

Stream:  $\sigma = \langle a_1, a_2, \dots, a_m \rangle$ , token  $a_i \in U = [n]$ .

Goal: process  $\sigma$  using space  $s$  st  $s \ll m \wedge s \ll n$ ,  $s = O(\min(m, n))$ , pref  $s = O(\log m + \log n)$ .

Sometimes best  $s = \text{polylog}(\min(m, n))$ ,  $\text{polylog} = O((\log(g(n)))^c)$  for  $c > 0$ .

Streams can only be access in sequence.

Multiplicative condition: Let  $A(\sigma)$  be random stream alg A on  $\sigma$ : Let  $\phi$  be target function. A is an  $\epsilon, \delta$ -approx alg of  $\phi$  if

$$\Pr\left[\left|\frac{A(\sigma)}{\phi(\sigma)} - 1\right| > \epsilon\right] \leq \delta$$

To strong a condition when  $\phi(\sigma)$  has values close to zero.

Additive-approx:  $(\epsilon, \delta)$ -additive-approx-alg A of  $\phi$  if

$$\Pr[|A(\sigma) - \phi(\sigma)|] \leq \delta$$

Often instrested in statistical properties of multiset in  $\sigma$ . Let vector  $\vec{F} = (f_1, \dots, f_n)^T$ ,  $f_j = |\{i : a_i = j\}|$ . So  $\phi(\vec{F})$  target.  $\vec{F}$  frequency vector.

Turnstile model:  $\sigma : [n] \times \{-L, \dots, L\}$  so  $f_j \sum_{\{i: a_i = j\}} l_i$ . Redefined m to max number tokens in multiset  $\|\vec{F}\|_1 \leq m$ .

Strict turnstile model:  $\vec{F} \geq 0$

Cash register model:  $\forall i. l_i > 0$

## 1 Frequency Problem

Majority problem: if  $\exists j : f_j > m/2$ , then output j else null. Frequency problem: For some k, output  $\{j : f_j > m/k\}$

Frequency-estimateion problem: For stream  $\sigma$  produce structure that can estimate  $\hat{f}_a$  for freq  $f_a$  for  $a \in [n]$ .

Misra-Gries Alg: Takes param k (same as freq problem). Maintain associative array, fx bbt

Process j:

```

If j in keys(A) then
  A[j] <- A[j]+1
else if |keys(A)| < k -1 then
  A[j] <- 1
else
  foreach l in keys(A)
    A[l] <- A[l] - 1
    if A[l] = 0 then remove l
output:  $\hat{f}_a = A[a]$  if keys(a) else 0

```

**Thrm** The Misra-Gries alg. with param k uses one pass and  $O(k(\log m + \log n))$  bits of space and fives estimate  $\hat{f}_j$  satisfying

$$\max(0, f_j - \frac{m}{k}) \leq \hat{f}_j \leq f_j$$

Proof: When  $A[j]$  decr then we decr k-1 other counters  $\Rightarrow$  decr witnessed (applied to) k tokens  $\Rightarrow \leq m/k$  decrs as  $|\sigma| = m$ , therefore  $\hat{f}_j \leq f_j - m/k$ .

## 2 The median trick

Event  $X = \sum_1^t x_i$ , X is bad if  $|X - f_x| \geq \epsilon \|f_x\| \geq \gamma$

$X_{t/2}$  (median) bad implies  $X_{<t/2}$  bad events.

$B_i \equiv X_i$  bad,  $B = \sum B_i$

$X_{t/2} \Rightarrow B \geq t/2$ ,  $\Pr[B_i] \leq 1/3 \Rightarrow E[B] \leq t/3$

$\Pr[X_{t/2} \text{ bad}] \leq \exp(-\Omega(t))$

### 3 Sketchs (Ran freq algs)

Sketch: Datastructure d st.  $d(\sigma_1 \cdot \sigma_2)$ ,  $\cdot$  is string concat, can be computed from  $d(\sigma_1)$  and  $d(\sigma_2)$  using space efficient comb alg  $C(d(\sigma_1), d(\sigma_2)) = d(\sigma_1 \cdot \sigma_2)$ .

Linear Sketch: Sketching alg. A st  $\forall \sigma \subseteq [n]$   $A(\sigma)$  takes value in vector space of dim  $l = l(n)$  and  $A(\sigma)$  is linear function of  $\vec{F}(\sigma)$ .  $l$  is the dim of linear sketch. Works under turnstile and strict turnstile model.

