CSC373

Algorithm Design, Analysis & Complexity

Nisarg Shah

Introduction

- Instructors
 - Nisarg Shah
 - o cs.toronto.edu/~nisarg, nisarg@cs, SF 2301C
 - o LEC 0101 and 0201
- TAs: Too many to list

Disclaimer!

- > Many things are up in the air, so expect a somewhat bumpy ride at the start, but hopefully, we'll get through together!
- > Use any of the feedback mediums (email, Piazza, ...) to let me know if you have any suggestions for improvement

Mostly useless this semester!

Course Information

- Course Page www.cs.toronto.edu/~nisarg/teaching/373f21/
 - > Not ready yet, but will be ready soon
- Discussion Board piazza.com/utoronto.ca/fall2021/csc373
- Grading MarkUs
 - > Link will be distributed after about a week or two
 - > LaTeX preferred, scans are OK!
- All times will be in the Eastern time zone

Lectures & Tutorials

Lectures for LEC 0101

- > Tue 1-3pm, Thu 2-3pm
- > BA 1190 or Zoom

Lectures for LEC 0201

- > Tue 3-4pm, Thu 3-5pm
- > KP 108 or Zoom

Tutorials for both sections

- ➤ Mon 4-5pm
- > In-person (room details TBA on the course webpage) or on Zoom

Delivery

First two weeks

- > Lectures, tutorials, and office hours will all be on Zoom
- > All Zoom links and other details TBA on the course webpage

Afterwards

- > To be decided!
- > Likely, lectures and office hours will be online, and tutorials will be in-person
 - FAS requires some regularly scheduled component of the course (i.e., either lectures or tutorials) to be in-person
 - o Zoom poll!

Lecture Format

- Delivered by me
- Will start at 10 minutes past the hour
 - > 10-minute break after 50 minutes of lecture in the 2-hour slot
- In-person/on Zoom: Ask questions by raising your hand
 - > On Zoom, you can also type your question in chat, but it may take me some time to see it

Tutorial Format

- Delivered by the TAs
- Think of them as preparation for assignments/exams
 - > Some of the tutorial problems may be easier than assignment/exam questions
- Problem sets & solutions
 - > Problem sets will be posted to the course webpage in advance of the tutorial
 - > Solutions will be posted to the course webpage after the tutorial
- What to do
 - > Please attempt the problems before coming to the tutorials
 - > During the tutorials, the TAs will go over the solutions and explain key ideas

Tutorial Format

Further details

- > Each section is divided into three parts (A,B,C)
- > Division by birth month: A = Jan-Apr, B = May-Aug, C = Sep-Dec
- > Feel free to attend a different tutorial than the one you're assigned
 - o EXCEPT when the tutorial slot is being used for a test
- > If the tutorial attendance is really low, the number of tutorials per section may be reduced

Office Hours

- Time & Place: Tue 4-5pm, Thu 10-11am, Zoom
 - > If you have a conflict with both these slots, feel free to schedule 1-1 office hours by emailing me

Details

- > I will conduct them
- Use the "raise hand" feature
- > When I call your name, unmute and ask the question
- > Try to phrase your question without giving away your approach/solution to an assignment problem
 - o If this is not possible, we will go to a breakout room

Tests

2 term tests, one end-of-term test (final exam/assessment)

- Time & Place
 - Mon 4-6pm (tutorial slot + subsequent hour)
 - > If you have a regular commitment that conflicts with Mon 5-6pm, reach out to me by Sep 30
 - > You will likely be offered an alternative time *before* the actual test (e.g., on the morning of the same Monday)
 - > If you do not reach out by then, I will only grant an exception for emergencies, at my discretion

Delivery method: TBD

Assignments

4 assignments, best 3 out of 4

Group work

- > In groups of up to three students
- > Best way to learn is for each member to try each problem

Questions will be more difficult

- > May need to mull them over for several days; do *not* expect to start and finish the assignment on the same day!
- > May include bonus questions
- Submission on MarkUs, more details later
 - > May need to compress the PDF

