

# Basics of Interactive Visual Analysis

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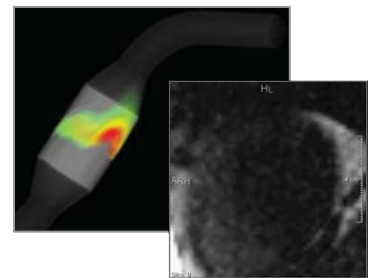


## Interactive Visual Analysis



- Given data –  
*too much* and/or *too complex* to be shown all at once:
- IVA is an **interactive visualization methodology** to facilitate
  - the **exploration** and/or **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**

- **IVA** (interactive visual analysis) **since 2000**
- **Tightly related** to visual analytics, of course, e.g., *integrating computational & interactive data analysis*
- **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**...



## 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  $\mathbf{d} \leftrightarrow$  **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:**
  - $\mathbf{d}$  is (at least to a certain degree) **continuous** wrt.  $\mathbf{x}$

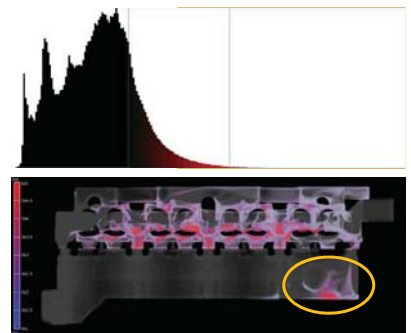
- **Interactive visual analysis** (as exemplified in this tutorial) **works really well with scientific data**, e.g.,
  - results from **numerical simulation** (spatiotemporal)
  - imaging / **measurements** (in particular multivariate)
  - sampled **models**
- When used to study scientific data, **IVA 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)



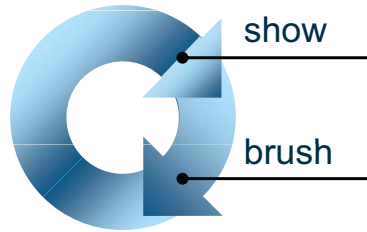
- **KISS-principle IVA:**

- 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)!
- brush this “something”
- linked F+C visualization
- first insight!

... leads to ...

degree of interest

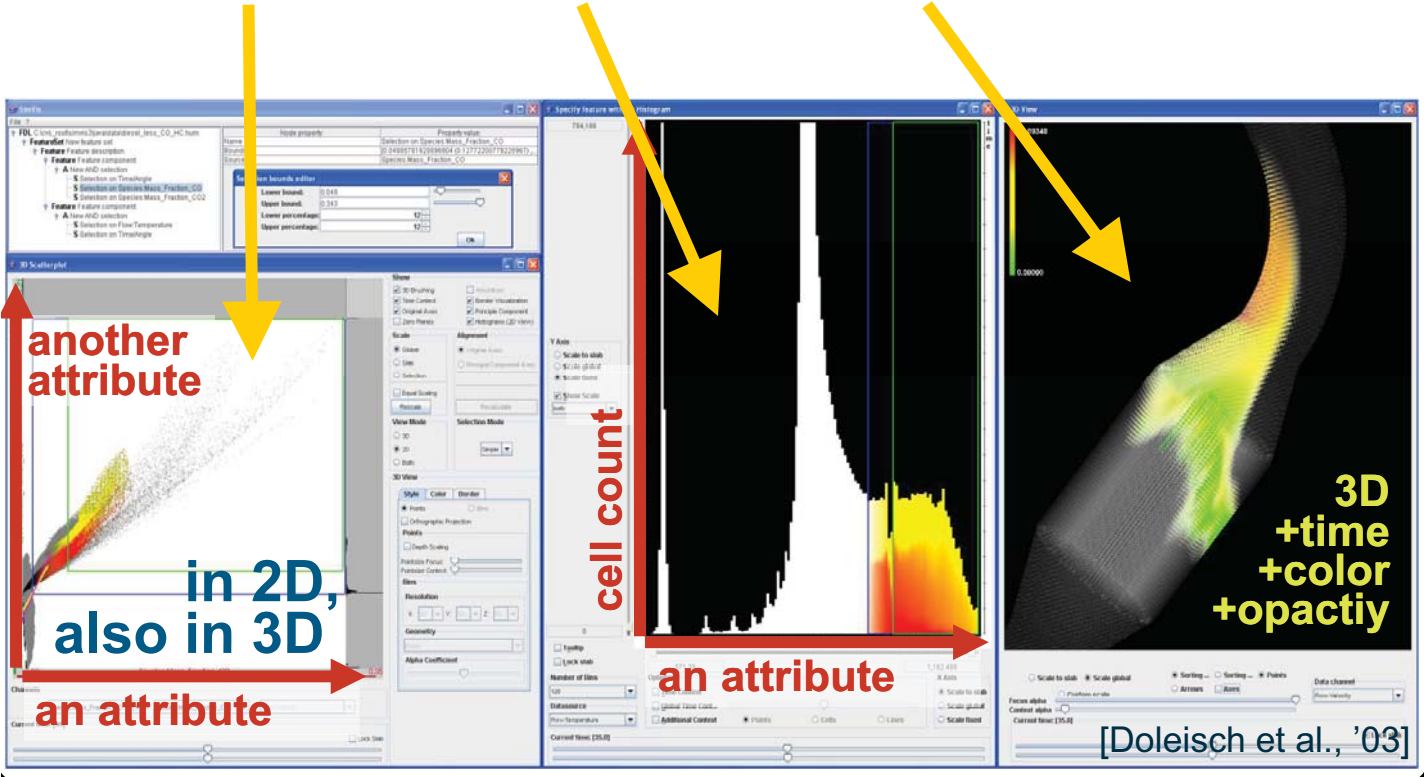
... requires ...

