

Basics of Interactive Visual Analysis

Helwig Hauser (Univ. of Bergen)

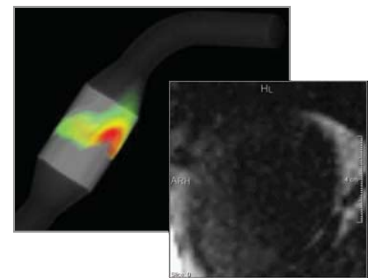


Interactive Visual Analysis



- 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**...



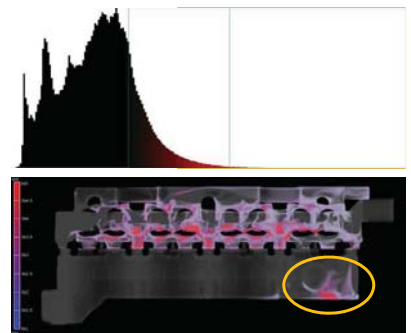
Target Data Model: “Scientific Data”

- **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

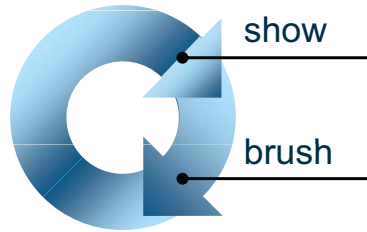
- Loop / bundling of *two complementary parts*:
 - **visualization** – show to the user!
Something new, or something due to interaction.
 - **interaction** – tell the computer!
What is interesting? What to show next?
- Basic example (**show – brush – show – ...**), cooling jacket context:
 1. show a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

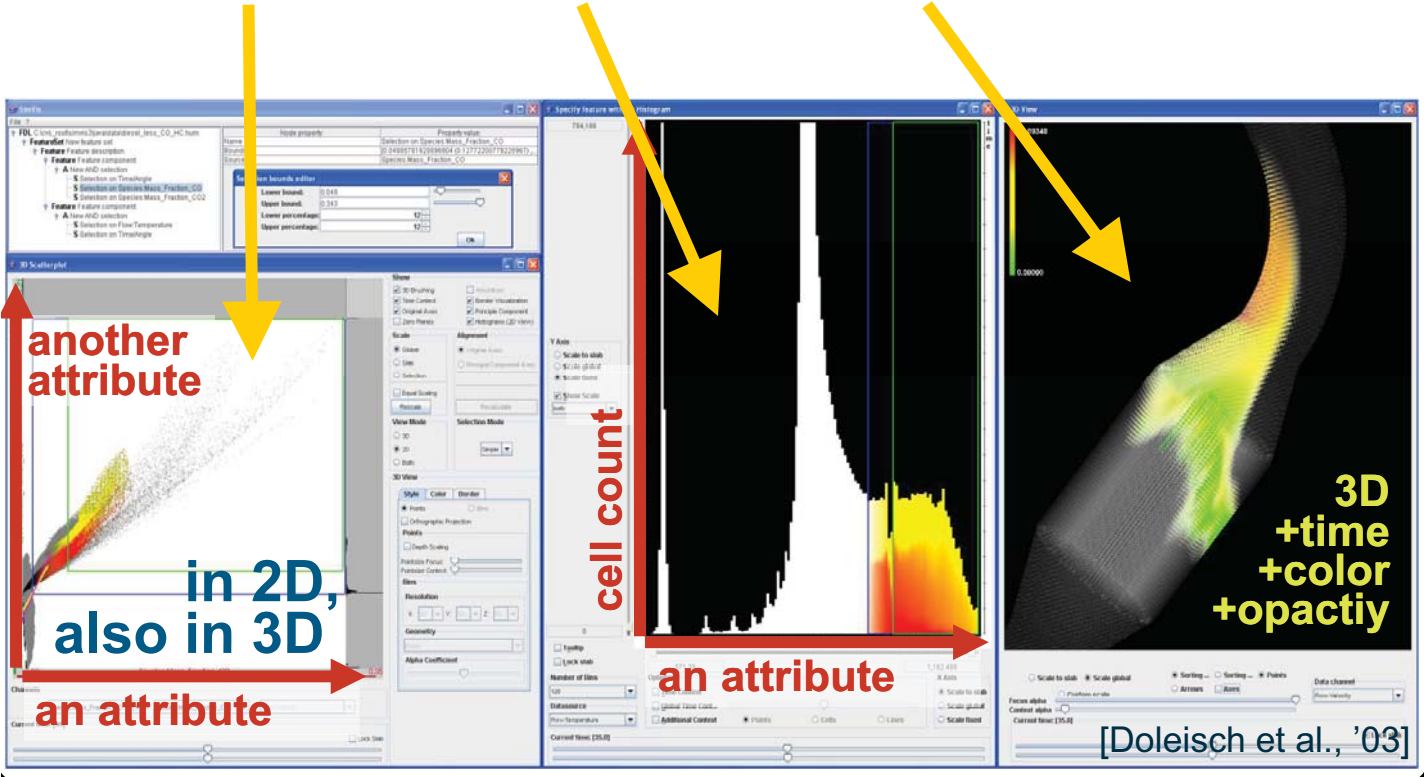
... requires ...

