

CS3000: Algorithms & Data

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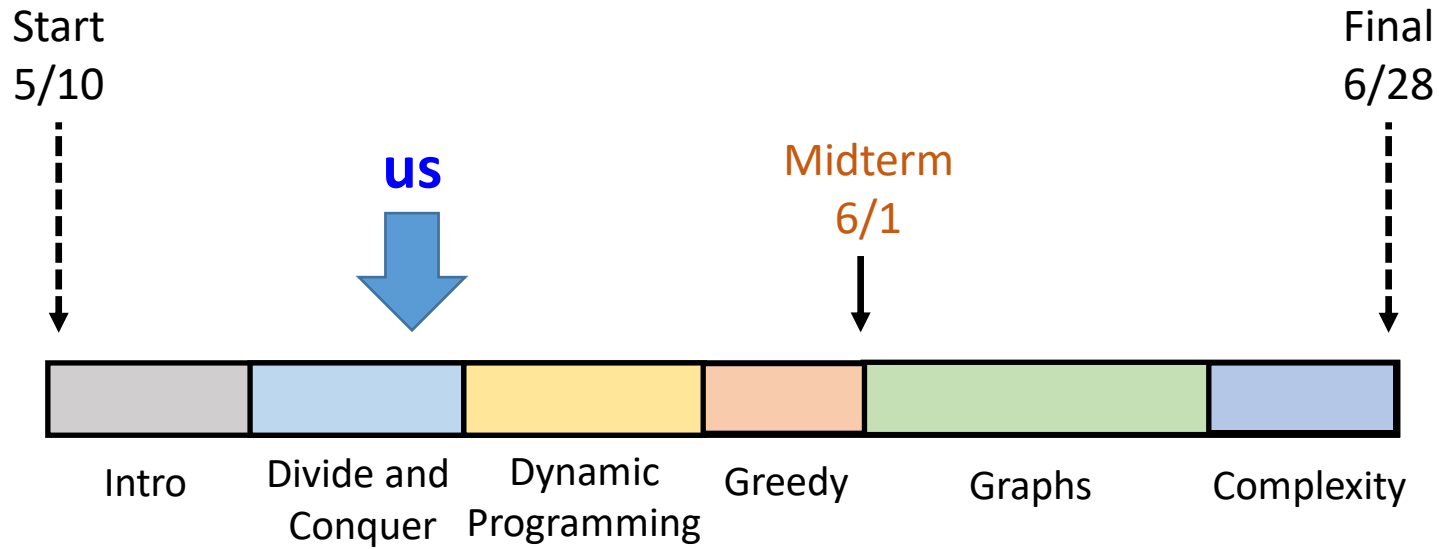
Lecture 7

- Binary Search
- Selection

May 19, 2021



Outline



Last class: divide and conquer: Karatsuba's, Master theorem

Next class: dynamic programming: Weighted Interval Scheduling



(Reduce)

Divide-and-Conquer:
Binary Search



Binary Search

↙ *target*
Is t in this list? If so, where?

Sorted:

2	3	8	11	<u>15</u>	17	28	42
---	---	---	----	-----------	----	----	----

A

Q: If t were in the list, which half would it be in?

$$t \geq 15$$

look in right half

15	17	28	42
----	----	----	----

$$t < 15$$

look in left half

2	3	8	11
---	---	---	----



Binary Search

Is 28 in this list? If so, where?

Sorted:

2	3	8	11	15	17	28	42
---	---	---	----	----	----	----	----

A

$n = 8$

$l = 1$

m

$r = n$

$$m = l + \left\lfloor \frac{r-l}{2} \right\rfloor$$

$$1 + \left\lfloor \frac{8-1}{2} \right\rfloor = 4$$

+ vs. $A[m]$

+ vs. $A[m]$

$28 > 11$ ↓

look in

15	17	28	42
----	----	----	----

$l = m+1, r = r$

$28 > 17$

28	42
----	----

$l = 7, r = 8$

$28 = 28$ ✓

return $m = 7$

① $+ > A[m] \rightarrow$ search in $A[m+1, \dots, r]$

② $+ < A[m] \rightarrow$ search in $A[1, \dots, m]$

③ $+ = A[m] \rightarrow$ return m



- Problems: counting students, stable matching, sorting, n-digit multiplication, **array searching**
- Alg. techniques: divide & conquer
- Analysis: asymptotic analysis, recursion trees, Master Thm.
- Proof techniques: (strong) induction, contradiction



