

HEAVY HITTERS
TAIL BOUNDS

ANNA KARLIN

STREAM MODEL

- Input elements (e.g. Google queries) enter/arrive one at a time.
- We cannot possibly store the stream.

Question: How do we make critical calculations about the data stream using a limited amount of memory?

SOURCES OF THIS KIND OF DATA

- Sensor data
 - E.g. millions of temperature sensors deployed in the ocean
- Image data from satellites or surveillance cameras
 - E.g. London
- Internet and web traffic
 - E.g. millions of streams of IP packets
- Web data
 - E.g. Search queries on Google, clicks on Bing, etc.

EXAMPLE APPLICATIONS

- Mining query streams
 - Google wants to know which queries are more frequent today than yesterday.
- Mining click streams
 - Facebook wants to know which of its ads are getting an unusual number of hits in the last hour.
- Mining social network news feeds
 - E.g., looking for trending topics on Twitter and Facebook, trending videos on TikTok

MORE APPLICATIONS

- Sensor networks
 - Many sensors feeding into a central controller.
- IP packets
 - Gather congestion information for optimal routing
 - Detect denial-of-service attacks

PROBLEM

- Input: sequence of n elements x_1, x_2, \dots, x_n from a known universe U (e.g., 8-byte integers).
- Goal: perform a computation on the input, in a single left to right pass where
 - Elements processed in real time
 - Can't store the full data. => minimal storage requirement to maintain working "summary"

HEAVY HITTERS: KEYS THAT OCCUR MANY TIMES

$$\frac{n}{4} = 3 \quad n=11$$

32, 12

32, 12, 14, 32, 7, 12, 32, 7, 32, 12, 4,

$$f_{32}^6 = 2$$

x_1, x_2, \dots, x_n

Applications:

- Determining popular products
- Computing frequent search queries
- Identifying heavy TCP

find all elts
that occur
more than $\frac{n}{100}$

$$\forall t \leq n$$

f_x^t : # times element x has
appeared in x_1, x_2, \dots, x_t

Goal: Find/output all elements with $f_x^n \geq \frac{n}{k}$
(These elts are "heavy hitters") (k=100)

Output has size $O(k)$

Provably impossible to solve this problem exactly with sublinear space

Modified goal: (solve ϵ -HH problem)

① If $f_x^n \geq \frac{n}{k}$, added to list we output.

② If some elt, say y , added to list, then w.p. $\geq 1 - \epsilon$

$$f_y^n \geq \frac{n}{k} - \epsilon n$$

How many elts can there be with $f_x^n \geq \frac{n}{10}$?

a) ∞
b) n

c) 100
d) 10

Suppose $k = 20$ $\epsilon = 0.01$

$$\epsilon = \frac{1}{100}$$

① $f_x^n \geq 0.05n$ times, then in our list

② for every y in list w.p. $\geq 1 - \frac{1}{100}$, $f_y^n \geq 0.05n - \frac{n}{100}$

$k, \epsilon, \delta.$



determine

$b, \ell.$

0.014
 $= 0.04\%$

COUNT-MIN SKETCH

- Maintain a short summary of the information that still enables answering queries.
- Cousin of the Bloom filter
 - Bloom Filter solves the “membership problem”.
 - We want to extend it to solve a counting problem.

COUNT-MIN SKETCH

Want:

① If $f_x^n \geq \frac{n}{k}$, added to ^{HH} list ✓ we output.

② If some elt, say y , added to list, then w.p. $\geq 1 - \delta$
 $f_y^n \geq \frac{n}{k} - \epsilon n$

designer specifies $k, \epsilon, \delta \Rightarrow \underbrace{b, \ell}$

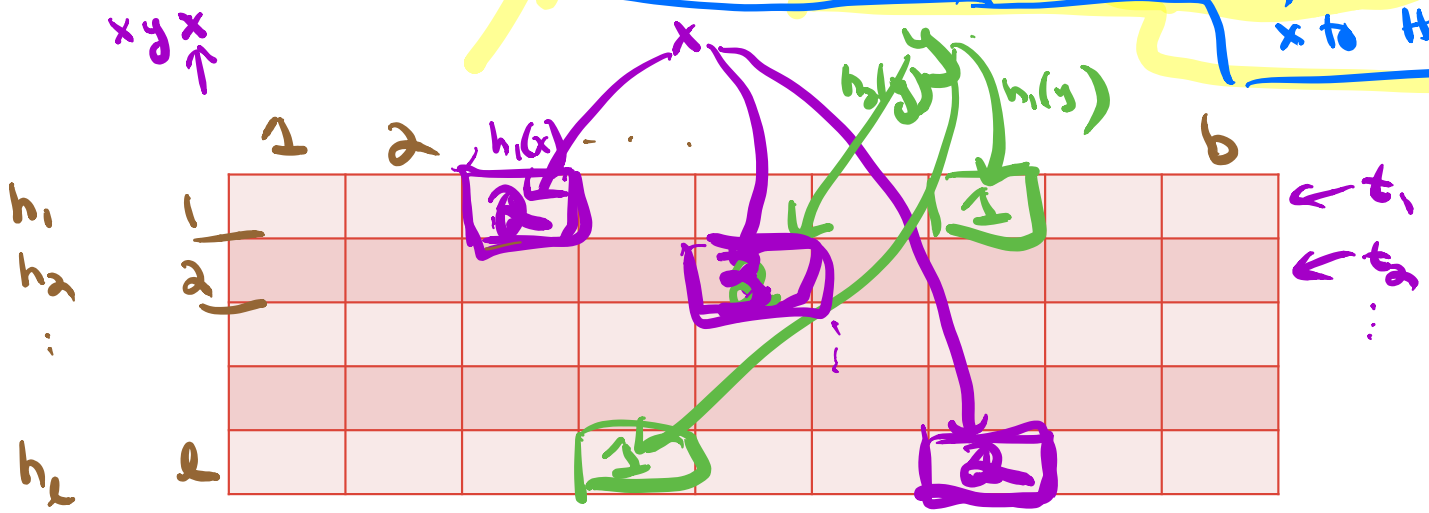
keep 2D array
 ℓ hash tables, each of size b .

