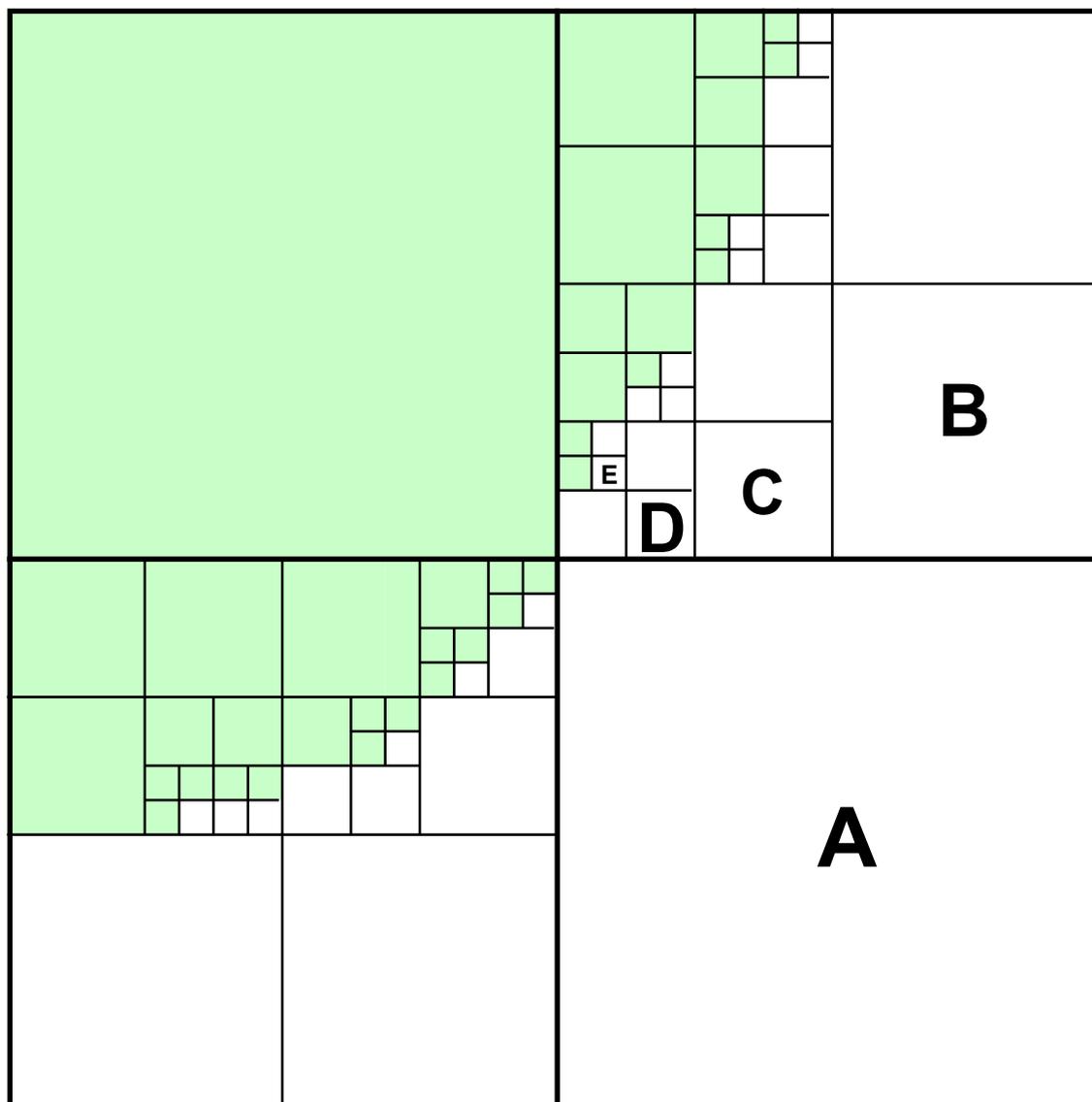
Elementary tasks II

- ◆ **interval queries** in 2D space (databases)
- ◆ **set operations** on map entities
 - areas, line objects, points
- ◆ **collision tests (interferences)**, minimal distances among planar objects (VLSI)
- ◆ **object processing in some geometric order**
 - increasing distance from a given center



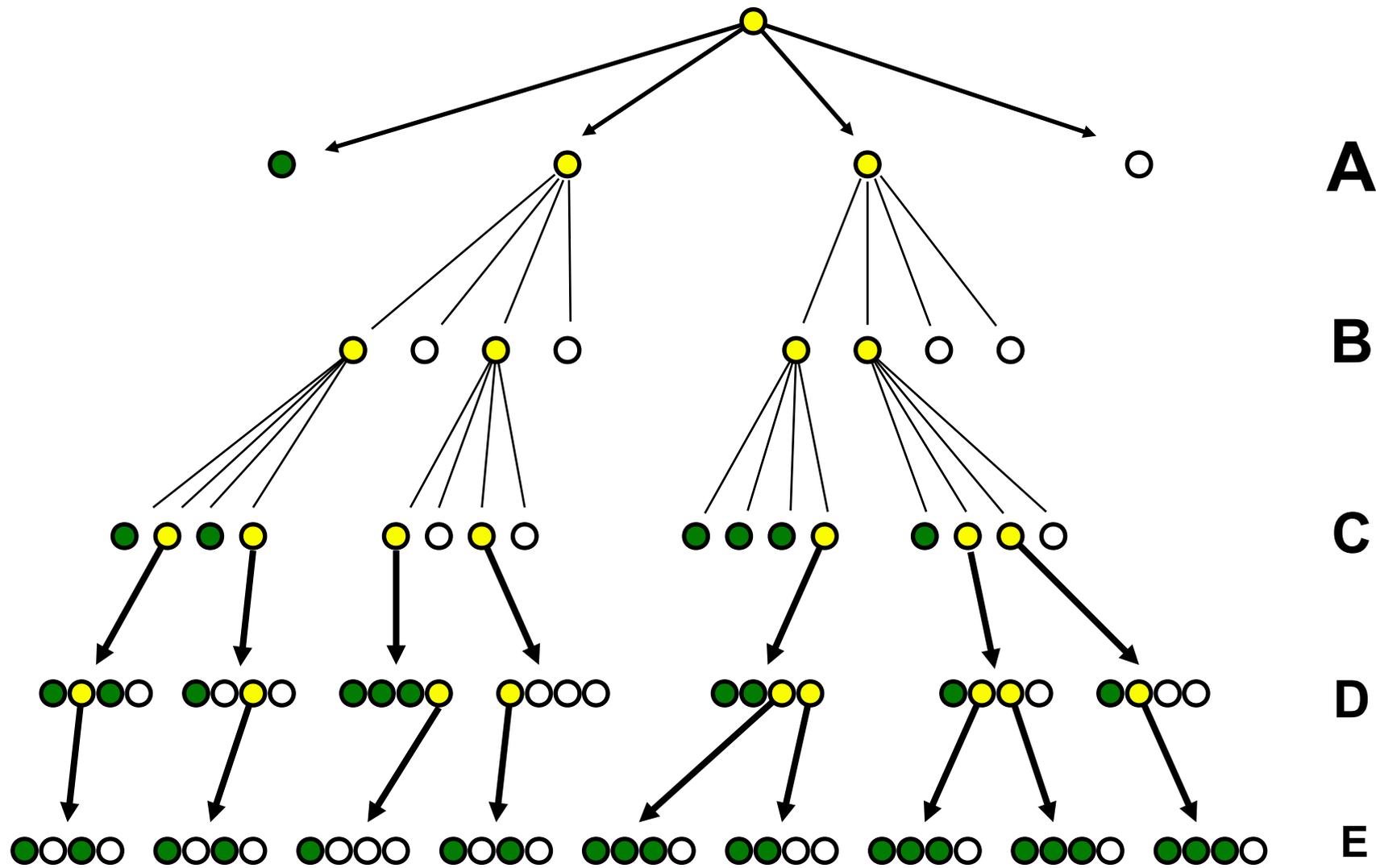
“Region quadtree”

- ◆ **representation of areal regions in 2D**
- ◆ **splitting exactly in the middle**
- ◆ **information is stored in leaf nodes only**





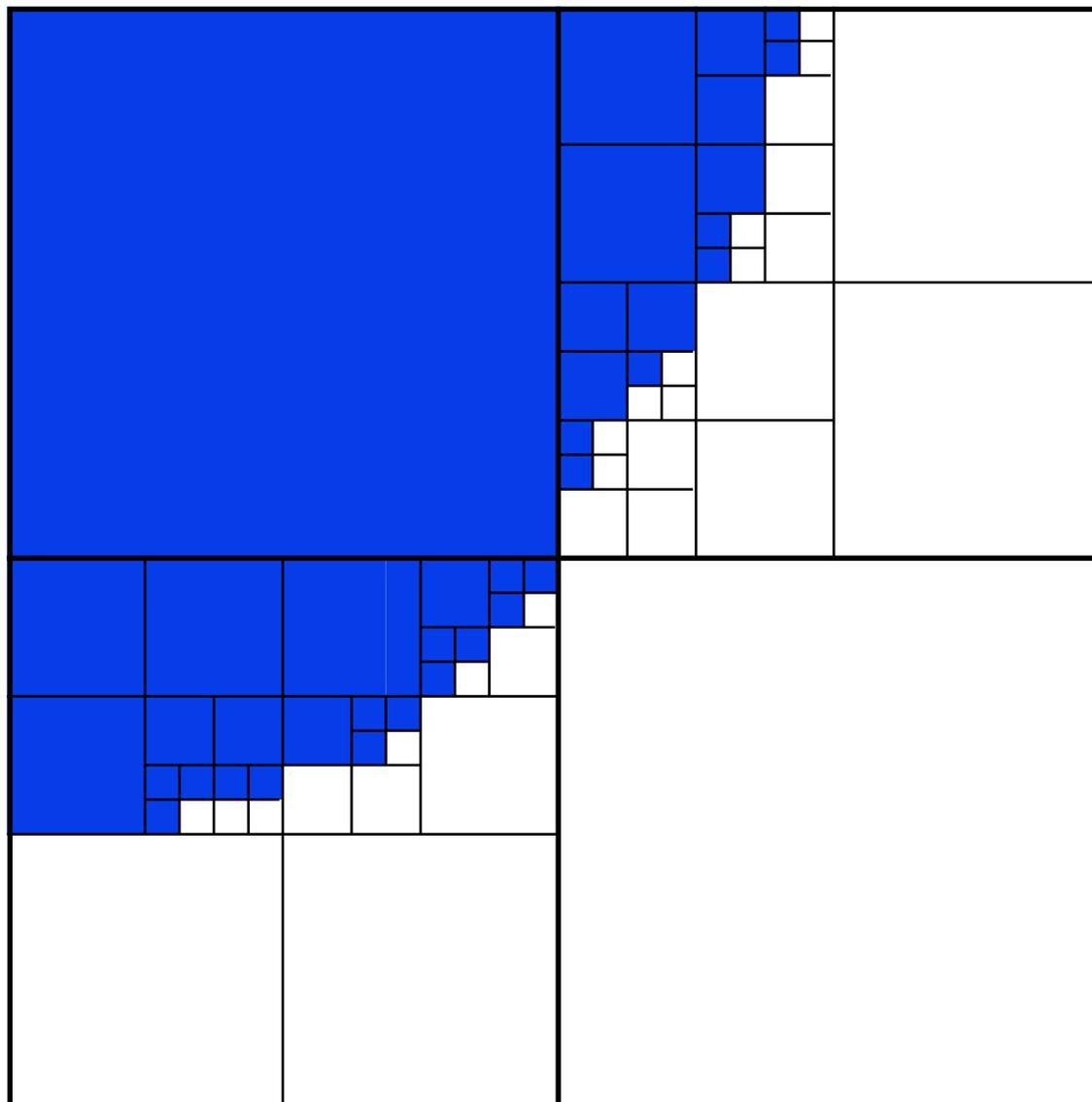
“Region quadtree”





