

Visual Analytics and Data Mining in S-T-applications

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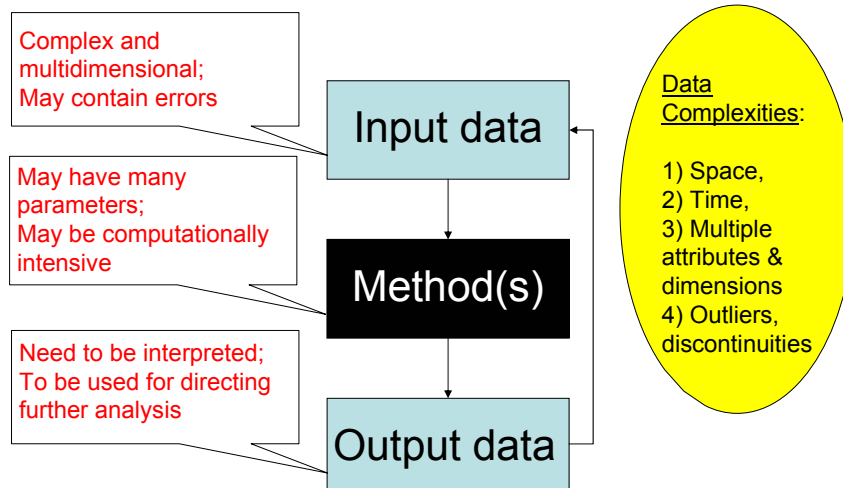


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A View on S-T Data Mining



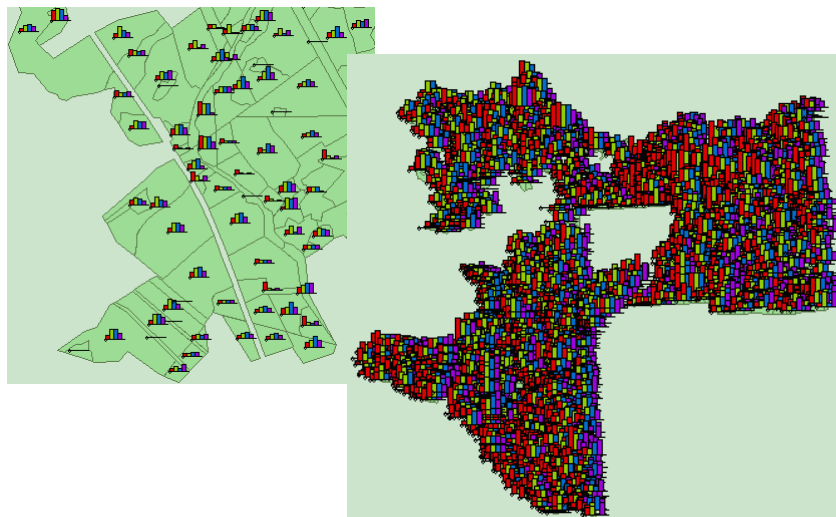
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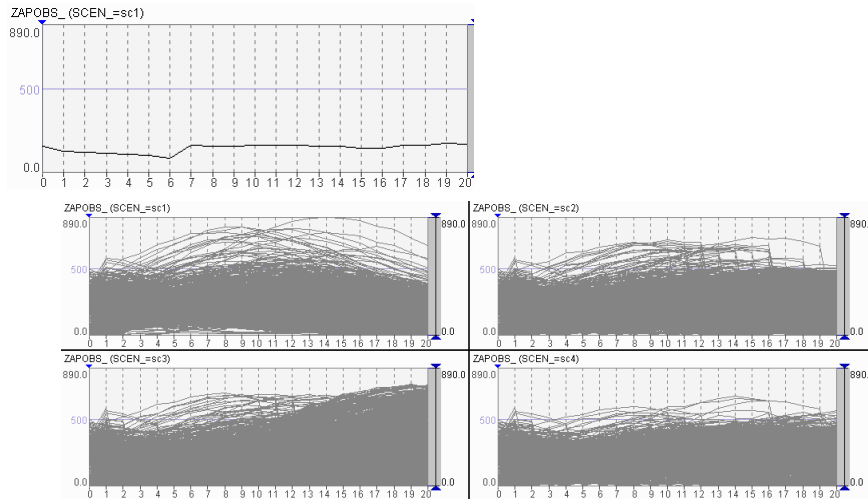
Complexities

- Number of attributes
- Length of time series
- Number of spatial objects
- High dimensionality
- Abrupt temporal changes
- Great variability of values

