

Visual Analytics to support research & applications for a sustainable future

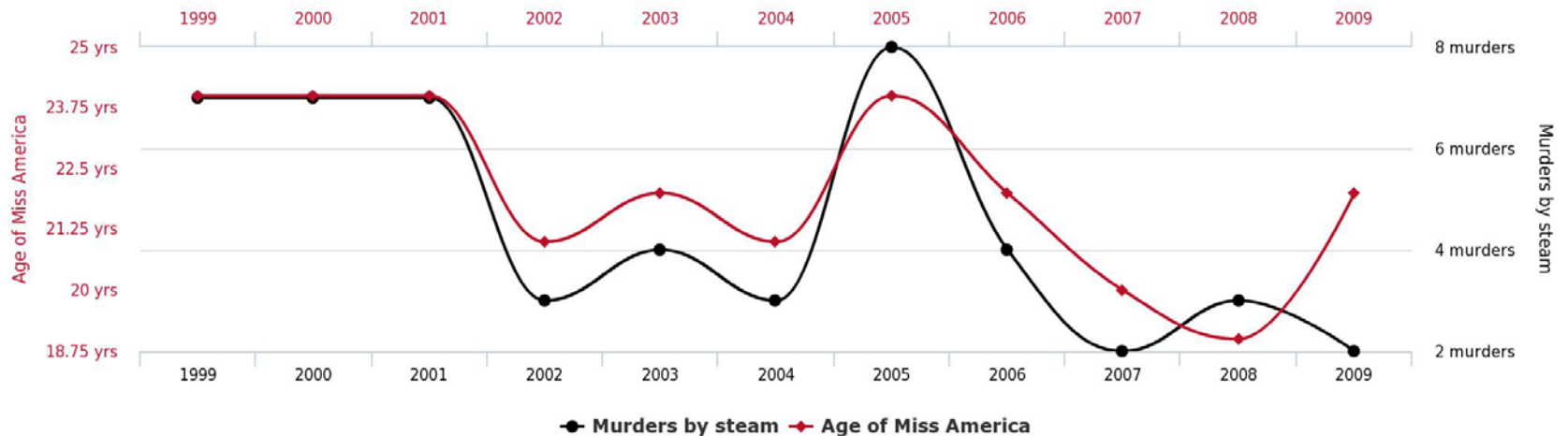
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INSTITUTE
OF SCIENCE
AND TECHNOLOGY



Evidence in data – case 1

Age of Miss America correlates with Murders by steam, hot vapours and hot objects



tylervigen.com

correlation \neq causation

Evidence in data – case 2

Anscombe's example*

Let's consider datasets I- IV with:

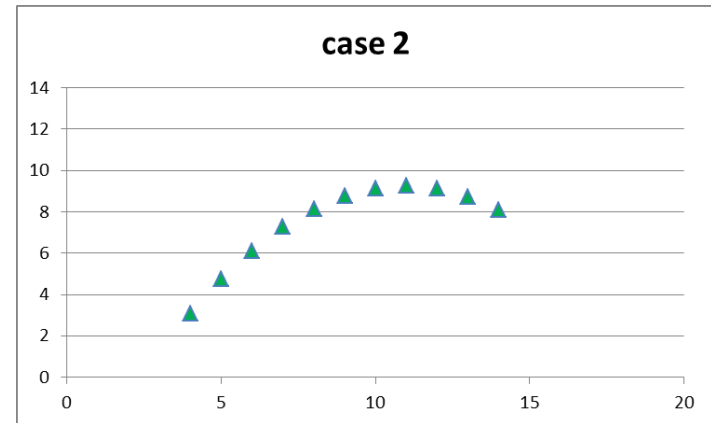
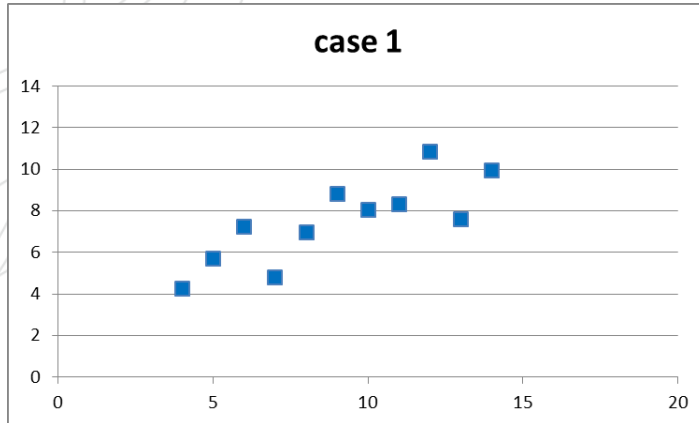
- ✓ same mean (x) = 9.0,
- ✓ same mean (y) = 7.5
- ✓ same x variance
- ✓ nearly same y variance
- ✓ identical (x,y) correlation
- ✓ identical regression lines

