

# Interactive Visual Analysis of Rich Scientific Data

Helwig Hauser  
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HH, [ii.UiB.no/vis](http://ii.UiB.no/vis)



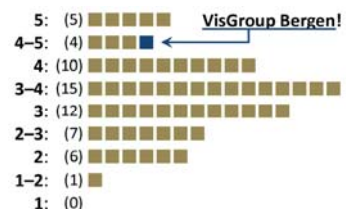
HH: prof. in visualization (vis)  
@ Dept. of Informatics (ii)  
@ Univ. of Bergen (UiB)

## UiB VisGroup

- 2007: group of 3:
- 2009: larger projects start
- 2011: EuroVis in Bergen



- 2013: new prof.:



[ranking from NFR's 10-year evaluation in 2011/2012]



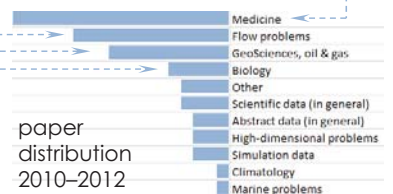
## ➤ Application-oriented basic research in visualization:

### 1. Researched visualization methodology (how to visualize)

- **Interactive Visual Analysis,  $nD$  data** (H. Hauser *et al.*)
- **Visual Knowledge Discovery, 3D data** (St. Bruckner *et al.*)
- **Illustrative Visualization** (I. Viola *et al.*)

### 2. Applications at which this research is oriented (for whom)

- **Medical Visualization** (partner in MedViz Bergen, *etc.*)
- **GeoSciences / Oil & Gas** (e.g., financed by Statoil's Akademiaavtale)
- **Biology / Bioinformatics** (with CBU@ii *et al.*)
- **Fluid Dynamics** (in collab. with FFI.no, for ex.)
- **Engineering** (visual analysis of simulation data)



## ii.UiB.no/vis PhDs (11 so far)

	Daniel Patel (Oct. 2009): Expressive Vis. & Rapid Interpr. of Seismic Volumes	
	Jean-Paul Balabanian (Jan. 2010): Multi-Aspect Vis.: from Linked to Integrated Views	
	Johannes Kehrer (May 2011): IVA of Multi-faceted Scientific Data	
	Ove Daae Lampe (Nov. 2011): IVA of Process Data	
	Armin Pobitzer (June 2012): IVA of Time-dependent Flows	
	Paolo Angelelli (June 2012): Visual Expl. of Human Physiology	
	Veronika Šoltészová (Oct. 2012): Perception-Augmenting Illumination	
	Åsmund Birkeland (May 2013): Ultrasonic Vessel Vis.: From Extraction to Perception	
	Endre Lidal (May 2013): Sketch-based Storytelling for Cognitive Problem Solving	
	Çağatay Turkey (Nov. 2013): Interactive Visual Analysis of High-dimensional Data	
	Mattia Natali (Sept. 2014): Sketch-based Mod. & Conceptual Vis. of Geomorphological Processes for ...	

# Interactive Visual Analysis

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## Interactive Visual Analysis (IVA)

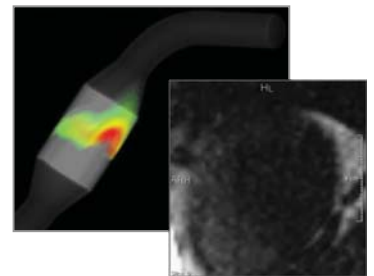


- Given data –  
*too much and/or too complex to be shown at once:*
- IVA is an **interactive visualization approach** to facilitate
  - the **exploration** and/or the **analysis** of data (not necessarily the presentation of data), including
    - **hypothesis generation & evaluation, sense making,**
    - **knowledge crystallization, etc.**
  - according to the **user's interest/task**, *for ex.*, by interactive feature extraction,
  - navigating between **overview** and **details**, *e.g.*, to enable interactive information drill-down [Shneiderman]
- through an **iterative & interactive visual dialog**

# Interactive Visual Analysis ↔ Visual Analytics



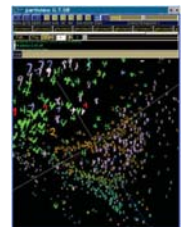
- **IVA** (“interactive visual analysis”) **since 2000**
- **Tightly related to visual analytics**, of course, *e.g., integrating computational & interactive data analysis*
- A **particular methodology** with specific components (*CMV, linking & brushing, F+C vis., etc.*)
- General enough to work in **many application fields**, but not primarily the VA fields (*national security, etc.*), in particular **“scientific data” fields...**



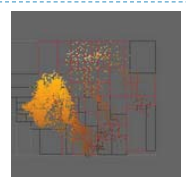
## Integrating Interaction & Computation



- **Goal:** to combine the *best of two worlds* [Keim et al.]:
  - data **exploration/analysis** by the **user**, based on **interactive visualization**
  - and **data analysis** by the **computer**, based on **statistics, machine learning, etc.**
- State of the art / **levels of integration**:
  - **mostly no integration**, still
  - some **vis. of results** of computations
  - also: making **comp. semi-interactive** (here called “**inner integration**”)
  - **rare: tight integration**
- **Outer integration** (here!): bundling **interaction & computation in a loop**



[Maniayar & Nabney, 2006]



[Williams & Munzner, 2004]

# Target Data Model: “Scientific Data”



- **Characterized** by a combination of
  - **independent variables**, like **space** and/or **time** (cf. **domain**)
  - and **dependent variables**, like **pressure**, **temp.**, etc. (cf. **range**)
- So we can think of this type of data as **given as  $d(\mathbf{x})$**  with  $\mathbf{x} \leftrightarrow$  **domain** and  $d \leftrightarrow$  **range** – examples:
  - **CT data**  $d(\mathbf{x})$  with  $\mathbf{x} \in \mathbb{R}^3$  and  $d \in \mathbb{R}$
  - **unstead 2D flow**  $\mathbf{v}(\mathbf{x}, t)$  with  $\mathbf{x} \in \mathbb{R}^2$ ,  $t \in \mathbb{R}$ , and  $\mathbf{v} \in \mathbb{R}^2$
  - **num. sim. result**  $\mathbf{d}(\mathbf{x}, t)$  with  $\mathbf{x} \in \mathbb{R}^3$ ,  $t \in \mathbb{R}$ , and  $\mathbf{d} \in \mathbb{R}^n$
  - **system sim.**  $\mathbf{q}(\mathbf{p})$  with  $\mathbf{p} \in \mathbb{R}^n$  and  $\mathbf{q} \in \mathbb{R}^m$
- **Common property:**
  - $d$  is (at least to a certain degree) **continuous** wrt.  $\mathbf{x}$

## Interactive Visual Analysis of Scientific Data



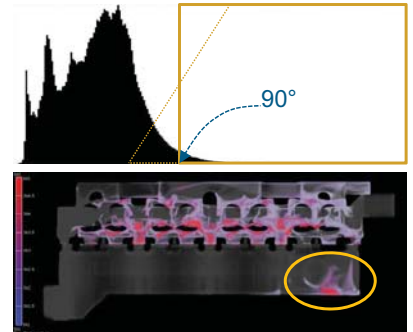
- **Interactive visual analysis** (as exemplified in this tutorial) **works really well with scientific data**, e.g.,
  - results from **numerical simulation** (spatiotemporal)
  - imaging / **measurements** (in particular multivariate)
  - sampled **models**
- When used to study scientific data, **IVA employs**
  - methods from **scientific visualization** (vol. rend., ...)
  - methods from **statistical graphics** (scatterplots, ...), **information visualization** (parallel coords., etc.)
  - **computational tools** (statistics, machine learning, ...)
- Applications include
  - **engineering, medicine, meteorology/climatology, biology, etc.**



# The Iterative Process of IVA

- Loop / bundling of *two complementary parts*:
  - **visualization** – *show to the user!*  
*Something new, or something due to interaction.*
  - **interaction** – *tell the computer!*  
*What is interesting? What to show next?*
- Basic example (**show – brush – show – ...**),  
cooling jacket context:

1. show a histogram of temperatures
2. brush high temperatures ( $>90^{\circ}[\pm 2^{\circ}]$ )
3. show focus+context vis. in 3D
4. locate relevant feature(s)

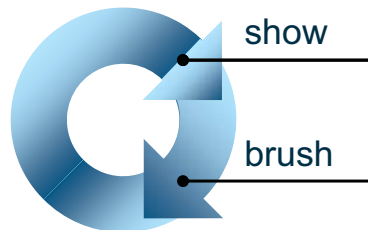


- **KISS-principle IVA:**

- linking & brushing, focus+context visualization, ...

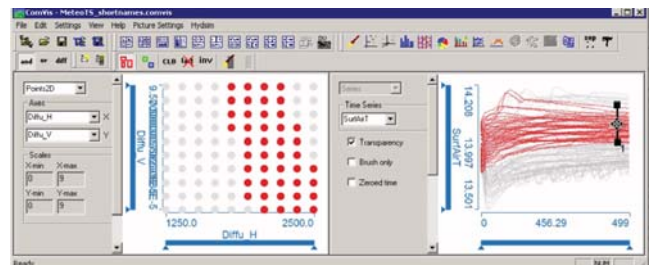
## ■ Tightest IVA loop

- **show data** (explicitly represented information)
- **one brush** (on one view, can work on >1 dims.)



## A typical (start into an) IVA session of this kind:

- bring up multiple views
  - at least one for  $x, t$
  - at least one for  $d_i$
- I see (something)!
- brush this “something”
- linked F+C visualization
- first insight!



## ■ Tightest IVA loop

- **show data** (explicitly represented information)
- **one brush** (on one view, can work on >1 dims.)

## ■ Requires:

- multiple views ( $\geq 2$ )
- interactive brushing capabilities on views (brushes should be editable)
- focus+context visualization
- linking between views

## A typical (start into an) IVA session of this kind:

- bring up multiple views
  - at least one for  $x, t$
  - at least one for  $d_i$
- I see (something)!
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- linked F+C visualization
- first insight!

