

Visualization and Visual Analysis of Multi-faceted Scientific Data: A Survey

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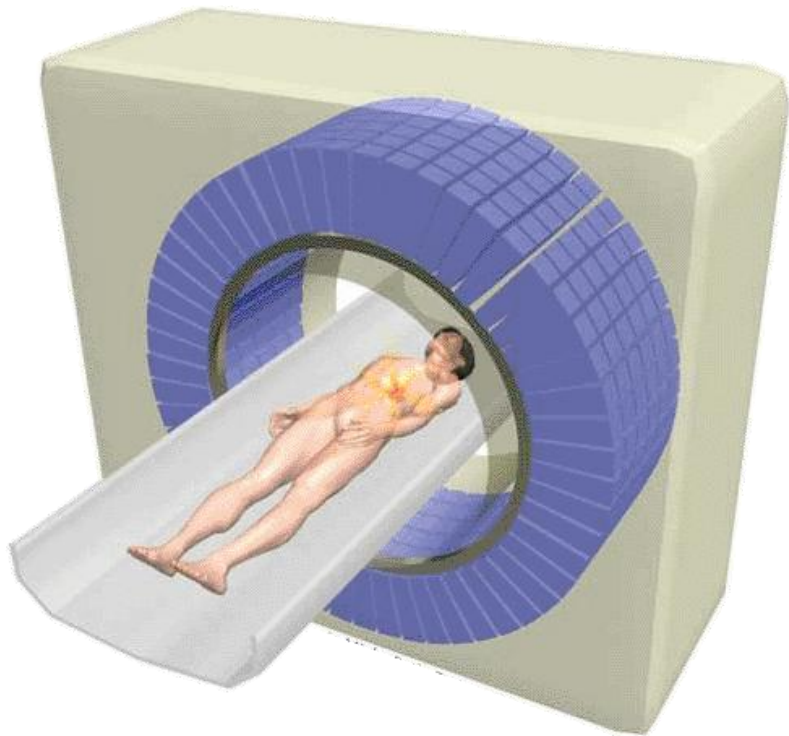
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Motivation

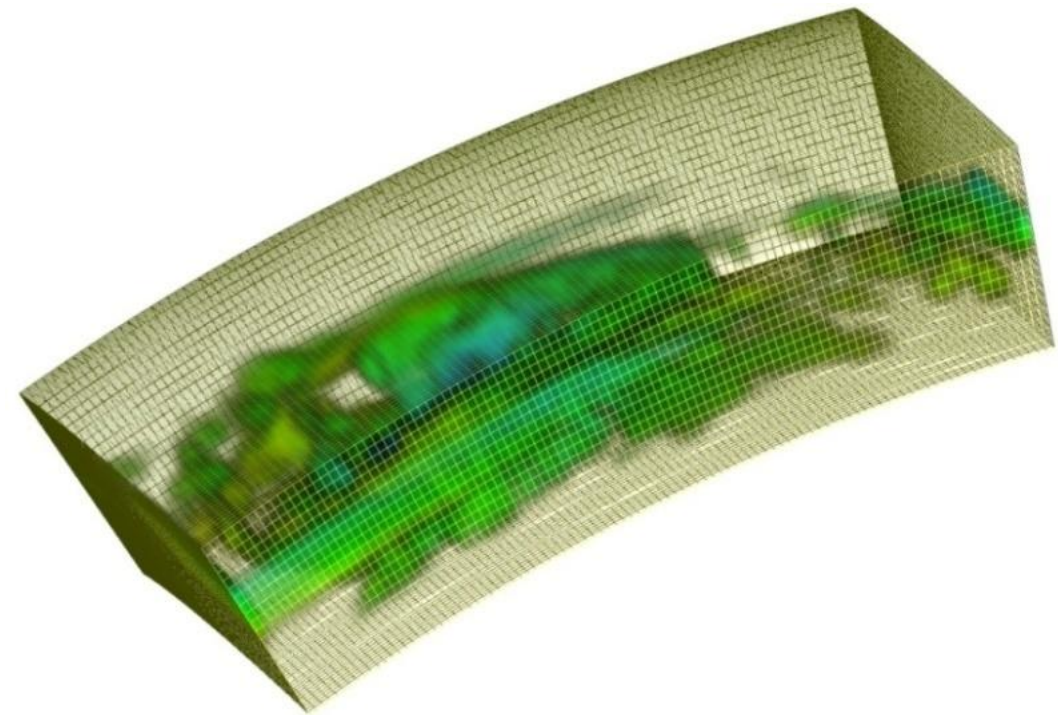
Increasing amounts of scientific data



medical scanner



computational simulation



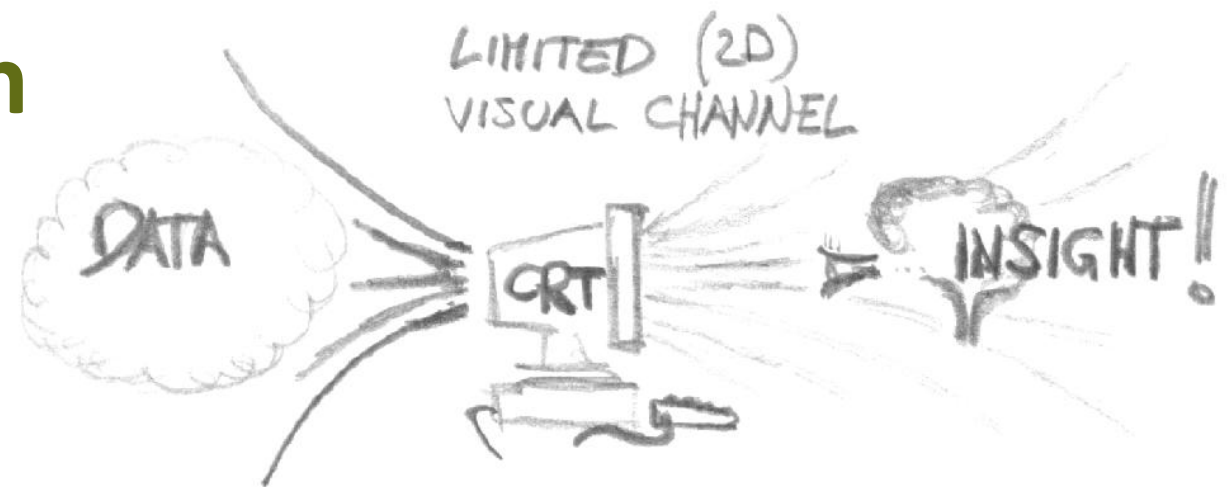
time-dependent
3D data

Hard to analyze and understand

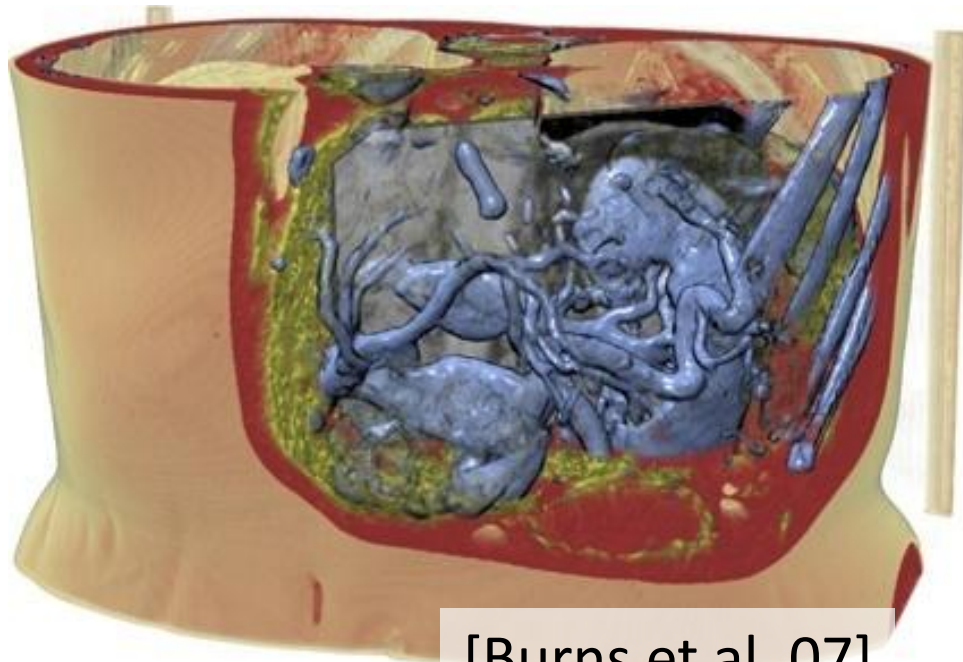
Visualization

“The purpose of visualization is insight, not pictures”

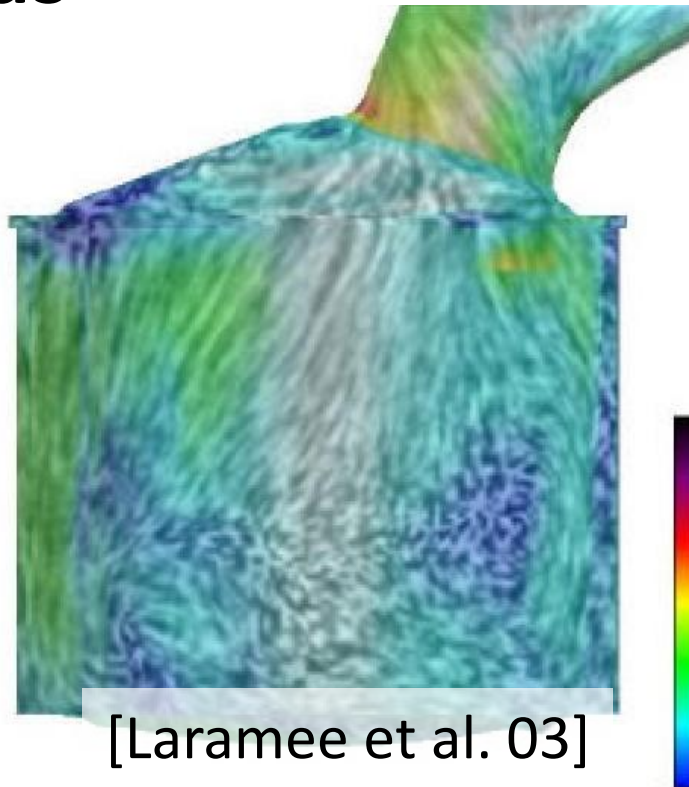
[Shneiderman '99]



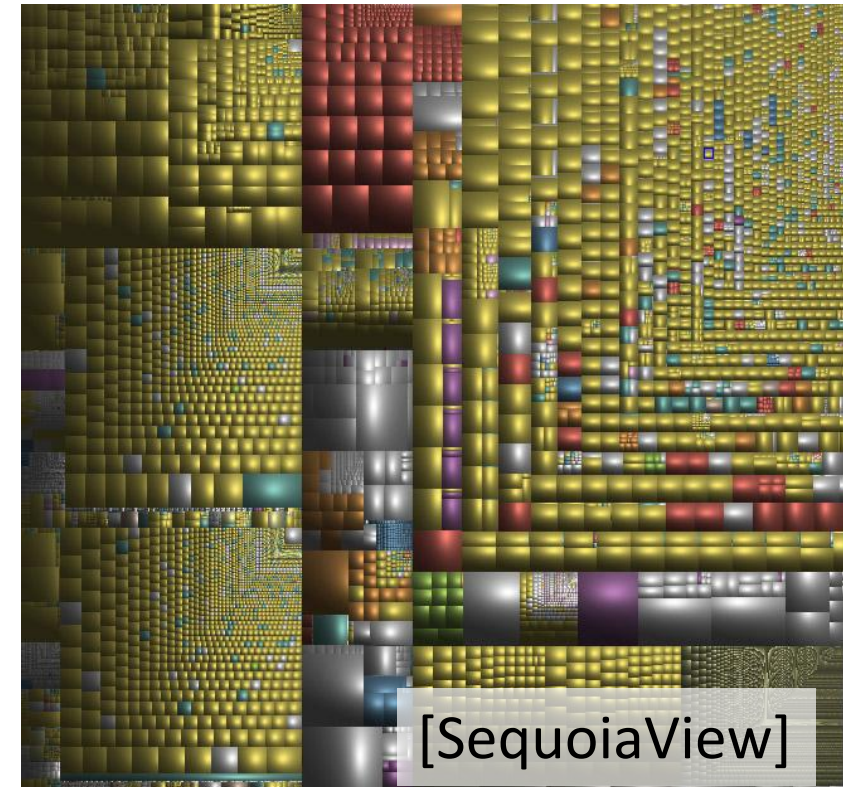
Different application areas



[Burns et al. 07]



[Laramée et al. 03]



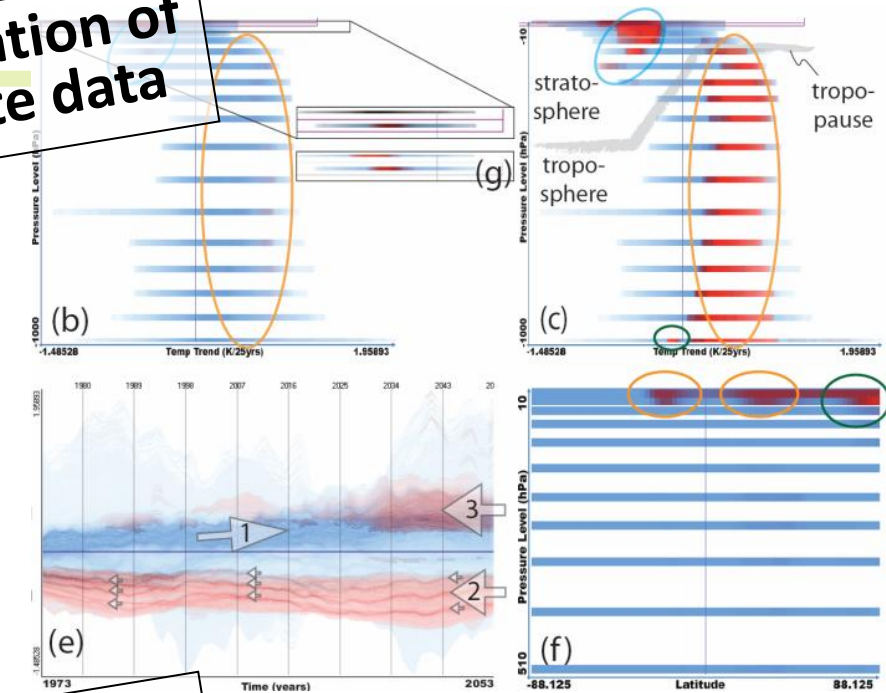
[SequoiaView]

