

Implementation of KD Trees and KNN Algorithm

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Introduction

- **What is a KD Tree?**

- A KD (K-dimensional) tree is a data structure used for organizing and searching points in a multi-dimensional space.

- It's a binary tree that partitions space into half-spaces, useful for efficiently indexing k-dimensional data.

- **Applications:**

- Used in computer graphics, vision, data mining, spatial databases, and other fields.

Problem and Objectives

- **Problem Statement:**

- With the growth of digital data, efficiently retrieving relevant information has become challenging.
- Nearest Neighbor (NN) search is a simple, effective way to locate data points closest to a query in high-dimensional space.

- **Objectives:**

- Implement the KD Tree data structure.
- Perform nearest neighbor searches for efficient data retrieval.

Insertion and Search in KD Trees

- **Insertion:**

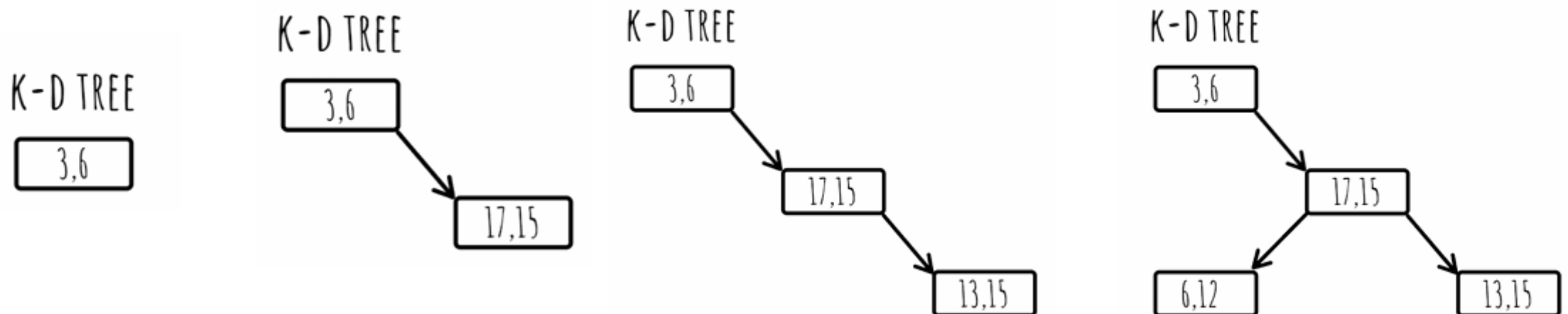
- The tree is partitioned based on alternating dimensions (e.g., x and y for a 2D KD Tree).
- Each node splits space based on its dimension's coordinate value, creating left and right subtrees.

- **Search:**

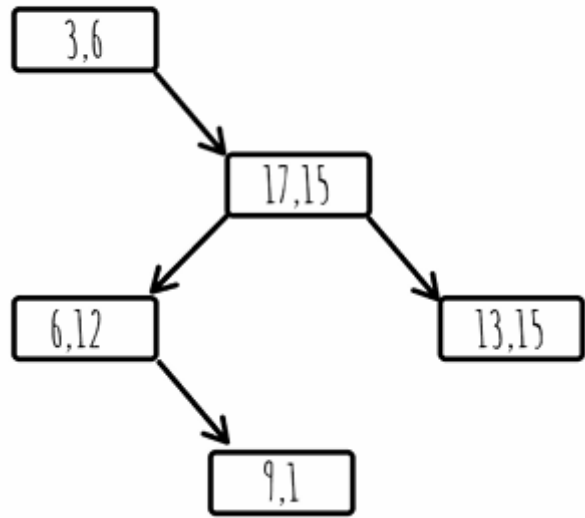
- Traverses similar to a binary tree, switching between dimensions at each level.
- Efficiently reduces search space by excluding large portions of irrelevant data points.

