# **Algorithms for Primitives of Stream Processing**

- Update stream model.
- Stream viewed as sequence (i, v),
  - 1. *i* is identity of record.
  - 2. v is change in frequency of i.
  - 3. v > 0: v insertions of i
  - 4. v < 0: v deletions of i.

#### **Frequency Sketches**

- Keys  $\in \{0, 1, \dots, N-1\}$ .
- $x_i$  is a random variable for each i = 0, 1, ..., N 1.
- $\mathbf{P}(x_i = 1) = \mathbf{P}(x_i = -1) = \frac{1}{2}$ .
- $x_i$ 's 4-wise independent

### Frequency Sketches contd.

• Sketch is a random variable X

$$X = \sum_{i=1}^{N-1} f_i x_i$$

- X is efficiently maintained: UPDATE(i, v):  $X := X + x_i \cdot v$
- $\bullet E[X \cdot x_i] = f_i$
- Efficient retrieval of top-k items [Charikar, Chen, Farach 2002].

## FM-sketches for Count Distinct Queries

- h is a random hash function from  $\{0, \ldots, N-1\} \rightarrow \{0, \ldots, N-1\}$ .
- $\bullet$  N is a power of 2.
- lsb(x): least significant bit of x.
- $i \rightarrow lsb(h(i)) (h(i))$ .
- $\mathbf{P}(h(i) = 1) = \frac{1}{2}, \mathbf{P}(h(i) = 2) = \frac{1}{4}$  $\mathbf{P}(h(i) = l) = \frac{1}{2l}$

#### FM-sketches contd.

- Let stream have n distinct items.
- let  $l = \lceil \log n \rceil$ .
- Expected number of items at level  $l = \frac{n}{2^l} \in \{\frac{1}{2}, 1\}$
- $\bullet$  Can be used to estimate n.
- Generalizes to all streaming models.