# Algorithms

ROBERT SEDGEWICK | KEVIN WAYNE

# Algorithms

**~** 

Robert Sedgewick | Kevin Wayne

http://algs4.cs.princeton.edu

## **3.5 SYMBOL TABLE APPLICATIONS**

▶ sets

dictionary clients

indexing clients

sparse vectors

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#### Mathematical set. A collection of distinct keys.

public	class SET <key extend<="" th=""><th>ds Comparable<key>&gt;</key></th></key>	ds Comparable <key>&gt;</key>
	SET()	create an empty set
void	add(Key key)	add the key to the set
boolean	<pre>contains(Key key)</pre>	is the key in the set?
void	remove(Key key)	remove the key from the set
int	size()	return the number of keys in the set
Iterator <key></key>	iterator()	iterator through keys in the set

#### Q. How to implement?

## **Exception filter**

- Read in a list of words from one file.
- Print out all words from standard input that are { in, not in } the list.



## **Exception filter applications**

- Read in a list of words from one file.
- Print out all words from standard input that are { in, not in } the list.

application	purpose	key	in list
spell checker	identify misspelled words	word	dictionary words
browser	mark visited pages	URL	visited pages
parental controls	block sites	URL	bad sites
chess	detect draw	board	positions
spam filter	eliminate spam	IP address	spam addresses
credit cards	check for stolen cards	number	stolen cards

## Exception filter: Java implementation

- Read in a list of words from one file.
- Print out all words from standard input that are in the list.



## Exception filter: Java implementation

- Read in a list of words from one file.
- Print out all words from standard input that are not in the list.



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## **Dictionary** lookup

#### Command-line arguments.

- A comma-separated value (CSV) file.
- Key field.
- Value field.

#### Ex 1. DNS lookup.



#### % more ip.csv

www.princeton.edu,128.112.128.15 www.cs.princeton.edu, 128.112.136.35 www.math.princeton.edu,128.112.18.11 www.cs.harvard.edu,140.247.50.127 www.harvard.edu,128.103.60.24 www.yale.edu,130.132.51.8 www.econ.yale.edu,128.36.236.74 www.cs.yale.edu,128.36.229.30 espn.com,199.181.135.201 yahoo.com,66.94.234.13 msn.com,207.68.172.246 google.com,64.233.167.99 baidu.com,202.108.22.33 yahoo.co.jp,202.93.91.141 sina.com.cn,202.108.33.32 ebay.com,66.135.192.87 adobe.com,192.150.18.60 163.com,220.181.29.154 passport.net,65.54.179.226 tom.com, 61.135.158.237 nate.com,203.226.253.11 cnn.com,64.236.16.20 daum.net,211.115.77.211 blogger.com,66.102.15.100 fastclick.com,205.180.86.4 wikipedia.org,66.230.200.100 rakuten.co.jp,202.72.51.22

## **Dictionary** lookup

#### Command-line arguments.

- A comma-separated value (CSV) file.
- Key field.
- Value field.
- Ex 2. Amino acids.



#### % more amino.csv

TTT, Phe, F, Phenylalanine TTC, Phe, F, Phenylalanine TTA, Leu, L, Leucine TTG,Leu,L,Leucine TCT, Ser, S, Serine TCC, Ser, S, Serine TCA, Ser, S, Serine TCG,Ser,S,Serine TAT, Tyr, Y, Tyrosine TAC, Tyr, Y, Tyrosine TAA, Stop, Stop, Stop TAG, Stop, Stop, Stop TGT,Cys,C,Cysteine TGC,Cys,C,Cysteine TGA, Stop, Stop, Stop TGG, Trp, W, Tryptophan CTT,Leu,L,Leucine CTC,Leu,L,Leucine CTA, Leu, L, Leucine CTG,Leu,L,Leucine CCT, Pro, P, Proline CCC, Pro, P, Proline CCA, Pro, P, Proline CCG, Pro, P, Proline CAT, His, H, Histidine CAC, His, H, Histidine CAA,Gln,Q,Glutamine CAG,Gln,Q,Glutamine CGT, Arg, R, Arginine CGC, Arg, R, Arginine

. . .

