

### **Evaluation & Usability**

# **Content :: Evaluation & Usability**

Information Visualization Evaluation

**Evaluation in Practice** 

in2vis DisCō **Stardinates** 

### Crucial InfoVis Challenges





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# **Content :: Evaluation & Usability**

# Information Visualization Evaluation

# Evaluation in Practice in2vis DisCō

Stardinates

# **Crucial InfoVis Challenges**

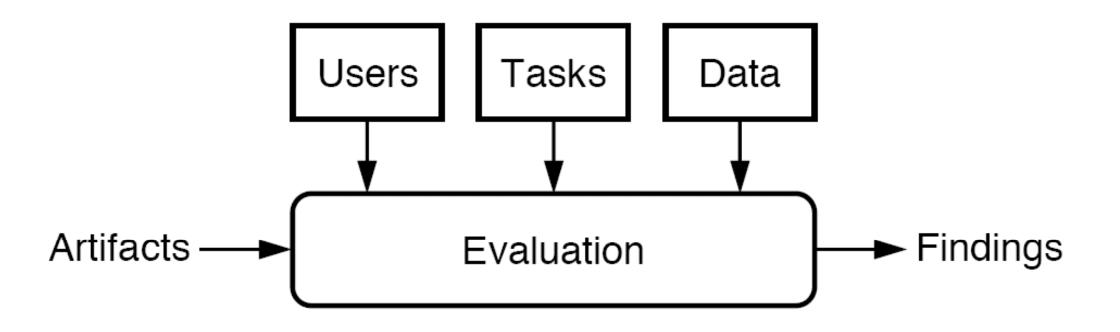








### The Main Ingredients of Evaluation [Keim, et al. 2010 - RoadMap]



For Example, *Artifact* :: scatterplots *Task* :: helpful to find clusters *Data* :: a limited number of real valued attributes *Users* :: training in the proper interpretation







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Can be professional well trained or lay persons Can be proficient with computers or not Can be young or old

### Difficult issues

Expert are well trained and know the tasks but their time is precious and they are scarce resources

Students as found in our labs will not exhibit the same kinds of performance as experts for real tasks Monika Lanzenberger/ Silvia Miksch





### Several levels

Low level: important but not "ecologically valid" and not sufficient

Can be done in clean lab settings



### Several levels

# Low Level Encodings

e.g., grey value vs. size

# **Component Level**

e.g., visualization/interaction technique

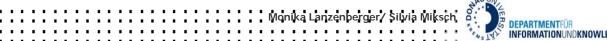
# System Level

e.g., system X vs. system Y

### **Environment Level**

e.g., integration of system X in environment Z





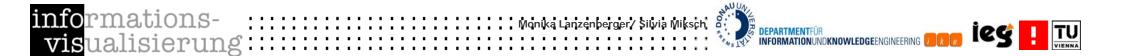


### Several levels

### Low level are homogeneous

Mid level are heterogeneous/multiple

# High level are dynamic, varying, under specified and noisy



# **Evaluation Areas**

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#### Controlled experiments comparing design elements

to compare specific widgets (e.g., alphaslider designs) or mappings of information to graphical display

#### Usability evaluation of a tool

to provide feedback on the problems users encountered with a tool to show how designers can refine the design

#### Controlled experiments comparing two or more tools

common type of study to compare a novel technique with the state of the art

#### Case studies of tools in realistic settings

least common type of studies

advantage

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report on users in their natural environment doing real tasks demonstrating feasibility and in-context usefulness

disadvantage

time consuming to conduct,

and results may not be replicable and generalizable

BELIV'06

Information Visualization

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### Time to completion

**Error** rates

### **GOMS** - Modeling and describing human task performance

GOMS = Goals, Operators, Methods, and Selection Rules Goals represent the goals that a user is trying to accomplish, usually specified in a hierarchical manner. Operators are the set of atomic-level operations with which a user composes a solution to a goal. Methods represent sequences of operators, grouped together to accomplish a single goal. Selection Rules are used to decide which method to use for solving a goal when several are applicable.

### **Benchmarks Repositories**

**Infovis Contest** 

http://www.cs.umd.edu/hcil/InfovisRepository/

**Visual Analytics Benchmark Repository** 

http://hcil.cs.umd.edu/localphp/hcil/vast/archive/d time and errors: BEVOID time and errors: novel evaLuation methods for

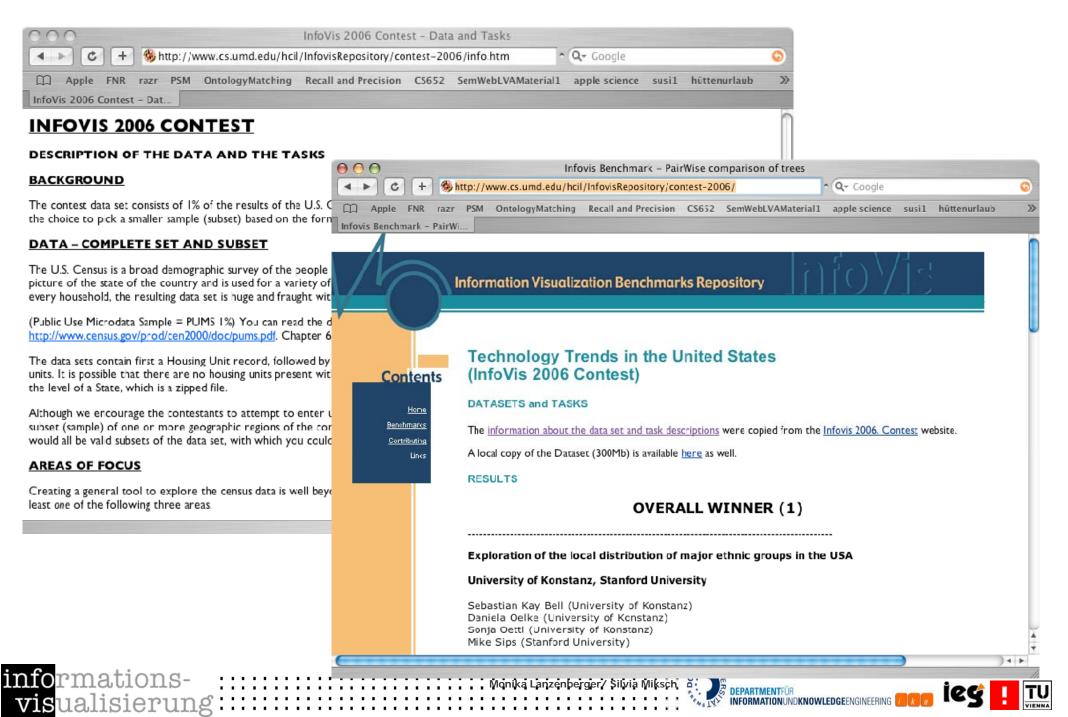
### Insights

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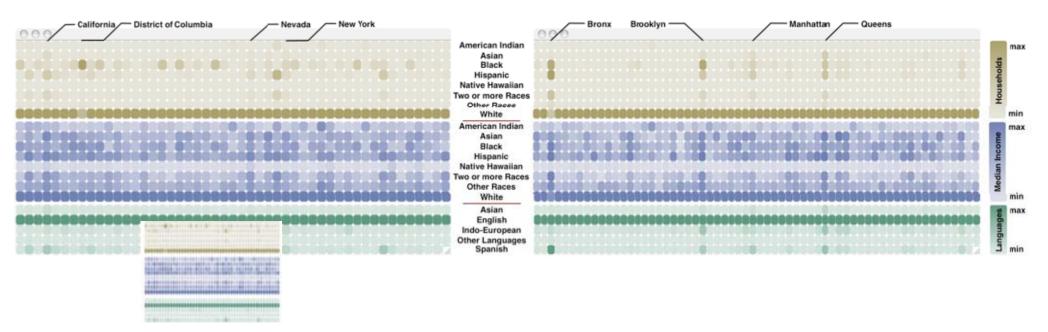
High level cognitive processes:

reasoning, causality, explanation, ... info

# InfoVis Contest 2006



InfoVis Contest 2006 Winners [Belle, et al. 2006] Exploration of the Local Distribution of Major Ethnic Groups in the USA



Visualization of the local distribution of major ethnic groups, their income and the regionally spoken languages. Geographical units are represented by columns, the data for the categories such as household, income, and language data by rows. Left: state level, middle: county level for state New York, bottom left: again state level, but with an iPod-resolution of 220x176 pixel (in comparison to the other screenshots having a resolution of approx. 800x400 pixel). (Column-by-column normalization strategy)

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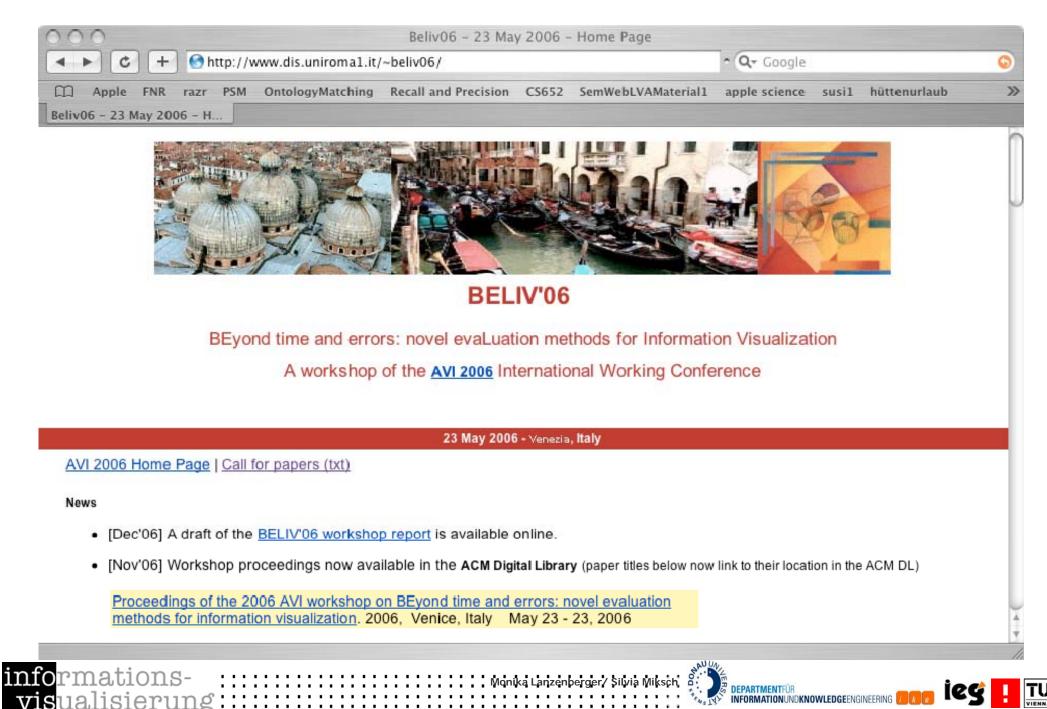




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# **BELIV'06 Workshop**



# **BELIV'06 Workshop Description**

'Controlled experiments remain the workhorse of evaluation but there is a growing sense that information visualization systems need new methods of evaluation, from longitudinal field studies, insight based evaluation and other metrics adapted to the perceptual aspects of visualization as well as the exploratory nature of discovery.'