... is realized via ...

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

IVA: Multiple Views

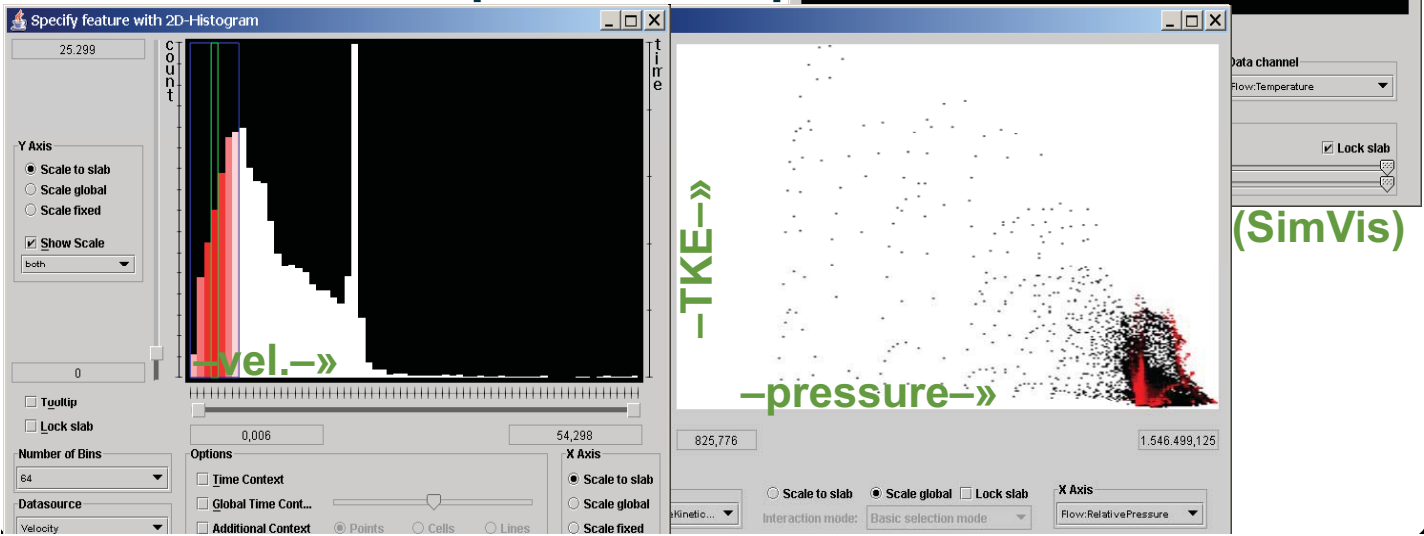
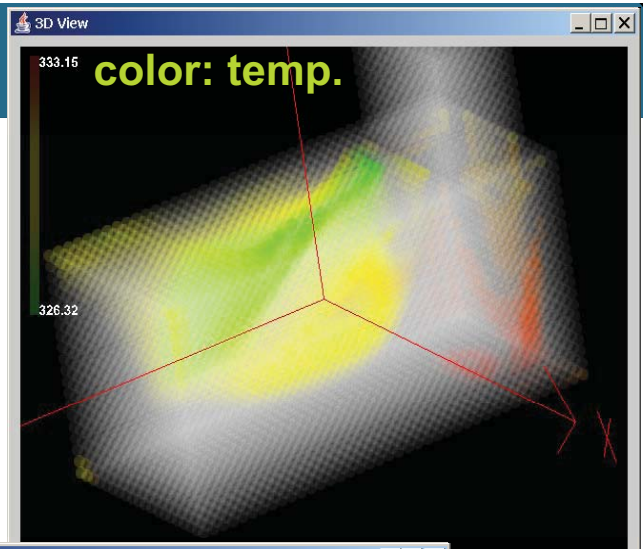
- One dataset, but multiple views
- Scatterplots, histogram, 3D(4D) view, etc.



