

DATA STREAMING : FLOW OF MASSIVE SEQUENCES OF DATA

- lots of data
- cannot store it

n elements
Polylog space in n

MOTIVATION

- IP traffic in a router
- telephone calls
- query logs

CHALLENGE

- simple problems (statistics) become difficult

example EASY : find Max : DIFFICULT : find
 QUANTILES
(most frequent)
top k , etc

Problems cannot be solved deterministically

FEW EXCEPTIONS: Find the missing element in a permutation of $1, 2, \dots, n$ {just $n-1$ elements arrive}

- sum $2^{\log n}$ bit
- XOR $\log n$ bit

COUNT-MIN SKETCH (Cormode - Muthu)

n items numbered 1...n

F frequency array $F[i] = \# \text{times } i \text{ appears}$
in the stream

operations: $F[i]++$, $F[i]--$, invariant: $F[i] \geq 0$

- ▷ cannot store F entirely, have only $O(\log n)$ bits
- ▷ find approximation \tilde{F} s.t.

$$\forall i : F[i] \leq \tilde{F}[i] \leq F[i] + \underbrace{\varepsilon \|F\|}_{\text{with probability } 1-\delta}$$

$\|F\| = \sum_{j=1}^n F[j]$

Need a couple of notions

- k-wise independence
 - random variables
 - hash functions
- Markov's inequality

K-wise limited independence

X_1, X_2, \dots, X_n random variables with support S_1, S_2, \dots, S_n

k-wise independent if

\forall choice $i_1 < i_2 < \dots < i_k \in [n]$ and $a_{ij} \in S_{i_j}$

$$\Pr[X_{i_1} = a_{i_1} \wedge X_{i_2} = a_{i_2} \wedge \dots \wedge X_{i_k} = a_{i_k}] = \Pr[X_{i_1} = a_{i_1}] \times \Pr[X_{i_2} = a_{i_2}] \times \dots \times \Pr[X_{i_k} = a_{i_k}]$$

Implication: $E[X_{i_1} X_{i_2} \dots X_{i_k}] = E[X_{i_1}] \times E[X_{i_2}] \times \dots \times E[X_{i_k}]$

$\{h\}_{h \in H}$ family of hash functions $[n] \rightarrow [b]$

$\Pr[h \in H] = \frac{1}{|H|}$, uniform distributions

Family H is k-wise independent if

$\forall x_1, x_2, \dots, x_k \in [n], b_1, b_2, \dots, b_k \in [b]$

$$\Pr_{h \in H}[h(x_1) = b_1 \wedge h(x_2) = b_2 \wedge \dots \wedge h(x_k) = b_k] = \Pr_{h \in H}[h(x_1) = b_1] \times \Pr_{h \in H}[h(x_2) = b_2] \times \dots \times \Pr_{h \in H}[h(x_k) = b_k]$$

$$\begin{bmatrix} x_{i,j} = h(x_j) \\ a_j = b_j \end{bmatrix}$$

Example

$$h(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{k-1} x^{k-1}$$

$$\Rightarrow |H| = |\mathbb{F}|^k \quad a_i \in \text{finite field } \mathbb{F}$$

\Rightarrow there $|\mathbb{F}|^{k-1}$ solutions to $h(x_i) = b_i$

$$a_0 = b_i - a_1 x_i - a_2 x_i^2 - \dots - a_{k-1} x_i^{k-1}$$

$$\Pr[h(x_1) = b_1 \wedge h(x_2) = b_2 \wedge \dots \wedge h(x_k) = b_k]$$

$\frac{1}{|F|^k}$

$$\Pr[h(x_1) = b_1] \wedge \Pr[h(x_2) = b_2] \wedge \dots \wedge \Pr[h(x_k) = b_k]$$

$\frac{1}{F}$ $\frac{1}{F}$ $\frac{1}{F}$

$$= \frac{1}{F^k} \quad \text{QED}$$

Space to store $h(\mathbf{x})$ is that of $2, 2, \dots, 2^{k-1}$
 $O(k \lg |F|)$ bits

Ex. Take prime $p \in [b-1..2n]$: show $h'(x) = (h(x) \bmod p) \bmod b$ is approximately k -wise indep.

