

Ocelot: Time Aggregation Visualization for Trace Overview

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Context: SoC-Trace project



Embedded system trace analysis problematic

- **Hardware** and **software** complexity
 - Trace **size** and **format** management
 - Analysis technique **scalability**

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Propositions

- **FrameSoC** infrastructure : storage, data-model, trace/tool/result management
 - **Analysis flow** : statistics, data-mining, **visualization**...

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Fil rouge: typical embedded system use case

Fil rouge: application tracing result

Only 20 second duration but...

- Almost **1500** different functions, 4 threads
 - **One million** of events
 - **100 Mo** trace (Pajé format)

- Same number of functions, but...
 - More than **30 millions** of events!
 - **3 GB** trace!

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For a 10 minute-long video

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 - More than **30 millions** of events!
 - **3 GB** trace!

We can easily obtain well bigger traces!!

How to represent this trace behavior over time?

Gantt Chart is the most common technique employed by analysts...

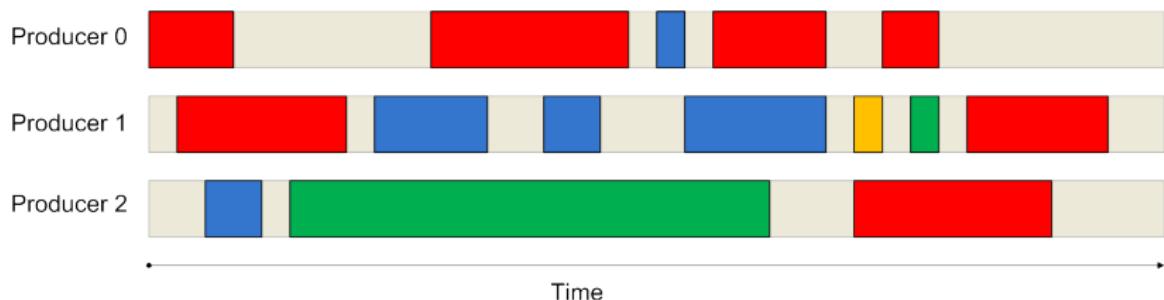


Figure 1: *Synthetic example of Gantt Chart*

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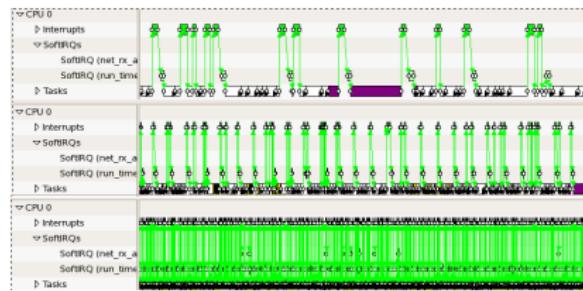


Figure 1: *KPTrace dezoom* : example of time axis scalability issues

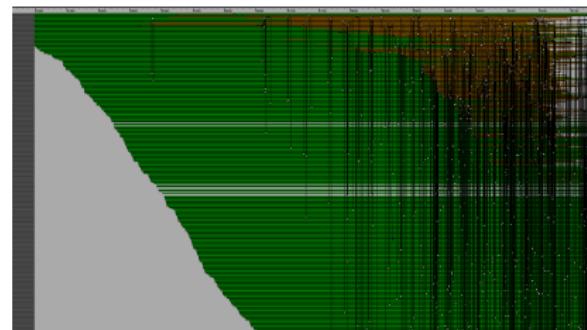


Figure 2: Example of space limitations :
Pajé trace with 700 producers

... but it does not scale to voluminous traces

Our proposal: Ocelotl

Fit to Schneiderman's methodology...

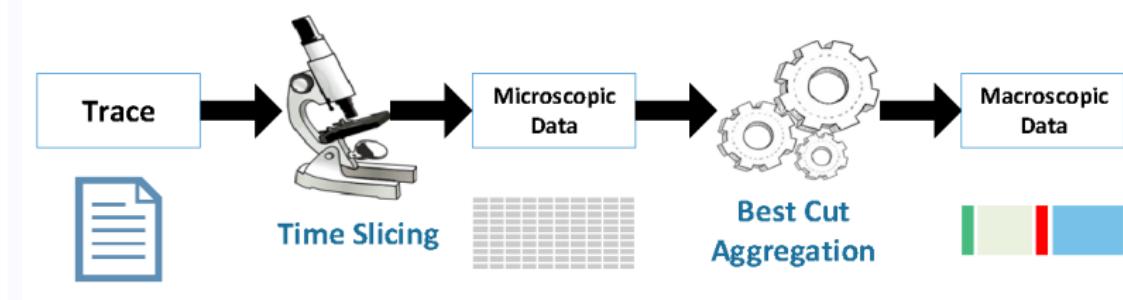
- **Overview** first, zoom and filter, then details on demand

Our proposal: Ocelotl

Fit to Schneiderman's methodology...

- **Overview** first, zoom and filter, then details on demand

... by providing a macroscopic description of the trace...