Typical Visualization Tasks

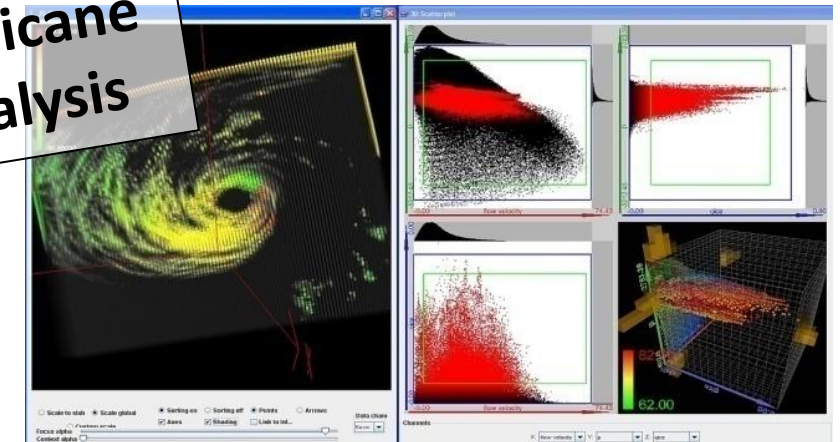
Visualization is good for

- **visual exploration**
 - find unknown/unexpected
 - generate new hypothesis
- **visual analysis** (confirmative vis.)
 - verify or reject hypotheses
 - information drill-down
- **presentation**
 - show/communicate results

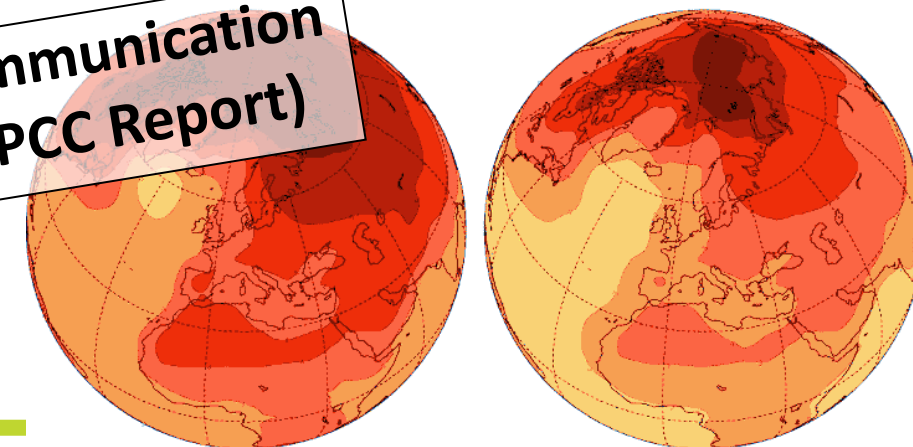
exploration of
climate data



hurricane
analysis

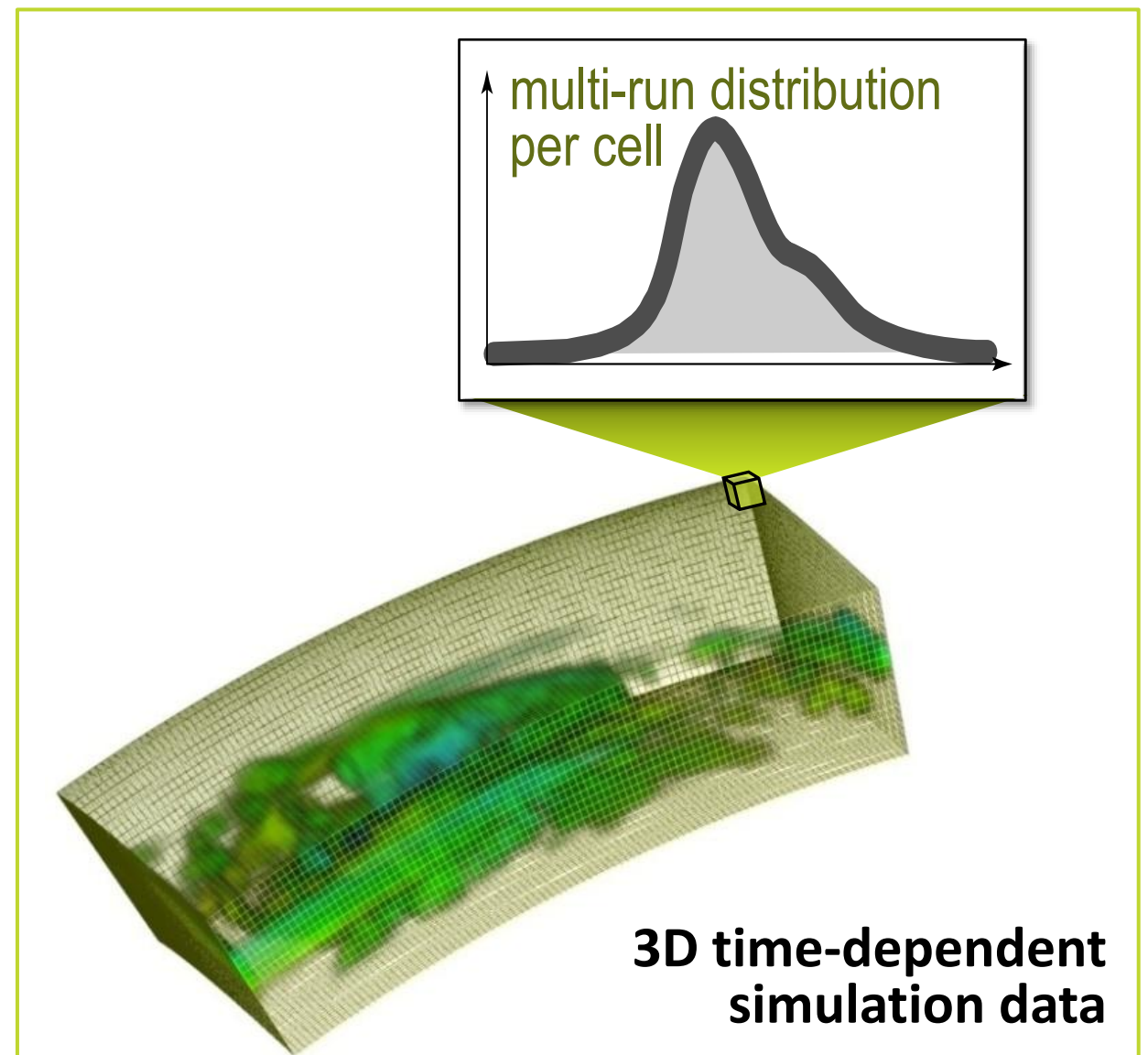
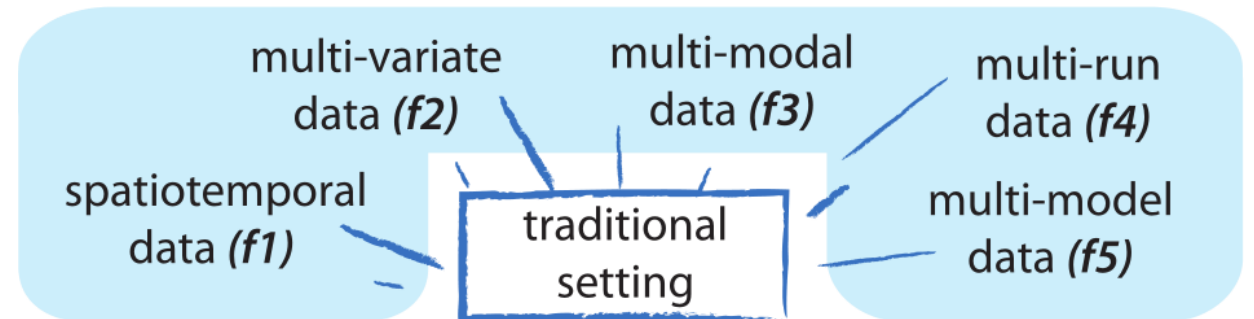


communication
(IPCC Report)

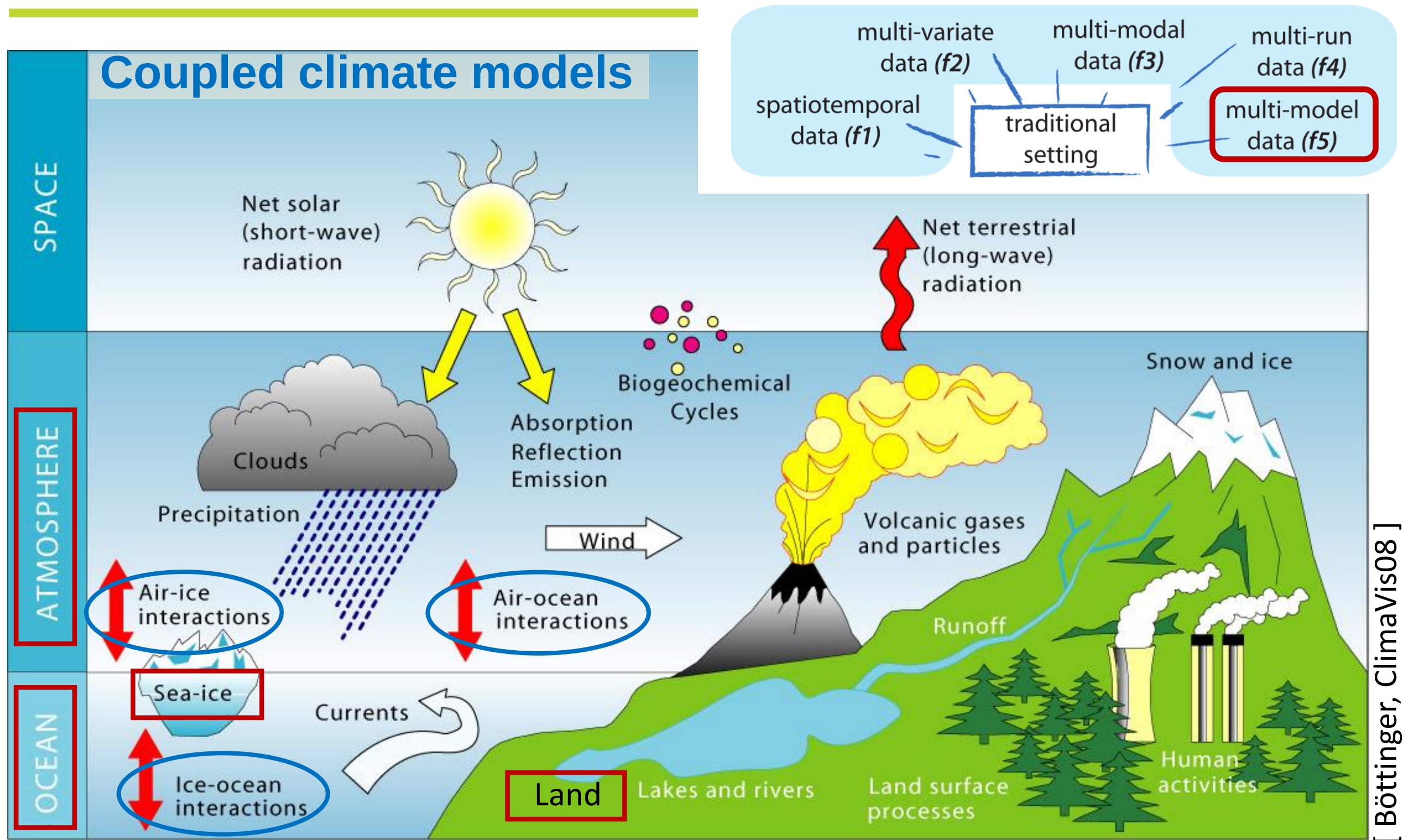


Multi-faceted Scientific Data

- **Spatiotemporal data**
- **Multi-variate/multi-field data**
(multiple data attributes, e.g., temperature or pressure)
- **Multi-modal data**
(CT, MRI, large-scale measurements, simulations, etc.)
- **Multi-run/ensemble simulations** (repeated with varied parameter settings)
- **Multi-model scenarios**
(e.g., coupled climate model)

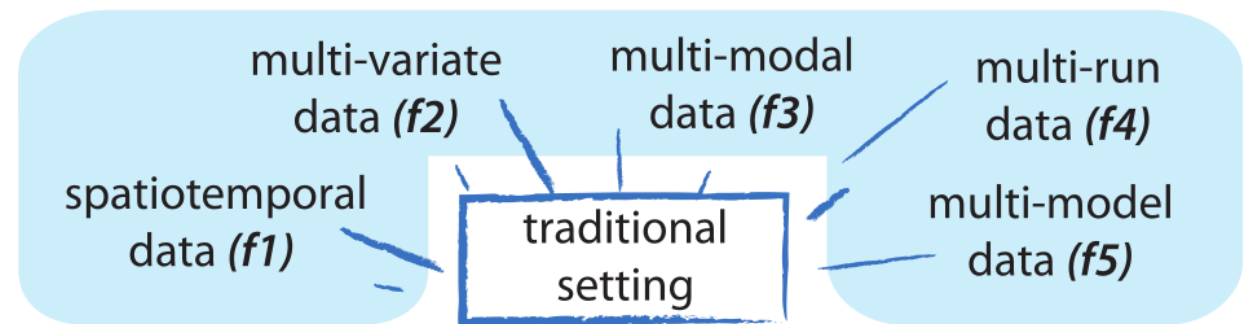


Multi-faceted Scientific Data



[Böttinger, ClimaVis08]

Categorization

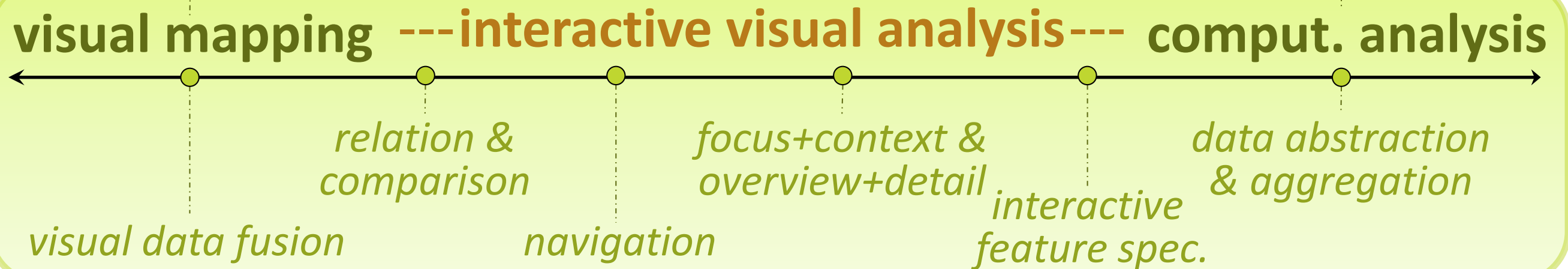


- Literature review of 200+ papers on scientific data
- How are vis., interaction, and comput. analysis combined?