## **Dictionary lookup**

#### Command-line arguments.

- A comma-separated value (CSV) file.
- Key field.
- Value field.
- Ex 3. Class list.



#### % more classlist.csv 13, Berl, Ethan Michael, P01, eberl 12, Cao, Phillips Minghua, P01, pcao 11, Chehoud, Christel, P01, cchehoud 10, Douglas, Malia Morioka, P01, malia 12, Haddock, Sara Lynn, P01, shaddock 12, Hantman, Nicole Samantha, P01, nhantman 11, Hesterberg, Adam Classen, P01, ahesterb 13, Hwang, Roland Lee, P01, rhwang 13, Hyde, Gregory Thomas, P01, ghyde 13,Kim,Hyunmoon,P01,hktwo 12,Korac,Damjan,P01,dkorac 11, MacDonald, Graham David, P01, gmacdona 10, Michal, Brian Thomas, P01, bmichal 12, Nam, Seung Hyeon, P01, seungnam 11, Nastasescu, Maria Monica, P01, mnastase 11, Pan, Di, P01, dpan 12, Partridge, Brenton Alan, P01, bpartrid 13, Rilee, Alexander, P01, arilee 13, Roopakalu, Ajay, P01, aroopaka 11, Sheng, Ben C, P01, bsheng 12,Webb,Natalie Sue,P01,nwebb

## Dictionary lookup: Java implementation



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Goal. Index a PC (or the web).

Spotlight	Spotlight searching challenge		
	Show All (200)		
Top Hit	🙀 10Hashing		
Documents	mobydick.txt movies.txt		
	<ul> <li>Papers/Abstracts</li> <li>score.card.txt</li> <li>Requests</li> </ul>		
Mail Messages	<ul> <li>Re: Draft of lecture on symb</li> <li>SODA 07 Final Accepts</li> <li>SODA 07 Summary</li> <li>Got-it</li> <li>No Subject</li> </ul>		
PDF Documents	<ul> <li>08BinarySearchTrees.pdf</li> <li>07SymbolTables.pdf</li> <li>07SymbolTables.pdf</li> <li>06PriorityQueues.pdf</li> <li>06PriorityQueues.pdf</li> </ul>		
Presentations	<ul> <li>10Hashing</li> <li>07SymbolTables</li> <li>06PriorityQueues</li> </ul>		

## File indexing

Goal. Given a list of files specified, create an index so that you can efficiently find all files containing a given query string.

% ls \*.txt
aesop.txt magna.txt moby.txt
sawyer.txt tale.txt

% java FileIndex \*.txt

freedom
magna.txt moby.txt tale.txt

whale
moby.txt

lamb
sawyer.txt aesop.txt

% ls \*.java BlackList.java Concordance.java DeDup.java FileIndex.java ST.java SET.java WhiteList.java

% java FileIndex \*.java

import
FileIndex.java SET.java ST.java

Comparator null

Solution. Key = query string; value = set of files containing that string.

## File indexing



#### Goal. Index for an e-book.

Abstract data type (ADT), 127-195 abstract classes, 163 classes, 129-136 collections of items, 137-139 creating, 157-164 defined, 128 duplicate items, 173-176 equivalence-relations, 159-162 FIFO queues, 165-171 first-class, 177-186 generic operations, 273 index items, 177 insert/remove operations, 138-139 modular programming, 135 polynomial, 188-192 priority queues, 375-376 pushdown stack, 138-156 stubs, 135 symbol table, 497-506 ADT interfaces array (myArray), 274 complex number (Complex), 181 existence table (ET), 663 full priority queue (PQfull), 397 indirect priority queue (PQi), 403 itcm (myItem), 273, 498 kcy (myKey), 498 polynomial (Poly), 189 point (Point), 134 priority queue (PQ), 375 queue of int (intQueue), 166