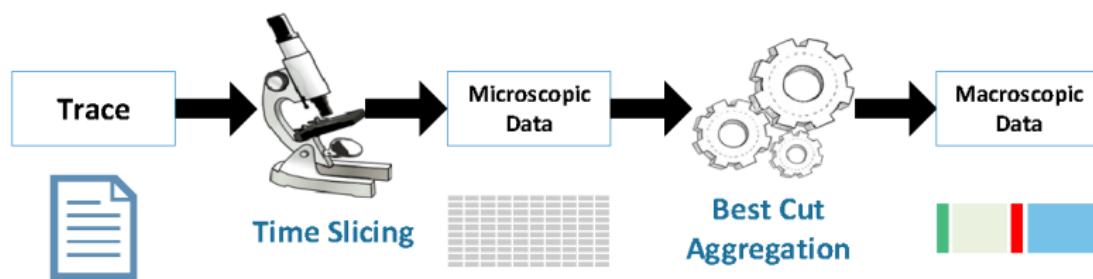


Our proposal: Ocelotl

Fit to Schneiderman's methodology...

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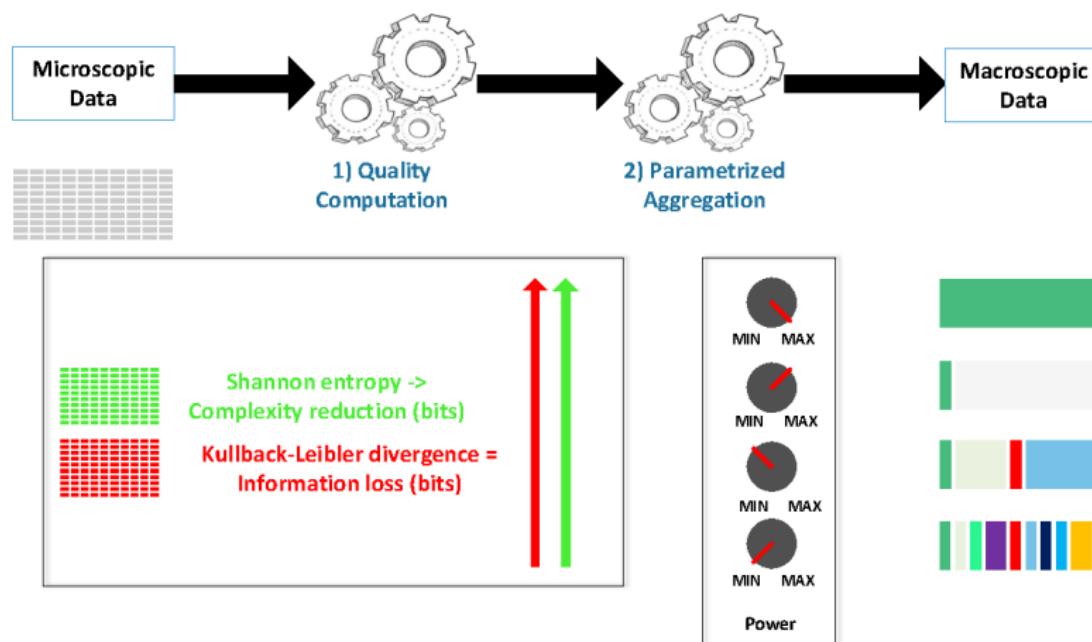


... build upon an algorithm proposed by Lamarche-Perrin

- Adapted to timestamped events using **time slicing**
 - Extended to **multiple event sources**

User controlled level of details

Parametrized aggregation enables interaction



Analysis with Ocelotl (Settings)

Analysis with Ocelotl (Overview, Qualities)

Analysis with Ocelotl (Zoom and details)

Information based time aggregation

- ... describes behavior by **highlighting phases** and **perturbations**
 - **Interaction** helps to focus on these points
 - **Performance:**
 - 20s to visualize 1M event trace
 - 2h to visualize 30M event trace (bounded by database query time)

- #### ■ Lack of space dimension representation

Results

Information based time aggregation

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But..