Pyramid

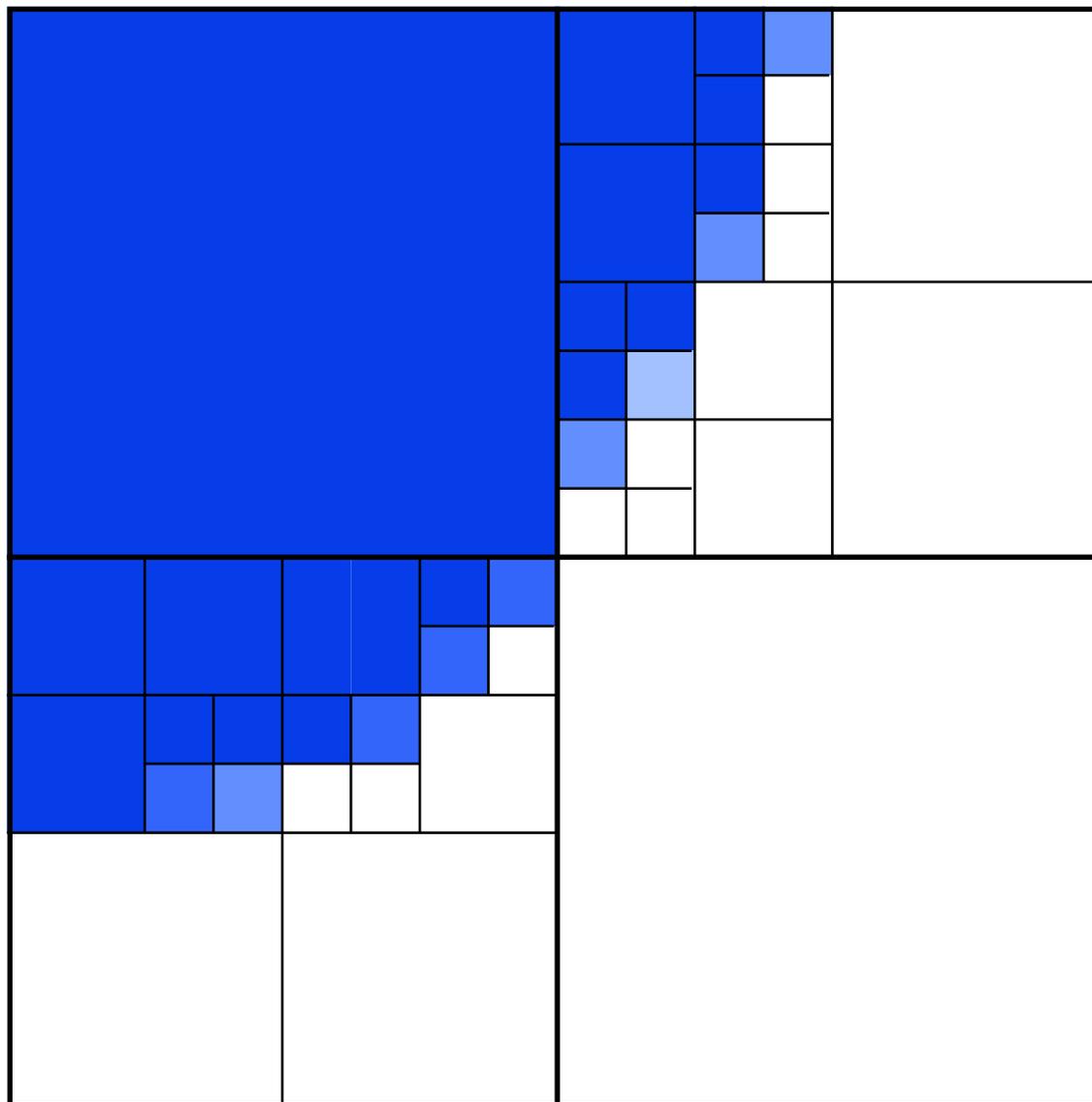
- ◆ **representation of areal regions in 2D**
- ◆ **additional summary info in inner nodes**
- ◆ **pyramid nodes up to level: **E****





