

Visualisierung – Aktuelle Themen und Trends

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



The vis-group



Austrian Academy
of Sciences



General Hospital
Vienna



PHILIPS
Medical Systems



Virtual Reality and Visualization
Research Center



University of Bergen
Norway



Upper Austria University of
Applied Sciences



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“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



Visualization – Three Major Areas

■ Three major areas

- ◆ Volume Visualization
- ◆ Flow Visualization

Scientific Visualization

Inherent spatial reference

3D

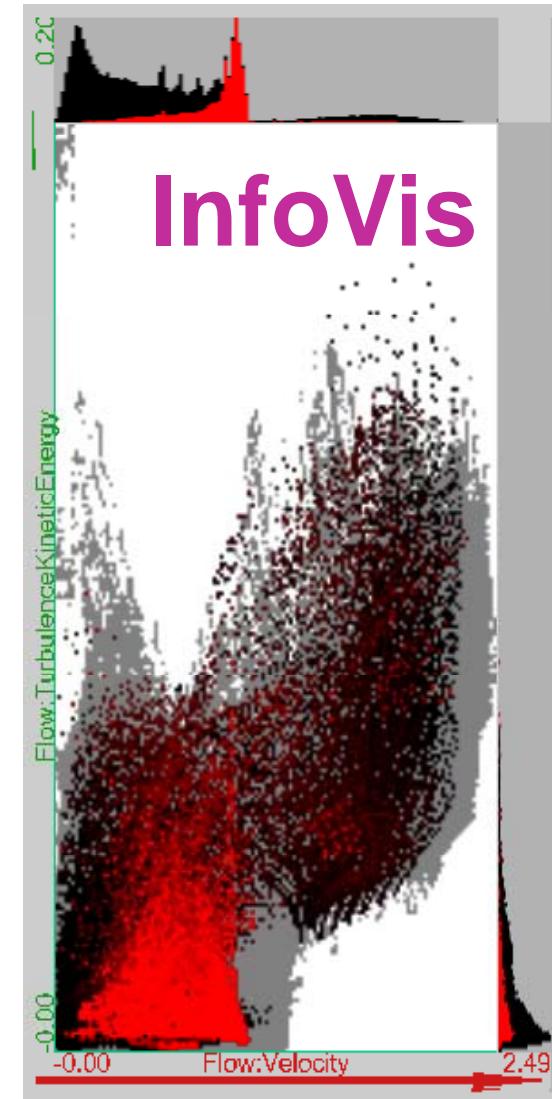
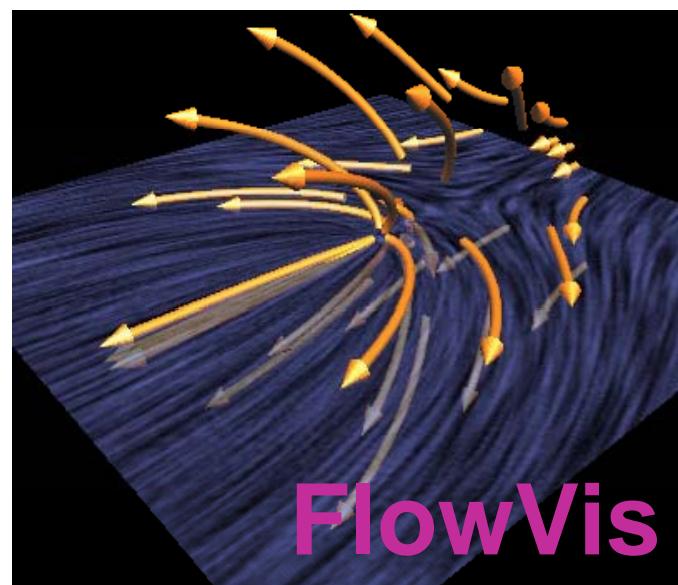
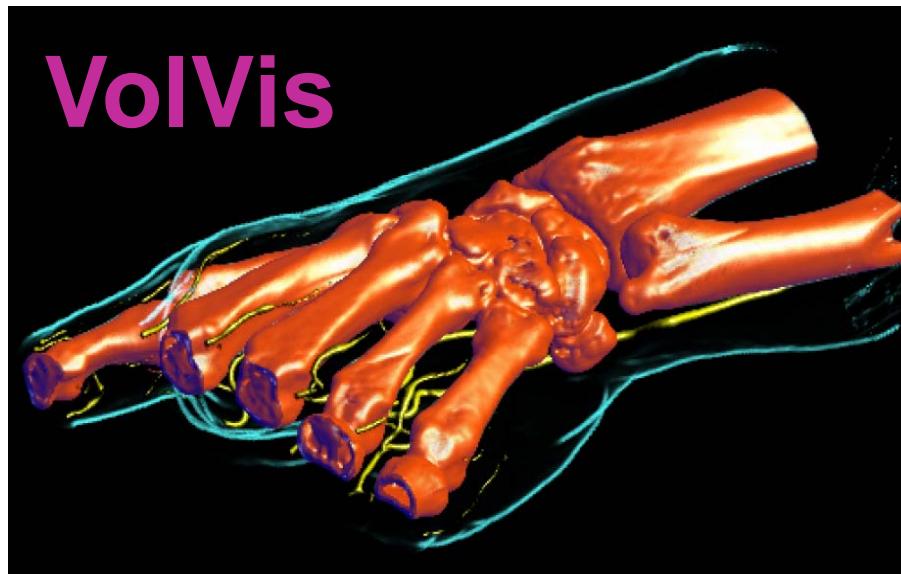
- ◆ Information Visualization

nD

Usually no spatial reference



Visualization Examples



- Abstract data
 - n-dimensional
 - Very important:
 - ◆ Visual metaphor
 - ◆ User interaction
 - ◆ **Exploration, Analysis, Presentation**
-
- Concrete Data
 - 2- oder 3-dimensional, time related?
 - Very important:
 - ◆ 3D-rendering
 - ◆ Fast rendering
 - ◆ **Analysis, Exploration, Presentation**



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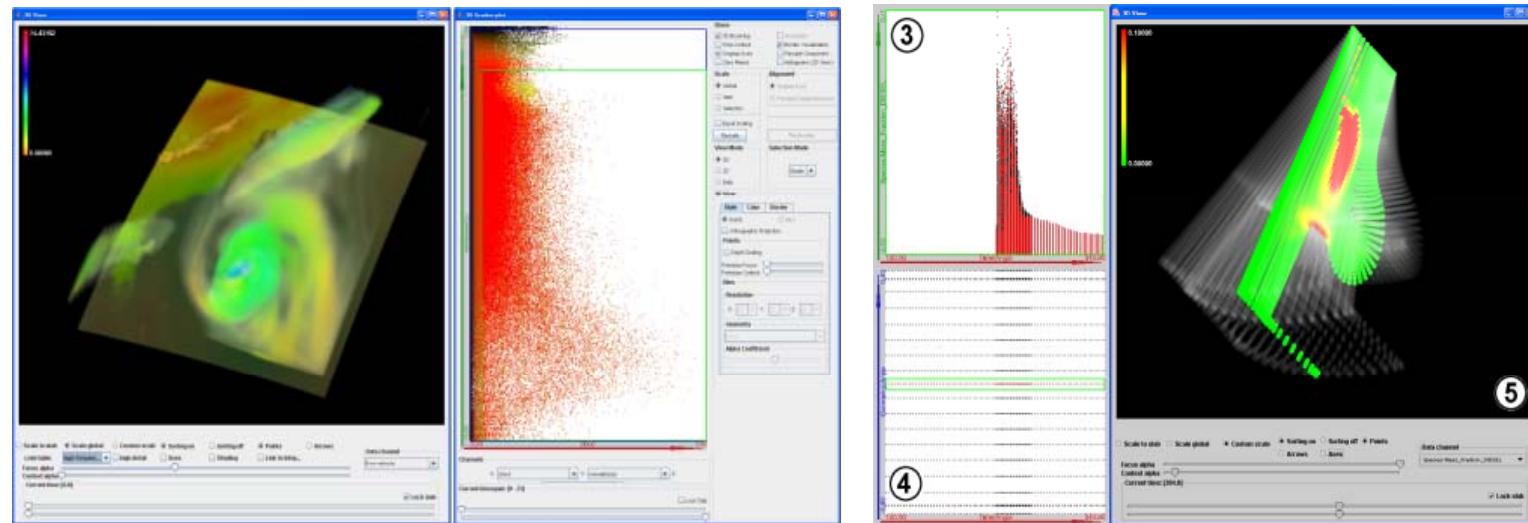


Challenges in Visualization

■ Scientific Visualization ↔ Information Visualization



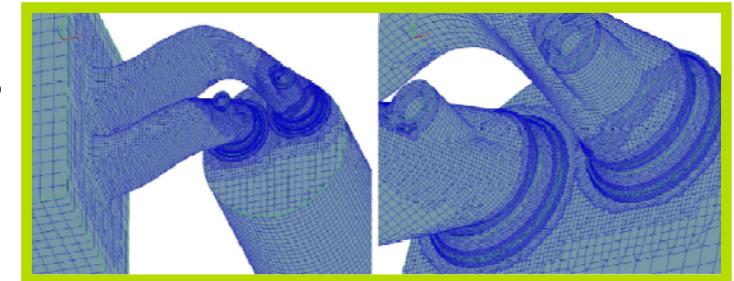
SimVis: Interactive Visual Analysis of Large & Complex Simulation Data



Dr. Helmut Doleisch
VRVis Research Center

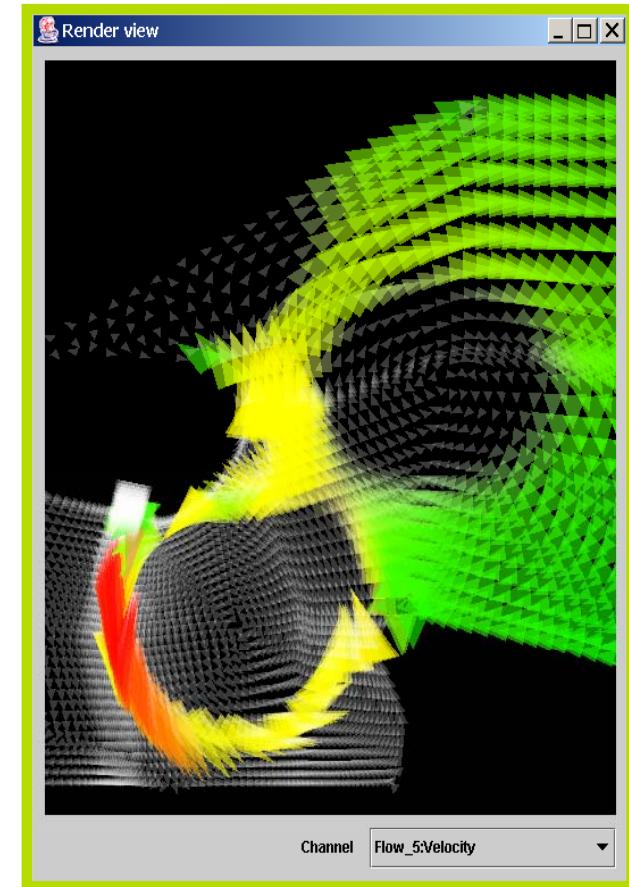
The Beginning: CFD Data

- 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



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



Interactive Data Handling

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

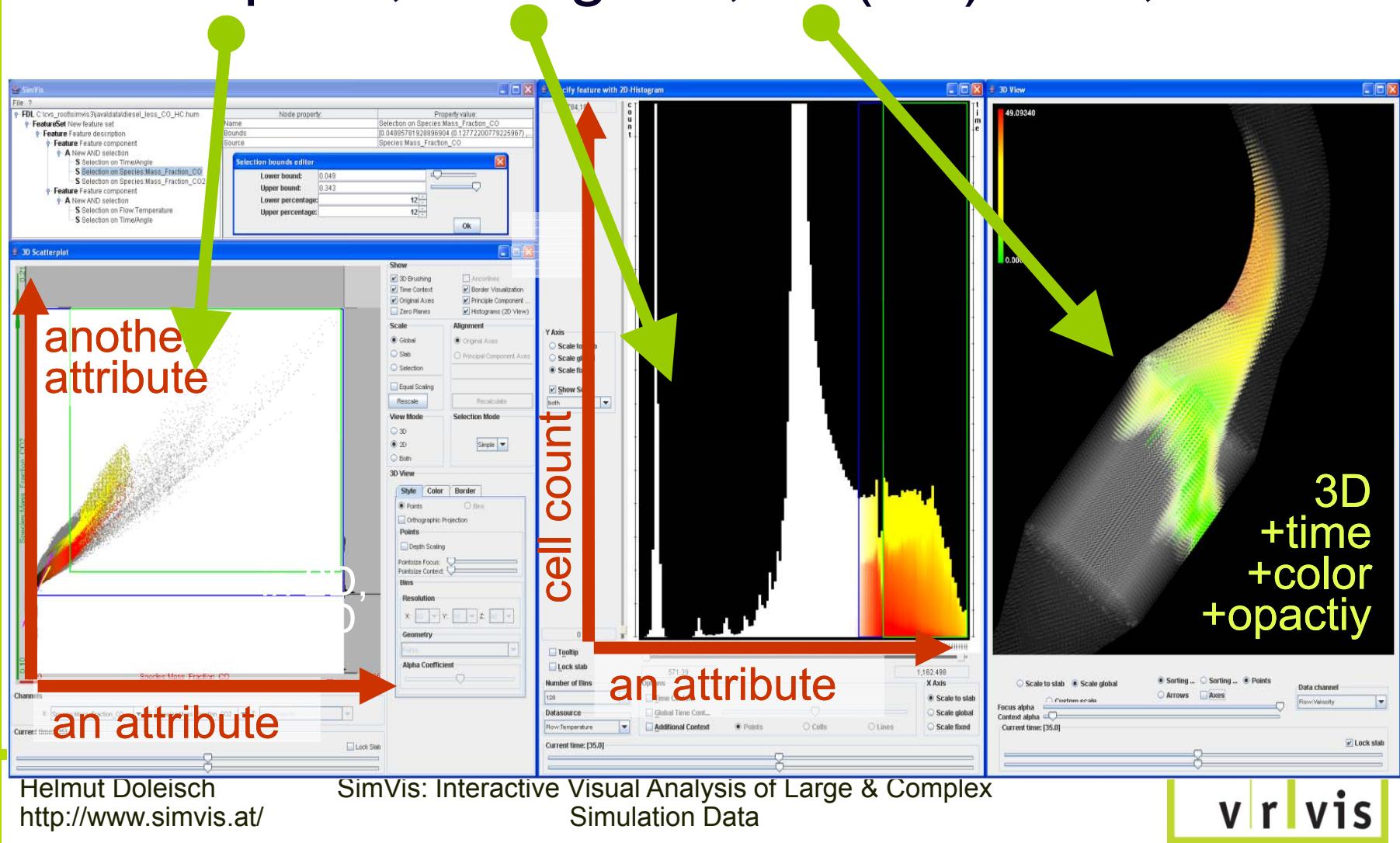
cells	timesteps	attributes	cells * timesteps	cells * timesteps * attributes
704.900	20	16	14.098.000	225.568.000
150.124	600	6	90.074.400	540.446.400
7.680.000	288	15	2.211.840.000	33.177.600.000

