



Data Structures:

Trees: Spatial Trees, Tries

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(Lecture by Youngmin Oh)

Spring 2022



Characterizing Trees



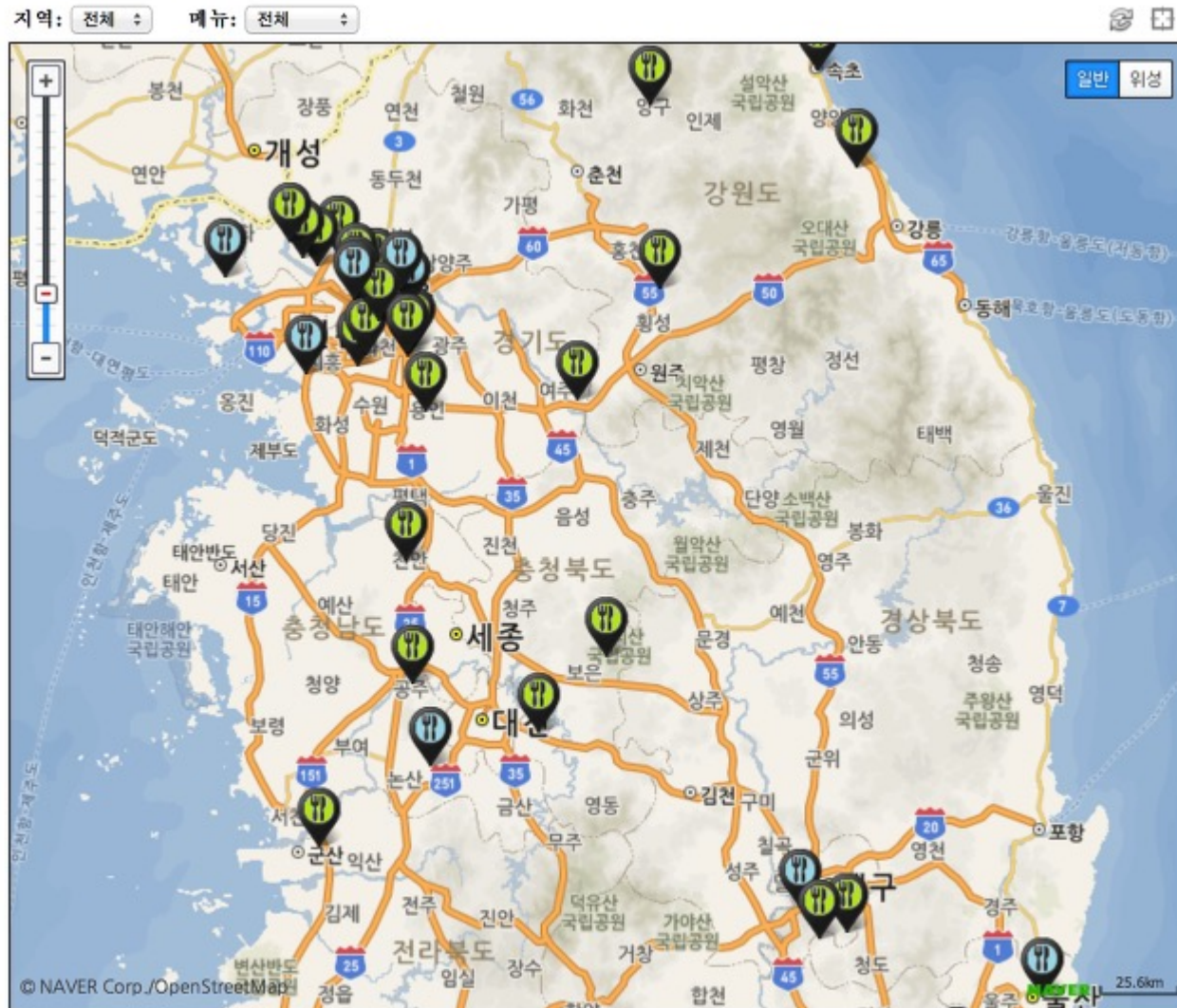
Tree Properties

- Degrees of Trees
 - unary tree
 - linked list
 - binary tree
 - binary search tree
 - m-way tree ($m > 2$)
- Height Balanced
 - unbalanced
 - balanced (but not perfectly)
 - AVL tree, T-tree
 - perfectly balanced
 - 2-3 tree
- Dimension
 - one dimension
 - AVL tree, T-tree
 - n-dimension ($n \geq 2$)
 - quad tree, kd tree
- Interior node has no value
 - trie, region quadtree



Spatial Data Structures

Storing and Searching for Popular Restaurants on a Map



- * 자료출처: 이영돈PD의 먹거리X파일
- * 사용법: 표시클릭 - 상세정보 보기, 더블클릭 - 확대.
- * 데이터 날짜: 2013.9.27.
- * 댓글 남기기: 바로가기



Spatial Data Structures

- QuadTree, OctTree
- k-d Tree, Grid File

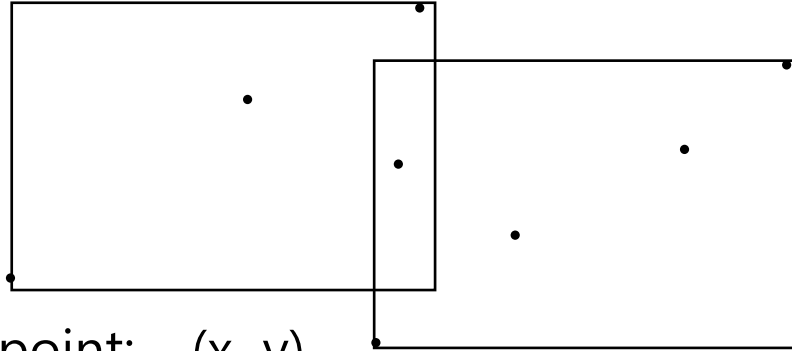


Spatial vs. Non-Spatial Data Structures

- Non-Spatial Data Structures
 - Indexing 1-dimensional data
 - Key value is a single data item
 - e.g., 250, "Hong Gil Dong",...
- Spatial Data Structures
 - Indexing multi-dimensional data
 - Key value is a set of data items
 - E.g., (75, 127), (15, 46, 93), ("Hong Gil Dong", 25, thief, "running on the roof")



Point and Bounding Rectangle



- point: (x, y)
- bounding rectangle (BR): $((x1, y1), (x2, y2))$
- Relationship between a BR and a point
 - A point is contained in the BR
 - A point lies on an edge of the BR
 - Distance of the point from the center of the BR
- Relationship between two BRs
 - One BR contains the other BR
 - The two BRs touch each other
 - Distance between the BRs

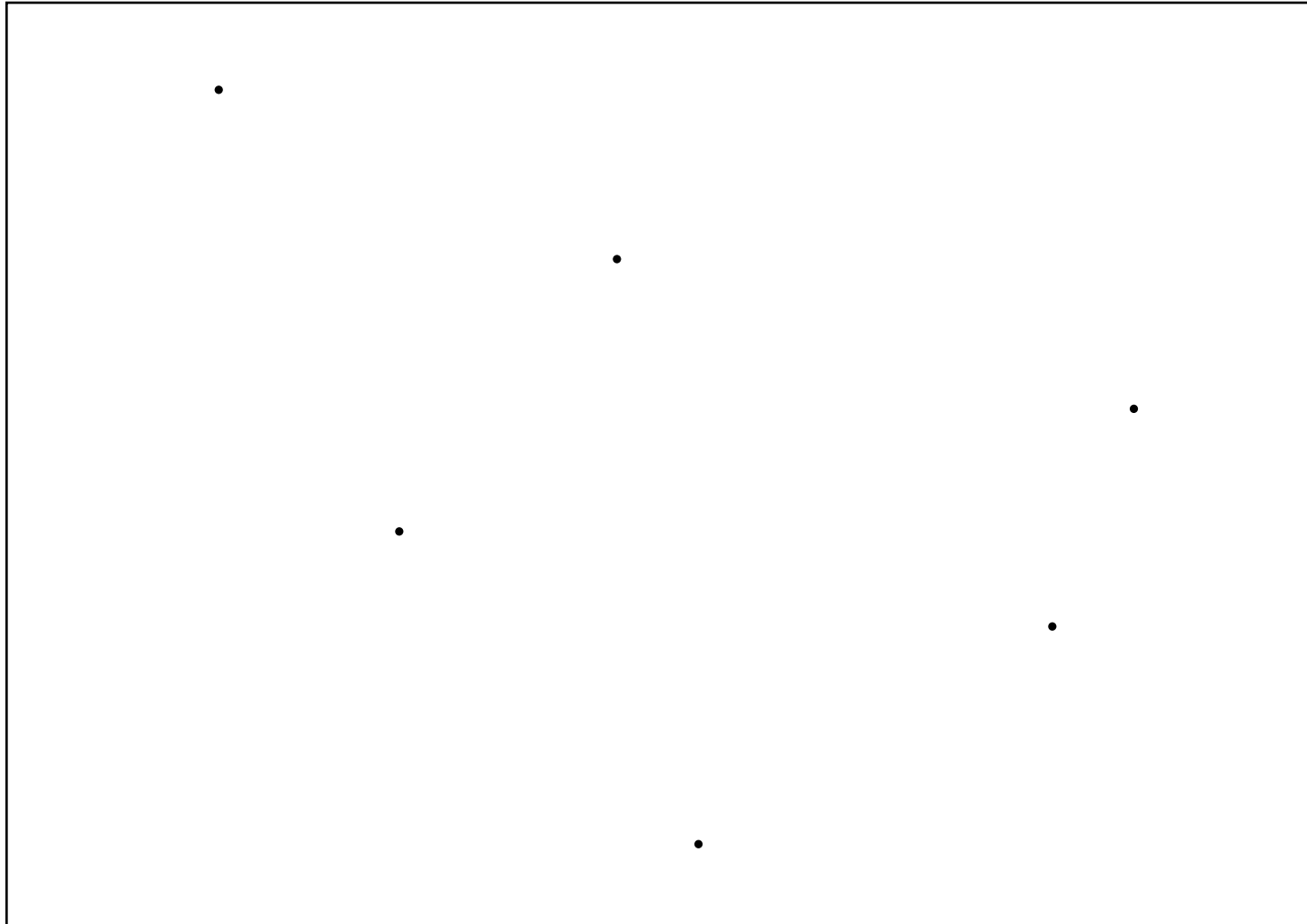


(Region) QuadTree

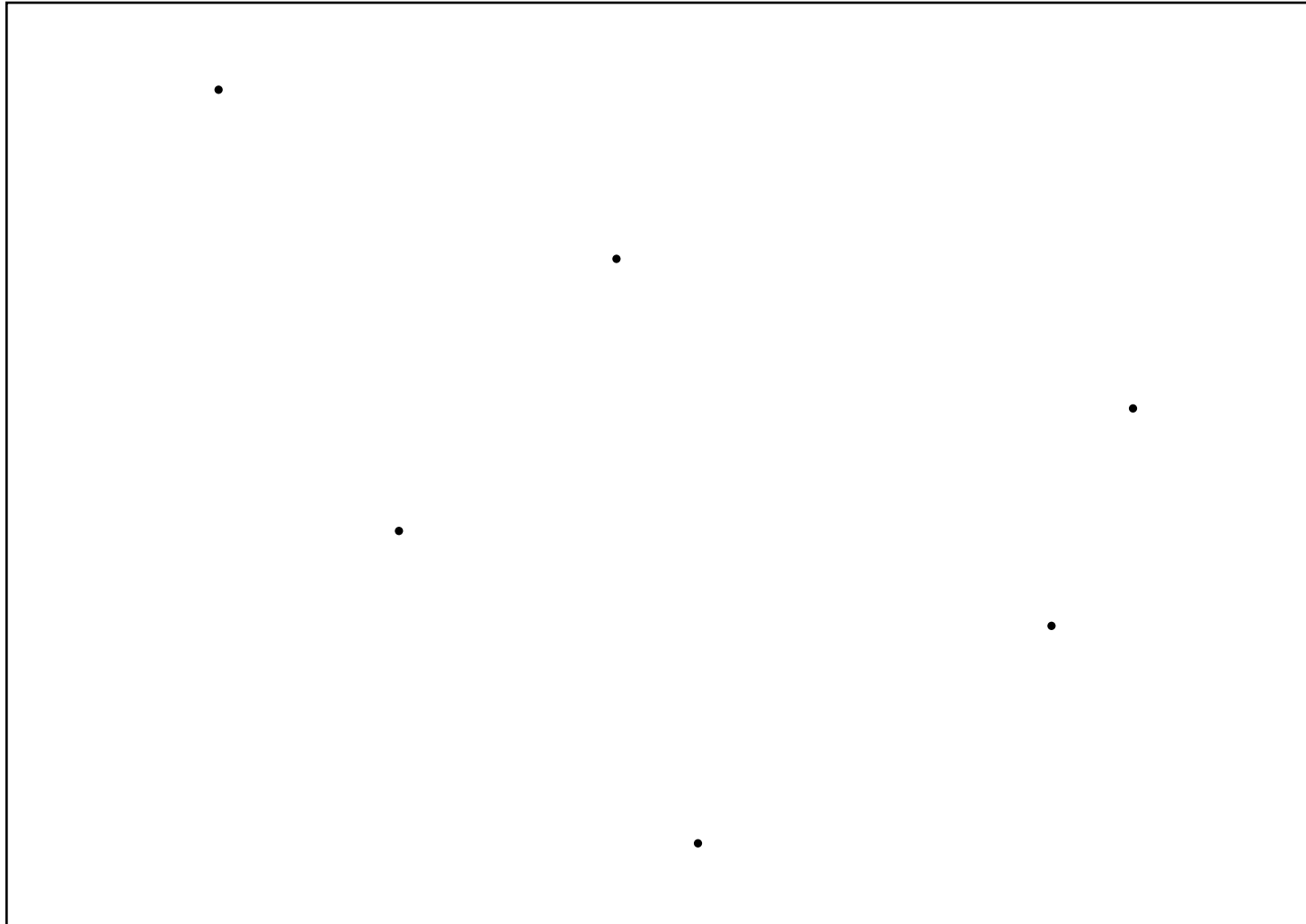
- Tree of Degree 4
- Performance: $O(\log_4 n)$
- Not Balanced
- Divides a 2-Dimensional Space into 4 Quadrants (Regions)
- Reading
 - <http://www.cs.ubc.ca/~pcarbo/cs251/welcome.html>

Example: Find Restaurants on the Map of a City

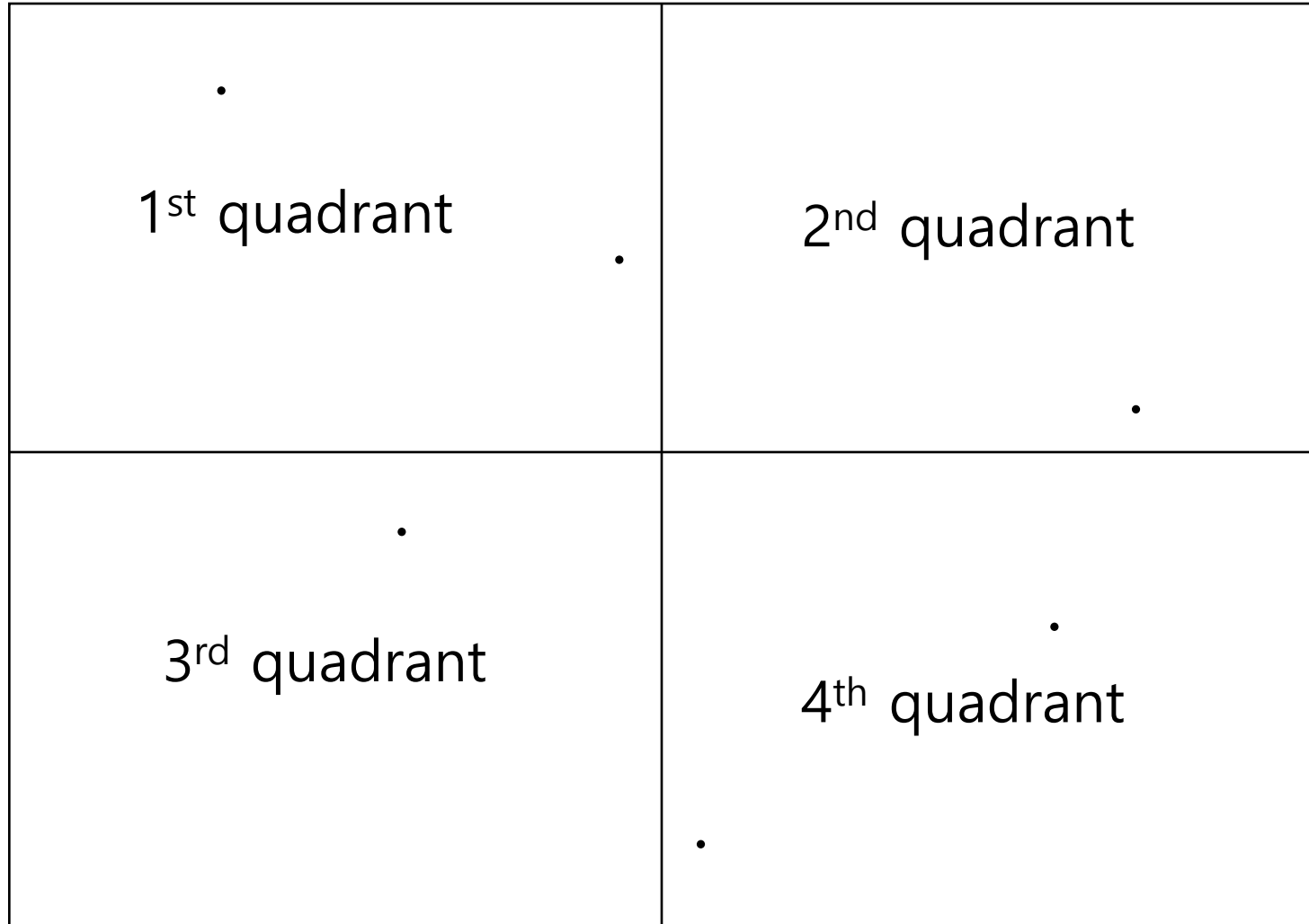
(Each Point Represents All Data About a Restaurant.)



Divide a 2-D Space into Successive Quadrants



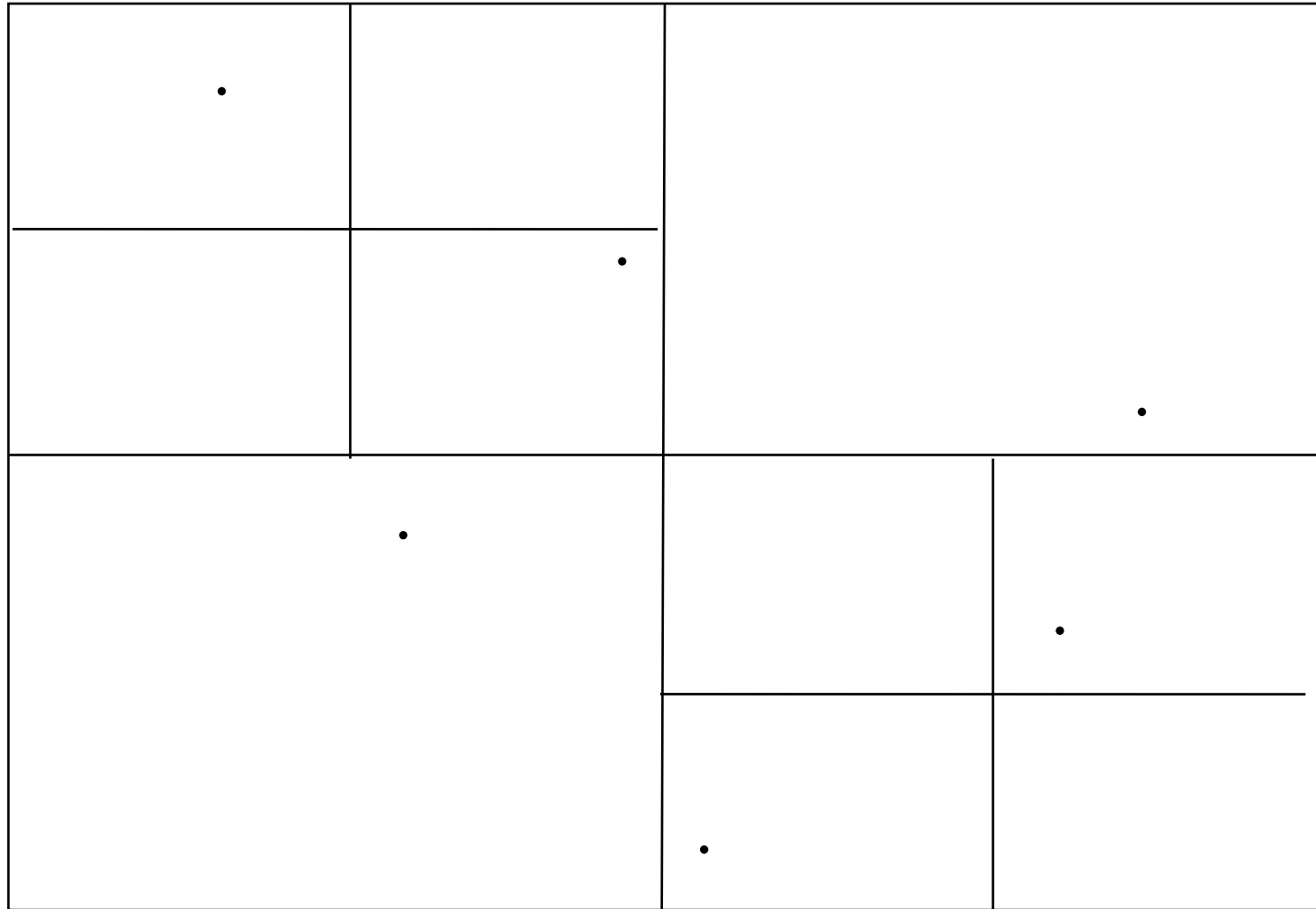
Divide a 2-D Space into Successive Quadrants



Divide a 2-D Space Into Successive

Quadrants

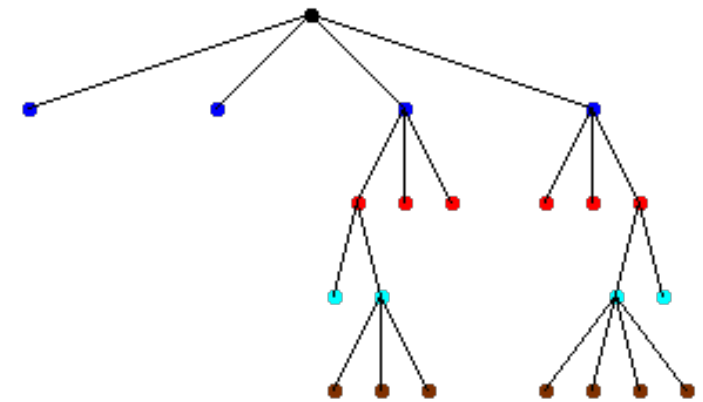
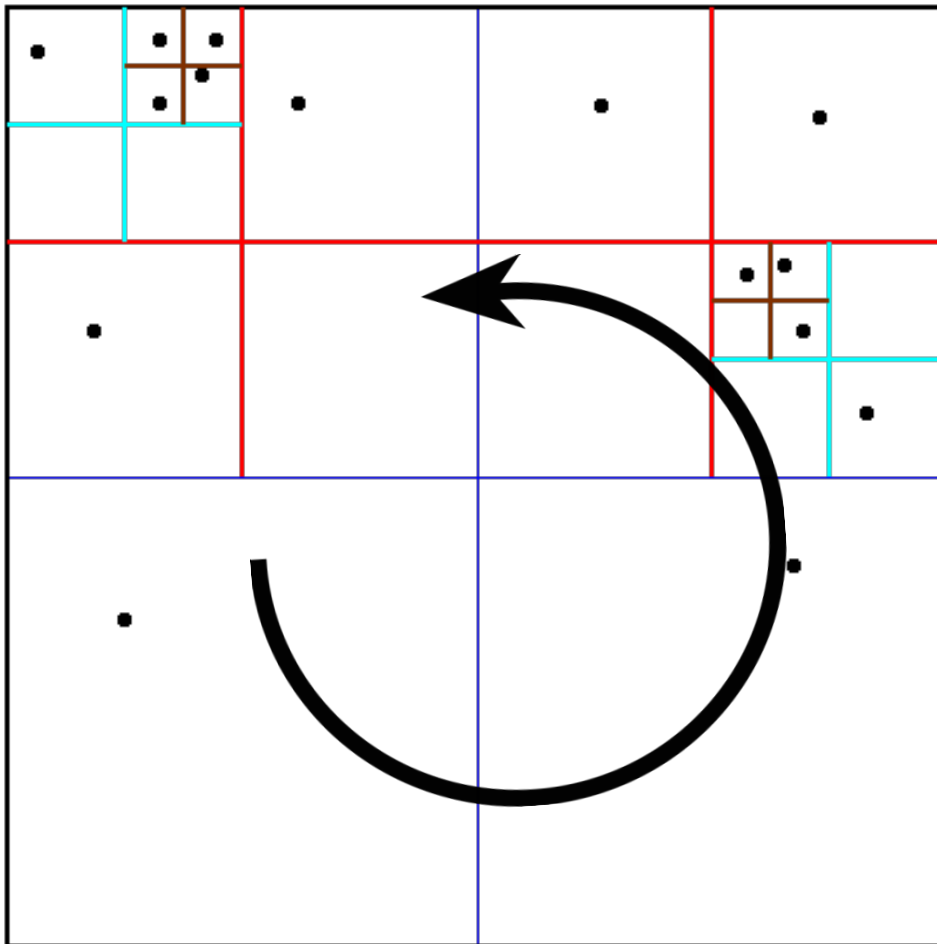
(until each quadrant has only one point in it)



Map the Quadrants to Nodes of a QuadTree:

specify the starting quadrant and direction of mapping

Adaptive quadtree where no square contains more than 1 particle



<counter-clockwise
from the 3rd quadrant>



Region QuadTree Nodes

- Region quadtree is a type of trie (to learn later).
- Interior nodes contain pointers to child nodes, but no data.
- Data are stored on the leaf nodes.



