

Lecture 5

CS325

Problem: Selection

Problem definition:

Input: A list of numbers A and integer k

Output: the k -th smallest number of A

For simplicity, we can assume all elements are distinct

A generalization of the **median** problem, which finds the 50th percentile of S .

Immediate solution for $O(n \log n)$?

Can we do solve it in $O(n)$?

A recursive solution: rough idea

- Randomly select an element (pivot)



- Partition** the array according to q



- Recurse to A_L or A_R based on r and k
 - If $k < r$, recurse on $(A_L, ?)$
 - If $k = r$, ?
 - If $k > r$, recurse on $(A_R, ?)$

Pseudocode

```
Partition(A[1,...,n], q)
  i=1, j=n
  for k=1 to n except v:
    if A[k]<A[v]:
      B[i] = A[k], i++
    else:
      B[j] = A[k], j--
  B[i] = A[v]
  copy B to A
  return i
```

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```

```
Select(A[1,...,n], k)
  v = a random # in {1,...,n}
  r = partition(A, v)
  if r=k:
    return A[r]
  if r<k:
    Select(A[1,...,r-1],k)
  if r>k:
    Select(A[r+1,...,n],k-r)
```

Runtime analysis

$$T(n) = O(n) + \max\{1, T(r - 1), T(n - r)\}$$

$$= O(n) + \max\{T(r - 1), T(n - r)\}$$

where r is a random number in $\{1, \dots, n\}$

Good case: $r = \frac{n}{2}$

Bad case: $r = 1$ or $n - 1$

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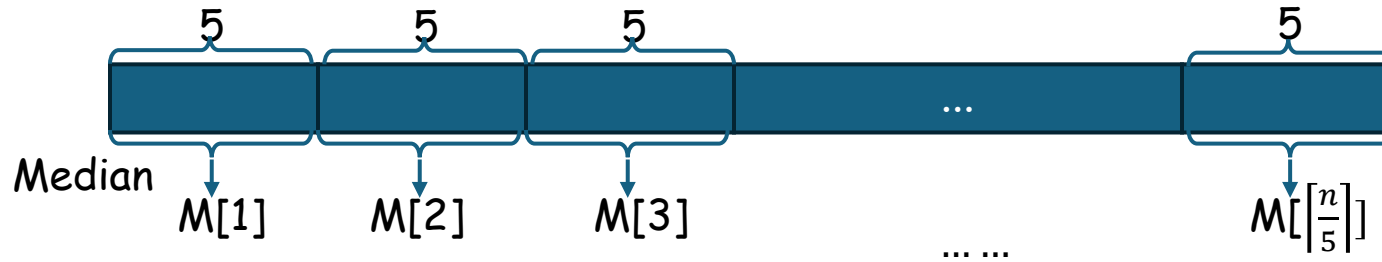
$$T(n) = O(n)$$

Bad case: $r = 1$ or $n - 1$

$$T(n) = O(n^2)$$

Smart* selection of pivot

- Break A into chunks of fives and find the median of each chunk



- Use the median of M to be the pivot

• Claim:

Median of M is larger than $\frac{3}{10}$ of A and smaller than $\frac{3}{10}$ of A



Select(A, k)

1. for $i=1, \dots, n/5$
2. $M[i] = \text{median}(A[5(i-1)+1, 5i])$
3. $MM = \text{Select}(M, n/10)$
4. $v = \text{index of } MM \text{ in } A$
5. $r = \text{partition}(A, v)$
6. if $k=r$: return $A[r]$
7. if $k < r$: **Select**($A[1, \dots, r-1]$, k)
8. if $k > r$: **Select**($A[r+1, \dots, n]$, $k-r$)

$$\begin{aligned} T(n) &\leq O(n) + T\left(\frac{n}{5}\right) + T\left(\frac{7}{10}n\right) \\ &\Rightarrow T(n) = O(n) \end{aligned}$$

Practical considerations

- The “median of median” method has a better theoretical runtime – worst case $O(n)$
- Practically it is not really used
 - The overhead of identifying the median of median is generally too big for it to be practically useful
 - The original randomized algorithm has average runtime $O(n)$ --- see DPV discussion on this
- Similar idea is behind the popular quicksort algorithm
- **Partition** can be implemented in place (without requiring additional memory allocation)
 - See figure 1.8 here for the algorithm
<https://jeffe.cs.illinois.edu/teaching/algorithms/book/01-recursion.pdf#page=9.10>

Example Problem: D & C

Input: an array A of sorted integers that have been shifted.

Goal: find the largest element in A

Example:

(40, 57, 89, 2, 8, 25, 30)
shifted 3 positions

Brute force?

Can we do better?

Goal: $O(\log n)$ run time

Q: how can we achieve $O(\log n)$ time?

A: Each recursion

- do some constant time operation
- shrink the input size by a constant factor. e.g. to $n/2$ like binary search

Rough idea:

(40, 57, 89, 2, 8, 25, 30)

- Identify the middle element,
- Do constant time operation on it
- Eliminate half of the array