Embedded EthiCS Module

Relatively new initiative

> You may have seen this in one of your courses during the last year

Goal

- > Help you learn how to reason about ethical issues, practice conveying your thoughts on such issues
- > In the context of a topic from the course

Will be during a 2-hour lecture slot of your section

- > Later in the course (likely end of November)
- > A lightweight assignment before & after the module
- > Three group activities during the module

Grading Policy

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    Best 3/4 homeworks * 10% = 30%
    2 term tests * 20% = 40%
    EthiCS Module * 5% = 5%
    Final exam * 25% = 25%
```

• NOTE: To pass, you must earn at least 40% on the final exam

Approximate Due Dates

> Assignment 1: Oct 5

> Assignment 2: Oct 19

> Assignment 3: Nov 16

> Assignment 4: Dec 10

> Midterm 1: Oct 25

> Midterm 2: Nov 22

Textbook

- Primary reference: lecture slides
- Primary textbook (required)
 - > [CLRS] Cormen, Leiserson, Rivest, Stein: Introduction to Algorithms.
- Supplementary textbooks (optional)
 - > [DPV] Dasgupta, Papadimitriou, Vazirani: Algorithms.
 - > [KT] Kleinberg; Tardos: Algorithm Design.

Other Policies

Collaboration

- > Free to discuss with classmates or read online material
- > Must write solutions in your own words
 - o Easier if you do not take any pictures/notes from discussions

Citation

- > For each question, must cite the peer (write the name) or the online sources (provide links), if you obtained a significant insight directly pertinent to the question
- > Failing to do this is plagiarism!

Other Policies

- "No Garbage" Policy
 - > Borrowed from: Prof. Allan Borodin (citation!)
 - > Applies to assignments (except for bonus questions) and tests
 - 1. Partial marks for viable approaches
 - 2. Zero marks if the answer makes no sense
 - 3. 20% marks if you admit to not knowing how to approach the question ("I do not know how to approach this question")
- 20% > 0%!!

Other Policies

Late Days

- > 4 total late days across all 4 assignments
- Managed by MarkUs
- > At most 2 late days can be applied to a single assignment
- > Already covers legitimate reasons such as illness, university activities, etc.
 - o Petitions will only be granted for circumstances which cannot be covered by this

Questions?

Enough with the boring stuff.

What will we study?

Why will we study it?



Muhammad ibn Musa al-Khwarizmi c. 780 - c. 850

Algorithms

- > Ubiquitous in the real world
 - o From your smartphone to self-driving cars
 - o From graph problems to graphics problems
 - 0 ...
- > Important to be able to design and analyze algorithms
- > For some problems, good algorithms are hard to find
 - o For some of these problems, we can formally establish complexity results
 - o We'll often find that one problem is easy, but its minor variants are suddenly hard

Algorithms

- > Algorithms in specialized environments or using advanced techniques
 - o Distributed, parallel, streaming, sublinear time, spectral, genetic...
- > Other concerns with algorithms
 - o Fairness, ethics, ...
- > ...mostly beyond the scope of this course

- Designing fast algorithms
 - > Divide and Conquer
 - > Greedy
 - > Dynamic programming
 - Network flow
 - > Linear programming
- Proving that no fast algorithms are likely possible
 - > Reductions & NP-completeness
- What to do if no fast algorithms are likely possible