IVA: Interactive Brushing

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

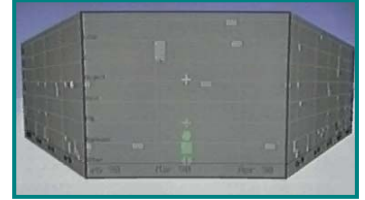
[Doleisch et al., '03]



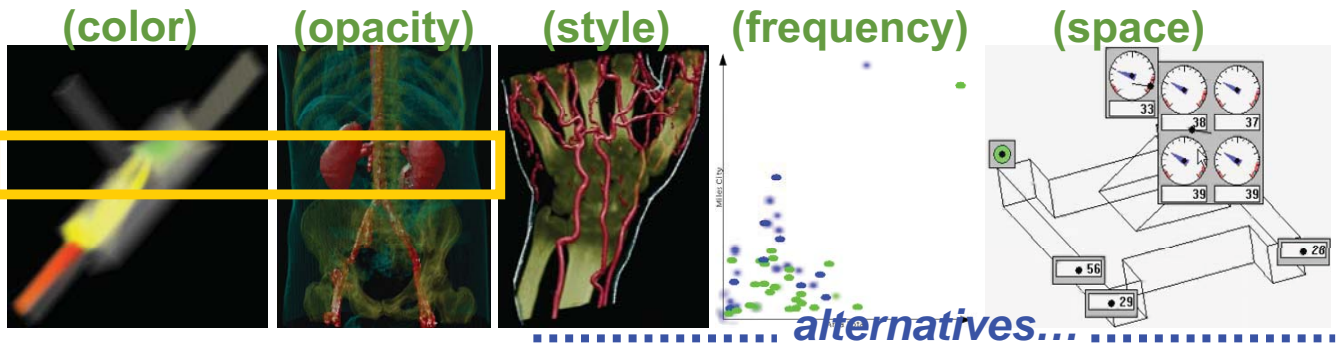
IVA: Focus+Context Visualization

- 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]

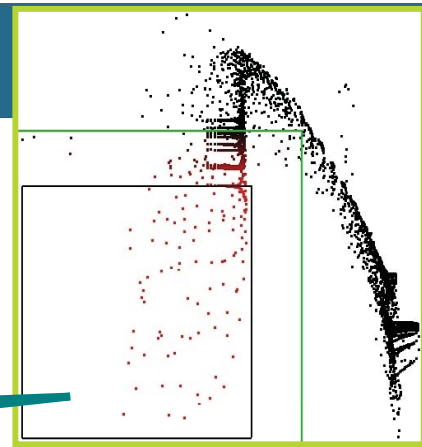


[Hauser... 2001, 2003]

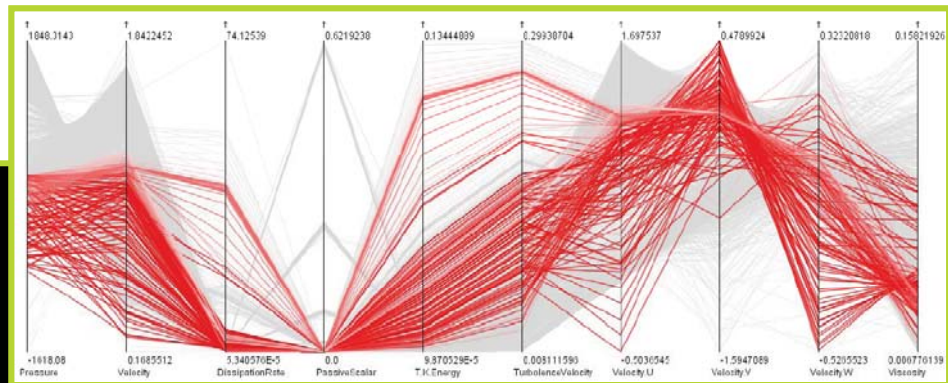


IVA: Linked Views

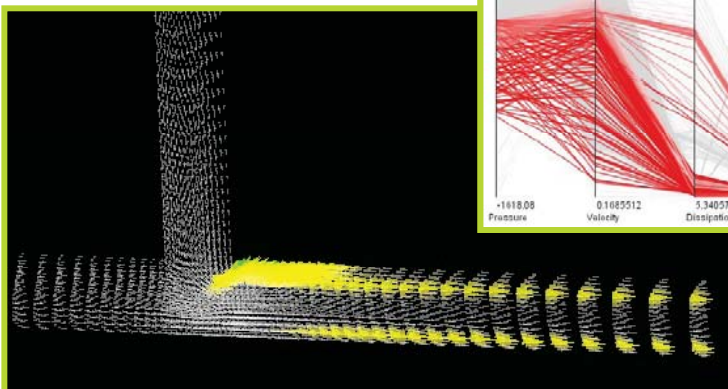
- Brushing: mark data subset as especially interesting
- Linking: enhance brushed data in linked views consistently (F+C)



(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: $0 \leq doi(tr_i) \leq 1$

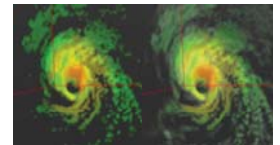
- several levels (0, low, med., ...)
- a continuous measure of interest
- a probabilistic definition of interest