Binary Search

Search(A, t) :

// A[1:n] sorted in ascending order

Return BS(A, 1, n, t)
 ℓ r

BS(A, ℓ , r , t) :

If ($\ell > r$) : return FALSE

$$m \leftarrow \ell + \left\lfloor \frac{r - \ell}{2} \right\rfloor$$

① If (A[m] = t) : return m

② ElseIf (A[m] > t) : return BS(A, ℓ , m-1, t)

③ Else: return BS(A, m+1, r, t)

Only make 1
of these
recursive calls

W/C
runtime of
BS on
list of
size n
↓
T(n):

$$T(n) = T\left(\frac{n}{2}\right) + O(1)$$

T(1):

$$O(1)$$



Running Time Analysis

$$\begin{aligned}T(n) &= T(n/2) + C \\T(1) &= C\end{aligned}$$

Master Thm.!

$$a = 1$$

$$b = 2$$

$$d = 0$$

$$\frac{1}{2^0} = 1$$

Case 2

$$T(n) = O(\log n)$$

or $\Theta(\log n)$ in w/c



Binary Search Wrapup

- Search a sorted array in time $O(\log n)$
- Divide-and-conquer approach
 - Find the middle of the list, recursively search half the list
 - **Key Fact:** eliminate half the list each time
- Prove correctness via induction
- Analyze running time via Master Thm.
 - $T(n) = T(n/2) + C$



Selection (Median)



Selection

↙ *unsorted*

- Given an array of numbers $A[1, \dots, \underline{n}]$, how quickly can I find the:

- Smallest number? $\rightarrow O(n)$

- Second smallest? $\rightarrow O(n)$

- k -th smallest? $\rightarrow O(kn)$

- median? $\rightarrow O(n^2)$

$\underbrace{n}_{1^{st} \text{ Pass}} + \underbrace{n-1}_{2^{nd} \text{ Pass}}$

let $k = \lceil \frac{n}{2} \rceil$

SELECTION
MEDIAN

median: $\lceil \frac{n}{2} \rceil$

11	3	42	28	17	8	2	15
----	---	----	----	----	---	---	----

A

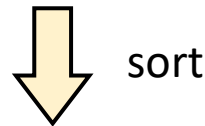


Selection

- **Fact:** can select the k -th smallest in $O(n \log n)$ time
 - Sort the list and look up $A[k]$ $O(n \log n)$

