

Visualisierung – Aktuelle Themen und Trends

Eduard Gröller

Institute of Computer Graphics and Algorithms
Vienna University of Technology



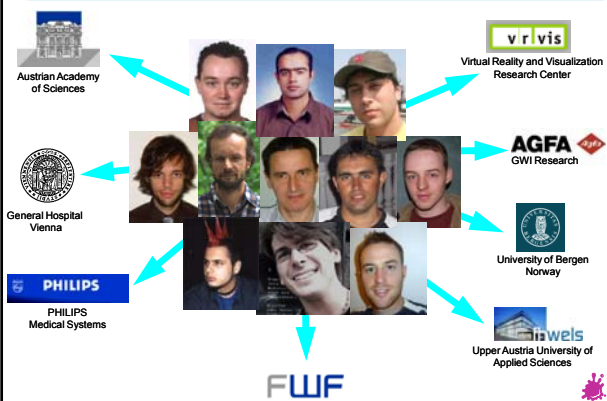
Outline




- Vis-group at Vienna University of Technology
- Brief Comments on Visualization
- Challenges in Visualization





The vis-group



Outline 


- Vis-group at Vienna University of Technology
- Brief comments on Visualization
- Challenges in Visualization




Scientific Visualization - Information Visualization 

“The use of computer-supported, interactive, visual representations of (abstract) data to amplify cognition”


- computer-based - new medium
- interactive - direct manipulation & animation
- visual representations - use human perception
- data - task specific
- amplify cognition - helping people to think

Eduard Gröller 


Visualization – Three Major Areas 

- Three major areas
 - ◆ Volume Visualization } **Scientific Visualization** Inherent spatial reference
 - ◆ Flow Visualization } 3D
 - ◆ **Information Visualization** nD

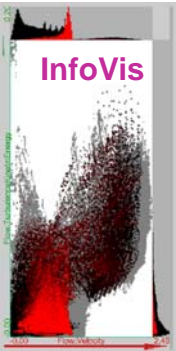
Usually no spatial reference

Eduard Gröller, Helwig Hauser 

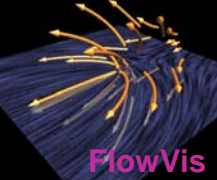
Visualization Examples TU
WIEN



VolVis



InfoVis



FlowVis

Eduard Gröller, Helwig Hauser TU
WIEN


InfoViz	vs.	SciViz
<ul style="list-style-type: none"> ■ Abstract data ■ n-dimensional ■ Very important: <ul style="list-style-type: none"> ◆ Visual metaphor ◆ User interaction ◆ Exploration, Analysis, Presentation 		<ul style="list-style-type: none"> ■ Concrete Data ■ 2- oder 3- dimensional, time related? ■ Very important: <ul style="list-style-type: none"> ◆ 3D-rendering ◆ Fast rendering ◆ Analysis, Exploration, Presentation

Eduard Gröller, Helwig Hauser TU
WIEN


Outline TU
WIEN

- Vis-group at Vienna University of Technology
- Brief comments on Visualization
- Challenges in Visualization

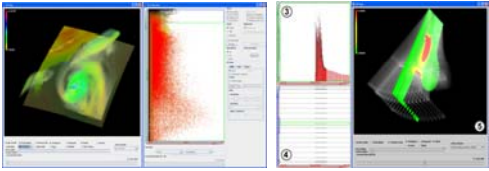
TU
WIEN

Challenges in Visualization 


- Scientific Visualization ↔ Information Visualization
-
-
-
-
-

Eduard Gröller 

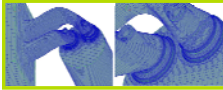
SimVis: Interactive Visual Analysis of Large & Complex Simulation Data




Dr. Helmut Doleisch
VRVis Research Center

<http://www.VRVis.at/> 

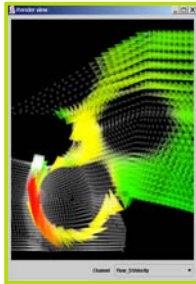
The Beginning: CFD Data

- computational fluid dynamics simulation
- data resulting from CFD:
 - grid-based geometry
 - scalar and vector data per grid element (cell or vertex)
 - time-dependent results
 - time-varying grid geometries
- data characteristics:
 - multi-dimensional data 
 - large data sets (#cells * #timesteps * #dim.)
 - data ranges differ by many magnitudes

