



Interactive Visual Exploration of Large Data Series

Anna Gogolou









What are time series?



Data sequences ordered along the dimension of time



Observations, measurements ordered along a dimension (time, angle, mass, position, etc.)



Why do we care about data series?

They occur in many scientific, medical, business, and social domains:

astronomy, biology, neuroscience, smart cities, nuclear power plants, finance, politics, ...

→ Our systolic/diastolic blood pressure

→ Donald Trump's popularity rating

→ The annual sunshine in Paris



Outline

Motivation

Our goal

Existing data series visualization tools

Scalability problems

State-of-the-art in data series management

Approximate vs exact search

Progressive visualizations

Motivation

How do we manage, explore and analyze large data series (order of terabytes)?

- More and more data series are produced every day:
 - 1 hour of ElectroCardioGraphy (ECG) data: 1 Gigabyte
 - Typical Weblog: 5 Gigabytes per week
 - EDF Database: ~100TB in total for all reactors (58), all sensors (10000 sensors per reactor), over all years and growing …

Data series analysis

Analysis Tasks:

Pattern Matching

Frequent Pattern Mining

Clustering

Classification

Outlier Detection

• • •

Pattern matching



User-defined pattern

~ Maken and and many many and and and and a share . may hy may a my har man how my how how was hard with the second of the and making the and the second the second of any many where we have been and the second of the second second and the second se a many property and the second of the stand ~ www.www.www.www.www.www.www.h.www.h.www.h.www.ly.w. - while when a ward when a man and the second of the secon m rywym m MWW www. which mak anna managed and many approximation of the second of the ~-www.www.www.www. DB Collection of data series

Pattern matching



User-defined pattern

~ Maker and the second war and the second and the second of the second o . way water water and the water water and the water and the second and walk walk and the second the second of the second and the seco and warden and the stand and the stand and the second and the seco a many property and the second of the stand - way and a har a har and and a har and a har a ~ while when a frank when when when when the when the when the when the when the second of the secon m rywym m MWW www. which mak anna managed and many approximation and the particular and the ~-www.www.www.www. Man Marine Man Marine Ma Marine Marin DB Collection of data series

Outline

Motivation

Our goal

Existing data series visualization tools

Scalability problems

State-of-the-art in data series management

Approximate vs exact search

Progressive visualizations

Our goal

Interactive visual exploration of large data series

Interdisciplinary topic in the areas of:

Human-Computer Interaction (HCI) & Information Visualization

Understand what analysts do with their data

Support interactive exploration of large datasets

Databases

Analyze efficiently large data series

Get answers quickly

Outline

Motivation

Our goal

Existing data series visualization tools

Scalability problems

State-of-the-art in data series management

Approximate vs exact search

Progressive visualizations

Existing data series visualization tools



Aggregation

BinX

High level of aggregation (Overview)



Lior Berry, Tamara Munzner, "Binx: Dynamic exploration of time series datasets across aggregation levels", *Information Visualization 2004. INFOVIS 2004. IEEE Symposium on pages p2-p2. IEEE*, vol. 5, 2004. [doi>10.1109/INFVIS.2004.11]

ChronoLenses

Data series chart panel



Jian Zhao, Fanny Chevalier, Emmanuel Pietriga, Ravin Balakrishnan, Exploratory Analysis of Time-Series with ChronoLenses, *IEEE Transactions* on Visualization and Computer Graphics, v.17 n.12, p.2422-2431, December 2011 [doi>10.1109/TVCG.2011.195]

Focus+Context

Line graph explorer

Line graph panel



Robert Kincaid, Heidi Lam, Line graph explorer: scalable display of line graphs using Focus+Context, *Proceedings of the working conference on Advanced visual interfaces*, May 23-26, 2006, Venezia, Italy [doi>10.1145/1133265_1133348]

Visual clustering

Viztree



Jessica Lin, Eamonn Keogh, Stefano Lonardi, Visualizing and discovering non-trivial patterns in large time series databases, *Information Visualization*, v.4 n.2, p.61-82, July 2005 [doi>10.1057/palgrave.ivs.9500089]

