

Grundlagen methodischen Arbeitens Informationsvisualisierung [WS0708 | 01]

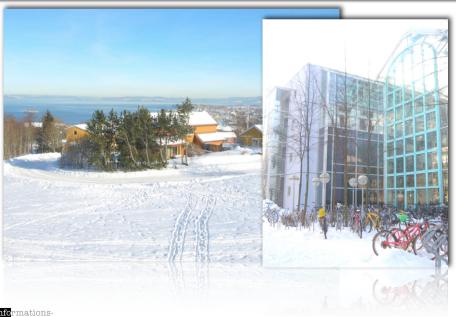
Monika Lanzenberger

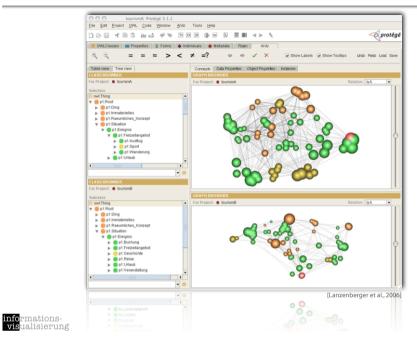
lanzenberger@ifs.tuwien.ac.at

17.10.2007

AlViz & Trondheim

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Topics VO.01

- Motivation Examples
- Definitions and Goals
- Knowledge Crystallization
- Exploration Techniques
- Visual Encoding Techniques
- Summary



Example 2: Chemical Elements

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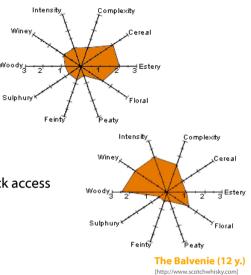
• Taste is very abstract

- 10 basic tastes
- Intensity [0, 3]

Wheel chart

Points - form a polygon

Polygon's properties give quick access to the represented taste

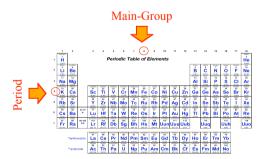




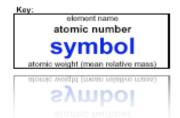
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Example 2: Chemical Elements

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Glenfiddich

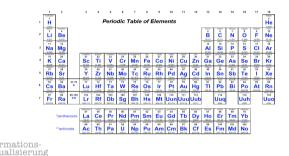


[Pictures: Miksch Slide]

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Periodic Table

- Invented by Dimitri Mendeleyev
- Structured and classified Representation of all chemical elements and their properties
- Predicted the existence of several elements before they were discovered

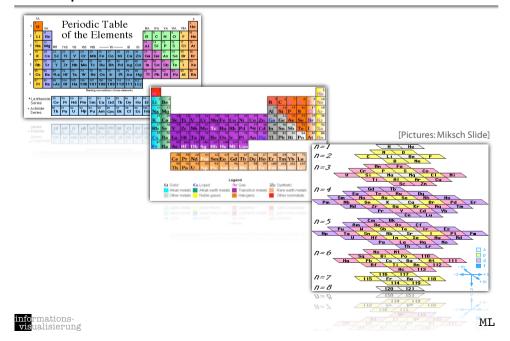




[Pictures: Miksch Slide]

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Example 2: Chemical Elements

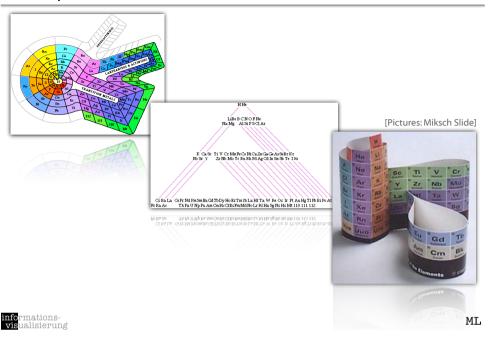




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Example 3: The Challenger Disaster



Example 3: The Challenger Disaster

 January 27, 1986 -

Space shuttle Challenger explodes 72 seconds after launch.

Sealing-rings in the right booster were damaged due to weather conditions.

Reliability-problems of the so called O-rings were known.

