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http://www.cs.princeton.edu/~wayne/kleinberg-tardos

# 5. DIVIDE AND CONQUER I

- mergesort
- counting inversions
- closest pair of points
- randomized quicksort
- median and selection

### Divide-and-conquer.

- Divide up problem into several subproblems.
- Solve each subproblem recursively.
- Combine solutions to subproblems into overall solution.

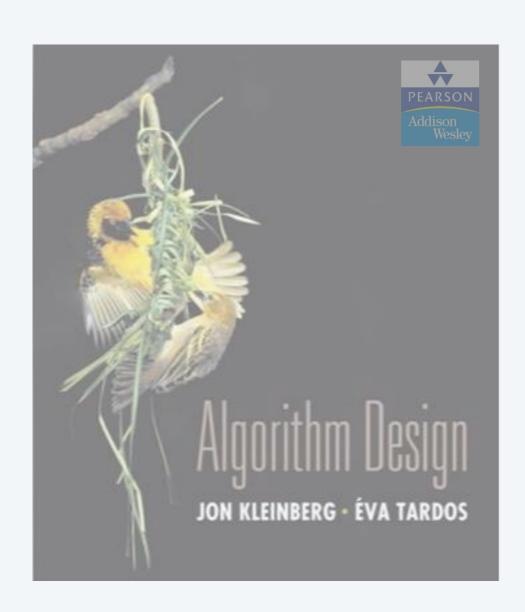
### Most common usage.

- Divide problem of size n into two subproblems of size n/2 in linear time.
- Solve two subproblems recursively.
- Combine two solutions into overall solution in linear time.

### Consequence.

- Brute force:  $\Theta(n^2)$ .
- Divide-and-conquer:  $\Theta(n \log n)$ .

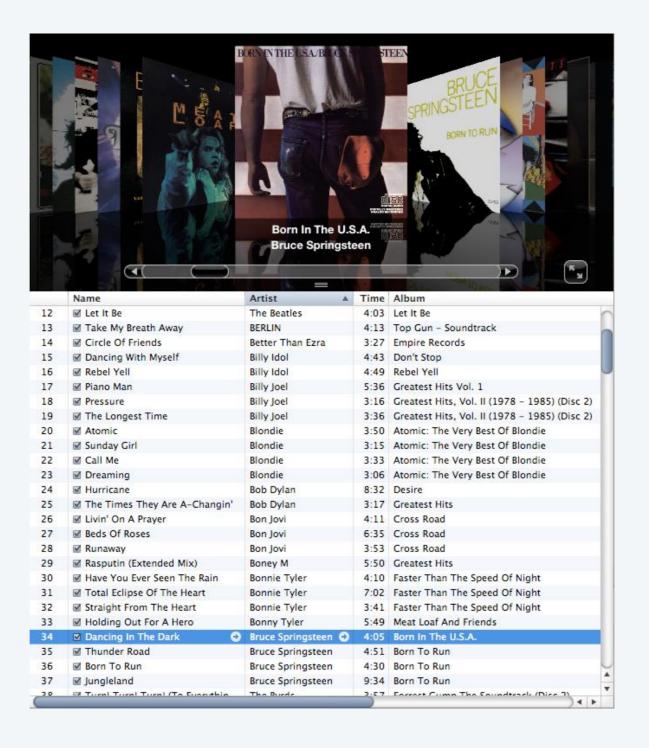




# 5. DIVIDE AND CONQUER

- mergesort
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Problem. Given a list of n elements from a totally-ordered universe, rearrange them in ascending order.



### Sorting applications

### Obvious applications.

- Organize an MP3 library.
- Display Google PageRank results.
- List RSS news items in reverse chronological order.

### Some problems become easier once elements are sorted.

- Identify statistical outliers.
- Binary search in a database.
- Remove duplicates in a mailing list.

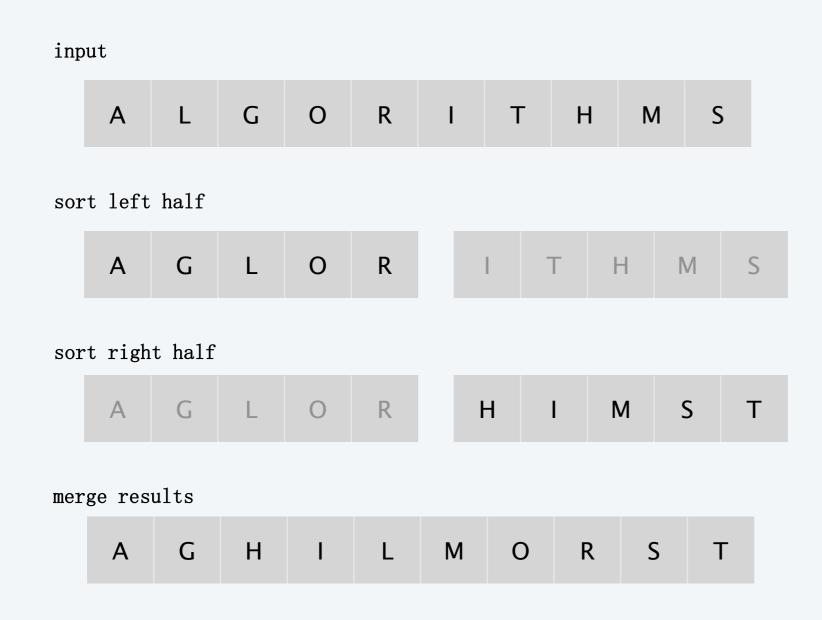
### Non-obvious applications.

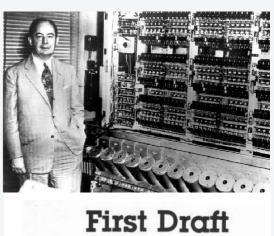
- Convex hull.
- Closest pair of points.
- Interval scheduling / interval partitioning.
- Minimum spanning trees (Kruskal's algorithm).
- Scheduling to minimize maximum lateness or average completion time.

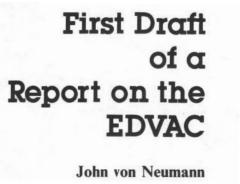
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### Mergesort

- Recursively sort left half.
- Recursively sort right half.
- Merge two halves to make sorted whole.







### Goal. Combine two sorted lists A and B into a sorted whole C.

0

- Scan A and B from left to right.
- Compare  $a_i$  and  $b_j$ .
- If  $a_i \leq b_j$ , append  $a_i$  to C (no larger than any remaining element in B).
- If  $a_i > b_j$ , append  $b_j$  to C (smaller than every remaining element in A).



Def.  $T(n) = \max$  number of compares to mergesort a list of size  $\leq n$ . Note. T(n) is monotone nondecreasing.

Mergesort recurrence.

$$T(n) \le \begin{cases} 0 & \text{if } n = 1 \\ T(\lceil n/2 \rceil) + T(\lceil n/2 \rceil) + n & \text{otherwise} \end{cases}$$

**Solution.** T(n) is  $O(n \log_2 n)$ .

Assorted proofs. We describe several ways to prove this recurrence. Initially we assume n is a power of 2 and replace  $\leq$  with =.

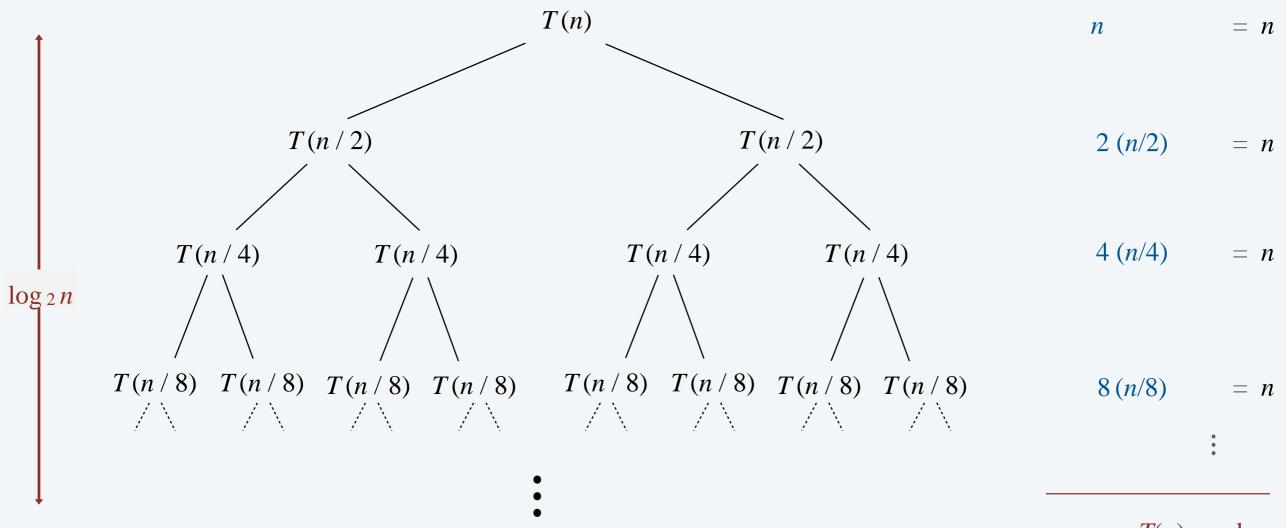
## Divide-and-conquer recurrence: proof by recursion tree

Proposition. If T(n) satisfies the following recurrence, then  $T(n) = n \log_2 n$ .

$$T(n) = \begin{cases} 0 & \text{if } n = 1 \\ 2 T(n/2) + n & \text{otherwise} \end{cases}$$

assuming n is a power of 2

Pf 1.



Proposition. If T(n) satisfies the following recurrence, then  $T(n) = n \log_2 n$ .

$$T(n) = \begin{cases} 0 & \text{if } n = 1 \\ 2 T(n/2) + n & \text{otherwise} \end{cases}$$

assuming n is a power of 2

### Pf 2. [by induction on *n*]

- Base case: when n = 1, T(1) = 0.
- Inductive hypothesis: assume  $T(n) = n \log_2 n$ .
- Goal: show that  $T(2n) = 2n \log_2 (2n)$ .

$$T(2n) = 2 T(n) + 2n$$
  
=  $2 n \log_2 n + 2n$   
=  $2 n (\log_2 (2n) - 1) + 2n$   
=  $2 n \log_2 (2n)$ .

