

# PRINCIPLES OF DATA ORGANISATION

Multidimensional Indexing : Introduction



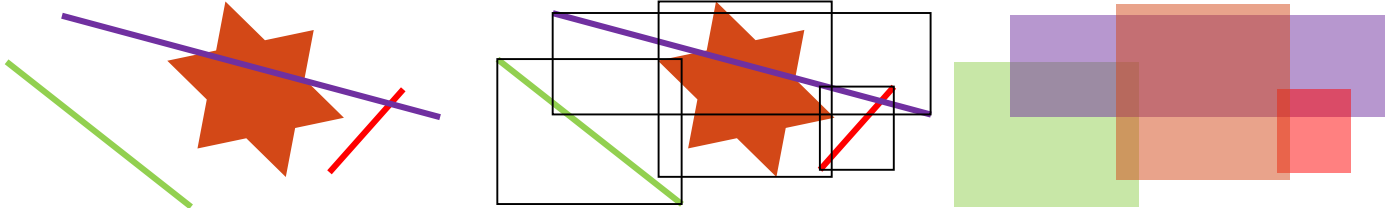
# MOTIVATION

- ❧ How to search effectively in more than one dimension?
  - ❧ We know 1D
- ❧ How to represent spatial object in the database?
- ❧ Single Dimension-Based Indexing has an issue with locality
  
- ❧ **Multidimensional indexes focus on** storing spatial objects in such a way that objects close to each other in the space are also close in the structure and on the disk, i.e., maintain **locality**
  - ❧ We will now focus on basics not the secondary memory as before



# OBJECTS APPROXIMATIONS

- General spatial objects are more complex than simple points
- To easily represent a possibly complex spatial object, we use approximation expressed by (Minimum) Bounding Rectangle/cube/box/object (**MBR**)
  - 2D: rectangle
- Comparison of objects is reduced to the comparison of their MBRs
  - Pros and cons



# GRID-BASED INDEXING

- ⌘  $N$ -dimensional grid covers the space and is not dependent on the data distribution in any way
  - ⌘ The grid is formed in advance
  - ⌘ We anchor the data
- ⌘ Every point object can be addressed by the grid address
- ⌘ Objects distribution in the grid does not have to be uniform → retrieval times for different grid cells may differ substantially for different parts of the space



# QUAD-TREE

🔗 Finkel, Bentley; 1974

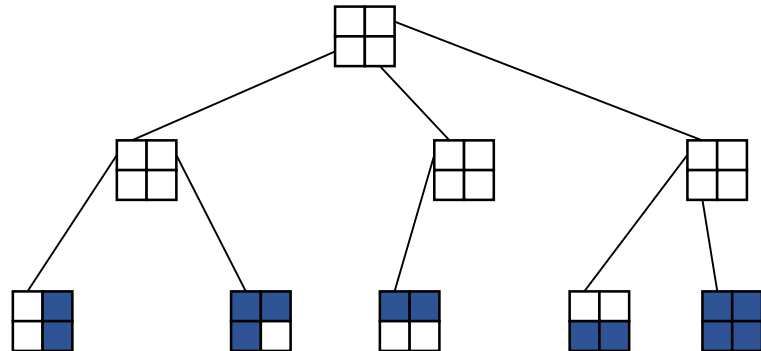
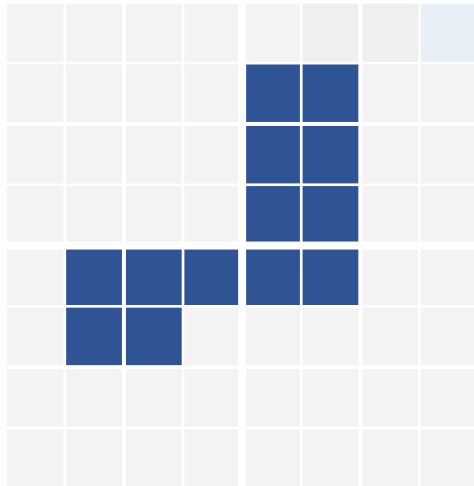
🔗 Tree structure representing recursive splitting of a space into quadrants

🔗 Quad = quadrant (4 regions)

