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## GEOMETRIC APPLICATIONS OF BSTs

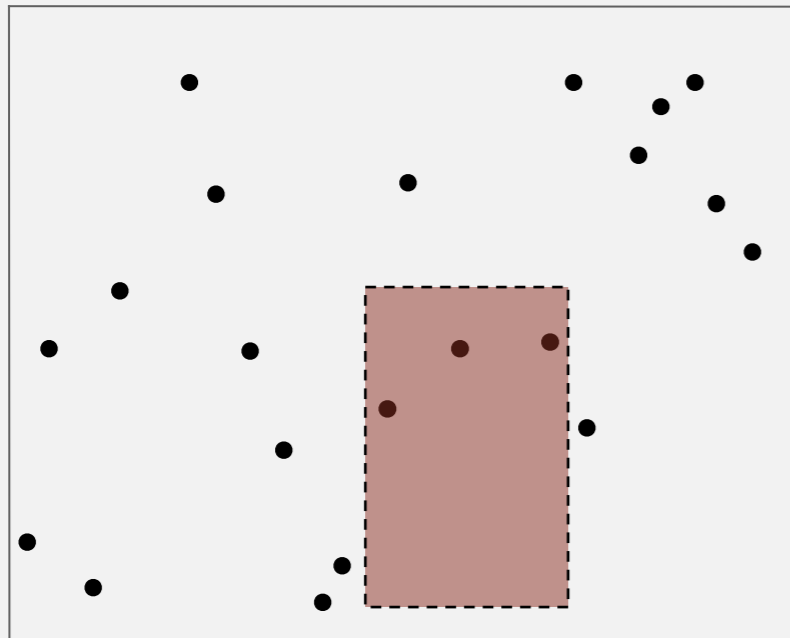
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- ▶ *1d range search*
- ▶ *line segment intersection*
- ▶ *kd trees*

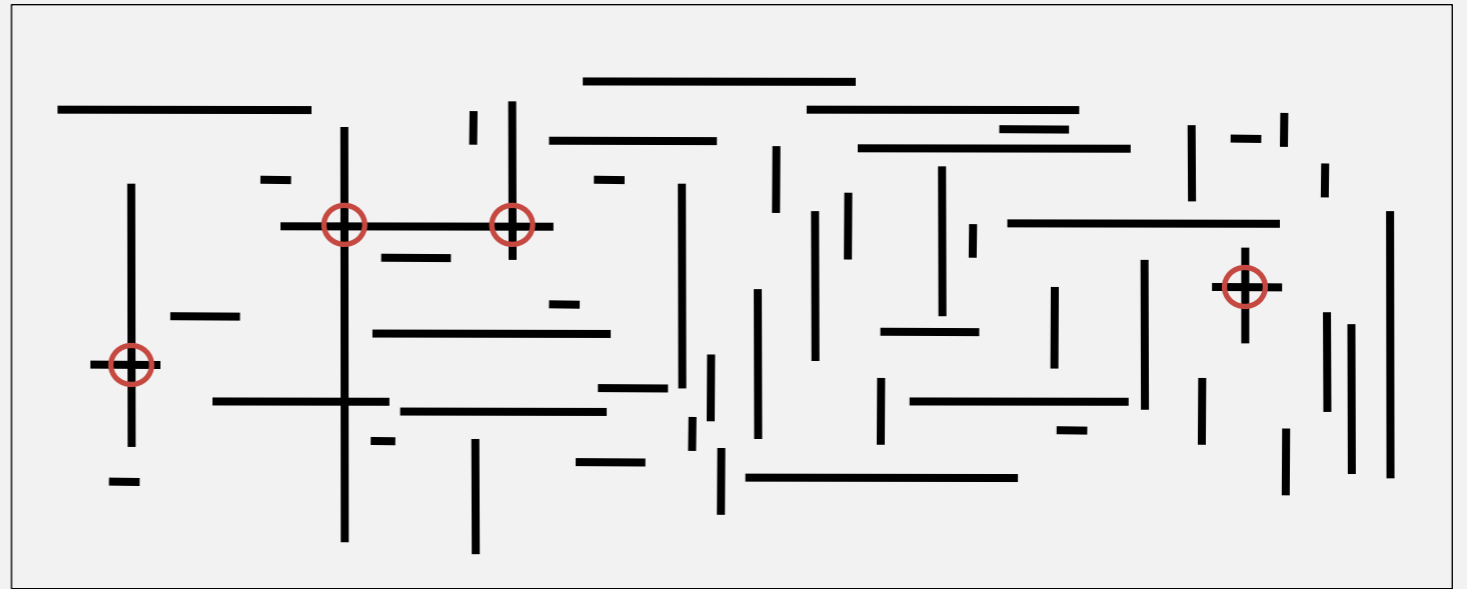
# Overview

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This lecture. Intersections among **geometric objects**.



2d orthogonal range search



line segment intersection


**Applications.** CAD, games, movies, virtual reality, databases, GIS, ....

**Efficient solutions.** **Binary search trees** (and extensions).

# Overview

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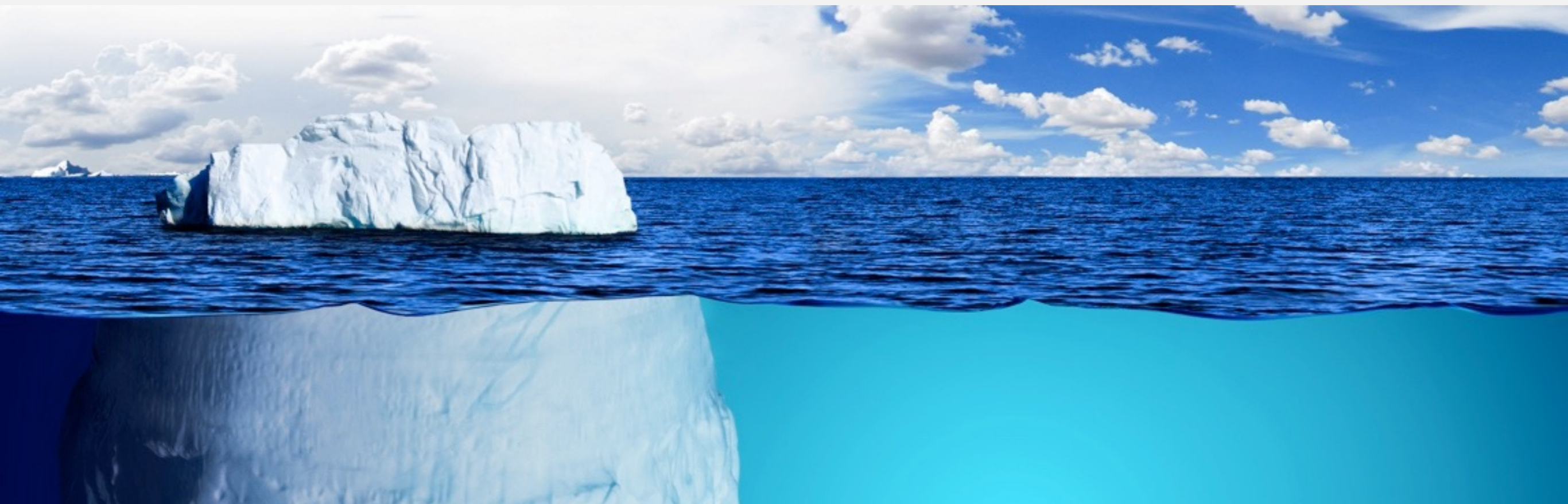
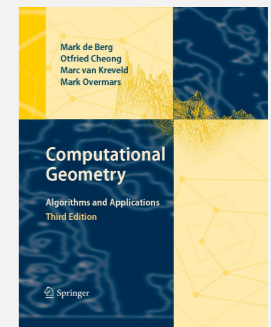
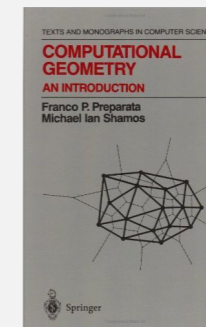
This lecture. Only the tip of the iceberg.



**Computer Science 451**  
**Computational Geometry**

[Bernard Chazelle](#)

[Princeton University](#)  
[Computer Science](#)  
[Department](#)



# GEOMETRIC APPLICATIONS OF BSTs

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- ▶ *1d range search*
- ▶ *line segment intersection*
- ▶ *kd trees*



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# 1d range search

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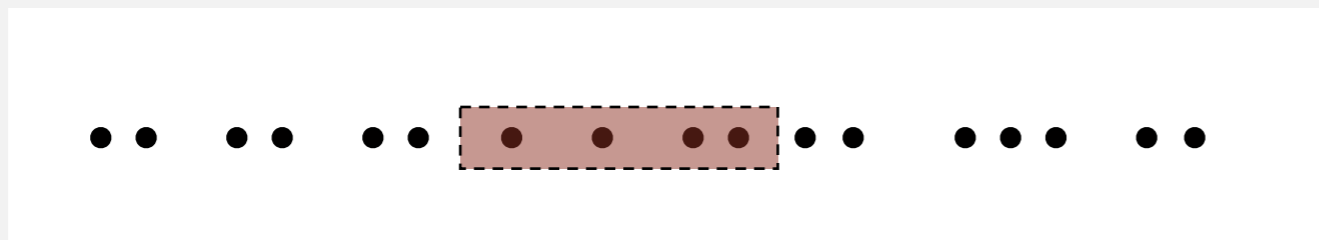
## Extension of ordered symbol table.

- Insert key-value pair.
- Search for key  $k$ .
- Delete key  $k$ .
- **Range search:** find all keys between  $k_1$  and  $k_2$ .
- **Range count:** number of keys between  $k_1$  and  $k_2$ .

**Application.** Database queries.

## Geometric interpretation.

- Keys are point on a **line**.
- Find/count points in a given **1d interval**.



```
insert B      B
insert D      B D
insert A      A B D
insert I      A B D I
insert H      A B D H I
insert F      A B D F H I
insert P      A B D F H I P
search G to K H I
count G to K  2
```

# Geometric search: quiz 1

---

Suppose that the keys are stored in a sorted array. What is the order of growth of the running time to perform **range count** as a function of  $n$  and  $R$ ?

$n$  = number of keys  
 $R$  = number of matching keys

- A.  $\log R$
- B.  $\log n$
- C.  $\log n + R$
- D.  $n + R$

# 1d range search: elementary implementations

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Ordered array. Slow insert; fast range search.

Unordered list. Slow insert; slow range search.

order of growth of running time for 1d range search

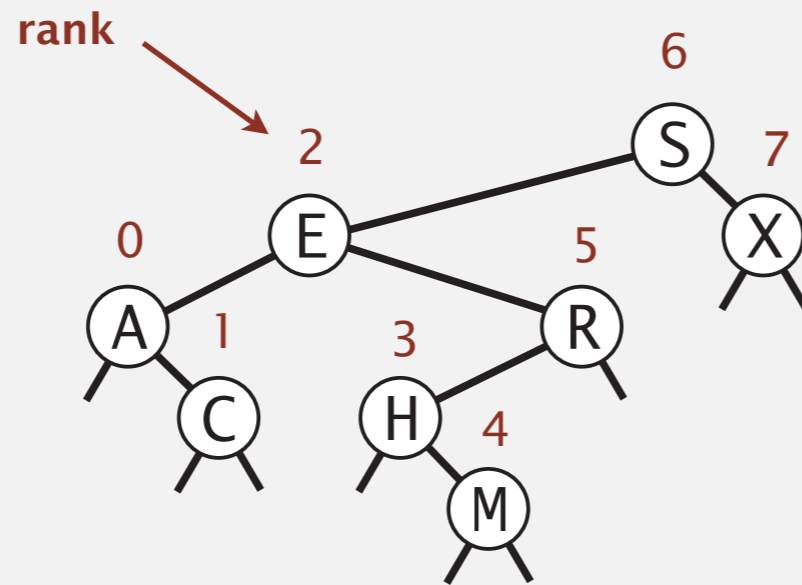
data structure	insert	range count	range search
ordered array	$n$	$\log n$	$R + \log n$
unordered list	$n$	$n$	$n$
goal	$\log n$	$\log n$	$R + \log n$

$n$  = number of keys

$R$  = number of keys that match

# 1d range count: BST implementation

1d range count. How many keys between  $l_o$  and  $h_i$  ?



**rangeCount(E, S)**

- rank(S) = 6
- rank(E) = 2
- 5 keys between E and S

```
public int size(Key lo, Key hi)
{
    if (contains(hi)) return rank(hi) - rank(lo) + 1;
    else               return rank(hi) - rank(lo);
}
```

← number of keys < hi

**Proposition.** Running time proportional to  $\log n$ . ← assuming BST is balanced

**Pf.** Nodes examined = search path to  $l_o$  + search path to  $h_i$ .

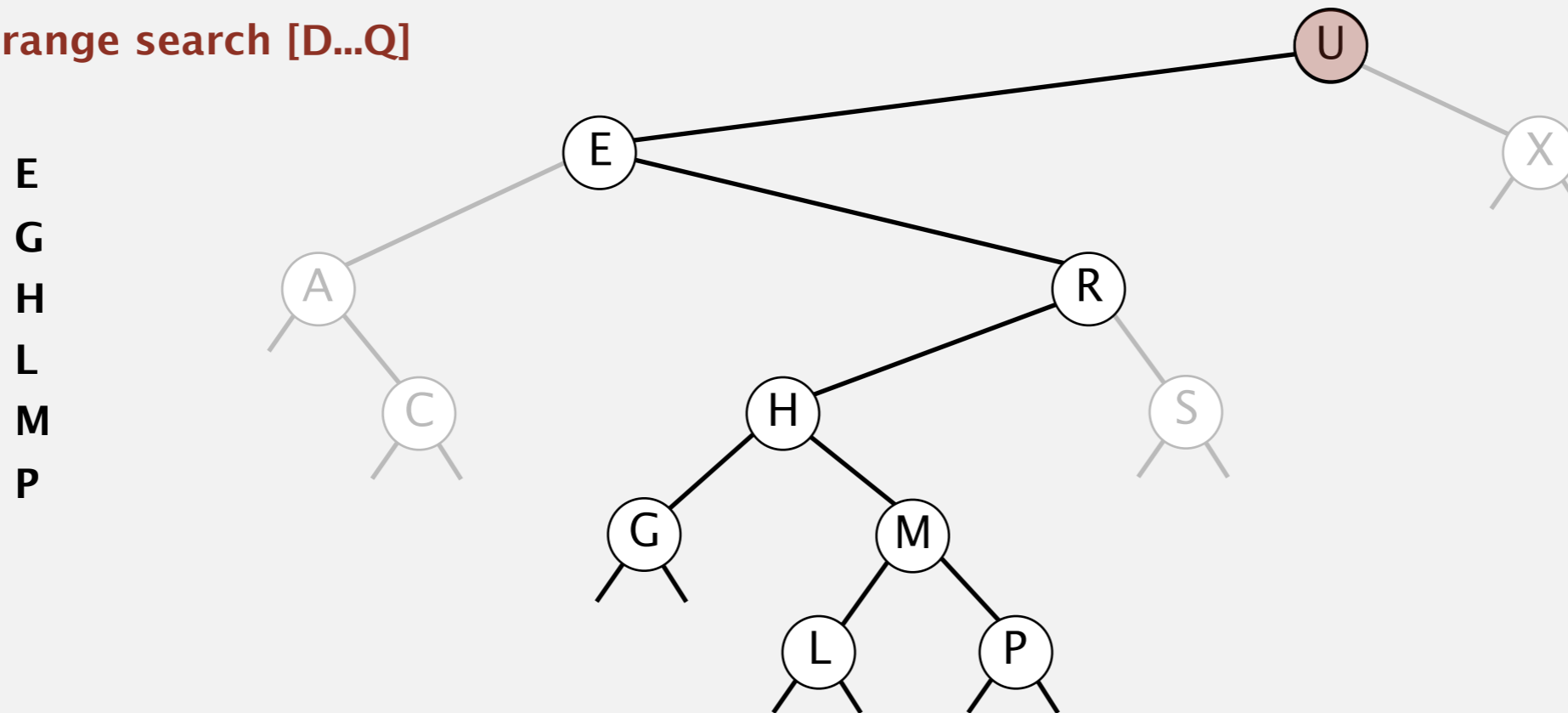


# 1d range search: BST implementation

1d range search. Find all keys between  $l_0$  and  $h_i$ .

