



# Spatial Data Structures and Hierarchies

CS535

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# Spatial Data Structures

- Store geometric information
- Organize geometric information
- Permit fast access to/of geometric information
- Applications
  - Heightfields
  - Collision detection (core to \*many\* uses)
  - Simulations (e.g., surgery, games)
  - Rendering (e.g., need to render fast!)

# Hierarchical Modeling

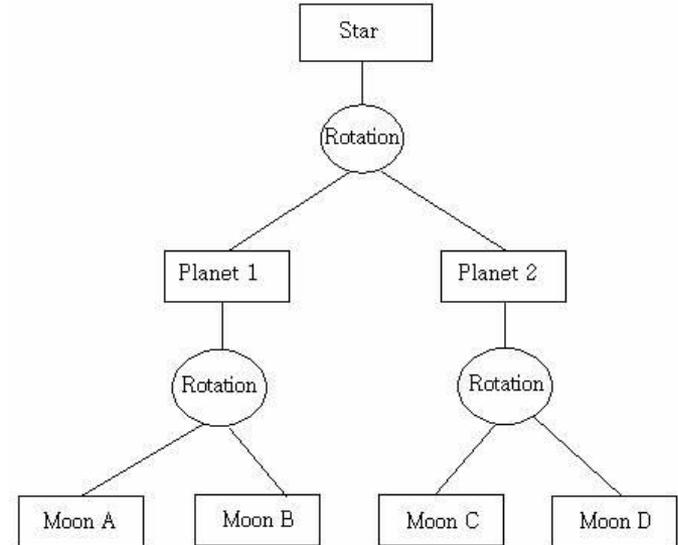
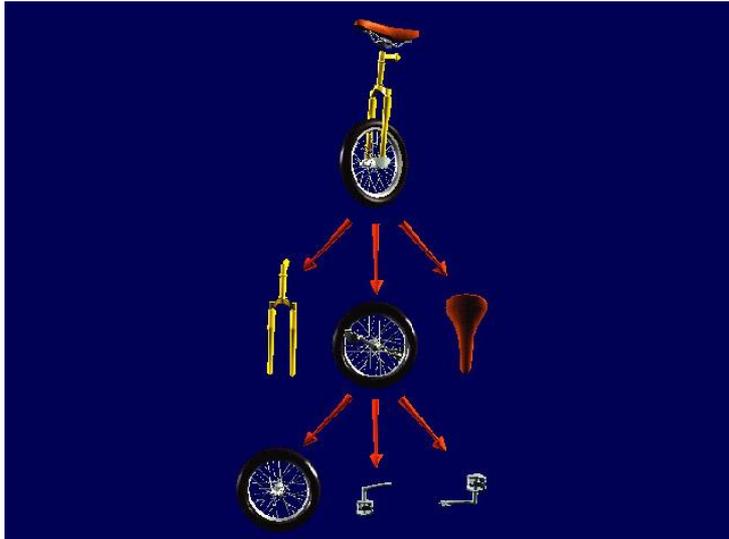


- Concept is old but fundamental
  - “Hierarchical geometric models for visible surface algorithms”, James Clark - 1976



