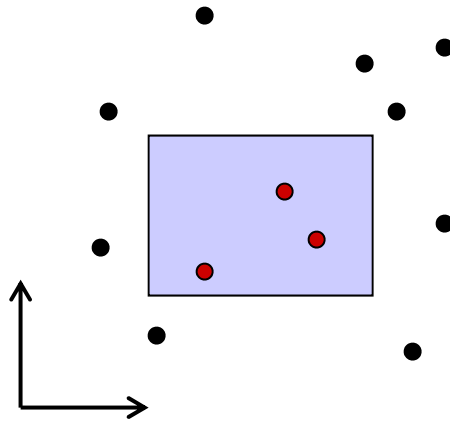


# CMPS 3130/6130 Computational Geometry Spring 2015



## *Orthogonal Range Searching*

**Carola Wenk**

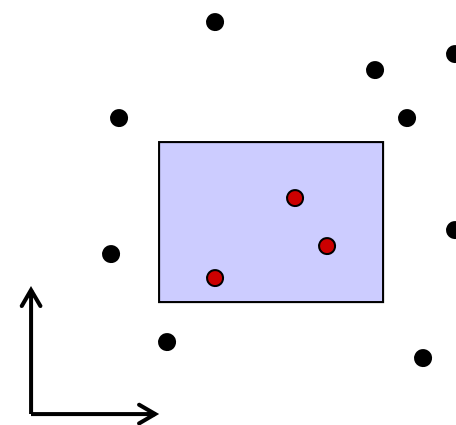
# Orthogonal range searching

**Input:**  $n$  points in  $d$  dimensions

- E.g., representing a database of  $n$  records each with  $d$  numeric fields

**Query:** Axis-aligned *box* (in 2D, a rectangle)

- Report on the points inside the box:
  - Are there any points?
  - How many are there?
  - List the points.



# Orthogonal range searching

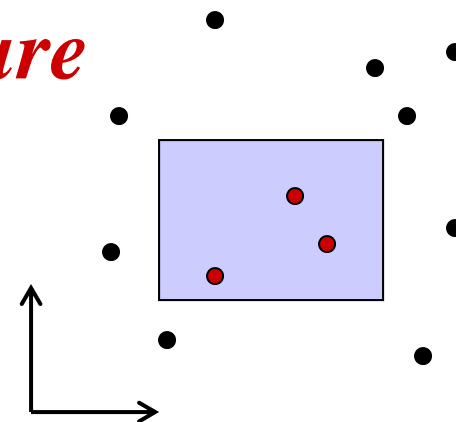
**Input:**  $n$  points in  $d$  dimensions

**Query:** Axis-aligned *box* (in 2D, a rectangle)

- Report on the points inside the box

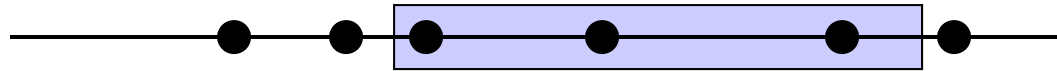
**Goal:** Preprocess points into a data structure to support fast queries

- Primary goal: *Static data structure*
- In 1D, we will also obtain a dynamic data structure supporting insert and delete



# 1D range searching

In 1D, the query is an interval:



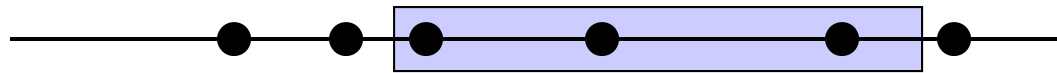
First solution:

- Sort the points and store them in an array
  - Solve query by binary search on endpoints.
  - Obtain a static structure that can list  $k$  answers in a query in  $O(k + \log n)$  time.

**Goal:** Obtain a dynamic structure that can list  $k$  answers in a query in  $O(k + \log n)$  time.

# 1D range searching

In 1D, the query is an interval:

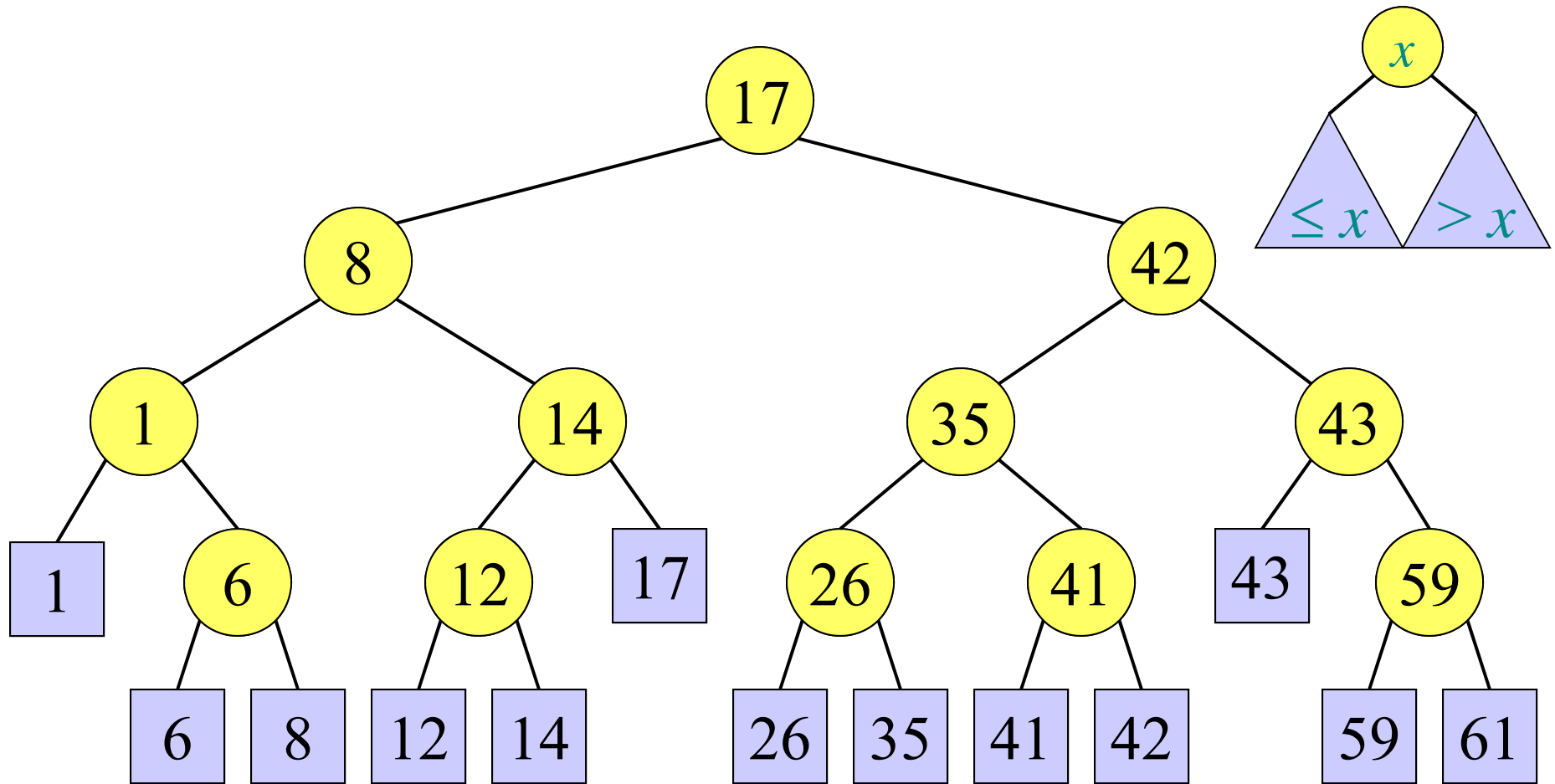


New solution that extends to higher dimensions:

- Balanced binary search tree
  - New organization principle:  
Store points in the *leaves* of the tree.
  - Internal nodes store copies of the leaves to satisfy binary search property:
    - Node  $x$  stores in  $key[x]$  the maximum key of any leaf in the left subtree of  $x$ .