Markov's Inequality

$X = \text{random variable } \geq 0, \forall \epsilon > 0$

$$\Pr[X \geq \bar{\alpha}] \leq \frac{\mathbb{E}[X]}{\bar{\alpha}}$$

Proof.

$$I = \text{indicator variable} = \begin{cases} 0 & \text{if } X < \bar{\alpha} \\ 1 & \text{if } X \geq \bar{\alpha} \end{cases}$$

$$\text{Fact } \mathbb{E}[I] = 0 \cdot \Pr[I=0] + 1 \cdot \Pr[I=1] \\ = \Pr[I=1] = \Pr[X \geq 2]$$

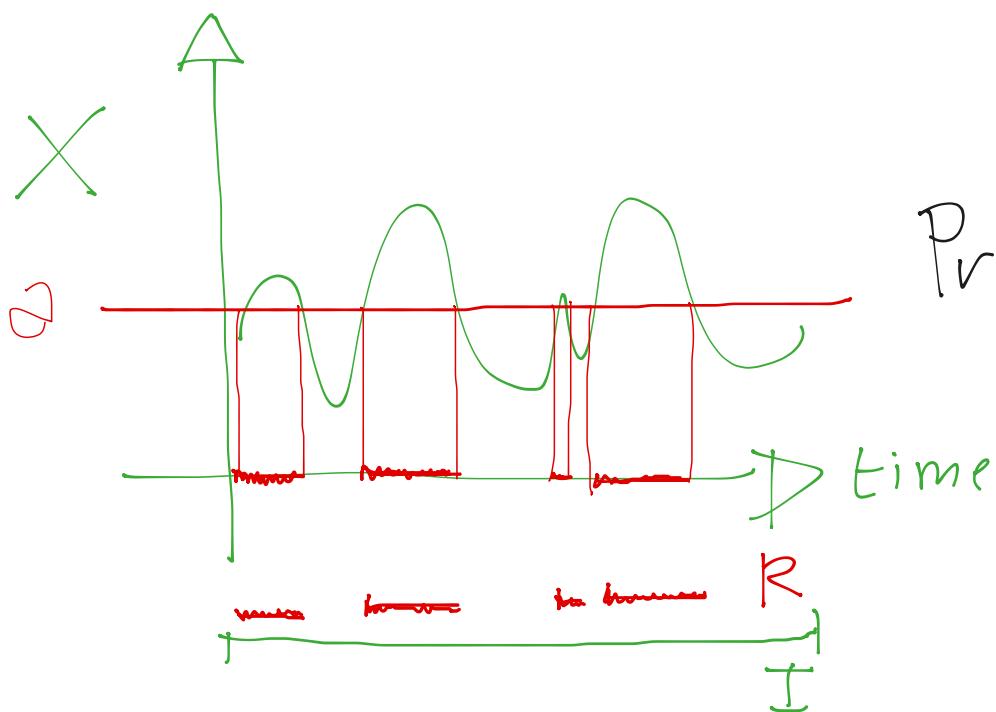
$$\triangleright 2I \leq X$$

$$\therefore X < 2 \Rightarrow I = 0$$

$$\therefore X \geq 2 \Rightarrow 2I = 2 \leq X$$

$$\Rightarrow \mathbb{E}[2I] \leq \mathbb{E}[X]$$

$$\therefore \mathbb{E}[I] = \frac{1}{2} \Pr[X \geq 2] \quad (\text{see Fact})$$



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$$\forall i : F[i] \leq \tilde{F}[i] \leq F[i] + \underbrace{\varepsilon ||F||}_{\text{with probability } 1-\delta}$$

$$||F|| = \sum_{i=1}^n F[i]$$

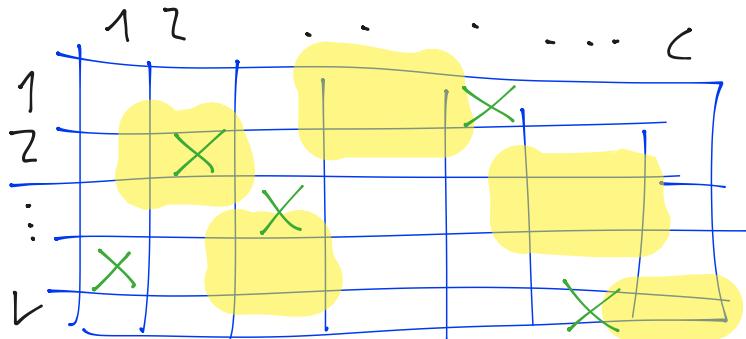
Algorithm

1. Let $r = \lceil \frac{1}{\delta} \rceil$ and $c = \frac{e}{\varepsilon}$
 $e = 2.71828$
Euler's constant
2. Let T be a table of $r \times c$ counters
initially set to zero

	1	2	.	.	.	c
1						
2						
:						
r						

3. Take a family \mathcal{H} of 2-wise independent hash functions
 $\text{eg } h \in \mathcal{H} \text{ iff } h(x) = [(ax+b) \bmod p] \bmod c + 1$
4. Choose h_1, h_2, \dots, h_r uniformly and independently from \mathcal{H}
 $\text{eg } r \text{ pairs } (a_1, b_1), \dots, (a_r, b_r)$
5. Given element i , associate r cells (one per row)
 $T(1, h_1(i)), T(2, h_2(i)), \dots, T(r, h_r(i))$

$i \neq i'$
 might have
 collisions



6. OPS
- $F[i]++ \Rightarrow$ increment by 1
 the r cells for i
- $F[i]-- \Rightarrow$ decrement, as above
7. $\tilde{F}[i] = \min_{1 \leq j \leq r} T(j, h_j(i))$

Fact 1

Storage is $O(r \cdot c) = O(\epsilon^{-1} \lg S^{-1})$
words of memory, where each word can store $\|F\|$.