🔗 Each node has from zero to four children

🔗 Typically the regions are squares

🔗 Any arbitrary shape is possible



# K-D-TREE

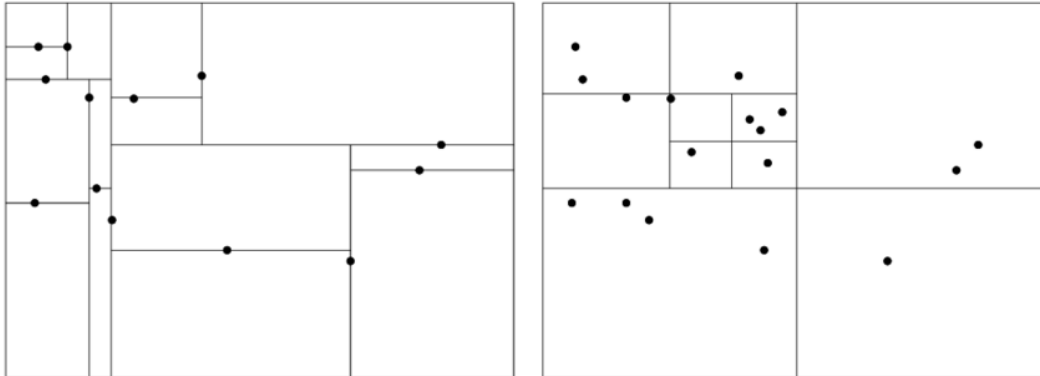
⌘ Bentley; 1975

⌘ Problem: quad tree can be unbalanced

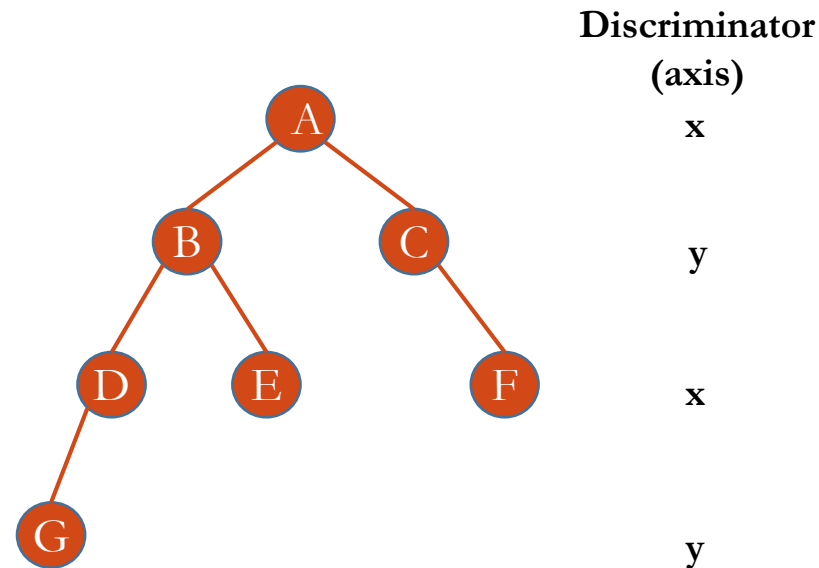
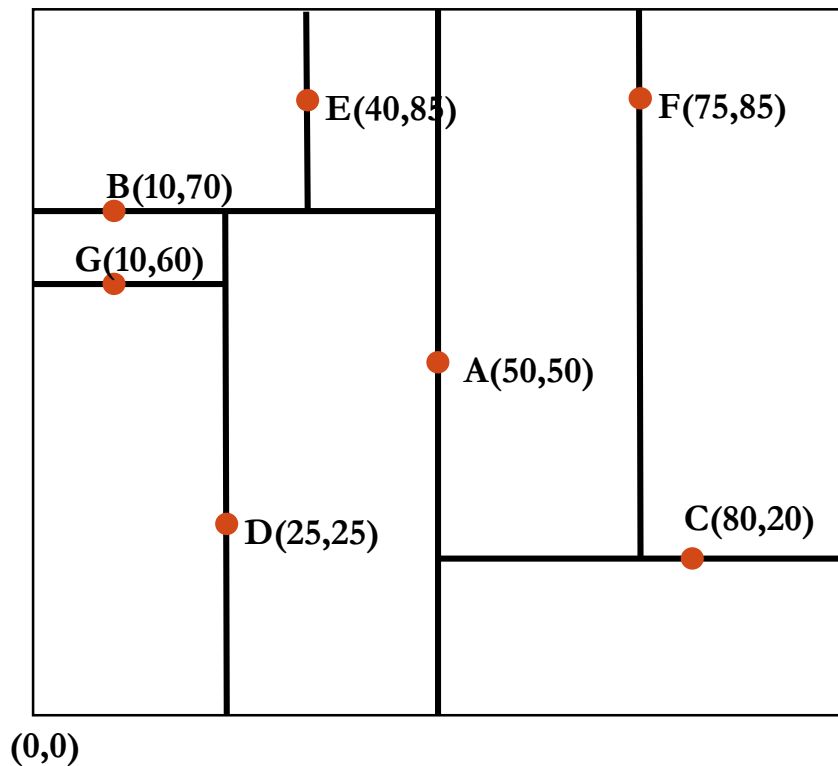
⌘ Objects in  $k$ -dimensional space

⌘ Binary search tree where inner nodes consist of a point, an axis identification (hyperplane in  $d$ -dimensions), and two pointers

⌘ Inner nodes correspond to hyperplanes splitting space into two parts where the location of the hyper plane is defined by the point



# K-D-TREE : EXAMPLE



# K-D-B-TREE

🔗 Robinson; 1981

🔗 Problem: **k-d-tree is designed for main memory**

✂ What if it does not fit there?

🔗 Combination of K-D-Tree and B-Tree

🔗 Each tree node is stored as a page, but unlike B-trees 50% utilisation can not be guaranteed

🔗 Each inner node contains multiple split axes to fill the node's capacity

🔗 Leaf nodes contain indexed records (points)

🔗 Like in redundant B-trees

🔗 Splitting and merging happens analogously to B-trees





# K-D-B-TREE : EXAMPLE

