

# Geometric range searching

# Where we are

## “Global” problems

- closest pair
- convex hull
- intersections
- ..

## Geometric search problems

- range searching
- nearest neighbor
- k-nearest neighbor
- find all roads within 1km of current location
- ..

## Techniques

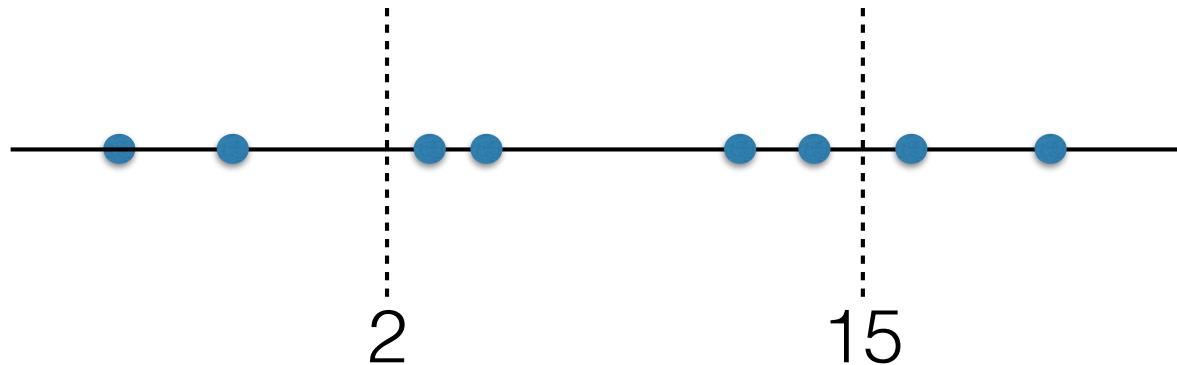
- divide-and-conquer
- incremental
- space decomposition
- plane sweep
- ..

next

We start with 1D range searching

# 1D Range searching

Given a set of  $n$  points on the real line and an interval  $[a,b]$ , find all points in  $[a,b]$



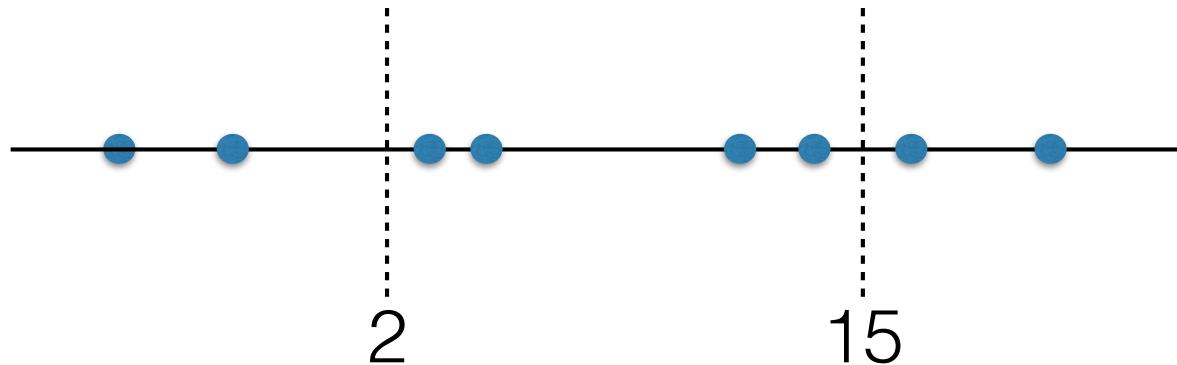
find all values in  $[2,15]$

Assume first that the points are **fixed**, i.e. don't change.

What can we do?

# 1D Range searching

Given a set of  $n$  points on the real line and an interval  $[a,b]$ , find all points in  $[a,b]$



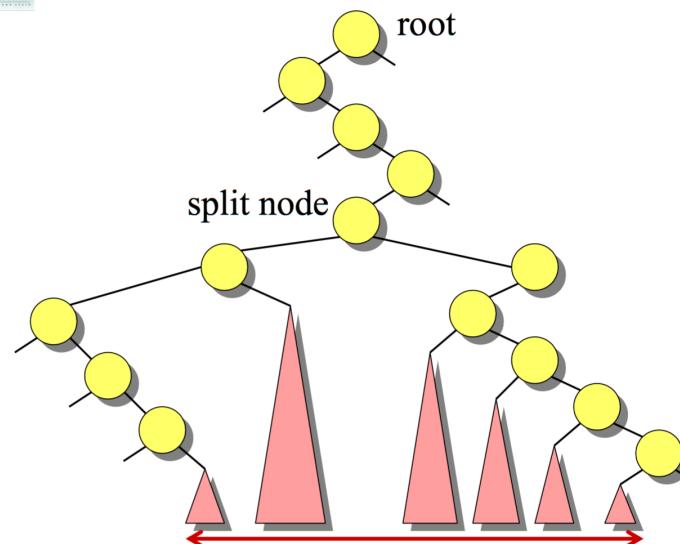
Assume now that the points are **dynamic**, i.e. in addition to range queries, we want to be able to **insert** and **delete** points.

# 1D Range searching

- A set of  $n$  points in 1D can be pre-processed into a BBST such that:
  - Build:  $O(n \lg n)$
  - Space:  $\Theta(n)$
  - Range queries:  $O(\lg n + k)$
  - Dynamic: points can be inserted/deleted in  $O(\lg n)$



## General 1D range query



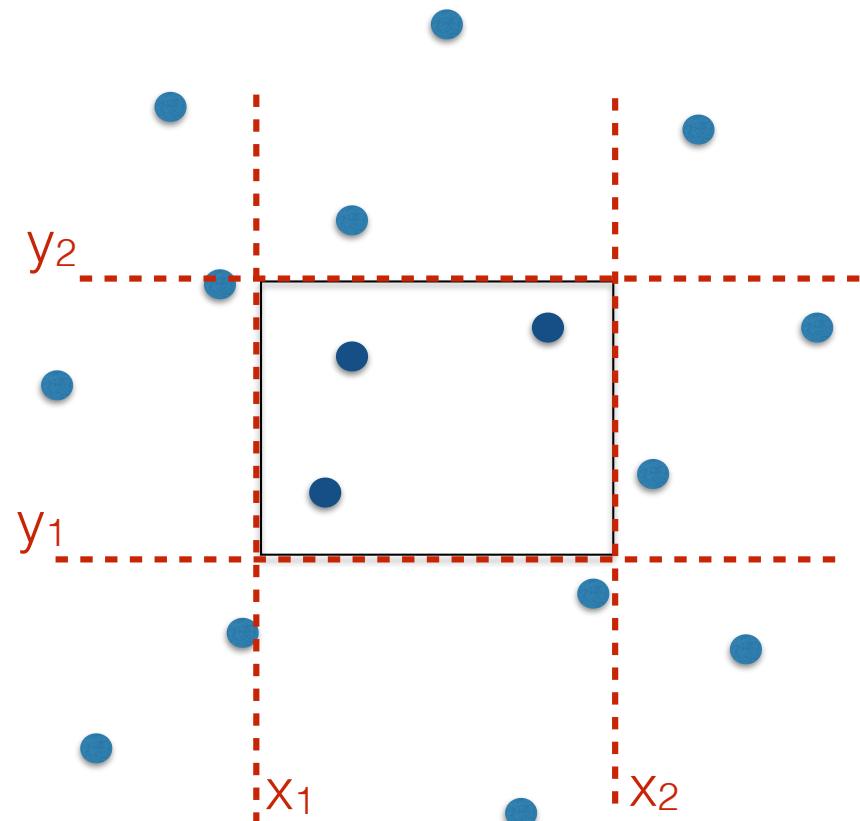
The  $k$  points in the range sit in  $O(\lg n)$  subtrees

