

COMP 5711: Advanced Algorithm
2018 Fall Semester
Midterm Exam Solutions

Problem 1 (20pts) We can use a linked list as the underlying data structure. The operations are as follows.

- **INSERT**(S, x): Insert x to the linked list. The cost is $O(1)$.
- **DELETE-LARGER-HALF**(S): Find the median of elements in the linked list using the linear-selection algorithm. Scan the linked list and delete the large half one by one according to the median. The cost is $O(|S|)$.

We use the accounting method. Charge each **INSERT** operation an amortized cost of $O(1)$, which pays for the actual cost of the **INSERT** operation, and stores $O(1)$ as credits. Half of the elements are deleted in **DELETE-LARGE-HALF** operation. The larger-half has credits $\Omega(|S|)$ which is enough to pay for the $O(|S|)$ cost of **DELETE-LARGE-HALF**.

Problem 2 (30pts) The operations are as follows.

- **ENQUEUE**(Q, x): Push x into the S_1 .
- **DEQUEUE**(Q): If S_2 is not empty, pop from S_2 . Otherwise, pop each element from S_1 and push it immediately into S_2 . Then pop from S_2 .

Define the potential for a queue as 2 times the number of elements in the S_1 . So we have $\phi D_0 = 0$ and $\phi D_i \geq 0$. For **ENQUEUE** operation, the potential change is 2 and the actual cost is 1. So the amortized cost of **ENQUEUE** is $O(1)$. For **DEQUEUE**, the potential decreases by $2|S_1|$ and the actual cost is $2|S_1|$. So the amortized cost of **DEQUEUE** is also $O(1)$.

Problem 3 (30pts) We independently and uniformly sample (with repetition) $k = \frac{2}{\delta} \ln m$ students from S . And let C be the set of sampled students. We show that, with probability $1 - \frac{1}{m}$, $C \cap G_i \neq \emptyset$ for all $i \in [1, m]$. Let s_1, \dots, s_k be the students we sampled. We fixed some i in $[1, m]$. Since all the students are sampled independently,

$$Pr[G_i \cap C = \emptyset] = Pr[(s_1 \notin G_i) \wedge \dots \wedge (s_k \notin G_i)] \leq \prod_{j=1}^k Pr[s_j \notin G_i].$$

G_i has at least δn members, so for any $j \in [1, k]$,

$$Pr[s_j \notin G_i] \leq \frac{n - \delta n}{n} = 1 - \delta.$$

Recall that $k = \frac{2}{\delta} \ln m$. We have

$$Pr[G_i \cap C = \emptyset] \leq \prod_{j=1}^k Pr[s_j \notin G_i] = (1 - \delta)^{\frac{2}{\delta} \ln m} \leq \left(\frac{1}{e}\right)^{2 \ln m} = \frac{1}{m^2}.$$

Finally, by union bound, we obtain

$$Pr[(C \cap G_1 = \emptyset) \wedge \dots \wedge (C \cap G_m = \emptyset)] \leq \sum_{i=1}^m Pr[G_i \cap C = \emptyset] \leq m \cdot \frac{1}{m^2} = \frac{1}{m}.$$

Problem 4 (20pts) We call a pivot "good" if it is between the 25% and 75%-percentile of the current elements, otherwise "bad". The probability of getting "good" pivot is then $1/2$. Note that each "good" pivot reduces the number of elements by at least $1/4$. So the number of "good" pivot is $O(\log n)$. Let X_i denote the number of pivots between the i -th good pivot (not including) and the $(i + 1)$ -th "good" pivot (including).

$$E[X_i] = \sum_{j=0}^{\infty} j \times \Pr(X_i = j) \leq \sum_{j=0}^{\infty} \frac{j}{2^j} = 2.$$

According to the linearity of expectations, the expected number of iterations (pivots) is $E[\sum_{i=1}^{O(\log n)} X_i] = \sum_{i=1}^{O(\log n)} E[X_i] = O(\log n)$.