Leaf Node

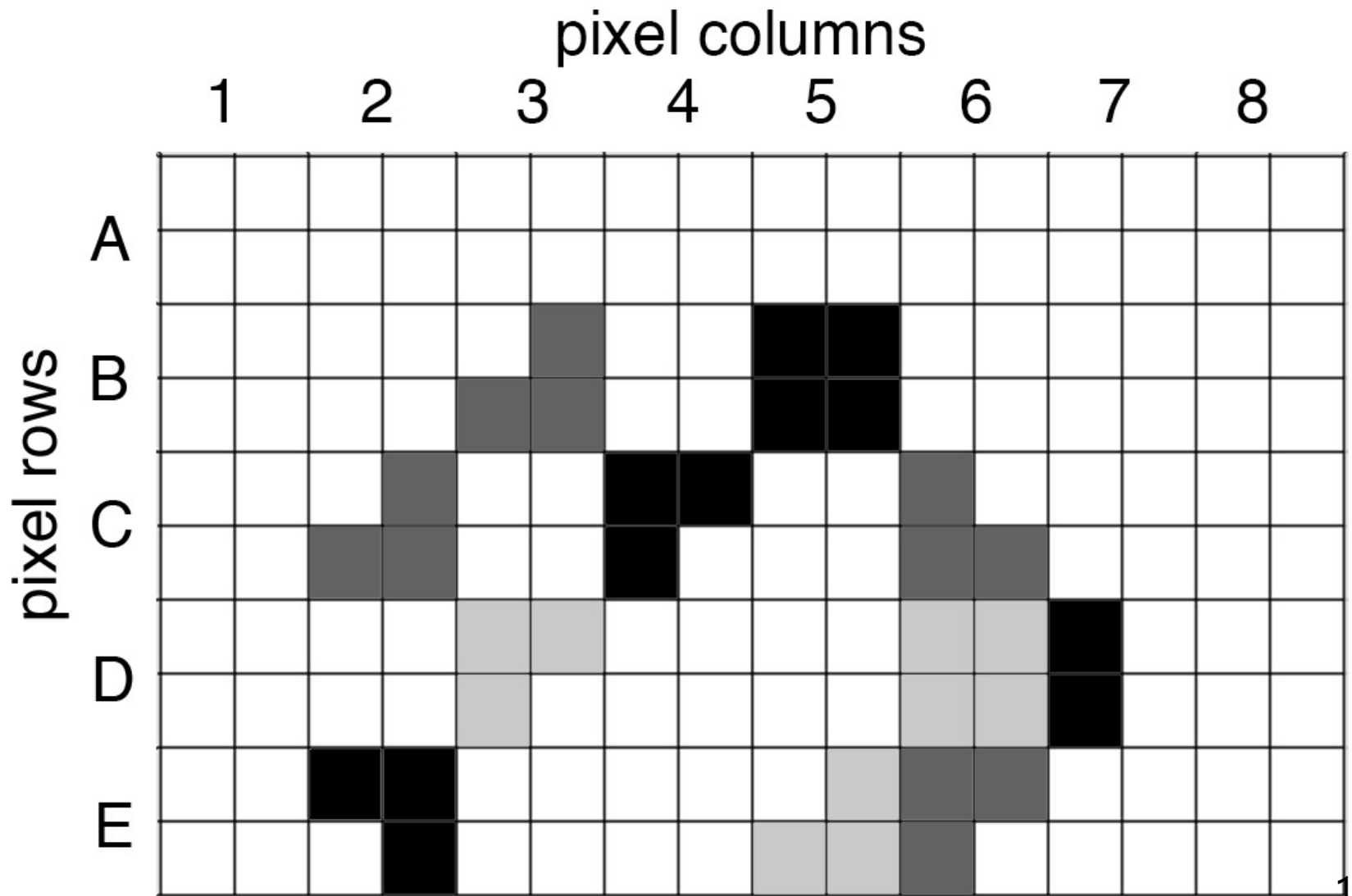
- Each leaf node contains data corresponding to a specific subregion, for example,
 - point, number, string, pointer to a file,...
 - line, shape, pixels, image elements,...
 - the latitude and longitude of a set of cities
 - the average temperature over the subregion it represents



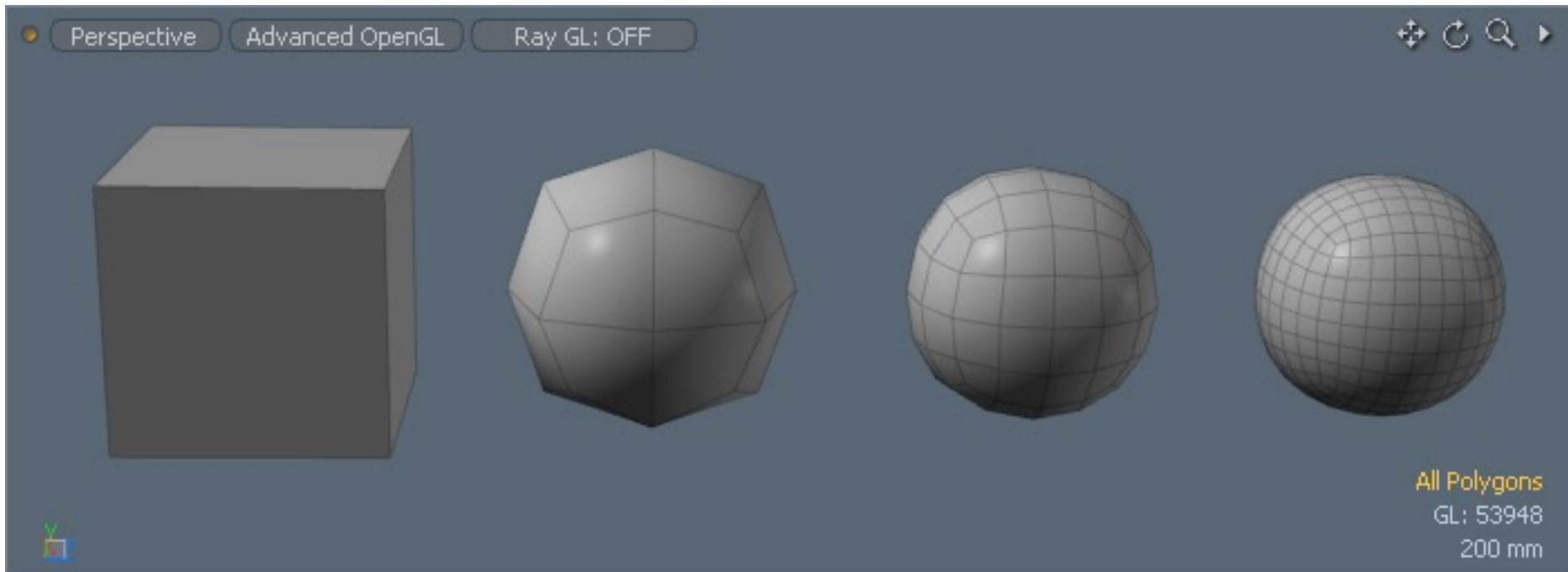
Applications

- software that searches for segments of stored color images
- software that searches for information about restaurants (schools, churches, car centers,...) within a region of a city (province, country)
- software that searches for parts of geometry objects
- supporting collision detection (intersection of two solid objects) in video games, physical simulations, computational geometry

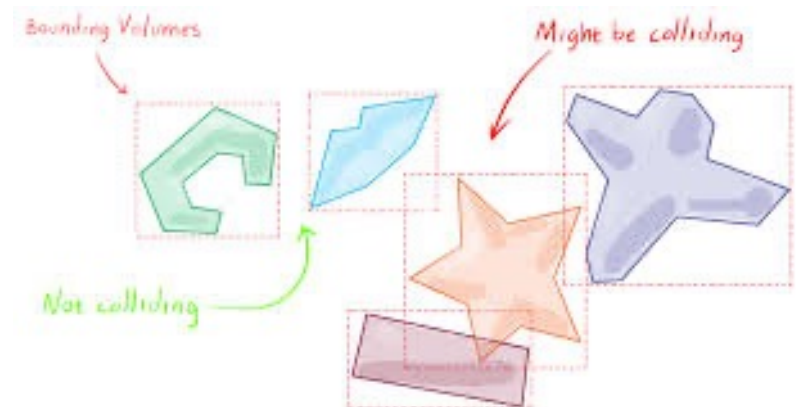
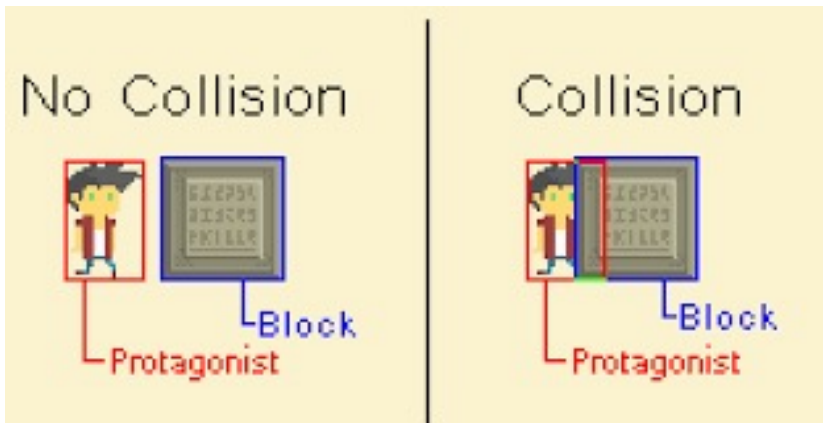
Image = Collection of Pixels (Picture Elements)



Subdividing Geometry Objects



Collision Detection (e.g., in Video Games)



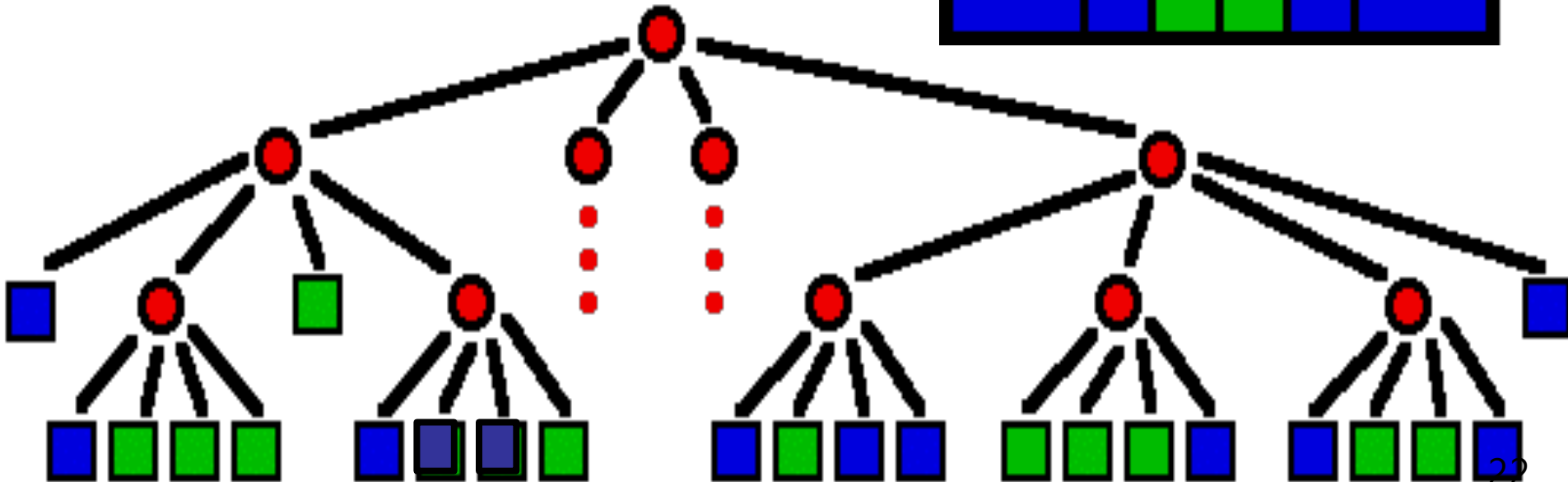
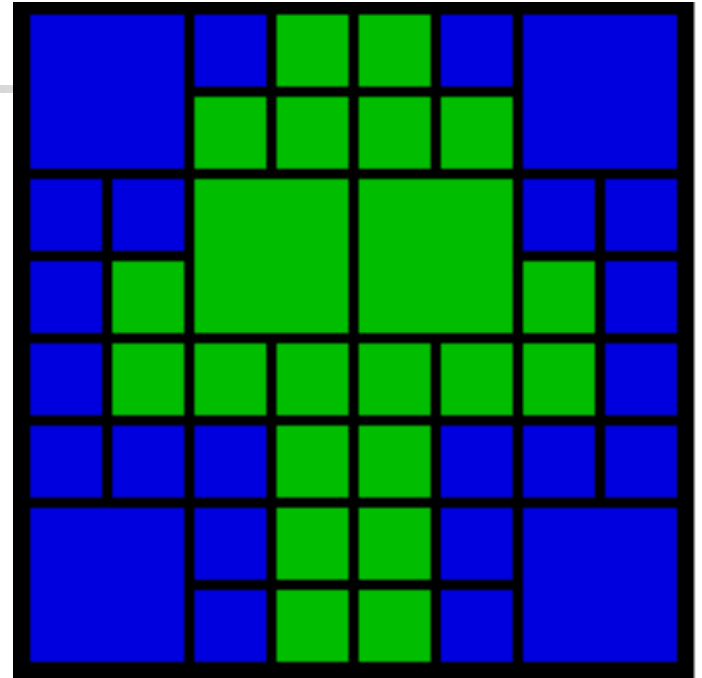


Applications: Representing an Image

- A region quadtree with a depth of n may represent an image consisting of $2^n \times 2^n$ pixels, where each pixel value is 0 or 1.
- The root node represents the entire image region. If the pixels in any region are not entirely 0s or 1s, it is subdivided.
- Each leaf node represents a block of pixels that are all 0s or all 1s.

A QuadTree for Storing 2-D Image Elements

<clockwise from the 1st quadrant>

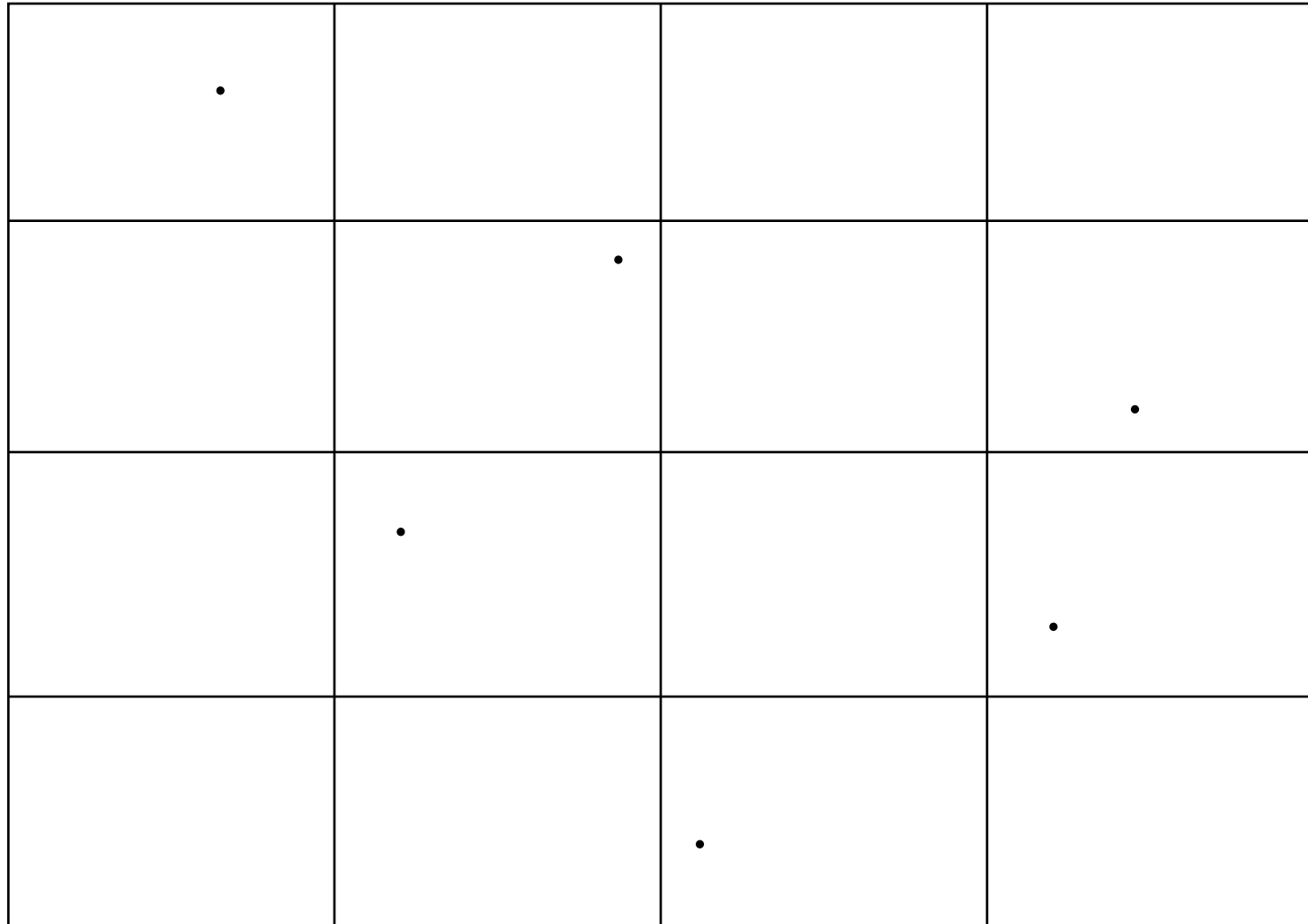




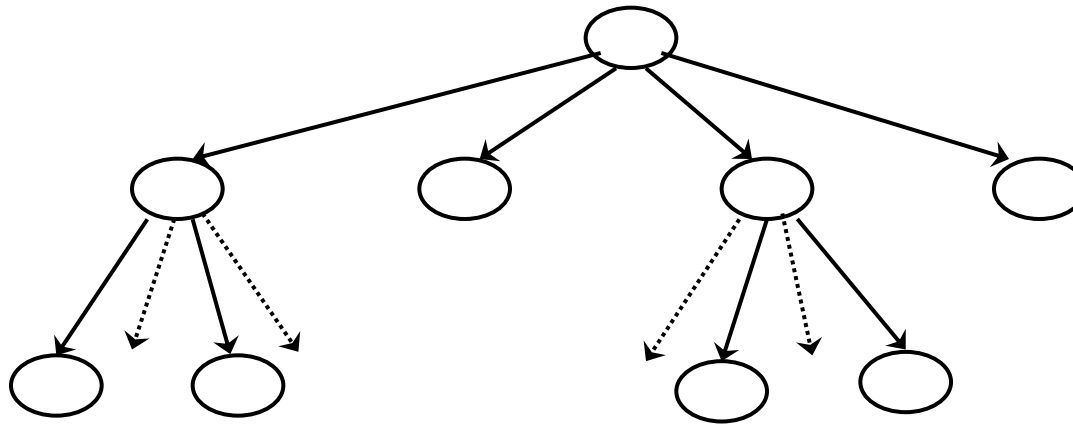
Storing and Accessing 2-D Image Elements

- Recursive subdividing of a quadrant
 - until each sub-quadrant has only one color
- Fast access to any region of an image at any level
 - any quadrant at any level of the tree
- Controlling the resolution of an image
 - raising or lowering the level of the tree

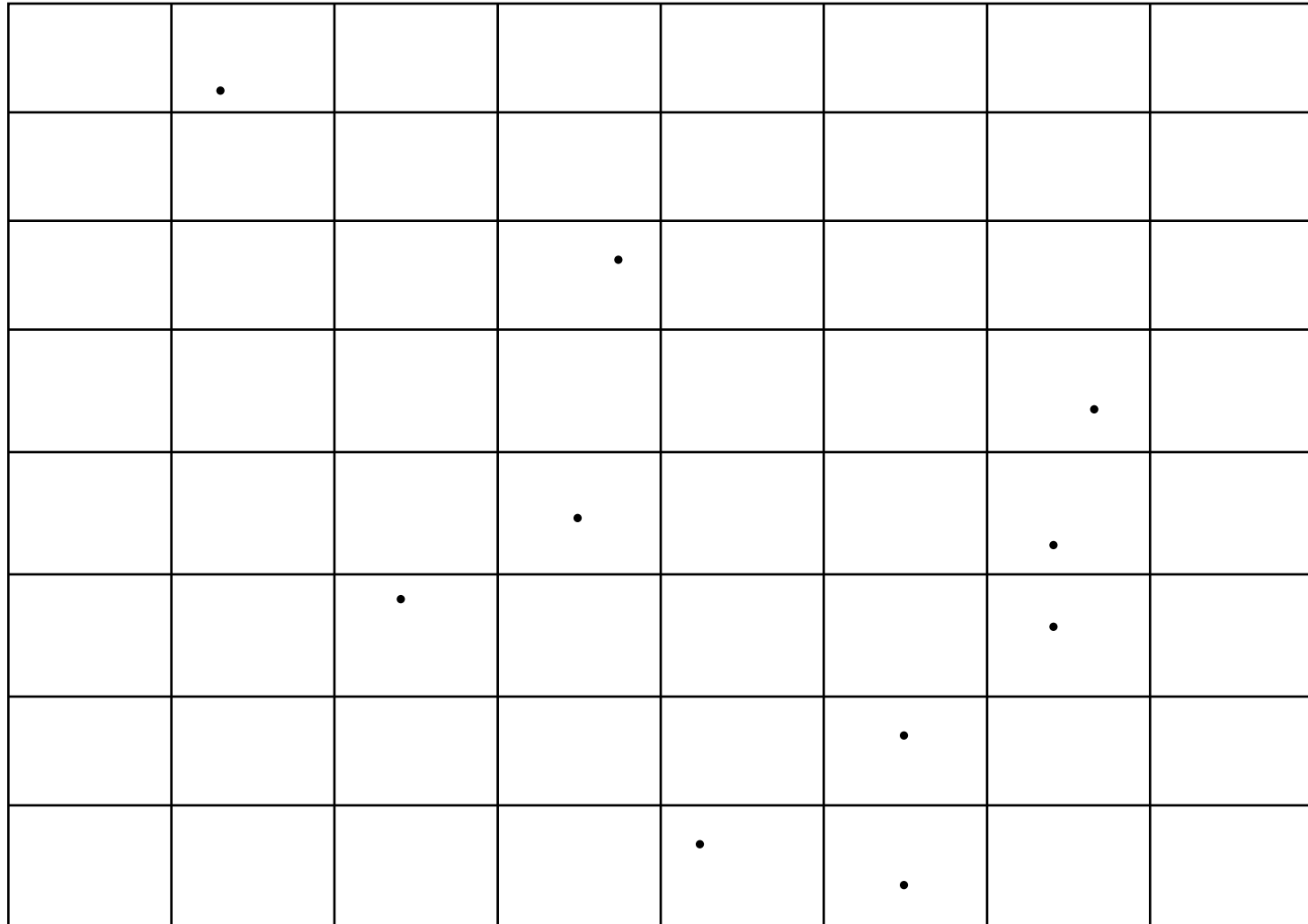
Exercise: Draw a Quad-Tree for the
Following Point Data (clockwise from the 1st
quadrant)



Exercise: Result

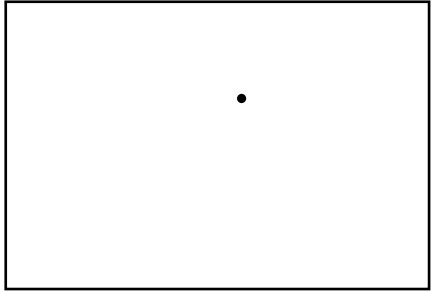


Exercise: Draw a Quad-Tree for the
Following Point Data (clockwise from the 1st
quadrant)





How to Determine Which Subtree to Go To?