 - > Approximation algorithms (if time permits)
 - > Randomized algorithms (if time permits)

- How do we know which paradigm is right for a given problem?
 - > A very interesting question!
 - > Subject of much ongoing research...
 - o Sometimes, you just know it when you see it...
- How do we analyze an algorithm?
 - > Proof of correctness
 - > Proof of running time
 - We'll try to prove the algorithm is efficient in the worst case
 - In practice, average case matters just as much (or even more)

- What does it mean for an algorithm to be efficient in the worst case?
 - > Polynomial time
 - > It should use at most poly(n) steps on any n-bit input
 - $o n, n^2, n^{100}, 100n^6 + 237n^2 + 432, ...$
 - \rightarrow If the input to an algorithm is a number x, the number of bits of input is $\log x$
 - This is because it takes $\log x$ bits to represent the input x in binary
 - \circ So the running time should be polynomial in $\log x$, not in x
 - > How much is too much?

Picture-Hanging Puzzles*

Erik D. Demaine[†] Martin L. Demaine[†] Yair N. Minsky[‡] Joseph S. B. Mitchell[§] Ronald L. Rivest[†] Mihai Pătrașcu[¶]

Theorem 7 For any $n \ge k \ge 1$, there is a picture hanging on n nails, of length $n^{c'}$ for a constant c',

Theorem 7 For any $n \ge k \ge 1$, there is a picture hanging on n nails, of length n^c for a constant c', that falls upon the removal of any k of the nails.

 $n^{6,100 \log_2 c}$. Using the $c \leq 1,078$ upper bound, we obtain an upper bound of $c' \leq 6,575,800$. Using

So, while this construction is polynomial, it is a rather large polynomial. For small values of n, we can use known small sorting networks to obtain somewhat reasonable constructions.

Better Balance by Being Biased: A 0.8776-Approximation for Max Bisection

Per Austrin*, Siavosh Benabbas*, and Konstantinos Georgiou†

has a lot of flexibility, indicating that further improvements may be possible. We remark that, while polynomial, the running time of the algorithm is somewhat abysmal; loose estimates places it somewhere around $O(n^{10^{100}})$; the running time of the algorithm of [RT12] is similar.

- What if we can't find an efficient algorithm for a problem?
 - > Try to prove that the problem is hard
 - > Formally establish complexity results
 - > NP-completeness, NP-hardness, ...
- We'll often find that one problem may be easy, but its simple variants may suddenly become hard
 - > Minimum spanning tree (MST) vs bounded degree MST
 - > 2-colorability vs 3-colorability

I'm not convinced.

Will I really ever need to know how to design abstract algorithms?

At the very least...

This will help you prepare for your technical job interview!

Real Microsoft interview question:

- Given an array a, find indices (i, j) with the largest j i such that a[j] > a[i]
- Greedy? Divide & conquer?

Disclaimer

- The course is theoretical in nature
 - > You'll be working with abstract notations, proving correctness of algorithms, analyzing the running time of algorithms, designing new algorithms, and proving complexity results.
- Something for everyone...
 - > If you're somewhat scared going into the course
 - > If you're already comfortable with the proofs, and want challenging problems

Related/Follow-up Courses

Direct follow-up

- > CSC473: Advanced Algorithms
- > CSC438: Computability and Logic
- > CSC463: Computational Complexity and Computability

Algorithms in other contexts

