



Universität Stuttgart

GeoVis

– From Terrain to Tweets and Movements

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 Institut für Visualisierung
und Interaktive Systeme

51. Photogrammetrische Woche 2007

Towards Gigapixel Displays for Data Visualization

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Conclusions

- Large displays offer new opportunities for
 - science, development, education, entertainment
 - to deal with the increasing amount of visual information
 - by visualization, interaction, and collaboration
- Large displays need
 - wide screens AND high resolution
 - enormous graphics power (GPU clusters)
 - new rendering and visualization algorithms
- GPU-based visualization techniques
 - allow for interactive exploration of huge data sets
 - are a fun research area

TE007

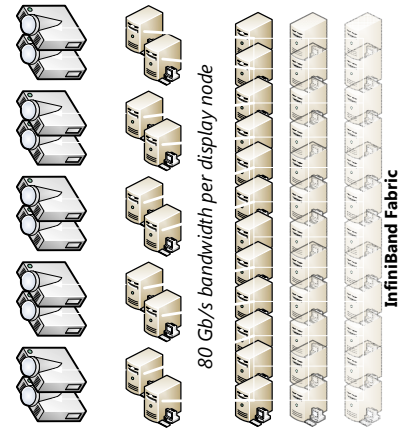
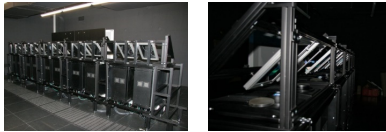
The Stuttgart PowerWall Project

- Visualization Research Center of the Univ. of Stuttgart (VISUS)
- Moved into remodeled „exotic“ building in 2010
- Funding for new “immersive visualization lab”
- Decision for high-res back projection stereo wall
- Visualization should benefit more from highres than from immersion
- Large Screen (display full car model): 6.0 m x 2.2 m
- Stereo for larger groups, no bevels
- Higher resolution than usual
 - (typical PowerWall pixel > 2 mm)
- Goal: monitor resolution - pixel size 0.5 mm (50 dpi)
 - requires close to 2 x 50 Megapixels (50 HD projectors)



Projection System

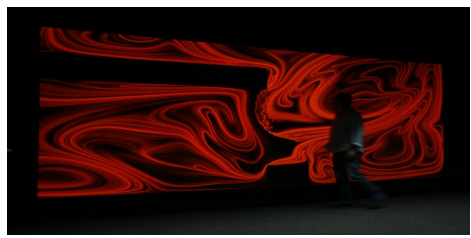
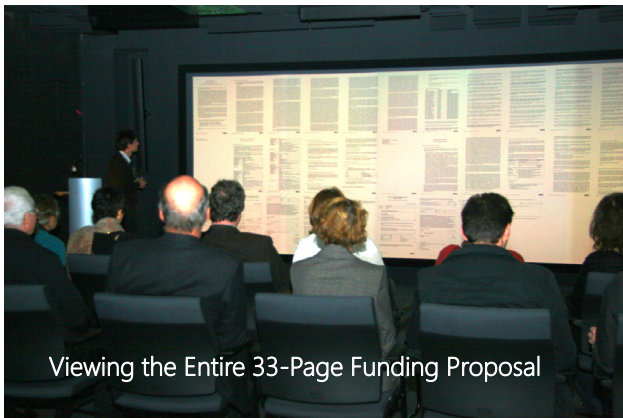
- Use as few projectors as possible and minimize seams
- 4K projectors (9 Megapixels via 4 video-in) in 5 vertical stripes
- 10x 4K projectors require 40 video-in and at least 20 Gbyte/s
- Gross 100 Megapixels must be generated
 - Only display nodes are connected to projectors
 - Render nodes perform off-screen rendering
 - 600 CPU cores
 - 150 GPUs



10x JVC DLA-SH4K
10x Display Nodes
 2 CPUs, 2 GPUs, Framelock

64x Render Nodes
 2 CPUs, 2 GPUs

Applications



TRR 161
 Transregional Collaborative Research Center
 Quantitative Methods for Visual Computing