The manufacturer of the boosters warned NASA before launch that the expected cold temperatures might be an extra risk.

NASA did not see any correlation between the failing of O-rings and the temperatures.

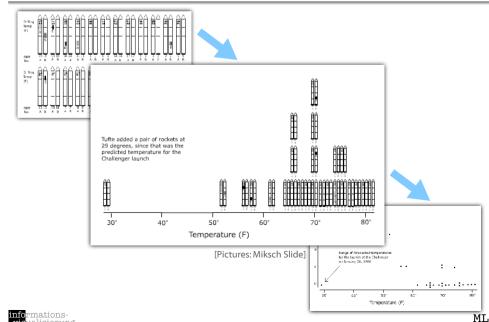




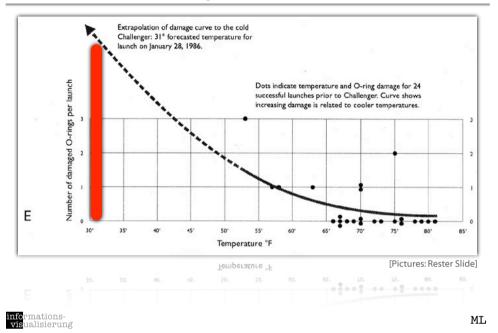
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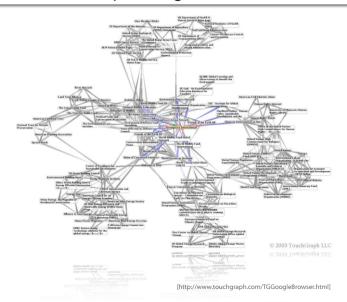
Example 3: The Challenger Disaster

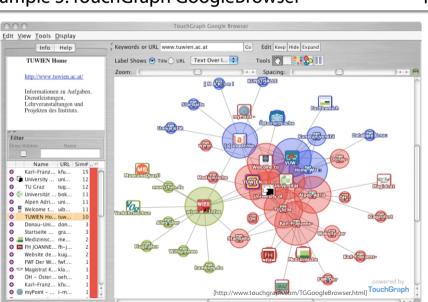


Example 4: inxight TableLens



Example 5:TouchGraph GoogleBrowser (Outdated) 15





WESTPARK San Jose 95124

 $[http://www.inxight.com/demos/tl_housing/tl_housing.html] \\$

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Example 5: TouchGraph GoogleBrowser

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Topics VO.01 17 Definitions ...

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Data

Information

Knowledge

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[Schreiber et al., 2000]

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Information Visualization

InfoVis

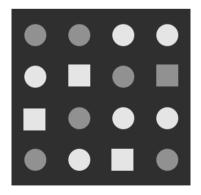
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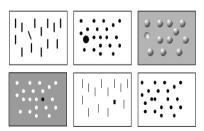
InfoVis is ...

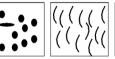
- ... the process of transforming data, information, and knowledge into visual form making use of humans' natural visual capabilities.
- ... the computer-assisted use of visual processing to gain understanding.
- ... providing the user with an overview first and then details on demand (<-> text).

... based on pre-attentive features (< 200ms).

Information Visualization is ...









... based on pre-attentive features (< 200ms).

"input signals to sensory and cognitive processes"

carry out tasks and to create new information"

"the whole body of data and information together with cognitive

machinery that people are able to exploit to decide how to act, to

"data with an associated meaning"

[Card & Mackinlay, 1997, Gershon, Eick, Card, 1998, Ware, 2000]







"It is important to distinguish information visualization from scientific visualization (SciVis).

In scientific visualization what is seen primarily relates to, and represents visually, something physical. Thus, the flow of a water in a pipe or the nature of the weather in a mountainous area [...] are displayed directly superimposed on or at least close to a realistic representation of the physical thing.