Helmut Doleisch <http://www.simvis.at/> SimVis: Interactive Visual Analysis of Large & Complex Simulation Data 

Motivation

- large data sets from simulation
- goal:** support **exploration** and **analysis** of results
 - analyze n-dim. data **interactively**
 - use **3D visualization**
 - overview, zoom** and **filter, detail on demand** (Shneidermans' information seeking mantra)
- challenge:**
 - occlusion
 - interactive data handling



Helmut Doleisch
<http://www.simvis.at/>

SimVis: Interactive Visual Analysis of Large & Complex
Simulation Data



Interactive Data Handling

- sample data set size:
 - 540 million data items
 - currently working to expand to billions

cells	timesteps	attributes	cells * timesteps	cells * timesteps * attributes
704.900	20	16	14.098.000	225.568.000
150.124	600	6	90.074.400	540.446.400
7.680.000	288	15	2.211.840.000	33.177.600.000

Helmut Doleisch
<http://www.simvis.at/>

SimVis: Interactive Visual Analysis of Large & Complex
Simulation Data



SimVis

- VRVis' solution for these challenges
- Feature-based visualization framework
- SimVis key features:
 - Multiple, linked views
 - Interactive feature specification
 - Focus+Context visualization
 - Smooth feature boundaries
 - Explicit feature representation
 - On-the-fly attribute derivation

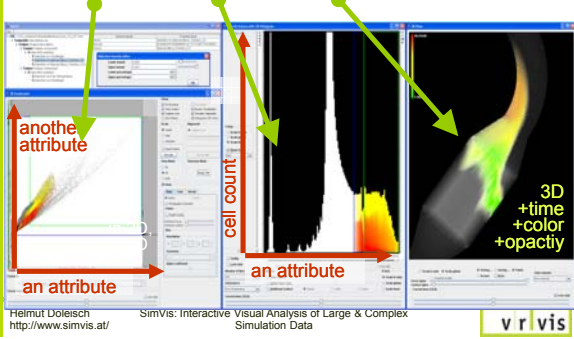
Helmut Doleisch
<http://www.simvis.at/>

SimVis: Interactive Visual Analysis of Large & Complex
Simulation Data



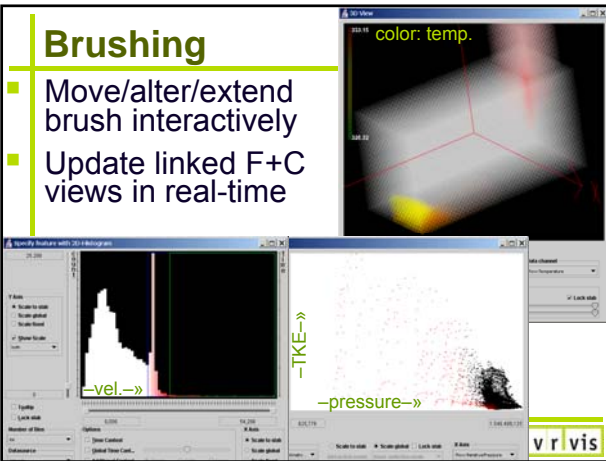
SimVis: Multiple Views

- Scatterplots, histogram, 3D(4D) view, etc.



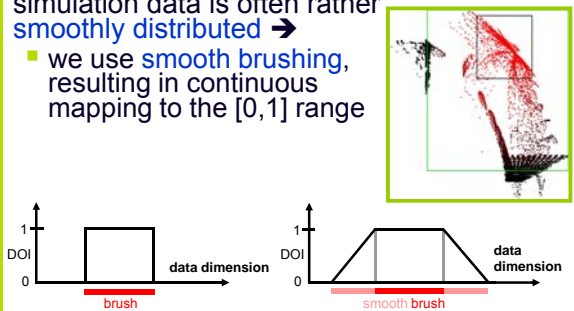
Brushing

- Move/alter/extend brush interactively
- Update linked F+C views in real-time

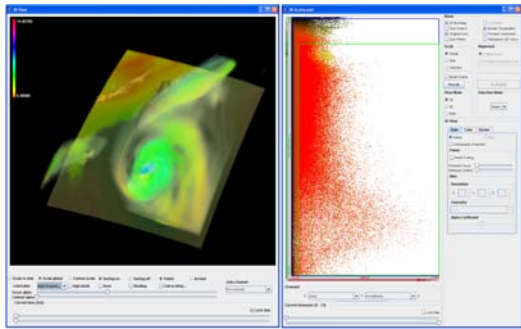


Brushing extensions: smooth brushing

- simulation data is often rather smoothly distributed →
 - we use smooth brushing, resulting in continuous mapping to the [0,1] range