Pyramid

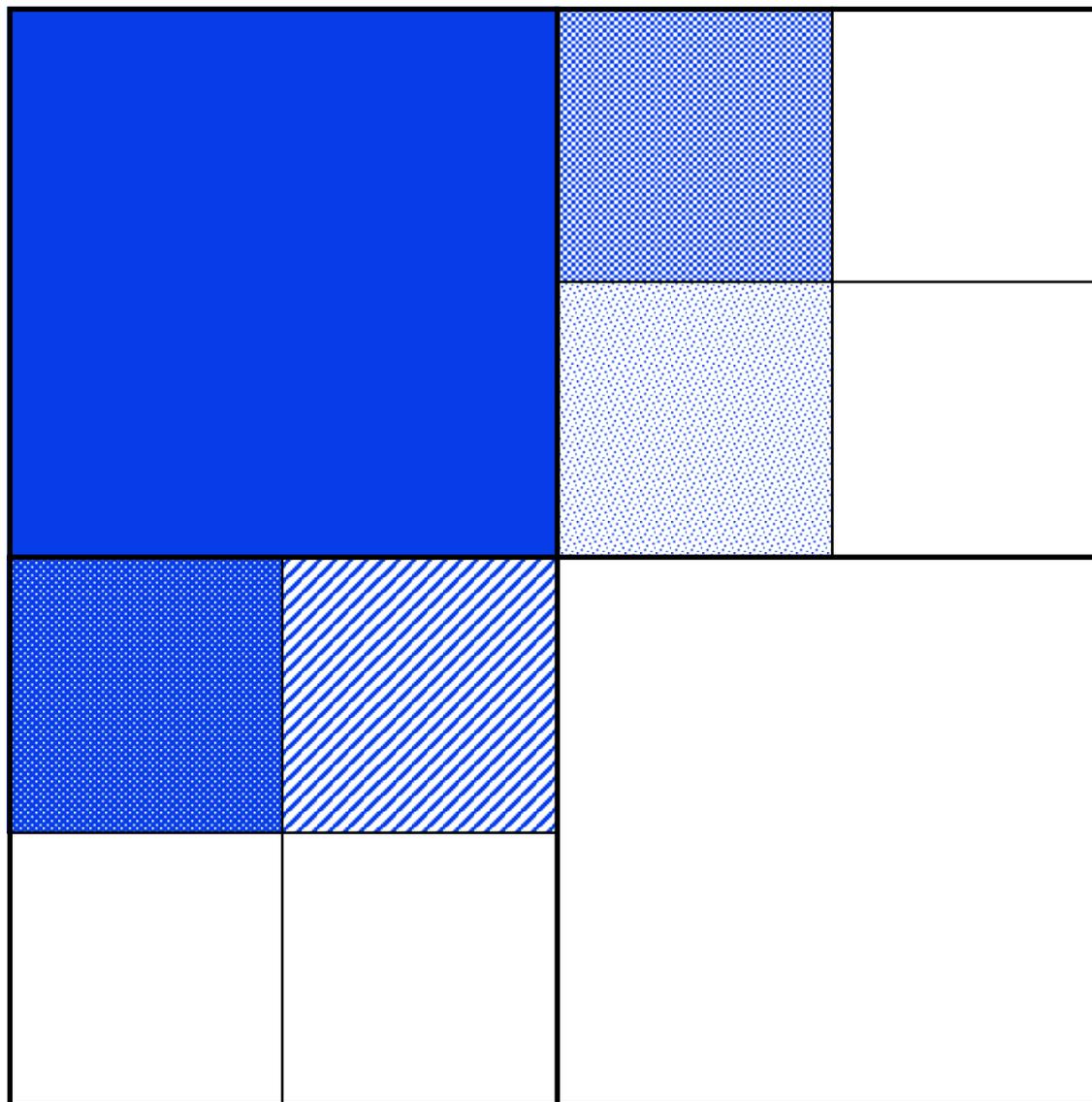
- ◆ pyramid nodes up to level: D





Pyramid

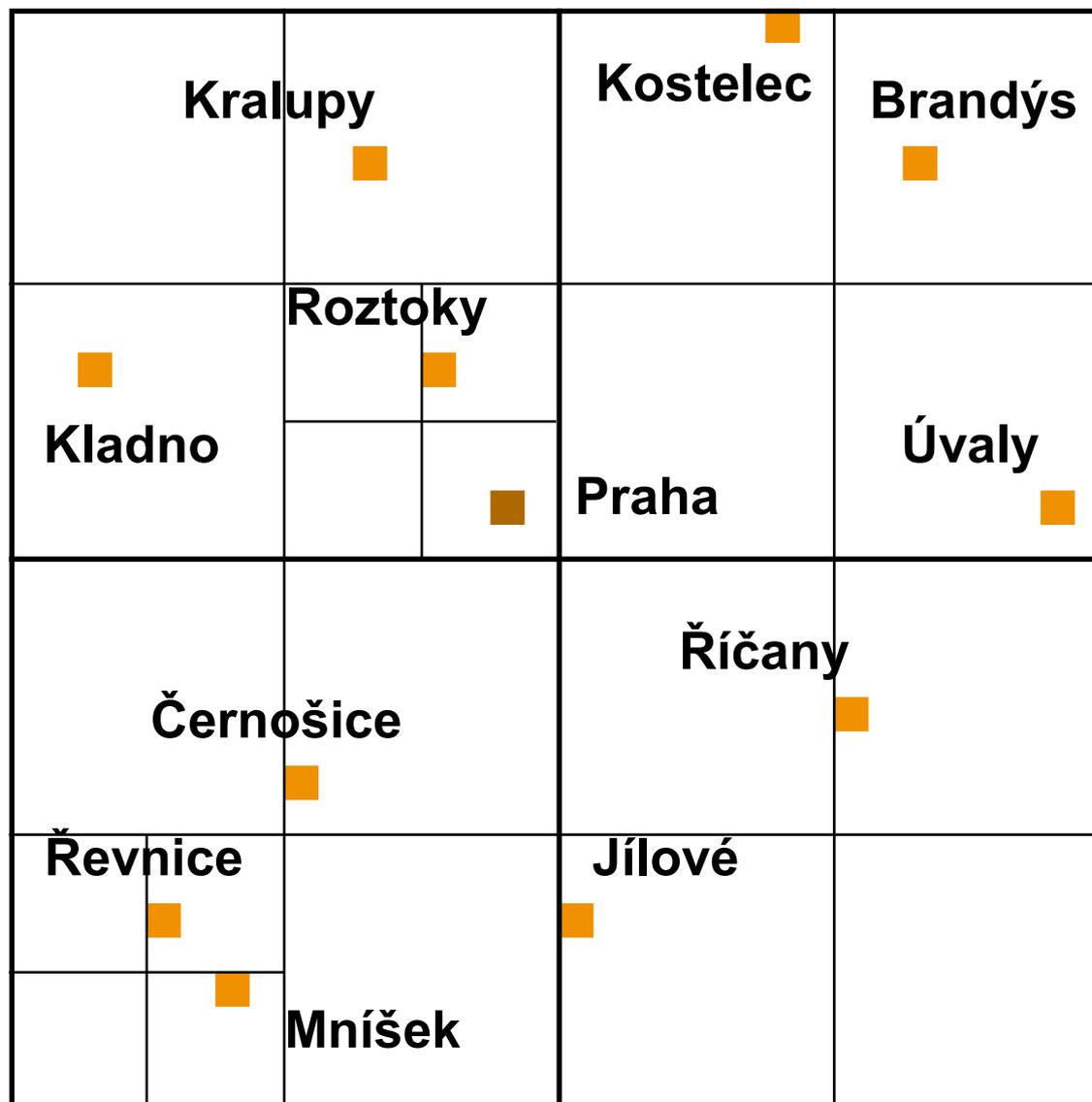
- ◆ pyramid nodes up to level: **B**





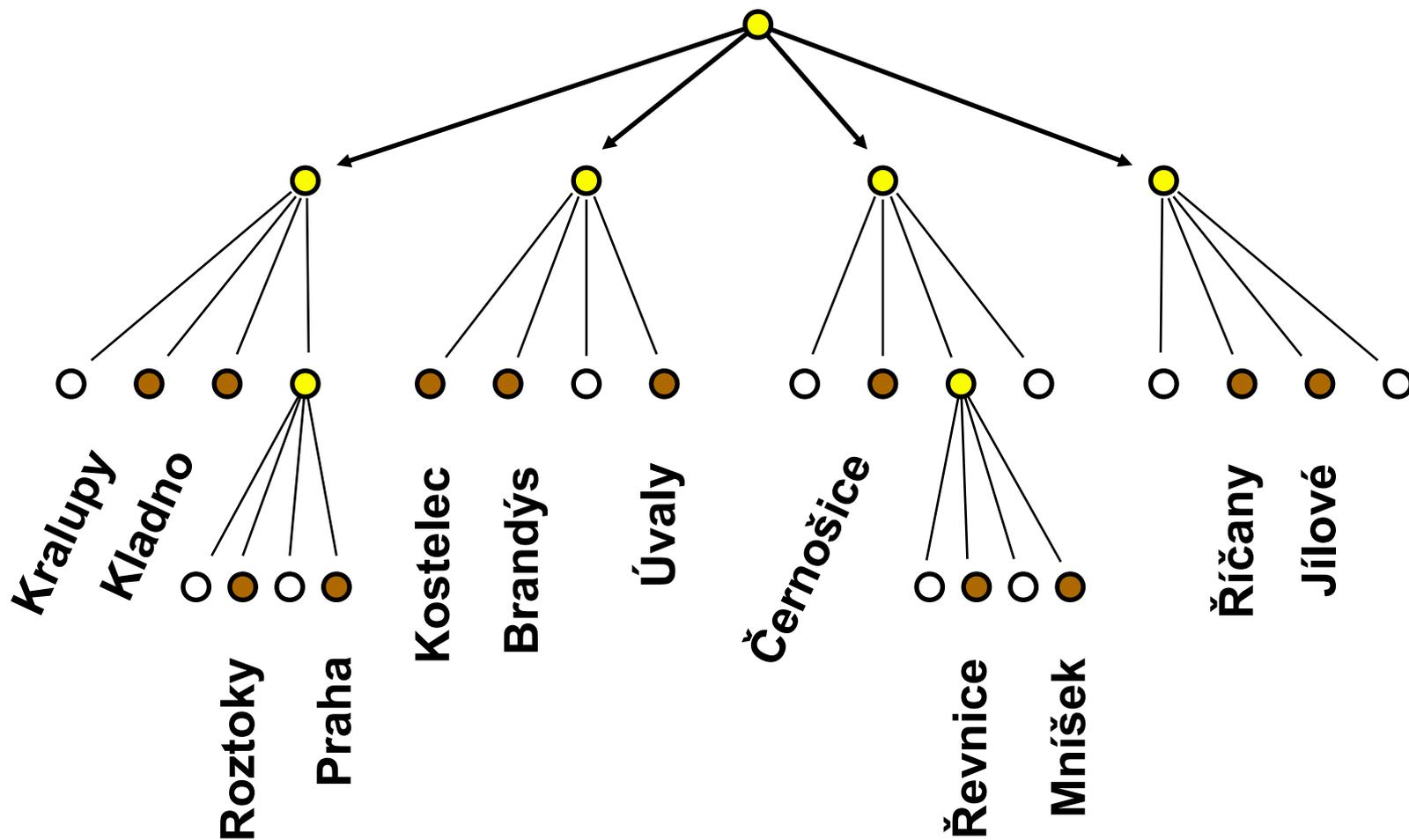
“PR quadtree” (Point-Region)

- ◆ **representation of point objects**
- ◆ **splitting exactly in the middle**
- ◆ **object information stored in leaf nodes (all levels, max. one object per node)**





“PR quadtree” (Orenstein)





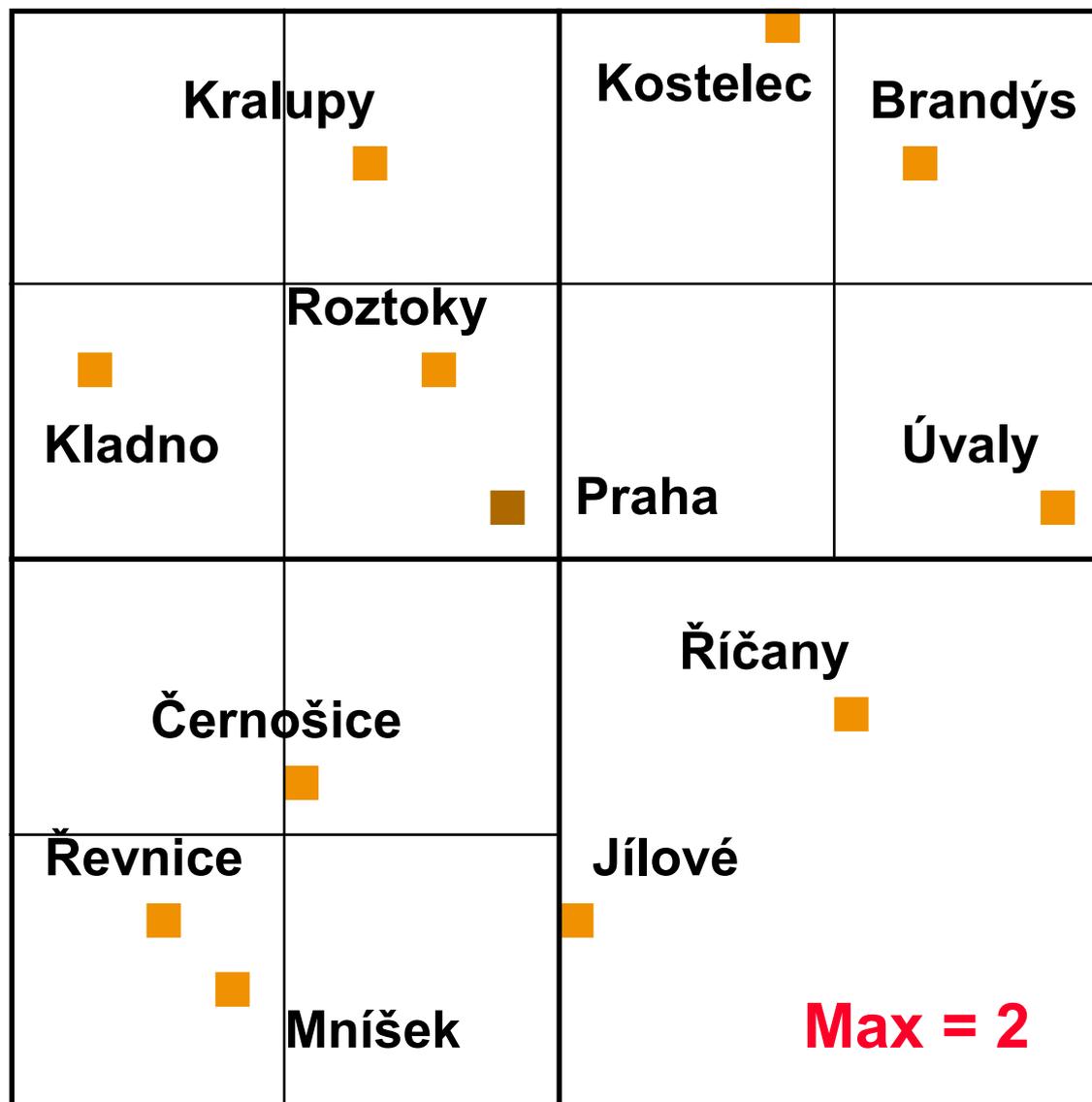
Bucket methods

- node can store information about **more than one object**
 - $0 < \text{object\#} \leq \text{Max}$**
 - **Max** is constant value tuned for storage efficiency (record size..)
- **memory & time efficiency**
 - less overhead in pointer management
- **AB-tree analogy**