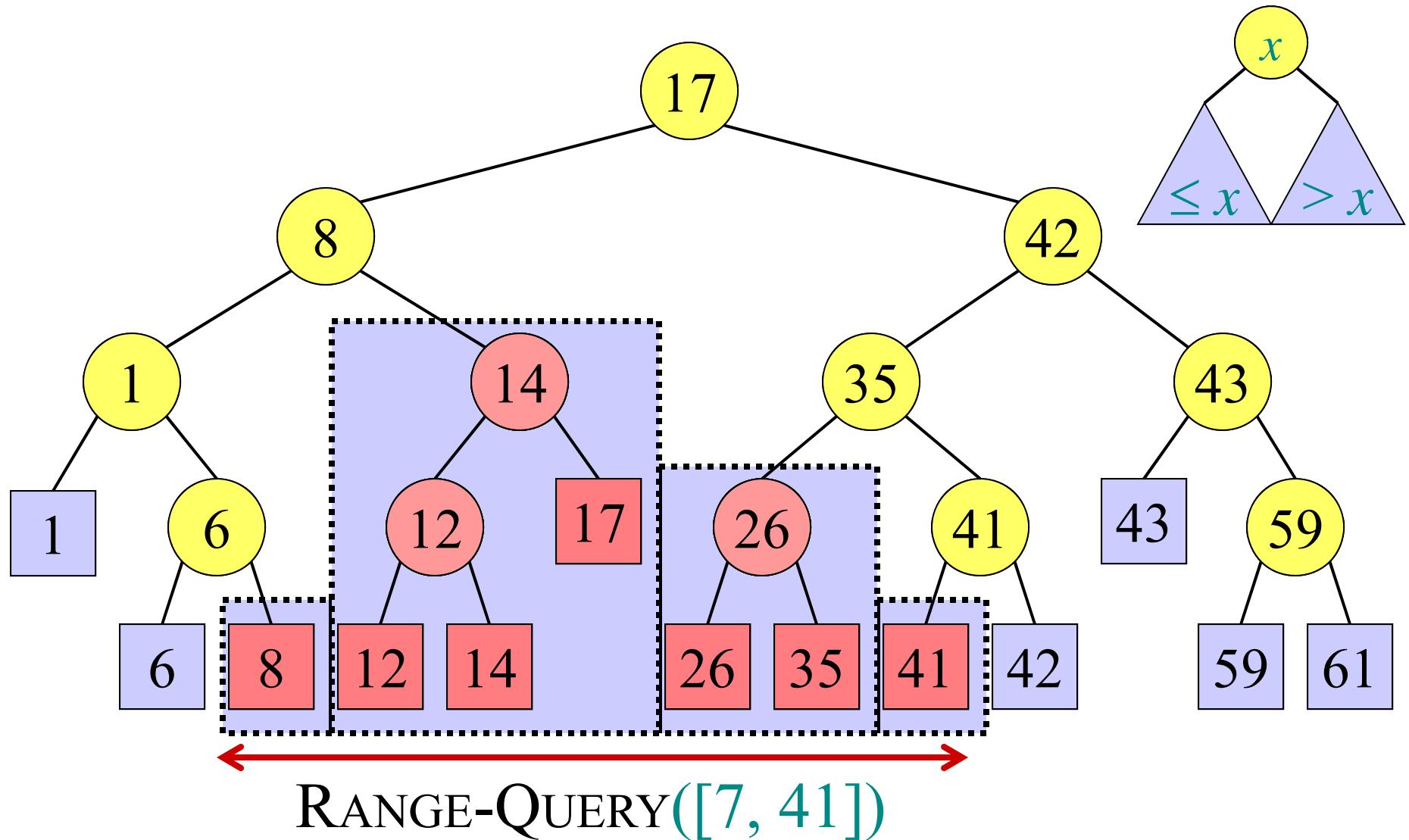


# Example of a 1D range tree



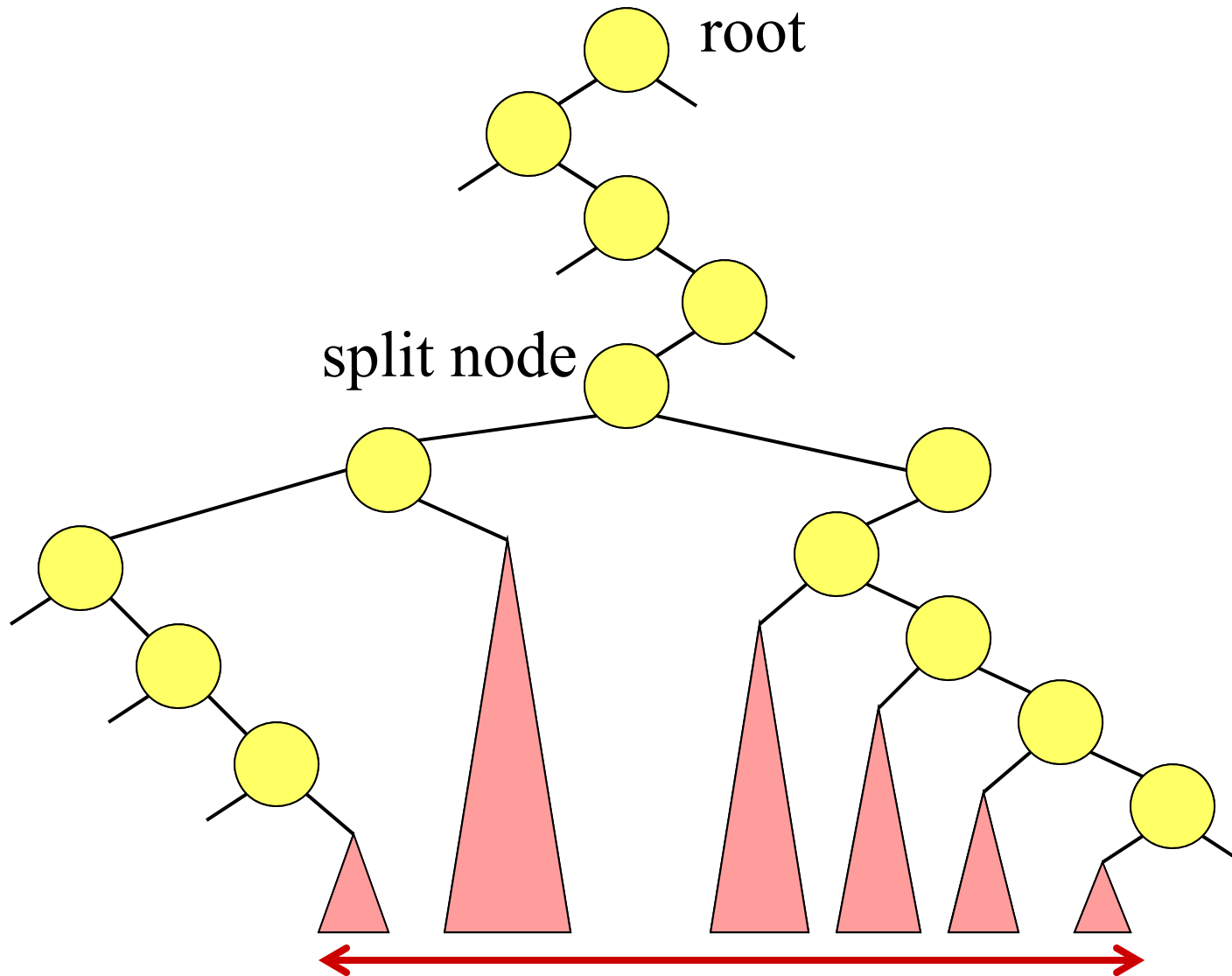
$key[x]$  is the maximum key of any leaf in the left subtree of  $x$ .

# Example of a 1D range query





# General 1D range query



# Pseudocode, part 1: Find the split node

1D-RANGE-QUERY( $T, [x_1, x_2]$ )

$w \leftarrow \text{root}[T]$

**while**  $w$  is not a leaf and  $(x_2 \leq \text{key}[w]$  or  $\text{key}[w] < x_1)$

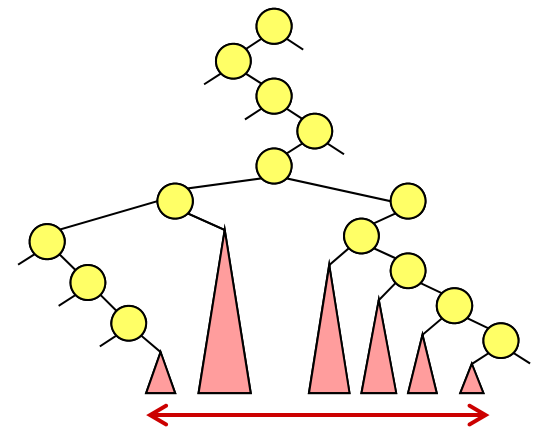
**do if**  $x_2 \leq \text{key}[w]$

**then**  $w \leftarrow \text{left}[w]$

**else**  $w \leftarrow \text{right}[w]$

//  $w$  is now the split node

*[traverse left and right from  $w$  and report relevant subtrees]*



# Pseudocode, part 2: Traverse left and right from split node

1D-RANGE-QUERY( $T, [x_1, x_2]$ )