Index

stack of int (intStack), 140 symbol table (ST), 503 text index (TI), 525 union-find (UF), 159 Abstract in-place merging, 351-353 Abstract operation, 10 Access control state, 131 Actual data, 31 Adapter class, 155-157 Adaptive sort, 268 Address, 84-85 Adjacency list, 120-123 depth-first search, 251-256 Adjacency matrix, 120-122 Ajtai, M., 464 Algorithm, 4-6, 27-64 abstract operations, 10, 31, 34-35 analysis of, 6 average-/worst-case performance, 35, 60-62 big-Oh notation, 44-47 binary search, 56-59 computational complexity, 62-64 efficiency, 6, 30, 32 empirical analysis, 30-32, 58 exponential-time, 219 implementation, 28-30 logarithm function, 40-43 mathematical analysis, 33-36, 58 primary parameter, 36 probabilistic, 331 recurrences, 49-52, 57 recursive, 198 running time, 34-40 search, 53-56, 498 steps in, 22-23 See also Randomized algorithm Amortization approach, 557, 627 Arithmetic operator, 177-179, 188, 191 Array, 12, 83 binary search, 57 dynamic allocation, 87

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#### Concordance

Goal. Preprocess a text corpus to support concordance queries: given a word, find all occurrences with their immediate contexts.

% java Concordance tale.txt cities tongues of the two \*cities\* that were blended in majesty their turnkeys and the \*majesty\* of the law fired me treason against the \*majesty\* of the people in of his most gracious \*majesty\* king george the third

princeton no matches

#### Concordance



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indexing clients

sparse vectors

sets

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## sparse vectors

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#### Matrix-vector multiplication (standard implementation)





## Sparse matrix-vector multiplication

Problem. Sparse matrix-vector multiplication.

Assumptions. Matrix dimension is 10,000; average nonzeros per row ~ 10.



#### Vector representations

#### 1d array (standard) representation.

- Constant time access to elements.
- Space proportional to N.

#### Symbol table representation.

- Key = index, value = entry.
- Efficient iterator.
- Space proportional to number of nonzeros.



## Sparse vector data type



#### Matrix representations

2D array (standard) matrix representation: Each row of matrix is an array.

- Constant time access to elements.
- Space proportional to N<sup>2</sup>.

Sparse matrix representation: Each row of matrix is a sparse vector.

- Efficient access to elements.
- Space proportional to number of nonzeros (plus N).



#### Sparse matrix-vector multiplication





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## GEOMETRIC APPLICATIONS OF BSTS

Id range search

- Ine segment intersection
- kd trees
- interval search trees
- rectangle intersection

This lecture. Intersections among geometric objects.





2d orthogonal range search

orthogonal rectangle intersection

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

Efficient solutions. Binary search trees (and extensions).

## **GEOMETRIC APPLICATIONS OF BSTS**

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## 1d range search

#### 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	В
insert D	B D
insert A	ABD
insert l	ABDI
insert H	ABDHI
insert F	ABDFHI
insert P	ABDFHIP
count G to K	2
search G to K	ΗI

## 1d range search: elementary implementations

Unordered list. Fast insert, slow range search.

Ordered array. Slow insert, binary search for  $k_1$  and  $k_2$  to do range search.

#### order of growth of running time for 1d range search

data structure	insert	range count	range search
unordered list	1	Ν	Ν
ordered array	Ν	log N	R + log 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 1o and hi?



**Proposition.** Running time proportional to log *N*.

Pf. Nodes examined = search path to 10 + search path to hi.

1d range search. Find all keys between 1o and hi.

- 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).

searching in the range [F...T]



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

Pf. Nodes examined = search path to 10 + search path to hi + matches.

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## Orthogonal line segment intersection

Given *N* horizontal and vertical line segments, find all intersections.



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

Nondegeneracy 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.



## 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.



## 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.



## 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.

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

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## 2-d orthogonal range search

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

- Insert a 2d key.
- Delete 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





## 2d orthogonal range search: grid implementation

#### Grid implementation.

- Divide space into *M*-by-*M* grid of squares.
- Create list of points contained in each square.
- Use 2d array to directly index relevant square.
- Insert: add (x, y) to list for corresponding square.
- Range search: examine only squares that intersect 2d range query.



choose M ~  $\sqrt{N}$ 

#### Space-time tradeoff.

- Space:  $M^2 + N$ .
- Time:  $1 + N/M^2$  per square examined, on average.