(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

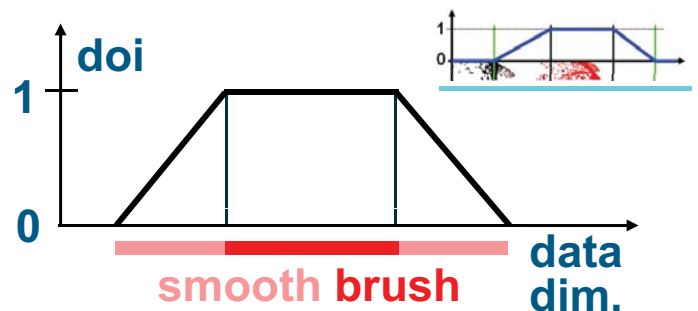
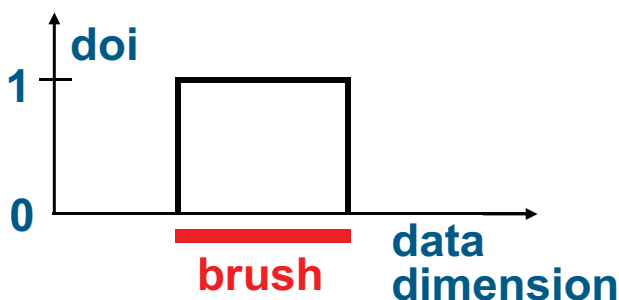
- Goes well with gradual focus+context vis. techniques (coloring, semitransparency)



Specification: **smooth brushing**

[Doleisch & Hauser, 2002]

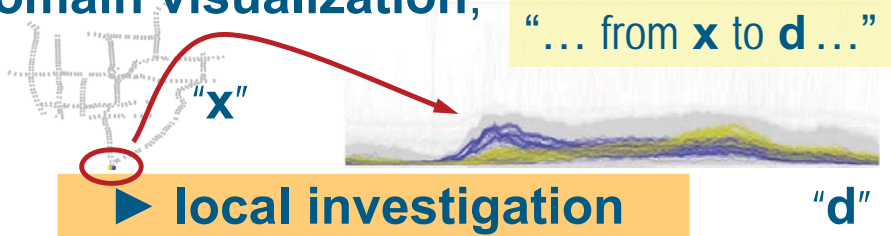
- “inner” range: all 100% interesting (DOI values of 1)
- between “inner” & “outer” range: fractional DOI values
- outside “outer” range: not interesting (DOI values of 0)



- Preliminary: domain x & range d visualized (≥ 2 views)

- **brushing on domain visualization,**

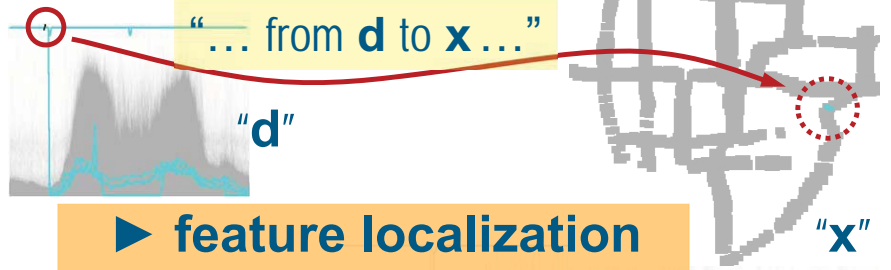
e.g., brushing special locations in the map view



