

Math 261C: Randomized Algorithms

Lecture topic: Floyd-Rivest Median Selection

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1. IMPROVED MEDIAN SELECTION: THE ALGORITHM

As sketched at the end of yesterday's lecture, we can improve our randomized median or (more generally) k^{th} element selection algorithm by skewing the pivot distribution towards elements that are close to the k^{th} element. The idea is to pick \sqrt{n} elements at random and pivot using the k^{th} element of this subset, which should be close to the k^{th} element of A . This algorithm is originally from [BFP⁺73] and we follow the treatment from [Kiw05]. See [Pat96] for a survey of median finding.

Some notation and parameters:

- a_i^* is the i^{th} sorted element of A , and we
- s is the “sample size,” the number of potential pivot points we sample from A
- g is the “gap” size

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Data:  $A, n, k$ 
Result: the  $k^{th}$  element of  $A$ 
if  $n = 1$  then
  | return  $a_0$ ;
else
  | Choose  $S \subset A$  of size  $s$  uniformly at random without replacement ;
  |  $j_u = \max\{k \cdot \frac{s}{n} - g, 0\}$ ;
  |  $j_v = \min\{k \cdot \frac{s}{n} + g, 0\}$ ;
  | ;
  |  $u = \text{FR-Select}(S, s, j_u)$ ;
  |  $v = \text{FR-Select}(S, s, j_v)$ ;
  | ;
  | /* Scan  $A$  sequentially to partition it as below                               */
  |  $\{u\}$ ;
  |  $\{v\}$ ;
  |  $L = \{a_i < u\}$ ;
  |  $M = \{u < a_i < v\}$ ;
  |  $U = \{a_i > v\}$ ;
  | ;
  | if  $|L| = k$  then
  | | return  $u$ ;
  | else if  $|L| + |M| + 1 = k$  then
  | | return  $v$ ;
  | else if  $|L| > k$  then
  | | return  $\text{FR-Select}(L, |L|, k)$ ;
  | else if  $(|L| + |M| + 1) > k$  then
  | | return  $\text{FR-Select}(M, |M|, k - |L| - 1)$ ;
  | else
  | | return  $\text{FR-Select}(U, |U|, k - (|L| + |M| + 2))$ ;
  | end
end

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Algorithm 1: The Floyd-Rivest k^{th} element selection algorithm

2. THE ANALYSIS

As before, we will measure our runtime by expected number of comparison operations. The runtime will be dominated by scanning A sequentially, so the algorithm is at least linear. Further, without lose of generality, we can assume that $k < \frac{n}{2}$, because we can reverse A to arrange for this. We require some notation, let:

i_u be the index of u such that $u = a_{i_u}^*$
 i_v be the index of v such that $v = a_{i_v}^*$

The algorithm compares first against u , and then against v if necessary.

We claim:

$$\begin{aligned} \mathbb{E}[\# \text{ comparisons}] &= n + i_v + (\# \text{ recursive calls}) \\ &\leq n + \frac{n}{2} + o(n) + (\# \text{ recursive calls}) \\ &= n + \frac{n}{2} + o(n) \end{aligned}$$

To begin, we will prove:

$$i_v \leq \frac{n}{2} + o(n)$$

The intuition here is that $i_v \approx j_v \cdot \frac{n}{s} \approx k + g \cdot \frac{n}{s}$, and $i_u \approx k - g \cdot \frac{n}{s}$. Specifically, we want to show that the following hold with high probability:

$$k - 2\frac{gn}{s} \underbrace{\leq}_{(1)} i_u \underbrace{\leq}_{(2)} k \underbrace{\leq}_{(3)} i_v \underbrace{\leq}_{(4)} k + 2\frac{gn}{s}$$

Note that:

- if (2) holds, we don't call FR-Select recursively on L
- if (2) and (3) hold, this implies that $|M| = o(n)$
- if (3) holds, we don't call FR-Select recursively on U
- if (4) holds, we have $i_v \leq \frac{n}{2} + o(1)$

Setting $s = \sqrt{n}$ and $g = n^{1/3}$, we have:

$$\begin{aligned}
|M| &= i_v - i_u - 1 \\
&\leq 2 \cdot \frac{gn}{s} \\
&= 2n^{1/3} \cdot \frac{n}{n^{1/2}} \\
&= 2n^{5/6} \\
&= o(n)
\end{aligned}$$

Lemma 1. *Prob[(3) fails] is $o(1)$*

Will prove Lemma 1 above, but similar arguments show that (1), (2), (4) and (5) also fail with probability $o(1)$.

Proof. Suppose $k > i_v$, then $v = j_v^{\text{th}}$ element of S and $v = a_{i_v}^*$, the i_v^{th} element of A in sorted order. Then $k > i_v$ iff the j_v^{th} element of S is greater than the k^{th} element of A , which occurs iff S has more than j_v many elements selected from the first k elements of A in sorted order.

We can describe the event above using a balls and urns model. The balls are members of A , and so there are n total balls. The red balls are $\{a_i : a_i < a_k^*\}$, and thus there are $\frac{k}{n}$ red balls total. To obtain the set S , we draw s balls from the urn. The *bad* event is that $> j_v$ of the balls drawn are red. Let's obtain an expression for j_v in terms of useful quantities:

$$\begin{aligned}
j_v &= k \cdot \frac{s}{n} + g \\
&= \left(k + \frac{gn}{s}\right) \frac{s}{n} \\
&= \frac{k}{n}
\end{aligned}$$

Now, we use the following lemma, which improves Chernoff bounds for a balls-and-urn model:

Lemma 2 (Chvatal Chernoff Improvement [Chv79]). *If N balls have pN red with or without replacement, and M balls are drawn, then:*

$$\text{Prob}[> (p + t)M \text{ balls in sample are red}] \leq e^{-2t^2M}$$

We apply the bounds with: $M \leftarrow s$, $p \leftarrow \frac{k}{n}$, and $t \leftarrow \frac{g}{s}$, so:

$$\begin{aligned}
\text{Prob}[(3)\text{ fails}] &\leq e^{-2t^2M} \\
&= e^{-2(g/s)^2s} \\
&= e^{-2(g^2/s)} \\
&= o(1) \qquad \text{by } s = n^{1/2} \text{ and } g = n^{1/2}
\end{aligned}$$

□

We write out the runtime $T(n)$ in number of comparisons:

$$\begin{aligned}
T(n) &\leq n + \frac{n}{2} + o(1) \\
&\quad + \text{Prob}[L \text{ or } U \text{ is recursed on}] \cdot T(\max\{|U|, |L|\}) \\
&\quad + \text{Prob}[M \text{ is recursed on}] \cdot T(|M|) \\
&\quad + T(\sqrt{n}) \\
&\leq \frac{3}{2}n + o(n) + o(1) + T(n) + O(2n^{5/6}) \\
&\leq \frac{3}{2}n + o(n)
\end{aligned}$$

This completes our argument.

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