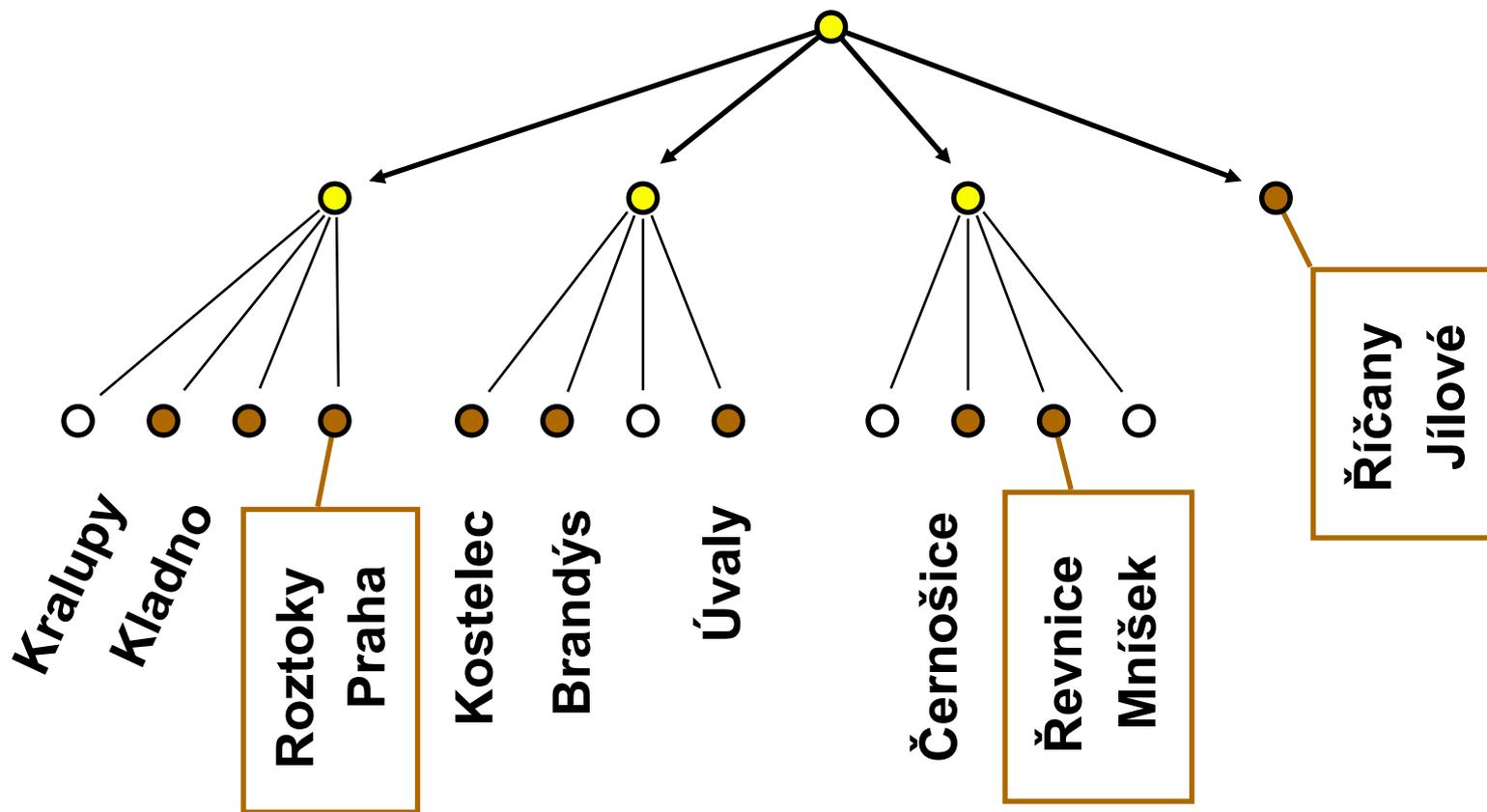
“Bucket PR quadtree”

- ◆ representation of point objects
- ◆ splitting exactly in the middle
- ◆ object information stored in leaf nodes (all levels, $\leq \text{Max}$ object per node)





“Bucket PR quadtree”

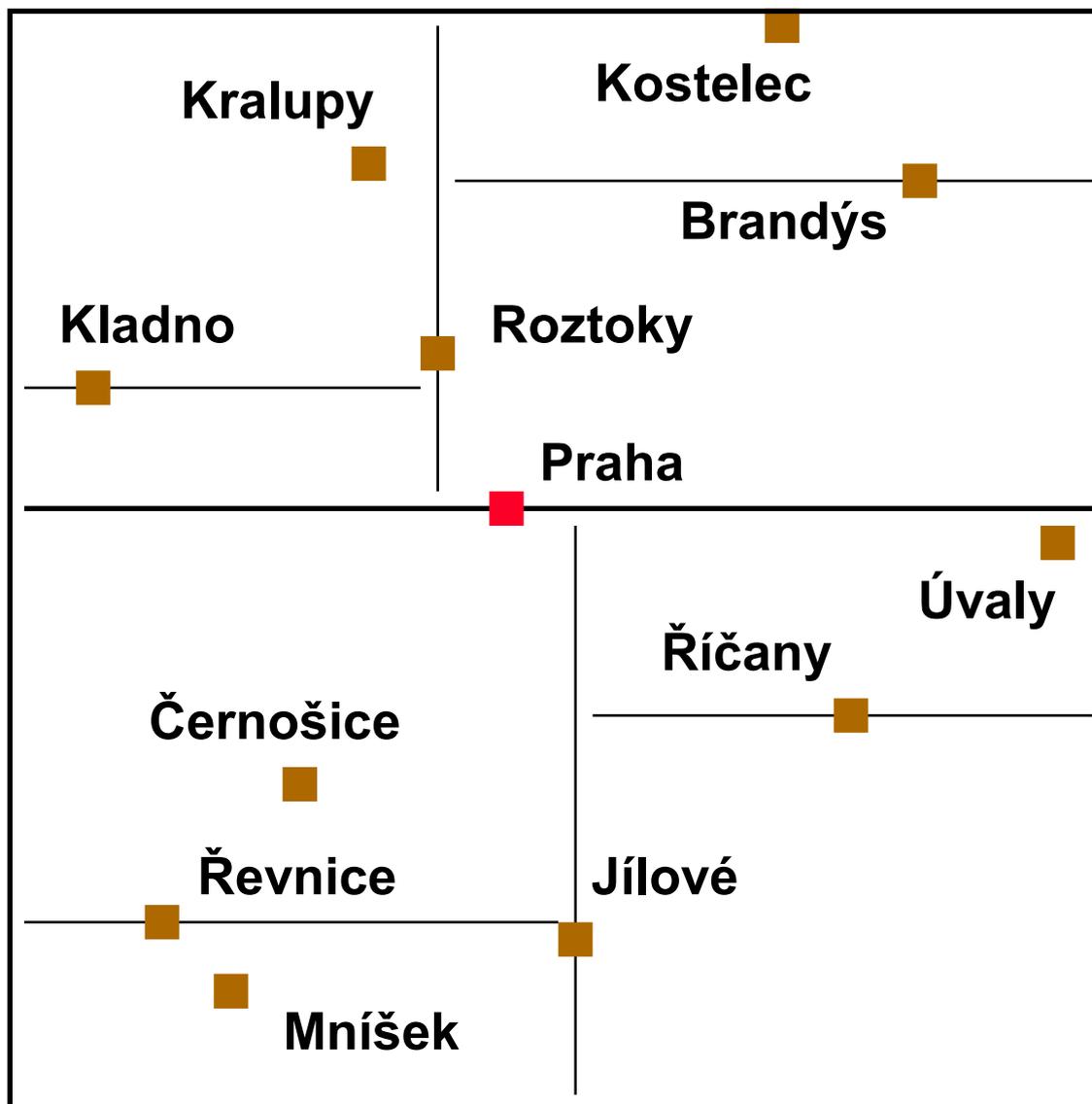


Max = 2

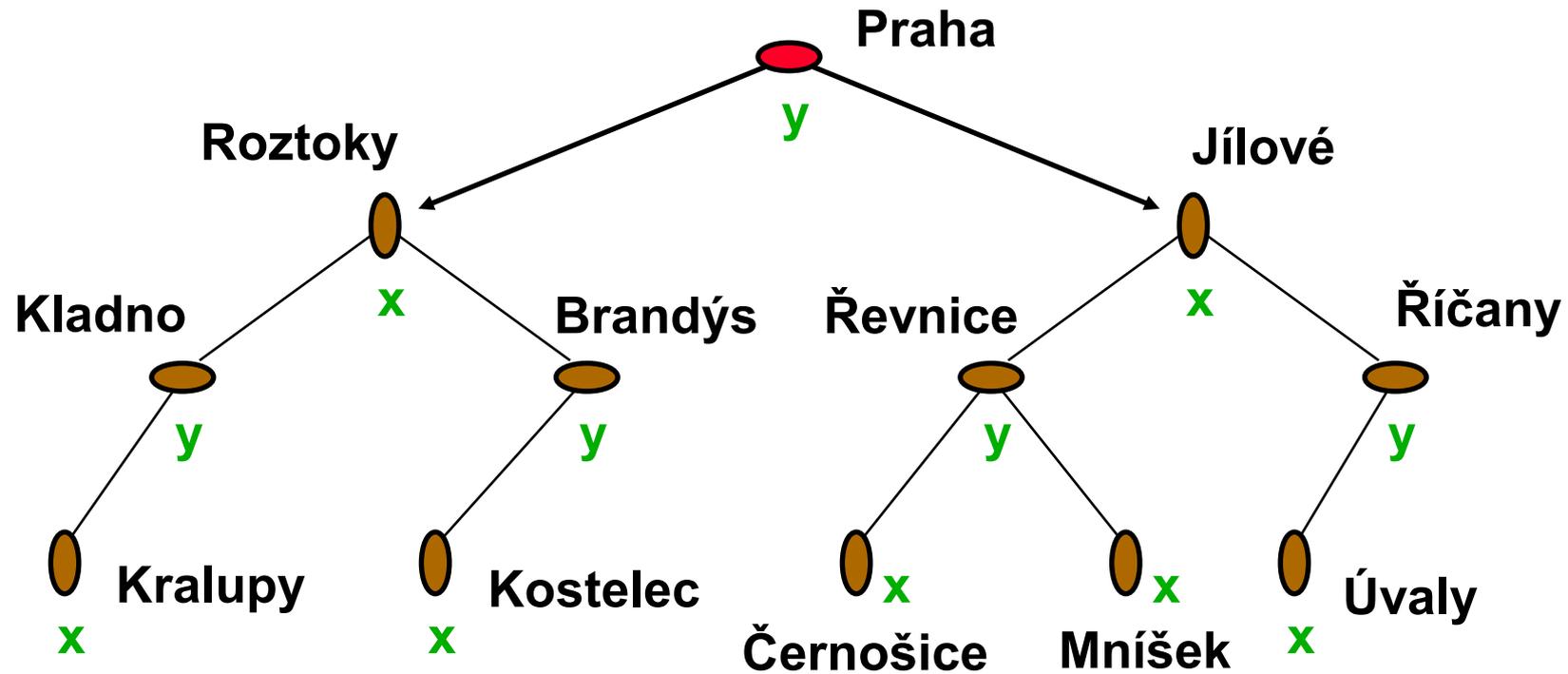


“K-D tree” (Bentley)

- ◆ **representation of point objects**
- ◆ **adaptive splitting - one coordinate at a time (binary tree).**
regular alternation of the coordinates
- ◆ **object information stored in all nodes**