Complexities: example 1



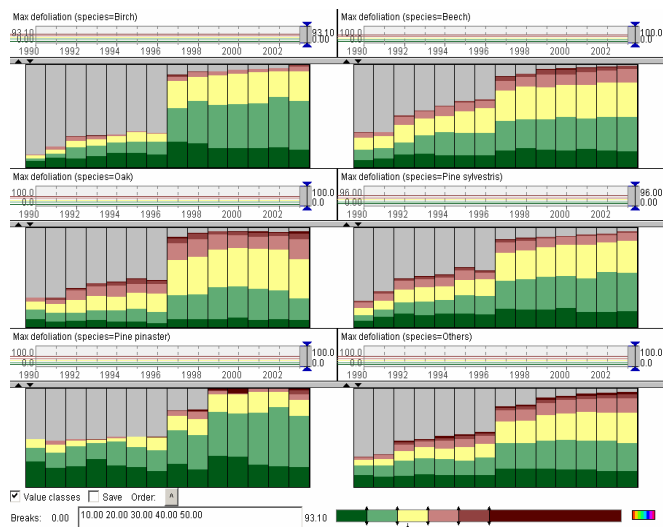
Complexities: example 2



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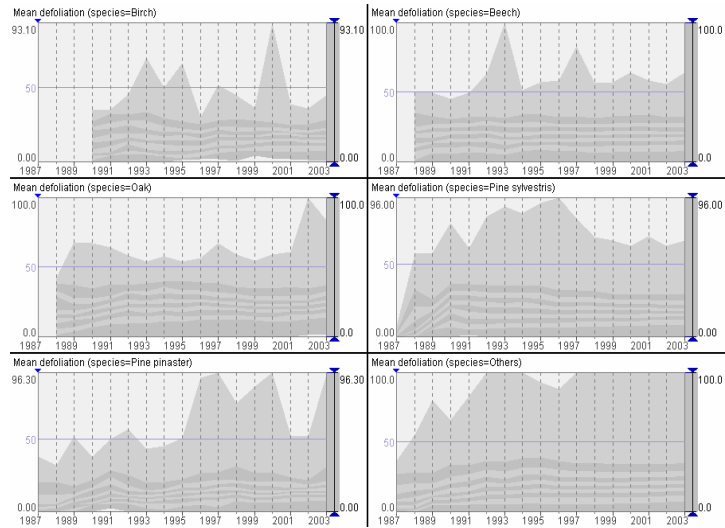
Aggregation method 1: by intervals



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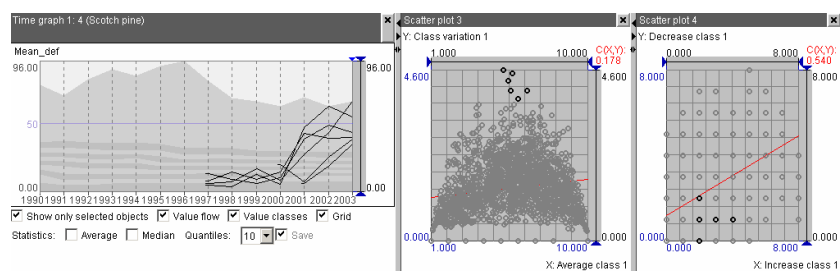
Aggregation method 2: by quantiles



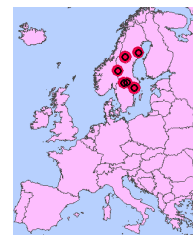
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Attend to particulars: high variation



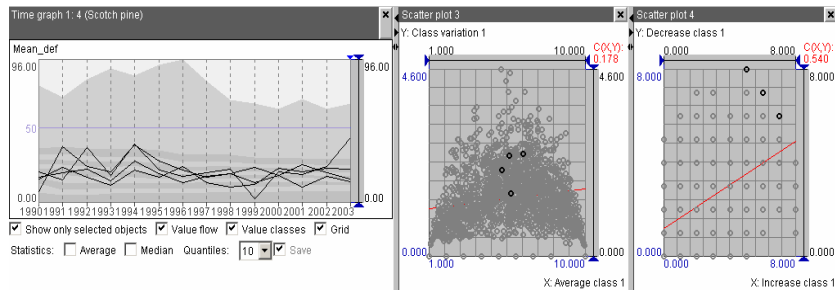
1. Aggregate time graph by quantiles
2. Save counts
3. Visualise e.g. on a scatter plot
4. Select items with high variation



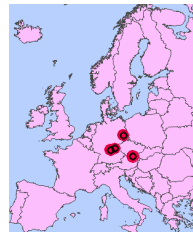
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Attend to particulars: high fluctuation



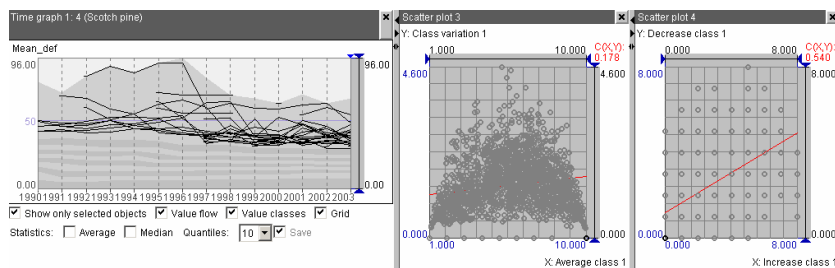
- Select items with maximal number of jumps between quantiles