when elt x shows up.

Update (x): $\forall 1 \leq j \leq \ell$ increment $t_j[h_j(x)]$

Count (x): return $\min_{1 \leq j \leq \ell} t_j[h_j(x)]$

if $\text{Count}(x) \geq \frac{b}{k}$, add x to HT list



after $x_1 \dots x_t$ arrived
 $t_j[h_j(x)]$ guaranteed

- a) $\leq f_x^+$
- b) $= f_x^+$
- c) $\geq f_x^+$
- d) I don't know

after $x_1 \dots x_t$ arrived
Count(x) guaranteed

- a) $\leq f_x^+$
- b) $= f_x^+$
- c) $\geq f_x^+$
- d) I don't know

Assumptions

① hash functions behave like random maps
 $h_1, \dots, h_\ell : \mathcal{U} \rightarrow \{0, 1, \dots, b-1\}$

$$\forall x \neq y \quad \Pr(h_j(x) = h_j(y)) = \frac{1}{b}$$

② hash fns h_1, \dots, h_ℓ
are indep of each other.

$$b \cdot \frac{1}{b^2} = \frac{1}{b}$$

If hash fn h
random, $x \neq y$
 $\Pr(h(x) = h(y)) =$

a) $\frac{1}{b}$

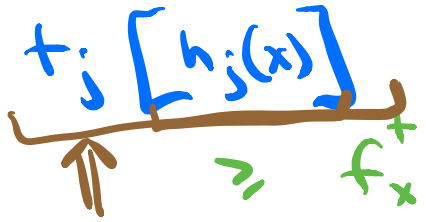
b) $\frac{1}{b^2}$

c) 0

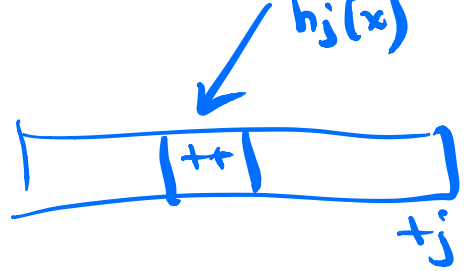
d) I don't know

Fix time t . (x_1, \dots, x_t have just arrived)

$$Z_j^t \equiv$$



(# of times x occurs in x_1, \dots, x_t)



$$Z_j^t = f_x^+ + \sum_{\substack{\text{distinct} \\ \text{els } y \neq x \\ y \in \{x_1, \dots, x_t\}}} f_y^+ w_{xy}$$

$$w_{xy} = \begin{cases} 1 & \text{if } h_j(x) = h_j(y) \\ 0 & \text{o.w.} \end{cases}$$

$$E(Z_j) = f_x^t + \underbrace{\sum_{\substack{\text{distinct } y \\ y \neq x}} f_y^+}_{\frac{t - f_x^t}{b}} E(w_{xy})$$

$t \leq t \leq n$

$$E(z_j) \leq f_x^+ + \boxed{\sigma/s} \quad \equiv$$

$$E(z_j - f_x^+) \leq \frac{\sigma}{s}$$

$$\Pr(z_j - f_x^+ \geq \frac{2\sigma}{s}) \leq \frac{\sigma}{s}$$

$$\Pr(X \geq \frac{2\sigma}{s})$$

$$\leq \Pr(X \geq 2E(X))$$



Markov's Inequality.

If $X \geq 0$ r.v.

$$\Pr(X \geq cE(X)) \leq \frac{1}{c}$$

COUNT-MIN SKETCH

- Elegant small space data structure.
- Space used is independent of n .
- Is implemented in several real systems.
 - AT&T used in network switches to analyze network traffic.
 - Google uses a version on top of Map Reduce parallel processing infrastructure and in log analysis.
- Huge literature on sketching and streaming algorithms (algorithms like Distinct Elements, Heavy Hitters and many many other very cool algorithms).

