

CSOR4231 Algorithm

HW2

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1 Problem I

1.1 Show that when $p = 1/n$, the probability our sample is good is larger than some positive constant (independent of n).

According to the description of the question, the probability of $X_1 = [\text{R contains at least 1 element of T}]$ as P_1 . is exactly the same as 1 minus the probability of the opposite event $[\text{R contains 0 element of S}]$:

$$P_1 = 1 - \left(\frac{n-1}{n}\right)^n$$

The probability of $X_2 = [\text{R contains no element of S}]$ as P_2 is :

$$P_2 = \left(\frac{n-1}{n}\right)^n$$

Because X_1 and X_2 is independent events, the probability of the sample is good should be :

$$P = P_1 * P_2 = \left(1 - \left(\frac{n-1}{n}\right)^n\right) * \left(\frac{n-1}{n}\right)^n = (1 - a_n) * a_n$$

where

$$1 > a_n = \left(\frac{n-1}{n}\right)^n > 0 \quad (\text{when } n > 1)$$

$$\frac{\partial \left(\frac{n-1}{n}\right)^n}{\partial n} = \frac{\left(1 - \frac{1}{n}\right)^{n-1}}{n} > 0 \quad (\text{when } n > 1)$$

from which we know that the $a_n(n=2) = \frac{1}{4}$ is the min value of a_n
According to a knowledge of $\lim_{n \rightarrow \infty} \left(1 + \frac{x}{n}\right)^n = e^x$, $\lim_{n \rightarrow \infty} \left(1 - \frac{1}{n}\right)^n = e^{-1}$, so the tight upper bound of a_n is e^{-1} . So a_n has a range: $[\frac{1}{4}, e^{-1})$

Since $P = (1 - a_n) * a_n$ is increasing before $a_n = 1/2$ and decreasing after $a_n = 1/2$, because $1/4 < e^{-1} < 1/2$, the min value of P happens when $a = 1/4$, $P = 3/16$

$$P = (1 - a_n) * a_n \geq \frac{3}{16} \quad (\text{for } n > 1)$$

Then we have the statement that the probability our sample is good is larger than some positive constant, for example, $3/16$.

2 Problem II

2.1 What is the running time of this algorithm?

Outside the for loop: The algorithm spends constant time $O(1)$.

Inside the for loop: The algorithm spends $O(n)$ time.

Then the running time of the algorithm is $O(n)$.

2.2 What kind of algorithm is Randomized Approximate Median and why? What is the success probability of this algorithm?

It is a randomized algorithm.

Because although the input is fixed, the algorithm randomly selects an item at the beginning and the output may be different. So it's a randomized algorithm.

The success event is to uniformly select an element at random and the element neither belongs to the smallest $1/4$ nor the largest $1/4$ of all the elements.

So the success probability is $1/2$.

2.3 How can you improve the success probability of the algorithm to over 99%? What is the running time of the new algorithm?

Pseudocode:

Algorithm 1 Function Advanced-RAM(S)

Select ($k = \log n$) items into set $A = b_1 \dots b_{\log n} \in S$ uniformly at random

Sort them by merge-sort.

Select the $(k/2)$ th item a_i from the set as the median element

return a_i

Success probability(Correctness):

Randomized algorithm should show its correctness by the success probability P .

This algorithm succeeds unless the median of the k items belongs to the smallest (or the biggest) $1/4$ items which equals to $P_{fail} = P[\text{more than half of the } k \text{ items belongs to the smallest (or the biggest) } 1/4 \text{ items}]$.

