

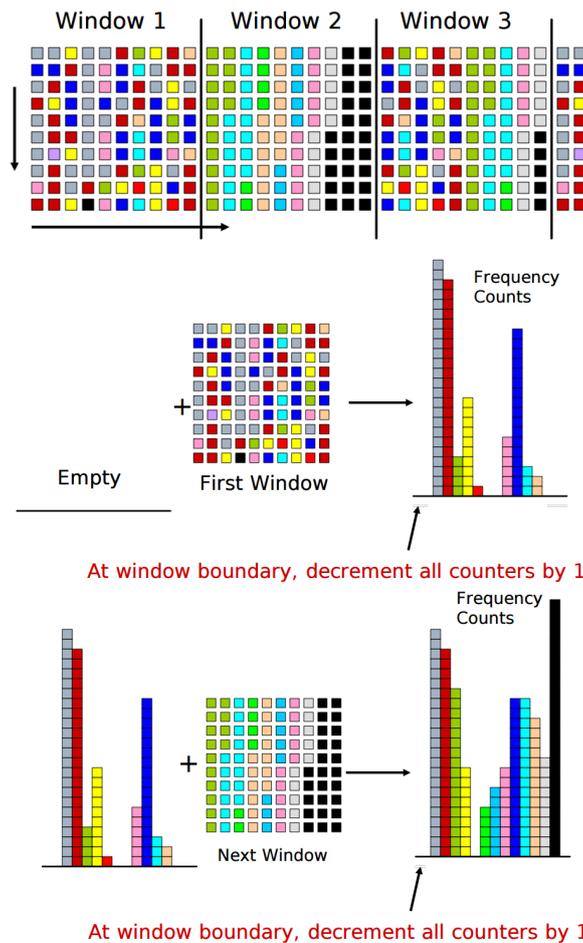
Knowledge Discovery in Databases II  
 SS 2019

Exercise 6: Correlation Clustering and Stream 1

Exercise 6-1 Lossy Counting

Before stream clustering, let's take a look at a more fundamental task in stream: count the occurrence of objects in a stream and output objects with a count larger or equal to some given threshold:  $minSup \times L$ , where  $L$  is the length of the stream up to now and  $minSup$  is the given threshold (minimum support).

Lossy Counting is one of the basic algorithms that solve this problem. Given the windows size as  $w = \frac{1}{\epsilon}$ , the lossy counting algorithm works as the follows: cut stream into windows, process one window a time and prune histogram entries with 0 counts at each window boundary. The illustration is given below:



Please prove that

- (a) the maximum count error (the maximum difference between the real count and the estimated count) of the lossy counting algorithm is  $\epsilon L$

(b) the memory consumption, i.e., the number of entries stored in the histogram, is  $O(\frac{1}{\epsilon} \log(\epsilon L))$ . (Optional)

Let  $W$  be the current window id. For each  $i \in [1, W]$ , let  $d_i$  denote the number of entries in the histogram  $H$  which corresponds to window  $W - i + 1$ .

Thus, the item in the stream corresponding to such entry must occur at least  $i$  times in window  $B - i + 1$  through  $W$ ; otherwise, it would have been deleted. Since the size of each window is  $w$ , we have:

$$\sum_{i=1}^j id_i \leq jw \quad \text{for } j = 1, 2, \dots, W$$

Now we want to prove:  $\sum_{i=1}^j d_i \leq \sum_{i=1}^j \frac{w}{i}$

By induction:

- For  $j = 1$ , this is true.
- Assume it is true for  $j = 1, 2, \dots, p - 1$ , then for  $j = p$ , adding the first inequality for  $j = p$  to all  $p - 1$  instances of the second inequality gives us:

$$\sum_{i=1}^p id_i + \sum_{i=1}^1 d_i + \sum_{i=1}^2 d_i + \dots + \sum_{i=1}^{p-1} d_i \leq pw + \sum_{i=1}^1 \frac{w}{i} + \sum_{i=1}^2 \frac{w}{i} + \dots + \sum_{i=1}^{p-1} \frac{w}{i}$$

(Here, the second inequality is used for  $p - 1$  times with  $j$  varies from 1 to  $p - 1$ . We can do this because we assume the second inequality is true for all  $j \in [1, p - 1]$ .)

Then we get  $p \sum_{i=1}^p d_i \leq pw + \sum_{i=1}^{p-1} \frac{(p-i)w}{i} = p \sum_{i=1}^j \frac{w}{i} \Rightarrow \sum_{i=1}^p d_i \leq \sum_{i=1}^j \frac{w}{i}$

(The process is:

$$\begin{aligned} pw + \sum_{i=1}^{p-1} \frac{(p-i)w}{i} &= \frac{pw}{p} + \frac{1w}{1} + \frac{2w}{2} + \dots + \frac{(p-1)w}{p-1} + \sum_{i=1}^{p-1} \frac{(p-i)w}{i} \\ &= \sum_{i=1}^p \frac{pw}{i} \end{aligned}$$

)

Thus the memory consumption at window  $W$  is  $|H| = \sum_{i=1}^W d_i \leq \sum_{i=1}^W \frac{w}{i} = \frac{1}{\epsilon} \log W = \frac{1}{\epsilon} \log \epsilon L$

(Here the inequality  $\sum_{i=1}^W \frac{1}{i} \leq \log W$  is used, as it is the harmonic series.)