- point: (x, y)
- bounding rectangle (BR): $((x1, y1), (x2, y2))$
- Each quadtree node has a corresponding BR.
- At each node, determine which of the child nodes (rectangles) can contain the search point or search rectangle.
- (The rectangle BR data may be stored in each non-leaf node or may be computed.)

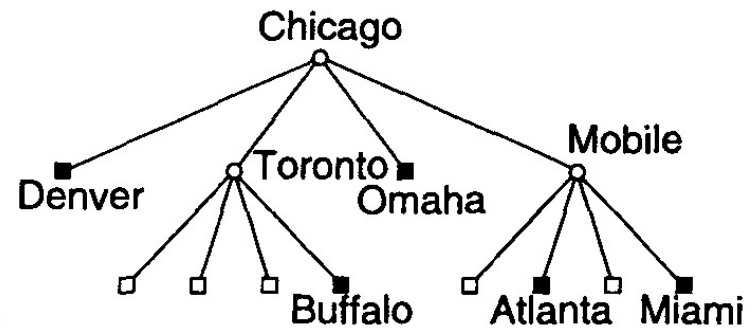
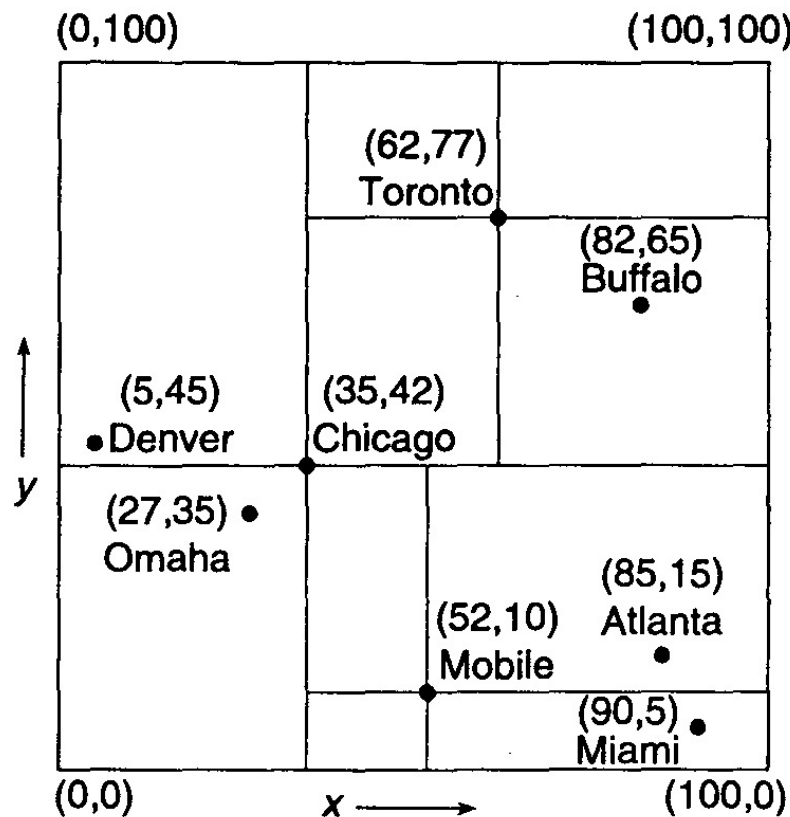


Point QuadTree

- Adaptation of a binary tree to represent 2-dimensional point data
- Each non-leaf node has 4 child nodes.
- Each node contains
 - 4 pointers (NW, NE, SW, SE)
 - key (x, y coordinates)
 - Value
- ** k-d tree is better (to be covered shortly)

Point Quadtree: Example

How do we determine the root node? (later..)



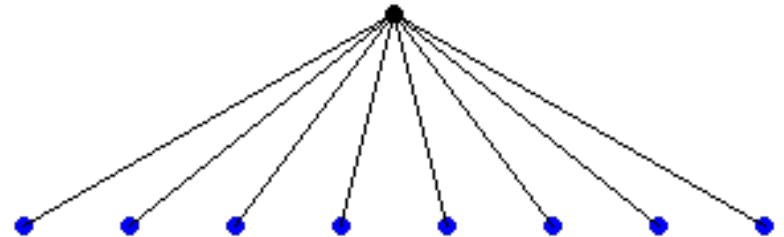
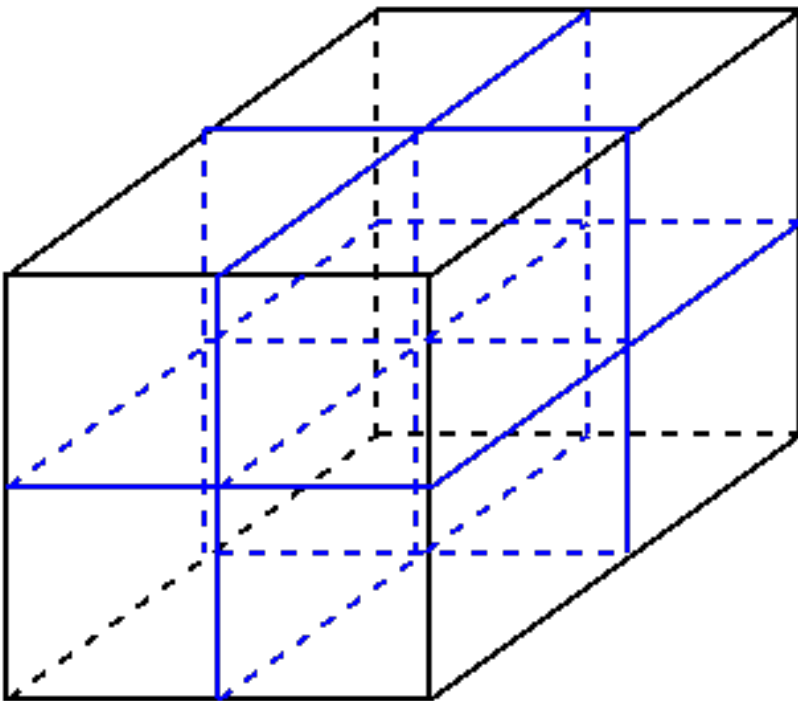


OctTree

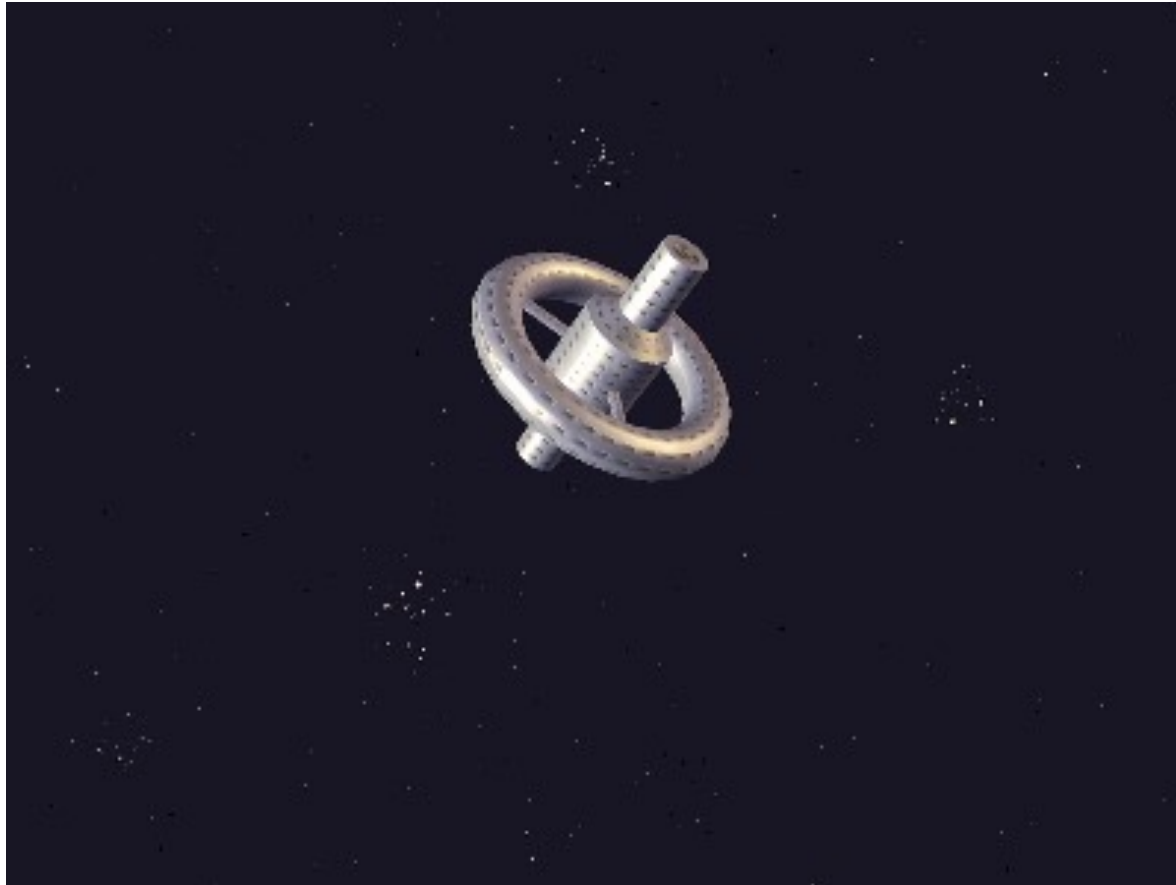
- Tree of Degree 8
- Divides a 3-dimensional space into 8 sub-cubes
- Performance: $O(\log_8 n)$
- Leaf nodes can store pointers to 3-D geometry objects, 3-D image elements
- Reading
 - <http://www.cs.berkeley.edu/~demmel/cs267/lecture26/lecture26.html>

OctTree

2 Levels of an Octree



Storing and Accessing 3-D Image Elements





K-Dimensional Trees



k-Dimensional Space

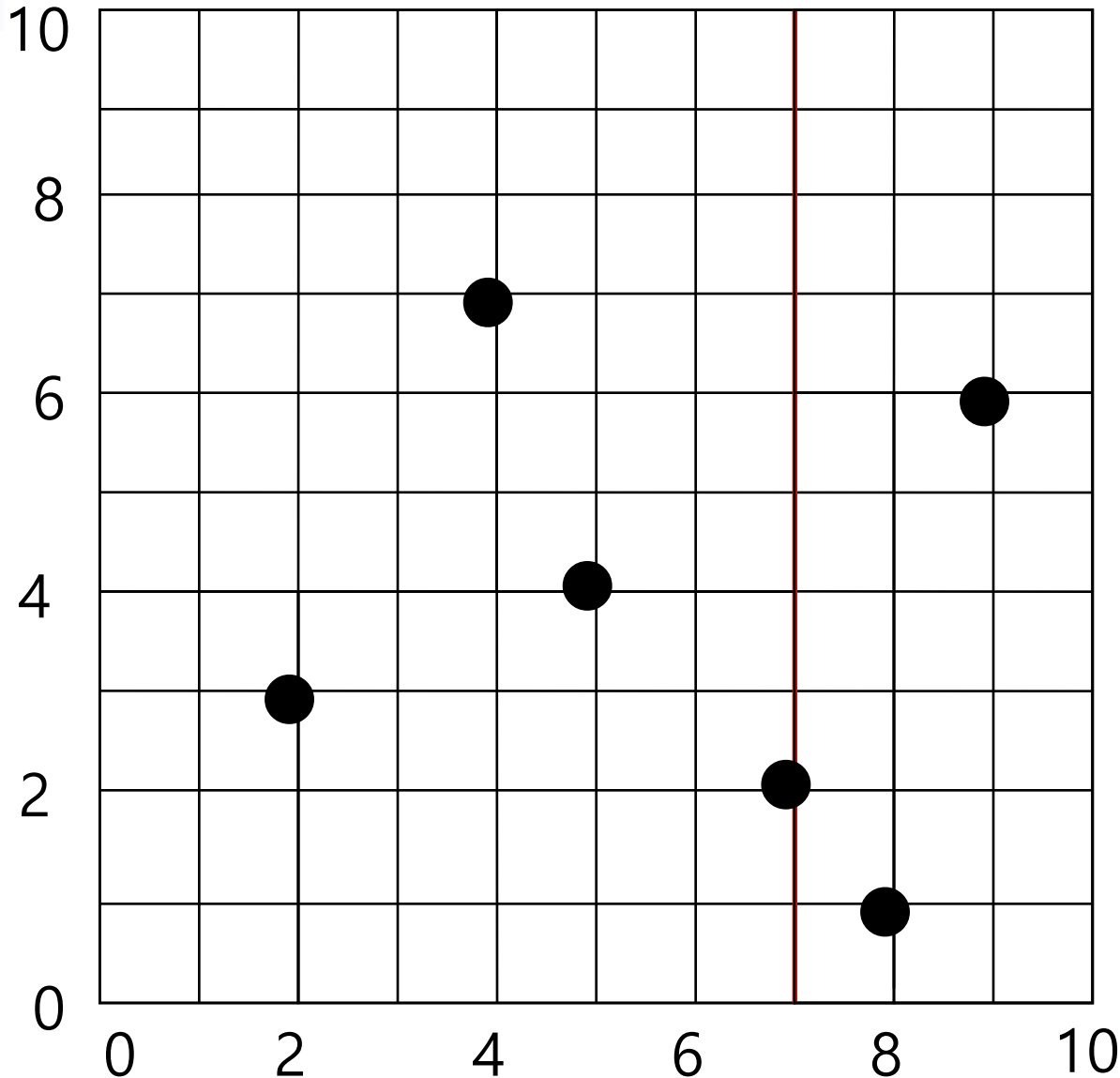
- 2-d
- 3-d
- 4-d
 - 3-d plus time
 - Person's (gender, age, education, ethnic origin)
- 5-d, 6-d, 7-d,...
 - Person's (gender, age, education, ethnic origin, religion, marital status, political party,...)



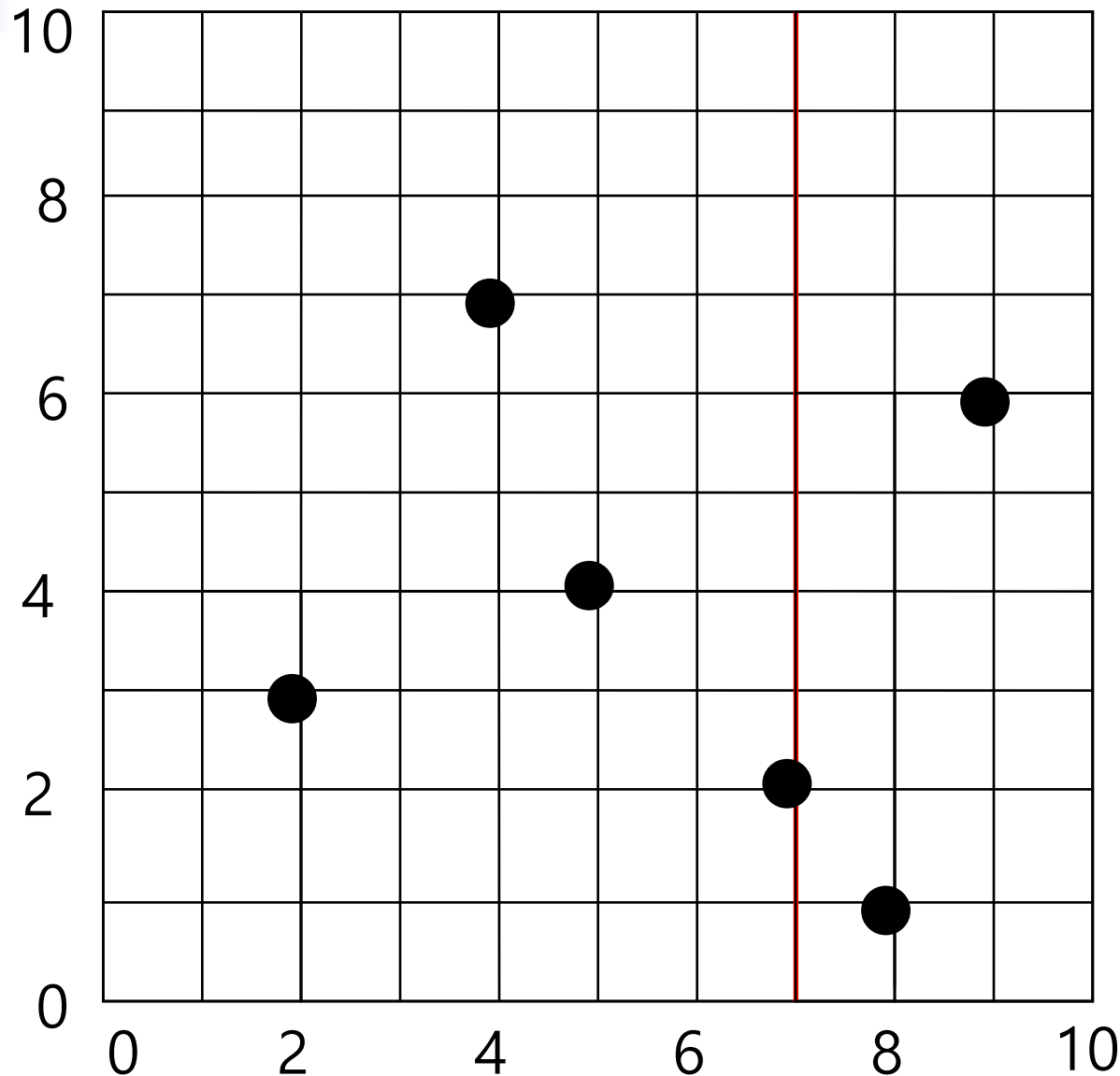
k-d Tree

- Store and search points in a k-dimensional space
 - $k \geq 2$
- Point data ($d_1, d_2, d_3, d_4, \dots$)
 - ex. (5, 4), (70, 55, 37)
- Reading
 - <http://www.answers.com/topic/kd-tree>
 - <http://www2.toki.or.id/book/AlgDesignManual/BOOK/BOOK3/NODE134.HTM>
 - <http://cis.poly.edu/~hakcan01/projects/kdtree/kdTree.html>

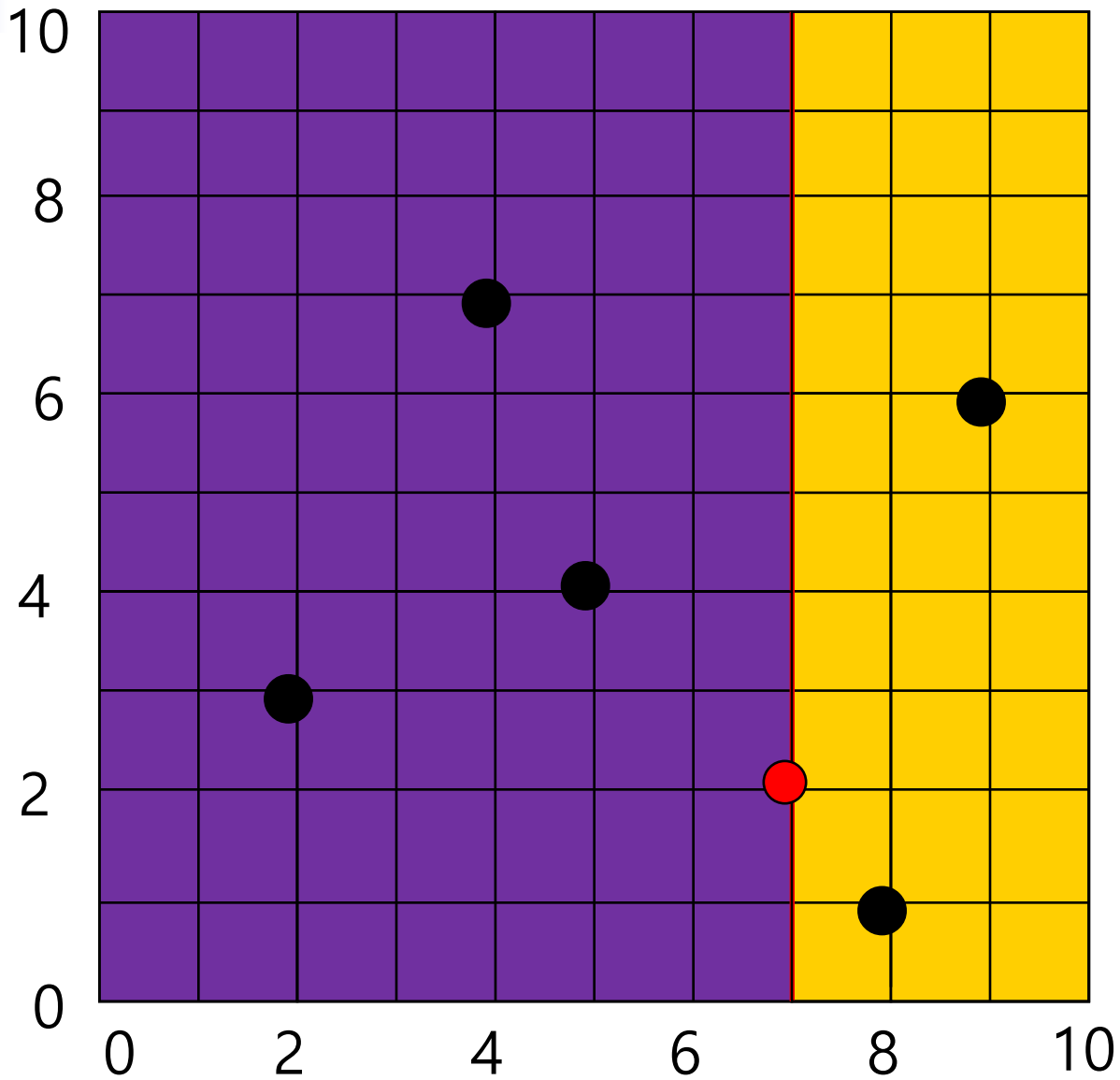
Successively Partition (Divide) a k-D Space Until No Partition Contains a Point



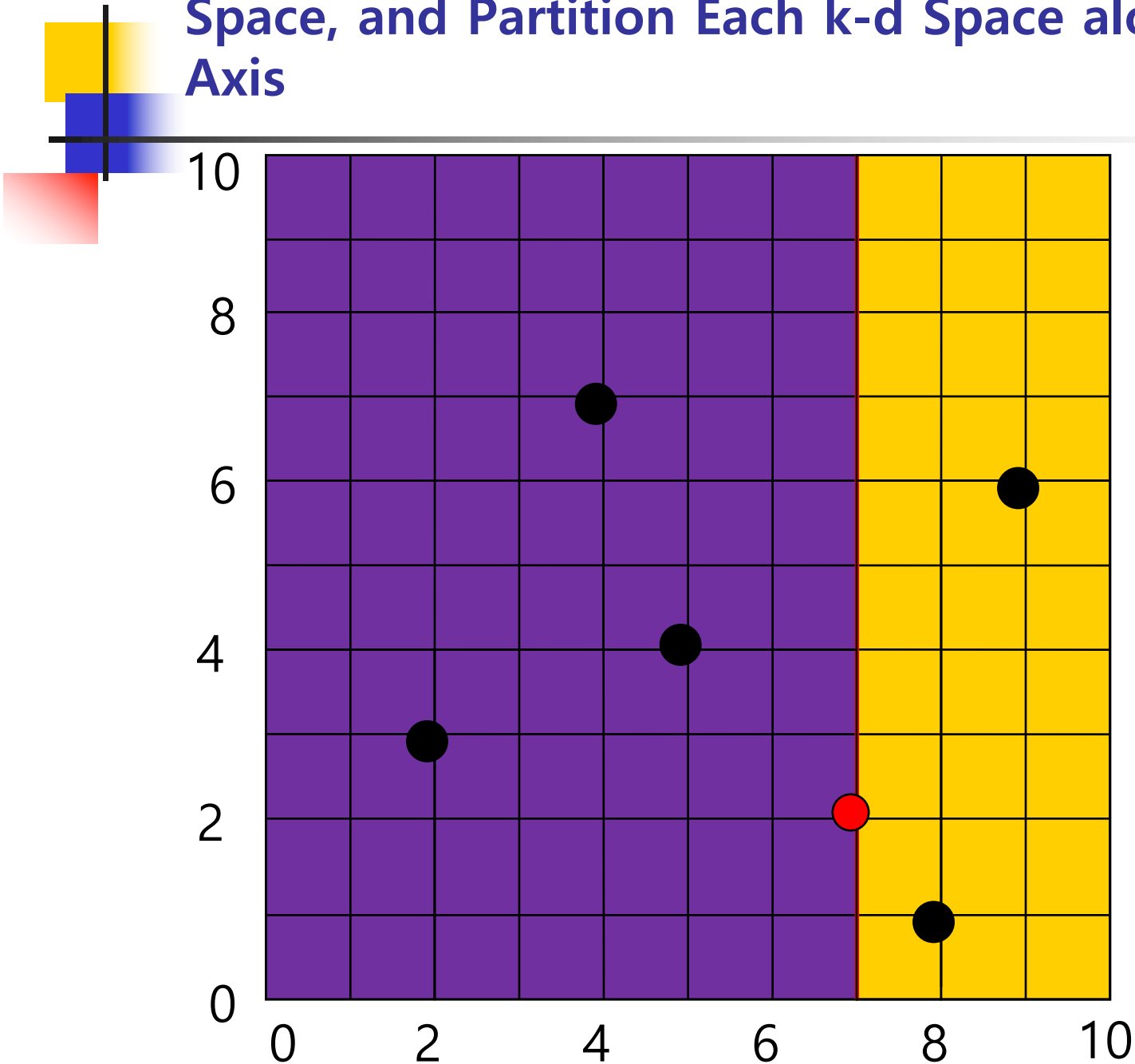
First, Select Point (7,2) and Partition the k-d Space along the Y Axis