Since we know each element is selected by probability of $1/n$ and the probability of the selected number belonging to $1/4$ part of S is $1/4$:

$$\begin{aligned}
P_{fail}/2 &= \sum_{i=k/2}^k P[i \text{ of } k \text{ belongs to the smallest part}] \\
&= \sum_{i=k/2}^k (1/4)^i * (3/4)^{k-i} * \binom{k}{i} \\
&= \sum_{i=k/2}^k (1/4)^i * (4/3)^i * (3/4)^k * \binom{k}{i} \\
&\leq \binom{k}{k/2} * (3/4)^k * \sum_{i=k/2}^k (1/3)^i \tag{1} \\
&\leq \binom{k}{k/2} * (3/4)^k * (1/3)^{\frac{k}{2}} * \frac{3}{2} \quad \text{since } \sum_{i=k/2}^k (1/3)^i = \frac{(1/3)^{\frac{k}{2}}(1 - 1/3^{\frac{k}{2}})}{1 - 1/3} \leq (1/3)^{\frac{k}{2}} * \frac{3}{2} \\
&\leq 2^{k/2} * (3/4)^k * (1/3)^{k/2} \quad \text{since } \binom{k}{k/2} = \frac{(k!)}{(\frac{k}{2}!)^2} \leq 2^{k/2} \\
&= 3^{k/2} * 2^{-3k/2} \\
&\leq 2^{-k} = 2^{-\log n} = \frac{1}{n}
\end{aligned}$$

So the success probability $P = 1 - P_{fail}$ where $P_{fail} = \frac{2}{n}$
When n is very large (larger than 200), $P > 99\%$

Running time:

The algorithm takes $O(k)$ time to select and $O(k \log k)$ time to merge-sort.
That is $O(\log n(\log \log n))$ time in total.

Space complexity:

The new assigned space is $\Theta(\log n)$
So the space complexity is $\Theta(\log n)$

3 Problem III

3.1 Suppose that in some round we have $k = \varepsilon n$ balls. At most how many balls should you expect to have in the next round?

The probability of a ball get discarded in an k -ball round is $P_f = \binom{n}{1} * \frac{1}{n} * (1 - \frac{1}{n})^{k-1} = (1 - \frac{1}{n})^{k-1}$
According to the Linearity of expectation, the expectation of the number of balls get discarded:

$$E[X] = \varepsilon n P_f = \varepsilon n (1 - \frac{1}{n})^{\varepsilon n - 1}$$

The expectation of the number of balls remained is:

$$\begin{aligned} E[Y] &= \varepsilon n - E[X] = \varepsilon n (1 - (1 - \frac{1}{n})^{\varepsilon n - 1}) \\ &= \varepsilon n (1 - e^{-\varepsilon + \frac{1}{n}}) \quad \text{since } \lim_{n \rightarrow \infty} (1 + \frac{x}{n})^n = e^x \\ &\leq \varepsilon n (1 - e^{-\varepsilon}) \\ &\leq \varepsilon n (1 - (1 - \varepsilon)) \quad \text{since } 1 + x \leq e^x \\ &= \varepsilon^2 n = (\varepsilon n)^2 / n \end{aligned} \tag{2}$$

Then the number of balls I expect to have in the next round is no more than $\varepsilon^2 n$

3.2 Assuming that everything proceeded according to expectation, prove that we would discard all the balls within $O(\log \log n)$ rounds.

Assume at the beginnig of round i the balls remained is x_i , then in the round x_{i+1} :

According to the former question:

$$x_{i+1} \leq x_i^2 / n$$

Also recall the formula in the former question $E[Y] = \varepsilon n (1 - e^{-\varepsilon})$, we can see after a constant $i \geq 3$ rounds there will only be less than $n/2$ balls left, then:

Then in the $i+t$ round:

$$\begin{aligned} x_{i+t} &\leq x_{i+t-1}^2 / n \\ &\leq x_i^{2^t} / n^{2^t - 1} \\ &\leq (n/2)^{2^t} / n^{2^t - 1} \\ &\leq n / 2^{2^t} \end{aligned} \tag{3}$$

When $t = \log \log n$, $x_{i+t} = 1$

$$t + i = \text{constant} + O(\log \log n) = O(\log \log n)$$

So it's proved that we would discard all the balls within $O(\log \log n)$ rounds.