- Recursively find all keys in left subtree (if any could fall in range).
- Check key in current node.
- Recursively find all keys in right subtree (if any could fall in range).

range search [D...Q]



**Proposition.** Running time proportional to  $R + \log n$ .

**Pf.** Nodes examined = search path to  $l_0$  + search path to  $h_i$  + matches.

# 1d range search: summary of performance

---

**Ordered array.** Slow insert; fast range search.

**Unordered list.** Slow insert; slow range search.

**BST.** Fast insert; fast range search.

order of growth of running time for 1d range search

data structure	insert	range count	range search
ordered array	$n$	$\log n$	$R + \log n$
unordered list	$n$	$n$	$n$
goal	$\log n$	$\log n$	$R + \log n$

$n$  = number of keys

$R$  = number of keys that match

# INTERVAL STABBING QUERY

**Goal.** Insert intervals (left, right) and support queries of the form “how many intervals contain  $x$  ?”

```
public class IntervalStab
```

```
IntervalStab()
```

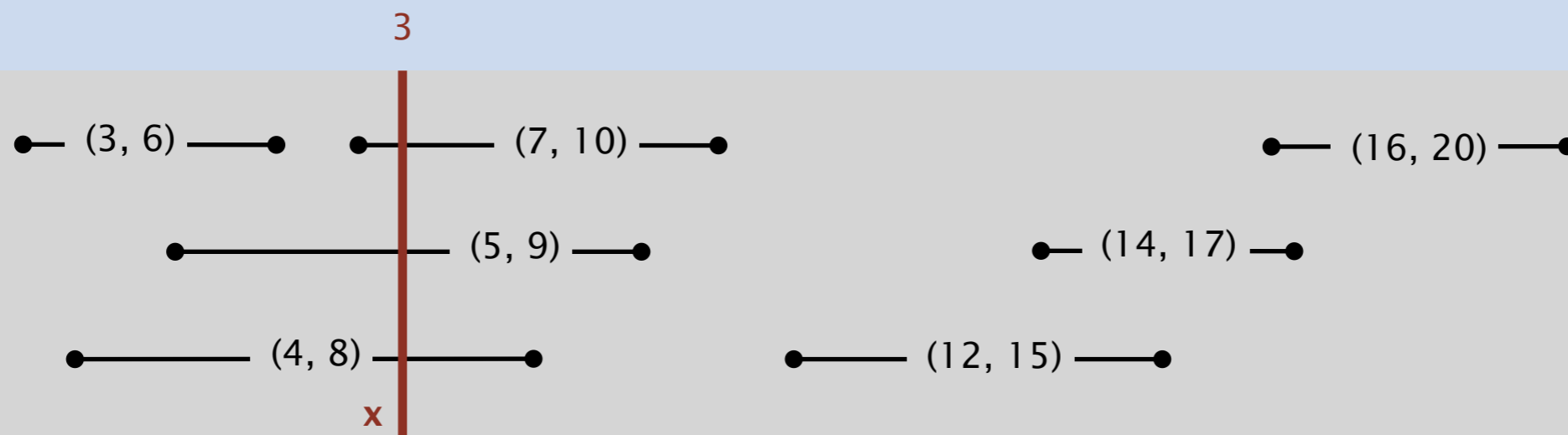
*create an empty data structure*

```
void insert(double left, double right)
```

*insert the interval (left, right)  
into the data structure*

```
int count(double x)
```

*number of intervals  
that contain  $x$*



# GEOMETRIC APPLICATIONS OF BSTs

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- ▶ *1d range search*
- ▶ *line segment intersection*
- ▶ *kd trees*

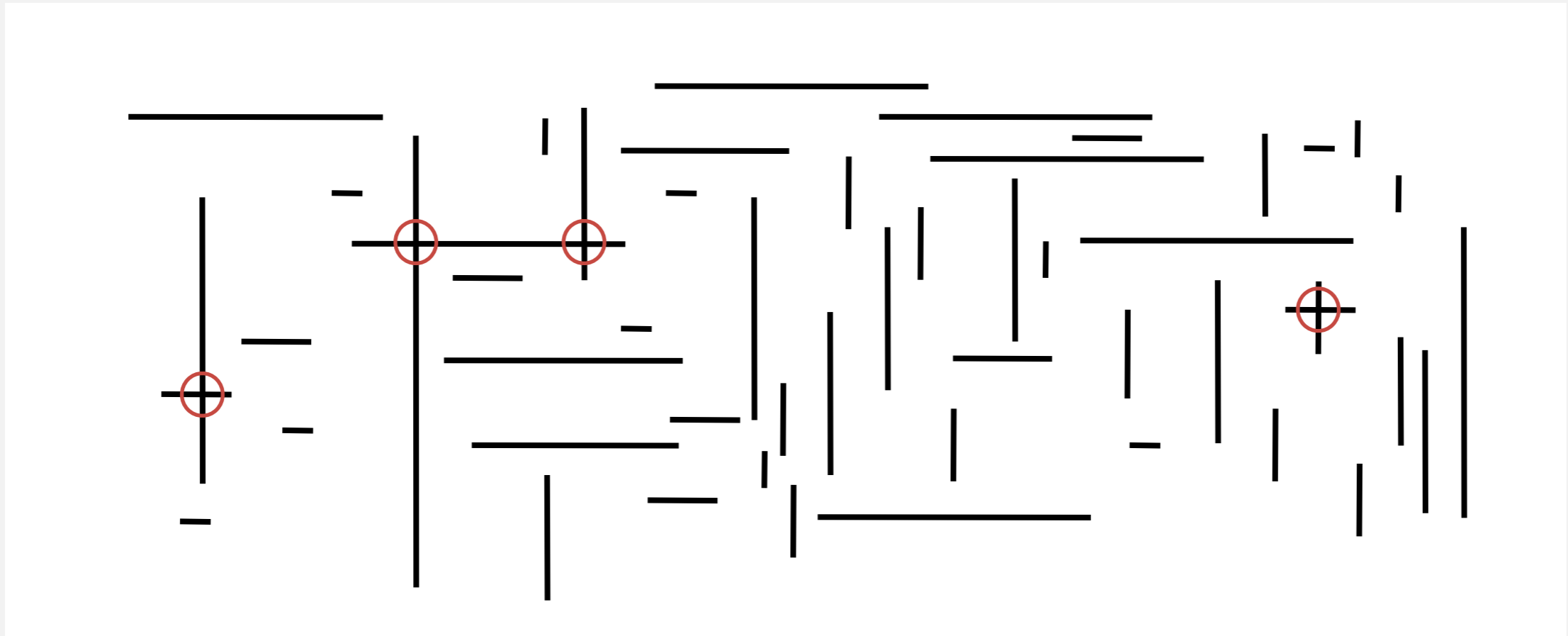


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# Orthogonal line segment intersection

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Given  $n$  horizontal and vertical line segments, find all intersections.



**Quadratic algorithm.** Check all pairs of line segments for intersection.

# Microprocessors and geometry

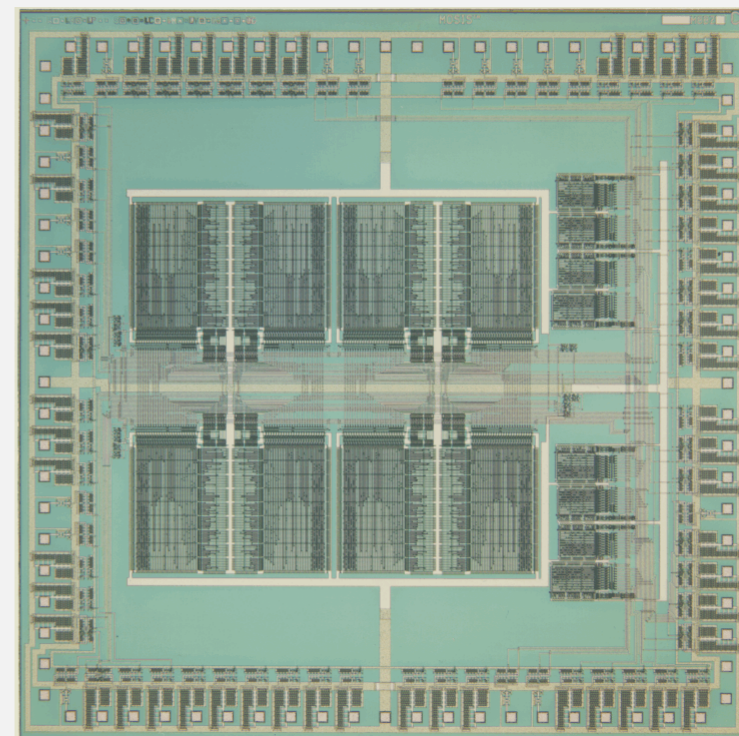
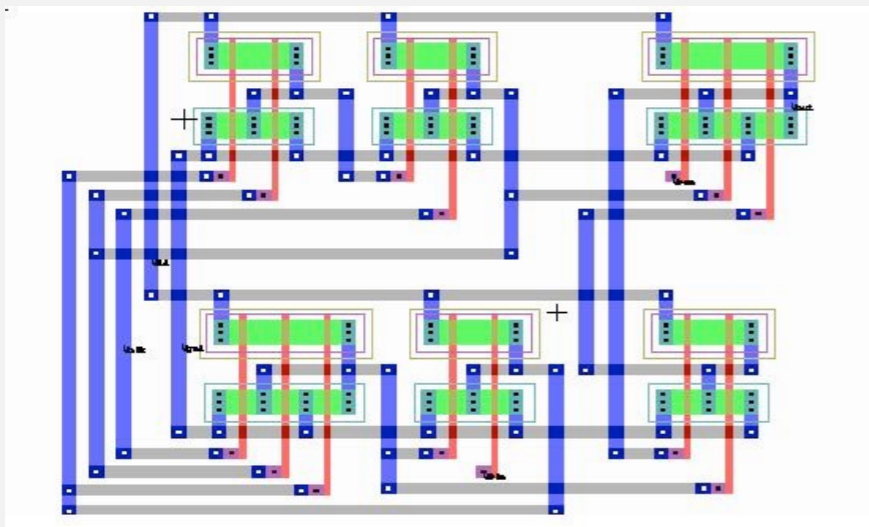
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Early 1970s. microprocessor design became a **geometric** problem.

- Very Large Scale Integration (VLSI).
- Computer-Aided Design (CAD).

Design-rule checking.

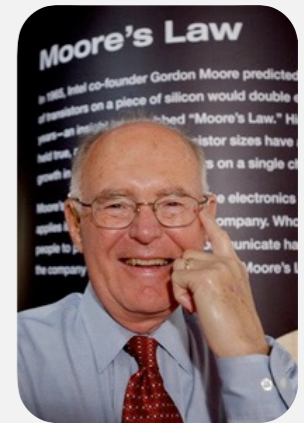
- Certain wires cannot intersect.
- Certain spacing needed between different types of wires.
- Debugging = line segment (or rectangle) intersection.



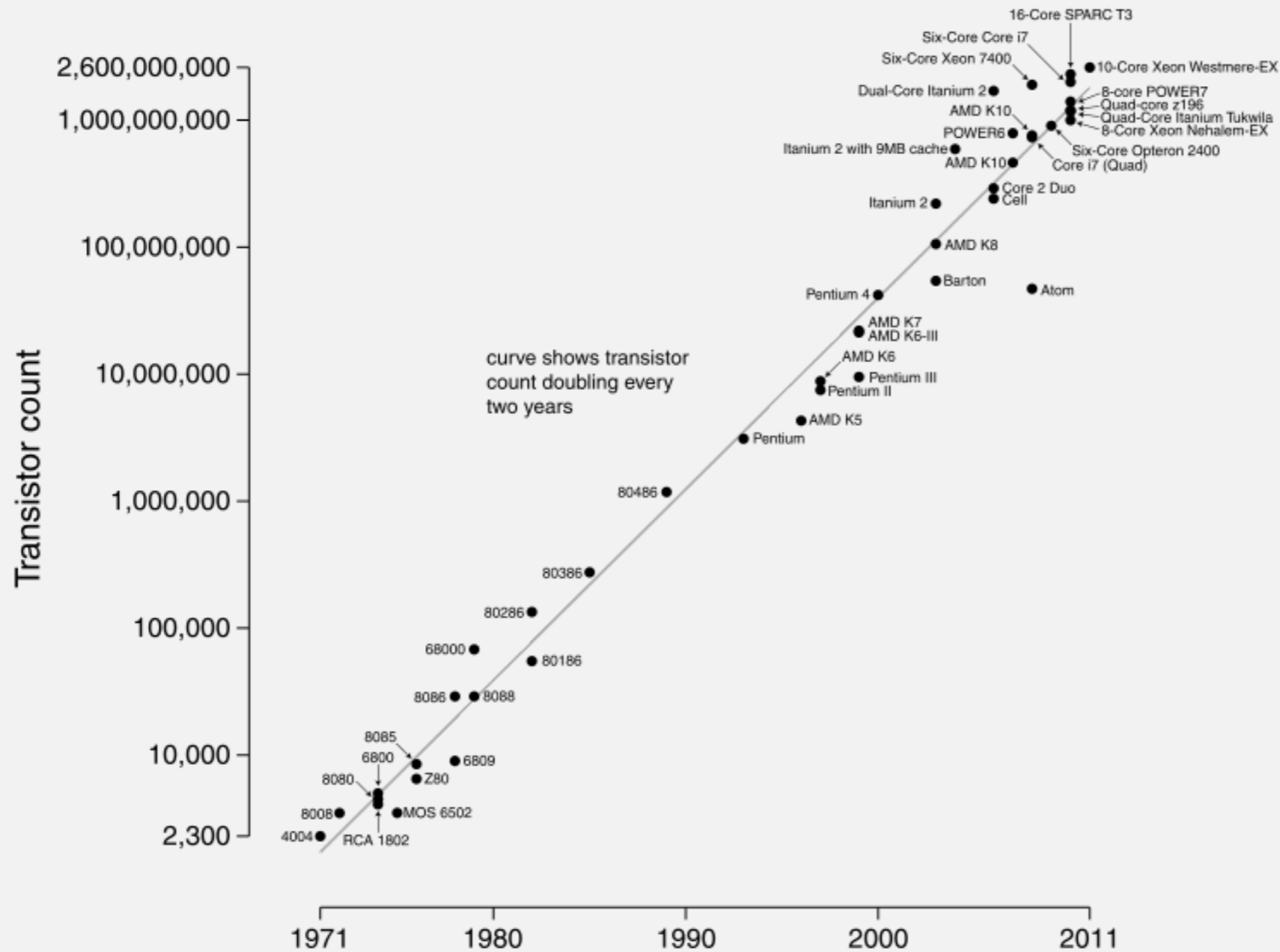


# Algorithms and Moore's law

Moore's law (1965). Transistor count doubles every 2 years.



