Divide and Conquer: Order Statistics

Lecture 6

Akshar Varma 13th July, 2023

CS3000 Algorithms and Data

Order Statistics

Good and Bad Pivots

Median-of-medians: MoMs

Quicksort: straightforward, MoMs, [OPTIONAL: randomized]

Two Data Structures: Priority Queues and Binary Search Trees

1. Order Statistics

"Chaos is a ladder." –G. R. R. M.

- Order statistics, as the name implies are statistical values that characterize order.
- You've heard of minimum and maximum; special cases of order statistics.
- The k^{th} order statistic of a collection is the k^{th} smallest value in it.
- Minimum is k = 1, maximum is k = n (where n is the size of the collection).
- You may also have heard of other special order statistics:
 - Quartile: 0.25n, 0.5n, 0.75n
 - Decile: $0.1n, 0.2n, 0.3n, \ldots, 0.8n, 0.9n$
 - Percentile: 0.1n, 0.33n, 0.68n, 0.95n, 0.99n, etc.
 - Median: $\lfloor \frac{n}{2} \rfloor$
- This has extensive applications in statistics and inference.
- We'll see the $1^{st}/n^{th}$ order statistic (minimum/maximum) show up a lot in the future.

- k = 1 (minimum) and k = n (maximum) can both be found in $\Theta(n)$ time.
- Note: k-selection is $\Omega(n)$ since we must look at all the data.
- Using $\Theta(k)$ space, we can do k-selection in $\Theta(n)$ time.
- Without space usage, we can find minimums repeatedly and finish in $\Theta(kn)$ time.
- All of the above assumed unordered collection/list/array.
- If sorted, then k-selection is doable in $\Theta(1)$ for any value of k (just index into array).
- Spending $O(n \log n)$ on sorting will allow k-selection in $\Theta(1)$.
- Can we do selection without sorting and in $\Theta(n)$ for all k?

- \cdot Pick a pivot element p. This can be any element, for example, the first one.
- Split array into everything < p (*Left*), everything = p (*Middle*), everything > p (*Right*).
- This is $\Theta(n)$.
- Let ℓ, m, r be the sizes of Left, Middle and Right.
- If $k \leq \ell$ recurse in *Left*.
- + If $\ell < k \leq \ell + m$, elements in the Middle, that is, p is the k^{th} order statistic.
- Otherwise, recurse on *Right*, to find k'^{th} order statistic where $k' = k \ell m$.

• Runtime: $T(n) = T(\max(\ell, r)) + \Theta(n)$.

2. Good and Bad Pivots

"Pivoting isn't Plan B; it is part of the process."

- + ℓ and r depend on how large p vs. other elements, that is, its order.
- If it is too small or too large, then $\max(\ell, r)$ may be $\Omega(n)$.
- In that case, the recursion does not give a $\Theta(n)$ runtime.
- $\cdot\,$ So we need to ensure that p is close to the median.
- Pivoting needs $\Omega(n)$ time irrespective of p.
- So we only have O(n) time to find a pivot p close to the median.

3. Median-of-medians: MoMs

"There are 2 hard problems in computer science: cache invalidation, naming things, and off-by-1 errors." – Leon Bambrick

Median of Medians Algorithm

- \cdot We don't need the exact median, only something close.
- $\cdot\,$ Sort every group of 5 elements (sorting is constant time), $\Theta(n)$ such sortings.
- Recurse with these $\lfloor \frac{n}{5} \rfloor$ medians. When ≤ 5 elements return true median.
- Once a median-ish pivot is found, continue with your k-selection.



Median of Medians guarantee

• k-selection recurrence becomes $T(n) = T\left(\frac{n}{5}\right) + T\left(\frac{7n}{10}\right) + \Theta(n) \implies T(n) = \Theta(n).$

4. Quicksort: straightforward, MoMs, [Optional: randomized]

"There are 2 hard problems in computer science: cache invalidation, naming things, and off-by-1 errors." – Leon Bambrick

- Pivoting also suggests a sorting algorithm.
- Perform pivoting, recurse on left and right halves.
- + Same problems as before wrt "good pivot". Worst-case complexity is $\Theta(n^2)$ (exercise).
- Resort to MoM to get a good pivot. Show that in this case runtime is $\Theta(n \log n)$.
- Alternative: Randomness!

Pick a random element as the pivot. Chance that it is bad is low.

- We won't do a proof, but this also gives a runtime of $\Theta(n\log n)$.
- In practice, due to large constants in MoM, randomized Quicksort is commonly used.

5. Two Data Structures: Priority Queues and Binary Search Trees

Two Data Structures: Priority Queues and Binary Search Trees

1. PRIORITY QUEUES:

- Keeps track of things in order of "priority".
- + $\Theta(1)$ access to the min or max priority element.
- Creation of data structure: O(n)
- + Deletion $O(\log n)$
- 2. (BALANCED) BINARY SEARCH TREES:
 - Keeps track of things in a binary tree with the following property: Left subtree has smaller values and right subtree has larger values.
 - Insertion, Deletion, Search, etc. are all $O(\log n)$ operations.
 - Many other things can be done.
 - Inorder, preorder, postorder traversals.
 - Depth and height of nodes.
 - Nodes can track subtree size to allow indexing and order statistics.