| I | | II | | III | | IV | |
|------|-------|------|------|------|-------|------|-------|
| x | y | x | y | x | y | x | y |
| 10.0 | 8.04 | 10.0 | 9.14 | 10.0 | 7.46 | 8.0 | 6.58 |
| 8.0 | 6.95 | 8.0 | 8.14 | 8.0 | 6.77 | 8.0 | 5.76 |
| 13.0 | 7.58 | 13.0 | 8.74 | 13.0 | 12.74 | 8.0 | 7.71 |
| 9.0 | 8.81 | 9.0 | 8.77 | 9.0 | 7.11 | 8.0 | 8.84 |
| 11.0 | 8.33 | 11.0 | 9.26 | 11.0 | 7.81 | 8.0 | 8.47 |
| 14.0 | 9.96 | 14.0 | 8.10 | 14.0 | 8.84 | 8.0 | 7.04 |
| 6.0 | 7.24 | 6.0 | 6.13 | 6.0 | 6.08 | 8.0 | 5.25 |
| 4.0 | 4.26 | 4.0 | 3.10 | 4.0 | 5.39 | 19.0 | 12.50 |
| 12.0 | 10.84 | 12.0 | 9.13 | 12.0 | 8.15 | 8.0 | 5.56 |
| 7.0 | 4.82 | 7.0 | 7.26 | 7.0 | 6.42 | 8.0 | 7.91 |
| 5.0 | 5.68 | 5.0 | 4.74 | 5.0 | 5.73 | 8.0 | 6.89 |

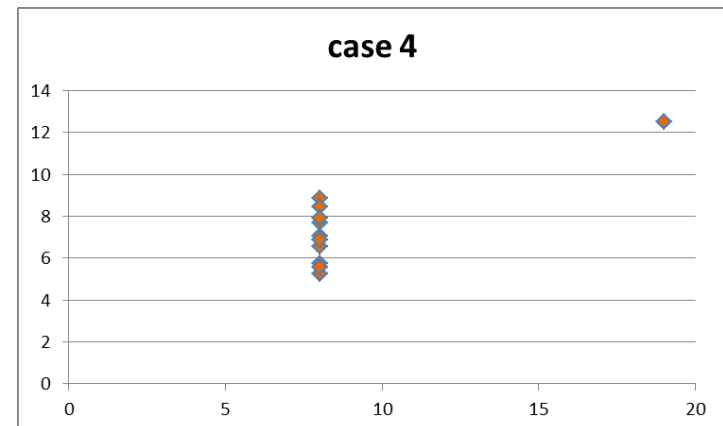
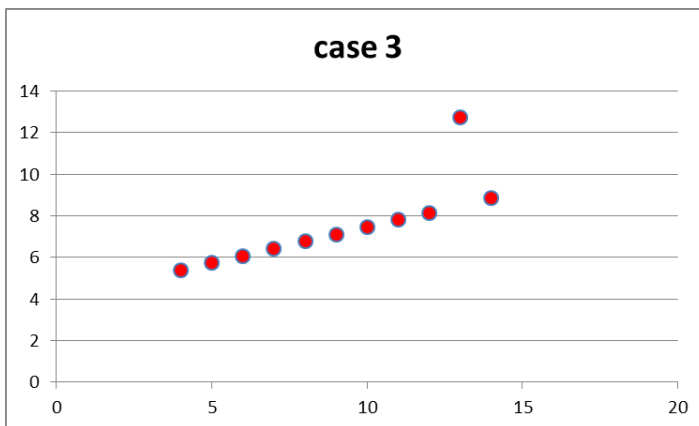
similarity between the 4 cases could be deducted

* The American Statistician, Vol. 27 (1), pp. 17-21 (1973)