1

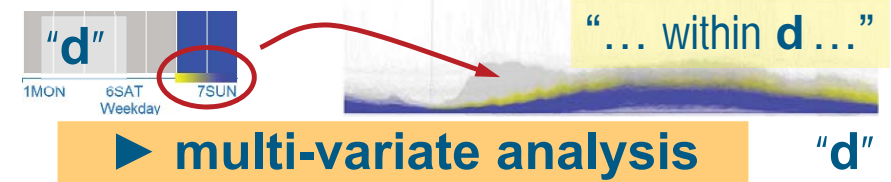
- **brushing on range visualization,**

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



2

relating multiple range variates



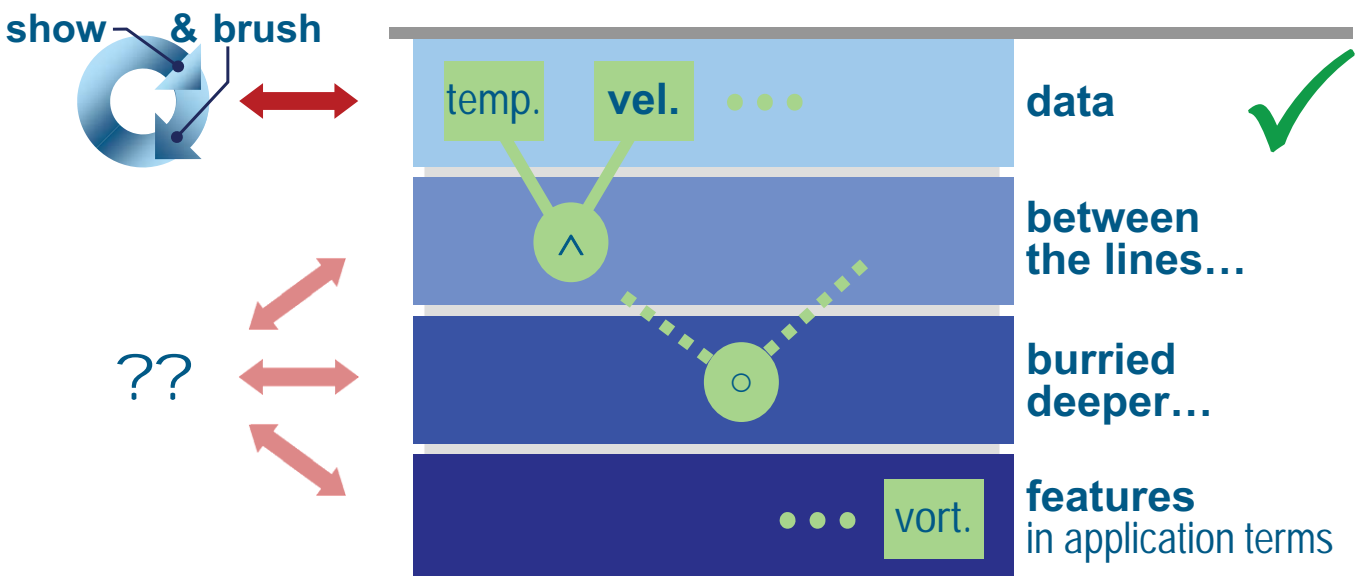
3

IVA – Levels of Complexity

(1/4)

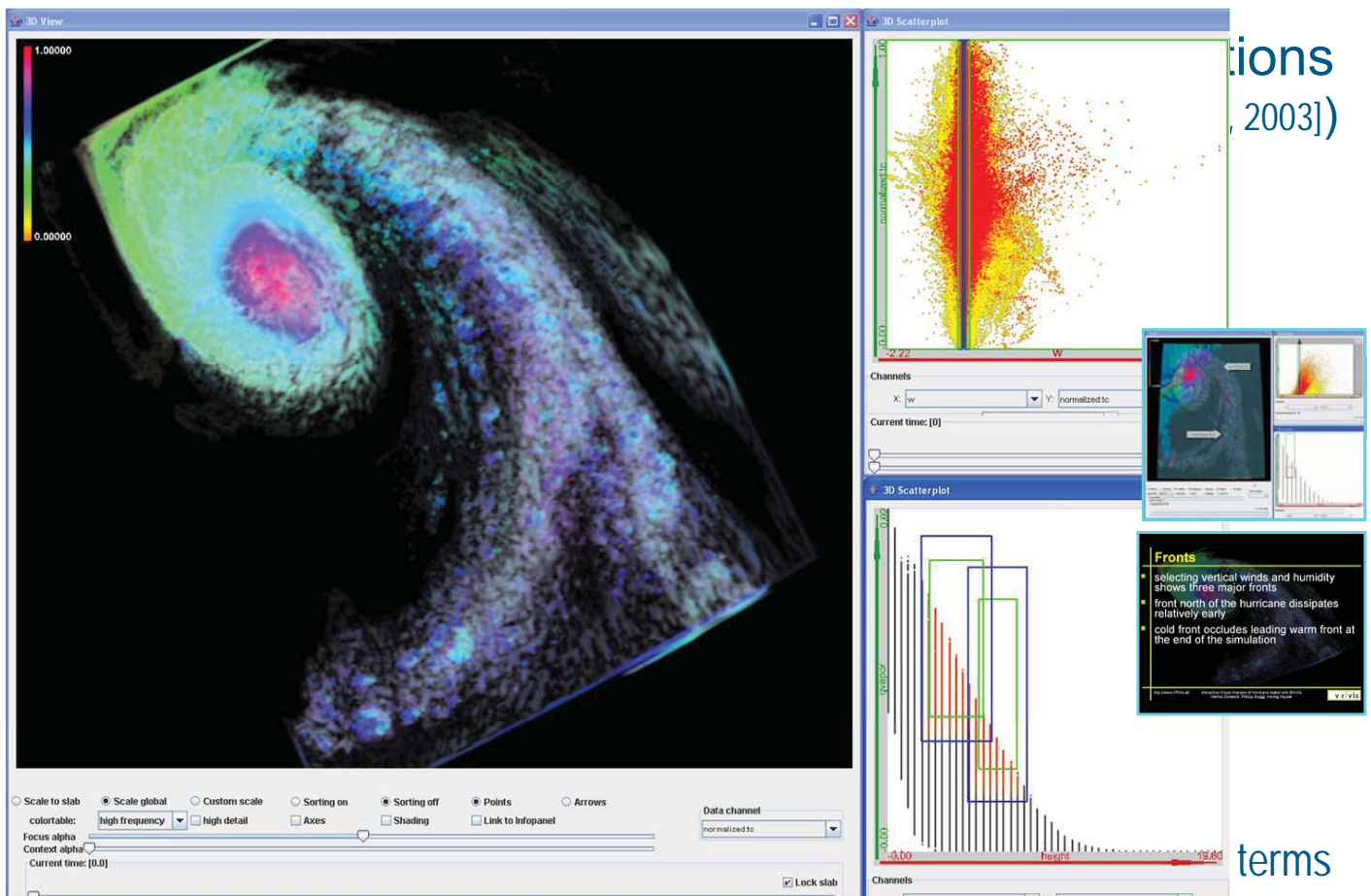
- A *lot* can be done with basic IVA, already! [pareto rule]
- We can consider a **layered information space**: from **explicitly** represented information (the **data**) to **implicitly** contained information, **features**, ...

