

CSCI 4560/6560 Computational Geometry

<https://www.cs.rpi.edu/~cutler/classes/computationalgeometry/S22/>

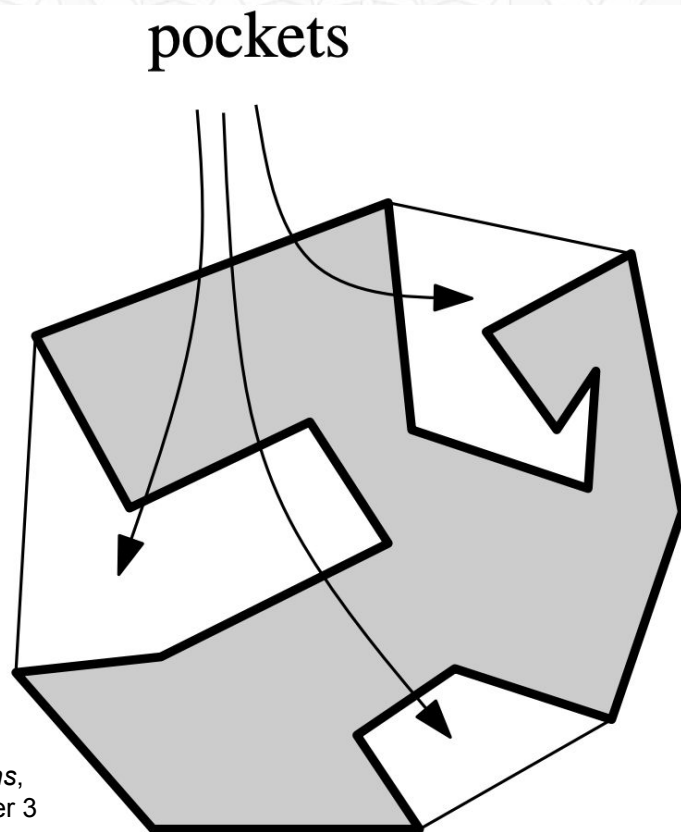
Lecture 8: Orthogonal Range Searching

Outline for Today

- Homework 3 Posted
- Last Time: Bounding Spheres & Randomized Incremental Construction
- Motivating Application: Database Queries
- Motivating Application: Graphics & Photon Mapping
- Data Structure Choices
 - Cost to construct, memory to construct, cost to query
- Review: (1D) Binary Search Trees
- 2D kD Trees & Higher dimension kD Trees
- 2D Range Trees & Higher Dimension Range Trees

Homework 3 - CGAL Programming Task

- Compute triangulation of input polygon & triangulation of “pockets” outside input polygon but inside convex hull
- Compute areas
- Compute changes to boundary edges
- Leverage CGAL libraries for convex hull & triangulation



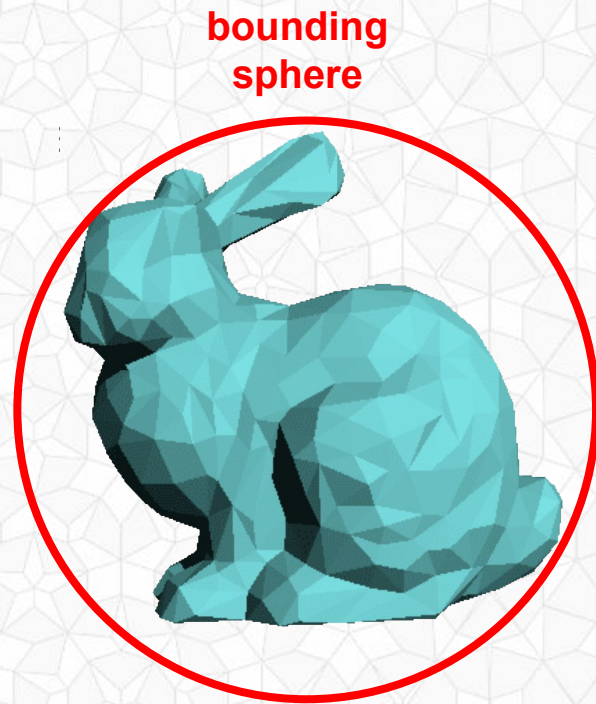
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Problem: Minimal Bounding ~~Sphere~~ Circle

- Input: n vertices in ~~3D~~ 2D
- Assume (for convenience):
 - “General Position”
 - No 3 points are collinear
 - No 4 points lie on the same circle
- Output: 3 of those vertices uniquely define a circle such that all other points lie inside of that circle

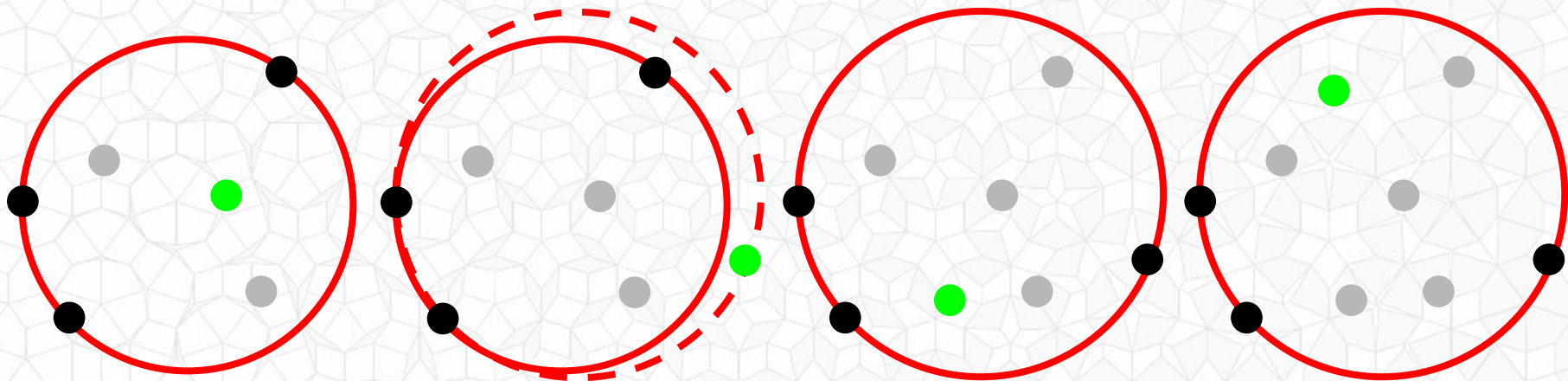
Note: In 3D, we would output 4 vertices (4 vertices uniquely define a sphere)



Randomized Incremental Construction

- We start with all n points and the optimal minimal bounding circle, which is defined by 3 of those points.
- Each step, we randomly choose one of n points to remove.

← *think backwards*



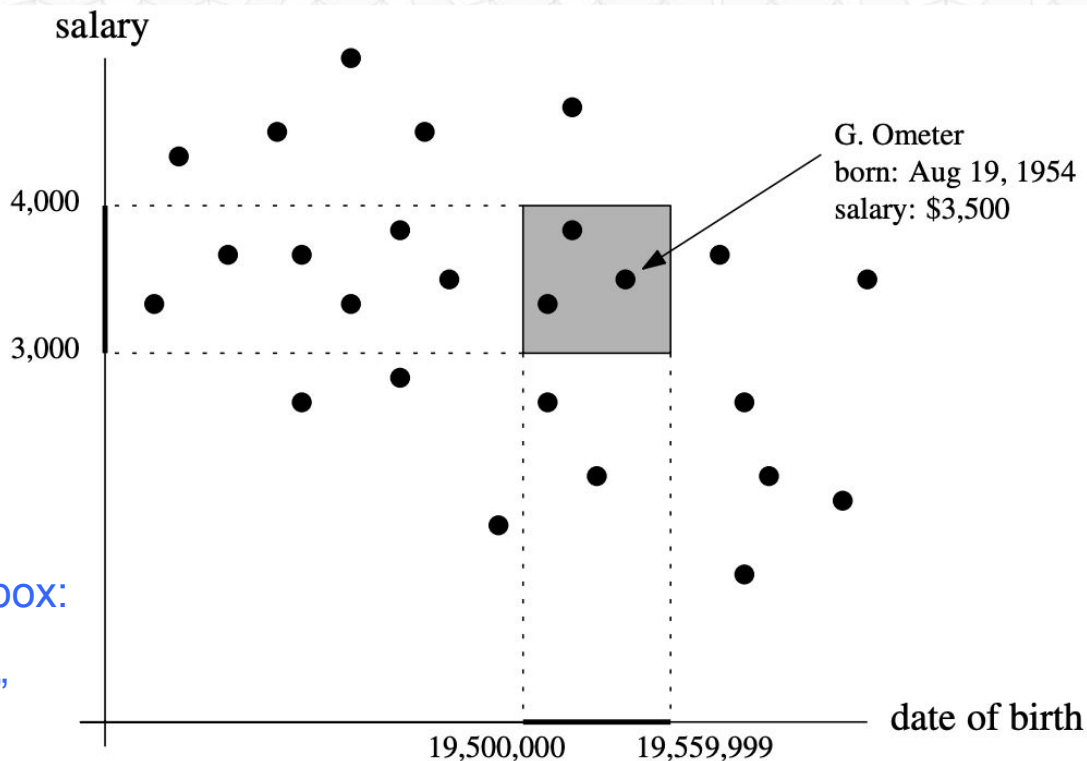
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Motivating Application: 2D Database Queries

- Return all data points with x value in range $[x_0, x_1]$ and y value in range $[y_0, y_1]$

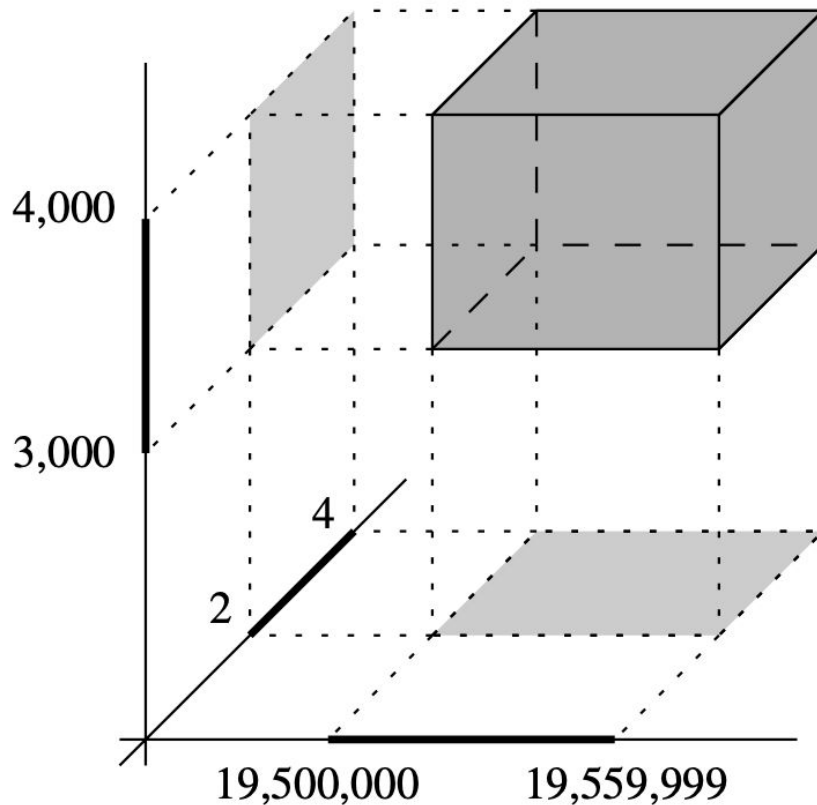
Find all values in an axis parallel box:
a “*rectangular range query*”
a.k.a. “*orthogonal range query*”



Higher Dimensional Database Queries

- Return all data points with
x value in
range $[x_0, x_1]$
and y value in
range $[y_0, y_1]$
and z value in
range $[z_0, z_1]$
and ...

