

Building stream summaries

Looking for a new approach of exploratory analysis of data streams
under time and memory constraints



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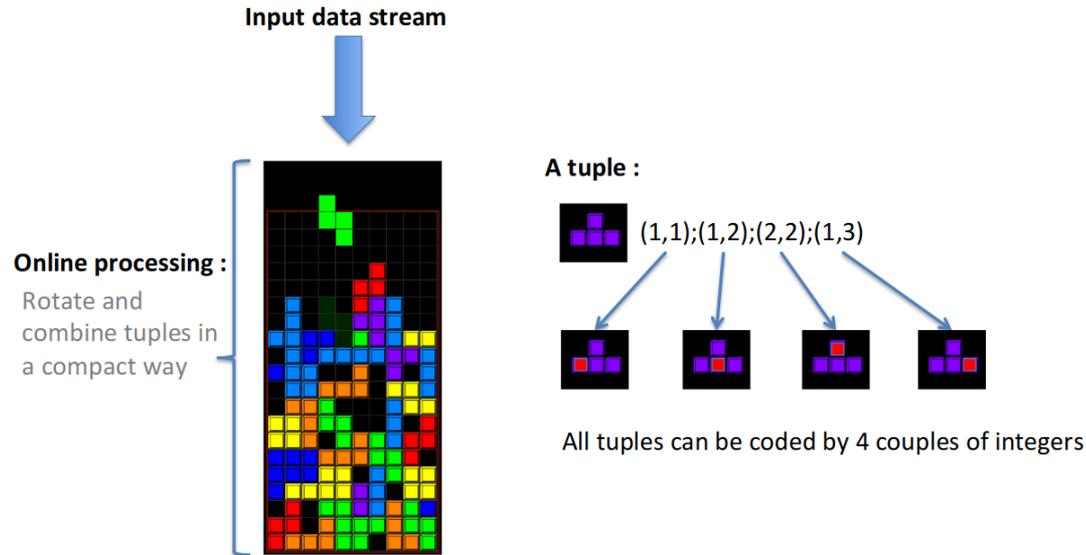
July 9th 2019

MOTIVATION

Data streams

Fraud detection, health monitoring, smart cities...

- Tuples (data units) arrive continuously and at high speed
- Data can be accessed only once
- Distributions are constantly changing



MOTIVATION

Summaries *Compact representations of past tuples that should allow to query or analyze the whole history of the data*

- Estimate the distribution with a controlled memory space and precision trade-off
- Making as few **assumptions** on the nature of the data as possible (*exploratory analysis*)
- Taking into account **time** and **memory** use constraints defined beforehand

Specific summaries

Answer a specific question



Flajolet-Martin Sketch : approximates the number of unique objects in a stream;

Count-Min Sketch : enumerates the number of elements with a particular value, or within an interval of values.

Generic summaries

Answer a range of questions



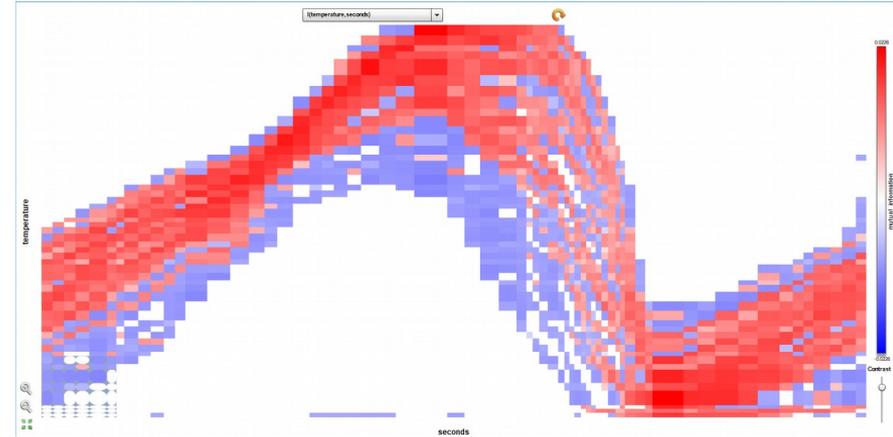
CluStream : micro-clusters over time and 'snapshots';

StreamSamp : successive windowing and sampling.

OUR APPROACH (SO FAR)

Adaptative grid models

- Piece-wise constant estimators : Time x Data
- Variable width grid cells
- Automatic best model selection with the *Minimum Description Length* principle



Temperature evolution of 65 sensors in one day (Intel Lab Sensors dataset)

MDL

- Information theory for statistical inference

$$\text{cost}(\text{Model}) = -\log(p(\text{Model}|\text{Data})) \propto -\log(p(\text{Model}) \times p(D|M))$$

Natural compromise between the precision and the robustness of the model

ROADMAP

- Extend the grid construction to the stream setting
 - As data arrives instead of batch mode
- Beware of change in distribution as it can affect the quality of the density estimation
 - Drift detection mechanism for streams
- Keep merging the produced summaries so we have the most informative yet compact representation
- **Quality of a general summary ?**
 - Compare performance of models that learned from the raw data and from the summaries ?

HPC IN ALL OF THIS

- The optimal MDL criterion is expensive to compute
 - Already parallelized by Orange team
- Streams are high volume and high dimensional by nature
 - process stream chunks
 - process attributes separately by distributing them over the processors available
- Optimise the memory use for grid construction and summary storage
 - Automatic resource allocation for streams ?