how to represent
the data

interaction concepts
(linking & brushing, zooming,
view reconfiguration, etc.)

what are main
characteristics /
features



[compare to Keim et al. 09;
Bertine & Lalanne 09]

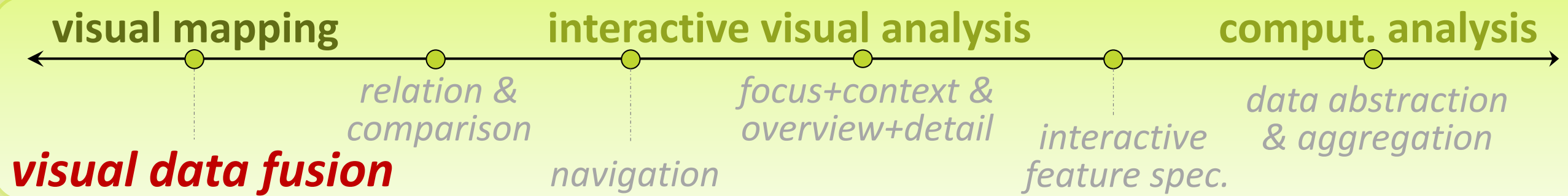
Visual vs. Computational Analysis

■ Interactive Visual Analysis

- + user-guided analysis possible
- + detect interesting features without looking for them
- + understand results in context
- + uses power of human visual system
- human involvement not always possible or desirable (expensive!)
- limited dimensionality
- often only qualitative results
- (still) often unfamiliar

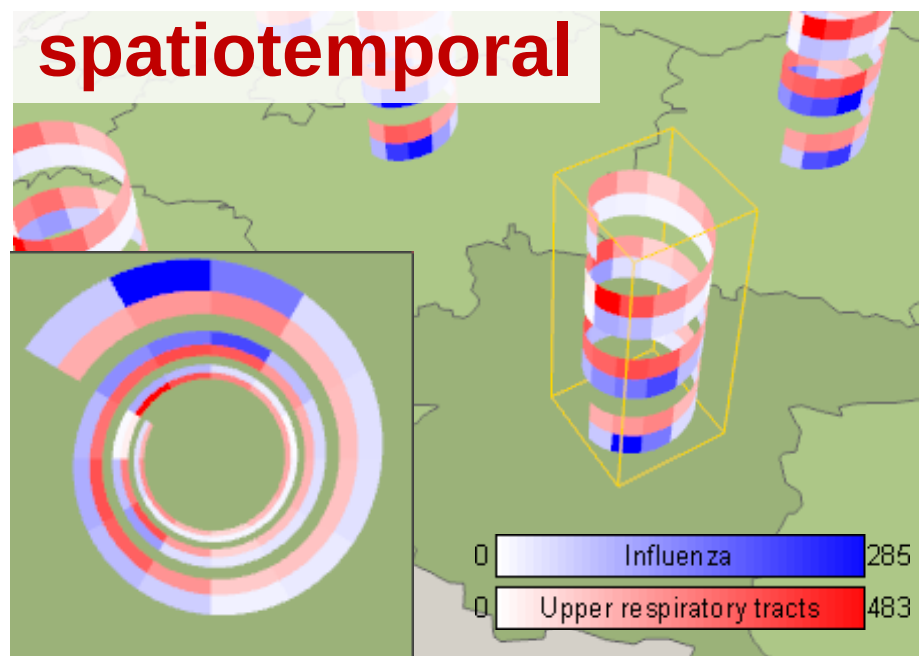
■ Automated Data Analysis

- needs precise definition of goals
- limited tolerance of data artifacts
- result without explanation
- computationally expensive
- + hardly any interaction required (after setup)
- + scales better w.r.t. many dimensions
- + precise results
- + long history (mostly statistics)

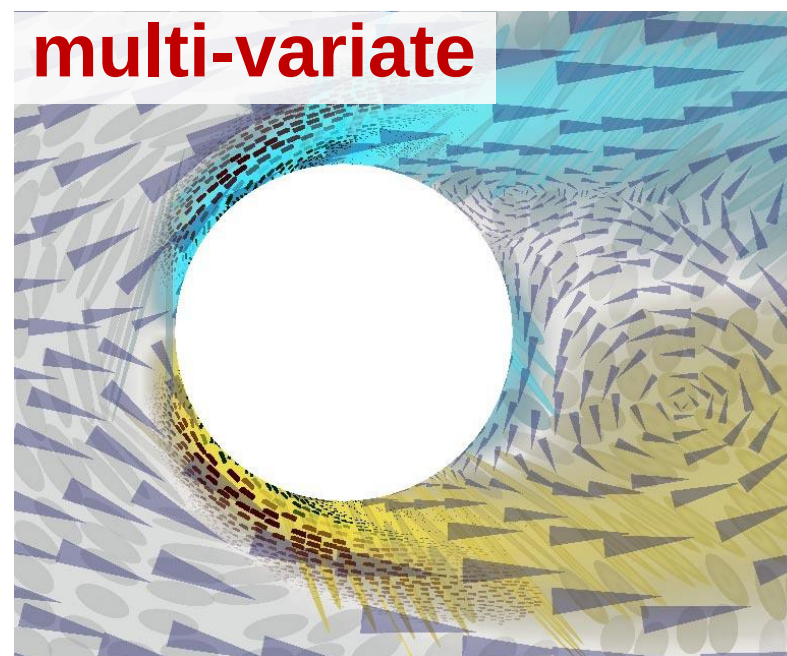


Fusion within a single visualization

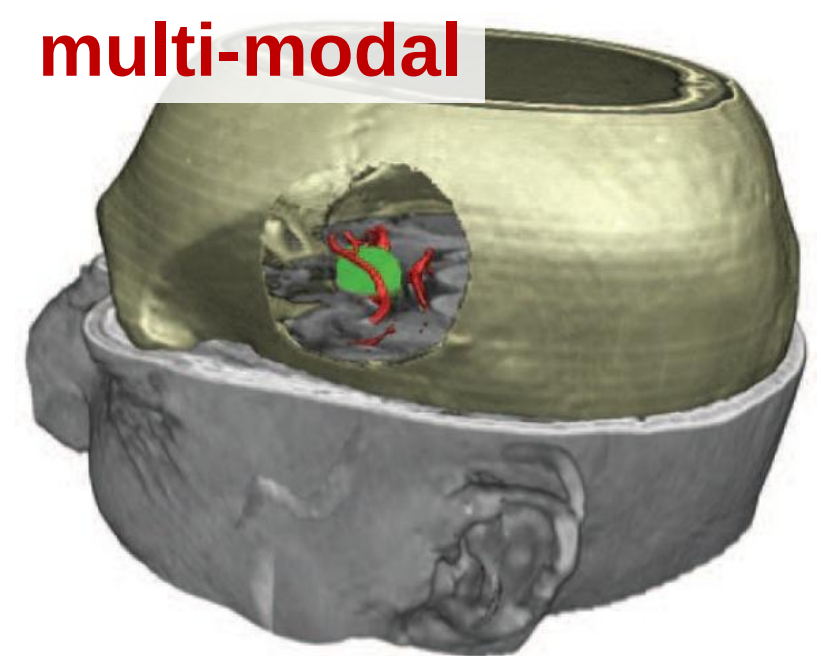
- common frame of reference
- layering techniques (e.g., glyphs, color, transparency)
- multi-volume rendering (coregistration, segmentation)



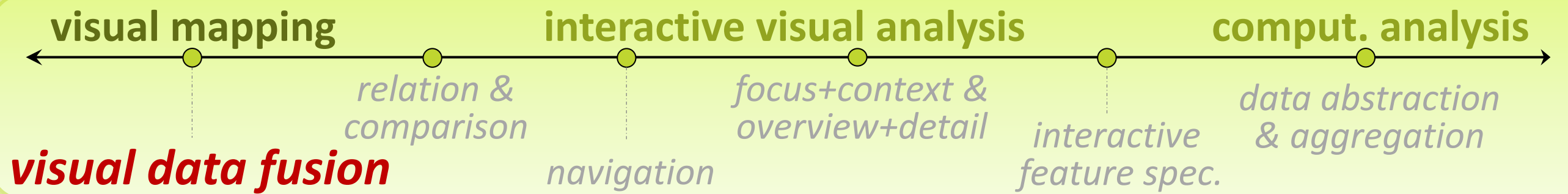
Helix glyphs [Tominski et al. 05]



Layering [Kirby et al. 99]

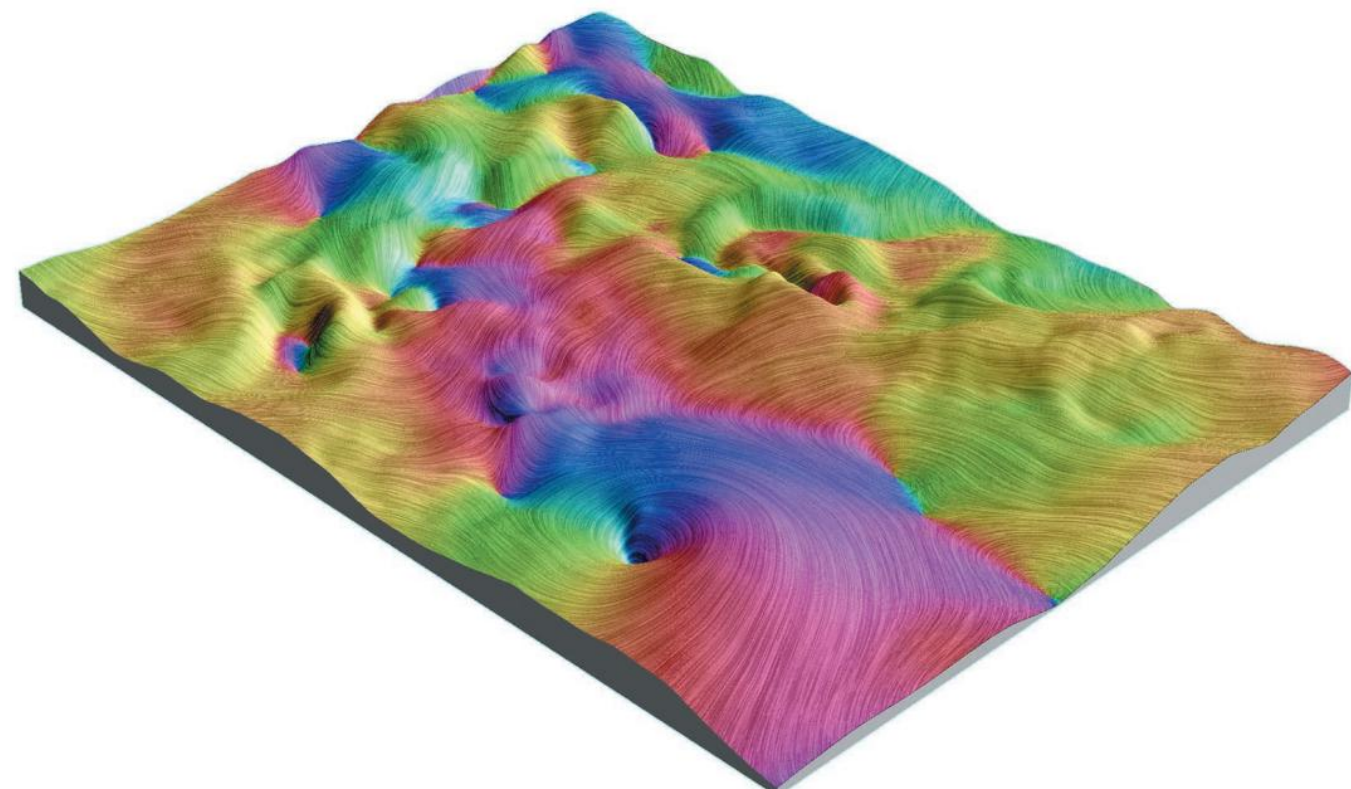
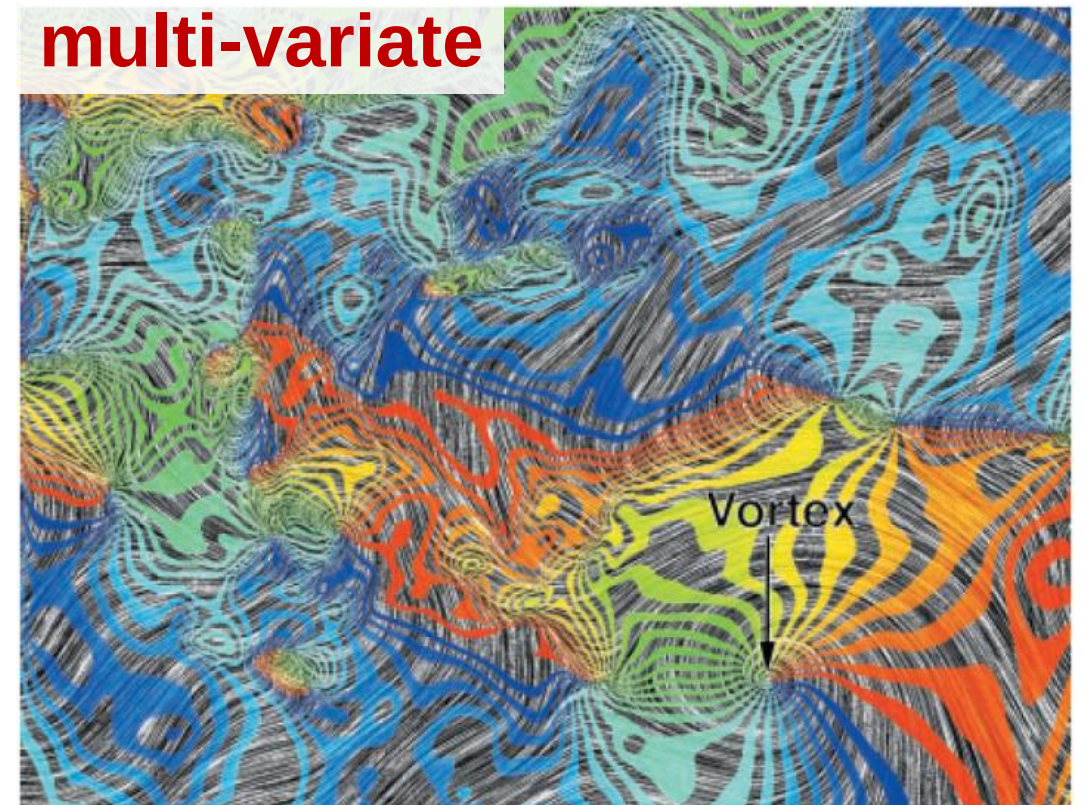
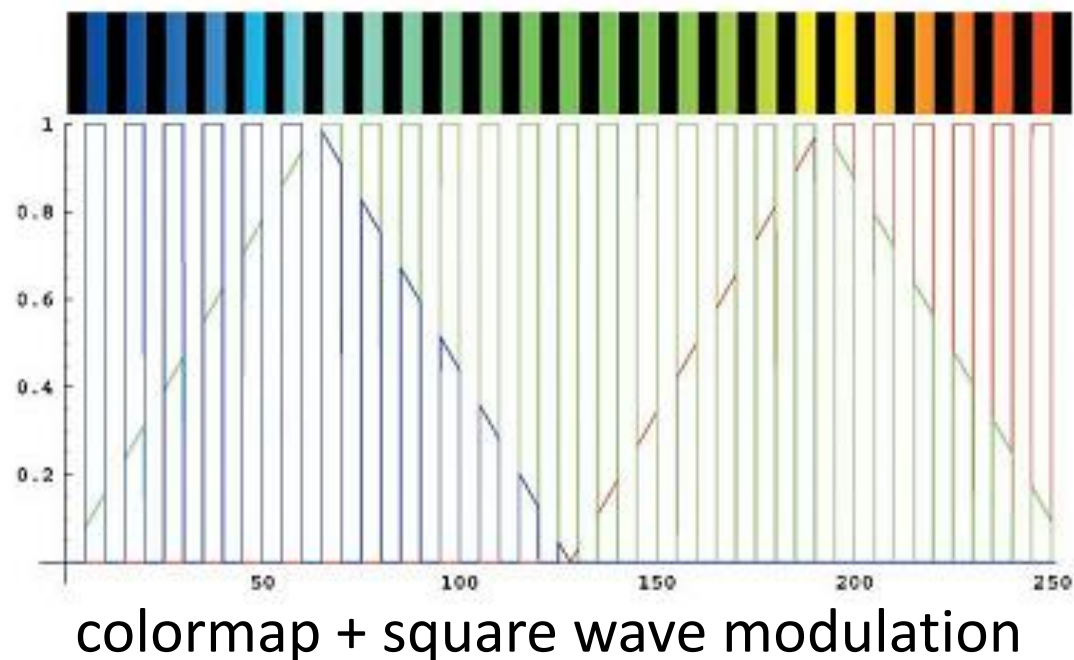