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

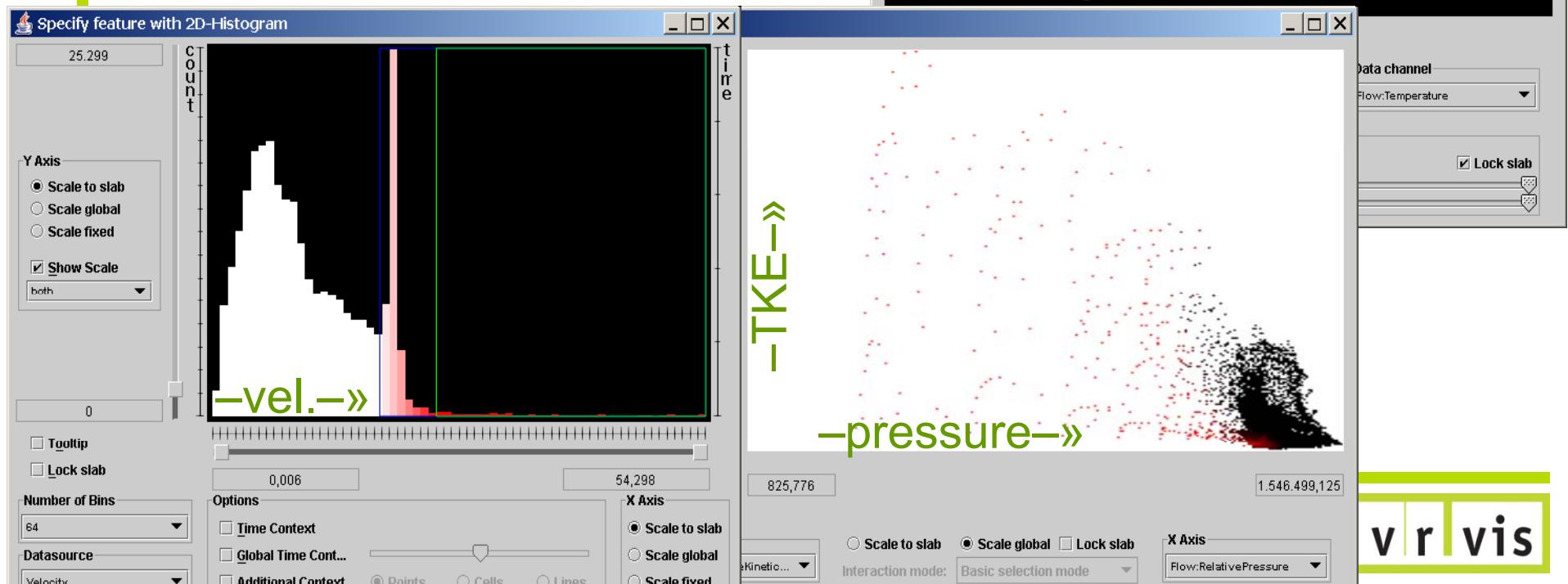
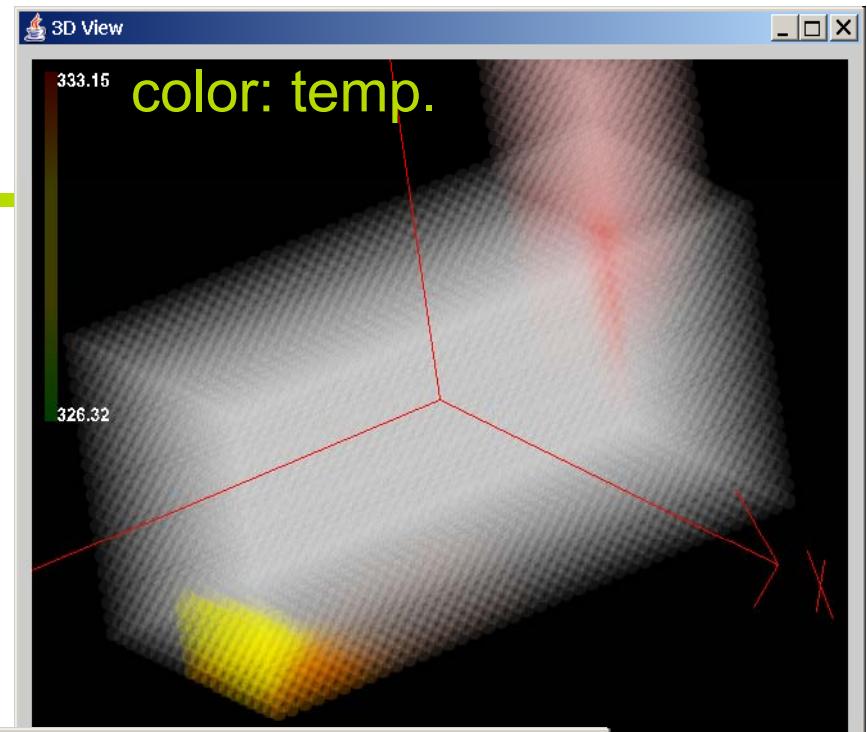
SimVis: Multiple Views

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



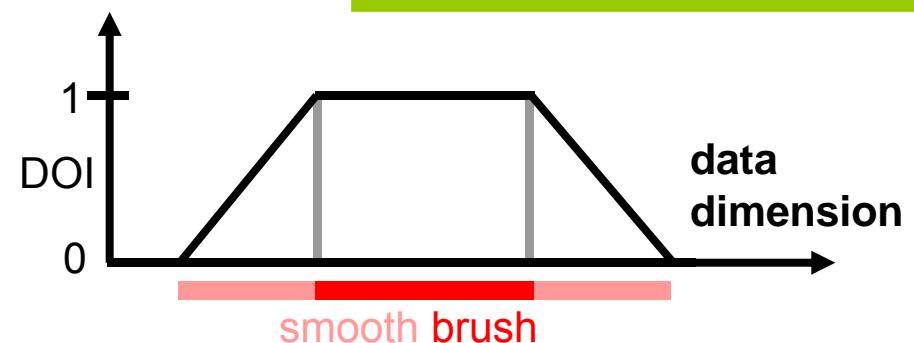
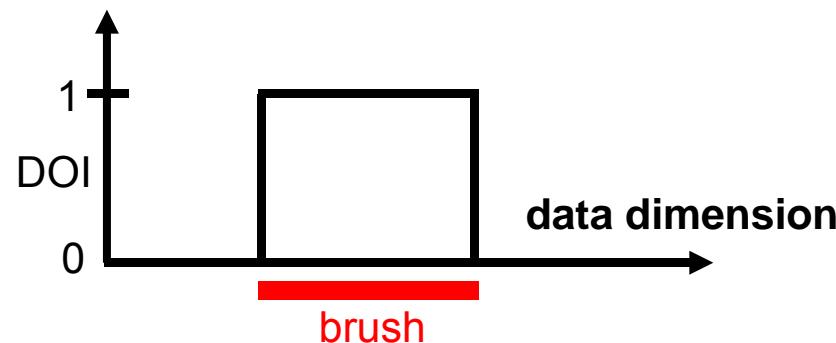
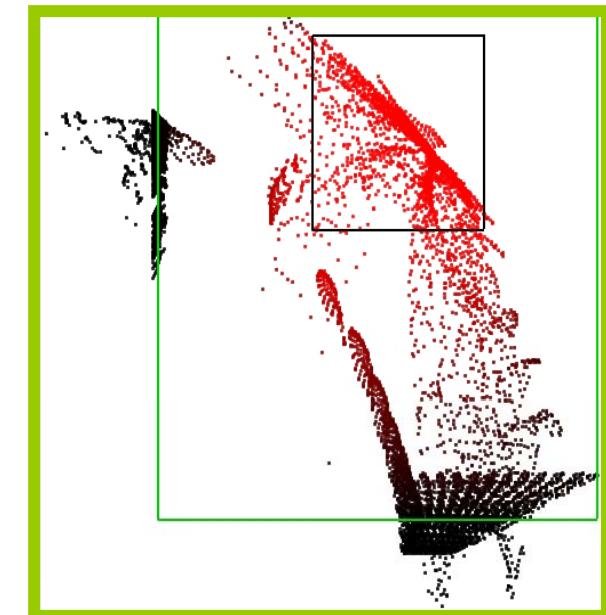
Brushing

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

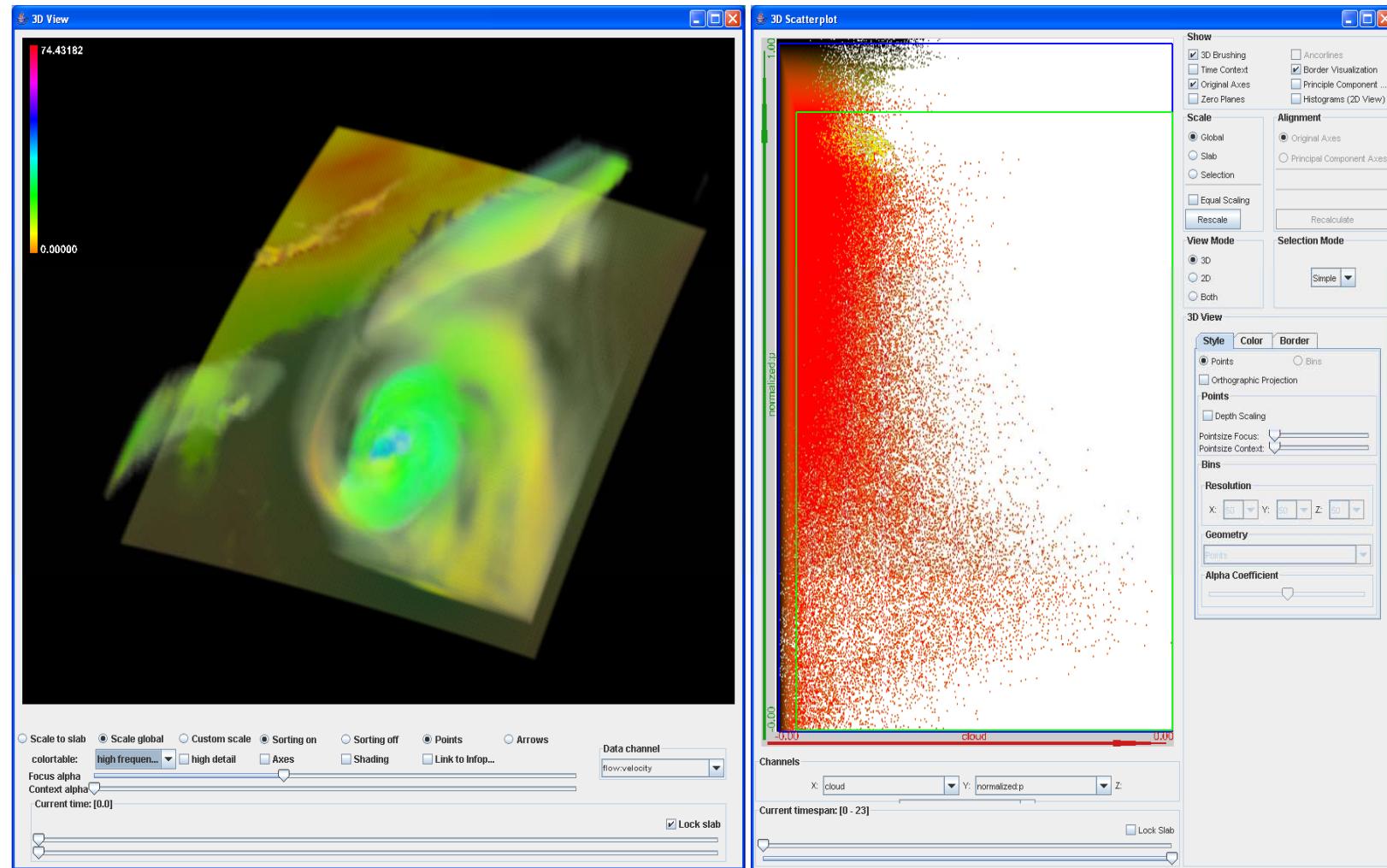


Brushing extensions: smooth brushing

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



Brushing extensions: smooth brushing

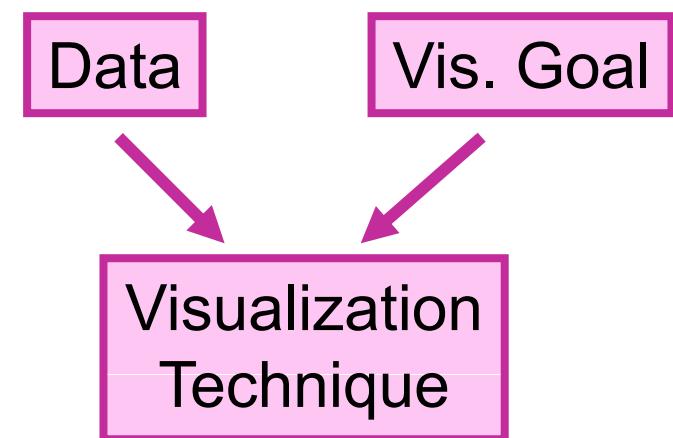


Challenges in Visualization

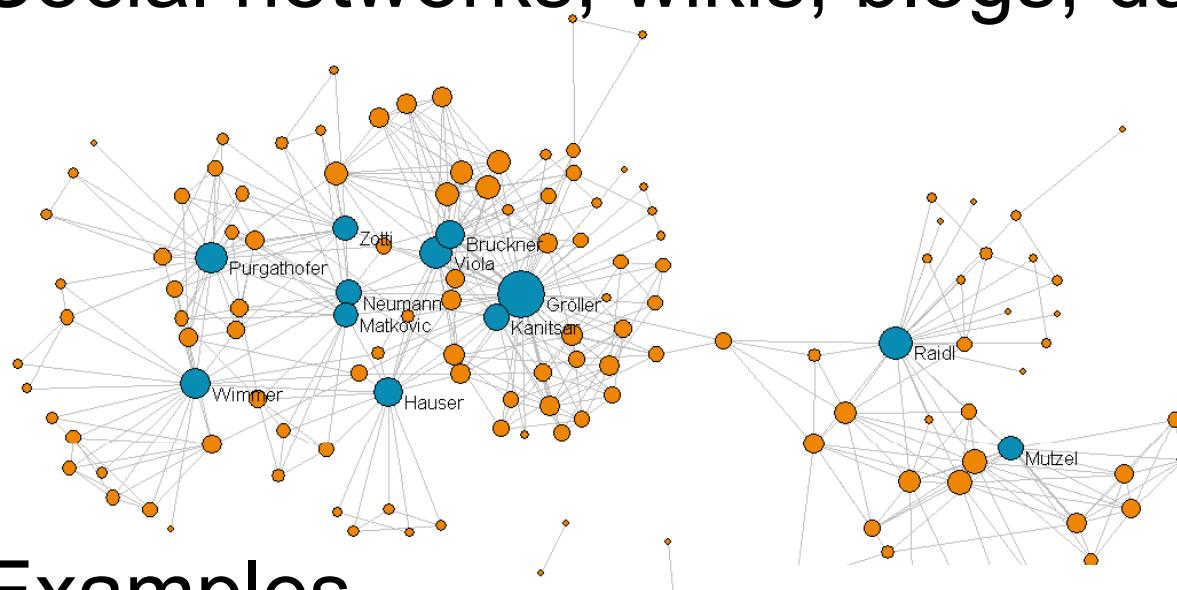
- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities



