

**Exercise 1 (Frequency Moments and Tug-Of-War) 3+4+1+3+1+2=14 points**

Consider the following stream of data elements over the universe  $\mathbb{U} = \{1, 2, \dots, 9\}$ :

$$\mathbf{a} = 1, 4, 4, 1, 5, 3, 8, 2, 2, 1, 5.$$

- a) Compute  $F_0(\mathbf{a})$ ,  $F_1(\mathbf{a})$  and  $F_2(\mathbf{a})$ .
- b) Assume we run the AMS-Estimator (slide 8.36) on  $\mathbf{a}$  and evaluate the variables at the end of the while loop (after line 8). Complete the following table with suitable values:

$i$	1	2	3	4	5	6	7	8	9	10	11
$a_i$	1			1							
$a$	1										2
$r$	1		2				2				2

- c) What is the estimated result for  $F_2(\mathbf{a})$  returned by the AMS-Estimator in b)?
- d) What is the result  $x^2$  returned by the Tug-of-War estimator on  $\mathbf{a}$  if the randomly chosen hash function is given by

$u$	1	2	3	4	5	6	7	8	9
$h(u)$	1	1	-1	1	-1	1	1	1	-1

- e) Is there a hash function  $h': \mathbb{U} \rightarrow \{-1, 1\}$  such that Tug-of-War returns a better (i. e. closer) estimate for  $F_2$  than it does in part d)? If yes, give such a hash function. If no, argue why not.
- f) Now think of any stream  $b$  with  $n$  elements, of which  $m$  are distinct. What are the minimum and maximum possible values of  $F_2(b)$  (as a function of  $m$  and  $n$ ).

**Solution:** \_\_\_\_\_

- a) We have  $F_0(\mathbf{a}) = 6$ ,  $F_1(\mathbf{a}) = 11$  and  $F_2(\mathbf{a}) = 9 + 4 + 1 + 4 + 4 + 0 + 0 + 1 + 0 + 0 = 23$ .

- b) The second row  $a_i$  is given by the stream.

$i$	1	2	3	4	5	6	7	8	9	10	11
$a_i$	1	4	4	1	5	3	8	2	2	1	5
$a$	1										2
$r$	1		2				2				2

The remaining values are enforced by:

- having to choose  $a = 2$  at  $i = 8$  (otherwise we can't have  $a = 2$  and  $r = 2$  at  $i = 11$ ).

- having to choose  $a = 4$  at  $i = 2$  (otherwise we can't have  $r = 2$  at  $i = 3$ )
- having to not change  $a = 4$  until  $i = 8$  (otherwise we can't have  $r = 2$  at  $i = 7$ )

This is the resulting table:

$i$	1	2	3	4	5	6	7	8	9	10	11
$a_i$	1	4	4	1	5	3	8	2	2	1	5
$a$	1	4	4	4	4	4	4	2	2	2	2
$r$	1	1	2	2	2	2	2	1	2	2	2

- c) We choose  $k = 2$  and return  $11 \cdot (2^2 - 1^2) = 33$ .
- d) We first need to compute  $x = \sum_{i \in [11]} h(a_i) = 1 + 1 + 1 + 1 - 1 - 1 + 1 + 1 + 1 + 1 - 1 = 5$ .  
The the Tug-Of-War estimator return  $x^2 = 25$ .
- e) We return a value of the form  $x^2$  where  $x$  is an integer and 25 is the closest square number to 23, so there can't be a hash function that returns a closer estimate.
- f) For the maximum second moment, we pick  $(m - 1)$ -many elements just once each and the remaining one element  $(n - m + 1)$  times. This gives us the function

$$1 \cdot (n - m + 1)^2 + (m - 1) \cdot 1^2.$$

For the minimum second moment, we want to distribute the elements as evenly as possible:  $(n \bmod m)$ -many elements of size  $\lceil n/m \rceil$ , and the remaining elements of size  $\lfloor n/m \rfloor$  (i.e. one smaller in size). This gives us the function

$$(n \bmod m) \cdot \lceil n/m \rceil^2 + (m - n \bmod m) \cdot \lfloor n/m \rfloor^2.$$

## Exercise 2 (Improve the Probability)

6 points

Consider an algorithm  $\mathfrak{A}(h)$  that uses a (truly) random hash function  $h \in \mathcal{H}$  and gives an estimate  $\hat{x} = \mathfrak{A}(h)$  of the true value  $x$  of some variable. Suppose that:

$$\Pr_{h \in \mathcal{H}} \left( \frac{x}{4} \leq \hat{x} \leq 4x \right) \geq 0.6.$$