Case 2 continued



Seeing the graphics makes the differences crystal clear

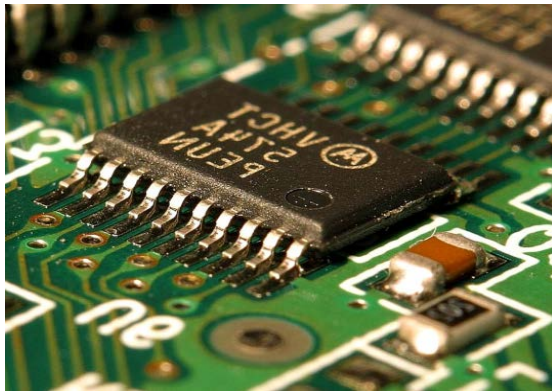


Data to knowledge

Making sense of data needs perspectives & domain knowledge (case 1)

Combining human & machine is successfully supported by visualisation (case 2)

Data Analytics Approaches
(statistics, machine learning...)



Data Visualisation Approaches
(InfoVis, 3D Visu., Maps...)

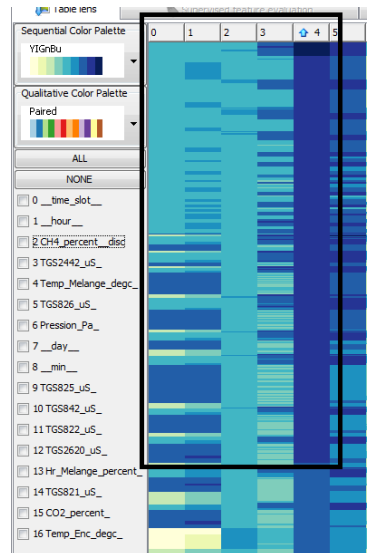


How to use Computer + Human Vision in the most efficient way?

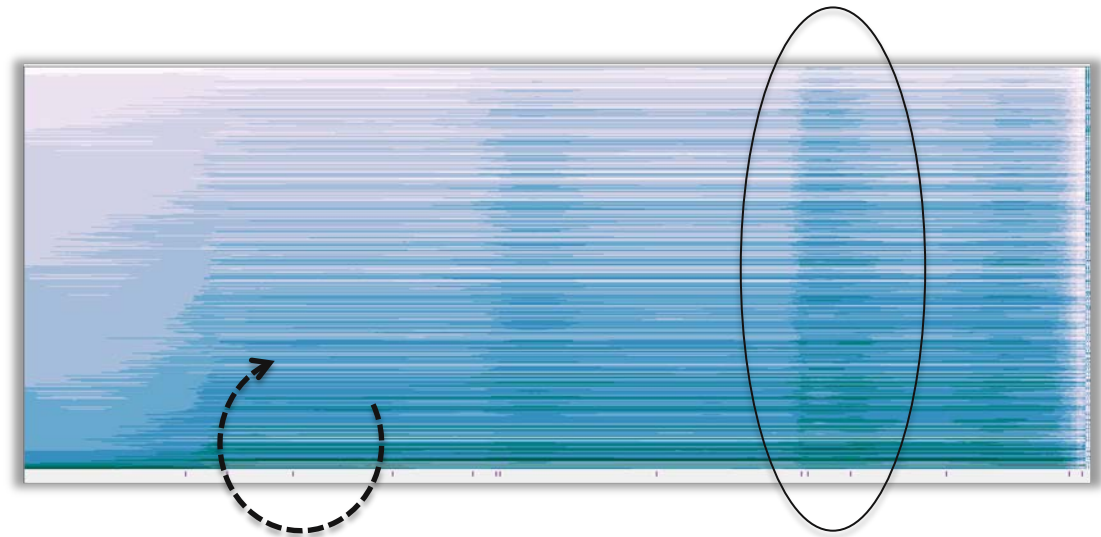
Human in the loop

Visual perception

- Most important input channel to communicate information to the brain
- Very efficient, hard-coded in the human brain
- Parallel processing of visual variables (e.g. colours, shapes, movement...)
- Intuitive, no need for teaching
- Still much more efficient than computer vision



Pattern recognition



Human vision support (real-time) interaction with graphics

Visual Analytics Paradigm

Visual analytics is the science
of analytical reasoning
facilitated by interactive visual interfaces*



* Illuminating the Path. The Research and Development Agenda for Visual Analytics, J.J .Thomas and K.A Cook, IEEE Editions, 2005.