- 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



■ Social networks, wikis, blogs, data warehouses

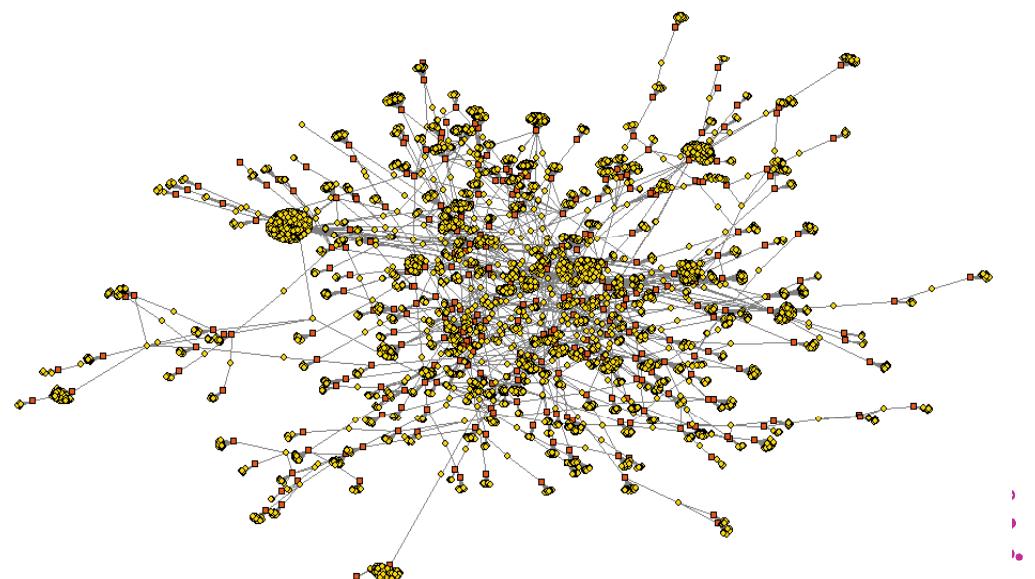


[Pfeffer 2007]

■ Examples

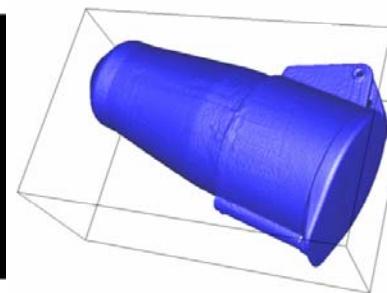
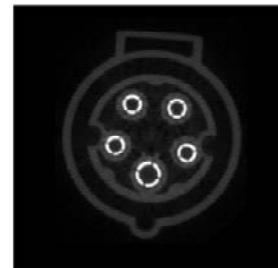
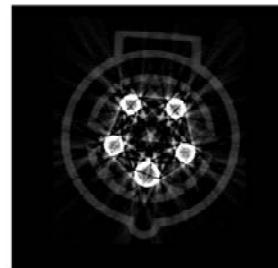
- ◆ MySpace
- ◆ LinkedIn
- ◆ Flickr
- ◆ YouTube

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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]



Challenges in Visualization

- Scientific Visualization ↔ Information Visualization
- New Data Sources - Novel Imaging Modalities
- Visual Analytics - Visual Computing – Knowledge Assisted Visualization
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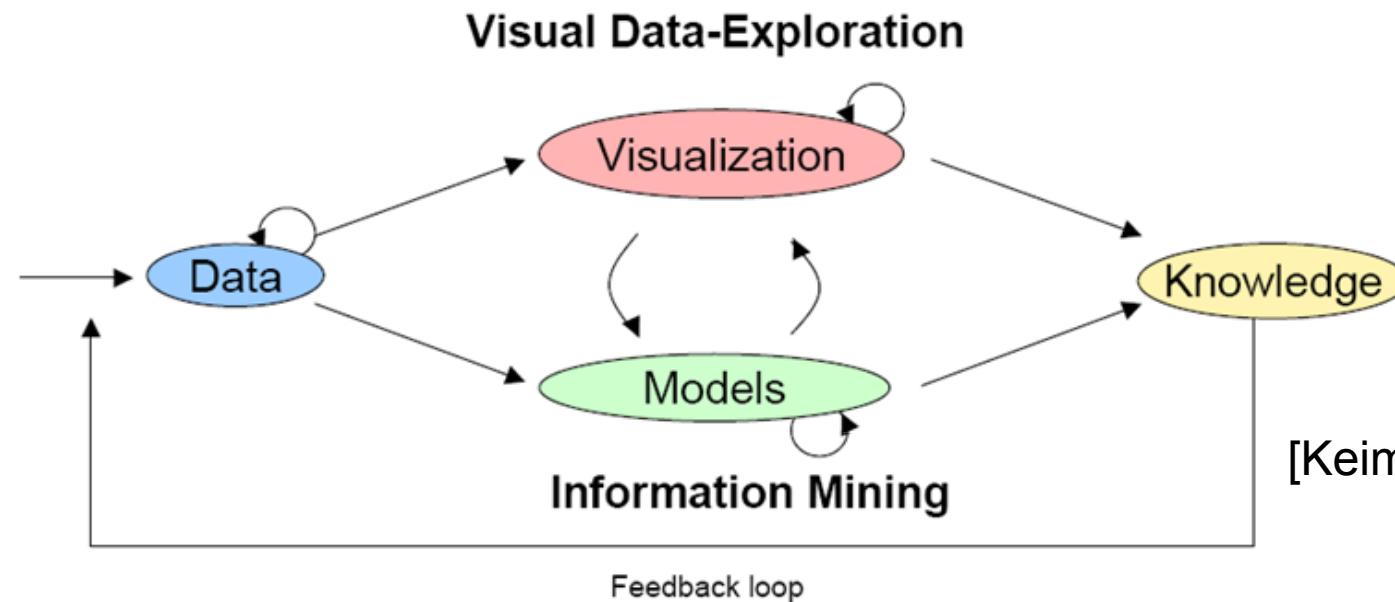
“Visual Analytics is the science of analytical reasoning facilitated by interactive visual interfaces”

What do we have?

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

What do we need?

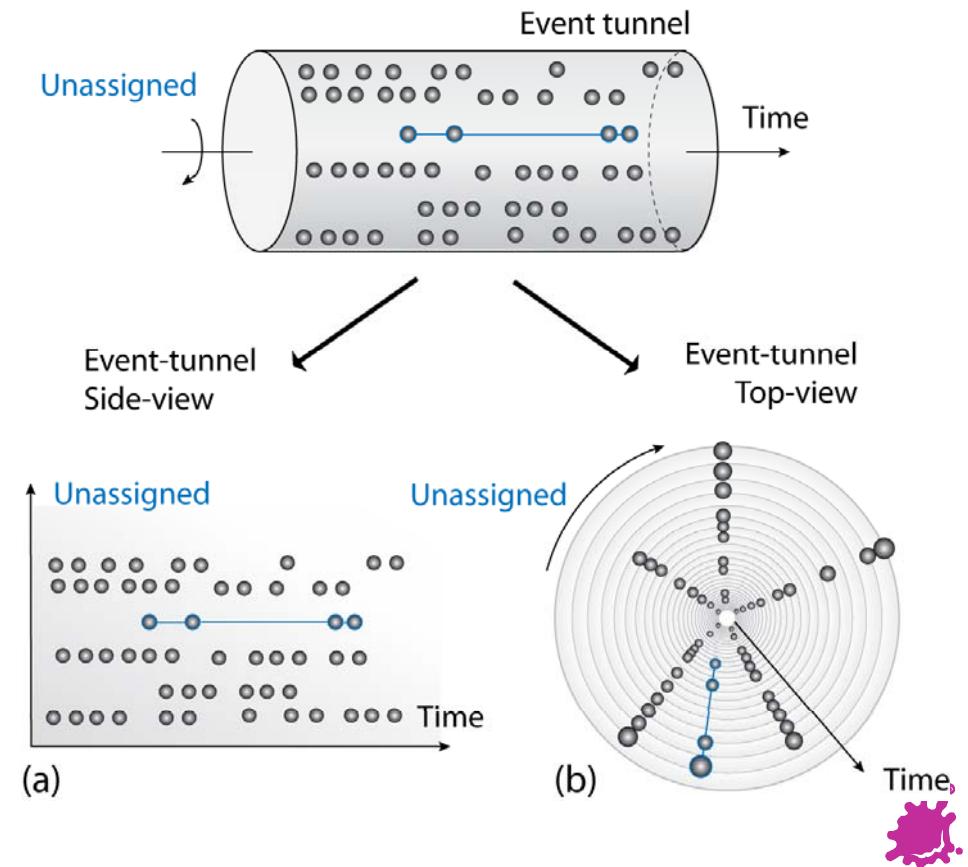
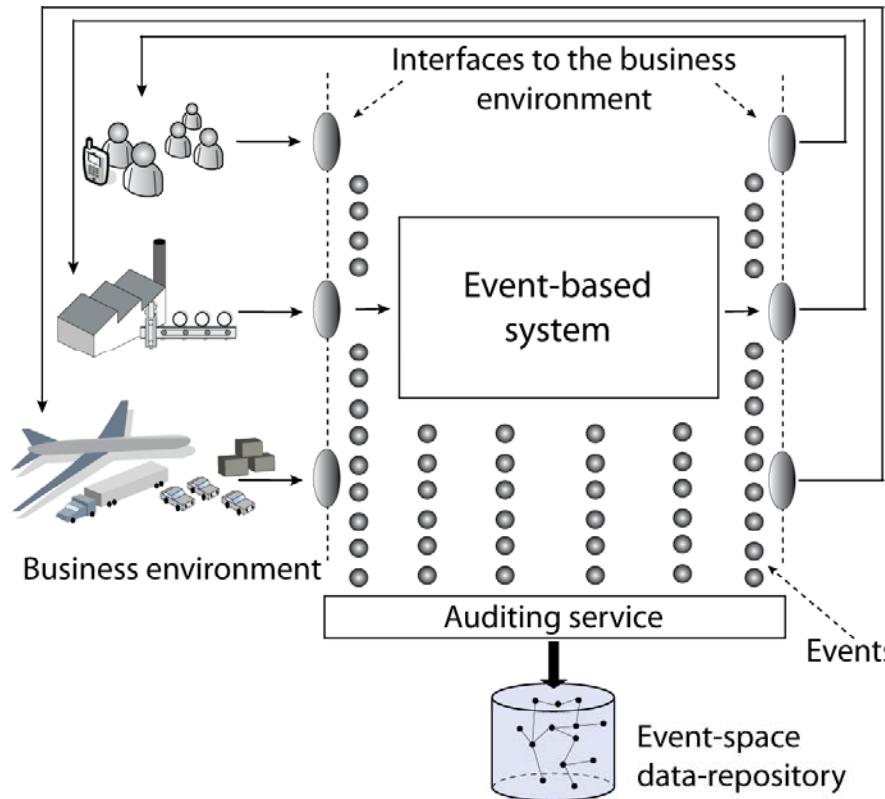
Tight Integration of Visual and Automatic Data Analysis Methods with Database Technology for a Scalable Interactive Decision Support



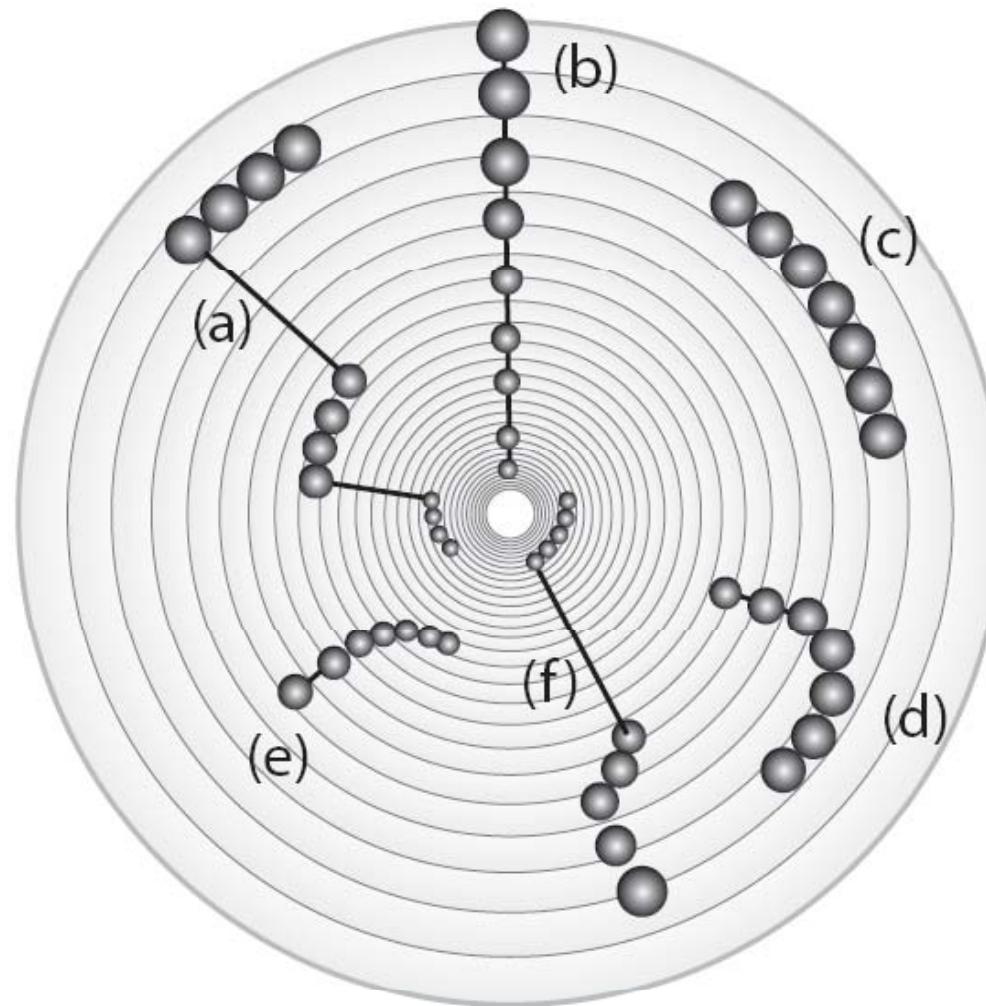
Visual Analytics – The Event Tunnel (1)