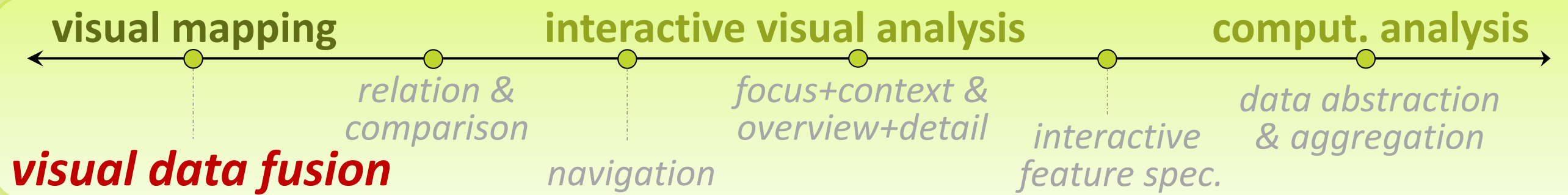
Multi-volume rendering
[Beyer et al. 07]



Layering techniques [Wong et al. 02]

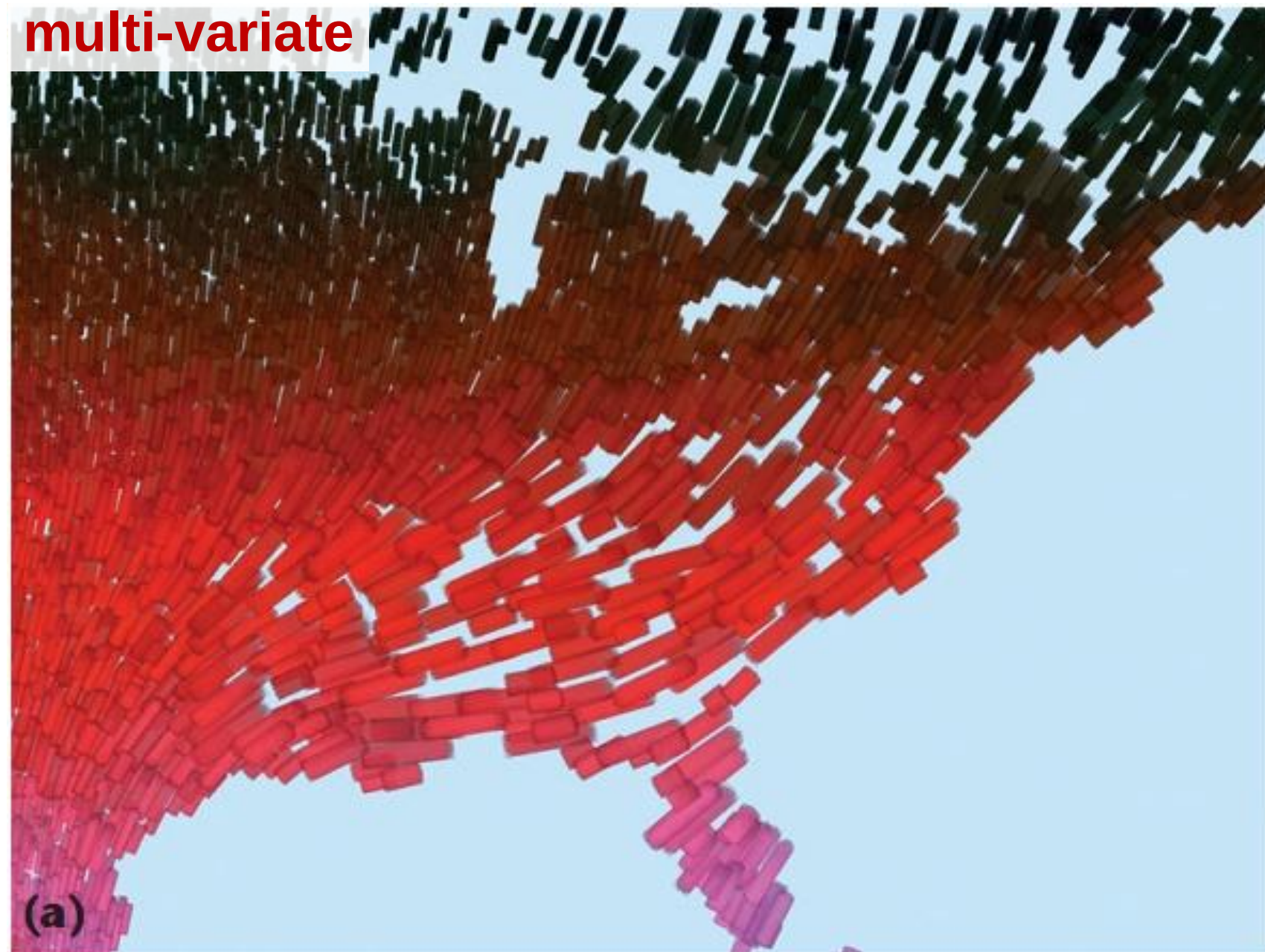
- opacity modulation
- filigreed
- colormap enhancement
- 2D heightmap

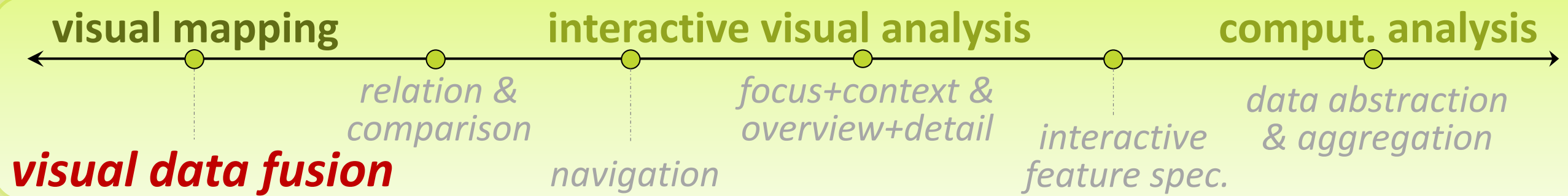




Preattentive Visual Features: Textures and Colors [Healey & Enns 02]

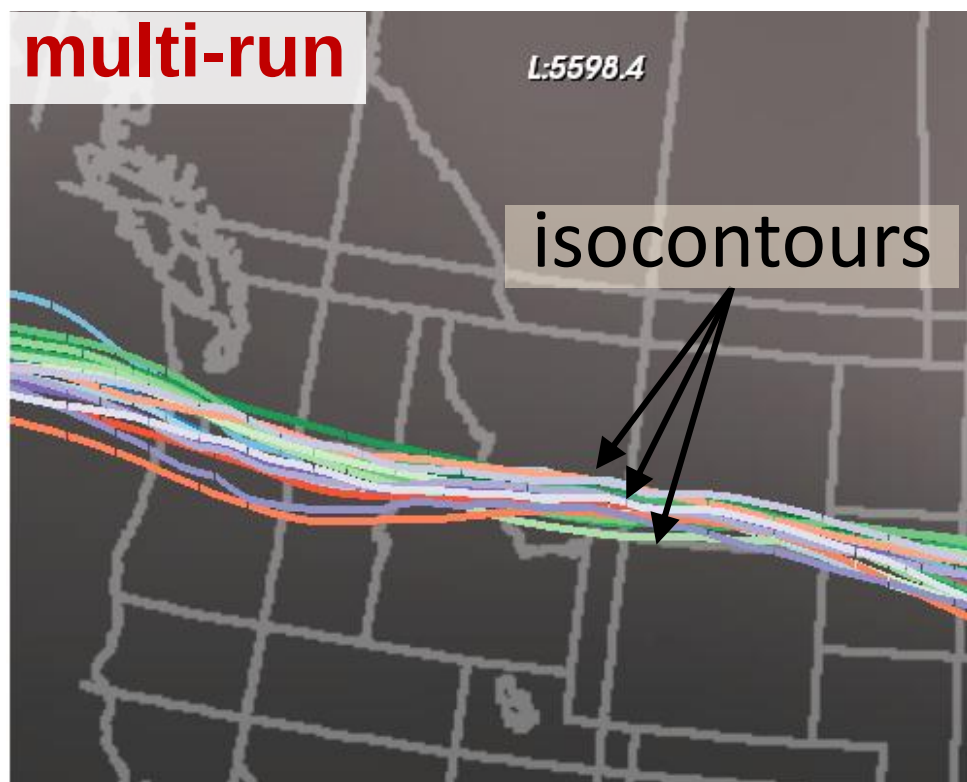
- temperature
→ color
- wind speed
→ coverage
- pressure
→ size
- precipitation
→ orientation