Visual Analytics in Environment

Model is not the reality

Model → Abstraction, simplification

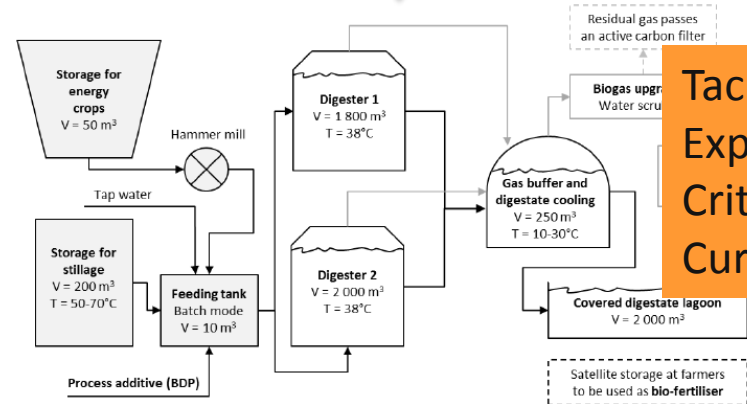


Figure from *

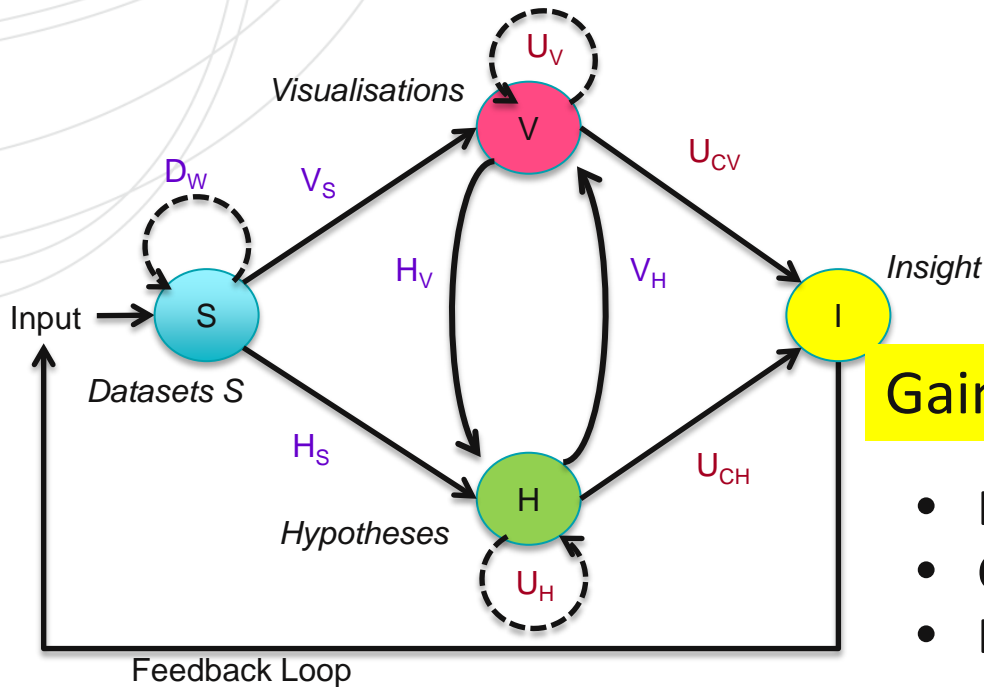
Tacit knowledge
Experience
Critical thinking
Curiosity

"I remember that I've already observed something similar"

*"This behaviour contradicts this well-known chemical rule,
we should check the data again"*



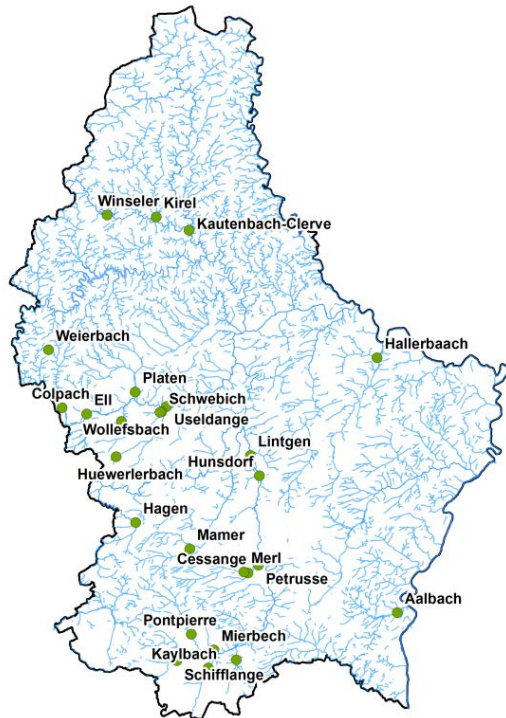
Visual Analytics in practice



Gaining Insight... for:

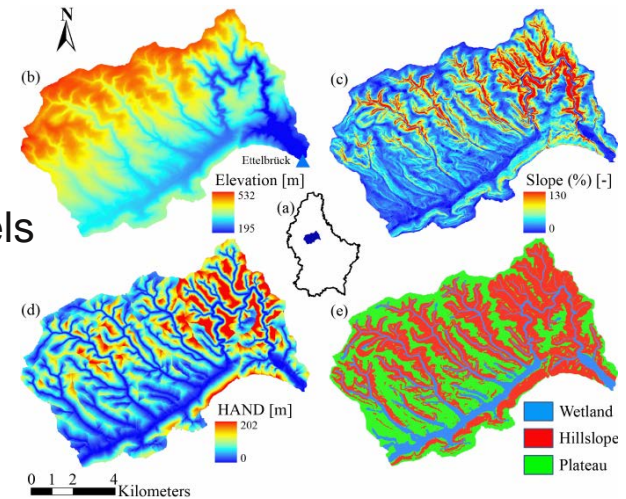
- Data pre-processing
- Complex system understanding
- Ensemble exploration (simulation)
- Crisis management, citizen science
- ...

Data pre-processing (hydrology)



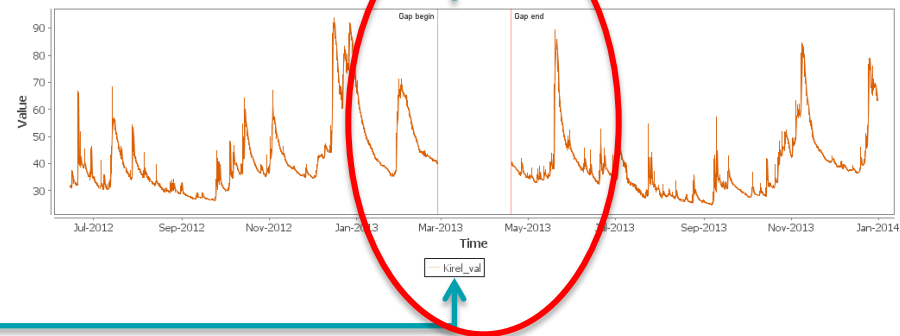
Raw Data
Collected from in-situ sensors

Hydrological models
Simulation results



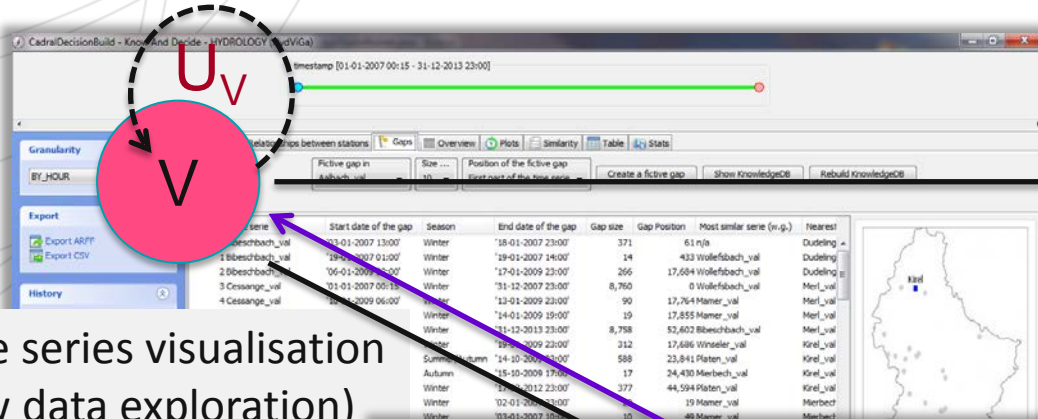
Pre-processed data

Missing Values



Hydrology continued

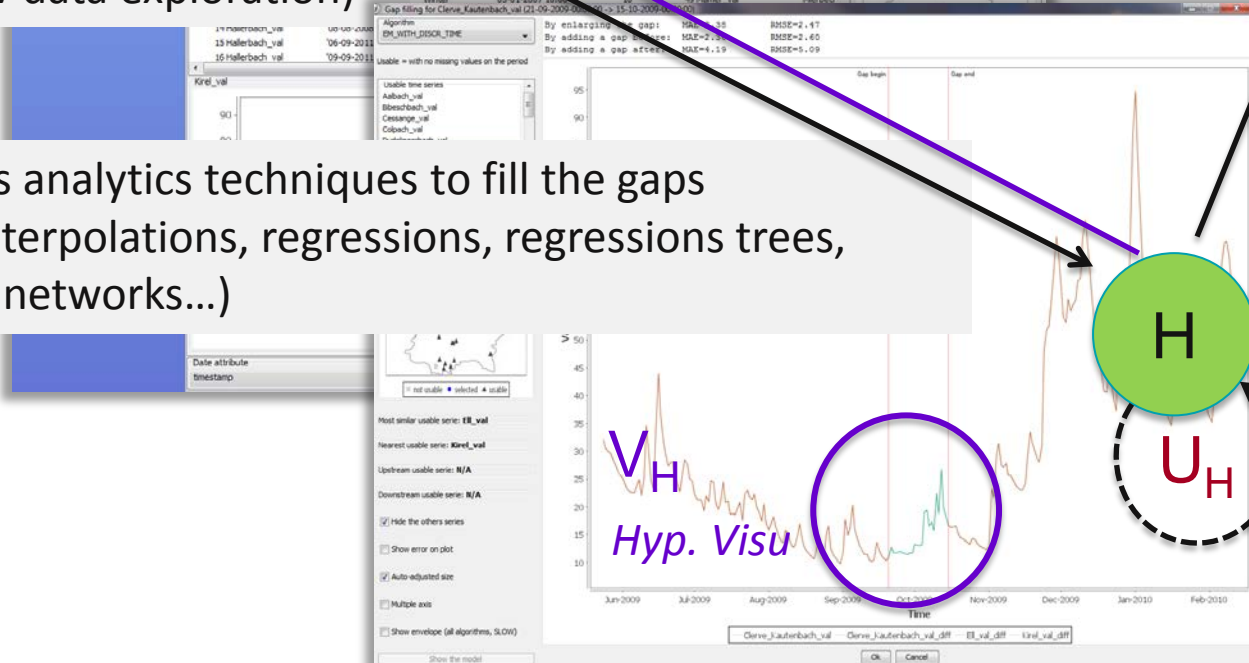
User Interact. with Visu.



Time series visualisation
(Raw data exploration)



Various analytics techniques to fill the gaps
(e.g. Interpolations, regressions, regressions trees, neural networks...)



User Interact. with Hyp.



Complex systems (viticulture)

Better understanding the relationship between the wine and the terroir where it is produced

Heterogeneous raw data: 108 parameters



Grape features



Soil characteristics



Wine-making parameters



Scientific analyses of wine



Wine tasting scores

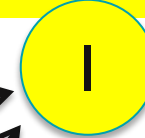
Viticulture continued

- Heterogeneous raw data: 108 parameters



Better characterization of various Lux. terroirs

User Interact. with Visu.