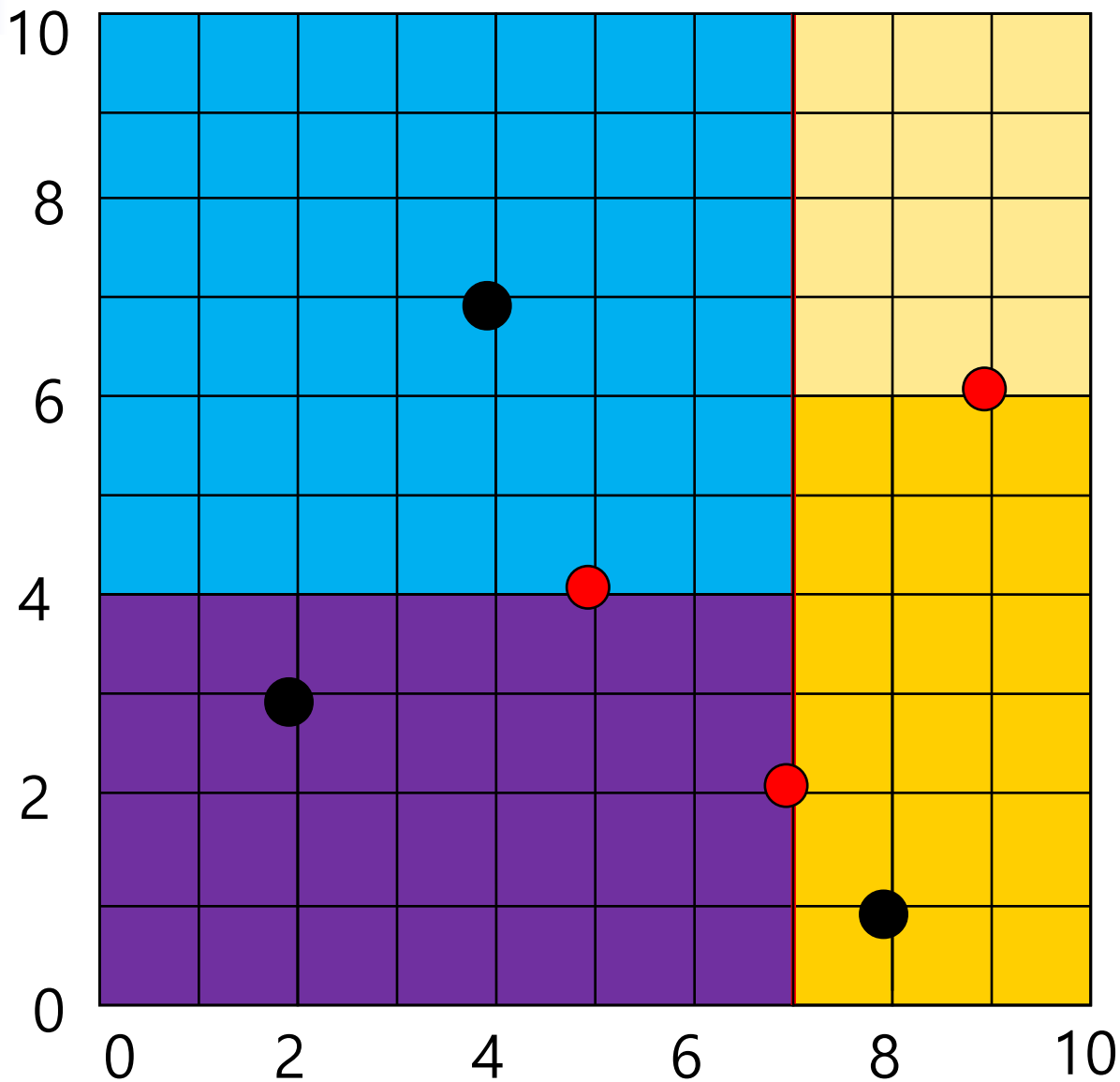
Result



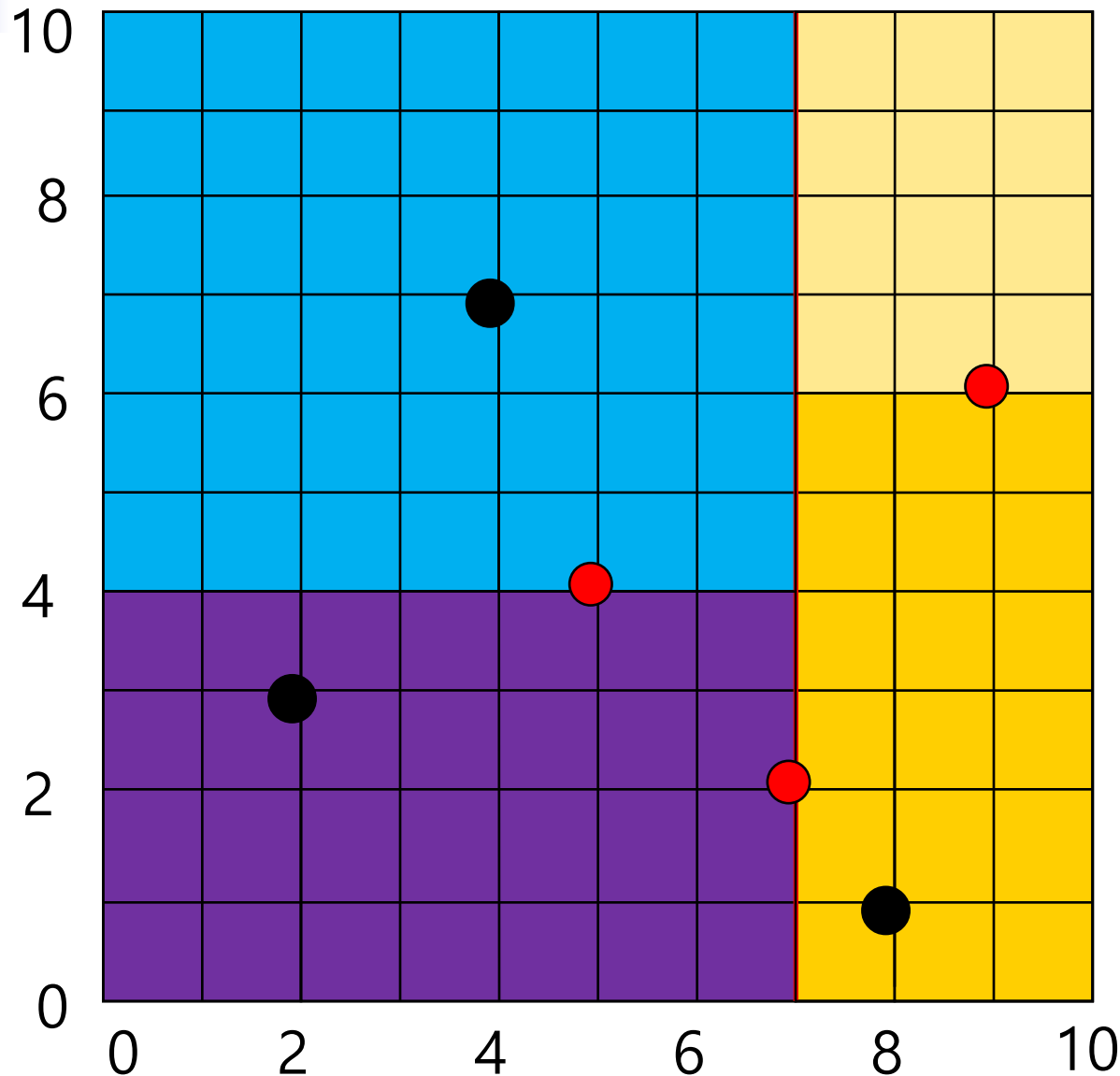
Second, Select Point (5,4) and (9,6) from Each k-d Space, and Partition Each k-d Space along the **X** Axis



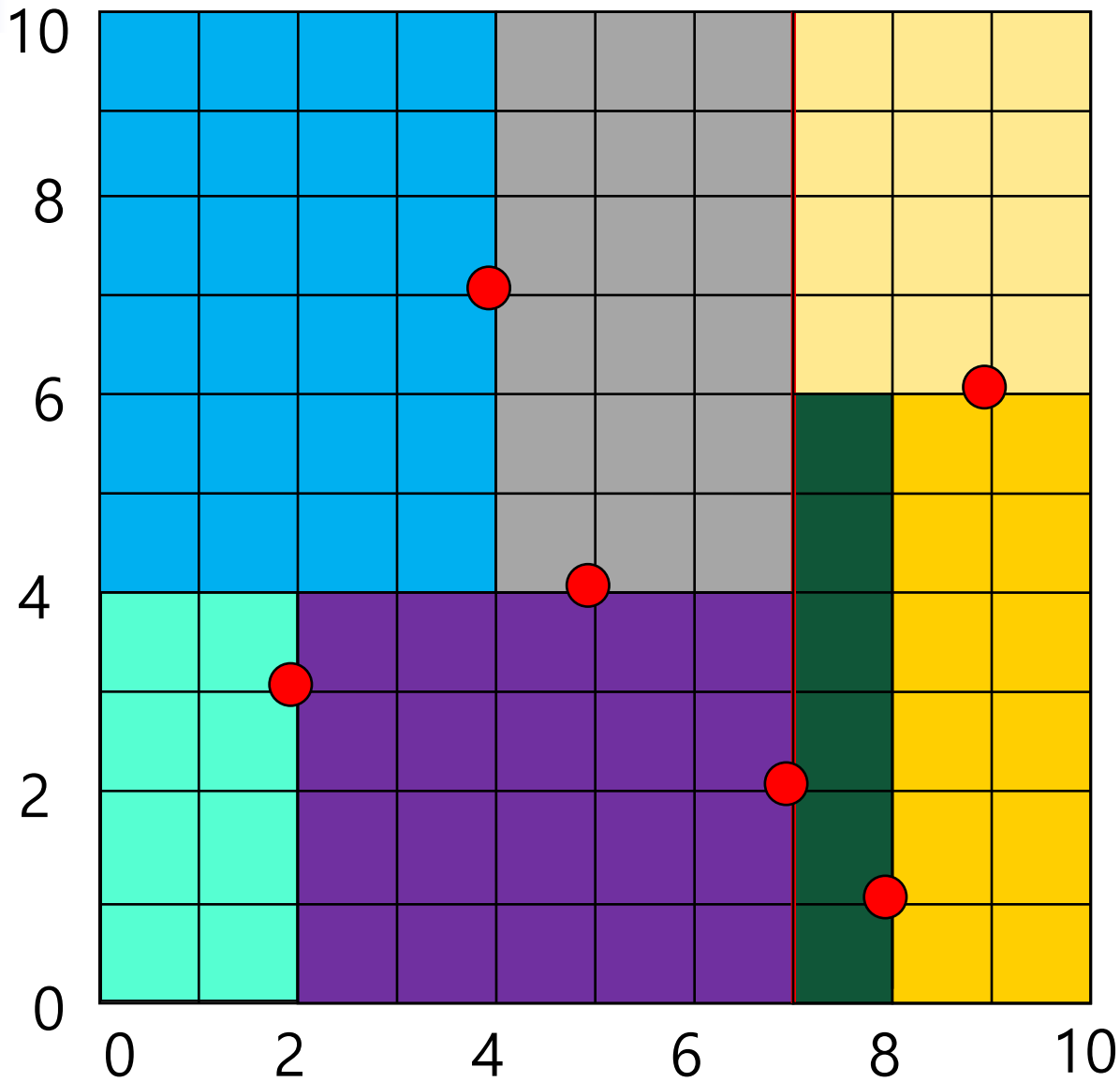
Result



Third, Select Point (2,3), (4,7), and (8,1) from Each k-d Space, and Partition Each k-d Space along the Y Axis

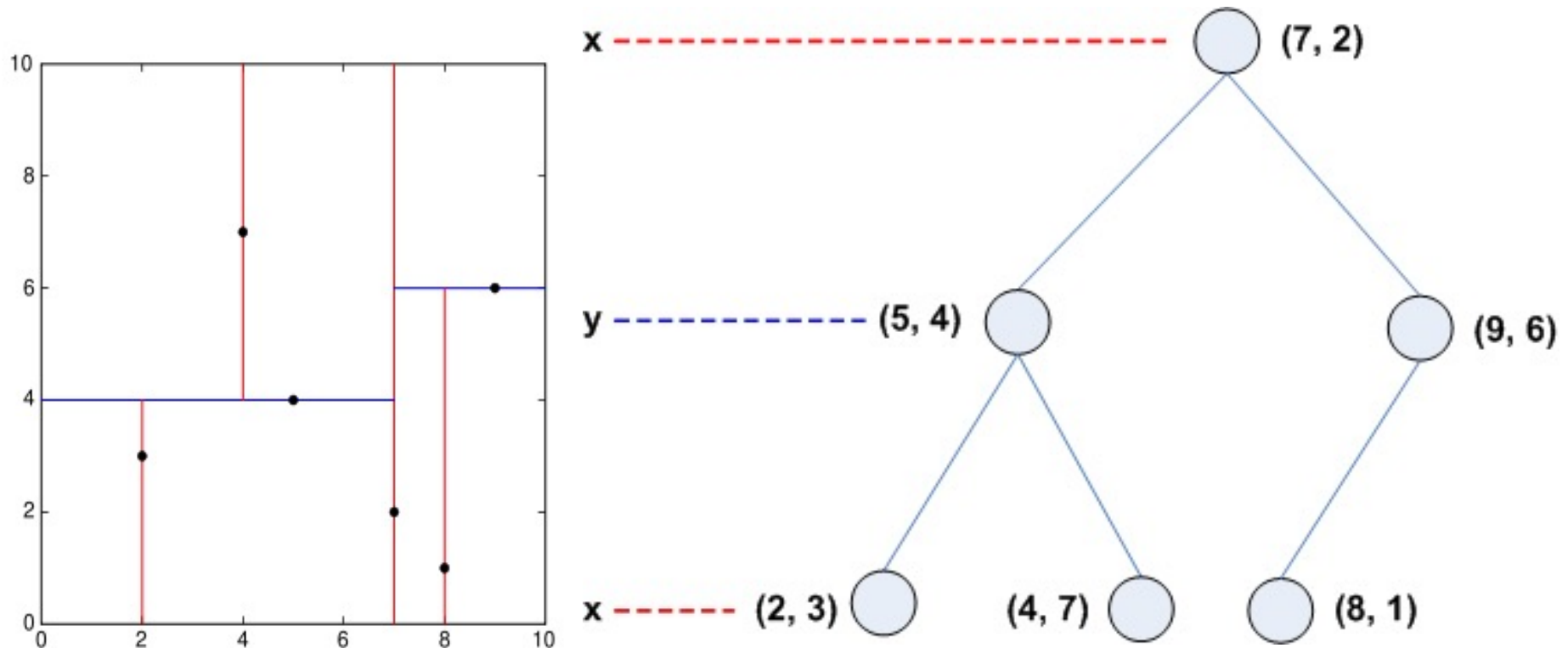


Result



Map Selected Points in a k-d Space to Nodes of a k-d Tree

How are these nodes selected? (We will see shortly...)





Constructing a k-d Tree

- In general, **static** construction with a set of given points in a k dimensional space.
- Each level of the tree represents a **partitioning axis**.
- The partitioning axis cycles through the k dimensions.
 - (e.g.) $k=2$ x, y, x, y, x, y, \dots
 - (e.g.) $k=3$ $x, y, z, x, y, z, x, y, z, \dots$
 - (e.g.) $k=4$ $x, y, z, w, x, y, z, w, x, y, z, w, \dots$
- **Discriminator** is the **median** key at each level of the k-d tree.
 - Median key is selected to distribute the data evenly on the tree



Computing the Median

- “Median”
 - equal number of values $<$ and $>$
 - (e.g.) 4 7 9 **10** 11 15 30
- If the numbers are different,
 - Select the Largest of the Smaller Group or the Smallest of the Larger Group.
 - (e.g.) 4 7 10 11 15 30
 - 4 7 **10** 11 15 30
 - 4 7 10 **11** 15 30
- If there are multiple identical values in a median candidate?
 - Results in a skewed tree
 - (e.g.) 4 7 10 10 10 10 30
 - 4 7 **10** **10** **10** **10** 30
- either “ \leq median” go to the left subtree
 - “ $>$ median” go to the right subtree
- or “ $<$ median” go to the left subtree
 - “ \geq median” go to the right subtree



Example (1/5)

set of points in 2-dimension

X Y

2,3

4,7

5,4

7,2

8,1

9,6

select the
median X value

Example (2/5)

select the median x value

2, 4, 5, 7, 8, 9

7,2

$X \leq 7$

X Y

2,3

4,7

5,4

select the
median Y value

X Y

8,1

9,6

$X > 7$

select the
median Y value

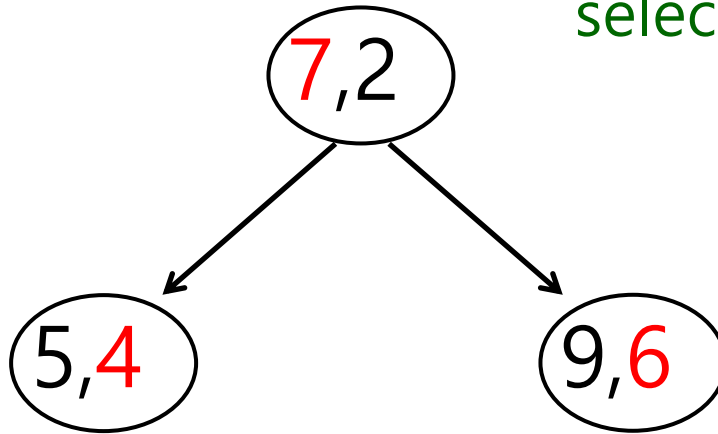
Example (3/5)

select the median x value

2, 4, 5, 7, 8, 9

x

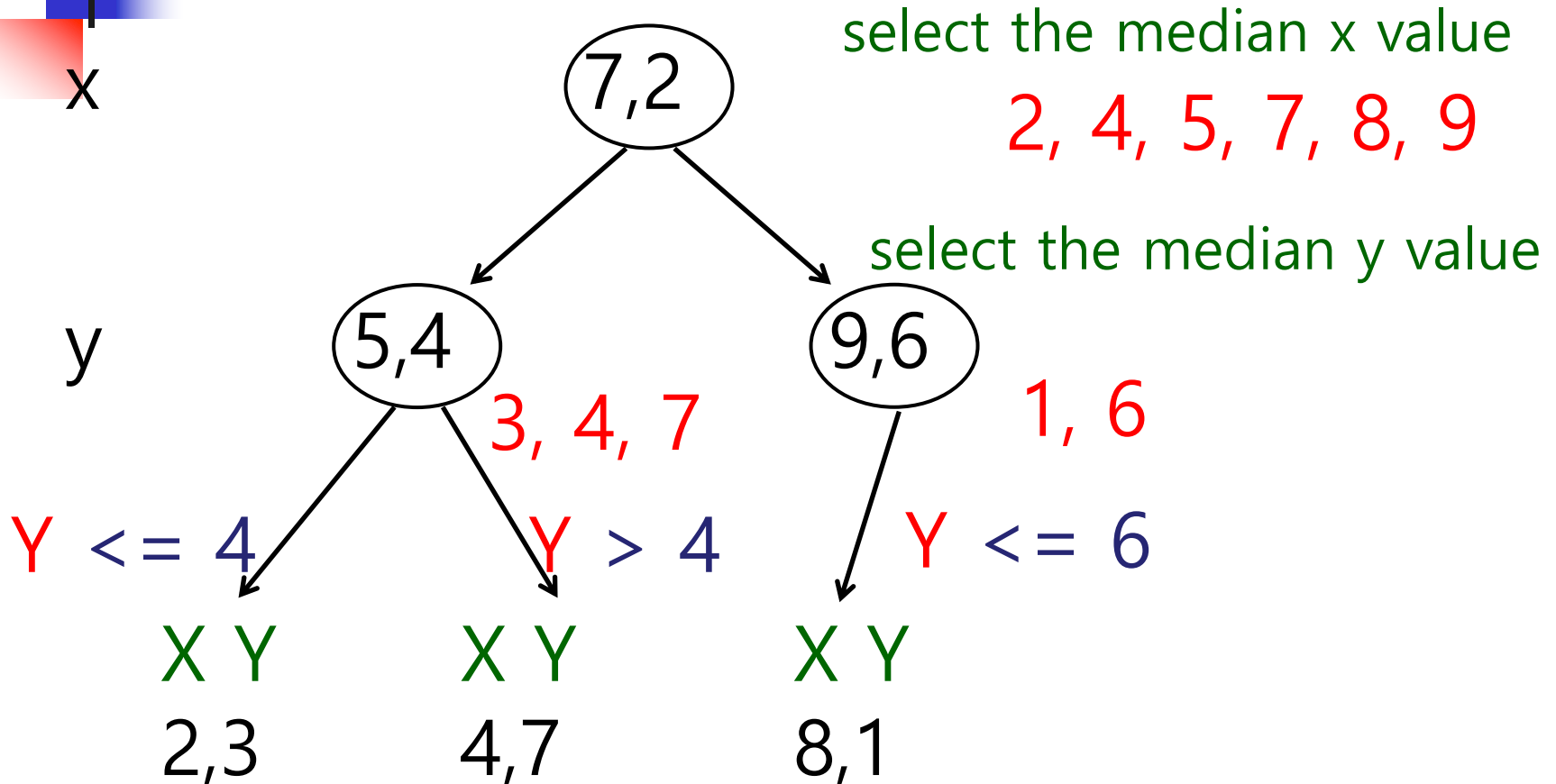
y



select the
median y value

select the
median y value

Example (4/5)



Example (5/5)