# Hierarchical Modeling

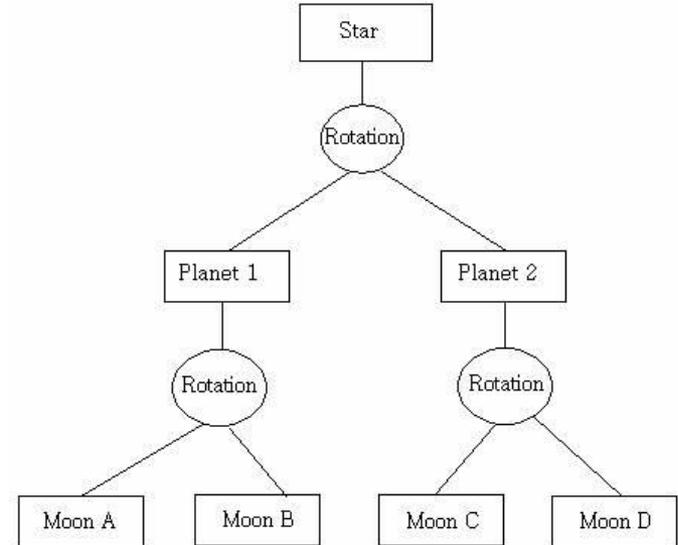
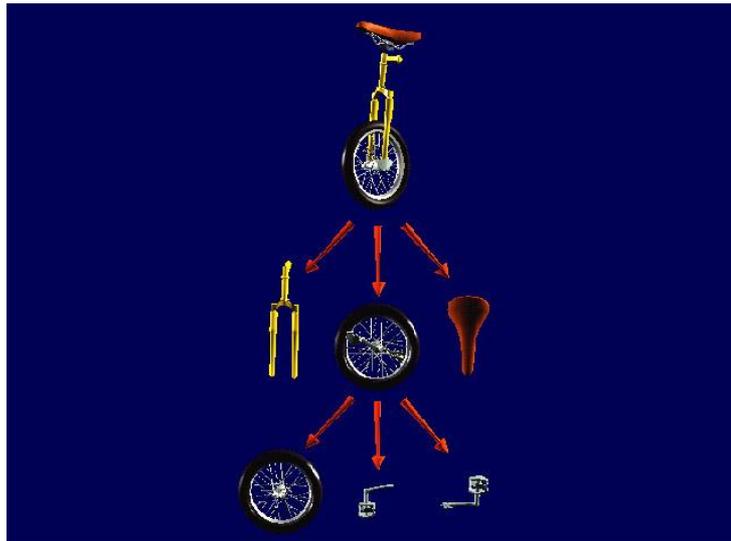
- Trees and Scene Graphs





# Hierarchical Modeling

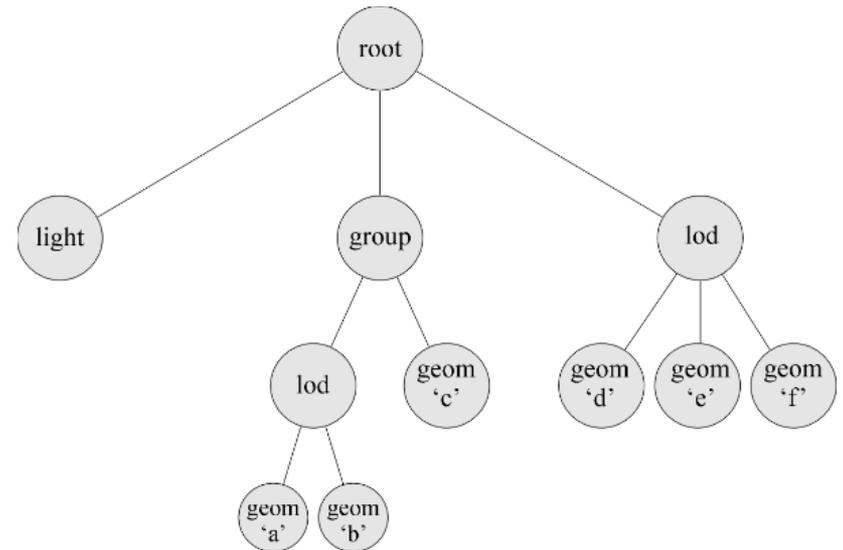
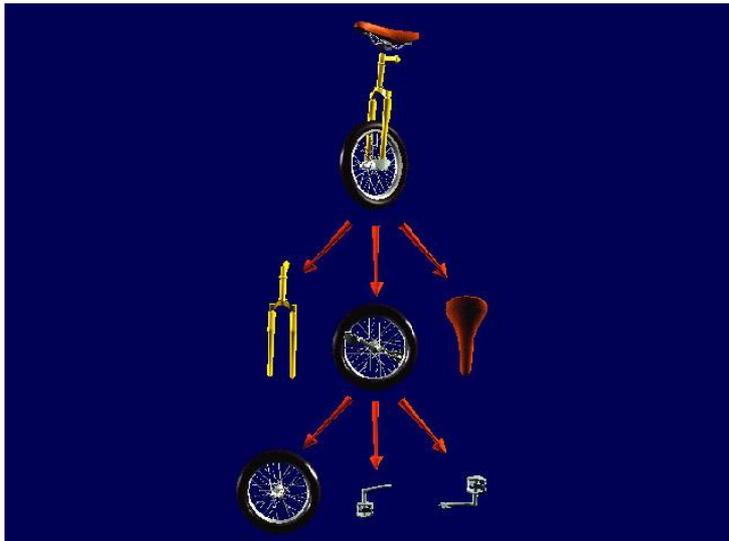
- Trees and Scene Graphs