4 Problem IV

4.1 Determine the probability that a fixed person i succeeds in accessing the computer during a specific step.

$$P_i = p * (1 - p)^{n-1}$$

4.2 How would you set p to maximize the above probability?

Maximizing P_i is equal to maximizing $\log P_i = \log p + (n - 1) \log(1 - p)$, so we need to have

$$\frac{\partial \log P_i}{\partial p} = \frac{1}{p} - (n - 1) \frac{1}{1 - p} = 0$$

$$p = \frac{1}{n}$$

$$P_i = \frac{1}{n} * (1 - \frac{1}{n})^{n-1}$$

4.3 For the choice of p in part(b), upper bound the probability that person i did not succeed to access the computer in any of the first $t = en$ steps.

$$P_1 = P[A \text{ person } i \text{ did not succeed to access the computer in a single round}] = 1 - P_i$$

$$\begin{aligned} P_1 &= 1 - \frac{1}{n} * (1 - \frac{1}{n})^{n-1} \\ &= 1 - \frac{1}{n} * \frac{n}{n-1} * (1 - \frac{1}{n})^n \\ &= 1 - \frac{1}{n-1} * e^{-1} \text{ since } n \rightarrow \infty \\ &\approx 1 - \frac{1}{n} * e^{-1} \text{ since } n \rightarrow \infty \\ &\leq e^{-\frac{1}{en}} \text{ since } 1 + x \leq e^x \end{aligned} \tag{4}$$

According to chain rules: $P_t = P_1^t \leq e^{-\frac{t}{en}} = e^{-1}$

4.4 What is the number of steps t required so that the probability that person i did not succeed to access the computer in any of the first t steps is upper bounded by an inverse polynomial in n ?

According to the former question, to satisfy the statement, we need $P_t = P_i^t \leq e^{-\frac{t}{en}} = O(\frac{1}{n^k})$

If $t = ken \log n$, $P_t \leq e^{-\frac{t}{en}} = e^{-\frac{ken \log n}{en}} = \frac{1}{n^k}$

So t need to satisfy $t \geq ken \log n$ (k is a positive constant).

4.5 How many steps are required to guarantee that all people succeeded to access the computer with probability at least $1/n$?

From former questions we know $1 - P_t = P[\text{A single person successfully access the computer in } t \text{ steps}]$
 So the $P_n = (1 - P_t)^n = P[\text{All people succeeded to access the computer in } t \text{ steps}]$

$$\begin{aligned} P_n &= \left(1 - \left(1 - \frac{1}{n-1} * e^{-1}\right)^t\right)^n \\ &\geq \left(1 - e^{-\frac{t}{en}}\right)^n \geq \frac{1}{n} \end{aligned} \tag{5}$$

$$e^{-\frac{t}{en}} \leq 1 - n^{-\frac{1}{n}} \leq e^{-n^{-\frac{1}{n}}}, \text{ meaning } \frac{t}{en} \geq n^{-\frac{1}{n}}$$

$$\text{So, } t \geq \frac{\log(1 - n^{-\frac{1}{n}})}{1 - \frac{1}{ne}} \geq -en \ln(1 - n^{-\frac{1}{n}}) \geq en^{1-\frac{1}{n}} \approx en \text{ is required}$$

5 Problem V

5.1 What is the expected time to find a good partitioning element?

Because a good partitioning element is greater than at least $n/4$ of the input items and smaller than at least $n/4$ of the input items. That means there are half of the elements can be considered as a good partitioning element.

So the expected number of executing the while loop is 2.

Inside the while loop the algorithm takes $O(n)$ time.

So the expected time to find a good partitioning element is $O(n)$, where $n = |S|$.

5.2 What is the expected time of Randomized Quicksort-v1 on a subproblem of size $|S|$, excluding the time spent on recursive calls?

When the size $|S| \leq 3$ the execution takes $O(1)$ time.

When the size $|S|$ is larger than 3, then:

Inside the while loop: the time in (a): $O(n)$, where $n = |S|$.

Outside the while loop: $O(1)$ time.

So in total it spends $O(n)$ time, where $n = |S|$.

5.3 We will say that a subproblem is of type j if its input consists of at most $n(\frac{3}{4})^j$ and at least $n(\frac{3}{4})^{j+1}$ items.

5.3.1 For a fixed j , how much time is spent on a subproblem of type j ?

According to problem (b), the expected time spent on subproblem of type j is : $O(\text{size}) = O(n(\frac{3}{4})^j)$

5.3.2 For a fixed j , how many subproblems of type j are there?

The algorithm divide the size of problem by $1/4$ to $3/4$. When we consider the expected number N of subproblem of size K , it should satisfy $N \cdot K = n$.