Claim. If T(n) satisfies the following recurrence, then  $T(n) \le n \log_2 n$ .

$$T(n) \le \begin{cases} 0 & \text{if } n = 1 \\ T(\lceil n/2 \rceil) + T(\lceil n/2 \rceil) + n & \text{otherwise} \end{cases}$$

## Pf. [by strong induction on n]

- Base case: n = 1.
- Define  $n_1 = \lfloor n/2 \rfloor$  and  $n_2 = \lceil n/2 \rceil$ .
- Induction step: assume true for 1, 2, ..., n-1.

$$T(n) \leq T(n_{1}) + T(n_{2}) + n \leq \left\lceil 2^{\lceil \log_{2} n \rceil} / 2 \right\rceil$$

$$\leq n_{1} \left\lceil \log_{2} n_{1} \right\rceil + n_{2} \left\lceil \log_{2} n_{2} \right\rceil + n = 2^{\lceil \log_{2} n \rceil} / 2$$

$$\leq n_{1} \left\lceil \log_{2} n_{2} \right\rceil + n_{2} \left\lceil \log_{2} n_{2} \right\rceil + n$$

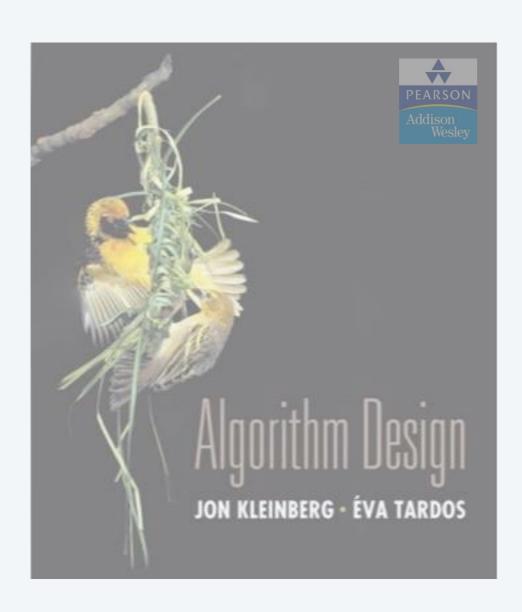
$$= n \left\lceil \log_{2} n_{1} - 1 \right\rceil + n$$

$$\leq n \left\lceil \log_{2} n \right\rceil - 1$$

$$\leq n \left\lceil \log_{2} n \right\rceil - 1 + n$$

$$= n \left\lceil \log_{2} n \right\rceil. \quad \bullet$$

 $n_2 = \lceil n/2 \rceil$ 



# 5. DIVIDE AND CONQUER

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## Counting inversions

### Music site tries to match your song preferences with others.

- You rank n songs.
- Music site consults database to find people with similar tastes.

Similarity metric: number of inversions between two rankings.

- My rank: 1, 2, ..., n.
- Your rank:  $a_1, a_2, ..., a_n$ .
- Songs *i* and *j* are inverted if i < j, but  $a_i > a_j$ .

	А	В	С	D	Е
me	1	2	3	4	5
you	1	3	4	2	5
2 inversions: 3-2, 4-2					

Brute force: check all  $\Theta(n^2)$  pairs.

## Counting inversions: applications

- Voting theory.
- Collaborative filtering.
- Measuring the "sortedness" of an array.
- Sensitivity analysis of Google's ranking function.
- Rank aggregation for meta-searching on the Web.
- Nonparametric statistics (e.g., Kendall's tau distance).

### Rank Aggregation Methods for the Web

Cynthia Dwork\* Ravi Kumar\* Moni Naor\* D. Sivakumar\*

#### ABSTRACT

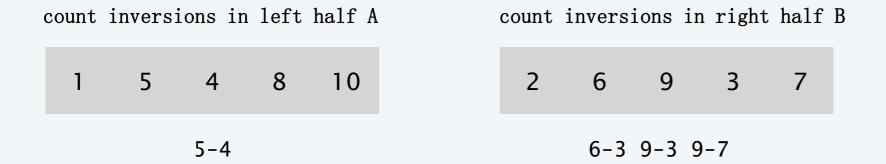
We consider the problem of combining ranking results from various sources. In the context of the Web, the main applications include building meta-search engines, combining ranking functions, selecting documents based on multiple criteria, and improving search precision through word associations. We develop a set of techniques for the rank aggregation problem and compare their performance to that of well-known methods. A primary goal of our work is to design rank aggregation techniques that can effectively combat "spam," a serious problem in Web searches. Experiments show that our methods are simple, efficient, and effective.

Keywords: rank aggregation, ranking functions, metasearch, multi-word queries, spam

## Counting inversions: divide-and-conquer

- Divide: separate list into two halves A and B.
- Conquer: recursively count inversions in each list.
- Combine: count inversions (a, b) with  $a \in A$  and  $b \in B$ .
- Return sum of three counts.





count inversions (a, b) with  $a \in A$  and  $b \in B$ 

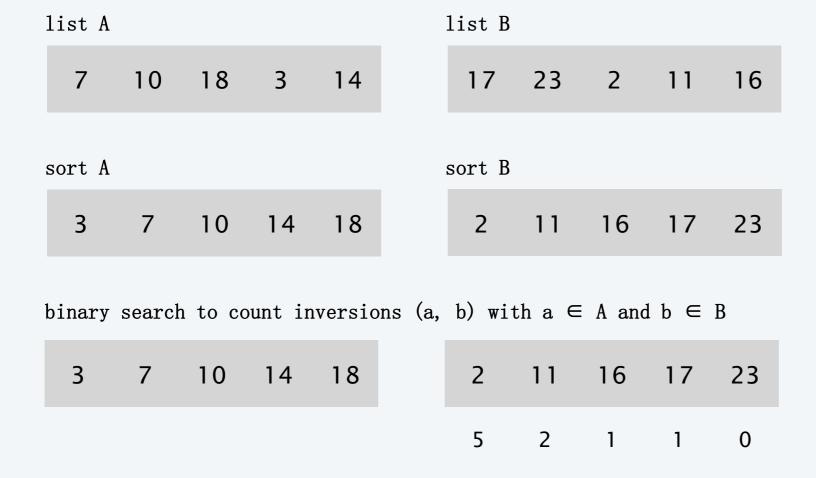
4-2 4-3 5-2 5-3 8-2 8-3 8-6 8-7 10-2 10-3 10-6 10-7 10-9

## Counting inversions: how to combine two subproblems?

- Q. How to count inversions (a, b) with  $a \in A$  and  $b \in B$ ?
- A. Easy if *A* and *B* are sorted!

## Warmup algorithm.

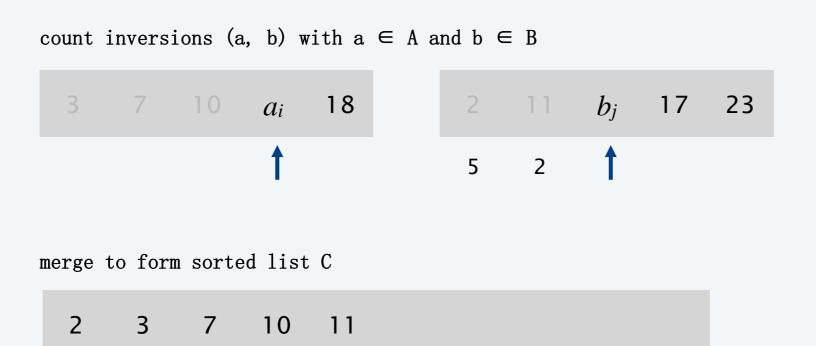
- Sort A and B.
- For each element  $b \in B$ ,
  - binary search in A to find how elements in A are greater than b.



## Counting inversions: how to combine two subproblems?

Count inversions (a, b) with  $a \in A$  and  $b \in B$ , assuming A and B are sorted.

- Scan A and B from left to right.
- Compare  $a_i$  and  $b_j$ .
- If  $a_i < b_j$ , then  $a_i$  is not inverted with any element left in B.
- If  $a_i > b_j$ , then  $b_j$  is inverted with every element left in A.
- Append smaller element to sorted list C.



Counting inversions: divide—and—conquer algorithm implementation

Input. List *L*.

Output. Number of inversions in L and sorted list of elements L'.

### SORT-AND-COUNT (*L*)

IF list L has one element

RETURN (0, L).

DIVIDE the list into two halves A and B.

$$(r_A, A) \leftarrow \text{SORT-AND-COUNT}(A)$$
.

$$(r_B, B) \leftarrow \text{SORT-AND-COUNT}(B)$$
.

$$(r_{AB}, L') \leftarrow \text{MERGE-AND-COUNT}(A, B).$$

RETURN 
$$(r_A + r_B + r_{AB}, L')$$
.

Proposition. The sort-and-count algorithm counts the number of inversions in a permutation of size n in  $O(n \log n)$  time.

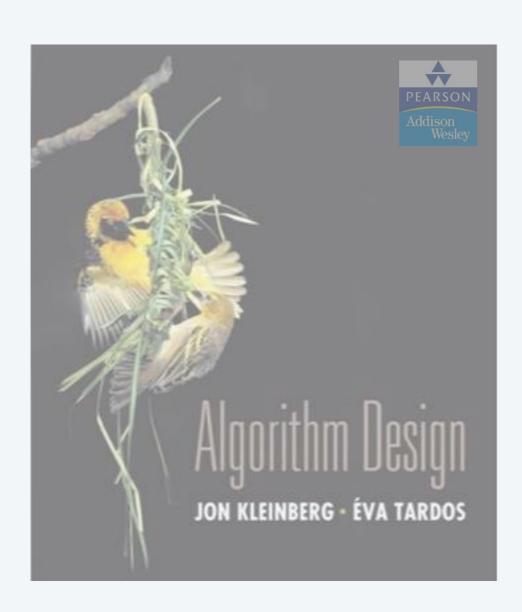
Pf. The worst-case running time T(n) satisfies the recurrence:

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ T(\lceil n/2 \rceil) + T(\lceil n/2 \rceil) + \Theta(n) & \text{otherwise} \end{cases}$$

Divide the list into two halves

A contains the first  $\lceil n/2 \rceil$  elements

B contains the remaining  $\lfloor n/2 \rfloor$  elements



# 5. DIVIDE AND CONQUER

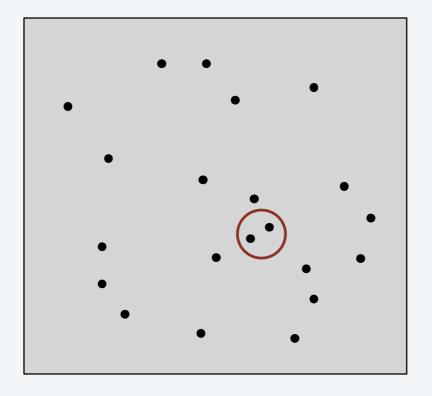
- mergesort
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- closest pair of points
- randomized quicksort
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Closest pair problem. Given n points in the plane, find a pair of points with the smallest Euclidean distance between them.