Brushing extensions: smooth brushing



Helmut Doleisch
<http://www.simvis.at/>

SimVis: Interactive Visual Analysis of Large & Complex
 Simulation Data



Challenges in Visualization



- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities
-
-
-
-

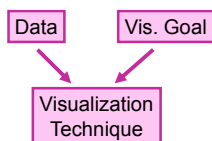
Eduard Gröller



New Data Sources - Novel Imaging Modalities




- Challenges
 - ◆ Very large (abstract) data sets
 - ◆ High-dimensional, multi-valued, multi-modal, heterogeneous
 - ◆ Time varying
 - Spatially sparse/dense, temporally sparse/dense
 - Need for registration
 - Need for feature extraction
- Examples
 - ◆ Web 2.0
 - ◆ Dual energy CT




Eduard Gröller



New Data Sources – Web 2.0 


- Social networks, wikis, blogs, data warehouses



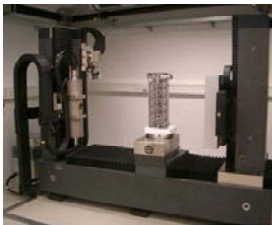
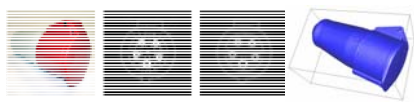
[Pfeffer 2007]

- Examples
 - ◆ MySpace
 - ◆ LinkedIn
 - ◆ Flickr
 - ◆ YouTube

Eduard Gröller


Novel Imaging Modalities – Dual Energy CT 

- Micro CT – Industrial CT
- Two X-ray sources
- Metrology and dimensional measurement
- Multi-materials
- Res: 508x523x61
- Voxelsize (μm) 200
- Data have complimentary strengths and weaknesses

[Heinzl et al. 2007]

Eduard Gröller

Challenges in Visualization 

- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities
- **Visual Analytics** - Visual Computing – Knowledge Assisted Visualization
-
-
-

Eduard Gröller

Visual Analytics



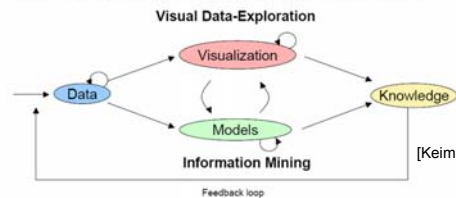
“Visual Analytics is the science of analytical reasoning facilitated by interactive visual interfaces“

What do we have?

- Automatic Knowledge Discovery & Information Mining
- Interactive Visual Data-Exploration

What do we need?

Tight Integration of Visual and Automatic Data Analysis Methods with Database Technology for a Scalable Interactive Decision Support



[Keim, Thomas 2007]

Eduard

Feedback loop

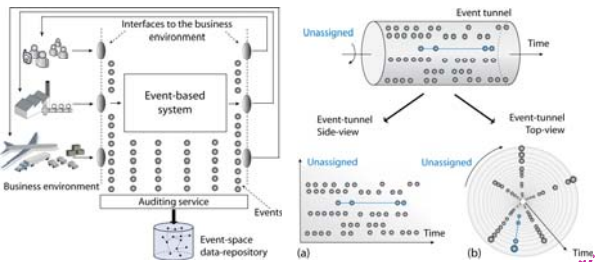


Visual Analytics – The Event Tunnel (1)



Interactive Visualization of Complex Event Streams for Business Process Pattern Analysis

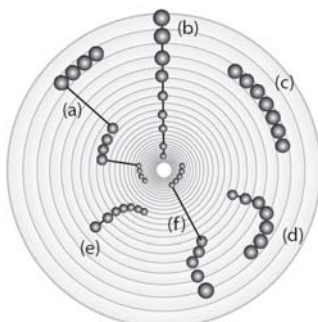
[Suntinger et al. 2008]