# 1D Range searching

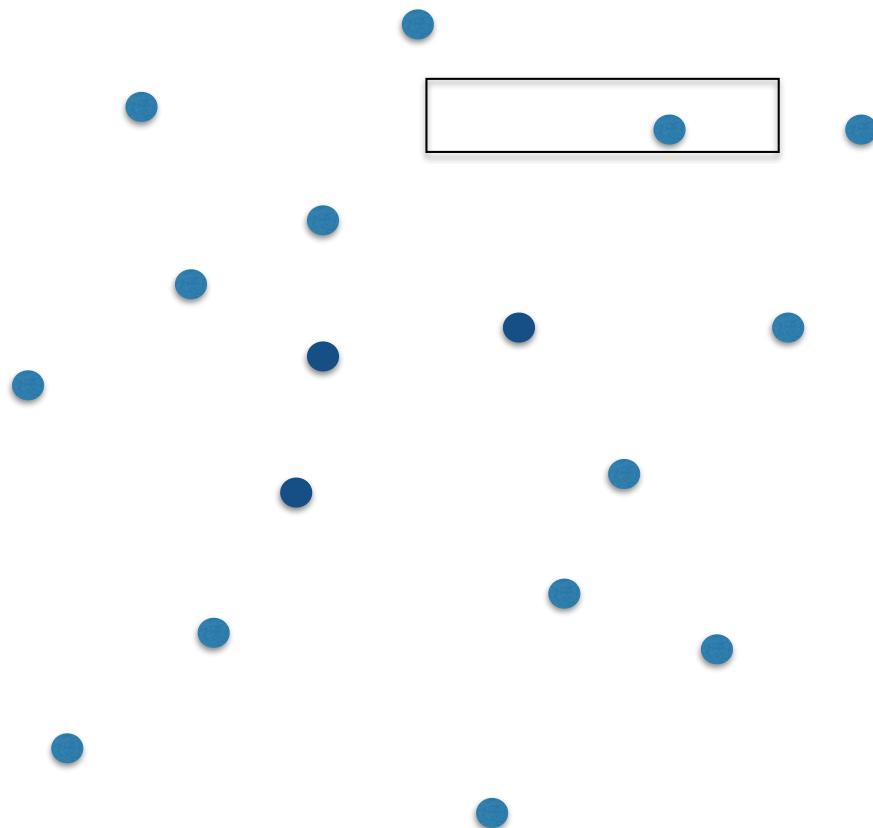
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## 2D Range searching

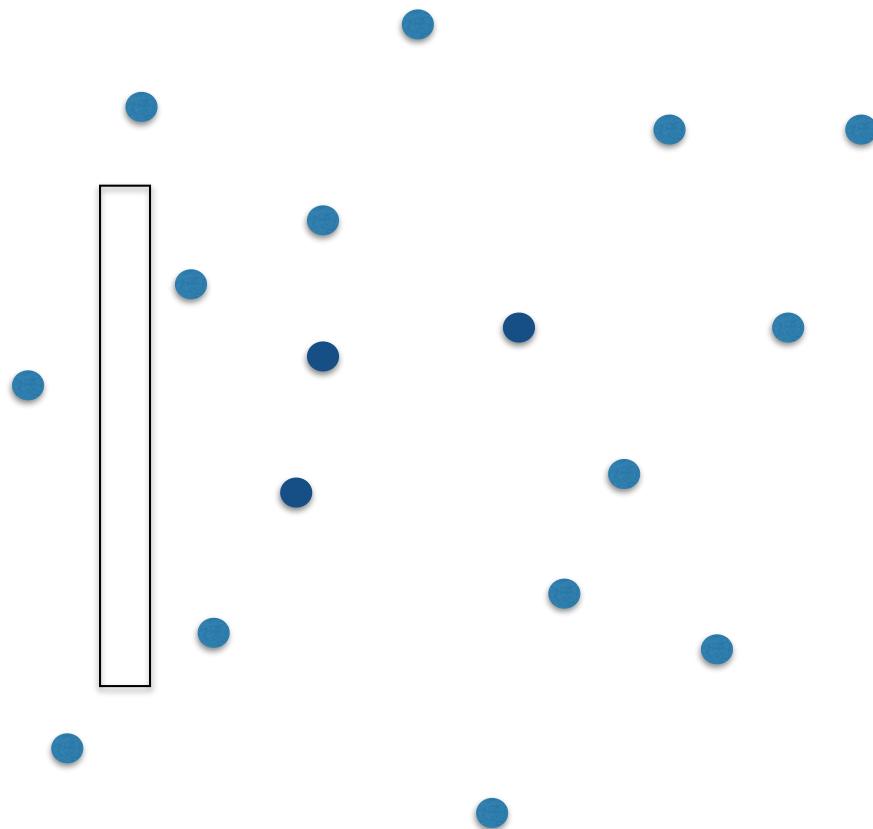
Given a set of  $n$  points in 2D and an arbitrary range  $[x_1, x_2] \times [y_1, y_2]$ , find all points in this range



