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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 paradigm

Divide-and-conquer.

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

Most common usage.

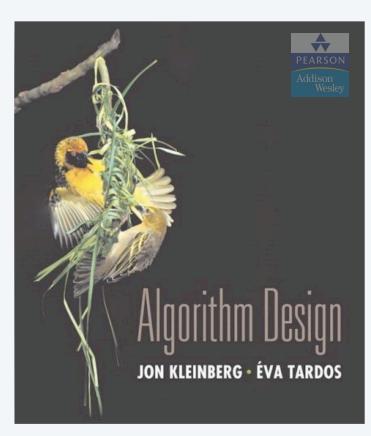
- Divide problem of size n into two subproblems of size n/2 in linear time.
- Solve two subproblems recursively.
- Combine two solutions into overall solution in linear time.

Consequence.

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



attributed to Julius Caesar



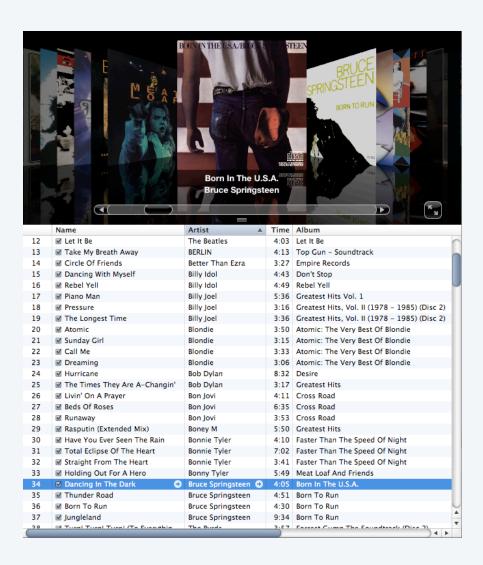
SECTION 5.1

5. DIVIDE AND CONQUER

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

Sorting problem

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.

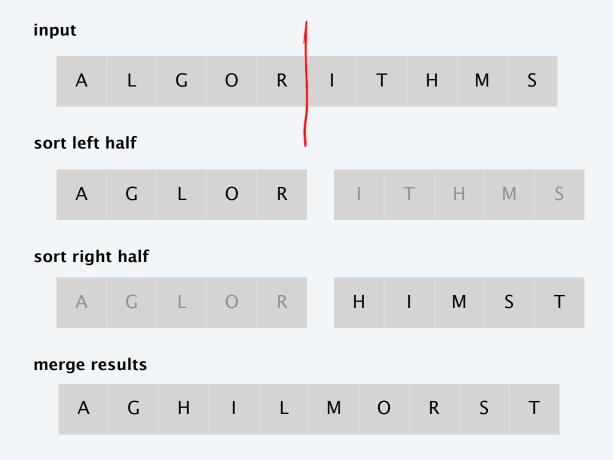
Non-obvious applications.

- 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.

• ...

Mergesort

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





Merging

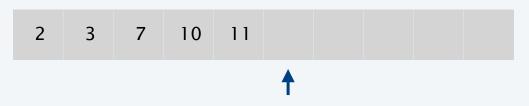
Goal. Combine two sorted lists *A* and *B* into a sorted whole *C*.

0

- Scan A and B from left to right.
- Compare a_i and b_j .
- If $a_i \le b_j$, 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).



merge to form sorted list C



A useful recurrence relation

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 =.

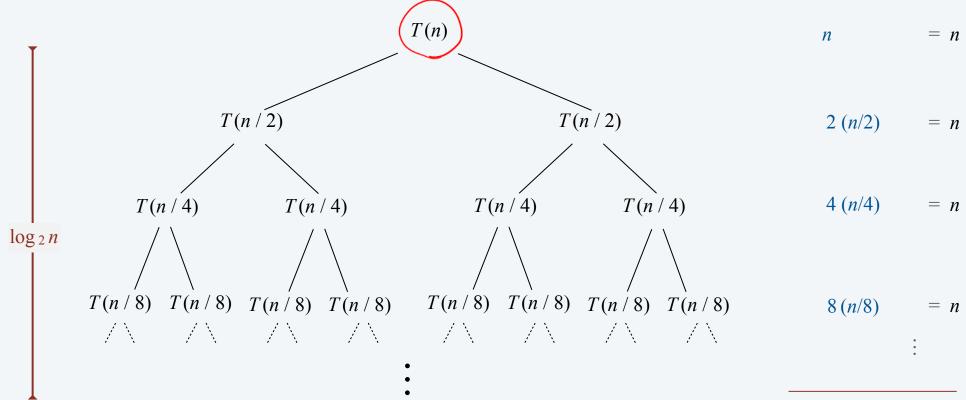
Divide-and-conquer recurrence: proof by recursion tree

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

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

assuming n is a power of 2

Pf 1.



Proof by induction

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

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

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 $T(2n) = 2n \log_2 (2n)$.

$$T(2n) = 2n \log_{2}(2n).$$

$$T(2n) = 2T(n) + 2n$$

$$= 2n \log_{2} n + 2n$$

$$= 2n (\log_{2}(2n) - 1) + 2n$$

$$= 2n \log_{2}(2n).$$

$$= 2 \log_{2}(2n) - \log_{2}(2n) - \log_{2}(2n)$$

$$= 2 \log_{2}(2n).$$

Analysis of mergesort recurrence

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

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

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.