The probability of the estimate is with respect to choice of the hash function. How would you compute an estimate  $x'$  that has an improved probability of:

$$\Pr \left( \frac{x}{4} \leq x' \leq 4x \right) \geq 0.8?$$

**Hint:** Since we do not know the variance, taking the **average** of multiple runs may not help.

**Solution:** \_\_\_\_\_

Let us choose some  $k$  and run the algorithm  $k' := 2k - 1$  times with  $k'$  different random hash functions that are drawn independently. Let  $\hat{x}_1, \dots, \hat{x}_{k'}$  be the sorted resulting estimations. Return the median  $x' := \hat{x}_k$  as the new estimation.

We can follow the proof for the Approximation Guarantee on Page 8.31 to find that  $58.06 \leq k$ .

**Alternatively** we can calculate a simple and better bound by hand:

Now let us assume that the new estimate  $x'$  is not within the range  $\frac{x}{4} \leq x' \leq 4x$ . This means we must have sampled a hash function  $h$  such that  $\hat{x}$  is outside of this range at least  $\frac{k'+1}{2}$  times (otherwise the median is in the range).

Conversely, if we sample outside of the range  $i < \frac{k'+1}{2}$  times, then the returned estimator  $x'$  is clearly also within the range  $\frac{x}{4} \leq x' \leq 4x$ . The probability for this is at least:

$$\Pr \left( \frac{x}{4} \leq x' \leq 4x \right) \geq \sum_{i=0}^{\frac{k'-1}{2}} \binom{k'}{i} \cdot 0.4^i \cdot 0.6^{(k'-i)}$$

For  $k' = 17$  or  $k = 9$  we have  $\Pr \left( \frac{x}{4} \leq x' \leq 4x \right) \geq 0.801$ .

**Exercise 3 (Minimum Memory for Distinct Elements Approximation) 0 points**

Show that any deterministic algorithm that even guarantees to **approximate** the number of distinct elements in a data stream over universe  $\mathbb{U} = \{1, \dots, m\}$  with error less than  $\frac{m}{16}$  must use  $\Omega(m)$  bits of memory. This is even the case for data streams of length less than  $2m$ .

**Hint:** There is a constant  $c > 0$ , for which it is possible to create  $2^{cm}$  subsets of  $\{1, \dots, m\}$ , each with  $m/2$  elements, such that no two of the subsets have more than  $3m/8$  elements in common. You can use this fact without proving it.

**Solution:** \_\_\_\_\_

Let  $\mathfrak{S}$  be the set of  $2^{cm}$  subsets of  $\{1, \dots, m\}$ , each with  $m/2$  elements, such that no two of the subsets have more than  $3m/8$  elements in common (as given by the hint). We denote the sets by  $S_i \in \mathfrak{S}$  and we define the stream  $s_i$  as a stream containing every element in  $S_i$  exactly once in ascending order.

We now show that any algorithm with a guaranteed error of less than  $\frac{m}{16}$  needs to use at least  $2^{cm}$  distinct states and therefore memory of at least  $c \cdot m$  bits.

Assume we use less than  $c \cdot m$  bits. Then there are two sets  $S_1, S_2 \in \mathfrak{S}$ , such that after reading the streams  $s_1$  and  $s_2$  we end in the same state. Now consider appending these streams by  $s_1$ . Then the resulting streams  $s_1 \cdot s_1$  and  $s_2 \cdot s_1$  end in the same state and we return the same result  $y$  for both streams. By definition of the sets,  $s_1 s_1$  has exactly  $y_1 = m/2$  distinct elements and  $s_2 s_1$  has at least  $y_2 \geq m/2 + m/8$  distinct elements. This is because the two sets  $S_1$  and  $S_2$  differ in at least  $m/8$  elements. The streams have a length of  $2 \cdot m/2 = m$  elements.

Now assume the errors are  $|y_1 - y| < m/16$  and  $|y_2 - y| < m/16$ , then  $|y_1 - y_2| < m/8$ . We know from the construction of the strings that  $|y_1 - y_2| \geq m/8$ , which leads to a contradiction.

Therefore we show that the algorithm must be able to distinguish between at least  $2^{cm}$  different states (representing the  $2^{cm}$  sets) and therefore must use at least  $\log(2^{cm}) = cm$  bits, which can be written as  $\Omega(m)$  bits.