Find all values in an axis parallel box:
a “*rectangular range query*”
a.k.a. “*orthogonal range query*”

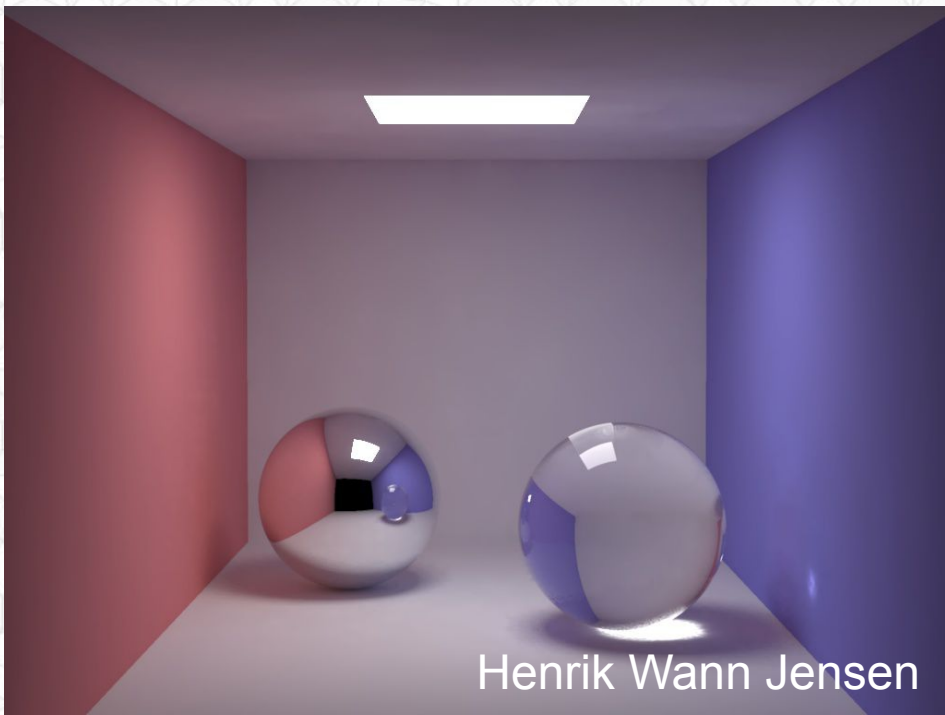
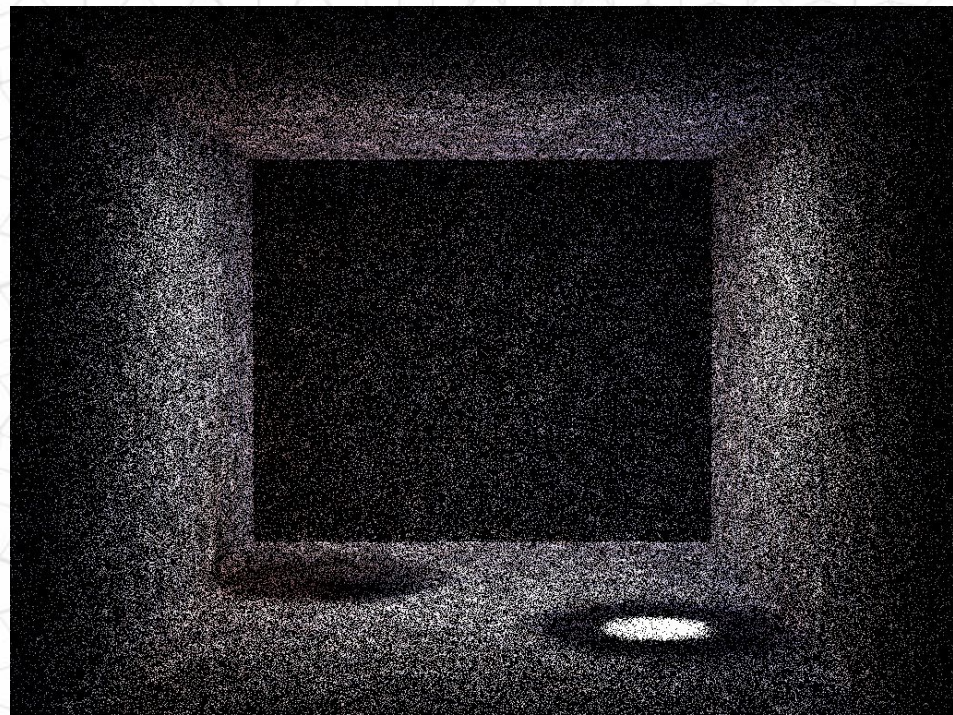


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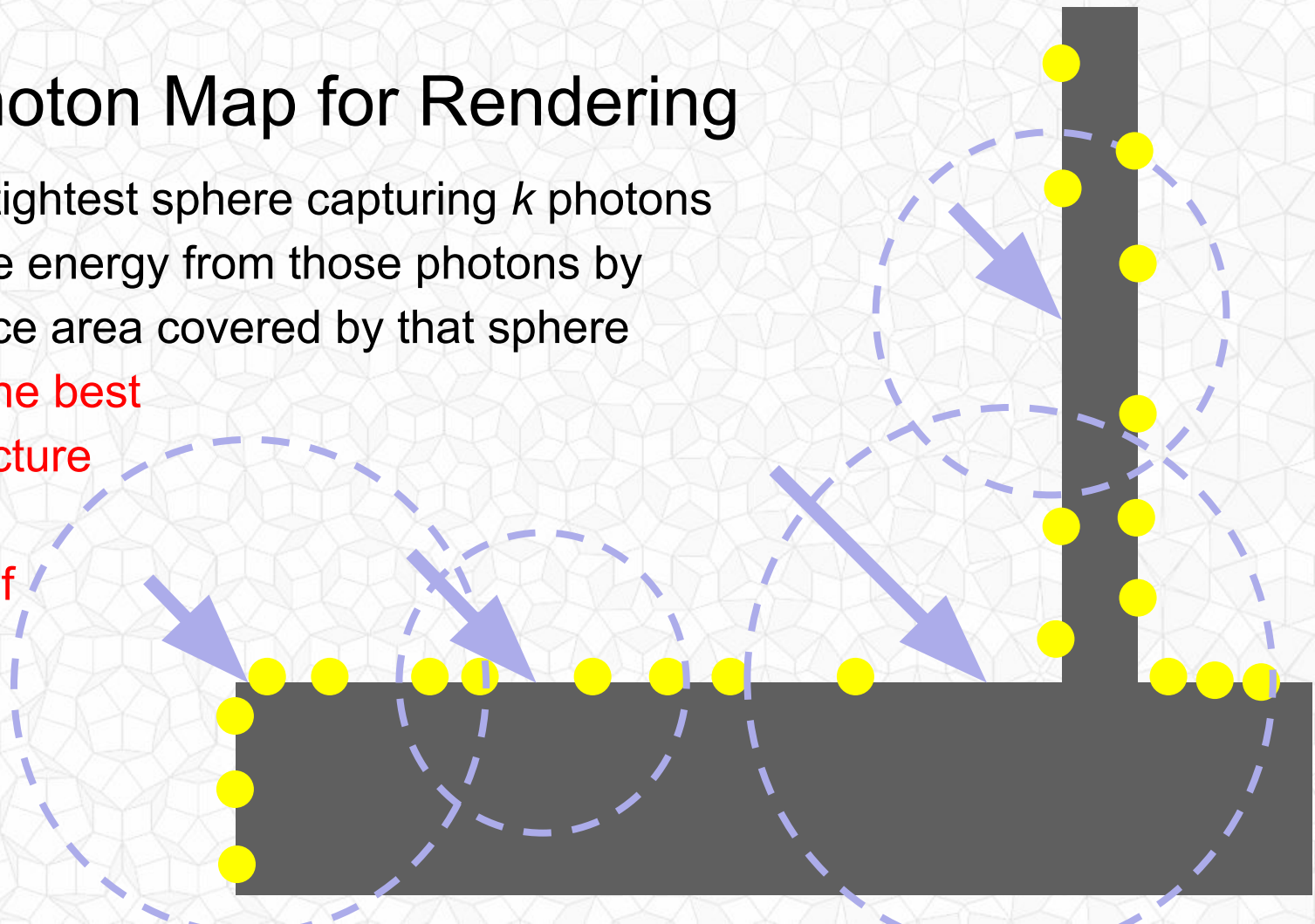
Motivating Application: Photon Mapping

- Photons bounce around room and stored on each surface they hit



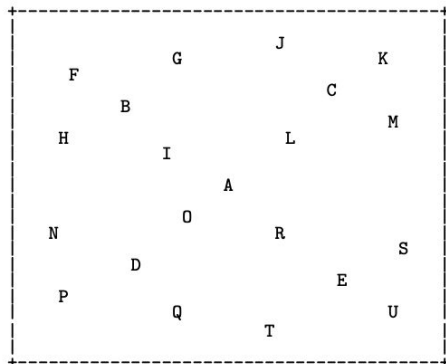
Using Photon Map for Rendering

- Find the tightest sphere capturing k photons
- Divide the energy from those photons by the surface area covered by that sphere
- What is the best data structure to store millions of photons?

