

# Operations on distributed data sketches

$$M = (S, m)$$

- $S$  - fundamental set
- $m: S \rightarrow \mathbb{N}_{\geq 1}$   
number of occurrences of an element

## 1) Data sketch

1, 2, 3, 1, 2, 3  
stream

$\rightsquigarrow$  SKETCH

$\downarrow$   
number of unique  
elements in the stream  
 $n = |S| = 3$

• memory required to have  
the exact answer is  $O(n)$   
(store each unique element)

• we look for a solution with  
memory of size like  $O(\log N)$  or  
even better  $O(\log \log N)$

$(1, w_1), (2, w_2), (3, w_3), (1, w_1), \dots$  - stream of pairs (val, size)

$\downarrow$   
SKETCH  $\rightsquigarrow$  total size of unique elements  $|S|_w \times w_1 + w_2 + w_3$   
where  $S = \{(1, w_1), (2, w_2), \dots\}$   
and  $|S|_w = \sum_{(i, w_i)} w_i$

## 2) Operations on data sketches

node A  $\rightsquigarrow$  set A  $\rightsquigarrow$  sketch A  
node B  $\rightsquigarrow$  set B  $\rightsquigarrow$  sketch B

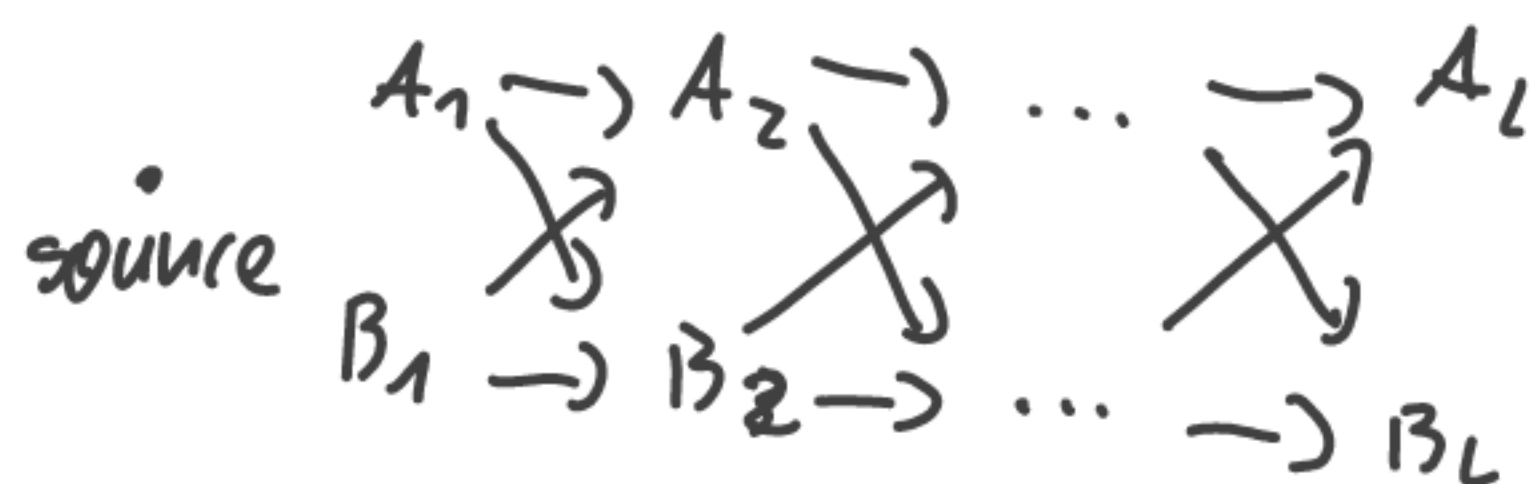
- what's  $|A \cap B|$ ,  $|A \cup B|$ ,  $|A \setminus B|$ ,  $|A \cap B|_w$ ,  $|A \setminus B|_w$  ...
- the idea: construct a solution that allows to estimate/minimize set theory operations based on data sketches and use any number of sketches and any sequence of operations on those sketches.