- Lack of **space dimension** representation

Current Focus

New features

- ## ■ Spatio-temporal aggregation

- ## ■ HPC/Distributed system relevant use-cases

New features

- ## ■ Spatio-temporal aggregation

Use-cases

- ### ■ HPC/Distributed system relevant use-cases

Links

My website

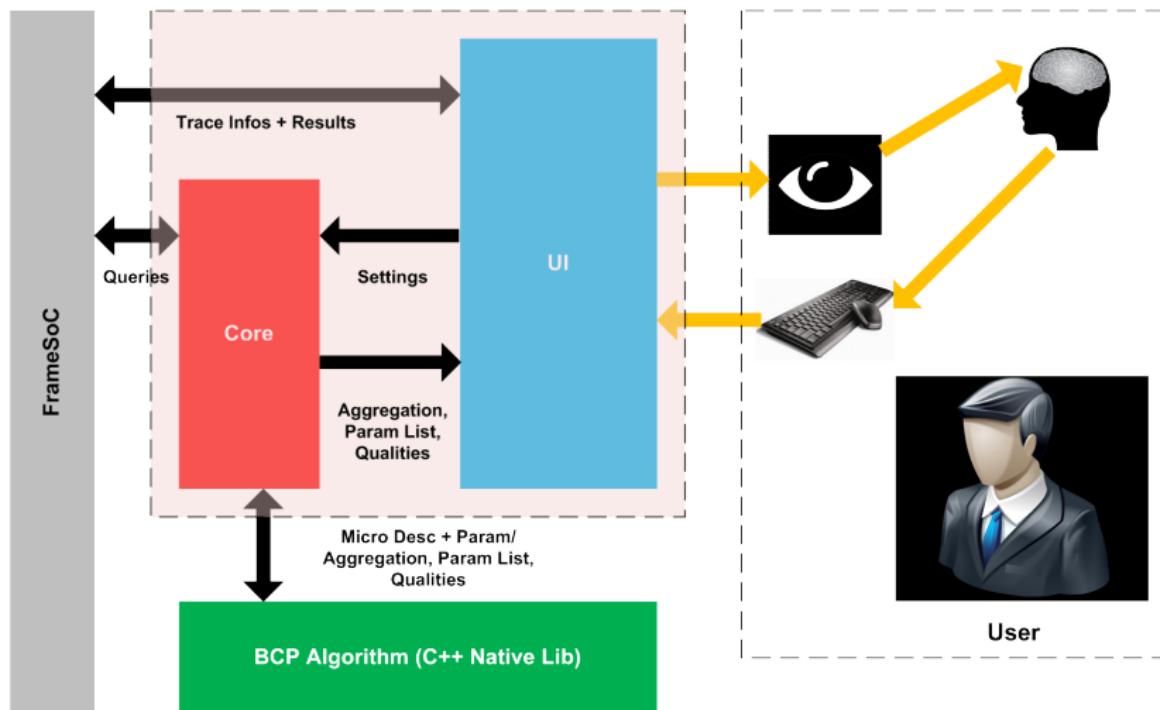
<http://moais.imag.fr/membres/damien.dosimont/>

Tools and libraries are available on my github

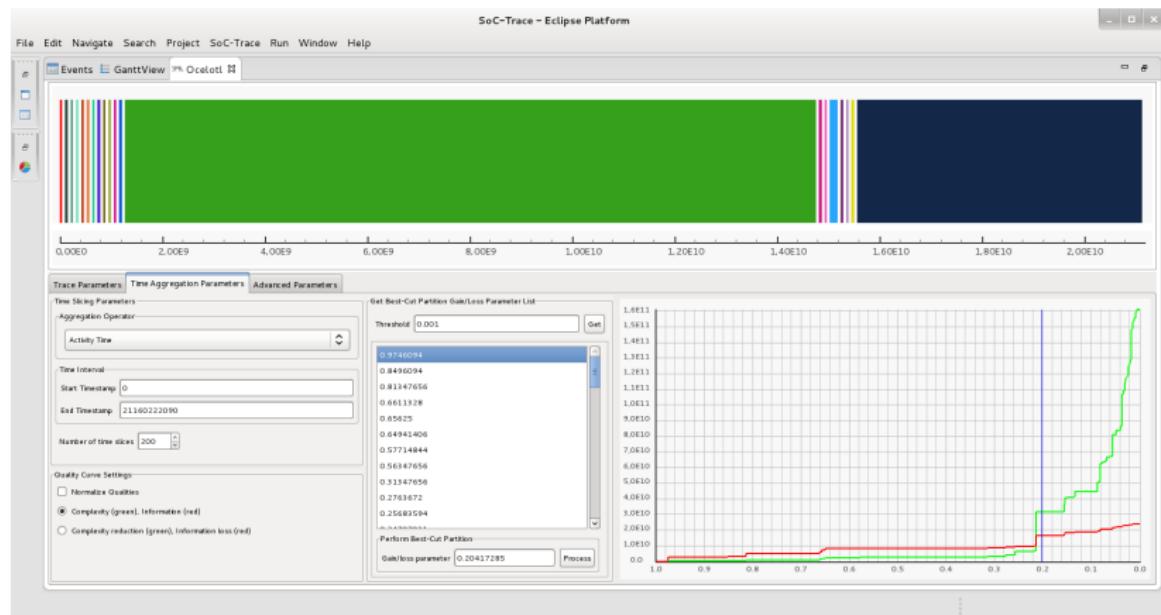
<http://github.com/dosimont>

Merci pour votre attention!

