



Mining Frequent Itemsets in a Stream

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Outline

- Motivation
- Max-Frequency
- Algorithm
 - for one itemset
 - mining all Frequent Itemsets
- Experiments
- Conclusion



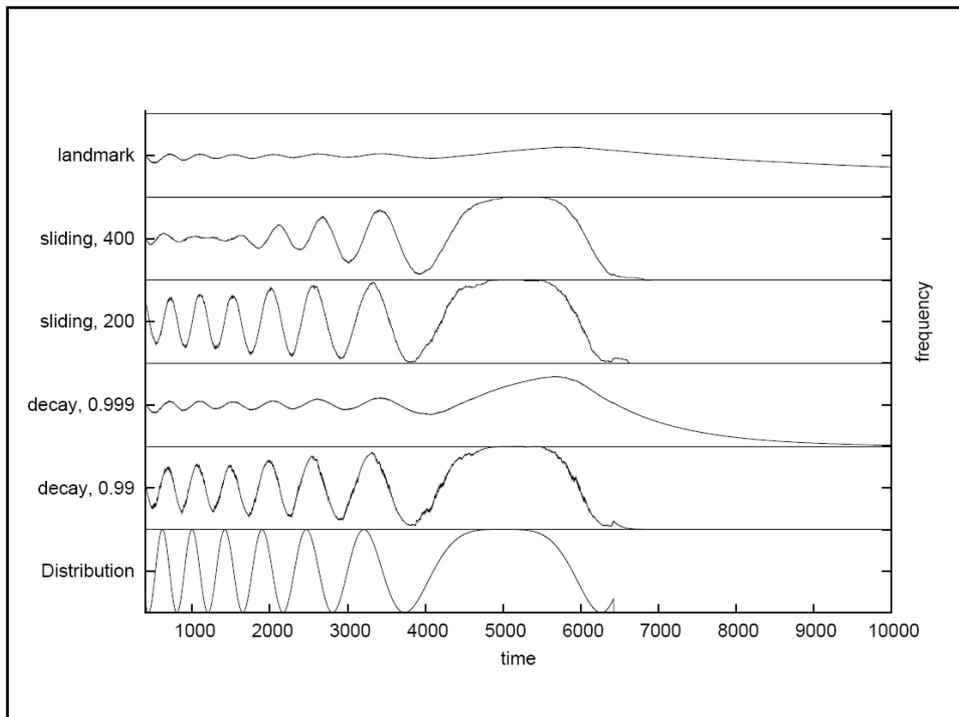
Motivation

- Model:
 - Every timestamp an itemset arrives
- Goal:
 - Find sets of items that frequently occur together
 - Take into account history,
 - Yet, recognize sudden bursts quickly



Motivation

- Most definitions of frequency rely heavily on the correct parameter settings
 - Sliding window length
 - Decay factor
 - ...
- Correct parameter setting is hard
 - Can be different for different items (not to mention sets!)



- ## ● ● ● | Outline
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Max-Frequency

Therefore, a new **frequency** measure:

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

Frequency is measured in the window where it is *maximal*.

Itemset gets the benefit of the doubt ...