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

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

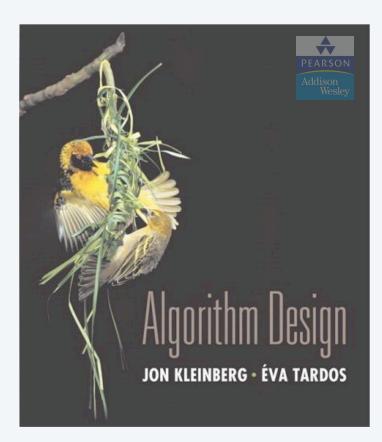
$$\leq n_{1} \lceil \log_{2} n_{2} \rceil + n_{2} \lceil \log_{2} n_{2} \rceil + n$$

$$= n \lceil \log_{2} n_{2} \rceil + n \qquad \qquad \log_{2} n_{2} \leq \lceil \log_{2} n \rceil - 1$$

$$\leq n (\lceil \log_{2} n \rceil - 1) + n$$

$$= n \lceil \log_{2} n \rceil. \quad \blacksquare$$

 $n_2 = \lceil n/2 \rceil$



SECTION 5.3

5. DIVIDE AND CONQUER

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

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§

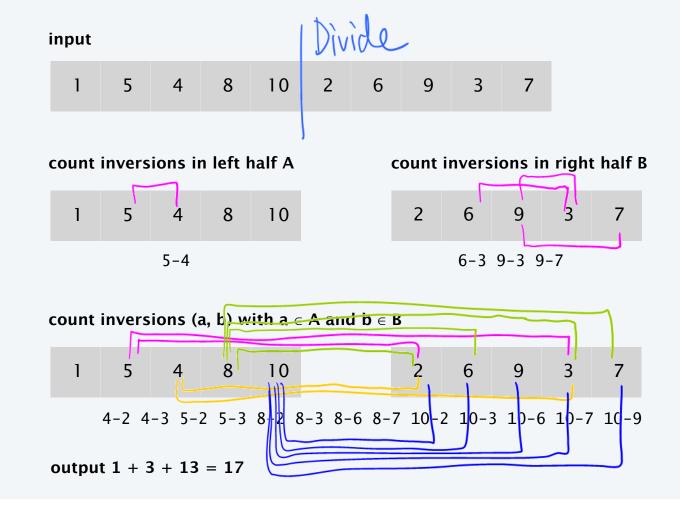
ABSTRACT

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.

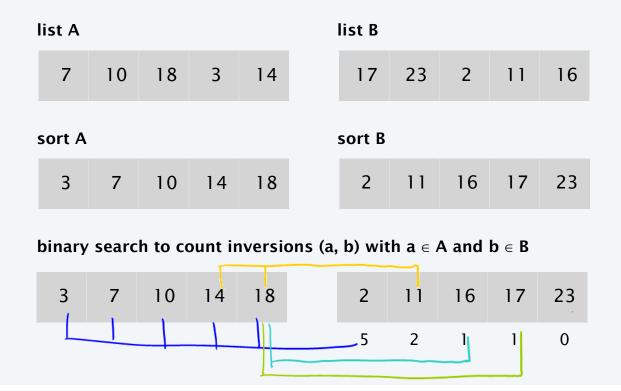


Counting inversions: how to combine two subproblems?

- 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?

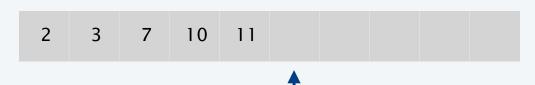
Count inversions (a, b) with $a \in A$ and $b \in B$, assuming A and B are sorted.

- 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 $a_i > b_i$, then b_i is inverted with every element left in A.
- Append smaller element to sorted list *C*.

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



merge to form 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'.

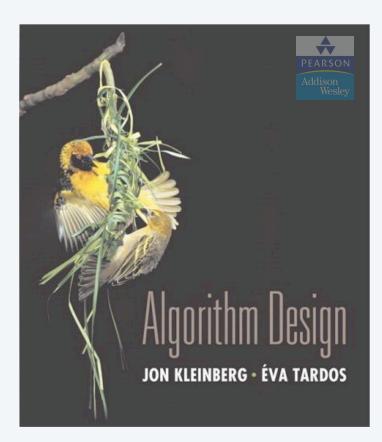
SORT-AND-COUNT (L)IF list L has one element RETURN (0, L). DIVIDE the list into two halves A and B. $(r_A, A) \leftarrow \text{SORT-AND-COUNT}(A)$. Recursine $(r_B, B) \leftarrow \text{SORT-AND-COUNT}(B)$. $(r_{AB}, L') \leftarrow \text{MERGE-AND-COUNT}(A, B).$ RETURN $(r_A + r_B + r_{AB}, L')$.

Counting inversions: divide-and-conquer algorithm analysis

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:

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



SECTION 5.4

5. DIVIDE AND CONQUER

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

Closest pair of points

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

N

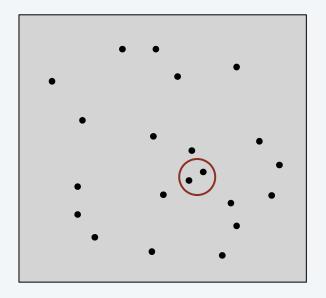
Closest pair of points

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.