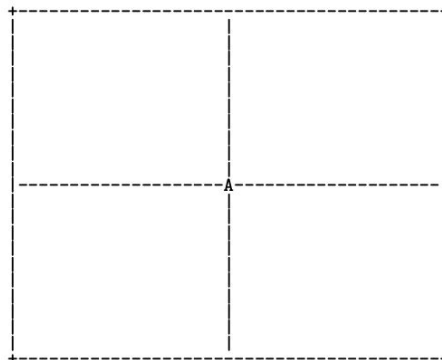


Data Structures Homework 8: Quad Tree

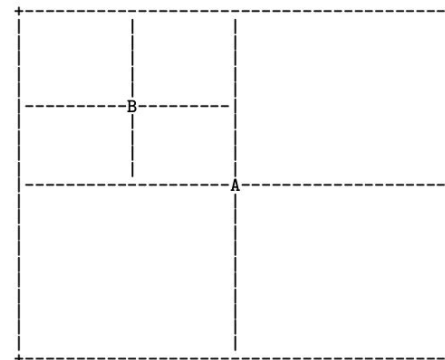
input points



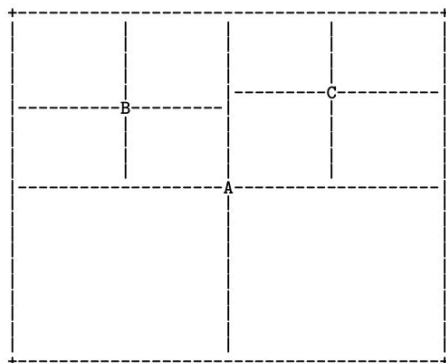
after adding the 1st point



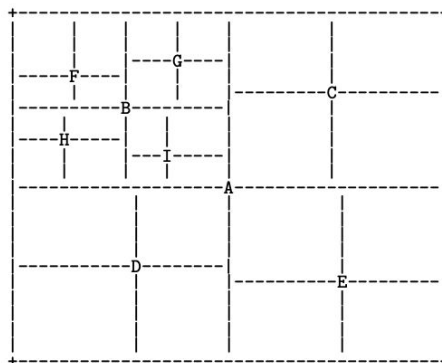
after adding the 2nd point



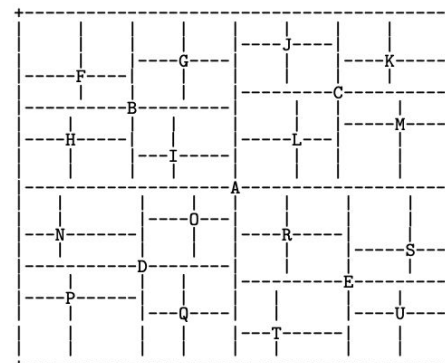
after adding the 3rd point



after adding 9 points

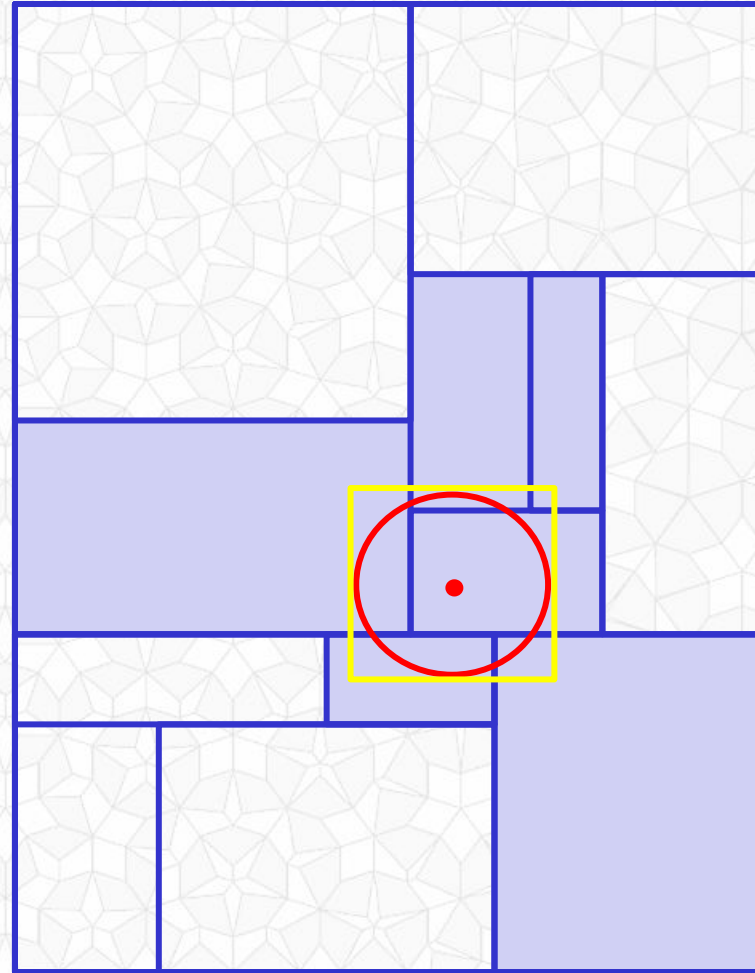


after adding all 21 points



Collecting Photons from a *kd* tree

- Query point, and initial guess for radius (red)
- Make a rectangular/orthogonal query to the *kD* tree (yellow)
- *kD* tree returns all cells that overlap with query box (blue)
- Further processing necessary to filter points inside red circle and find smallest circle capturing exactly k photons



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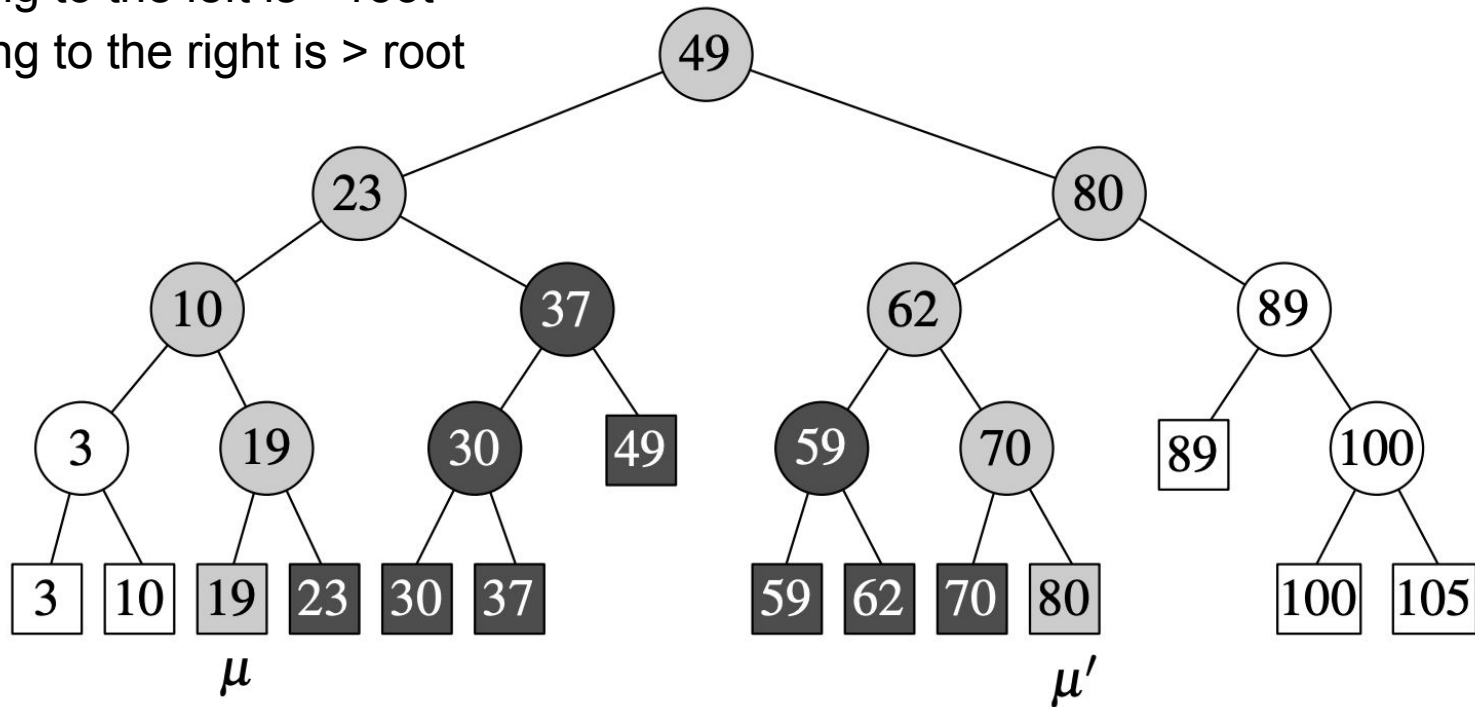
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Review: 1 Dimensional Binary Search Trees

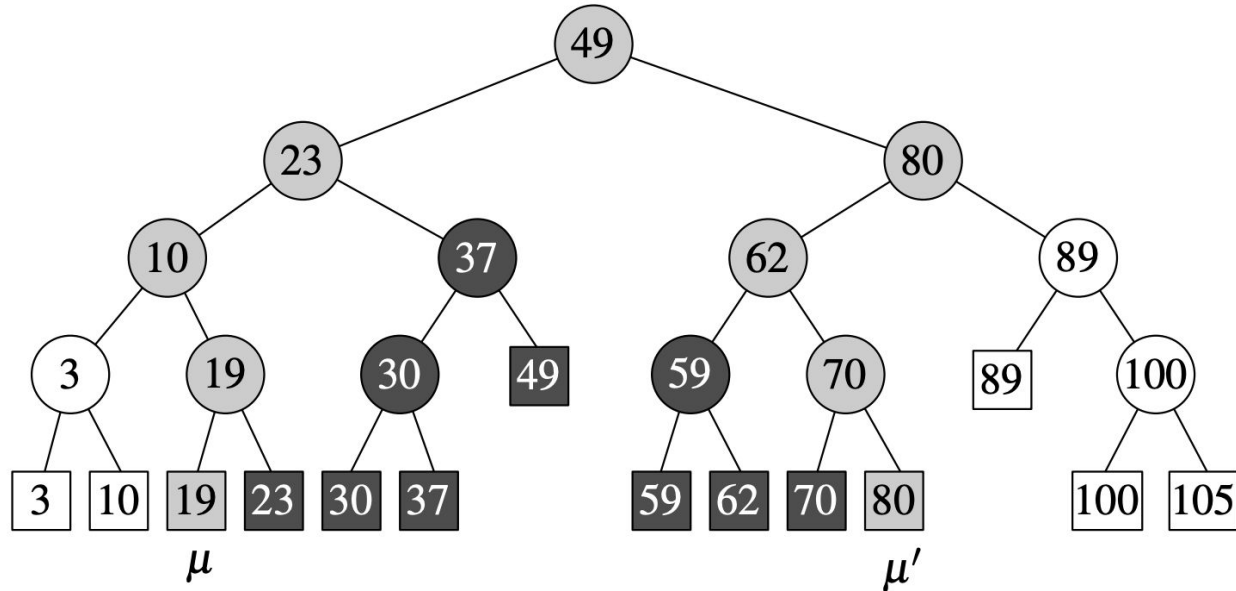
- Everything to the left is \leq root
- Everything to the right is $>$ root