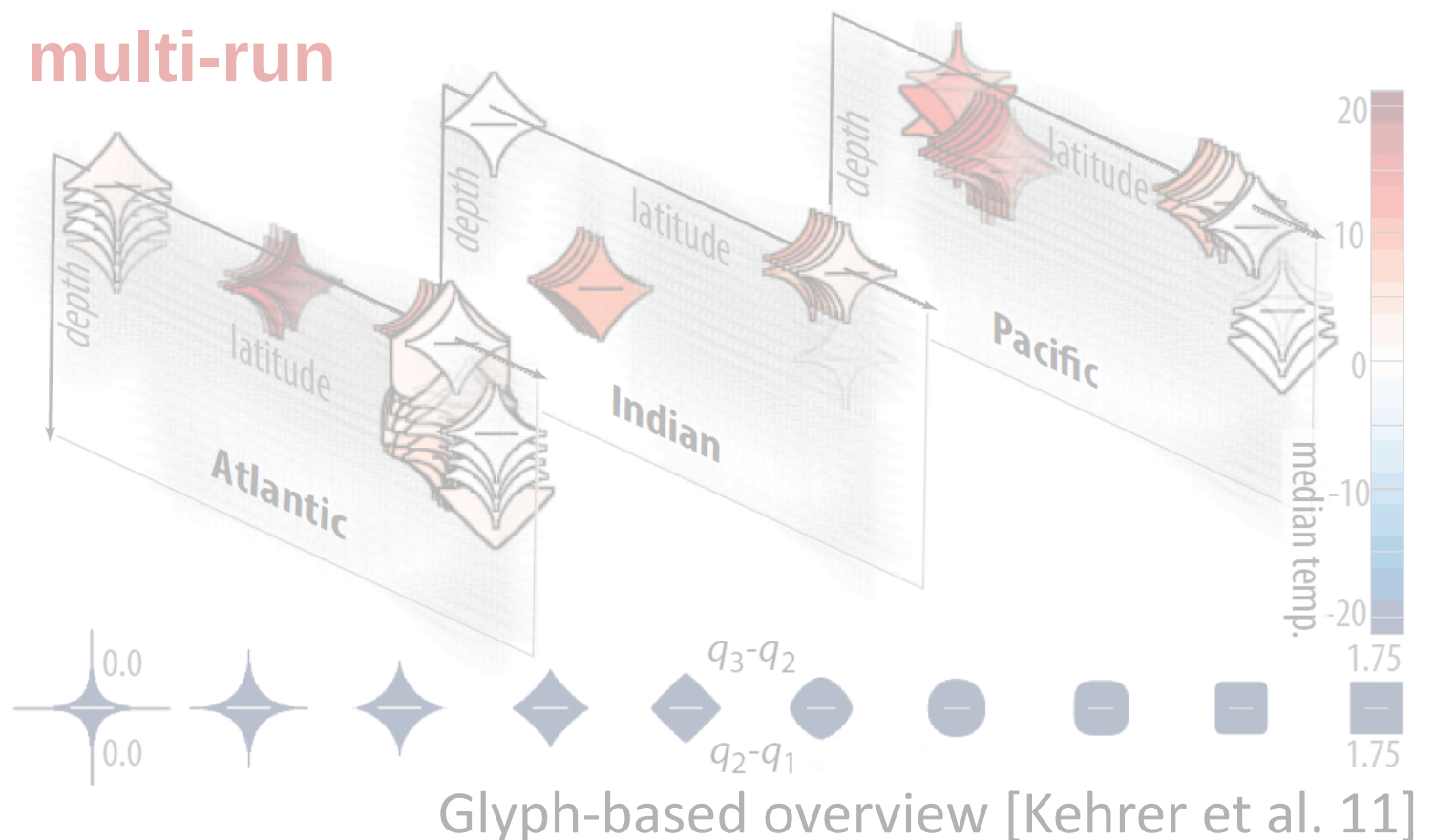


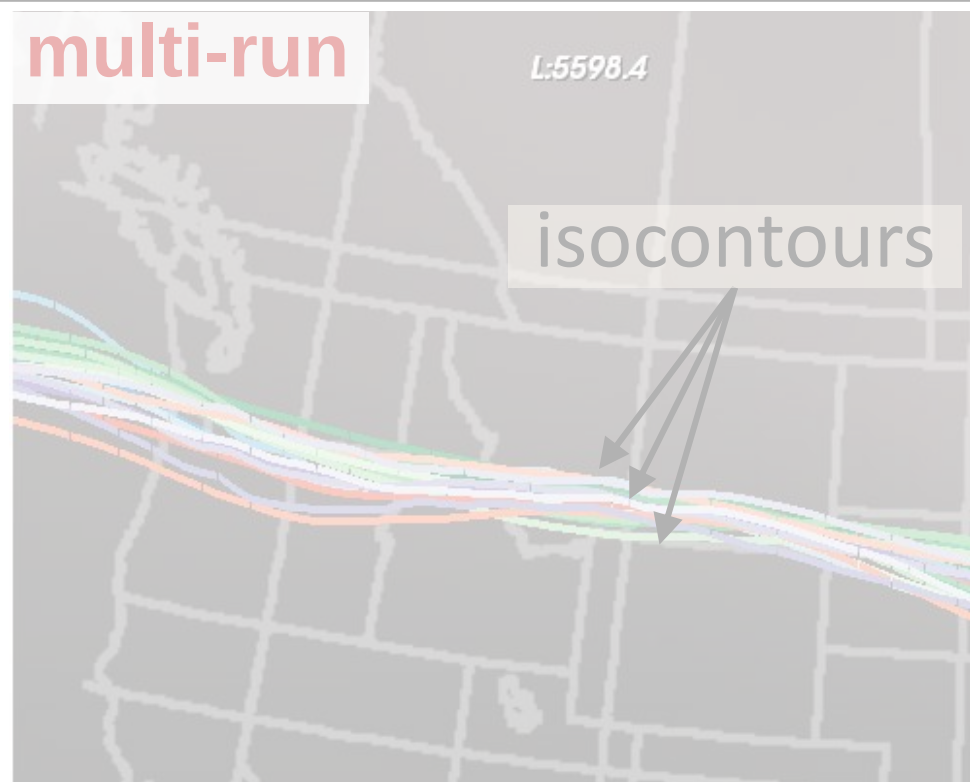
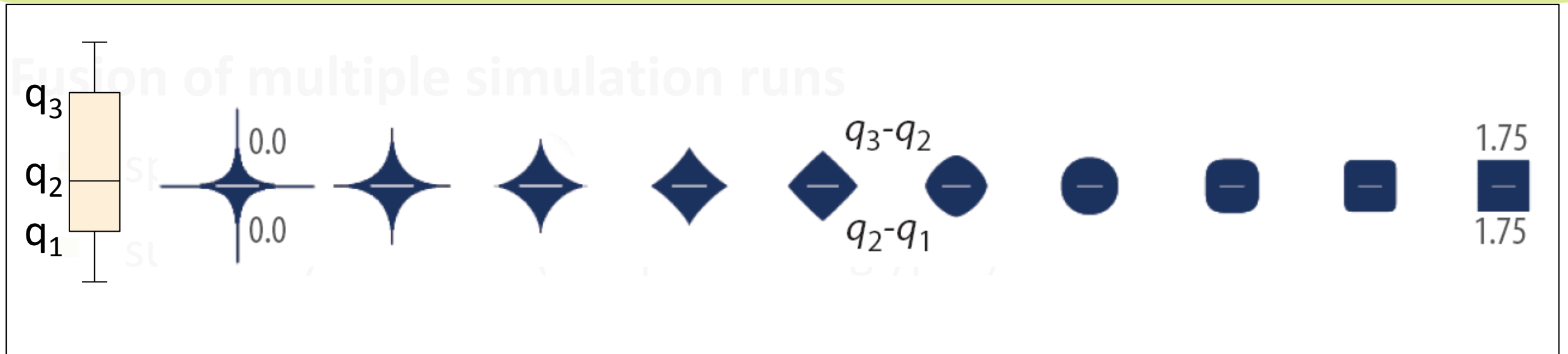
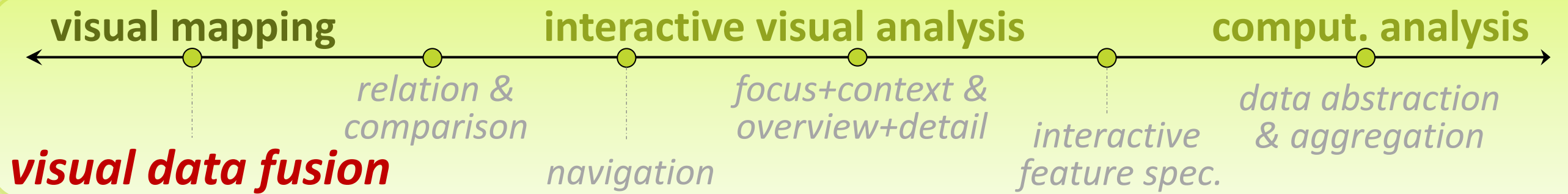
Fusion of multiple simulation runs

- spaghetti plots [Diggle et al. 02]
- summary statistics (box plots and glyphs)

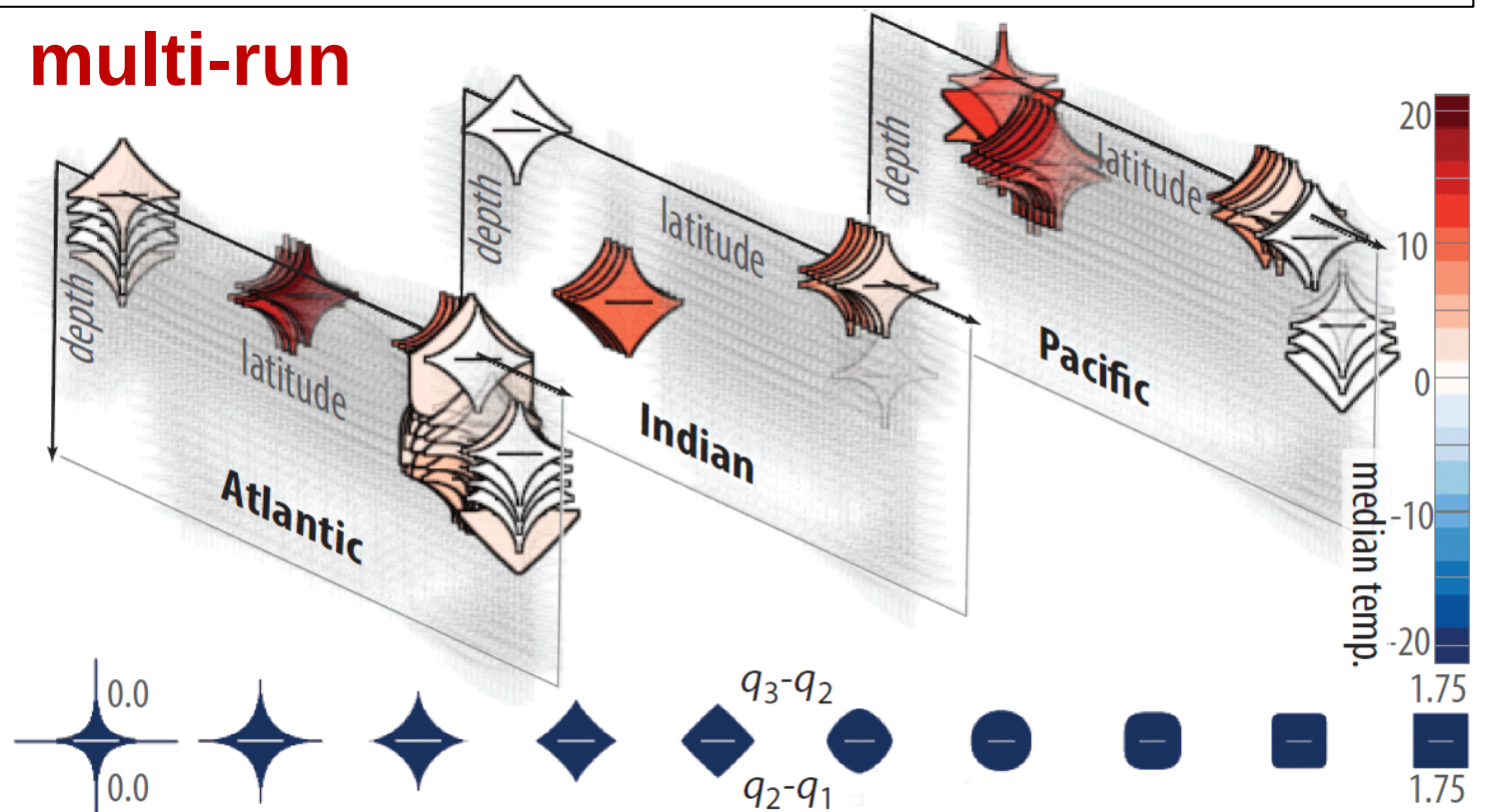


EnsembleVis [Potter et al. 09]

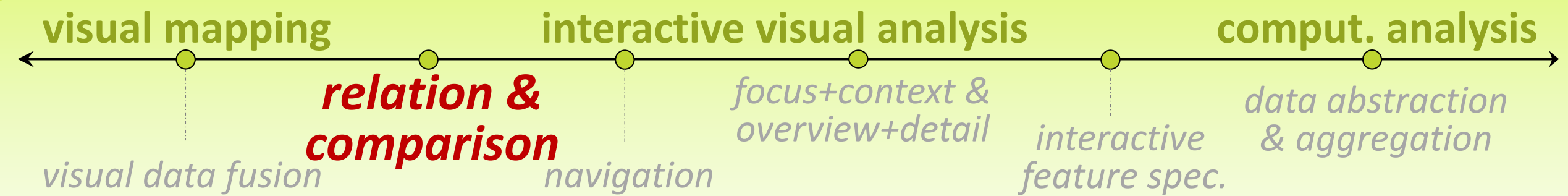




EnsembleVis [Potter et al. 09]

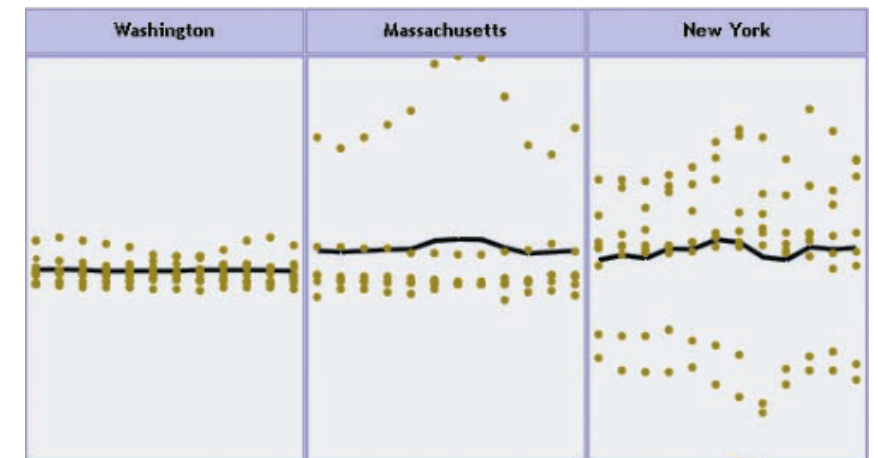


Glyph-based overview [Kehrer et al. 11]

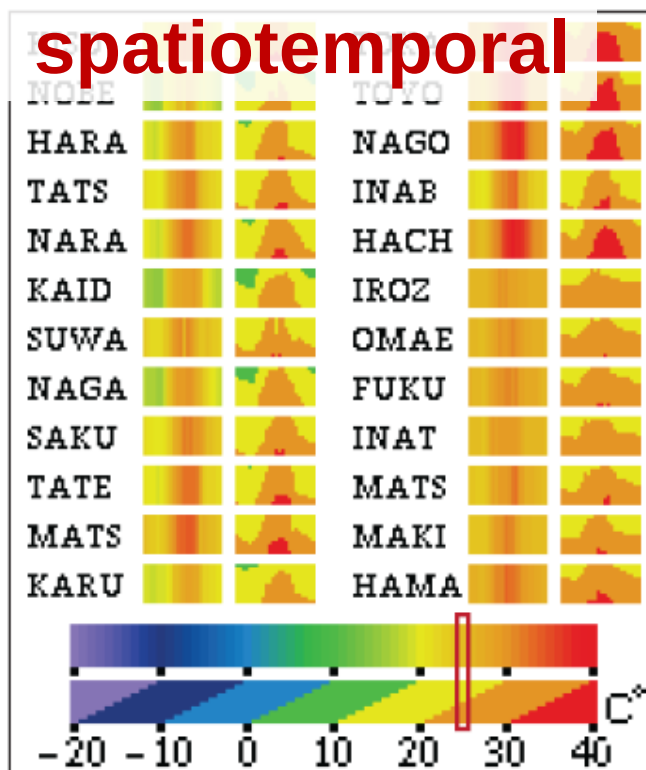


Taxonomy [Gleicher et al. 11]