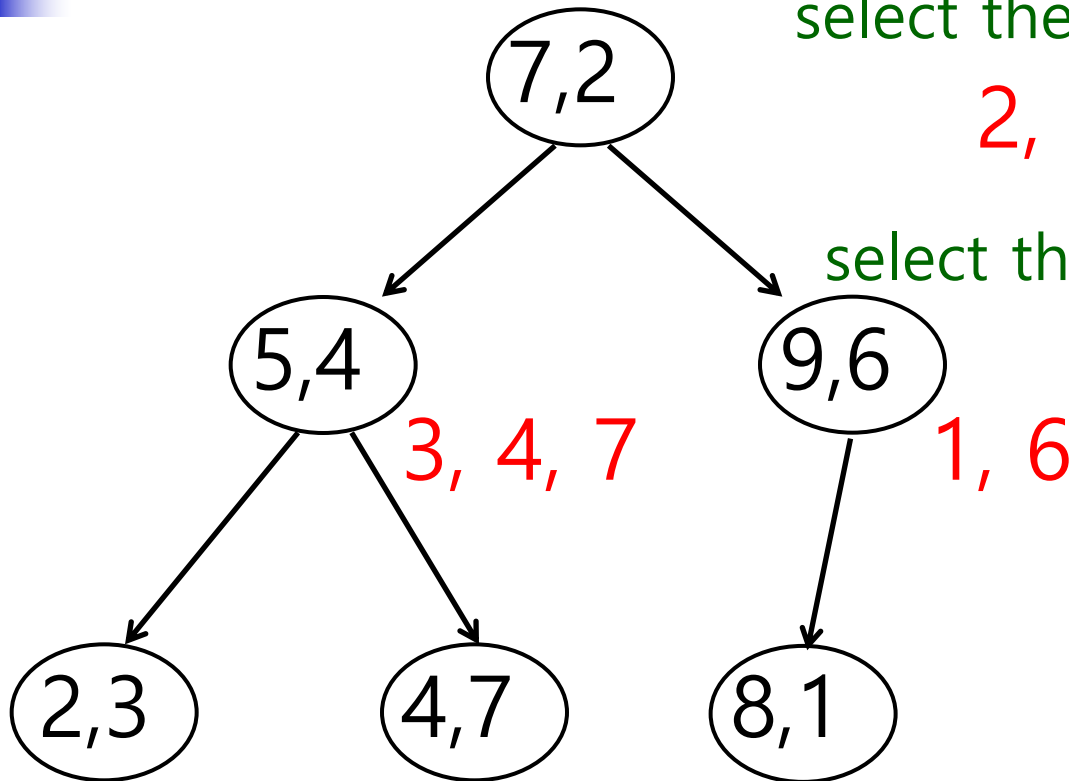
select the median x value

2, 4, 5, 7, 8, 9

select the median y value

y

x





Example (1/6)

set of points in 2-dimension

X Y

2,3

5,1

5,2

5,4

5,6

5,7

9,8

10,6

Example 2 (2/6)

x

5,4

select the median x value

2,5,5,5,5,5,9,10

X Y

2,3

5,1

5,2

5,4

5,6

5,7

9,8

10,6

Example 2 (3/6)

x

5,4

select the median x value

2,5,5,5,5,5,9,10

X Y

2,3

5,1

5,2

5,6

5,7

X Y

9,8

10,6

X Y

2,3

5,1

5,2

5,4

5,6

5,7

9,8

10,6

Example 2 (4/6)

x

5,4

select the median x value

2,5,5,5,5,5,9,10

X Y

5,1

5,2

2,3

5,6

5,7

X Y

10,6

9,8

X Y

2,3

5,1

5,2

5,4

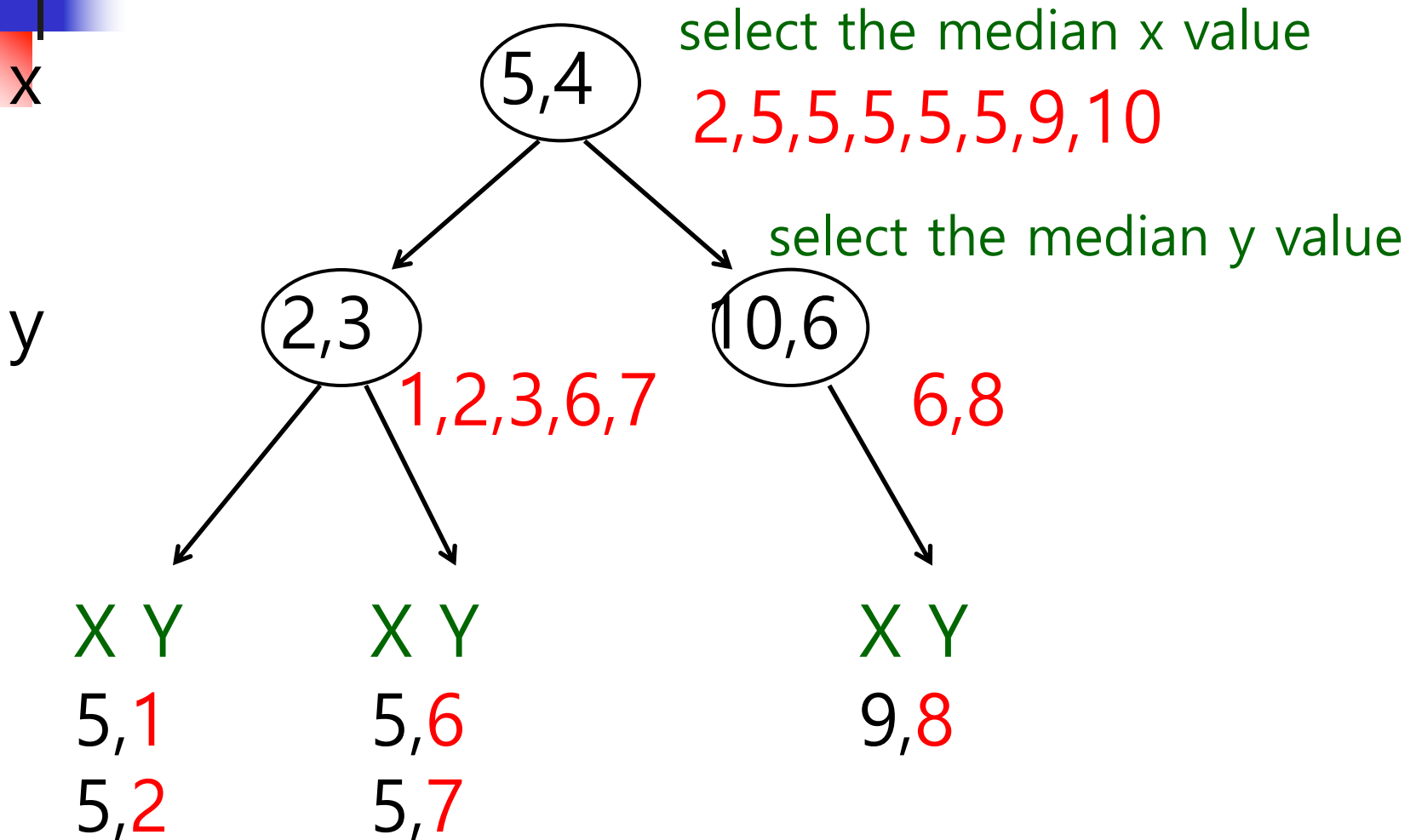
5,6

5,7

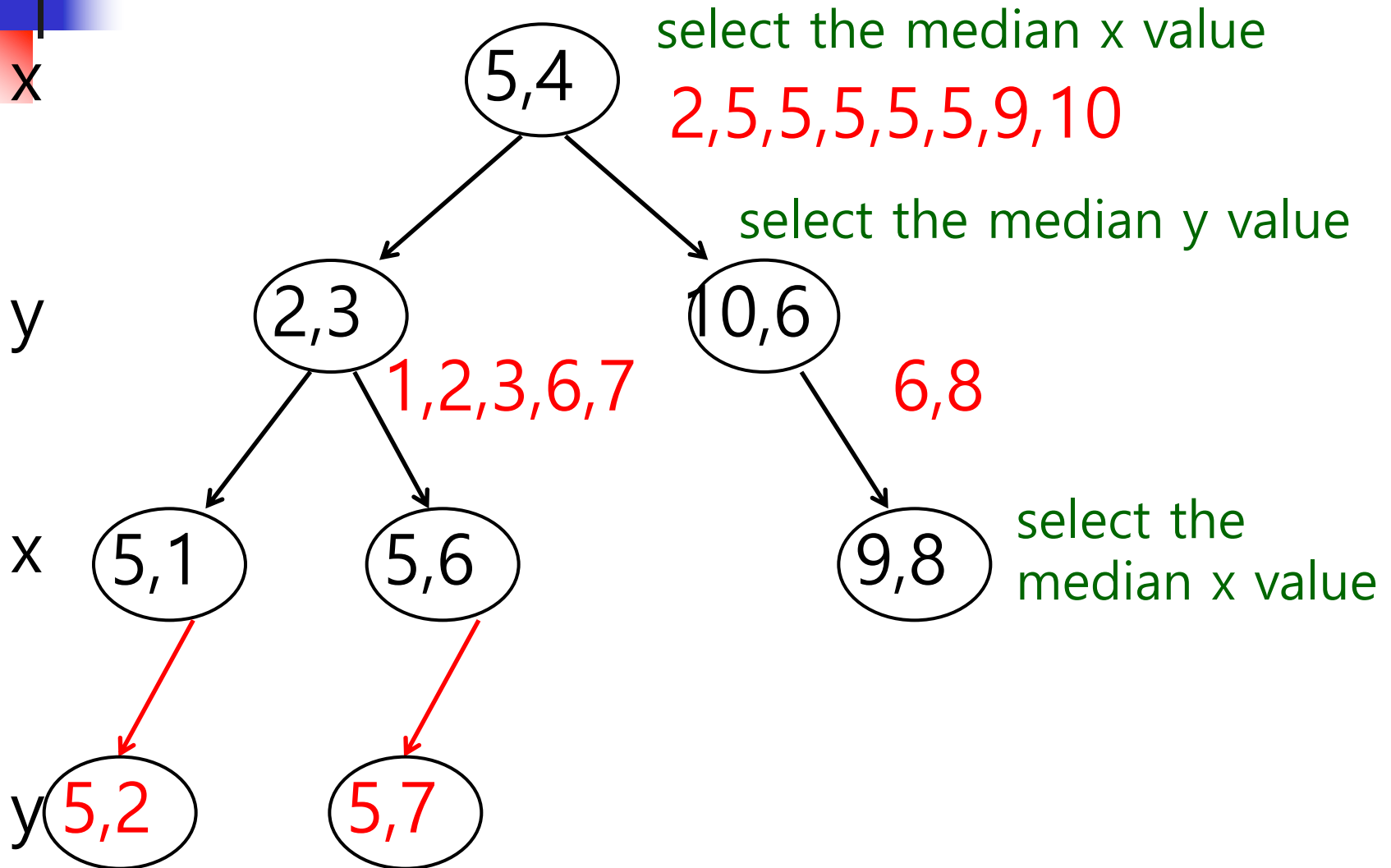
9,8

10,6

Example 2 (5/6)



Example 2 (6/6)





Exercise

set of points in 3-dimension

X Y Z

1,4,5

2,3,11

3,7,4

4,10,6

5,12,9

6,2,8

7,8,4

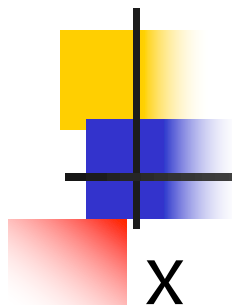
9,5,3

10,6,5

11,11,11

12,1,1

Searching a k-d Tree: (similar to constructing a k-d tree)

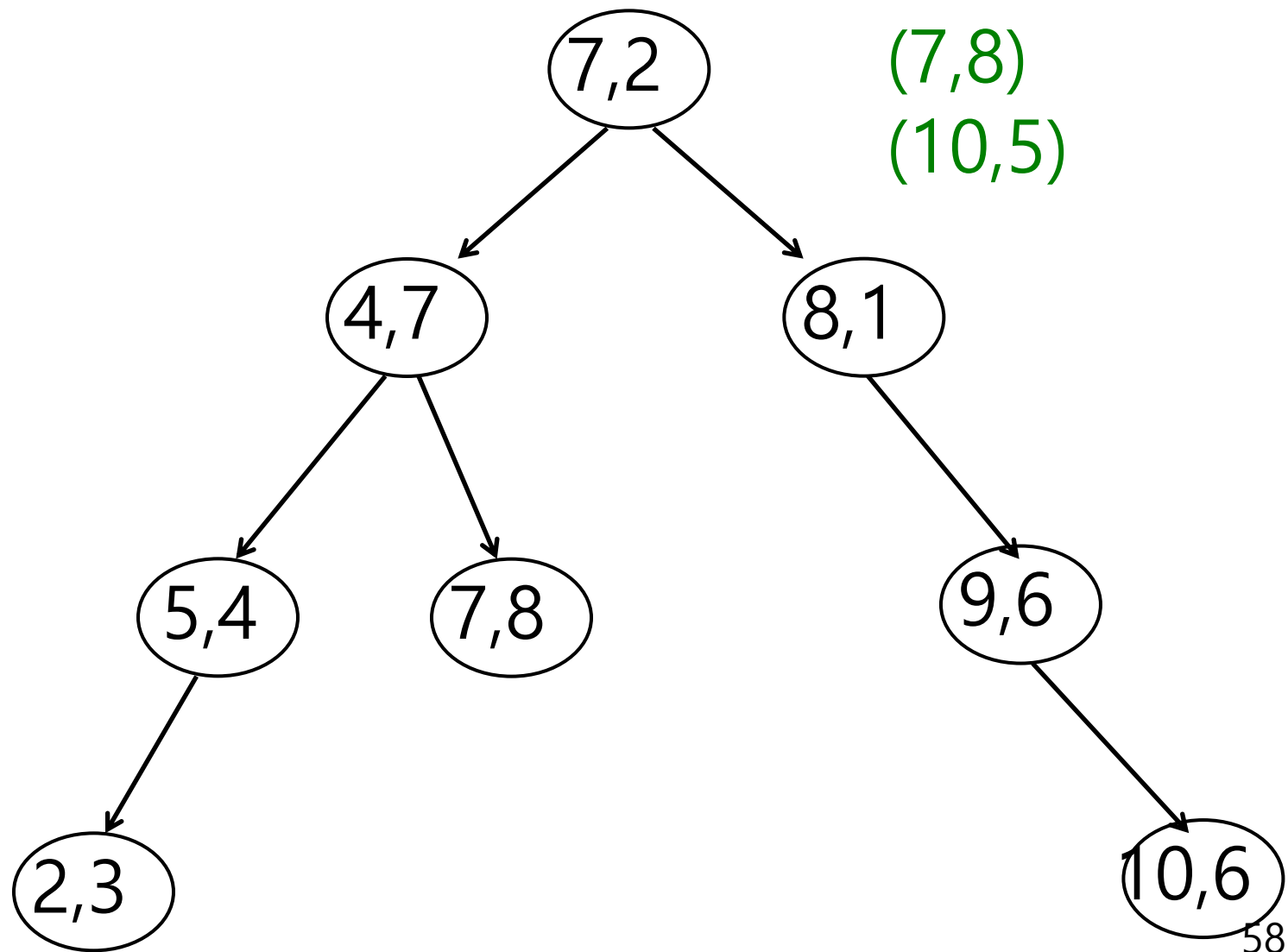


x

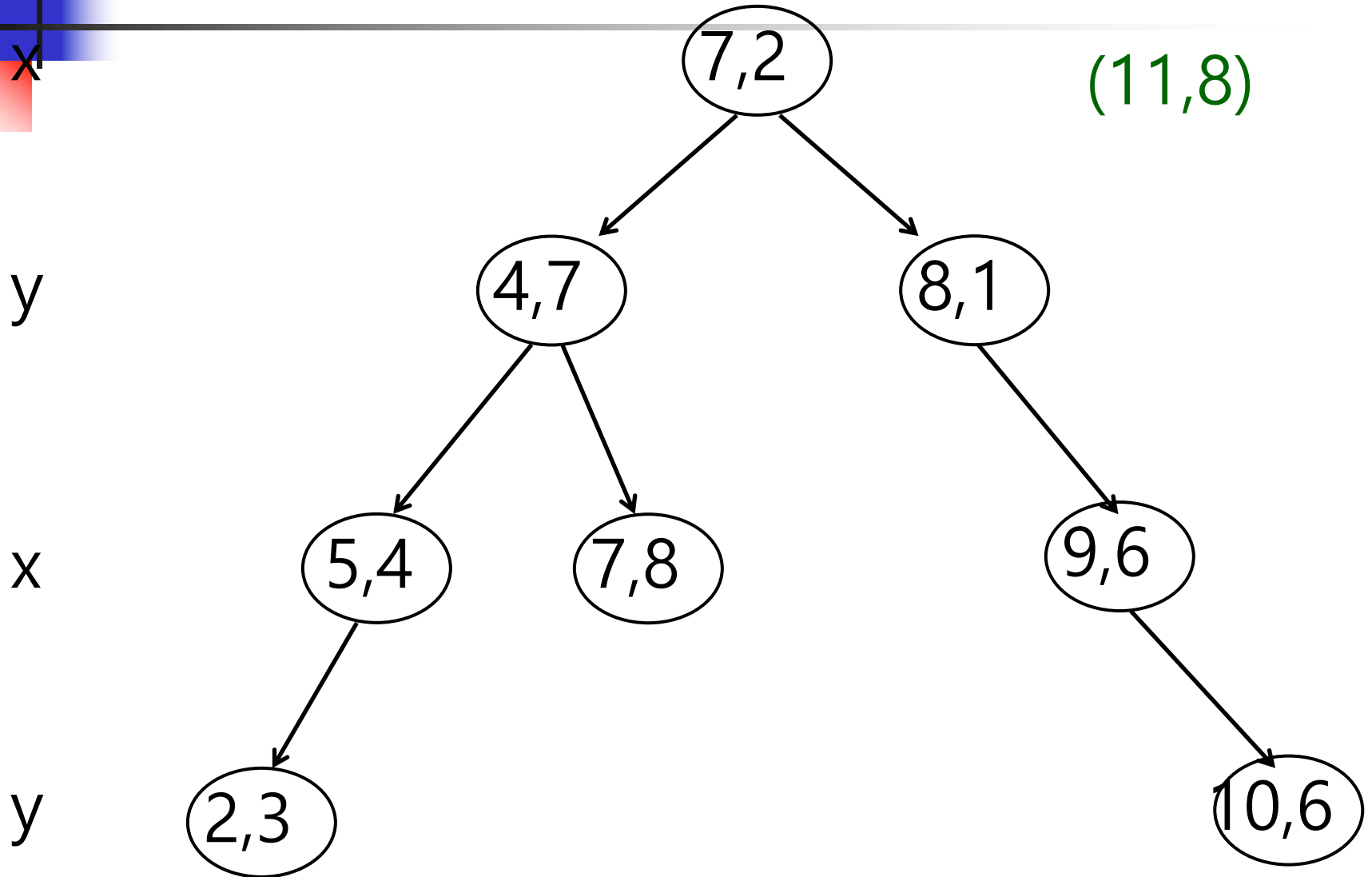
y

x

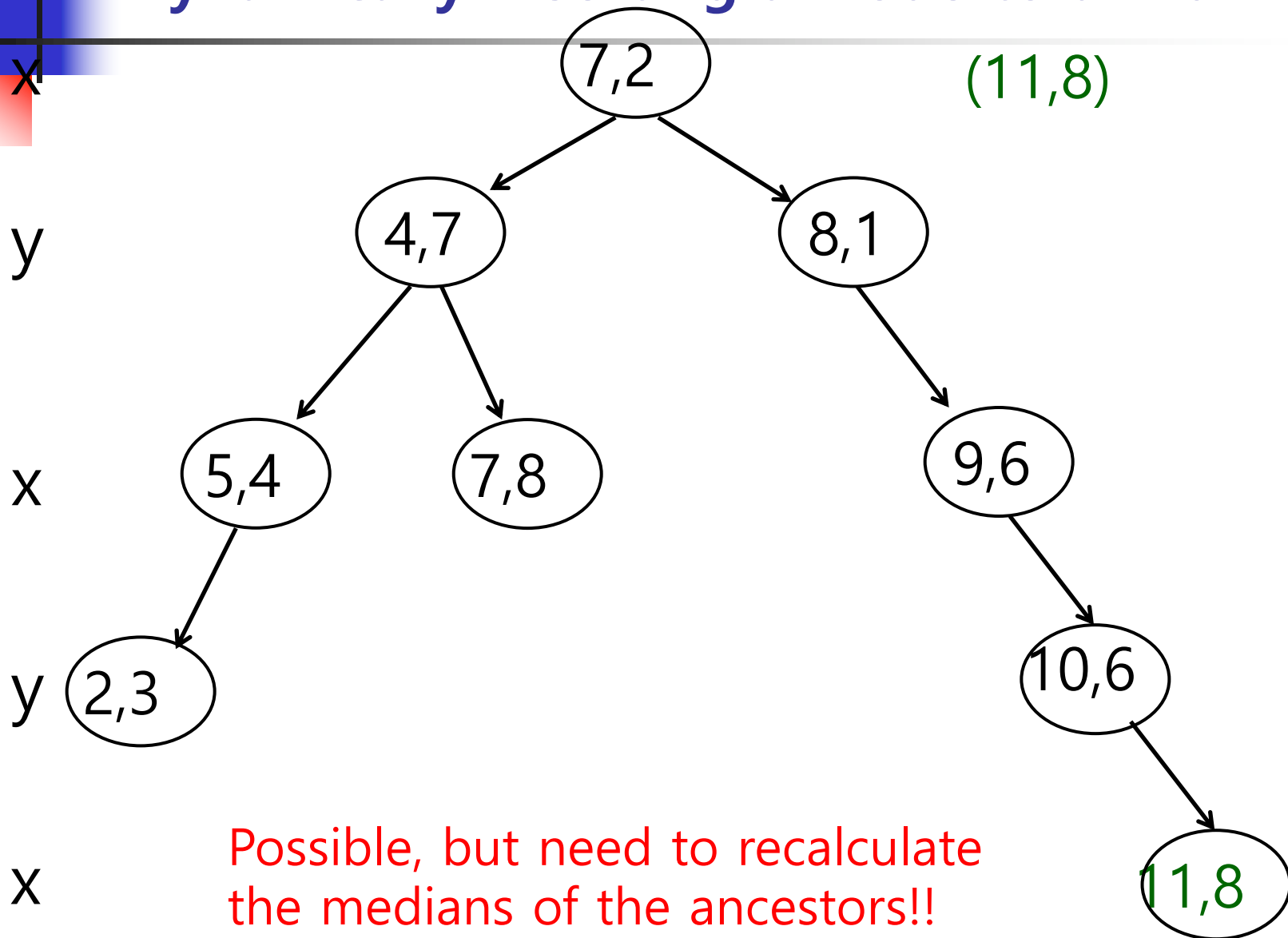
y



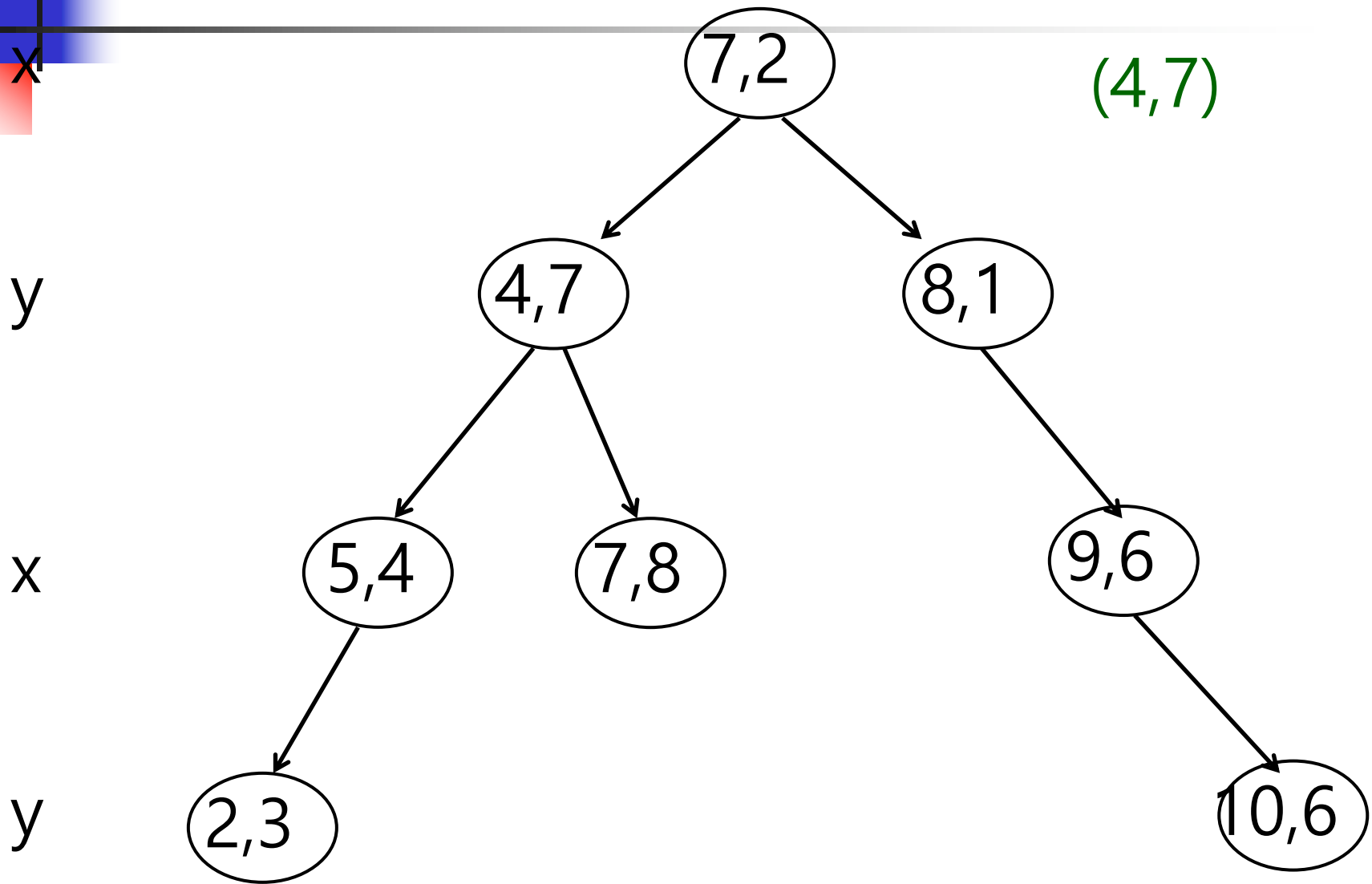
Dynamically Inserting a Node to a k-d Tree



