

# Homework Assignment #4

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Due February 28, 2019, by 5:30 p.m.

## Question 1.

Zijin Zhang wrote the solution to this question and Quan Xu read this solution to verify its clarity and correctness.

(a)

The algorithm will return *TRUE* in the first iteration if and only if the first  $i$  it picks satisfies  $A[i] = x$ . As it given, we know that there are  $k$  copies of  $x$  in  $A[1, \dots, n]$ . So the probability of finding an  $x$  is  $\frac{k}{n}$ .

(b)

Since the algorithm will either return *TRUE* or return *FALSE*, the probability of returning *TRUE*,  $P_{TRUE}$  is equal to  $1 - P_{FALSE}$  that  $P_{FALSE}$  represent the probability of returning *FALSE*. Now consider the scenario that the algorithm will return *FALSE*. The algorithm will return *FALSE* if and only if it fail to find  $x$  in the first  $r$  number of trials. We know that the probability of finding  $x$  in each trail is  $\frac{k}{n}$ , so the probability of not finding  $x$  in each trail is  $1 - \frac{k}{n}$  and we can get the probability of not finding  $x$  for  $r$  trails is  $(1 - \frac{k}{n})^r$ . Therefore, the probability of returning *TRUE* is  $1 - (1 - \frac{k}{n})^r$ .

(c)

For this part, we want to find the number of iteration for finding  $x$  in  $A[1, \dots, n]$ . To be more specific, we want to find the number of failures before the first success, which can be consider as a geometric distribution. Let  $r$  be the number of fail iteration until we finally find  $x$  in  $A[1, \dots, n]$  for the first time. As we proved in part (a), the probability of finding  $x$  in each iteration is  $\frac{k}{n}$ . So we have  $r \sim Geo(\frac{k}{n})$ . Now, we want to find the expected value of  $r$ . Since we know that  $r \sim Geo(\frac{k}{n})$ ,

$$E(r) = \frac{1 - \frac{k}{n}}{\frac{k}{n}} = \frac{\frac{n-k}{n}}{\frac{k}{n}} = \frac{n-k}{k}$$

So the expected number of failures is  $\frac{n-k}{k}$ . Therefore, the total number of iteration with the last success iteration is  $\frac{n-k}{k} + 1 = \frac{n}{k}$ .

## Question 2.

Quan Xu wrote the solution to this question and Zijin Zhang read this solution to verify its clarity and correctness.

Let  $i, j \in \mathbb{N}, 1 \leq i \neq j \leq n$

Assume  $n, m \in \mathbb{N}, n \geq 1$  and  $m \geq n$ .

We will use disjoint set(forest structure) and the operations  $Union()$  with WU(weight union by size) and  $Find\_set()$  with PC(path compression) to design the algorithm:

- Step 1: Loop over all  $n$  distinct variables,  $x_1, x_2, \dots, x_n$ . For  $i^{th}$  loop iteration, make a copy of variable  $x_i$  and using  $Make\_set()$  operation to make each variable as a singleton set, and sets each variable as the representative of its singleton set for now, and sets the size of each set as 1 for now.

- Step 2: Loop over all the  $m$  constraints and only do the following instructions for all **equality constraints**:

- For each *equality* constraints, find the representative of the two variables,  $x_i$  and  $x_j$ , of each *equality* constraints by using  $Find\_set()$  operation with PC to find the sets that contains  $x_i$  and  $x_j$  respectively, and their representative of their own sets respectively. Then union the two sets by using  $Union()$  with WU, iff the two representatives are different. After we union the two sets, the new set we created by  $Union()$  with Wu has size = sum of the two size of two old sets, and the new representative is one of the two old representatives who has larger size.

- Step 3: Loop over all the  $m$  constraints and only do the following instructions for all **inequality constraints**:

- For each *inequality* constraints, we have two variables  $x_i, x_j$ . Using the  $Find\_set()$  operation with PC to find the two sets that contains  $x_i$  and  $x_j$  respectively, and their representatives of their own set respectively. If the two representatives are the same, then the algorithm return NIL. Otherwise, continuous.

- Step 4: If no NIL return, then print an assignment of integers to variables( $x_1, x_2, \dots, x_n$ ).

Now consider the running time of the algorithm in the worst case:

Assume  $a, b, c, d \in \mathbb{R}^+$ .

- Step 1: There are  $n$  distinct variable and  $n \geq 1$ ,  $Make\_set()$  of each variable takes constant time. And assigning representative and updating size take also constant time. There are  $n$  variables, so total runtime is  $nC \in \mathcal{O}(n)$ .

- Step 2: For each loop iteration, finding each set that contains the variables and the representative of the set takes  $\mathcal{O}(\log^*n)$  by using  $Find\_set()$  operation with PC when it is an *equality* constraints. And the  $Union()$  operation with WU takes  $\mathcal{O}(1)$ , and at most  $n - 1$   $Union()$  operations. There are at most  $m$  iterations of loop, therefore, in total  $m(2\log^*n + C) \in \mathcal{O}(m\log^*n)$ .

- Step 3: For each iteration of loop, finding the two set that contains  $x_i, x_j$  by using  $Find\_set()$  with PC takes  $\mathcal{O}(\log^*n)$  OR it will return NIL which takes  $\mathcal{O}(1)$ . So for worst-case, we never run "return NIL". And the loop will at most iterates  $m$  times. So in total runtime is  $m(2\log^*n + C) \in \mathcal{O}(m\log^*n)$ .

- Step 4: To get each value of the each variable and print it take constant time, and there are  $n$  variables. So in total  $nC \in \mathcal{O}(n)$ .

Therefore, total runtime of worst-case scenario is :

$$(an + bm\log^*n + cm\log^*n + dn + C) \in \mathcal{O}(n + m\log^*n) \in \mathcal{O}(m\log^*n) \text{ (\# since } m \geq n) \in \mathcal{O}(mn)$$

So the algorithm is asymptotically better than  $\mathcal{O}(mn)$ .