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** (((b₀ op₁ b₁) op₂ b₂) op₃ b₃) ...)

IVA (level 2) Example



tions
(2003)

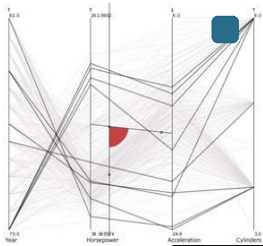
terms

- 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
 - 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

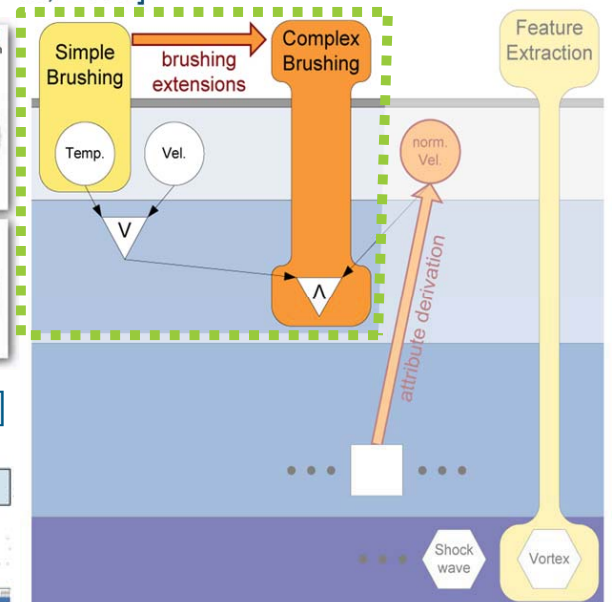
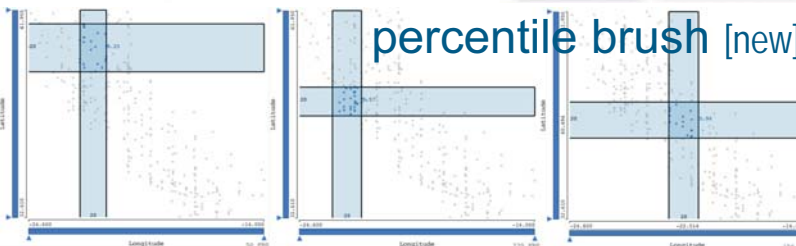
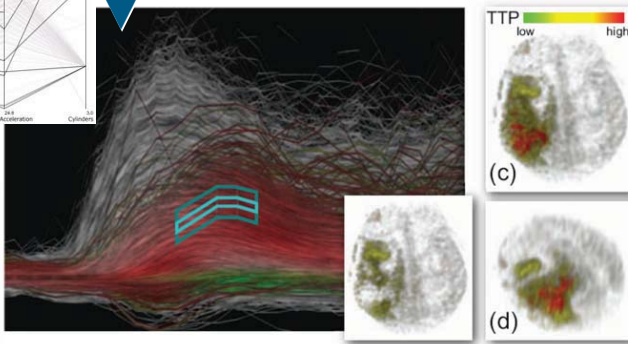
IVA (level 3): Advanced Brushing

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

Examples:

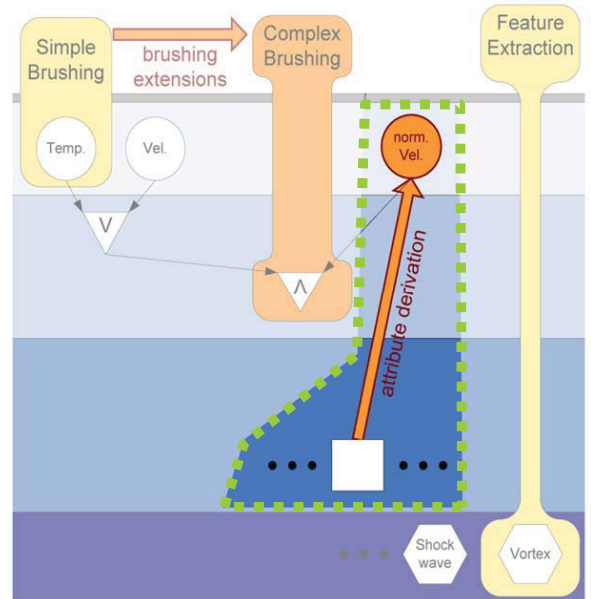


- angular brushing [Hauser et al., 2002]
- similarity brushing [Muigg et al., 2008]