- > CSC304: Algorithmic Game Theory and Mechanism Design (self promotion!)
- > CSC384: Introduction to Artificial Intelligence
- > CSC436: Numerical Algorithms
- > CSC418: Computer Graphics

Divide & Conquer

History?

- Maybe you saw a subset of these algorithms?
 - \rightarrow Mergesort $O(n \log n)$
 - \succ Karatsuba algorithm for fast multiplication $O(n^{\log_2 3})$ rather than $O(n^2)$
 - \triangleright Largest subsequence sum in O(n)
 - **>** ...
- Have you seen some divide & conquer algorithms before?
 - > Maybe in CSC236/CSC240 and/or CSC263/CSC265
 - > Write "yes"/"no" in chat

Divide & Conquer

- General framework
 - > Break (a large chunk of) a problem into two smaller subproblems of the same type
 - > Solve each subproblem recursively and independently
 - > At the end, quickly combine solutions from the two subproblems and/or solve any remaining part of the original problem
- Hard to formally define when a given algorithm is divide-and-conquer...
- Let's see some examples!

Problem

 \triangleright Given an array a of length n, count the number of pairs (i,j) such that i < j but a[i] > a[j]

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)

- Problem
 - \triangleright Count (i, j) such that i < j but a[i] > a[j]
- Brute force
 - > Check all $\Theta(n^2)$ pairs
- Divide & conquer
 - Divide: break array into two equal halves x and y
 - Conquer: count inversions in each half recursively
 - > Combine:
 - \circ Solve (we'll see how): count inversions with one entry in x and one in y
 - Merge: add all three counts

• From Kevin Wayne's slides

SORT-AND-COUNT (L)

If list L has one element RETURN (0, L).

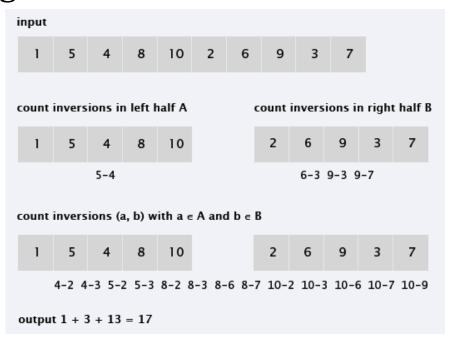
DIVIDE the list into two halves A and B.

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

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

$$(r_{AB}, L') \leftarrow \text{Merge-And-Count}(A, B).$$

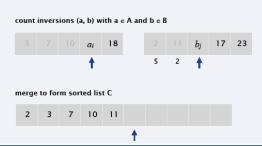
RETURN
$$(r_A + r_B + r_{AB}, L')$$
.



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

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.



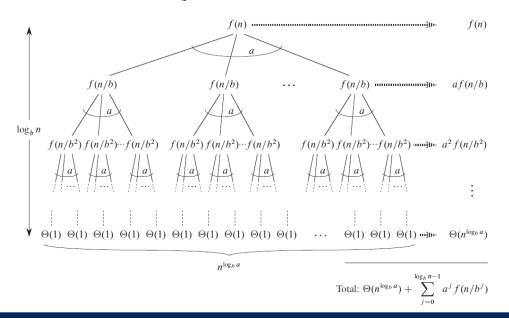
- How do we formally prove correctness?
 - > Induction on *n* is usually very helpful
 - > Allows you to assume correctness of subproblems
- Running time analysis
 - \rightarrow Suppose T(n) is the worst-case running time for inputs of size n
 - > Our algorithm satisfies $T(n) \le 2 T(n/2) + O(n)$
 - \rightarrow Master theorem says this is $T(n) = O(n \log n)$

Master Theorem

- Here's the master theorem.
 - > Useful for analyzing divide-and-conquer running time
 - > If you haven't already seen it, please spend some time understanding it
 - ▶ Theorem: Let $a \ge 1$ and b > 1 be constants, f(n) be a function, and T(n) be defined on nonnegative integers by the recurrence $T(n) \le a \cdot T\left(\frac{n}{b}\right) + f(n)$, where n/b can be $\left\lceil \frac{n}{b} \right\rceil$. Let $d = \log_b a$. Then:
 - o If $f(n) = O(n^{d-\epsilon})$ for some constant $\epsilon > 0$, then $T(n) = O(n^d)$.
 - \circ If $f(n) = O(n^d \log^k n)$ for some $k \ge 0$, then $T(n) = O(n^d \log^{k+1} n)$.
 - \circ If $f(n) = O(n^{d+\epsilon})$ for some constant $\epsilon > 0$, then T(n) = O(f(n)).

Master Theorem

Intuition: Compare f(n) with $n^{\log_b a}$. The larger determines the recurrence solution.



Problem:

 \triangleright Given n points of the form (x_i, y_i) in the plane, find the closest pair of points.