## 2D Range searching



## 2D Range searching

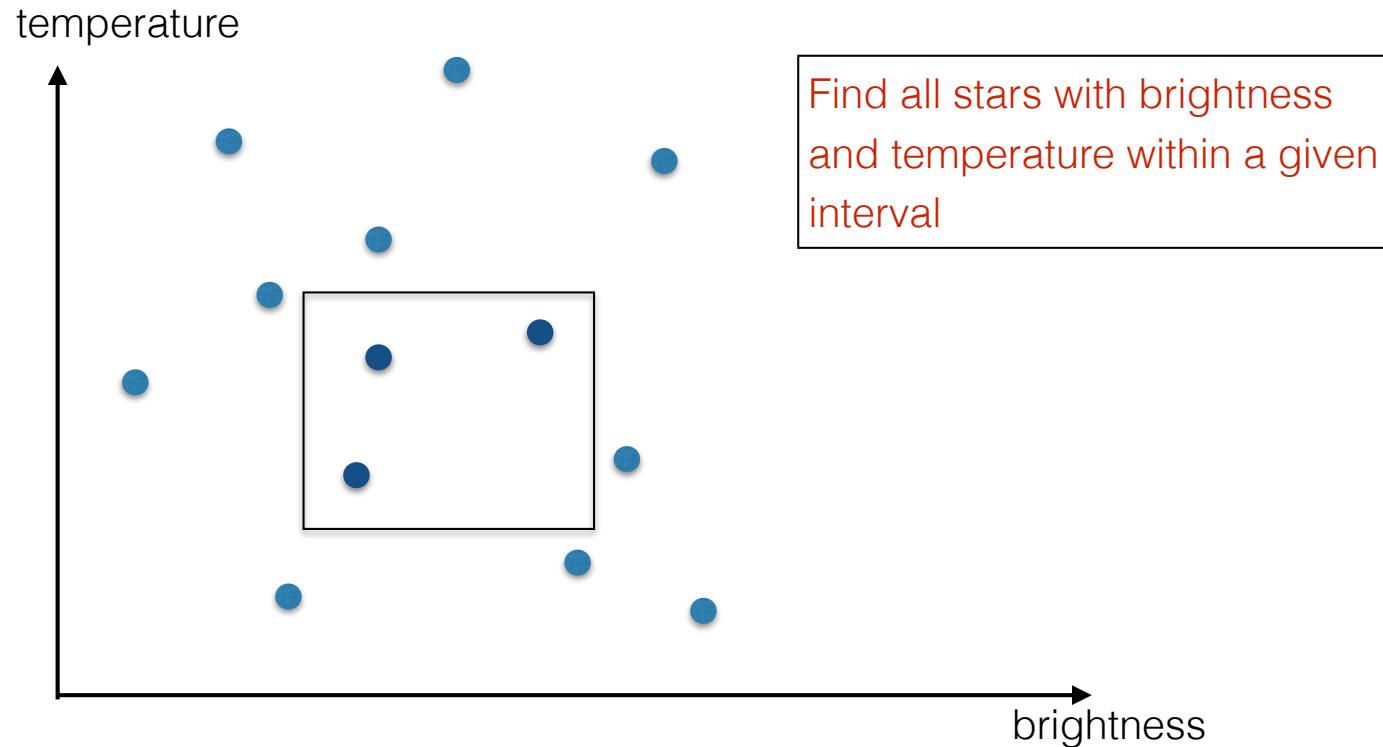


# Why range searching?

Searching is a fundamental operation. This is the multi-dimensional version of the “report all points in this interval”

Interestingly, it comes up in settings that are not geometrical

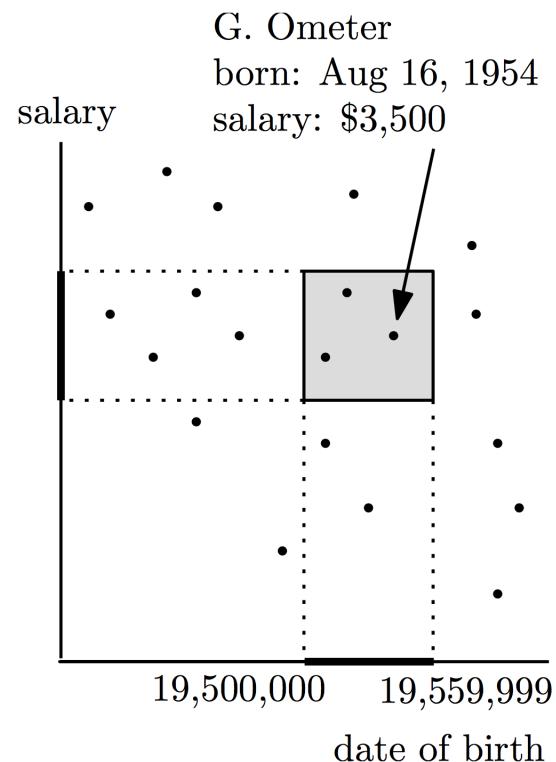
e.g. Database of stars. A star = (brightness, temperature,.....)



# Why range searching?

e.g. Database of employees. An employee = (age, salary,.....)

A database query may ask for all employees with age between  $a_1$  and  $a_2$ , and salary between  $s_1$  and  $s_2$

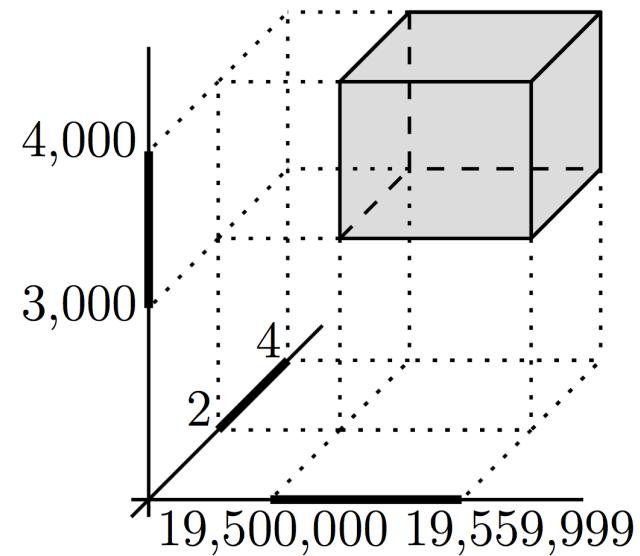


screenshot from Mark van Kreveld slides at <http://www.cs.uu.nl/docs/vakken/ga/slides5a.pdf>

# Why range searching?

3d-range searching, etc

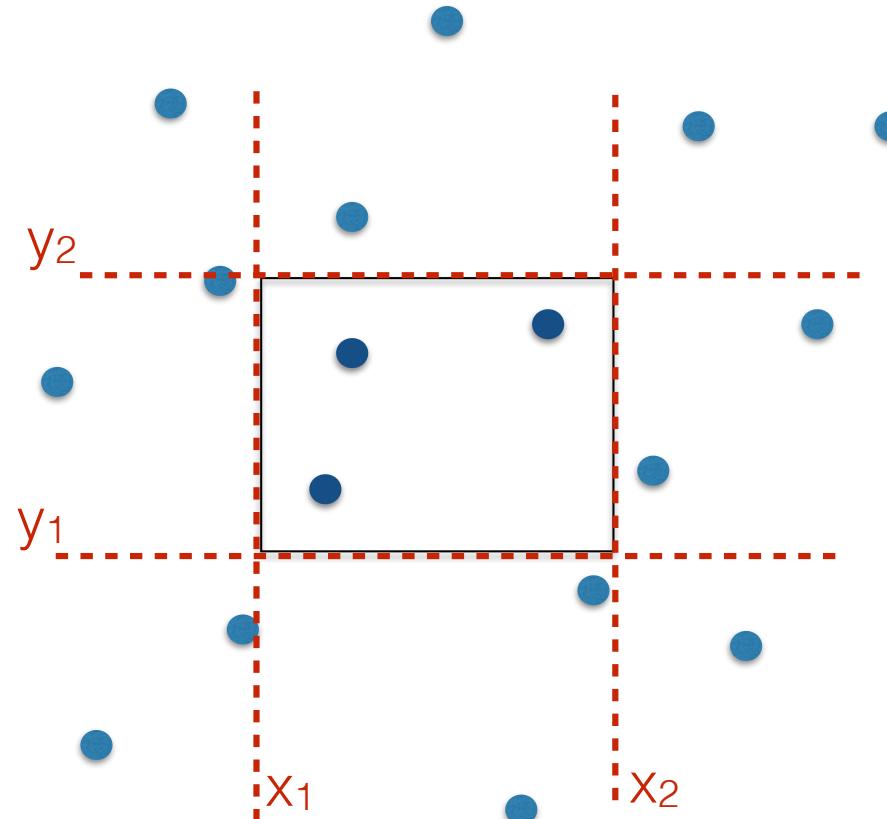
Example of a 3-dimensional (orthogonal) range query:  
children in  $[2, 4]$ , salary in  $[3000, 4000]$ , date of birth in  $[19,500,000, 19,559,999]$