... leads to ...

degree of interest

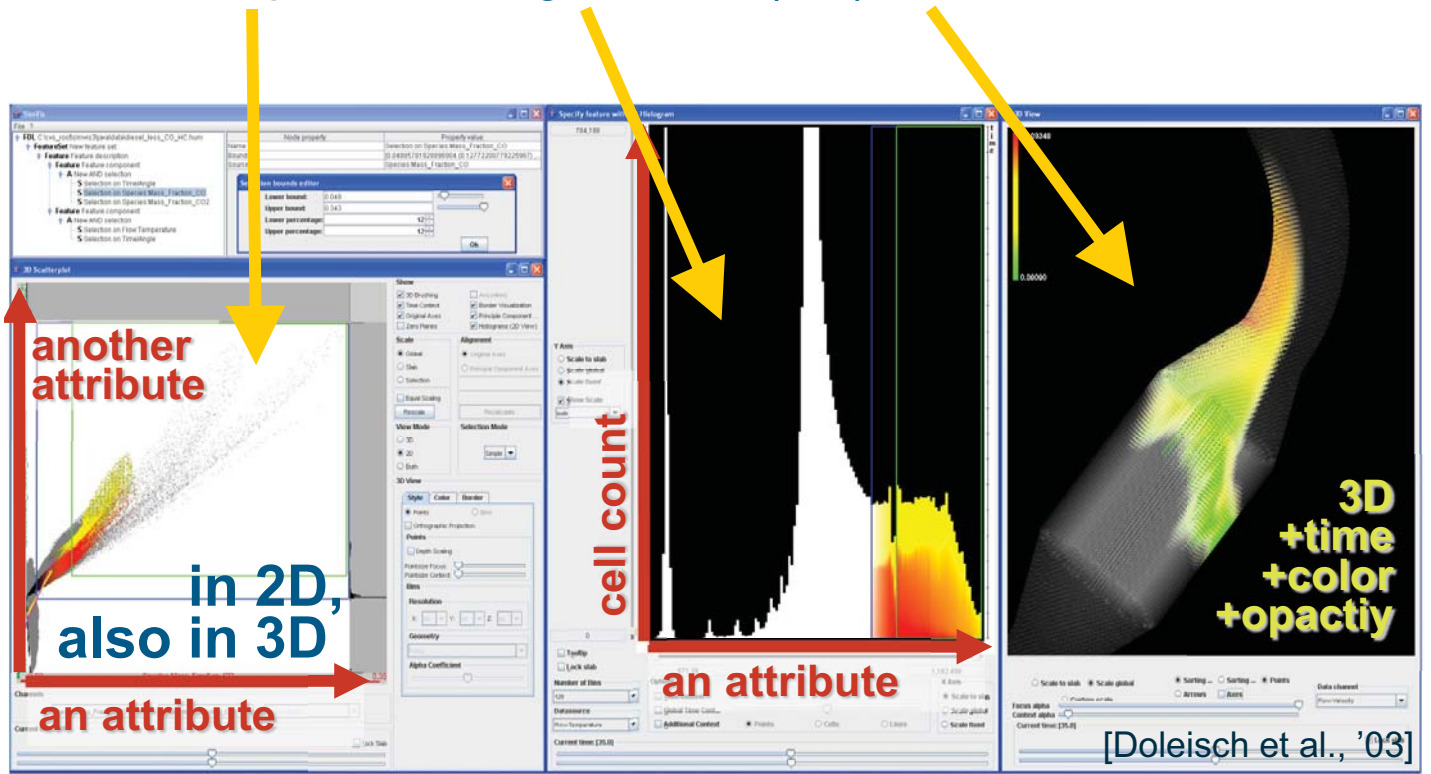
... requires ...

... is realized via ...

- Allows for **different IVA patterns** (wrt. domain & range)

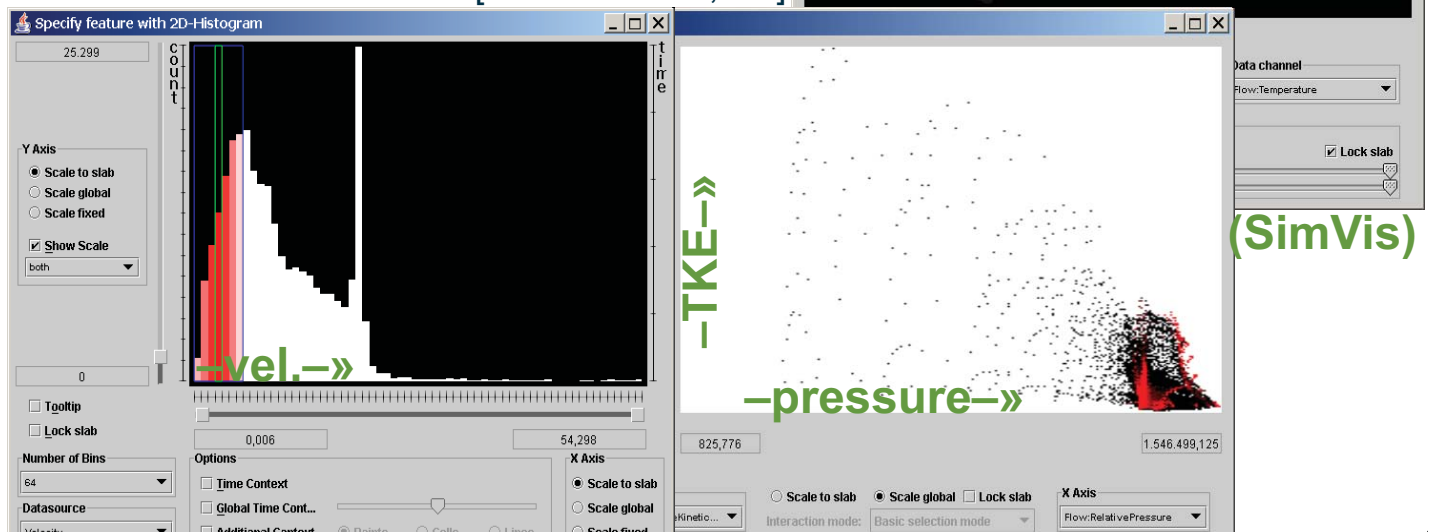
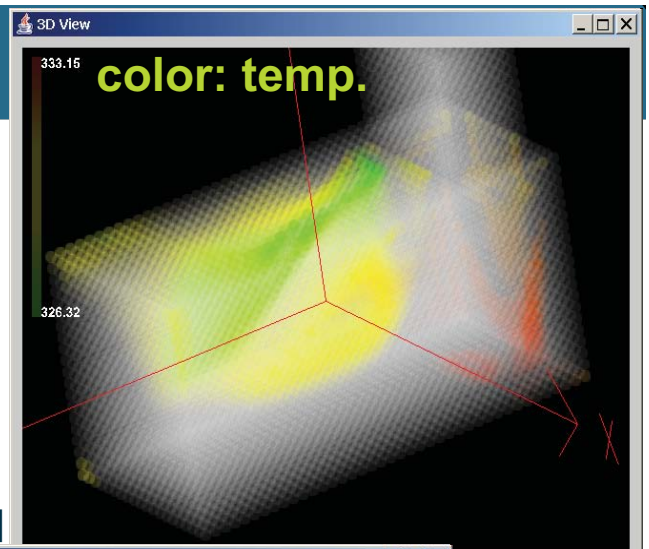
# IVA: Multiple Views

- One dataset, but multiple views
- Scatterplots, histogram, 3D(4D) view, etc.



# Interactive Brushing

- Move/alter/extend brush interactively
- Interactively explore/analyze multiple variates

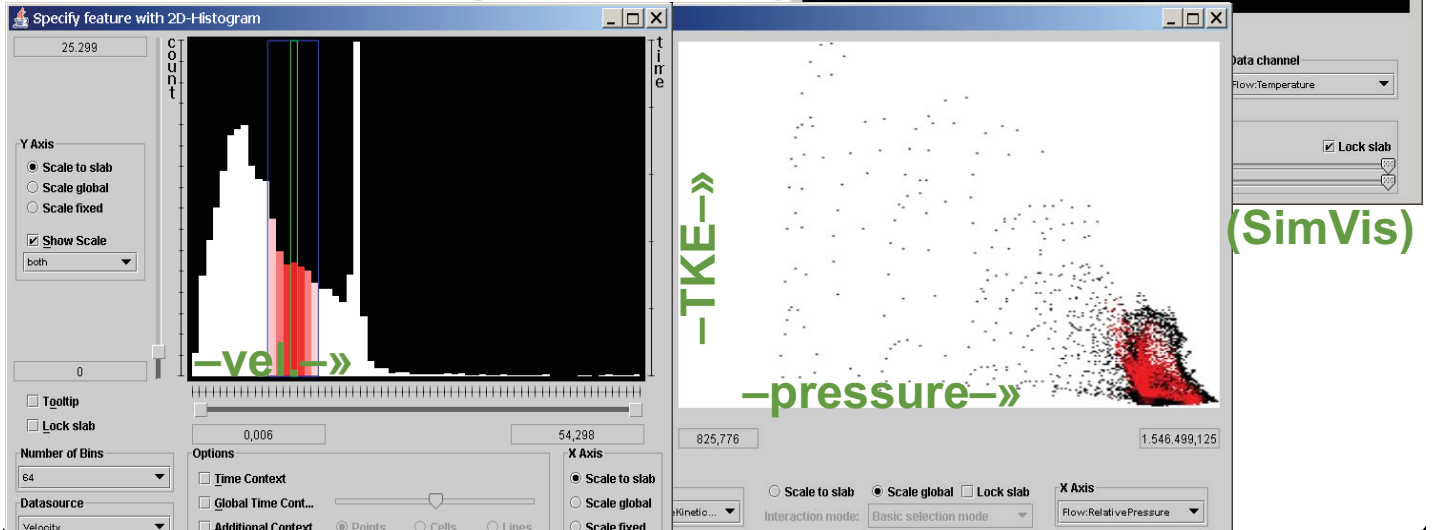
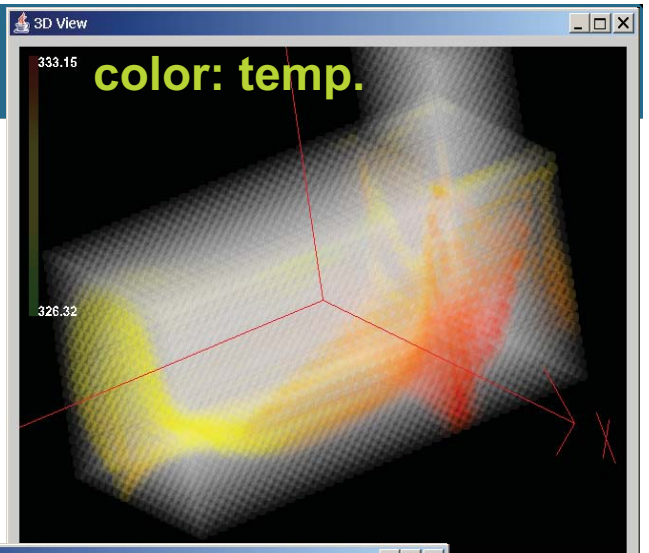




