Homework 1

Algorithms for Big Data CS498ABD Spring 2019

Due: 10am, Wednesday, Feb 6th

Instructions:

- Each home work can be done in a group of size at most two. Only one home work needs to be submitted per group. However, we recommend that each of you think about the problems on your own first.
- Homework needs to be submitted in pdf format on Gradescope. See https://courses.engr.illinois.edu/cs374/fa2018/hw-policies.html for more detailed instructions on submitting on Gradescope.
- Follow academic integrity policies as laid out in student code. You can consult sources but cite all of them including discussions with other class mates. Write in your own words. See the site mentioned in the preceding item for more detailed policies.

Exercise 1: Sampling, Chebyshev vs Chernoff. Suppose you want to estimate the average of n numbers via sampling (for example the heights of students in the class). The average can be very skewed by outliers. However, we can obtain an accurate estimate if we assume that the numbers are within some limited range. Assume the input numbers z_1, z_2, \ldots, z_n are from [a, b] where $a, b \in \mathbb{R}$ with $a \leq b$. Suppose you sample k input numbers (with replacement) and output their average as the estimate for the true average $\alpha = (\sum_i a_i)/n$. Let X be the random variable denoting the output value.

(a) Using Chebyshev's inequality, show that for $k \geq \frac{(b-a)^2}{\delta \epsilon^2}$, we have

$$P[|X - \alpha| \ge \epsilon] \le \delta.$$

(b) Using the Chernoff inequality, show that there exists a constant c>0 such that for $k \geq \frac{c(b-a)^2 \log(1/\delta)}{\epsilon^2}$, we have

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$$P[|X - \alpha| \ge \epsilon] \le \delta.$$

Exercise 2: QuickSort Given an array A of n numbers (which we assume are distinct for simplicity), the algorithm picks a pivot x uniformly at random from A and computes the rank of x. If the rank of x is between n/4 and 3n/4 (call such a pivot a good pivot), it behaves like the normal QuickSort in partitioning the array A and recursing on both sides. If the rank of x does not satisfy the desired property (the pivot picked is not good), the algorithm simply repeats the process of picking a pivot until it finds a good one. Note that in principle the algorithm may never terminate!

- (a) Write a formal description of the algorithm.
- (b) Prove that the expected run time of this algorithm is $O(n \log n)$ on an array on n numbers.
- (c) Prove that the algorithm terminates in $O(n \log n)$ time with high probability.

Exercise 3,4: To be announced.

Additional exercises (not to be submitted)

Exercise 5. In class, we proved a powerful tail inequality called the "(multiplicative) Chernoff bound" that we will use time and time again. In this exercise, we rewrite the Chernoff in a convenient form that is a little more interpretable and easier to apply.

Recall the Chernoff inequality, as follows. Let $X_1, X_2, \ldots, X_n \in [0, 1]$ be n independent, nonnegative, and uniformly bounded random variables. Let

$$\mu = E\left[\sum_{i=1}^{n} X_i\right] = \sum_{i=1}^{n} E[X_i]$$

be the expected value of the sum. The Chernoff inequality states that for any $\delta > 0$, we have

$$P\left[\sum_{i=1}^{n} X_i \ge (1+\delta)\mu\right] \le \left(\frac{e^{\delta}}{(1+\delta)^{1+\delta}}\right)^{\mu} \text{ and } P\left[\sum_{i=1}^{n} X_i \le (1-\delta)\mu\right] \le \left(\frac{e^{-\delta}}{(1-\delta)^{1-\delta}}\right).$$

Now let X_1, \ldots, X_n and μ be as above.

(a) Show that for $x \ge 0$ sufficiently small, we have

$$x - (1+x)\ln(1+x) \le -\frac{x^2}{3}$$
.

Hint: Consider the Taylor expansion $\ln(1+x) = x - x^2 + \frac{x^3}{3} - \frac{x^4}{4} + \frac{x^5}{5} - \cdots$ for $x \in (-1,1]$.

(b) Show that for $\epsilon \in [0, 1]$,

$$P\left[\sum_{i=1}^{n} X_i \ge (1+\epsilon)\mu\right] \le e^{-\epsilon^2\mu/3}.$$

(c) Show that for $x \in [0, 1]$, we have

$$x + (1 - x)\ln(1 - x) \ge \frac{x^2}{2}$$
.

(d) Show that for $\epsilon \in [0, 1]$,

$$P\left[\sum_{i=1}^{n} X_i \le (1 - \epsilon)\mu\right] \le e^{-\epsilon^2 \mu/2}.$$

Exercise 6, 7. Exercises 2 and 3 from HW 4 of the 2016 algorithms course (https://courses.engr.illinois.edu/cs473/fa2016/Homework/hw4.pdf)

Exercise 8: Reservoir sampling. Show that sample-without-replacement (below) outputs uniformly random sample of k elements without replacement from a stream.

sample-without-replacement(k)

- 1. $S[1..k] \leftarrow \text{null}$
- 2. $m \leftarrow 0$
- 3. While (stream is not done)
 - A. $m \leftarrow m+1$
 - B. e_m is current item in stream
 - C. If $(m \le k)$ $S[m] \leftarrow x_m$
 - D. else
 - i. select integer r uniformly at random from $\{1,2,\ldots,m\}$
 - ii. If $(r \leq k)$ $S[r] \leftarrow x_m$
- 4. Output ${\cal S}$