Gordon Moore



[http://commons.wikimedia.org/wiki/File%3ATransistor\\_Count\\_and\\_Moore's\\_Law\\_-\\_2011.svg](http://commons.wikimedia.org/wiki/File%3ATransistor_Count_and_Moore's_Law_-_2011.svg)

# Algorithms and Moore's law

---

## Sustaining Moore's law.

- Problem size doubles every 2 years. ← problem size = transistor count
- Processing power doubles every 2 years. ← get to use faster computer
- How much \$ do I need to get the job done with a quadratic algorithm?

$$T_n = a n^2 \quad \text{running time today}$$

$$\begin{aligned} T_{2n} &= (a/2)(2n)^2 && \text{running time in 2 years} \\ &= 2T_n \end{aligned}$$

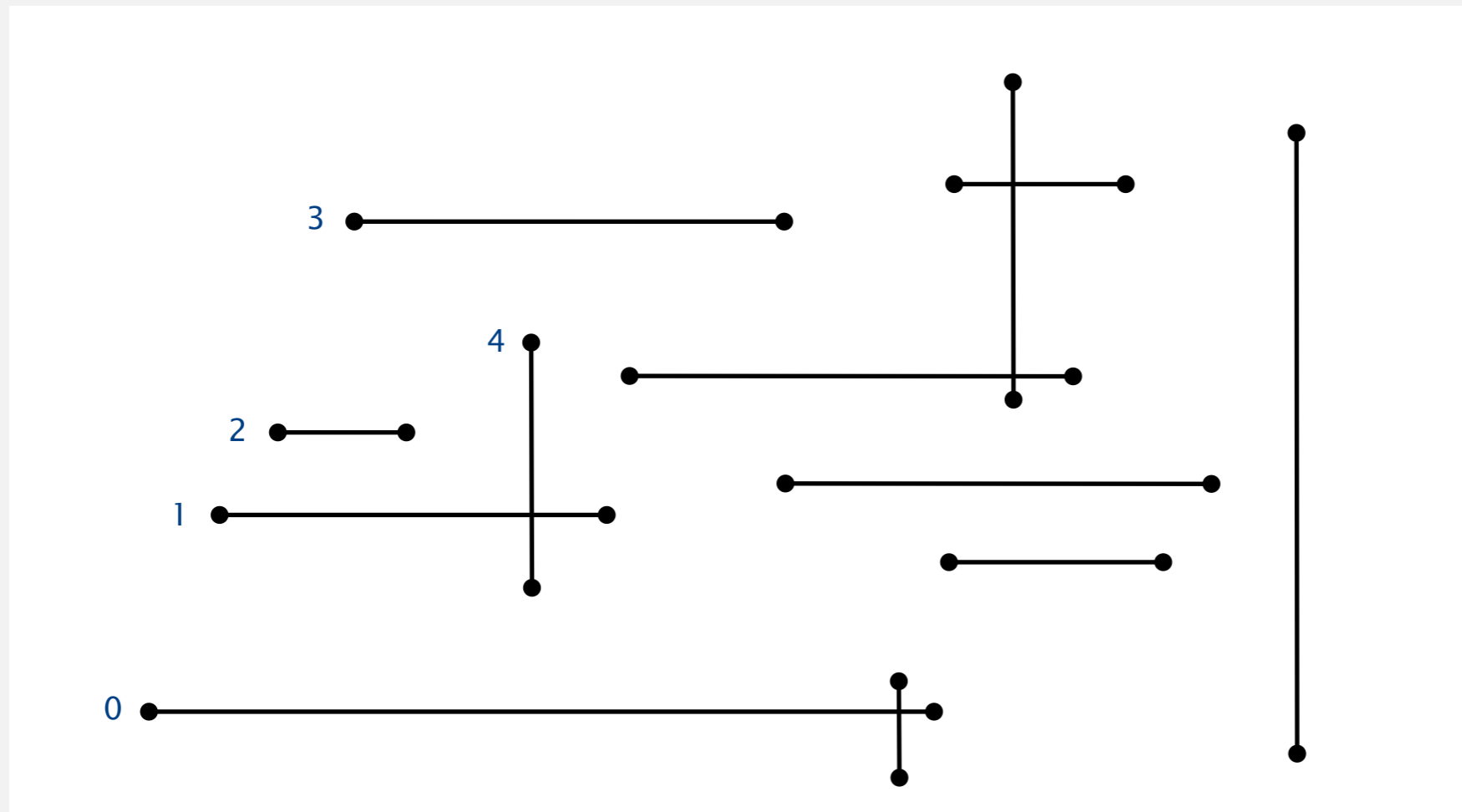
running time	1970	1972	1974	2000
$n$	\$ $x$	\$ $x$	\$ $x$	\$ $x$
$n \log n$	\$ $x$	\$ $x$	\$ $x$	\$ $x$
$n^2$	\$ $x$	\$ $2x$	\$ $4x$	\$ $2^{15}x$

**Bottom line.** Linearithmic algorithm is **necessary** to sustain Moore's Law.

# Orthogonal line segment intersection: sweep-line algorithm

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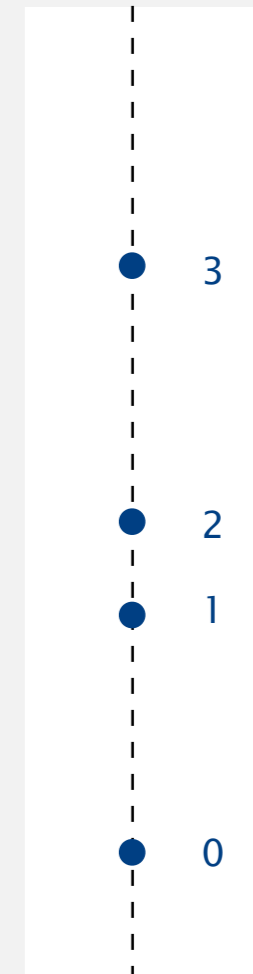
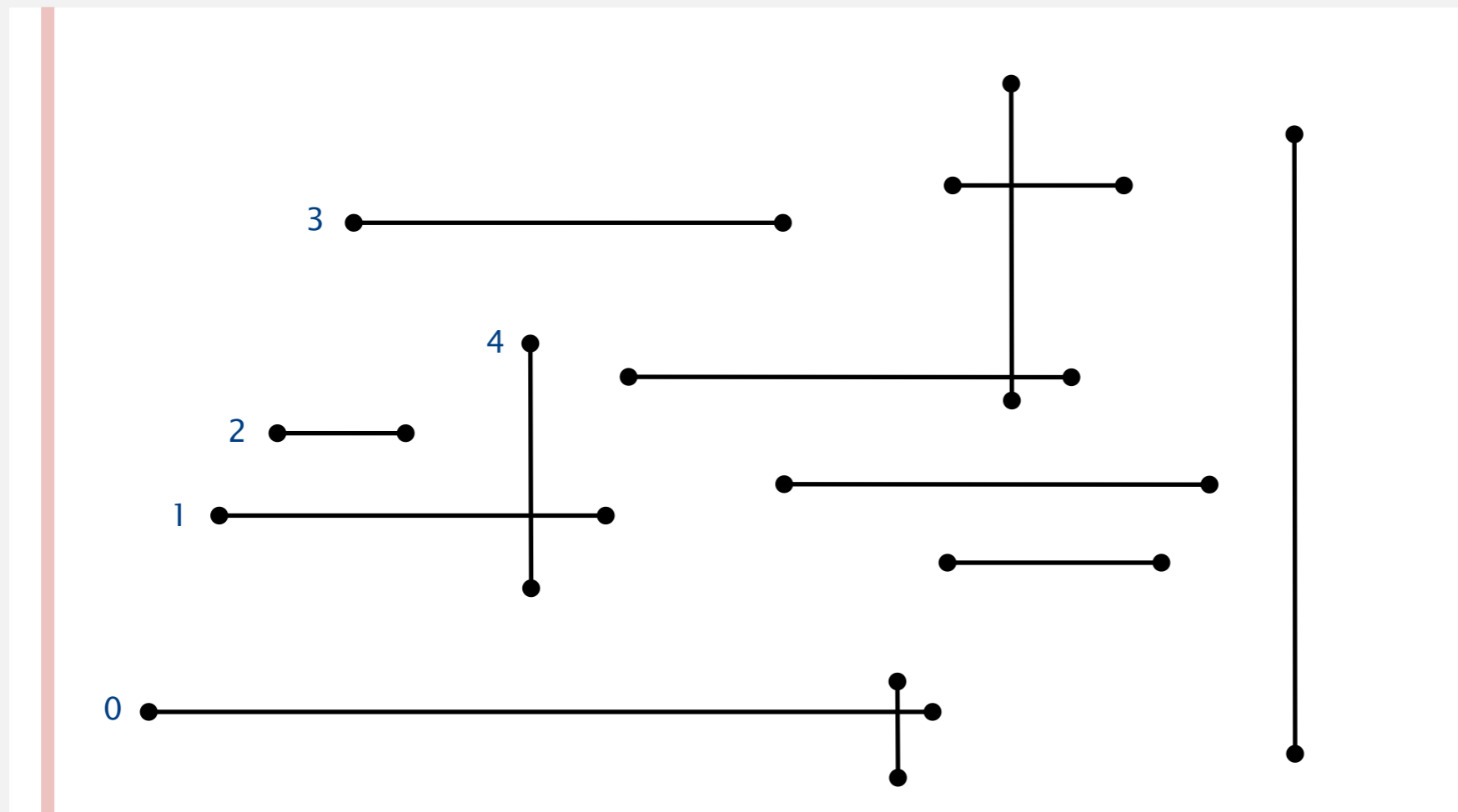
Non-degeneracy assumption. All  $x$ - and  $y$ -coordinates are distinct.



# Orthogonal line segment intersection: sweep-line algorithm

Sweep vertical line from left to right.

- $x$ -coordinates define events.
- $h$ -segment (left endpoint): insert  $y$ -coordinate into BST.



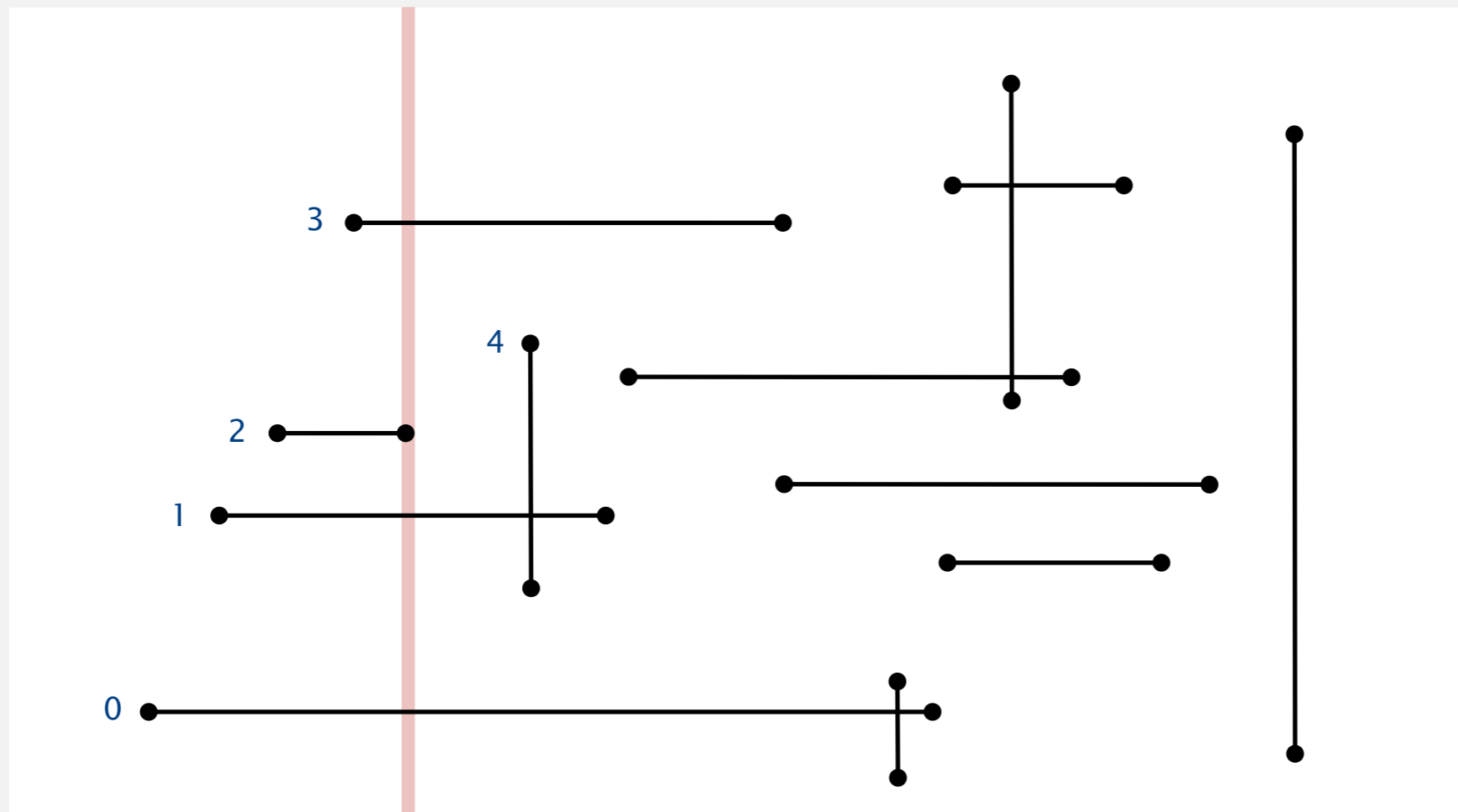
nondegeneracy assumption: all  $x$ - and  $y$ -coordinates are distinct

$y$ -coordinates

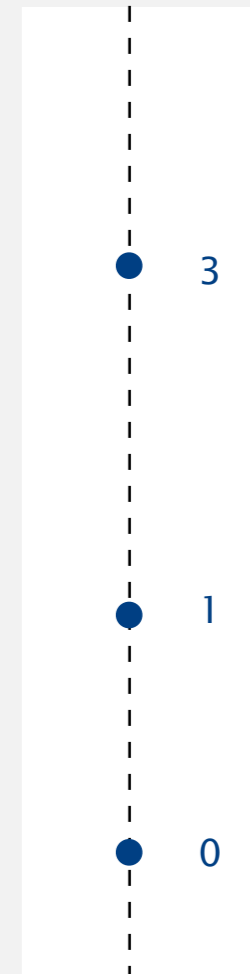
# Orthogonal line segment intersection: sweep-line algorithm

Sweep vertical line from left to right.

- $x$ -coordinates define events.
- $h$ -segment (left endpoint): insert  $y$ -coordinate into BST.
- $h$ -segment (right endpoint): remove  $y$ -coordinate from BST.



nondegeneracy assumption: all  $x$ - and  $y$ -coordinates are distinct

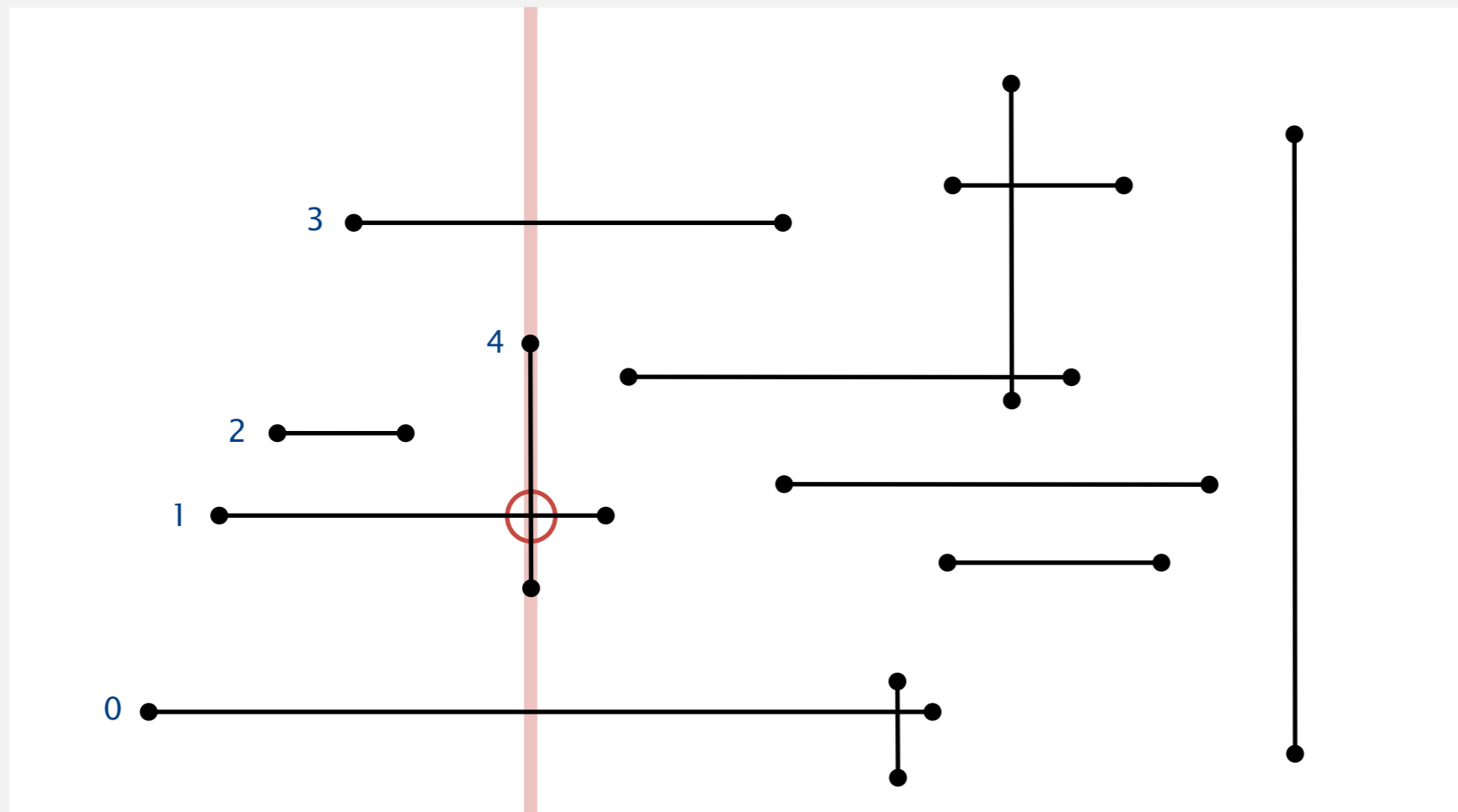


$y$ -coordinates

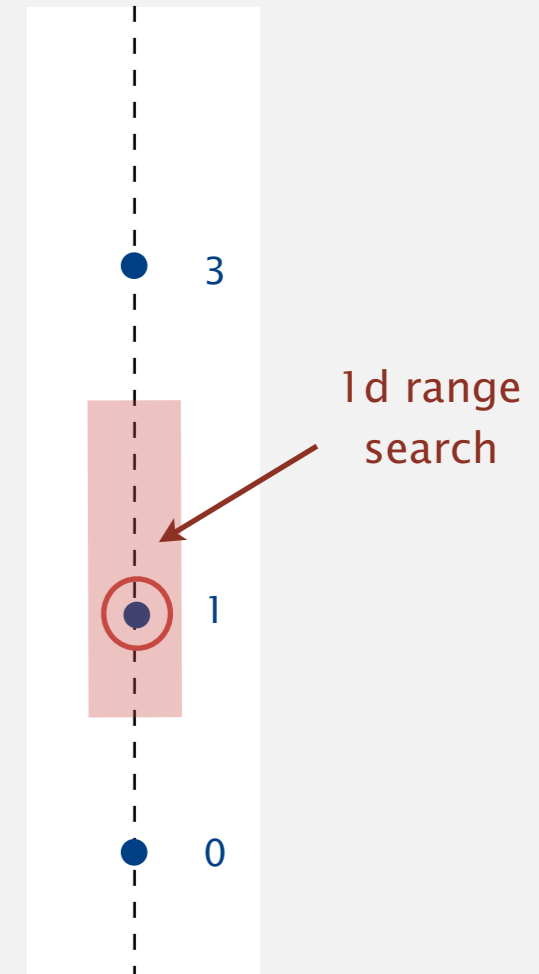
# Orthogonal line segment intersection: sweep-line algorithm

Sweep vertical line from left to right.

