# M4: A Visualization-Oriented Time Series Data Aggregation

Uwe Jugel, Zbigniew Jerzak,

Gregor Hackenbroich

SAP AG

Chemnitzer Str. 48, 01187 Dresden, Germany

Volker Markl
Technische Universitat Berlin
Strae des 17. Juni 135
10623 Berlin, Germany
volker.markl@tu-berlin.de

#### I. Introduction

Visualization of large scale time series data is a crucial need of modern exploratory bigdata analysis [1]. But the huge size of the data is a barrier to visualization [2], [3], [4]. To address this challenge of bigdata different data reduction and sampling strategies are used to overcome the barrier [5], [6]. But for preserving the semantics of trend line of time series data these sampling strategies show huge limitations [7].

In this review paper we present a review of the paper [7] which address this issue of preserving the semantic of time series data and present some related works in the line. The paper appeared in the Proceedings of the VLDB Endowment, 2014.

The authors present M4, an aggregation based time series data reduction strategy that guarantees error free visualization of time series data as line chart as well as higher rate of data reduction. The approach is generic to any visualization system as long as the visualization systems uses RDBMS as data source.

### II. CONTROBUTIONS OF THE PAPER

The authors of the paper rewrite visualization queries Q using data reduction operator  $M_R$  such that the visualization of the original data from query Q and the visualization from the query  $Q_R = M_R(Q)$  are similar and error free. As

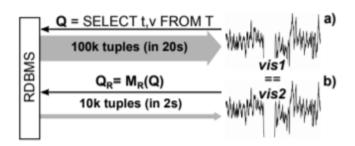


Figure 1. Time series visualization: a) based on original data; b) Using data reduction operator;

shown in Figure 1  $Q_R$  produced the same visualization as Q with almost 10 times less tuples and 10 times reduced time.

A. Query Rewriting
TODO Query rewriting

B. Time Series Visualization
TODO

C. Data Reduction Operators
TODO

D. Time Series Data Reduction

III. EVALAUTION

IV. RELATED WORKS

**TODO** 

V. OUR PROPOSAL

**TODO** 

## VI. CONCLUSION

The conclusion goes here, this is more of the conclusion

### REFERENCES

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