Eduard Gröller



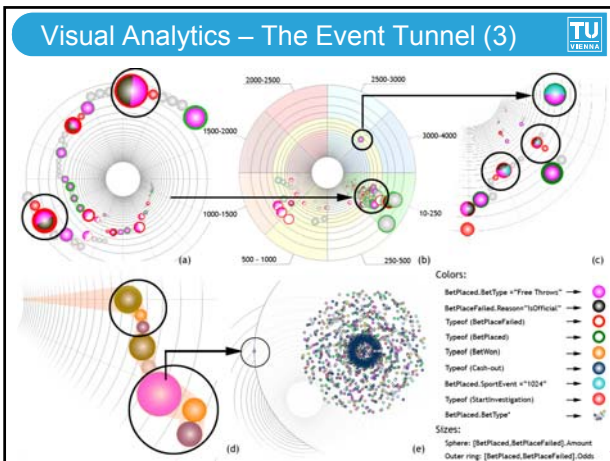
Visual Analytics – The Event Tunnel (2)



- (a) Stair pattern
Process with several idle times
- (b) Non-interfering chain
Process with regular steps
- (c) Parallel chain
Fast process without idle times
- (d) Acceleration worm
Process execution accelerated continuously
- (e) Deceleration worm
Process execution decelerated continuously
- (f) Rattlesnake
Process with one extreme idle time

Eduard Gröller





Challenges in Visualization

- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities
- Visual Analytics - **Visual Computing** – Knowledge Assisted Visualization
-
-
-

Eduard Gröller

Visual Computing - Computational Sciences

Computational Sciences

```

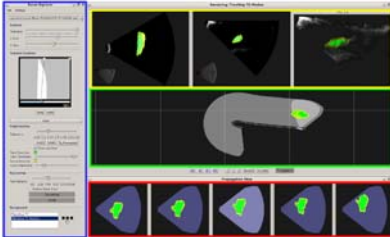
    graph LR
      A[Data Acquisition] --> B[Data Enhancement]
      B --> C[Visualization Mapping]
      C --> D[Quantitative Analysis]
  
```

- Visual Computing
 - Scientific visualization
 - Computer vision
 - Human computer interaction

Eduard Gröller

Visual Computing – Sonar Explorer (1)

- 4D sonar data
- Cones with res: 25x20x1319
- Ping rate 1 Hz
- 2 GB/ping
- Time steps overlapping
- Highly anisotropic
- Noisy
- Signal strength reduced with spreading and absorption

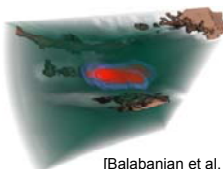


[Balabanian et al. 2007]

Eduard Gröller

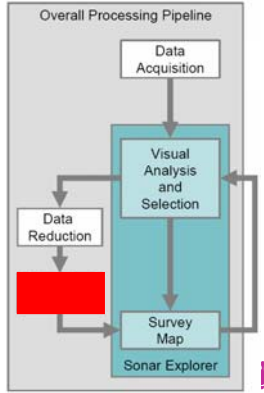
Visual Computing – Sonar Explorer (2)

- Fish school monitoring
 - Size of school
 - Center of gravity
 - Shape parameters
 - Motion characteristics



[Balabanian et al. 2007]

Eduard Gröller



Challenges in Visualization

- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities
- Visual Analytics - Visual Computing – Knowledge Assisted Visualization
-
-
-

Eduard Gröller

Challenges in Visualization



- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities
- Visual Analytics - Visual Computing – Knowledge Assisted Visualization
- Scalability
-
-

Eduard Gröller



Scalability



- Challenges [Keim, Thomas 2007]
 - ◆ amount of data and dimensionality
 - ◆ numbers of data sources and heterogeneity
 - ◆ data quality and data resolution
 - ◆ dynamicity and novelty
 - ◆ data representation and visual resolution
- Examples
 - ◆ Focus+Context
 - ◆ Aggregation
 - ◆ Abstraction and Illustration

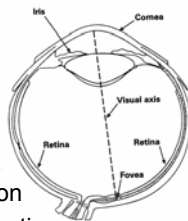
Eduard Gröller



Scalability - Focus+Context Principle



- Basic idea of Focus+Context Visualization:
- Important regions in great detail (focus)
 - Global view with reduced detail (context)
 - Dynamic integration