Assumptions

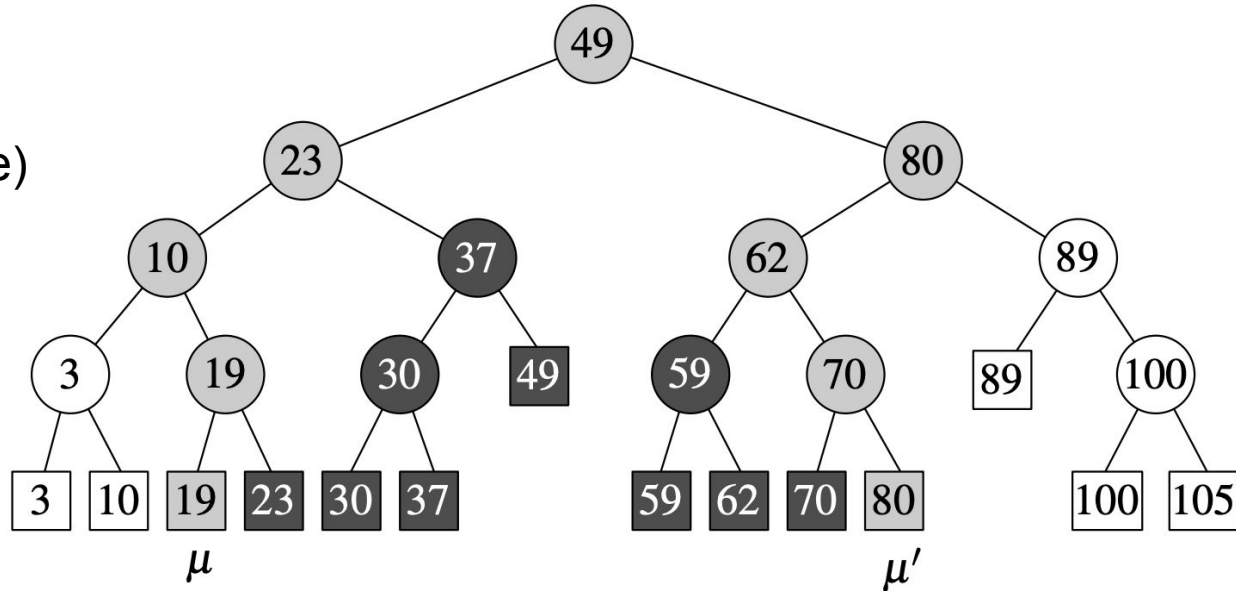
- No 2 data points have the same value in any dimension
... for convenience, there are workarounds
- We are given all of the data points at the start,
allowing us to sort the data and construct well-balanced trees

1D BST Construction Algorithm



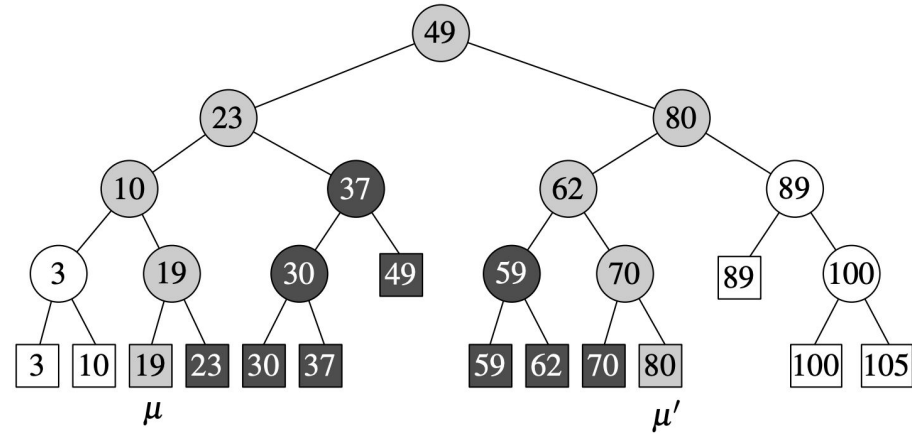
1D BST Construction Algorithm

- Sort the data by x value
- Put the median (middle) value at the root
- Create 2 sublists for left & right
- Recurse



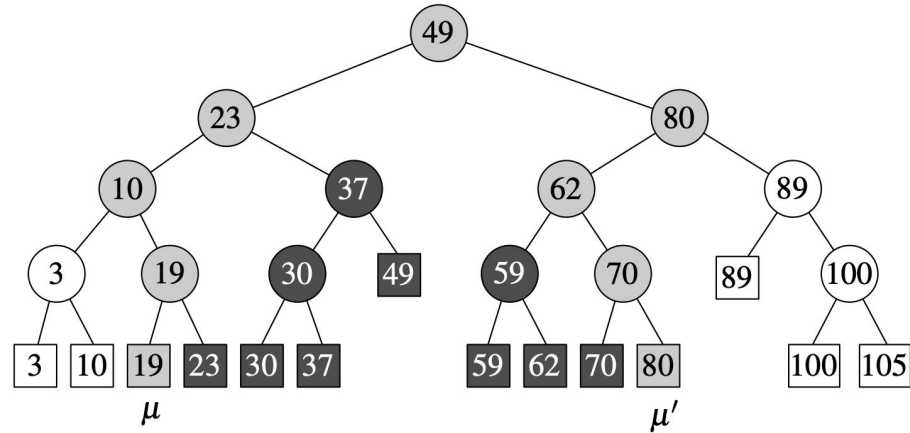
1D BST Query Algorithm

- Given a desired range $[\mu, \mu']$



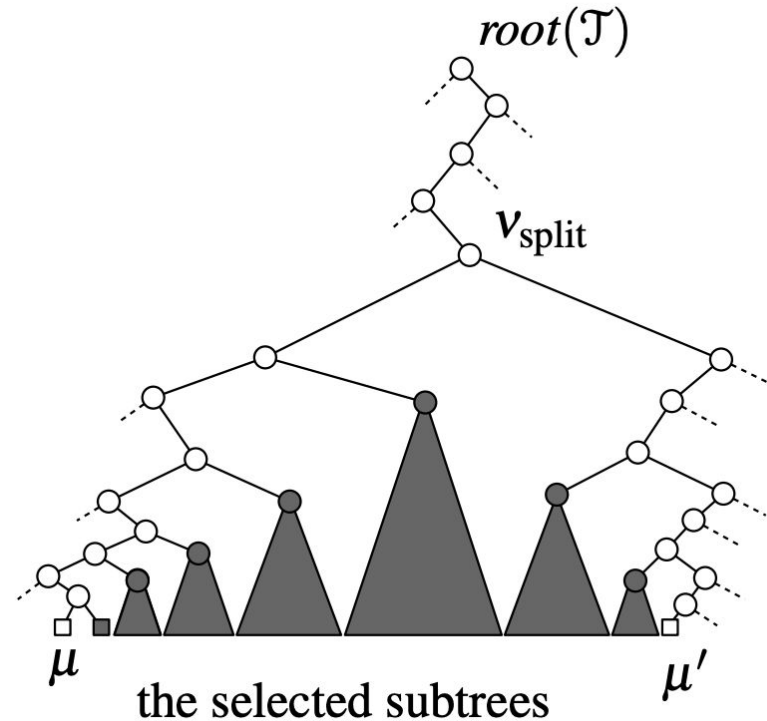
1D BST Query Algorithm

- Given a desired range $[\mu, \mu']$
- Locate the leaf storing $\mu \rightarrow O(\log n)$
- Locate the leaf storing $\mu' \rightarrow O(\log n)$
- Increment from $\mu \rightarrow \mu'$
 - `Operator++`
 $\rightarrow O(1)$ expected time
 - `Operator++` from $\mu \rightarrow \mu'$
 $\rightarrow k * O(1) = O(k)$ expected



1D BST Query Algorithm

- Given a desired range $[\mu, \mu']$
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- Increment from $\mu \rightarrow \mu'$
 - `Operator++`
 $\rightarrow O(1)$ expected time
 - `Operator++` from $\mu \rightarrow \mu'$
 $\rightarrow k * O(1) = O(k)$ expected
- *Equivalently:* Find all subtrees between the leaves, return all values in those subtrees



Analysis: 1D Binary Search Tree

Starting with n values..

- Memory to store:
 - # of leaf nodes:
 - # of intermediate nodes:
 - Height of tree:
- Time to construct:
 - Sort the data:
 - Place middle value at root, recurse on left & right sublists:
- Time to query:
 - For search target / output returning k values

Analysis: 1D Binary Search Tree

Starting with n values..

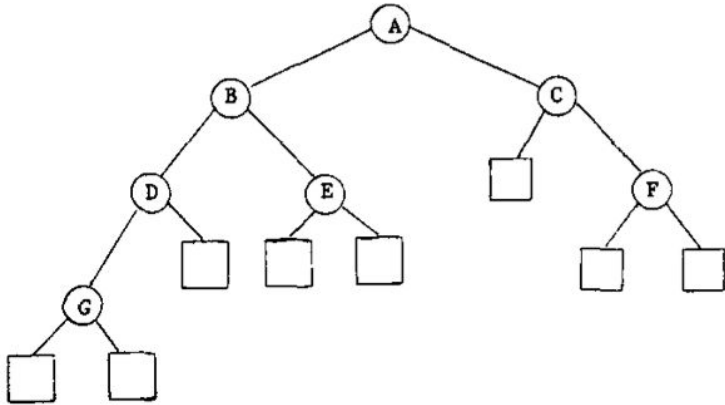
- Memory to store: $\rightarrow O(n)$
 - # of leaf nodes: n
 - # of intermediate nodes: $n-1$
 - Height of tree: $\log n$
- Time to construct: $\rightarrow O(n \log n)$
 - Sort the data: $O(n \log n)$
 - Place middle value at root, recurse on left & right sublists: $O(n)$
- Time to query: $\rightarrow O(\log n + k)$
 - For search target / output returning k values