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[...]

'e.g. new ways of conducting user studies, definition and assessment of infovis effectiveness through the formal characterization of perceptual and cognitive tasks and insights, definition of quality criteria and metrics. Case study and survey papers are also welcomed when clearly presenting general guidelines, practical advices, and lessons learned.' Monika Lanzenberger/ Şilvia Miksch 🏅

# **Evaluation - Specification of Goals**

What to investigate? What are the research questions? How to investigate in order to get answers?

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Domain knowledge helps to identify relevant research questions

# Example: E-learning system

Question 1: Did the participants learn the content? Method: Exam

**Question** 2: Did the participants like to use the system? **Method**: Interviews

**Question** 3: Is the system easy to use? **Methods**: Observation, Software logs



# Select and find participants for the study (subjects)

# Laboratory setting

- + clear conditions allow for good identification of causality
- simulated and restricted setting could yield irrelevant statements

# Field study

- + lifelike and informative
- identification of valid statements is difficult because of the complexity (high number of variables)



# Types of Evaluation (1)

### Formative evaluation

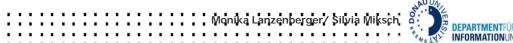
evaluation and development are done in parallel (iterative development process) feedback about usability and utility results cause improvement of the tool

### Summative evaluation

development of the tool is finished assessment of efficacy and features (e.g., comparative evaluation) results may support buyers' decisions

'When the cook tastes the soup, that's formative; when the guests taste the soup, that's summative.'





# Types of Evaluation (2)

# Quick-and-dirty

informal and non-systematic

small number (2 to 10) subjects use the product and

tell what they think about it

usually conducted during product development

low cost

### Scientific evaluation

elaborated process

definition and validation of scientific hypotheses

minimum of 20 subjects for quantitative studies

standardized evaluation methods: quantitative or qualitative

Monika Lanzenberger/ Şilvia Miksch

conducted to investigate core questions of a product or research topic, e.g., command-line interaction versus direct manipulation of objects





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### **Evaluation Methods**

Interviews / focus groups Questionnaire Observation Software logs Thinking Aloud



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# Interviews / Focus Groups

# Interviews

can give a differentiated idea of the usability and efficacy of a tool

subjects cannot always report their behavior, since some cognitive processes are automatic and unconscious

subjects' intentions can provide reasons for measurements and objective data

allows for in-depth analysis

based on guidelines

### Focus groups

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discussions with groups

sometimes a problem to ensure equal participation

group situation could influence topics

based on guidelines for discussion and moderation

# Questionnaire

In contrast to interviews questionnaires allow for studying large groups of people (quantitative evaluation)

Can yield representative data

Should avoid bias

Difficult to prevent misunderstandings because of different interpretations

Simple questions

Closed questions: given answer categories

Open questions: free answers, etc.

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# Observation

Collection of information does not depend on subjects' reports (sometimes subjects can give no information about their activities)

Subjective falsifications are impossible

Problem to understand why persons set certain actions.

No guarantee that the observed person behaves naturally (Hawthorne effect)

Observations can take place in laboratories or in real-world situations

Yields an abundance of data

Difficult to select relevant data

Based on guidelines (what to observe)

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# Software logs

Monitoring tool collects data about computer and user activities, e.g., about number and location of clicks or type of keyboard input

Observes only a limited number of activities

- Delivers high amount of data
- Procedure is not visible for user
- Does not intervene user's activities

Activity sequences yield more information than single step

Analysis of activity sequences is difficult

Software logs do not register the intentions or goals of the users

# **Thinking Aloud**

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Mixes observation and questioning

Subjects are asked to describe their thoughts while using the product

Gives more details than interviews, because information filtering is reduced

Thinking aloud could impede the interaction processes

It is difficult to express the thoughts if interaction with the tool requires attention

Sometimes crucial situations are not reported

Provides with highly relevant and interesting data

Monika Lanzenberger/ Silvia Miksch

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# Usability Evaluation

### Guidelines checklist

Broad principles, empirically-derived results, established conventions **Cognitive walkthrough** 

Based on specific tasks: 'simulation' of a user (model) How difficult is it for the user to identify and operate the interface element most relevant to their current subgoal?

#### Pluralistic walkthrough

Users + developers + HCI experts: Identify primary tasks, step through those tasks

Different Stakeholders adopt different goals / perspectives

=> more usability problems are identified

#### **Consistency inspection**

Quality control technique: consistency in: design, graphics, text, interaction

#### User testing

4-10++ "users", series of tasks, observation, thinking aloud, log files, ...

#### Performance measurement

Efficiency of use, task completion times; useful for comparative studies Monika Lanzenberger/ Silvia Miksch, a

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# **Usability Evaluation: Relevant Links**

www.useit.com www.jnd.org www.nngroup.com www.asktog.com www.usabilityfirst.com Jakob Nielsen Don Norman Nielsen Norman Group Bruce Tognazzini Diamond Bullet Design







# Heuristic Evaluation (1)

A small number of trained evaluators (typically 3 to 5) separately inspect a user interface by applying a set of **'heuristics'**, **broad guidelines** that are generally relevant

Use more evaluators if usability is critical or evaluators aren't domain experts

#### Go through interface at least twice:

- 1. Get a feeling for the flow of the interaction
- 2. Focus on specific interface elements

#### Write reports

Reference rules, describe problem, one report for each problem.

#### **Don't communicate** before all evaluations are completed!

Observer assists evaluators

Use additional usability principles

Provide typical usage scenario for domain-dependent systems

Conduct a debriefing session (provides design advice)

#### Phases: pre-evaluation training / evaluation / debriefing / severity rating

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# **Heuristic Evaluation (2): Rules**

### Visibility of system status

The system should always keep users informed about what is going on, through appropriate feedback within reasonable time.

#### Match between system and the real world

The system should speak the users' language, with words, phrases, and concepts familiar to the user, rather than system-oriented terms. Follow real-world conventions, making information appear in a natural and logical order.

#### **User control and freedom**

Users often chose system functions by mistake and will need a clearly marked "emergency exit" to leave the unwanted state without having to go through an extended dialogue. Support undo and redo.

#### **Consistency and standards**

Users should not have to wonder whether different words, situations, or actions mean the same thing. Follow platform conventions.

### **Error prevention**

Even better than good error messages is a careful design which prevents a problem from occurring in the first place.







# Heuristic Evaluation (3): Rules

### **Recognition rather than recall**

Make objects, actions, and options visible. The user should not have to remember information from one part of the dialogue to another. Instructions for use of the system should be visible or easily retrievable whenever appropriate.

### Flexibility and efficiency of use

Accelerators — unseen by the novice user — may often speed up the interaction for the expert user to such an extent that the system can carter to both inexperienced and experienced users. Allow users to tailor frequent actions.

### Aesthetic and minimalist design

Dialogues should not contain information which is irrelevant od rarely needed. Every extra unit of information in a dialogue competes with the relevant units of information and diminishes their relative visibility.

### Help users recognize, diagnose, and recover from errors

Error messages should be expressed in plain language (no codes), precisely indicate the problem, and constructively suggest a solution.

### Help and documentation

Even though it is better if the system can be used without documentation, it may be necessary to provide help and documentation. Any such information should be easy to search, focused on the user's task, list concrete steps to be carried out, and not be too







# Heuristic Usability Evaluation (1) Forsell & Johansson, 2010

A new set of 10 heuristics out of 63 heuristics (from 6 earlier published heuristic sets)

Especially tailored to the evaluation of common and important usability problems in *Information Visualization techniques* 



# Heuristic Usability Evaluation (1) Forsell & Johansson, 2010

1. **B5**. **Information coding**. Perception of information is directly dependent on the mapping of data elements to visual objects. This should be enhanced by using realistic characteristics/techniques or the use of additional symbols.

2. E7. Minimal actions. Concerns workload with respect to the number of actions necessary to accomplish a goal or a task.

3. E11: Flexibility. Flexibility is reflected in the number of possible ways of achieving a given goal. It refers to the means available to customization in order to take into account working strategies, habits and task requirements.

4. **B7**: **Orientation and help**. Functions like support to control levels of details, redo/undo of actions and representing additional information.

5. **B3: Spatial organization**. Concerns users' orientation in the information space, the distribution of elements in the layout, precision and legibility, efficiency in space usage and distortion of visual elements.



# Heuristic Usability Evaluation (1) Forsell & Johansson, 2010

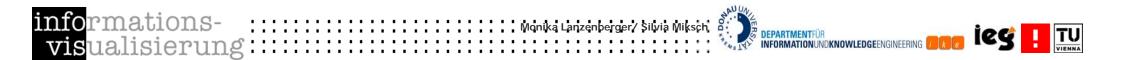
6. E16: Consistency. Refers to the way design choices are maintained in similar contexts, and are different when applied to different contexts.

7. C6: Recognition rather than recall. The user should not have to memorize a lot of information to carry out tasks.

8. E1: Prompting. Refers to all means that help to know all alternatives when several actions are possible depending on the contexts

9. D10: Remove the extraneous. Concerns whether any extra information can be a distraction and take the eye away from seeing the data or making comparisons.

10. **B9: Data set reduction**. Concerns provided features for reducing a data set, their efficiency and ease of use





# **Jean-Daniel Fekete**

# Newer Methodologies

Recently, Ronald Rensink advertised using "Vision Science" methods to evaluate visualizations

Ronald A. Rensink, and Gideon Baldridge, The Perception of Correlation in Scatterplots. Computer Graphics Forum, 29: 1203-1210. 2010.

#### Instead of counting insights, look at decisions on sample datasets (decision theory)

- Expressing insight is a high-level complex process
- Decision is much more direct, does not need verbal expression

#### Use Log/Trace analysis for longitudinal studies

- Instrument programs (at the right levels) and analyze the logs (use visualization to explore)
- Nathalie Henry, Niklas Elmqvist and Jean-Daniel Fekete. A Methodological Note on Setting Up Logging and Replay Mechanisms in InfoVis Systems. In BELIV'08, a workshop at the ACM CHI 2008 conference, April 2008.

#### Use MRI or BCI to study brain response to VA systems

- Detect insight?
- Measure cognitive load and fatigue

#### Use Eye Tracking to study attention and cognitive load

 Chris Weaver. "Look Before You Link: Eye Tracking in Multiple Coordinated View Visualization". BELIV '10, Atlanta, GA, April 2010.