■ Interactive Visualization of Complex Event Streams for Business Process Pattern Analysis

[Suntlinger et al. 2008]



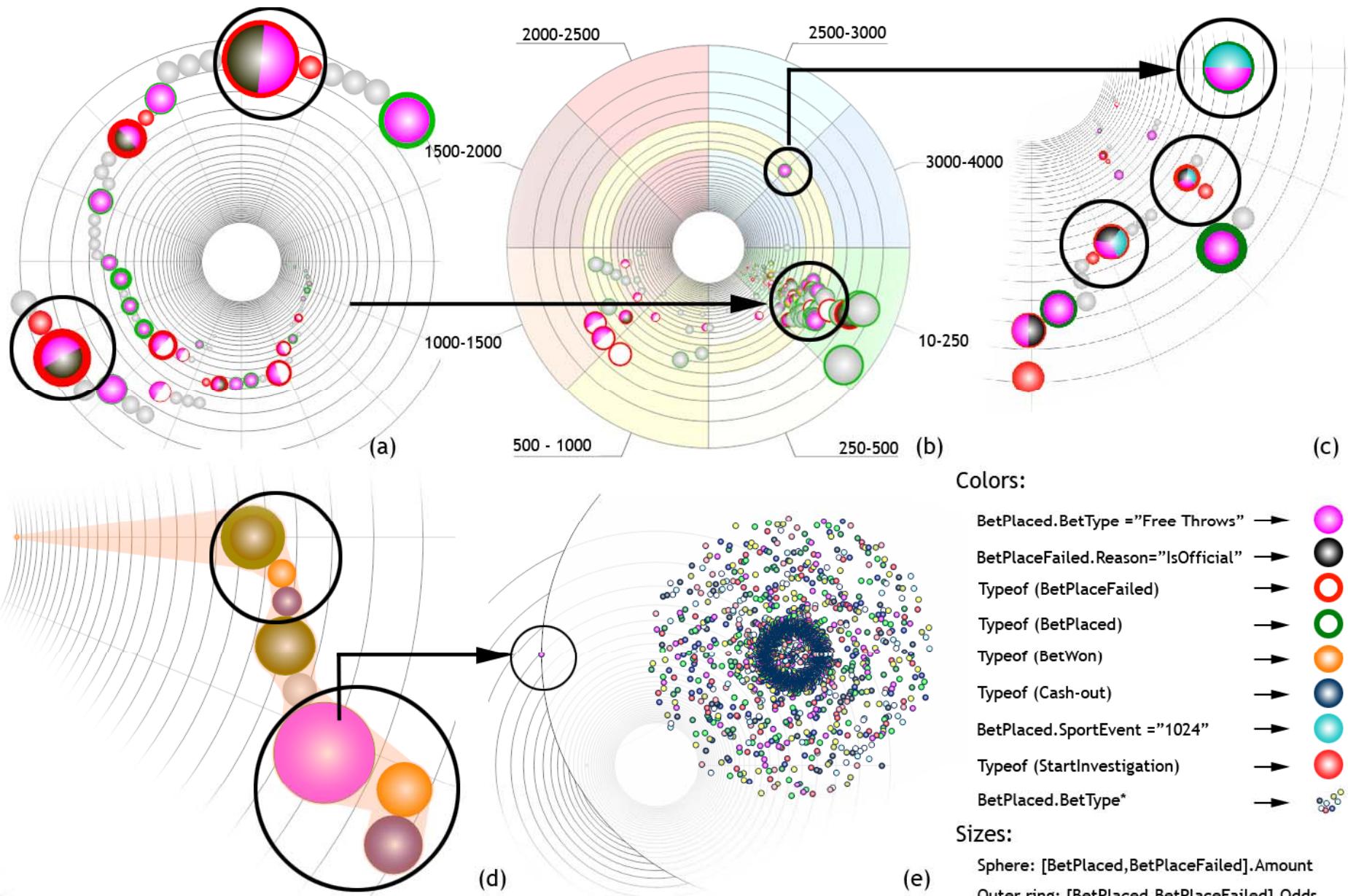
Visual Analytics – The Event Tunnel (2)



- (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



Visual Analytics – The Event Tunnel (3)

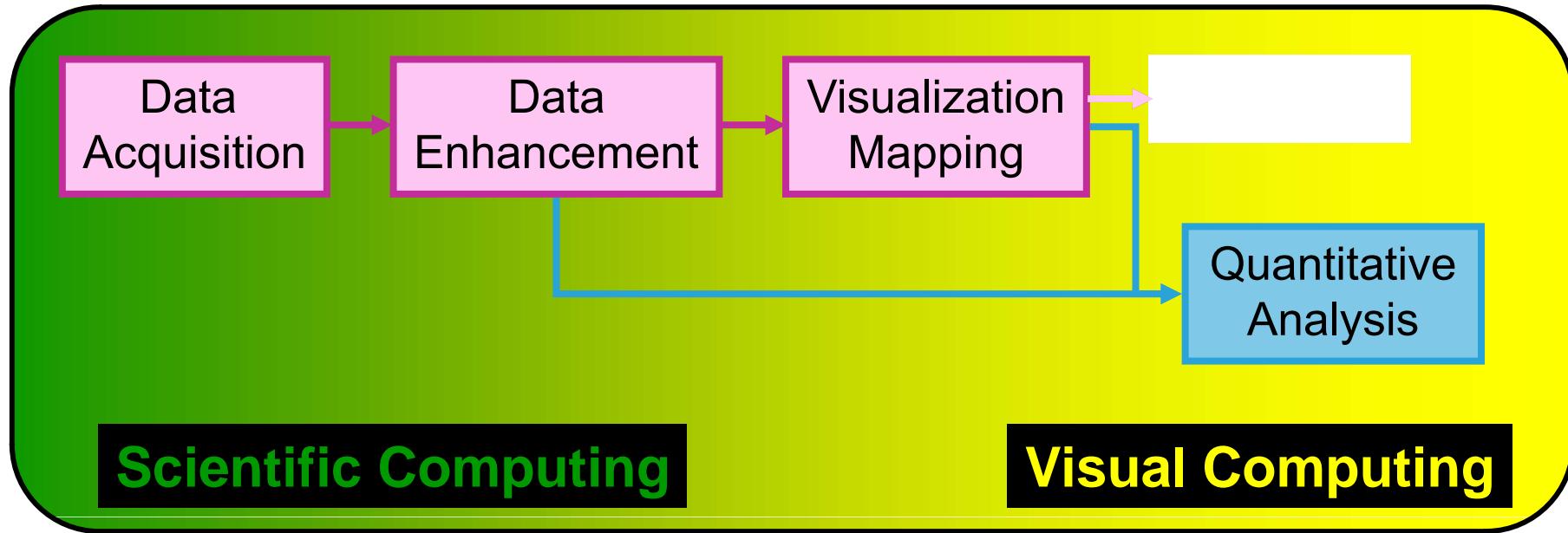


Challenges in Visualization

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

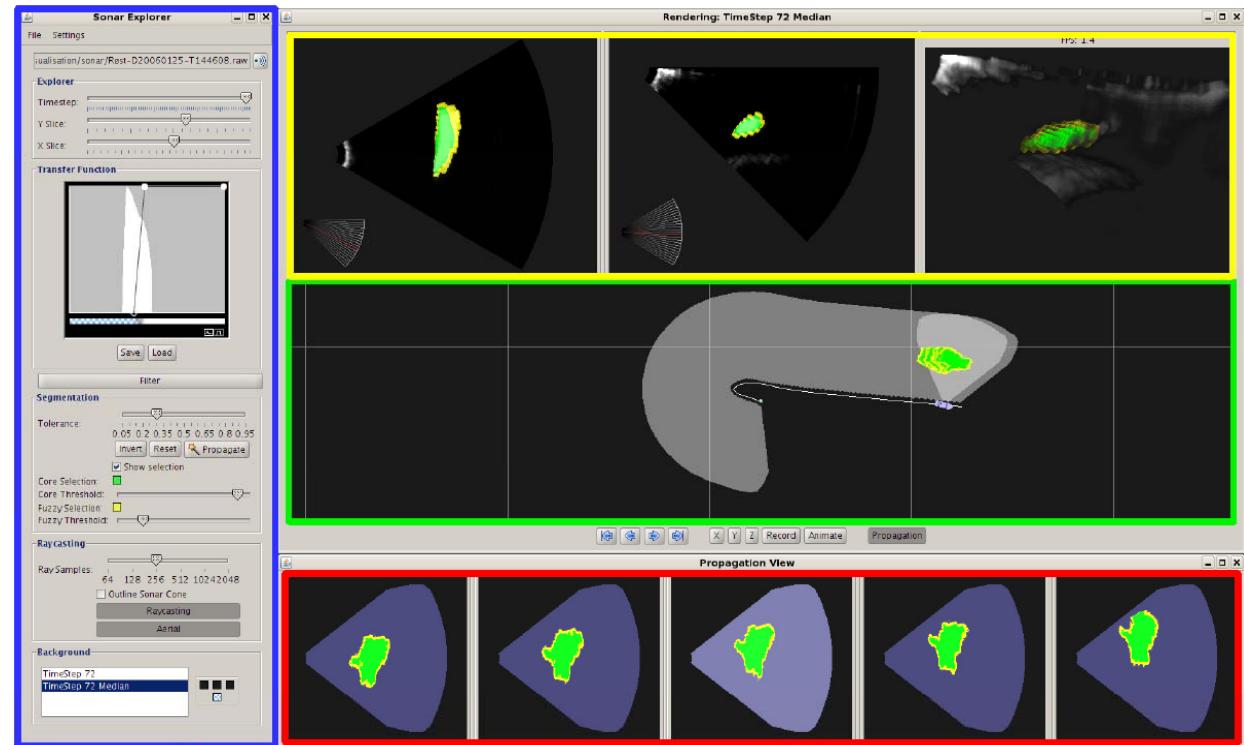


- Visual Computing
 - ◆ Scientific visualization
 - ◆ Computer vision
 - ◆ Human computer interaction

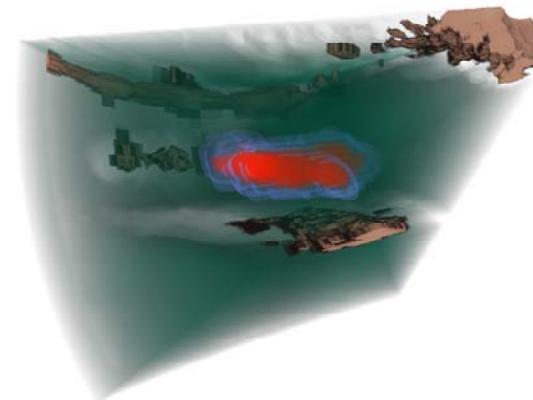


Visual Computing – Sonar Explorer (1)

- 4D sonar data
- Cones with res:
 $25 \times 20 \times 1319$
- Ping rate 1 Hz
- 2 GB/ping
- Time steps
overlapping
- Highly
anisotropic
- Noisy
- Signal strength reduced with
spreading and absorption

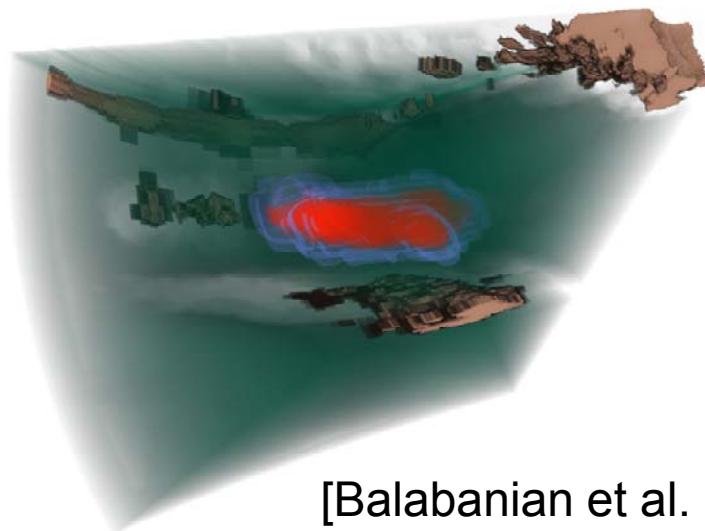


[Balabanian et al. 2007]