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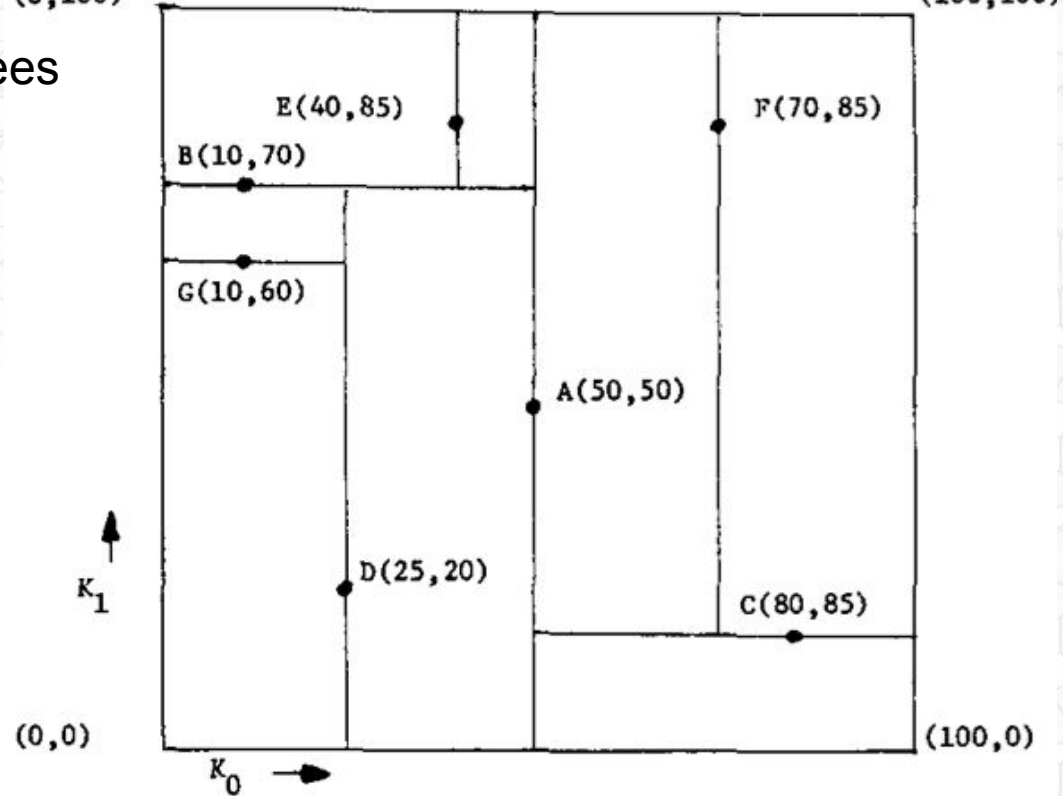
What is a k -d Tree?

"Multidimensional Binary Search Trees
Used for Associative Searching",
Communications of the ACM,
Bentley 1975

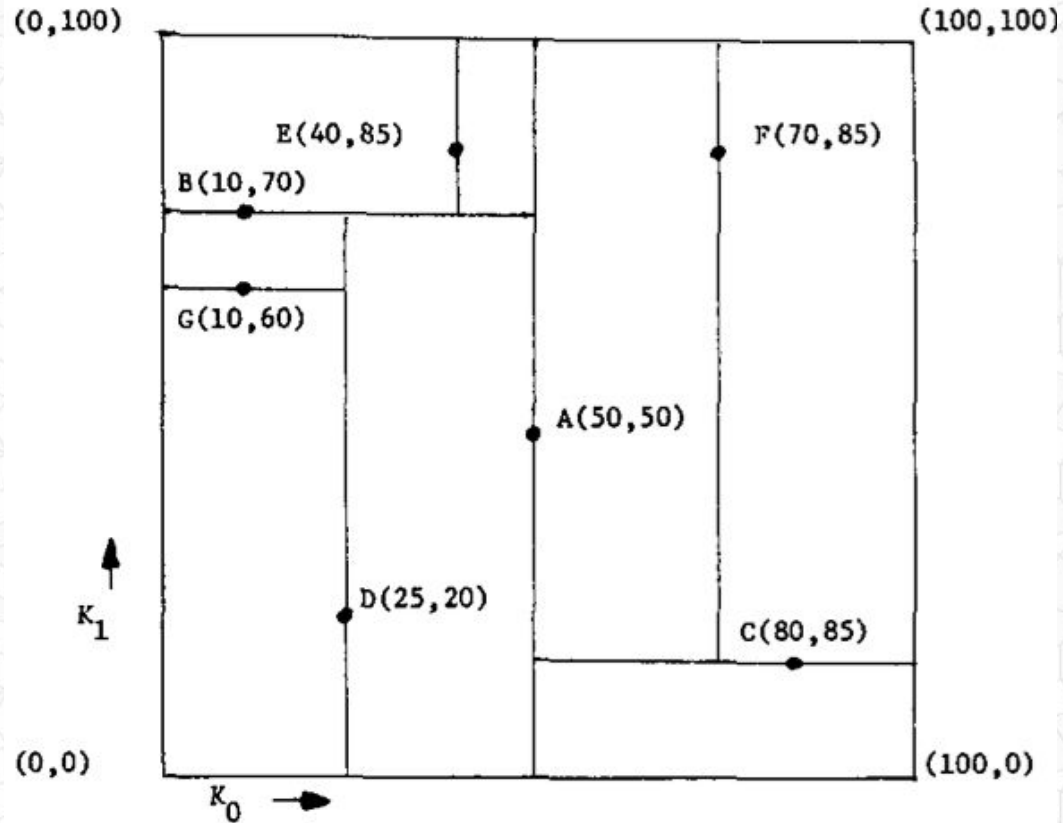


(0,100)

(100,100)

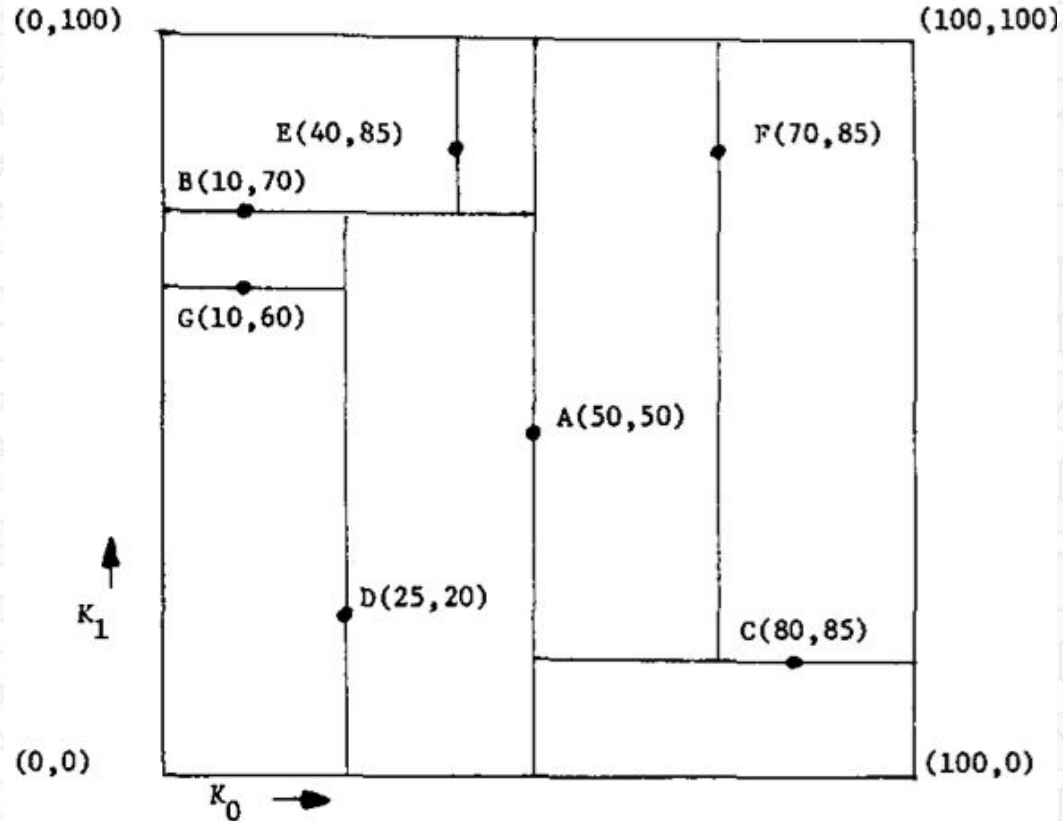


2D kd Tree Construction



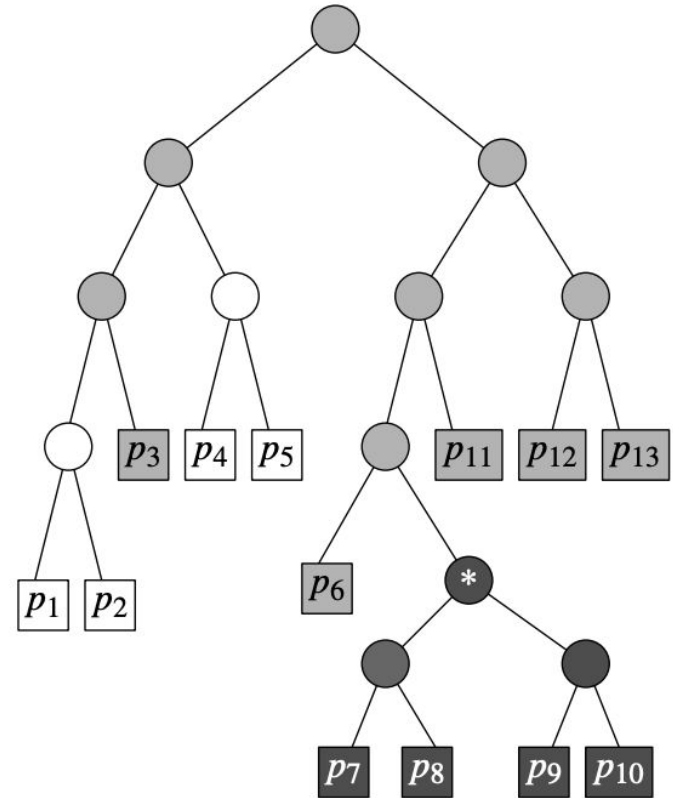
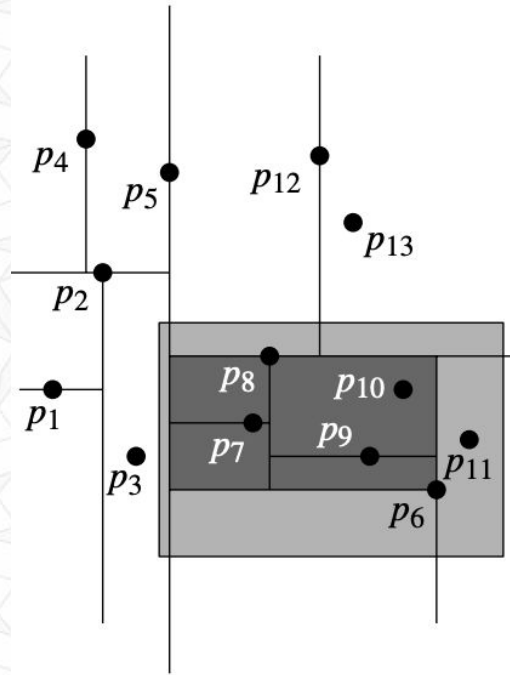
2D kd Tree Construction

