

Recurrences, Closest Pair and Selection

Lecture 6

September 16, 2014

Part I

Recurrences

Solving Recurrences

Two general methods:

- 1 Recursion tree method: need to do sums
 - 1 elementary methods, geometric series
 - 2 integration
- 2 Guess and Verify
 - 1 guessing involves intuition, experience and trial & error
 - 2 verification is via induction

Recurrence: Example I

- 1 Consider $T(n) = 2T(n/2) + n/\log n$. i th level has 2^i nodes, and problem size at each node is $n/2^i$ and hence work at each node is $\frac{n}{2^i} / \log \frac{n}{2^i}$.

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- 3 Summing over all levels

$$\begin{aligned} T(n) &= \sum_{i=0}^{\log n - 1} 2^i \left[\frac{(n/2^i)}{\log(n/2^i)} \right] \\ &= \sum_{i=0}^{\log n - 1} \frac{n}{\log n - i} \\ &= n \sum_{j=1}^{\log n} \frac{1}{j} = nH_{\log n} = \Theta(n \log \log n) \end{aligned}$$

Recurrence: Example II

- 1 Consider...
- 2 What is the depth of recursion?
 $\sqrt{n}, \sqrt{\sqrt{n}}, \sqrt{\sqrt{\sqrt{n}}}, \dots, \mathbf{O(1)}$.
- 3 Number of levels: $n^{2^{-L}} = 2$ means $\mathbf{L = \log \log n}$.
- 4 Number of children at each level is $\mathbf{1}$, work at each node is $\mathbf{1}$
- 5 Thus, $\mathbf{T(n) = \sum_{i=0}^L 1 = \Theta(L) = \Theta(\log \log n)}$.

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Recurrence: Example IV

- 1 Consider $T(n) = T(n/4) + T(3n/4) + n$.
- 2 Using recursion tree, we observe the tree has leaves at different levels (a *lop-sided* tree).
- 3 Total work in any level is at most n . Total work in any level without leaves is exactly n .
- 4 Highest leaf is at level $\log_4 n$ and lowest leaf is at level $\log_{4/3} n$
- 5 Thus, $n \log_4 n \leq T(n) \leq n \log_{4/3} n$, which means $T(n) = \Theta(n \log n)$

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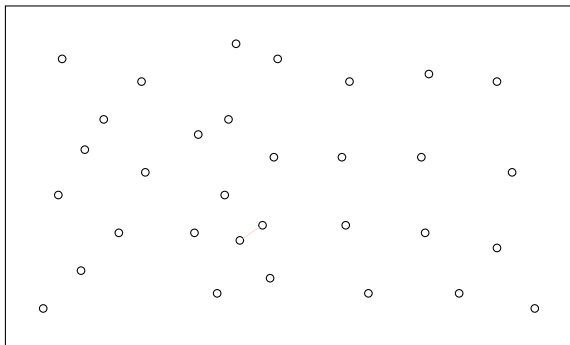
Part II

Closest Pair

Closest Pair - the problem

Input Given a set S of n points on the plane

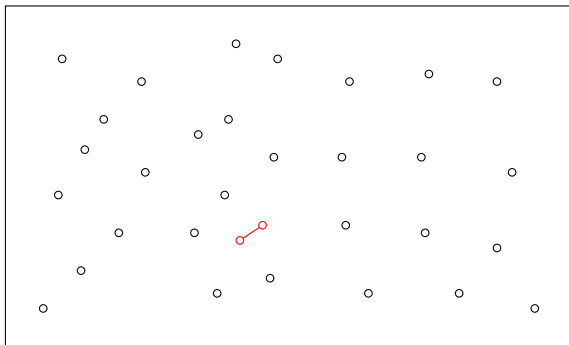
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Applications

- ① Basic primitive used in graphics, vision, molecular modelling
- ② Ideas used in solving nearest neighbor, Voronoi diagrams, Euclidean MST

Algorithm: Brute Force

- 1 Compute distance between every pair of points and find minimum.
- 2 Takes $O(n^2)$ time.
- 3 Can we do better?