Consider an example of insert in the 2-D plane:

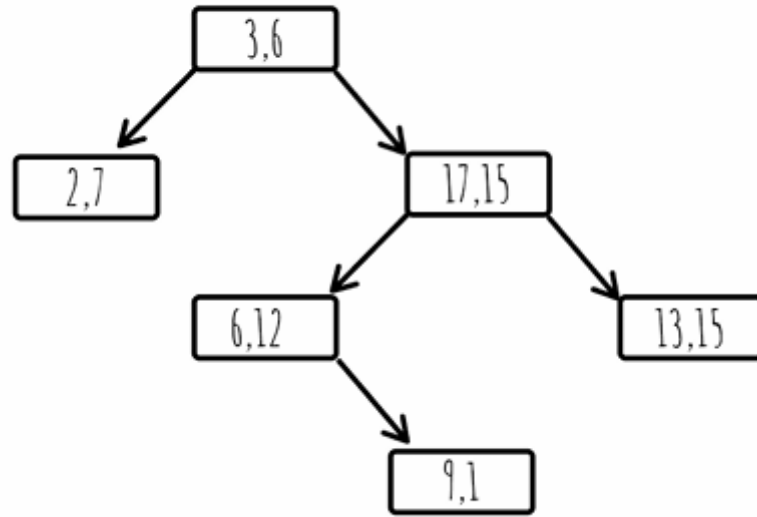
(3, 6), (17, 15), (13, 15), (6, 12), (9, 1), (2, 7), (10, 19)



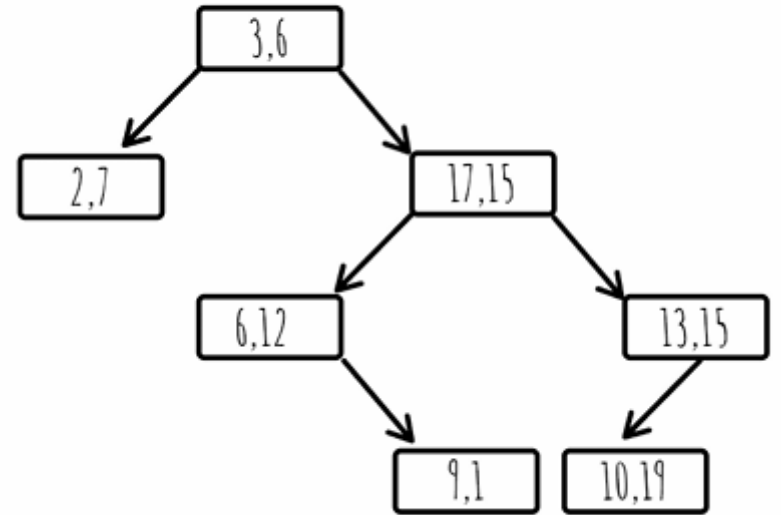