Implementation



Interface Overview



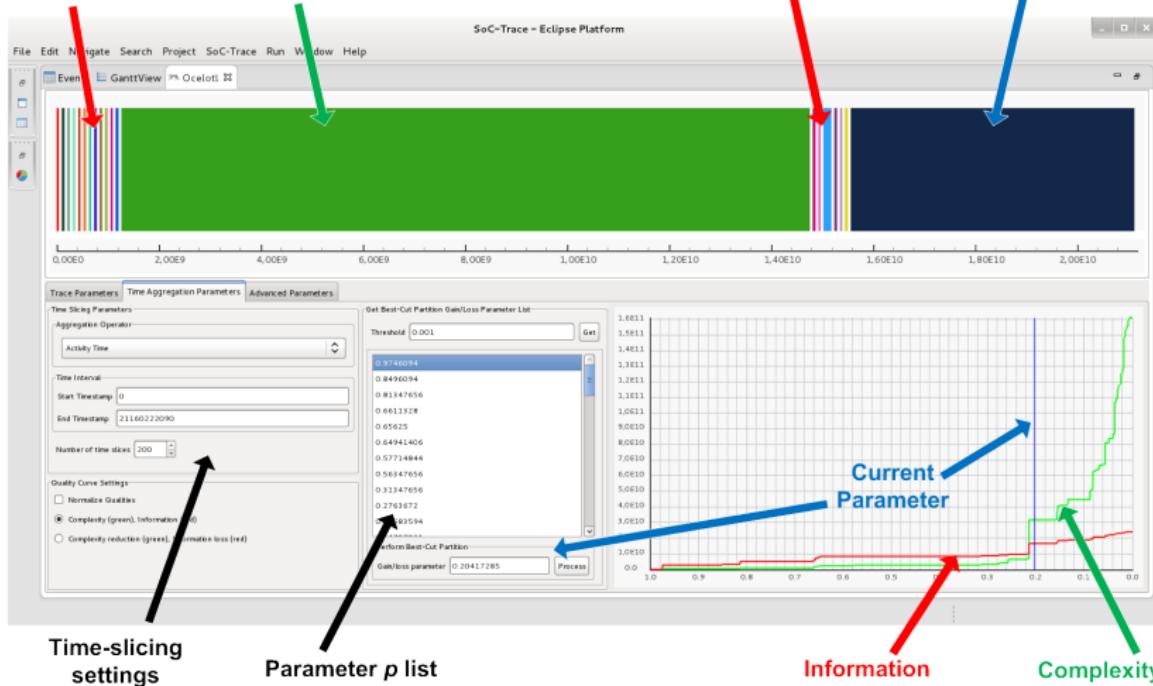
Interface Overview

Initialization

Steady State

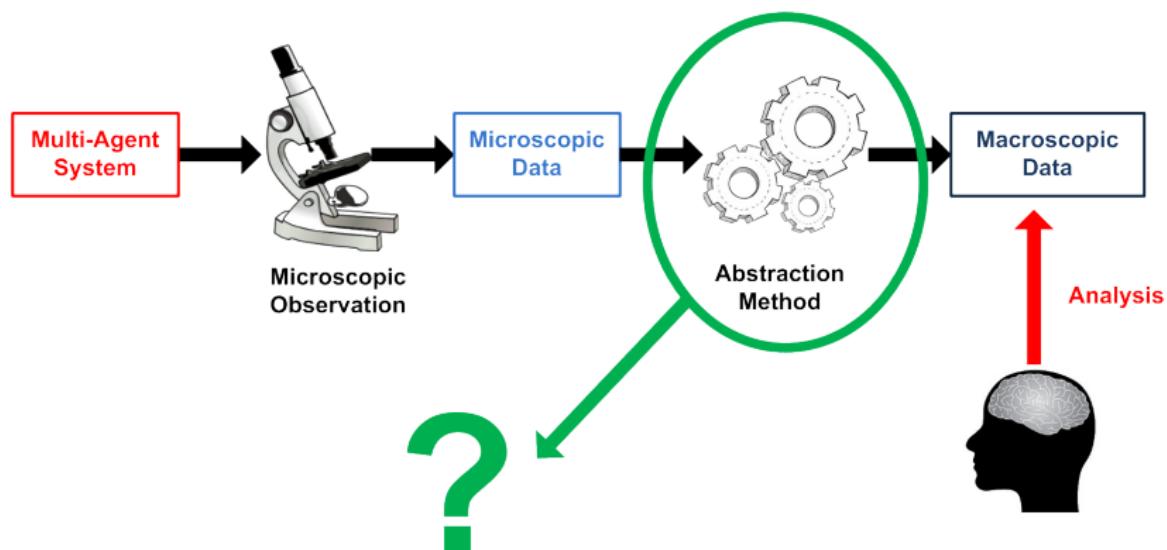
Disruption

Steady State

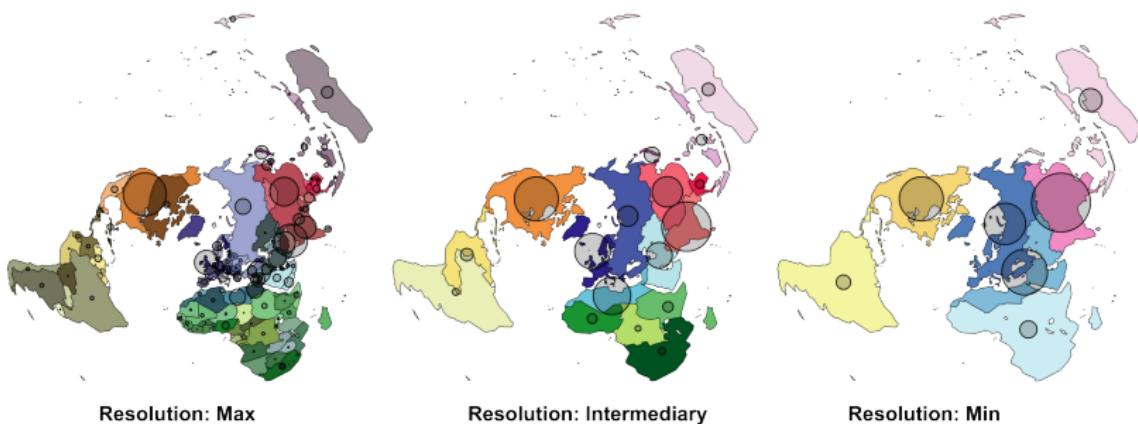
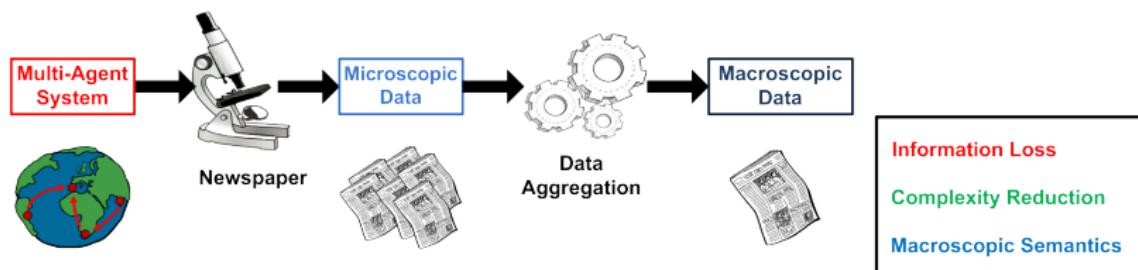