Rationale

- Zooming hides the context
- Two separate displays split attention
- Human vision has both fovea and retina

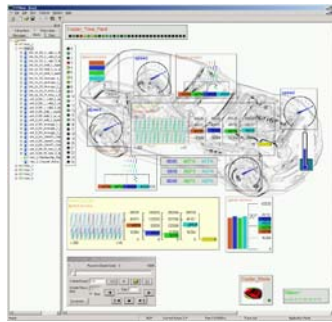
Eduard Gröller



Scalability - Process Visualization (1)



- Improving singular instruments
 - ◆ History encoding
 - ◆ Multi-instruments
 - ◆ Levels of detail (LOD)
- Improving the monitoring system
 - ◆ Focus+Context (F+C) rendering
 - ◆ Collision avoidance



[Matković et al. 2002]

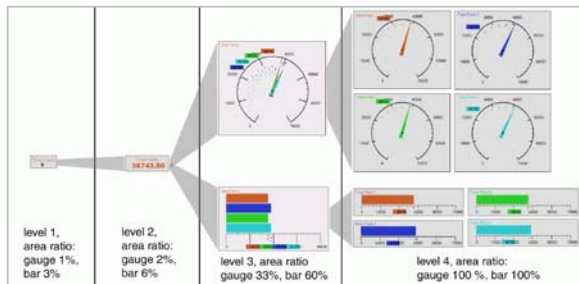
Eduard Gröller



Scalability - Process Visualization (2)



- Various instruments can be used to construct Levels of Detail (LODs)



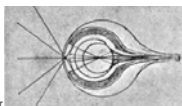
Eduard Gröller



Scalability - Illustration



- An illustration is a picture with a communicative intent
- Conveys complex structures or procedures in an easily understandable way
- Uses abstraction to prevent visual overload – allows to focus on the essential parts
- Abstraction is visualized through distinct stylistic choices



Stefan Bruckner



Scalability - Abstraction



- Fundamental for creating an expressive illustration
- Introduces a distortion between visualization and underlying model
- Different degrees of abstraction introduced at different levels
- Task of an illustrator: find the necessary abstraction for the intent of the illustration



„As detailed as necessary - as simple as possible“

Stefan Bruckner

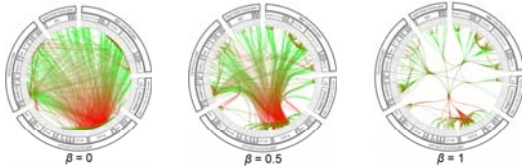


Scalability – Illustration Examples

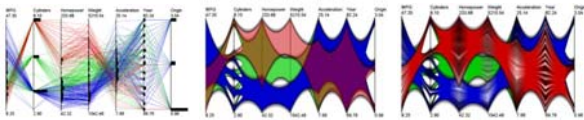


- Hierarchical Edge Bundles

[Holten 2006]



- Illustrative Parallel Coordinates



Eduard Gröller

[McDonnell, Mueller 2008]

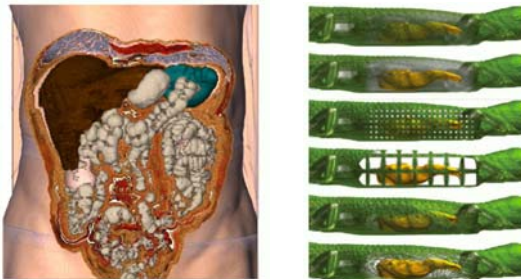


Scalability – Smart Visibility (1)




- Importance-driven feature enhancement

[Viola et al. 2004, 2005]

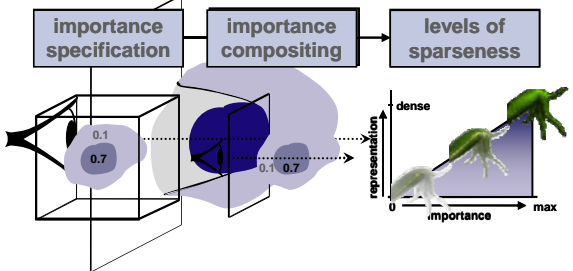


Stefan Bruckner





Scalability – Smart Visibility (2) 