K-D TREE



K-D TREE



K-D TREE



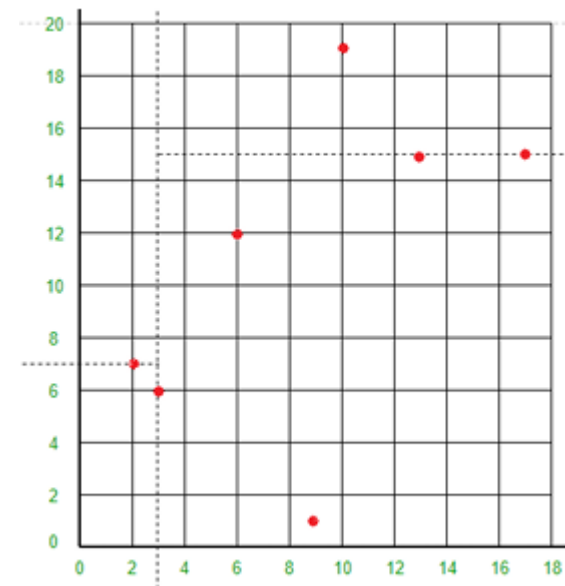
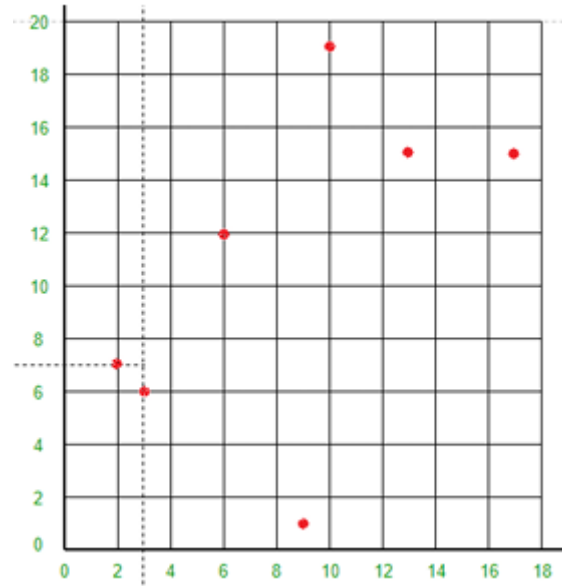
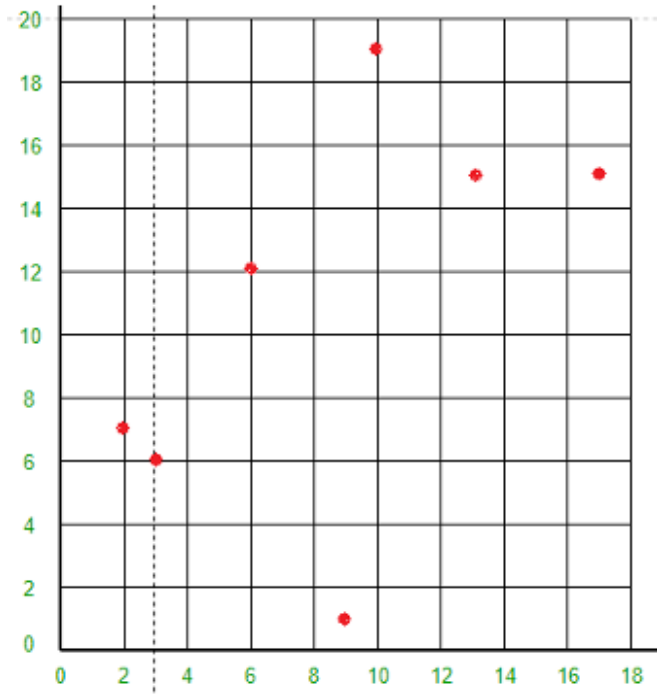
Insertion Pseudo code