#### Choose grid square size to tune performance.

- Too small: wastes space.
- Too large: too many points per square.
- Rule of thumb:  $\sqrt{N-by}-\sqrt{N}$  grid.

#### Running time. [if points are evenly distributed]

- Initialize data structure: N.
- Insert point: 1.
- Range search: 1 per point in range.



Grid implementation. Fast, simple solution for evenly-distributed points.

Problem. Clustering a well-known phenomenon in geometric data.

- Lists are too long, even though average length is short.
- Need data structure that adapts gracefully to data.



Grid implementation. Fast, simple solution for evenly-distributed points.

Problem. Clustering a well-known phenomenon in geometric data.



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

Grid. Divide space uniformly into squares.

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

Quadtree. Recursively divide space into four quadrants.

BSP tree. Recursively divide space into two regions.



## 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.





## 2d tree construction

Recursively partition plane into two halfplanes.



### 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.



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 demo

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).  $R + \sqrt{N}$ .



## Nearest neighbor search in a 2d tree demo



### Nearest neighbor search in a 2d tree demo

- 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 search in a 2d tree analysis

Typical case. log *N*. Worst case (even if tree is balanced). *N*.



## Flocking birds

Q. What "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]

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. 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

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=ua7Y1N4eL\_w

## Appel's algorithm for N-body simulation

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.

SIAM J. SCI. STAT. COMPUT. Vol. 6, No. 1, January 1985 © 1985 Society for Industrial and Applied Mathematics 008

#### 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

# Id range search

interval search trees

rectangle intersection

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### 1d interval search

1d interval search. Data structure to hold set of (overlapping) intervals.

- Insert an interval ( *lo*, *hi* ).
- Search for an interval ( *lo*, *hi* ).
- Delete an interval ( *lo*, *hi* ).
- Interval intersection query: given an interval (*lo*, *hi*), find all intervals (or one interval) in data structure that intersects (*lo*, *hi*).

Q. Which intervals intersect (9, 16)?A. (7, 10) and (15, 18).



public class	IntervalST <key< th=""><th>extends</th><th>Comparab</th><th>le<key>,</key></th><th>Value&gt;</th></key<>	extends	Comparab	le <key>,</key>	Value>
--------------	--	---------	----------	-----------------	--------

IntervalST()

void put(Key lo, Key hi, Value val)

Value get(Key lo, Key hi)

void delete(Key lo, Key hi)

Iterable<Value> intersects(Key lo, Key hi)

create interval search tree

put interval-value pair into ST

value paired with given interval

delete the given interval

all intervals that intersect the given interval

Nondegeneracy assumption. No two intervals have the same left endpoint.
#### Interval search trees

Create BST, where each node stores an interval ( lo, hi ).

- Use left endpoint as BST key.
- Store max endpoint in subtree rooted at node.



#### Interval search tree demo

To insert an interval (*lo*, *hi*):

- Insert into BST, using *lo* as the key.
- Update max in each node on search path.



#### insert interval (16, 22)



#### Interval search tree demo

To search for any one interval that intersects query interval (*lo*, *hi*):

- If interval in node intersects query interval, return it.
- Else if left subtree is null, go right.
- Else if max endpoint in left subtree is less than lo, go right.
- Else go left.



#### Search for an intersecting interval implementation

To search for any one interval that intersects query interval ( lo, hi ):

- If interval in node intersects query interval, return it.
- Else if left subtree is null, go right.
- Else if max endpoint in left subtree is less than lo, go right.
- Else go left.

```
Node x = root;
while (x != null)
{
    if (x.interval.intersects(lo, hi)) return x.interval;
    else if (x.left == null) x = x.right;
    else if (x.left.max < lo) x = x.right;
    else x = x.left;
}
return null;
```

#### Search for an intersecting interval analysis

To search for any one interval that intersects query interval (*lo*, *hi*):

- If interval in node intersects query interval, return it.
- Else if left subtree is null, go right.
- Else if max endpoint in left subtree is less than *lo*, go right.
- Else go left.