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Closest Pair: 1-d case

Input Given a set \mathbf{S} of \mathbf{n} points on a line

Goal Find $\mathbf{p, q} \in \mathbf{S}$ such that $\mathbf{d(p, q)}$ is minimum

Algorithm

- 1 Sort points based on coordinate
- 2 Compute the distance between successive points, keeping track of the closest pair.

Running time $\mathbf{O(n \log n)}$

Can we do this in better running time?

Can reduce Distinct Elements Problem (see lecture 1) to this problem in $\mathbf{O(n)}$ time. Do you see how?

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Generalizing 1-d case

Can we generalize **1**-d algorithm to **2**-d?

Sort according to **x** or **y**-coordinate??

No easy generalization.

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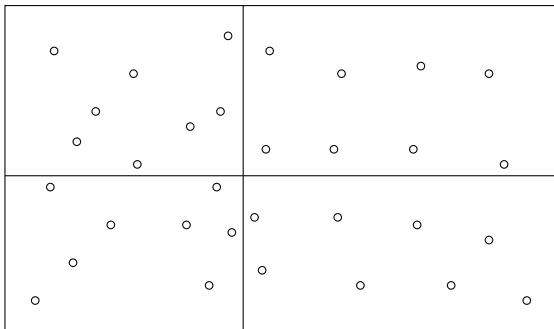
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First Attempt

Divide and Conquer I

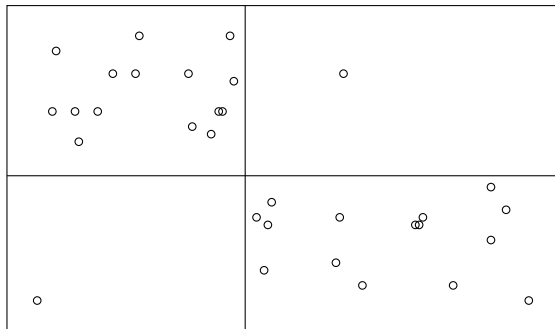
- 1 Partition into 4 quadrants of roughly equal size.
- 2 Find closest pair in each quadrant recursively
- 3 Combine solutions



First Attempt

Divide and Conquer I

- 1 Partition into 4 quadrants of roughly equal size. Not always!
- 2 Find closest pair in each quadrant recursively
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Finding a negative cycle in a graph...

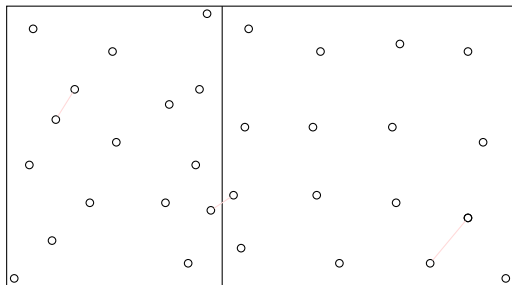
Let $G = (\mathbf{V}, \mathbf{E})$ be a directed graph with \mathbf{n} vertices, \mathbf{m} edges, and (positive or negative) weights on the edges. Assume G has at least one negative cycle. Outputting one of these negative cycles can be done in

- (A) You can only detect if a negative cycle exists. There is no way to compute it.
- (B) This takes exponential time (i.e., this problem is **NP-Hard**).
- (C) $O(\mathbf{nm})$ time.
- (D) $O(\mathbf{n} + \mathbf{m})$.
- (E) IDK (25%).

New Algorithm

Divide and Conquer II

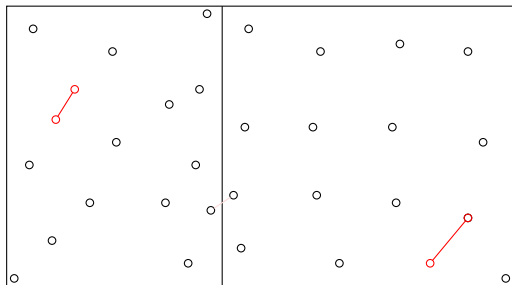
- 1 Divide the set of points into two equal parts via vertical line
- 2 Find closest pair in each half recursively
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- 4 Return the best pair among the above 3 solutions