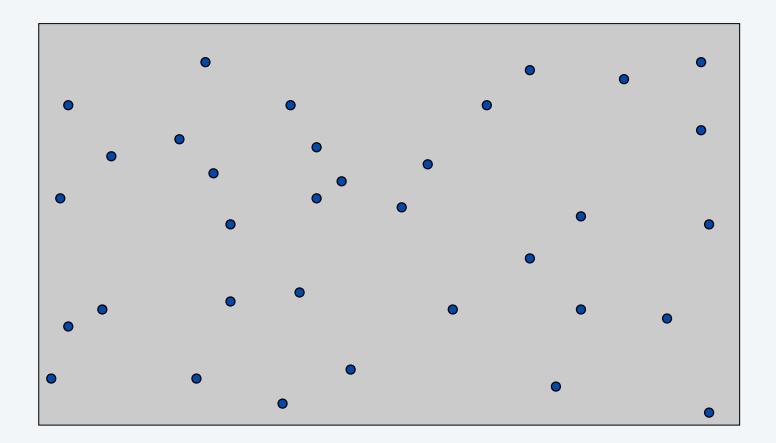
Nondegeneracy assumption. No two points have the same *x*-coordinate.



Closest pair of points: first attempt

Sorting solution.

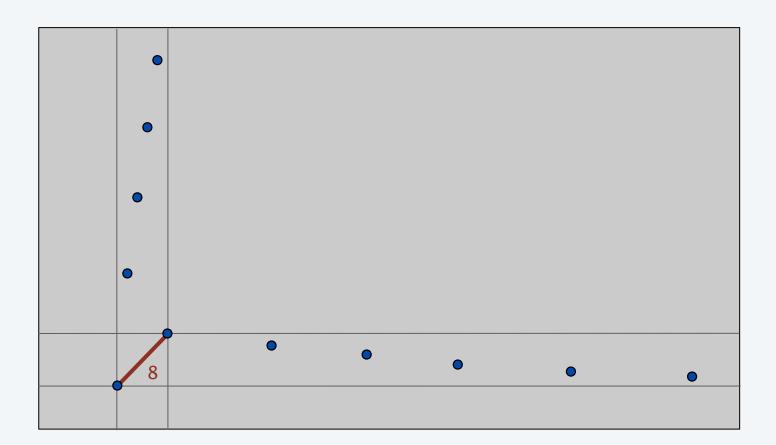
- Sort by *x*-coordinate and consider nearby points.
- Sort by *y*-coordinate and consider nearby points.



Closest pair of points: first attempt

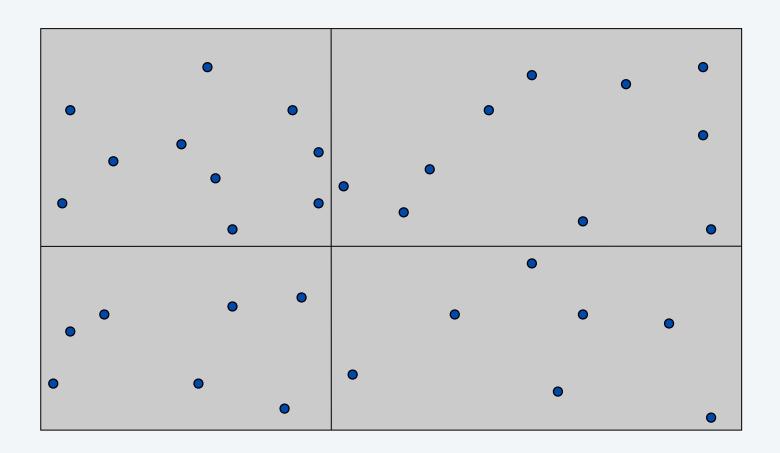
Sorting solution.

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



Closest pair of points: second attempt

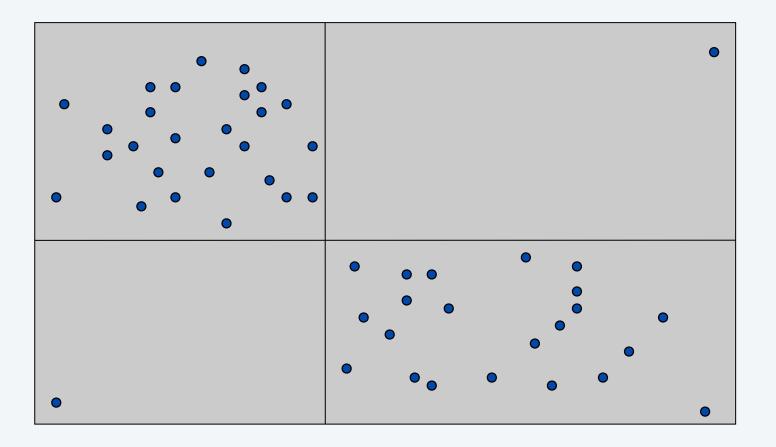
Divide. Subdivide region into 4 quadrants.