screenshot from Mark van Kreveld slides at <http://www.cs.uu.nl/docs/vakken/ga/slides5a.pdf>

## 2D Range searching

Given a set of  $n$  points in 2D and an arbitrary range  $[x_1, x_2] \times [y_1, y_2]$ , find all points in this range



Build a structure to  
answer this efficiently

## 1D

- A set of  $n$  points in 1D can be pre-processed into a BBST such that:
  - Build:  $O(n \lg n)$
  - Space:  $\Theta(n)$
  - Range queries:  $O(\lg n + k)$
  - Dynamic: points can be inserted/deleted in  $O(\lg n)$

## 2D

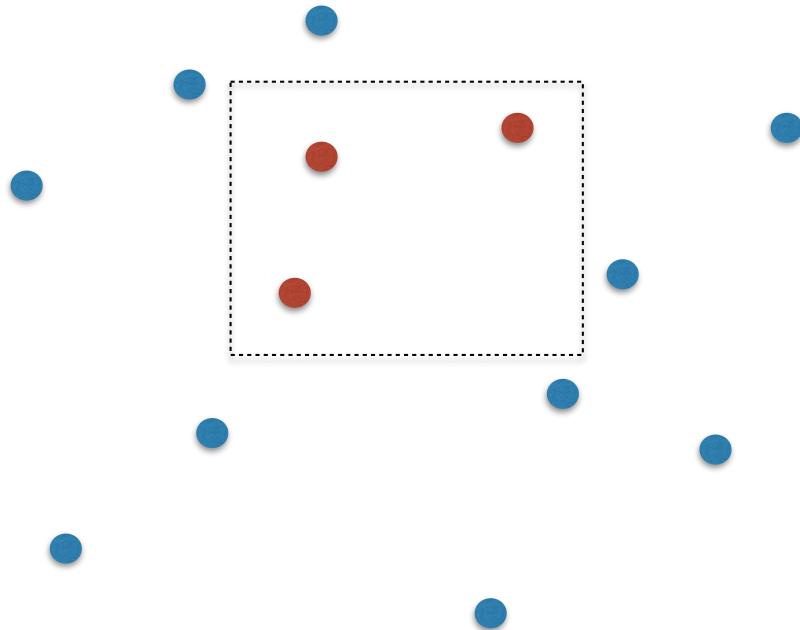
- A set of  $n$  points in 2D can be pre-processed in a ??2d-BBST??? such that
  - Build:  $O(n \lg n)$
  - Space:  $\Theta(n)$
  - Range queries:  $O(\lg^2 n + k)$

These bounds would be nice

But how?

## 2D Naive Approach

- n: size of the input (number of points)
- k: size of output (number of points inside range)



Points are static or dynamic?

We'll assume static (it's hard enough)

The naive approach: just traverse and check in  $O(n)$

Analysis:

Build: none

Space: none

2d range query:  $O(n)$

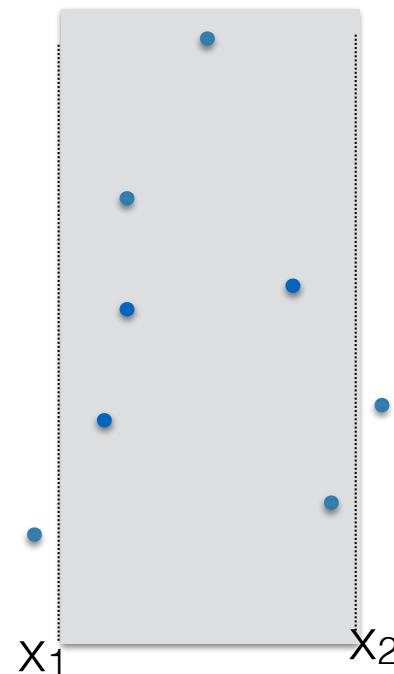
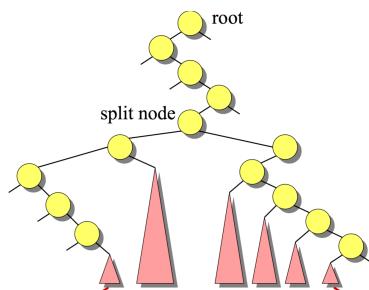
← We want  $O(\lg^2 n + k)$

query  $[x_1, x_2] \times [y_1, y_2]$

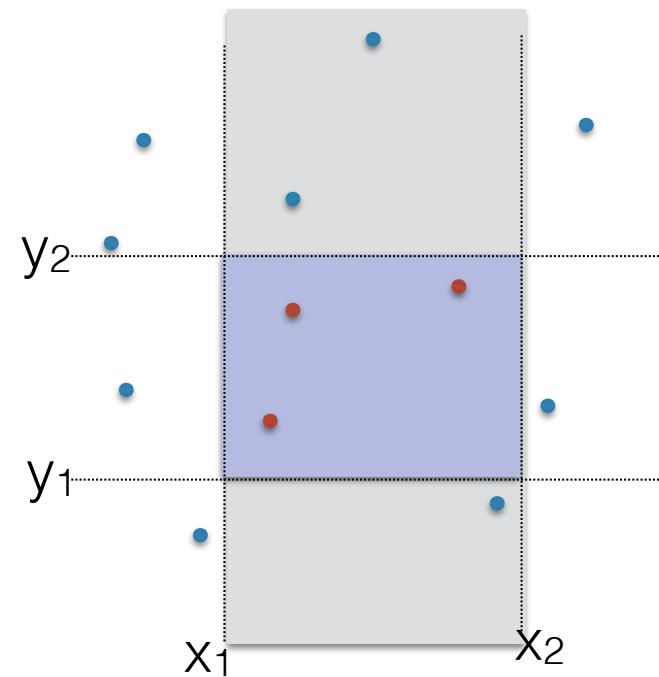
How about this:

1. Find all points with the x-coords in  $[x_1, x_2]$

BBST in x-order

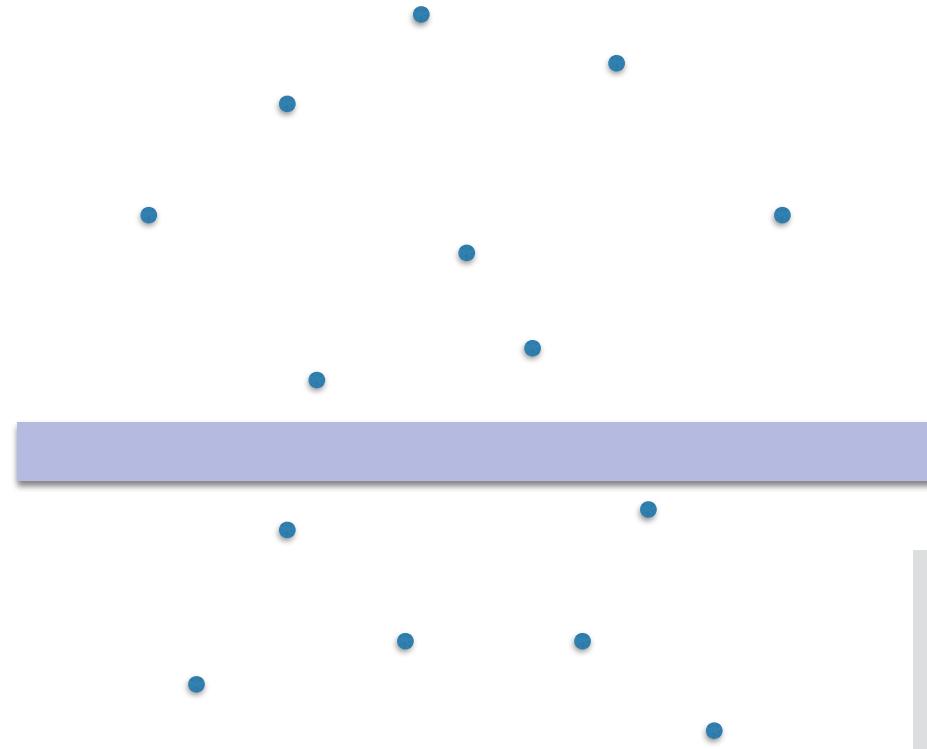


2. Traverse these points and find those with y-coord in  $[y_1, y_2]$



- This works, but it's worst case is slow
- $O(\lg n + k')$  to find the points in the vertical strip, and then  $O(k')$  to traverse and find the points in  $[y_1, y_2]$

- The problem is that the nb. of points in  $[x_1, x_2]$  can be large, and the nb. of points in  $[x_1, x_2] \times [y_1, y_2]$  may be small
- Worst case:  $k' = n, k = 0$



to find the points in this range takes  $O(\lg n + n) = O(n)$  time

# Searching via space partition structures

We'll partition the space, store it in a data structure and use it to speed up searching

[the grid heuristic](#)

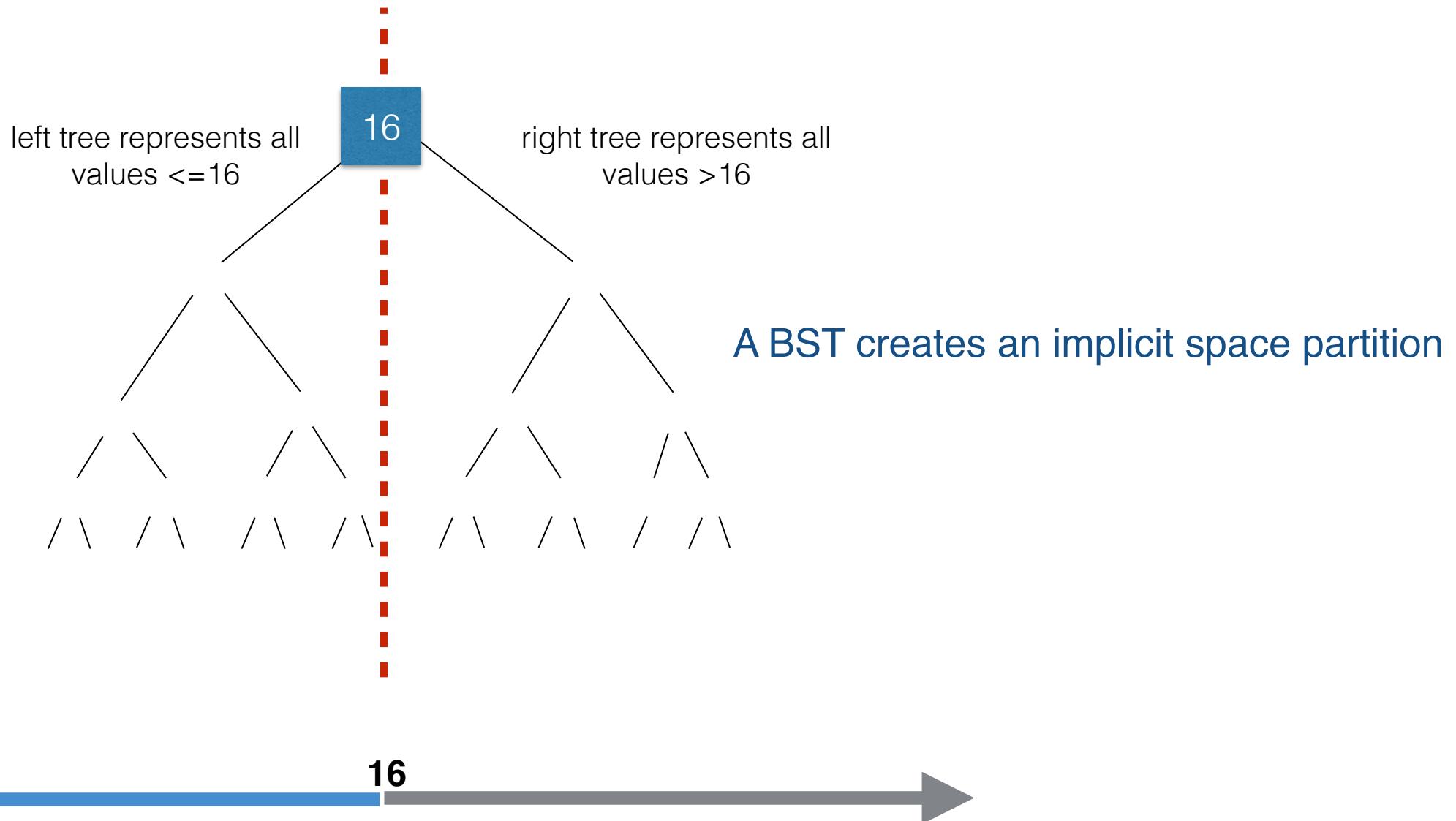
[kd-trees](#)

[range-trees](#)