IVA (level 3): Attribute Derivation

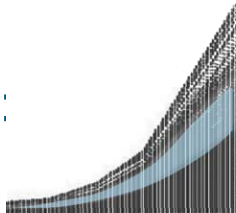
- **Principle** (in the context of iterative IVA):
 - see some data feature Φ of interest in a visualization
 - identify a **mechanism T** to describe Φ
 - **execute** (interactively!) an **attribute derivation step** to represent Φ explicitly (as new, synthetic attribute[s] d_ϕ)
 - brush d_ϕ to get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

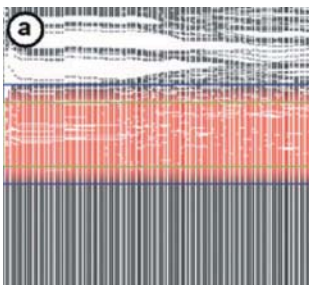


Attribute Derivation \leftrightarrow User Task / example

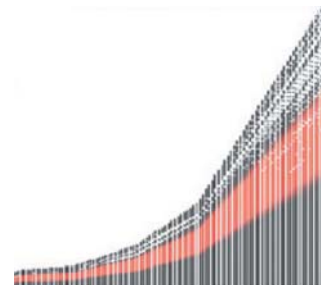
- The tools T, available in an IVA system, must reflect/match the **analytical steps of the user**:

■ Example:

- first vis.:  \leftrightarrow user wishes to select the "band" in the middle
- so? :-) an advanced brush? a lasso maybe?
- ah! \rightarrow 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 ⇒ **calculus**
 - *etc.* ⇒ **data mining**
(fast enough?)
- **Common characteristic** here:
 - **questions/tools generic**, not application-dependent!

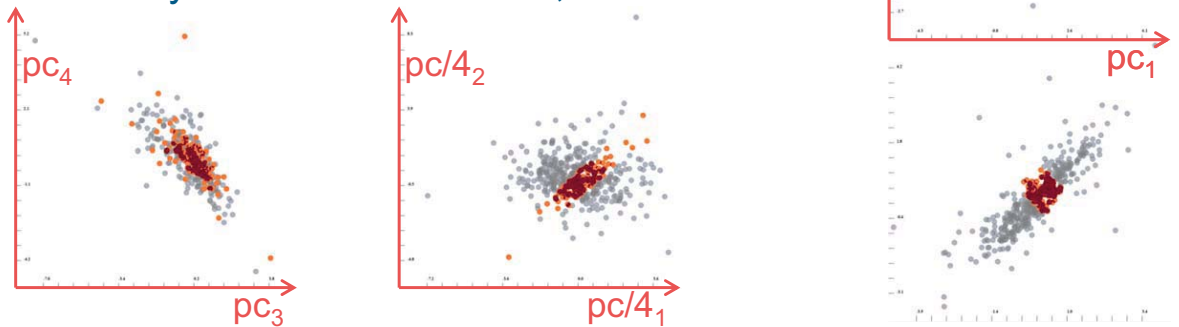
- From **analysis, calculus, num. math**:
 - **linear filtering** (convolve the data with some linear filter on demand, e.g., to smooth, for derivative estimation, etc.)
 - **calculus** (use an interactive formula editor for computing simple relations between data attributes; +, -, ·, /, etc.)
 - **gradient 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 3rd-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_{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 τ_{change}** (computed like above) is then less sensitive to temperature fluctuations

■ The Iterative Process of 3rd-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

