

# CountMin and Count Sketches

Lecture 10

February 14, 2019

# Heavy Hitters Problem

**Heavy Hitters Problem:** Find all items  $i$  such that  $f_i > m/k$  for some fixed  $k$ .

Heavy hitters are **very** frequent items.

We saw Misra-Gries deterministic algorithm that in  $O(k)$  space finds the heavy hitters assuming they exist. Two pass algorithm correctly identifies heavy hitters.

# (Strict) Turnstile Model

- Turnstile model: each update is  $(i_j, \Delta_j)$  where  $\Delta_j$  can be positive or negative
- Strict turnstile: need  $x_i \geq 0$  at all time for all  $i$

In terms of frequent items we want additive error to  $x_i$

# Basic Hashing/Sampling Idea

**Heavy Hitters Problem:** Find all items  $i$  such that  $f_i > m/k$ .

- Let  $b_1, b_2, \dots, b_k$  be the  $k$  heavy hitters
- Suppose we pick  $h : [n] \rightarrow [ck]$  for some  $c > 1$
- $h$  spreads  $b_1, \dots, b_k$  among the buckets ( $k$  balls into  $ck$  bins)
- In ideal situation each bucket can be used to count a separate heavy hitter

# Part I

## CountMin Sketch

# CountMin Sketch

[Cormode-Muthukrishnan]

**CountMin-Sketch( $w, d$ ):**

$h_1, h_2, \dots, h_d$  are pair-wise independent hash functions  
from  $[n] \rightarrow [w]$ .

While (stream is not empty) do

$e_t = (i_t, \Delta_t)$  is current item

    for  $\ell = 1$  to  $d$  do

$C[\ell, h_\ell(i_j)] \leftarrow C[\ell, h_\ell(i_j)] + \Delta_t$

    endWhile

For  $i \in [n]$  set  $\tilde{x}_i = \min_{\ell=1}^d C[\ell, h_\ell(i)]$ .

Counter  $C[\ell, j]$  simply counts the sum of all  $x_i$  such that  $h_\ell(i) = j$ .

That is,

$$C[\ell, j] = \sum_{i: h_\ell(i)=j} x_i.$$

# Intuition

- Suppose there are  $k$  heavy hitters  $b_1, b_2, \dots, b_k$
- Consider  $b_i$ : Hash function  $h_\ell$  sends  $b_i$  to  $h_\ell(b_i)$ .  $C[\ell, h(b_i)]$  counts  $x_{b_i}$  and also other items that hash to same bucket  $h(b_i)$  so we always overcount (since strict turnstile model)
- Repeating with many hash functions and taking *minimum* is right thing to do: for  $b_i$  the goal is to avoid other heavy hitters colliding with it

# Property of CountMin Sketch

## Lemma

Let  $d = \Omega(\log \frac{1}{\delta})$  and  $w > \frac{2}{\epsilon}$ . Then for any fixed  $i \in [n]$ ,  $x_i \leq \tilde{x}_i$  and

$$\Pr[\tilde{x}_i \geq x_i + \epsilon \|x\|_1] \leq \delta.$$



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- Unlike Misra-Greis we have over estimates
- Actual items are not stored (requires work to recover heavy hitters)
- Works in strict turnstile model and hence can handle deletions
- Space usage is  $O(\frac{\log(1/\delta)}{\epsilon})$  counters and hence  $O(\frac{\log(1/\delta)}{\epsilon} \log m)$  bits

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$$\Pr[Z_\ell] \geq x_i + \epsilon \|x\|_1 \leq 1/2$$

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Since the  $d$  hash functions are independent

$$\Pr[\min_\ell Z_\ell \geq x_i + \epsilon \|x\|_1] \leq 1/2^d \leq \delta$$

# Summarizing

## Lemma

Let  $d = \Omega(\log \frac{1}{\delta})$  and  $w > \frac{2}{\epsilon}$ . Then for any fixed  $i \in [n]$ ,  $x_i \leq \tilde{x}_i$  and

$$\Pr[\tilde{x}_i \geq x_i + \epsilon \|x\|_1] \leq \delta.$$

Choose  $d = 2 \ln n$  and  $w = 2/\epsilon$ : we have

$$\Pr[\tilde{x}_i \geq x_i + \epsilon \|x\|_1] \leq 1/n^2.$$

By union bound, with probability  $(1 - 1/n)$ , for all  $i \in [n]$ ,

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Total space  $O(\frac{1}{\epsilon} \log n)$  counters and hence  $O(\frac{1}{\epsilon} \log n \log m)$  bits.

# CountMin as a Linear Sketch

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Recall that for  $1 \leq \ell \leq d$  and  $1 \leq s \leq w$ :

$$C[\ell, s] = \sum_{i: h_\ell(i)=s} x_i$$

Thus, once hash function  $h_\ell$  is fixed:

$$C[\ell, s] = \langle u, x \rangle$$

where  $u$  is a row vector in  $\{0, 1\}^n$  such that  $u_i = 1$  if  $h_\ell(i) = s$  and  $u_i = 0$  otherwise

Thus, once hash functions are fixed, the counter values can be written as  $Mx$  where  $M \in \{0, 1\}^{wd \times n}$  is the sketch matrix

# Part II

## Count Sketch

# Count Sketch

[Charikar-Chen-FarachColton]

**Count-Sketch( $w, d$ ):**

$h_1, h_2, \dots, h_d$  are pair-wise independent hash functions  
from  $[n] \rightarrow [w]$ .