Closest pair of points: second attempt

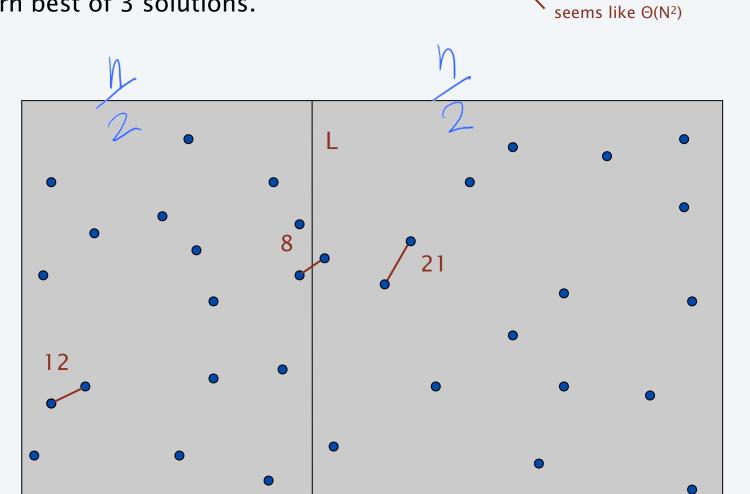
Divide. Subdivide region into 4 quadrants.

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



Closest pair of points: divide-and-conquer algorithm

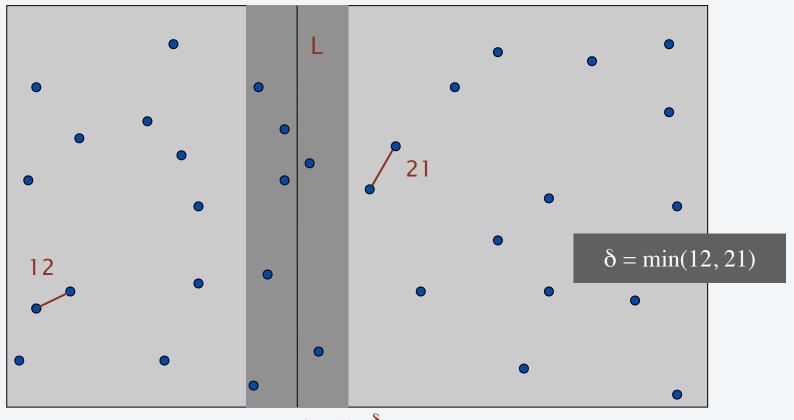
- Divide: draw vertical line L so that n/2 points on each side.
- Conquer: find closest pair in each side recursively.
- Combine: find closest pair with one point in each side.
- Return best of 3 solutions.



How to find closest pair with one point in each side?

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

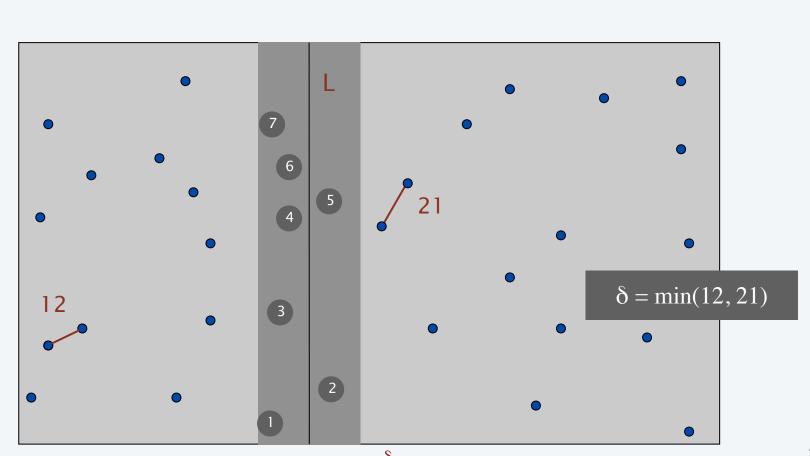
• Observation: only need to consider points within δ of line L.



How to find closest pair with one point in each side?

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

- Observation: only need to consider points within δ of line L.
- Sort points in 2δ -strip by their *y*-coordinate.
- Only check distances of those within 11 positions in sorted list!



How to find closest pair with one point in each side?

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

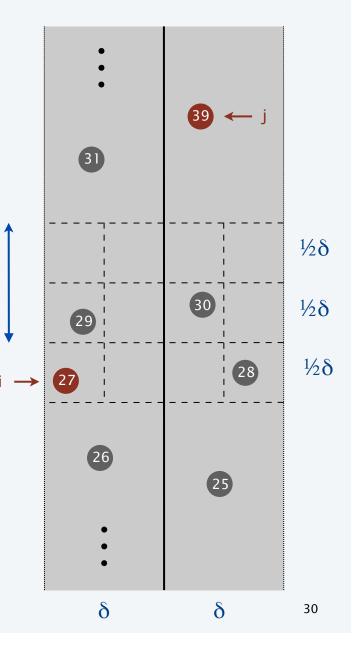
2 rows

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

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 of points: divide-and-conquer algorithm

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 δ 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 δ , update δ .

RETURN δ .



$$\longleftarrow 2 T(n/2)$$

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

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

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

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.

