

# Multi-Dimensional Data Sets

- Current data structures such as search trees can work only with 1-dimensional data.
- A big inherent problem with multi-dimensional data is that they are not comparable.
- In a 2-d setting, which is bigger? (10, 15) or (22, 8)?
- So need new data structures that can impose an order on multi-dimensional data.



# The Case of 1-Dimension

- Identifying the relevant points in a 1-dimensional setting is possible.
- Think of a search tree  $T$ , that can locate all the values between  $x$  and  $y$  (both inclusive).
- Can be done in time  $O(k + \log n)$  where  $k$  is the number of such elements.
- How?



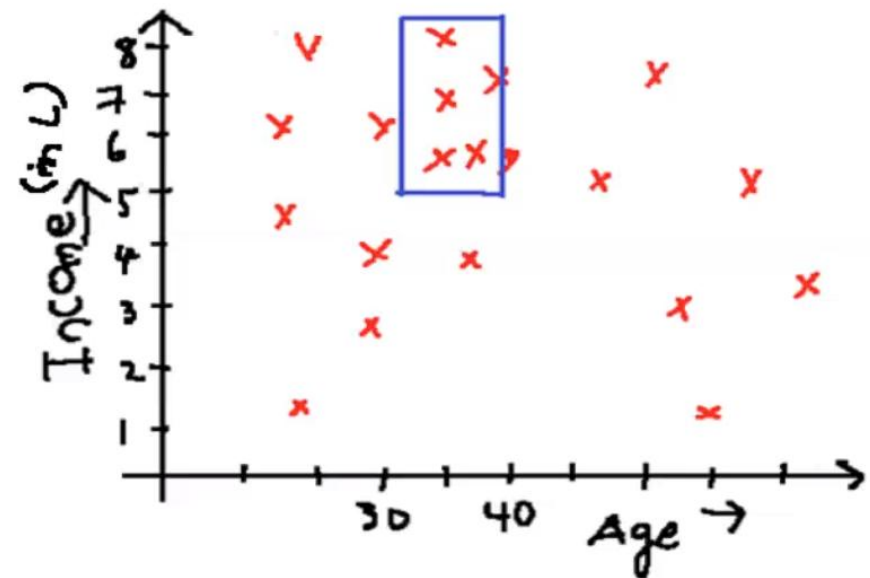
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  - Need not filter the  $n$  points which takes  $O(n)$  time.
- How?
- Need a similar technique and a data structure for higher dimensional data sets.
- A solution that is faster than  $O(n)$ .



# Three Solutions

- We will study **three** different data structures in this context.
- We will also seek to solve a standard query:
- Given a rectangular region  $q = [a,b] \times [c,d]$ , find all the points of  $P$  that are inside  $q$ .
- This is called as a range query.



# Some Solution Ideas

- In each case, suppose there are  $n$  points in a  $d$ -dimensional space.
- Can consider all these points and arrive at the result.
- Typically, however, the region of interest, or the query region, has far fewer points. Can find the result on **these** subset of points.
- But, identifying these points itself may take time.



# A New Way for Data Structures

- Preprocess the points and create a suitable data structure.
- This data structure can then be used repeatedly for answering any query.
- Some parameters of efficiency:
  - Space used by the data structure
  - Time taken to build the data structure
  - Time to answer a query
- Query time is lower bounded by the size of the output. This is denoted  $k$ .
- So, we seek query time of  $O(k + \text{polylog}(n))$  so that the result is practically fast.



# A New Way for Data Structures

- Seemingly easy with one dimensional data.
- Build a balanced binary search tree of the  $n$  points.
- Exercise: Given two values  $x$  and  $y$  with  $x < y$ , can find the values in the input set that are between  $x$  and  $y$ .
- Can do this in time  $O(k + \log n)$ .
- Is it possible to do similar things in two and further dimensions?