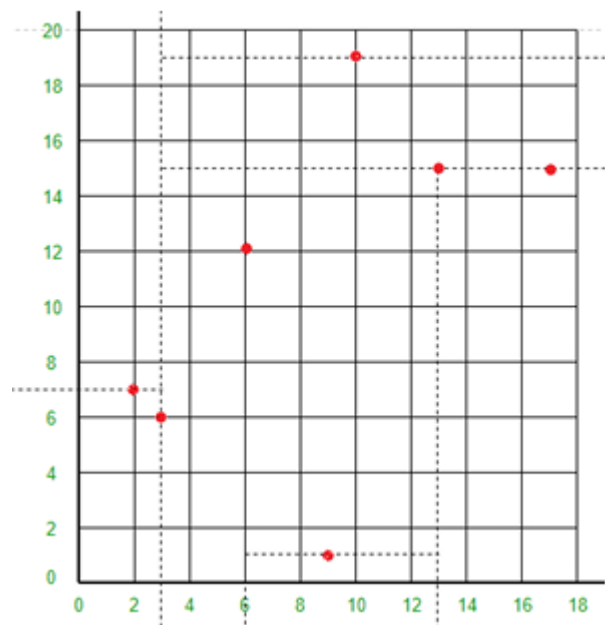
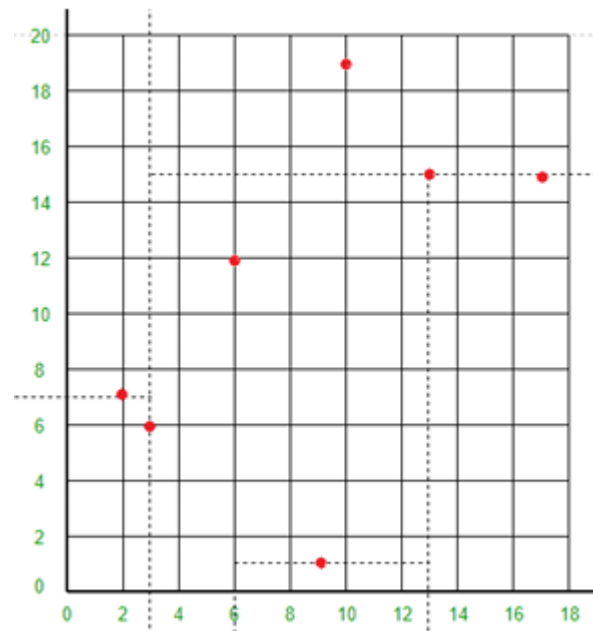
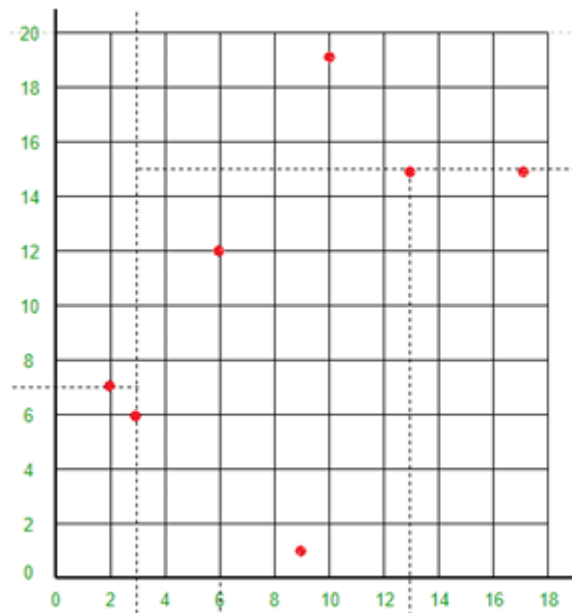
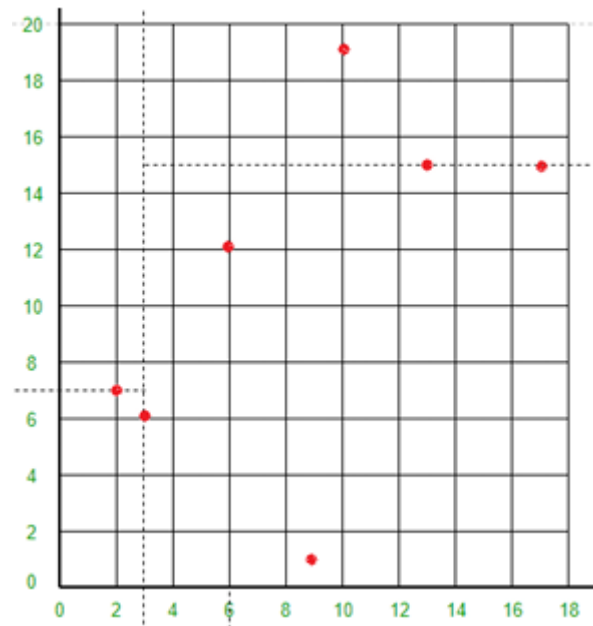
Algorithm 1 Insertion

```
1: node* Insert(node* x,int point[],int depth)
2: if  $x$  is NULL then
3:   temp=Allocate memory()
4:   for  $i = 0$  upto  $k - 1$  do
5:      $temp.point_i = point_i$ 
6:   end for
7:    $x = temp$ 
8:   return temp
9: end if
10: if  $point_{depth \% k} < x.point_{depth \% k}$  then
11:   return Insert( $x.left$ ,point,depth)
12: else
13:   return Insert( $x.right$ ,point,depth)
14: end if
15: return  $x$ 
```