Generate Hyp. From Visu

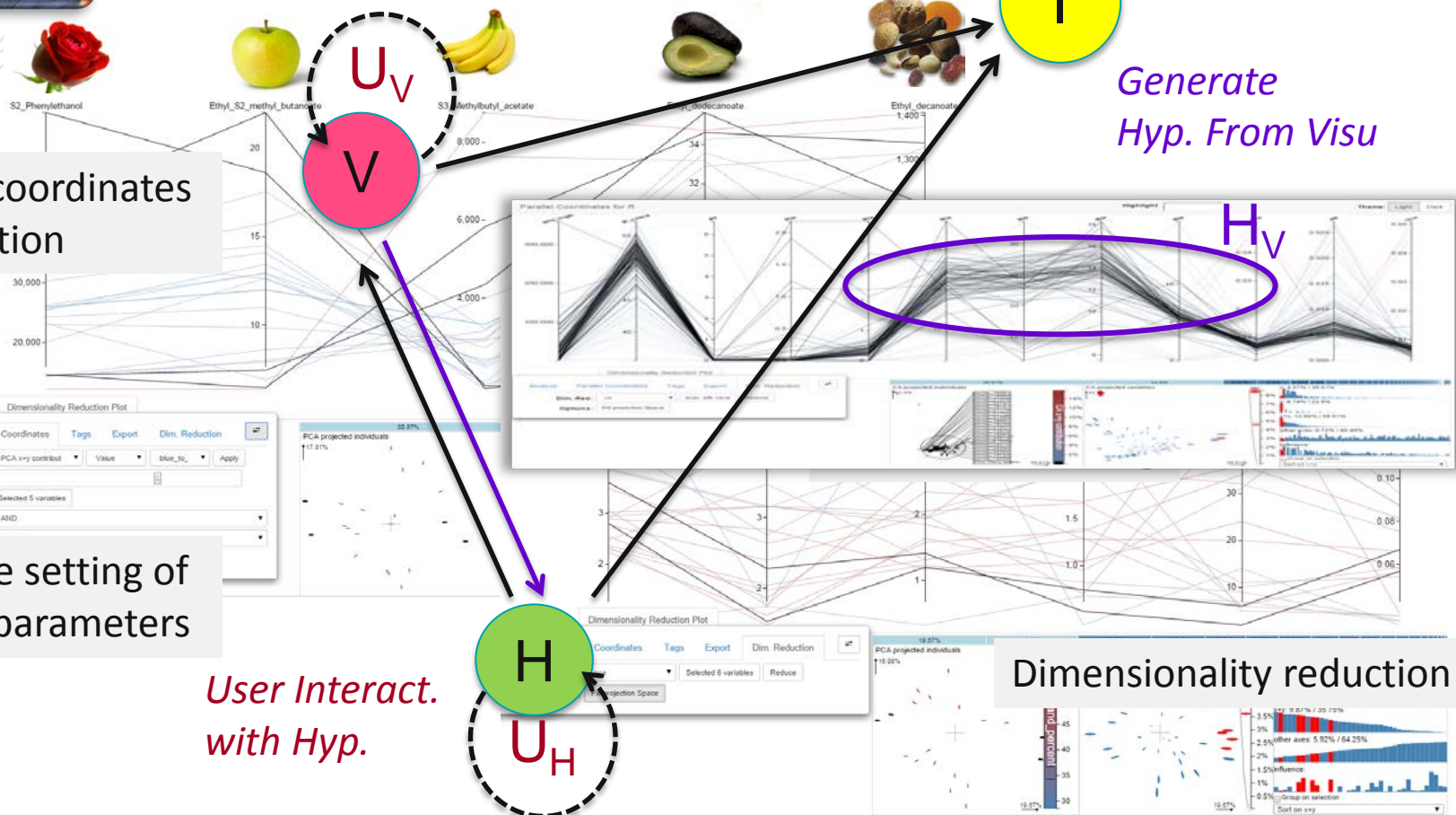


Parallel coordinates visualisation

Interactive setting of analytics parameters

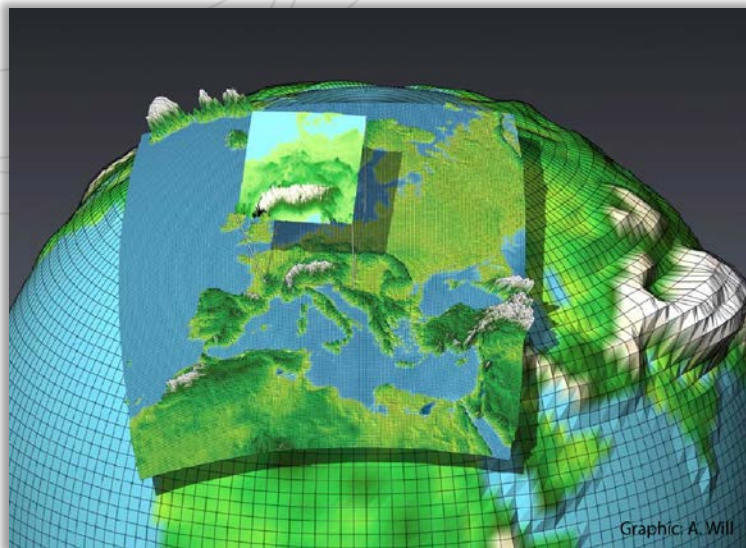
User Interact. with Hyp.