Proof

Each entry stores an integer in $[\|F\|]$

Fact 2 $F[i] \leq \tilde{F}[i]$

Proof Let $T(j, h_j(i)) = \tilde{F}[i]$

By construction, $\exists i_1, i_2, \dots, i_l$ s.t. one of them is i and

$$T(j, h_j(i)) = \sum_{k=1}^l F[i_k] = F[i] + X_{ji}$$

where $X_{ji} \geq 0$ is the excess (as $F[i_k] \geq 0$)

obs To model the excess (due to hash collisions) use indicator variable QED

$$I_{jik} = \begin{cases} 1 & \text{if } k \neq i \text{ and } h_j(i) = h_j(k) \\ 0 & \text{otherwise} \end{cases}$$

$$\text{obs } X_{j,i} = \sum_{k=1}^n I_{j,i,k} \cdot F[k]$$

follows from the fact that $I_{j,i,k} = 1$ iff there is a collision in cell $T(j, h_j(i))$

$$\text{obs } E[I_{j,i,k}] = \frac{\epsilon}{c}$$

$$\text{proof } E[I_{j,i,k}] = 0 \cdot \Pr[I_{j,i,k}=0] + 1 \cdot \Pr[I_{j,i,k}=1]$$

$$= \Pr[I_{j,i,k}=1]$$

$$= \Pr[\exists d \in [C] : k \neq i \wedge h_j(i)=d \wedge h_j(k)=d]$$

$$= \sum_{d \in [C]} \Pr[k \neq i \wedge h_j(i)=d \wedge h_j(k)=d]$$

use the fact that h_j is 2-wise independent

$$= \sum_{d \in [C]} \Pr[h_j(i)=d] \times \underbrace{\Pr[k \neq i \wedge h_j(k)=d]}_{\frac{1}{C}}$$

"almost" $\frac{1}{C}$

since one element in the domain is missing

$$= \sum_{d \in [C]} \frac{1}{C^2} = \frac{1}{C} = \frac{\epsilon}{c}$$

Fact 3

$$\tilde{F}[i] \leq F[i] + \varepsilon \|F\|$$

Proof

with probability $\geq 1 - \delta$

Let $\tilde{F}[i] = F[i] + X_{ji}$, where $X_{ji} \geq 0$ is the excess

$$\text{D } \mathbb{E}[X_{ji}] = \mathbb{E}\left[\sum_k I_{jik} F[k]\right] = \sum_k \mathbb{E}[I_{jik} F[k]] \leq \sum_k (F[k] \mathbb{E}[I_{jik}]) = \sum_k (F[k] \cdot \frac{\varepsilon}{e}) = \frac{\varepsilon}{e} \|F\|$$

$$\mathbb{E}[X_{ji}] = \frac{\varepsilon}{e} \|F\| \Leftrightarrow \varepsilon \|F\| = e \mathbb{E}[X_{ji}]$$

$$\text{D } \Pr[\tilde{F}[i] > F[i] + \varepsilon \|F\|] =$$

$$\Pr[\tilde{F}[i] + X_{ji} > F[i] + \varepsilon \|F\|] =$$

$$\Pr[\forall j_i X_{ji} > \underbrace{\varepsilon \|F\|}_{e \mathbb{E}[X_{ji}]}] =$$

$$\Pr[\forall j_i X_{ji} > e \mathbb{E}[X_{ji}]] =$$

h_1, h_2, \dots, h_r chosen uniformly and independently

$$\Pr[X_{j_1} > e \mathbb{E}[X_{j_1}]] \times \Pr[X_{j_2} > e \mathbb{E}[X_{j_2}]] \times \dots \times \Pr[X_{j_r} > e \mathbb{E}[X_{j_r}]]$$

$$\hookrightarrow \text{Markov's inequality} \leq \frac{\mathbb{E}[X_{ji}]}{e \mathbb{E}[X_{ji}]} = \frac{1}{e} < \frac{1}{2} \quad 1 \leq i \leq r$$

$$\leq \left(\frac{1}{2}\right)^r = \left(\frac{1}{2}\right)^{\log_2 \frac{1}{\delta}} = S$$

Q.E.D

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