• Applications:

- > Basic primitive in graphics and computer vision
- > Geographic information systems, molecular modeling, air traffic control
- > Special case of nearest neighbor

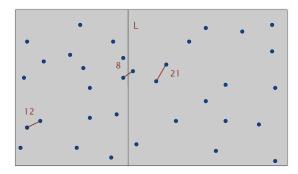
• Brute force: $\Theta(n^2)$

Intuition from 1D?

- In 1D, the problem would be easily $O(n \log n)$
 - > Sort and check!
- Sorting attempt in 2D
 - > Find closest points by x coordinate
 - > Find closest points by y coordinate
 - > Doesn't work! (Exercise: come up with a counterexample)
- Non-degeneracy assumption
 - > No two points have the same x or y coordinate

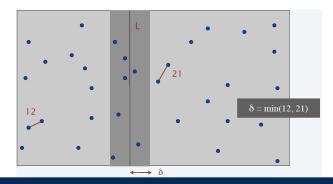
- Let's try divide-and-conquer!
 - ➤ Divide: points in equal halves by drawing a vertical line L
 - Conquer: solve each half recursively
 - ➤ Combine: find closest pair with one point on each side of L
 - > Return the best of 3 solutions





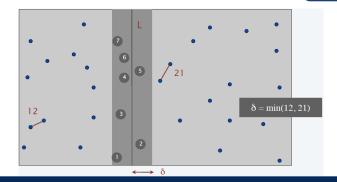
Combine

> We can restrict our attention to points within δ of L on each side, where δ = best of the solutions within the two halves



- Combine (let δ = best of solutions in two halves)
 - \triangleright Only need to look at points within δ of L on each side,
 - > Sort points on the strip by y coordinate
 - > Only need to check each point with next 11 points in sorted list!

Wait, what? Why 11?



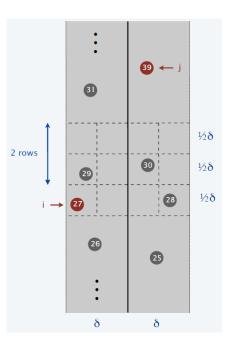
Why 11?

• Claim:

> If two points are at least 12 positions apart in the sorted list, their distance is at least δ

• Proof:

- > No two points lie in the same $\delta/2 \times \delta/2$ box
- > Two points that are more than two rows apart are at distance at least δ



Running Time Analysis

- Running time for the combine operation
 - \triangleright Finding points on the strip: O(n)
 - > Sorting points on the strip by their y-coordinate: $O(n \log n)$
 - \rightarrow Testing each point against 11 points: O(n)
- Total running time: $T(n) \le 2T\left(\frac{n}{2}\right) + O(n\log n)$
- By the Master theorem, this yields $T(n) = O(n \log^2 n)$
 - \triangleright Can be improved to $O(n \log n)$ by doing a single global sort by y-coordinate at the beginning

Recap: Karatsuba's Algorithm

- Fast way to multiply two n digit integers x and y
- Brute force: $O(n^2)$ operations
- Karatsuba's observation:
 - > Divide each integer into two parts

$$0 x = x_1 * 10^{n/2} + x_2, y = y_1 * 10^{n/2} + y_2$$

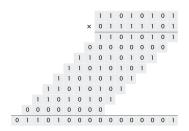
$$0 xy = (x_1y_1) * 10^n + (x_1y_2 + x_2y_1) * 10^{n/2} + (x_2y_2)$$

 \rightarrow Four $^{n}/_{2}$ -digit multiplications can be replaced by three

$$\circ x_1 y_2 + x_2 y_1 = (x_1 + x_2)(y_1 + y_2) - x_1 y_1 - x_2 y_2$$

> Running time

$$\circ T(n) \le 3 T(n/2) + O(n) \Rightarrow T(n) = O(n^{\log_2 3})$$



Strassen's Algorithm

- Generalizes Karatsuba's insight to design a fast algorithm for multiplying two $n \times n$ matrices
 - > Call *n* the "size" of the problem

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} * \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

 \triangleright Naively, this requires 8 multiplications of size n/2

$$\circ A_{11} * B_{11}, A_{12} * B_{21}, A_{11} * B_{12}, A_{12} * B_{22}, \dots$$

> Strassen's insight: replace 8 multiplications by 7

$$\circ$$
 Running time: $T(n) \le 7 T(n/2) + O(n^2) \Rightarrow T(n) = O(n^{\log_2 7})$

Strassen's Algorithm

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} * \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