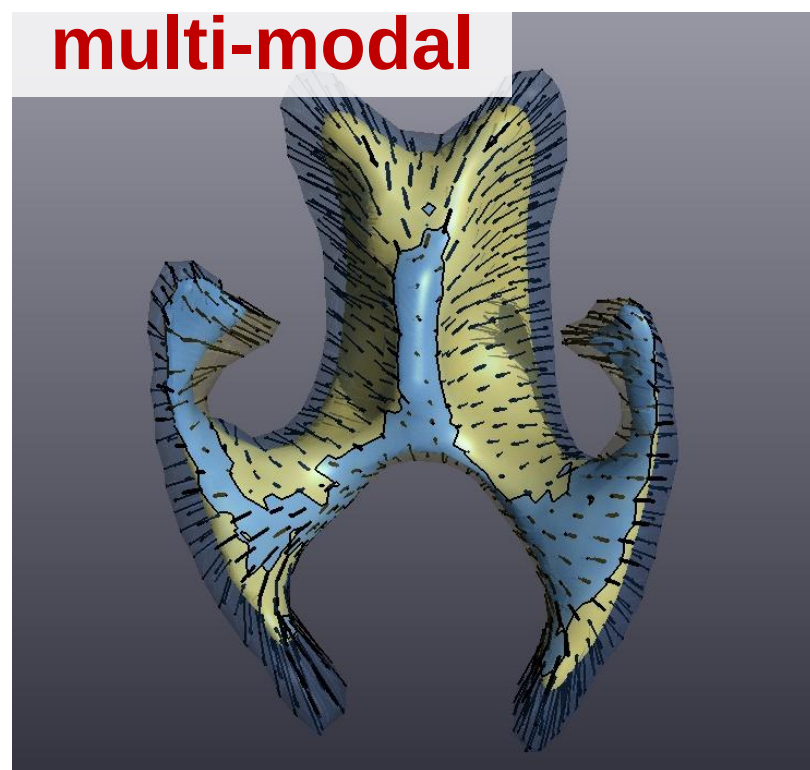
- side-by-side comparison
- overlay in same coordinate system
- explicit encoding of differences / correlations



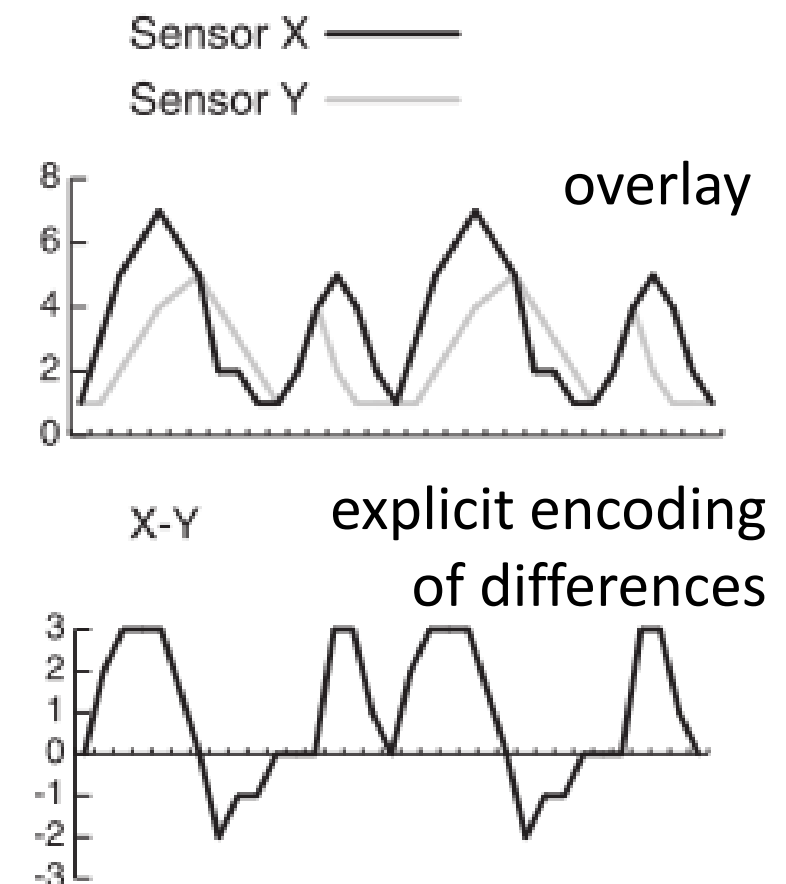
side-by-side comp.



2-tone coloring [Saito et al. 05]



Nested surfaces [Buskin et al. 11]



visual mapping

interactive visual analysis

comput. analysis

**relation &
comparison**

visual data fusion

navigation

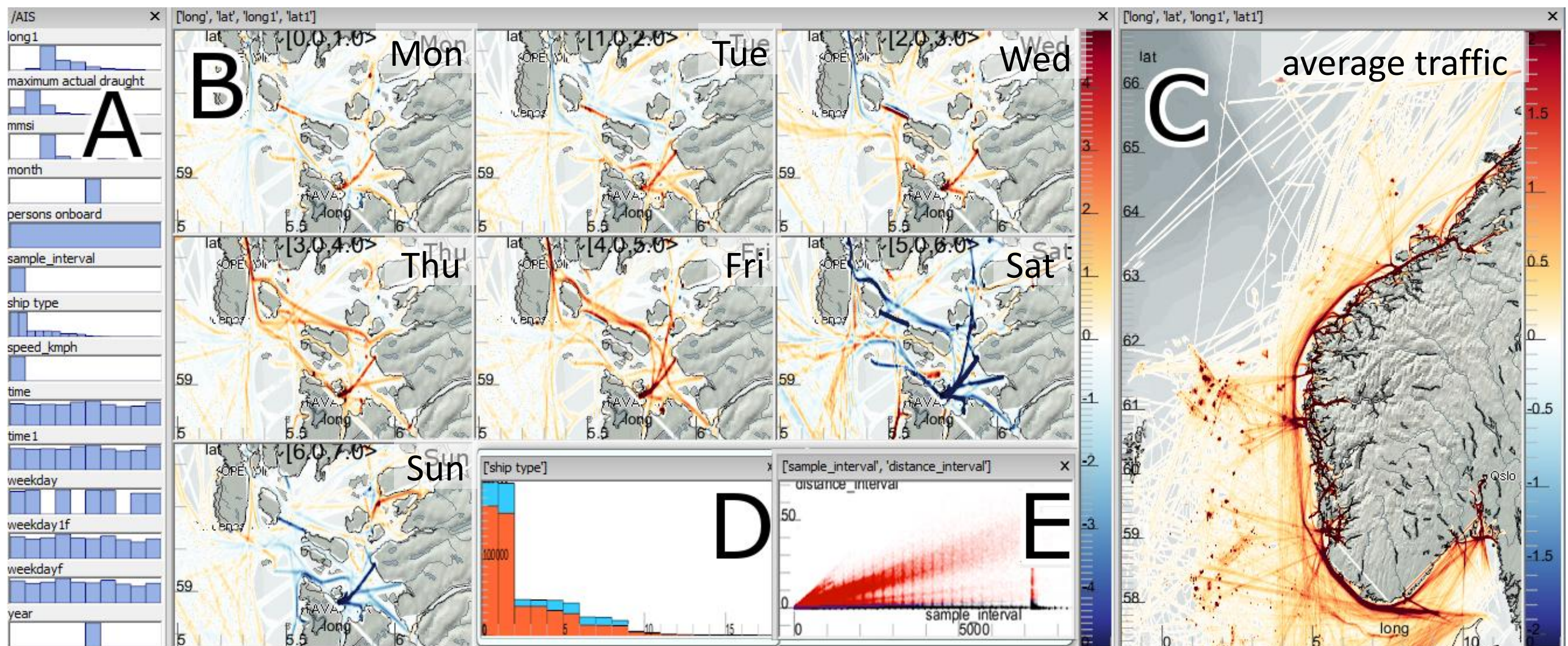
focus+context &
overview+detail

interactive
feature spec.

data abstraction
& aggregation

Difference Views [Daae Lampe et al. 10]

spatiotemporal



visual mapping

interactive visual analysis

comput. analysis

visual data fusion

relation &
comparison

navigation

focus+context &
overview+detail

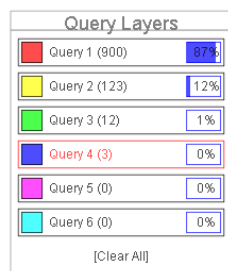
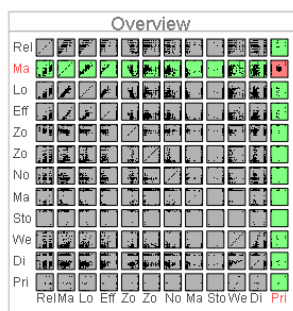
interactive
feature spec.

data abstraction
& aggregation

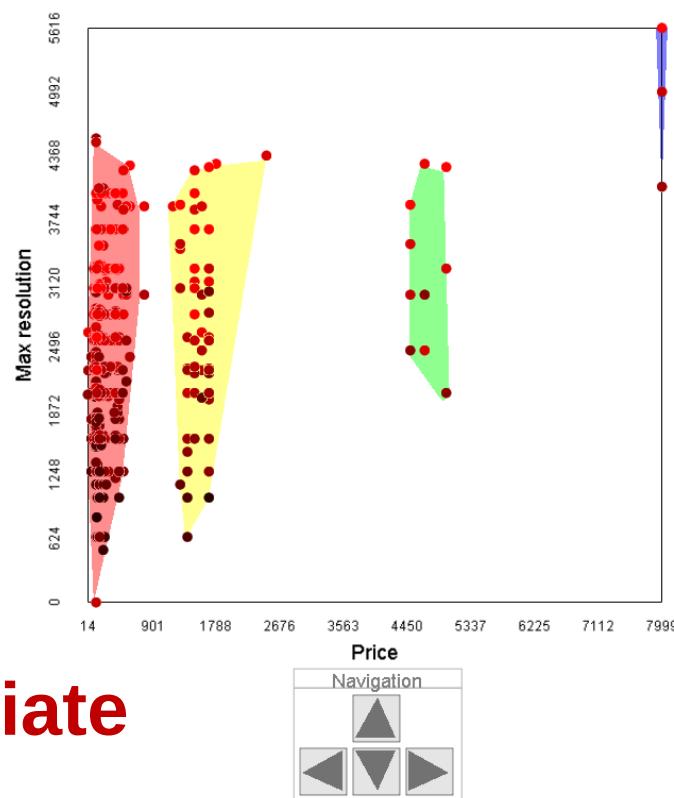
■ Interactive search, zooming, and panning

■ Scatterplot Matrix Navigation [Elmqvist et al. 08]

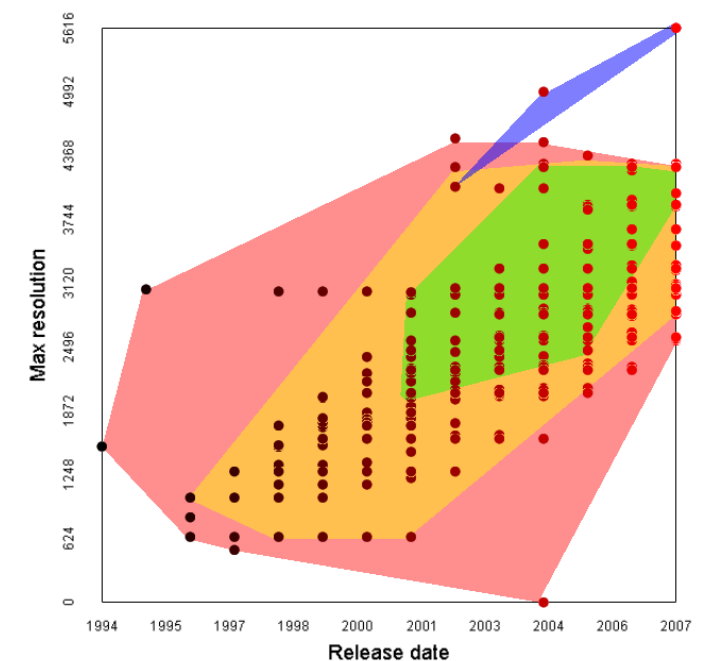
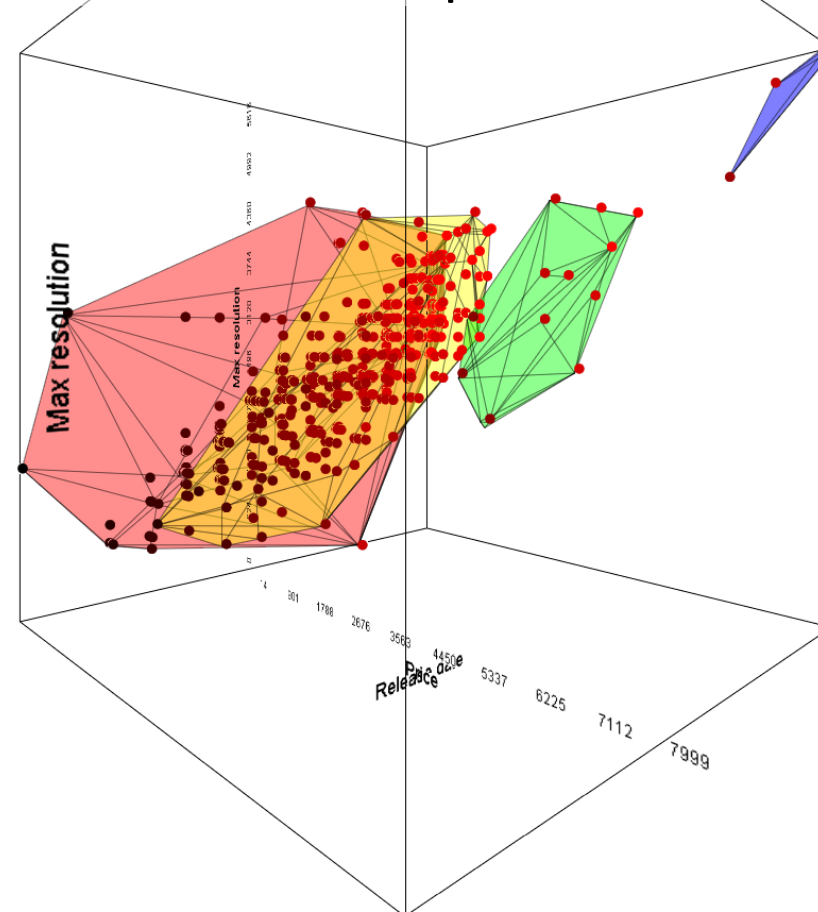
scatterplot
matrix