# **Content :: Evaluation & Usability**

# **Information Visualization Evaluation**

# Evaluation in Practice in2vis

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# **Crucial InfoVis Challenges**







# in2vis Project: Visualization



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### Domain: therapy of anorectic young women Support psychotherapists

during therapy a large amount of highly complex data is collected patients and parents have to fill in numerous questionnaires (before, during, and after the therapy)

### Statistical methods are insufficient

small sample size ( $\sim 27$  patients in three years)

high number of variables (~40 different questionnaires with ~40 items. some of them every week, others every 3 months)

time-oriented data

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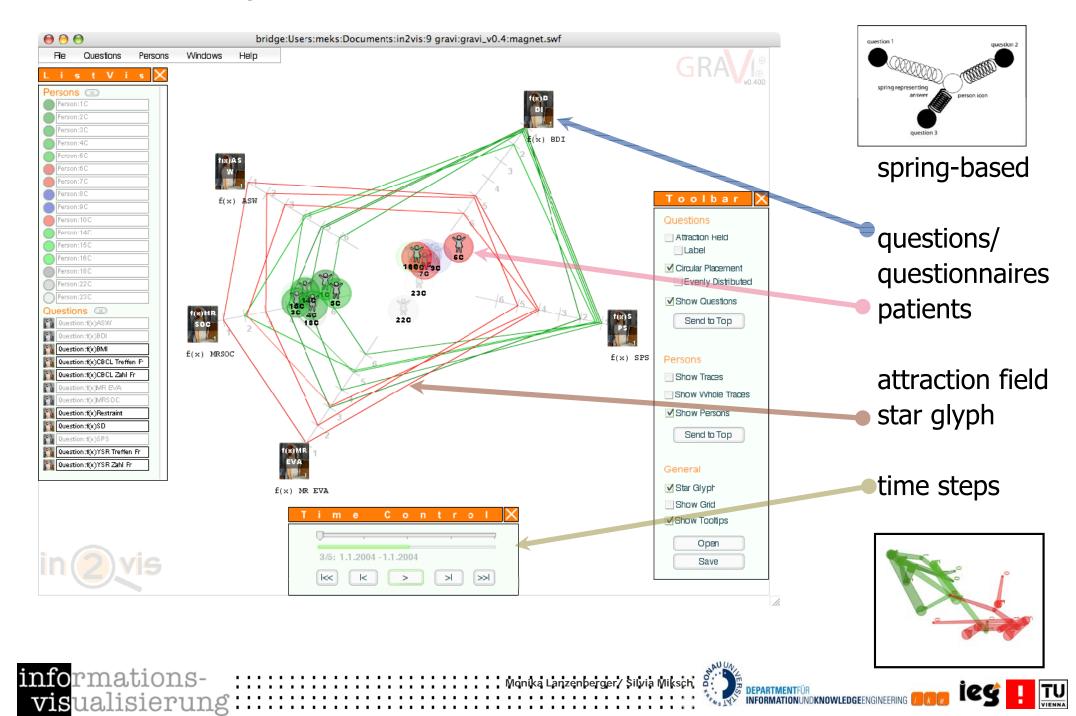
### Aims of the therapists

predict success or failure of the therapy for the individual patients analyze the factors influencing anorexia nervosa

reduce the number of questionnaires the patients have to fill out Monika Lanzenberger/ Silvia Miksch

### in2vis Project: Visualization

#### [in2vis]



### in2vis Project: Evaluation

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Stage	Method	Subjects	Aim	Collected Material	
	usability inspection	1 usability expert	spot most obvious glitches	31 usability problems	
Usability	heuristic evaluation	27 semi-experts	in depth testing	447 reports documenting 576 problems (221 different)	
focus groups	focus groups	in usability	additional usability assessment	no new problems BUT different perspective	
Insight Study	insight reports		patterns of insight & cognitive strategies	876 reports documenting 2166 insights	
(Gravi++, EDA, Machine	log files	33 domain novices	used vis. options & exploration strategies	56055 log file entries	
Learning)	focus groups		relativize findings & aids correct interpretation	transcription of 3x 100min	
Case Chudu	interviews		feasibility & usefulness	transcription of 1x 60min	
Case Study	thinking aloud	2 real users	in real life	notes on 1x 180min	
Transferability	interviews	14 experts of other domains	usefulness in other domains	transcription of 14x 60min	
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# in2vis Project: Usability Evaluation Setting 39

[Rester, et al. 2006]

### Motivation

improve visualization application

preclude mix-up of usability problems with weaknesses of visualization method as such

### Sample

27 students of informatics-related studies semi-experts



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### Methods

informal usability inspection / guideline review heuristic evaluation

Monika Lanzenberger/

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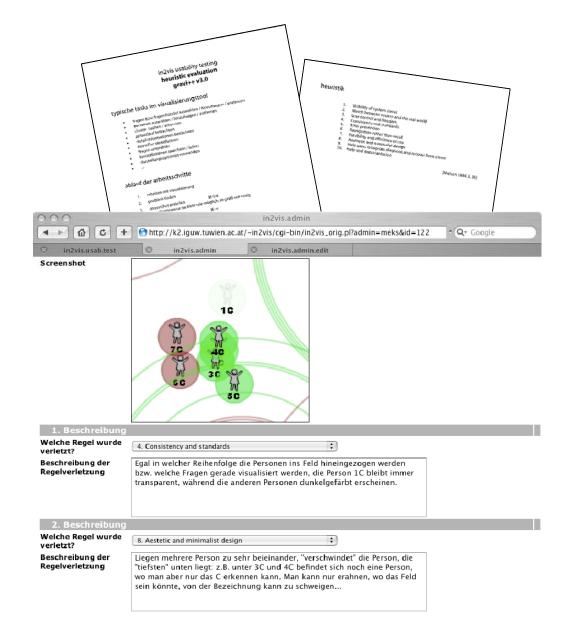




### in2vis Project: Usability Evaluation Setting

[Rester, et al. 2006]

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# Handouts

typical tasks detailed procedure heuristics (outline)

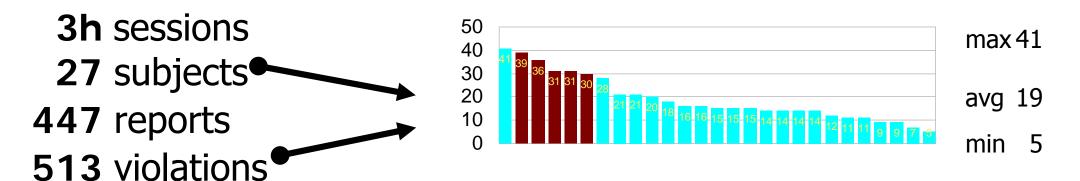
Report system screenshot upload violated rule(s) description(s)

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[Rester, et al. 2006]

# in2vis Project: Usability Evaluation Results 41



Rule	Mentions	Percentage
1. Visibility of system status	63	12.28
2. Match between system and the real world	40	7.80
3. User control and freedom	59	11.50
4. Consistency and standards	105	20.47
5. Error prevention	23	4.48
6. Recognition rather than recall	19	3.70
7. Flexibility and efficiency of use	32	6.24
8. Aesthetic and minimalist design	52	10.14
9. Help users recognize, diagnose, and recover from errors	12	2.34
10. Help and documentation	33	6.43
11. Other Rule	75	14.62
	513	100.00
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## in2vis Project: Usability Evaluation Results 42

Frequency of assigned principles is affected amongst others by:

Rule				
1. Visibility of system status		Monti		
	$\neg$	Mention: 63	reicentage	
3. User control and freedom 4. Consistency and		40	12.28	
<ol> <li>Consistency and standards</li> <li>Error prevention</li> </ol>	$\bot$	59	7.80 11.50	
0. Recognition rather in	+	105	20.47	
	+	23	4.48	
8. Aesthetic and minimalist design 9. Help users room	+	<u>19</u> 32	3.70	
	+	52	6.24	
10. Help and documentation 11. Other Rule		12	10.14	
Nui C		33	2.34 6.43	
		<u>'5</u>	0.43 14.62	
oncoc	51	.3	100.00	

quantity of true existences

comprehension of the principles by subjects

difficulty of tracking down violations of the different principles

domain knowledge needed to find problems of different categories



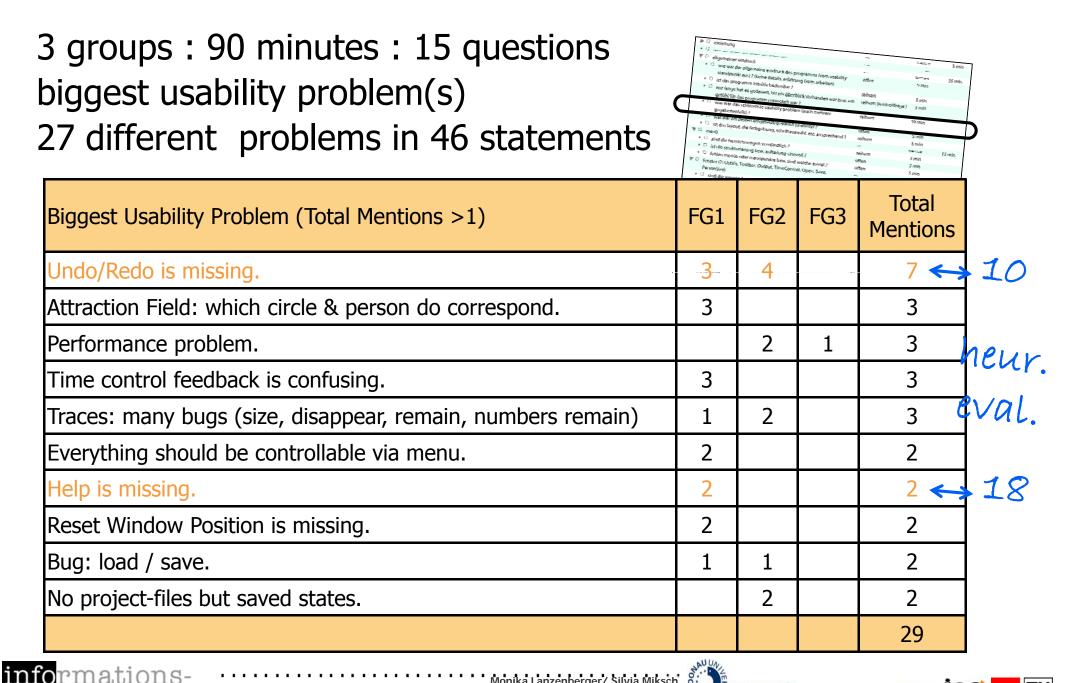




[Rester, et al. 2006]

in2vis Project: Usability Evaluation Results (Focus Groups)

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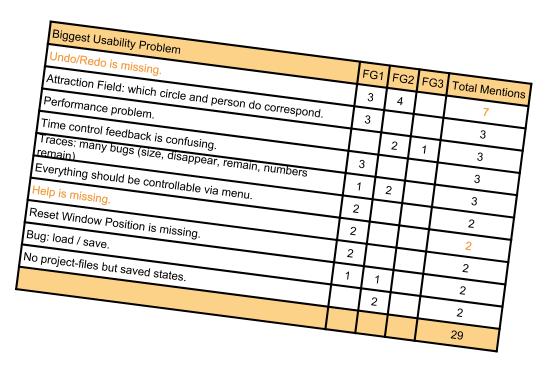


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### in2vis Project: Usability Evaluation Results (Focus Groups)

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A problem's importance may be assessed among others by:



total number of mentions within all groups number of groups in which it is stated distribution of the total number across groups



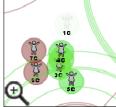




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## in2vis Project: Usability Evaluation Results 45





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#### 4. Consistency and standards

Egal in welcher Reihenfolge die Personen ins Feld hineingezogen werden bzw. welche Fragen gerade visualisiert werden, die Person 1C bleibt immer transparent, während die anderen Personen dunklegefärbt erscheinen.