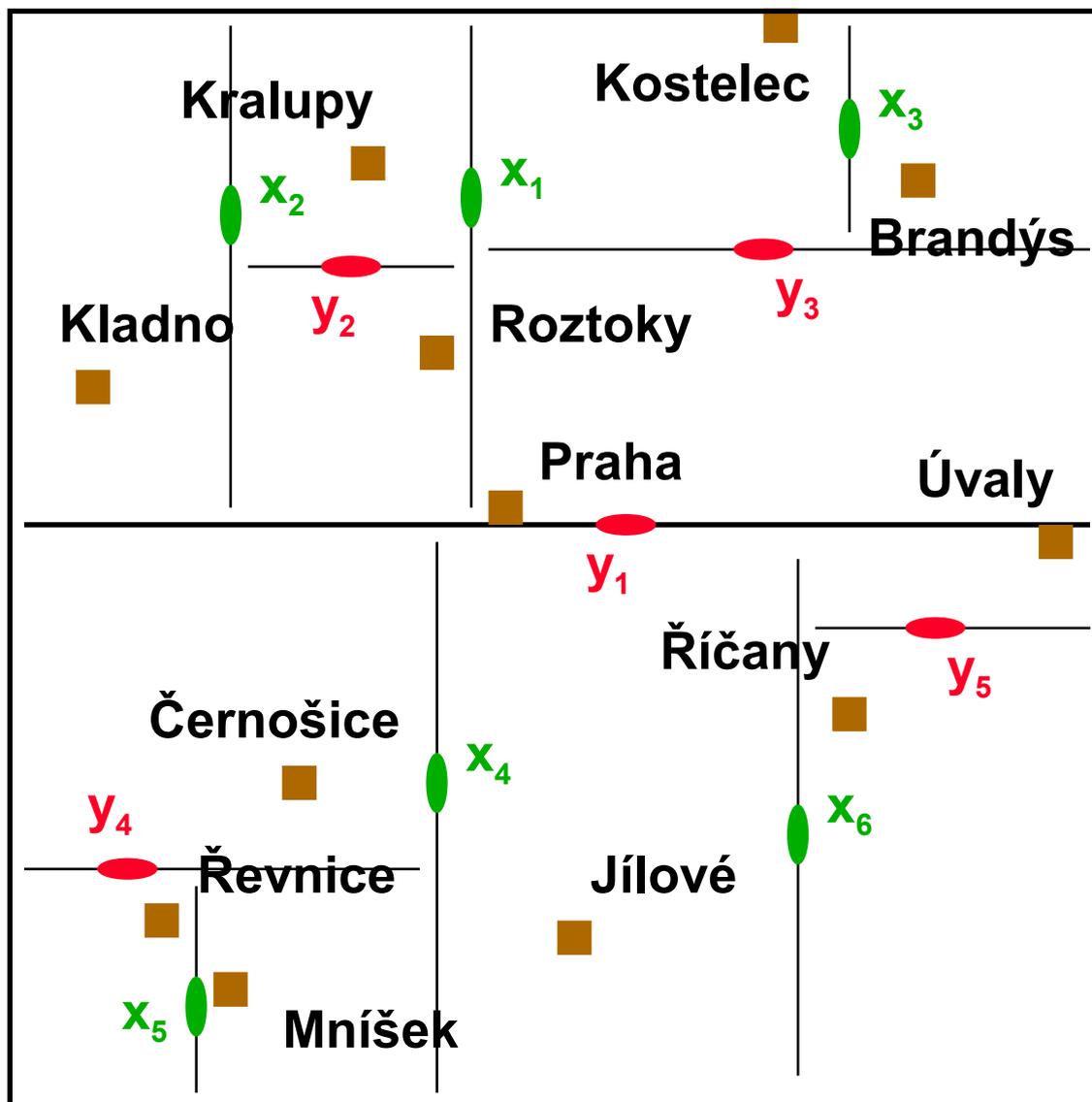
“K-D tree”





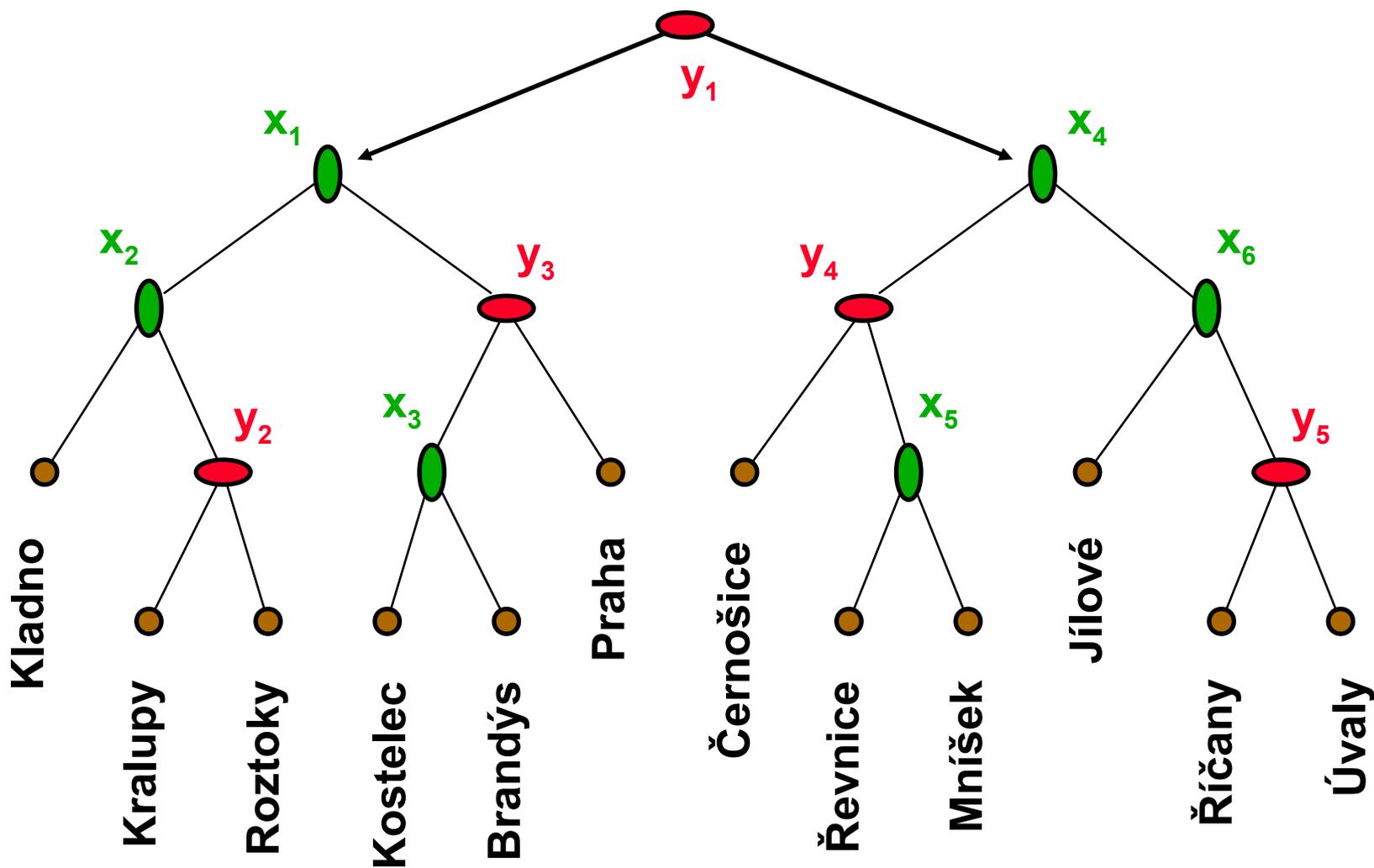
“Adaptive K-D tree” (Friedman)

- ◆ representation of point objects
- ◆ adaptive splitting - one coordinate at a time (binary tree).
Exact object coordinates not used
- ◆ object information stored in leafs only





“Adaptive K-D tree”





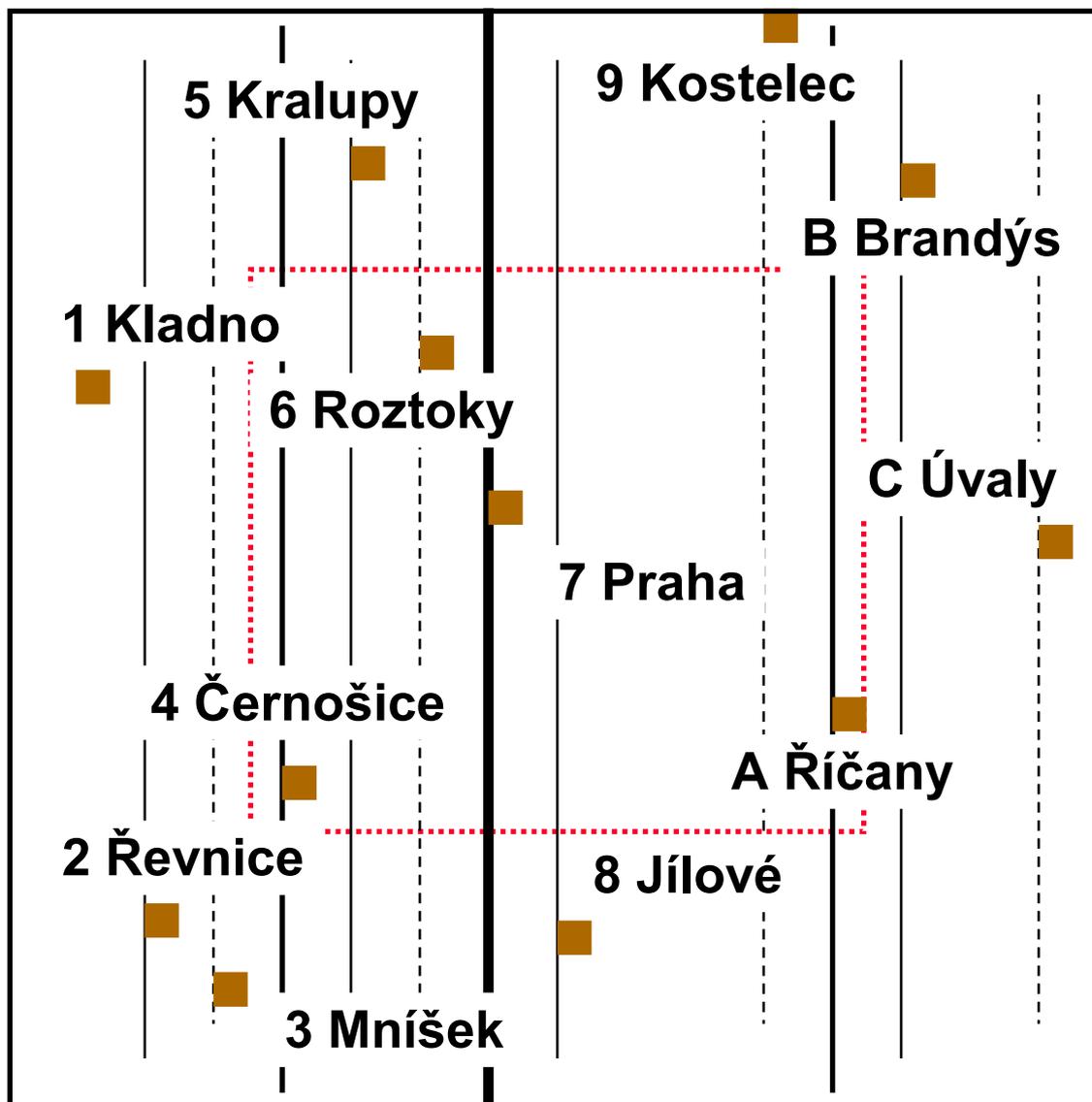
“Range trees”