Lamarche-Perrin Works: Multi-Agent Systems

How to Build a Meaningful Macroscopic Description?



Example: Geomedia Project

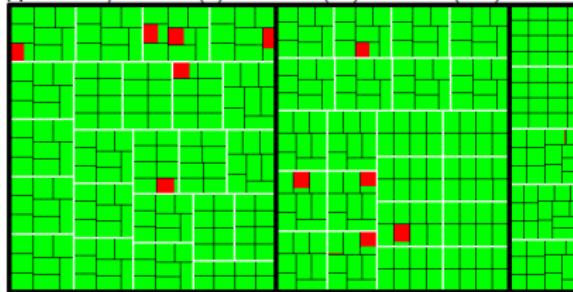




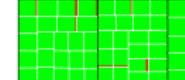
Example: Viva

Represent Hierarchical Structure according to Value Heterogeneity

A Hierarchy: Cluster (3) - Machine (50) - Process (433)



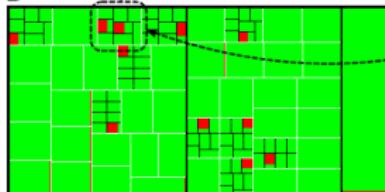
A.1 Machine level



A.2 Cluster level

A.3 Full aggregation

B Ratio Gain/Loss with P = 10%



C Ratio Gain/Loss with P = 30%



Information Loss

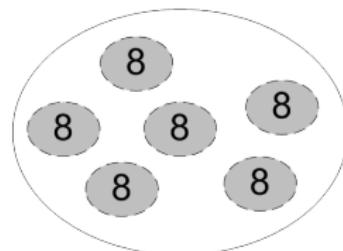
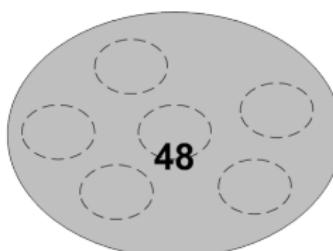
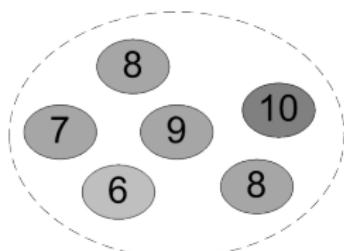
Aggregation



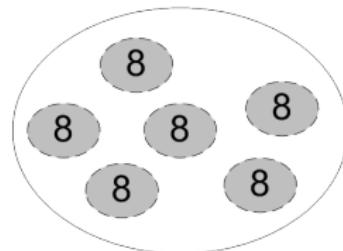
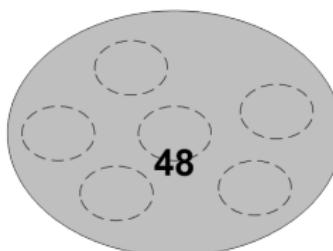
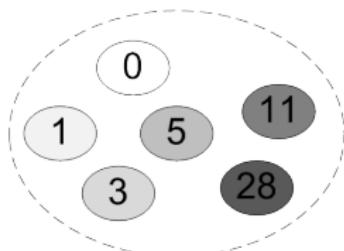
Interpretation?