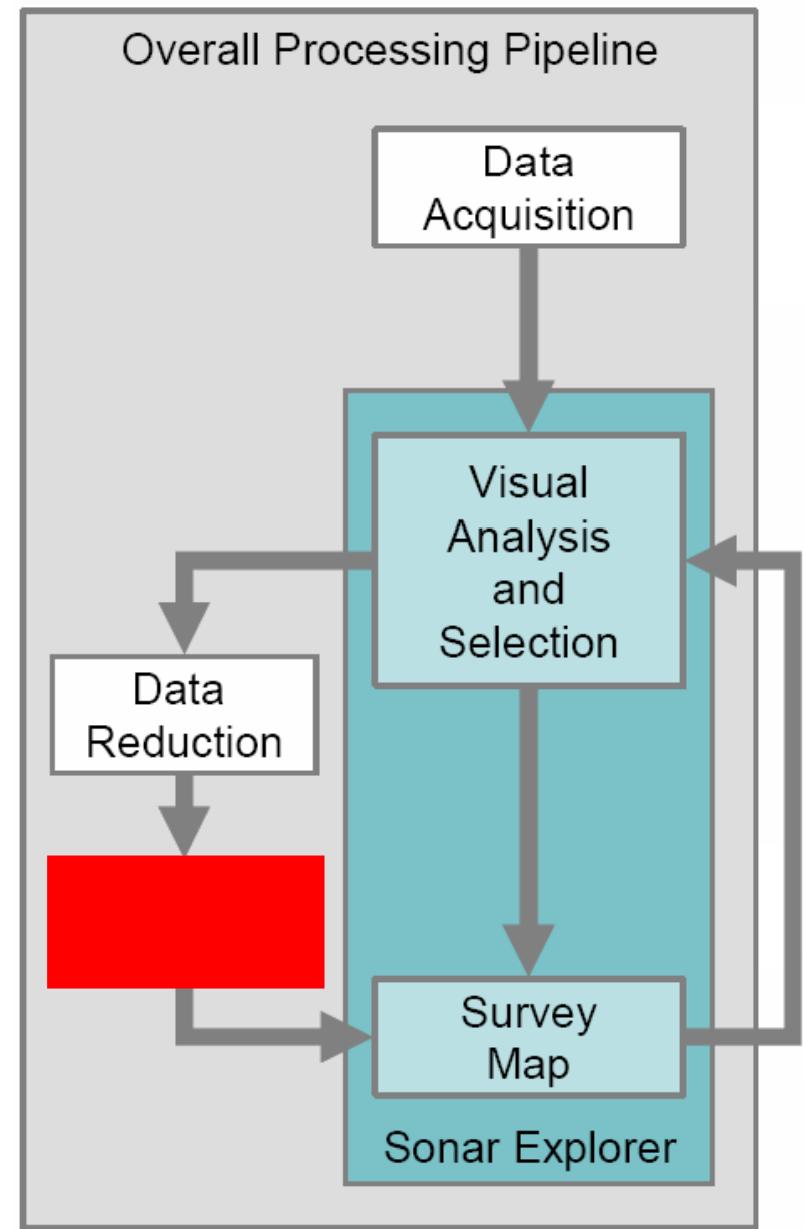
Visual Computing – Sonar Explorer (2)

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



[Balabanian et al. 2007]

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Challenges in Visualization

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



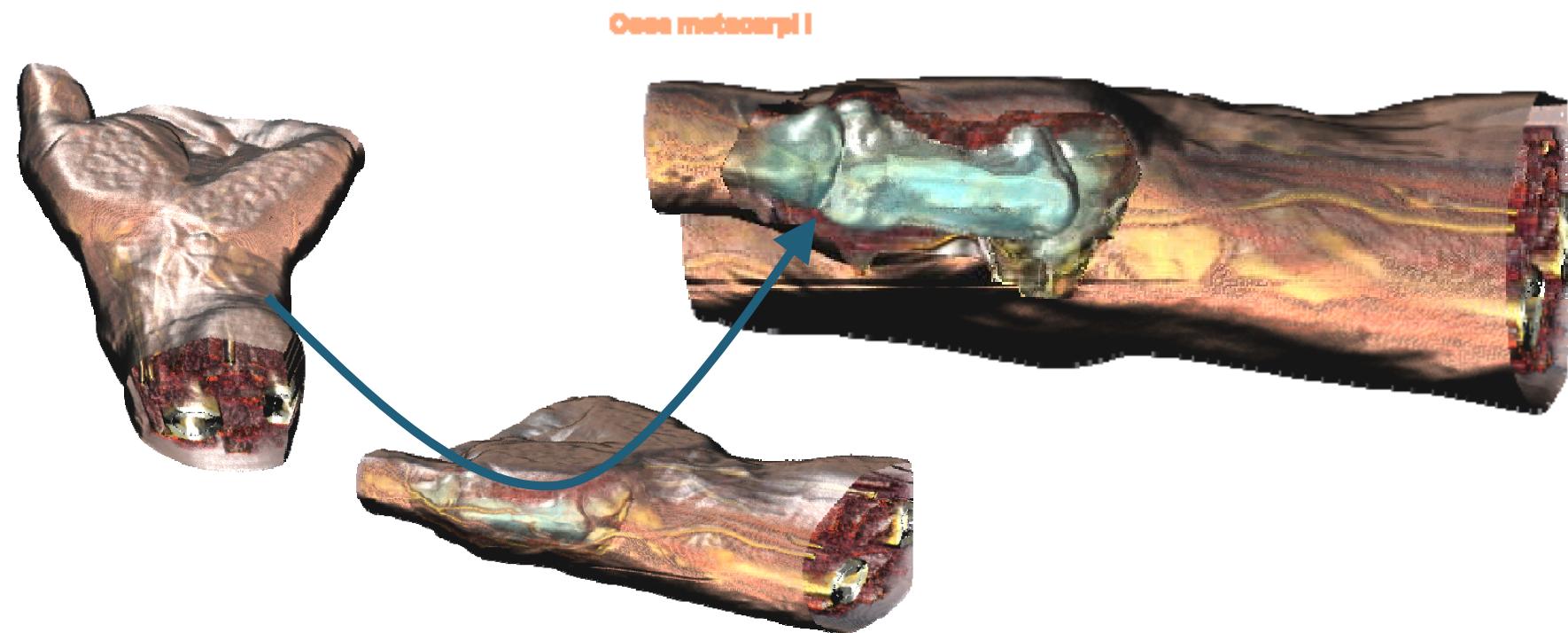
“Utilize knowledge and information derived from the process of scientific visualization or from abstract data analysis”

- Challenges
 - ◆ Metadata visualization
 - ◆ Visualization enabled by
 - topological information of the data
 - statistical information of the data
 - geometric information of the data
 - semantic information of the data
 - ◆ Visualization via learning
 - ◆ Visualization via shared knowledge in a collaborative setting
 - ◆ Knowledge representation for visualization
- Example
 - ◆ Automatic view point selection
 - ◆ Automatic reporting

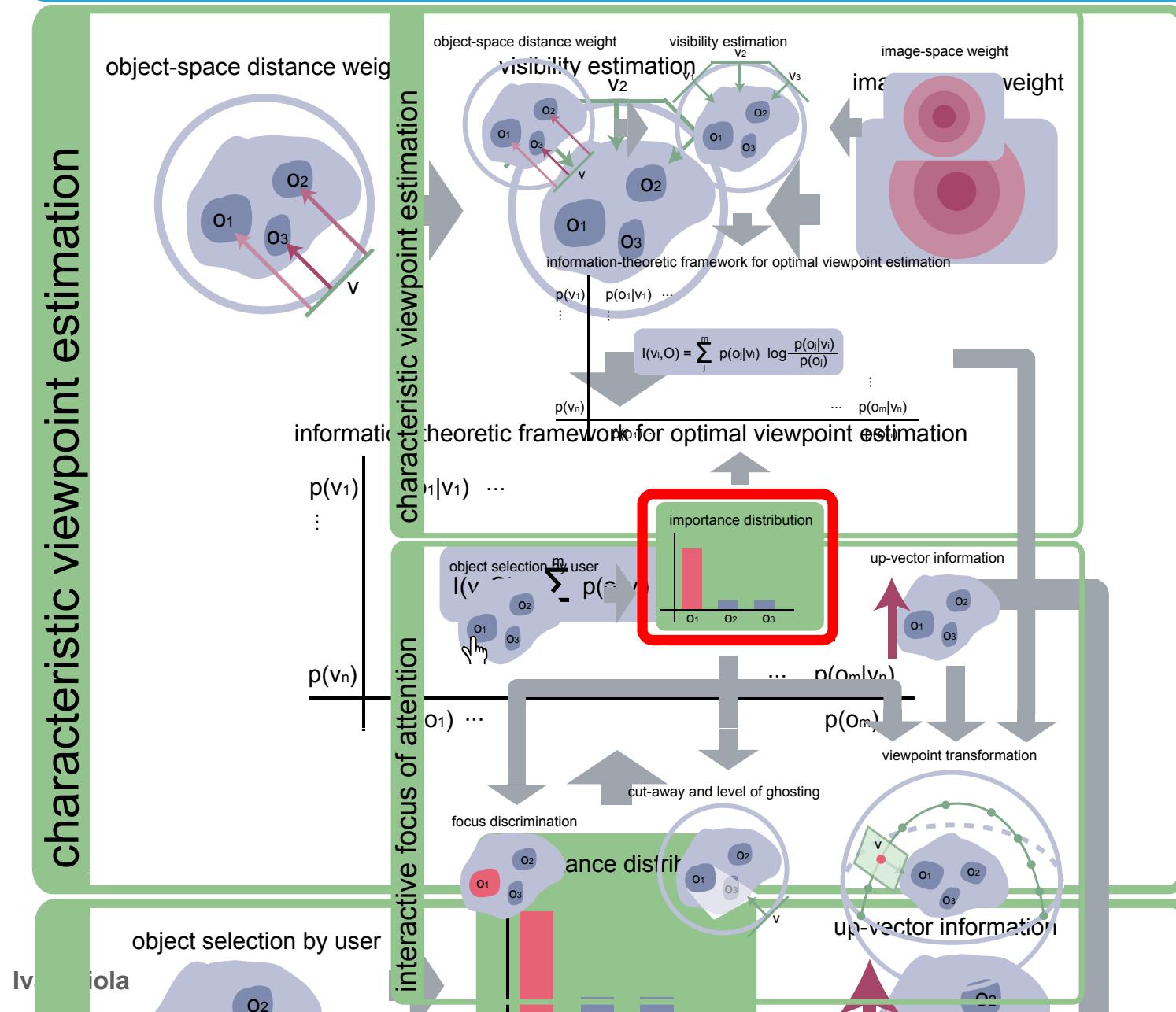


- Guided navigation between characteristic views

[Viola et al. 2006]



KAU - Importance-Driven Focus of Attention (2)



Challenges in Visualization

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

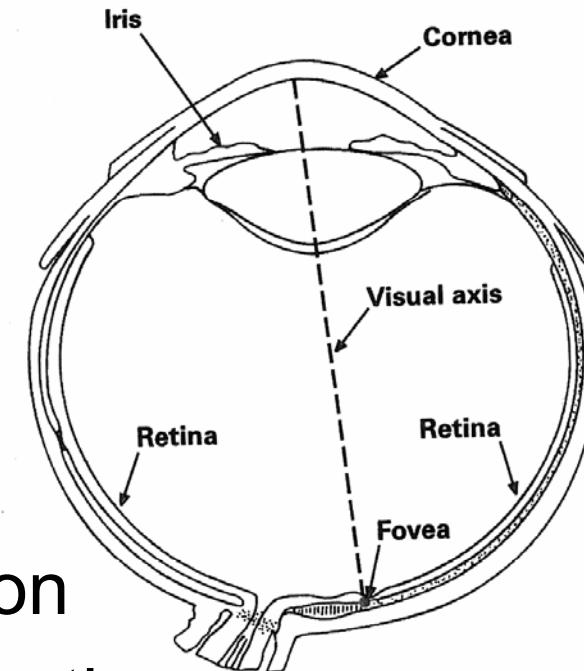


Basic idea of Focus+Context Visualization:

- Important regions in great detail (focus)
- Global view with reduced detail (context)
- Dynamic integration

Rationale

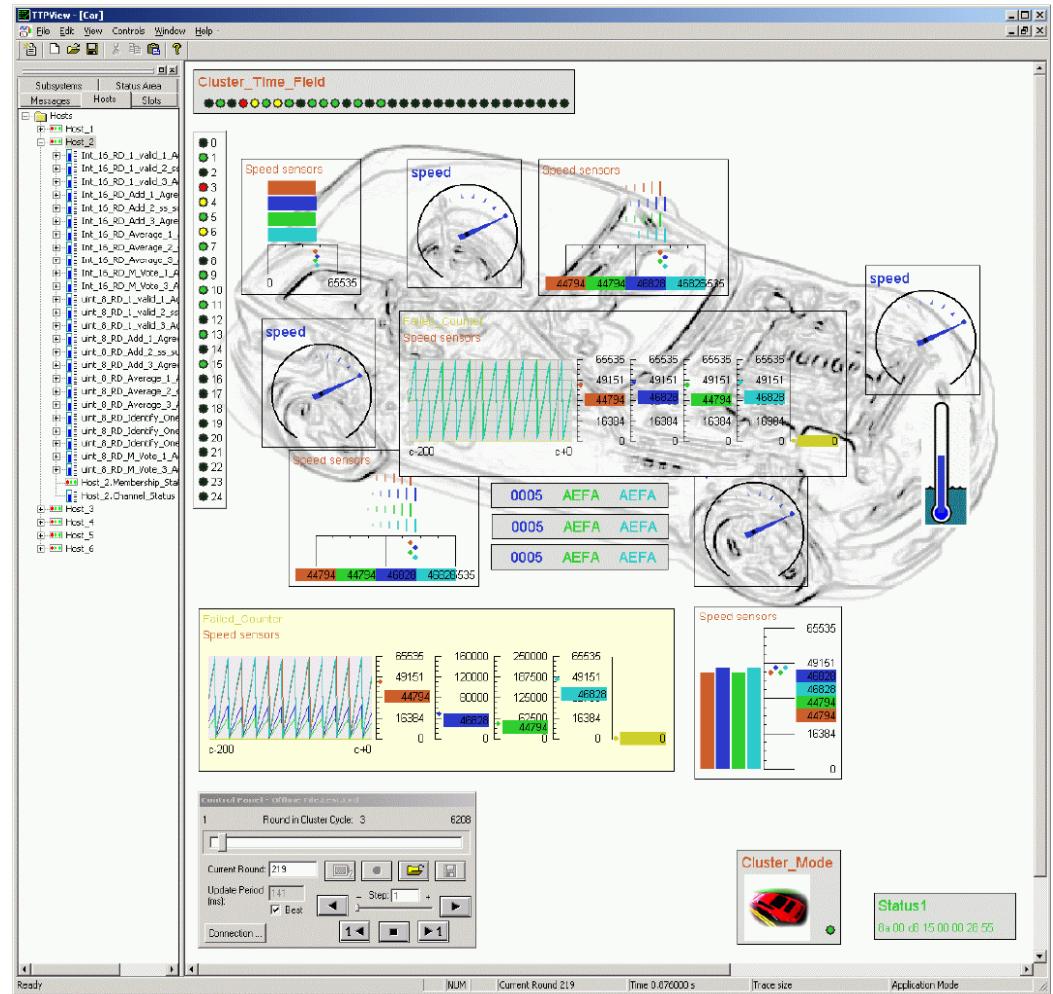
- Zooming hides the context
- Two separate displays split attention
- Human vision has both fovea and retina