# Interactive Brushing

- Move/alter/extend brush interactively
- Interactively explore/analyze multiple variates

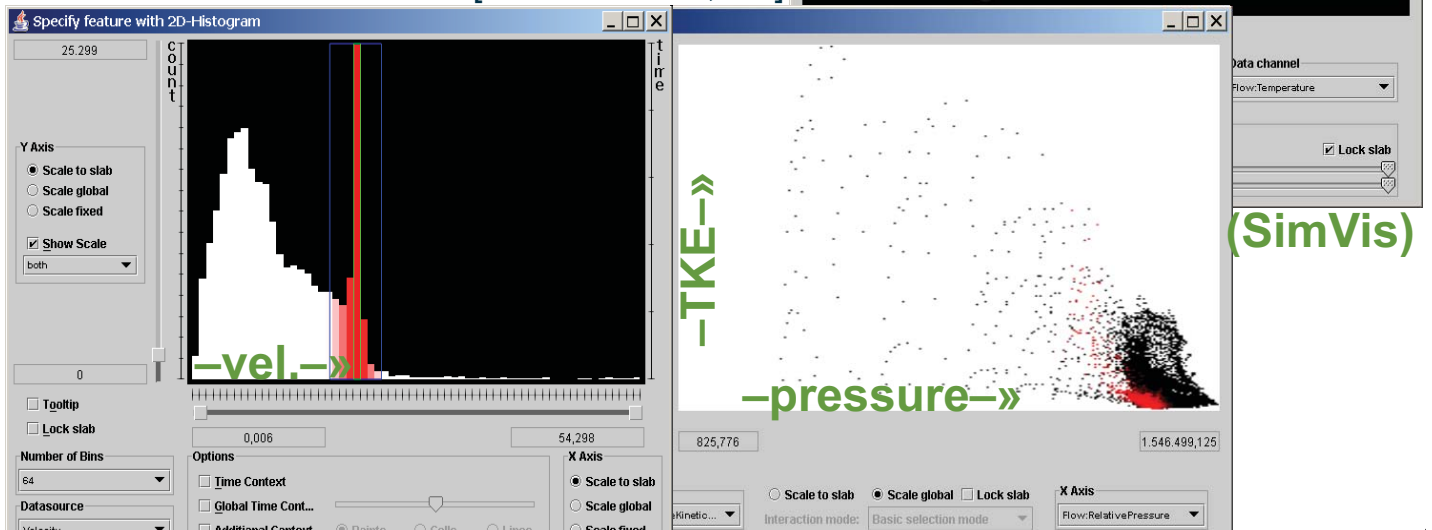
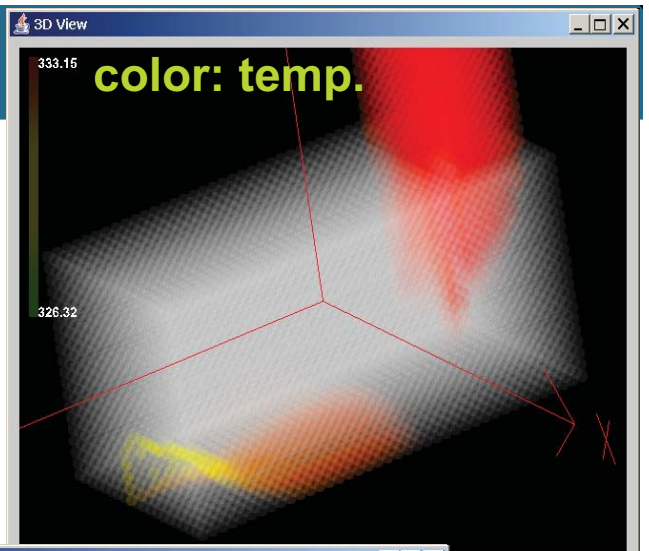
[Doleisch et al., '03]



# Interactive Brushing

- Move/alter/extend brush interactively
- Interactively explore/analyze multiple variates

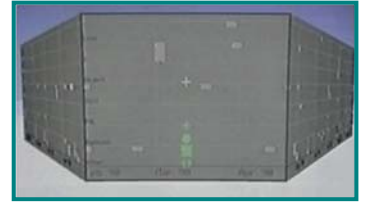
[Doleisch et al., '03]



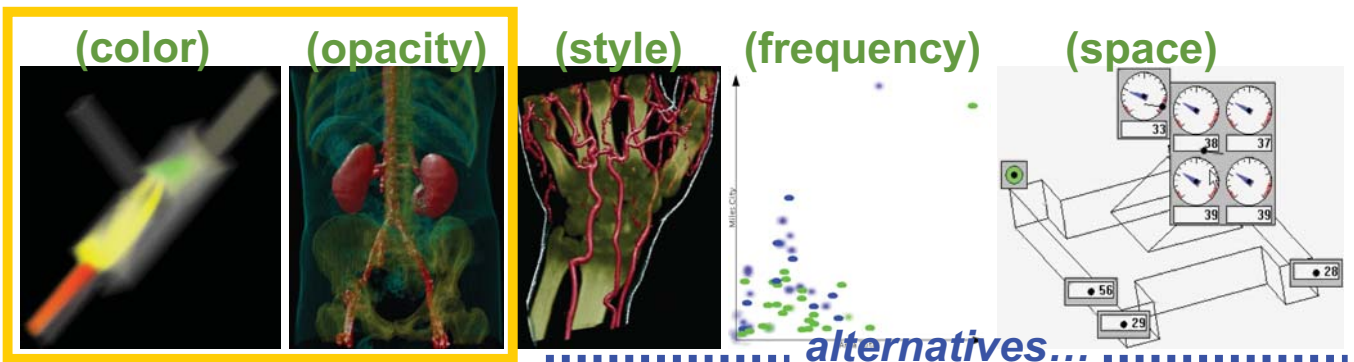
# IVA: Focus+Context Visualization

- Traditionally space distortion
  - more space for data of interest
  - rest as context for orientation
- Generalized F+C visualization
  - emphasize data in focus (color, opacity, ...)
  - differentiated use of visualization resources

[Mackinlay et al. 1991]



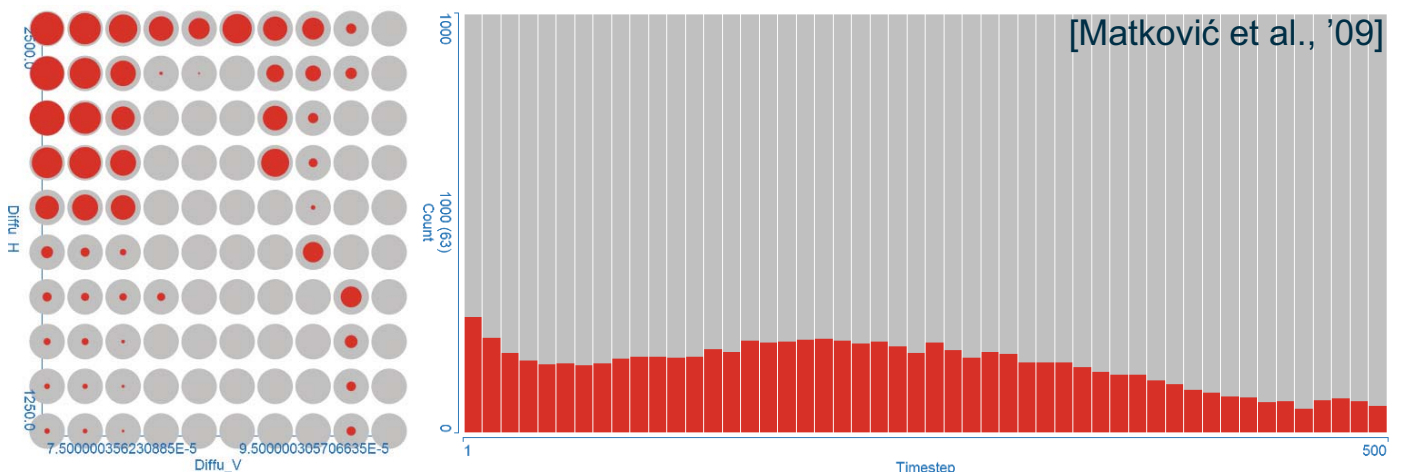
[Hauser... 2001, 2003]



## F+C Visualization in IVA Views

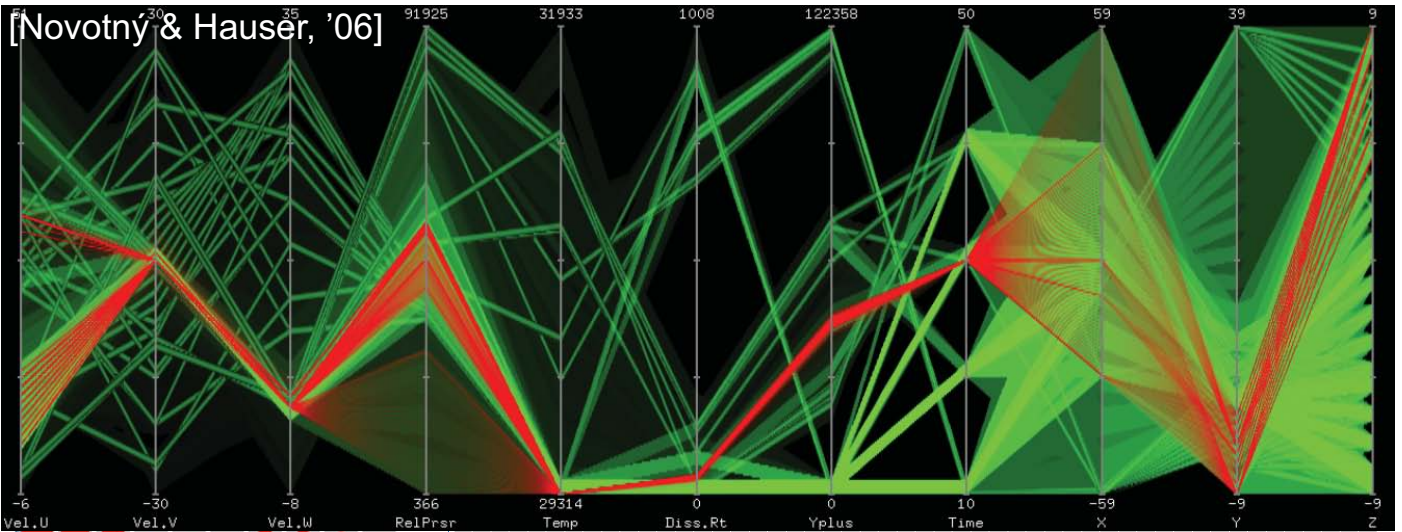
- Colored vs. gray-scale visualization
- Opaque vs. semi-transparent visualization

In a scatterplot (left) or histogram (right): brushed data in red...