```
STRASSEN(n, A, B)
                        IF n=1 RETURN A \times B.
assume n is
                        Partition A and B into 2-by-2 block matrices.
a power of 2
                        P_1 \leftarrow \text{STRASSEN}(n / 2, A_{11}, (B_{12} - B_{22})).
                        P_2 \leftarrow \text{STRASSEN}(n / 2, (A_{11} + A_{12}), B_{22}).
                                                                                                keep track of indices of submatrices
                                                                                                      (don't copy matrix entries)
                        P_3 \leftarrow \text{STRASSEN}(n/2, (A_{21} + A_{22}), B_{11}).
                        P_4 \leftarrow \text{STRASSEN}(n/2, A_{22}, (B_{21} - B_{11})).
                        P_5 \leftarrow \text{STRASSEN}(n / 2, (A_{11} + A_{22}) \times (B_{11} + B_{22})).
                        P_6 \leftarrow \text{STRASSEN}(n / 2, (A_{12} - A_{22}) \times (B_{21} + B_{22})).
                        P_7 \leftarrow \text{STRASSEN}(n / 2, (A_{11} - A_{21}) \times (B_{11} + B_{12})).
                        C_{11} = P_5 + P_4 - P_2 + P_6
                        C_{12} = P_1 + P_2.
                        C_{21} = P_3 + P_4.
                        C_{22} = P_1 + P_5 - P_3 - P_7
                        RETURN C.
```

Median & Selection

Selection:

- \triangleright Given array A of n comparable elements, find kth smallest
- > k = 1 is min, k = n is max, $k = \lfloor (n+1)/2 \rfloor$ is median
- > O(n) is easy for min/max

What about k-selection?

- > O(nk) by modifying bubble sort
- $> O(n \log n)$ by sorting
- $> O(n + k \log n)$ using min-heap
- $> O(k + n \log k)$ using max-heap
- Q: What about just O(n)?
- A: Yes! Selection is easier than sorting.

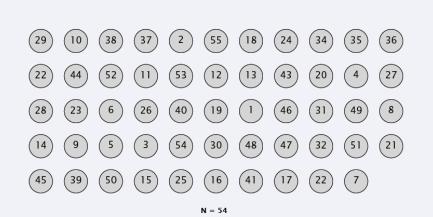
QuickSelect

- Find a pivot p
- Divide A into two sub-arrays
 - A_{less} = elements $\leq p$, A_{more} = elements > p
 - > If $|A_{less}| \ge k$, return k-th smallest in A_{less} , otherwise return $(k |A_{less}|)$ -th smallest element in A_{more}

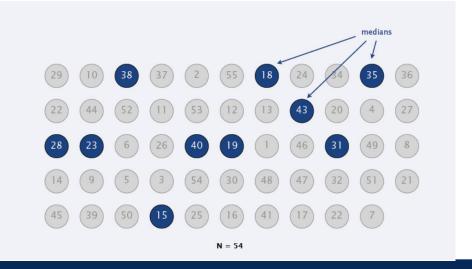
Problem?

- > If pivot is close to the min or the max, then we basically get $T(n) \le T(n-1) + O(n)$, which only gives us $T(n) = O(n^2)$
- \rightarrow We want to reduce n-1 to a fraction of n (e.g., n/2, 5n/6, etc)

• Divide n elements into n/5 groups of 5 each

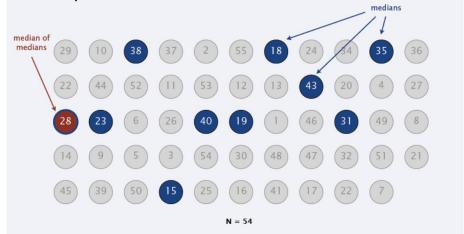


- Divide n elements into n/5 groups of 5 each
- Find the median of each group



- Divide n elements into n/5 groups of 5 each
- Find the median of each group

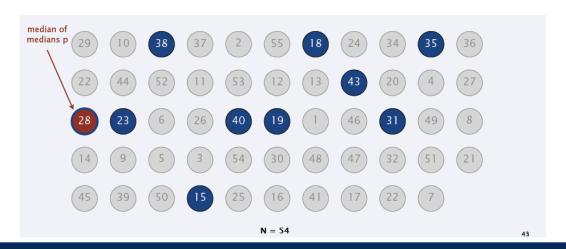
• Find the median of n/5 medians



- Divide n elements into n/5 groups of 5 each
- Find the median of each group
- Find the median of n/5 medians
- Use this median of medians as the pivot in quickselect

Q: Why does this work?

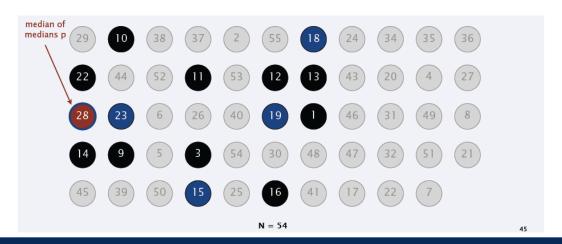
- How many elements can be $\leq p^*$?
 - > Out of n/5 medians, n/10 are $> p^*$



- How many elements can be $\leq p^*$?
 - > Out of n/5 medians, n/10 are $> p^*$