Group A



Group B



Agent level

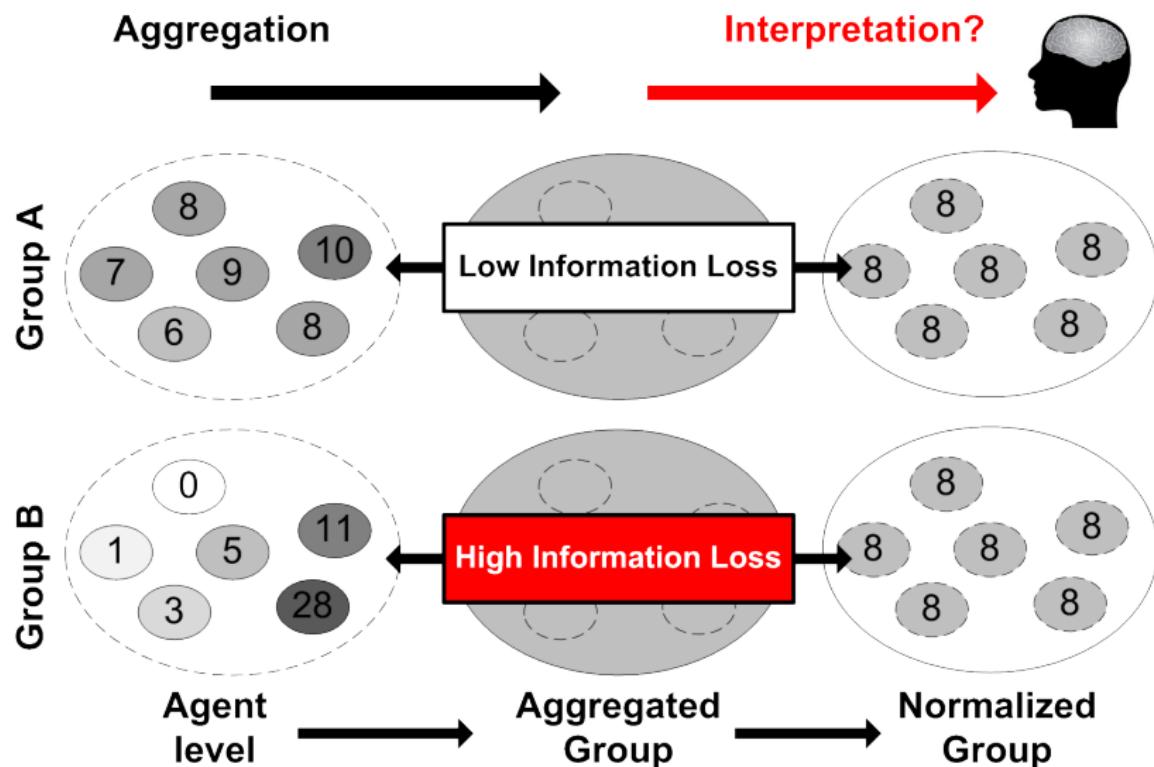


Aggregated Group



Normalized Group

Information Loss



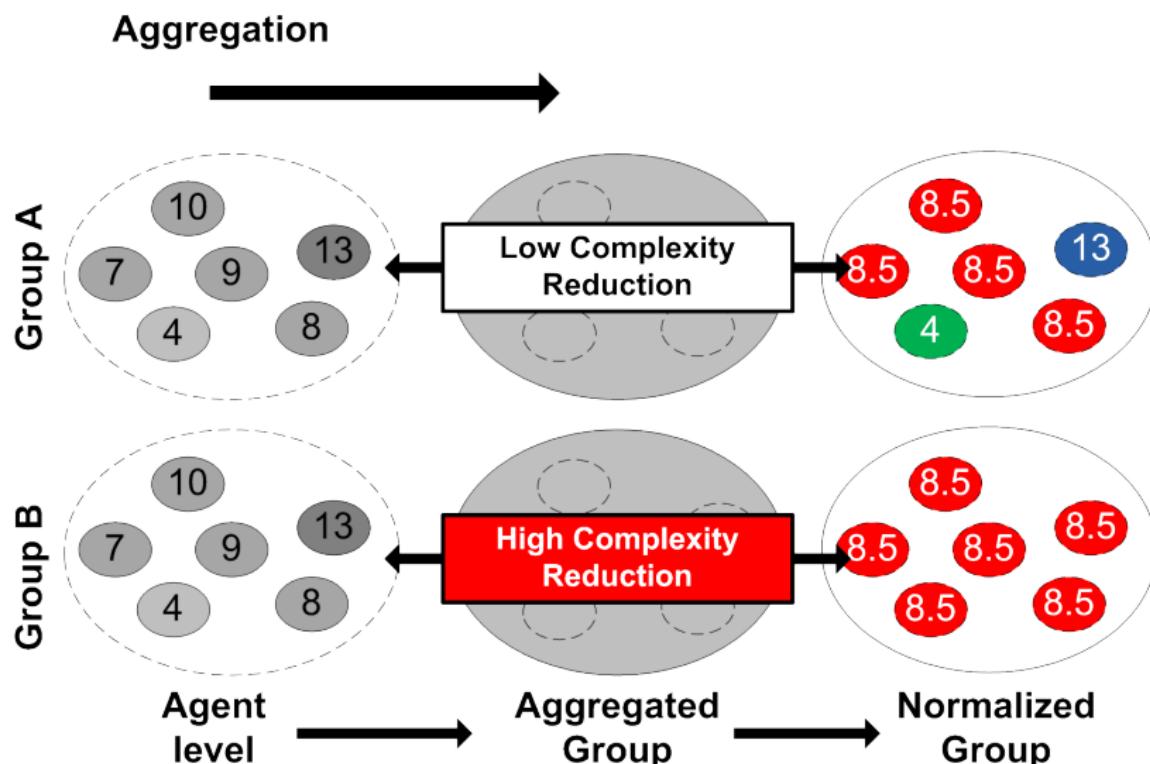
Information Loss Measure

Kullback-Leibler Divergence

$$\text{loss}(A||e) = \sum_{e \in A} v(e) \times \log_2 \left(\frac{v(e)}{v(A)} \right) \text{ in bits/x}$$

- Quantity of information than one **loses** by using an **aggregated description** instead of the **microscopic description**