- ◆ efficient implementation of **range queries**
 - $\langle \mathbf{x}_{\min}, \mathbf{x}_{\max} \rangle \times \langle \mathbf{y}_{\min}, \mathbf{y}_{\max} \rangle$ ve 2D
 - complexity: $O(\log_2 N + F)$ in 1D, $O(\log_2^2 N + F)$ in 2D
- ◆ “**1D range tree**”
 - balanced binary search tree, leaves are connected by a double-linked list
- ◆ “**2D range tree**”
 - balanced binary search tree for the \mathbf{x} coordinate
 - every inner node contains “1D range tree” (\mathbf{y} coord.) for all points in the relevant region



“2D range tree”

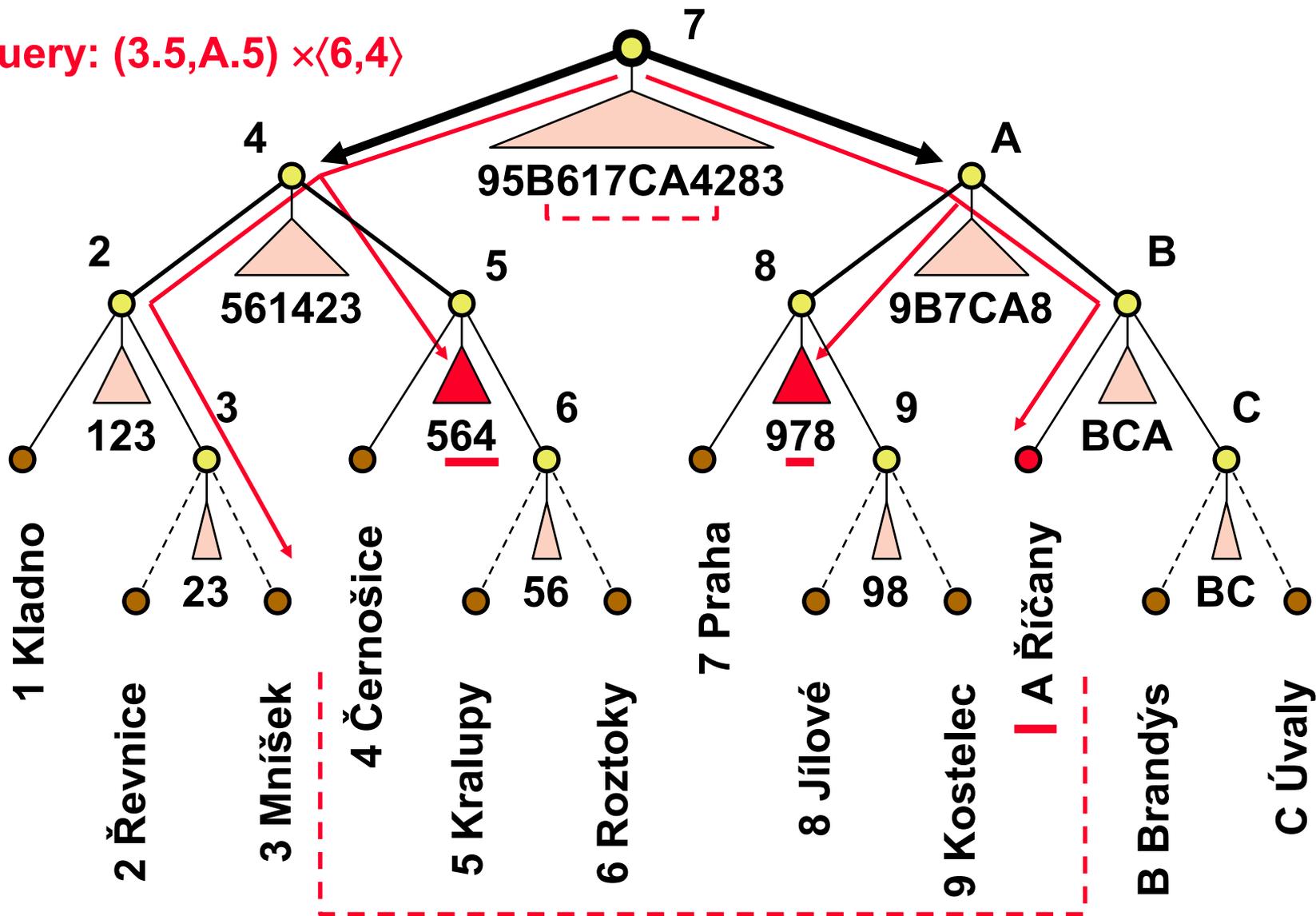
- ◆ **balanced binary tree for x-coordinate**
- ◆ **object information in leaves only**
- ◆ **inner nodes: 1D range trees for y-coordinate**





“2D range tree”

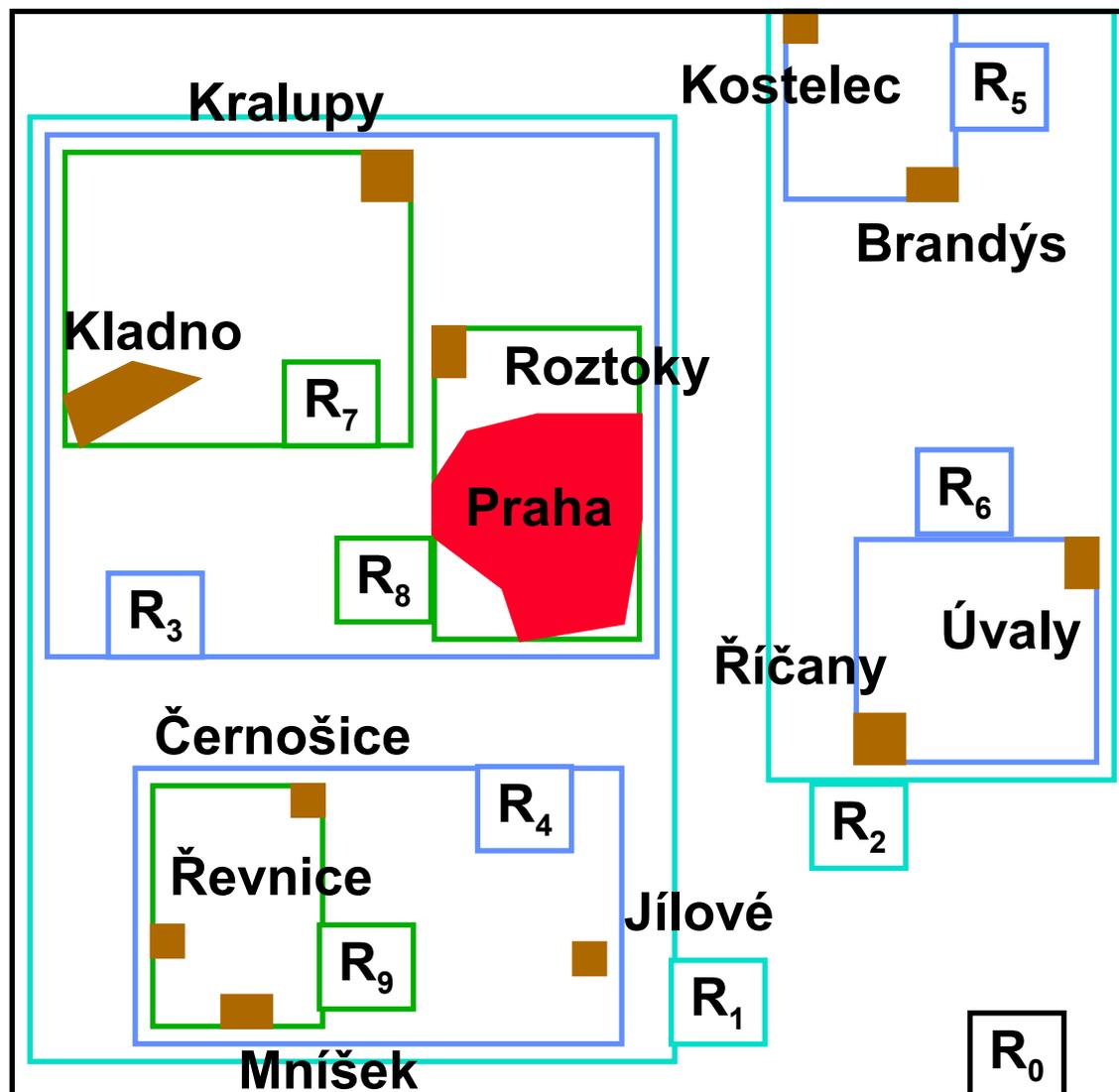
Query: $(3.5, A.5) \times \langle 6, 4 \rangle$





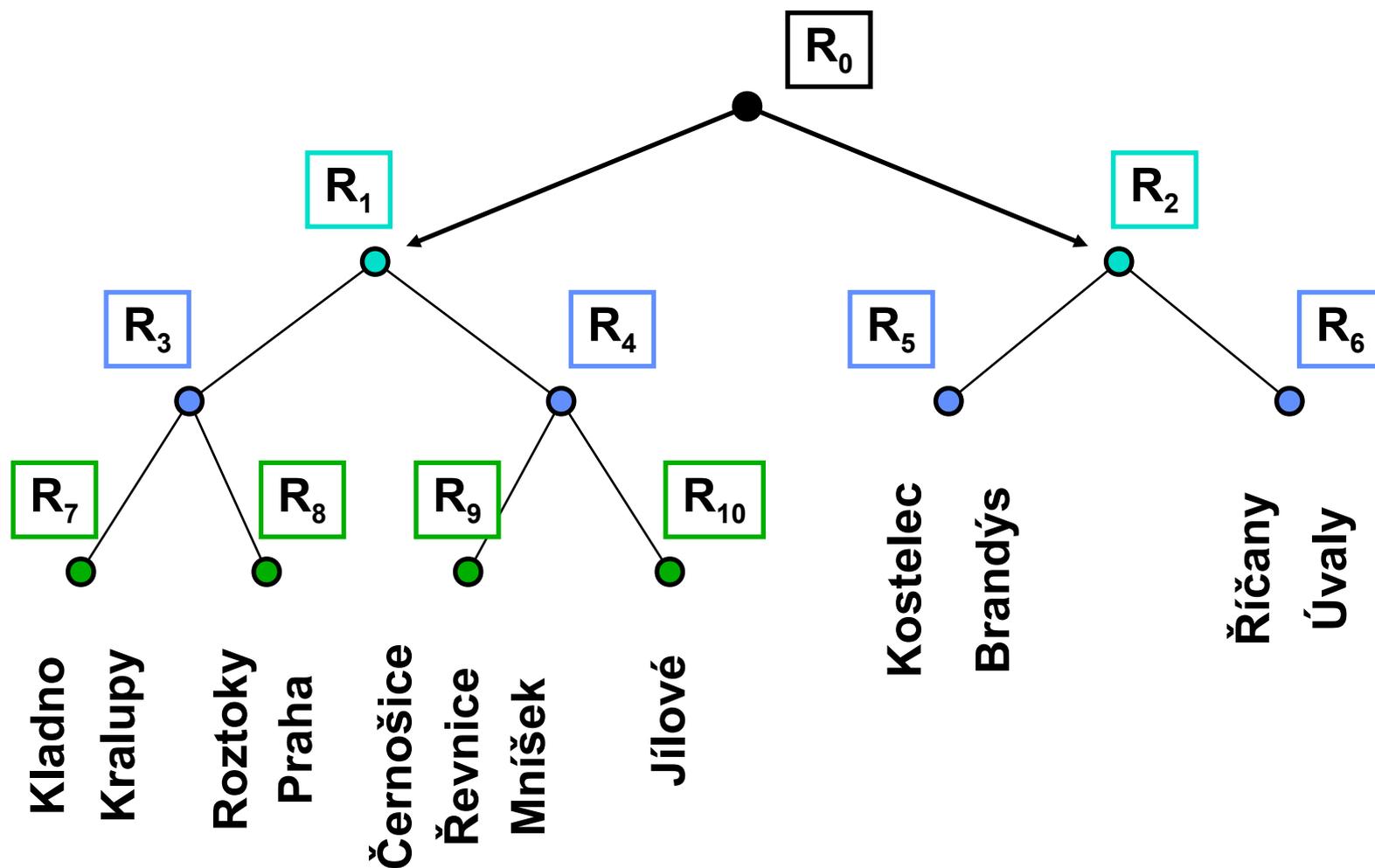
“R-tree” (Guttman)