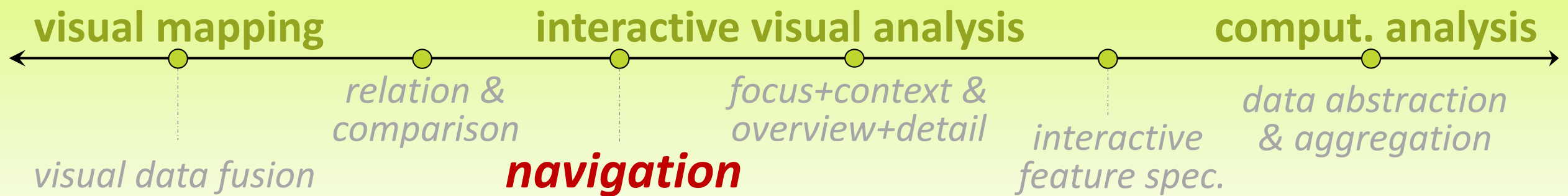


multi-variate

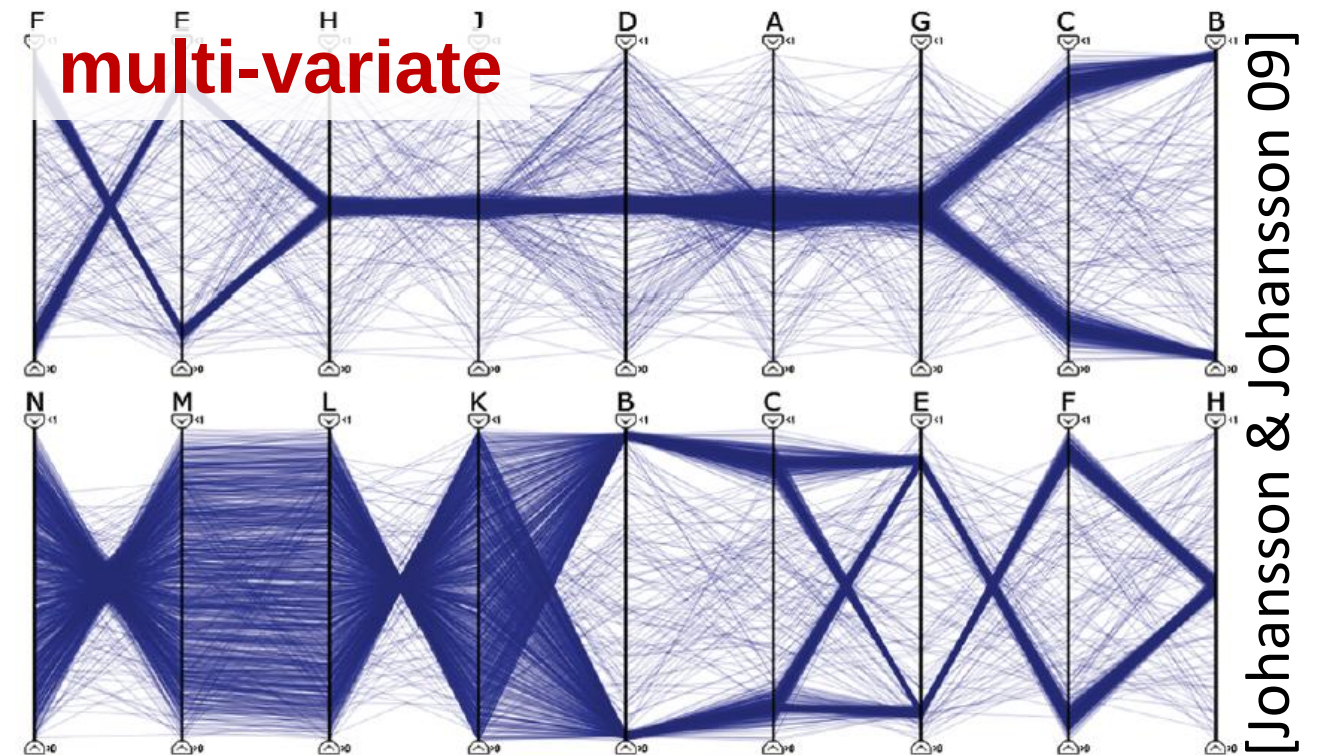


3D transition between
2 scatterplots

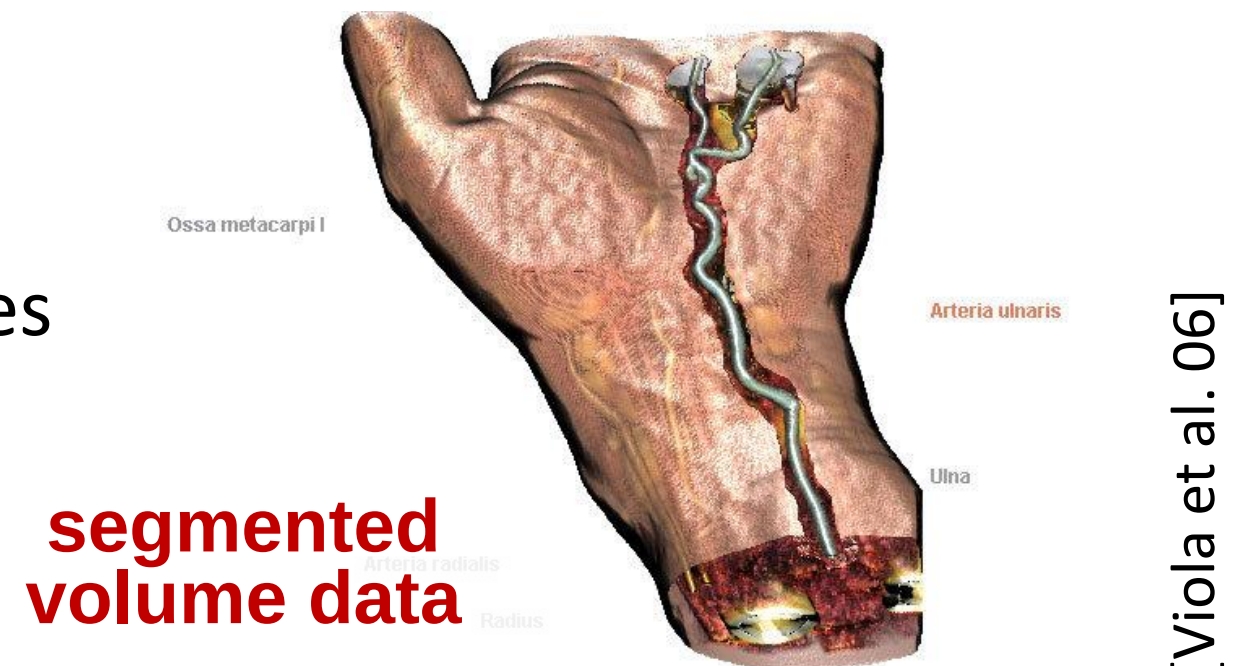




- **Ranking/quality metrics**
[Bertini et al. 2011]
 - clustering, correlations, outliers, image quality, etc.



- **Automated viewpoint selection**
 - information-theoretic measures



visual mapping

interactive visual analysis

comput. analysis

visual data fusion

relation &
comparison

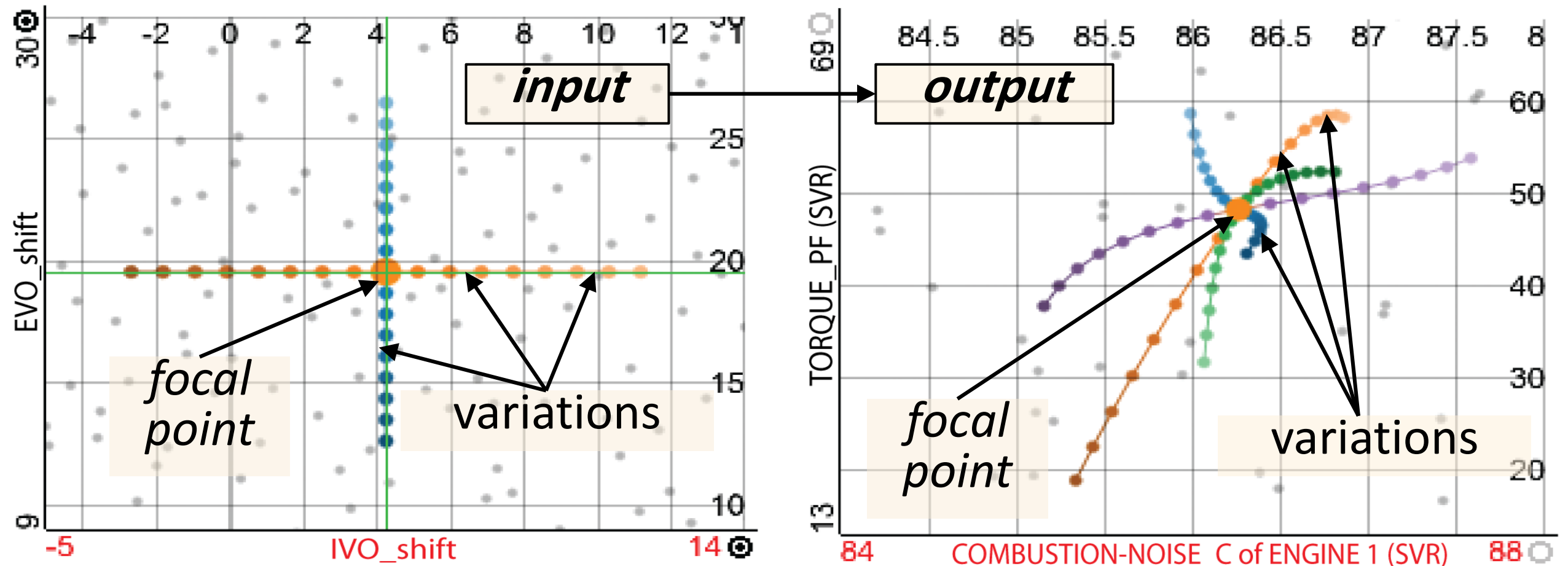
navigation

focus+context &
overview+detail

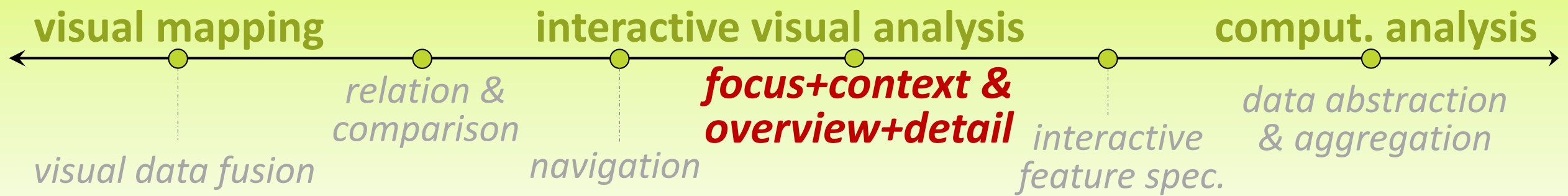
interactive
feature spec.

data abstraction
& aggregation

Parameter space navigation (multi-run data)



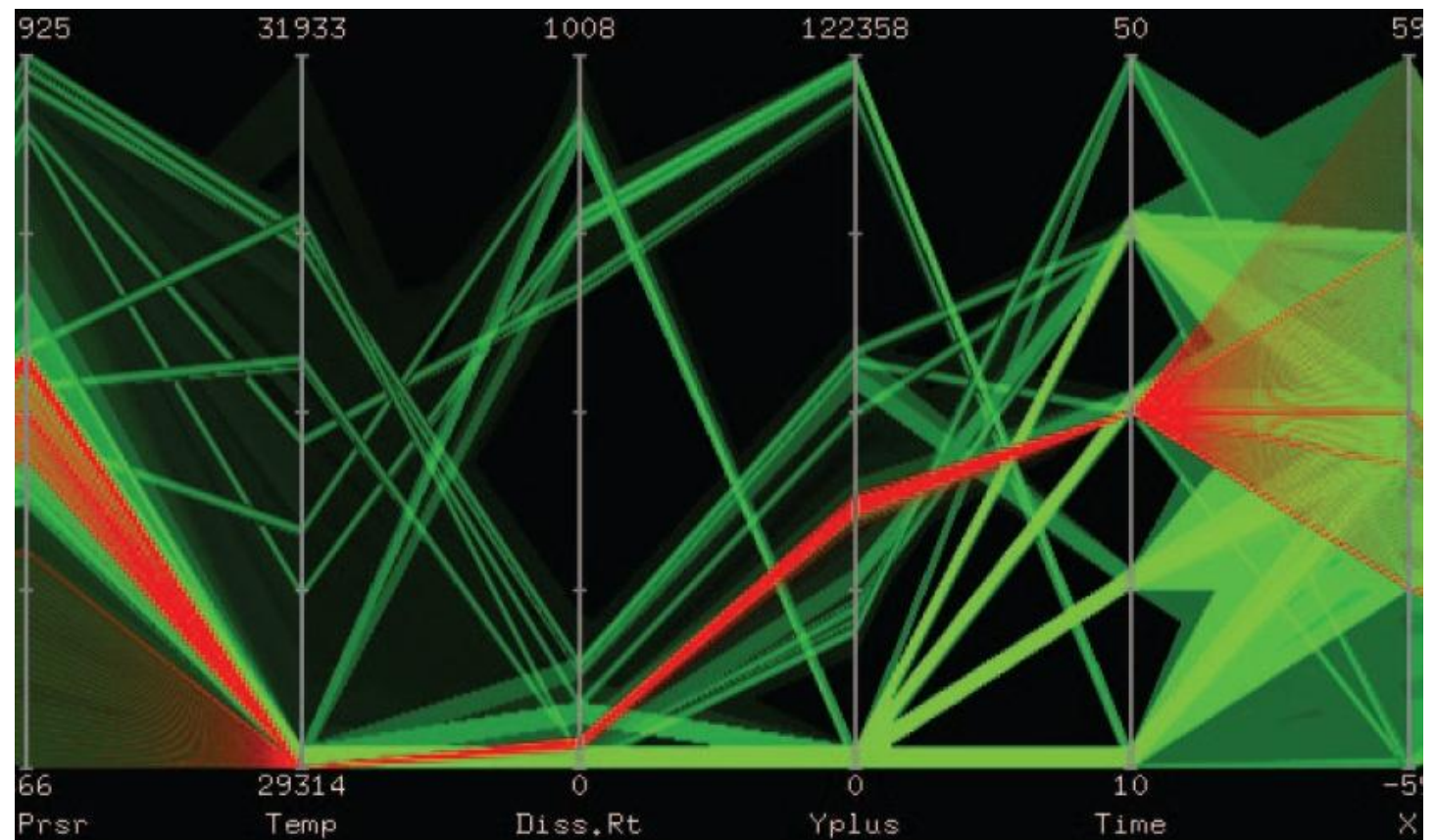
[Berger et al. 11]



