# Mining Frequent Items in a Stream using Flexible Windows

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- What
- New Frequency
- Properties
- Algorithm
- Worst Case
- Further Work

### What...?

#### Finding frequent items in a continuous stream of items

#### abbaacdeebcdaababaacdbabaacadaa

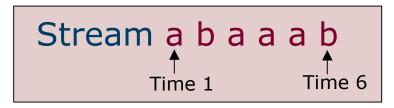
 $\uparrow$ timestamp t=1

- → New Frequency Measure: Max-Frequency
- → Incremental Algorithm
- $\rightarrow$  Worst-Case Analysis

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Timestamp 6 Target item a





mfreq(a, abaaab) = max(freq(a, last(k, abaaab)))k=1..6  $= \max(0/1, 1/2, 2/3, 3/4, 3/5, 4/6)$ 4/6 = 3/43/5 3/4 2/3 1/2 ⊢ 0/1 b b a а a a Universiteit Antwerpen 3

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### New Frequency: Definition

For each item, we consider the window in which it has the highest probability:

Max-Frequency:

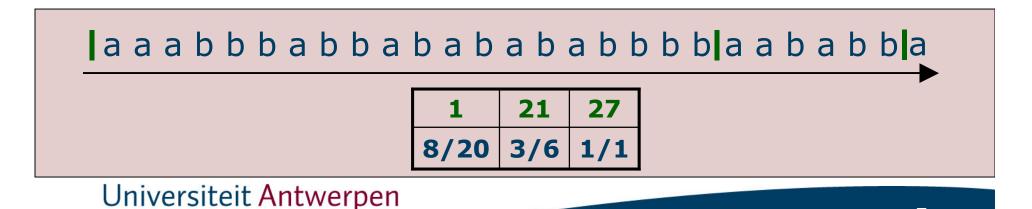
 $mfreq(i, S) := \max_{k=1..|S|} (freq(i, last(k, S)))$ 

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#### Properties

Checking all possible windows to find the maximal one: **infeasible** 

BUT: not every point needs to be checked  $\downarrow$ Only some special points = the borders



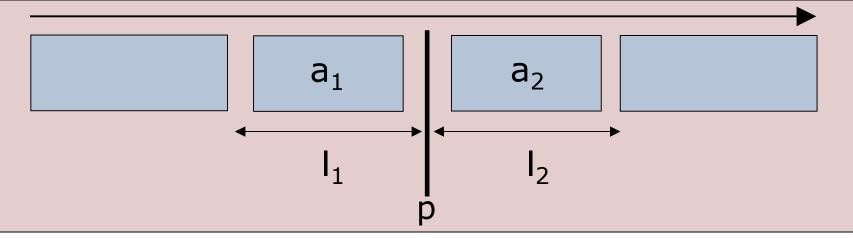
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# How to find the borders?

#### Target item a

#### **a**<sub>i</sub> = # occurrences of **a** in that block



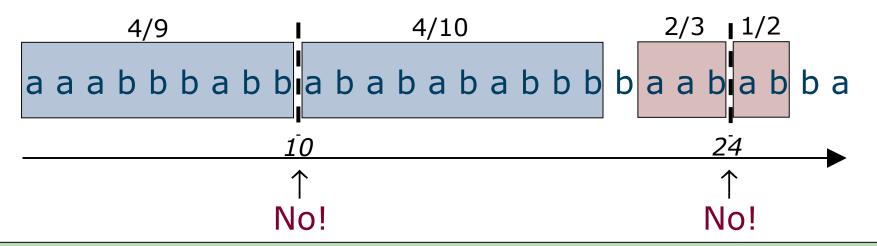
#### If $a_1/l_1 \ge a_2/l_2$ , position p is never the border again! Very powerful pruning criterion!

If a position **p** is **not** a border in *S*, then it neither can be a border in **any extension** from *S*.