- ◆ objects can be areal
- ◆ bounding boxes in inner nodes (the whole subtree must fit into the bounding box)
- ◆ object references in leaf nodes





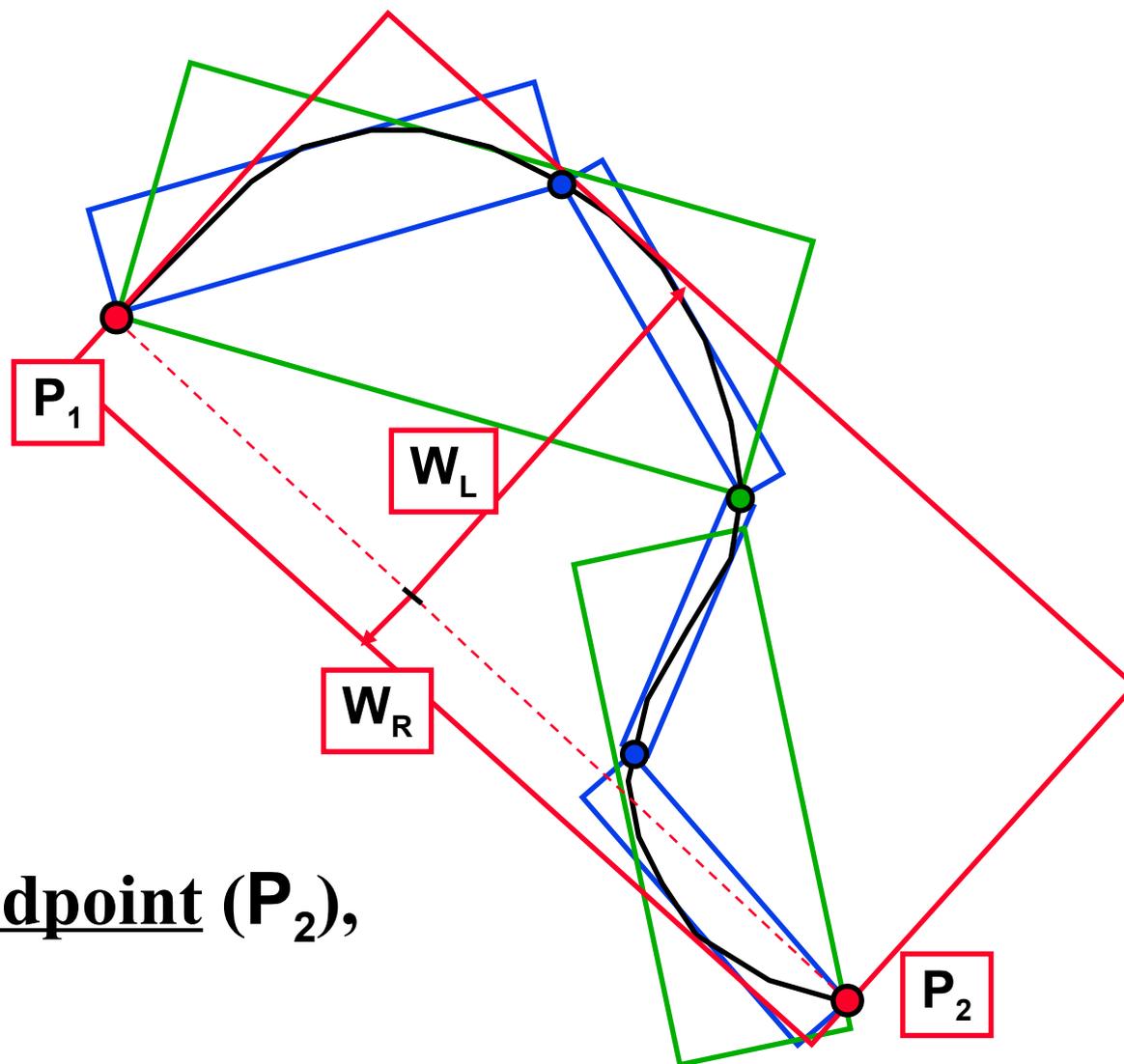
“R-tree”





“Strip tree” (Ballard)

- ◆ planar curve
(polyline)
- ◆ adaptive splitting
induced by the
curvature
- ◆ oriented bounding
rectangle defined by:
starting point (P_1), endpoint (P_2),
widths (W_L , W_R)

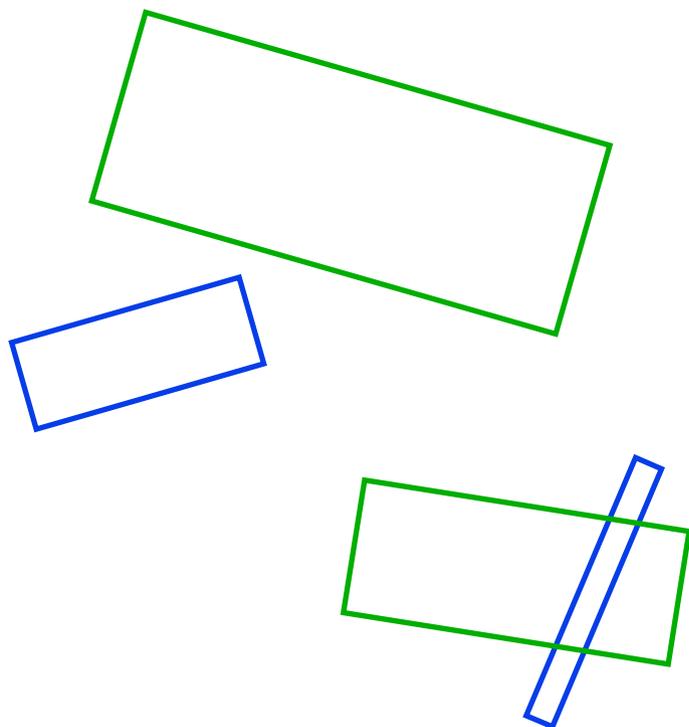




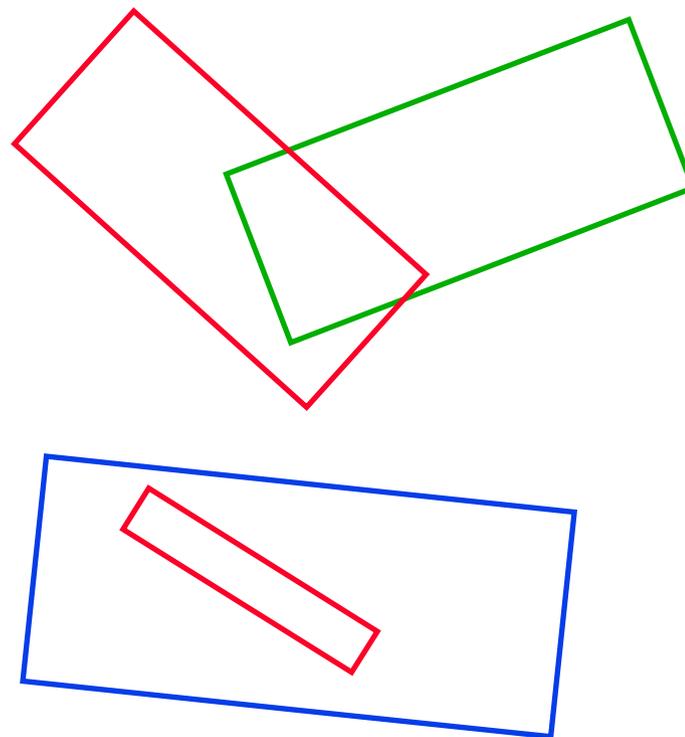
Intersection of two curves

◆ "strip trees" were built for both curves

◆ trivial cases



◆ subdivision is necessary





Bounding system hierarchies

- ➔ **”Sphere tree”** (Palmer, Grimsdale, 1995)
 - simple tests & transform, approximation not so good
- ➔ **”AABB tree”** (Held, Klosowski, Mitchell, 1995)
 - simple test, more complicated transform
- ➔ **”OBB tree”** (Gottschalk, Lin, Manocha, 1996)
 - simple transform, more complicated test, good approximation
- ➔ **”K-dop tree”** (Klosowski, Held, Mitchell, 1998)
 - complicated transform & test, excellent approximation



Directional pass

- ◆ data pass using specific **directional order**
 - visibility (front-to-back or vice versa) in orthographic projection
 - ”plane sweep” pass (“sweeping”)
- ◆ pass from the **center point**
 - visibility (form-factors, ..) in perspective projection
 - kNN search (“k-Nearest Neighbors”)



Presumption

- ◆ **hierarchical** space decomposition
 - inclusion conditions on objects only, not needed for bounding boxes (e.g. "strip tree")
- ◆ effective computation of **minimal distance of each cell / box ... $d(\mathbf{B})$**
 - distance from sweep plane or center point of the pass
 - distance need not be an infimum (but has to be an lower bound)