Complexity Reduction



Complexity Reduction Measure

Shannon Entropy

$$H(v) = \sum (v(i) \times \log_2 v(i)) \text{ in bits/x}$$

Entropy Reduction

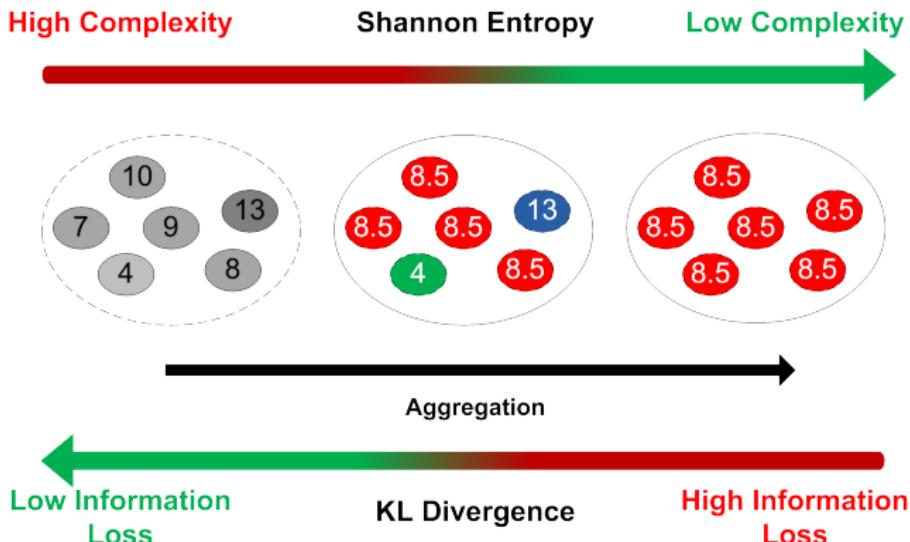
$$\text{gain}(A|e) = H(A) - H(e) \text{ in bits/x}$$

- Quantity of information than one **saves** by encoding the **aggregated description** instead of the **microscopic description**

Compromise Finding between Information Loss and Complexity Reduction

Parametrized Information Criterion

$$\text{pIC}(\mathcal{A}) = p \times \text{gain}(\mathcal{A}) - (1 - p) \times \text{loss}(\mathcal{A})$$

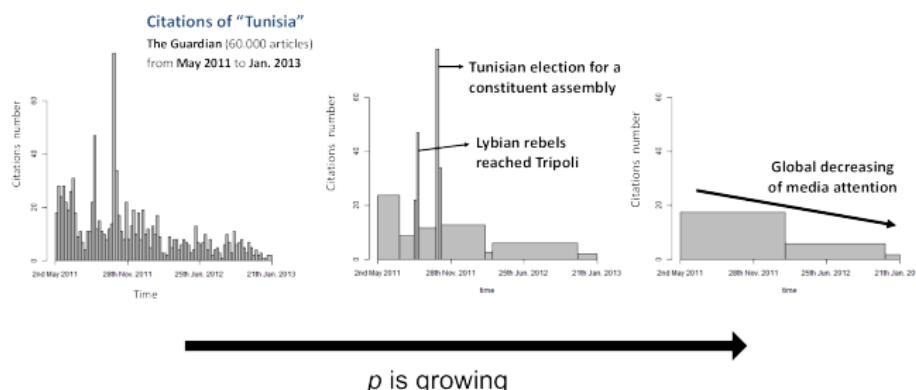


Temporal Aggregation

Temporal Aggregation principle

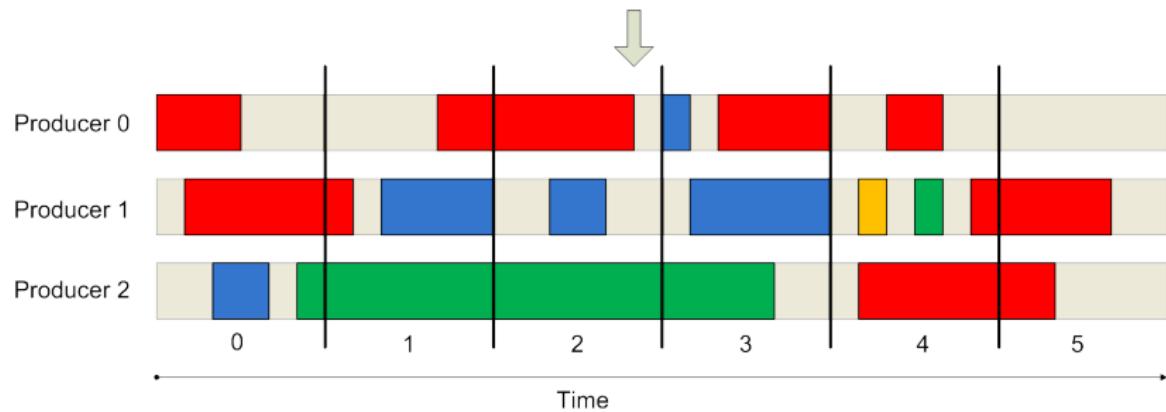
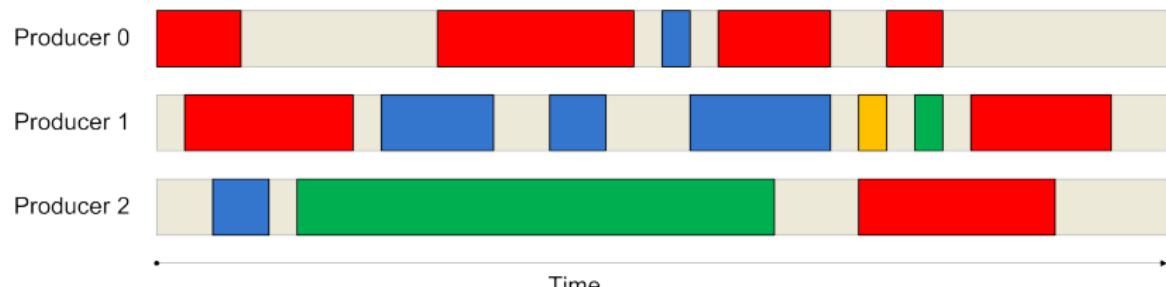
- Same principle but only consecutive data can be aggregated