It has an average of $O(\log n)$
time complexity and $O(n)$
space complexity.

Search





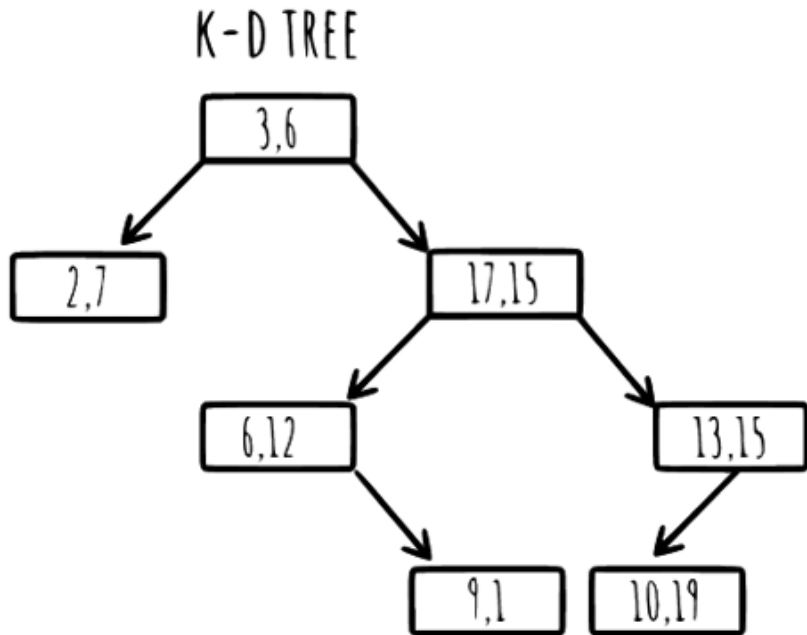
Search Pseudocode

Algorithm 2 Search

```
1: node* Search(node* x,int point[ ],int depth)
2: if  $x$  is NULL then
3:   return NULL
4: end if
5: if  $point_{depth \% k} = x.point_{depth \% k}$  then
6:   if  $point = x.point$  then
7:     return  $x$ 
8:   else if  $point_{(depth+1) \% k} < x.point_{(depth+1) \% k}$  then
9:     return Search( $x.left$ ,point,depth+1)
10:  else
11:    return Search( $x.right$ ,point,depth+1)
12:  end if
13: else if  $point_{depth \% k} < x.point_{depth \% k}$  then
14:   return Search( $x.left$ ,point,depth)
15: else
16:   return Search( $x.right$ ,point,depth)
17: end if
```

It has an average of $O(\log n)$ time complexity.

Traversal



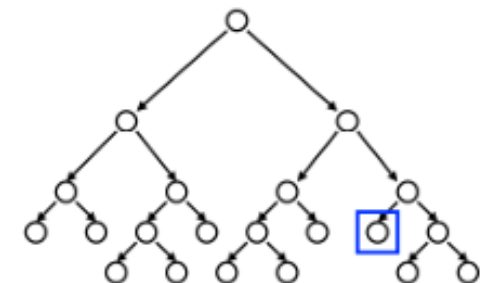
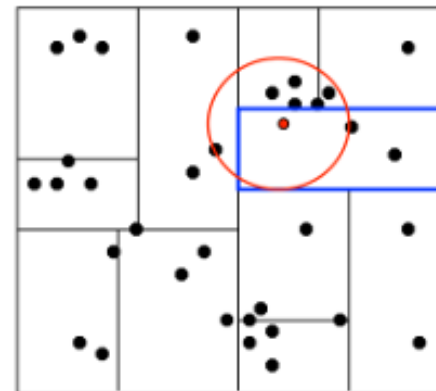
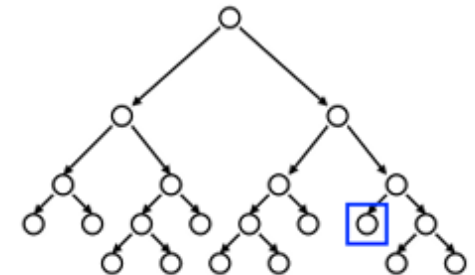
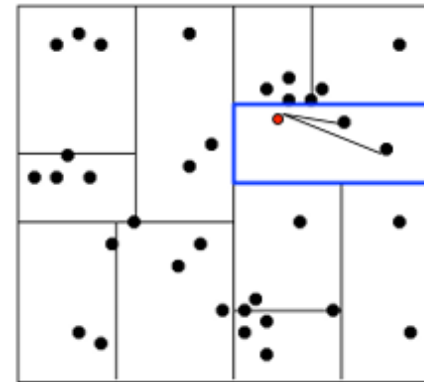
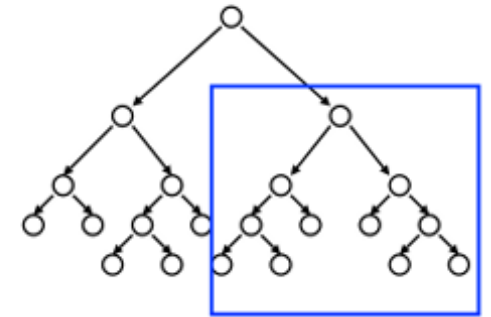
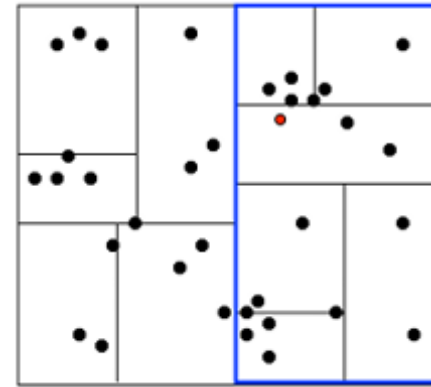
INORDER: (2,7),(3,6),(6,12),(9,1),(17,15),(10,19),(13,15)

Algorithm 3 Inorder Traversal

```
1: void Inorder(node* x)
2: if x is NULL then
3:   return NULL
4: end if
5: Inorder(x.left)
6: print x.point
7: Inorder(x.right)
```