### Fundamental geometric primitive.

- Graphics, computer vision, geographic information systems, molecular modeling, air traffic control.
- Special case of nearest neighbor, Euclidean MST, Voronoi.

fast closest pair inspired fast algorithms for these problems

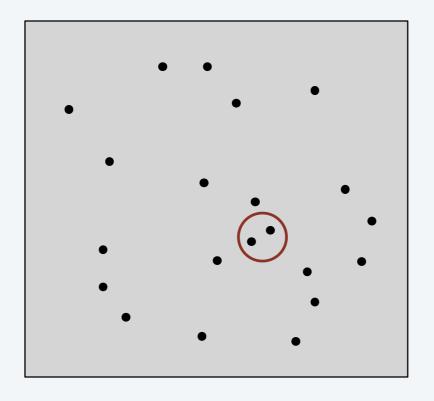


Closest pair problem. Given *n* points in the plane, find a pair of points with the smallest Euclidean distance between them.

Brute force. Check all pairs with  $\Theta(n^2)$  distance calculations.

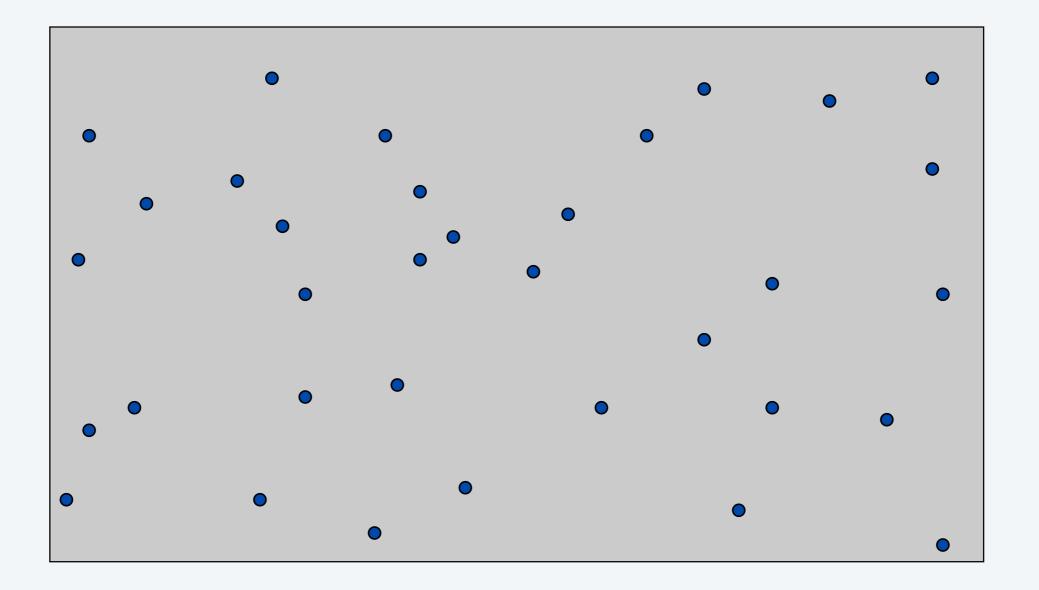
1d version. Easy  $O(n \log n)$  algorithm if points are on a line.

Nondegeneracy assumption. No two points have the same *x*-coordinate.



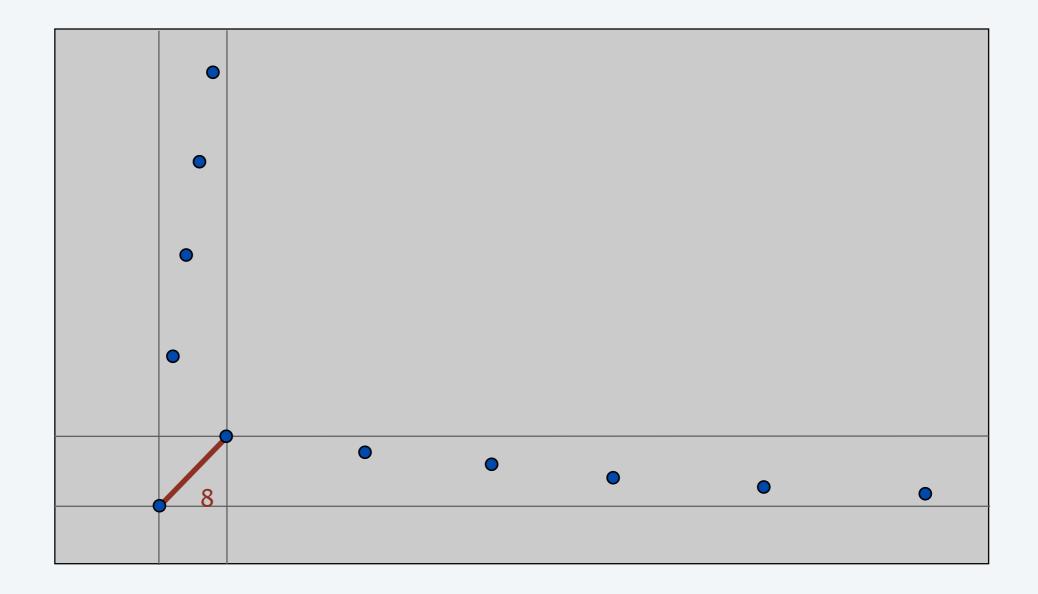
## Sorting solution.

- Sort by *x*-coordinate and consider nearby points.
- Sort by *y*-coordinate and consider nearby points.

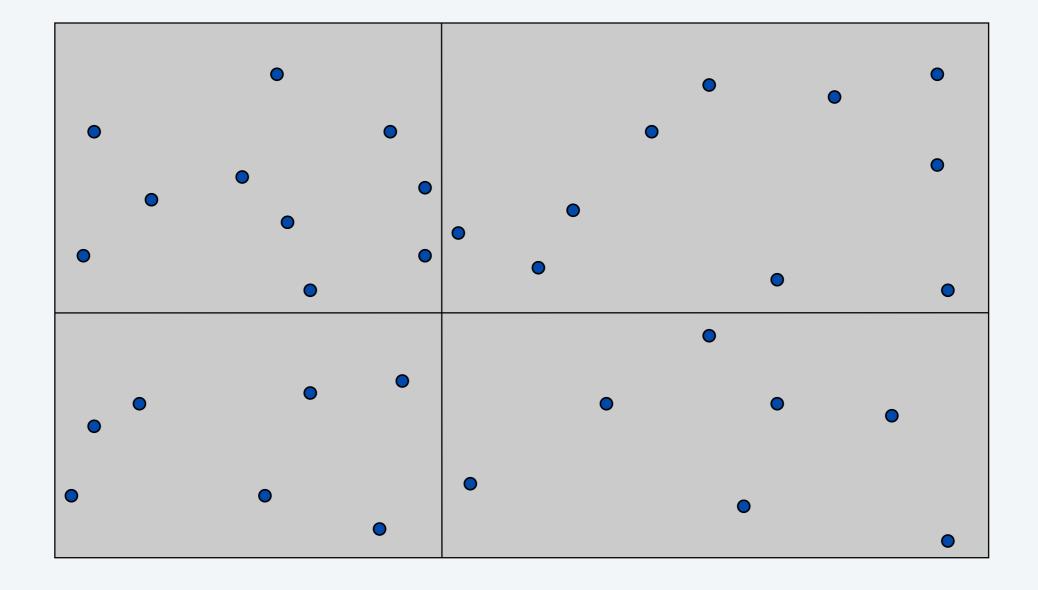


## Sorting solution.

- Sort by *x*-coordinate and consider nearby points.
- Sort by y-coordinate and consider nearby points.

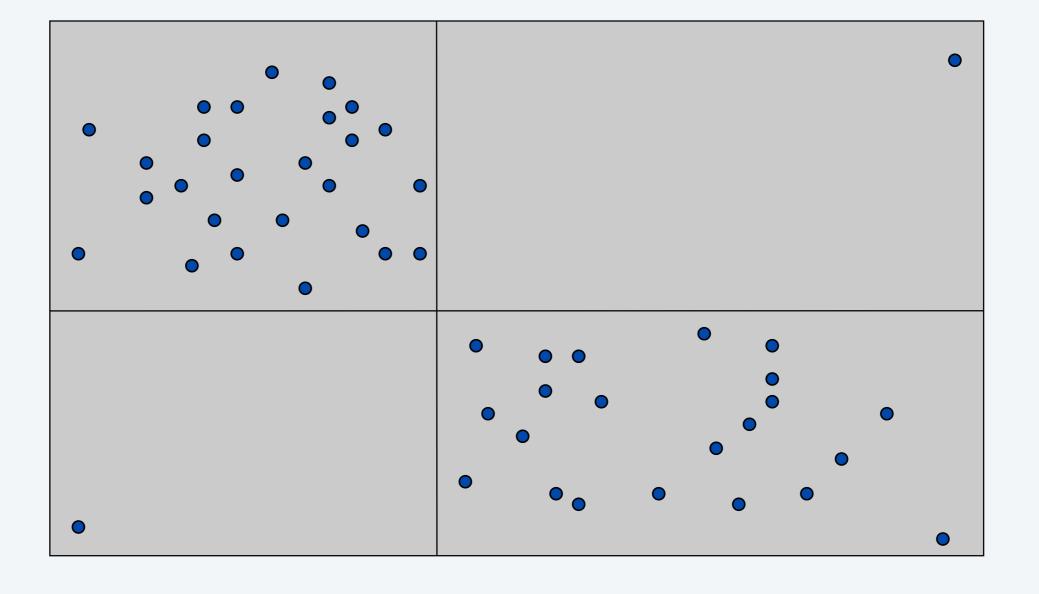


Divide. Subdivide region into 4 quadrants.



