

CountMin and Count Sketches

Lecture 09

September 22, 2020

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.

- Identifies correct heavy hitters if they exist but can make a mistake if they don't and need second pass to verify
- Cannot handle deletions

(Strict) Turnstile Model

- Turnstile model: each update is (i_j, Δ_j) where Δ_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
- Use multiple independent hash functions to improve estimate

Part I

CountMin Sketch

CountMin Sketch: Offline view

- d independent hash functions h_1, h_2, \dots, h_d . Each hash function is pair-wise independent
- Each $h_\ell : [n] \rightarrow [w]$ (hence maps to w buckets)
- Store one number per bucket and hence total of dw numbers which can be viewed as d -day array (d rows, w columns).
 $C[\ell, s]$ is the counter for bucket s for hash function h_ℓ .
- Let $x \in \mathbb{R}^n$ be the given vector. For $1 \leq \ell \leq d$, $1 \leq s \leq w$

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

hence it keeps track of sum of all coordinates that h_ℓ maps to bucket s

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

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

Intuition

- Suppose there are k heavy hitters b_1, b_2, \dots, b_k
- Consider b_i : Hash function h_ℓ 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

Consider strict turnstile mode ($x \geq 0$). 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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$$\mathbb{E}[Z_\ell] = x_i + \sum_{i' \neq i} \Pr[h_\ell(i') = h_\ell(i)] x_{i'}$$

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By pairwise-independence

$$\mathbf{E}[Z_\ell] = x_i + \sum_{i' \neq i} x_{i'} / w \leq x_i + \epsilon \|x\|_1 / 2$$

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Via Markov applied to $Z_\ell - x_i$ (we use strict turnstile here)

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

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Via Markov applied to $Z_\ell - x_i$ (we use strict turnstile here)

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

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 $d > \log \frac{1}{\delta}$ 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 $w = 2/\epsilon$. Then

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

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

$$\tilde{x}_i \leq x_i + \epsilon \|x\|_1$$

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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

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

Corollary

With $d = \Omega(\ln n)$ and $w = 2/\epsilon$, with probability $(1 - \frac{1}{n})$ for all $i \in [n]$:

$$\tilde{x}_i \leq x_i + \epsilon \|x\|_1$$

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

Question: Why is CountMin 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_ℓ 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

- Similar to CountMin use d hash functions each with w buckets each and hence array of dw counters
- Inspired by F_2 estimation use additional $\{-1, 1\}$ hash functions which creates negative values
- Use median estimate

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_ℓ spreads the elements across w buckets
- The has function g_ℓ 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.

Property of Count Sketch

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

Analysis

Fix an $i \in [n]$ and $\ell \in [d]$. Let $Z_\ell = g_\ell(i)C[\ell, h_\ell(i)]$.

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

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

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$$Z_\ell = g_\ell(i)C[\ell, h_\ell(i)] = g_\ell(i) \sum_{i'} g_\ell(i') x_{i'} Y_{i'}$$

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$$Z_\ell = g_\ell(i)C[\ell, h_\ell(i)] = g_\ell(i) \sum_{i'} g_\ell(i') x_{i'} Y_{i'}$$

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_ℓ and $Y_{i'}$ is independent of $g_\ell(i)$ and $g_\ell(i')$.

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

Analysis

$$Z_\ell = g_\ell(i)C[\ell, h_\ell(i)].$$

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

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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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Using Chebyshev:

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Via the Chernoff bound,

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

Summarizing

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

Corollary

With $d = \Theta(\ln n)$ and $w = 3/\epsilon^2$, with probability $(1 - \frac{1}{n})$ for all $i \in [n]$:

$$|\tilde{x}_i - x_i| \leq \epsilon \|x\|_2.$$

Total space: $O(\frac{1}{\epsilon^2} \log n)$ counters and hence $O(\frac{1}{\epsilon^2} \log n \log m)$ bits.