#### 8. Aestetic and minimalist design

Liegen mehrere Person zu sehr beieinander, "verschwindet" die Person, die "tiefsten" unten liegt: z.B. unter 3C und 4C befindet sich noch eine Person, wo man aber nur das C erkennen kann. Man kann nur erahnen, wo das Feld sein könnte, von der Bezeichnung kann zu schweigen...

Meta		
Location	workspace	=
	persons	;
		;
Classification	persons: ausgrauen unklar	=
	persons: unübersichtliche überdeckungen	•
		;
Delta	0 ;	
Important		
Report abschließen		

# 3-tier location Unique classification(s)

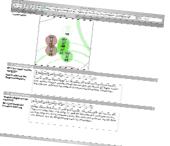
### Some results

221 unique problems 576 documentations (513) top-evaluator(s): 47 (41)easy to spot problems many bugs (20%) feature requests (15%) person-icons (9%) inconsistencies (6%) question-icons (5%) menu (5%)

### in2vis Project: Usability Evaluation - Summary

### Informal usability inspection identifies obvious weaknesses increases quality of heuristic evaluation

### Heuristic evaluation proper method



Focus groups reveal overall view of evaluators of efficiently identify dramatic problems



3 methods give a different perspective on usability issues complement each other to a broader view





[Rester, et al. 2006]

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# in2vis Project: Insight Study

### Tools used by subjects gravi++ interactive infovis eda explorative data analysis

ml machine learning

in	60 min		
	30 min		
	30 min		
i	30 min		
9 subj.	12 subj.	12 subj.	
ml	gravi	eda	60 min
eda	ml	gravi ml	60 min
gravi	eda	ml	60 min

# Comparative study

scenarios (data subset): undirected exploration
concrete tasks (data subset + question):
still argument required

# Goals

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types of insight gained with different tools different insights by varying orders of used tools? patterns of insight & cognitive strategies

Moniką Lanzenberger/ Silvia Miksch

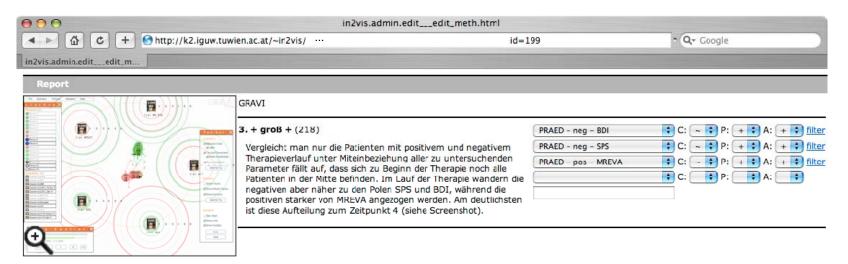
#### 47

### in2vis Project: Report System (1)

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meth_altered]		
	laufn: 5 ID: 203 Matrikelnummer: Datum: 2005-11-25 11:19:31 Status: finished method: GRAVI used material: attributes: 1 insights: missing data - t3 - 8 pat data - pos - t3 - 5 pat data - neg - t3 - 3 pat classified: 000 discuss: 000 ===done MR: es gibt ja noch 3 weitere grüne. => P: nur comment: auf BMI) zu den 50% gerechnet -> würde P: + MR: BMI ka transparent. MR: ich halte die aussage fuer mittelmaessig	ommt nicht vor. die leigen brav am workspace und sin
	confidence: ~ mittel ~ Bei Zeitperiode 3 erkennt man: 50% gibt es Daten, 5 Leute haben die Therapie	
	laufnr: 6 ID: 204 Matrikelnummer: Datum: 2005-11-25 11:19:31 Status: finished method: GRAVI used material: attributes: 1 insights: PRAED - pos - ASW PRAED - pos - BDI PRAED - neg - ASW PRAED - neg - ASW PRAED - neg - BDI classified: 000 discuss: 000 comment: ====done MR: BDI != SPS falsche begruendung ===eof	
	confidence: + groß + Bereits im zweiten Zeitschritt ist ein deutlicher Unterschied zwischen den pos. der Therapie sind die Werte für ASW und BDI für alle Pat. ähnlich): Positiv: hoh geringe Selbsteinschätzung, hohe soziale Unsicherheit	
	laufnr: 7 ID: <b>205</b>	

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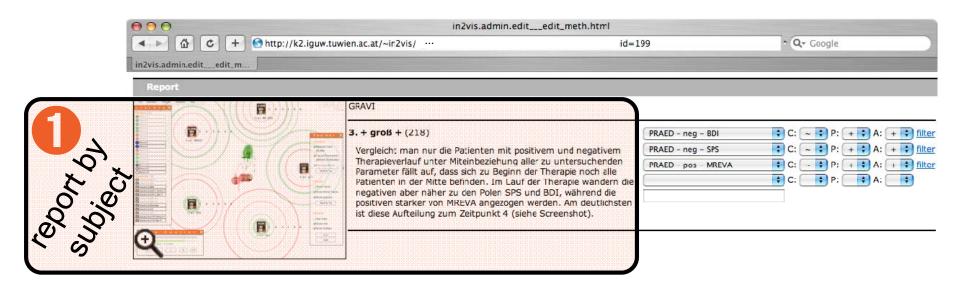
### in2vis Project: Report System (2)



Meta		
Discuss	INT MP EXT	
Classified	✓ MR □ SW	
Comment	===done MR: genuegt das fuer ein + bei argument? (nur hinweis auf t4) SW: ich wuerde es als + bewerten. ===eof	
Report abschließen		abschließen



### in2vis Project: Report System (3)



Meta		
Discuss	INT MP EXT	
Classified	<ul> <li>✓ MR</li> <li>□ SW</li> </ul>	
Comment	===done MR: genuegt das füer ein + bei argument? (nur hinweis auf t4) SW: ich wuerde es als + bewerten. ===eof	
Report abschließen		abschließen

### report generated by subject including

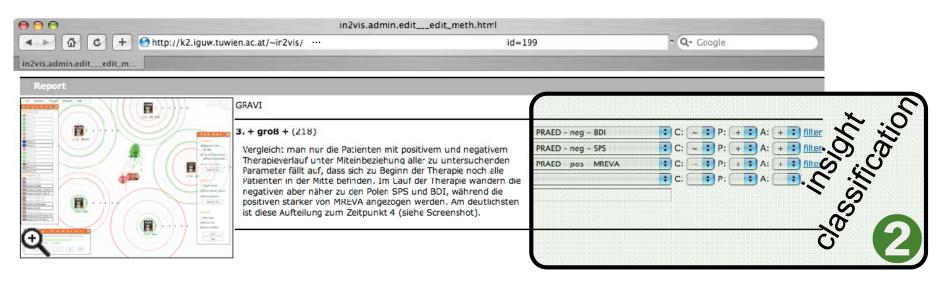
uploaded screenshot confidence rating ( high | mid | low ) insight description







### in2vis Project: Report System (4)



Discuss	🔲 INT	
	MP	
	EXT	
Classified	MR	
	SW SW	
Comment	===done	]
	MR: genuegt das fuer ein + bei argument? (nur hinweis auf t4) SW: ich wuerde es als + bewerten.	
	===eof	
Report abschließen		abschließen
	ication including	

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insight identifier complexity ( complex | regular | trivial )

plausability (high | mid | low)

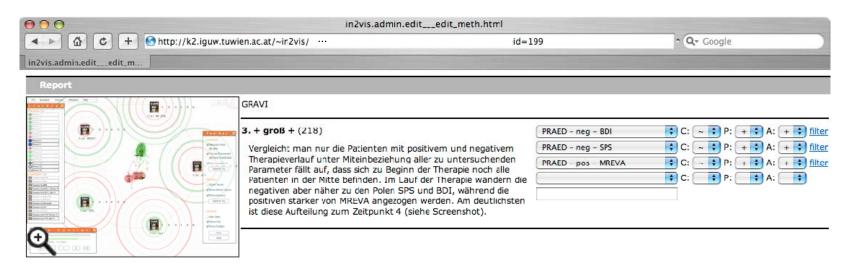
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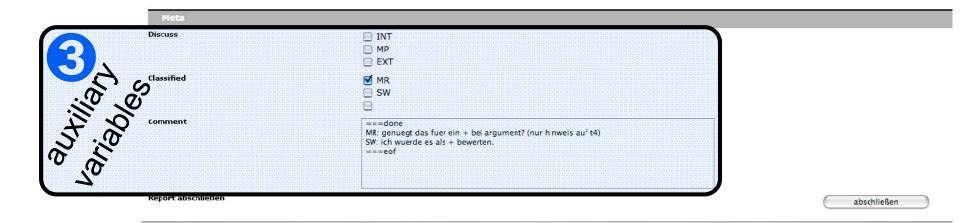
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argument ( correct | missing | wrong )

### in2vis Project: Report System (5)





Rester, et al: 2

### 3 auxiliary variables including

various to-discuss flags (e.g., between investigators, with domain experts) classification status (proofread by a 'second set of eyes') comment/discussion field for investigators







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### in2vis Project: Evaluation Issues

53

### Insight reports

Should long reports be split in basic insights or are they a unique occurrence of a complex insight?

Are they simply a cumulative documentation from a subject who did not adhere to the test procedure of reporting insights immediately after having them?

 $\rightarrow$  for comparability splitting is necessary.

### Log files

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How should one account for the learning curve? Log file chunks between later insights will probably not reflect the explorative interactions leading to an insight.