Divide. Subdivide region into 4 quadrants.

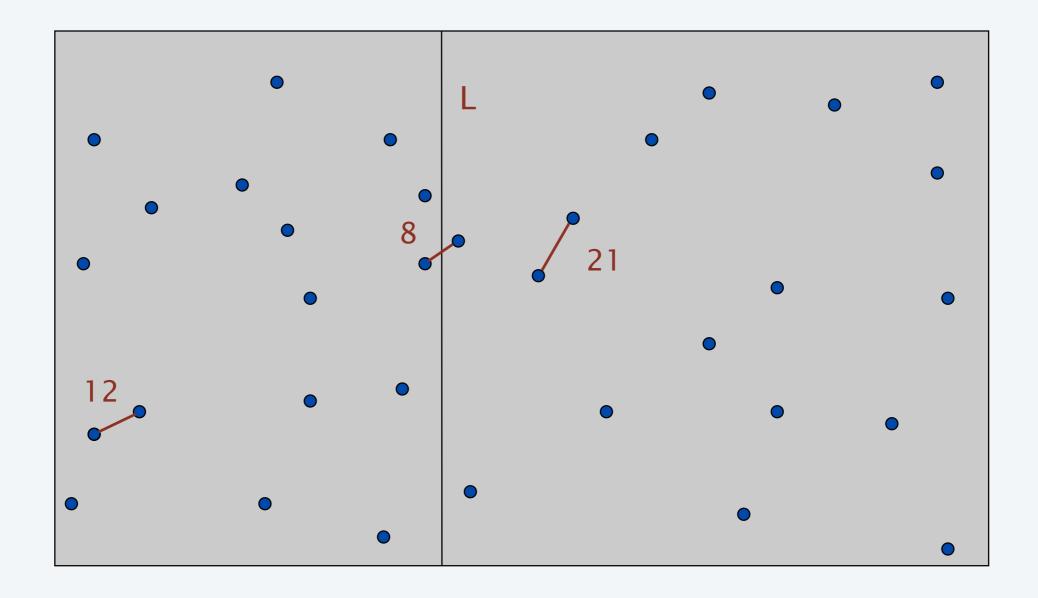
Obstacle. Impossible to ensure n/4 points in each piece.



## Closest pair of points: divide-and-conquer algorithm

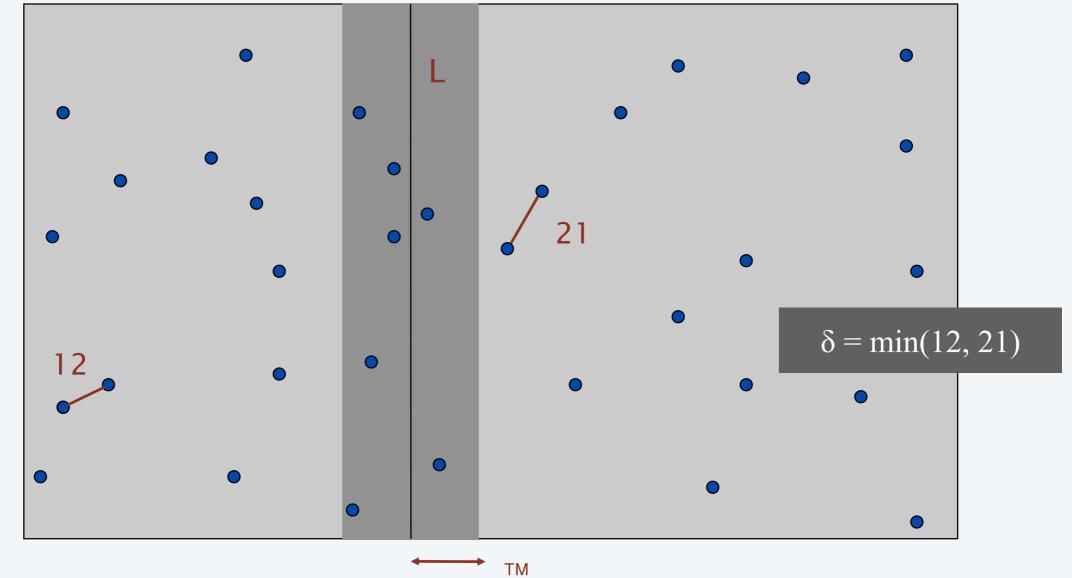
- Divide: draw vertical line L so that n/2 points on each side.
- Conquer: find closest pair in each side recursively.
- Combine: find closest pair with one point in each side.
- Return best of 3 solutions.

seems like  $\Theta(N^2)$ 



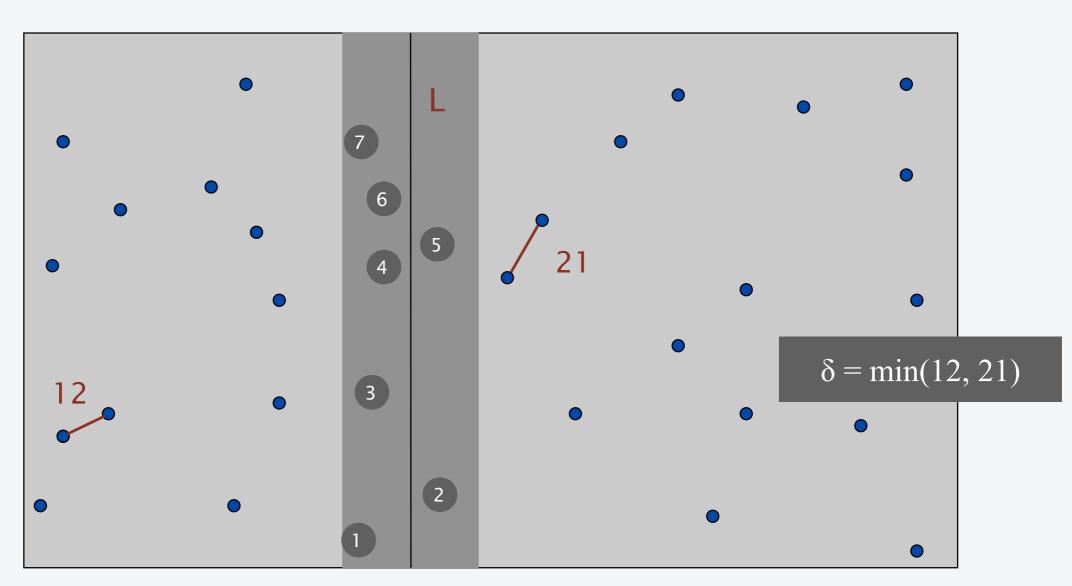
Find closest pair with one point in each side, assuming that distance  $< \delta$ .

• Observation: only need to consider points within  $\delta$  of line L.



Find closest pair with one point in each side, assuming that distance < ™.

- Observation: only need to consider points within  $^{\text{\tiny TM}}$  of line L.
- Sort points in  $2\delta$ -strip by their *y*-coordinate.
- Only check distances of those within 11 positions in sorted list!



TM

why 11?

Def. Let  $s_i$  be the point in the  $2\delta$ -strip, with the  $i^{th}$  smallest y-coordinate.

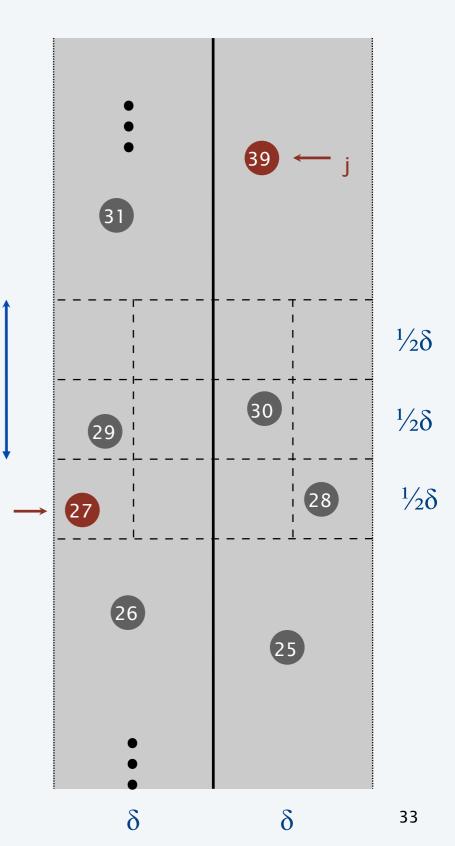
2 rows

Claim. If  $|i-j| \ge 12$ , then the distance between  $s_i$  and  $s_j$  is at least  $\delta$ .

Pf.

- No two points lie in same  $\frac{1}{2} \delta$ -by- $\frac{1}{2} \delta$  box.
- Two points at least 2 rows apart have distance  $\geq 2 (\frac{1}{2} \delta)$ .

Fact. Claim remains true if we replace 12 with 7.



### CLOSEST-PAIR $(p_1, p_2, ..., p_n)$

Compute separation line L such that half the points are on each side of the line.

 $\delta_1 \leftarrow \text{CLOSEST-PAIR}$  (points in left half).

 $\delta_2 \leftarrow \text{CLOSEST-PAIR}$  (points in right half).

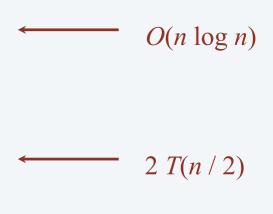
 $\delta \leftarrow \min \{ \delta_1, \delta_2 \}.$ 

Delete all points further than  $\delta$  from line L.

Sort remaining points by *y*-coordinate.

Scan points in y-order and compare distance between each point and next 11 neighbors. If any of these distances is less than  $\delta$ , update  $\delta$ .

Return  $\delta$ .



$$O(n)$$

$$O(n \log n)$$

$$\leftarrow$$
  $O(n)$ 

Closest pair of points: analysis

Theorem. The divide-and-conquer algorithm for finding the closest pair of points in the plane can be implemented in  $O(n \log^2 n)$  time.