Ex: Tunisia citation

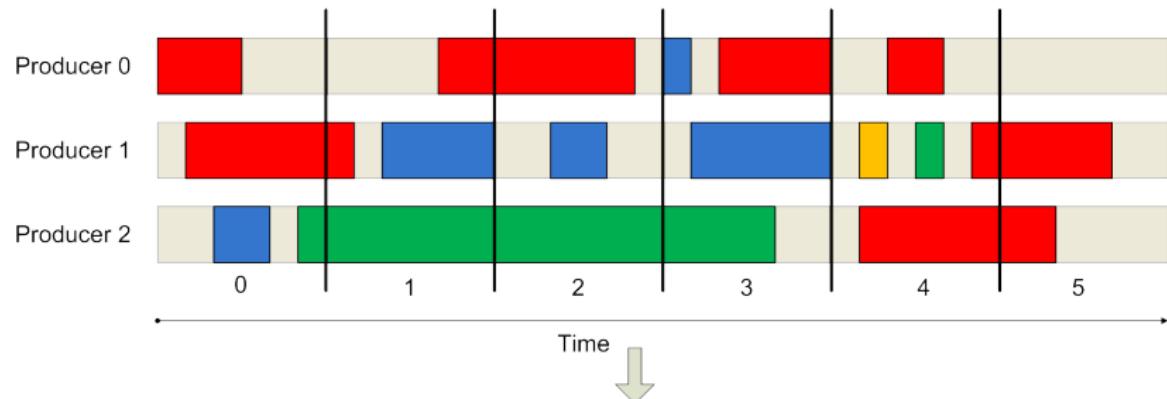


Need of a microscopic level description

Microscopic Level: Time-Slicing



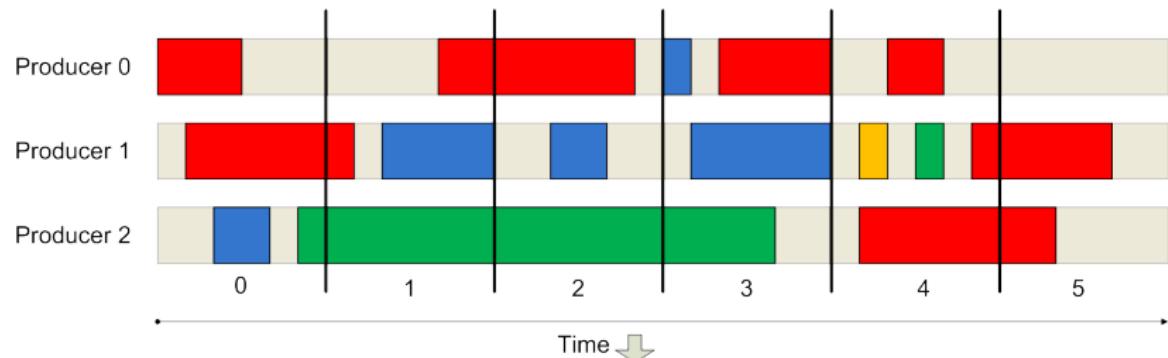
Microscopic Level: Producer Activity Time Matrix



Producer 0	0.5	0.33	0.83	0.83	0.33	0
Producer 1	0.83	0.83	0.33	0.83	0.33	0.66
Producer 2	0.5	1	1	0.66	0.83	0.33

Part number ↓

Microscopic Level: State Activity Time Cubic Matrix



	Part number					
	0	1	2	3	4	5
Producer 0	0.5	0.33	0.83	0.66	0.33	0
	0	0	0	0.16	0	0
	0	0	0	0	0	0
	0	0	0	0	0	0
Producer 1	0.83	0.16	0	0	0.16	0.66
	0	0	0.33	0.83	0	0
	0	0	0	0	0.16	0
	0	0	0	0	0.16	0
Producer 2	0	0	0	0	0.83	0.33
	0.33	0	0	0	0.33	0
	0.16	1	1	0.66	0.33	0
	0	0	0	0	0	0

Quality Computation

Gain and loss formulas: originally for scalars

012345						
01234	12345					
0123	1234	2345				
012	123	234	345			
01	12	23	34	45		
0	1	2	3	4	5	

Adaptation for time-sliced description

- Vector (ex: activity time per process)
 $\text{quality}(A) = \sum_{i \in n} \text{quality}(A[i])$
- Matrix (ex: activity time per state type)
 $\text{quality}(A) = \sum_{i \in n} (\sum_{j \in m} \text{quality}(A[i][j]))$

Best-Cut Partition for a given p