So the expected number of subproblems of type j is at most $(\frac{4}{3})^{j+1}$, that is $O((\frac{4}{3})^j)$

5.3.3 For a fixed j , how much time is spent on all subproblems of type j

The total time spent on subproblems of type j = the number of problems of type j * the expected time spent on subproblem of type j .

That is

$$O((\frac{4}{3})^j) * O(n(\frac{3}{4})^j) = O(n)$$

5.4 What is the expected running time of Randomized Quicksort-v1?

Since we have the time spent on all subproblems of type j , we only need to know the total number of subproblem types.

We know the smallest type of subproblem is 3, so: $n * (\frac{3}{4})^J = 3$,

expected number of types $J = \log_{\frac{3}{4}}(\frac{n}{3}) = O(\log n)$.

So the expected running time is $O(n \log n)$

6 Problem VI

6.1 Use the ideas from the previous problem to design and analyze the expected running time of a recursive randomized algorithm that returns the k -th smallest number in a set S of n distinct integers, for any k .

For example, for $k = \lfloor n/2 \rfloor$, your algorithm will return the median item.

Algorithm 2 Function $kThSmallestNumber(S, k)$

```
if  $k > |S|$  then
    return None
end if
if  $|S| \leq 3$  then
    sort  $S$ 
    return the  $k$ th value
end if
for no good partitioning element has been found do
    Select an element  $a_i \in S$  uniformly at random
    for each element  $a_j \in S$  do
        Put  $a_j$  in  $S^-$  if  $a_j < a_i$ 
        Put  $a_j$  in  $S^+$  if  $a_j > a_i$ 
    end for
    if  $|S^-| \geq |S|/4$  and  $|S^+| \geq |S|/4$  then
         $a_i$  is a good partitioning element
    end if
end for
if  $|S^-| \geq k$  then
    return  $kThSmallestNumber(|S^-|, k)$ 
else
    return  $kThSmallestNumber(|S^+|, k - 1 - |S^-|)$ 
end if
```

Correctness:

Set the size of S $|S| = n$, then:

Base case: $n = 1, 2, 3$ The algorithm will sort and output the k th value.

Induction hypothesis: Assume that the statement is true for case $n \in [1, 2, 3, \dots, i]$ $i \geq 3$.

Inductive step: Show it true for case $n=i+1$:

In the algorithm, if the k th value is directly returned, the statement is correct;

otherwise after partitioning the list, this case can be converted to :

1. $n_1 > k$, Discarding the largest $(n - n_1 - 1)$ elements which do not contain the k th smallest value then finding a k th smallest value for the remained list of size $n_1 \in [1, 2, 3, \dots, i]$ $i \geq 3$, the returned value is correct according to the induction hypothesis

or

2. $n_1 < k$, Discarding the smallest (n_1) elements which do not contain the k th smallest value then finding a k th smallest value for the remained list of size $n_1 \in [1, 2, 3, \dots, i]$ $i \geq 3$, the returned value is correct according to the induction hypothesis.

So the case $n=i+1$ can also return the correct value.

Conclusion:

It follows that the statement is true for all n since we can apply the inductive step for $n = 4; 5; 6; \dots$

Running time:

For a subproblem of size n , the running time excluding the recursion step should be :

In the for loop: The expected time to find a good partition element is $O(n)$.

Outside the for loop: $O(1)$ time.

So in total is $O(n)$ time.

In a single recursion, the problem of size n is downsized into a problem of size $1/4n$ to $3/4n$:

$$T(n) = T(3/4n) + O(n)$$

So according to the master theorem, the total running time is $O(n)$.

Space complexity:

For a subproblem of size a , the new assigned space is $\Theta(a)$;

The total recursion depth is $\log n$, the size of subproblem for depth i is at most $n * (\frac{3}{4})^i$.

So the space complexity is $\sum_{i=1}^{\log n} \Theta(n * (\frac{3}{4})^i) = \Theta(n)$