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ T(\lceil n/2 \rceil) + T(\lceil n/2 \rceil) + O(n \log n) & \text{otherwise} \end{cases}$$

$$(x_1-x_2)^2+(y_1-y_2)^2$$

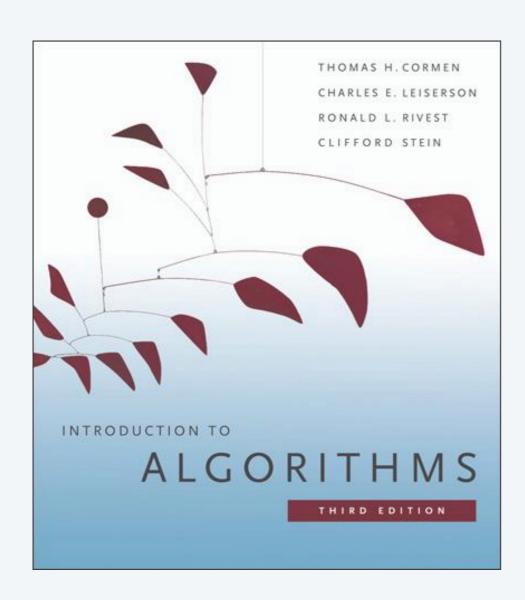
Lower bound. In quadratic decision tree model, any algorithm for closest pair (even in 1D) requires  $\Omega(n \log n)$  quadratic tests.

- Q. How to improve to  $O(n \log n)$ ?
- A. Yes. Don't sort points in strip from scratch each time.
  - Each recursive returns two lists: all points sorted by *x*-coordinate, and all points sorted by *y*-coordinate.
  - Sort by merging two pre-sorted lists.

Theorem. [Shamos 1975] The divide-and-conquer algorithm for finding the closest pair of points in the plane can be implemented in  $O(n \log n)$  time.

Pf. 
$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ T(\lceil n/2 \rceil) + T(\lceil n/2 \rceil) + \Theta(n) & \text{otherwise} \end{cases}$$

Note. See Section 13.7 for a randomized O(n) time algorithm.



CHAPTER 7

# 5. DIVIDE AND CONQUER

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## Randomized quicksort

### 3-way partition array so that:

- Pivot element p is in place.
- Smaller elements in left subarray L.
- Equal elements in middle subarray M.
- Larger elements in right subarray *R*.

#### the array A



the partitioned array A

Recur in both left and right subarrays.

### RANDOMIZED-QUICKSORT (A)

IF list *A* has zero or one element RETURN.

Pick pivot  $p \in A$  uniformly at random.

 $(L, M, R) \leftarrow \text{PARTITION-3-WAY}(A, a_i).$ 

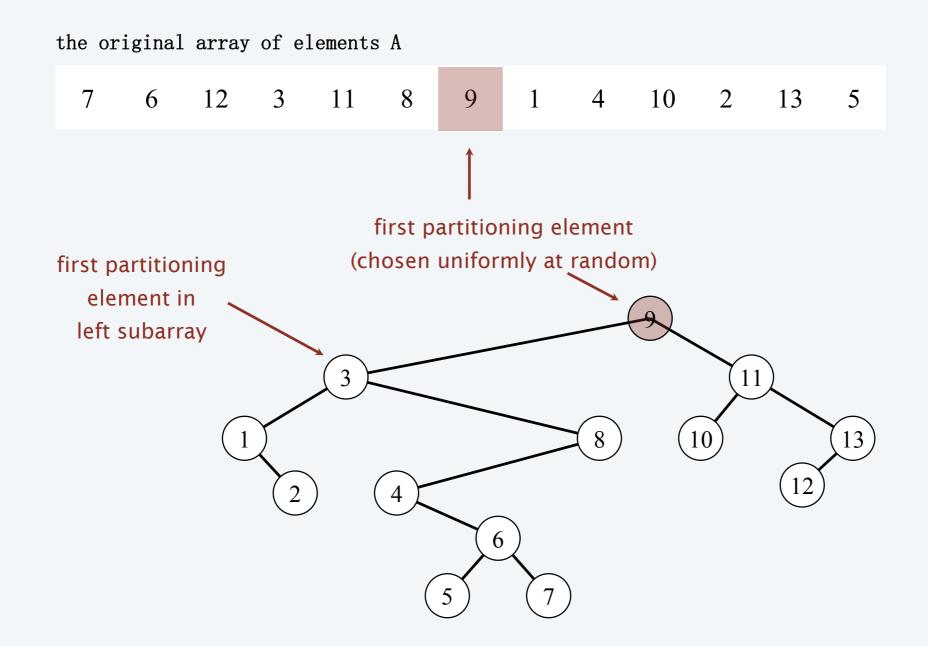
RANDOMIZED-QUICKSORT(L).

RANDOMIZED-QUICKSORT(R).

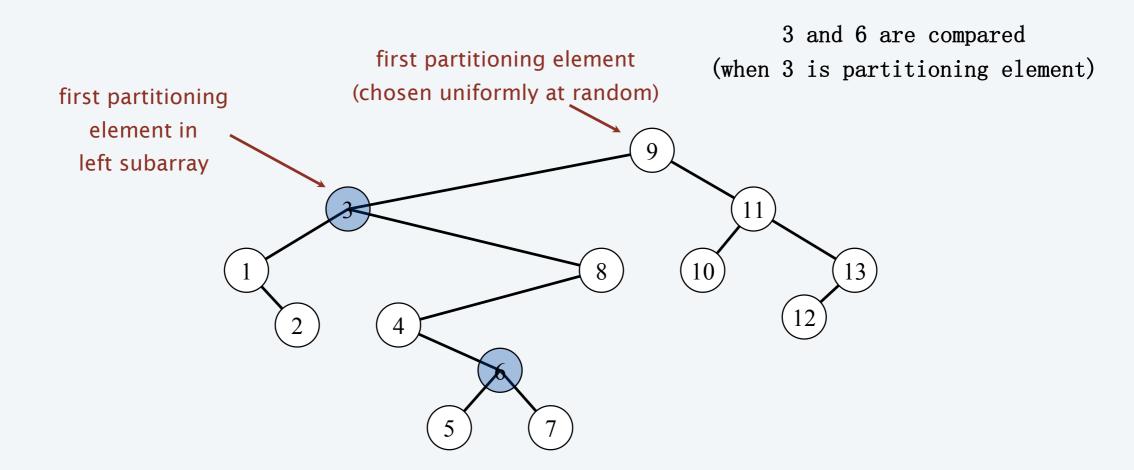
3-way partitioning can be done in-place (using n-1 compares)

Proposition. The expected number of compares to quicksort an array of n distinct elements is  $O(n \log n)$ .

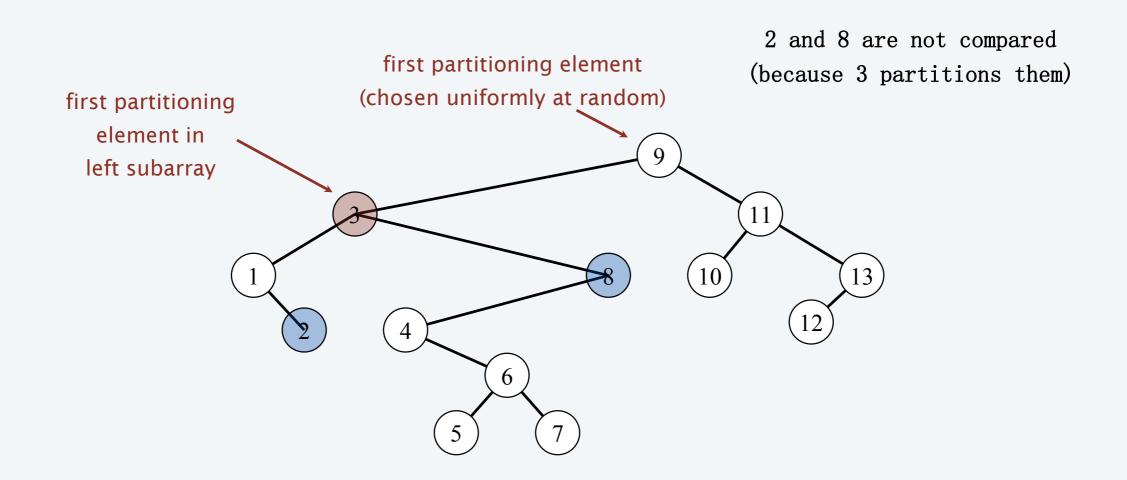
Pf. Consider BST representation of partitioning elements.



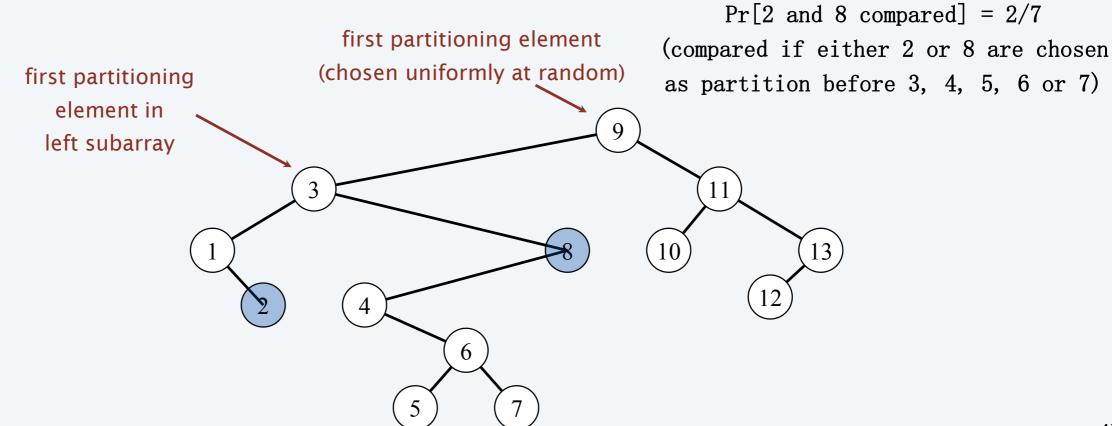
- Pf. Consider BST representation of partitioning elements.
  - An element is compared with only its ancestors and descendants.