### **Content :: Evaluation & Usability**

### **Information Visualization Evaluation**

### **Evaluation in Practice**

in2vis DisCō Stardinates

**Crucial InfoVis Challenges** 







### Project DisCō (lat. ich lerne)

### visual DIScovery and COmmunication of complex time patterns in non regularly gathered multigranular and multivariate data

FIT-IT [ Visual Computing

Silvia Miksch, Wolfgang Aigner, Alessio Bertone, Tim Lammarsch, Thomas Turic

> Johannes Gärtner, Dieter Punzengruber, Sabine Wahl

Hanna Risku, Eva Mayr, Michael Smuc











(lat. I learn)

#### Data

time-oriented, irregularly sampled multivariate, multigranular

### Goals

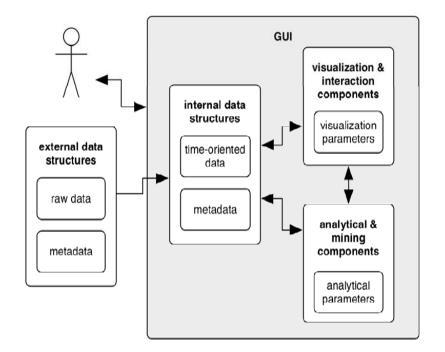
Interactive visualization of data and results with visual parameterization

Analytical methods for analyzing timeoriented data

Ensuring usability and utility of developed methods via User-Centered Design

#### Task

Discovery of complex patterns and relationships



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### **Research and Development Process**

### (1) Task & user analysis

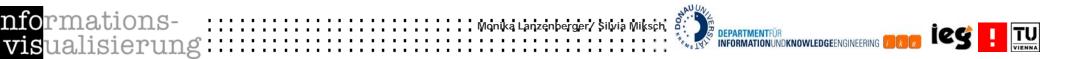
In-depth interviews: users tasks, needs & goals

### (2) Iterative process & user-driven design

Iterative design, Usability-inspection, focusgroups

### (3) Usability testing & data analysis

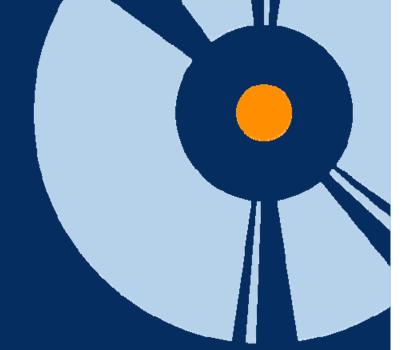
Usability-evaluation





Visual **DIS**covery and **CO**mmunication of complex time patterns

FIT-IT [



### Visualizations at first sight. Do insights require training?

Michael Smuc, Eva Mayr, Tim Lammarsch, Alessio Bertone, Wolfgang Aigner, Hanna Risku and Silvia Miksch



### DisCō: Insight Study

### Insights

### Insight Study: Visualizations at First Sight.

Material: Cycleplot & Multiscale

- Method
- **Insight Counters**
- Insight Visualization

### **Discussion: Do Insights Require Training?**







## Insights

 the generation of new knowledge by individuals out of visualization for data analysis.
 (Low granularity – single observations)



### Insight Study: Visualizations at First Sight

### **Research Questions:**

Can users generate insights without prior knowledge about the visualization?

Can users generate insights without domain knowledge?



### Method

### Mockup-interviews

Think-aloud technique

Instruction:

", Take a look at this visualization and think aloud while exploring it"

### Analysis:

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Transcription of interviews Segmentation Coding of insights



### **Insight Categories**

Integration of Prior Knowledge		"It decreases until 6 in the morning, to a minimum. I assume this is due to [], to my knowledge, change of shift."
Visualization Insights	How-insight	"The more green the less assignments, the more blue the more assignments."
lineignee	Meta-insight	"Okay, first I'm looking at the days, if I can detect any patterns."
	Improvement- insight	"It would be good to be able to filter out one day."
Data Insights	Cycleplot: Cycle	"Starting in the morning it rises to a peak around 10, 11 am. Then it calms down at noon with a second peak around 4, 5 pm. Then it falls down again."
	Cycleplot: Trend	"The first Monday is high, descending on the second, and rising again on the third and forth."
	Multiscale: Overview	"Sundays are rather low, on average."
	Multiscale: Detail	"Especially at noon it's higher than before or after noon. It's always darkest then."

### **Innovations: Highlight 1**

#### Goal

Development of **methods and measures** for the Usability of visualizations and visualization tools

#### Problem

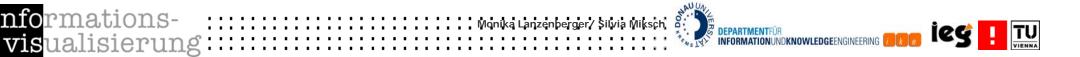
Benefits of classical Usability measures like **completion time** and **errors** are limited, esp. for design of Visual Analytics tools

#### State of the Art

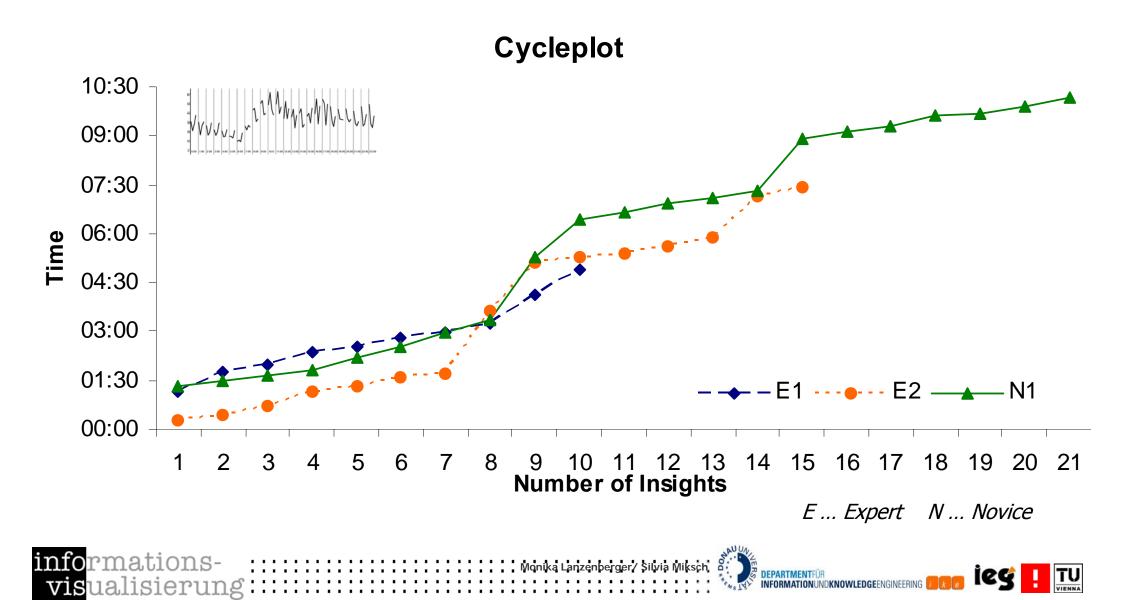
Productivity measures like counting the number of insights [North, 06]

#### **Our Solution**

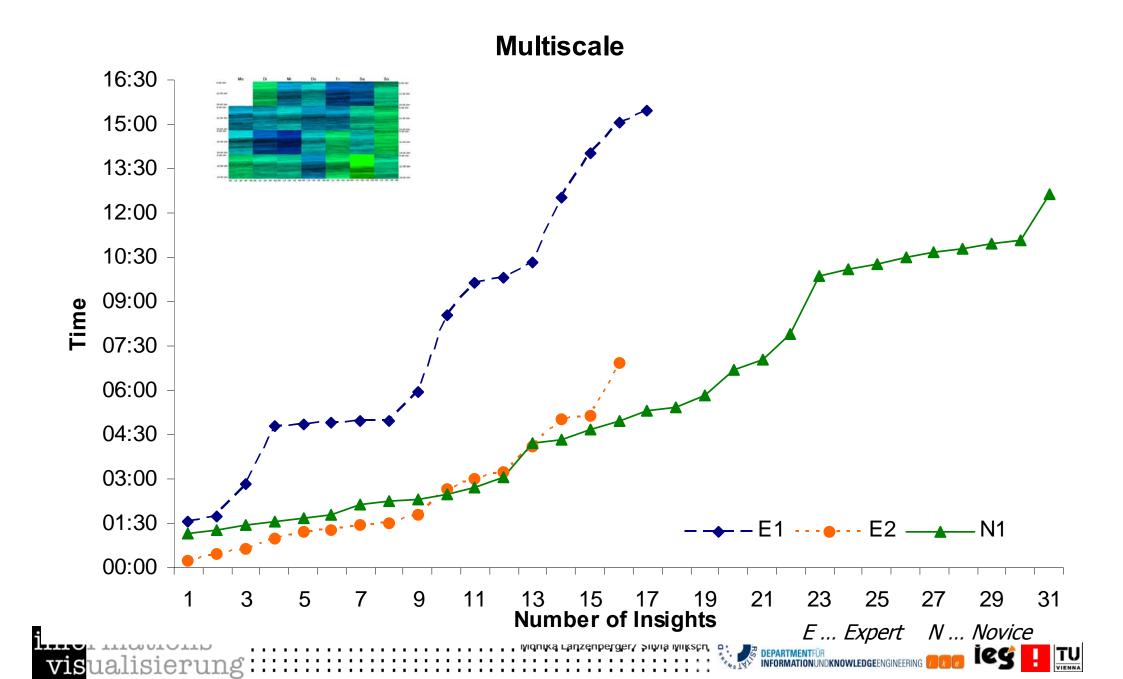
Development of the **Relational Insight Organizer** (RIO) optimized for iterative design [Smuc et al., 2008]



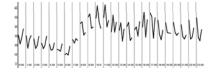
### **Insight Counters**

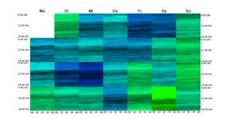


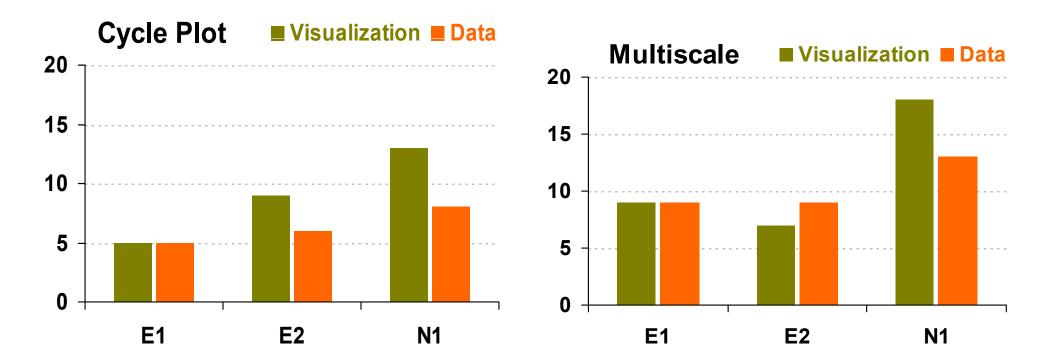
### **Insight Counters**



### **Insight Counters**







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E ... Expert N ... Novice

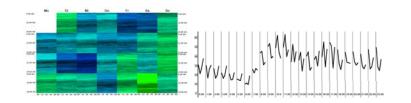
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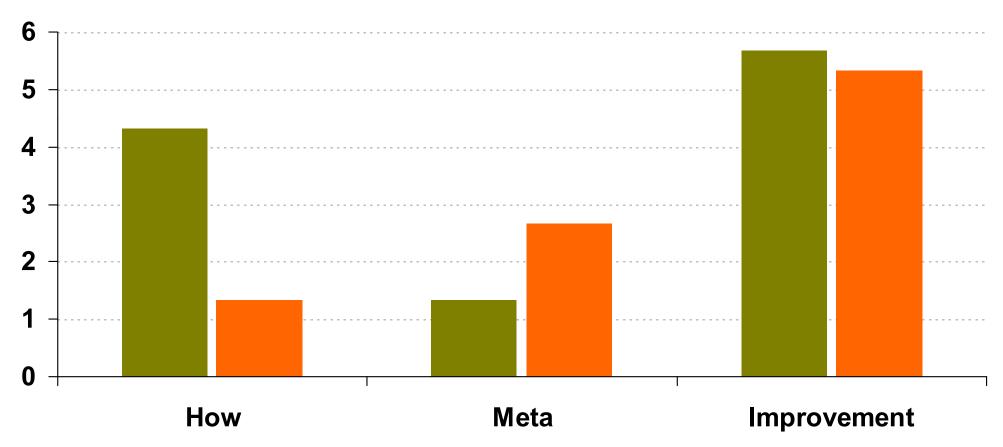