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

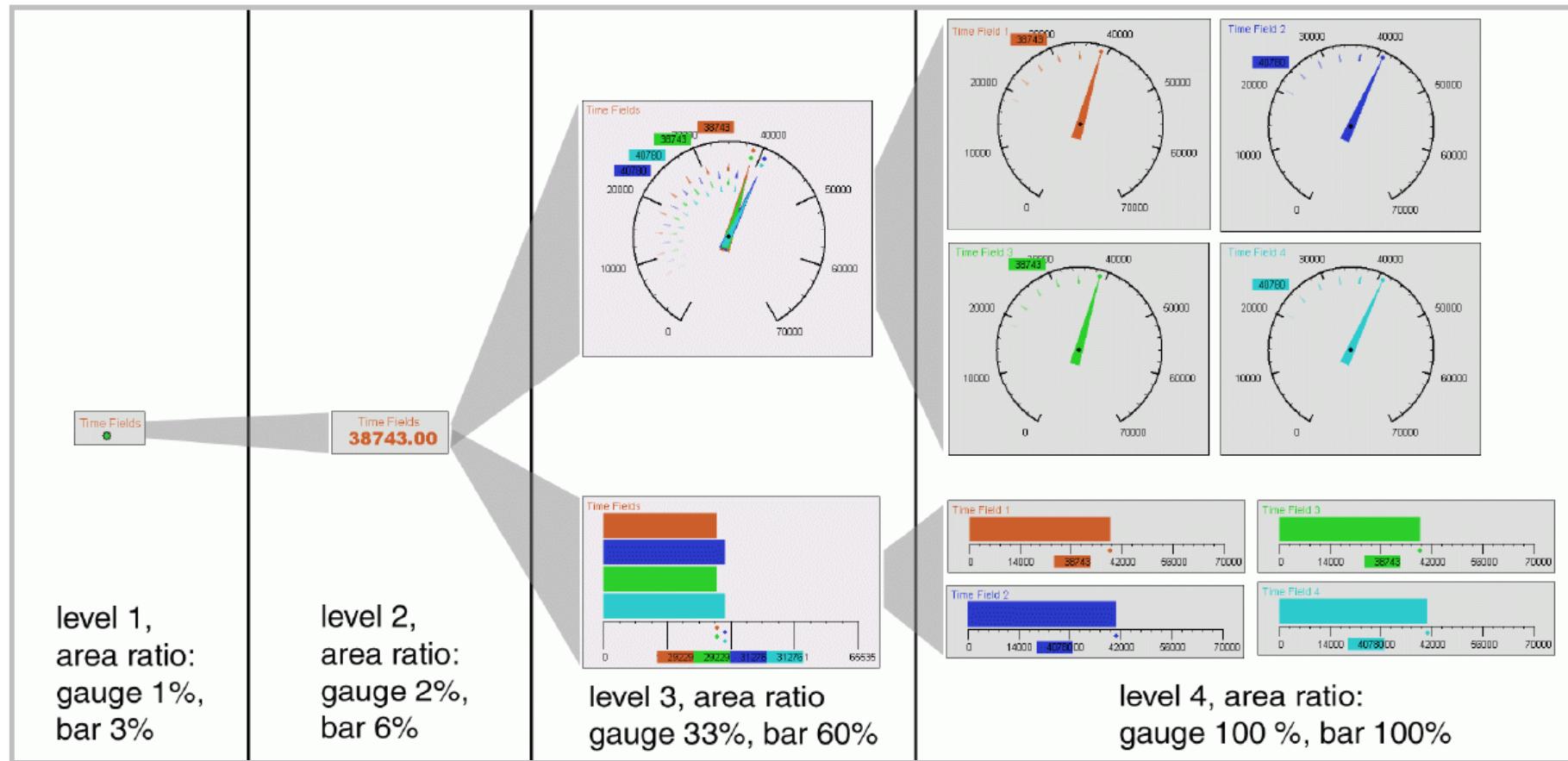


[Matković et al. 2002]



Scalability - Process Visualization (2)

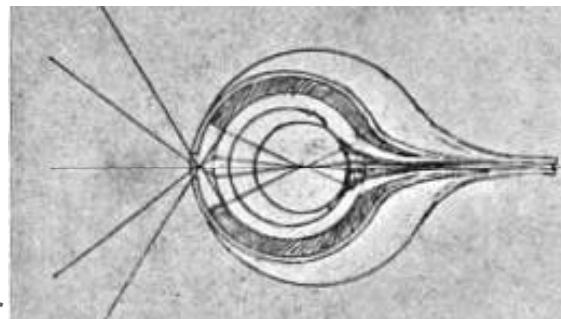
- Various instruments can be used to construct Levels of Detail (LODs)



Scalability - Illustration

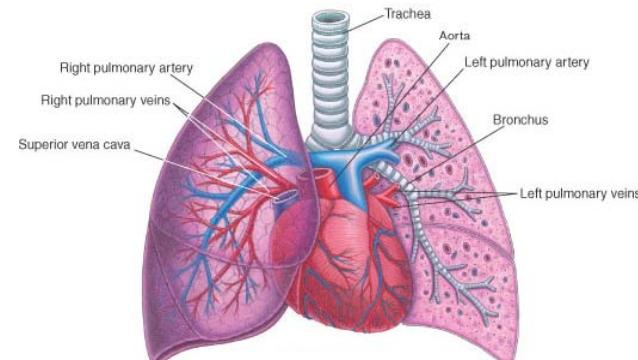
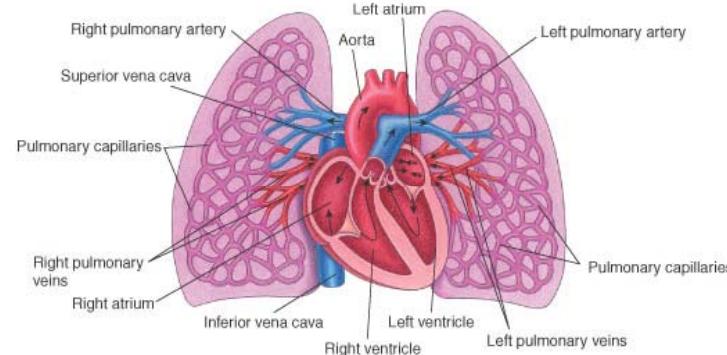
- An illustration is a picture with a communicative intent
- Conveys complex structures or procedures in an easily understandable way
- Uses abstraction to prevent visual overload – allows to focus on the essential parts
- Abstraction is visualized through distinct stylistic choices

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



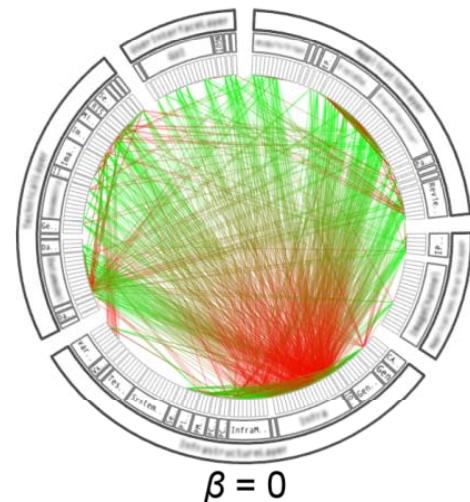
„As detailed as necessary - as simple as possible“



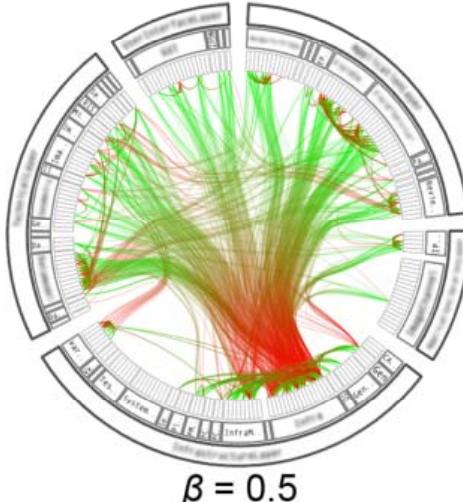
Scalability – Illustration Examples

■ Hierarchical Edge Bundles

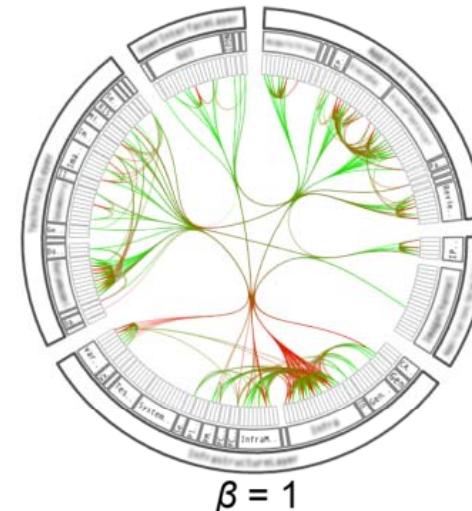
[Holten 2006]



$\beta = 0$

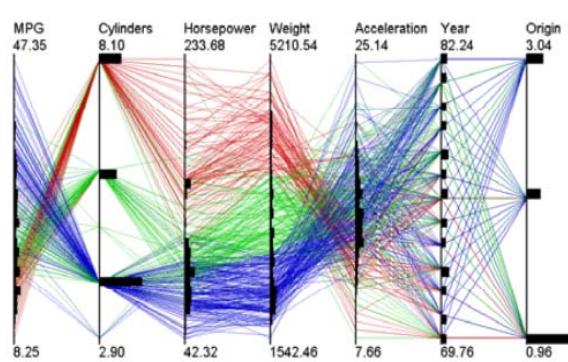


$\beta = 0.5$

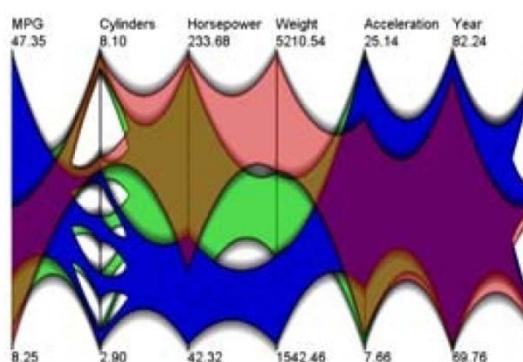


$\beta = 1$

■ Illustrative Parallel Coordinates



Eduard Gröller



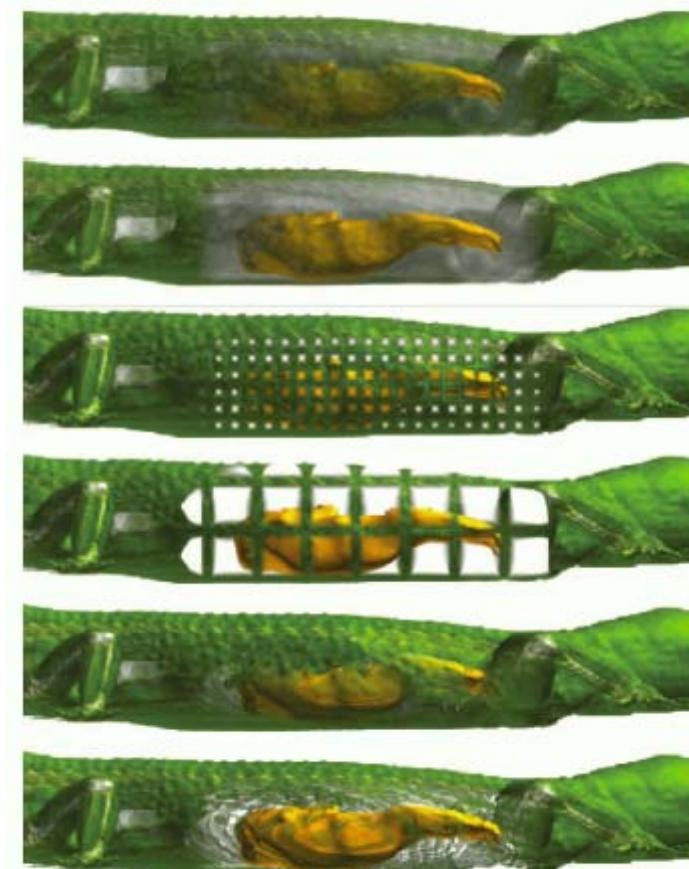
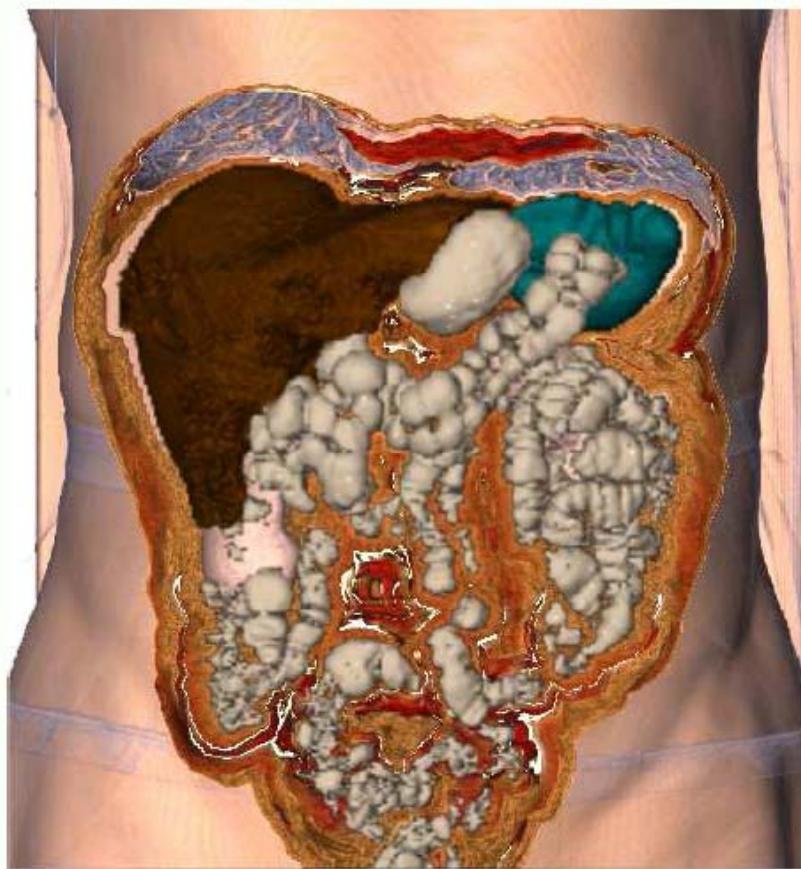
[McDonnell, Mueller 2008]