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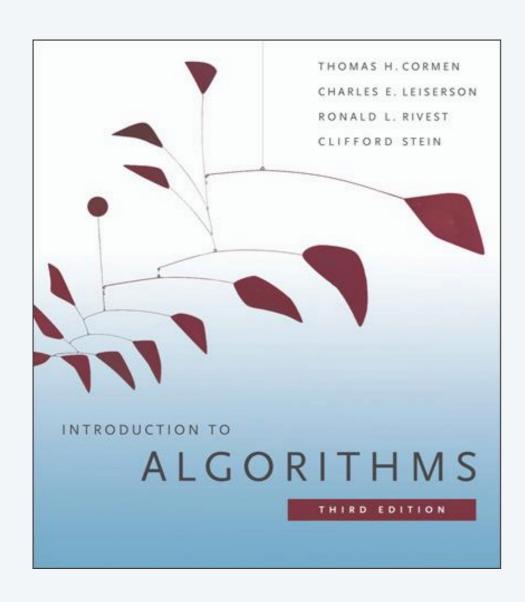
- Pf. Consider BST representation of partitioning elements.
  - An element is compared with only its ancestors and descendants.
  - Pr [  $a_i$  and  $a_j$  are compared ] = 2 / |j-i+1|.



- Pf. Consider BST representation of partitioning elements.
  - An element is compared with only its ancestors and descendants.
  - Pr [  $a_i$  and  $a_j$  are compared ] = 2 / |j-i+1|.

• Expected number of compares 
$$=\sum_{i=1}^N\sum_{j=i+1}^N\frac{2}{j-i+1}=2\sum_{i=1}^N\sum_{j=2}^{N-i+1}\frac{1}{j}$$
  $\leq 2N\sum_{j=1}^N\frac{1}{j}$   $\sim 2N\int_{x=1}^N\frac{1}{x}\,dx$   $=2N\ln N$ 

Remark. Number of compares only decreases if equal elements.



CHAPTER 9

# 5. DIVIDE AND CONQUER

- mergesort
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Selection. Given n elements from a totally ordered universe, find k<sup>th</sup> smallest.

- Minimum: k = 1; maximum: k = n.
- Median:  $k = \lfloor (n+1)/2 \rfloor$ .
- O(n) compares for min or max.
- $O(n \log n)$  compares by sorting.
- $O(n \log k)$  compares with a binary heap.

Applications. Order statistics; find the "top k"; bottleneck paths, ...

- Q. Can we do it with O(n) compares?
- A. Yes! Selection is easier than sorting.

#### 3-way partition array so that:

- Pivot element p is in place.
- Smaller elements in left subarray L.
- Equal elements in middle subarray M.
- Larger elements in right subarray R.

Recur in one subarray—the one containing the  $k^{th}$  smallest element.

```
QUICK-SELECT (A, k)

Pick pivot p \in A uniformly at random.

(L, M, R) \leftarrow \text{PARTITION-3-WAY}(A, p). \leftarrow \text{can be done in-place (using n-1 compares)}

If k \leq |L| RETURN QUICK-SELECT (L, k).

ELSE IF k > |L| + |M| RETURN QUICK-SELECT (R, k - |L| - |M|)

ELSE RETURN p.
```



Intuition. Split candy bar uniformly  $\Rightarrow$  expected size of larger piece is  $\frac{3}{4}$ .

$$T(n) \leq T(\sqrt[3]{n}) + n \Rightarrow T(n) \leq 4n$$

Def. T(n, k) = expected # compares to select k<sup>th</sup> smallest in an array of size  $\leq n$ .

Def.  $T(n) = \max_k T(n, k)$ .

Proposition.  $T(n) \leq 4n$ .

**Pf.** [by strong induction on *n*]

- Assume true for 1, 2, ..., n-1.
- T(n) satisfies the following recurrence:

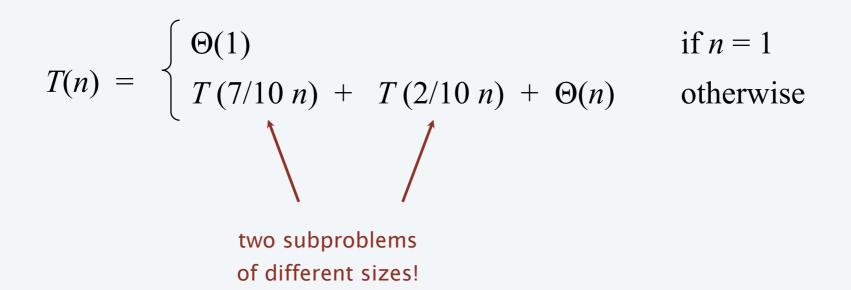
can assume we always recur on largest subarray since T(n) is monotonic and we are trying to get an upper bound

$$T(n) \le n + 2/n [T(n/2) + ... + T(n-3) + T(n-2) + T(n-1)]$$
  
 $\le n + 2/n [4n/2 + ... + 4(n-3) + 4(n-2) + 4(n-1)]$   
 $= n + 4(3/4 n)$   
 $= 4 n.$  • tiny cheat: sum should start at  $T(\lfloor n/2 \rfloor)$ 

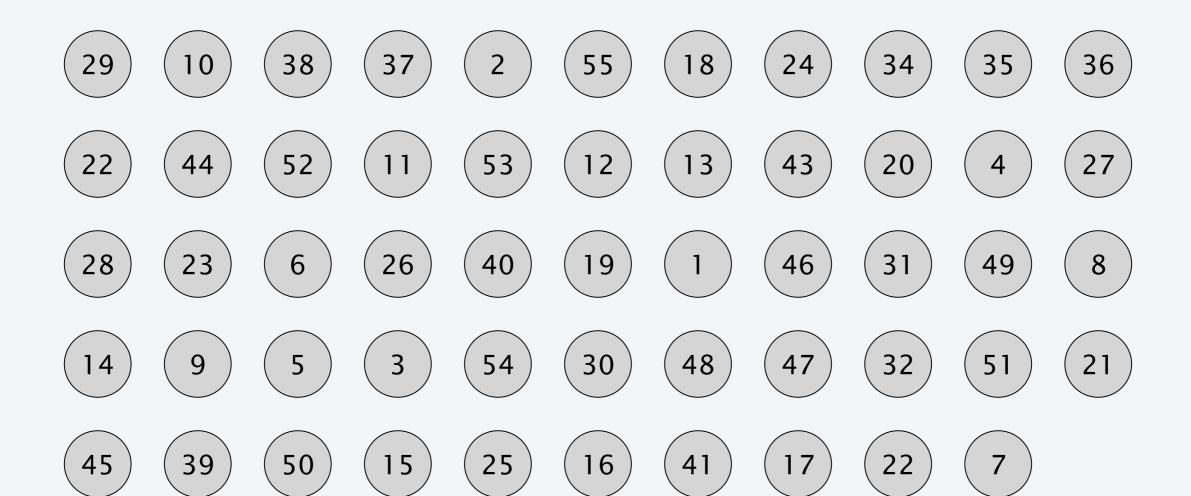
48

Goal. Find pivot element p that divides list of n elements into two pieces so that each piece is guaranteed to have  $\leq 7/10 n$  elements.

- Q. How to find approximate median in linear time?
- A. Recursively compute median of sample of  $\leq 2/10 n$  elements.



• Divide n elements into  $\lfloor n/5 \rfloor$  groups of 5 elements each (plus extra).