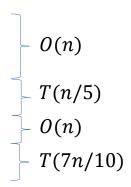


- n/10 of the n/5 medians are $\leq p^*$
 - > For each such median, there are 3 elements $\leq p^*$
 - > So there can be at most $^{7n}/_{10}$ elements that can be $> p^*$



- Thus, $|A_{more}| \le ^{7n}/_{10}$
 - \rightarrow Similarly, $|A_{less}| \leq ^{7n}/_{10}$
 - > (These are rough calculations...)
- How does this factor into overall algorithm analysis?

- Divide n elements into n/5 groups of 5 each
- Find the median of each group
- Find p^* = median of n/5 medians
- Create A_{less} and A_{more} according to p^*
- Run selection on one of A_{less} or A_{more}



•
$$T(n) \le T(n/5) + T(7n/10) + O(n)$$

- Note: $\frac{n}{5} + \frac{7n}{10} = \frac{9n}{10}$
 - > Only a fraction of n, so using a similar analysis to the one in the Master theorem, T(n) = O(n)

- Lower bounds on the worst-case running time
 - > Note that we only derived *upper bounds* on the worst-case running time of the form $T(n) = O(n^2)$ or T(n) = O(n)
 - > If we want to claim that our algorithm does not run *faster* than what is claimed in this upper bound, we have to produce a matching *lower bound*, e.g., $T(n) = \Omega(n^2)$ or $T(n) = \Omega(n)$
 - > This is typically done by producing a *family of examples*, one for each value of n, such that the algorithm's running time on these examples grows like n^2 or n as the value of n grows

- Best algorithm for a problem?
 - > Typically hard to determine
 - \rightarrow We still don't know best algorithms for multiplying two n-digit integers or two $n \times n$ matrices
 - Integer multiplication
 - Breakthrough in March 2019: first $O(n \log n)$ time algorithm
 - It is conjectured that this is asymptotically optimal
 - Matrix multiplication
 - 1969 (Strassen): $O(n^{2.807})$
 - 1990: $O(n^{2.376})$
 - 2013: $O(n^{2.3729})$
 - 2014: $O(n^{2.3728639})$

- Best algorithm for a problem?
 - > Usually, we design an algorithm and then analyze its running time
 - > Sometimes we can do the reverse:
 - o E.g., if you know you want an $O(n^2 \log n)$ algorithm
 - o Master theorem suggests that you can get it by

$$T(n) = 4T(n/2) + O(n^2)$$

 \circ So maybe you want to break your problem into 4 problems of size n/2 each, and then do $O(n^2)$ computation to combine

Access to input

- > For much of this analysis, we are assuming random access to elements of input
- > So we're ignoring underlying data structures (e.g. doubly linked list, binary tree, etc.)

Machine operations

- > We're only counting the number of comparison or arithmetic operations
- > So we're ignoring issues like how real numbers are stored in the closest pair problem
- > When we get to P vs NP, representation will matter

Size of the problem

- > Can be any reasonable parameter of the problem
- \triangleright E.g., for matrix multiplication, we used n as the size
- \triangleright But an input consists of two matrices with n^2 entries
- > It doesn't matter whether we call n or n^2 the size of the problem
- > The actual running time of the algorithm won't change