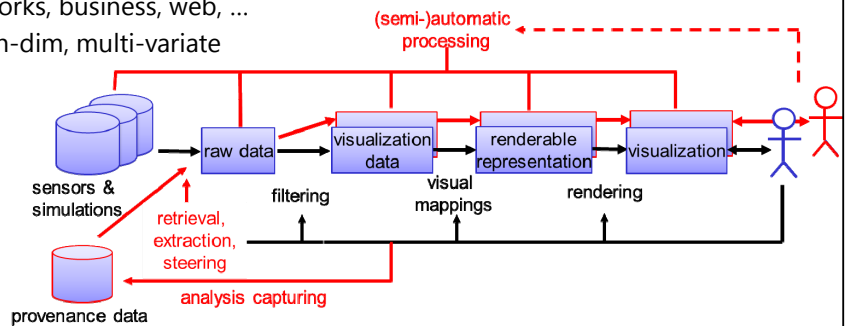


Visualization – Diversity after 25 Years

- Scientific Visualization
 - Science and engineering, medical imaging, ...
 - Spatial data, fields, particles, underlying physics
- Information Visualization
 - Databases, networks, business, web, ...
 - Non-spatial, high-dim, multi-variate
- Visual Analytics
 - Data analysis, interaction, visualization

National Science Foundation ViSC Report 1987:

Visualization is a method of computing. It transforms the symbolic into the geometric, enabling researchers to observe their simulations and computations. Visualization offers a method for seeing the unseen.



GeoVis – An Obvious Combination

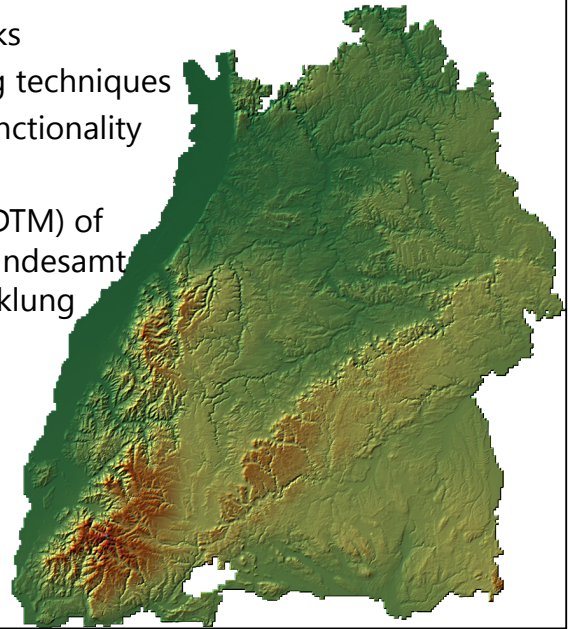
- Spatial data exhibits an outstanding value in visualization: “Everything is related to everything, but near things are more related than distant things.” [Tobler, 1970]
- If any feature can be visualized spatially, visualization will implement a map view for finding patterns, trends, and outliers in the data
- Three examples for visualization approaches for spatial data
 - Terrain visualization
 - Geo-referenced social media
 - E-bikes trajectories



Terrain Visualization



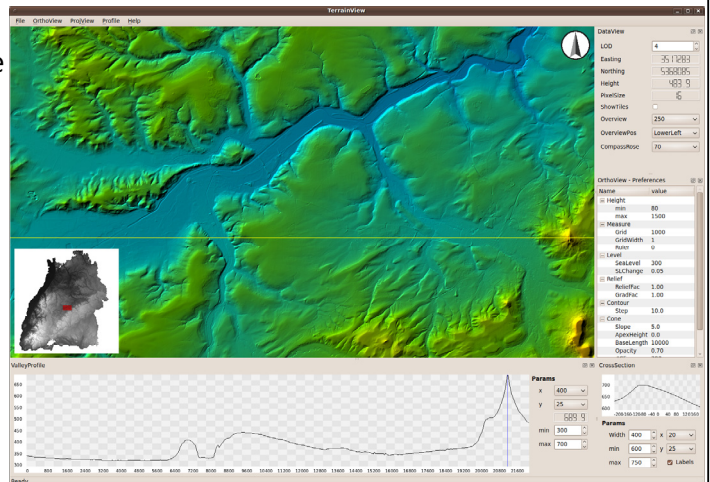
- Is this still interesting with Google Earth being available?
- Yes, for large and high-resolution data
- Yes, for advanced visual analysis tasks
- Yes, for advanced rendering/shading techniques
- Yes, for exploiting advanced GPU functionality
- LiDAR-based digital terrain model (DTM) of Baden-Württemberg provided by Landesamt für Geoinformation und Landentwicklung
 - covers about 360 00 km²
 - horizontal resolution: 1 m
 - vertical resolution: 15 cm
 - raw dataset:
 - about 37 000 tiles
 - à 1001 m x 1001 m => 1 TB