... is realized via ...

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

# IVA: Multiple Views

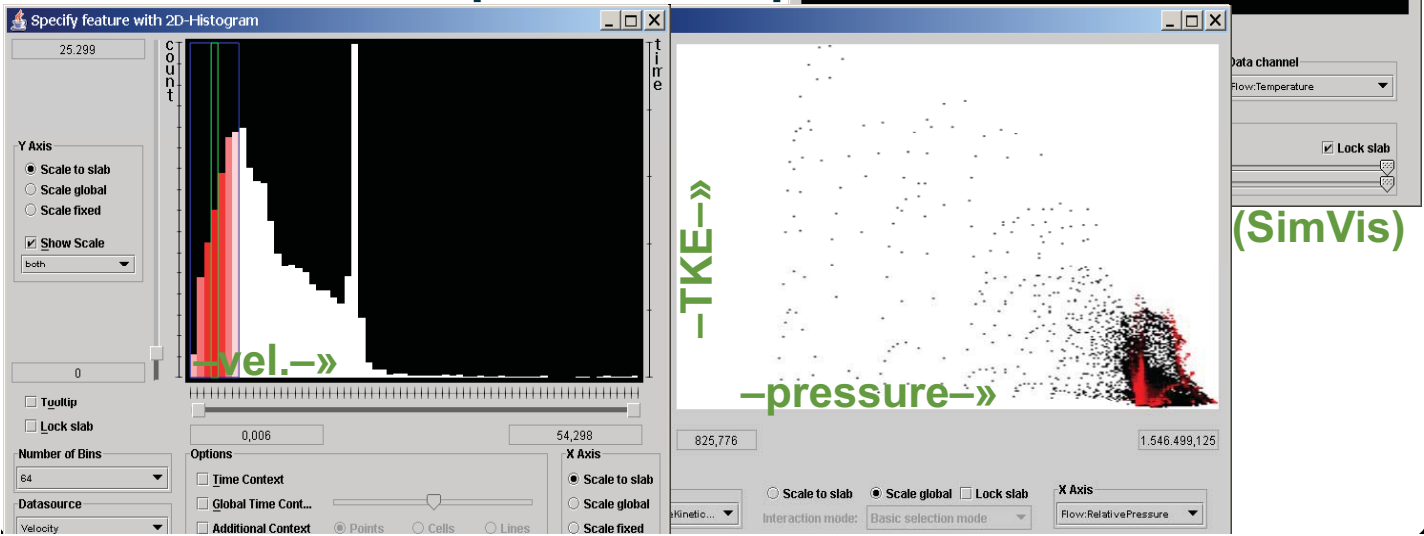
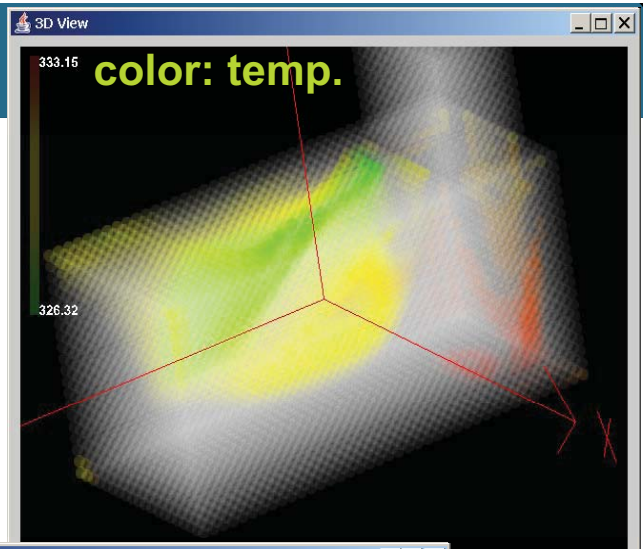
- One dataset, but multiple views
- Scatterplots, histogram, 3D(4D) view, etc.



# IVA: Interactive Brushing

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

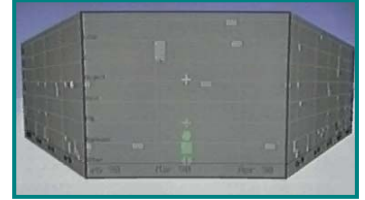
[Doleisch et al., '03]