- $x$ -coordinates define events.
- $h$ -segment (left endpoint): insert  $y$ -coordinate into BST.
- $h$ -segment (right endpoint): remove  $y$ -coordinate from BST.
- $v$ -segment: range search for interval of  $y$ -endpoints.



nondegeneracy assumption: all  $x$ - and  $y$ -coordinates are distinct



$y$ -coordinates



# Orthogonal line segment intersection: sweep-line analysis

---

**Proposition.** The sweep-line algorithm takes time proportional to  $n \log n + R$  to find all  $R$  intersections among  $n$  orthogonal line segments.

**Pf.**

- Put  $x$ -coordinates on a PQ (or sort). ←  $n \log n$
- Insert  $y$ -coordinates into BST. ←  $n \log n$
- Delete  $y$ -coordinates from BST. ←  $n \log n$
- Range searches in BST. ←  $n \log n + R$

**Bottom line.** Sweep line reduces 2d orthogonal line segment intersection search to 1d range search.

# Sweep-line algorithm: context

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The **sweep-line algorithm** is a key technique in computational geometry.

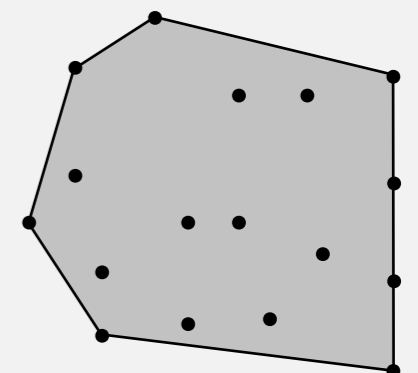
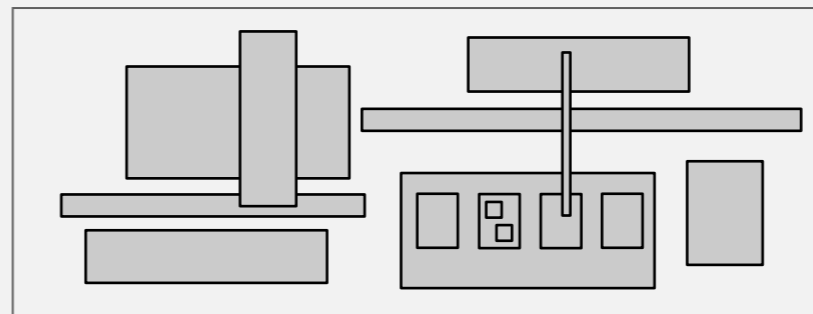
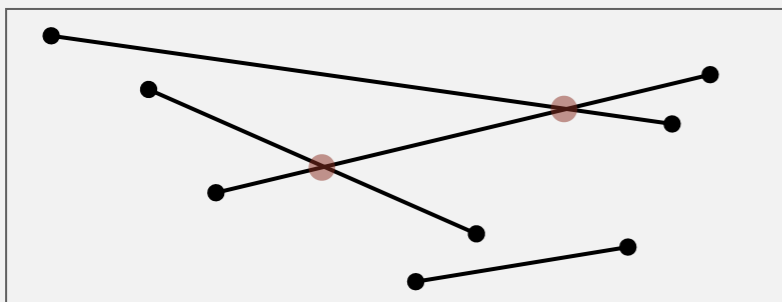
## Geometric intersection.

- General line segment intersection.
- Axis-aligned rectangle intersection.
- ...



## More problems.

- Andrew's algorithm for convex hull.
- Fortune's algorithm Voronoi diagram.
- Scanline algorithm for rendering computer graphics.
- ...





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## GEOMETRIC APPLICATIONS OF BSTs

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- ▶ *1d range search*
- ▶ *line segment intersection*
- ▶ *kd trees*

# 2-d orthogonal range search

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## Extension of ordered symbol-table to 2d keys.

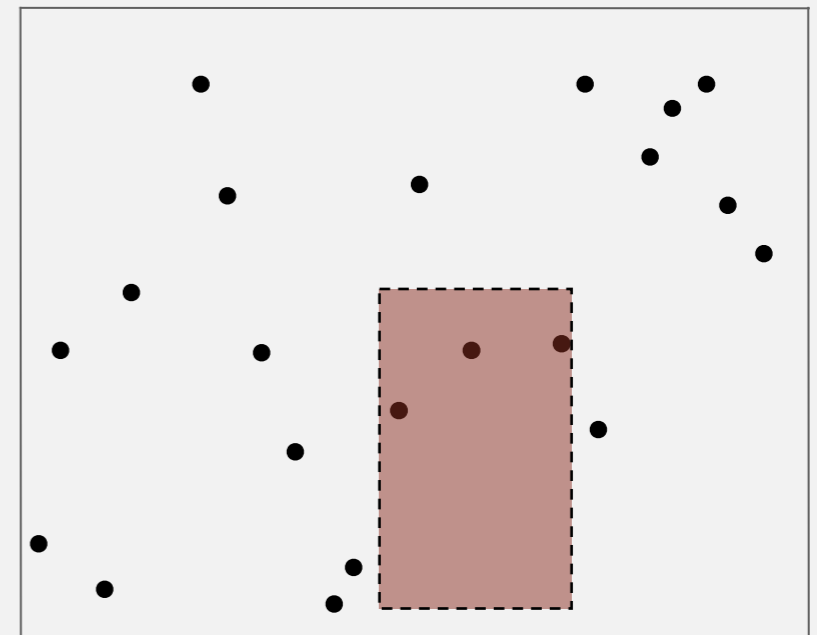
- Insert a 2d key.
- Search for a 2d key.
- **Range search:** find all keys that lie in a 2d range.
- **Range count:** number of keys that lie in a 2d range.

**Applications.** Networking, circuit design, databases, ...

## Geometric interpretation.

- Keys are point in the **plane**.
- Find/count points in a given  **$h-v$  rectangle**

↑  
rectangle is axis-aligned



# Space-partitioning trees

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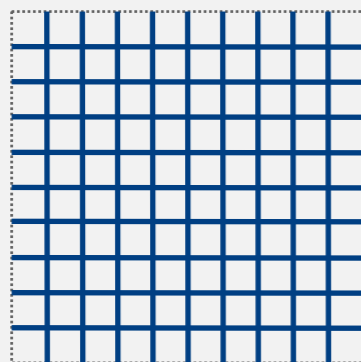
Use a **tree** to represent a recursive subdivision of 2d space.

**Grid.** Divide space uniformly into squares.

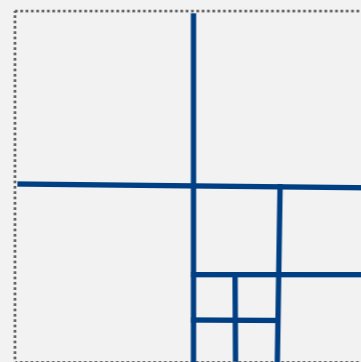
**Quadtree.** Recursively divide space into four quadrants.

**2d tree.** Recursively divide space into two halfplanes.

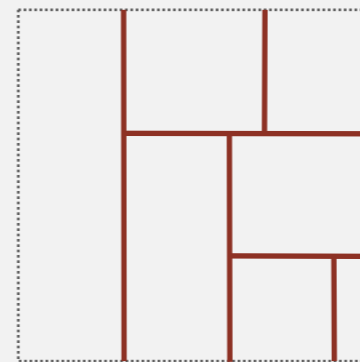
**BSP tree.** Recursively divide space into two regions.



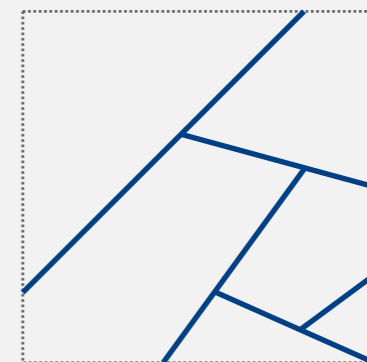
Grid



Quadtree



2d tree

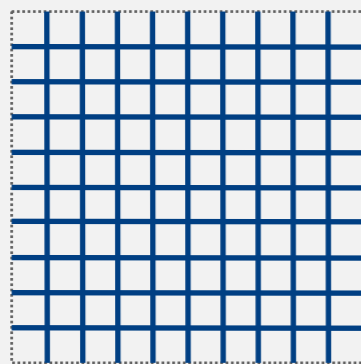


BSP tree

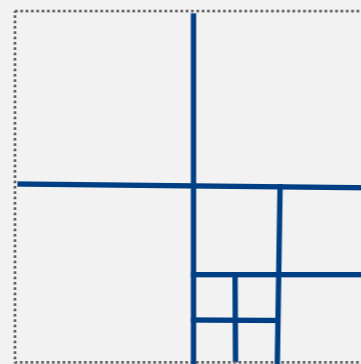
# Space-partitioning trees: applications

## Applications.

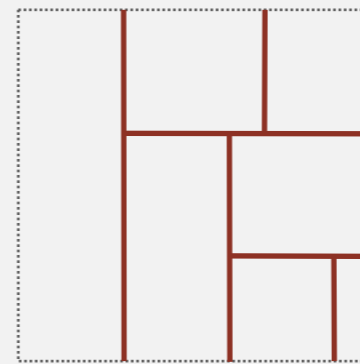
- Ray tracing.
- **2d range search.**
- Flight simulators.
- N-body simulation.
- Collision detection.
- Astronomical databases.
- **Nearest neighbor search.**
- Adaptive mesh generation.
- Accelerate rendering in Doom.
- Hidden surface removal and shadow casting.



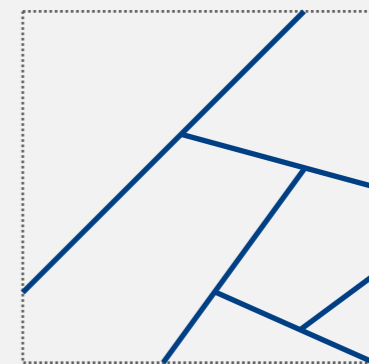
Grid



Quadtree



**2d tree**

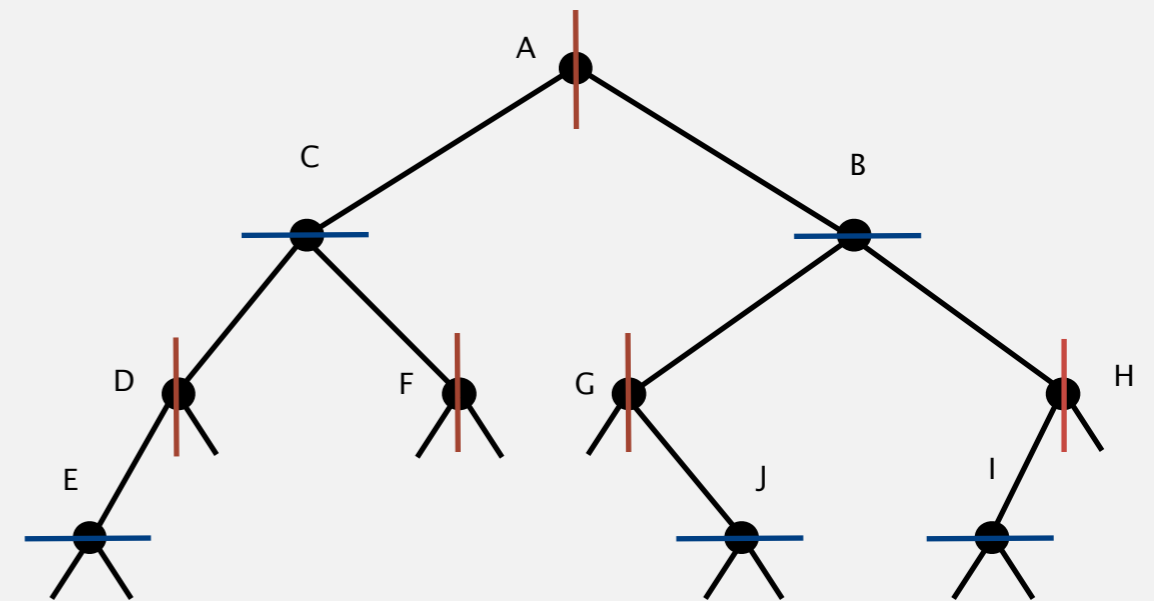
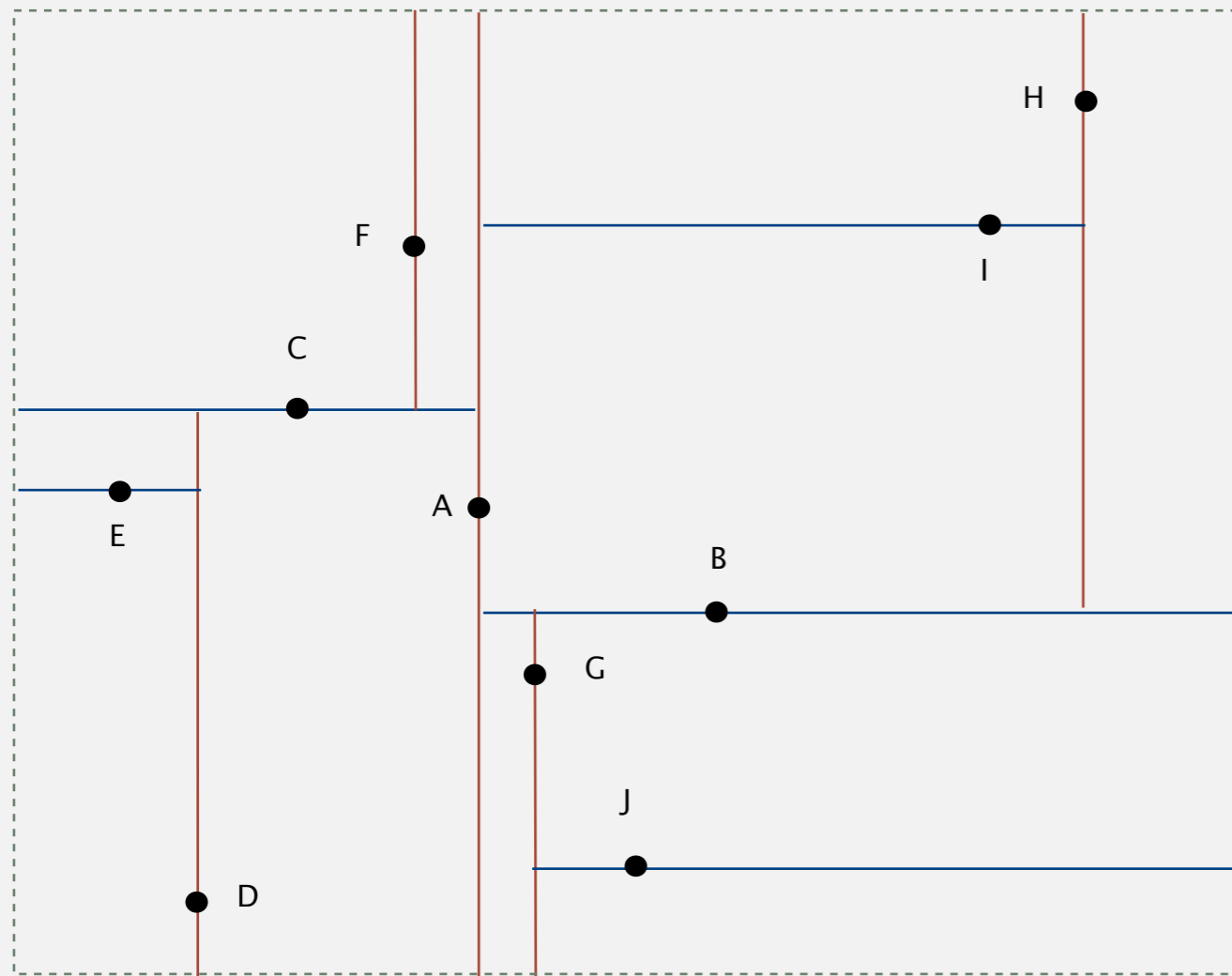


BSP tree



# 2d tree construction

Recursively partition plane into two halfplanes.