#### 3.1 Count Sketch

Basic sketch parameter  $\epsilon$  (desired accuracy)

Init

```
C[1,...,k] <- 0,...,0
h: [n] -> [k] 2-universal hash function
g: [k] -> {-1,1} 2-universal hash function
```

Process(j,c)

```
C[h(j)] += cg(j)
```

Query(a in [N])

```
f^_a = g(a)C[h(a)]
```

Analysis of Basic Sketch Estimate:

Fix a, consider  $X = \hat{f}_a$ , let  $Y_j = 1$  iff  $h(j) = h(a)$ , j contributes to counter  $C[h(a)] \Leftrightarrow h(j) = h(a)$ , contributes sign  $g(j)$  and freq  $f_j$ .

Therefore,

$$X = g(a) \sum_1^n f_j g(j) Y_j = f_a + \sum_{j \neq a} f_j g(a) g(j) Y_j$$

$$E[g(j) Y_j] = E[g(j)] E[Y_j] = (1/2 - 1/2) E[Y_j] = 0$$

first equality uses g is independent of h, above implies

$$E[X] = f_a + \sum f_j g(a) E[g(j) Y_j] = f_a$$

$f_j$  and  $g(a)$  are constants. Hence  $X = \hat{f}_a$  is an unbiased estimator for  $f_a$ .

Second moment of Basic sketch:

By 2-universality of  $h \in \mathcal{U}(h)$  we have  $\forall j \in [n] - a$

$$E[Y_j^2] = E[Y_j] = \Pr[h(j) = h(a)] = 1/k$$

By 2-universality of g and independence of g and h  $\forall i, j \in [n], i \neq j$

$$E[g(j) g(i) Y_i Y_j] = E[g(j)] E[g(i)] E[Y_i Y_j] = 0 * 0 * E[Y_i Y_j] = 0$$

This implies that

$$\text{var}[X] = 0 + g(a)^2 \text{Var}[\sum_{j \neq a} f_j g(j) Y_j] \tag{1}$$

$$= E[\sum_{j \neq a} f_j^2 Y_j^2 + \sum_{i, j \neq a, i < j} f_i f_j g(i) g(j) Y_i Y_j] - (\sum_{j \neq a} f_j E[g(j) Y_j])^2 \tag{2}$$

$$= \sum_{j \neq a} f_j^2 / k + 0 + 0 = \|f\|_2^2 - f_a^2 / k \tag{3}$$

where  $f$  is frequency distribution determine by  $\sigma$ .

From Chebyshev inequality

$$\Pr[|\hat{f}_a - f_a| \geq \epsilon \sqrt{\|f\|_2^2 - f_a^2}] = \Pr[|X - E[X]| \geq \epsilon \sqrt{\|f\|_2^2 - f_a^2}] \tag{4}$$

$$\leq \frac{\text{var}[x]}{\epsilon^2 (\|f\|_2^2 - f_a^2)} \tag{5}$$

$$= 1/k\epsilon^2 = 1/3 \tag{6}$$

For  $j \in [n]$  let  $f_{-j}$  denote the  $(n-1)$  dimensional vector obtained by dropping  $j$ 'th entru of  $f$  then

$$\Pr[|\hat{f}_a - f_a| \geq \epsilon \|f_a\|_2^2] \leq 1/3$$

### 3.2 Count sketch alg

Init

```
C[1..t][1..k] <- 0, k = 3/epsilon, t = O(log(1/delta))
choose t indep 2-universal hash funsc h_1,...,h_t:[n]->[k]
choose t indep 2-universal hash funcs g_1,...,g_t:[k]->{-1,1}
```

Process( $j, c$ )

```
for i=1 to t C[i][h_i(j)] += c g_i(j)
```

Query( $a$ )

```
f_a = median_{1 \leq i \leq t} C[i][h_i(a)]g_i(a)
```

Chernoff bound argument proves  $\hat{f}_a$  satiesfies

$$\Pr[|\hat{a}_a - f_a| \geq \epsilon \|f_{-a}\|] \leq \delta$$

Hash fuction stored in  $O(t \log n)$  space. each of the tk countes uses  $\log m$  space. This gives space bound

$$O(t \log n + tk \log m) = O(1/\epsilon^2 \log 1/\delta (\log n + \log m))$$

## 4 Count-Min Sketch