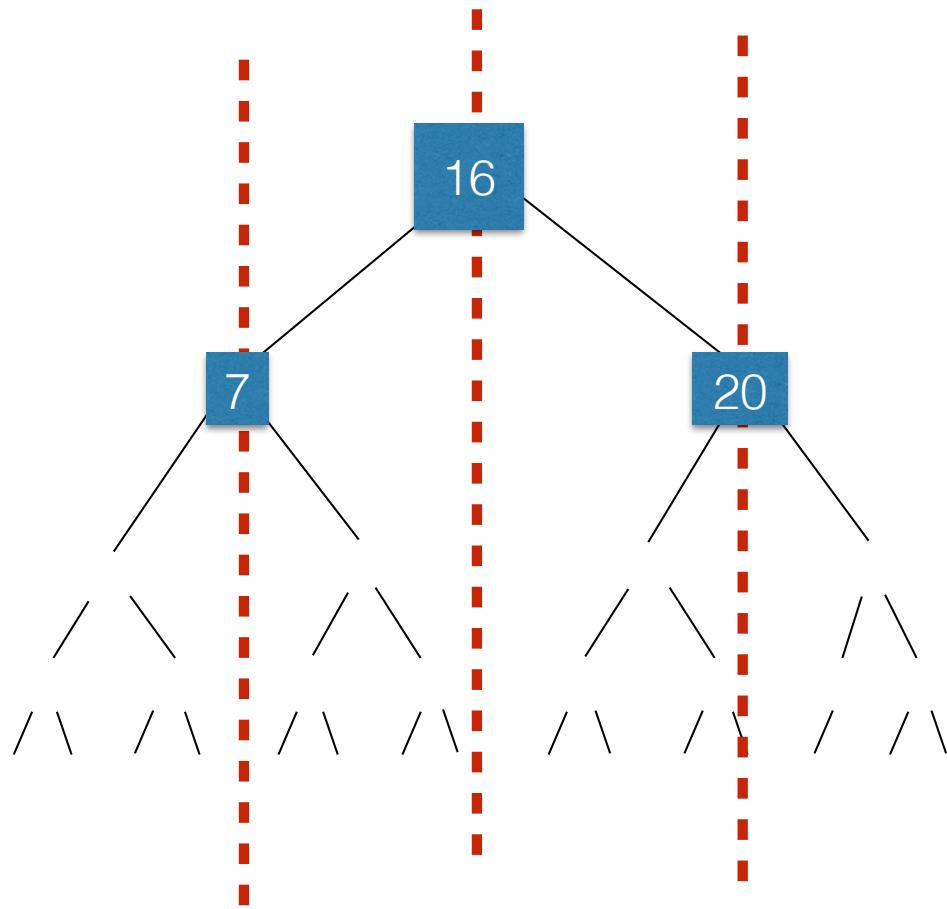
First, let's look in 1D:

The BBST as a space partition structure

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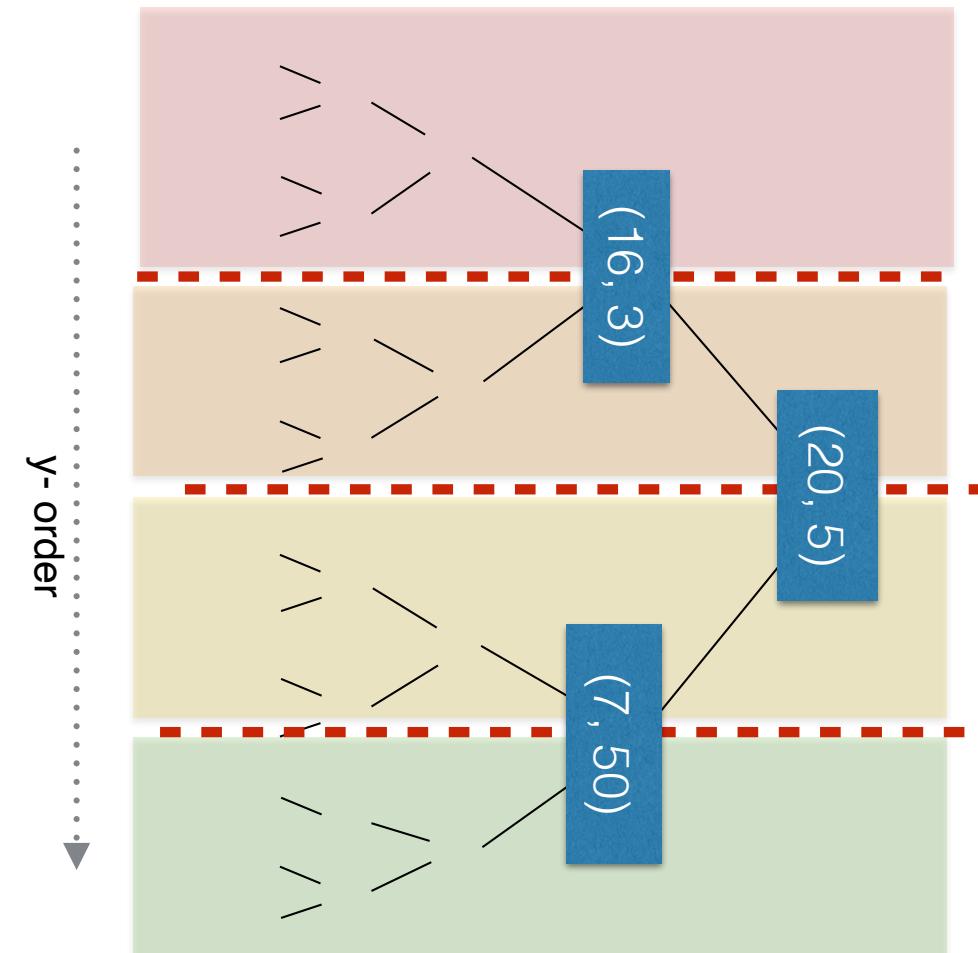
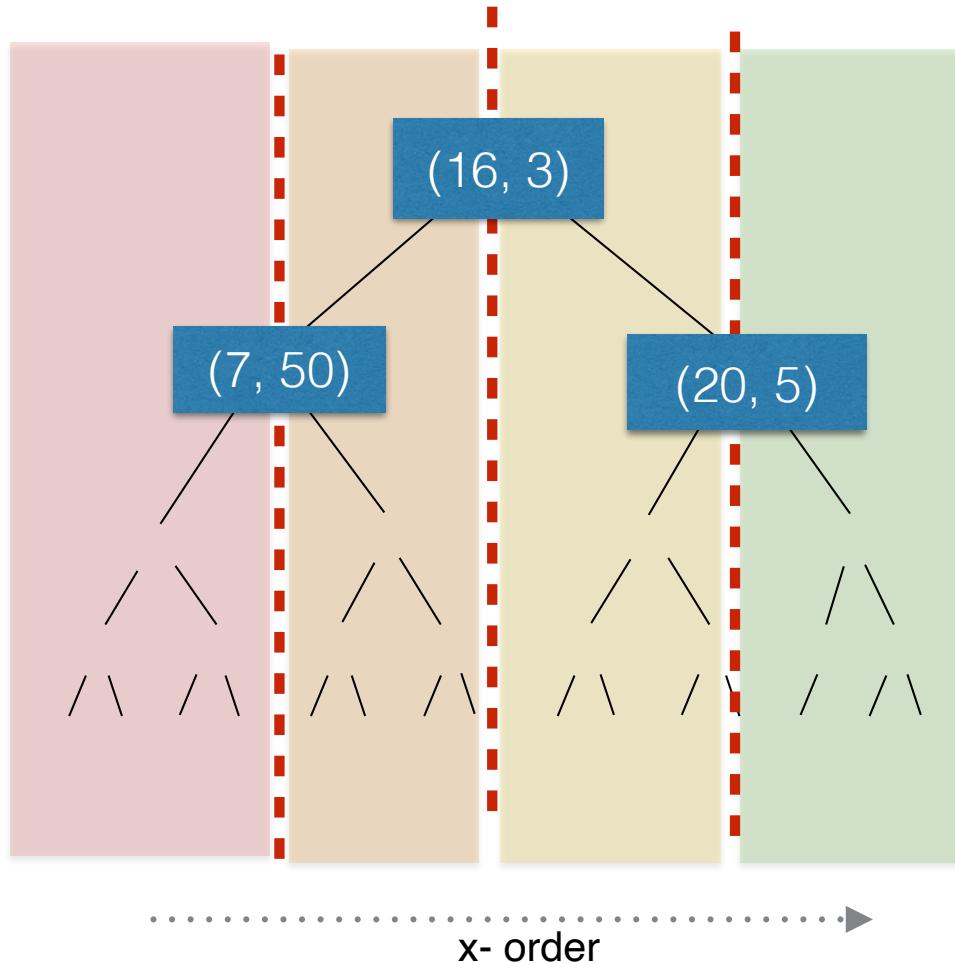