# Hierarchical Modeling

- Trees and Scene Graphs





# Bounding Volumes

- Problem:
  - Suppose you need to intersect rays with a scene...
  - Suppose you have a scene divided into objects...
- Solution: bottom-up
  - Wrap complex objects into simple ones
    - Boxes, spheres, other shapes...
  - Organize into a tree

# Bounding Volume Hierarchy (BVH)



- How to building an axis aligned bounding box (AABB) BVH?
- How to intersect?
- Complexity? Problem cases?



# AABB BVH

- Example construction
  - Given  $M$  2D points, use k-means clustering to determine clusters
  - Then group nearby clusters (e.g., use Voronoi diagram or Delaunay triangulation)
  - And iteratively form a tree from the bottom-up
  - In each node, approximate the contained points using an axis-aligned bounding box
    - e.g.,  $\text{box} = [\text{min of all contained pts}, \text{max of all contained pts}]$

# Bounding Volume Hierarchy (BVH)



- How to building an oriented bounding box (OBB) BVH?
- How to intersect?
- Complexity? Problem cases? Advantages over axis-aligned?



# OBB BVH

- Example construction
  - Similar to AABB BVH but “fit” an oriented box to the points within each cluster/node of the tree
  - Methods:
    - Sample possible rotations and sizes in order to pick the best box
    - Compute distance of points to a line and optimize the line equation parameters until finding the line that best approximates all points
    - Then compute a box width – consider the benefit/cost of the box size
      - e.g., totally containing all points might make the box very large; could also choose to mostly contain the points – however, what does this mean with regards to operations using the BVH?



# Space Subdivision

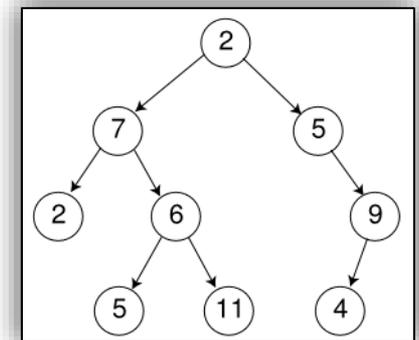
- Binary tree / Quadtree / Octree
- k-D tree
- Binary Space Partitioning (BSP) Tree



# Binary Tree

- A directed edge refers to the link from the parent to the child (the arrows in the picture of the tree).
- The root node of a tree is the node with no parents; there is at most one root node in a rooted tree.
- A leaf is a node that has no children.
- The depth of a node is the length of the path from the root to the node. The root node is at depth zero.
- The height of a tree is the depth of its furthest leaf. A tree with only a root node has a height of zero.
- Siblings are nodes that share the same parent node