*[find the split node]*

//  $w$  is now the split node

**if**  $w$  is a leaf

**then** output the leaf  $w$  if  $x_1 \leq \text{key}[w] \leq x_2$

**else**  $v \leftarrow \text{left}[w]$

// Left traversal

**while**  $v$  is not a leaf

**do if**  $x_1 \leq \text{key}[v]$

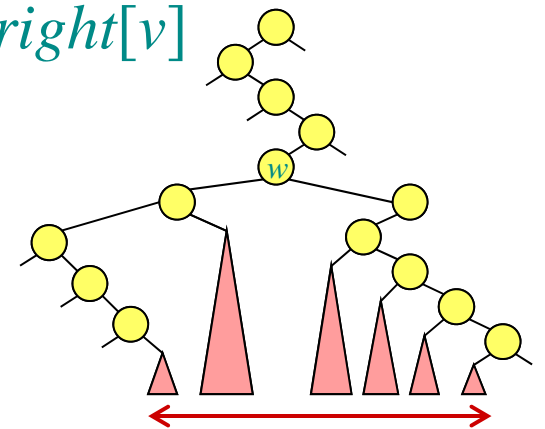
**then** output the subtree rooted at  $\text{right}[v]$

$v \leftarrow \text{left}[v]$

**else**  $v \leftarrow \text{right}[v]$

output the leaf  $v$  if  $x_1 \leq \text{key}[v] \leq x_2$

*[symmetrically for right traversal]*



# Analysis of 1D-RANGE-QUERY

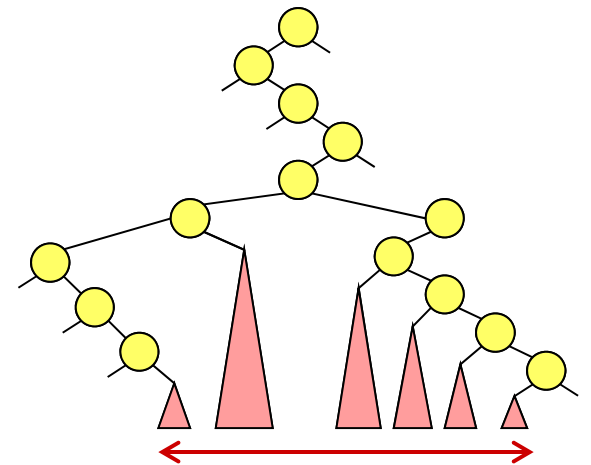
**Query time:** Answer to range query represented by  $O(\log n)$  subtrees found in  $O(\log n)$  time.

Thus:

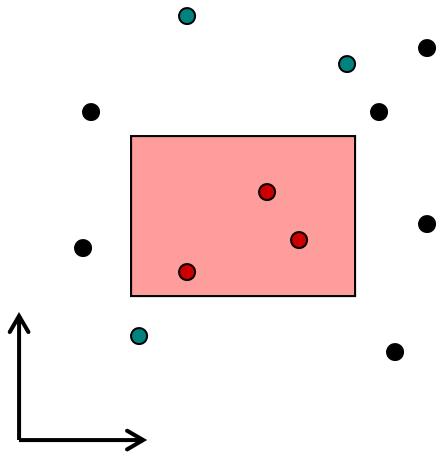
- Can test for points in interval in  $O(\log n)$  time.
- Can report all  $k$  points in interval in  $O(k + \log n)$  time.
- Can count points in interval in  $O(\log n)$  time

**Space:**  $O(n)$

**Preprocessing time:**  $O(n \log n)$



# 2D range trees

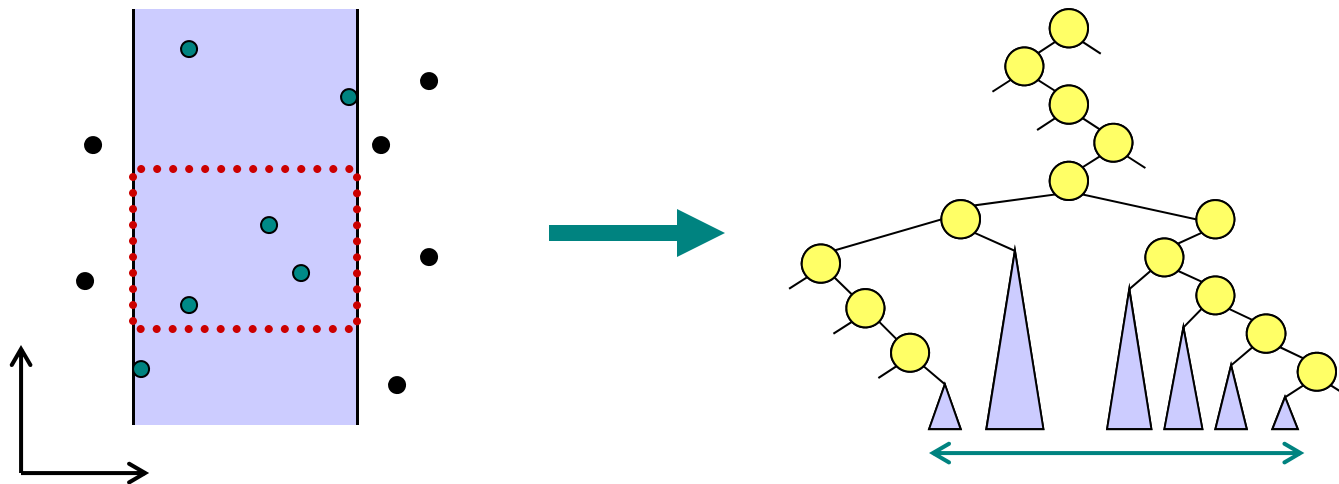


# 2D range trees

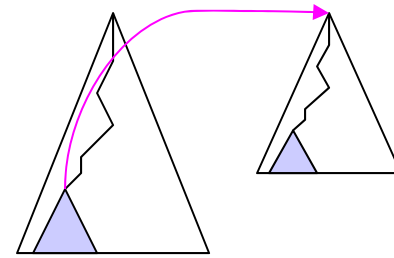
Store a *primary* 1D range tree for all the points based on  $x$ -coordinate.