### 3) Applications

- network traffic analysis

$A_j, B_j$  - denote the set of packets observed by  $A_j, B_j$

Broad-chain



size of packets that go through a given path:

$$|A_1 \cap A_2 \cap \dots \cap A_L|_w$$

- Query to distributed database

sketch  $A \cup B \rightarrow$  estimate  $\pm 10\%$  for up to  $4 \cdot 10^8$  element (?)

Hospital A  
 $p_1$ -cov  
 $p_2$ -cov

Hospital B  
 $p_1$ -cov  
 $p_3$ -cov

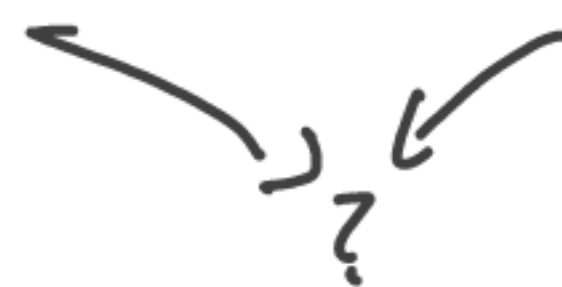


Hospital A

Hospital B

sketch A

sketch B





#### 4) Problem formalization

- Let  $M = (S, n)$  be a multiset, where  $S$  is a fundamental set and  $m: S \rightarrow \mathbb{N}_{\geq 1}$  a multiplicity function
- Problem:** We observe element of  $M$  sequentially and we try to estimate  $n = |S|$  using memory of size  $o(n)$ , e.g.  $O(\log n)$  or  $O(\log \log(n))$
- MinCount**, **HyperLogLog**  $\leadsto$  counting problem (number of unique elements)  
1990 2007, 2016
- ExpSketches**  $\leadsto$  total weights of unique elements.  
2009, 2017

#### 5) MinCount( $k, h, M$ )

$M \leftarrow (1, 1, \dots, 1) \quad |M| = k$   
 for each  $s \in M$   
 if  $h(s) \notin M \wedge h(s) < M[k]$   
      $M[k] \leftarrow h(s)$   
     sort( $M$ ) // increasing order

anytime we can estimate  $n = |S|$

if  $M[k] = 1$  return  $|\{i: M[i] \neq 1\}|$   
 else return  $\hat{n} = \frac{k-1}{M[k]} \leadsto E[\hat{n}] = n$

//  $k \geq 3$

$M = (S, m)$

$h: S \rightarrow \{0, 1\}^B \in [0, 1]$

$h(s_i) \sim \text{Uniform dist}$

$h(s_1), h(s_2), \dots$  independent

$h$  is hash

#### 6) Memory consumption

- we have a fixed size of array  $M$ ,  $k$  doesn't depend on  $n$ .
- what is the length  $B$  of hash value to avoid collisions?

Birthday paradox  $\sqrt{2^B} = n$ , we have small probability of collision  
 $\leadsto B = 2 \log_2 n$

- $k$  hash values of length  $B = 2 \log_2 n \rightarrow \text{memory } O(\log n)$

## 7) Precision of estimate

### Definition:

Let  $X_1, \dots, X_n$  be a sequence of any random variables.  
We sort their realisations in an ascending order:

$$X_{1:n} \leq \dots \leq X_{n:n}$$

Variable  $X_{i:n}$  we call the  $i$ -th order statistic.