### **Insight Counters**



Multiscale Cycleplot

#### **Visualisation Insights**

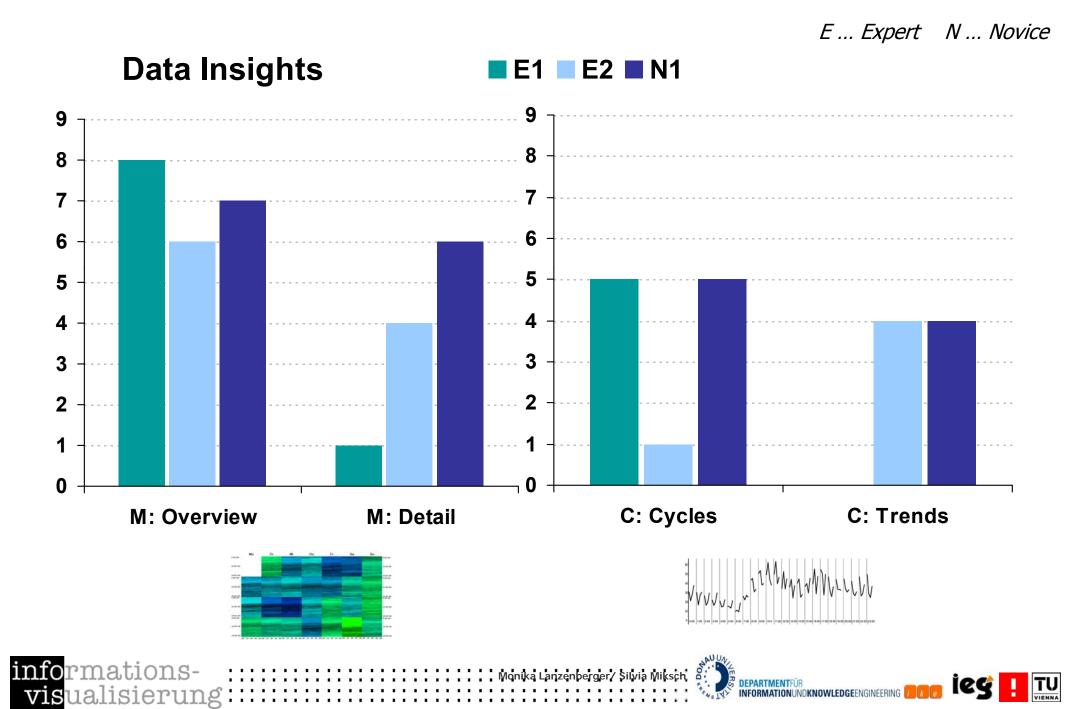




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### **Insight Counters**

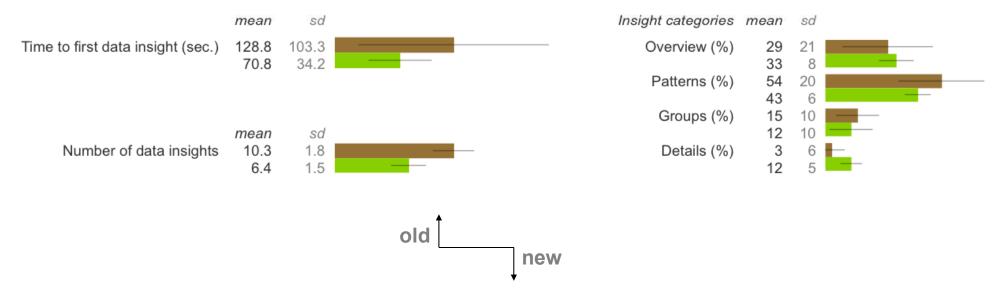


#### Multi Scale Plot vs. Cycle Plot

#### [Smuc et al., 2008]

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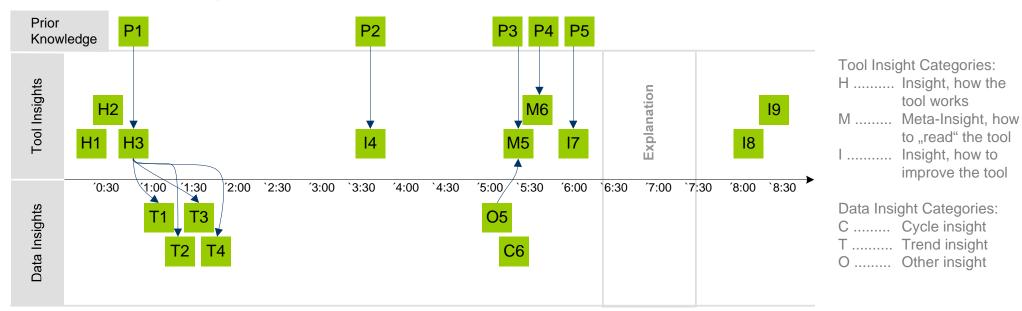
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**RIO of user 3 for Cycle Plot** 

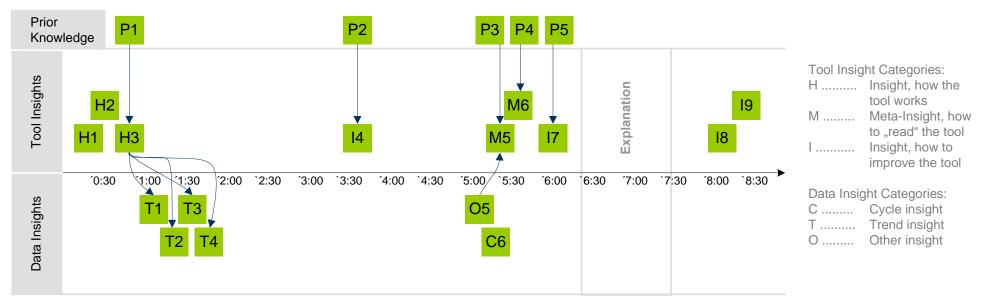
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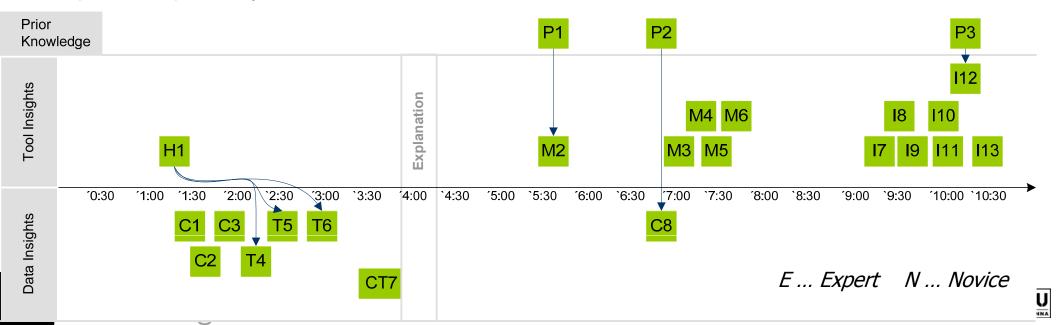


### **Insight Visualization**

#### E2: Uptake Graph for Cycle Plot



#### N1: Uptake Graph for Cycle Plot



### Discussion

### Do insights require training?

Participants were able to generate insights from the start

Domain knowledge was not necessary for insights

Insights into the visualization were needed prior to data insights, but no full understanding

### Is expert knowledge beneficial?

Not necessarily

Prior knowledge was used to interpret data

Experts' existing cognitive scripts maybe hindered more flexible analysis







# **Discussion of Methodology**

Similar insights by expert and novice users

Mockup tests did generate complex data insights

Insight counters provide limited findings for iterative design, rather qualitative analysis of insights is needed

Small sample can provide useful ideas for improvements

#### Limitations

Open task Sample size







# **Content :: Evaluation & Usability**

## **Information Visualization Evaluation**

## **Evaluation in Practice**

in2vis DisCō Stardinates

### Crucial InfoVis Challenges

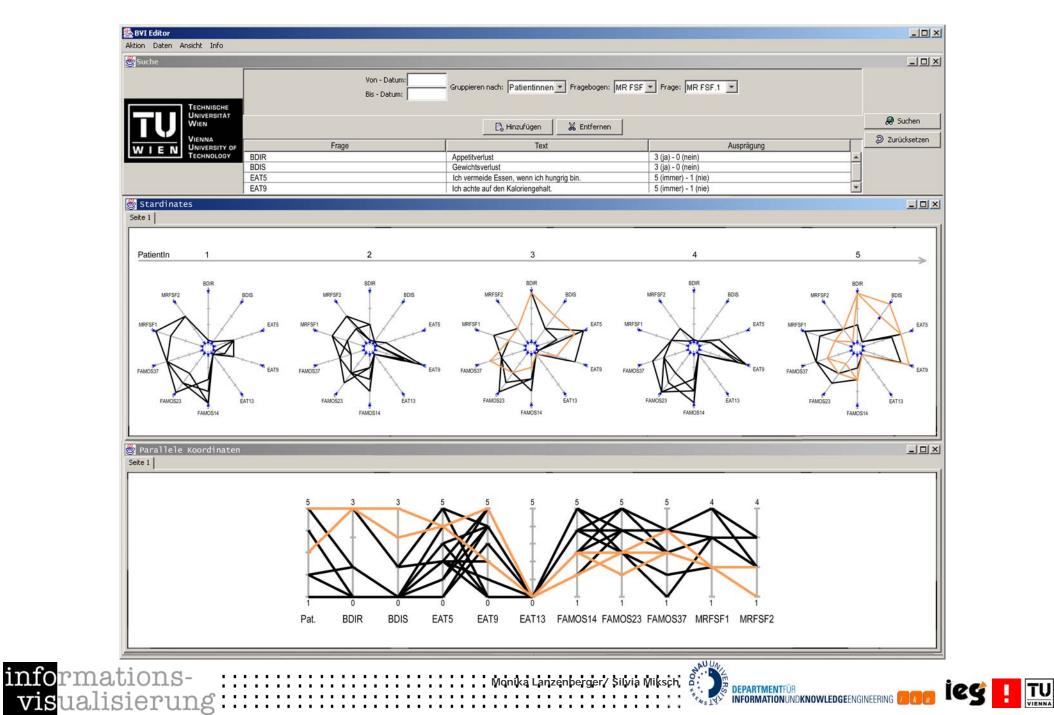






[Lanzenberger 2003]