Thus in  $O(\log n)$  time we can find  $O(\log n)$  subtrees representing the points with proper  $x$ -coordinate.

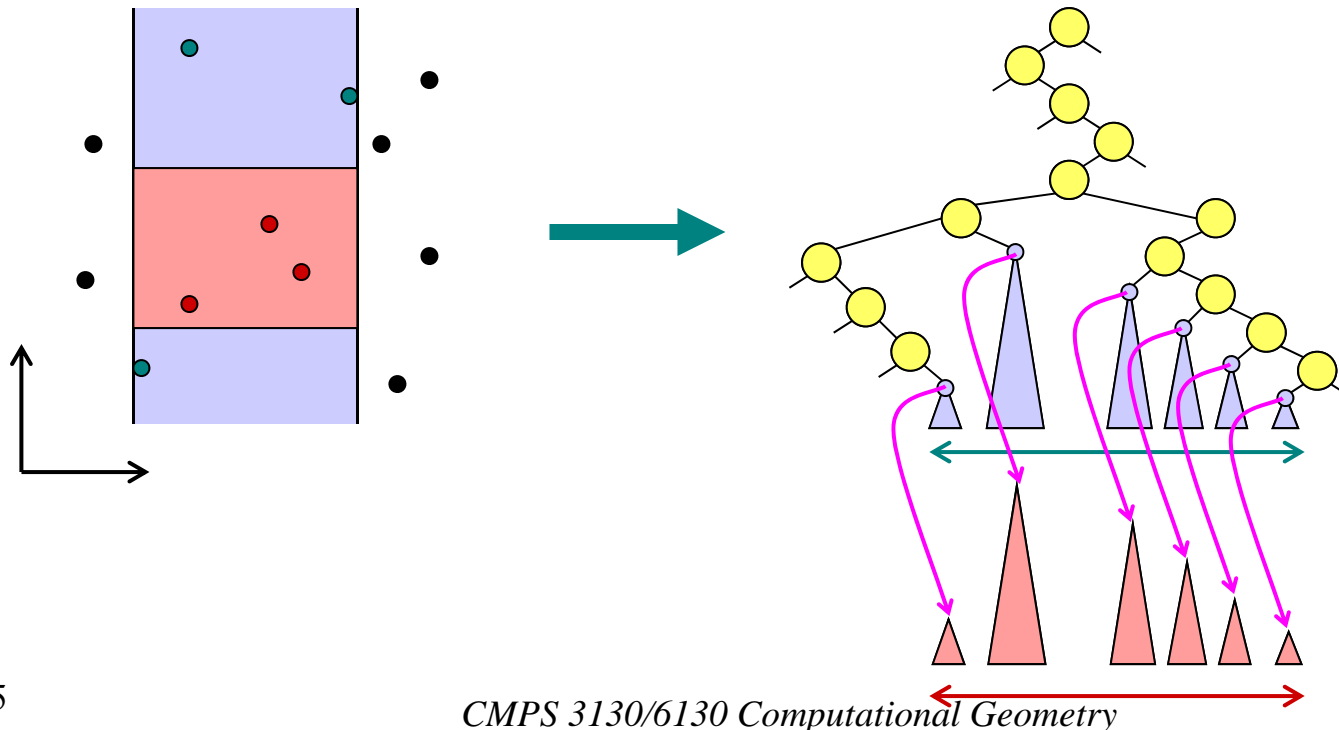
How to restrict to points with proper  $y$ -coordinate?