Size = 9  
Height = 3  
Root node = 2





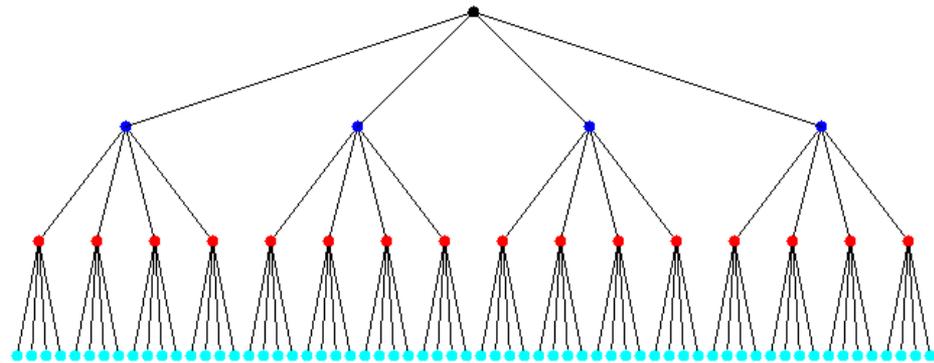
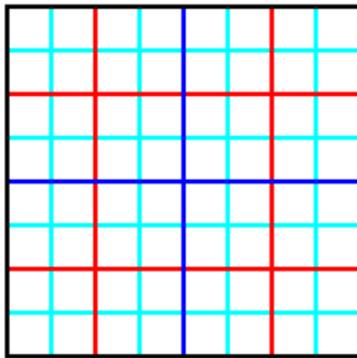
# Binary Tree

- Operations
  - Search
  - Insert
  - Delete



# Quadtree

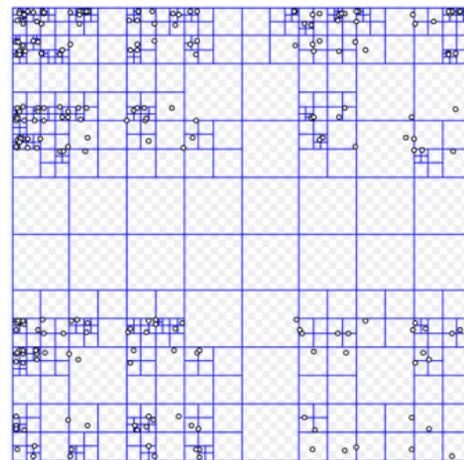
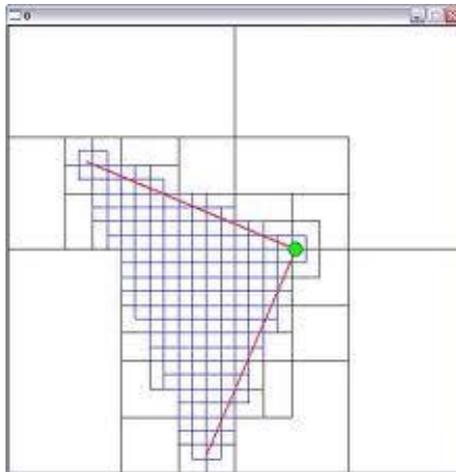
- Similar to binary-tree, but have 4 children per node
- Each node corresponds to one of four rectangular regions of the current quad





# Quadtree

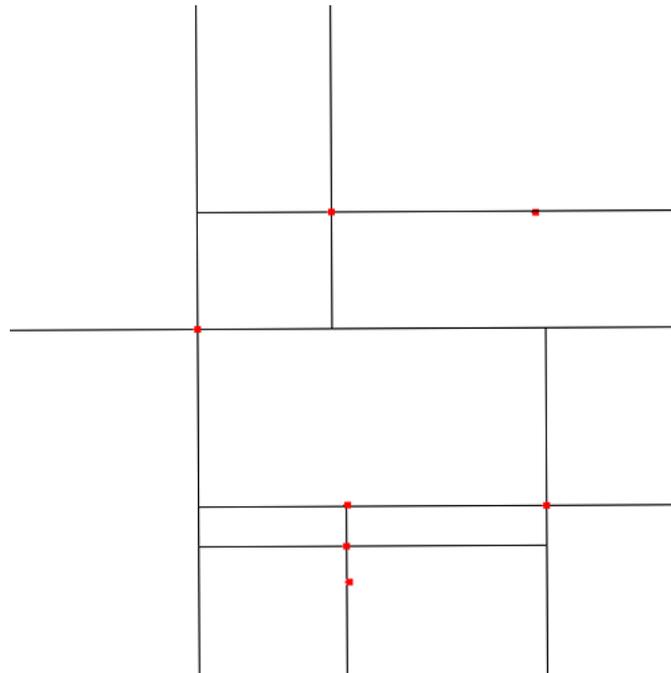
- Similar to binary-tree, but have 4 children per node
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# Quadtree