New Algorithm

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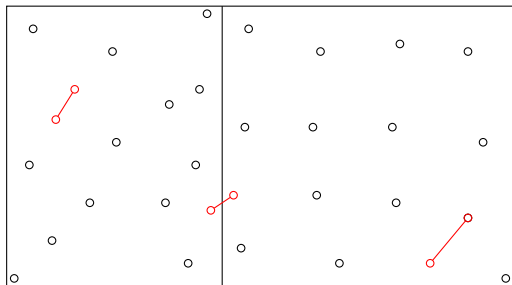
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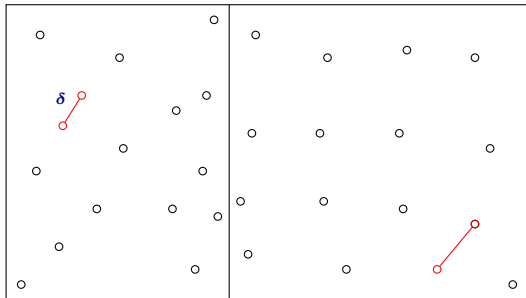
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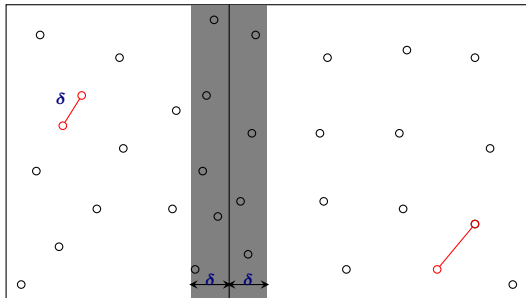
Combining Partial Solutions

- 1 Does it take $O(n^2)$ to combine solutions?
- 2 Let δ be the distance between closest pairs, where both points belong to the same half.

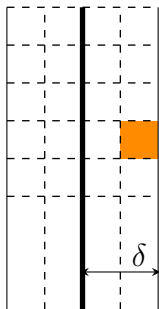


Combining Partial Solutions

- 1 Let δ be the distance between closest pairs, where both points belong to the same half.
- 2 Need to consider points within δ of dividing line



Sparsity of Band



Divide the band into square boxes of size $\delta/2$

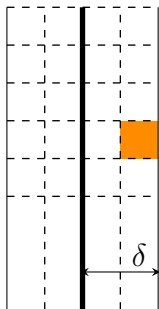
Lemma

Each box has at most one point

Proof.

If not, then there are a pair of points (both belonging to one half) that are at most $\sqrt{2}\delta/2 < \delta$ apart! □

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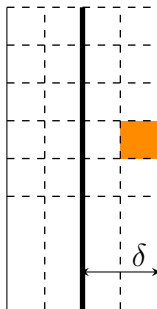
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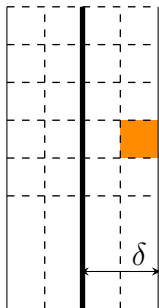
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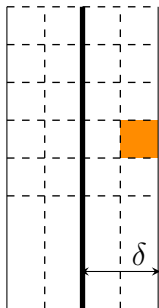
Lemma

Suppose \mathbf{a}, \mathbf{b} are both in the band
 $d(\mathbf{a}, \mathbf{b}) < \delta$ then \mathbf{a}, \mathbf{b} have at most two rows
of boxes between them.

Proof.

Each row of boxes has height $\delta/2$. If more
than two rows then $d(\mathbf{a}, \mathbf{b}) > 2 \cdot \delta/2!$ \square

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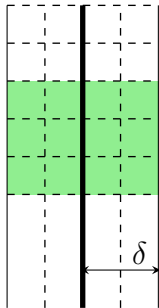
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Searching within the Band

Corollary

Order points according to their y -coordinate. If \mathbf{p}, \mathbf{q} are such that $d(\mathbf{p}, \mathbf{q}) < \delta$ then \mathbf{p} and \mathbf{q} are within **11** positions in the sorted list.



Proof.