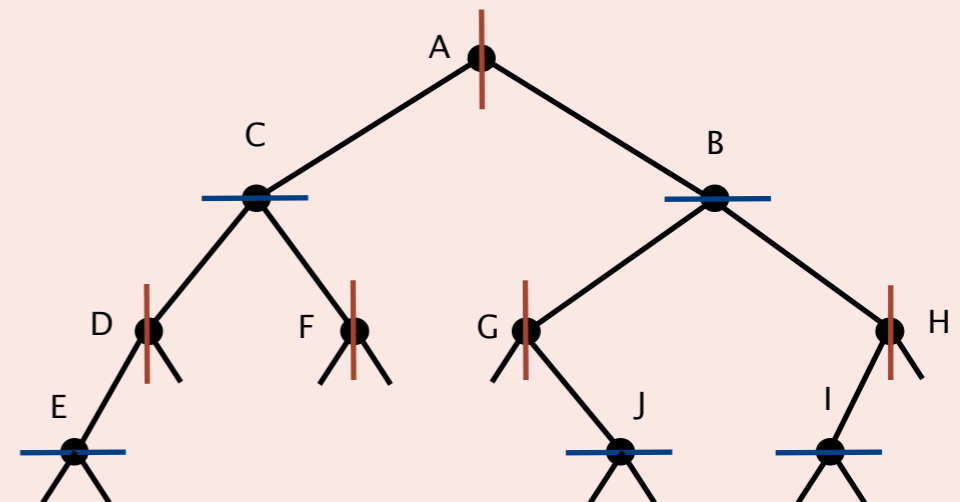
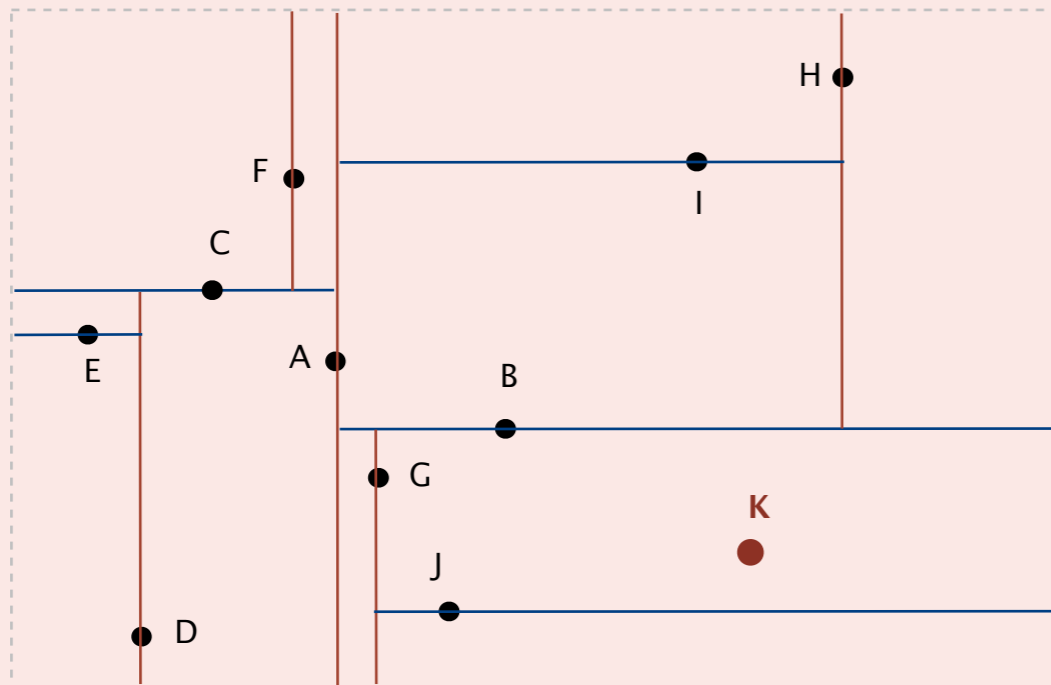


# Geometric search: quiz 2

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Where would point K be inserted in the 2d tree below?

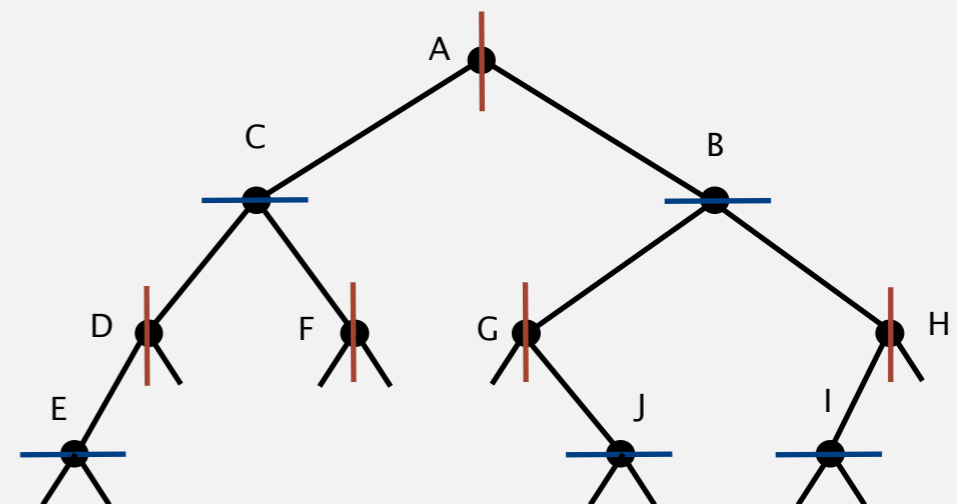
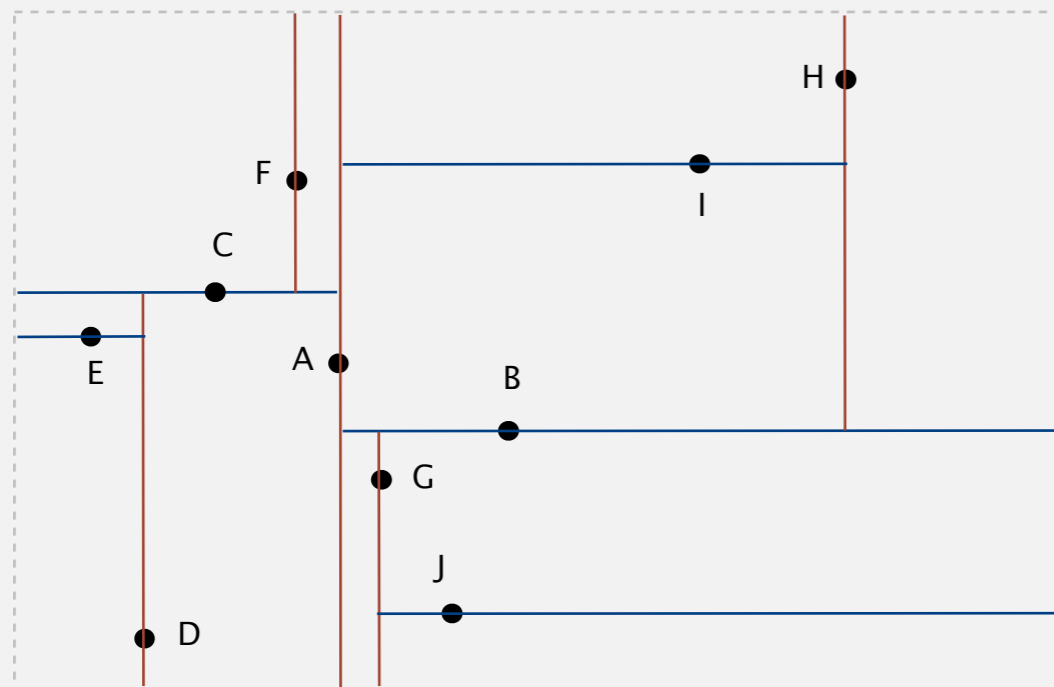
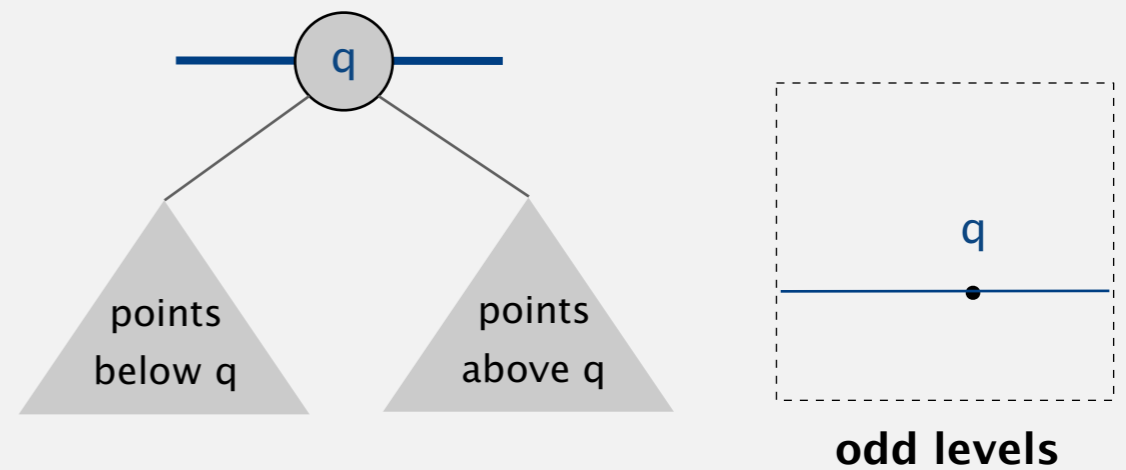
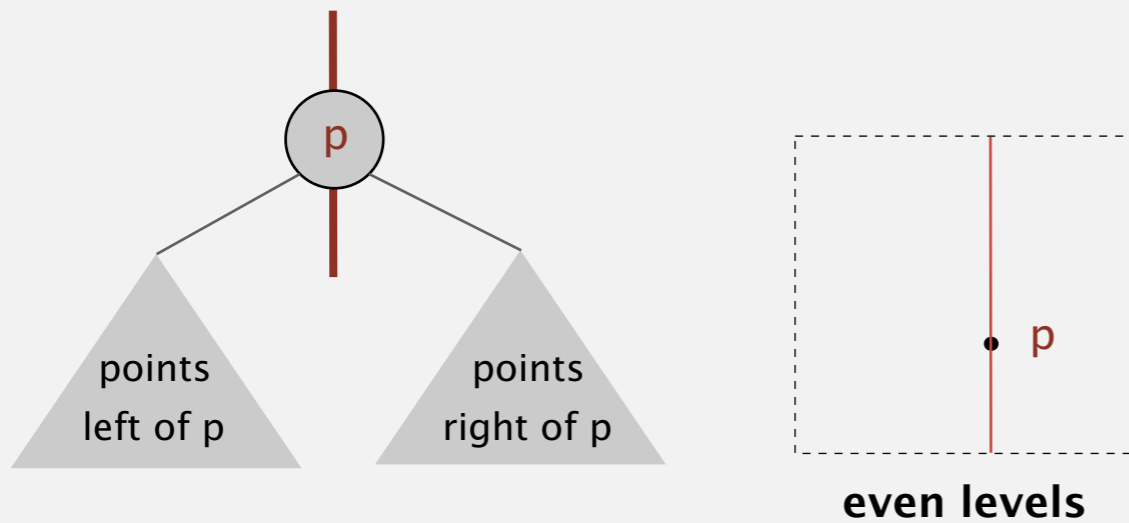
- A. Left child of G.
- B. Left child of J.
- C. Right child of J.
- D. Right child of I.



# 2d tree implementation

**Data structure.** BST, but alternate using  $x$ - and  $y$ -coordinates as key.

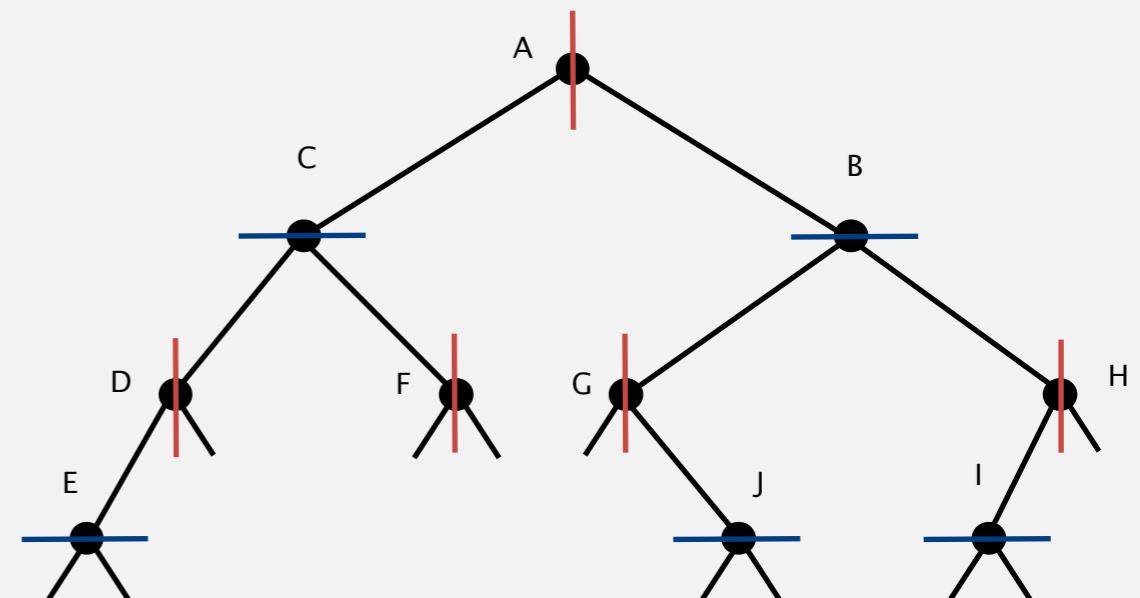
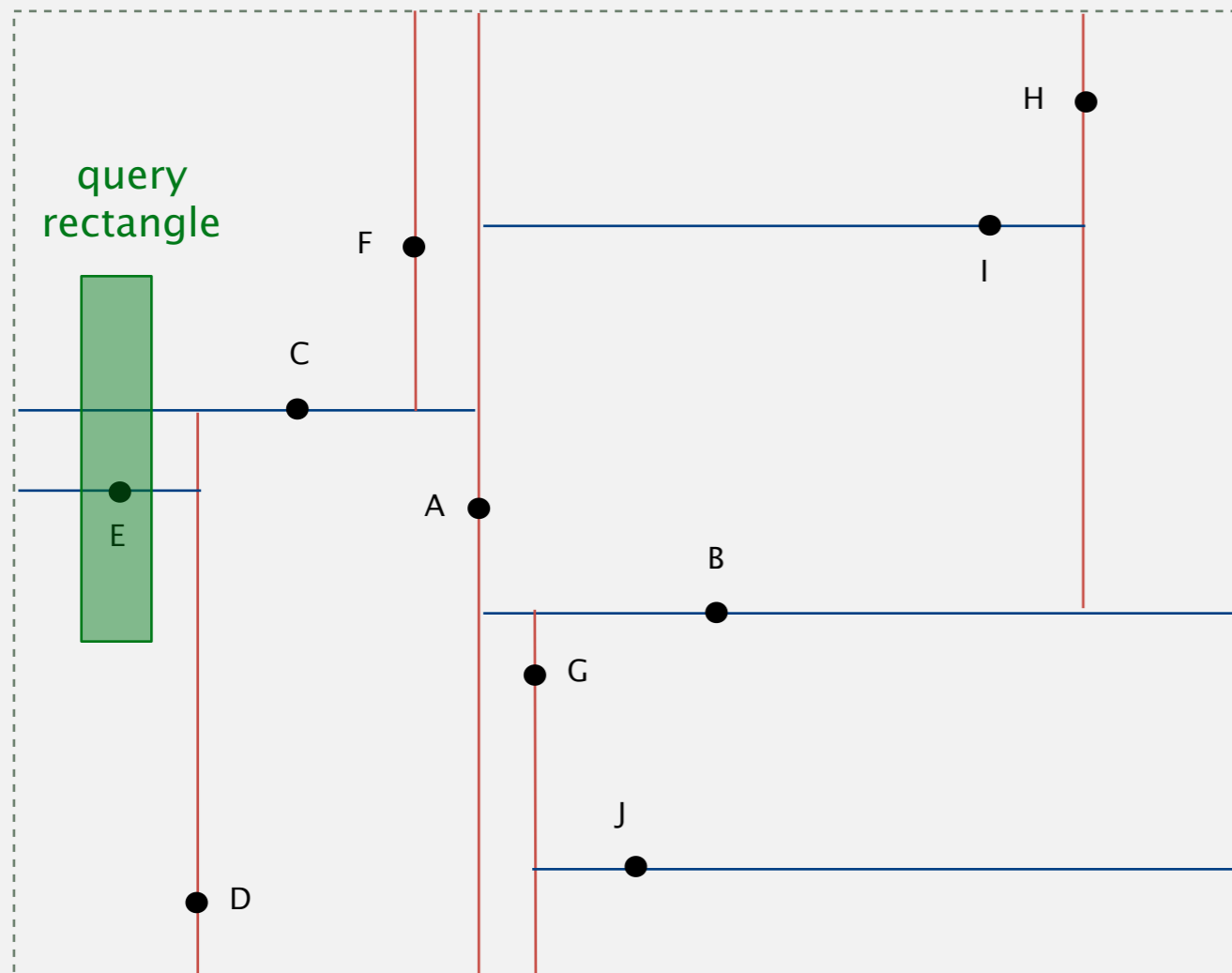
- Search gives rectangle containing point.
- Insert further subdivides the plane.