- Make 2 sorted lists, by x value and by y value
- Alternate dimensions (first split by x then by y)
- Find the median value
- Make a copy of the sorted lists, removing values from the other side
- Recurse



2D kd Tree Query Algorithm

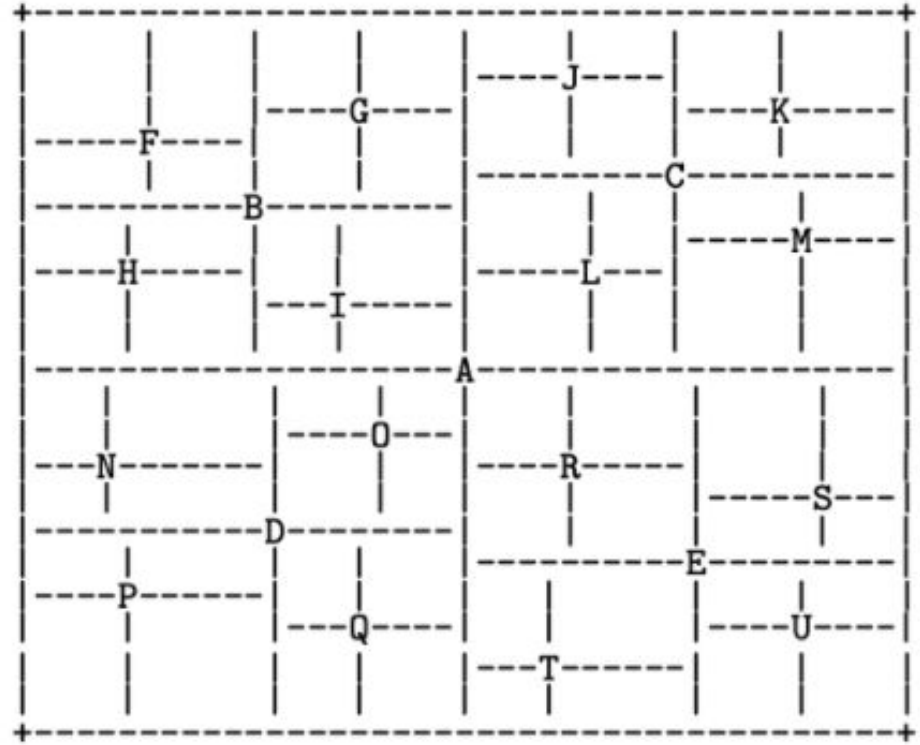
- At each split point
- Determine if the query box overlaps the split line
- Recurse down one or both branches
- If a subtree lies complete inside the box, return all items in that subtree
- Perform filtering in the leaves as necessary



Ignore that this is quadtree
pretend it is kd tree

2D kd Tree Query Analysis

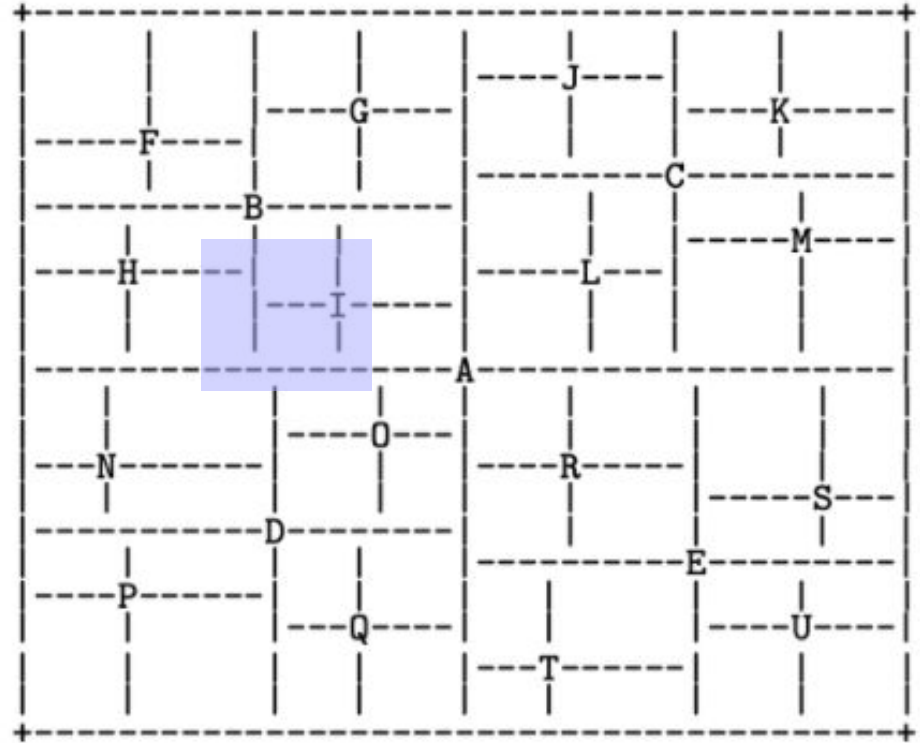
- 1 item is stored per leaf node
- For a query that will collect k items



Ignore that this is quadtree
pretend it is kdtree

2D kd Tree Query Analysis

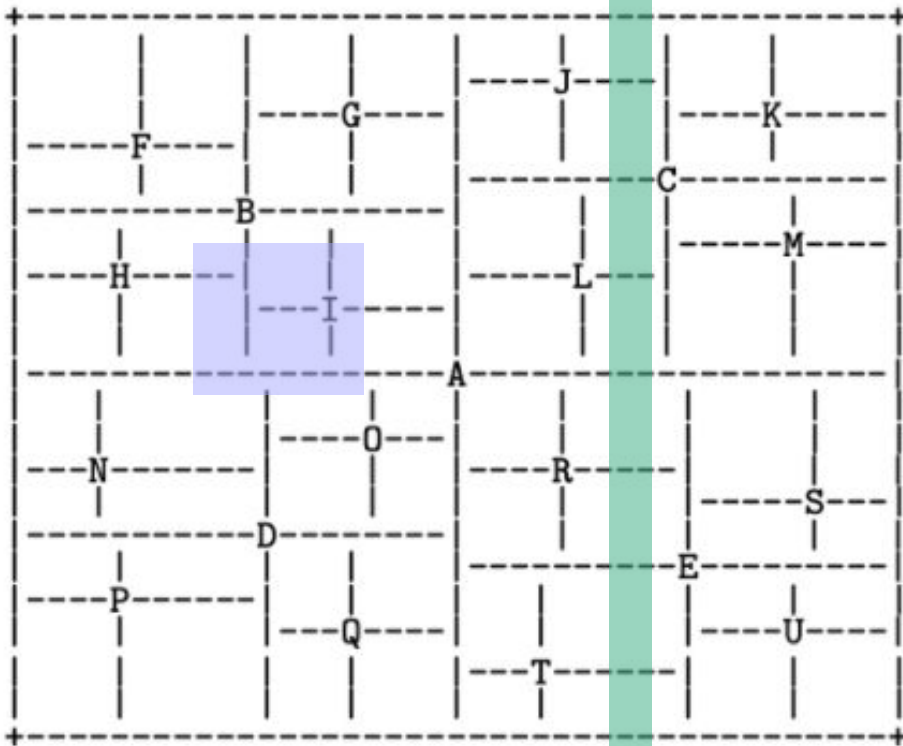
- 1 item is stored per leaf node
- For a query that will collect k items
- Best/Average(?) Case:
An approximately square query
(equal width & height)
 - touches/overlaps $O(k)$ leaves
 - gathering leaves $O(\log n + k)$
 - Overall $\rightarrow O(\log n + k)$



2D kd Tree Query Analysis

Ignore that this is quadtree
pretend it is kd tree

- 1 item is stored per leaf node
- For a query that will collect k items
- Best/Average(?) Case:
An approximately square query
(equal width & height)
 - touches/overlaps $O(k)$ leaves
 - gathering leaves $O(\log n + k)$
 - Overall $\rightarrow O(\log n + k)$
- Worst Case Query:
For a skinny / lopsided query box
 - touches/overlaps - $\sqrt{n} + k$ leaves
 - gathering leaves $O(\sqrt{n} + k)$
 - Overall $\rightarrow O(\sqrt{n} + k)$



Analysis: 2D kd Tree

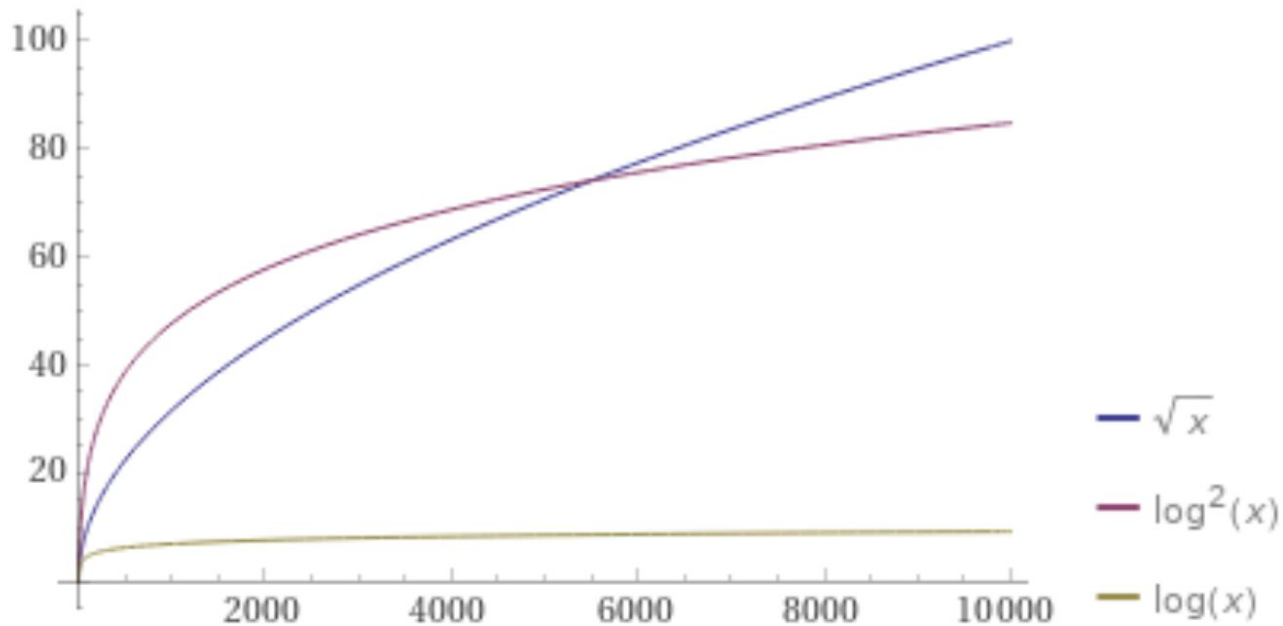
Starting with n values..