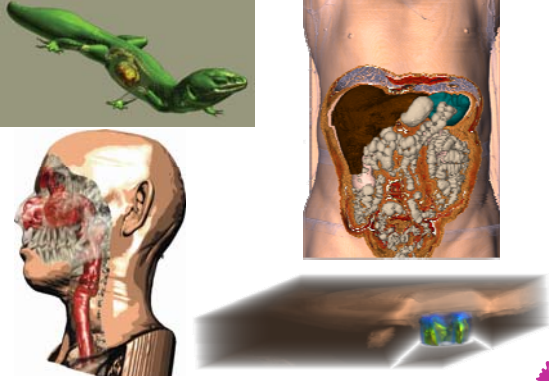
[Viola et al. '04 '05]





importance-driven feature enhancement

I. Viola and E. Gröller 

Scalability – Smart Visibility (3) 



I. Viola and E. Gröller 

Scalability 

*Do not fight complexity
with complexity*

Challenges in Visualization



- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities
- Visual Analytics - Visual Computing – Knowledge Assisted Visualization
- Scalability
- Visualization Yes ! – Interaction No ?
-

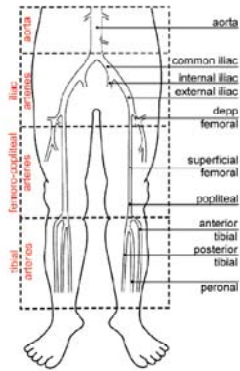
Eduard Gröller



Visualization Yes ! – Interaction No ?



- Problems
 - ◆ Interaction is very time-consuming
 - ◆ Interaction prevents comparisons
 - ◆ Interaction hampers reporting
- Challenges
 - ◆ Provide standardized views
 - ◆ Algorithms highly parameterized – provide sensible default settings
 - ◆ Support automatic parameter tuning
 - ◆ Provide navigational aids
- Examples
 - ◆ Automatic view point selection
 - ◆ Focus of attention
 - ◆ Automatic light placement (inconsistent lighting)
 - ◆ Automatic reporting
 - ◆ Dynamic poster - automatic storytelling



Eduard Gröller



Context-Preserving Rendering (1)



[Bruckner et al. 2006]

gradient magnitude $|g_x|$

shading intensity $\alpha(P)$

eye distance $|P - E|$

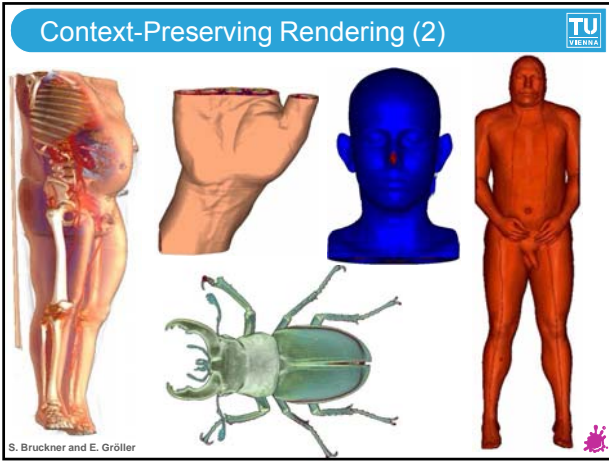
provisionally normalized opacity $\alpha_{1,2}$

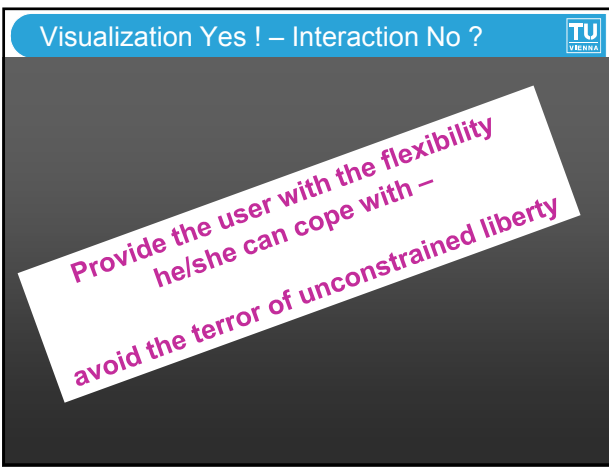
$$m(P_i) = \|g_{P_i}\| \left(\alpha_{1,2}(P_i) \times (1 - |P_i - E|) \times (1 - \alpha_{1,2}) \right)^{\alpha_{1,2}}$$