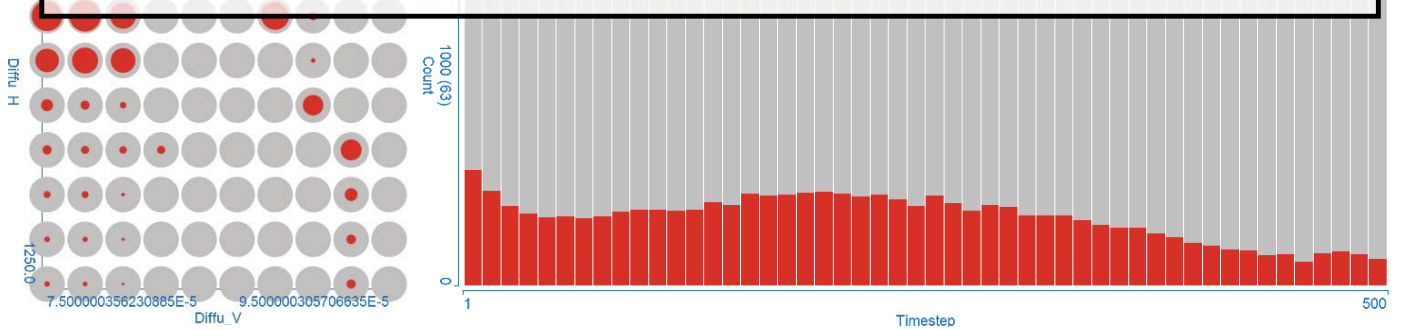


[Matković et al., '09]

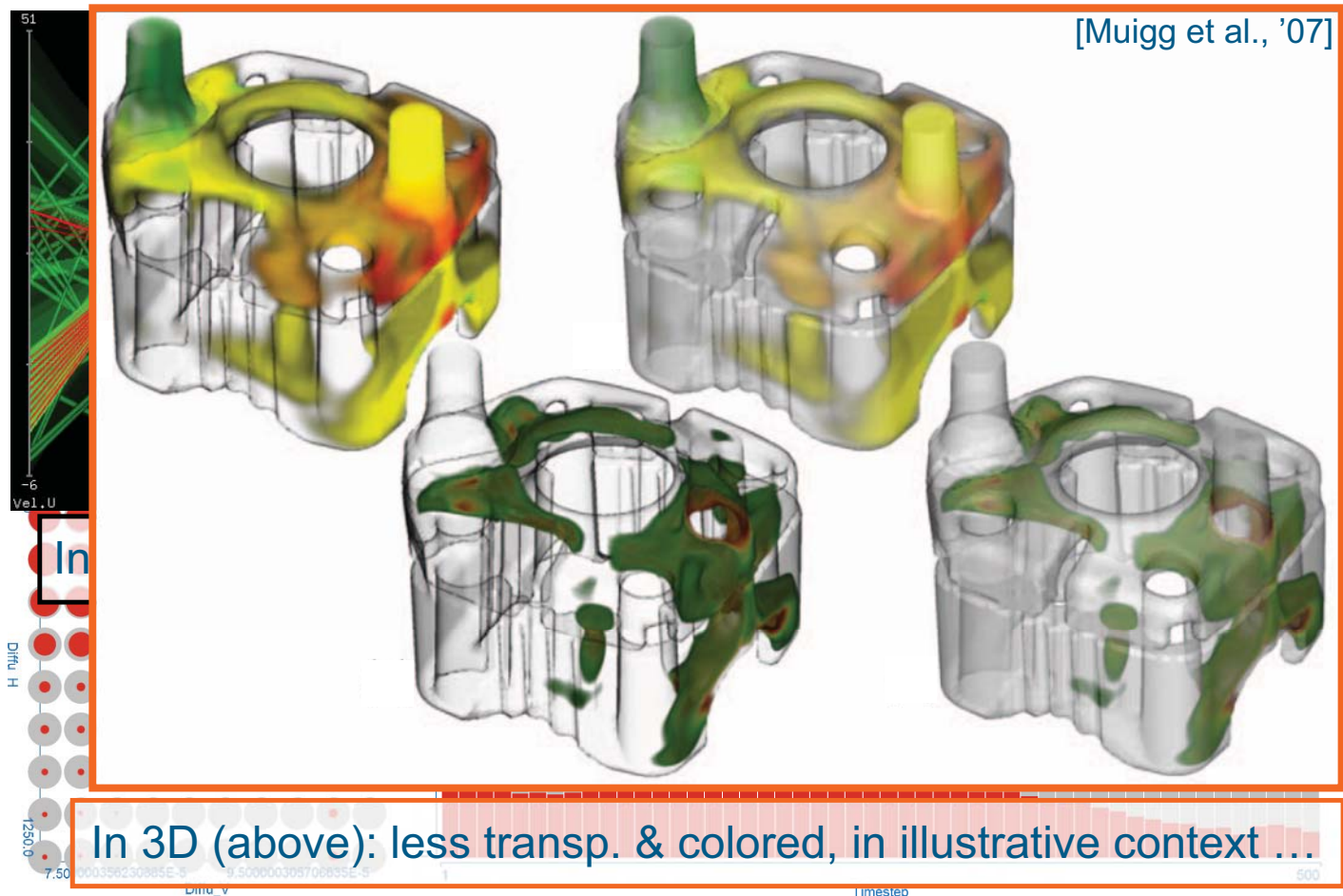
# F+C Visualization in IVA Views



In parallel coordinates (above): brushed data in red & over ...



# F+C Visualization in IVA Views

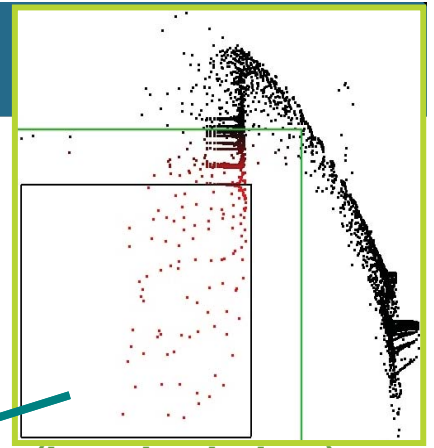


In 3D (above): less transp. & colored, in illustrative context ...

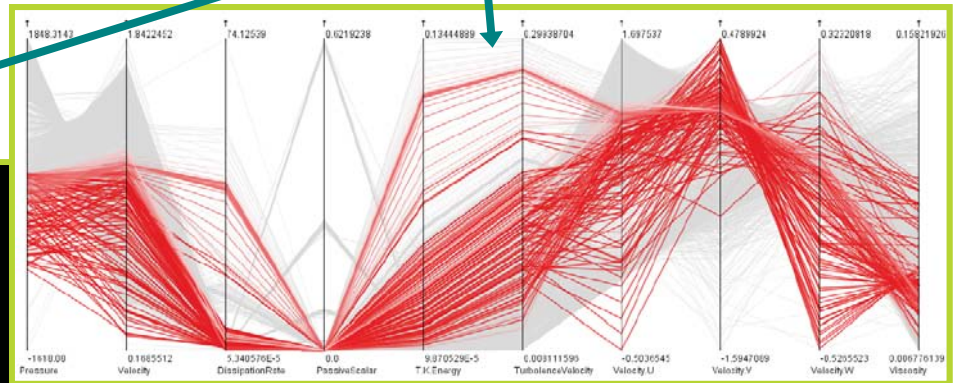


# IVA: Linked Views

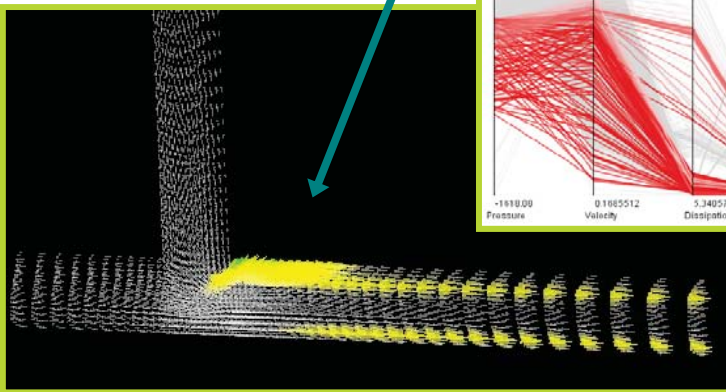
- Brushing: mark data subset as especially interesting
- Linking: enhance brushed data in linked views consistently (F+C)



(brushed view)



(linked views)



[Doleisch & Hauser, '02]

# IVA: Degree of Interest (DOI)

- $doi(.)$ : data items  $tr_i$  (table rows)  $\rightarrow$  degree of interest  
 $doi(tr_i) \in [0, 1]$

- $doi(tr_i) = 0 \Rightarrow tr_i$  not interesting ( $tr_i \in$  context)
- $doi(tr_i) = 1 \Rightarrow tr_i$  100% interesting ( $tr_i \in$  focus)

- Specification

- explicit, e.g., through direct selection
- implicit, e.g., through a range slider



x	y	d1	d2	doi
0	0	17,20	-0,22	0,00
1	0	12,10	0,10	0,00
2	0	7,70	0,45	0,00
3	0	2,10	0,90	0,00
0	1	24,10	0,02	0,00
1	1	21,90	0,36	0,00
2	1	15,50	0,87	0,74
3	1	11,10	1,20	1,00
0	2	27,20	0,12	0,00
1	2	24,10	0,66	0,18
2	2	17,30	1,35	1,00
3	2	12,10	2,20	0,60
0	3	35,50	0,67	0,00
1	3	30,90	1,30	0,00
2	3	24,50	2,10	0,10
3	3	20,80	2,90	0,00

- Fractional DOI values:  $0 \leq doi(tr_i) \leq 1$

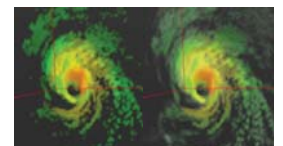
- several levels (0, low, med., ...)
- a continuous measure of interest
- a probabilistic definition of interest

(cont'd on next slide)

# IVA: Smooth Brushing $\rightarrow$ Fractional DOI

- Fractional DOI values esp. useful wrt. **scientific data**: (quasi-)continuous nature of data  $\leftrightarrow$  smooth borders