For example:

$$X_{1:n} = \min(X_1, \dots, X_n)$$

$$X_{n:n} = \max(X_1, \dots, X_n)$$

We assume that  $h(\epsilon_i) = V_i \sim U(0,1)$

$V_1, V_2, \dots, V_n$  - are independent

### Theorem

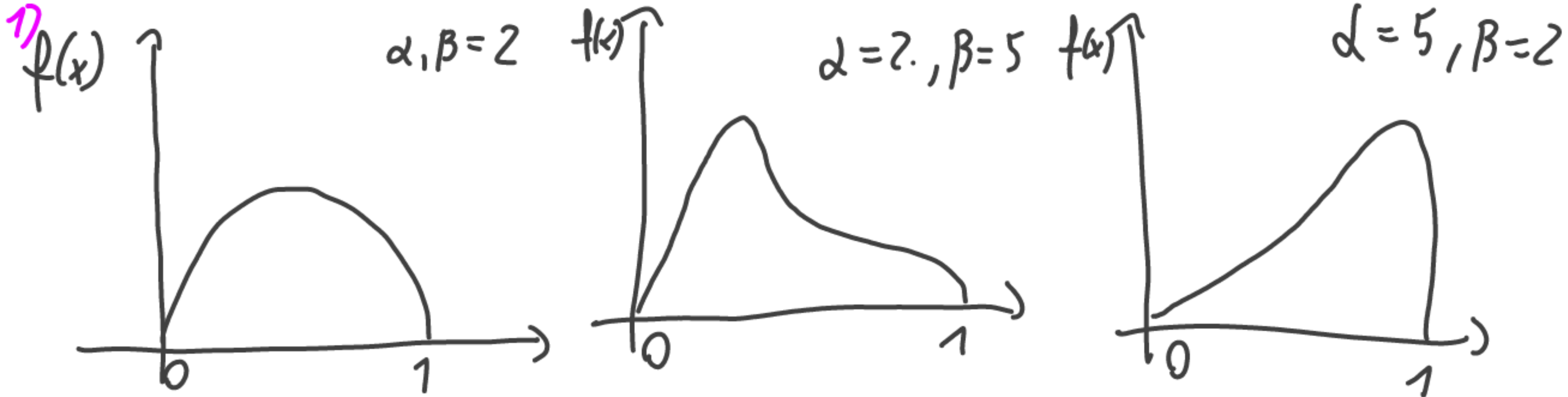
Let us consider  $V_1, \dots, V_n$ ,  $V_i \sim U(0,1)$ , then variable  $V_{k:n}$  has beta distribution  $\text{Beta}(k, n+1-k)$ .

$$V_{k:n} \sim M[k] \rightarrow \frac{k-1}{M[k]}$$

Proof: exercise 13 and 14

$$1) X \sim \text{Beta}(\alpha, \beta) \text{ has density } f(x, \alpha, \beta) = \frac{x^{\alpha-1} (1-x)^{\beta-1}}{B(\alpha, \beta)} \quad x \in (0,1) \\ \alpha, \beta > 0$$





2)  $B(\alpha, \beta)$  - beta function

$$B(\alpha, \beta) = \int_0^1 t^{\alpha-1} (1-t)^{\beta-1} dt, \quad \operatorname{Re}(\alpha), \operatorname{Re}(\beta) > 0$$

$$B(\alpha, \beta) = \frac{\Gamma(\alpha) \Gamma(\beta)}{\Gamma(\alpha + \beta)}, \quad \Gamma(s) = \int_0^\infty e^{-t} t^{s-1} dt, \quad \operatorname{Re}(s) > 0$$

$$n \in \mathbb{N} : \Gamma(n+1) = n!$$

3) Ex 13:  $X_1, \dots, X_n \sim f(x)$ , if  $F(x)$  have the same density  $f(x)$  (dist)

$$\text{then } X_{k:n} \sim f_k(x) = \frac{F^{k-1}(x) [1-F(x)]^{n-k} \cdot f(x)}{B(k, n+1-k)} \quad *$$

4) Ex 14: Use  $*$  and  $f(x), F(x)$  for uni dist. to show that

$$U_{k:n} \sim \text{Beta}(k, n+1-k)$$