# 2d tree demo: range search

**Goal.** Find all points in a query axis-aligned rectangle.

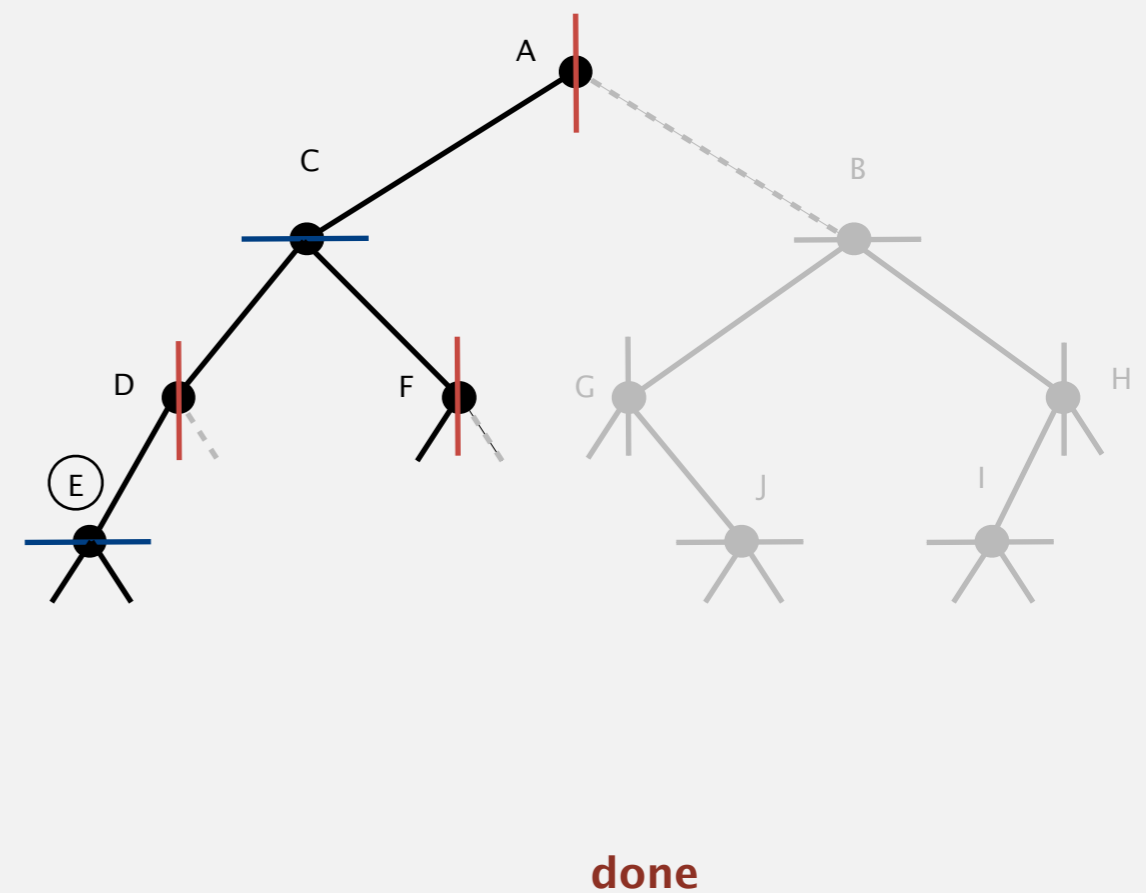
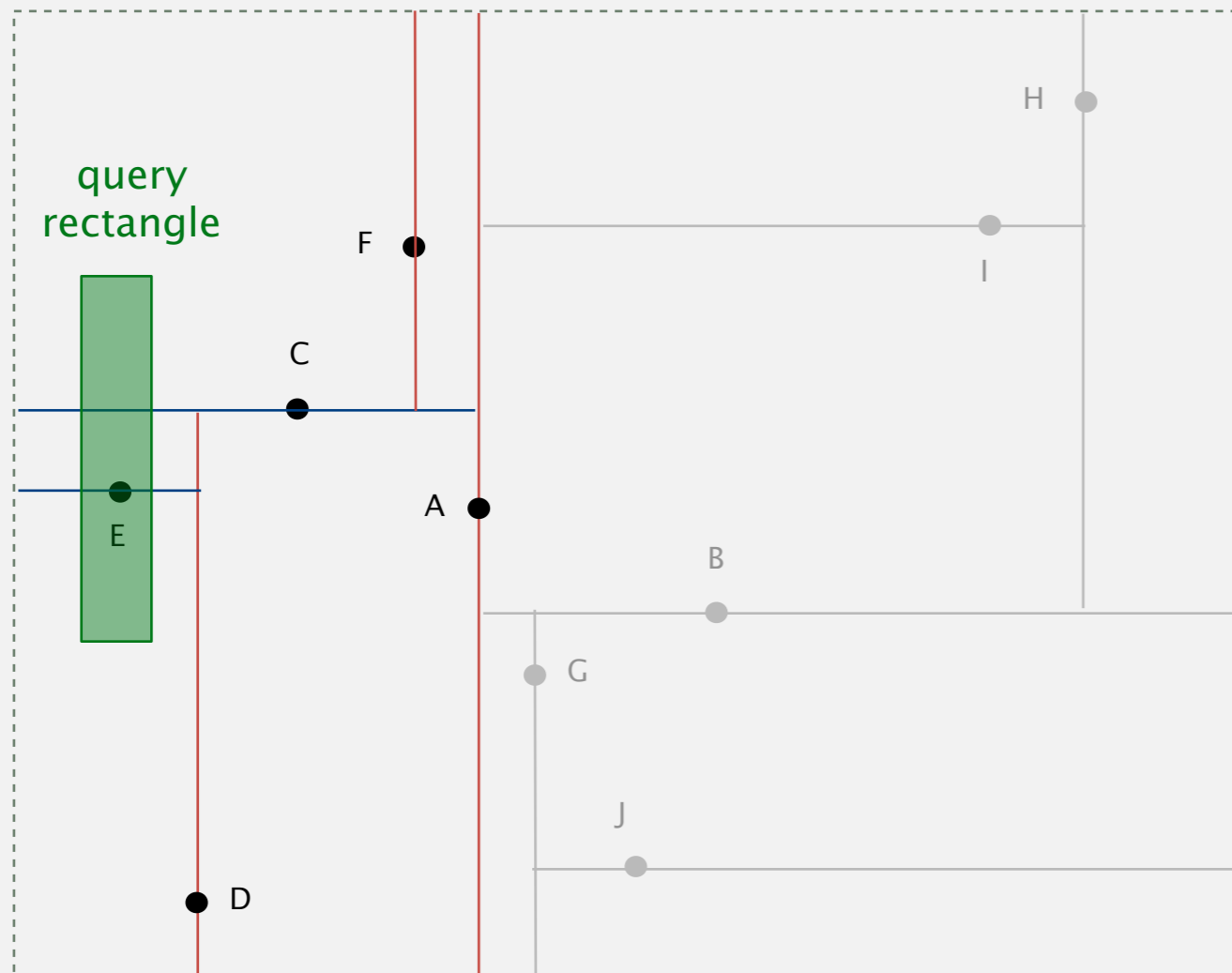
- Check if point in node lies in given rectangle.
- Recursively search left/bottom (if any could fall in rectangle).
- Recursively search right/top (if any could fall in rectangle).



## 2d tree demo: range search

**Goal.** Find all points in a query axis-aligned rectangle.

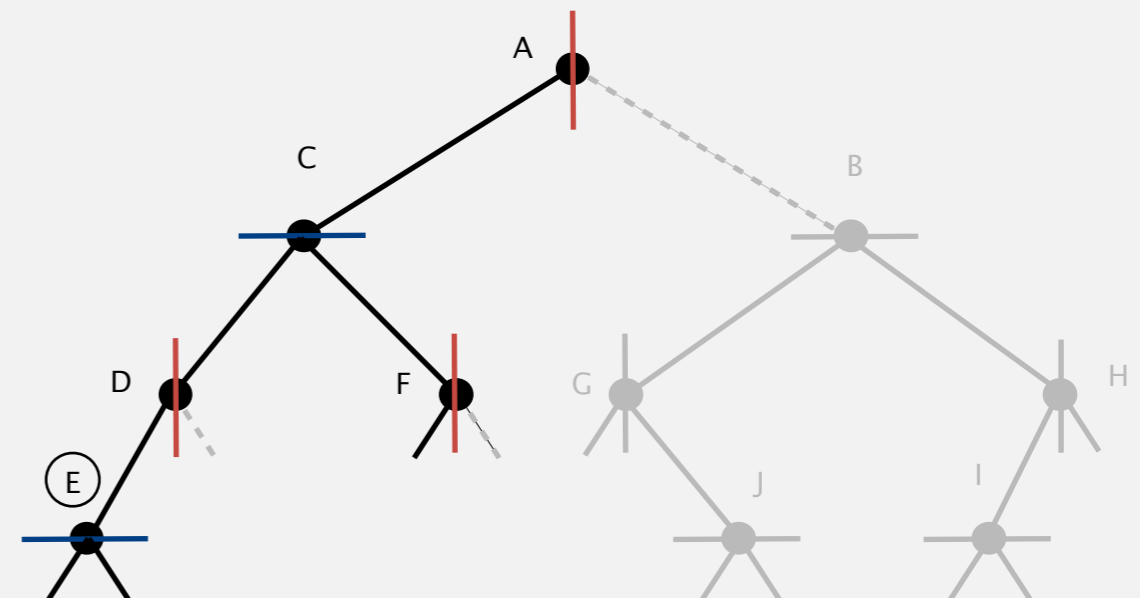
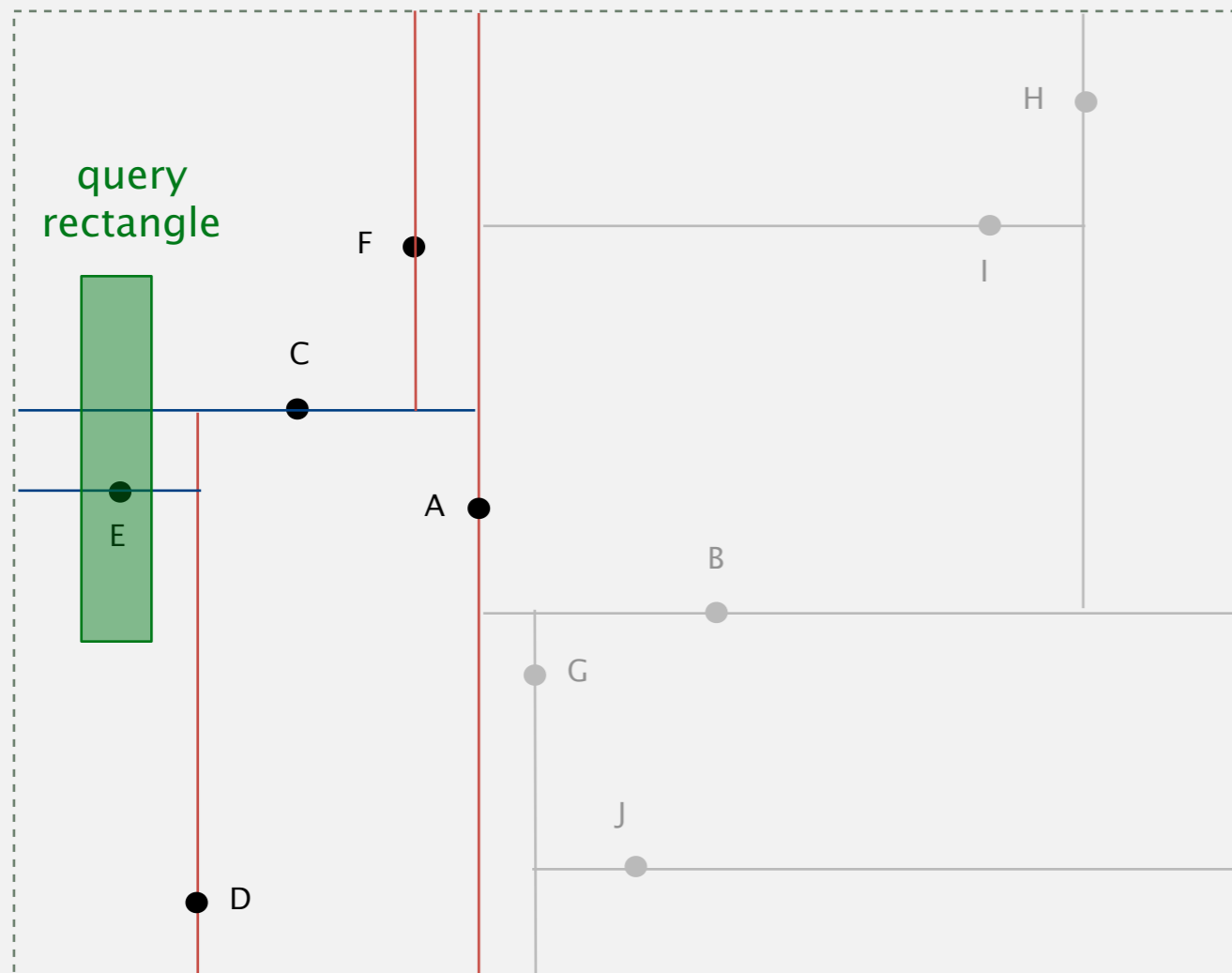
- Check if point in node lies in given rectangle.
- Recursively search left/bottom (if any could fall in rectangle).
- Recursively search right/top (if any could fall in rectangle).