- 1 ≤ 2 points between them if \mathbf{p} and \mathbf{q} in same row.
- 2 ≤ 6 points between them if \mathbf{p} and \mathbf{q} in two consecutive rows.
- 3 ≤ 10 points between if \mathbf{p} and \mathbf{q} one row apart.
- 4 \implies More than ten points between them in the sorted y order than \mathbf{p} and \mathbf{q} are more than two rows apart.
- 5 $\implies d(\mathbf{p}, \mathbf{q}) > \delta$. A contradiction. ■

The Algorithm

ClosestPair(**P**):

1. Find vertical line **L** splits **P** into equal halves: **P**₁ and **P**₂.
2. $\delta_1 \leftarrow \text{ClosestPair}(\mathbf{P}_1)$.
3. $\delta_2 \leftarrow \text{ClosestPair}(\mathbf{P}_2)$.
4. $\delta = \min(\delta_1, \delta_2)$
5. Delete points from **P** further than δ from **L**
6. Sort **P** based on **y**-coordinate into an array **A**
7. **for** $i = 1$ to $|\mathbf{A}| - 1$ **do**
 for $j = i + 1$ to $\min\{i + 11, |\mathbf{A}|\}$ **do**
 if ($\text{dist}(\mathbf{A}[i], \mathbf{A}[j]) < \delta$) update δ and closest pair

- 1 Step 1, involves sorting and scanning. Takes $O(n \log n)$ time.
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- 1 Step 1, involves sorting and scanning. Takes $O(n \log n)$ time.
- 2 Step 5 takes $O(n)$ time.
- 3 Step 6 takes $O(n \log n)$ time

The Algorithm

ClosestPair(P):

1. Find vertical line L splits P into equal halves: P_1 and P_2 .
2. $\delta_1 \leftarrow \text{ClosestPair}(P_1)$.
3. $\delta_2 \leftarrow \text{ClosestPair}(P_2)$.
4. $\delta = \min(\delta_1, \delta_2)$
5. Delete points from P further than δ from L
6. Sort P based on y -coordinate into an array A
7. **for** $i = 1$ **to** $|A| - 1$ **do**
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- 4 Step 7 takes $O(n)$ time.

Running Time

The running time of the algorithm is given by

$$T(n) \leq 2T(n/2) + O(n \log n)$$

Thus, $T(n) = O(n \log^2 n)$.

Improved Algorithm

Avoid repeated sorting of points in band: two options

- 1 Sort all points by **y**-coordinate and store the list. In conquer step use this to avoid sorting
- 2 Each recursive call returns a list of points sorted by their **y**-coordinates. Merge in conquer step in linear time.

Analysis: $T(n) \leq 2T(n/2) + O(n) = O(n \log n)$

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Part III

Selecting in Unsorted Lists

Quick Sort

Quick Sort [Hoare]

- 1 Pick a pivot element from array
- 2 Split array into 3 subarrays: those smaller than pivot, those larger than pivot, and the pivot itself. Linear scan of array does it. Time is $O(n)$
- 3 Recursively sort the subarrays, and concatenate them.

Example:

- 1 array: 16, 12, 14, 20, 5, 3, 18, 19, 1
- 2 pivot: 16
- 3 split into 12, 14, 5, 3, 1 and 20, 19, 18 and recursively sort
- 4 put them together with pivot in middle

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Time Analysis

- 1 Let k be the rank of the chosen pivot. Then,

$$T(n) = T(k - 1) + T(n - k) + O(n)$$

- 2 If $k = \lceil n/2 \rceil$ then

$$T(n) = T(\lceil n/2 \rceil - 1) + T(\lfloor n/2 \rfloor) + O(n) \leq 2T(n/2) + O(n).$$

Then, $T(n) = O(n \log n)$.

- 3 Theoretically, median can be found in linear time.

- 3 Typically, pivot is the first or last element of array. Then,

$$T(n) = \max_{1 \leq k \leq n} (T(k - 1) + T(n - k) + O(n))$$

In the worst case $T(n) = T(n - 1) + O(n)$, which means $T(n) = O(n^2)$. Happens if array is already sorted and pivot is always first element.