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

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

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

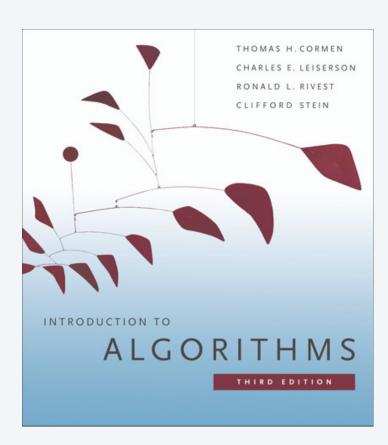
Improved closest pair algorithm

- 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

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

Randomized quicksort

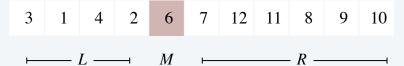
3-way partition array so that:

- 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.



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).

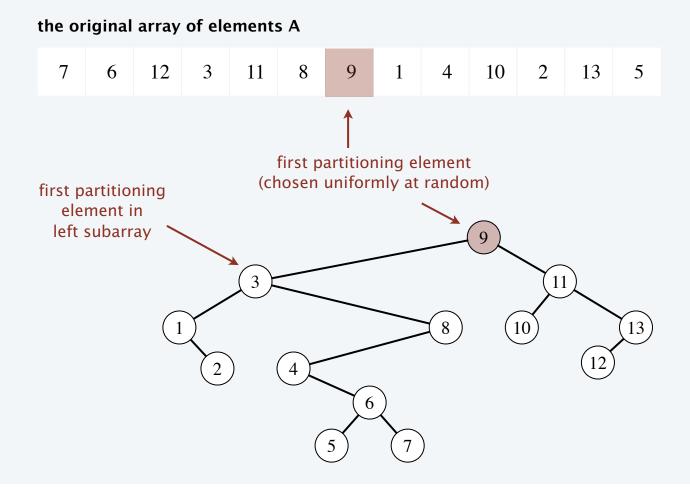
RANDOMIZED-QUICKSORT(R).

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

Analysis of randomized quicksort

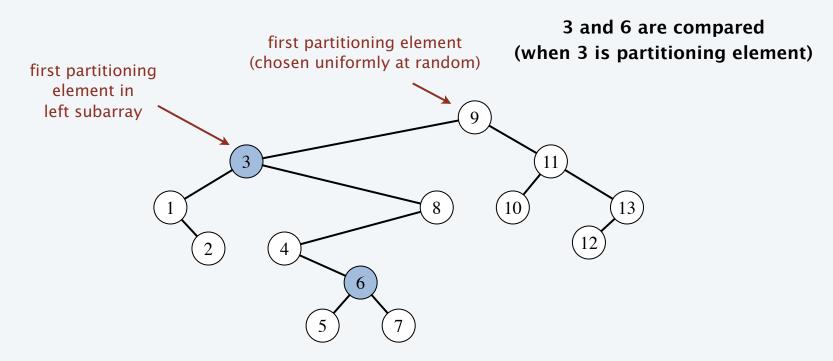
Proposition. The expected number of compares to quicksort an array of n distinct elements is $O(n \log n)$.

Pf. Consider BST representation of partitioning elements.



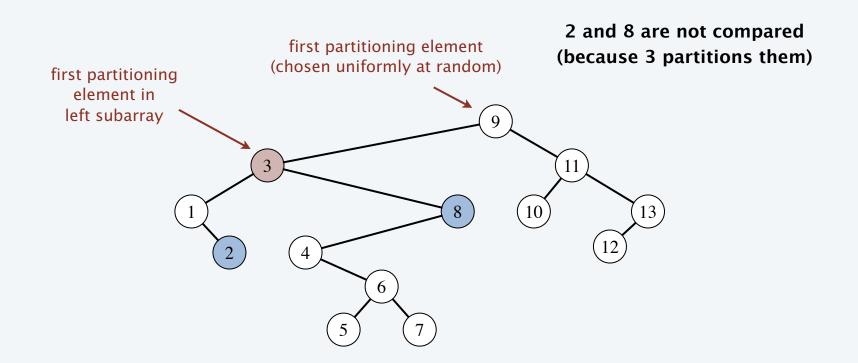
Proposition. The expected number of compares to quicksort an array of n distinct elements is $O(n \log n)$.

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



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

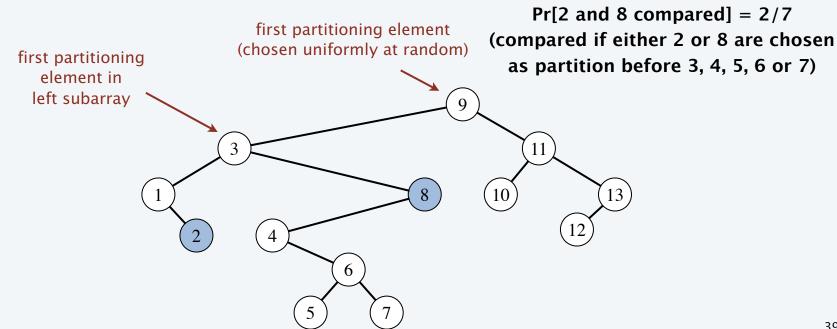
- Pf. Consider BST representation of partitioning elements.
 - An element is compared with only its ancestors and descendants.