# First Solution – A Quad Tree

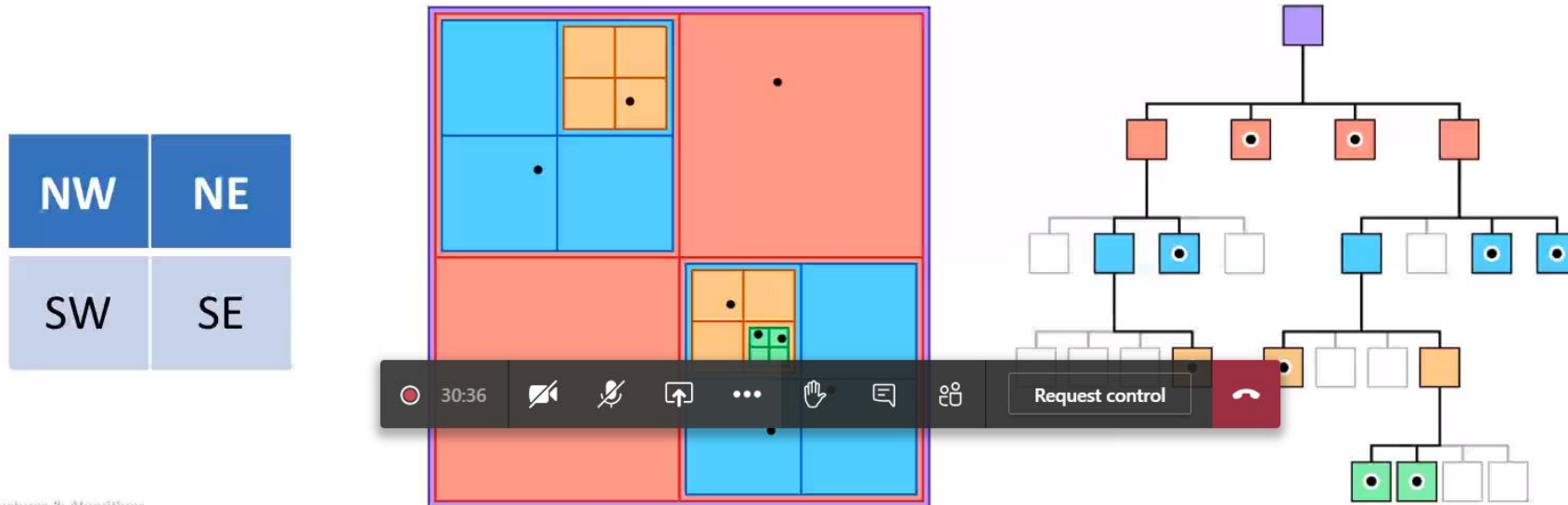
- We will first generalize a binary search tree to be useful for 2-d points.
- A node now has four children, labeled NE, NW, SE, and SW.
- At a node  $u$ , the points in the subtree rooted at the NE child have  $x$ - and  $y$ - coordinates larger than that of  $u$ .