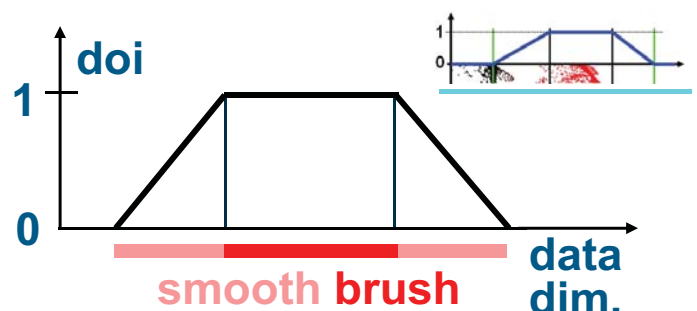
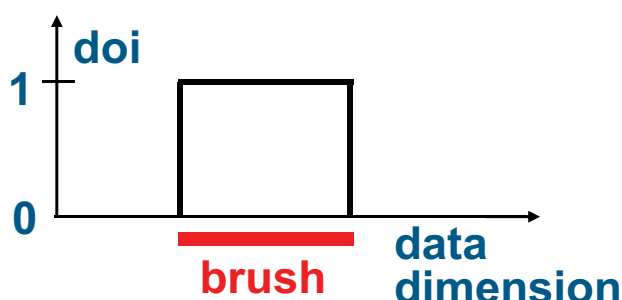
- Goes well with gradual focus+context vis. techniques (coloring, semitransparency)



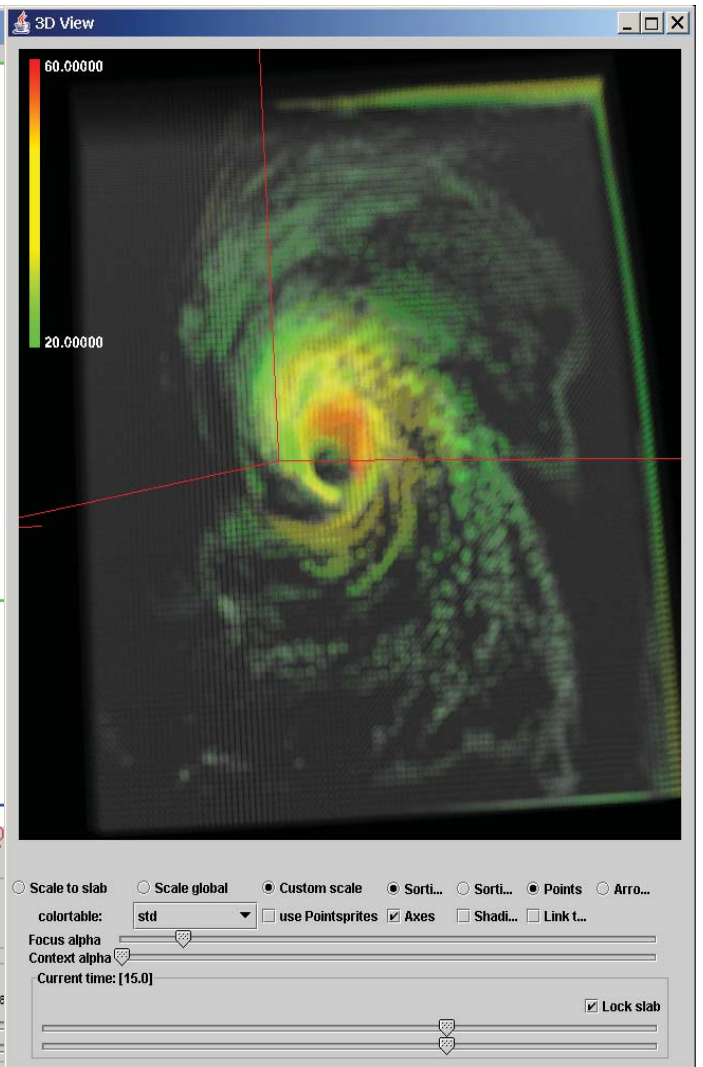
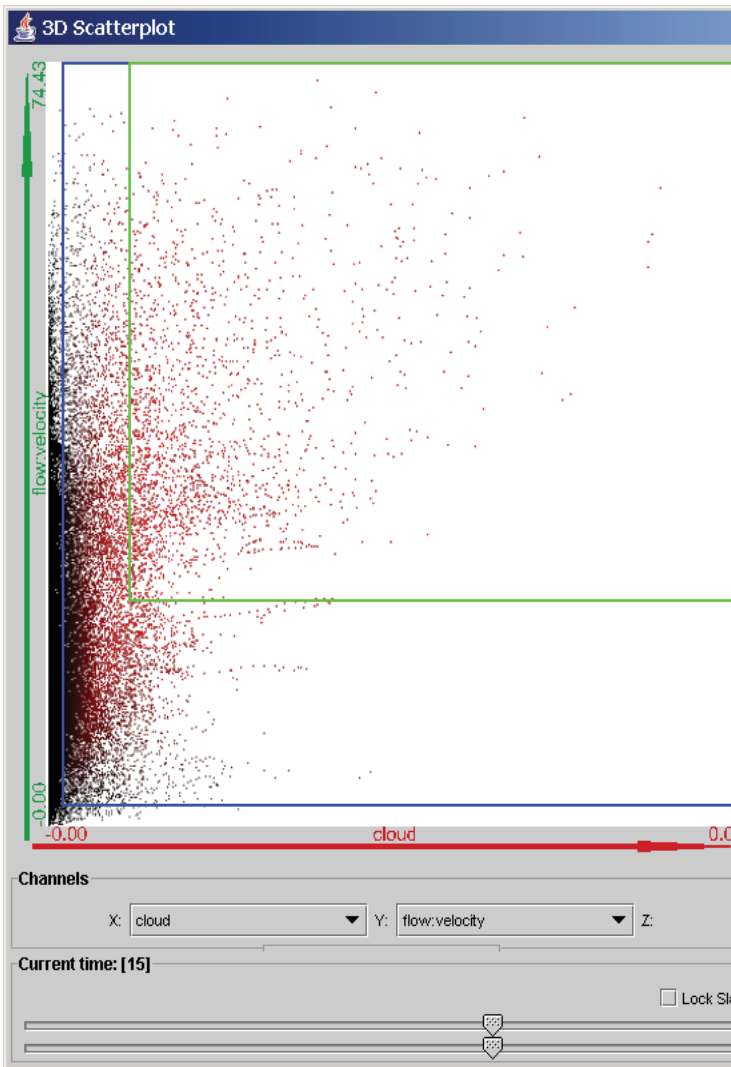
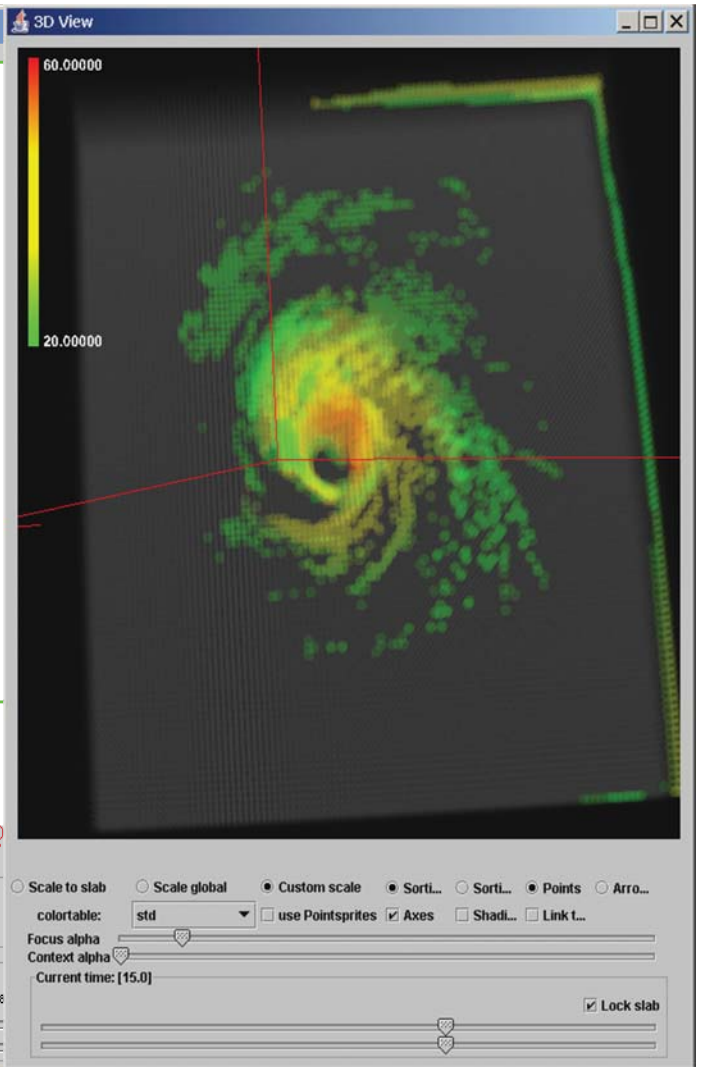
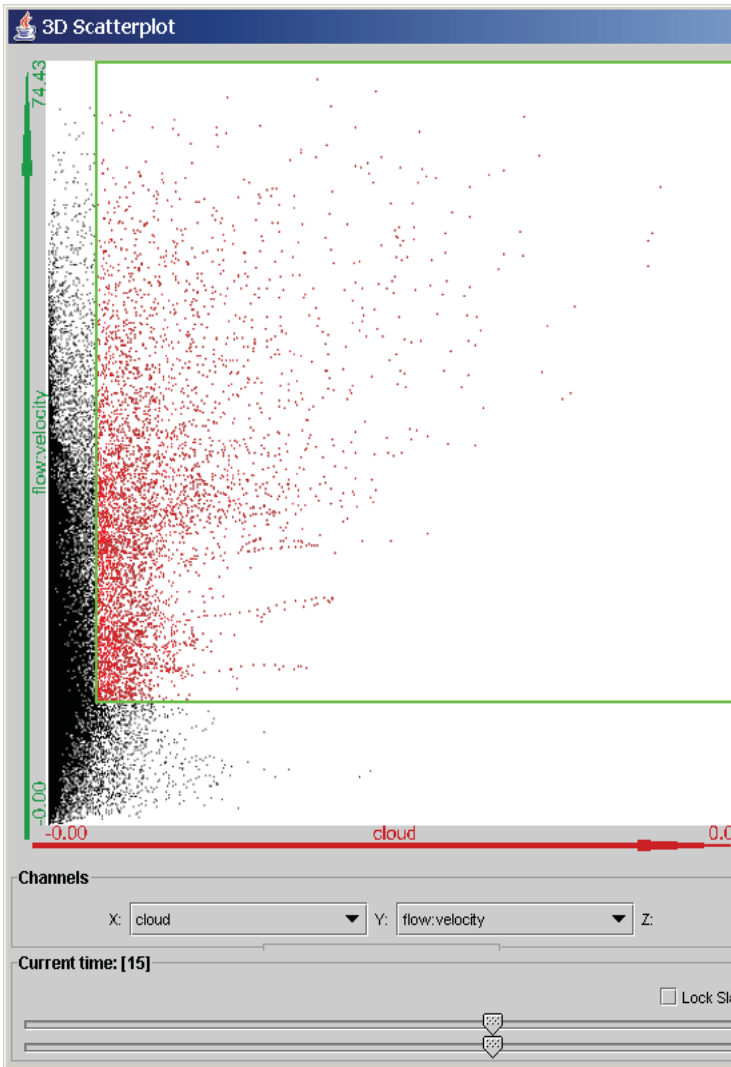
- Specification: **smooth brushing**