- Point quadtree
  - Partitions depend on the data
  - The quad is divided using the previous point within it





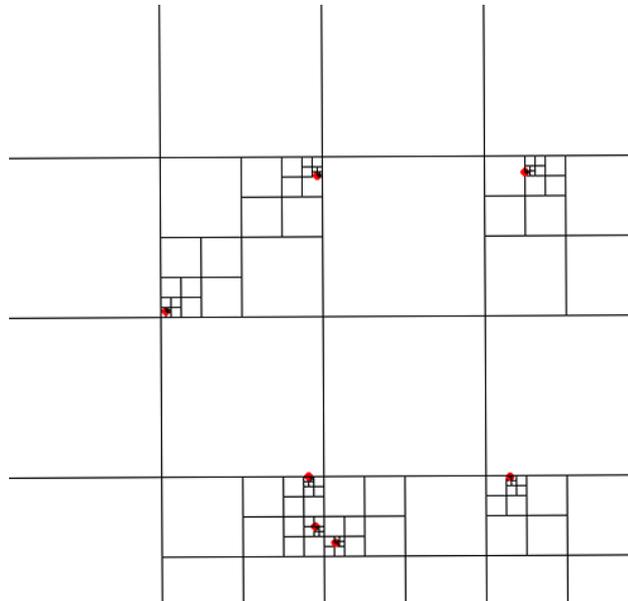
# Quadtree

- Point quadtree
  - Partitions depend on the data
  - The quad is divided using the previous point within it
- Advantage
  - Data dependent subdivision reduces (unnecessary) number of quads
- Disadvantage
  - Quads do not tightly approximate region surrounding the point



# Quadtree

- Matrix (MX) quadtree
  - Location of partition lines independent of the data
  - The occupied nodes are all subdivided until a tight fitting box





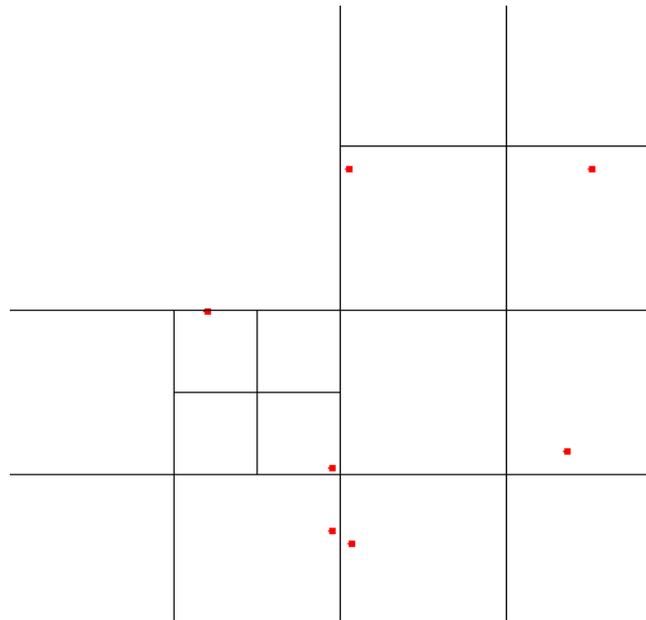
# Quadtree

- MX quadtree
  - Location of partition lines independent of the data
  - The occupied nodes are all subdivided until a tight fitting box
- Advantage
  - Quads leaf nodes always tightly approximate region surrounding the point
- Disadvantage
  - Potentially lots of levels from root to a point



# Quadtree

- Point Region (PR) quadtree
  - Location of partition lines independent of the data
  - The nodes are all subdivided until  $p$  or less points per node (e.g.,  $p=1$ )