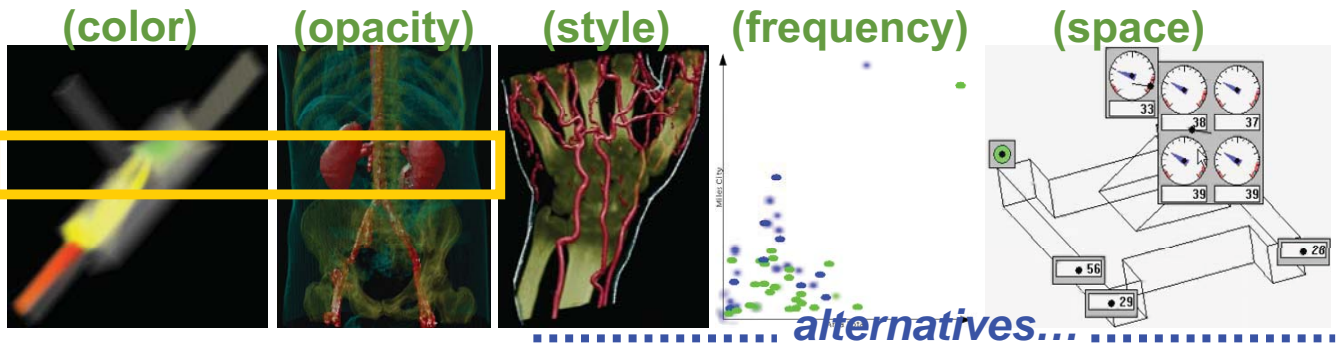
# IVA: Focus+Context Visualization

- 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]

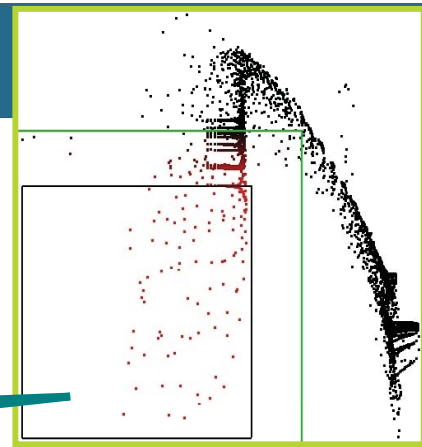


[Hauser... 2001, 2003]

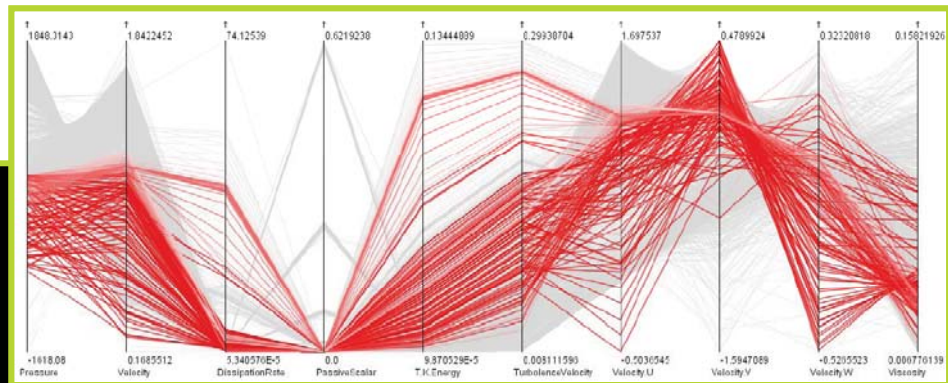


## IVA: Linked Views

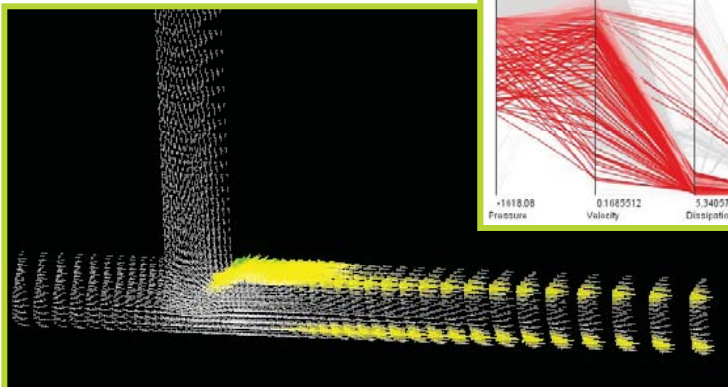
- Brushing: mark data subset as especially interesting
- Linking: enhance brushed data in linked views consistently (F+C)



(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)

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

## Specification

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



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

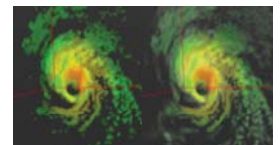
- several levels (0, low, med., ...)
- a continuous measure of interest
- a probabilistic definition of interest

(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

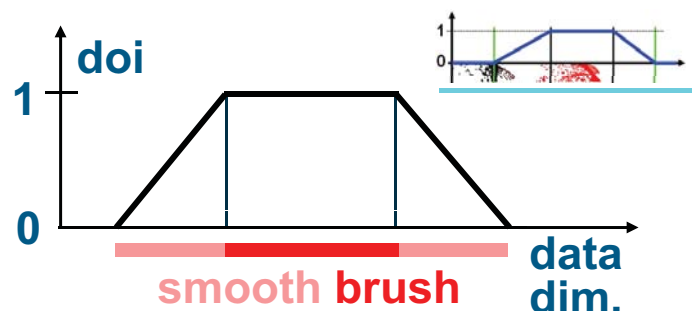
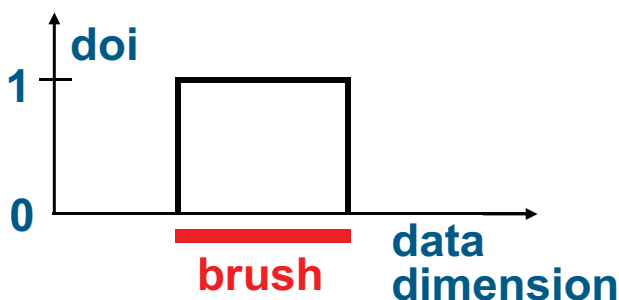
- Goes well with gradual focus+context vis. techniques (coloring, semitransparency)