By contrast, information visualization tends to deal with abstract quantities such as baseball scores, connections between known criminals, fluctuating exchange rates and electrical voltages.,

[Spence: Information Visualization, 2001]



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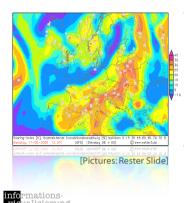
InfoVis: Using space

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- Visualization of abstract data (e.g., financial transactions, insurance risks, etc.) means to find spatial representations (2D, 3D).
- No inherent spatial structure available, so the designer / user needs to decide which dimensions are represented by space: Mapping.

 deals with physical data (e.g., human body, tourist maps, molecules, weather forecast, ...)

- abstract data may be involved
- · spatial reference is determined



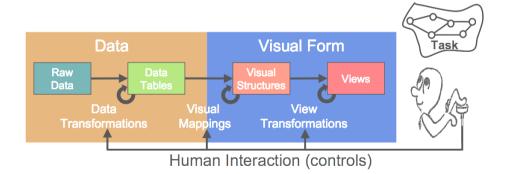




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Mapping

Visualization Reference Model



[Mackinlay, 2000][Card et al.,1999]





- Entities (e.g., people, terms) and relations (e.g., part-of, is-a)
- Both can have sets of attributes (duration, color, time, etc.)
- Types of attributes
 - 1. nominal, ordinal, interval, ratio
 - Category data (nominal), integer data (ordinal), real-number data (interval & ratio)
- High-frequency versus high-structural

[Ware: Information Visualization, 2000]



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Types of InfoVis

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Visualizations are characterized by their purpose for ...

Exploration

the user searches for structures and unknown relations which provide her or him with new insights about the data under investigation.

Analysis

starting with certain hypotheses about the data the user tries to prove them by goal-oriented investigations.

Presentation

static visualization of facts which are fixed a priori.

[Schumann et al., 2000]

Multi-Dimensionality

... contain more than three dimensions and are multi-variate

Multi-Modality

... a combination of data from different sources

Structural Complexity

... ranging from low-structured (simple data structure, but many instances, e.g., flow data, volume data) to high-structured data (complex data structure, but only a few instances, e.g., business data)

Disparity

... contain different types of information in the different dimensions

Largeness

... consist of at least hundreds of thousands of data points

Spatiality

... contain at least one (non-scalar) spatial component and non-spatial data

Time-Dependency

... data is given at several points in time

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InfoVis & Cognition

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Visualization can facilitate cognition by ...

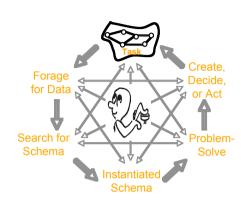
- ... increasing the memory and processing resources available to the user.
- ... reducing the search for information.
- ... using visual representations to enhance the detection of patterns.
- ... enabling perceptual inference operations.
- ... using perceptual attention mechanisms for monitoring.
- ... encoding information in a manipulable medium.

[Card, Mackinlay, Shneiderman: Readings in Information Visualization, 1999]





- Motivation Examples
- Definitions and Goals
- Knowledge Crystallization
- Exploration Techniques
- Visual Encoding Techniques
- Summary



[Card, Mackinlay & Shneiderman, 1999]



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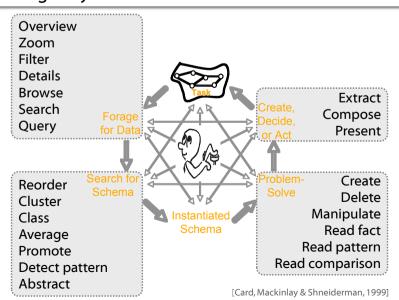
Knowledge Crystallization Sub-tasks

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Topics VO.01

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[Keim, 2001]

human abilities

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Visual Information Seeking Mantra

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"There are many visual design guidelines but the basic principle might be summarized as the Visual Information Seeking Mantra:

Overview first, zoom and filter, then details-on-demand! Overview first, zoom and filter, then details-on-demand!