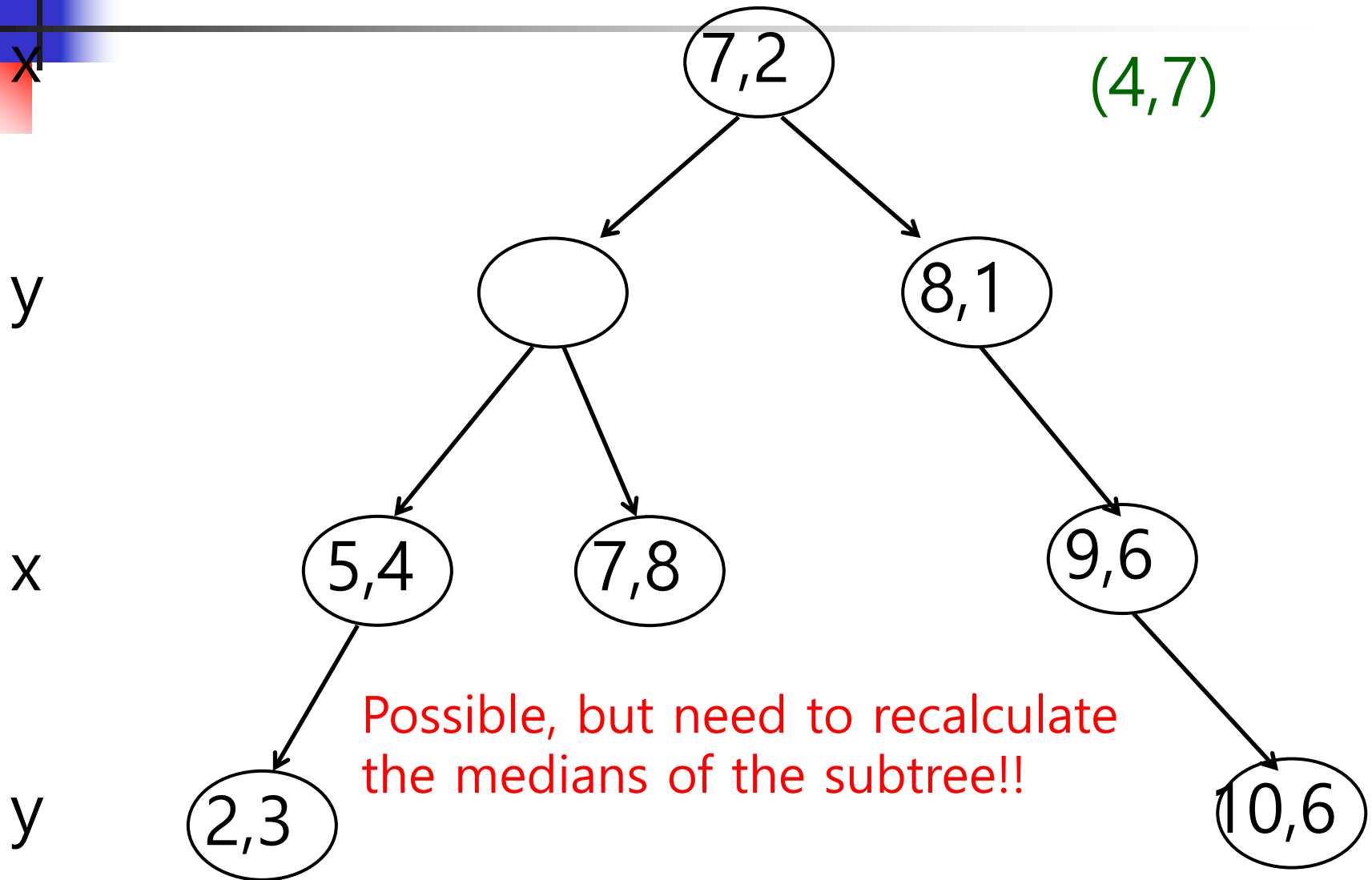
Dynamically Inserting a Node to a k-d Tree



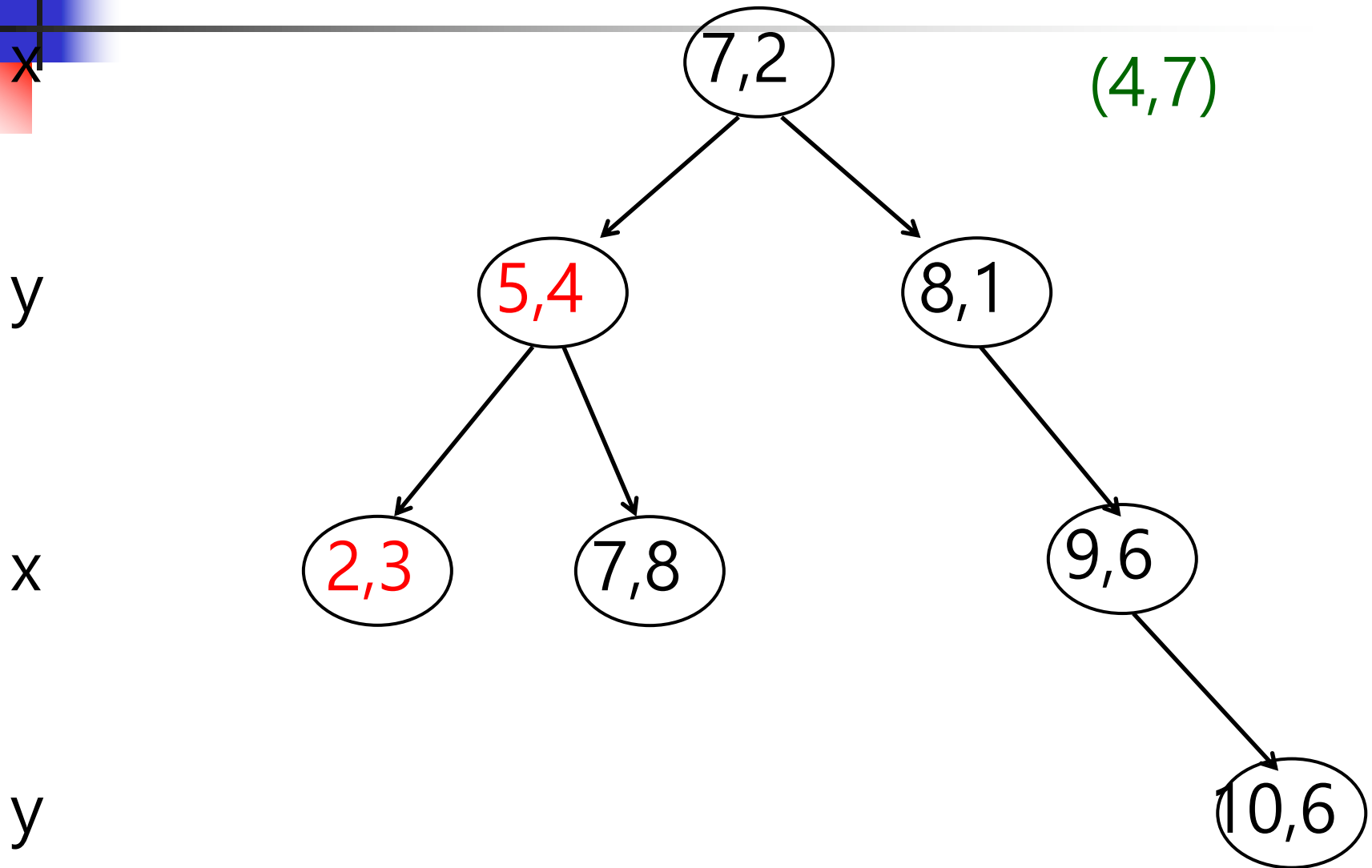
Dynamically Deleting a Node from a k-d Tree



Dynamically Deleting a Node from a k-d Tree



Dynamically Deleting a Node from a k-d Tree





k-d Tree: Properties

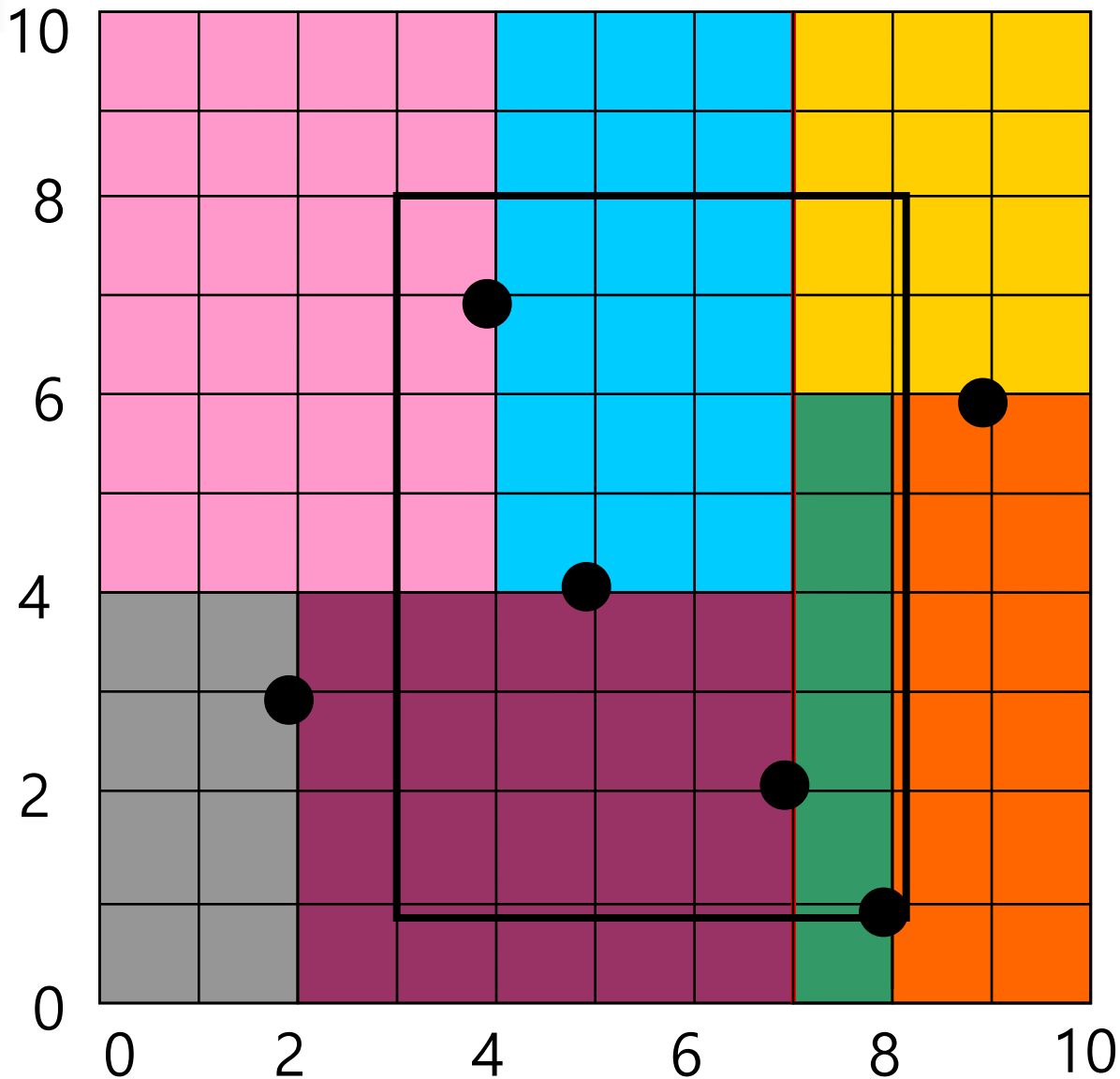
- It is a binary search tree for k-d keys.
- It is not height-balanced
- It can be badly skewed if there are many identical median values at any level.
- It is a static tree for searches; (i.e.,) it is constructed once for repeated searches.
- Dynamic insertion and deletion of keys can require partial reorganizations of the tree.



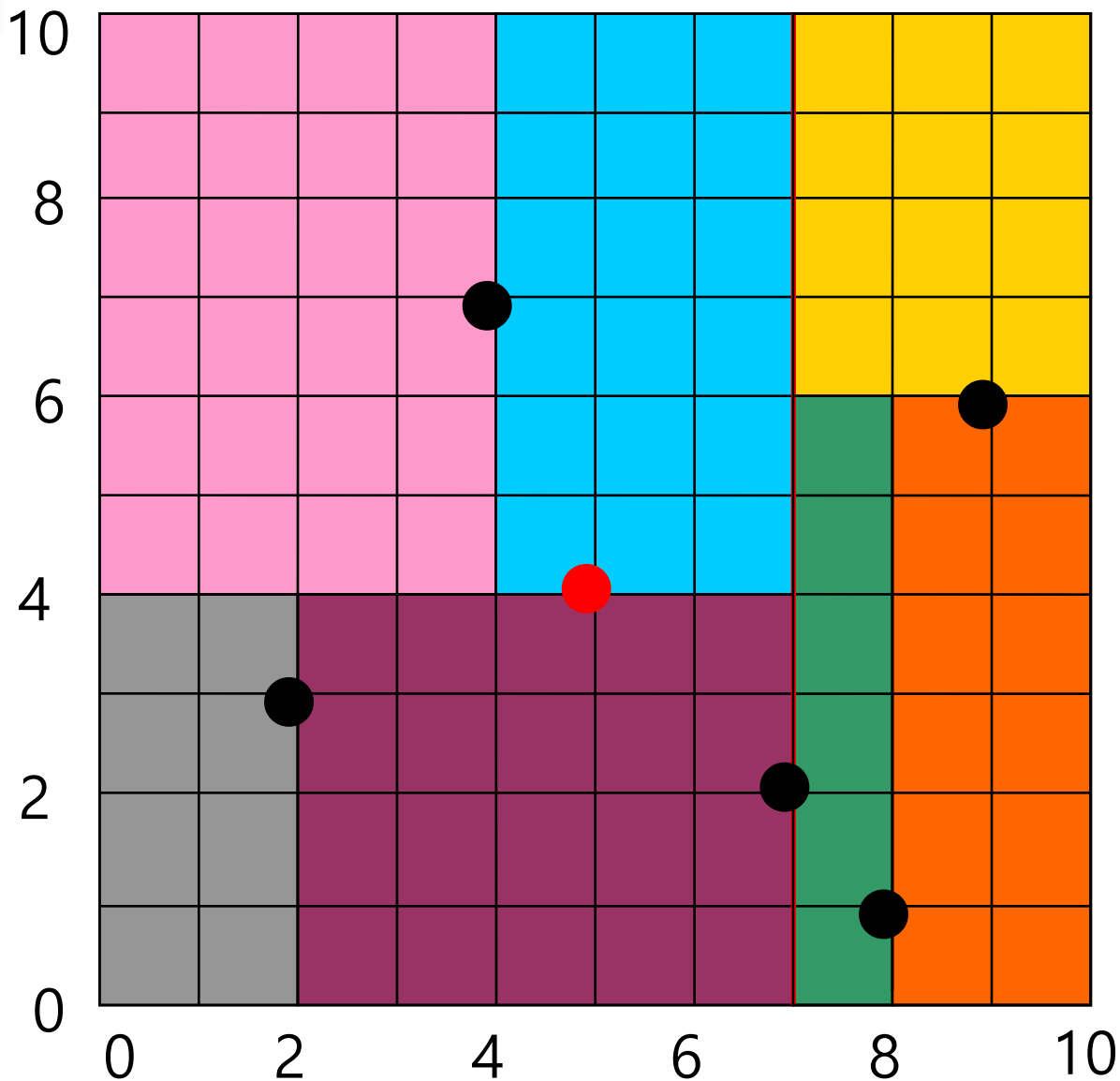
k-d Tree: Uses

- Point search
- Range (region) search
- Nearest neighbor search
- Partial key search

Example: Range Search



Example: Nearest Neighbor Search



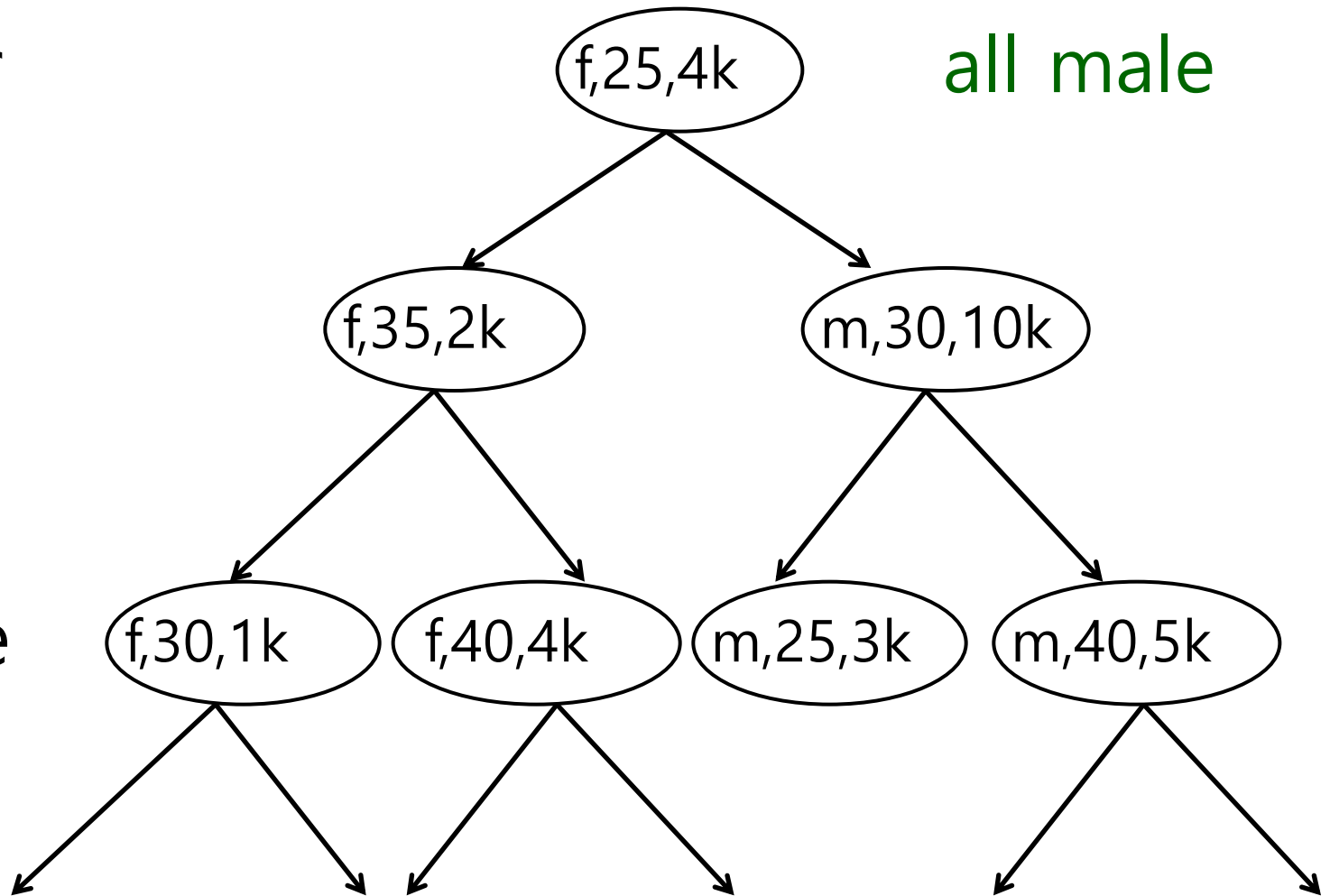
Example: Partial Key Search

gender

all male

age

income





Tries



Binary Trie

- Retrieval
- A binary search tree
 - Not height-balanced
- Can be used as a replacement for hashing
- Reading:
 - pp. 561-564



Binary Trie

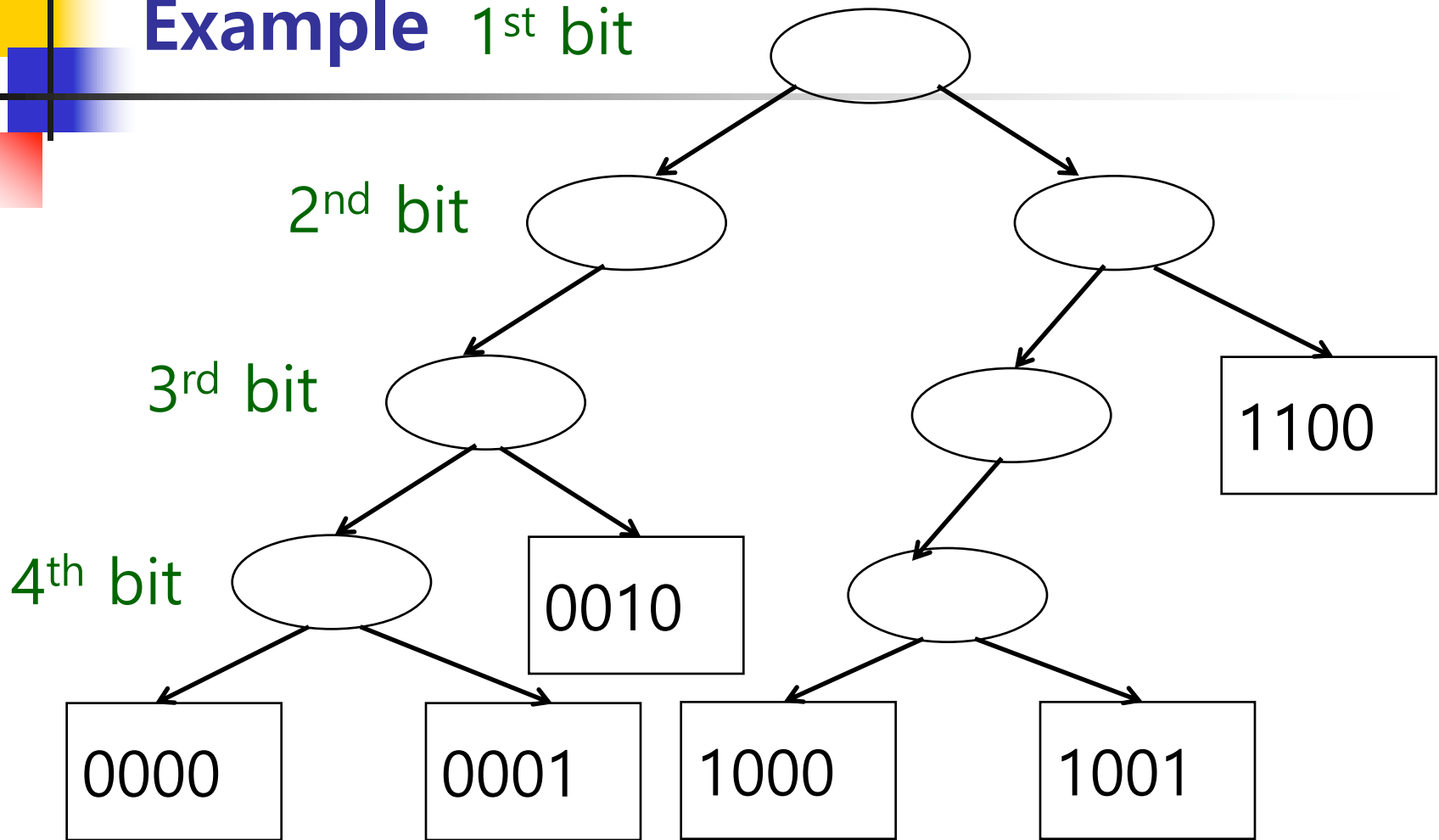
- Two Types of Node
 - Branch (link, interior, non-leaf) node
 - left-child ptr, right-child ptr
 - no data
 - Element (data, leaf) node
 - contains the full key



Binary Trie: Search

- For a Search Key k
 - At level i
 - if k 's i^{th} bit is 0, move to the left subtree;
 - if it is 1, move to the right subtree.
 - At a leaf node, the search key is compared with the data stored in the node.
- End of Search
 - success: at a leaf node
 - failure: at a leaf node or a NULL pointer