## Specification: smooth brushing

[Doleisch & Hauser, 2002]

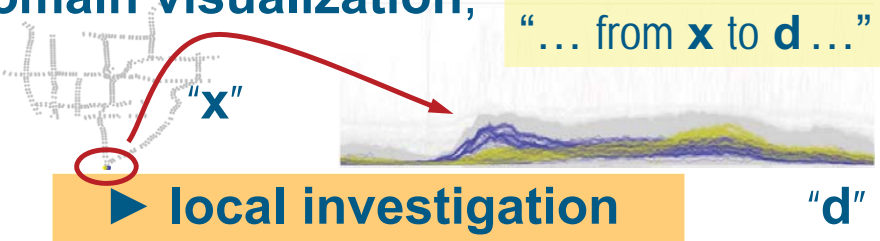
- “inner” range: all 100% interesting (DOI values of 1)
- between “inner” & “outer” range: fractional DOI values
- outside “outer” range: not interesting (DOI values of 0)



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

- **brushing on domain visualization,**

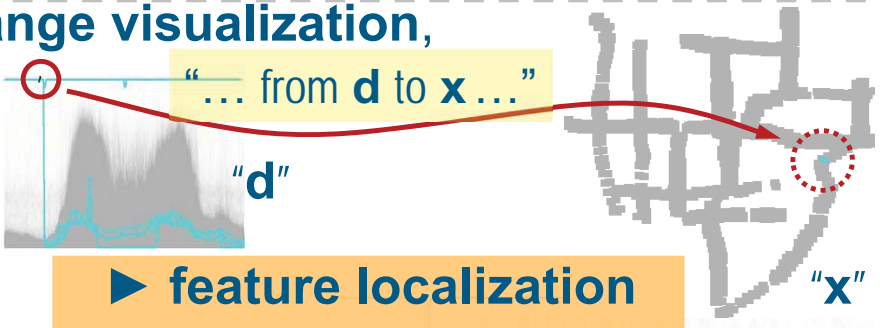
*e.g.*, brushing special locations in the map view