Scalability – Smart Visibility (1)

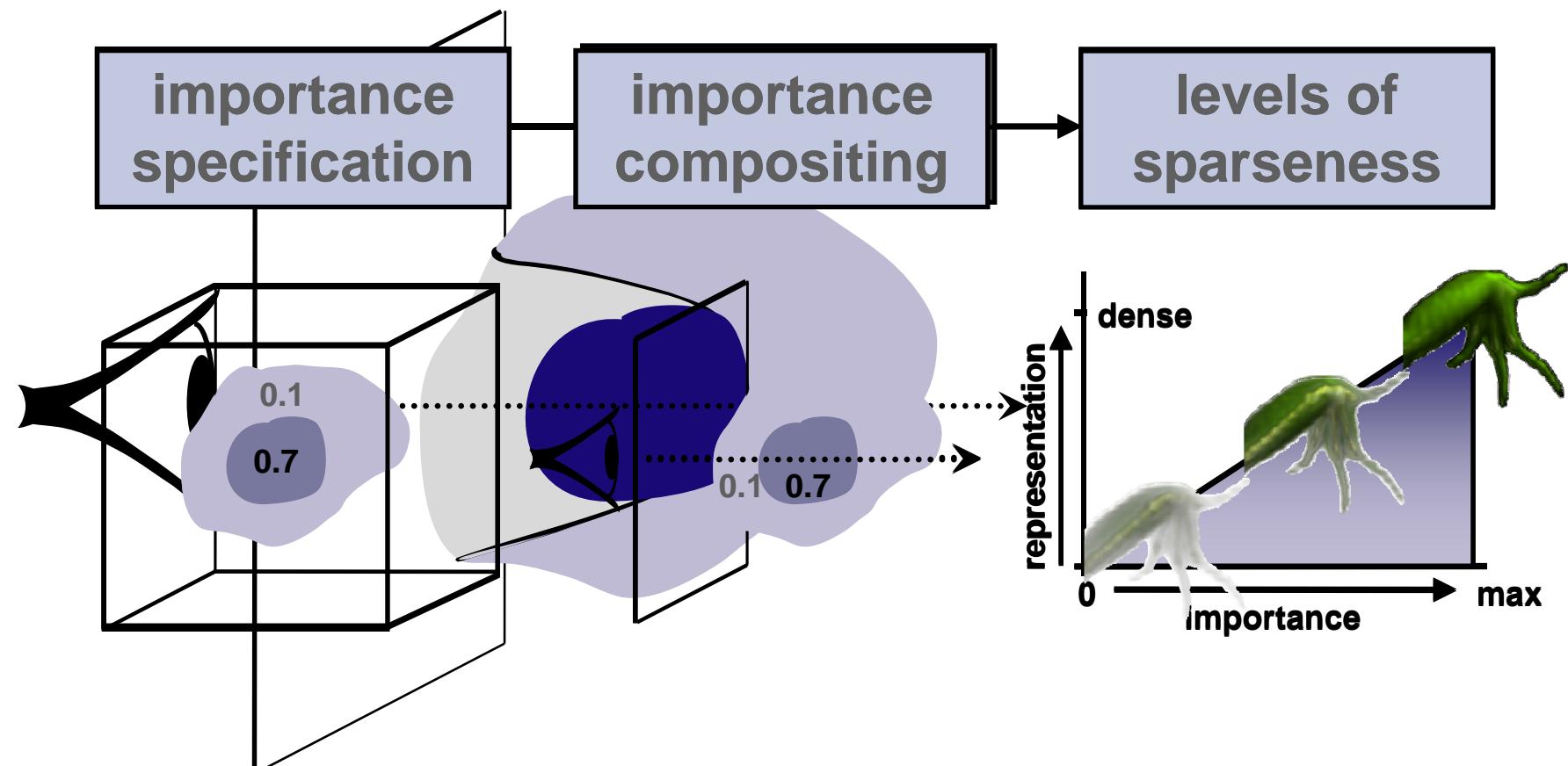
■ Importance-driven feature enhancement

[Viola et al. 2004, 2005]



Scalability – Smart Visibility (2)

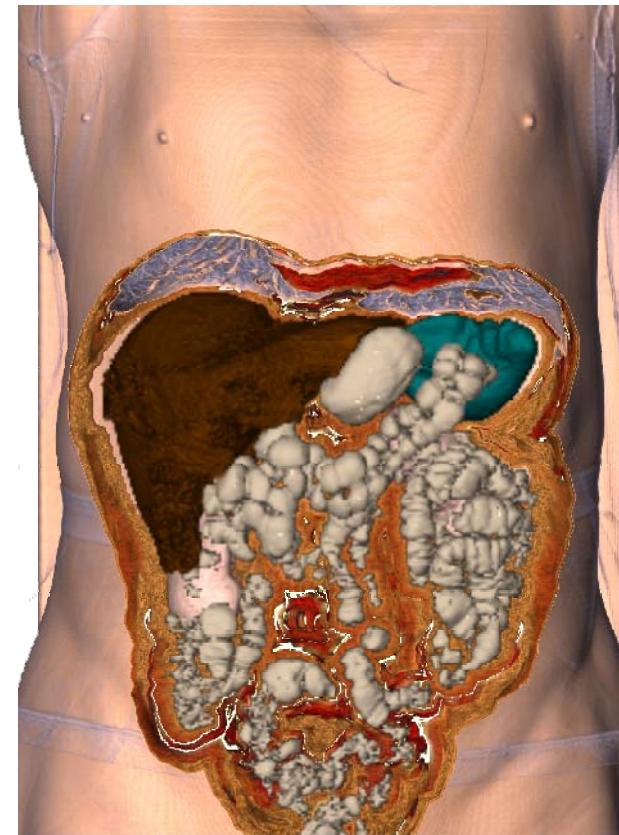
[Viola et al. '04 '05]



importance-driven feature enhancement



Scalability – Smart Visibility (3)



Scalability

**Do not fight complexity
with complexity**

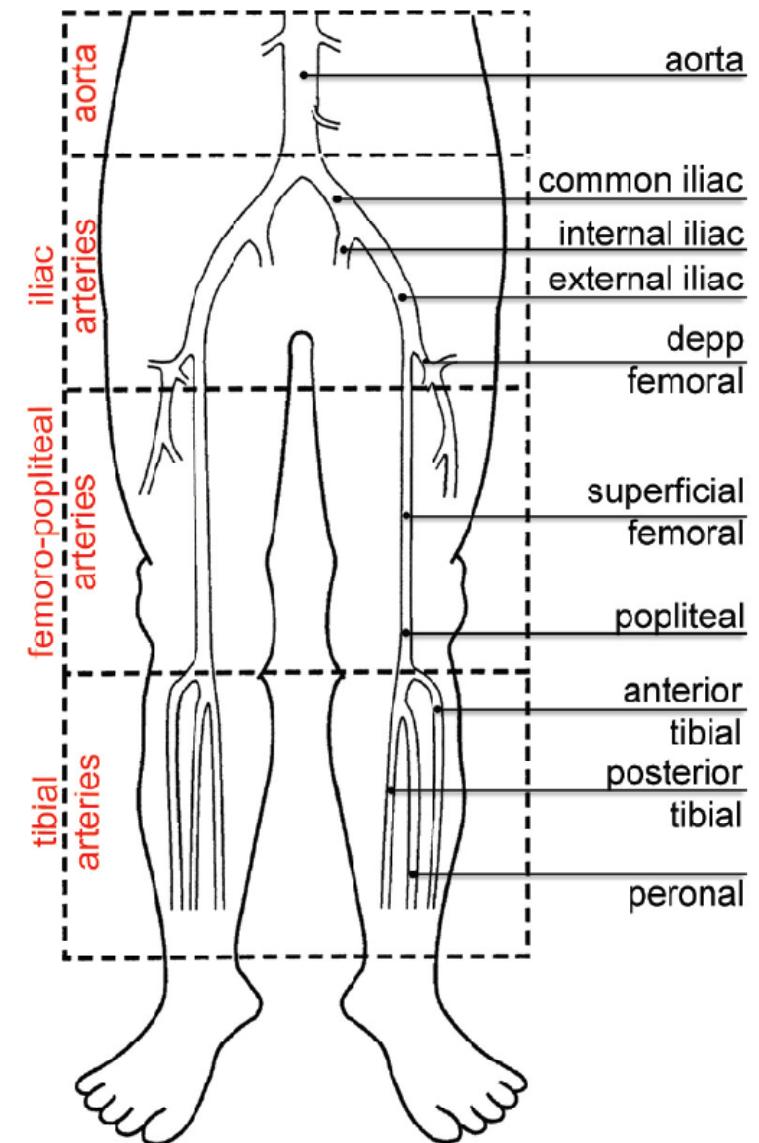
Challenges in Visualization

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

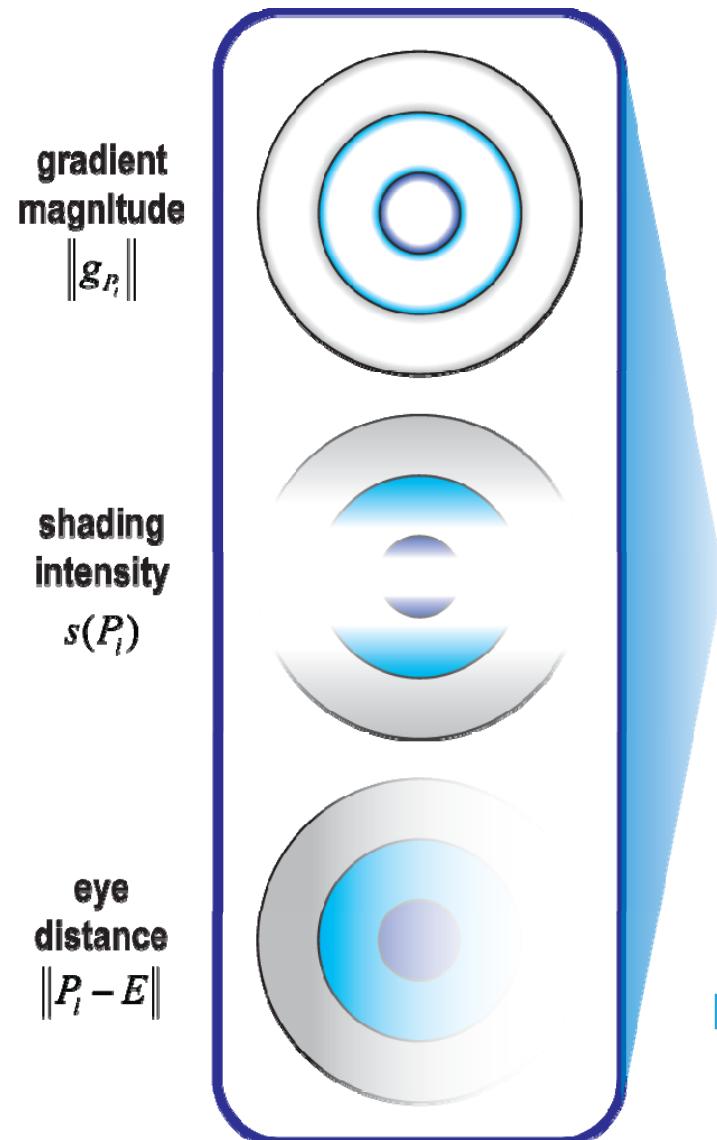


Visualization Yes ! – Interaction No ?

- 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

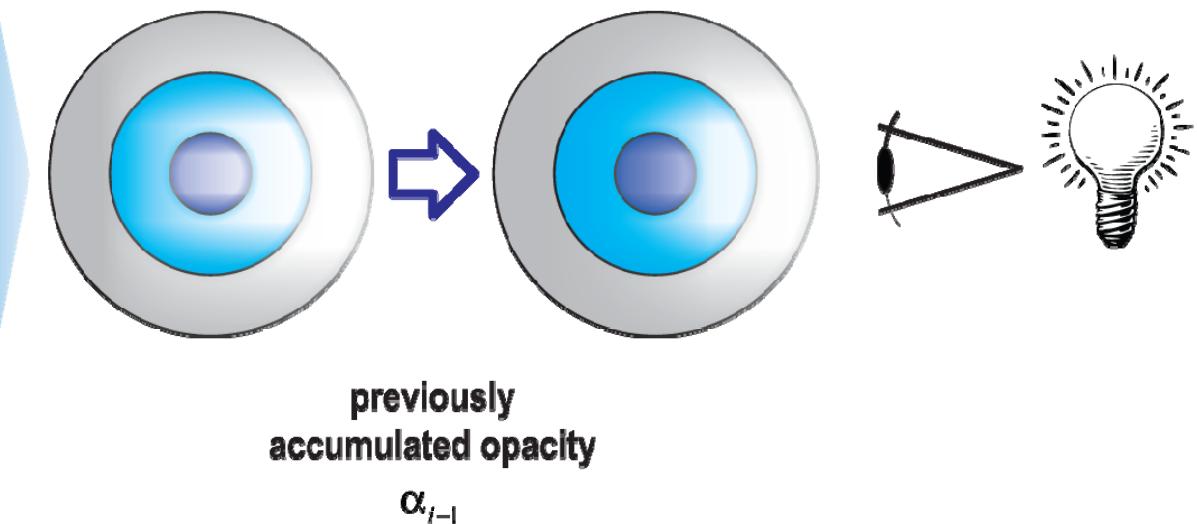


Context-Preserving Rendering (1)



[Bruckner et al. 2006]

