

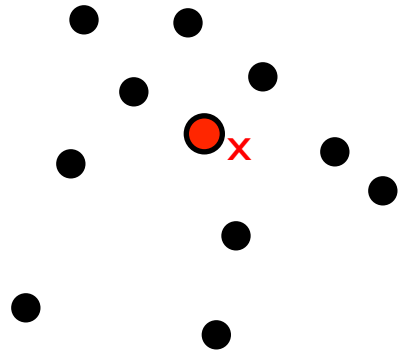
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# K-Dimensional Trees



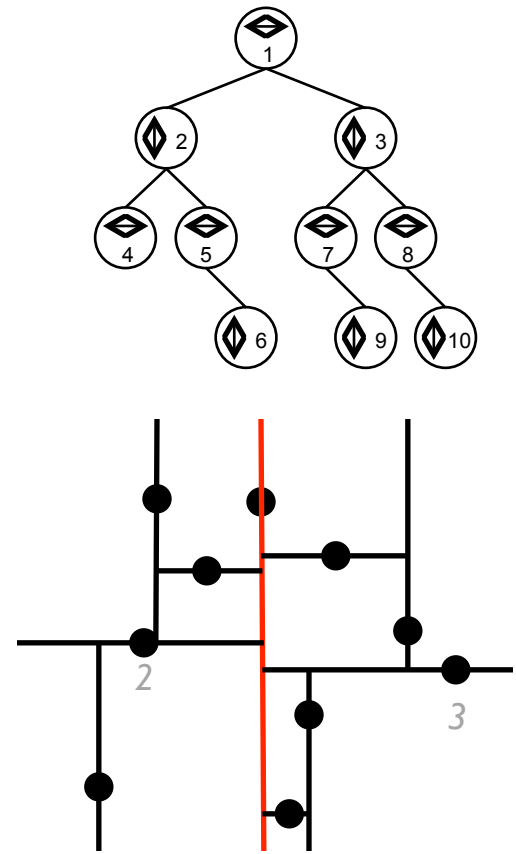
# Two-Dimensional Graphical Problem

- Given a collection of 2-dimensional points
  - Which point is closest to query point  $x$ ?
- Applications include
  - Evaluate coverage of fire stations
  - Big Data clustering analysis
- Obvious solution is  $O(n)$ 
  - Can we do better?



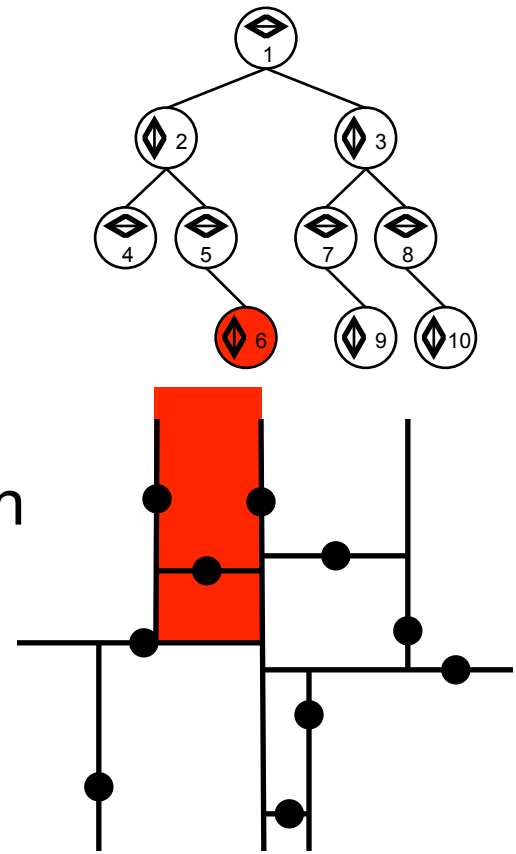
# Two-Dimensional Graphical Problem

- To achieve  $O(\log n)$  performance
  - Must find some way to divide points
  - Inspiration from Binary Search Tree
- KD-tree data structure
  - Nodes represents partitions of two-dimensional space
  - Easily extends to  $k$  dimensions



# Two-Dimensional Node Structure

- Partition is rectangular region
- Root node “covers” infinite region
  - Child node covers half of parent’s region
  - Alternating levels “flip” orientation
- Highlighted region associated with  $P_6$ 
  - Contains point
  - Prepared to subdivide further