# 2D range trees

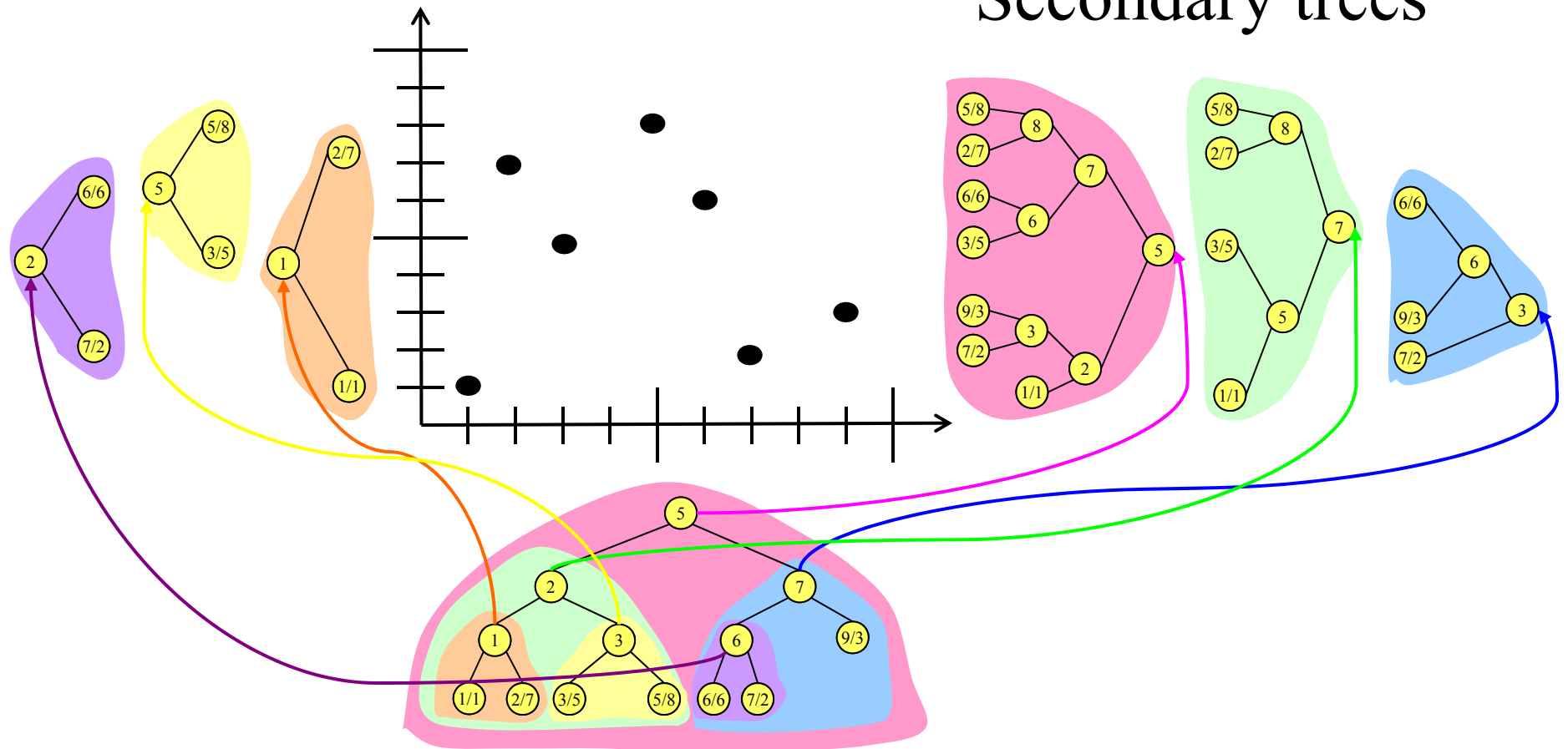


**Idea:** In primary 1D range tree of  $x$ -coordinate, every node stores a *secondary* 1D range tree based on  $y$ -coordinate for all points in the subtree of the node. Recursively search within each.



# 2D range tree example

Secondary trees



Primary tree



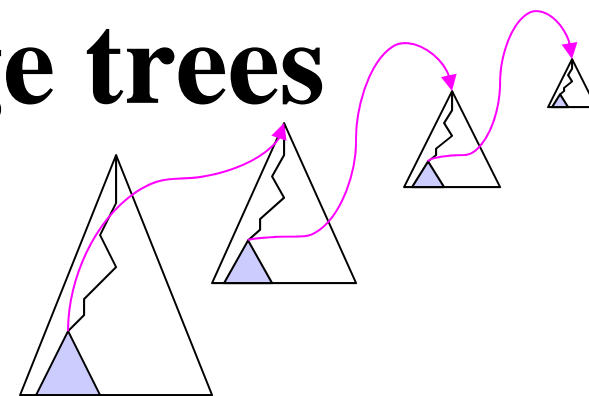
# Analysis of 2D range trees

**Query time:** In  $O(\log^2 n) = O((\log n)^2)$  time, we can represent answer to range query by  $O(\log^2 n)$  subtrees. Total cost for reporting  $k$  points:  $O(k + (\log n)^2)$ .

**Space:** The secondary trees at each level of the primary tree together store a copy of the points. Also, each point is present in each secondary tree along the path from the leaf to the root. Either way, we obtain that the space is  $O(n \log n)$ .

**Preprocessing time:**  $O(n \log n)$

# $d$ -dimensional range trees



Each node of the secondary  $y$ -structure stores a tertiary  $z$ -structure representing the points in the subtree rooted at the node, etc.

Save one  $\log$  factor using fractional cascading

**Query time:**  $O(k + \log^d n)$  to report  $k$  points.

**Space:**  $O(n \log^{d-1} n)$

**Preprocessing time:**  $O(n \log^{d-1} n)$

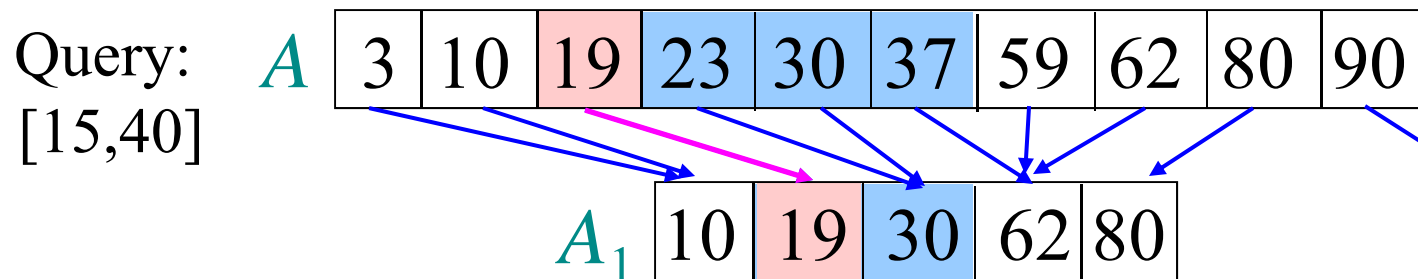
# Search in Subsets

**Given:** Two sorted arrays  $A_1$  and  $A$ , with  $A_1 \subseteq A$   
A query interval  $[l, r]$

**Task:** Report all elements  $e$  in  $A_1$  and  $A$  with  $l \leq e \leq r$

**Idea:** Add pointers from  $A$  to  $A_1$ :  
→ For each  $a \in A$  add a pointer to the smallest element  $b \in A_1$  with  $b \geq a$