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Prune and search

Consider an algorithm **alg** that is given an input of size **n**. In **$O(n)$** time, it either solves the problem, or solve it by calling recursively on an input of size, say, $\leq (15/16)n$. The running time of **alg** is

- (A) **$O(n^2)$**
- (B) **$O(n \log n)$**
- (C) **$O(n)$**
- (D) Undefined - there are not enough details.
- (E) I know the answer but unfortunately I was sworn to secrecy and can not share it with you.

Problem - Selection

Input Unsorted array **A** of **n** integers

Goal Find the **j**th smallest number in **A** (*rank j* number)

Example

A = {4, 6, 2, 1, 5, 8, 7} and **j** = 4. The **j**th smallest element is **5**.

Median: $j = \lfloor (n + 1)/2 \rfloor$

Algorithm 1

- 1 Sort the elements in **A**
- 2 Pick **j**th element in sorted order

Time taken = **$O(n \log n)$**

Do we need to sort? Is there an **$O(n)$** time algorithm?

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Algorithm II

If j is small or $n - j$ is small then

- 1 Find j smallest/largest elements in A in $O(jn)$ time. (How?)
- 2 Time to find median is $O(n^2)$.

Divide and Conquer Approach

- 1 Pick a pivot element \mathbf{a} from \mathbf{A}
- 2 Partition \mathbf{A} based on \mathbf{a} .
 $\mathbf{A}_{\text{less}} = \{x \in \mathbf{A} \mid x \leq \mathbf{a}\}$ and $\mathbf{A}_{\text{greater}} = \{x \in \mathbf{A} \mid x > \mathbf{a}\}$
- 3 $|\mathbf{A}_{\text{less}}| = \mathbf{j}$: return \mathbf{a}
- 4 $|\mathbf{A}_{\text{less}}| > \mathbf{j}$: recursively find \mathbf{j} th smallest element in \mathbf{A}_{less}
- 5 $|\mathbf{A}_{\text{less}}| < \mathbf{j}$: recursively find \mathbf{k} th smallest element in $\mathbf{A}_{\text{greater}}$
where $\mathbf{k} = \mathbf{j} - |\mathbf{A}_{\text{less}}|$.

Time Analysis

- 1 Partitioning step: $O(n)$ time to scan A
- 2 How do we choose pivot? Recursive running time?

Suppose we always choose pivot to be $A[1]$.

Say A is sorted in increasing order and $j = n$.

Exercise: show that algorithm takes $\Omega(n^2)$ time

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A Better Pivot

Suppose pivot is the ℓ th smallest element where $n/4 \leq \ell \leq 3n/4$.

That is pivot is *approximately* in the middle of \mathbf{A}

Then $n/4 \leq |\mathbf{A}_{\text{less}}| \leq 3n/4$ and $n/4 \leq |\mathbf{A}_{\text{greater}}| \leq 3n/4$. If we apply recursion,

$$T(n) \leq T(3n/4) + O(n)$$

Implies $T(n) = O(n)$!

How do we find such a pivot? Randomly? In fact works!
Analysis a little bit later.

Can we choose pivot deterministically?

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Divide and Conquer Approach

A game of medians

Idea

- 1 Break input \mathbf{A} into many subarrays: $\mathbf{L}_1, \dots, \mathbf{L}_k$.
- 2 Find median \mathbf{m}_i in each subarray \mathbf{L}_i .
- 3 Find the median \mathbf{x} of the medians $\mathbf{m}_1, \dots, \mathbf{m}_k$.
- 4 Intuition: The median \mathbf{x} should be close to being a good median of all the numbers in \mathbf{A} .
- 5 Use \mathbf{x} as pivot in previous algorithm.

But we have to be...

More specific...

- 1 Size of each group?
- 2 How to find median of medians?

Choosing the pivot

A clash of medians

- 1 Partition array \mathbf{A} into $\lceil n/5 \rceil$ lists of 5 items each.
 $L_1 = \{A[1], A[2], \dots, A[5]\}$, $L_2 = \{A[6], \dots, A[10]\}$, \dots ,
 $L_i = \{A[5i + 1], \dots, A[5i + 5]\}$, \dots ,
 $L_{\lceil n/5 \rceil} = \{A[5\lceil n/5 \rceil - 4], \dots, A[n]\}$.
- 2 For each i find median b_i of L_i using brute-force in $O(1)$ time.
Total $O(n)$ time
- 3 Let $\mathbf{B} = \{b_1, b_2, \dots, b_{\lceil n/5 \rceil}\}$
- 4 Find median b of \mathbf{B}

Lemma

Median of \mathbf{B} is an approximate median of \mathbf{A} . That is, if b is used a pivot to partition \mathbf{A} , then $|A_{less}| \leq 7n/10 + 6$ and $|A_{greater}| \leq 7n/10 + 6$.