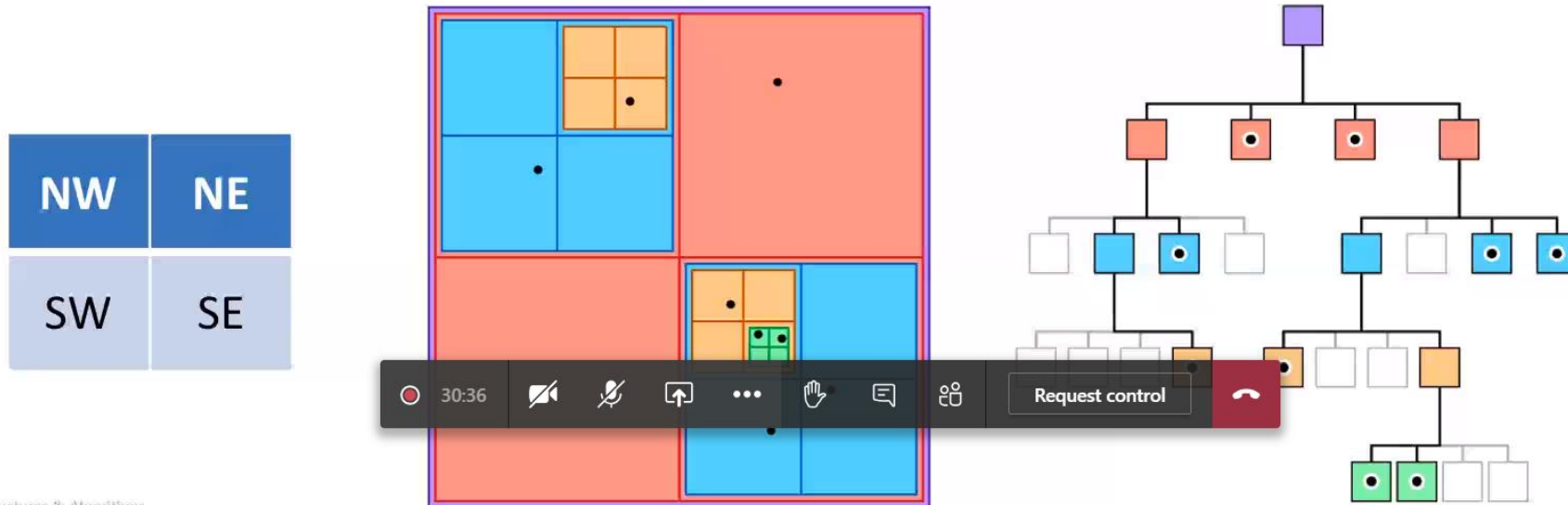
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