[Doleisch & Hauser, 2002]

- “inner” range: all 100% interesting (DOI values of 1)
- between “inner” & “outer” range: fractional DOI values
- outside “outer” range: not interesting (DOI values of 0)







# Three Patterns of SciData IVA

- Preliminary: domain  $x$  & range  $d$  visualized ( $\geq 2$  views)

1 ■ **brushing on domain visualization**,  
e.g., brushing special locations in the map view

► **local investigation** "d"

2 ■ **brushing on range visualization**,  
e.g., brushing outlier curves in a function graph view

► **feature localization** "x"

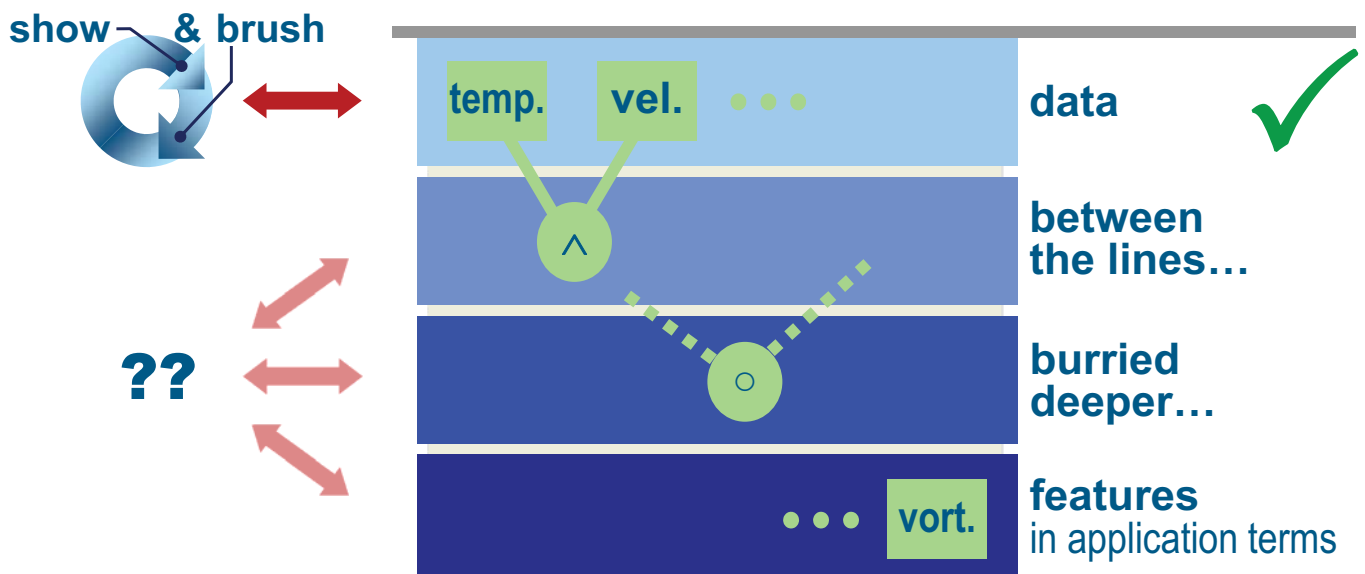
3 relating multiple range variates

► **multi-variate analysis** "d"

## IVA – Levels of Complexity

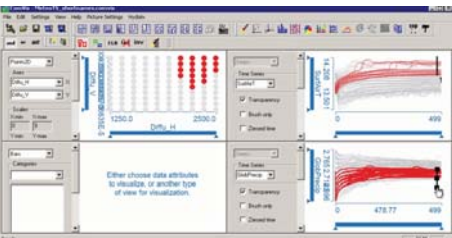
(1/4)

- A *lot* can be done with basic IVA, already! [pareto rule]
- We can consider a **layered information space**:  
from **explicitly** represented information (the **data**)  
to **implicitly** contained information, **features**, ...

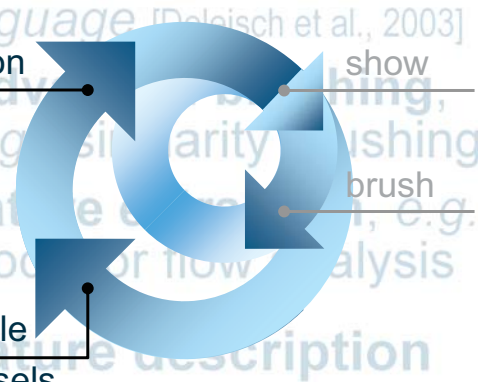




- A **lot** can be done with basic IVA, already! ✓ [parent rule]
- For more advanced exploration/analysis tasks, we extend it (in several steps):
  - IVA, level 2: **logical combinations of brushes**, e.g., utilizing the *feature definition language* [Doleisch et al., 2003]
  - IVA, l. 3: **attribute derivation**; **advanced brushing**, with interactive formula editor; e.g., similarity brushing
  - IVA, l4: **application-specific feature extraction**, e.g., based on vortex extraction methods for flow analysis
- Level 2: like **advanced verbal feature description**
  - ex.: “**hot flow, also slow, near boundary**” (cooling j.)
  - brushes comb. with **logical operators** (AND, OR, SUB)
  - in a **tree**, or **iteratively** (((b<sub>0</sub> op<sub>1</sub> b<sub>1</sub>) op<sub>2</sub> b<sub>2</sub>) op<sub>3</sub> b<sub>3</sub>) ...)



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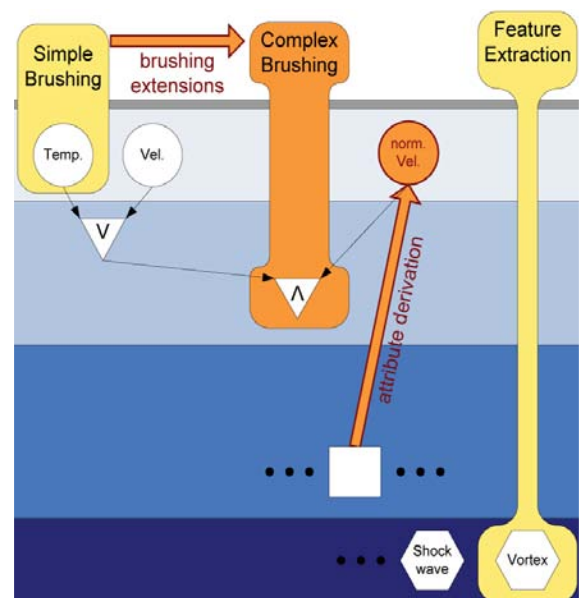




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- Level 3: using **general info extraction** mechanisms, two (partially complementary) approaches:
  1. **derive additional attribute(s)**, then show & brush
  2. use an **advanced brush** to select “hidden” relations



- A **lot** can be done with basic IVA
- For more advanced exploration/analysis tasks, we extend it (in several steps):
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  - IVA, l4: **application-specific feature extraction** based on vortex extraction methods



- Level 3: using **general info extraction** mechanisms, two (partially complementary) approaches:
  1. **derive additional attribute(s)**, then show & brush
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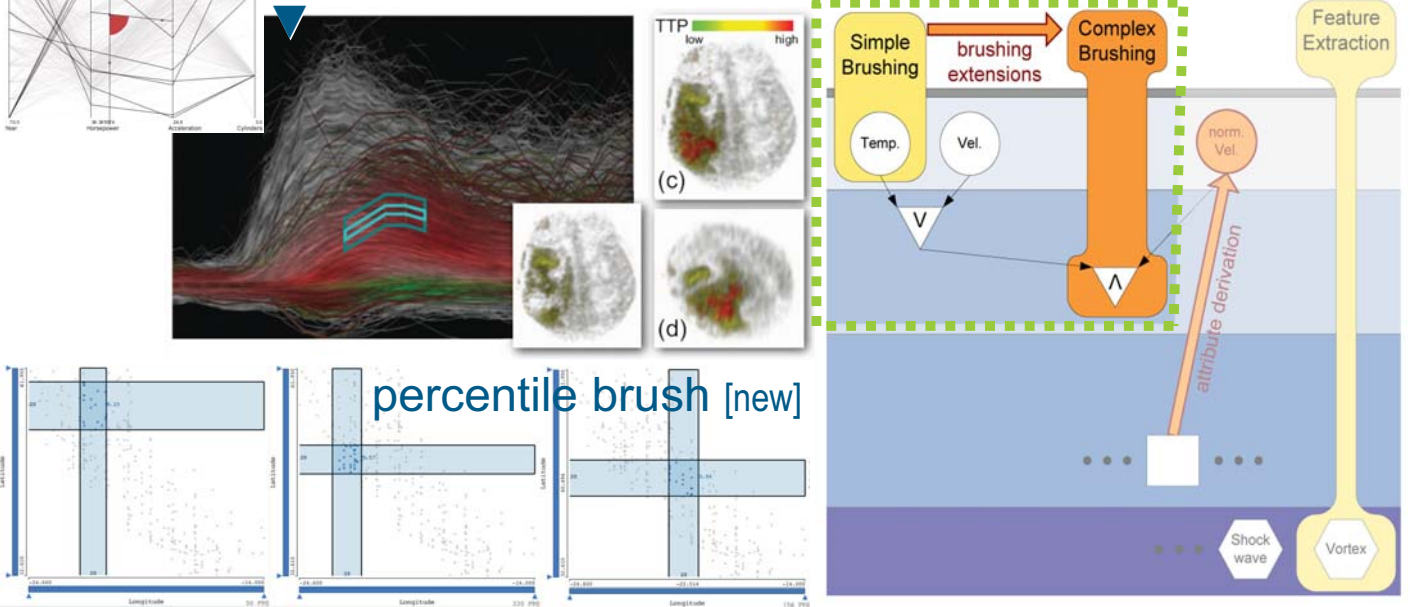


# IVA (level 3): Advanced Brushing

- **Std. brush:** brush 1:1 what you see
- **Adv. brush:** executes additional function (“intelligent?”)

