CS 498ABD: Algorithms for Big Data

Heavy Hitters

Lecture 08 September 17, 2020

Models

Richer model:

- Want to estimate a function of a vector x ∈ ℝⁿ which is initially assume to be the all 0's vector.
- Each element e_j of a stream is a tuple (i_j, Δ_j) where i_j ∈ [n] and Δ_i ∈ ℝ is a real-value: this updates x_{ij} to x_{ij} + Δ_j. (Δ_j can be positive or negative)

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- $\Delta_j > 0$: cash register model. Special case is $\Delta_j = 1$.
- Δ_i arbitrary: *turnstile* model
- Δ_j arbitrary but $x \ge 0$ at all times: *strict turnstile* model
- *Sliding window* model: interested only in the last *W* items (window)

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

Heavy hitters are **very** frequent items.

Majority element problem:

- Offline: given an array/list *A* of *m* integers, is there an element that occurs more than *m*/2 times in *A*?
- Streaming: is there an *i* such that $f_i > m/2$?

```
Streaming-Majority:
     c = 0, s \leftarrow null
     While (stream is not empty) do
          If (e_i = s) do
              c \leftarrow c + 1
          ElseIf (c = 0)
               c = 1
               s = e_i
          Else
               c \leftarrow c - 1
     endWhile
     Output s, c
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Claim: If there is a majority element *i* then algorithm outputs s = i and $c \ge f_i - m/2$.

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Claim: If there is a majority element *i* then algorithm outputs s = i and $c \ge f_i - m/2$. **Caveat:** Algorithm may output incorrect element if no majority element. Can verify correctness in a second pass.

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Misra-Gries Algorithm

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

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MisraGreis(k):
     D is an empty associative array
    While (stream is not empty) do
          e; is current item
          If (e; is in keys(D))
               D[e_i] \leftarrow D[e_i] + 1
          Else if (|keys(A)| < k - 1) then
          D[e_i] \leftarrow 1
         Else
              for each \ell \in keys(D) do
                    D[\ell] \leftarrow D[\ell] - 1
          Remove elements from D whose counter values are 0
endWhile
For each i \in keys(D) set \hat{f}_i = D[i]
For each i \notin keys(D) set \hat{f}_i = 0
```

Analysis

Space usage O(k).

Theorem

For each
$$i \in [n]$$
: $f_i - \frac{m}{k+1} \leq \hat{f}_i \leq f_i$.

Corollary

Any item with $f_i > m/k$ is in D at the end of the algorithm.

A second pass to verify can be used to verify correctness of elements in D.

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Alternative view of algorithm:

- Maintains counts *C*[*i*] for each *i* (initialized to 0). Only *k* are non-zero at any time.
- When new element e_j comes
 - If $C[e_j] > 0$ then increment $C[e_j]$
 - Elself less then k positive counters then set $C[e_j] = 1$
 - Else decrement all positive counters (exactly *k* of them)

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• Output $\hat{f}_i = C[i]$ for each i

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• Hence total number of times α increases is at most ℓ .

Deterministic to Randomized Sketches

Cannot improve O(k) space if one wants additive error of at most m/k. Nice to have a deterministic algorithm that is near-optimal

Why look for randomized solution?

- Obtain a sketch that allows for deletions
- Additional applications of sketch based solutions
- Will see Count-Min and Count sketches

Basic Hashing/Sampling Idea

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

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