Example

$\text{mfreq}(a, \text{ac abc ab ac ab bc})$

ac bc ab ac ab bc 0

ac bc ab ac ab bc 1/2

ac bc ab ac ab bc 2/3

ac bc ab ac ab bc 3/4 ←

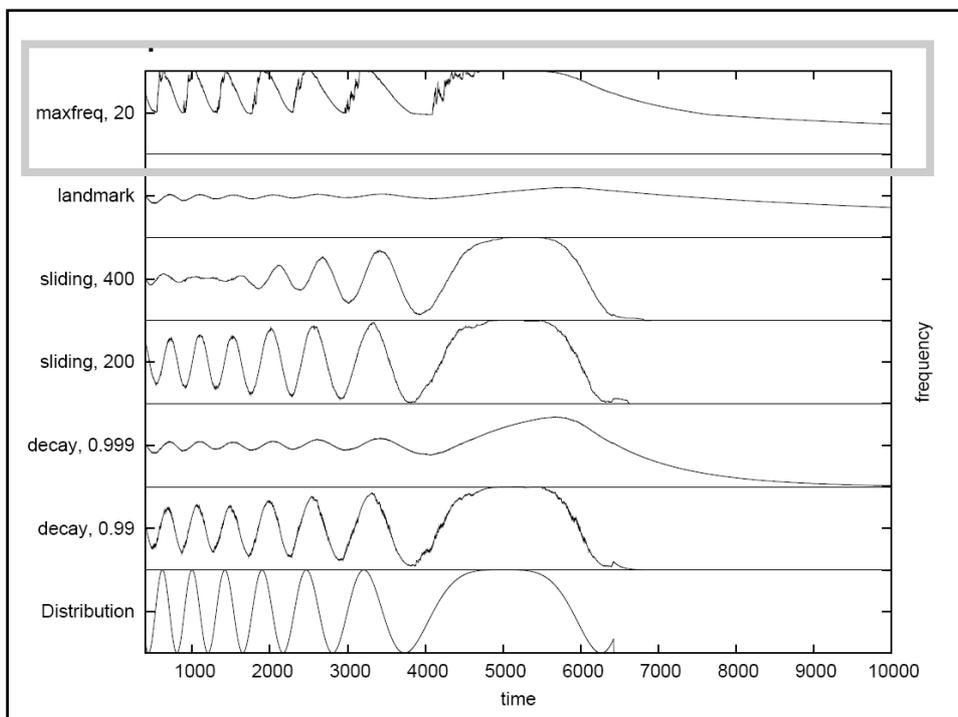
ac bc ab ac ab bc 3/5

ac bc ab ac ab bc 4/6



Properties of Max-Freq

- + Detects sudden bursts
- + Takes into account the past
- When target itemset arrives: sudden jump to a frequency of 1
- + Solution: minimal window length





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Algorithm

1. How to do it for **one** itemset?
2. How to do it for a **frequent** itemset?
3. How to do it for **all** frequent itemsets?

Maintain a **summary** of the stream that allows to find the frequencies immediately.

● ● ● | Properties (one itemset)

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

BUT: not every point needs to be checked

↓
Only some **special points** = the **borders**

| a a a b b b a b b a b a b a b b b b | a a b a b b | a

timestamp	1	21	27
# targets	8	3	1

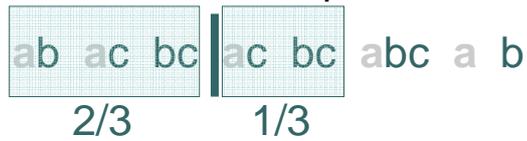
● ● ● | How to find a border?

- Target set a
- Is the marked position a border?

ab ac bc | ac bc abc a b

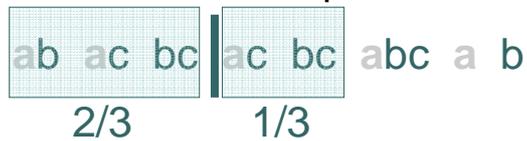
● ● ● | How to find a border?

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● ● ● | How to find a border?

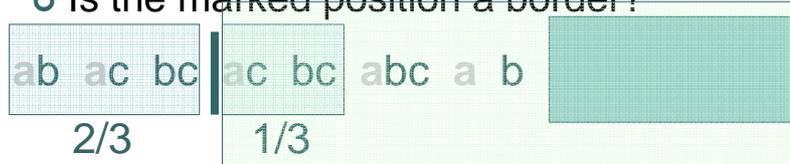
- Target set a
- Is the marked position a border?



NO

● ● ● | How to find a border?

- Target set a
- Is the marked position a border?

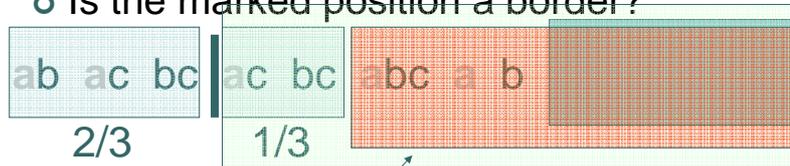


NO

> 2/3

● ● ● | How to find a border?

- Target set a
- Is the marked position a border?



NO

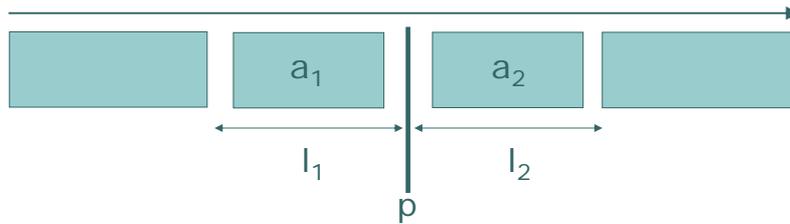
> 2/3

even bigger



How to find the borders?

- This is true in general:



If $a_1/l_1 \geq a_2/l_2$, position p is never the border again!

Very powerful pruning criterion!



The summary

- Summary only keeps counts for the borders.

| ab ac bc ac bc | abc a b

1	6
3	2

● ● ● | The summary

- Summary only keeps counts for the borders.

ab ac bc ac bc	abc a b	1	6
		3	2

- Frequencies always increasing
 - Thus: max-frequency in last cell
- Block with largest frequency before border p_i = always block from p_{i-1}

● ● ● | Updating the Summary

- When a new itemset arrives, the summary is updated.
 - borders need to be checked again

ab ac bc ac bc	abc a b	T
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Updating the Summary

- When a new itemset arrives, the summary is updated.