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Algorithm for Selection

A storm of medians

select(**A**, **j**):

Form lists $L_1, L_2, \dots, L_{\lceil n/5 \rceil}$ where $L_i = \{A[5i - 4], \dots, A[5i]\}$

Find median b_i of each L_i using brute-force

Find median b of $B = \{b_1, b_2, \dots, b_{\lceil n/5 \rceil}\}$

Partition **A** into A_{less} and A_{greater} using **b** as pivot

if ($|A_{\text{less}}| = j$) **return b**

else if ($|A_{\text{less}}| > j$)

return select(A_{less} , **j**)

else

return select(A_{greater} , $j - |A_{\text{less}}|$)

How do we find median of **B**?

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How do we find median of B ? Recursively!

Running time of deterministic median selection

A dance with recurrences

$$T(n) = T(\lceil n/5 \rceil) + \max\{T(|A_{\text{less}}|), T(|A_{\text{greater}}|)\} + O(n)$$

From Lemma,

$$T(n) \leq T(\lceil n/5 \rceil) + T(\lfloor 7n/10 + 6 \rfloor) + O(n)$$

and

$$T(1) = 1$$

Exercise: show that $T(n) = O(n)$

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Median of Medians: Proof of Lemma

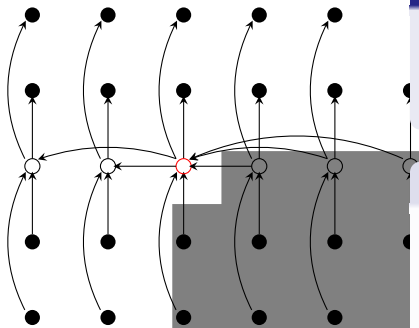


Figure : Shaded elements are all greater than **b**

Proposition

There are at least $3n/10 - 6$ elements greater than the median of medians **b**.

Proof.

At least half of the $\lceil n/5 \rceil$ groups have at least 3 elements larger than **b**, except for last group and the group containing **b**. So **b** is less than

$$3(\lceil (1/2)\lceil n/5 \rceil \rceil - 2) \geq 3n/10 - 6$$

□

Median of Medians: Proof of Lemma

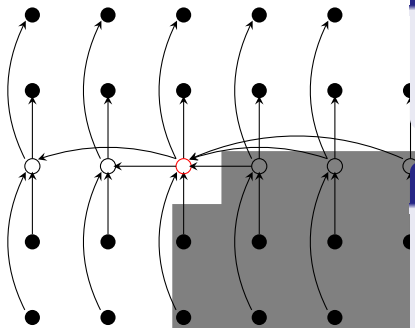


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Median of Medians: Proof of Lemma

Proposition

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Corollary

$$|A_{\text{less}}| \leq 7n/10 + 6.$$

Via symmetric argument,

Corollary

$$|A_{\text{greater}}| \leq 7n/10 + 6.$$

Questions to ponder

- 1 Why did we choose lists of size **5**? Will lists of size **3** work?
- 2 Write a recurrence to analyze the algorithm's running time if we choose a list of size **k**.

Median of Medians Algorithm

Due to:

M. Blum, R. Floyd, D. Knuth, V. Pratt, R. Rivest, and R. Tarjan.

“Time bounds for selection”.

Journal of Computer System Sciences (JCSS), 1973.

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Takeaway Points

- 1 Recursion tree method and guess and verify are the most reliable methods to analyze recursions in algorithms.
- 2 Recursive algorithms naturally lead to recurrences.
- 3 Some times one can look for certain type of recursive algorithms (reverse engineering) by understanding recurrences and their behavior.

Notes

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