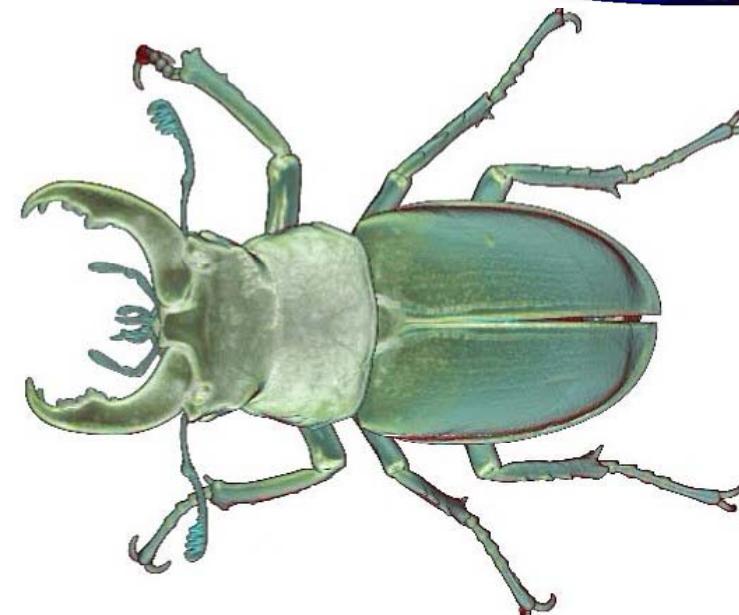
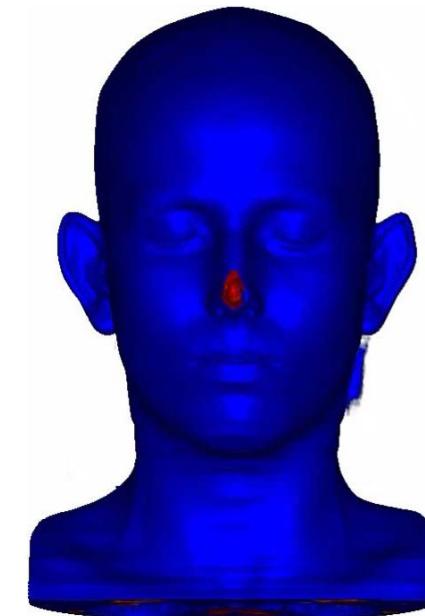
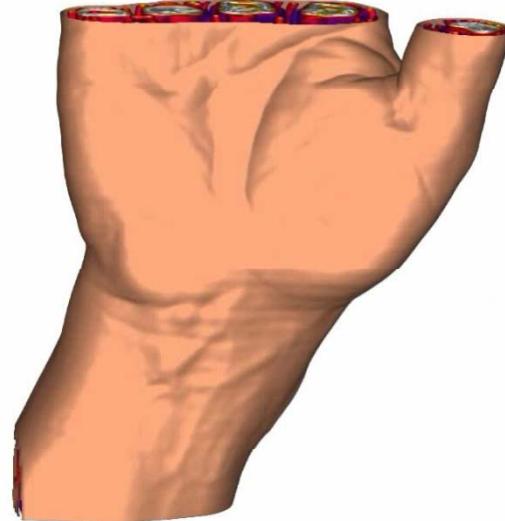
$$m(P_i) = \|g_{P_i}\|^{(\kappa_t \cdot s(P_i) \cdot (1 - \|P_i - E\|) \cdot (1 - \alpha_{i-1}))^{\kappa_s}}$$



- Integrate various focus+context approaches with only few parameters



Context-Preserving Rendering (2)



S. Bruckner and E. Gröller



Visualization Yes ! – Interaction No ?



Provide the user with the flexibility
he/she can cope with –
avoid the terror of unconstrained liberty

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- Interaction Yes ! – BUT User centric !

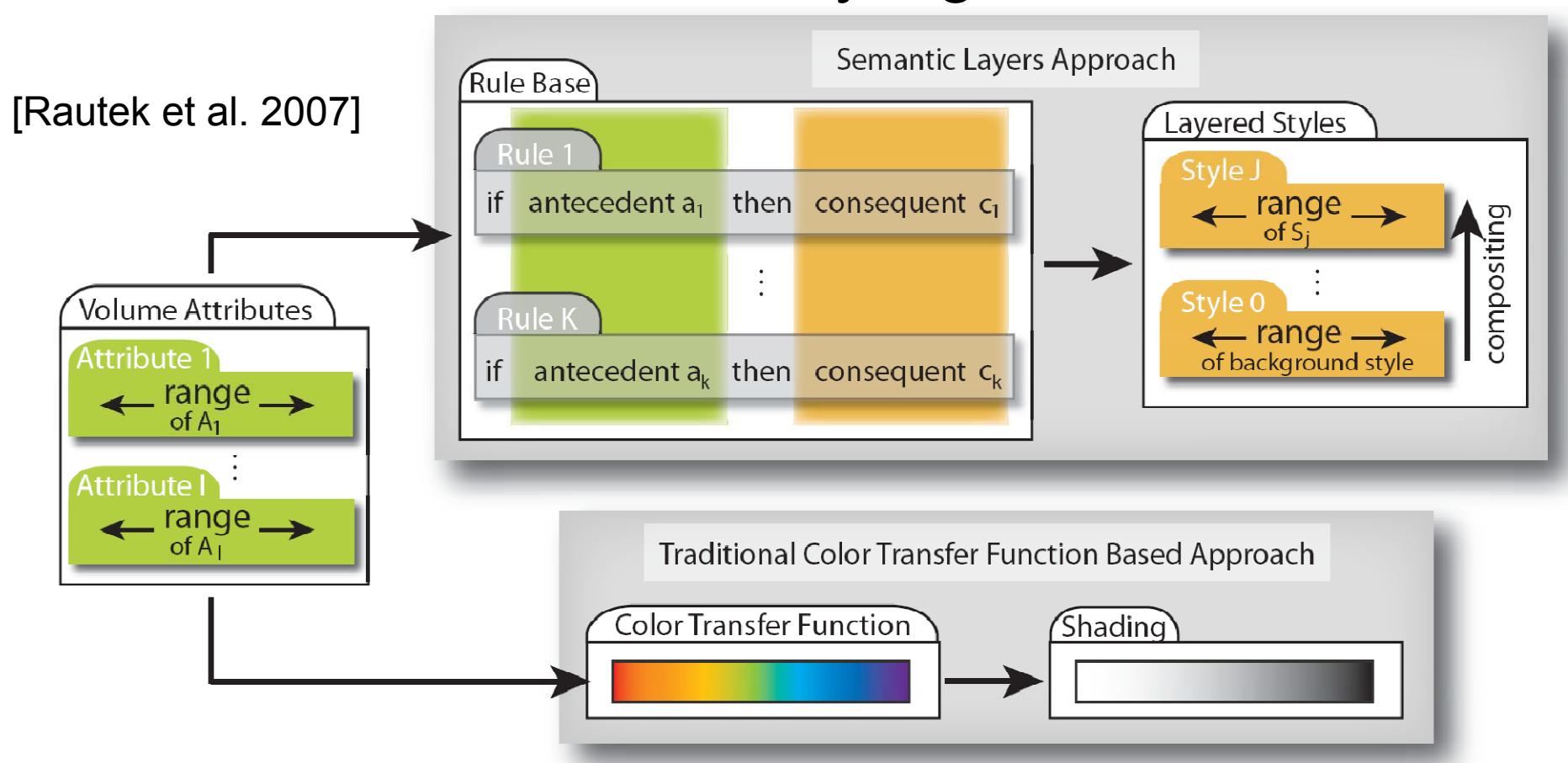


- Problems
 - ◆ Medical doctors do not (want to) know transfer functions
 - ◆ Complex 3D interaction is complex
- Challenges
 - ◆ Include user model (novice, experienced, expert)
 - ◆ Include motifs
 - ◆ Include user preferences
 - ◆ 2D+ navigation (instead of 3D navigation)
- Examples
 - ◆ Semantic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



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

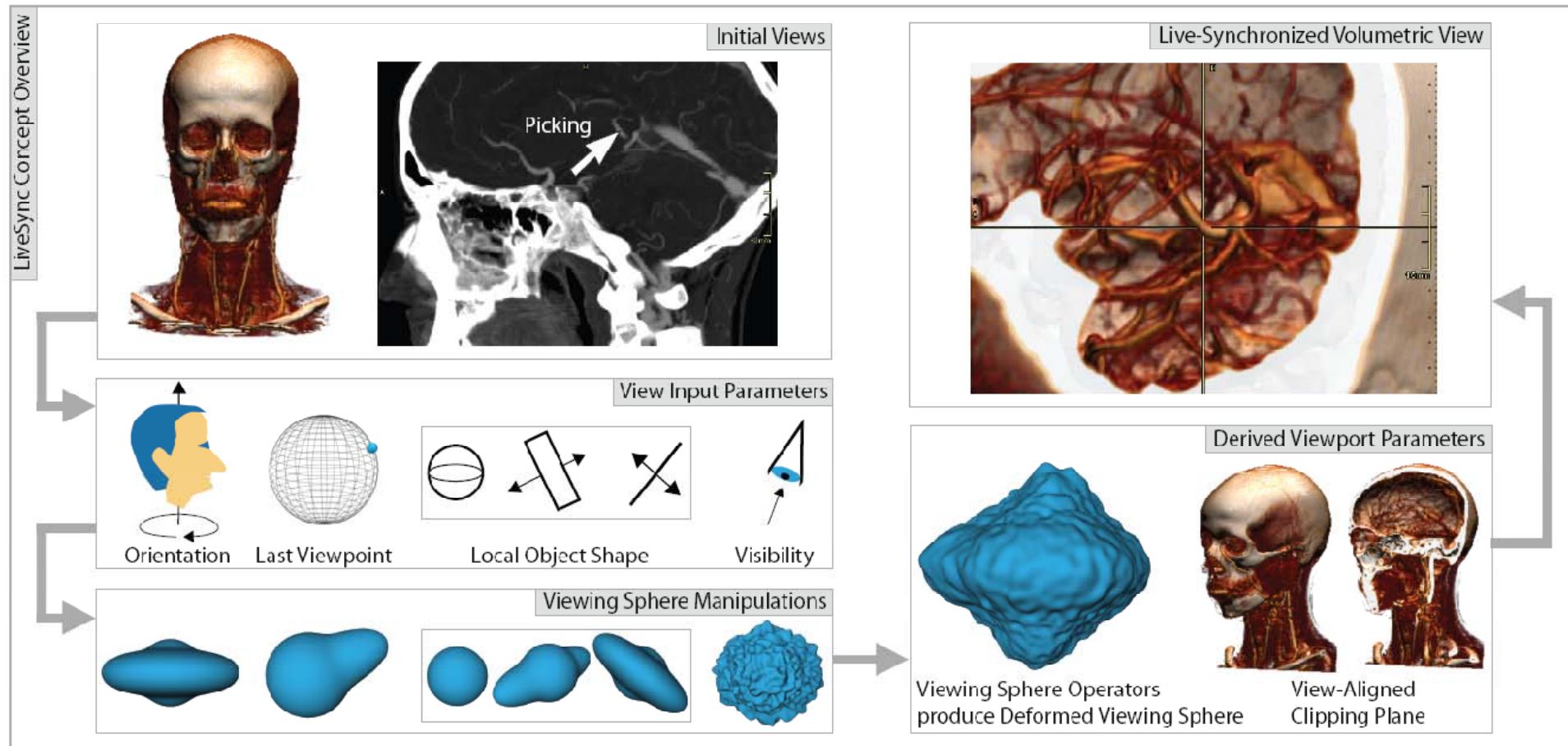


**Die Grenzen meiner Sprache bedeuten
die Grenzen meiner Welt**

[Ludwig Wittgenstein]

Knowledge-Based Navigation

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



[Kohlmann et al. 2007]

Video

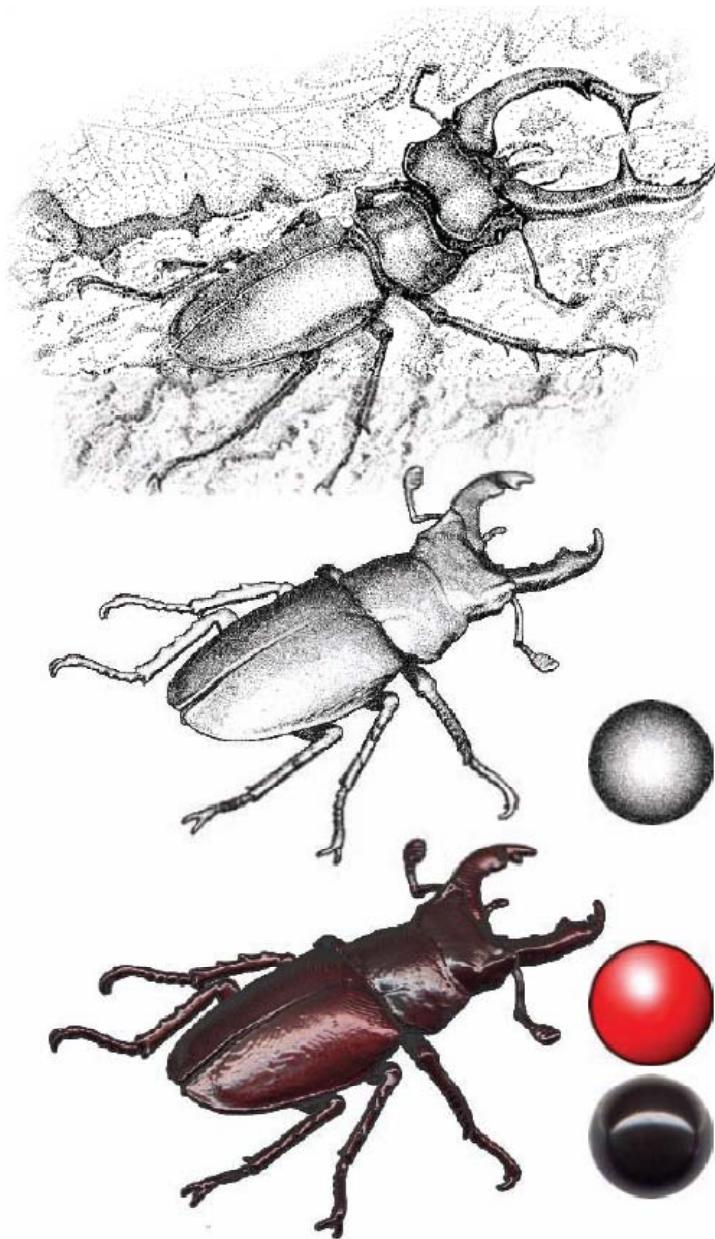


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Bring **visualization** into the workflow of users!!



Thank You for Your Attention



Questions ?
Comments?

Acknowledgments

Jean-Paul Balabanian
Stefan Bruckner
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Helwig Hauser
Christoph Heinzl
Krešimir Matković
Hannes Obweger
Peter Rautek
Martin Suntinger,
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