- borders need to be checked again

ab ac bc ac bc | abc a b T

- no new « before » - blocks
- only one new « after » - block
- maximal block before: always previous border

Updating the Summary

- When a new itemset arrives, the summary is updated.

- borders need to be checked again

ab ac bc ac bc | abc a b T

- no new « before » - blocks
- only one new « after » - block
- maximal block before: always previous border



Mining Frequent Itemsets

- Only interested in itemsets that are frequent.
- We can throw away any border with a frequency lower than the minimal frequency.

ab ac bc ac bc | abc a b | ab

6	9
2	1

minfeq = 2/3



Mining All Frequent Itemsets

- We only need to maintain the summaries for the frequent itemsets
- Can still be a lot, though ...
 - every subset of the most recent transaction
 - ...
 - minimal window length reduces this problem
- FUTURE WORK: reduce this number; rely, e.g., on approximate counts



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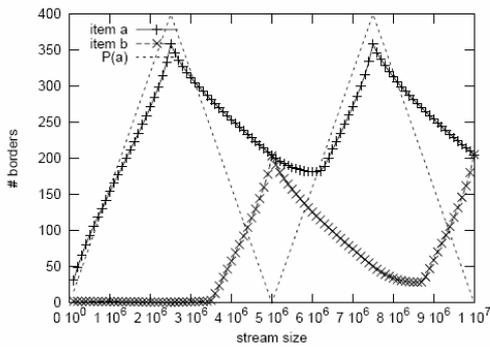
Experiments

- Size of the summaries
 - number of borders for random data
 - average, maximal number of borders in real-life data
- Theoretical worst case

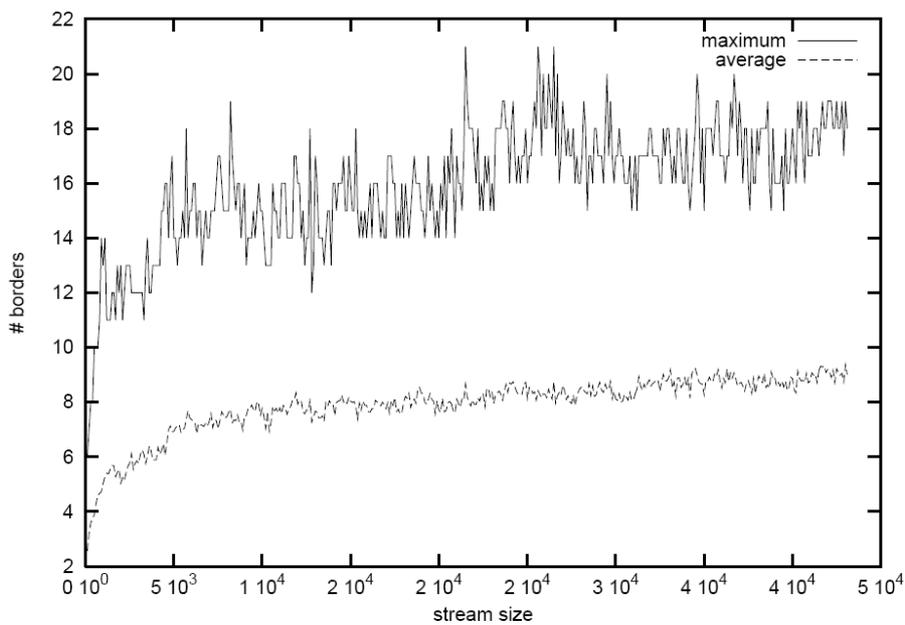
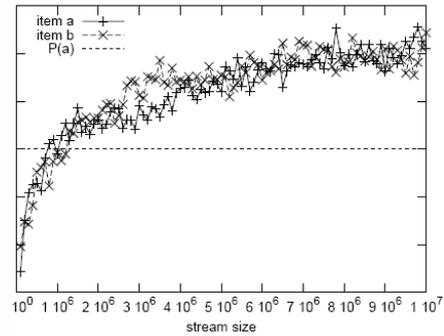
$$N = \left(\frac{\pi^2 L}{2} \right)^{2/3} \frac{3}{\pi^2}$$

Experiments

Twin Peaks distribution



Uniform Distribution





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Conclusions

- New frequency measure
- Summary for one itemset
 - small
 - easy to maintain
 - only few updates
- Mining all frequent itemsets
 - only need summary for frequent itemsets