1

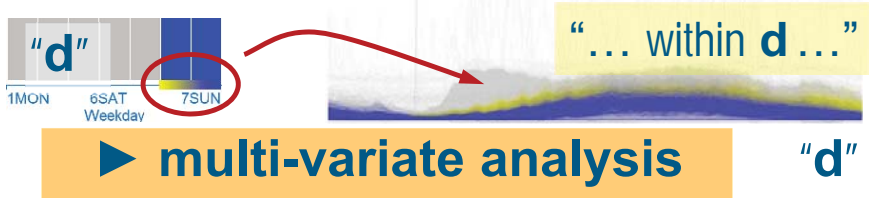
- **brushing on range visualization,**

*e.g.*, brushing outlier curves in a function graph view



2

relating multiple range variates



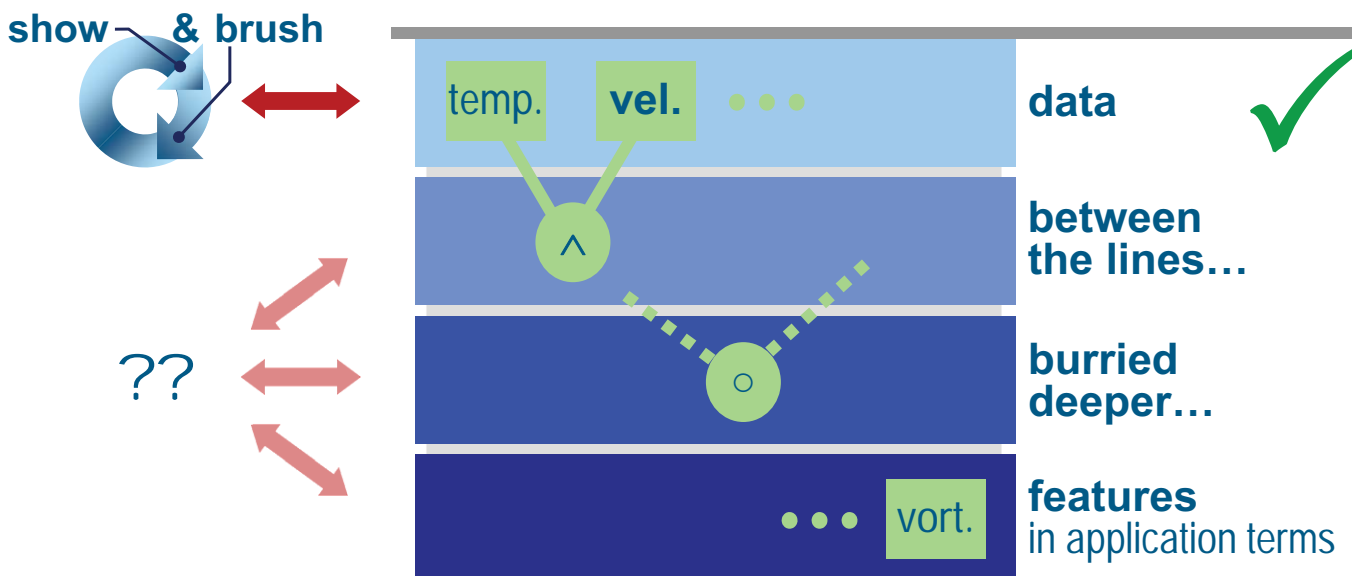
3

## 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**, ...

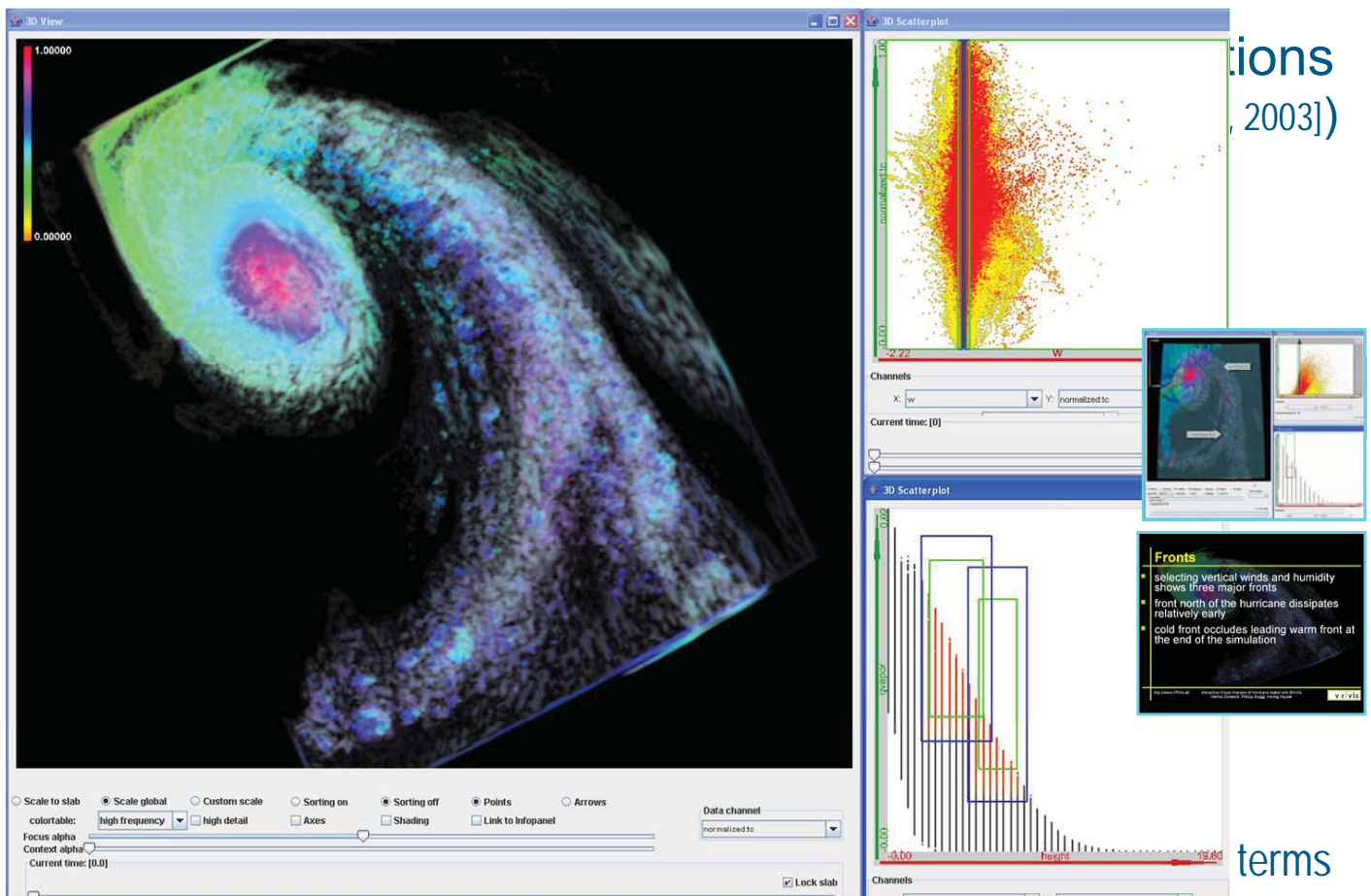
show & brush





- A lot can be done with KISS-principle IVA! ✓ (pareto 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>) ...)