# LinkStar



[Gärtner, et al. 2002, Lanzenberger 2003]

## **Evaluation of the Interactive Stardinates** 76

# ViCo - Metric to measure the complexity of visualization

Analyze the tasks of the users Define basic operations (e.g., Read, Compare, Highlight) Develop an algorithm

# Compare Parallel Coordinates and the Stardinates by calculating the complexity of their algorithms





#### **Complexity of Interactive Stardinates**

	Task	Code
	Task 1:	Read Shape & Decide(Op1)
	The Algorithm	/*One may be able to recognize the relation if it is strong
		enough and the changes over time occur in a homogeneous way */
		IF no clear Relation THEN
		FOR MANY Data Bundles (a*#P)
		Read Shape of Data Bundle & Decide(Op2)
		IF still no clear Relation THEN
		FOR MANY Lines (a*#T)
		Highlight(Op4)
		Read Area Shape (Op5)
		Compare Area (Op7)
		IF still no clear Relation THEN
		FOR EACH Data Point (#P*#T*#B)
		Read Data Point (Op6)
		Compare Data Point (Op8)
	Complexity for Task 1	Best Case:
		Op1
		Middle Cases:
		Op1 + a*#P*(Op2)
		or: Op1 + a*#P*(Op2)
		+ a*#T(Op4+Op5+Op7)
		Worst Case:
		Op1+ a*#P*(Op2) + a*#T(Op4+Op5+Op7)
		+#P*#T*#B (Op6+Op8)
informat	ions	Op1+ a*#P*(Op2) + a*#T(Op4+Op5+Op7) +#P*#T*#B (Op6+Op8)

[Lanzenberger 2003]

#### **Complexity of Parallel Coordinates**

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Task 1:	FOR each Patient (#P)
The Algorithm	Highlight (Op4) /*Select every patient*/
	/* Read Shape of one patient's data */
	Read Shape & Decide(Op1)
	IF no clear Relation THEN
	FOR MANY Data Bundles (a*#P)
	Highlight (Op4) /*Select every patient*/
	Read Line (b*Op5)
	Compare Line (b*Op7)
	IF still no clear Relation THEN
	FOR MANY Lines (a*#T*#P)
	Highlight (Op4)
	Read Line Shape (b*Op5)
	Compare Line (b*Op7)
	IF still no clear Relation THEN
	FOR EACH Data Point (#P*#T*#B)
	Highlight (Op4)
	Read Data Point (Op6)
	Compare Data Point (Op8)
Complexity for Task	1 Best Case:
	#P*(0p4+0p1)
	Middle Cases:
	#P*(Op4+Op1) + a*#P*(Op4+b*Op5+b*Op7)
	or: $#P*(Op4+Op1) + a*#P*(Op4+b*Op5+b*Op7)$
	+ a*#T*#P*(Op4+b*Op5+b*Op7)
	Worst Case:
	#P*(Op4+Op1) + a*#P*(Op4+b*Op5+b*Op7)
	+ a*#T*#P*(0p4+b*0p5+b*0p7)
rmations-	+ #P*#T*#B* (Op4+Op6+Op8)



# **Concept Testing**

Comparative study (Controlled experiment) with 22 participants (35 participants for each visualization method), 2 examples

#### Research questions:

Are the users able to find information at the first glance? Are the users able to find the crucial information? Which visualization supports the creation of hypotheses?

# **Evaluation:**

Time measurements, questionnaires Classification of strategies (categories) Expert defined 'Key Statements'

Age	# of Sub.
- 20	1
21 - 25	6
26 - 30	6
31 - 35	4
36 - 40	1
41 - 45	3
45 -	1
Total	22





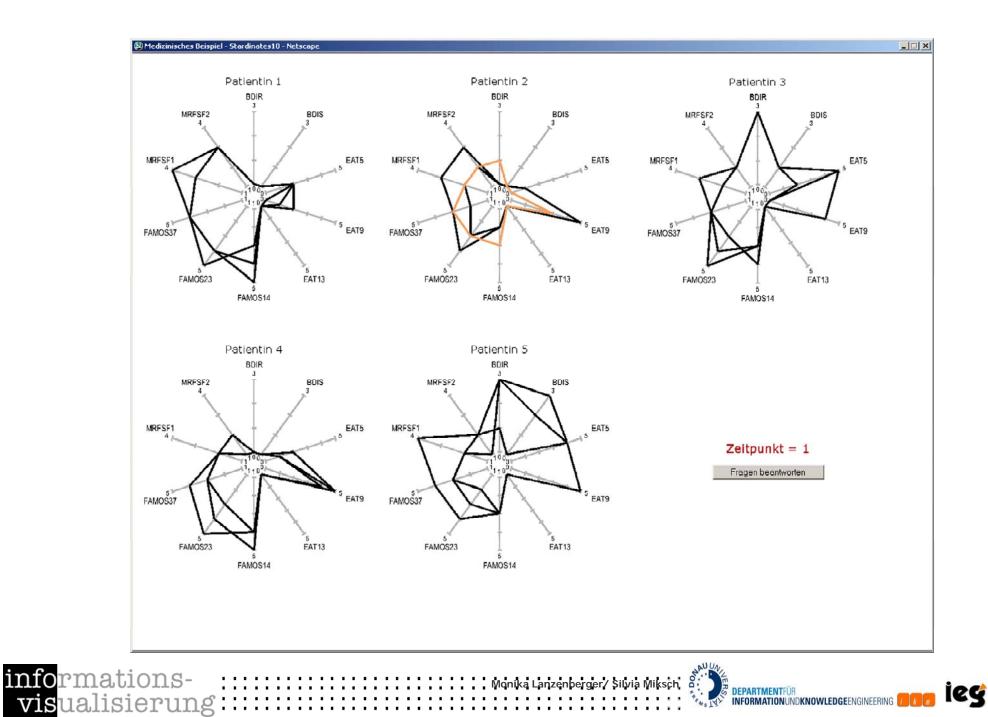


#### **Visualization Method: Parallel Coordinates**





#### **Visualization Method: Stardinates**



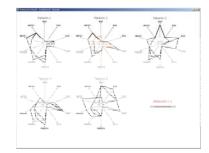


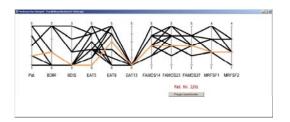
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#### **Evaluation Results: Time Measurement**







	Stardinates		Parallel Coordinates			
Task	Average	Minimum	Maximum	Average	Minimum	Maximum
Viewing 1.Example	2:28	0:10	7:56	4:17	0:11	11:51
1. Answering	9:29	2:14	18:32	9:09	2:43	22:16
Viewing 1.Example (Correct answer)	2:31	0:10	7:56	3:37	0:11	9:58
1. Answering (Correct answer)	8:57	2:14	18:25	11:32	4:24	22:16
Viewing 1.Example (Incorrect answer)	2:09	0:38	5:01	5:28	0:24	11:51
1. Answering (Incorrect answer)	12:50	8:13	18:32	4:49	2:53	8:30
Total time 1. Example	11:57	5:57	20:31	13:26	3:24	26:40
Viewing 2.Example	4:58	0:14	16:56	3:57	0:08	16:43
2. Answering	20:41	6:47	46:02	16:08	3:03	41:34
Total time 2. Example	25:39	7:46	53:50	20:05	4:43	48:33
Total time (1. + 2. Example)	37:36	16:37	1:14:21	33:31	12:42	1:04:11

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# **Evaluation** Example 1 - Aircraft Collision

#### Questions:

Ist deiner Meinung nach eine Kollision aufgetreten?

Wenn ja, welche Flugzeuge waren beteiligt?

Bei welcher Graphik (welchen Graphiken) konntest du etwas ablesen? Wenn ja, was hast du dort abgelesen?

Welche Probleme / Schwierigkeiten

hattest du bei der Interpretation?

#### **Results:**

Parallel Coordinates:

63.6% (14 subjects) **correct** answer,

22.7% (5 subjects) incorrect answer, 13.6% (3 subject) no answer.

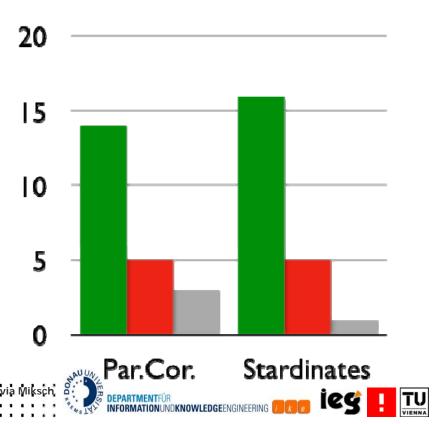
Stardinates:

72.7% (16 subjects) correct answer, 22.7% (5 subjects) incorrect answer, 4.5% (1 subject) no answer.

Two strategies with the Stardinates:

Compare triangles (shapes)

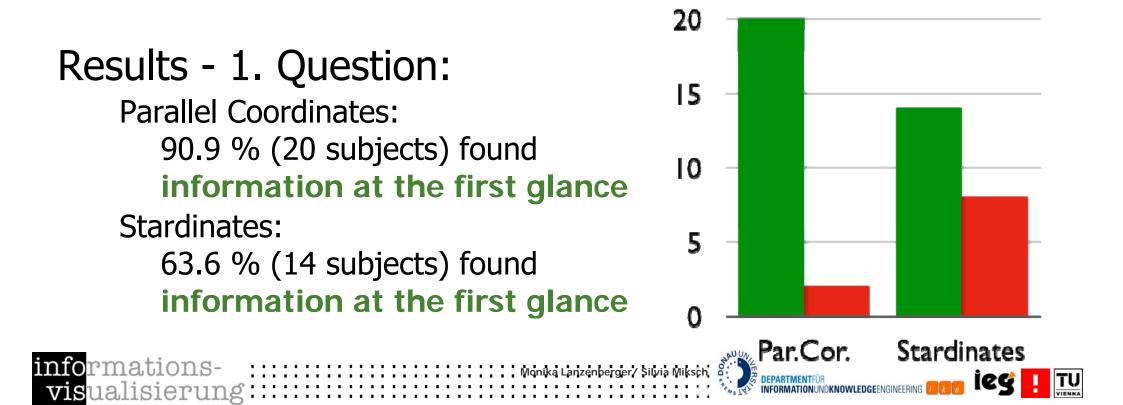
Read exact values info visualisieru



## Evaluation Example 2 – Psychotherapeut. Data 84

## Questions:

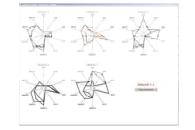
Gibt es Aussagen, die auf den ersten Blick auffallen?Bei welcher Graphik (welchen Graphiken) konntest du etwas ablesen? Wenn ja, was hast du dort abgelesen?Welche Probleme / Schwierigkeiten hattest du bei der Interpretation?