11	3	42	28	17	8	2	15
----	---	----	----	----	---	---	----

A



2	3	8	11	15	17	28	42
---	---	---	----	----	----	----	----

- **Today:** select the k -th smallest in $O(n)$ time

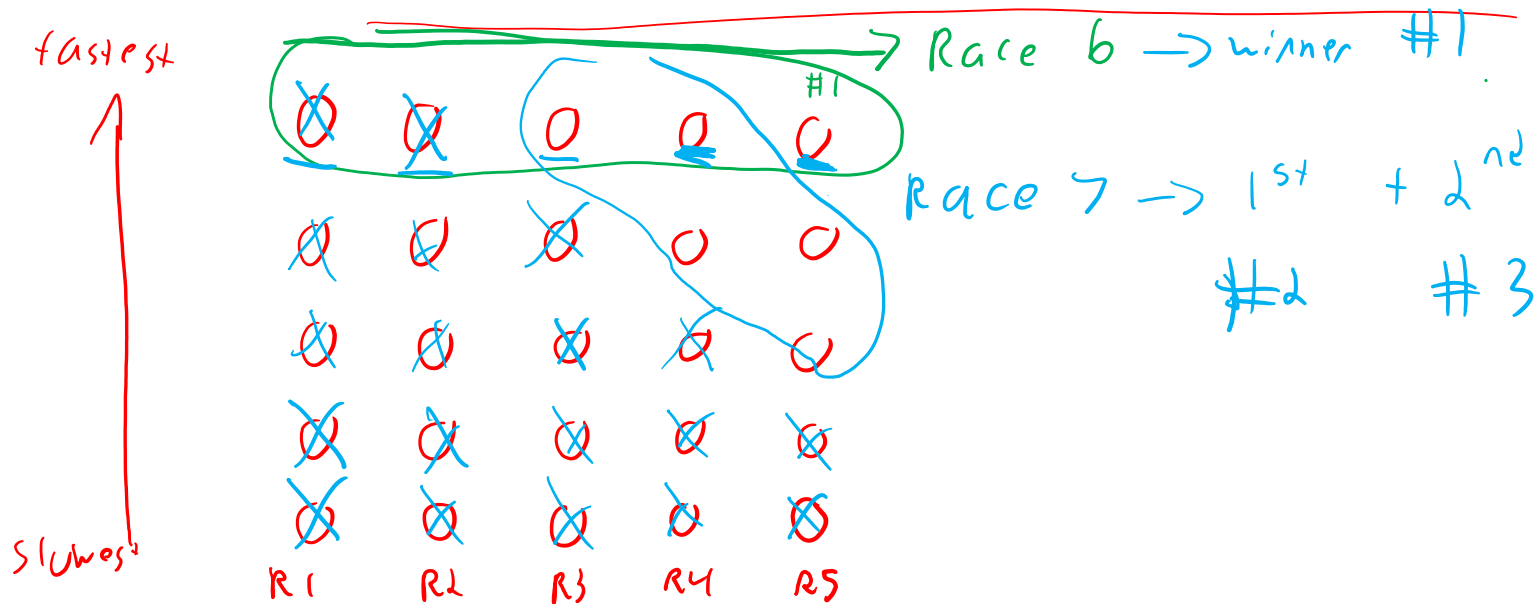


- Problems: counting students, stable matching, sorting, n-digit multiplication, array searching, **selection**
- Alg. techniques: divide & conquer
- Analysis: asymptotic analysis, Master Thm.
- Proof techniques: (strong) induction, contradiction



Warmup

- You have 25 horses and want to find the 3 fastest
- You have a racetrack where you can race 5 at a time
 - In: {1, 5, 6, 18, 22} Out: (6 > 5 > 18 > 22 > 1)
 - You don't have a stopwatch
 - Each horse always has the same finish time
- **Problem:** find the 3 fastest with only seven races



Median Algorithm: Take I

$p = 17$

17	3	42	11	28	8	2	15	13
----	---	----	----	----	---	---	----	----

A

$r = 7$

11	3	15	13	2	8	17	28	42
----	---	----	----	---	---	----	----	----

`Select(A[1:n], k):`

`If (n = 1): return A[1]`

`Choose a pivot $p = A[1]$`

`Partition around the pivot, let r = indexOf(A, p)`

`If (k = r): return A[r]`

`ElseIf (k < r): return Select(A[1:r-1], k) ←`

`ElseIf (k > r): return Select(A[r+1:n], k-r) ←`



Median Algorithm: Take I

$p=1$

1	2	3	4	5	6	7	8	9
---	---	---	---	---	---	---	---	---

A

$p=2$

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

Partitions around pivot: $O(n)$

Require $n/2$ iterations $\rightarrow O(n^2)$



Median Algorithm: Take II

- **Problem:** we need to find a good pivot element

Best pivot element: Median!

* "close" to the median
is good enough



Median of Medians

MOM(A[1:n]) :

Let $m \leftarrow \lfloor n/5 \rfloor$

For $i = 1, \dots, m$:

$M[i] \leftarrow \text{median}\{A[5i-4], \dots, A[5i]\}$

$p \leftarrow \text{Select}(M[1:m], \lfloor m/2 \rfloor)$

splitting into
groups of size 5
+
finding median

MOM
↑
list of medians
↑
median

Finding median in list of length 5:

$O(1)$

Number of times we find median:

$\frac{n}{5} = O(n)$

of ops. excluding recursive call:

$\frac{n}{5} \cdot c = O(n)$

MoM running time:

$c n + \text{time to run Select (find median)}$
on list of length $m = n/5$



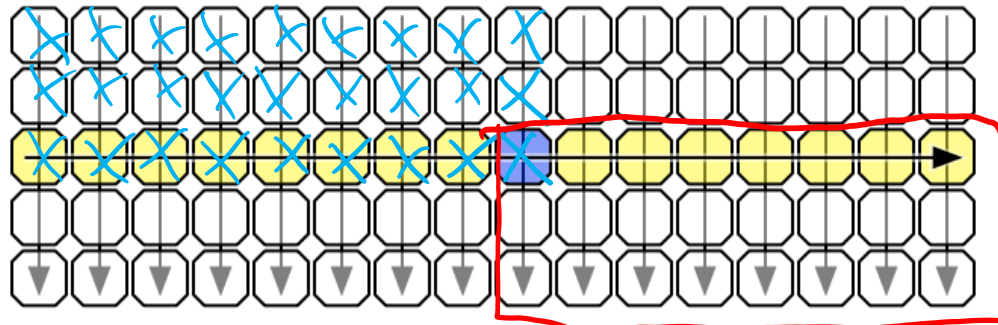
Median of Medians

- **Claim:** For every A there are at least $\frac{3n}{10}$ items that are ^{larger} smaller than **MOM(A)** _{at least}

$n/5$ groups, in $\frac{1}{2}$ groups MOM is ≥ 3 elements

\therefore in $n/10$ groups MOM is ≥ 3 elements

\rightarrow MOM is $\geq \frac{3n}{10}$ elems.



Visualizing the median of medians

Also MOM

is $\leq \frac{3n}{10}$ elems.



Selection Algorithm: Take II

$$\underline{T(n): T\left(\frac{n}{5}\right) + T\left(\frac{7n}{10}\right) + cn}$$

$$\underline{T(1): O(1)}$$

\downarrow
MOMSelect(A[1:n], k):

If (n ≤ 25): sort A and return A[k] $O(1)$

Let p = MOM(A)

Partition around the pivot, let r = IndexOf(A, p)

If (k = r): return A[r]

ElseIf (k < r): return MOMSelect(A[1:r-1], k)

ElseIf (k > r): return MOMSelect(A[r+1:n], k-r)

$\left. \begin{array}{l} \text{ElseIf (k < r): return MOMSelect(A[1:r-1], k) \\ \text{ElseIf (k > r): return MOMSelect(A[r+1:n], k-r) \end{array} \right\} T\left(\frac{7n}{10}\right)$

Time of MOM: $\Theta(n) + T(n/5)$



Recursion Tree

$$T(n) = T\left(\frac{7n}{10}\right) + T\left(\frac{2n}{10}\right) + \underline{Cn}$$

$$\underline{T(1) = C}$$

work @ level

$$Cn$$

level

largest piece

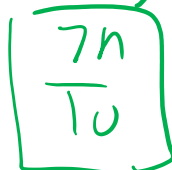
0

n



1

$\frac{7n}{10}$

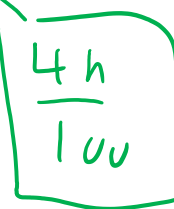
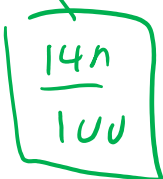


$$C \frac{7n}{10} + C \frac{2n}{10}$$

$$= C \frac{9n}{10}$$

2

$\frac{49n}{100}$



$$C \frac{81n}{100}$$

i

$$\left(\frac{7}{10}\right)^i n$$

$$\left(\frac{7}{10}\right)^i n = 1$$

$$C \left(\frac{9}{10}\right)^i n$$

$$\log_{\frac{10}{7}}(n)$$

$$i = \log_{\frac{10}{7}}(n)$$

$$Cn \sum_{i=0}^{\log_{\frac{10}{7}}(n)} \left(\frac{9}{10}\right)^i = Cn \underline{\underline{C}} = O(Cn)$$



Ask the Audience

- If we change MOM so that it uses $\frac{n}{3}$ blocks of size 3,
how many items can we eliminate?
- What is the new running time of the algorithm?



Selection Wrapup

- Find the k -th largest element in $O(n)$ time
 - Selection is strictly easier than sorting!
- Divide-and-conquer approach
 - Find a pivot element that splits the list roughly in half
 - **Key Fact:** median-of-medians-of-five is a good pivot
- Can sort in $O(n \log n)$ time using same technique
 - Algorithm is called Quicksort
- Analyze running time via recurrence
 - Master Theorem does not apply
- **Fun Fact:** a random pivot is also a good pivot in expectation!



