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1. Hash functions for sampling

(a) Prove that $p \leq Pr[hm(x)/m < p] \leq 1.01p$. For this we will use few things.

 $h(x) = h_m(x)/m$ is Strong Independent Hash Function;

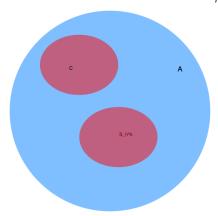
$$p \ge 100/m \Rightarrow p/100 \ge 1/m;$$

 $hm(x)/m \le p \Rightarrow hm(x) \le mp;$

(b)

2. Bottom-k sampling

Prove that $E[|C \cap S_h^k(A)|/k] = |C|/|A|$, assuming that $S_h^k(A)$ is a uniformly random size- $k \in A$. We assume that $C \subset A$ and $S_h^k(A) \subset A$ are independent, as shown on Venn diagram bellow:



Seeing this we can say:

$$\begin{aligned} ⪻[x \in S_h^k(A)] = \frac{k}{|A|} \\ ⪻[x \in C] = \frac{C}{|A|} \end{aligned}$$

$$E[|C \cap S_h^k(A)|/k]$$

$$= \frac{1}{k} E[|C \cap S_h^k(A)|]$$

$$= \frac{1}{k} \sum_{\forall a \in A} E[a \in C \land a \in S_h^k(A)]$$

$$= \frac{1}{k} \sum_{\forall a \in A} Pr[a \in C \land a \in S_h^k(A)]$$

$$= \frac{1}{k} \sum_{\forall a \in A} Pr[a \in C] \cdot Pr[a \in S_h^k(A)]$$

$$= \frac{1}{k} \sum_{\forall a \in A} \left(\frac{C}{|A|} \cdot \frac{k}{|A|}\right)$$

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$$= \frac{1}{k} \frac{C}{|A|} \cdot \frac{k}{|A|} \sum_{\forall a \in A} 1$$

$$= \frac{1}{k} \frac{C}{|A|} \cdot \frac{k}{|A|} |A|$$

$$= \frac{C}{|A|}$$

(From independence)

(Substitution from above)

(Substitution from above)

(Elements in sum are not dependant on $a \in A$)

2.1 Frequency estimation

Exercise 2

- (a) As a data structure we would use binary heap (max heap). Binary heaps are a one of the ways of implementing priority queues. In the heap we would only store k smallest keys.
- (b) To process another key from the stream it would take $O(log_2k)$.
- 2.2 Similarity estimation
- (a)
- (b)
- (c)
- 3 Bottom-k sampling with strong universality

3.1 A union bound

exercise 5

3.2 Upper bound with 2-independence

exercise 6

exercise 7