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### Example

#### On timestamp 27, we have $S_{27}$ :



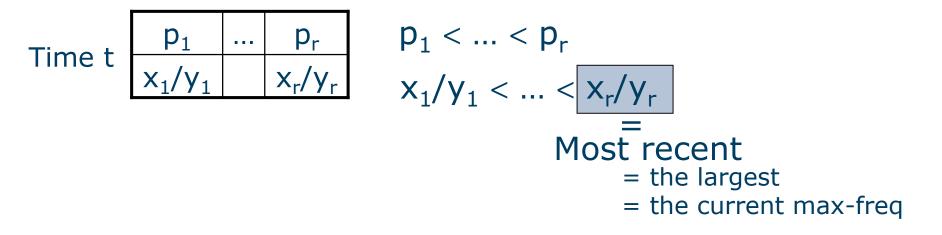
The only borders that need to be remembered:

1	21	27
8/20	3/6	1/1

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### Algorithm

#### **Output**: on every timestamp t: Summary(S<sub>t</sub>)



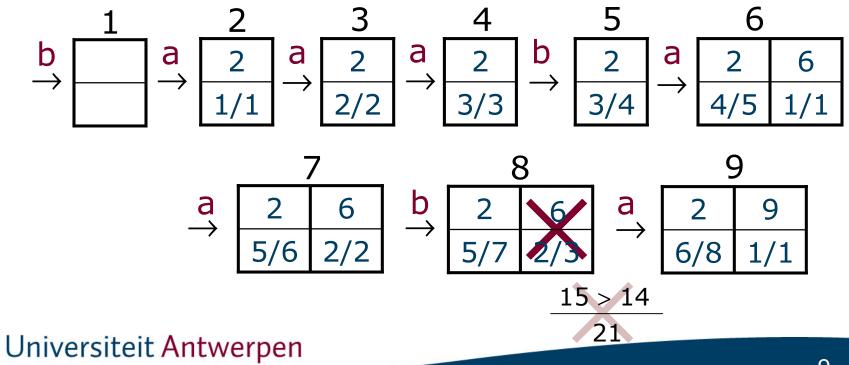
# How: on every timestamp, the algo adjusts the stored values based on the newly entered item

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### Example



Target item = a



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### Worst-Case Analysis

For a specific streamlength *I*, we will identify a stream of length *I* that maximizes the number of borders: the Farey stream.

The idea is to have as many blocks as possible, causing as many borders as possible

$$\begin{vmatrix} a_1 & b_1 \\ \bullet \\ I_1 & I_2 \\ \bullet \\ a_1/I_1 & I_2 \\ a_2/I_2 & I_r \\ \bullet \\ a_r/I_r \\ \bullet \\ a_r/I_r \end{vmatrix}$$

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What Farey has to do with it

$$a_1/l_1 < a_2/l_2 < ... < a_r/l_r$$

The challenge is for each streamlength  $k = I_1 + I_2 + ... + I_r$  to find such an increasing array of fractions

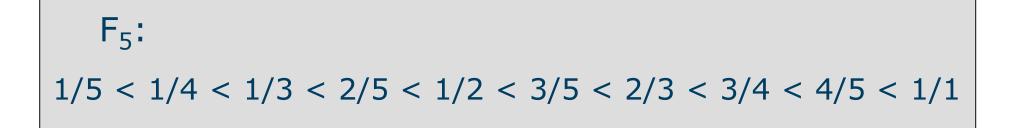
Solution: Farey sequences  

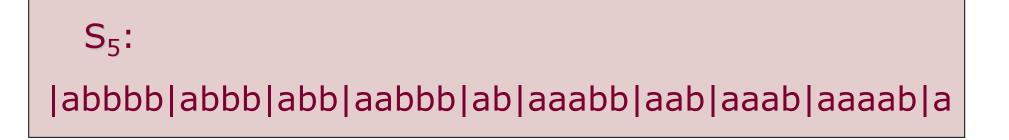
$$F_1 = 1/1$$
  
 $F_2 = 1/2, 1/1$   
 $F_3 = 1/3, 1/2, 2/3, 1/1$   
 $F_4 = 1/4, 1/3, 1/2, 2/3, 3/4, 1/1$ 

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#### Farey Streams

#### The Farey Sequence $F_n$ introduces the Farey Stream $S_n$ .





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### Most Important Result

## Theorem: For streams of length L, the maximal number of borders is given by N:

$$N = \left(\frac{\pi^2 L}{2}\right)^{2/3} \frac{3}{\pi^2}$$

#### Remark: Experiments show that the worst case never happens!

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### Further Work

- Minimum Window Length
- Focus on multiple targets in the stream
- Make the extension to itemset mining