A BST creates an implicit space partition

- To search for a value we need to find the region of space that would contain this value

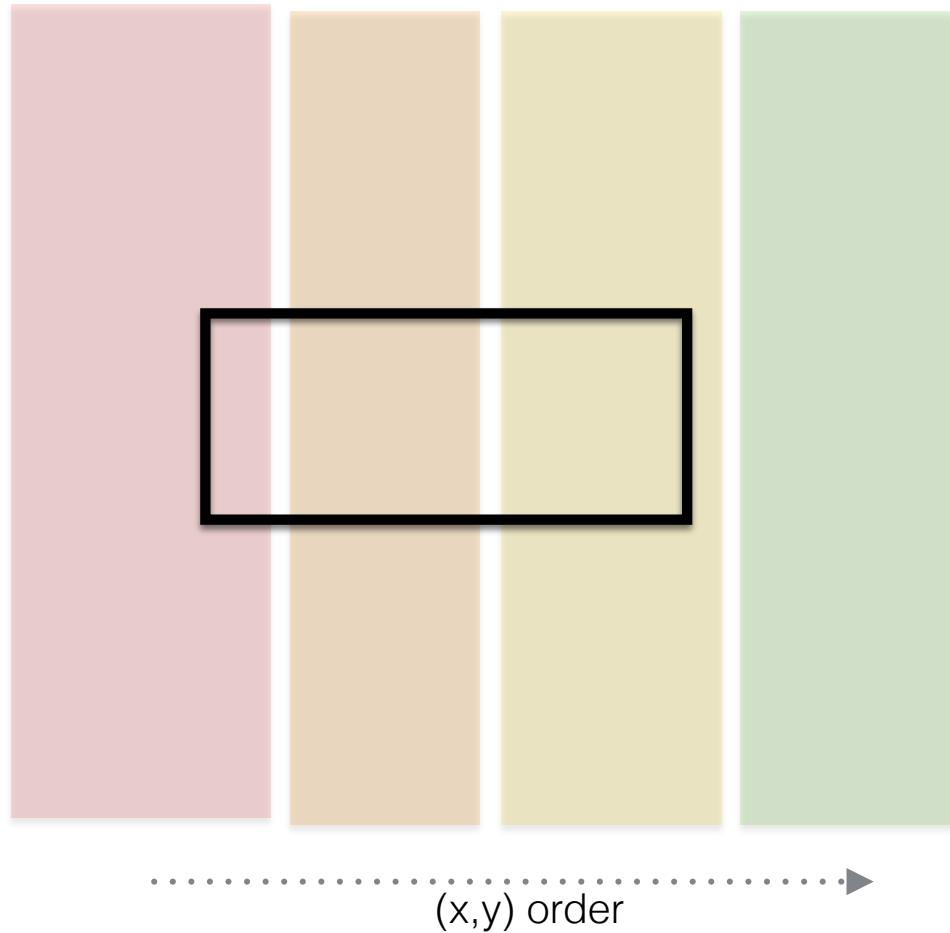


## Using a BST on 2d points



Partitions the space in vertical/horizontal stripes

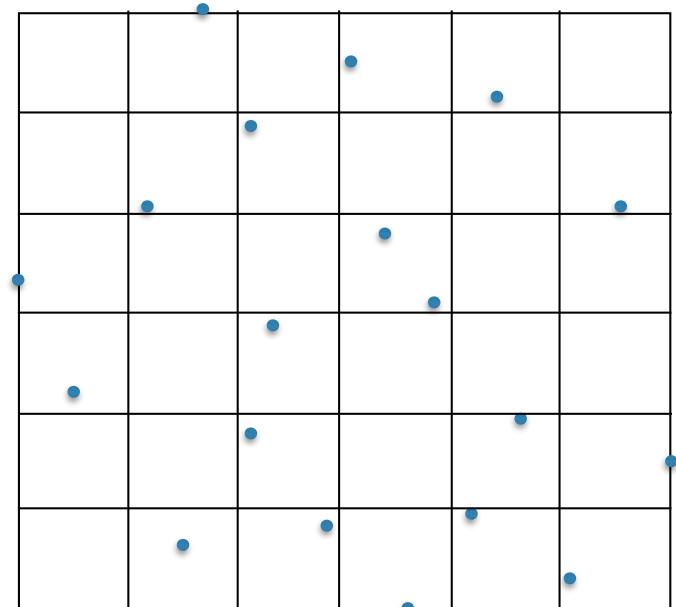
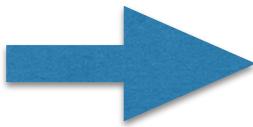
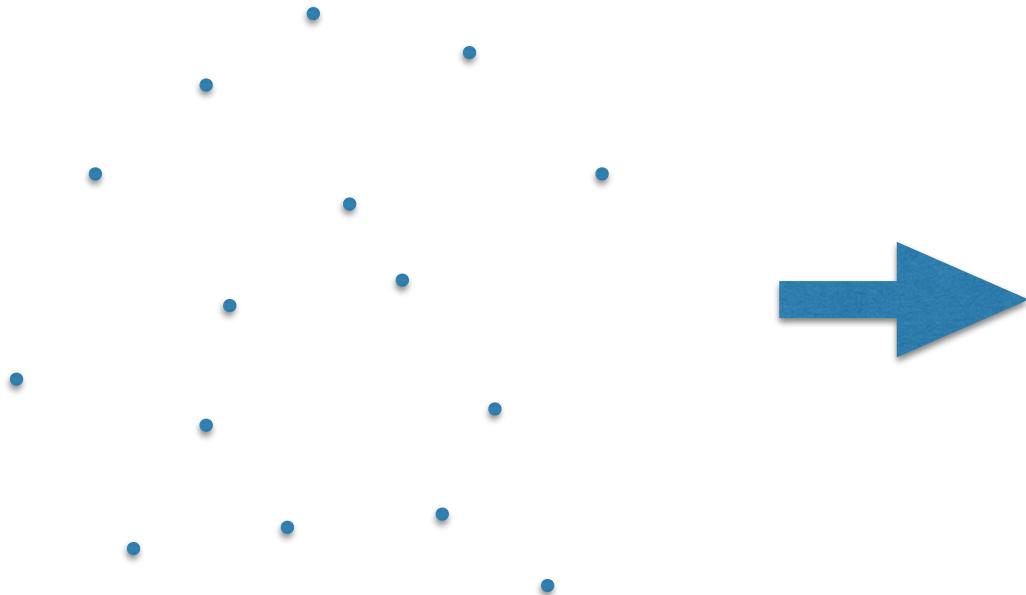
## Not a good partition for range-searching!