- Memory to store: $\rightarrow O(n)$
 - # of leaf nodes: n
 - # of intermediate nodes: $n-1$
 - Height of tree: $\log n$
- Time to construct: $\rightarrow O(n \log n)$
 - pre-sort the data, separately in x and in y : $O(n \log n)$
 - Alternate axes - place middle value at root, recurse on the two sublists: $O(n \log n)$
- Time to query: $\rightarrow O(n^{1/2} + k) = O(\sqrt{n} + k)$
 - For search target / output returning k values

Is Query Time = $O(\sqrt{n} + k)$ a problem?

Is Query Time = $O(\sqrt{n} + k)$ a problem?

- $O(1) < O(\log n) < O(\log^2 n) < O(\sqrt{n}) < O(n)$



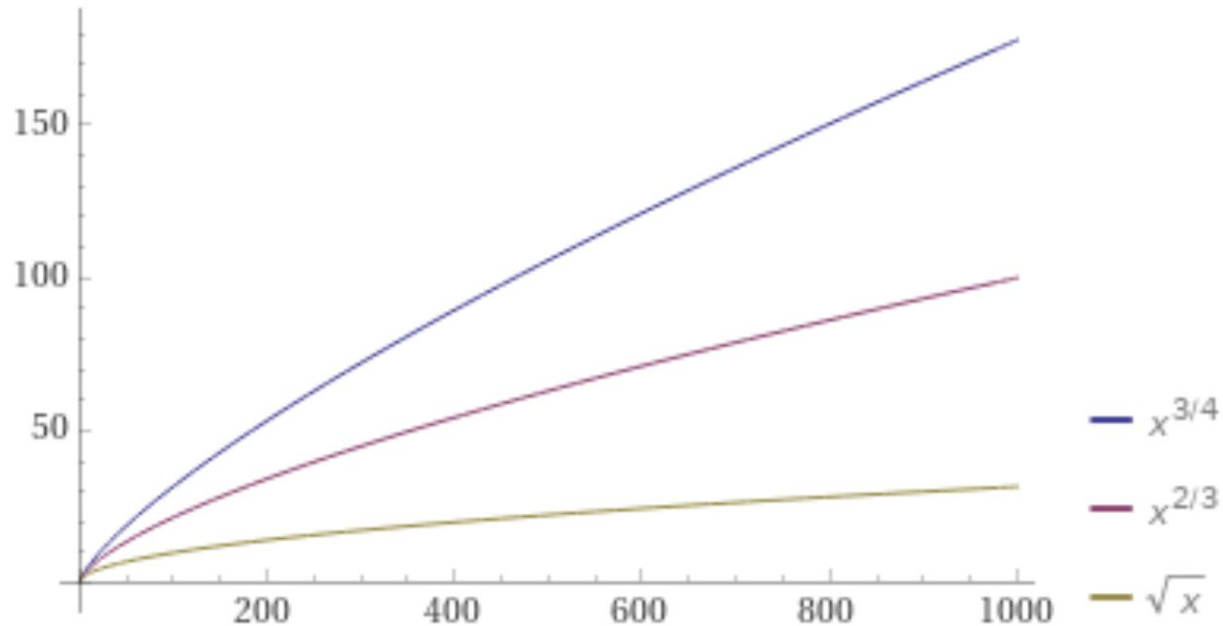
Analysis: 3D kd Tree and higher dimensions

Starting with n values..

- Memory to store: $\rightarrow O(n)$
 - # of leaf nodes: n
 - # of intermediate nodes: $n-1$
 - Height of tree: $\log n$
- Time to construct: $\rightarrow O(n \log n)$
 - pre-sort the data, separately in x and in y and in z : $O(n \log n)$
 - Rotate through axes (x, y, z, x, \dots) – place middle value at root, recurse on two sublists: $O(n \log n)$
- Time to query: $\rightarrow O(n^{2/3} + k) \rightarrow O(n^{(1-1/d)} + k)$
 - For search target / output returning k values

Is Query Time = $O(n^{(1-1/d)} + k)$ a problem?

- Yeah, this is a problem as dimensions increase
- Common for complex databases



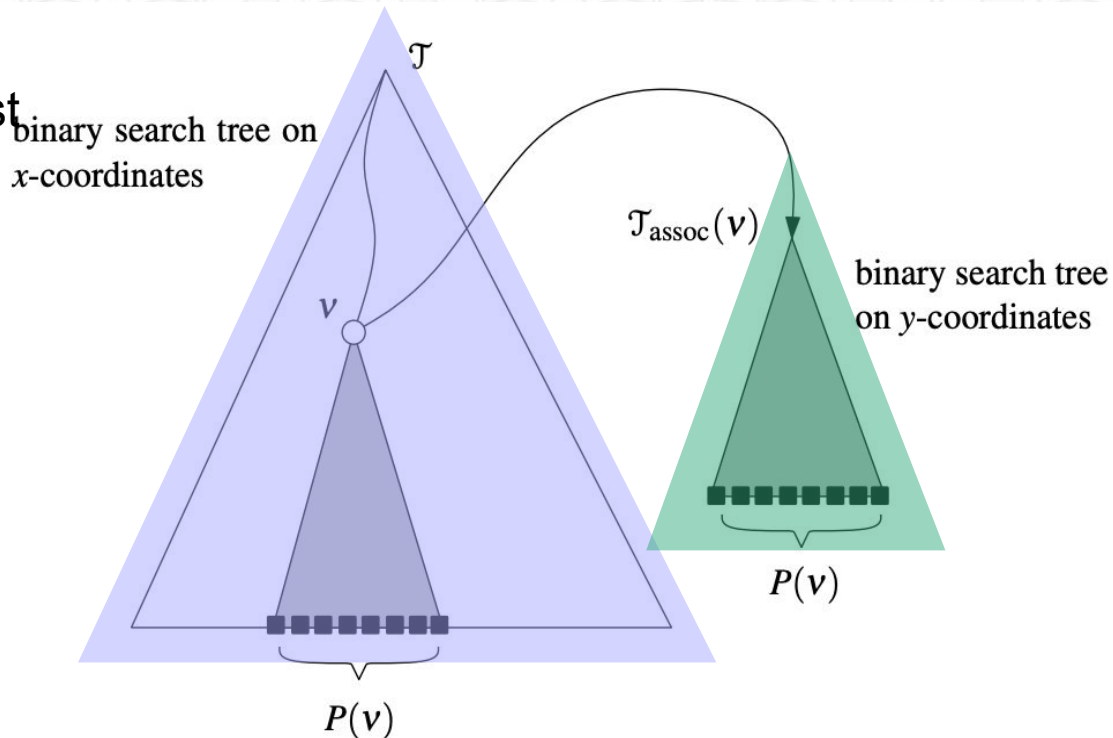
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- **2D Range Trees & Higher Dimension Range Trees**

What is a Range Tree?

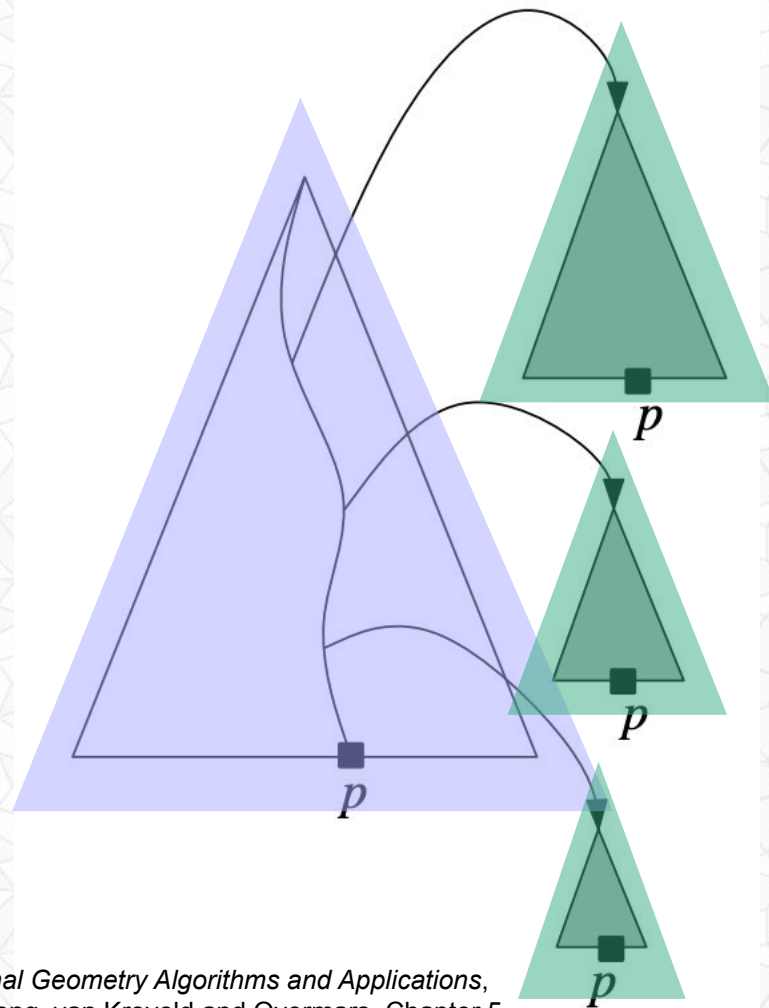
- Idea: If we use more memory, can we reduce worst case query time of kD tree?

- *First we organize the data in a BST by x value*
- *At every node in the tree, we store a pointer to a BST with the same data, but organized by y value*



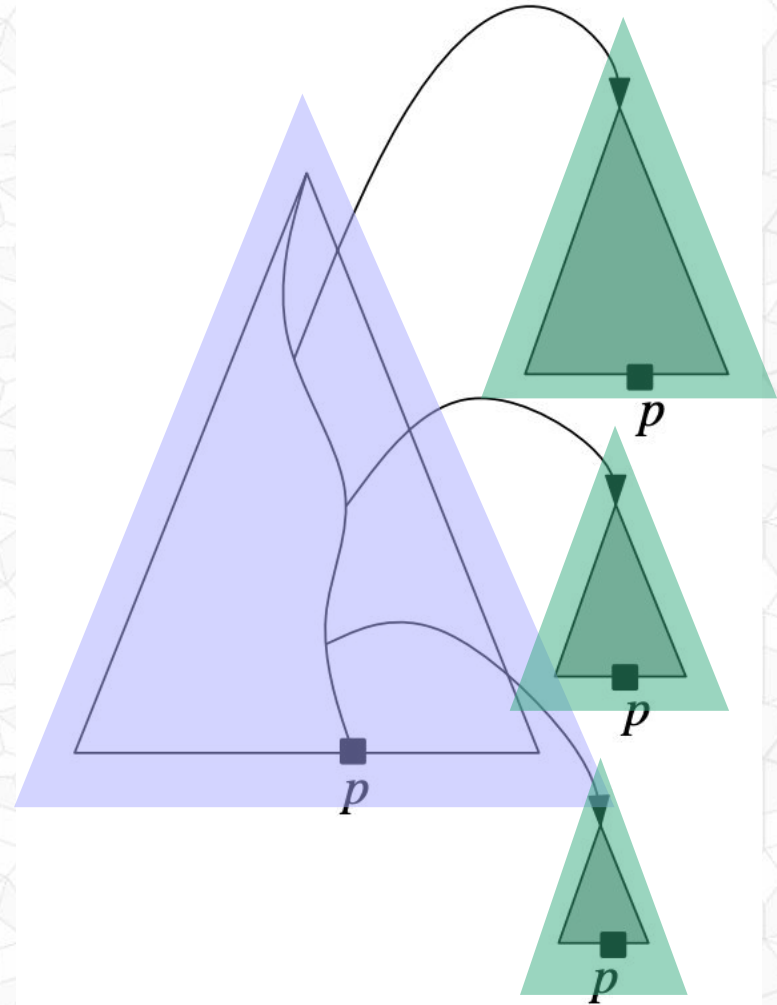
What is a Range Tree?