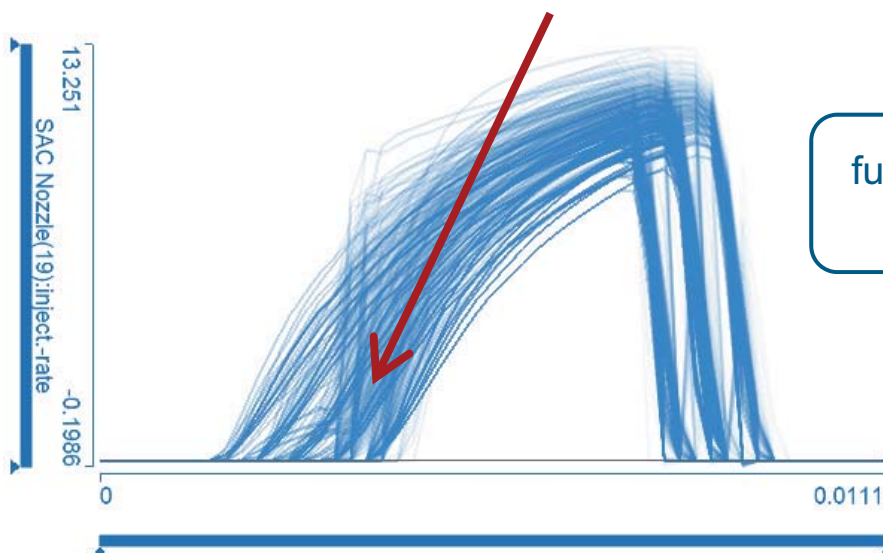
## Examples:

- angular brushing [Hauser et al., 2002]
- similarity brushing [Muigg et al., 2008]



# 3<sup>rd</sup> level IVA, adv. brushing example

- Considering a visualization of a family of function graphs:
  - select the steeply rising graphs



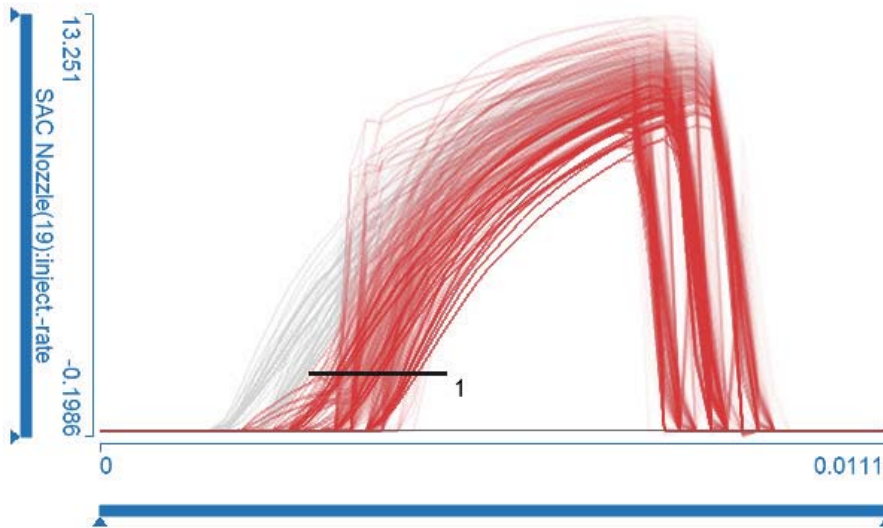
fuel injection simulation  
“injection rate”



# 3<sup>rd</sup> level IVA, adv. brushing example



- A simple line brush is not enough

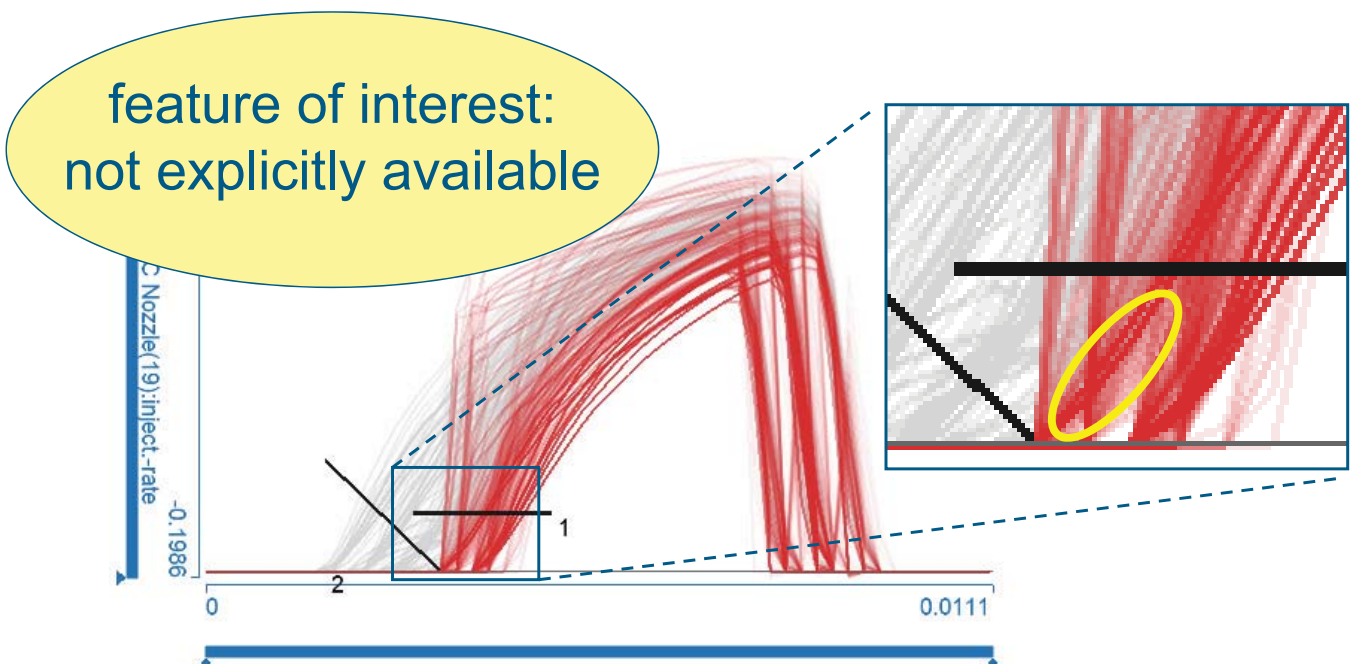


example prepared by Konyha, Zoltan

# 3<sup>rd</sup> level IVA, adv. brushing example



- A simple line brush is not enough
- Combining line brushes does not work, either

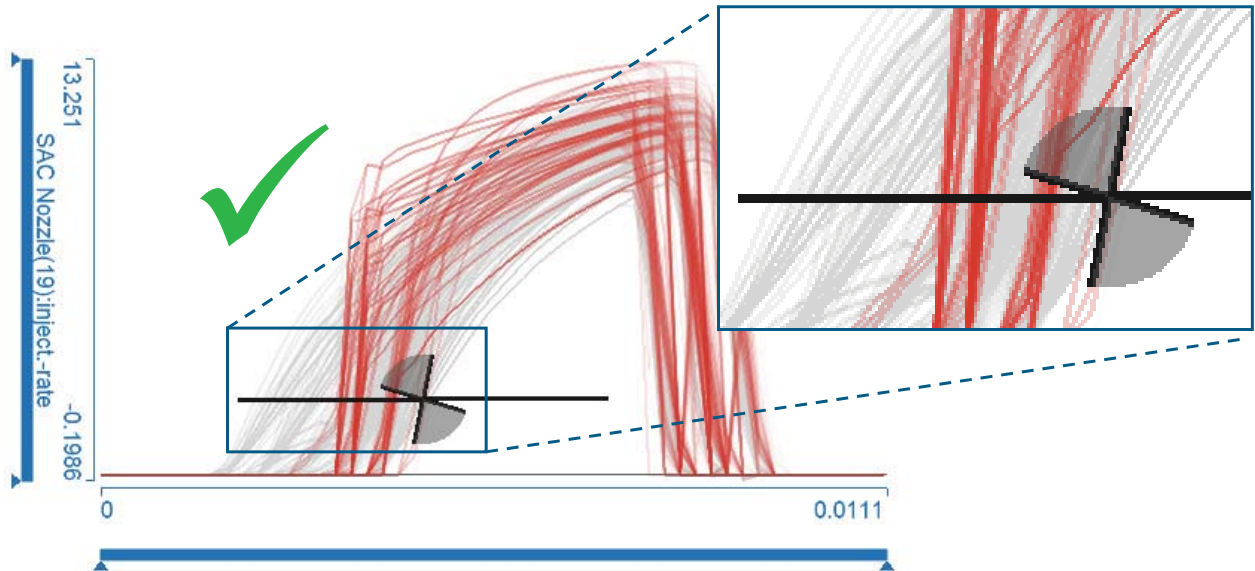


example prepared by Konyha, Zoltan

# 3<sup>rd</sup> level IVA, adv. brushing example



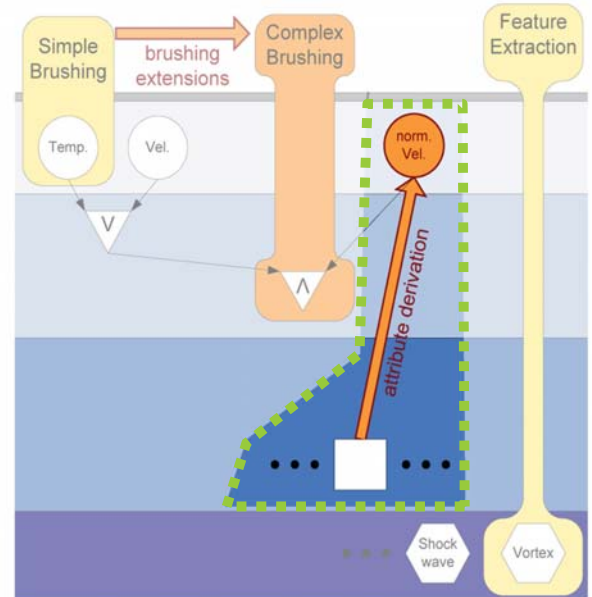
- The *angular line brush* (a specialized brush) selects the intended function graphs
  - that it intersects, and
  - the angle is in a given threshold



example prepared by Konyha, Zoltan

# IVA (level 3): Attribute Derivation

- Principle** (in the context of iterative IVA):
  - see some data feature  $\Phi$  of interest in a visualization
  - identify a **mechanism T** to describe  $\Phi$
  - execute** (interactively!) an **attribute derivation step** to represent  $\Phi$  explicitly (as new, synthetic attribute[s]  $d_\phi$ )
  - brush**  $d_\phi$  to get  $\Phi$
- Tools T** to describe  $\Phi$  from:
  - numerical mathematics
  - statistics, data mining
  - etc.*
  - **scientific computing**
- IVA w/ T ↔ visual computing**