Dynamic Programming



Dynamic Programming

- Don't think too hard about the name
 - *I thought dynamic programming was a good name. It was something not even a congressman could object to. So I used it as an umbrella for my activities. –Richard Bellman*
- Dynamic programming is careful & smarter recursion
 - Break the problem up into small pieces & recursively solve (like Divide & Conquer)
 - Reuse solutions as necessary when subproblems repeat
 - Often the only poly. time algorithm (D&C doesn't work)



- Problems: counting students, stable matching, sorting, n-digit multiplication, array searching, selection
- Alg. techniques: divide & conquer, **dynamic programming**
- Analysis: asymptotic analysis, recursion trees, Master Thm.
- Proof techniques: (strong) induction, contradiction

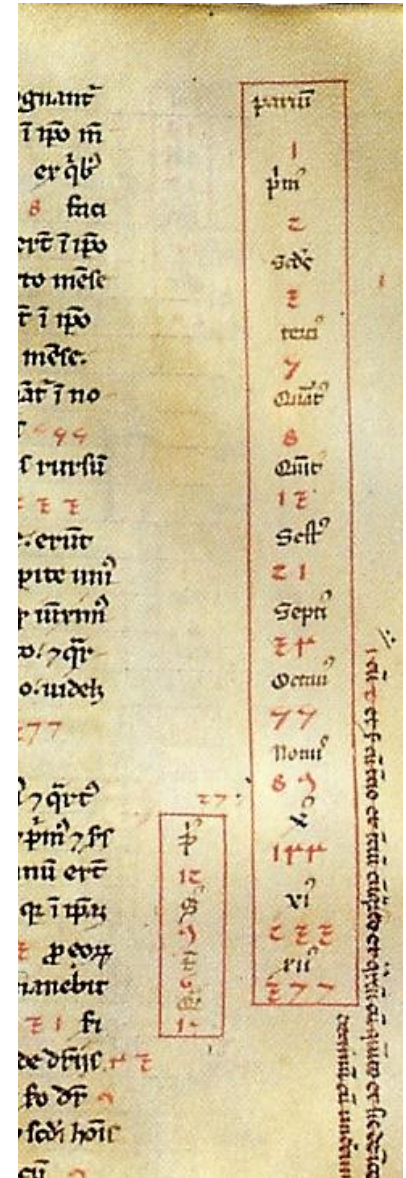


Intro: Fibonacci Numbers



Fibonacci Numbers

- $0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, \dots$
- $F(1) = 0, F(2) = 1,$
 $F(n) = F(n-1) + F(n-2)$
- $F(n) \rightarrow \phi^n \approx 1.62^n$
- $\phi = \left(\frac{1+\sqrt{5}}{2}\right)$ is the **golden ratio**



Fibonacci Numbers: Take I

```
FibI(n) :  
  If (n = 1): return 0  
  ElseIf (n = 2): return 1  
  Else: return FibI(n-1) + FibI(n-2)
```

- How many calls does **FibI(n)** make?
 - $T(n) = \# \text{ of calls by } \text{FibI}(n)$



Fibonacci Numbers: Take II (“Top down”)

```
M ← empty array, M[0] ← 0, M[1] ← 1
FibII(n):
  If (M[n] is not empty): return M[n]
  ElseIf (M[n] is empty):
    M[n] ← FibII(n-1) + FibII(n-2)
    return M[n]
```

- How many recursive calls does **FibII (n)** make?



Fibonacci Numbers: Take III (“Bottom up”)

```
FibIII(n) :  
  M[1] ← 0, M[2] ← 1  
  For i = 3,...,n:  
    M[i] ← M[i-1] + M[i-2]  
  return M[n]
```

- What is the # of loops of **FibIII**(n) ?



Fibonacci Numbers

- 0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, ...
- $F(n) = F(n - 1) + F(n - 2)$
- Solving the recurrence recursively takes $\Omega(1.62^n)$ time
 - Problem: Recompute the same values $F(i)$ many times
- Two ways to improve the running time
 - Remember values you've already computed ("top down")
 - Iterate over all values $F(i)$ ("bottom up")
- **Fact:** Fastest algorithms solve in logarithmic time