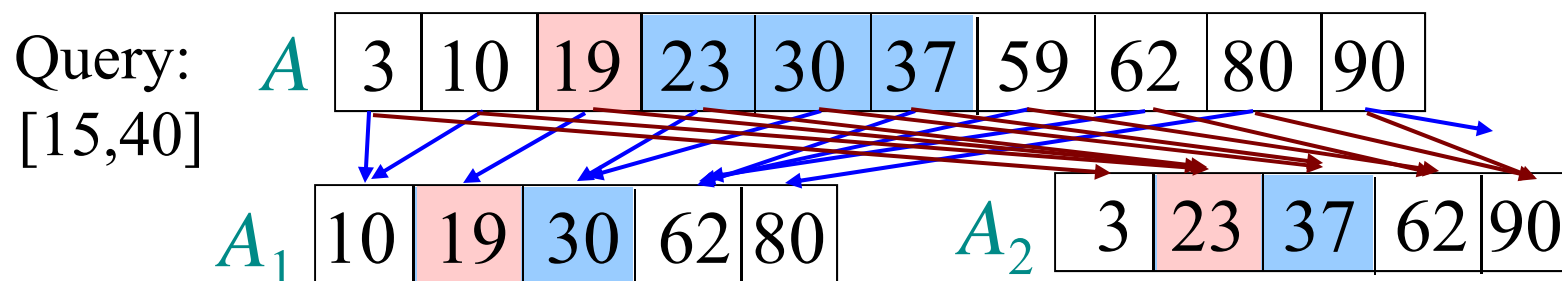
**Query:** Find  $l \in A$ , follow pointer to  $A_1$ . Both in  $A$  and  $A_1$  sequentially output all elements in  $[l, r]$ .



**Runtime:**  $O((\log n + k) + (1 + k)) = O(\log n + k)$

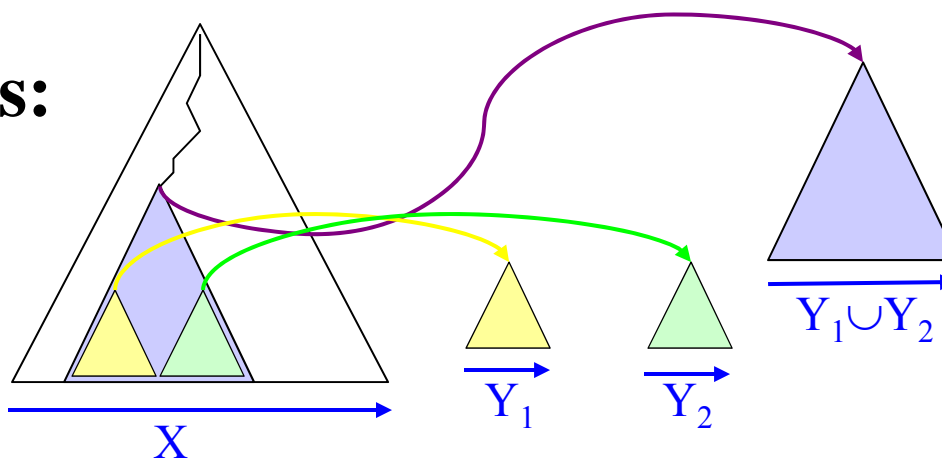
# Search in Subsets (cont.)

**Given:** Three sorted arrays  $A_1, A_2$ , and  $A$ ,  
with  $A_1 \subseteq A$  and  $A_2 \subseteq A$



**Runtime:**  $O((\log n + k) + (1+k) + (1+k)) = O(\log n + k)$

**Range trees:**



# Fractional Cascading: Layered Range Tree

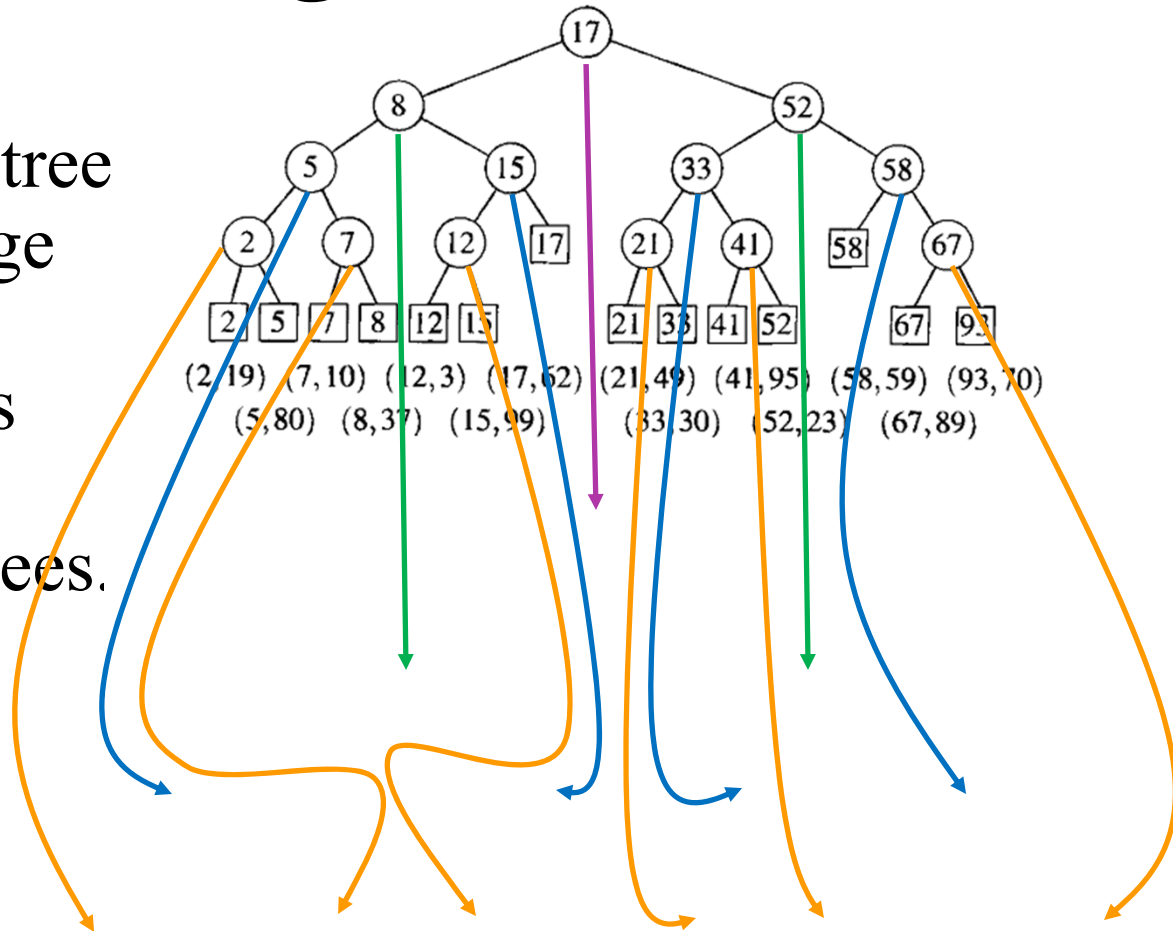
Replace 2D range tree with a layered range tree, using sorted arrays and pointers instead of the secondary range trees.