TerrainView

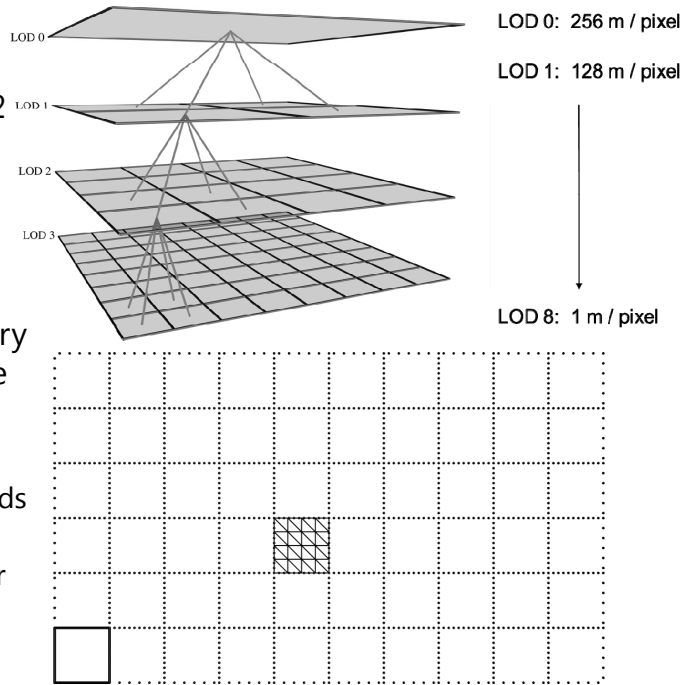
Thomas Müller, et al.

- Project collaboration with Prof. Hartmut Seyfried on identification and interpretation of young (Würmian) glacial geomorphology
- Commercial tools can handle only a subset of the data at once
- Goal: interactive visualization of entire dataset on desktop PC
- TerrainView functionality – all in “real-time”
 - Terrain visualization orthographic/perspective
 - coloring, relief shading (variable light directions), gradient representation, sea level, contours
 - calculate valley profiles along polylines for geomorphological investigs.
 - new cone model to estimate glacial loads



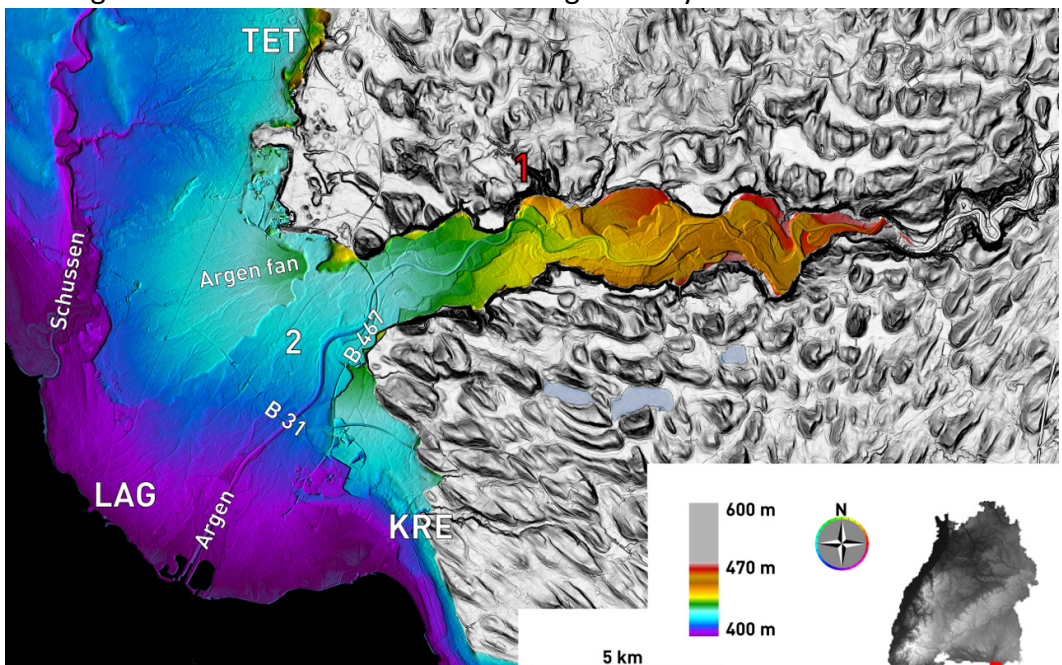
TerrainView Techniques

- Preprocessing step: resample into quadtree structure: $1024 \times 1024 \text{ m}^2$
- Heavy use of GPU features
- New shaders change the way how we do graphics
- Instead of sending geometry to the graphics card, let the GPU generate the triangles
 - Send one quad to GPU
 - Replicate into 112×80 quads by instancing
 - Tessellation control shader -> triangle mesh
 - Tessellation eval. shader -> adjust height