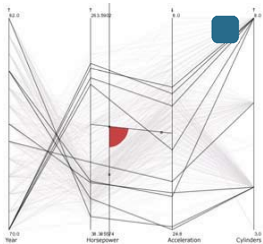
## IVA (level 2) Example



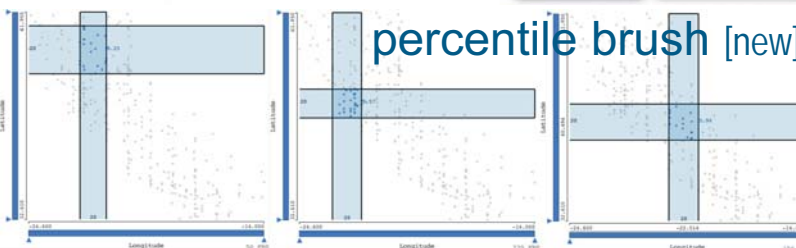
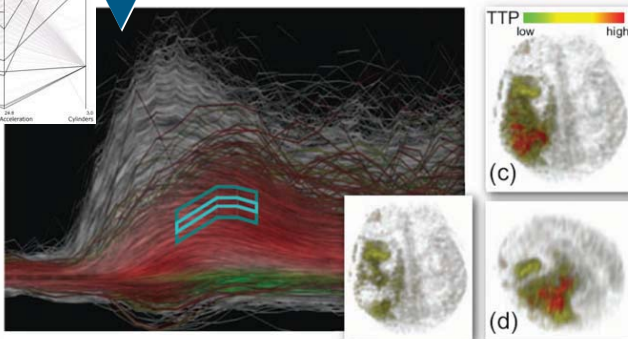
- A lot can be done with KISS-principle IVA! (pareto rule) ✓
- For more advanced exploration/analysis tasks, we extend it (in several steps):
  - IVA, level 2: **logical combinations of brushes** utilizing the *feature definition language* [Dreisch et al., 2009] ✓
  - IVA, l. 3: **attribute derivation; 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 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

## IVA (level 3): Advanced Brushing

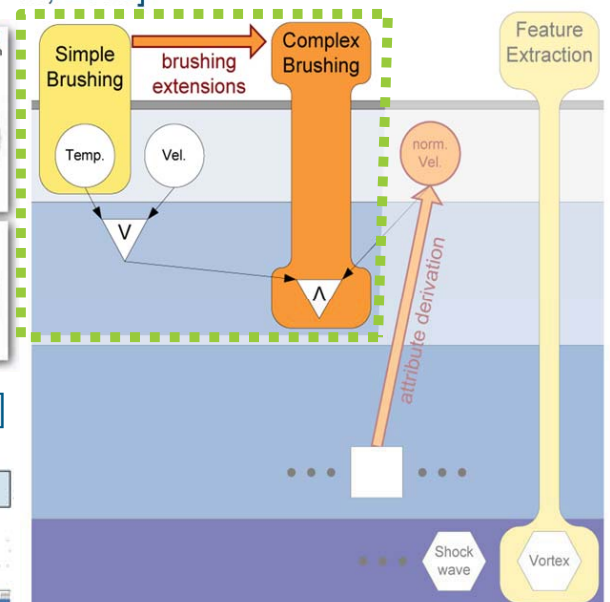
- **Std. brush:** brush 1:1 what you see
- **Adv. brush:** executes additional function (“intelligent?”)
- **Examples:**



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

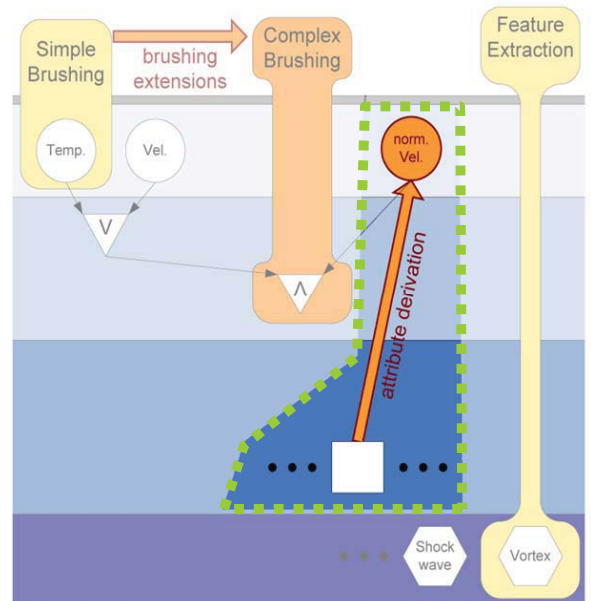


percentile brush [new]



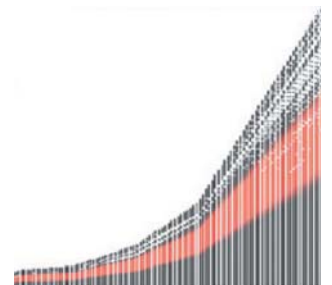
# IVA (level 3): Attribute Derivation

- 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!

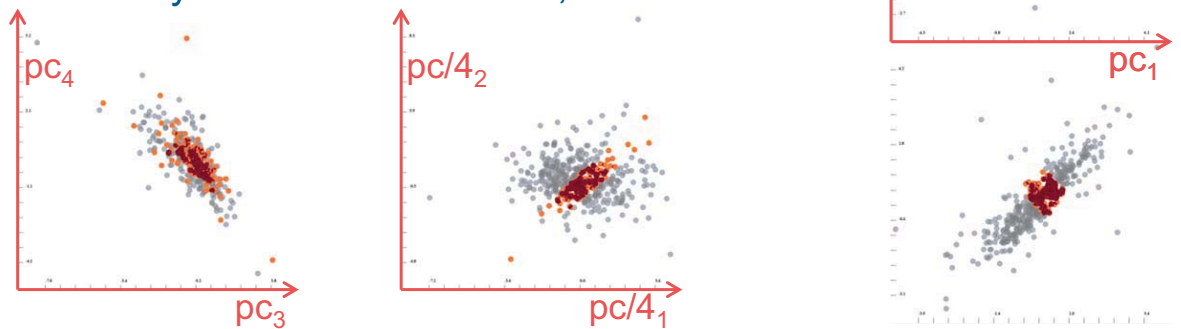
- From **analysis, calculus, num. math**:
  - **linear filtering** (convolve the data with some linear filter on demand, e.g., to smooth, for derivative estimation, etc.)
  - **calculus** (use an interactive formula editor for computing simple relations between data attributes; +, -, ·, /, etc.)
  - **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.