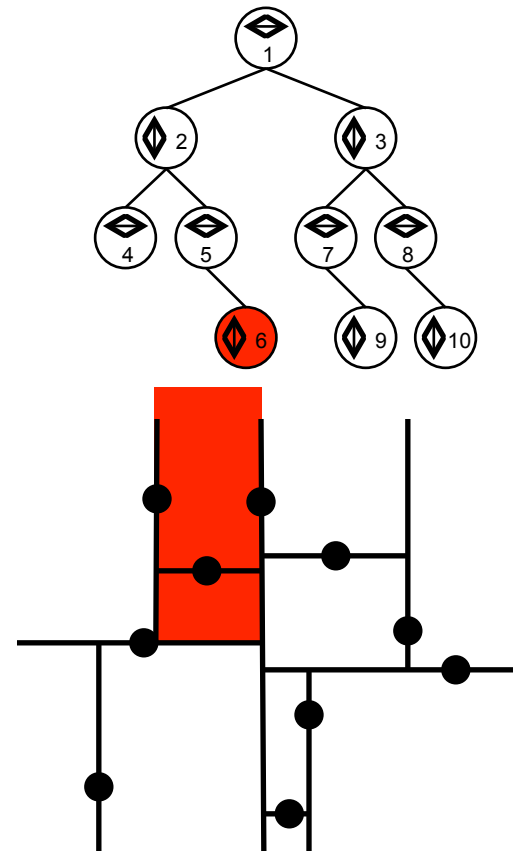
# Two-Dimensional Node Structure

- KD-tree has Binary Tree Structure
  - HORIZONTAL or VERTICAL orientation
  - Root arbitrarily VERTICAL

```
class KNode {  
    2D_Point    point  
    int         orient  
    Region      region  
    KNode       above  
    KNode       below  
}
```

```
class KDTree {  
    KNode    root  
}
```

```
class Region {  
    int xmin, ymin  
    int xmax, ymax  
}
```



# NEAREST NEIGHBOR Algorithm

Locate region where **X** would have been inserted

- Good place to start
- Confirm by traversing from root to see whether another point was actually closer

## NEAREST NEIGHBOR

Best	Average	Worst
$O(\log n)$	$O(\log n)$	$O(n)$



KD-tree

Recursion

```
def nearest (T, x)
```

```
    n = find parent node where x would have been inserted
```

```
    min = distance from x to n.point
```

```
    better = nearest (T.root, min, x)
```

```
    if (better found) { return better } else { return n }
```

```
def nearest (node, min, x)
```

```
    d = distance from x to node.point
```

```
    if (d < min) then
```

```
        min = d; result = node
```

```
    if “too close to call” then
```

```
        result = closer of nearest (node.above, min, x) and nearest (node.below, min, x)
```

```
    else
```

```
        if (node is above x) then
```

```
            pt = nearest (node.above, min, x)
```

```
        else
```

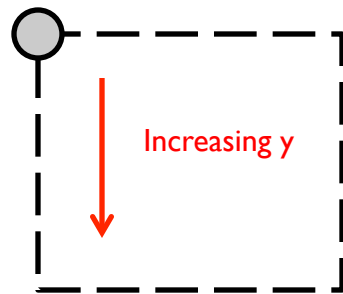
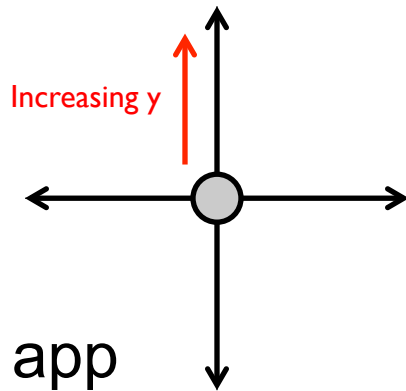
```
            pt = nearest (node.below, min, x)
```

```
        if (pt exists) then return pt
```

```
    return result
```

# Applications with KD-Trees

- Understand graphical coordinate systems
  - Cartesian coordinates different from computer coordinates
  - Origin in different location
  - Concept of up and down is different
- Code must reflect this
  - Let's write nearest neighbor app





# KD-Tree Summary

- Derives efficiency from recursion inspired by Binary Trees
  - Other applications abound
  - “What points are contained in query rectangle?”
- Binary Space Partitioning structures
  - Foundation for efficient 3D-graphics