Example 1st bit



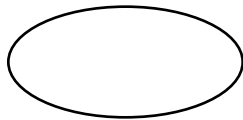
○ branch node

□ leaf node



Building a Binary Trie

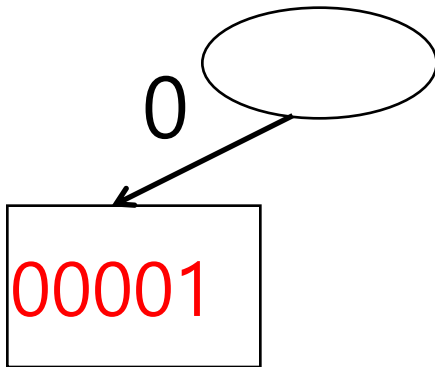
Insert 00001





Building a Binary Trie

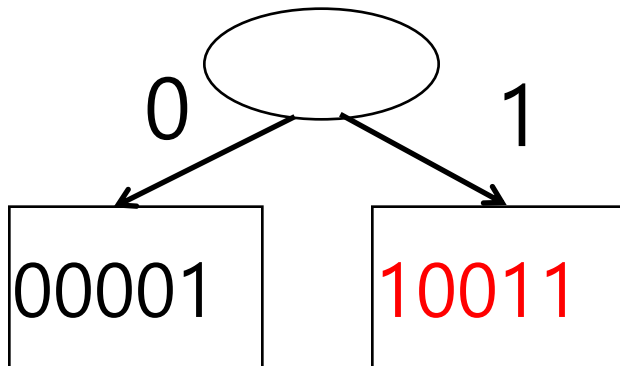
Insert 10011



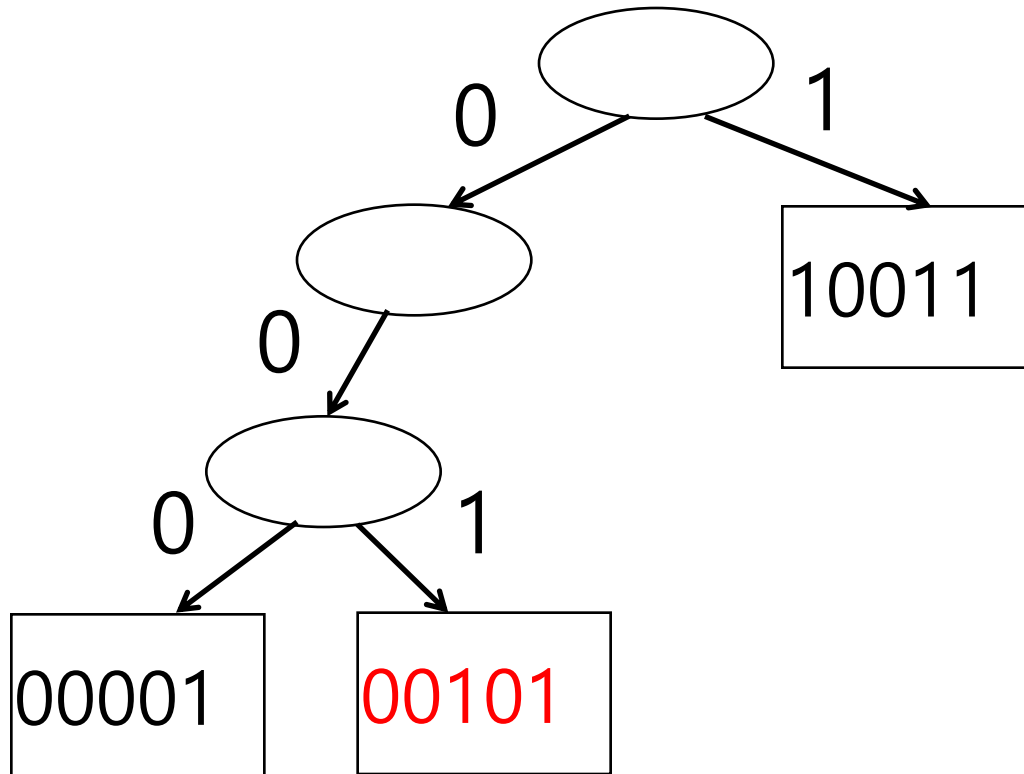


Building a Binary Trie

00101

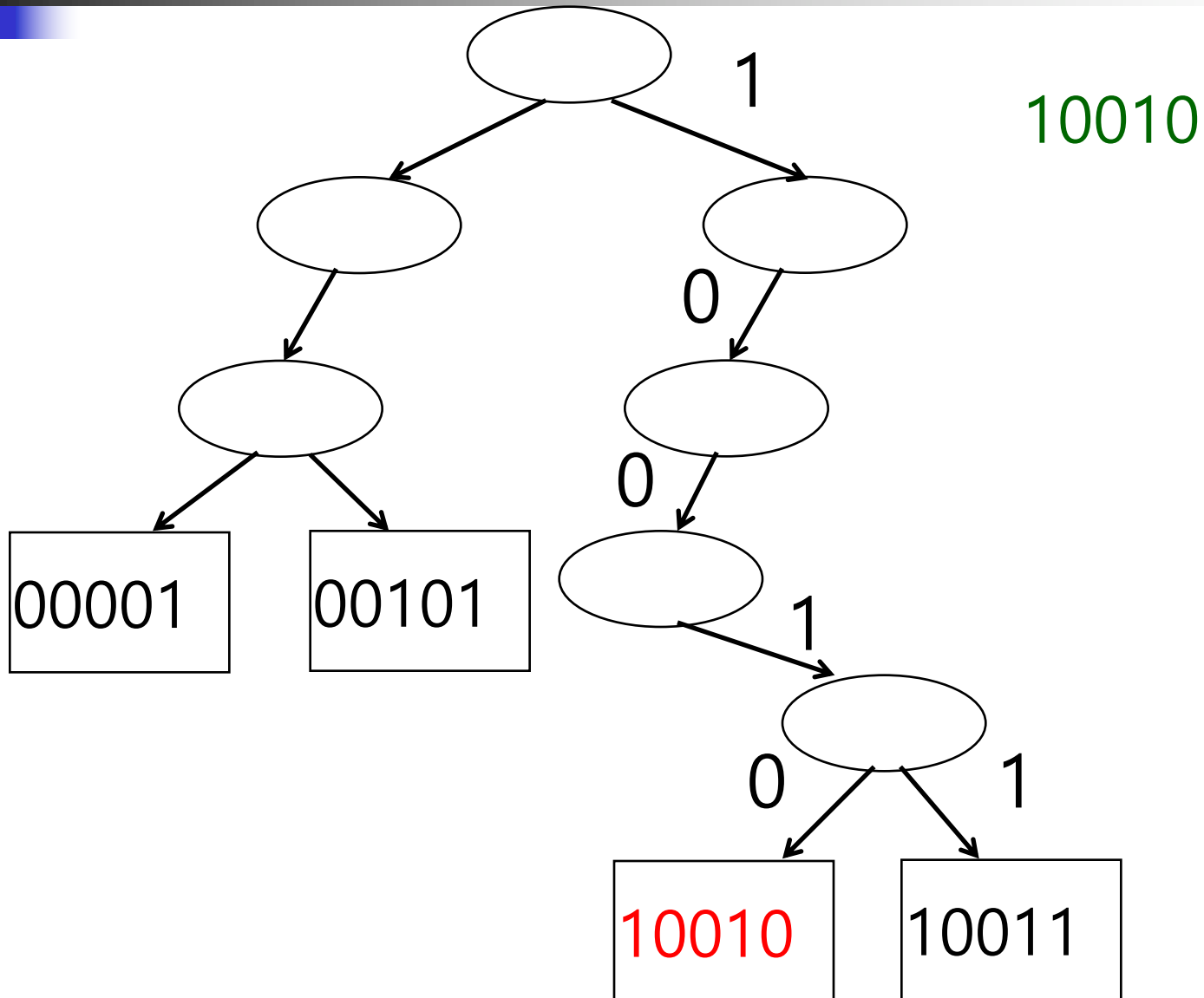


Building a Binary Trie – cont'd



10010

Building a Binary Trie – cont'd





Exercise

Insert 10010

10011

00001

00101

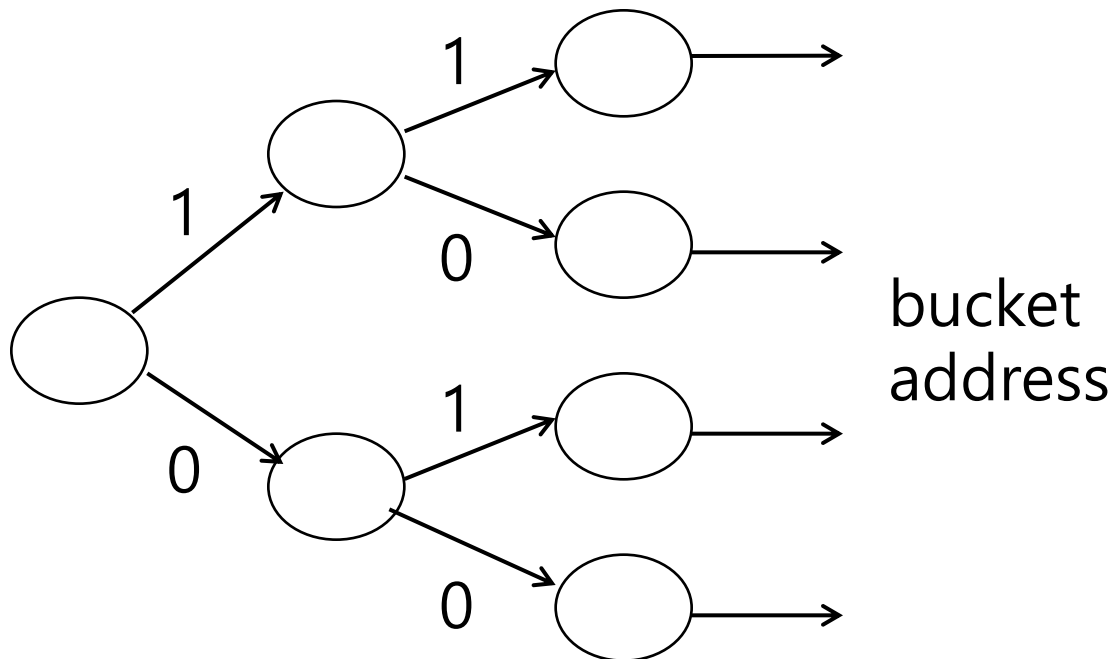


Binary Trie: Performance and Properties

- Avg. performance: $O(k)$
 - K is the maximum number of bits in the keys
- One comparison in a search
 - Only at the leaf node
- Efficient for exact key match and partial key match.
- Unbalanced tree
- The shape of the tree is independent of the order of insertions.

Binary Trie In Use

- Bucket directory in extendible hashing (to learn later)

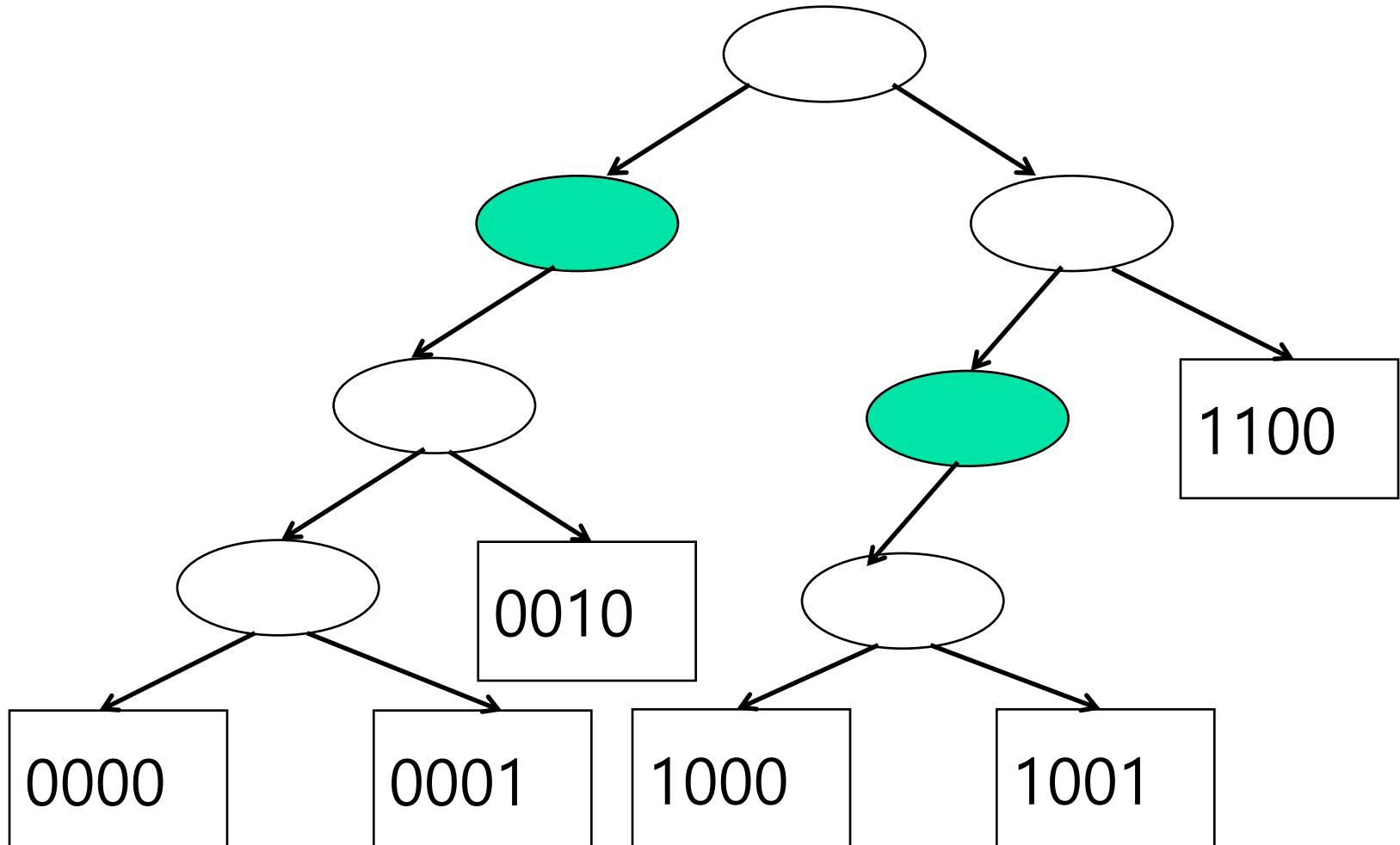




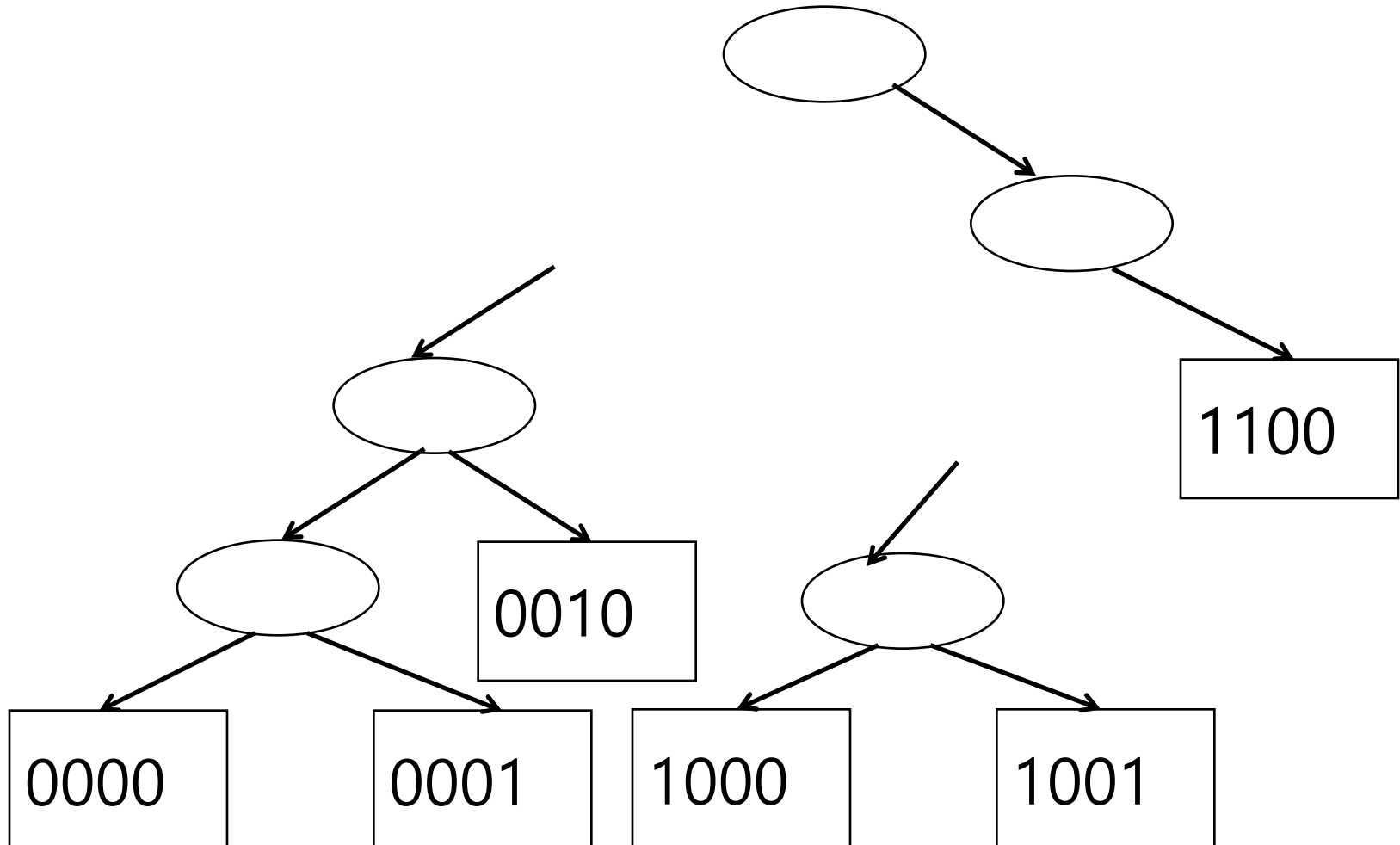
Compressed Binary Trie

- Branch nodes in a binary trie may be degree 1.
- Compressing a binary trie
 - Binary trie with degree-1 branch nodes eliminated.
 - Tree height is reduced.
- (On a compressed binary trie) each branch node is degree 2
- Branch node
 - left-child ptr, right-child ptr, bit-number

Eliminating Degree 1 Branch Nodes

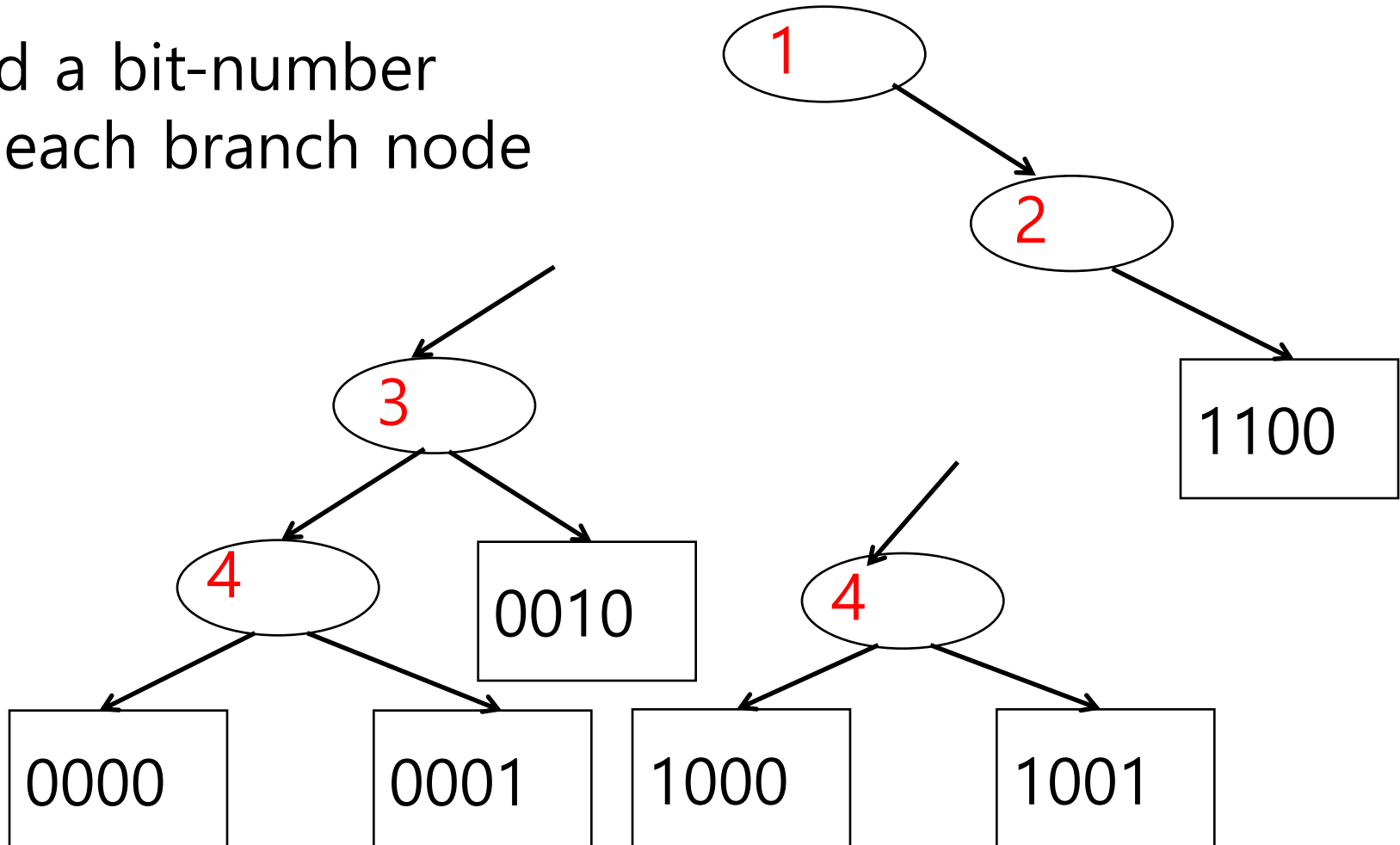


Eliminating Degree 1 Branch Nodes (cont'd)

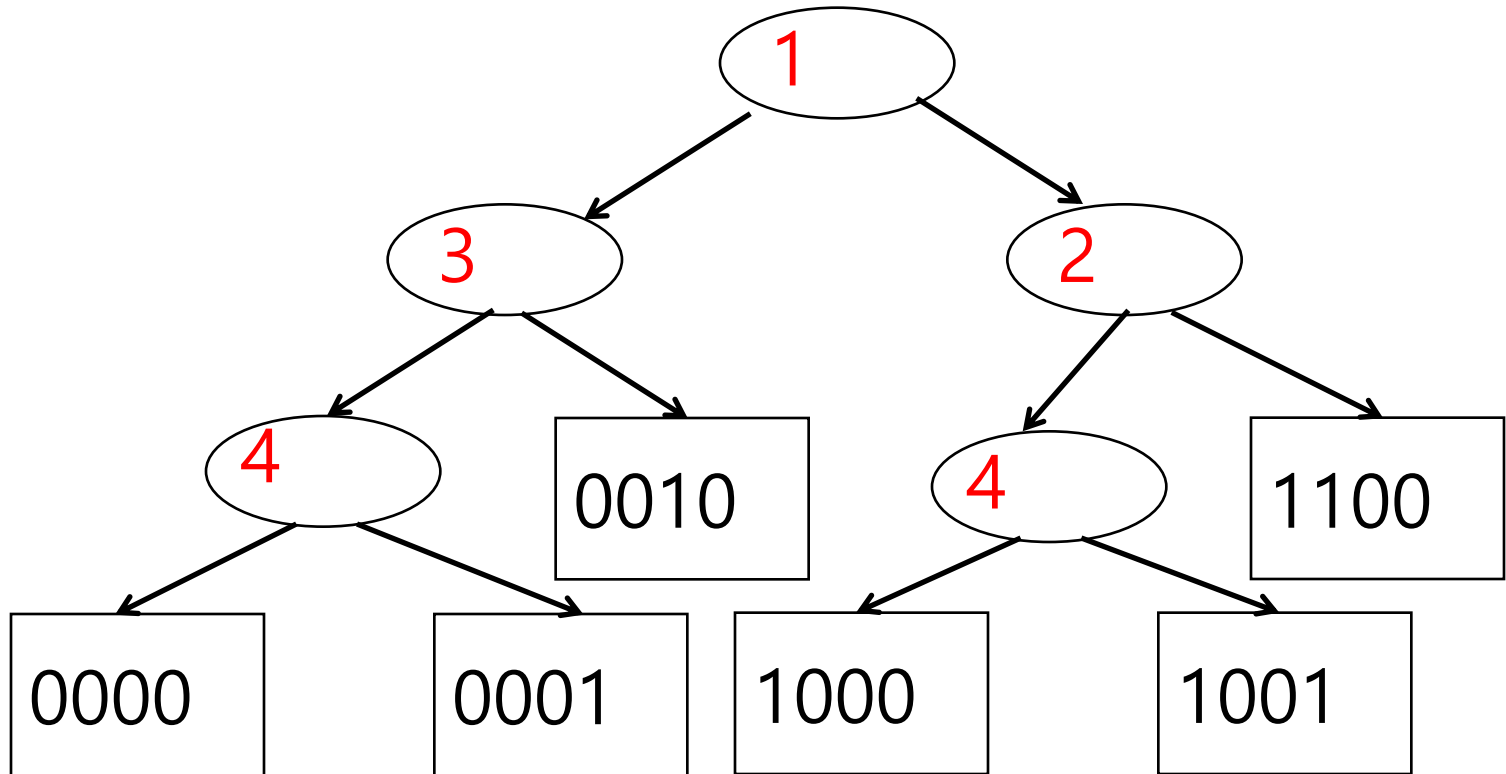


Eliminating Degree 1 Branch Nodes (cont'd)