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.

- An element is compared with only its ancestors and descendants.
- **Pr** [a_i and a_j are compared] = 2 / |j i + 1|.



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.

- An element is compared with only its ancestors and descendants.
- **Pr** [a_i and a_j are compared] = 2 / |j i + 1|.

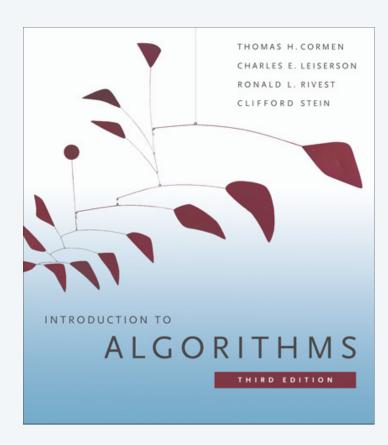
• Expected number of compares
$$= \sum_{i=1}^N \sum_{j=i+1}^N \frac{2}{j-i+1} = 2\sum_{i=1}^N \sum_{j=2}^{N-i+1} \frac{1}{j}$$

$$\leq 2N \sum_{j=1}^N \frac{1}{j}$$

$$\sim 2N \int_{x=1}^N \frac{1}{x} \, dx$$

$$= 2N \ln N$$

Remark. Number of compares only decreases if equal elements.



CHAPTER 9

5. DIVIDE AND CONQUER

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

Median and selection problems

Selection. Given n elements from a totally ordered universe, find kth 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.

Quickselect

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).