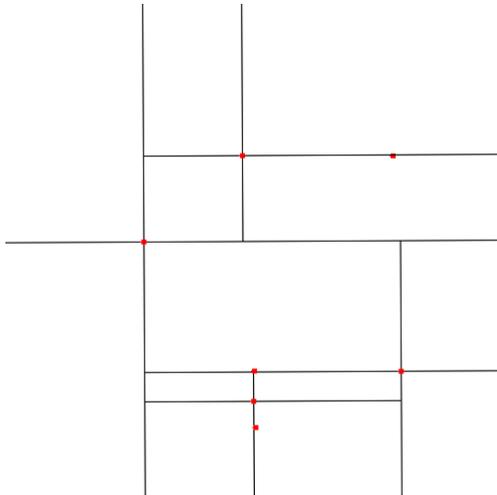
# Quadtree

- PR quadtree
  - Location of partition lines independent of the data
  - The nodes are all subdivided until  $p$  or less points per node (e.g.,  $p=1$ )
- Advantage
  - Partition lines are known and paths from root to point is only as long as needs to be
- Disadvantage
  - Quads do not tightly approximate region surrounding the point

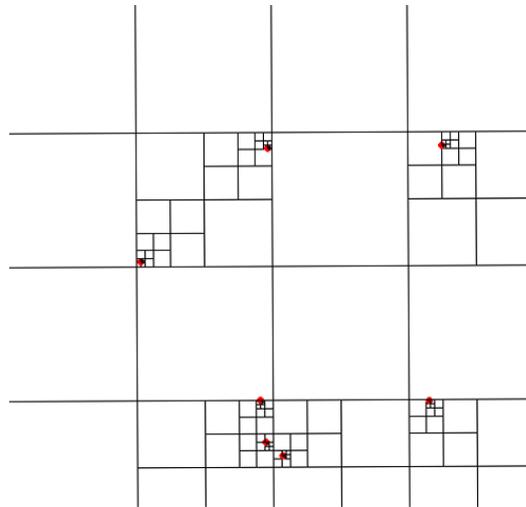


# Quadtree

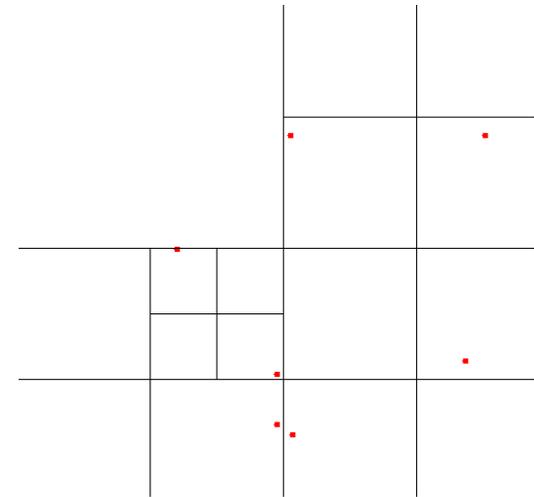
- Comparison



Point QT



MX QT



PR QT



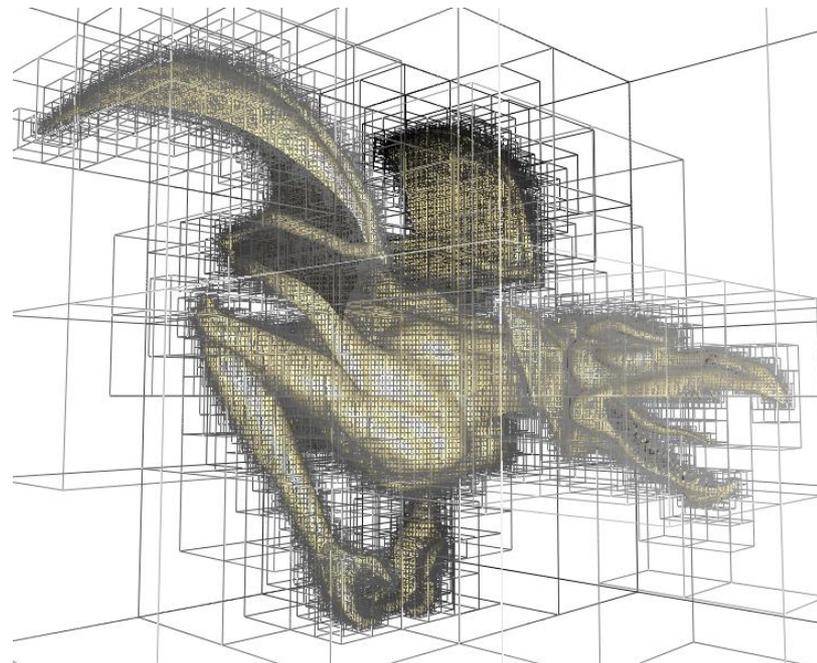