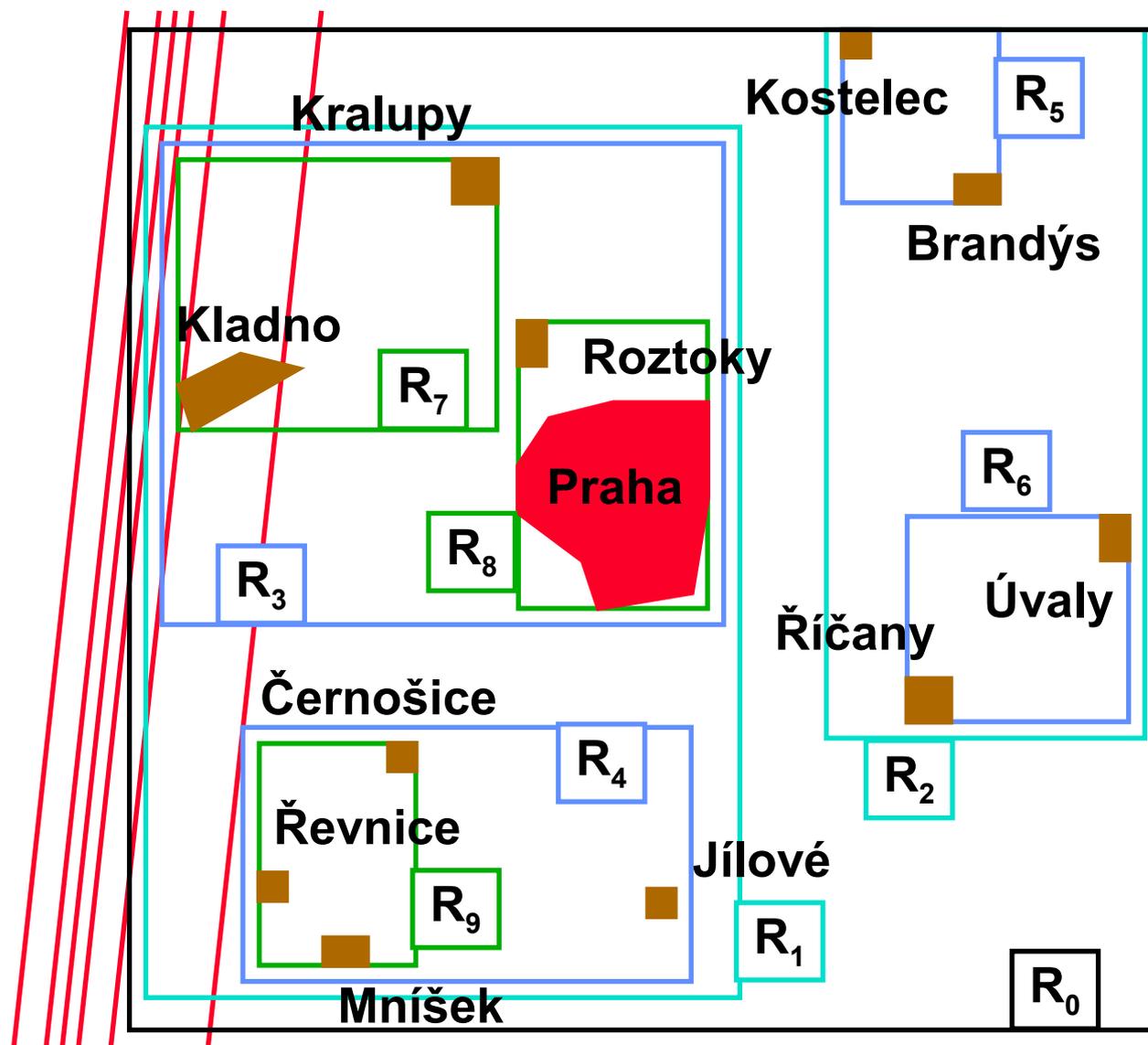
Algorithm

- ◆ auxiliary data structure **H** (heap)
 - efficient operations: **Min**, **DeleteMin**, **Insert**
- ① put the root node into **H**
 - **d(B)** is used for **heap sorting**
- ② if **Min(H)** is **an object**, process it and remove it
- ③ if **Min(H)** is **a hierarchy cell**, remove it and put all its children back into the heap
- ④ repeat steps ② and ③ until the heap **H** is empty or until all required output objects are processed (kNN)



Example I

- ◆ $\{R_0\} \rightarrow \{R_1, R_2\}$
- ◆ $\{R_1, R_2\} \rightarrow \{R_3, R_4, R_2\}$
- ◆ $\{R_3, R_4, R_2\}$
 $\rightarrow \{R_7, R_4, R_8, R_2\}$
- ◆ $\{R_7, R_4, R_8, R_2\}$
 $\rightarrow \{\text{Kladno}, R_4,$
 $\text{Kralupy}, R_8, R_2\}$
- ① Kladno
- ◆ $\{R_4, \text{Kralupy}, R_8, R_2\}$
 $\rightarrow \{R_9, \text{Kralupy}, R_8,$
 $\text{Jílové}, R_2\}$





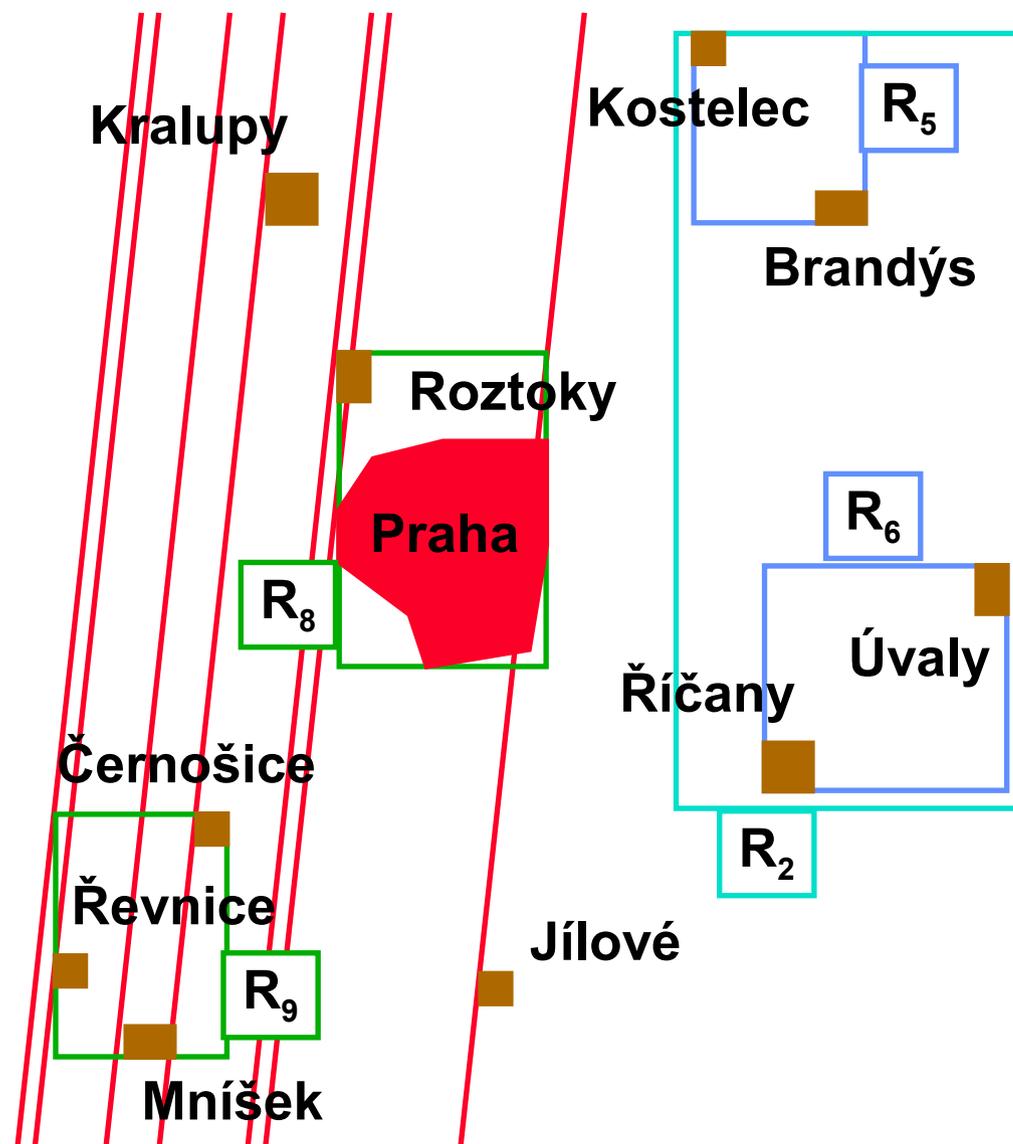
Example II

◆ $\{R_9, \text{Kralupy}, R_8, \text{Jílové}, R_2\}$
→ $\{\text{Řevnice}, \text{Mníšek},$
 $\text{Kralupy}, \text{Černošice},$
 $R_8, \text{Jílové}, R_2\}$

②-⑤ $\text{Řevnice}, \text{Mníšek},$
 $\text{Kralupy}, \text{Černošice}$

◆ $\{R_8, \text{Jílové}, R_2\}$
→ $\{\text{Roztoky}, \text{Praha},$
 $\text{Jílové}, R_2\}$

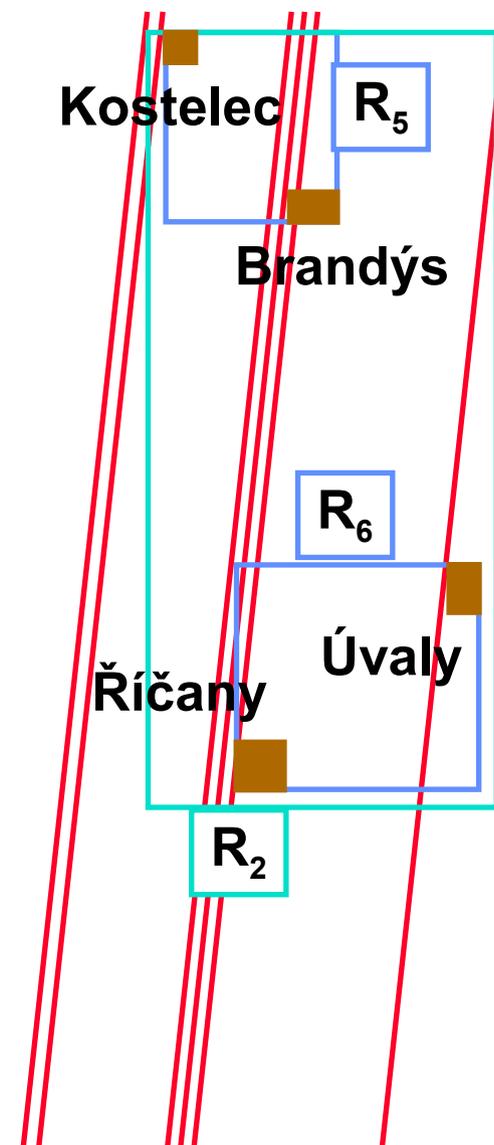
⑥-⑧ $\text{Roztoky}, \text{Praha},$
 Jílové





Example III

- ◆ $\{R_2\} \rightarrow \{R_5, R_6\}$
- ◆ $\{R_5, R_6\} \rightarrow \{\text{Kostelec}, R_6, \text{Brandýs}\}$
- ⑨ Kostelec
- ◆ $\{R_6, \text{Brandýs}\} \rightarrow \{\text{Brandýs}, \text{Říčany}, \text{Úvaly}\}$
- ⑩-①② Brandýs, Říčany, Úvaly





The End

More information:

- **H. Samet: *The Design and Analysis of Spatial Data Structures*, Addison-Wesley, 1990**
- **H. Samet: *Foundations of Multidimensional and Metric Data Structures*, Morgan Kaufmann, 2006**
- **F. Preparata, M. Shamos: *Computational Geometry, An Introduction*, Springer-Verlag, 1985**