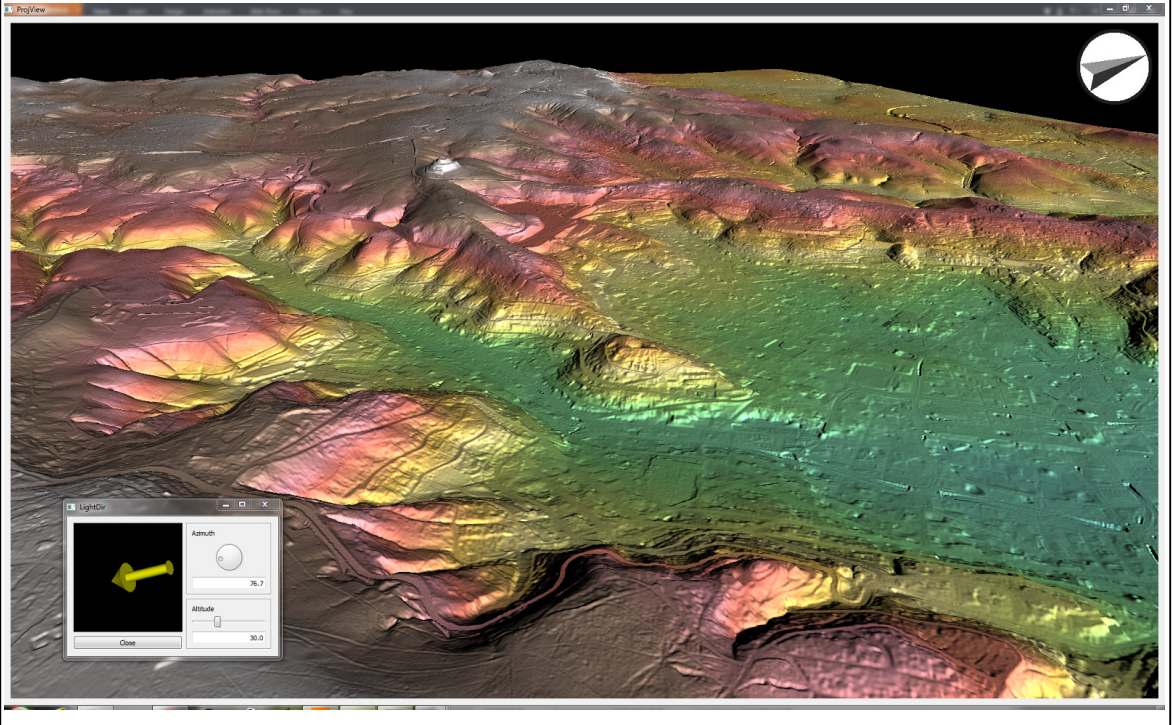


TerrainView - applications

- Post-glacial recessional terraces in the Argen valley



TerrainView - applications



GeoData for Situation Awareness

- We live in a world where cybersocial and cyberphysical systems produce billions of geo-related data records on a daily basis:
 - Twitter Posts
 - Instagram Photos
 - Facebook Posts
 - Flight Records
 - Hotel Check-Ins
 - Passenger Info
 - Traffic Data
 - Crime Records
 - Census
 - Weather data
 - Public Webcams
 - etc ...
- Accessing, extracting, and combining this information would significantly help to inform situation awareness, intelligence, and decision-making -> a typical Visual Analytics task
- Technical challenges in utilizing the information are commonly referred to as the four V's of Big Data:



Volume



Velocity



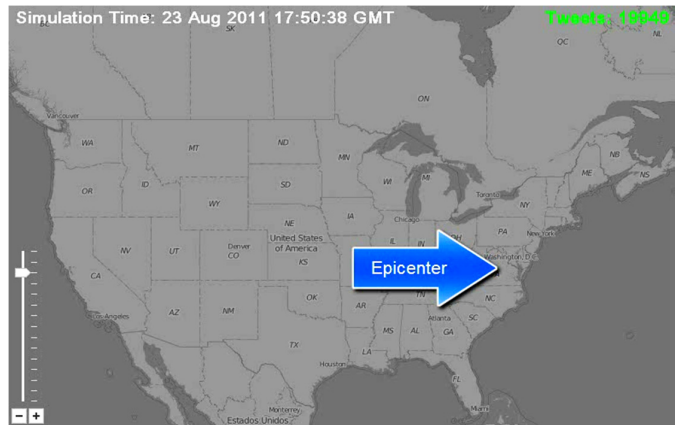
Variety



Veracity

Twitter for Situation Awareness

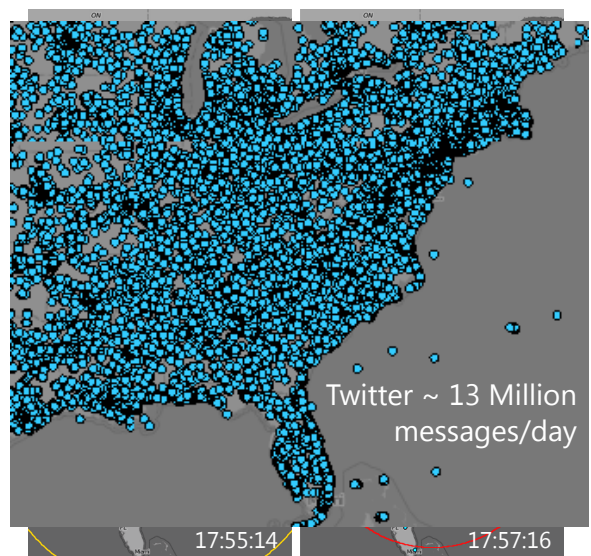
- Simulation of Virginia Earthquake 2011
 - Yellow = P-Wave
 - Red = S-Wave
 - Blue = Tweets
- Event demonstrates the high **timeliness** and **distribution** of social media reactions



Twitter for Situation Awareness

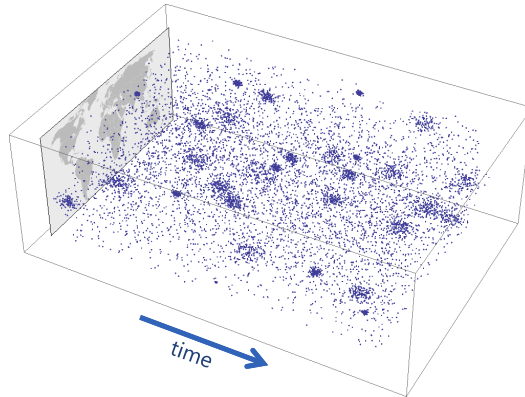
- Information extracted from the data could be of great value for decision makers in
 - Disaster Management
 - Public Safety
 - Disease Control
- Research question
How to **identify relevant information** and produce **meaningful situation overviews** from millions of data items in real-time?

Dennis Thom, et al.



Spatiotemporal Anomalies

- **Assumption 1:** Event is the most central entity in situation awareness
- **Assumption 2:** Events generate spatiotemporal clusters of similar tweets



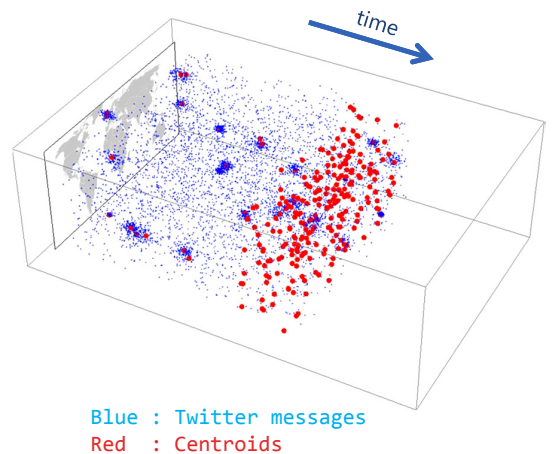
Spatiotemporal Anomalies

- Traditionally, analysts would enter **keyword queries** to find this kind of clusters
- **Idea:** Revert the process by **detecting such clusters** in the data and use them to generate a **visual overview** of what should be searched for



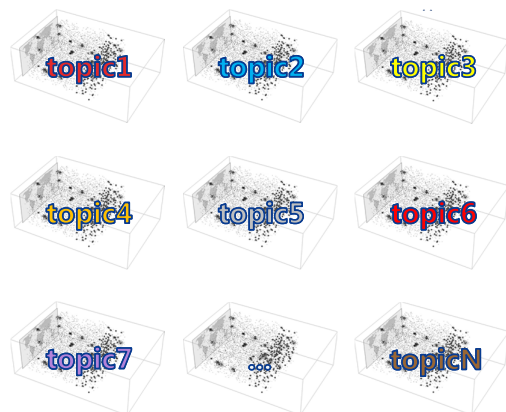
Anomaly Discovery

- Streaming-enabled cluster analysis based on K-Means
 - Instead of a fixed number of centroids (means), a **splitting mechanism** is employed
 - A sliding window is used to **evaluate/discard clusters** once they turn stale



