## 11<sup>TH</sup> TUTORIAL ON RANDOMIZED ALGORITHMS Streaming algorithms: KMV and Count-Min sketches

Even simpler count distinct: K Minimum Values (KMV) sketch. We would like to count the number of distinct elements, i.e., estimate set cardinality. We look at a different approach (which is actually quite popular in practice): For a parameter kand a hash function  $h: [N] \to (0,1]$ , store the k smallest hash values of the distinct stream elements, i.e., we store k pairs (item j, h(j)). When queried for cardinality, return  $(k-1)/v_k$ , where  $v_k$  is the k-th smallest hash value (the largest one stored).

1.

- a) Analyze the algorithm assuming h is fully random and prove that given  $\varepsilon \in (0, 1)$ , for  $k \geq c/\varepsilon^2$  (where c is a large enough constant) the algorithm gives an  $\varepsilon$ approximation of  $F_0$  = the number of distinct elements with constant probability. Focus on bounding the probability of  $(k-1)/v_k > (1+\varepsilon) \cdot F_0$ ; the other inequality is similar.
- b) What is wrong with h being fully random? What kind of hash functions would be sufficient for the analysis?

2. Count-Min sketch for frequency estimation. We would like to estimate frequencies and find heavy hitters under both insertions and deletions (similarly as CountSketch but with a different guarantee). We will assume that all frequencies are non-negative at the end. We use the following sketch for estimating frequencies  $f_i$  (screenshot from lecture notes by A. Chakrabarti):

## Algorithm 9 Count-Min Sketch Initialize:

- 1:  $C[1...t][1...k] \leftarrow \vec{0}$ , where  $k := 2/\varepsilon$  and  $t := \lfloor \log(1/\delta) \rfloor$
- 2: Choose *t* independent hash functions  $h_1, \ldots, h_t : [n] \to [k]$ , each from a 2-universal family

**Process** (token (j,c)): 3: for  $i \leftarrow 1$  to t do 4:  $C[i][h_i(j)] \leftarrow C[i][h_i(j)] + c$ 

Output (query *a*): 5: report  $\hat{f}_a = \min_{1 \le i \le t} C[i][h_i(a)]$ 

a) Using the assumption that all frequencies are non-negative at the end, derive lower and upper bounds on the estimator of a single row. That is, for any  $a \in [n]$ and row  $i \in [t]$  show that

$$\left| f_a - C[i][h_i(a)] \right| \le \varepsilon \cdot \|\mathbf{f}\|_1.$$
(1)

with a constant probability.

- b) Show a high probability bound for the final estimator  $\hat{a}_j$  for frequency  $f_a$ .
- c) Compare CountSketch (from the lecture) and Count-Min sketch, both in terms of their description and their properties.
- d) Can you derive a more refined bound on the error of Count-Min? That is, replace  $\|\mathbf{f}\|_1$  by a smaller quantity in (1).
- e) Count-Min is a linear sketch, that is, it can be viewed as a linear map of the frequency vector  $\mathbf{f}$  to a much smaller dimension. What are the properties of the matrix of this linear map?