CSE 417 Binary Search (pt 1)

UNIVERSITY of WASHINGTON



Reminders

> HW1 is due next Wednesday

– fixed a typo & added some clarification

> Lecture videos available on Canvas

- student questions cannot be heard
- back of heads can be seen

> Overloading sign up: https://goo.gl/forms/clZu6Wpy3xro69s13

Outline for Today

> Binary search on arrays
> Implementing binary search



Input: sorted array A and a value x

- array can store any ordered set: ints, floats, strings, etc.

Output: index i in [0, A.length] s.t. A[i-1] <= x < A[i]

> (partly vacuously true if i = 0 or i = A.length)

- returns where x would be inserted to maintain ordering
- if x appears multiple times, this returns the *last* one I
 > move "<=" to the right to get the *first* one

Maintain a region of the array that is unexplored



- > Start with all white region
- > Each iteration reduces size of the white region
- > Finish with no white region

> Can reduce the size by looking at any element i:



> Can reduce the size by looking at any element i:



> **Binary Search**: look at the *middle* element of the white region



- > This ensures size is *cut in half* each time
- > Size of white region approximately n / 2^k after k iterations
- > Done after k = lg n iterations
 - O(1) per iteration, so O(log n) time



Linear search algorithm takes O(n) time

- start at i = 0
- increase i by 1 each time
- stop when $A[i] \le x \le A[i+1]$

Binary search is an **exponential speedup**

- … since exp(log n) = n …
- hard to overstate the importance of this



Simple application

> Suppose we want to find the index i such that u A[i] + v = x

- > Q: How do we solve this?
 - cannot compute B[i] = u * A[i] + v
 - that would be exponentially slower!
- > A: Binary search for (x-v) / u since u A[i] + v = x => A[i] = (x - v) / u



Simple application 2

- > Suppose we want to find the index i such that tanh(A[i]) = x
 - (or sigmoid or any other monotonic function)
- > Q: How do we solve this?
- > A: Search for A[i] = tanh⁻¹(x)
- > Q: What if the function is not easily invertible?
 - come back to this later...

- > Problem: find the N-th largest number of the form 2^a3^b5^c for any a, b, c >= 0
 - i.e., numbers not divisible by anything other than 2, 3, or 5
- > **Idea**: generate the numbers *in order*, stopping at N
- > **Sub-problem**: find the (m+1)-st number of this form *given* first m
 - we can use the first m numbers to find the next one

> **Sub-problem 1**: find the (m+1)-st number of this form given first m

- > If (m+1)-st number is 2^a3^b5^c, then 2^{a-1}3^b5^c is smaller and also of this form (assuming a > 0), so 2^{a-1}3^b5^c is **one of the first m** numbers
 - likewise for b and c
 - next number is 2, 3, or 5 times an earlier number
- > In particular, if (m+1)-st is 2 x an earlier number, then it must be the **smallest** 2 x earlier not in the list



- > Sub-problem 2: given a list A[0..m-1] of the first m numbers, in increasing order, find the smallest i such that 2 A[i] > A[m-1]
 - equivalently, $2 A[i] \ge A[m-1] + 1$
- > Q: how do we do that?
- > A: binary search for (A[m-1]+1) / 2

- > Start with A = [1]
- > **Algorithm:** for m = 1 to N-1
 - binary search to find smallest i s.t. 2 * A[i] > A[m-1]
 - ... likewise for 3 and 5
 - add the smallest of these three numbers to the list
- > Q: total running time?
- > A: O(N log N)



> Can we improve this further?

- > Compare smallest i s.t. 2 * A[i] > A[m-1] and smallest j s.t. 2 * A[j] > A[m]
 - binary search for (A[m-1]+1) / 2 and (A[m]+1) / 2
 - these two indices should be close
- > Seems like we're doing too much work by
 - could restrict to A[i..m-1] with i from last search
 - but we can do better...



- > Compare smallest i s.t. 2 * A[i] > A[m-1] and smallest j s.t. 2 * A[j] > A[m]
- > **Observation**: they must be equal or differ by 1
- > Since A[m] is smallest of 2x, 3x, 5x... either A[m] = 2 * A[i] or A[m] < 2 * A[i]
 - if A[m] < 2 * A[i], then j = i since i is still big enough</p>
 - if A[m] = 2 * A[i], then A[m] < 2 * A[i+1] and j = i+1</p>



- > Algorithm 2: maintain indexes of smallest i, j, k s.t. 2 A[i], 3 A[j], 5 A[k] > A[m-1]
 - for m = 1 to N-1
 - add the smallest of 2 A[i], 3 A[j], 5 A[k] to the list
 - increment i, j, and/or k appropriately
- > Q: total running time?
- > A: O(N)



- > An example of how the fastest algorithm can be produced by starting with a basic technique + a lot of elbow grease
- > Binary search disappears in the final answer even though we used to get there
 - we will see other examples of this
 - presentations of the best algorithm will often just describe the optimal solution directly without any binary search
 - > makes you sound smarter if you do it that way



- > Find x in a sorted n x n table
 - every row and every column is sorted
- > **Algorithm 1**: binary search every row
 - O(n log n) time





- > Find x in a sorted n x n table
 - every row and every column is sorted
- > Once again, the binary searches are doing too much work
 - usually the next answer is about the same as the previous one
- > Not the case that the indexes only increase by 0 or 1 each time!
- > BUT the total increase over n searches is at most n:
 - we only move left each time
 - so we can take at most n steps all together





- > Find x in a sorted n x n table
 - every row and every column is sorted
- > **Algorithm 1**: linear every row starting from answer on previous
 - O(n) time
- > Again, binary search disappears