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

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

ELSE RETURN p.
```



Quickselect analysis

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

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

Def. $T(n, k) = \text{expected} \# \text{compares to select } k^{\text{th}} \text{ smallest in an array of size } \leq n.$ 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

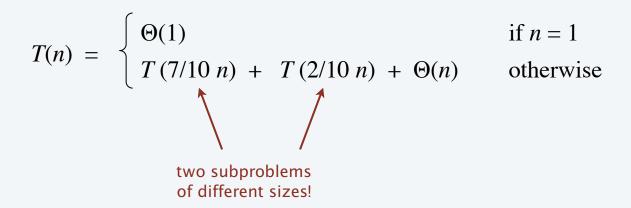
$$T(n) \le n + 2 / n [T(n/2) + ... + T(n-3) + T(n-2) + T(n-1)]$$

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

Selection in worst case linear time

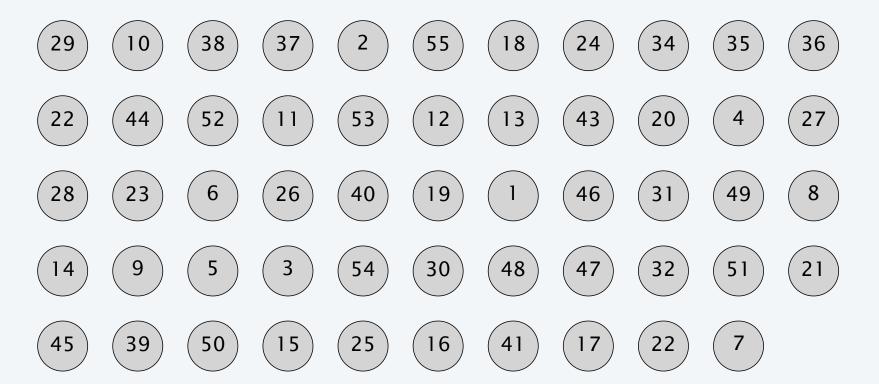
Goal. Find pivot element p that divides list of n elements into two pieces so that each piece is guaranteed to have $\leq 7/10 n$ elements.

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



Choosing the pivot element

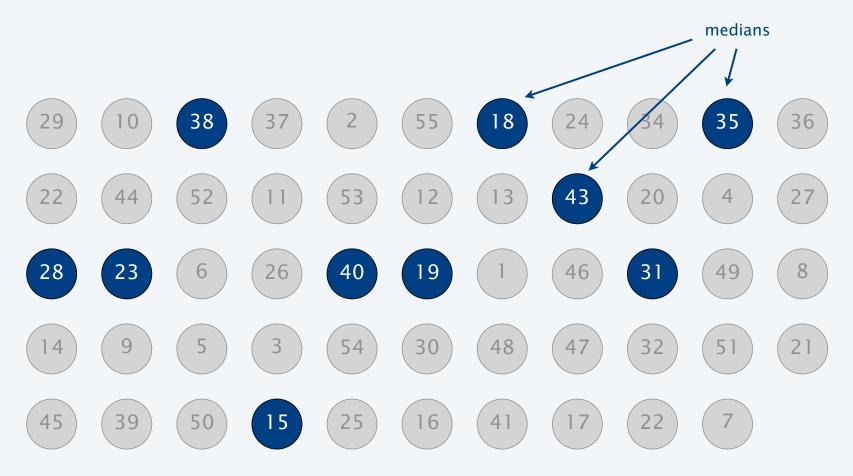
• Divide n elements into $\lfloor n/5 \rfloor$ groups of 5 elements each (plus extra).



N = 54

Choosing the pivot element

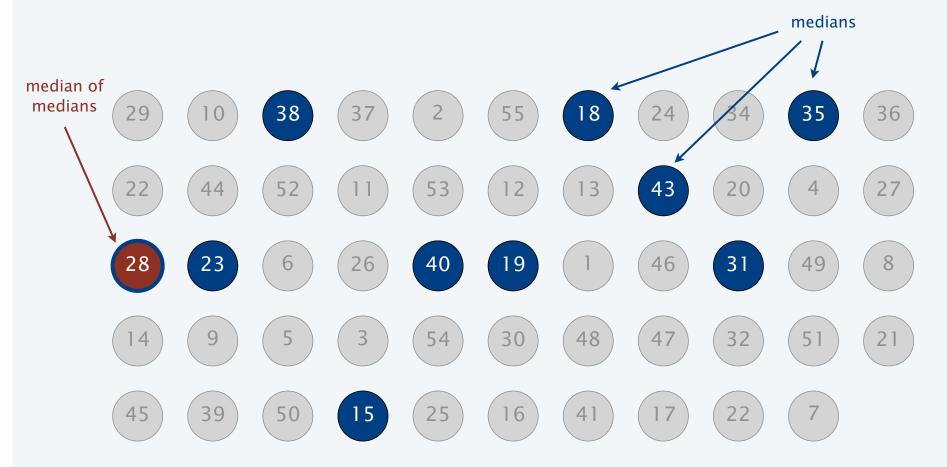
- Divide n elements into $\lfloor n/5 \rfloor$ groups of 5 elements each (plus extra).
- Find median of each group (except extra).



N = 54

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.

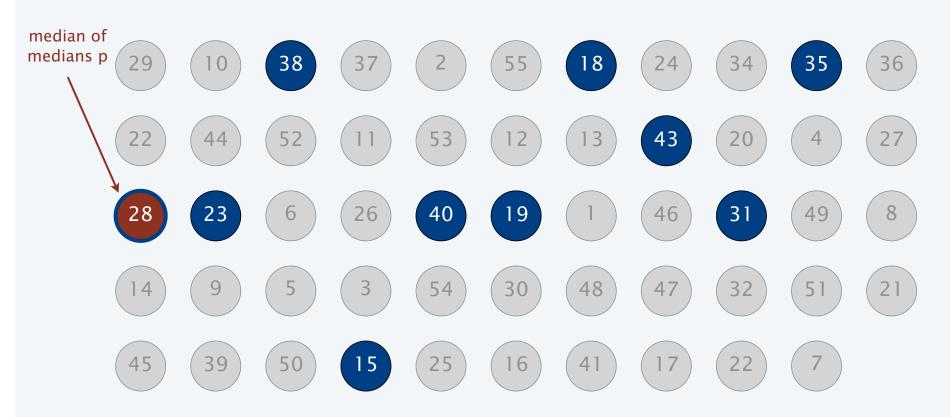


N = 54

Median-of-medians selection algorithm

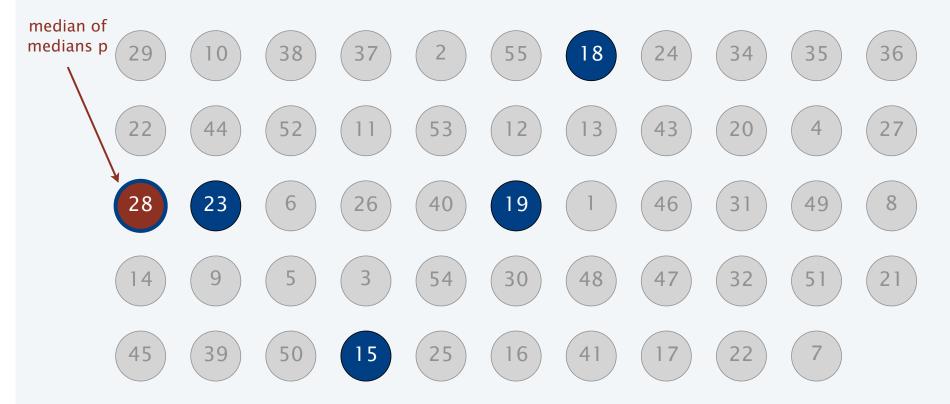
```
Mom-SELECT (A, k)
n \leftarrow |A|.
IF n < 50 RETURN k^{th} smallest of element of A via mergesort.
Group A into \lfloor n/5 \rfloor groups of 5 elements each (plus extra).
B \leftarrow median of each group of 5.
p \leftarrow \text{MOM-SELECT}(B, \lfloor n / 10 \rfloor) \leftarrow \text{median of 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 $\leq p$.