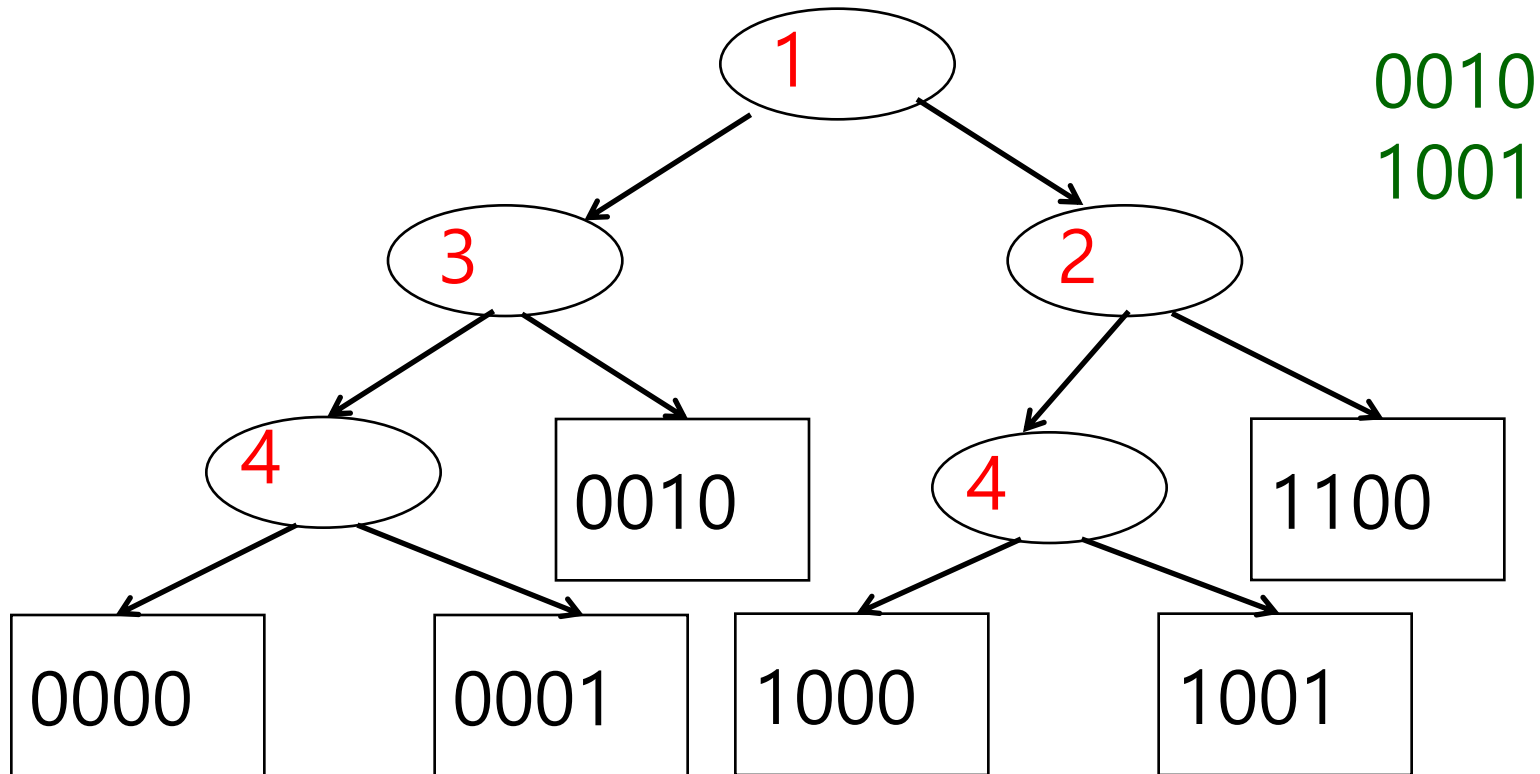
add a bit-number
to each branch node



Eliminating Degree 1 Branch Nodes (cont'd)



Searching a Compressed Binary Trie





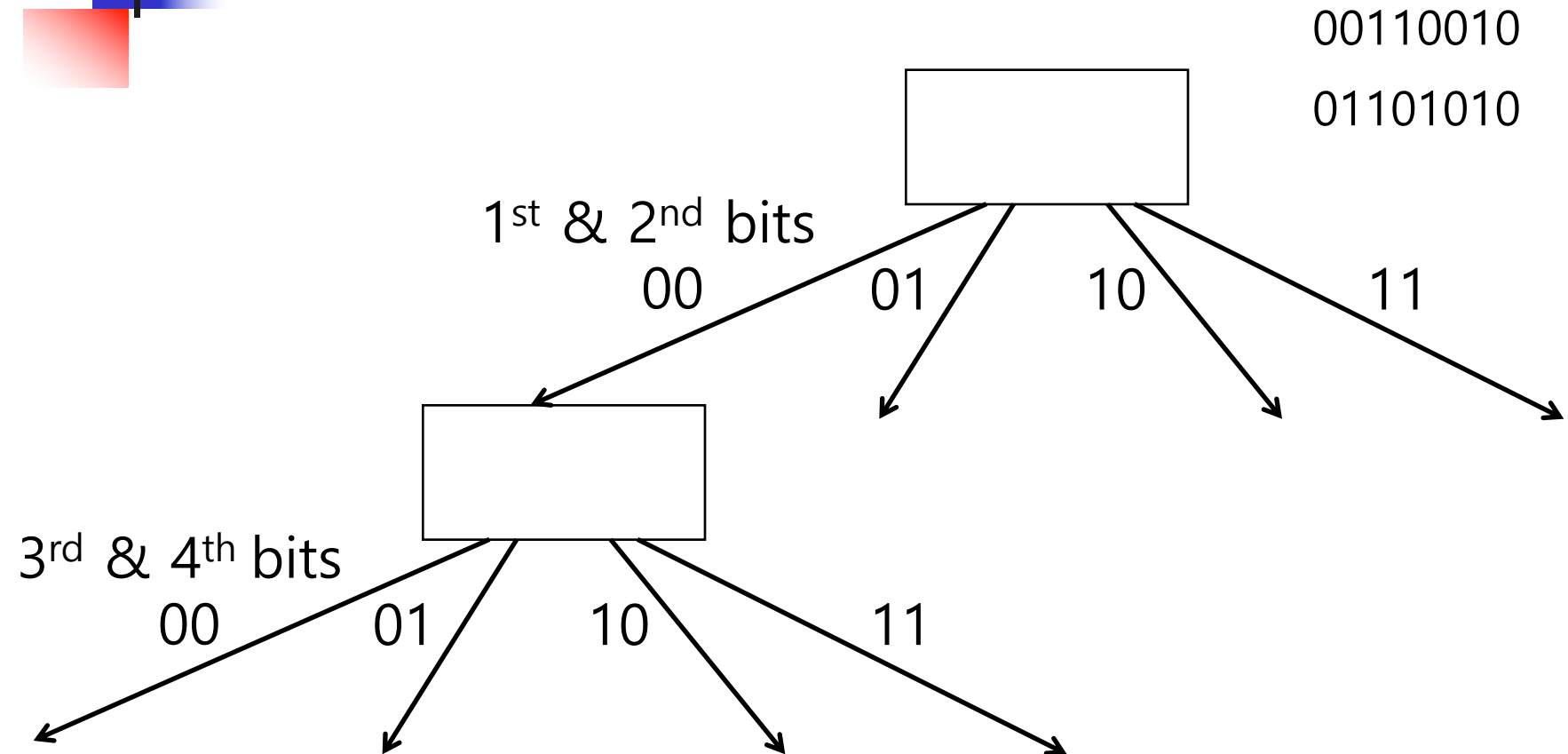
Multi-Way Tries



Multi-Way Trie

- Trie of degree $m \geq 2$
- Tree height is reduced (vs. Binary trie)
- Branching is based on a portion of the key.
 - May be a bit combination or digit or an alphabet
- Reading
 - pp. 571-578

A 4-Way Trie

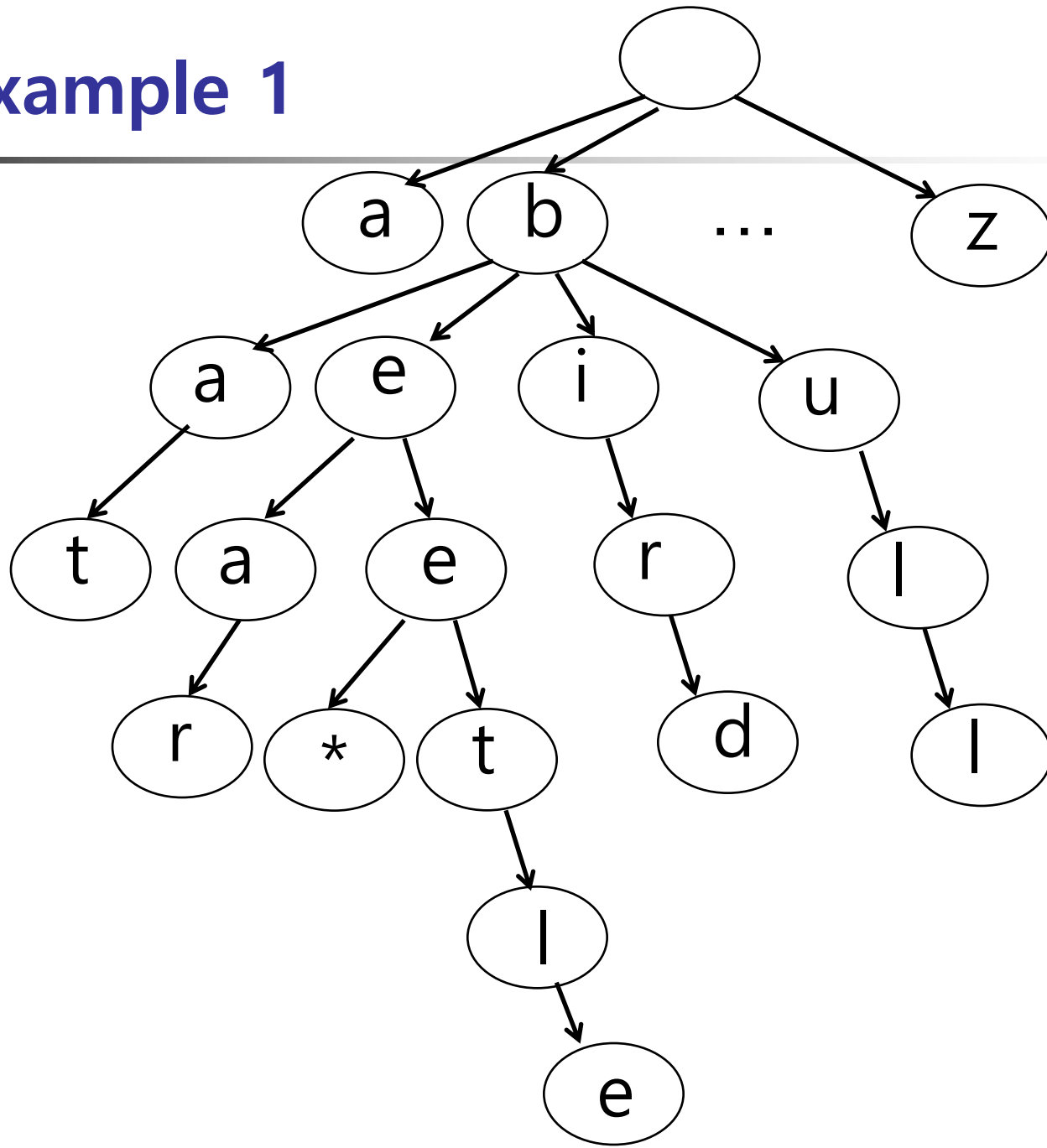




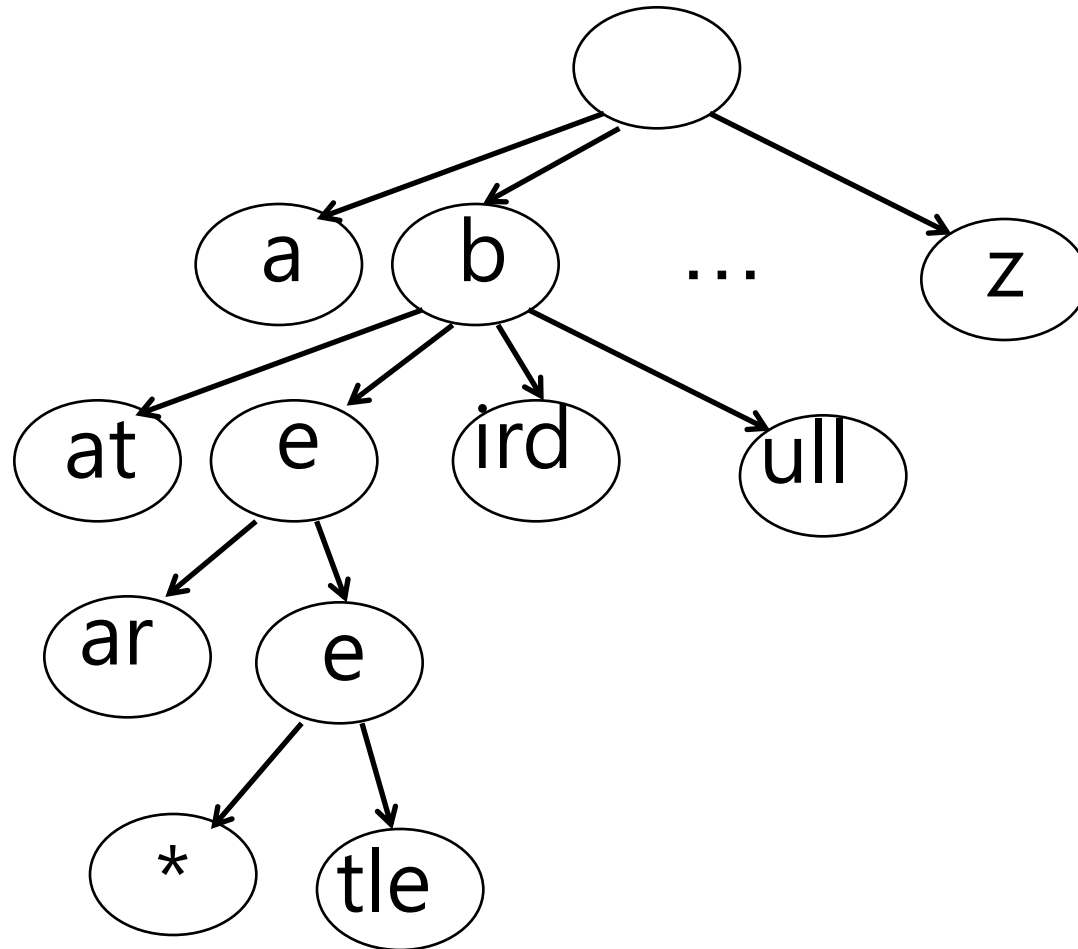
Multi-Way Trie: Use Example

- Stores a string over the alphabet across a chain of nodes.
- Usage
 - Storing a dictionary of words
 - Auto-complete dictionary, spell-checking and hyphenation software
- Keys at any level are the alphabet.
 - $M = 26 + 1$
- Keys are not stored in the branch nodes.
- Final portion of the keys are stored in the leaf nodes.

Example 1



Example 2



Implementation

bee beetle

root



2nd level

a b c

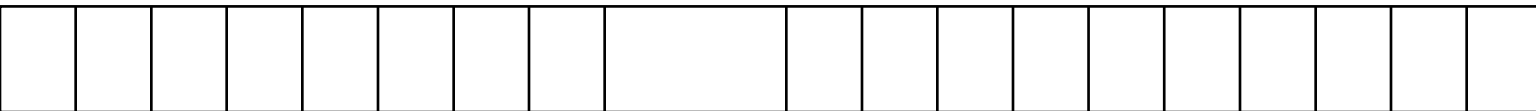
z



3rd level

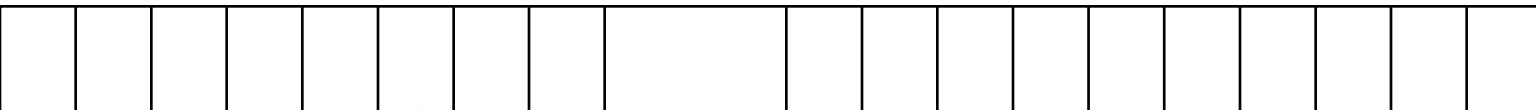
e

u



4th level

e



bee

t₉₄



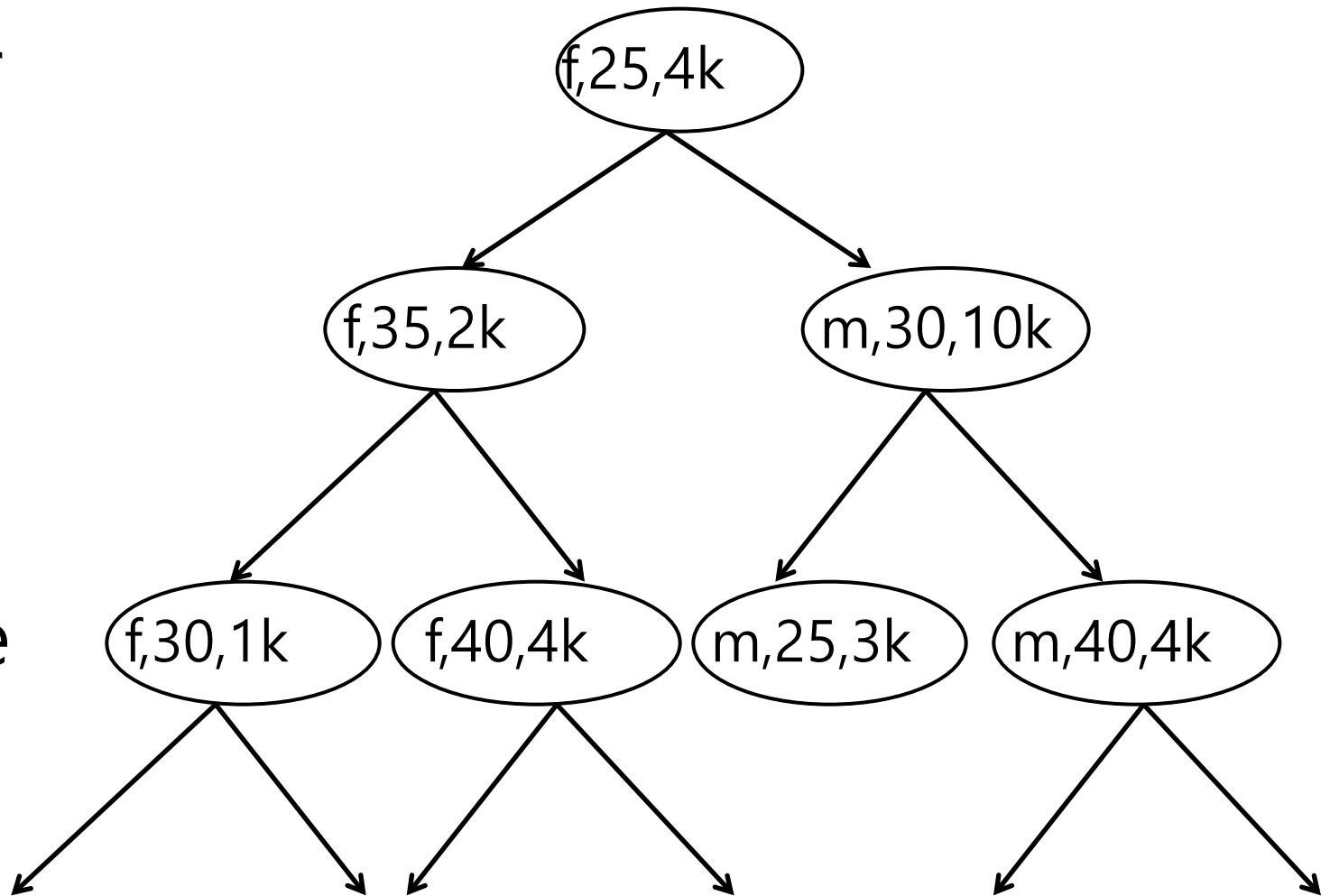
Data Structures for Partial Key Search

k-d Tree: Partial Key Search

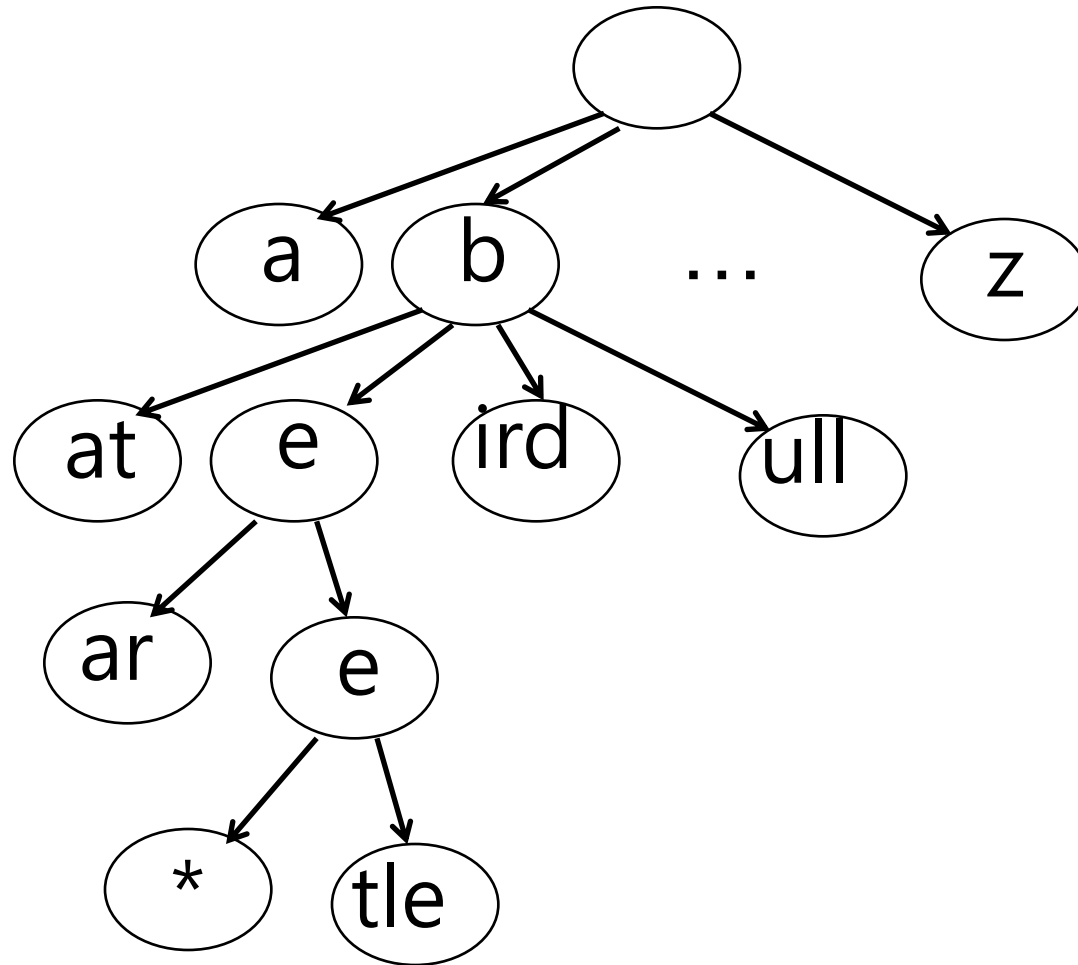
gender

age

income



M-Way Trie





Reprised

- Binary Search Trees
 - AVL tree, binary trie
 - k-d tree, T-tree
- Multi-Way Search Trees
 - m-way trie, quad tree, oct tree
- Height-Balanced Search Trees
 - AVL tree, T-tree, red-black tree
- Perfectly Height-Balanced Search Trees
 - 2-3 tree, (B tree, R-tree)
- Spatial Search Trees
 - quad tree, oct tree, k-d tree
- Trees that Support **Partial Searches**
 - all search trees
- **Trees that Support Partial Key Searches**
 - k-d trees, tries (binary, m-way)



Search Performance: Best, Avg, Worst (1)

- Array
 - $O(1)$, $O(n)$, $O(n)$
- Binary Search
 - $O(1)$, $O(\log_2 n)$, $O(\log_2 n)$
- Binary Search Tree
 - $O(1)$, $O(\log_2 n)$, $O(n)$
- AVL, Red-Black (Binary Search) Tree
 - $O(1)$, $O(\log_2 n)$, $O(\log_2 n)$
- Quad Tree
 - $O(1)$, $O(\log_4 n)$, $O(n)$
- k-d Tree
 - $O(1)$, $O(\log_2 n)$, $O(n)$
- T-Tree (with m-element array)
 - $O(1)$, $O(\log_2 n/m)$, $O(\log_2 n/m)$



Search Performance: Best, Avg, Worst (2)

- Binary Trie (with k -bit keys)
 - $O(1)$, $O(k)$, $O(k)$
- M-Way Trie (with k -digit keys)
 - $O(1)$, $O(k)$, $O(k)$



End of Lecture