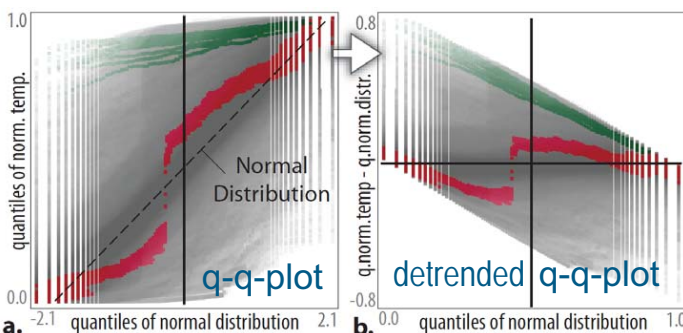
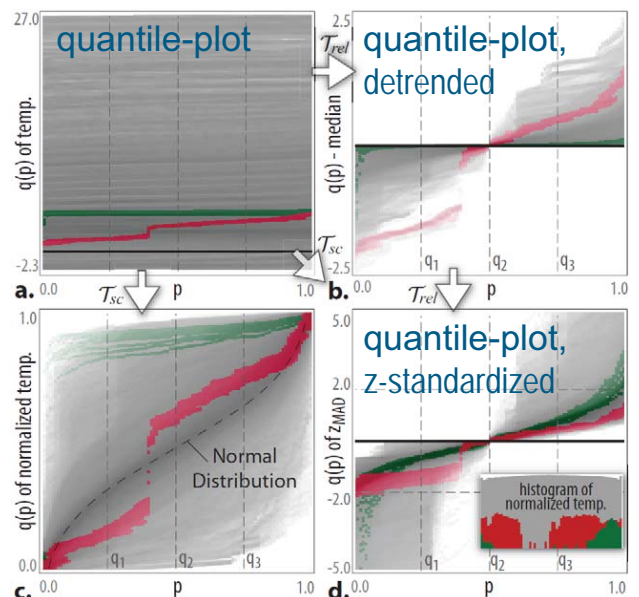
[Kehrer et al., TVCG 2011]

■ Useful statistical measures include:

- moments (μ, σ, \dots), **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):

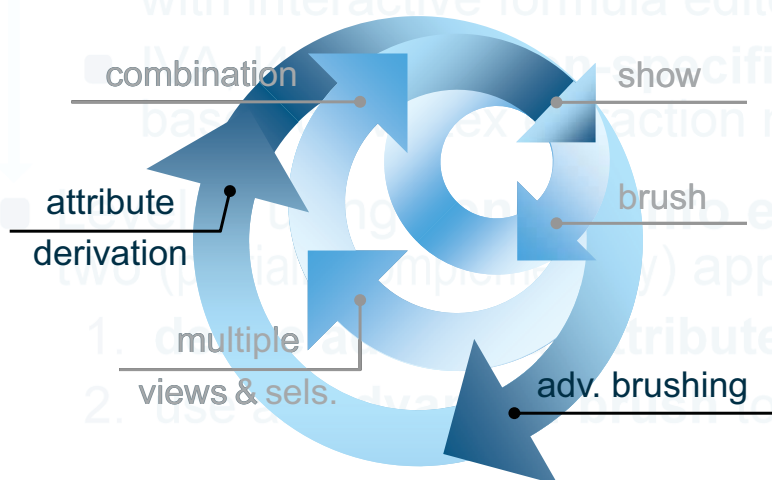
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feature extraction, e.g., methods for flow analysis

traction mechanisms, approaches:

1. show & brush select “hidden” relations

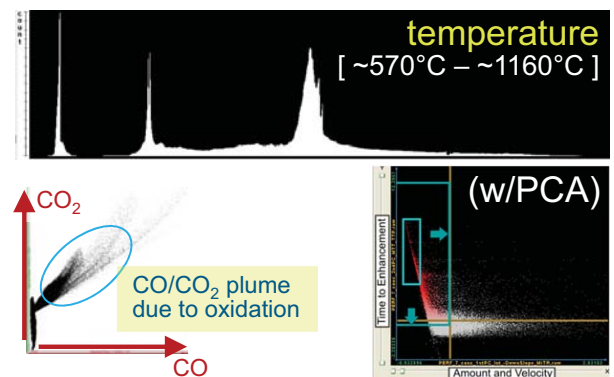


- A lot can be done with KISS-principle IVA! (pareto rule) ✓
- For more advanced exploration/analysis tasks, we extend it (in several steps):
 - IVA, level 2: **logical combinations of brushes** utilizing the *feature definition language* [Dreisch et al., 2009] ✓
 - IVA, l. 3: **attribute derivation;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

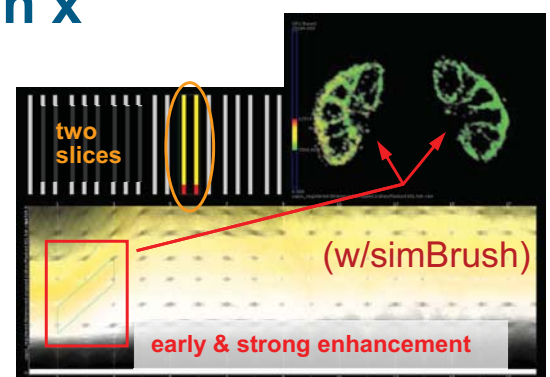
Understanding data wrt. range d

- 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.*!
- Collaboration partners (St. Oeltze, Fl. Ladstädter, G. Weber, *et al.*)
- All around SimVis and ComVis and ...
- Funding partners (FFG, AVL, EU, UiB, ...)