

# HW #4 - Problem 2

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## Problem 2

The recurrence relation of quick select is given by,

$$\mathbb{E}[X_n] = \frac{1}{n} \sum_{i=1}^{k-1} (\mathbb{E}[X_{n-i}] + An) + \frac{1}{n} \sum_{i=k+1}^n (\mathbb{E}[X_{i-1}] + An) + \frac{1}{n} An$$

Above, the first summation accounts for the case when  $i < k$ . In this case the algorithm partition to the right contains the  $k^{th}$  smallest number and is checked. The second summation accounts for the case when  $i > k$ . In this case the left partition is further analyzed. The final term  $(\frac{1}{n}An)$  accounts for the case when  $i == k$ . In this case the  $k^{th}$  smallest number has been found. Each term in the two summations and this final term  $(\frac{1}{n}An)$  each have an equal  $\frac{1}{n}$  chance of occurring.

Before I prove by induction that this is bounded by  $Cn$  I will first simplify this recurrence relation and show it is that given in the homework sheet.

$$\begin{aligned} \mathbb{E}[X_n] &= \frac{1}{n} \sum_{i=1}^{k-1} (\mathbb{E}[X_{n-i}] + An) + \frac{1}{n} \sum_{i=k+1}^n (\mathbb{E}[X_{i-1}] + An) + \frac{1}{n} An \\ &= \frac{1}{n} \sum_{i=1}^{k-1} \mathbb{E}[X_{n-i}] + \frac{1}{n} \sum_{i=k+1}^n \mathbb{E}[X_{i-1}] + \frac{1}{n} \sum_{i=1}^n An + A \\ &= \frac{1}{n} \sum_{i=1}^{k-1} \mathbb{E}[X_{n-i}] + \frac{1}{n} \sum_{i=k+1}^n \mathbb{E}[X_{i-1}] + An + A \\ &= \frac{1}{n} \sum_{i=1}^{k-1} \mathbb{E}[X_{n-i}] + \frac{1}{n} \sum_{i=k+1}^n \mathbb{E}[X_{i-1}] + n + 1 & A=1 \\ &= \frac{1}{n} \sum_{i=1}^{k-1} \mathbb{E}[X_{n-i}] + \frac{1}{n} \sum_{i=k+1}^n \mathbb{E}[X_{i-1}] + n & O(n+1)=O(n) \end{aligned}$$

This is the recurrence relation given in the homework sheet. This is the one that will be used in the following inductive proof.

*Proof by Induction*

**Induction Hypothesis:**  $\mathbb{E}[X_k] \leq Cn$  for all  $k < n$  where  $X_n$  is a r.v. corresponding to the running time of quick select on an array of size  $n$ .

**Base cases:**

$n=1$

$$\mathbb{E}[X_1] = A \leq 1 * A$$

In the case when there is only one element in the array the running time is constant. This is due to the fact that only simple comparisons are needed. If  $n = 1$  the only valid value of  $k$  is 1. In other words, if the array is only of size 1, only the smallest number can be requested, **not** the second smallest number since there is only one number in the array. Thus, the algorithm only needs to compare  $k$  to one. If  $k$  is indeed one the algorithm returns the only element in the array in constant time, else it throws an error. Thus, if we call this constant return time  $A$ , a  $C$  can be chosen,  $C = A$  more specifically, such that  $\mathbb{E}[X_k] \leq Cn$  holds.

A similar argument for the case when  $n = 0$  can also be made. In this case, when  $n = 0$ , the algorithm returns in constant time because there is no items in the array to return.