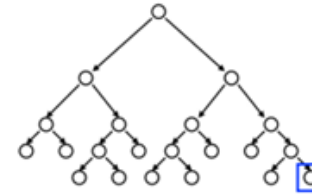
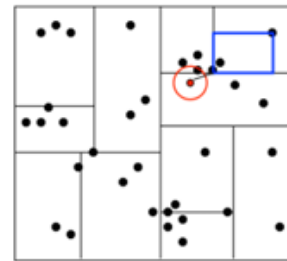
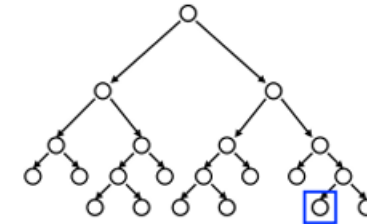
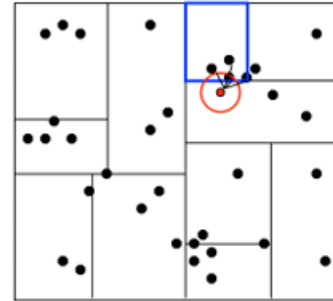
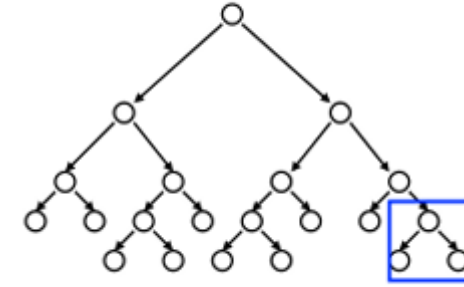
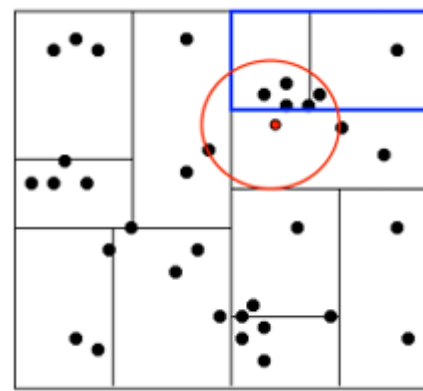
Nearest Neighbour

- 1. We traverse the tree looking for the nearest neighbor of the query point.
- 2. Explore the branch of the tree that is closest to the query point first.
- 3. When we reach a leaf node, compute the distance to each point in the node.

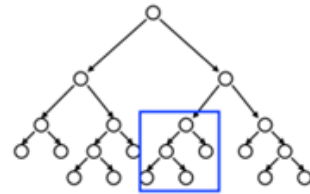
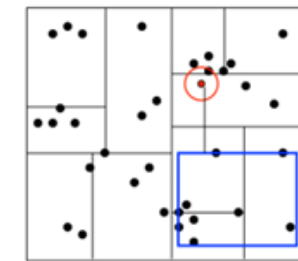


Nearest Neighbour

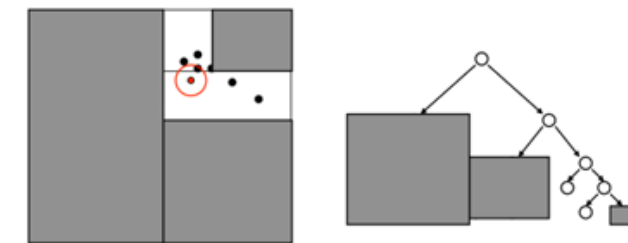
- 4. Then we can backtrack and try the other branch at each node visited.
- 5. Each time a new closest node is found, we can update the distance bounds.
- 6. Using the distance bounds and the bounds of the data below each node, we can prune parts of the tree that could NOT include the nearest neighbour.



(a)



(b)



(c)

Time Complexity Analysis

- **Insertion**

- Similar to a Binary Tree: Compares with root and inserts in left/right subtree.
- **Time Complexity:** $O(\log N)$ on average, where N is the number of nodes.

- **Search**

- Similar to insertion; traverses tree to locate point if it exists.
- **Time Complexity:** $O(\log N)$ on average.

- **Traversal**

- In-order traversal visits each node once.
- **Time Complexity:** $O(N)$.

- **Nearest Neighbor (NNS)**

- Traverses to leaf node, backtracks for pruning.
- **Complexity:** Ranges from $O(\log N)$ (best case, with pruning) to $O(N)$ (worst case, no pruning).

Test Case