- The **Iterative Process of 3<sup>rd</sup>-level IVA**:
  - Example 1:
    - you look at some *temp. distribution over some region*
    - you are *interested raising temperatures, but not temperature fluctuations*
    - you use a **temporal derivate estimator**, for ex., central differences  $t_{\text{change}} = (t_{\text{future}} - t_{\text{past}}) / \text{len}(\text{future} - \text{past})$
    - you plot  $t_{\text{change}}$ , e.g., in a **histogram** and **brush** what ever change you are interested in
    - maybe you see that 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

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

### ■ Example 2:

- you bring up a scatterplot of  $d_1$  vs.  $d_2$ :  
(from an ECG dataset [Frank, Asuncion; 2010])
- obviously,  $d_1$  and  $d_2$  are correlated,  
our interest: the **data center** wrt. the **main trend**
- we ask for a (local) **PCA** of  $d_1$  and  $d_2$ :
- then we **brush the data center**
- we get the wished selection
- from here further steps are possible...,  
incl. study of other PCA-results, etc.



## Visualizing / analyzing lots of statistics

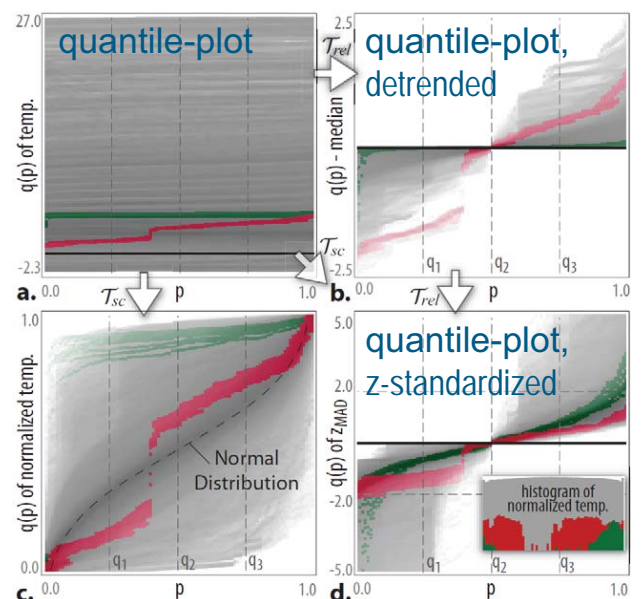
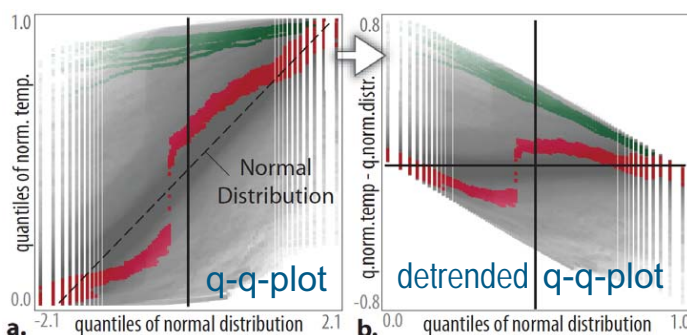
[Kehrer et al., TVCG 2011]

### ■ Useful statistical measures include:

- moments ( $\mu$ ,  $\sigma$ , ...), **robust versions** (median, IQR, ...)
- **quartiles, octiles, and quantiles**  $q(p)$

### ■ Useful views allow the interactive visual analysis

- **quantile-plot**  $q(p)$  vs.  $p$ ,  
here for numerous  $x$
- **detrending** (e.g.,  $-q_2$ ),  
**normalization** (e.g.,  $z$ )



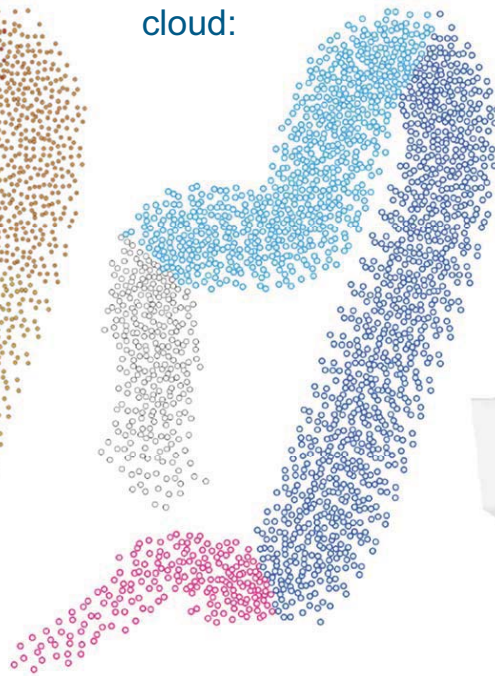
# Brushing of Attribute Clouds for the Visualization of Multivariate Data