### Inductive step:

Let  $n > k$

$$\begin{aligned}
\mathbb{E}[X_n] &= \frac{1}{n} \sum_{i=1}^{k-1} \mathbb{E}[X_{n-i}] + \frac{1}{n} \sum_{i=k+1}^n \mathbb{E}[X_{i-1}] + n & * \\
&= \frac{1}{n} \sum_{i=1}^{k-1} \mathbb{E}[X_{n-i}] + \frac{1}{n} \sum_{i=k+1}^n \mathbb{E}[X_{i-1}] + n & ** \\
&\leq \frac{C}{n} \sum_{i=1}^{k-1} (n-i) + \frac{C}{n} \sum_{i=k+1}^n (i-1) + n \\
&\leq \frac{C}{n} (n \sum_{i=1}^{k-1} 1 - \sum_{i=1}^{k-1} i + \sum_{i=k+1}^n i - \sum_{i=k+1}^n 1) + n \\
&\leq \frac{C}{n} (n(k-1) - \frac{(k-1)(k)}{2} + \sum_{i=1}^n i - \sum_{i=1}^k i - (n-k)) + n \\
&\leq \frac{C}{n} (n(k-1) - \frac{(k-1)(k)}{2} + \frac{(n)(n+1)}{2} - \frac{(k)(k+1)}{2} - n + k) + n \\
&\leq \frac{C}{n} (nk - 2n - \frac{(k-1)(k)}{2} + \frac{(n)(n+1)}{2} - \frac{(k)(k+1)}{2} + k) + n \\
&\leq \frac{C}{2n} (2nk - 4n - (k-1)(k) + (n)(n+1) - (k)(k+1) + 2k) + n \\
&\leq \frac{C}{2n} (2nk - 4n + 2k - k^2 + k + n^2 + n - k^2 - k) + n \\
&\leq \frac{C}{2n} (2nk - 3n + 2k - 2k^2 + n^2) + n
\end{aligned}$$

\* By recurrence relation

\*\* By Induction Hypothesis

At this point we have proven

$$\mathbb{E}[X_n] \leq \frac{C}{2n} f(k) + n$$

where  $f(x) = 2nk - 3n + 2k - 2k^2 + n^2$ .

If  $\mathbb{E}[X_n]$  is strictly less than a function for all values of  $k$  it must also be less than that function when it is at its maximum. We can find this maximum by finding the critical point of  $f(k)$ . This is equivalent to thinking of the running time of the algorithm in the worst case, when  $k$  has been chosen to maximize the expected running time.

*Critical Points of  $f(k)$*

$$f(k) = 2nk - 3n + 2k - 2k^2 + n^2$$

$$f'(k) = 2n + 2 - 4k$$

Setting  $f'(k)$  equal to zero to find the critical points yields,

$$2n + 2 - 4k = 0$$

$$4k = 2n + 2$$

$$k = \frac{n+1}{2}$$

Since  $f(k)$  is a downward facing parabola, this critical point must be a maximum. Plugging this expression of  $k$  into  $\mathbb{E}[X_n]$  gives,

$$\begin{aligned} \mathbb{E}[X_n] &\leq \frac{C}{2n}(n^2 - 2(\frac{n+1}{2})^2 + n(\frac{n+1}{2}) - 3n + 2n\frac{n+1}{2}) + n \\ &\leq \frac{C}{2n}(n^2 - \frac{n^2+2n+1}{2} + n + 1 - 3n + n^2 + n) + n \\ &\leq \frac{C}{4n}(2n^2 - n^2 - 2n - 1 + 2n + 2 - 6n + 2n^2 + 2n) + n \\ &\leq \frac{C}{4n}(3n^2 - 4n + 1) + n \\ &\leq \frac{3Cn}{4} + n - C + \frac{C}{4n} \end{aligned}$$

Since  $n$  and  $C$  are positive numbers  $-C + \frac{C}{4n}$  is a negative quantity. This means  $\frac{3Cn}{4} + n - C + \frac{C}{4n} < \frac{3Cn}{4} + n$ . Therefore,

$$\mathbb{E}[X_n] \leq \frac{3Cn}{4} + n$$

Now it needs to be shown that there is some value of  $C$  such  $\mathbb{E}[X_n]$  is bounded by  $Cn$ . More specifically, a value of  $C$  greater than needs to be found such that the following holds:

$$\mathbb{E}[X_n] \leq \frac{3Cn}{4} + n \leq Cn$$

Any value of  $C \geq 4$  makes this above inequality hold. The computation for this is shown below.

$$\frac{3Cn}{4} + n \leq Cn$$

$$n(\frac{3C}{4} + 1) \leq Cn$$

$$\frac{3C}{4} + 1 \leq C$$

$$1 \leq C - \frac{3C}{4}$$

$$1 \leq \frac{1}{4}C$$

$$4 \leq C$$

$$C \geq 4$$

Let's chose  $C = 4$ . By induction we have  $\mathbb{E}[X_n] \leq 4n = O(n)$