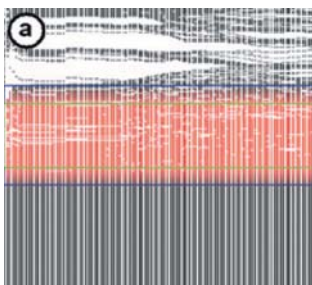


## Attribute Derivation ↔ User Task / example

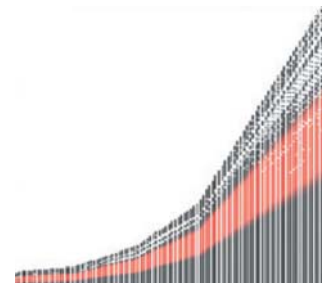
- The tools T, available in an IVA system, must reflect/match the **analytical steps of the user**:

### Example:

- first vis.:** ↔ user wishes to select the “band” in the middle
- so?** an advanced brush? a lasso maybe?
- ah!** → let's normalize y and then brush (a)



- leading to the wished selection:**



# What user wishes to reflect?

- Many **generic wishes** – users interest in:
  - something **relative** (instead of some absolute values),  
example: show me the *top-15%*
  - **change** (instead of current values),  
ex.: show me *regions with increasing temperature*
  - some **non-local property**,  
ex.: show me regions with *high average temperature*
  - **statistical properties**,  
ex.: show me *outliers*
  - **ratios/differences**,  
ex.: show me population per area, difference from trend
  - *etc.*
- **Common characteristic** here:
  - **questions/tools generic**, not application-dependent!

# How to reflect these user wishes?

- Many **generic wishes** – users interest in:
  - something **relative** (instead of some absolute values),  
example: show me the *top-15%* ⇒ **use, e.g., normalization**
  - **change** (instead of current values),  
ex.: show me *regions with increasing temperature* ⇒ **derivative estimation**
  - some **non-local property**,  
ex.: show me regions with *high average temperature* ⇒ **numerical integration**
  - **statistical properties**,  
ex.: show me *outliers* ⇒ **descriptive statistics**
  - **ratios/differences**,  
ex.: show me population per area, difference from trend ⇒ **calculus**
  - *etc.* ⇒ **data mining**  
(fast enough?)
- **Common characteristic** here:
  - **questions/tools generic**, not application-dependent!

- From **analysis, calculus, num. math**:
  - **linear filtering** (convolve the data with some linear filter on demand, e.g., to smooth, for derivative estimation, etc.)
  - **calculus** (use an interactive formula editor for computing simple relations between data attributes; +, -, ·, /, etc.)
  - **gradient estimation, numerical integration** (e.g., wrt. space and/or time) ⇒ example
  - **fitting/resampling** via **interpolation/approximation**
- From **statistics, data mining**:
  - **descriptive statistics** (compute the statistical moments, also robust, measures of outlyingness, detrending, etc.) ⇒ example
  - **embedding** (project into a lower-dim. space, e.g., with PCA for a subset of the attribs., etc.) ⇒ example
- **Important**: executed on demand, after prev. vis.

## 3<sup>rd</sup>-level IVA – Sample Iterations

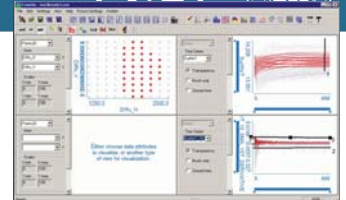
(1/2)



### ■ The Iterative Process of 3<sup>rd</sup>-level IVA:

#### ■ Example 1:

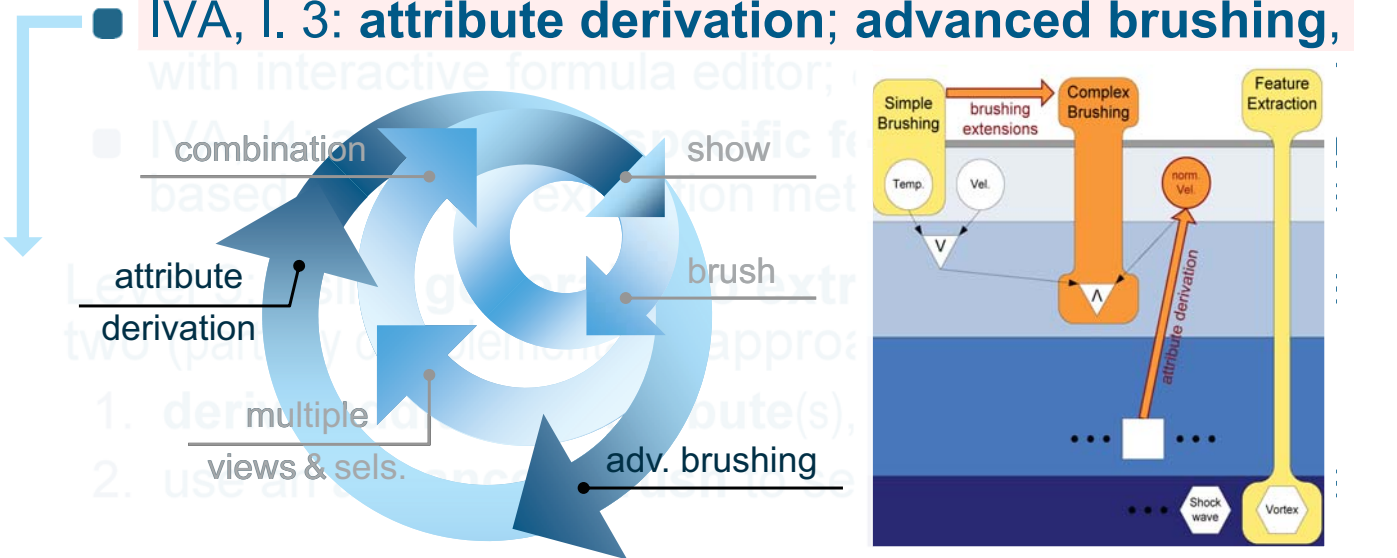
- you look at some *temp. distribution over some region*
- you are *interested in raising temperatures, but not temperature fluctuations*
- you use a **temporal derivate estimator**, for ex., central differences  $t_{\text{change}} = (t_{\text{future}} - t_{\text{past}}) / \text{len}(\text{future-past})$
- you plot  $t_{\text{change}}$ , e.g., in a **histogram** and **brush** whatever change you are interested in
- maybe you see some frequency amplification due to derivation, **so you go back** and
- **use an appropriate smoothing filter** to *remove high frequencies from the temp. data*, leading to a derived new  $\tau = t_{\text{smooth}}$  data attribute
- selecting from a **histogram** of  $\tau_{\text{change}}$  (computed like above) is then less sensitive to temperature fluctuations







- A **lot** can be done with basic IVA, already! ✓ [parent rule]
- For more advanced exploration/analysis tasks, we extend it (in several steps):
  - IVA, level 2: **logical combinations of brushes** utilizing the *feature definition language* [Dreisch et al., 2005]
  - IVA, l. 3: **attribute derivation; advanced brushing**,

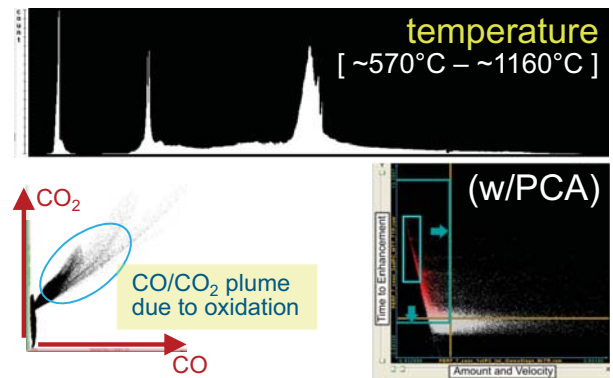


- A **lot** can be done with basic IVA, already! ✓ [parent rule]
- For more advanced exploration/analysis tasks, we extend it (in several steps):
  - IVA, level 2: **logical combinations of brushes** utilizing the *feature definition language* [Dreisch et al., 2005]
  - IVA, l. 3: **attribute derivation; advanced brushing**, with interactive formula editor; e.g., similarity
  - IVA, l4: **application-specific feature extraction** based on vortex extraction methods for flow analysis

- Level 4: **application-specific procedures**
  - tailored solutions (for a specific problem)
  - “deep” information drill-down
  - *etc.*

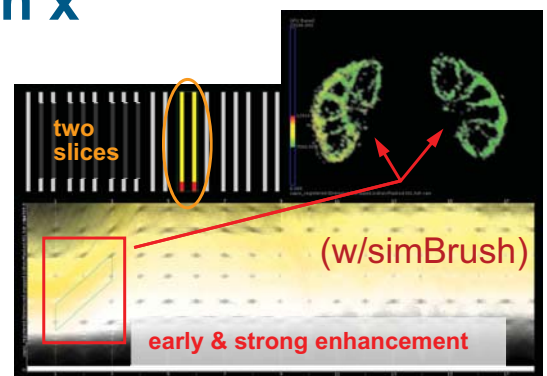
## ■ Understanding data wrt. range $d$

- which distribution has data attribute  $d_i$ ?
- how do  $d_i$  and  $d_j$  relate to each other? (**multivariate analysis**)
- which  $d_k$  discriminate data features?



## ■ Understanding data wrt. domain $x$

- **where** are relevant features? (**feature localization**)
- **which** values at specific  $x$ ? (**local analysis**)
- how are they **related to parameters**?



# The Iterative Process of IVA...

- ...is a **very useful methodology** for **data exploration & analysis**
- ...is **very general** and can be (has already been) applied to **many different application fields** (in this talk the focus was on scientific data)
- ...**meets scientific computing** as a complementary methodology (with the **important difference** that in IVA the **user** with his/her **perception/cognition** is **in the loop** at **different frequencies**, also many fps)
- ...is **not yet fully implemented** (we've done something, e.g., in the context of **SimVis**, **ComVis**, etc.) – from here: different possible paths, incl. InteractiveVisualMatlab, IVR, etc.)

## You!

Alexander Lundervold and HIB!

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*et al.*

Funding agencies!