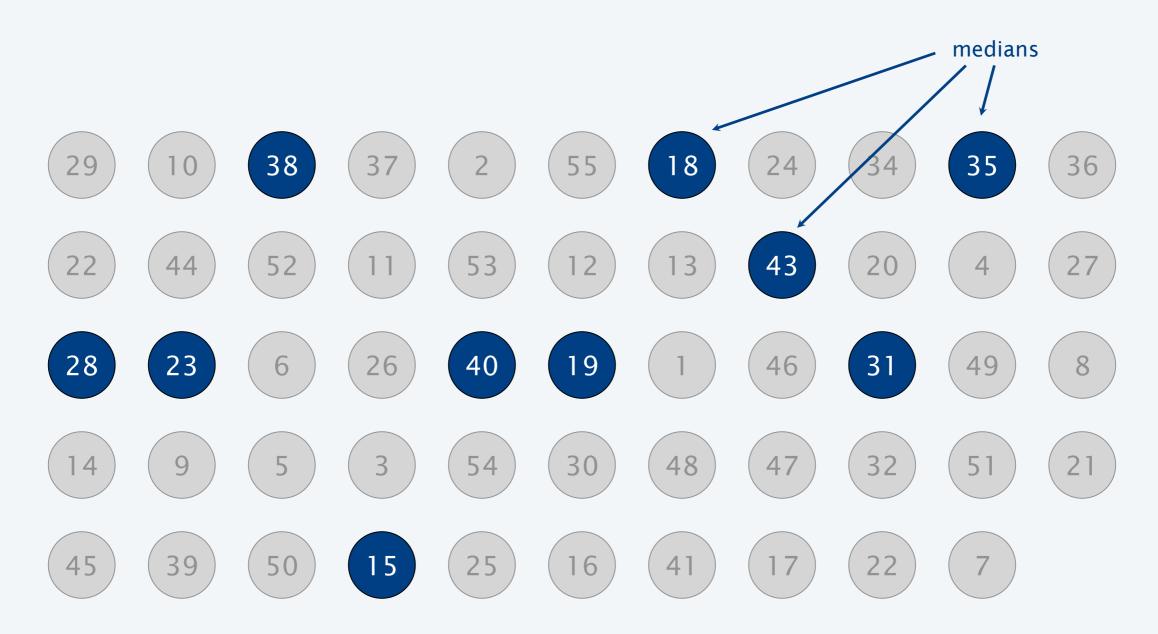


N = 54

50

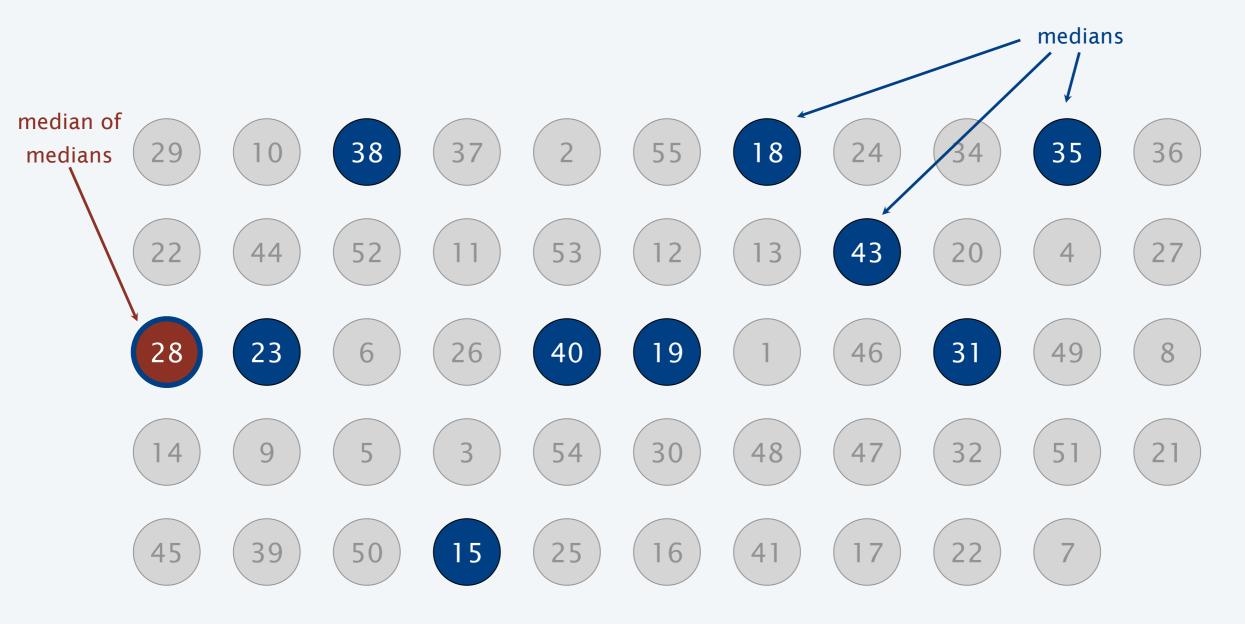
### Choosing the pivot element

- Divide n elements into  $\lfloor n/5 \rfloor$  groups of 5 elements each (plus extra).
- Find median of each group (except extra).



### Choosing the pivot element

- Divide n elements into  $\lfloor n/5 \rfloor$  groups of 5 elements each (plus extra).
- Find median of each group (except extra).
- Find median of  $\lfloor n/5 \rfloor$  medians recursively.
- Use median-of-medians as pivot element.



N = 54

52

#### Mom-Select (A, k)

$$n \leftarrow |A|$$
.

IF n < 50 RETURN  $k^{th}$  smallest of element of A via mergesort.

Group A into  $\ln / 5$  groups of 5 elements each (plus extra).

 $B \leftarrow$  median of each group of 5.

$$p \leftarrow \text{MOM-SELECT}(B, \lfloor n / 10 \rfloor) \leftarrow \text{median of medians}$$

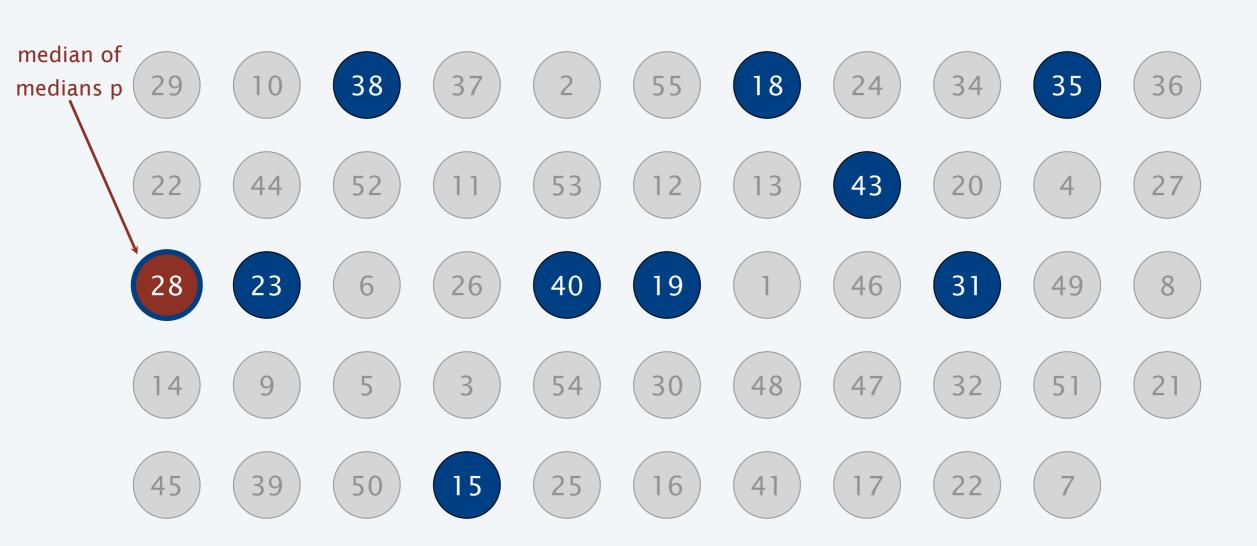
$$(L, M, R) \leftarrow \text{PARTITION-3-WAY}(A, p).$$

If 
$$k \le |L|$$
 RETURN MOM-SELECT  $(L, k)$ .

ELSE IF 
$$k > |L| + |M|$$
 RETURN MOM-SELECT  $(R, k - |L| - |M|)$ 

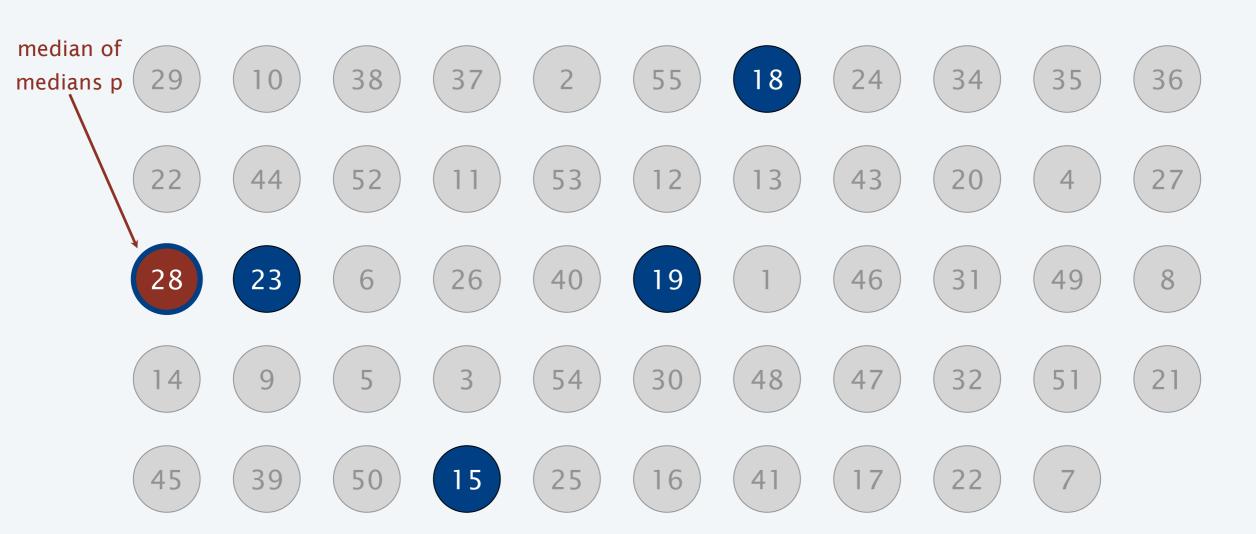
ELSE RETURN p.

• At least half of 5-element medians  $\leq p$ .



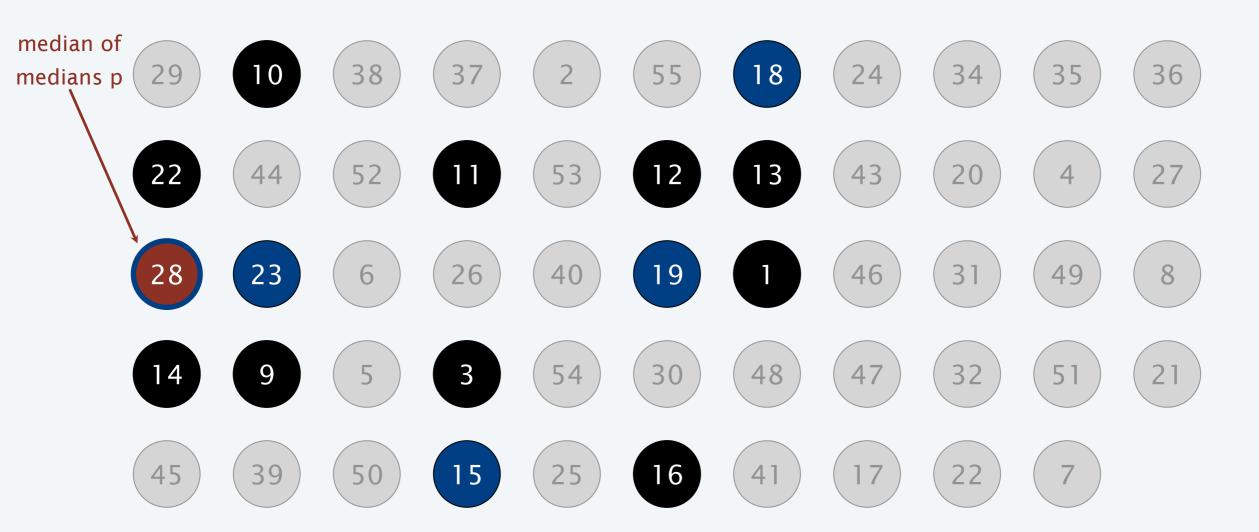
## Analysis of median-of-medians selection algorithm

- At least half of 5-element medians  $\leq p$ .
- At least  $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$  medians  $\leq p$ .



