

CS7960 L9 : Streaming I Heavy Hitters = Approximate Counts

Streaming Algorithms

Stream : $A = \langle a_1, a_2, \dots, a_m \rangle$

a_i in $[n]$ size $\log n$

Compute $f(A)$ in $\text{poly}(\log m, \log n)$
space

Let $f_j = |\{a_i \text{ in } A \mid a_i = j\}|$

MAJORITY: if some $f_j > m/2$, output
 j

else, output
NULL

one-pass requires $\Omega(\min\{m, n\})$
space

Simpler:

FP-MAJORITY: if some $f_j > m/2$,
output j

else,

output anything

How good w/ $O(\log m + \log n)$ (one
counter c + one location l)?

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c = 0, l = X
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for (a_i \in A)
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    if (a_i = l) c += 1
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    else          c -= 1
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    if (c <= 0)  c = 1, l = a_i
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return l
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Analysis: if $f_j > m/2$, then

if ($l \neq j$) then c decremented at
most $< m/2$ times, but $c > m/2$

if ($l == j$) can be decremented $< m/$

2, but is incremented $> m/2$
if $f_j < m/2$ for all j , then any
answer ok.

----- another view of analysis -----

Let $f_j > m/2$, and $k = m - f_j$.

After s steps, let $g_s =$ unseen
elements of index j

let $k_s =$ unseen
elements \neq index j

let $c_s = c$ if $l \neq j$,
and $-c$ if $l = j$

Claim: $g_s > c + k_s$

base case ($s=0$, or even $s=1$) easily
true.

Inductively 4 cases:

$a_i = l = j$: (g_s decremented, c
decremented)

$a_i = l \neq j$: (c incremented, k_s
decremented)

$a_i \neq l \neq j$: (c decremented, k_s
decremented)

$a_i \neq l = j$: (k_s decremented,

maybe c incremented)

Since at the end $g_s = k_s = 0$, then
 $0 > c + 0$, implies $c < 0$, and
 $l=j$.

FREQUENT: for k , output the set $\{j : f_j > m/k\}$
also hard.

k -FREQUENCY-ESTIMATION: Build data structure S .

For any j in $[n]$, $\hat{f}_j = S(j)$
s.t.

$$f_j - m/k \leq \hat{f}_j \leq f_j$$

aka ϵ -approximate ϕ -HEAVY-HITTERS:

Return all f_j s.t. $f_j > \phi$

Return no f_j s.t. $f_j < \phi - \epsilon * m$

(any f_j s.t. $\phi - \epsilon \cdot m < f_j < \phi$
is ok)

Misra-Gries Algorithm [Misra-Gries
'82]

Solves k -FREQUENCY-ESTIMATION in
 $O(k(\log m + \log n))$ space.

Let C be array of k counters $C[1],$
 $C[2], \dots, C[k]$

Let L be array of k locations $L[1],$
 $L[2], \dots, L[k]$

#####

Set all $C = 0$

Set all $L = X$

for (a_i in A)

 if (a_i in L) <at index j >

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    C[j] += 1
else      <a_i !in L>
    if (|L| < k)
        C[j] = 1
        L[j] = a_i
    else
        C[j] -= 1 forall j in [k]
for (j in [k])
    if (C[j] <= 0) set L[j] = X
#####
On query q in [n]
    if (q in L {L[j]=q}) return hat{f}
_q = C[j]
    else      return hat{f}
_q = 0
#####

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Analysis

A counter $C[j]$ representing $L[j] = q$ is only incremented if $a_i = q$

$$\hat{f}_q \leq f_q$$

If a counter $C[j]$ representing $L[j] = q$ is decremented,
then $k-1$ other counters are also decremented.

This happens at most m/k times.

A counter $C[j]$ representing $L[j] = q$ is decremented at most m/k times.

$$f_q - m/k \leq \hat{f}_q$$

How do we get an additive ϵ -approximate FREQUENCY-ESTIMATION ?
i.e. return \hat{f}_q s.t.

$$|f_q - \hat{f}_q| \leq \epsilon * m$$

Set $k = 2/\epsilon$, return $C[j] + (m/k)/2$

Space $O((1/\epsilon) (\log m + \log n))$

Also:

ϵ -approximate ϕ -HEAVY-HITTERS for
any $\phi > m \cdot \epsilon$ in
space $O((1/\epsilon) (\log m + \log n))$

Can solve k -FREQUENT optimally in two
passes w/ $O(k(\log n + \log m))$ space.
Run M-G algorithm w/ k counters.
For each stored location, make second
pass and count exactly.