Case 1. If search goes right, then no intersection in left.

- Pf. Suppose search goes right and left subtree is non empty.
  - Max endpoint *max* in left subtree is less than *lo*.
  - For any interval (*a*, *b*) in left subtree of *x*,



To search for any one interval that intersects query interval (*lo*, *hi*):

- If interval in node intersects query interval, return it.
- Else if left subtree is null, go right.
- Else if max endpoint in left subtree is less than lo, go right.
- Else go left.

Case 2. If search goes left, then there is either an intersection in left subtree or no intersections in either.

Pf. Suppose no intersection in left.

- Since went left, we have  $lo \leq max$ .
- Then for any interval (*a*, *b*) in right subtree of *x*,



max

#### Interval search tree: analysis

#### Implementation. Use a red-black BST to guarantee performance.

easy to maintain auxiliary information using log N extra work per op

operation	brute	interval search tree	best in theory
insert interval	1	log N	log N
find interval	Ν	log N	log N
delete interval	Ν	log N	log N
find any one interval that intersects ( <i>lo</i> , <i>hi</i> )	Ν	log N	log N
find all intervals that intersects ( <i>lo</i> , <i>hi</i> )	Ν	R log N	R + log N

order of growth of running time for N intervals

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rectangle intersection

interval search trees

line segment intersection

1d range search

kd trees

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#### Orthogonal rectangle intersection

Goal. Find all intersections among a set of *N* orthogonal rectangles.

Quadratic algorithm. Check all pairs of rectangles for intersection.



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

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 = orthogonal rectangle intersection search.





#### Algorithms and Moore's law

"Moore's law." Processing power doubles every 18 months.

- 197*x*: check *N* rectangles.
- 197(*x*+1.5): check 2*N* rectangles on a 2*x*-faster computer.



Gordon Moore

**Bootstrapping.** We get to use the faster computer for bigger circuits.

But bootstrapping is not enough if using a quadratic algorithm:

- 197*x*: takes *M* days.
- 197(x+1.5): takes (4M)/2 = 2M days. (!)

quadratic algorithm

X 2x-faster computer

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

#### Orthogonal rectangle intersection: sweep-line algorithm

#### Sweep vertical line from left to right.

- *x*-coordinates of left and right endpoints define events.
- Maintain set of rectangles that intersect the sweep line in an interval search tree (using *y*-intervals of rectangle).
- Left endpoint: interval search for *y*-interval of rectangle; insert *y*-interval.
- Right endpoint: remove *y*-interval.



y-coordinates

#### Orthogonal rectangle intersection: sweep-line analysis

**Proposition.** Sweep line algorithm takes time proportional to  $N \log N + R \log N$ to find *R* intersections among a set of *N* rectangles.

#### Pf.

- Put *x*-coordinates on a PQ (or sort). N log N
- Insert *y*-intervals into ST. — Nlog N
- Delete *y*-intervals from ST. — Nlog N
- Interval searches for y-intervals.  $\leftarrow$  N log N + R log N

Bottom line. Sweep line reduces 2d orthogonal rectangle intersection search to 1d interval search.

### Geometric applications of BSTs

problem	example	solution
1d range search	•• •• •• •• • <mark>•• • •</mark> • •• •• •• •	BST
2d orthogonal line segment intersection		sweep line reduces to 1d range search
kd range search		kd tree
1d interval search		interval search tree
2d orthogonal rectangle intersection		sweep line reduces to 1d interval search

## GEOMETRIC APPLICATIONS OF BSTS

# Algorithms

rectangle intersection

interval search trees

line segment intersection

1d range search

kd trees

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# Algorithms

ROBERT SEDGEWICK | KEVIN WAYNE

# Algorithms

 $\checkmark$ 

Robert Sedgewick | Kevin Wayne

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## GEOMETRIC APPLICATIONS OF BSTS

Id range search

- Ine segment intersection
- kd trees
- interval search trees
- rectangle intersection