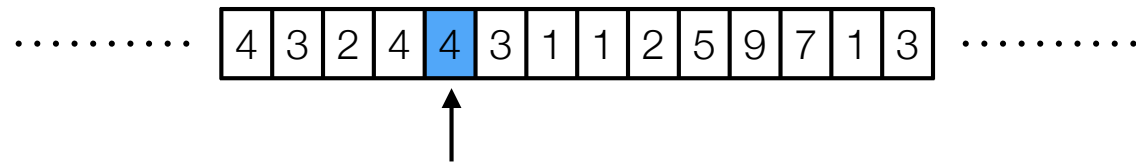


Streaming

Inge Li Gørtz

Streaming model (one-pass)

- **Stream.**
 - Elements a_1, a_2, \dots, a_m from the universe $[n] = \{1, 2, \dots, n\}$.
 - Elements arrive one by one.
 - Must process element a_i before we see a_{i+1} .



- **Space.** Measured in bits.
- **Goal.** Small space (sublinear/polylogarithmic).
- **Example.** What can we do in $O(\log n + \log m)$ space?

Frequent elements

Frequent elements

- **Heavy Hitters Problem.** Find all elements i that occurs more than m/k times for some fixed k .
- **Example.** Return all elements that occur more than $21/3$ times = 7.

4	4	1	2	4	4	3	1	1	2	5	9	7	4	1	3	4	1	4	4	1
---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---

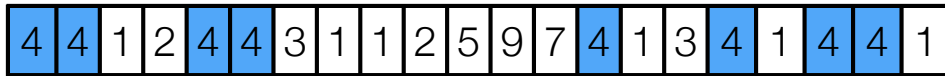
Frequent elements

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4	4	1	2	4	4	3	1	1	2	5	9	7	4	1	3	4	1	4	4	1
---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---

Frequent elements

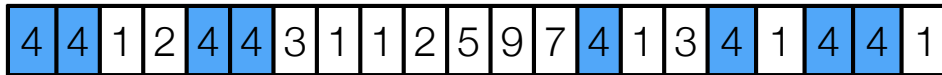
- **Heavy Hitters Problem.** Find all elements i that occurs more than m/k times for some fixed k .
- **Example.** Return all elements that occur more than $21/3$ times = 7.



- **Bad news.** Need $\Omega(n)$ space for one-pass algorithm.
- **Good news.**
 - Can estimate the frequency.
 - Can do better if we allow one-sided error:
 - Output all elements that occur more than m/k times.
 - Might also output other elements.

Frequent elements

- **Heavy Hitters Problem.** Find all elements i that occurs more than m/k times for some fixed k .
- **Example.** Return all elements that occur more than $21/3$ times = 7.



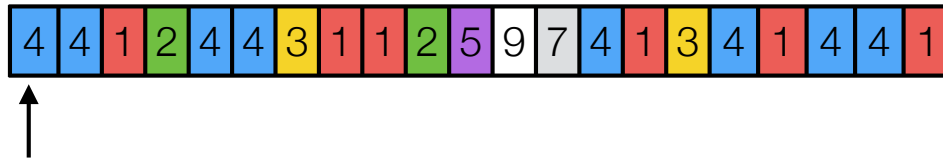
- **Misra-Gries.**

```
Keep  $k-1$  counters in an associative array  $A$ .
while (stream is not empty) do
  if  $j \in \text{keys}(A)$  then
     $A[j] \leftarrow A[j] + 1$ 
  else if  $|\text{keys}(A)| < k - 1$  then
     $A[j] \leftarrow 1$ 
  else
    Decrement all counters by 1.
    Remove all elements with counter 0.
Output all elements in  $\text{keys}(A)$ 
```

- **Space.** $O(k \cdot (\log n + \log m))$.

Frequent elements

- **Example.** $k = 3$.



counter 1: 4 , 2

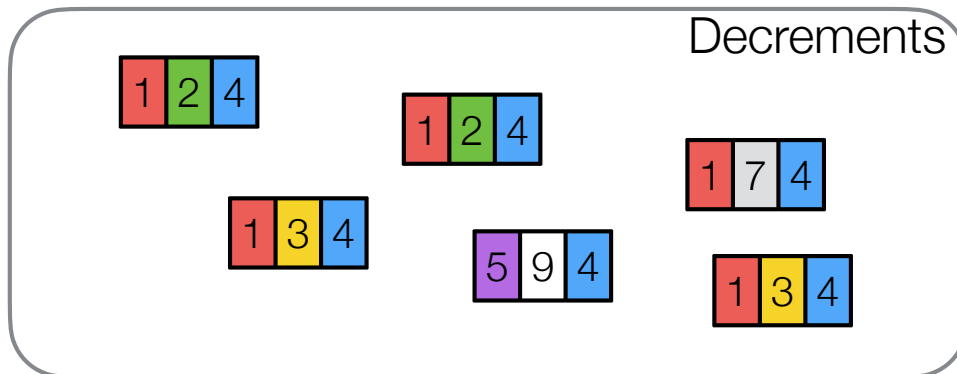
counter 2: 1 , 1

```
Keep  $k-1$  counters in an associative array  $A$ .
while (stream is not empty) do
  if  $j \in \text{keys}(A)$  then
     $A[j] \leftarrow A[j] + 1$ 
  else if  $|\text{keys}(A)| < k - 1$  then
     $A[j] \leftarrow 1$ 
  else
    Decrement all counters by 1.
    Remove all elements with counter 0.
Output all elements in  $\text{keys}(A)$ 
```


Misra-Gries Analysis

- **Lemma.** Any item with frequency more than m/k is in A by the end of the algorithm.
- **Lemma.** Let \hat{f}_i be the estimate of the frequency of element i . Then

$$f_i - \frac{m}{k} \leq \hat{f}_i \leq f_i .$$



Reservoir Sampling

Reservoir Sampling

- **Algorithm.**

```
put the first  $k$  elements into a “reservoir”  $R = \{r_1, r_2, \dots, r_k\}$ .  
for  $i > k$  until the stream is empty do  
    with probability  $k/i$  replace a random entry of  $R$  with  $a_i$   
Return  $R$ .
```

- **Claim.** For all $t \geq i$, $P[a_i \in R_t] = k/t$, where R_t denotes the reservoir after time t .
- **Proof.** By induction on t .
 - Base case: $t = k$ (assume stream has length at least k).

$$P[x_i \in R_t] = 1 = k/t.$$

Reservoir Sampling

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- **Claim.** For all $t \geq i$, $P[a_i \in R_t] = k/t$, where R_t denotes the reservoir after time t .

- **Proof.** By induction on t . Assume $t > k$.

- Case 1. $i = t$.

$$P[a_i \in R_t] = P[a_t \in R_t] = k/t.$$

Reservoir Sampling

- **Algorithm.**

```
put the first  $k$  elements into a “reservoir”  $R = \{r_1, r_2, \dots, r_k\}$ .  
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Return  $R$ .
```

- **Claim.** For all $t \geq i$, $P[a_i \in R_t] = k/t$, where R_t denotes the reservoir after time t .
- **Proof.** By induction on t . Assume $t > k$. Case 2. $i < t$.

$$\begin{aligned} P[x_i \in R_t] &= P[x_i \in R_{t-1}] \cdot P[x_i \text{ not replaced at time } t \mid x_i \in R_{t-1}] \\ &= \frac{k}{t-1} \cdot P[x_i \text{ not replaced at time } t \mid x_i \in R_{t-1}] \\ &= \frac{k}{t-1} \left(1 - \frac{k}{t} \cdot \frac{1}{k} \right) = \frac{k}{t} \end{aligned}$$