[Shneiderman: The eyes have it, 1996]

High-level Tasks

1	overview	gain an overview of the entire set of data
2	zoom	adjust the size of items of interest
3	filter	remove uninteresting items
4	details-on-demand	select one or more items and get details
5	relate	identify relationships between items
6	history	keep a history of actions to support undo/redo
7	extract	extract subsets of items for separate analysis

[Shneiderman: The eyes have it, 1996]

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Classification

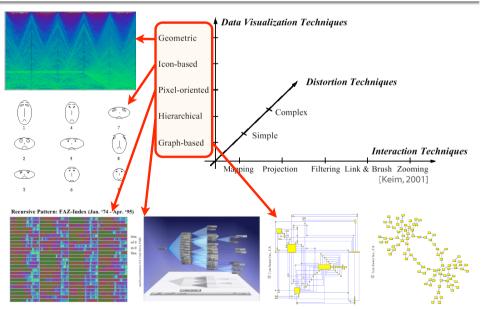




Figure 1. Classification of visual data exploration techniques. Data to Be Visualized One-dimensional Two-dimensional Visualization Technique Multidimensional Stacked Display Dense Pixel Display Text Web Iconic Display Hierarchies Graphs Geometrically Transformed Display Algorithm/software Standard 2D/3D Display Zoom Distortion Interaction and Distortion Technique LinkBrush

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[Keim, 2001]

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Topics VO.01

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Motivation - Examples

- **Definitions and Goals**
- **Knowledge Crystallization**
- **Exploration Techniques**
- Visual Encoding Techniques
- Summary

Coupling views by:

Slaving

movements in one view are automatically propagated in the other views

Linking

connects the data items of one view with the data items of the other views e.g., done by brushing: user selects and highlights items in one view and the corresponding items are highlighted automatically

[Baldonado, 2000]

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Visual Encoding Techniques

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Different ways in encoding information visually:

Space

(See details next slide)

Marks (in space)

Points, lines, areas, volumes

- Connections & enclosures
- Retinal properties

Crispness, shape, resolution, transparency, color, grayscale

- Temporal changes
- Viewpoint transformations

[Card, Mackinlay & Shneiderman, 1999]

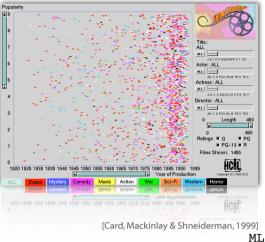




Composition

The orthogonal placement of axes,

creating a 2D metric space





Visual Encoding Techniques

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Composition

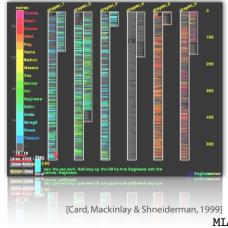
The orthogonal placement of axes, creating a 2D metric space

Alignment

The repetition of an axis at a different position in the space

Folding

The continuation of an axis in an orthogonal direction

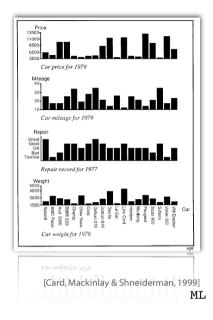


Composition

The orthogonal placement of axes, creating a 2D metric space

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The repetition of an axis at a different position in the space





Visual Encoding Techniques

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• Composition

The orthogonal placement of axes, creating a 2D metric space

Alignment

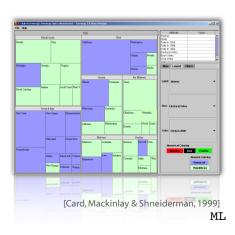
The repetition of an axis at a different position in the space

Folding

The continuation of an axis in an orthogonal direction

Recursion

The repeated subdivision of space







Motivation - Examples

Definitions and Goals

Knowledge Crystallization

• Visual Encoding Techniques

• Exploration Techniques

Composition

The orthogonal placement of axes, creating a 2D metric space

Alignment

The repetition of an axis at a different position in the space

Folding

The continuation of an axis in an orthogonal direction

Recursion

The repeated subdivision of space

Overloading

The reuse of the same space



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Summary: InfoVis ...

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Thanks to ...

Summary

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- is a very complex task.
- can help to get insight into data more quickly.
- is a kind of abstraction.
- requires preparation and sensible handling of the information.
- should make use of the properties of human visual perception.
- requires sensible handling, relative to the task.
- is a big challenge, if you want to do it good.

- ... Silvia Miksch and
- ... Markus Rester

for making nice slides of previous classes available.