Anomaly Discovery

- Clustering performed separately for any observed topic to account for „clusters within clusters“
- Each new message is assigned to topics and **only the corresponding cluster branches** are updated



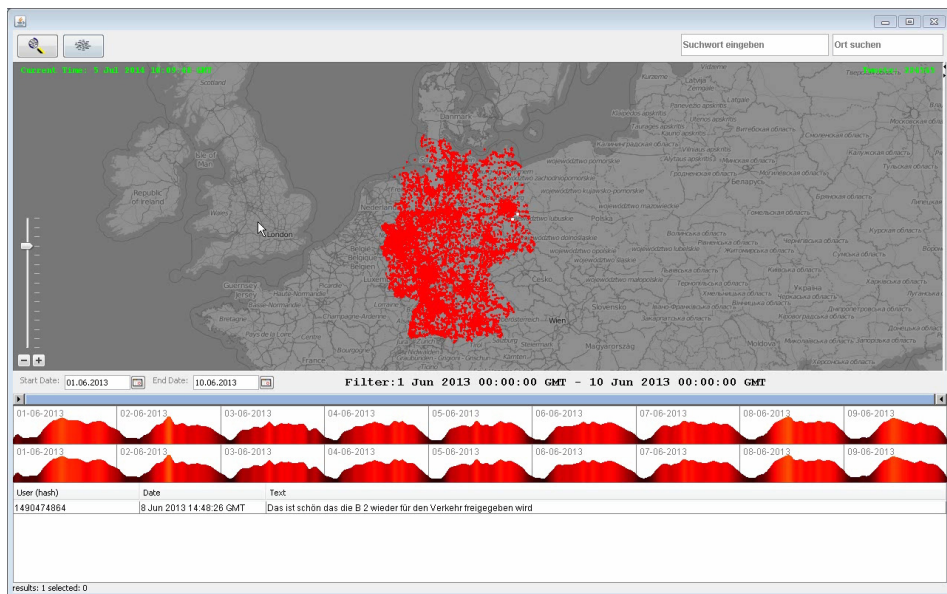
Anomaly Visualization



- Detected clusters are placed as **labels on the map** – collision resolution produces tagcloud layout
- **Similar overlapping labels** are aggregated to counteract overfitting and allow adaptive semantic zoom

earthquake
earthquake

ScatterBlogs Visual Analytics

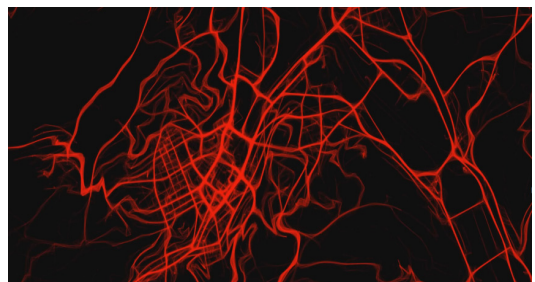


Data Enrichment

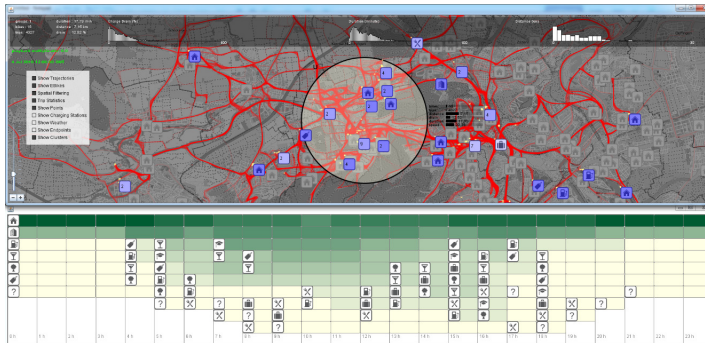
- In services like Twitter and Foursquare, users also provide information about relevant Points of Interest (POI) in an environment.
- Traditional geographic data can be interactively enriched with such knowledge from Web 2.0 data sources.
- For example: To better understand urban dynamics, but also to enable consumer acceptance analysis, there is an increasing interest to look into movement reasons.