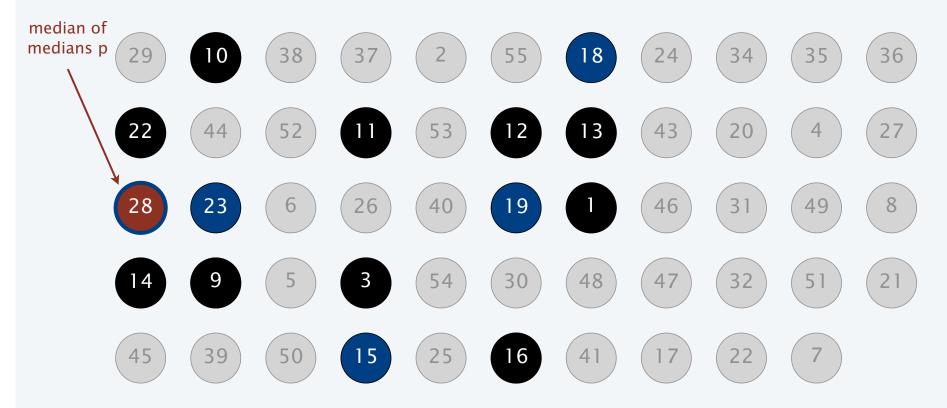
N = 54

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



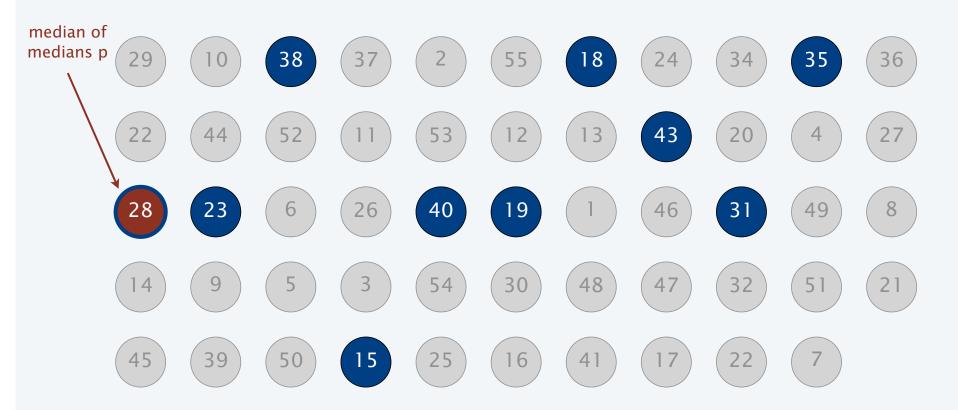
N = 54

- 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 $\leq p$.



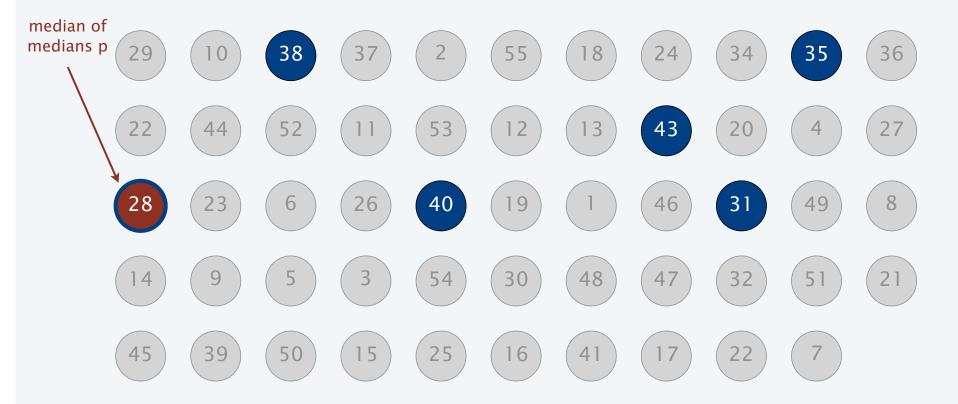
N = 54

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



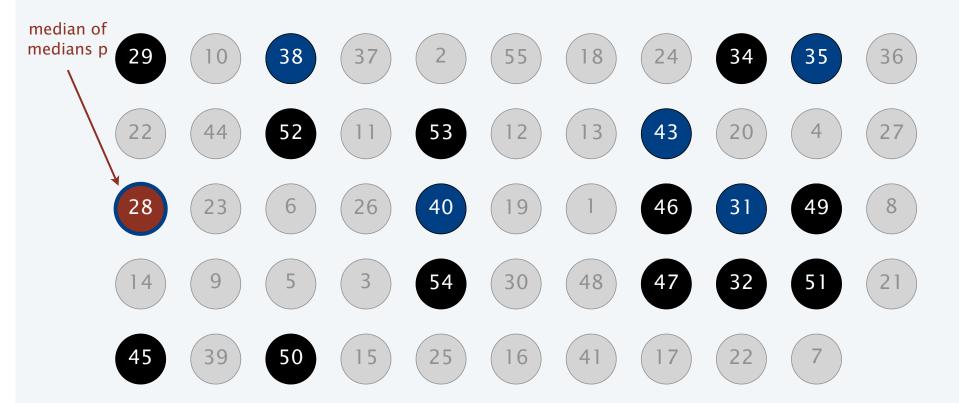
N = 54

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



N = 54

- 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 $\geq p$.



N = 54

Median-of-medians selection algorithm recurrence

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

$$C(n) \le C(\lfloor n/5 \rfloor) + C(n-3\lfloor n/10 \rfloor) + \frac{11}{5}n$$

median of recursive computing median of 5 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?

Median-of-medians selection algorithm recurrence

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:

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

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

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

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

Linear-time selection postmortem

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

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 5.4305 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]: $\geq (2 + \epsilon) n$ compares.

Linear-time selection postmortem

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

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 5.4305 n comparisons are ever required. This bound is improved for

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

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