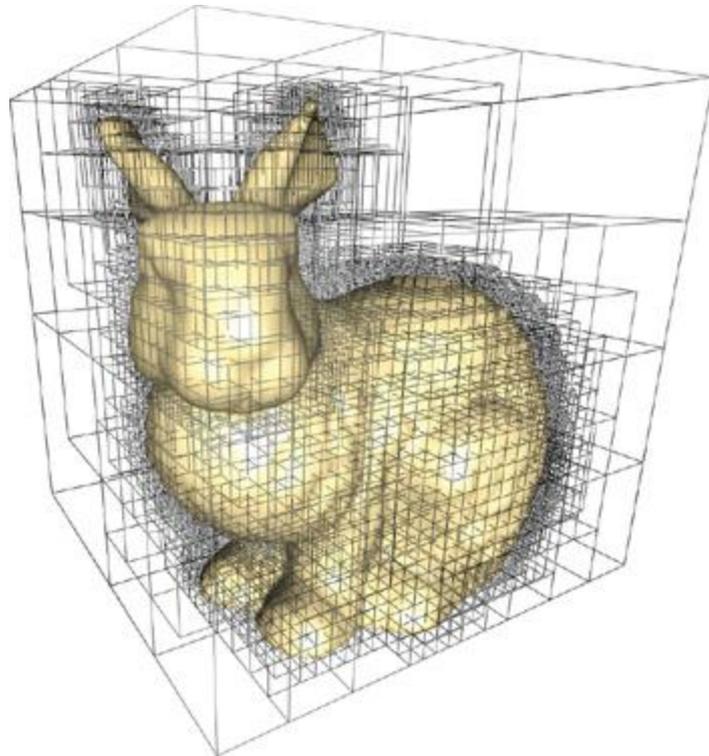
# Demo

- <http://donar.umiacs.umd.edu/quadtree/>



# Octree

- Analogous to Quadtree but extended to 3D
- Each node is divided into eight subboxes





# Octree

- Analogous to Quadtree but extended to 3D
- Each node is divided into eight subboxes
- Similar, there are
  - Point octrees
  - MX octrees
  - PR octrees

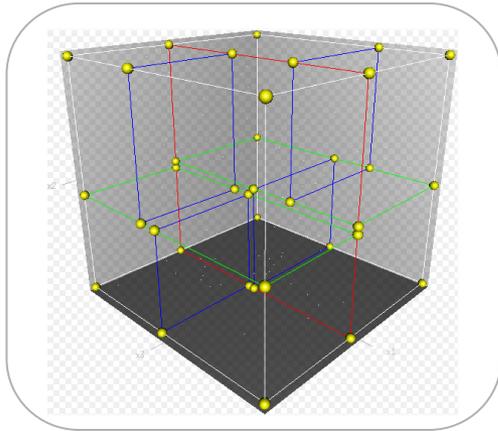


# K-D tree

- Partition each dimension in a cyclical fashion
  - Thus, can be applied to 2D, 3D, or higher dimensions
- Each node stores a next partitioned “half-space” of data points (or of the data space)