#### **Evaluation Results: Key Statements**

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	Stardinates		Parallel Coordinates	
Key Statement	# of Sub.	% of Sub.	# of Sub.	% of Sub.
Patients do not feel sick after eating.	12	54.55%	16	72.72%
Pat. 1: good starting basis.	15	68.18%	5	22.73%
Pat. 2: unstable.	6	27.27%	2	9.09%
Pat. 3: contradicting answers.	3	13.64%	3	13.64%
Pat. 4: positive progress in therapy. Cares	3	13.64%	0	0%
more about herself.				
Pat. 5: significantly positive progress in	12	54.55%	3	13.64%
therapy between second and third time point.				

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### **Evaluation Results: Key Statements**

### Are the users able to find the crucial information?

Although unfamiliar with psychotherapeutic data, users were able to find crucial insights.

# Statistical analysis:

Stardinates were significantly better for finding crucial information (represented by the key statements). Mean number of key statements:

- 2.32 with the Stardinates,
- 1.32 with the Parallel Coordinates.

(*t*=2.687, *df*=21, level of significance: 5%).

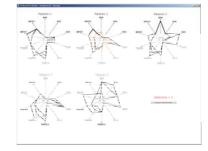
Monika Lanzenberger/ Şilvia Miksch, 🕈

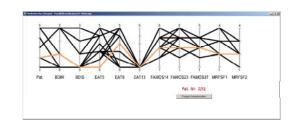
Parallel Coordinates showed a high result in the first category, which is based on one dimension (EAT13) only, but did not perform significantly better



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### **Evaluation Results: Classification of Strategies**





	Stard	linates	Parallel Coordinates		
Category	# of Sub.	% of Sub.	# of Sub.	% of Sub.	
1: Comparing Patients	15	68.18%	10	45.45%	
2: Overview	15	68.18%	5	22.73%	
3: Changes over Time	15	68.18%	10	45.45%	
<ol><li>Examining Single Axes</li></ol>	14	63.64%	19	86.36%	
5: General Conclusions	5	22.73%	1	4.55%	
6: Causal dependency	8	36.37%	1	4.55%	
7: None	0	0%	1	4.55%	

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#### **Evaluation Results: Classification of Strategies**

# Which visualization supports the creation of hypotheses?

Subjects produced significantly more statements with the Stardinates than with the Parallel Coordinates.

They did not need more time when using the Stardinates.

#### Statistical Analysis:

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Mean number of statements 3.27 with the Stardinates and 2.14 with the Parallel Coordinates (*t*=3.504, *df*=21, level of significance: 5%)

## **Content :: Evaluation & Usability**

**Information Visualization Evaluation** 

Evaluation in Practice

in2vis DisCō Stardinates

### **Crucial InfoVis Challenges**







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# **Crucial InfoVis Challenges: Top 10 Problems**

Usability Understanding perceptual-cognitive tasks Prior knowledge Education and training	user- centered perspective
Quality measures Scalability Aesthetics	technical challenges
Paradigm shift from structures to dynamics Causality, visual inference, and predictions Knowledge domain visualization	disciplinary challenges

informationsvisualisierung



# Top 10 Problems: Usability

Relevant for researchers and developers

compare Spotfire (http://www.spotfire.com) and Inspire (http://in-spire.pnl.gov) usercentered perspective

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InfoVis is growing much faster than its usability research

Lack of low-cost or open source InfoVis tools

Usability studies need to address critical details specific to InfoVis

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e.g., recognition of the intended patterns or interaction with possible cognitive paths in a network visualization



# Top 10 Problems: Perceptual-cognitive tasks<sup>[Chen 2005]</sup>

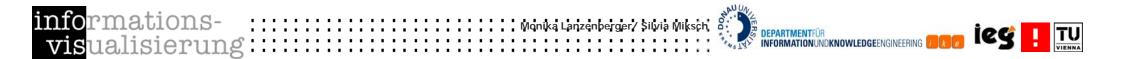
Evaluation of the usefulness of InfoVis components is done:



Identifying & decoding visualized objects, preattentive perception

#### But evaluation of high-level user tasks is needed:

Browsing, searching, recognition of clusters, identification of trends, discovery of previously unknown connections, insightful discovery



# Top 10 Problems: Prior knowledge



# Two types of prior knowledge:

the **knowledge of how to operate the device**, such as a telescope, a microscope, or, in our case, an InfoVis system, and the **domain knowledge** of how to interpret the content

Good usability and utility can reduce the dependence on the first type of prior knowledge

Distinguish perception, cognition and learning







# Top 10 Problems: Education and training



[Chen 2005]

Learn and share various principles and skills of visual communication and semiotics

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Language of InfoVis must become comprehensible

Potential beneficiaries outside the immediate field of InfoVis to see the value and how it might contribute in practice

# **Top 10 Problems: Quality measures**

Quantifiable measures of quality, benchmarks are missing



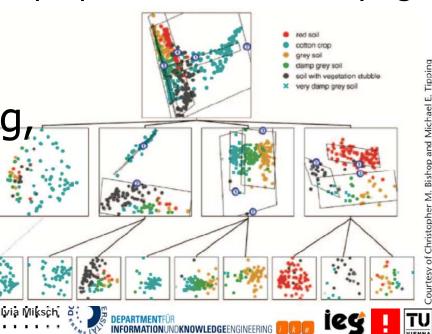
Simplifies development and evaluation of algorithms

Answer key questions such as:

- To what extent does an InfoVis design represent the underlying data faithfully and efficiently?
- To what extent does it preserve intrinsic properties of the underlying phenomenon?

Integrating machine learning for topic detection, trend tracking adaptive information filtering, and detecting concept drifts in streaming data

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# Top 10 Problems: Scalability (1)



[Chen 2005]

Long-lasting challenge for InfoVis

Unlike to scientific visualization, supercomputers have not been the primary source of data suppliers

Parallel computing and other high-performance computing techniques are not used

Visualization of data streams and the urgency to understand its contents







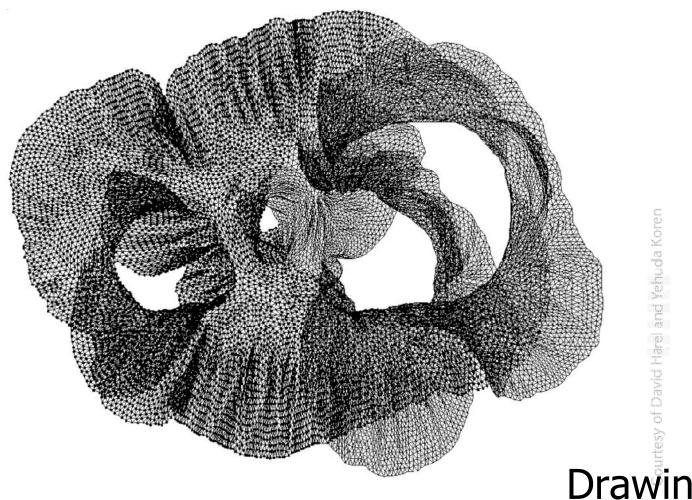
#### Top 10 Problems: Scalability (2)

[Chen 2005]

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# Drawing a 15,606-vertex and 45,878-edge graph within a matter of seconds Interaction?

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[Chen 2005], Graph: [Rester and Pohl 2005]

# **Top 10 Problems: Aesthetics**

Insights, not just pretty pictures

Goal is to enhance utility

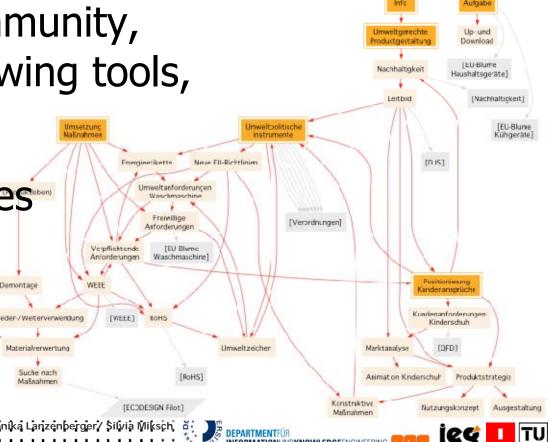
Understand how insights and aesthetics interact

Active graph-drawing community, e.g., automatic graph-drawing tools,

But often focuses on graph-theoretical properties and rarely involves the semantics associated with the data

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Sitema

# Top 10 Problems: Paradigm shift



- In 1990s most InfoVis tools dealt with structures such as cone tree, treemap, and hyperbolic views
- Paradigm shifts to dynamic visualization
- Changes over time and thematic trends
- Draw users' attention to changes and trends: built-in trend detection mechanisms
- Collaboration with data mining and artificial intelligence communities







[Chen 2005]

#### Top 10 Problems: Causality, visual inference, & predictions

Visual thinking, reasoning, and analytics: InfoVis powerful medium for finding causality, forming hypotheses, and assessing available evidence



Tufte's re-visualization of the data from the challenger space shuttle disaster and Snow's map of cholera deaths

Challenge is to resolve conflicting evidence and suppress background noises

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Freely interact with raw data as well as with its visualizations to find causality

Potential areas: evidence-based medicine, technology forecasting, collaborative recommendation, intelligence analysis, and patent examination

[Chen 2005]

disciplinary

Top 10 Problems: Knowledge domain visualization

Difference between knowledge and information can be seen in terms of the role of social construction

Knowledge involves interpretations and decisions

Interacting with InfoVis can be more than retrieving individual items of information

Entire body of domain knowledge is subject to the rendering

The KDViz problem is rich in detail, large in scale, extensive in duration, and widespread in scope



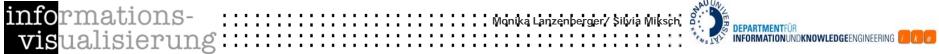




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- [Belle, et al.] Sebastian Kay Belle, Daniela Oelke, Sonja Oettl, and Mike Sips: Exploration of the local distribution of major ethnic groups in the USA, 2006. http://www.cs.umd.edu/hcil/InfovisRepository/contest-2006/files/Belle.pdf (checked online 6. Jan. 2009).
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