# Range search in a 2d tree analysis

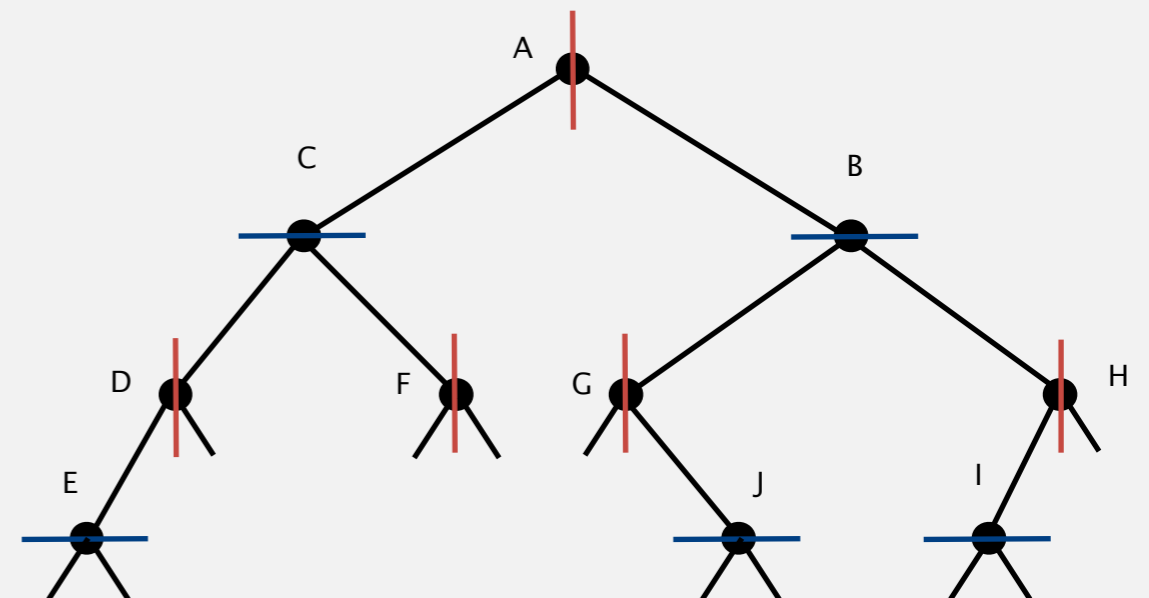
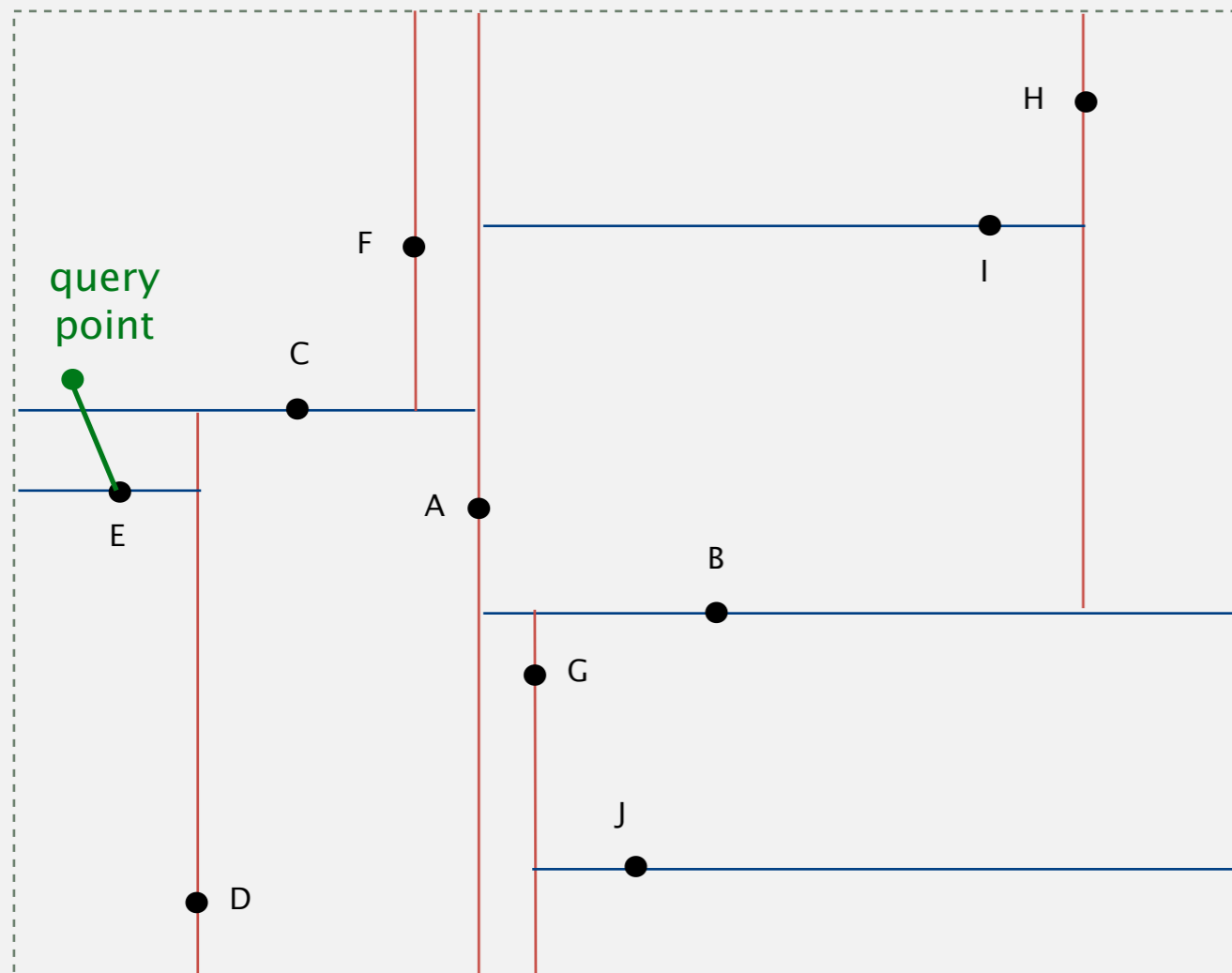
Typical case.  $R + \log n$ .

Worst case (assuming tree is balanced).  $R + \sqrt{n}$ .



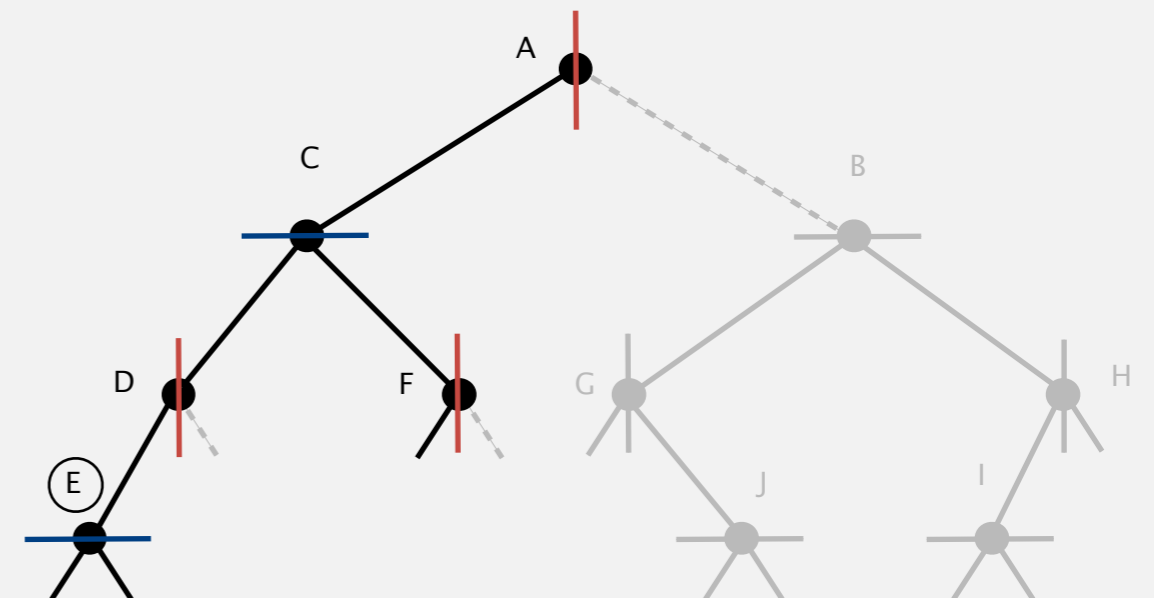
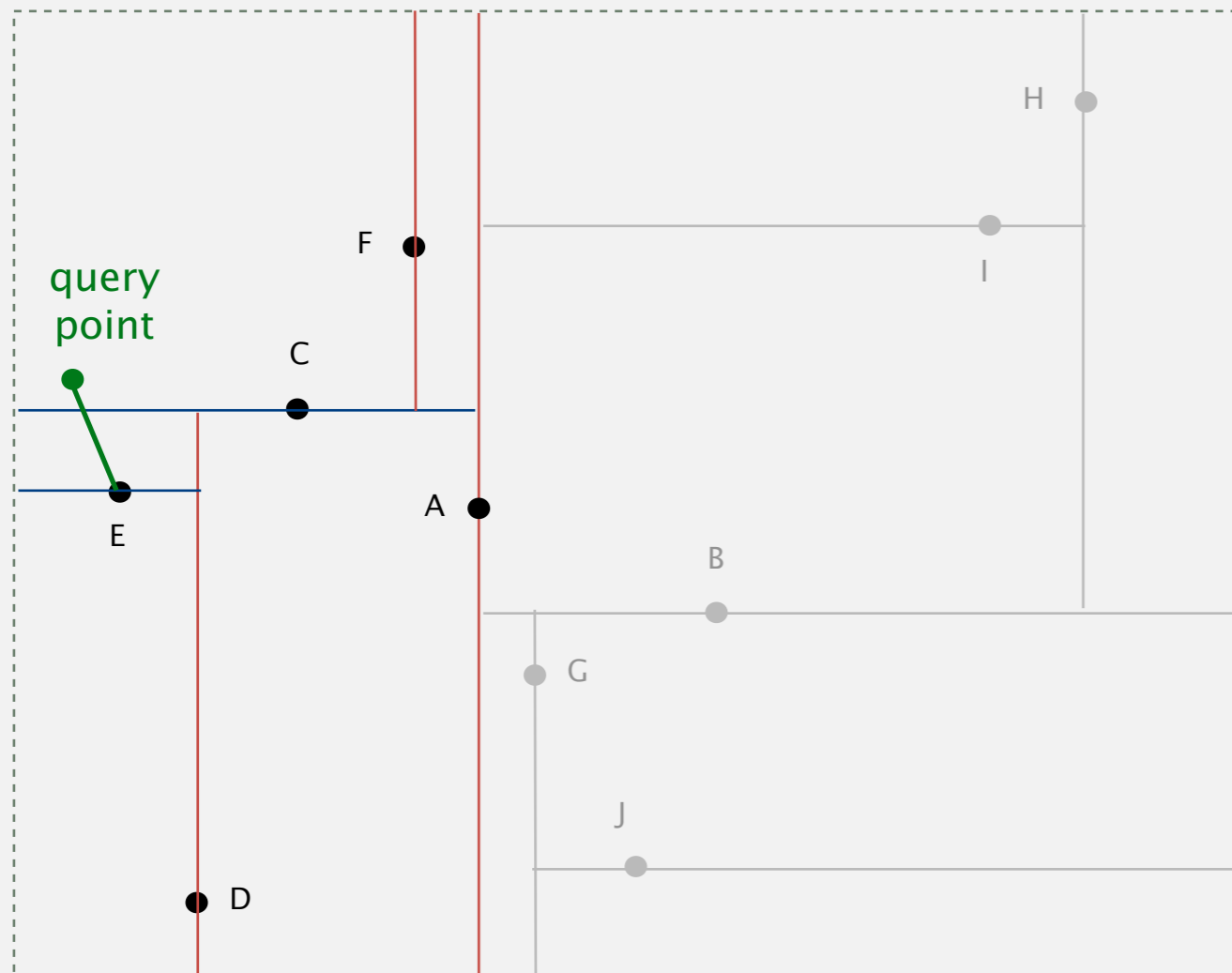
# 2d tree demo: nearest neighbor

**Goal.** Find closest point to query point.



## 2d tree demo: nearest neighbor

- Check distance from point in node to query point.
- Recursively search left/bottom (if it could contain a closer point).
- Recursively search right/top (if it could contain a closer point).
- Organize method so that it begins by searching for query point.



nearest neighbor = E



# Geometric search: quiz

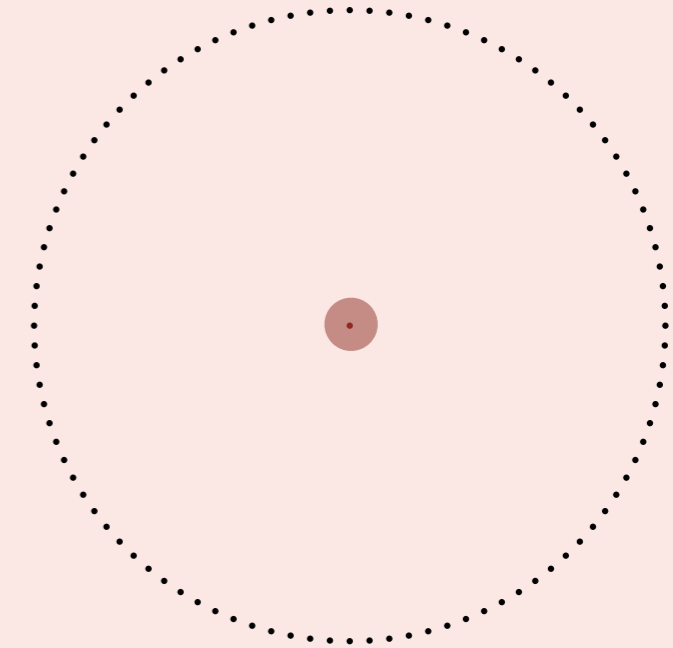
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Which of the following is the worst case for nearest-neighbor search?

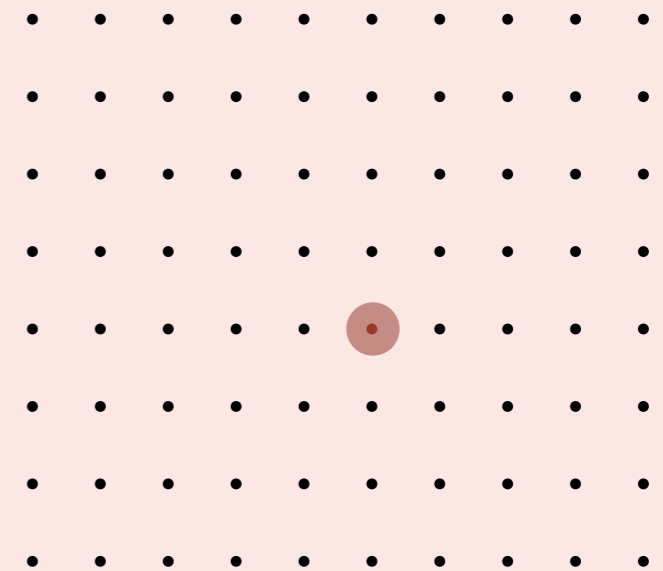
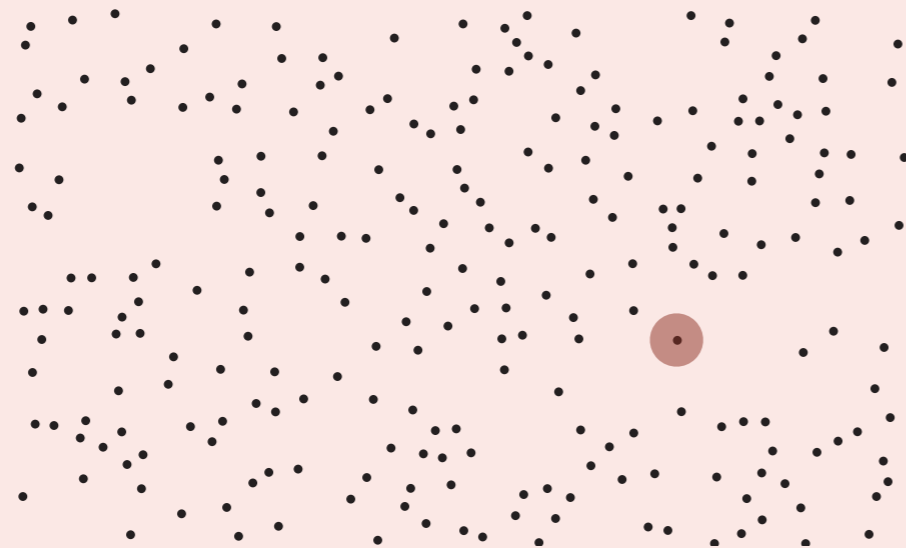
A.



C.