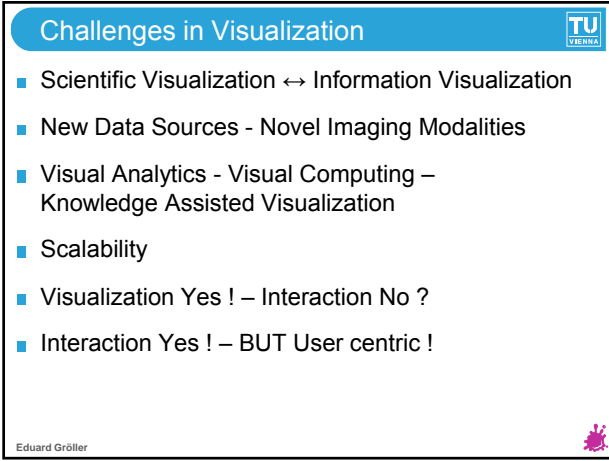
- Integrate various focus+context approaches with only few parameters

S. Bruckner and E. Gröller









Interaction Yes ! – BUT User Centric !



- Problems
 - ◆ Medical doctors do not (want to) know transfer functions
 - ◆ Complex 3D interaction is complex
- Challenges
 - ◆ Include user model (novice, experienced, expert)
 - ◆ Include motifs
 - ◆ Include user preferences
 - ◆ 2D+ navigation (instead of 3D navigation)
- Examples
 - ◆ Semantic layers for illustrative volume rendering
 - ◆ Knowledge-based navigation

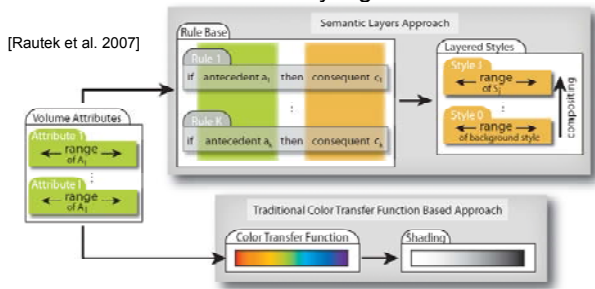
Eduard Gröller



Semantic Layers for Illustrative Volume Rendering (1)



- Mapping volumetric attributes to visual styles
- Use natural language of domain expert (rules)
- Rules evaluated with fuzzy logic arithmetics




Semantic Layers for Illustrative Volume Rendering (2)

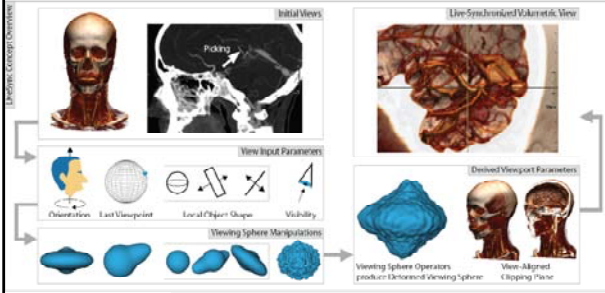



Die Grenzen meiner Sprache bedeuten
die Grenzen meiner Welt


[Ludwig Wittgenstein]

Knowledge-Based Navigation 

- Interaction with 2D slices
- Automatic generation of expressive 3D views





[Kohlmann et al. 2007] Video 

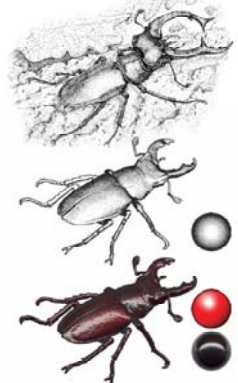
Challenges in Visualization 

- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities
- Visual Analytics - Visual Computing – Knowledge Assisted Visualization
- Scalability
- Visualization Yes ! – Interaction No ?
- Interaction Yes ! – BUT User centric !

Bring visualization into the workflow of users!!

Eduard Gröller 

Thank You for Your Attention 



**Questions ?
Comments?**

Acknowledgments
 Jean-Paul Balabanian
 Stefan Bruckner
 Helmut Doleisch
 Helwig Hauser
 Christoph Heinzl
 Krešimir Matković
 Hannes Obwegger
 Peter Rautek
 Martin Suntinger,
 Ivan Viola
 ...