- Idea: If we use more memory, can we reduce worst case query time of kD tree?
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How to Construct the 2D Range Tree?

How much memory does it use?

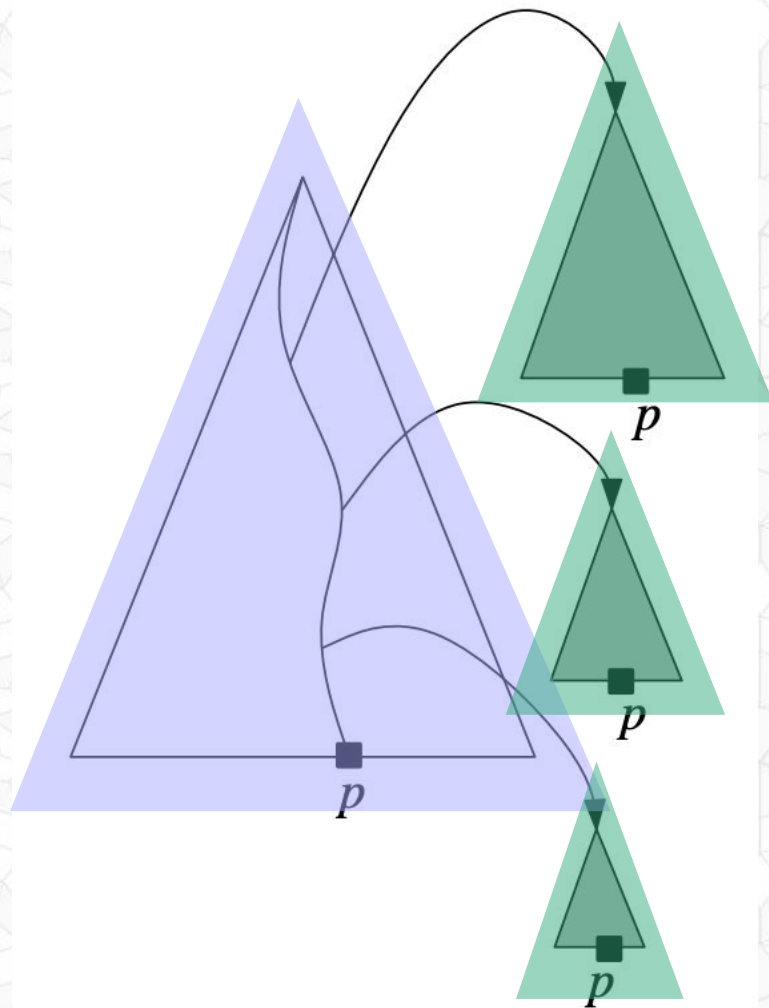


How to Construct the 2D Range Tree?

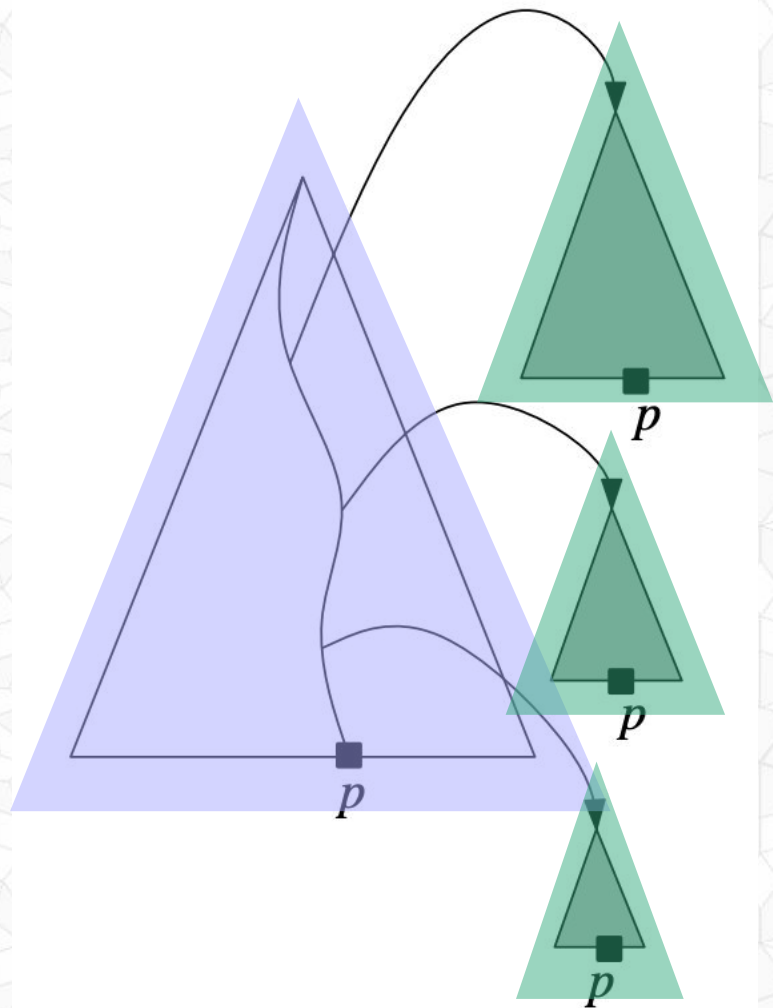
How much memory does it use?

- Each point p is stored once in the level 1 (organized by x) tree
- And many times in level 2 (organized by y) trees
- How many level 2 trees? And how big are they?
 - 1 tree with n values
 - 2 trees with $n/2$ values
 - 4 trees with $n/4$ values
 - ...
 - n trees with 1 values

→ $O(n \log n)$ memory



How to Query 2D Range Tree?



Analysis: 2D Range Tree

Starting with n values..

- Memory to store: $\rightarrow O(n \log n)$
- Time to construct: $\rightarrow O(n \log n)$
- Time to query: $\rightarrow O(\log^2 n + k)$

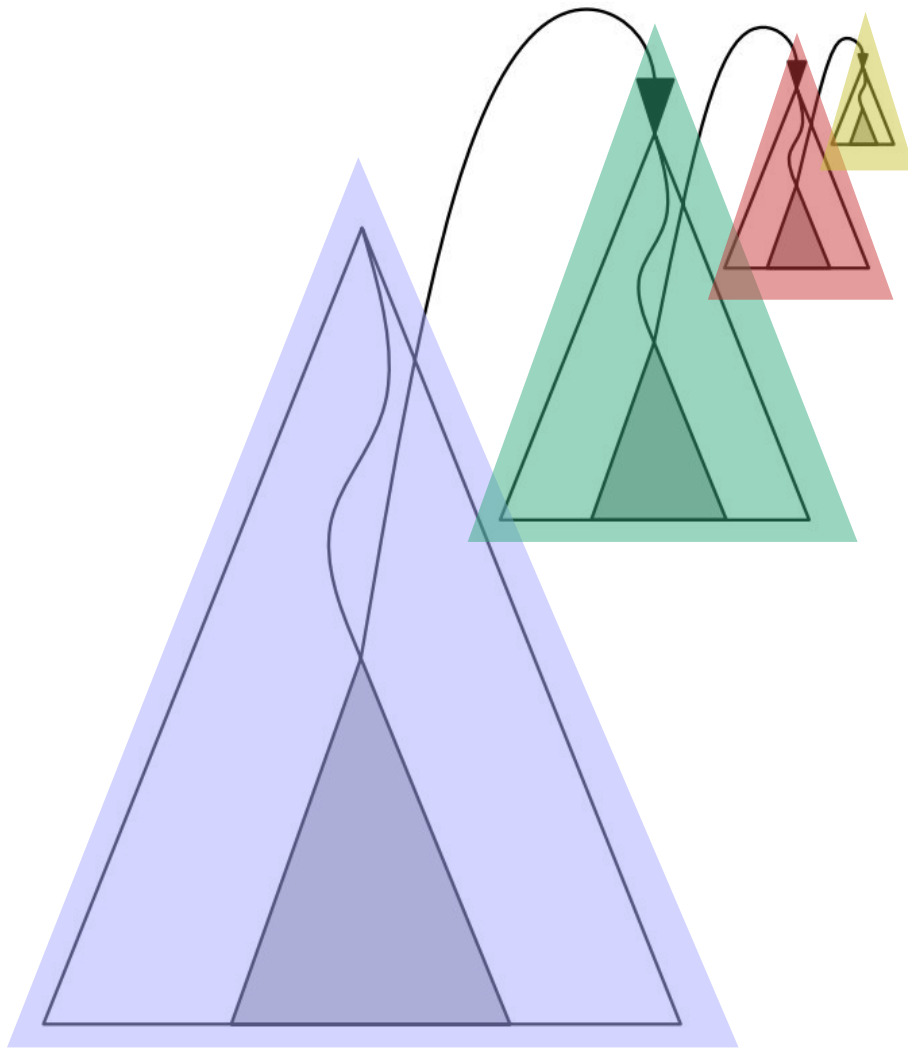
Analysis: 2D Range Tree

Starting with n values..

- Memory to store:
- Time to construct:
- Time to query:

Higher Dimensional Range Tree

- ... and can be extended to arbitrarily higher dimensions



Analysis: 3D *kd* Tree and higher dimensions

Starting with n values..

- Memory to store:
- Time to construct:
- Time to query:

Analysis: 3D *kd* Tree and higher dimensions

Starting with n values..

- Memory to store: $\rightarrow O(n \log^{d-1} n)$
- Time to construct: $\rightarrow O(n \log^{d-1} n)$
- Time to query: $\rightarrow O(\log^d n + k)$

Summary Comparison

- **kD tree**
 - Construction time
 - Memory
 - Query time

- **Range tree**
 - Construction time
 - Memory
 - Query time

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