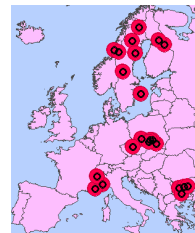
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Attend to particulars: stable extremes



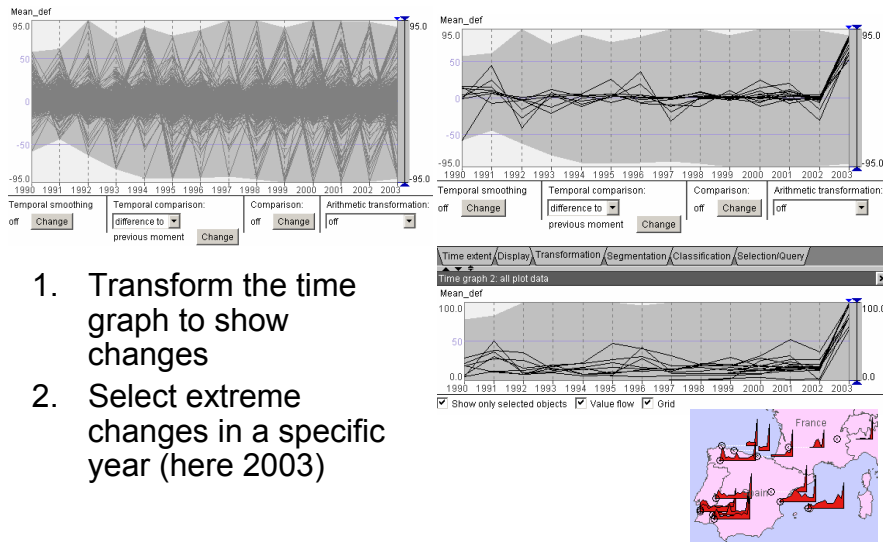
- Select items being always in the topmost 10%



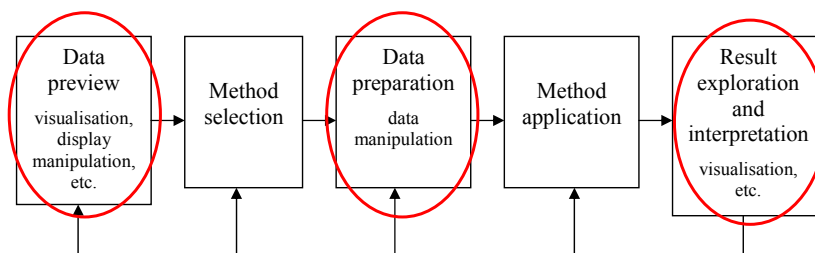
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Attend to particulars: extreme changes



Visual Analytics in Data Mining



Requirements to Visual Analytics

- Space- and Time-awareness
- Work with complex multidimensional data
- Support for uncertain and missing data
- Scalability
- Support and encouraging of several complementary views on the same data
- Dynamic linking and coordination of several data displays
- From the overall view to particulars of interest
- From idea generation to hypothesis testing using statistical methods, followed by reporting

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Potentially useful tools for MSTD

- Information visualisation tools, for example, HCE & TimeSearcher from HCIL, Univ. Maryland
- Geovisualisation tools, for example GeoVistaStudio (Penn State Univ.) and Descartes/CommonGIS (Fraunhofer Institute AIS)
- Graphical statistics tools, for example, Manet & Mondrian (Augsburg Univ.)
 - ☞ Usually such systems are research prototypes that implement innovative ideas, but provide restricted functionality and limited user support

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Still open issues (for all tools!)

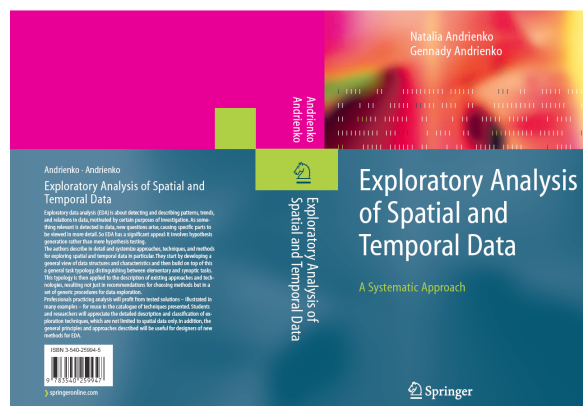
- Work with qualitative (non-numeric) data
- Work with fuzzy, uncertain, and missing data
- Continue scalability efforts
- Support in processing and management of findings: recording, structuring, browsing, searching, checking, combining, interpreting...
- Help in visual communication of derived data, constructed knowledge, and recommended decisions
- Adaptability to user, data, tasks, and hardware
- Embedding intelligence into software for helping users and avoiding cognitive overload

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EDA: from Practice to Practical Theory

- ✓ Data
- ✓ Tasks
- ✓ Tools
- ✓ Principles



to appear ~ end 2005

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Visual Analytics & Data Mining

1. Do they need each other?
2. How to benefit from combining two scientific disciplines and related technologies?
3. How to develop each of two scientific disciplines for achieving a synergy?