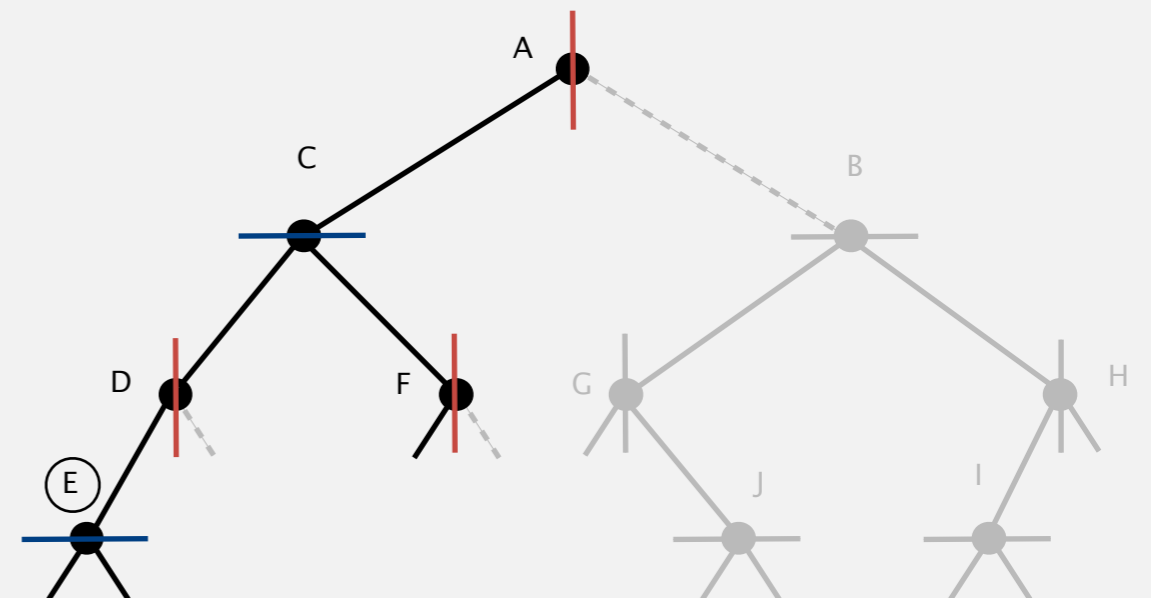
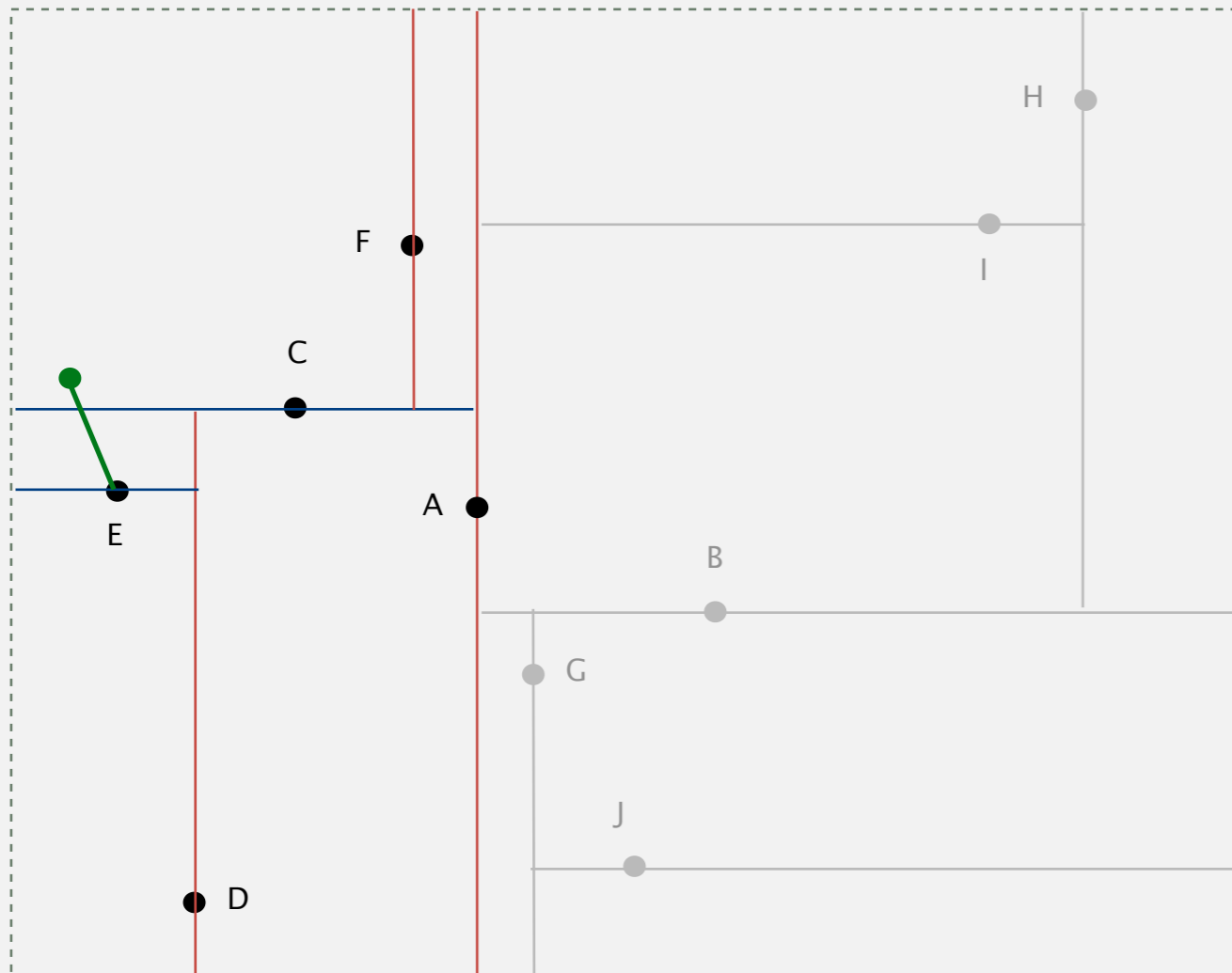
B.



# Nearest neighbor search in a 2d tree analysis

Typical case.  $\log n$ .

Worst case (even if tree is balanced).  $n$ .



nearest neighbor = E

# Flocking birds

---

Q. Which “natural algorithm” do starlings, migrating geese, starlings, cranes, bait balls of fish, and flashing fireflies use to flock?



<http://www.youtube.com/watch?v=XH-groCeKbE>

# Flocking boids [Craig Reynolds, 1986]

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**Boids.** Three simple rules lead to complex emergent flocking behavior:

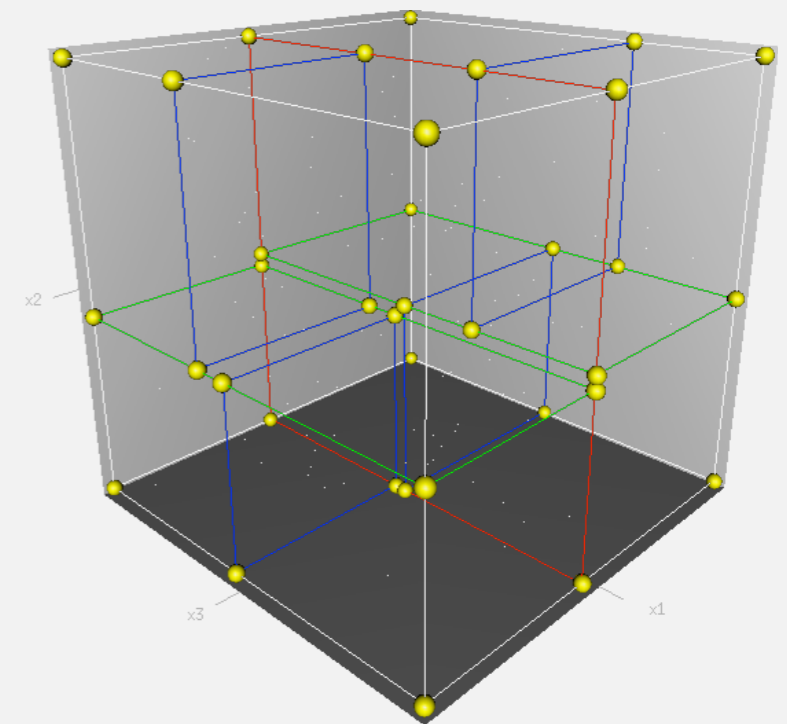
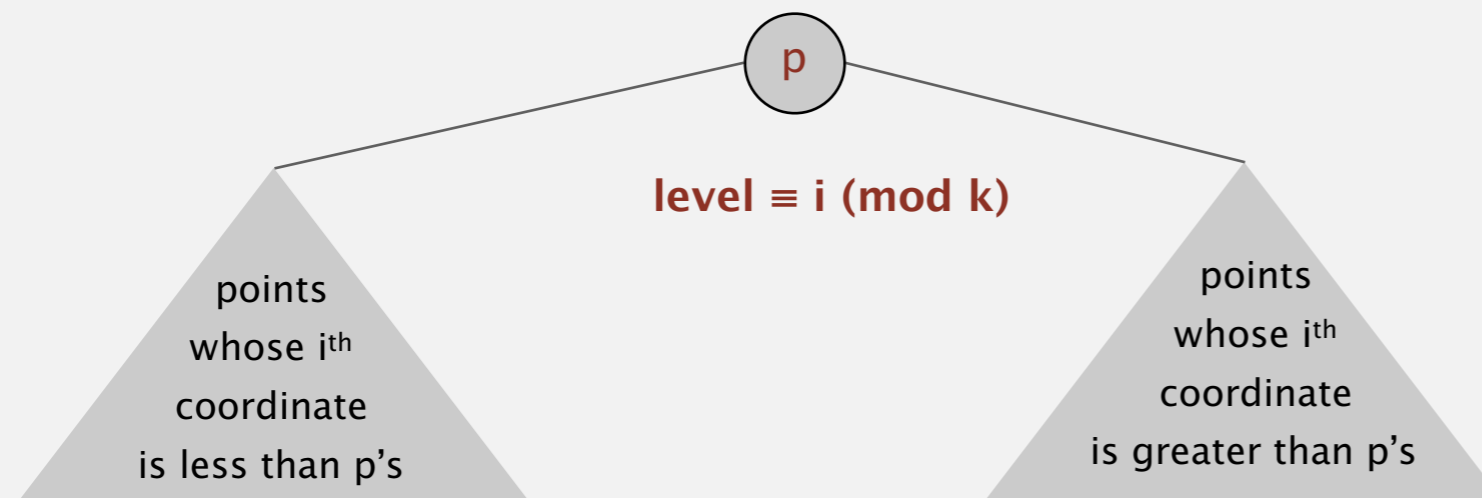
- Collision avoidance: point away from **k-nearest** boids.
- Flock centering: point towards the center of mass of **k-nearest** boids.
- Velocity matching: update velocity to the average of **k-nearest** boids.



# Kd tree

**Kd tree.** Recursively partition  $k$ -dimensional space into 2 halfspaces.

**Implementation.** BST, but cycle through dimensions ala 2d trees.



Efficient, simple data structure for processing  $k$ -dimensional data.

- Widely used.
- Adapts well to high-dimensional and clustered data.
- Discovered by an undergrad in an algorithms class!



Jon Bentley

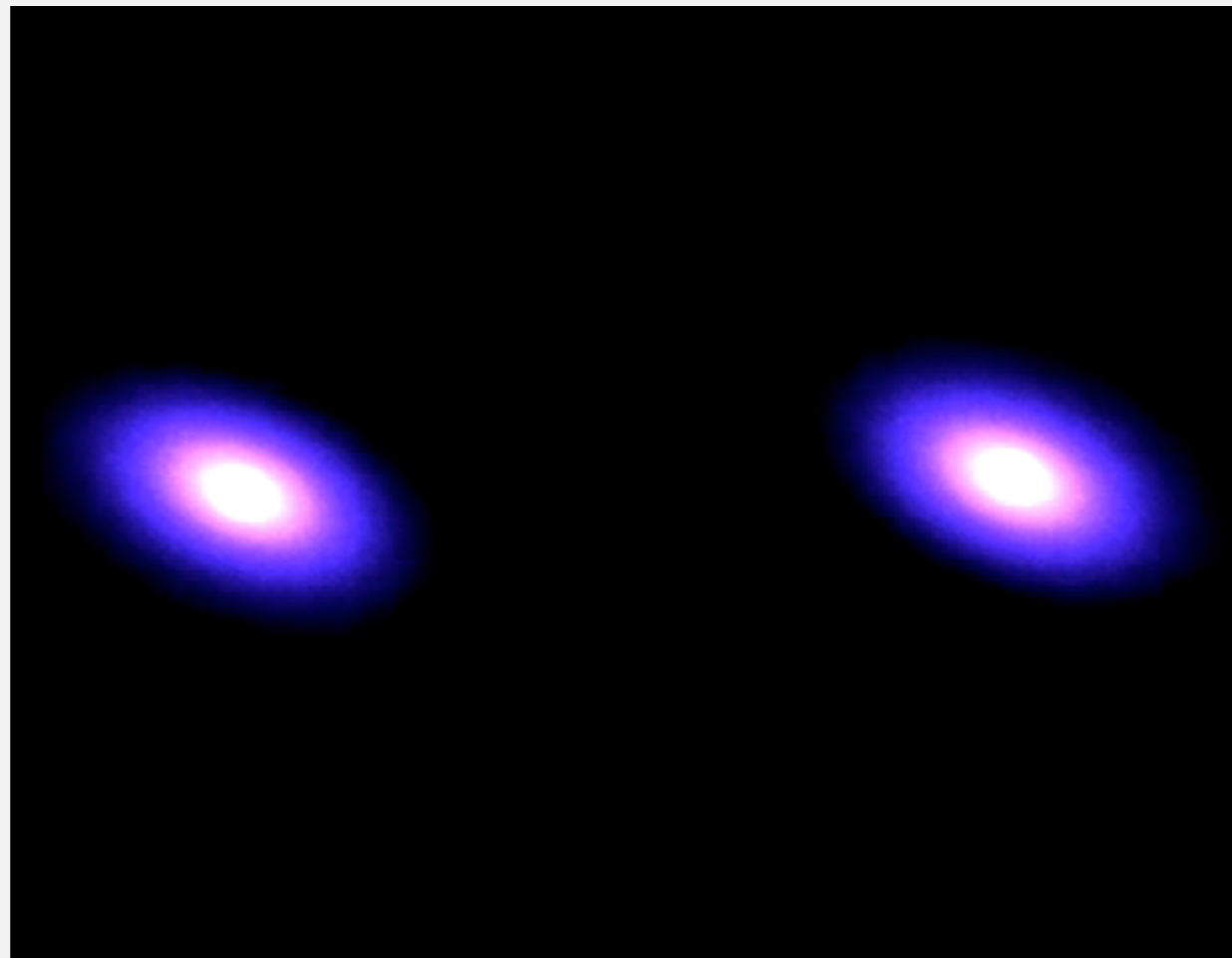
# N-body simulation

---

**Goal.** Simulate the motion of  $n$  particles, mutually affected by gravity.

**Brute force.** For each pair of particles, compute force:  $F = \frac{G m_1 m_2}{r^2}$

**Running time.** Time per step is  $n^2$ .



[http://www.youtube.com/watch?v=ua7YIN4eL\\_w](http://www.youtube.com/watch?v=ua7YIN4eL_w)



# Appel's algorithm for n-body simulation

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**Key idea.** Suppose particle is far, far away from cluster of particles.

- Treat cluster of particles as a single aggregate particle.
- Compute force between particle and **center of mass** of aggregate.



# Appel's algorithm for n-body simulation

---

- Build 3d-tree with  $n$  particles as nodes.
- Store center-of-mass of subtree in each node.
- To compute total force acting on a particle, traverse tree, but stop as soon as distance from particle to subdivision is sufficiently large.

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## AN EFFICIENT PROGRAM FOR MANY-BODY SIMULATION\*


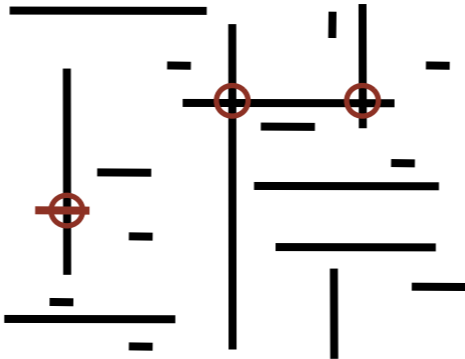
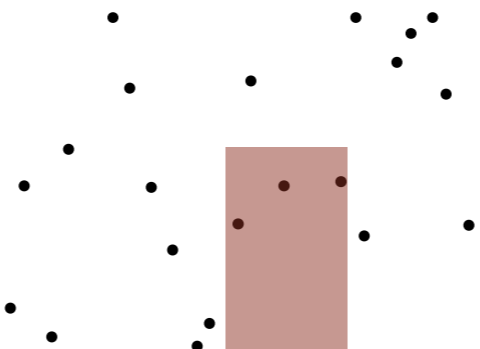
ANDREW W. APPEL†

**Abstract.** The simulation of  $N$  particles interacting in a gravitational force field is useful in astrophysics, but such simulations become costly for large  $N$ . Representing the universe as a tree structure with the particles at the leaves and internal nodes labeled with the centers of mass of their descendants allows several simultaneous attacks on the computation time required by the problem. These approaches range from algorithmic changes (replacing an  $O(N^2)$  algorithm with an algorithm whose time-complexity is believed to be  $O(N \log N)$ ) to data structure modifications, code-tuning, and hardware modifications. The changes reduced the running time of a large problem ( $N = 10,000$ ) by a factor of four hundred. This paper describes both the particular program and the methodology underlying such speedups.

**Impact.** Running time per step is  $n \log n \Rightarrow$  enables new research.



# Geometric applications of BSTs

problem	example	solution
1d range search		<i>binary search tree</i>
2d orthogonal line segment intersection		<i>sweep line reduces problem to 1d range search</i>
2d range search kd range search		<i>2d tree kd tree</i>