We have to search all vertical strips that intersect the range, which could have a lot of points outside the range.

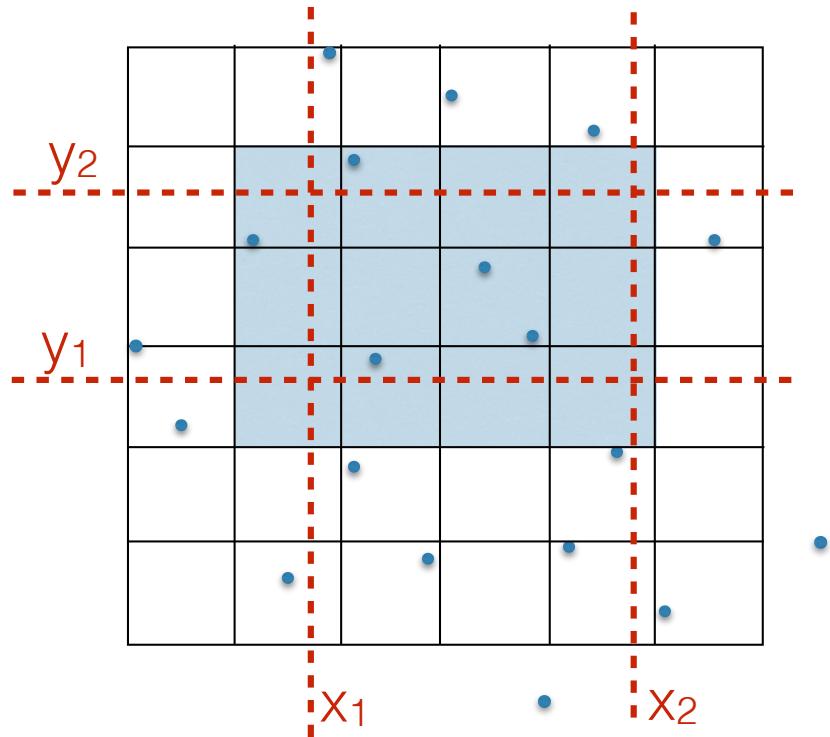
# The grid

The simplest space decomposition is a grid



- Build:  $O(n)$
- Space:  $O(n)$

## 2D range searches with a grid



- 2d range queries: traverse all cells that intersect the range
- Exact bound depends on how many points are in the cells
- Choose grid size  $m = O(\sqrt{n})$  and hope for  $O(1)$  points per cell. In this case, a range query takes  $O(k)$
- Worst case is bad: points are not uniformly distributed, a range query could take  $O(n)$  even if no points are reported

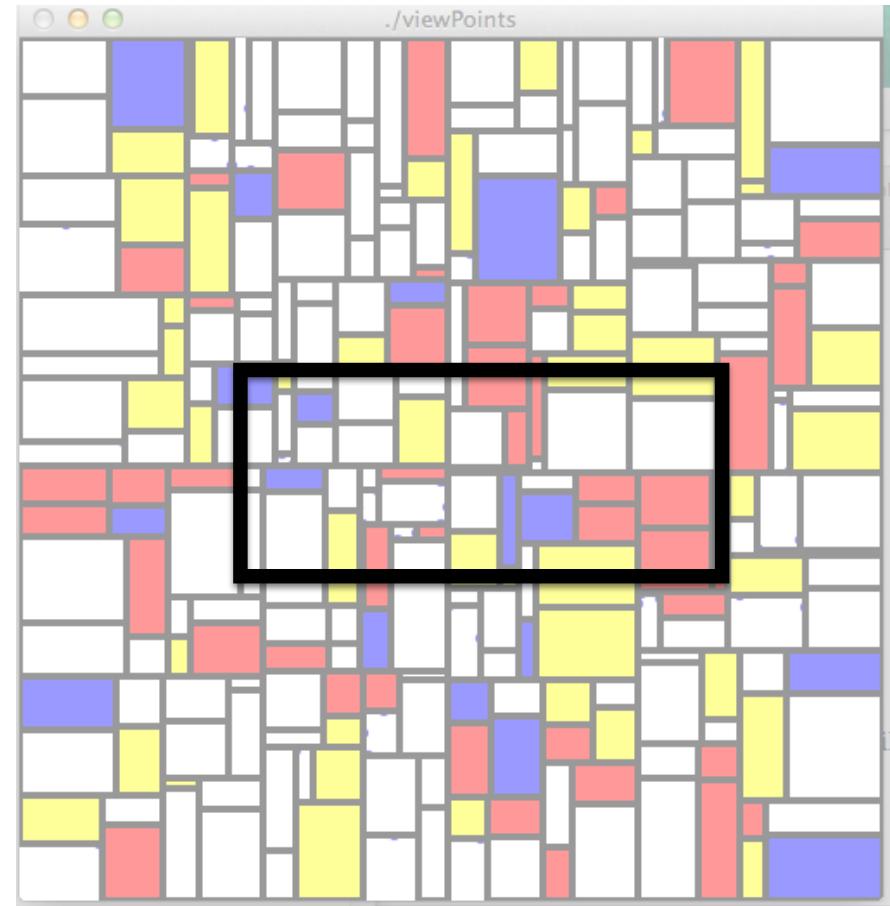
## The grid method

- + Simple to implement
- + Perform well if points are uniformly distributed
- + Can be used for many other problems besides range searching (e.g. [closest pair](#), neighbor queries)
- Gridding gives no guarantee on bounds. It's an heuristic.

Next we'll see two structures that extend the BST

# kd-trees

k-dimensional search trees



# range-trees