# Analysis of median-of-medians selection algorithm

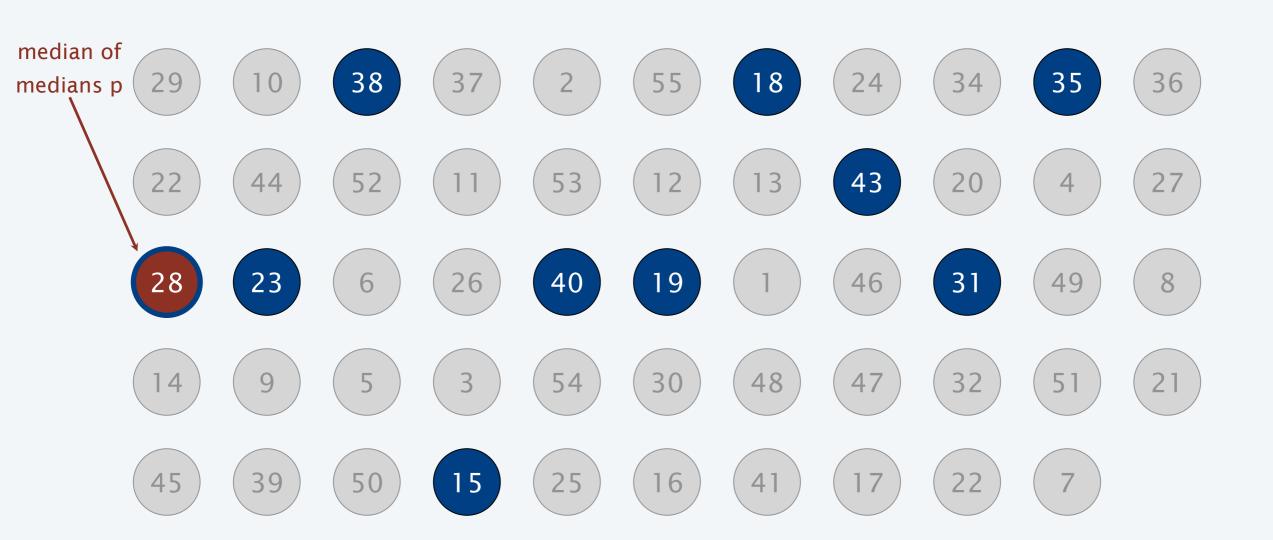
- At least half of 5-element medians  $\leq p$ .
- At least  $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$  medians  $\leq p$ .
- At least  $3 \lfloor n/10 \rfloor$  elements  $\leq p$ .



N = 54

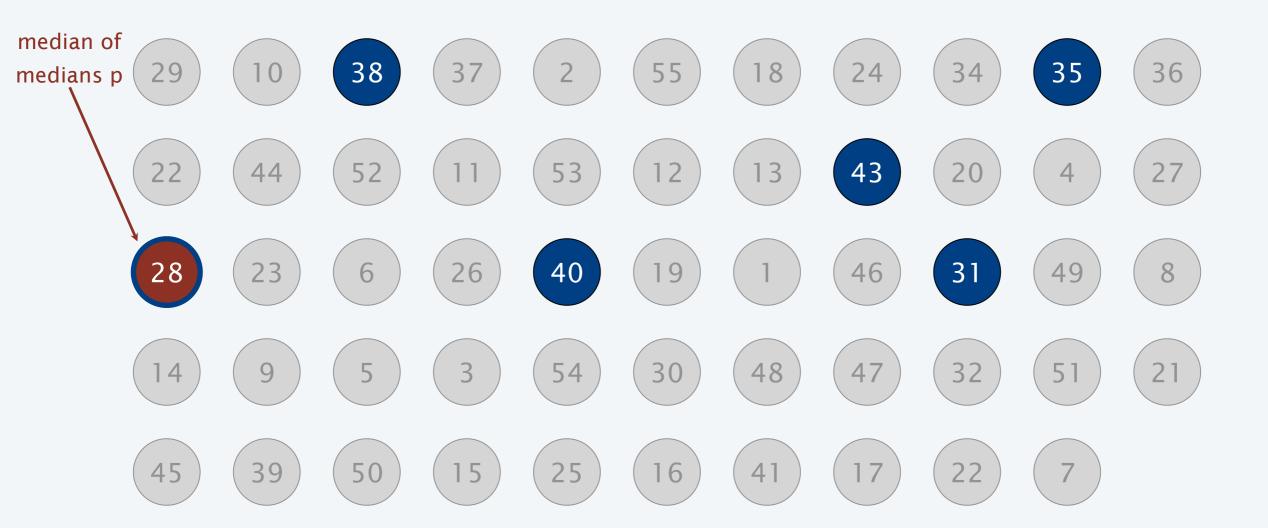
56

• At least half of 5-element medians  $\geq p$ .



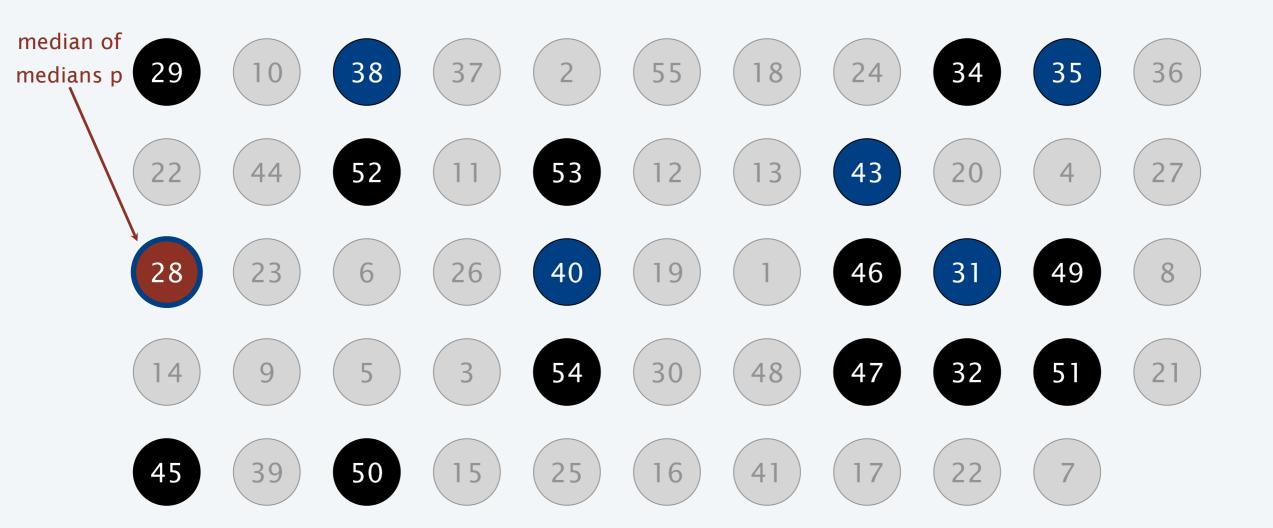
### Analysis of median-of-medians selection algorithm

- At least half of 5-element medians  $\geq p$ .
- Symmetrically, at least  $\lfloor n/10 \rfloor$  medians  $\geq p$ .



# Analysis of median-of-medians selection algorithm

- At least half of 5-element medians  $\geq p$ .
- Symmetrically, at least  $\lfloor n/10 \rfloor$  medians  $\geq p$ .
- At least  $3 \lfloor n/10 \rfloor$  elements  $\geq p$ .



### Median-of-medians selection algorithm recurrence.

- Select called recursively with  $\lfloor n/5 \rfloor$  elements to compute MOM p.
- At least  $3 \lfloor n/10 \rfloor$  elements  $\leq p$ .
- At least  $3 \lfloor n/10 \rfloor$  elements  $\geq p$ .
- Select called recursively with at most  $n 3 \lfloor n/10 \rfloor$  elements.

Def.  $C(n) = \max \# \text{ compares on an array of } n \text{ elements.}$ 

$$C(n) \le C(\lfloor n/5 \rfloor) + C(n-3 \lfloor n/10 \rfloor) + \frac{1}{5} n$$

median of recursive computing median of 5 medians select (6 compares per group)

partitioning (n compares)

### Now, solve recurrence.

- Assume n is both a power of 5 and a power of 10?
- Assume C(n) is monotone nondecreasing?

#### Analysis of selection algorithm recurrence.

- $T(n) = \max \# \text{ compares on an array of } \le n \text{ elements.}$
- T(n) is monotone, but C(n) is not!

$$T(n) \le \begin{cases} 6n & \text{if } n < 50 \\ T(\lfloor n/5 \rfloor) + T(n - 3\lfloor n/10 \rfloor) + \frac{11}{5}n & \text{otherwise} \end{cases}$$

#### Claim. $T(n) \leq 44 n$ .

- Base case:  $T(n) \le 6n$  for n < 50 (mergesort).
- Inductive hypothesis: assume true for 1, 2, ..., n-1.
- Induction step: for  $n \ge 50$ , we have:

$$T(n) \leq T(\lfloor n/5 \rfloor) + T(n-3 \lfloor n/10 \rfloor) + 11/5 n$$

$$\leq 44 (\lfloor n/5 \rfloor) + 44 (n-3 \lfloor n/10 \rfloor) + 11/5 n$$

$$\leq 44 (n/5) + 44 n - 44 (n/4) + 11/5 n \qquad \text{for } n \geq 50, \ 3 \lfloor n/10 \rfloor \geq n/4$$

$$= 44 n. \quad \blacksquare$$

Proposition. [Blum-Floyd-Pratt-Rivest-Tarjan 1973] There exists a compare-based selection algorithm whose worst-case running time is O(n).

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Time Bounds for Selection

by .

Manuel Blum, Robert W. Floyd, Vaughan Pratt,
Ronald L. Rivest, and Robert E. Tarjan

Abstract

The number of comparisons required to select the i-th smallest of n numbers is shown to be at most a linear function of n by analysis of a new selection algorithm -- PICK. Specifically, no more than

5.4505 n comparisons are ever required. This bound is improved for
```

#### Theory.

- Optimized version of BFPRT:  $\leq 5.4305 n$  compares.
- Best known upper bound [Dor-Zwick 1995]:  $\leq 2.95 n$  compares.
- Best known lower bound [Dor-Zwick 1999]:  $\geq (2 + \varepsilon) n$  compares.

Proposition. [Blum-Floyd-Pratt-Rivest-Tarjan 1973] There exists a compare-based selection algorithm whose worst-case running time is O(n).

```
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Abstract

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```

Practice. Constant and overhead (currently) too large to be useful.

Open. Practical selection algorithm whose worst-case running time is O(n).