Heike Jänicke, Michael Böttinger, and Gerek Scheuermann, *Member, IEEE*

2D embedding:  
the attribute cloud

brushed  
cloud:


corresponding  
feature(s):



## IVA – Levels of Complexity

(4/4)



- A lot can be done with KISS-principle IVA! ✓ (pareto rule) 
- For more advanced exploration/analysis tasks, we extend it (in several steps):

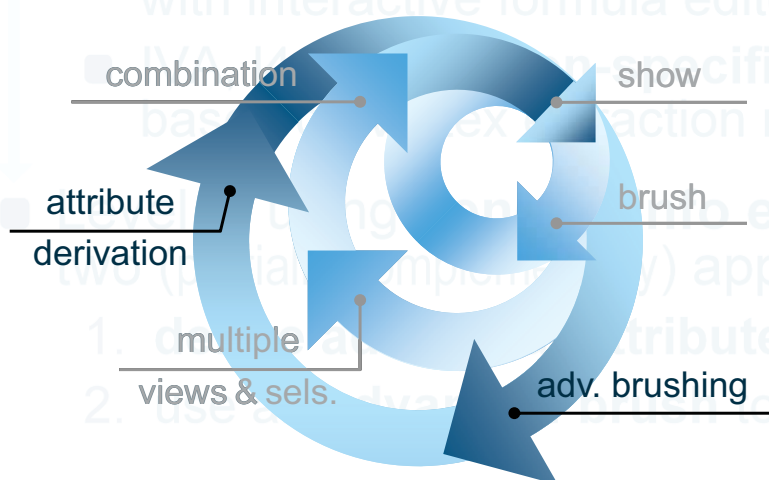
- IVA, level 2: **logical combinations of brushes** utilizing the *feature definition language* [Dreisch et al., 2005] ✓ 

- IVA, l. 3: **attribute derivation; advanced brushing,** with interactive formula editor; e.g., similarity brushing

**feature extraction, e.g.,** methods for flow analysis

**traction mechanisms,** approaches:

1. show & brush  
2. select “hidden” relations

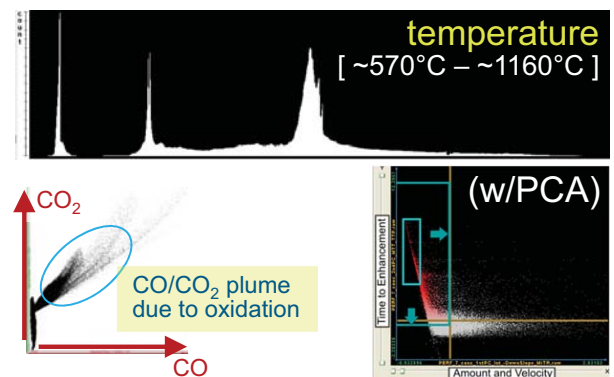


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- For more advanced exploration/analysis tasks, we extend it (in several steps):
  - IVA, level 2: **logical combinations of brushes** utilizing the *feature definition language* [Dreisch et al., 2009] ✓
  - IVA, l. 3: **attribute derivation; advanced brushing**, with interactive formula editor; e.g., similarity [Dreisch et al., 2009] ✓
  - IVA, l4: **application-specific feature extraction**, e.g., based on vortex extraction methods for flow analysis ✓
- Level 4: **application-specific procedures**
  - tailored solutions (for a specific problem)
  - “deep” information drill-down
  - *etc.*

## Interactive Visual Analysis – delivery

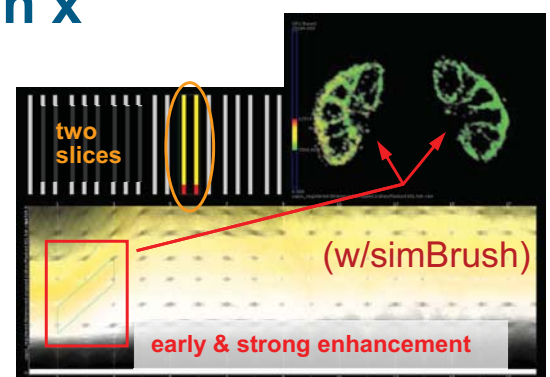
### ■ Understanding data wrt. range $d$

- 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...

...leads to an **interactive & iterative** workbench for **visual data exploration & analysis** (compare to **visual computing**, again)

■ Different **levels of complexity** (show & brush, logical combinations, advanced brushing & attribute derivation, *etc.*)...

...lead to according **iteration frequencies**: comment on human time constants

- on level 1: **smooth interactions, many fps**, for example during linking & brushing
- on level 2: **interleaved fast steps of brush ops.**, for example when choosing a logical op. to cont. with
- on level 3: **occasionally looking at a progress bar**, for example when computing some PCA, *etc.*

■ These frequencies **limit the spectrum** of usable tools

➤ New res. work will help to **extend this spectrum!**

# 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!

## ■ Krešimir Matković & Giuseppe Santucci!

- Helmut Doleisch, Raphael Fuchs, Johannes Kehrer, Çağatay Turkey, *et al.*!
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