Periodic patterns

Spirals



Christian Tominski, Heidrun Schumann, Enhanced interactive spiral display, *Proceedings of the annual SIGRAD conference*, pp.53–56 (2008)

Compact visualization

Horizon charts



Jeffrey Heer, Nicholas Kong, Maneesh Agrawala, Sizing the horizon: the effects of chart size and layering on the graphical perception of time series visualizations, *Proceedings of the SIGCHI Conference on Human Factors in Computing Systems*, April 04-09, 2009, Boston, MA, USA [doi>10.1145/1518701.1518897]

Sketching a query

The semantics of sketch

The matching panel



Michael Correll, Michael Gleicher, The Semantics of Sketch: A Visual Query System For Times Series Data, *Proceedings of the 2016 IEEE Conference on Visual Analytics Science and Technology (VAST)*, October 2016

Relaxed queries

SoftSelect

The graphdisplay and main-interaction area



Christian Holz, Steven Feiner, Relaxed selection techniques for querying time-series graphs, *Proceedings of the 22nd annual ACM symposium on User interface software and technology*, October 04-07, 2009, Victoria, BC, Canada [doi>10.1145/1622176.1622217]

Outline

Motivation

Our goal

Existing data series visualization tools

Scalability problems

State-of-the-art in data series management

Approximate vs exact search

Progressive visualizations

What about handling billions of data series?

None of these tools support scalability to terabytes of data:

How do we visualize them?

Limited number of pixels

Limited human cognitive resources

How do we interact with them?

Days to answer a single query

- Visual scalability

Interactive response time scalability

Limit: < 100 ms J. Nielsen, Response times: The 3 important limits, <u>https://</u>

www.nngroup.com/articles/response-times-3-important-limits/, January 1, 1993

Outline

Motivation

Our goal

Existing data series visualization tools

Scalability problems

State-of-the-art in data series management

Approximate vs exact search

Progressive visualizations



A raw data series T



Piecewise Aggregate Approximation (PAA)



Symbolic Aggregate approXimation (SAX)



Symbolic Aggregate approXimation (SAX)

Indexing

How can queries be answered faster?

Using index structures (usually trees), which improve the speed of database operations on database tables (insert, select, delete, ...).



Data series indexes

Data series indexes follow a similar tree structure.

They are built on top of data series summarizations.

ADS+¹ is the state-of-the-art data series index. It is based on:

SAX representation

Euclidean Distance (ED) between SAX representations



Response times

Finding the most similar pattern on 1 **billion** data series:



Outline

Motivation

Our goal

Existing data series visualization tools

Scalability problems

State-of-the-art in data series management

Approximate vs exact search

Progressive visualizations

Approximate vs exact search

Approximate search:

Returns answer not guaranteed to be the best one Associated with an approximation error

Very fast (order of milliseconds)

Exact search:

Always returns exact correct answer

Slow (order of minutes)

Approximate search



Exact search



Outline

Motivation

Our goal

Existing data series visualization tools

Scalability problems

State-of-the-art in data series management

Approximate vs exact search

Progressive visualizations

Progressive visualizations

How can users progressively move from approximate to exact results?



Visualizing approximation error



Danyel Fisher, Steven M. Drucker, Arnd Christian Koenig, Exploratory Visualization Involving Incremental, Approximate Database Queries and Uncertainty, *Journal IEEE Computer Graphics and Applications*, Volume 32, Issue 4, July 2012, Pages 55-62, [doi>10.1109/MCG.2012.48]



Provide quickly partial results

Visualize approximation error

Support iteration and refinement over approximate & progressive results

Easily focus on subspaces of interest

Conclusions

Open problems:

Scalability to terabytes of data

Current visualization tools cannot handle large data series collections

Need techniques that:

effectively visualize large volumes of data

have interactive response times (< 100ms)

Support for iteration and refinement of approximate & progressive results

Currently no such support for data series analytics

Need techniques that:

visually inform users on progress of task

convey information on accuracy of current results