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- > Find x in a sorted n x n table
 - every row and every column is sorted
- > **Algorithm 1**: linear every row starting from answer on previous
 - O(n) time
- > Worth noting: O(n log n) is not much slower than O(n)
 - remember that log n is exponentially smaller than n
- > Worth pointing out O(n log n) algorithm in an interview



Practical example: web search

- > For each word, make record of all the web pages with that word
 - many terabytes of data
 - too much data for one machine...
- > Partition web pages randomly across machines
- > Each machine records where word appears in its own pages
 - have enough machines that each gets, say, 100 GB of data
- > To look up all pages where word occurs:
 - send request to all machines
 - concatenate the lists they return

Practical example: web search

- > Q: How does each machine get the pages for a word?
- > A: sorting and binary search
 - each machine sorts its records on disk
 - look up a word by using binary search
- > Algorithm works fine if A is on disk
 - only need the ability to look up A[i] for any i
 - can do this in Java using FileChannel instead of FileInputStream

> Cost is time for lg n disk seeks



Practical example: web search

- > **Key lesson**: not always necessary to use dynamic data structures
 - don't always need hash tables and AVL trees (or B+ trees on disk)
 - they are more complex, slower, and user more memory (by constant factors)
- > Only need them to support intermixed updates and searches
 - sorting and binary search are fine if all the updates come first
- > Sorting also works fine if the data only changes occasionally
 - at one point, web indexes were only changed *nightly*
 - > this is not uncommon in practice
 - can add new data and re-sort at night when not in use



Outline for Today

> Binary search on arrays

> Implementing binary search 🦛



Implementing Algorithms

Key Idea: invariants — facts that are always true

- critical to correct implementation (and often run time analysis also)
- often most of the hard work is getting these right
- > **Rep invariant**: claim about data structures
 - always true (except briefly when mutating the data structures)
 - ex: in AVL trees, heights of two subtrees at any node differ by at most 1
- > **Loop invariant**: claim about method state
 - always true at the top of the loop



Implementing Algorithms

- > **Loop invariant**: claim about method state
 - always true at the *top* of the loop
- > To prove that a loop is correctly implemented, check:
 - invariant is true initially
 - invariant remains true each time the loop body executes
- > Then know the invariant is true after the loop
 - choose the invariant so that, when the loop exits, you have enough information to return a correct answer



> Method state: indexes i and j

> Loop invariant: A[0], ..., A[i-1] <= x and x < A[j], ..., A[n-1]



Notes on notation:

- A[0], ..., A[i-1] <= x means A[0] <= x and ... and A[i-1] <= x</p>
- vacuously true if i <= 0
 - > only making claims about indexes >= 0 and <= i-1

int i = 0, j = n;

true initially

```
// Invariant: A[0], ..., A[i-1] <= x and x < A[j], ..., A[n-1]
while (i < j) {
    int m = (i + j) / 2;
    if (A[m] <= x)
        i = m + 1;
    else
        j = m;
}</pre>
```



int i = 0, j = n;

// Invariant: A[0], ..., A[i-1] <= x and x < A[j], ..., A[n-1]
while (i < j) {
 int m = (i + j) / 2;
 if (A[m] <= x)
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 A[i-1] = A[m] <= x
 else
 j = m;
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int i = 0, j = n;

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int i = 0, j = n;

// Invariant: A[0], ..., A[i-1] <= x and x < A[j], ..., A[n-1]
while (i < j) {
 int m = (i + j) / 2;
 if (A[m] <= x)
 i = m + 1;
 A[i-1] = A[m] <= x
 else
 j = m;
 A[i-1] = A[m] <= x
 Q: What about A[i], ..., A[m-1]?
 A: Also <= x since A is sorted
}</pre>

int i = 0, j = n;



int i = 0, j = n;



// Invariant: A[0], ..., A[i-1] <= x and x < A[j], ..., A[n-1]
while (i < j) { ... }</pre>

- > When we exit the loop, we have
 - i = j and A[0], ..., A[i-1] <= x and x < A[j], ..., A[n-1]</p>
 - > (actually, we only obviously have i >= j, but a more careful check shows i = j)
 - in other words, A[0], ..., A[i-1] <= x and x < A[i], ..., A[n-1]</p>
 - thus, the problem specification says i is exactly what we promised to return
 - > this is not uncommon...
 - > loop invariants are often "weakened" versions of output promise



> When we exit the loop, we return the right answer

- > Q: Do we actually exit the loop?
 - usually you get that from the run time analysis,
 - but we did this somewhat sloppily....
- > A: Yes, provided that i increases or j decreases every time.
 > Q: Do they?

- > We set m = (i + j) / 2
 - this means that i <= m <= j</p>
- > If we set i = m + 1, then $i_{new} \ge i_{old} + 1$
 - looks good!
- > If we set j = m, then $j_{new} \le j_{old}$ - we are in trouble if $m = j_{old}!$



- > Can we have m = j when we set m = (i + j) / 2?
- > Can see that m will be closer to j when i is closer to j... so consider i = j – 1 (the worst case)
- > Then m = (j 1 + j) / 2 = (2j 1) / 2...
- > In Java, this integer division will **truncate** to j 1
 - we got lucky!
- > But Math.round((i + j) / 2.0) could loop forever!



Lessons:

- 1. invariants are critical to implementing complex algorithms & data structures
- 2. implementing algorithms correctly requires careful attention to detail
 - > easy to make mistakes on this, one of the easiest algorithms we will see!
 - > this comes up in interviews too
- 3. if a library implementation is available, use it!
 - > don't waste your time or risk releasing buggy code



- > For the most part, we will stick to pseudocode from here on
 - you'll still need to write code in the HWs
- > From a theory perspective, this wasn't a real problem
 - we only run into it when j i is small
 - we could switch to linear search when j i < 1000 asymptotic complexity would be the same