Data and Task

- Data: Large electric mobility dataset
 - 500 electric scooters,
155.000 trips, 8 mio
measurements
- Task: Usage analysis
 - *Can e-mobility find its way into everyday's life?*
 - *When, where, what for ...
... do costumers use their
scooters?*



Find Reasons of Movement



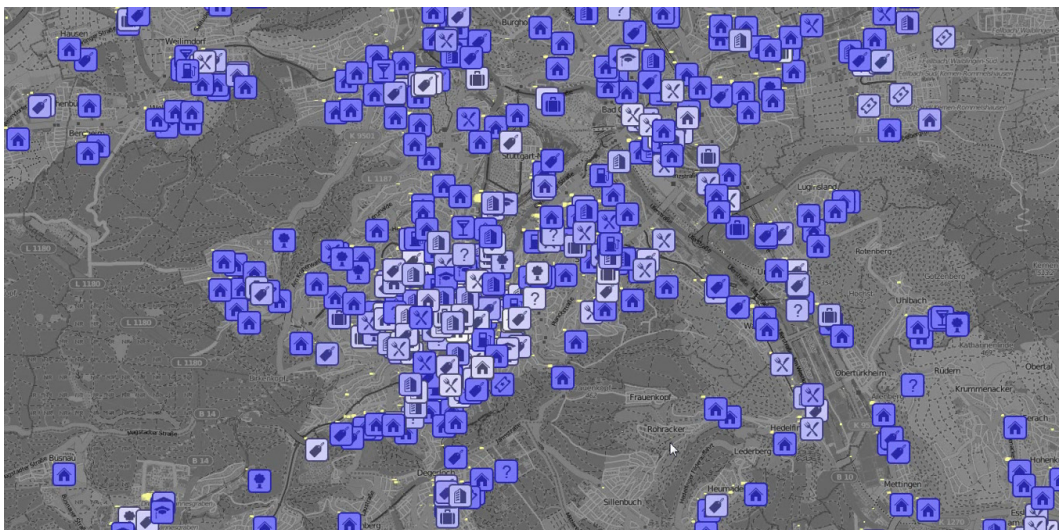
- Use data from Foursquare to annotate areas of interest with reasons of movement



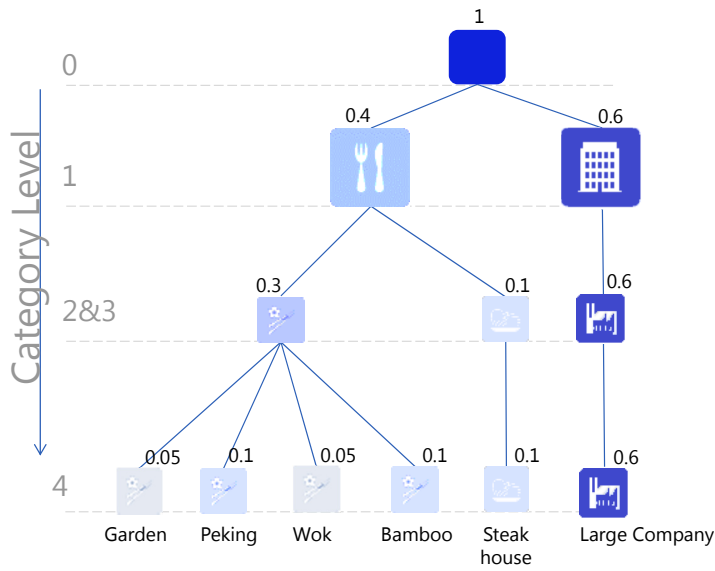
Visual Analysis of Movement Behavior using Web Data for Context Enrichment [Krueger et al. 2014]

Interactive Map

- Plotting POIs near clusters -> information overload and visual clutter



Certainty Aggregation



■ Certainty Criteria:

Distance

Checkins

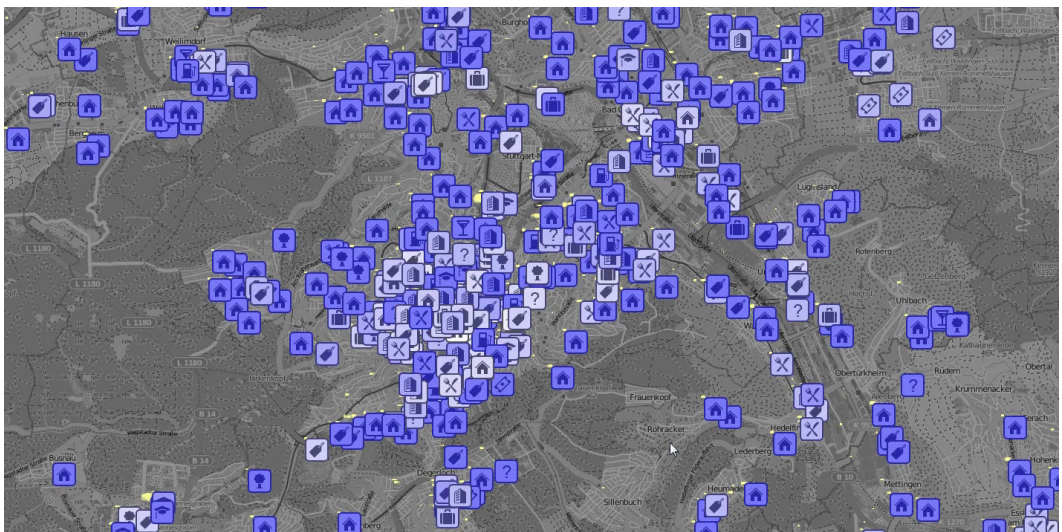
Users

Combination

Color legend:

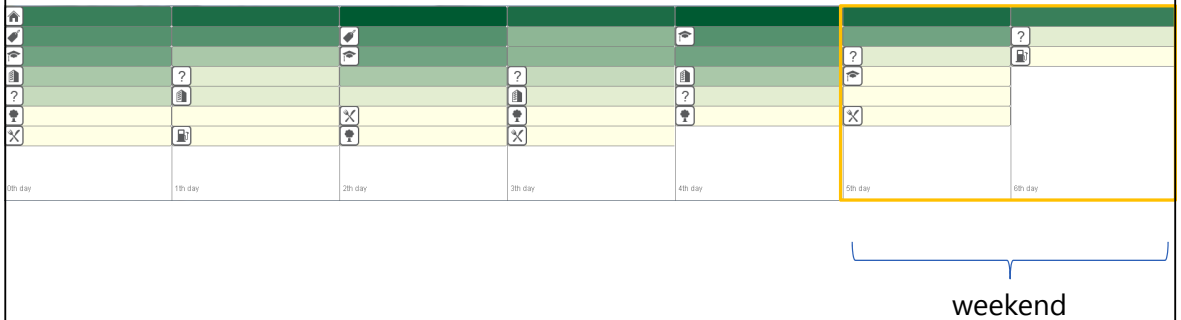


Certainty Aggregation



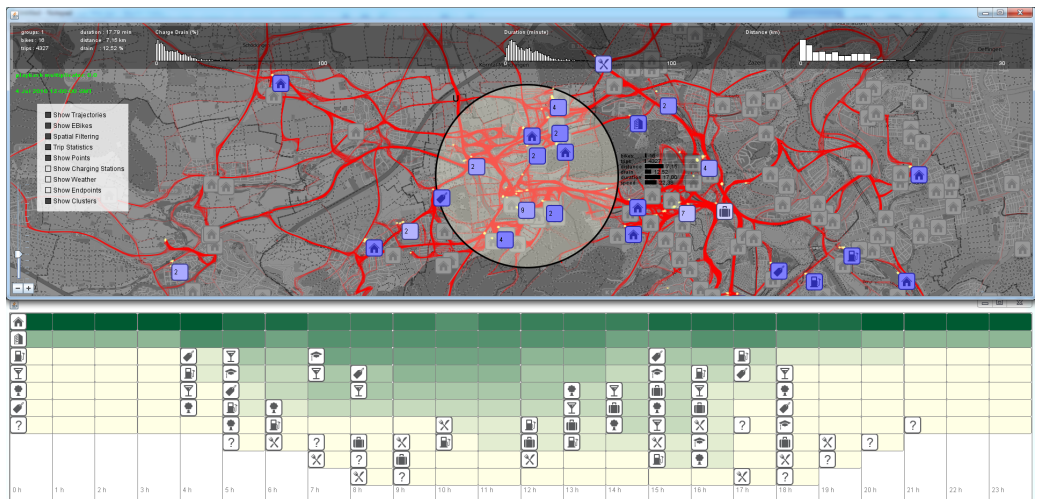
Temporal Analysis

- Evaluation of cyclic behavior (week: Monday to Sunday)



Analysis Examples

- Use Case: Area related usage



Conclusion

- Cybersocial and cyberphysical systems will continue generate enormous amounts of data, many of which will have a spatial relation
- Visualization plays an important role in the exploitation of this data
 - Terrain visualization continues to provide research challenges like correlations of geologic and geomorphic phenomena.
 - Spatial properties of data can be used to enable new visualization forms of abstract data, such as events in social media.
 - Context knowledge in Web 2.0 sources can be employed to enrich traditional geographic data.
- Visualization research can help to enable space-time indexed data exploration, the discovery of unknown correlations, and deep insights about the semantic realm entangled with our geospatial environment.