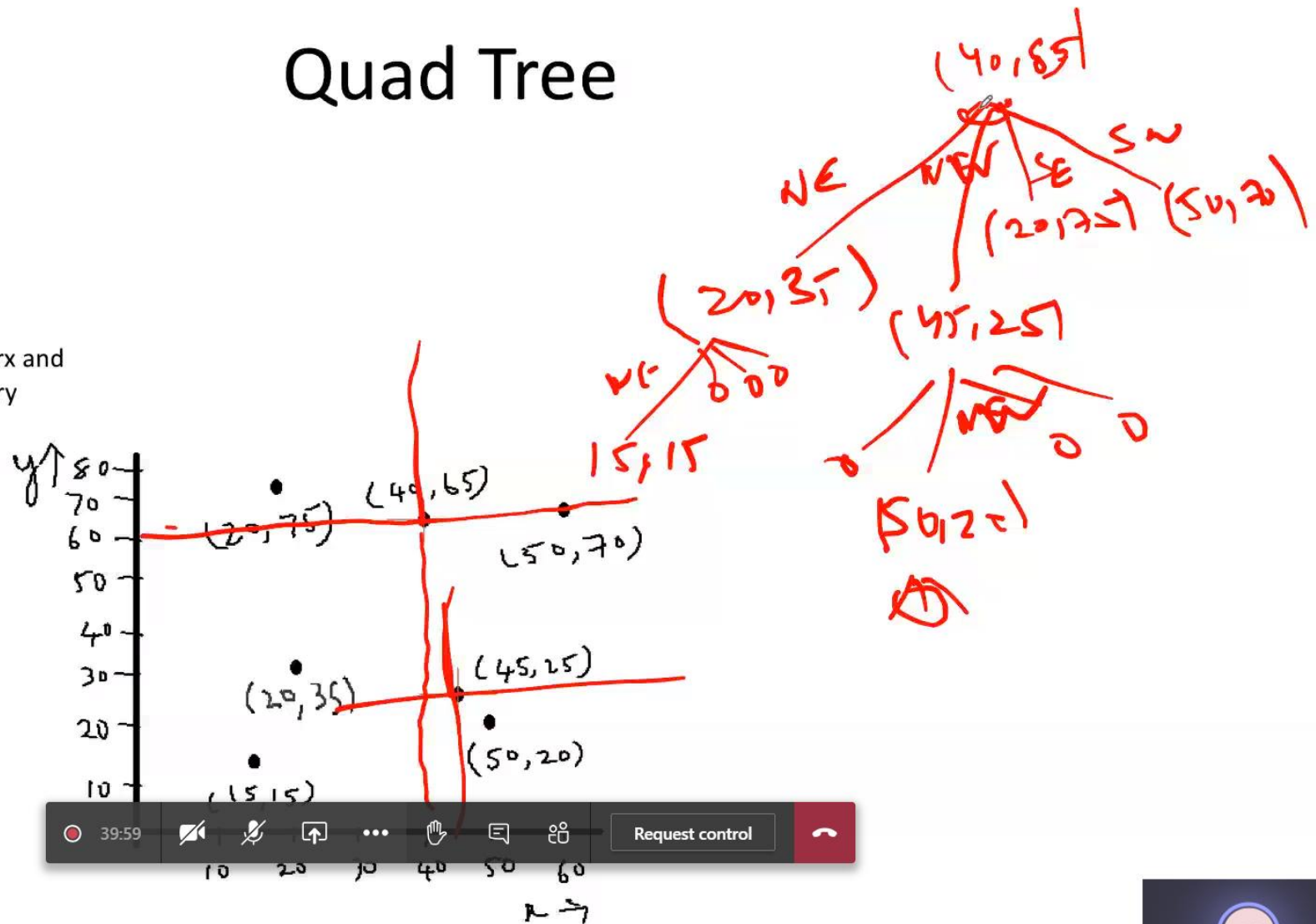
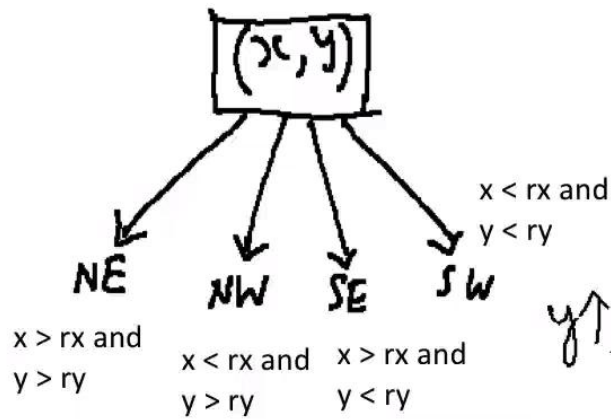