Dimensionality reduction

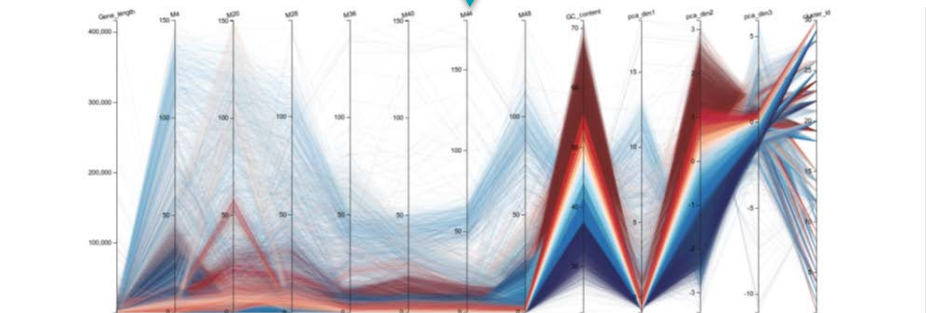


Simulation , exploration (climatology)

Sensitivity analysis of COSMO-CLM model



- Exploring the results of multiple simulation runs
- Find best range for model parameters
- ...



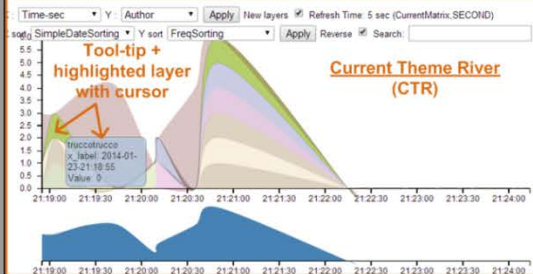
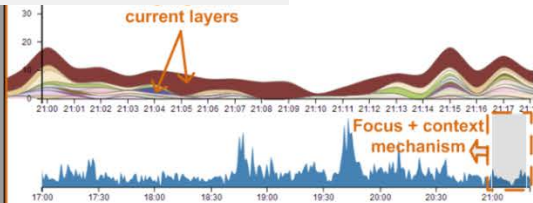
- VISUAL CLASSIFICATION METHOD FOR A LARGE SET OF SENSITIVITY RUNS IN THE FRAMEWORK OF CCLM 5.0 MODEL EVALUATION.
 - By Ferrone and Broeksema. In CLM Community Assembly, 2015

Unstructured data (events)

Crisis management, citizen science



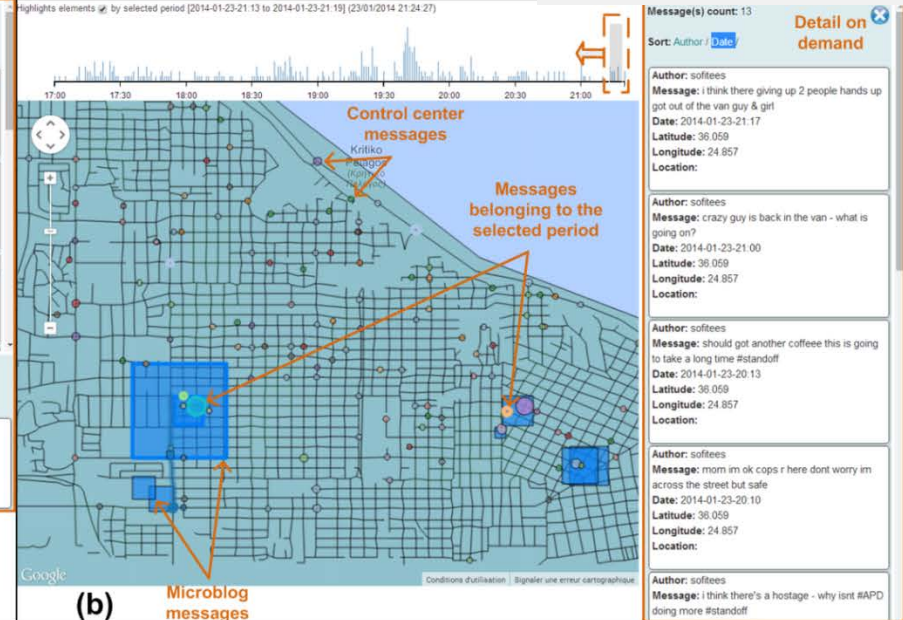
ThemeRiver
visualisation of
topics



Text analytics
in background



Map with events



Raw Data: Tweets, geo-located messages and events

Conclusion & Perspectives

Visual Analytics = Analytics + Visualisation + Human Factors

- Great potential for larger use in Environmental Informatics

Generic method

- Energy, Water, Agriculture, Remote Sensing, Biodiversity...
- Data pre-processing, Simulation, System analysis...

Challenges

- Interdisciplinary teams with expertise in Analytics AND Visualisation AND Human Factors together with Domain experts
- User Acceptance depends on background and mind openness

Contact

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