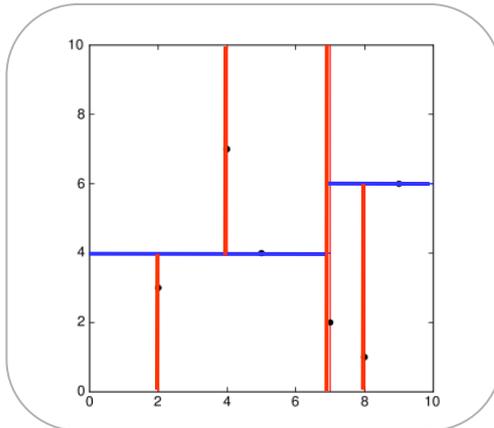


# k-D tree

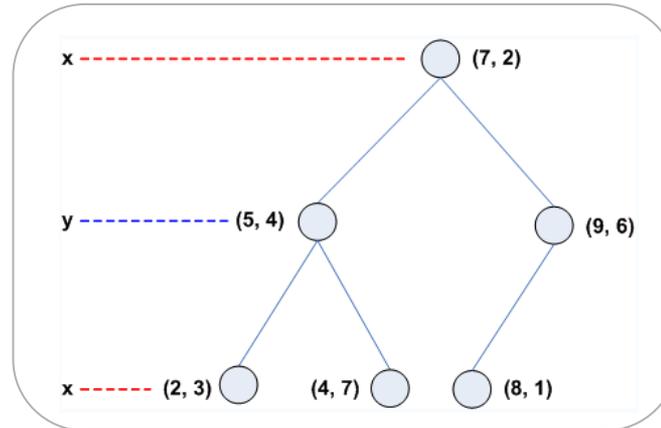


A 3-dimensional kd-tree

- The first split (**red**) cuts the root cell (**white**) into two
- Each of which is then split (**green**) into two subcells
- Each of those four is split (**blue**) into two subcells
- The final eight called leaf cells
- The **yellow** spheres represent the tree vertices



The resulting kd-tree decomposition



The resulting kd-tree



# Demo

- <http://donar.umiacs.umd.edu/quadtree/>

# Binary Space Partitioning (BSP)

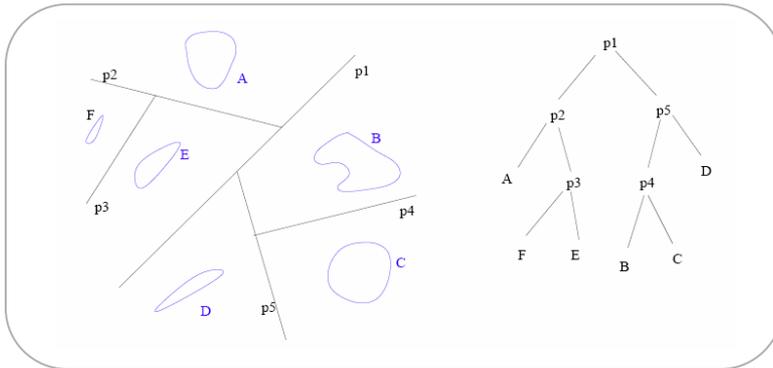


- Similar to k-D tree but splitting lines/planes are not necessarily axis-aligned
- Can adapt better to data
- Was algorithm used for visibility sorting...

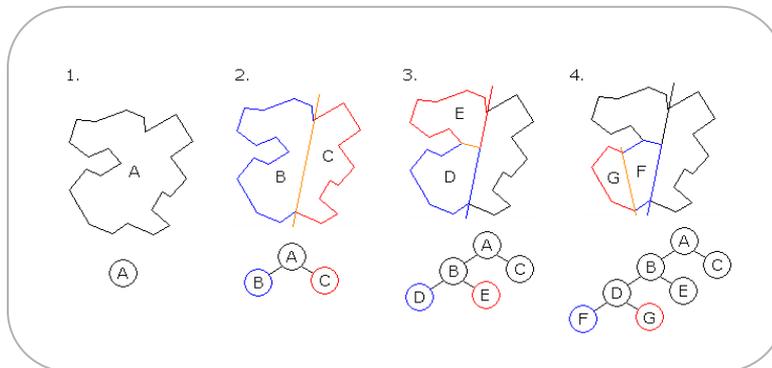
# Binary Space Partitioning (BSP)



- Suitable for any number of dimensions



Separating planes are shown in black and objects in blue)



BSP trees



# Demo

- More stuff at
  - <http://donar.umiacs.umd.edu/quadtree>
  - R-tree?
- See
  - H. Samet, *Foundations of Multidimensional and Metric Data Structures*, Morgan-Kaufmann, San Francisco, 2006