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# Quad Tree



# Quad Tree

- Similar routines to that of a binary search tree.
- To Insert( $p = (x, y)$ ) into a quad tree  $Q$ , do the following:
  - Start from the root of the tree.
  - Let the point at the root be  $r = (r_x, r_y)$ .
  - Four cases:
    - If  $x > r_x$  and  $y > r_y$  – Insert  $p$  in the NE child.
    - If  $x > r_x$  and  $y < r_y$  – Insert  $p$  in the SE child
    - If  $x < r_x$  and  $y > r_y$  – Insert  $p$  in the NW child.
    - If  $x < r_x$  and  $y < r_y$  – Insert  $p$  in the SW child.



# Quad Tree

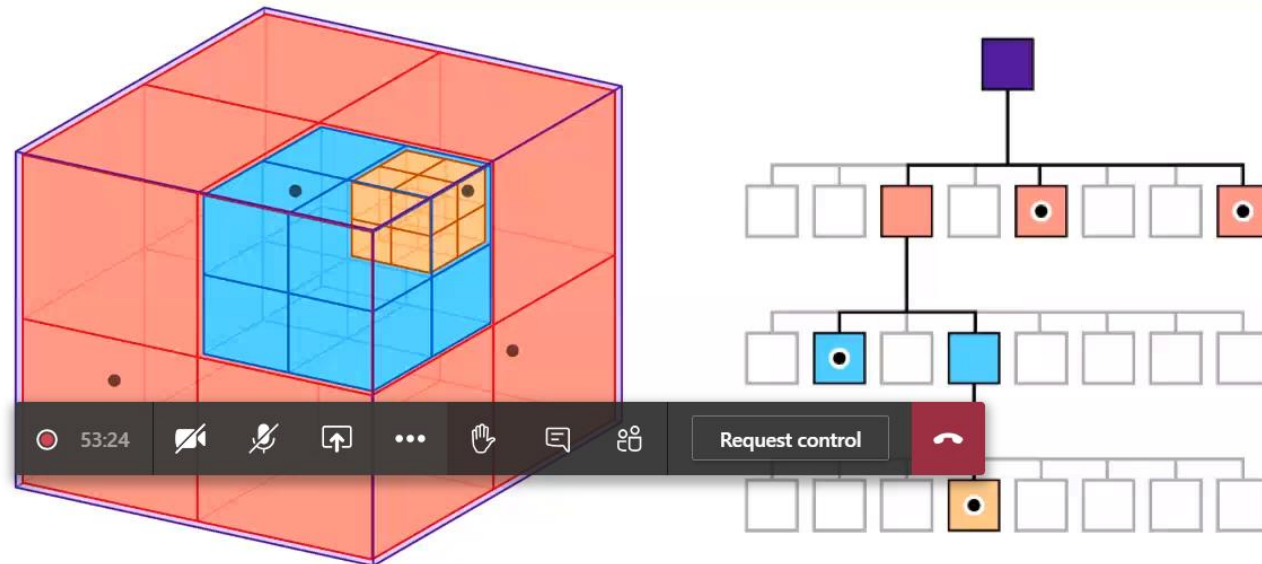
- Similar routines to that of a binary search tree.
- A Delete(p) routine can also be designed akin to the Delete routine of a binary search tree.
- A Find(p) routine is also similar. Start from the root, and depending on the x and y coordinates, search one of the four subtrees.