$g_1, g_2, \dots, g_d$  are pair-wise independent hash functions  
from  $[n] \rightarrow \{-1, 1\}$ .

While (stream is not empty) do

$e_t = (i_t, \Delta_t)$  is current item

    for  $\ell = 1$  to  $d$  do

$C[\ell, h_\ell(i_t)] \leftarrow C[\ell, h_\ell(i_t)] + g_\ell(i_t)\Delta_t$

    endWhile

For  $i \in [n]$

    set  $\tilde{x}_i = \text{median}\{g_1(i)C[1, h_1(i)], \dots, g_d(i)C[d, h_d(i)]\}$ .

Like CountMin, Count sketch has  $wd$  counters. Now counter values can become negative even if  $x$  is positive.

# Intuition

- Each hash function  $h_\ell$  spreads the elements across  $w$  buckets
- The hash function  $g_\ell$  induces cancellations (inspired by  $F_2$  estimation algorithm)
- Since answer may be negative even if  $x \geq 0$ , we take the median

**Exercise:** Show that Count sketch is also a linear sketch.

# Count Sketch Analysis

## Lemma

Let  $d \geq 4 \log \frac{1}{\delta}$  and  $w > \frac{3}{\epsilon^2}$ . Then for any fixed  $i \in [n]$ ,  $\mathbf{E}[\tilde{x}_i] = x_i$  and

$$\Pr[|\tilde{x}_i - x_i| \geq \epsilon \|x\|_2] \leq \delta.$$

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## Comparison to CountMin

- Error guarantee is with respect to  $\|\mathbf{x}\|_2$  instead of  $\|\mathbf{x}\|_1$ . For  $\mathbf{x} \geq \mathbf{0}$ ,  $\|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_1$  and in some cases  $\|\mathbf{x}\|_2 \ll \|\mathbf{x}\|_1$ .
- Space increases to  $O(\frac{1}{\epsilon^2} \log n)$  counters from  $O(\frac{1}{\epsilon} \log n)$  counters



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For  $i' \in [n]$  let  $Y_{i'}$  be the indicator random variable that is **1** if  $h_\ell(i) = h_\ell(i')$ ; that is  $i$  and  $i'$  collide in  $h_\ell$ .

$E[Y_{i'}] = E[Y_{i'}^2] = 1/w$  from pairwise independence of  $h_\ell$ .

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Therefore,

$$E[Z_\ell] = x_i + \sum_{i' \neq i} E[g_\ell(i)g_\ell(i')Y_{i'}]x_{i'} = x_i,$$

because  $E[g_\ell(i)g_\ell(i')] = 0$  for  $i \neq i'$  from pairwise independence of  $g_\ell$  and  $Y_{i'}$  is independent of  $g_\ell(i)$  and  $g_\ell(i')$ .

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$$\begin{aligned}\text{Var}(Z_\ell) &= \mathbf{E}[(Z_\ell - x_i)^2] \\ &= \mathbf{E}\left[\left(\sum_{i' \neq i} g_\ell(i)g_\ell(i')Y_{i'}x_{i'}\right)^2\right] \\ &= \mathbf{E}\left[\sum_{i' \neq i} x_{i'}^2 Y_{i'}^2 + \sum_{i' \neq i''} x_{i'}x_{i''} g_\ell(i')g_\ell(i'')Y_{i'}Y_{i''}\right] \\ &= \sum_{i' \neq i} x_{i'}^2 \mathbf{E}[Y_{i'}^2] \\ &\leq \|\mathbf{x}\|_2^2/w.\end{aligned}$$

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We have seen:  $\mathbf{E}[Z_\ell] = x_i$  and  $\mathbf{Var}(Z_\ell) \leq \|x\|_2^2/w$ .

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Using Chebyshev:

$$\Pr[|Z_\ell - x_i| \geq \epsilon \|x\|_2] \leq \frac{\mathbf{Var}(Z_\ell)}{\epsilon^2 \|x\|_2^2} \leq \frac{1}{\epsilon^2 w} \leq 1/3.$$



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Via the Chernoff bound,

$$\Pr[|\text{median}\{Z_1, \dots, Z_d\} - x_i| \geq \epsilon \|x\|_2] \leq e^{-cd} \leq \delta.$$

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## Lemma

Let  $d \geq 4 \log \frac{1}{\delta}$  and  $w > \frac{3}{\epsilon^2}$ . Then for any fixed  $i \in [n]$ ,  $\mathbf{E}[\tilde{x}_i] = x_i$  and  $\Pr[|\tilde{x}_i - x_i| \geq \epsilon \|x\|_2] \leq \delta$ .

Choose  $d = \theta(\ln n)$  and  $w = 3/\epsilon^2$ : we have

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