Preprocessing:

$$O(n \log n)$$

Query:

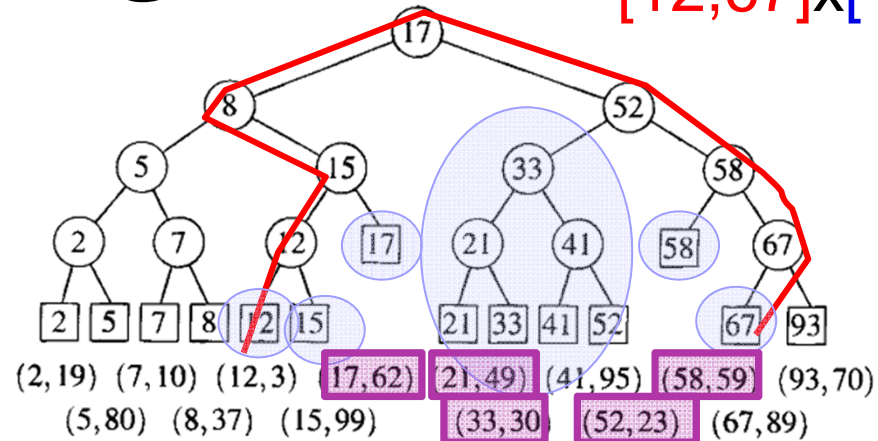
$$O(\log n + k)$$



# Fractional Cascading: Layered Range Tree

$[12,67] \times [19,70]$

Replace 2D range tree with a layered range tree, using sorted arrays and pointers instead of the secondary range trees.

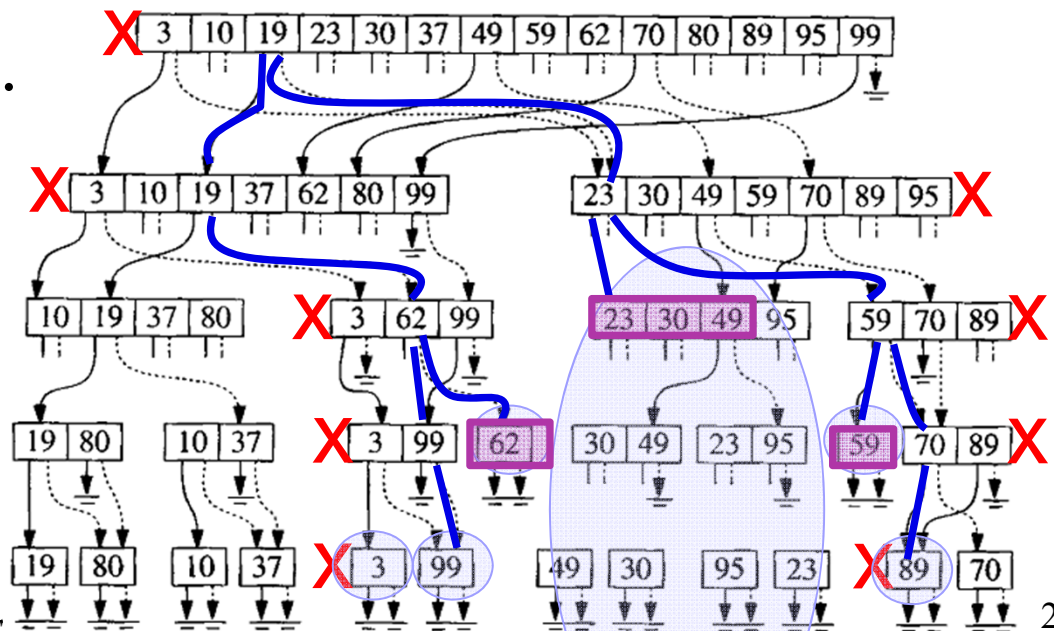


Preprocessing:

$$O(n \log n)$$

Query:

$$O(\log n + k)$$



# $d$ -dimensional range trees

**Query time:**  $O(k + \log^{d-1} n)$  to report  $k$  points,  
uses fractional cascading in the  
last dimension

**Space:**  $O(n \log^{d-1} n)$

**Preprocessing time:**  $O(n \log^{d-1} n)$

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**Best data structure to date:**

**Query time:**  $O(k + \log^{d-1} n)$  to report  $k$  points.

**Space:**  $O(n (\log n / \log \log n)^{d-1})$

**Preprocessing time:**  $O(n \log^{d-1} n)$