# Quad Tree

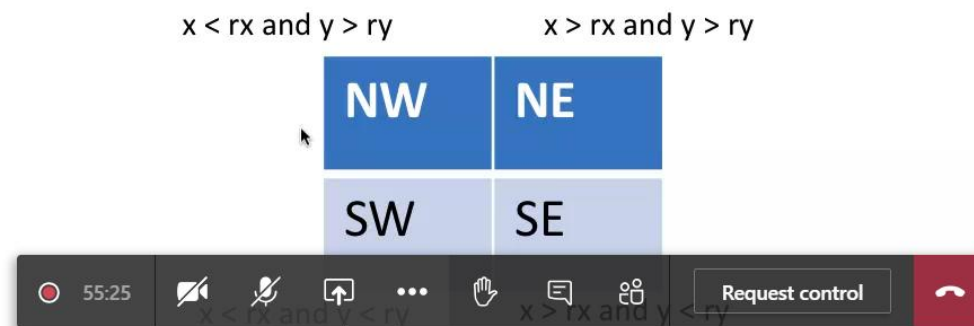
- But there are several disadvantages:
- Height balance is difficult to achieve on insertions and deletions.
- Number of children grows exponentially with the number of dimensions.

Octree  
for 3 dimensions



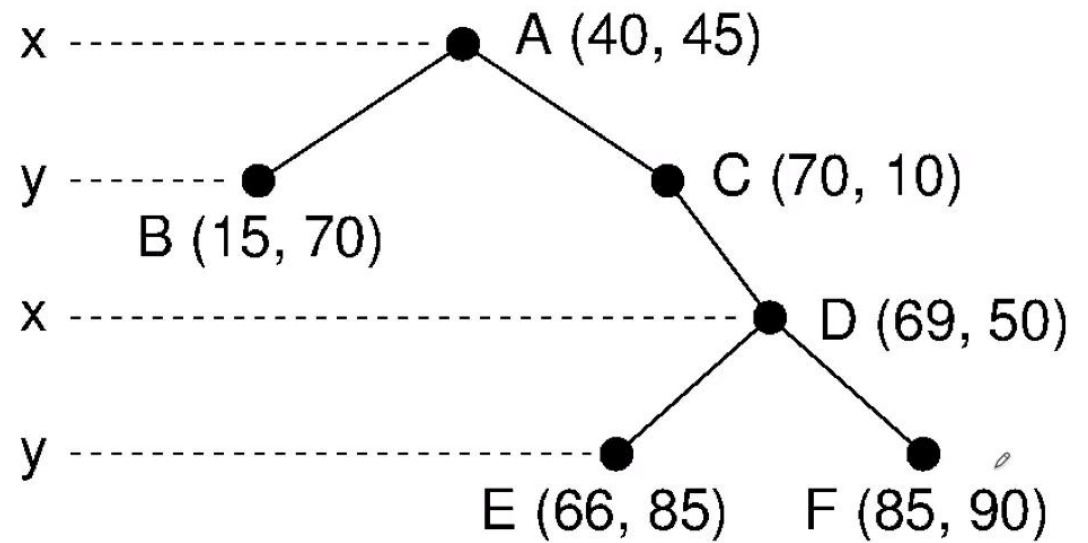
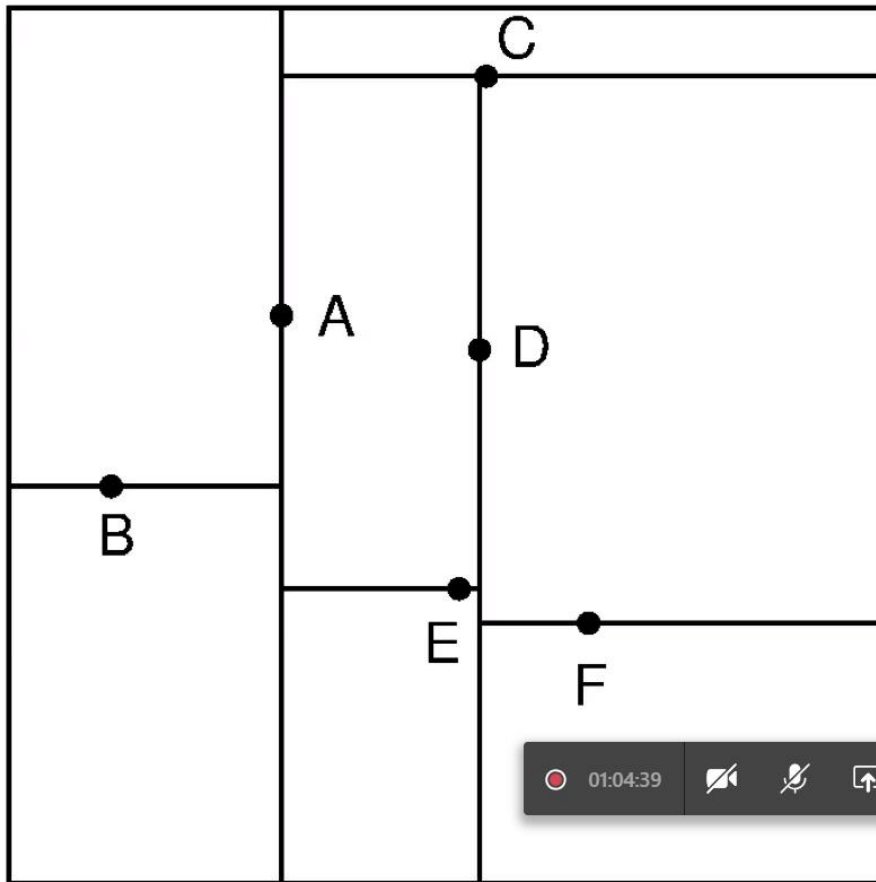
# Quad Tree

- Build a quad tree for the following set of 2-dimensional points. (Insert points into an initially empty quad tree in the same order)
- (15, 35), (20, 40), (25, 25), (10, 20), (50,10), (30, 30), (40, 20), and (45, 55)





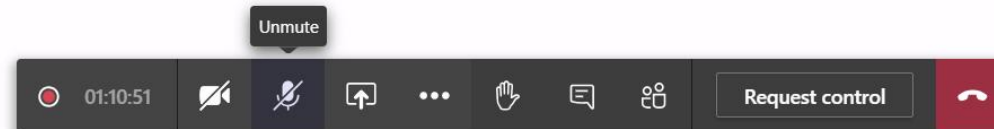
## A Better Solution – A 2-d Tree





# A Better Solution – A 2-d Tree

- Inserting into the 2-d tree is straight forward.
- Proceed from the root of the tree, at each level comparing the appropriate coordinate value.
- For higher dimensions, cycle over all the  $d$  dimensions over  $d$  levels.



# A Better Solution – A 2-d Tree

- Insert the following points into an initially empty 2-d tree.
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