■ Focus+context visualization

- different graphical resources (space, opacity, color, etc.)
- focus specification (e.g., by pointing, brushing or querying)

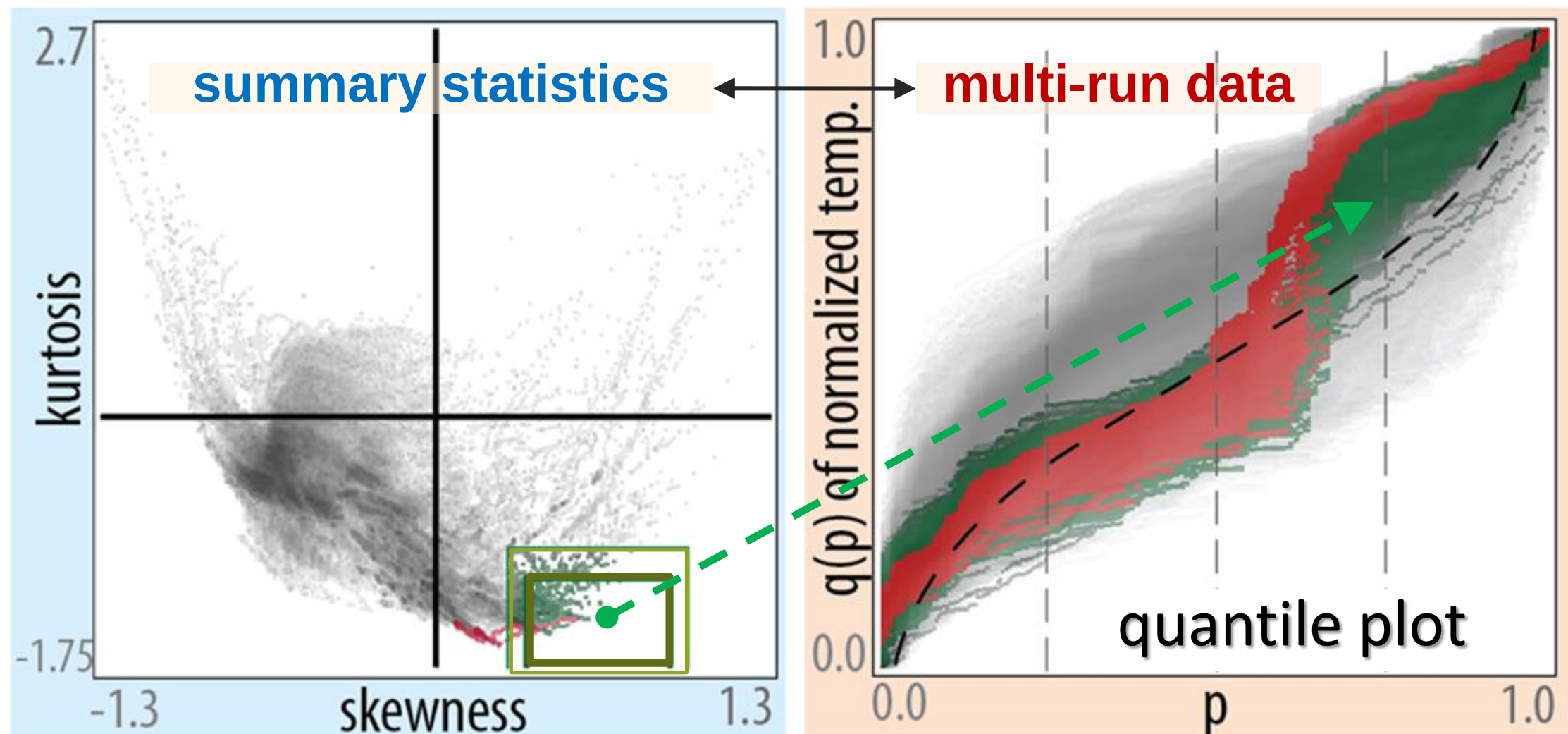
■ Clustering & outlier preservation



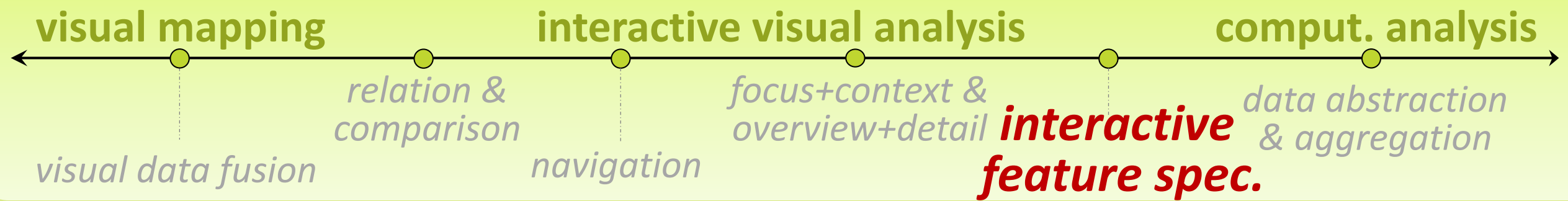
Outlier-preserving focus+context [Novontný & Hauser 06]



■ Overview+detail representation of multi-run data

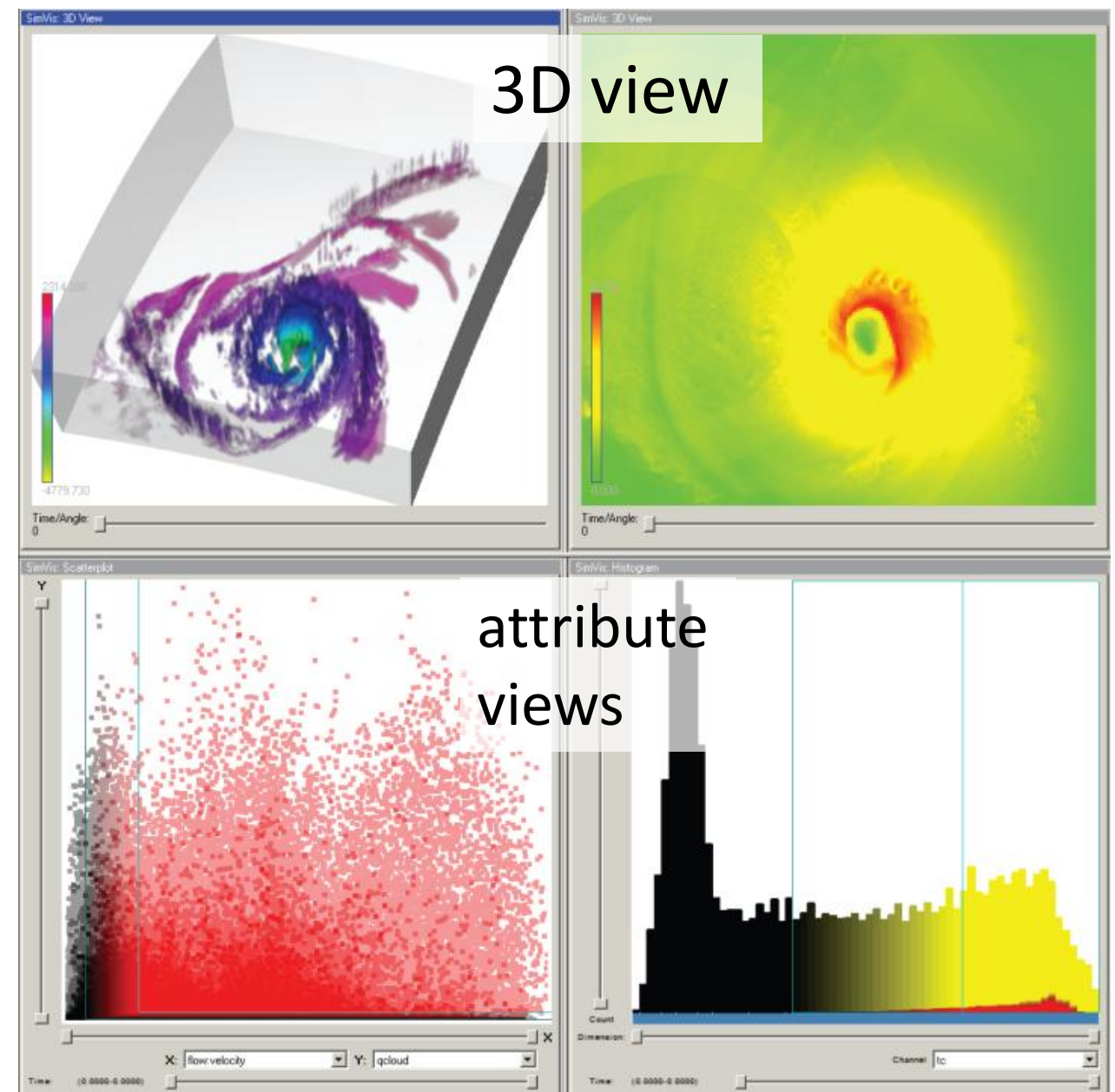


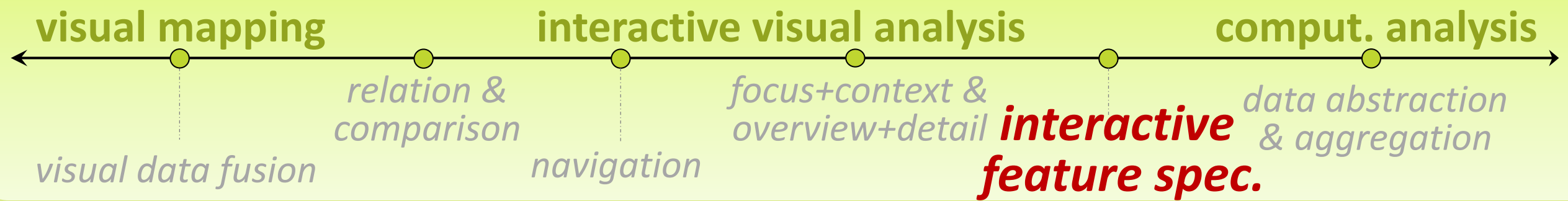
Brushing statistical moments [Kehrer et al. 10]



■ Brushing in multiple linked views

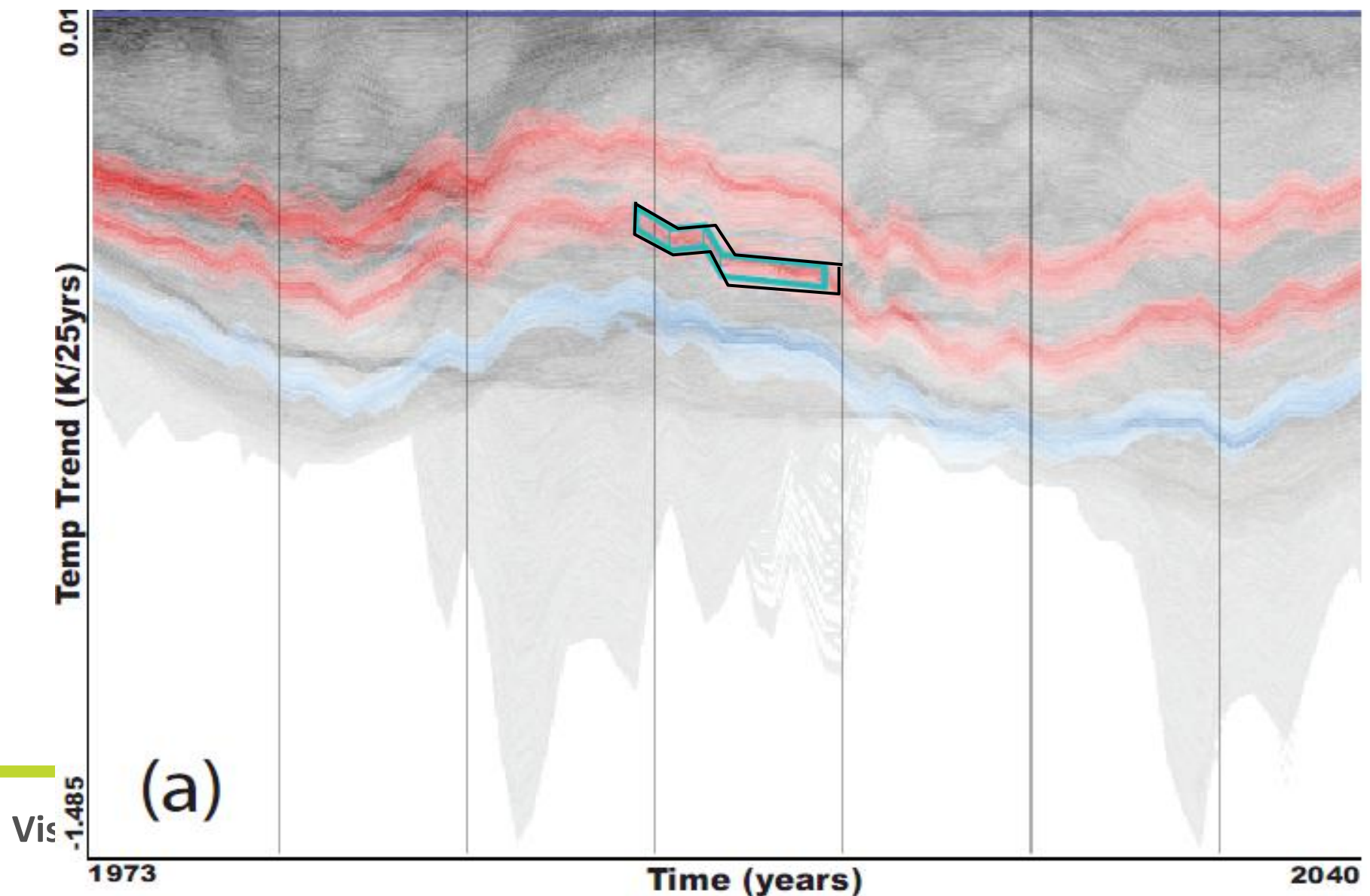
■ SimVis [Doleisch et al. 03, 04]

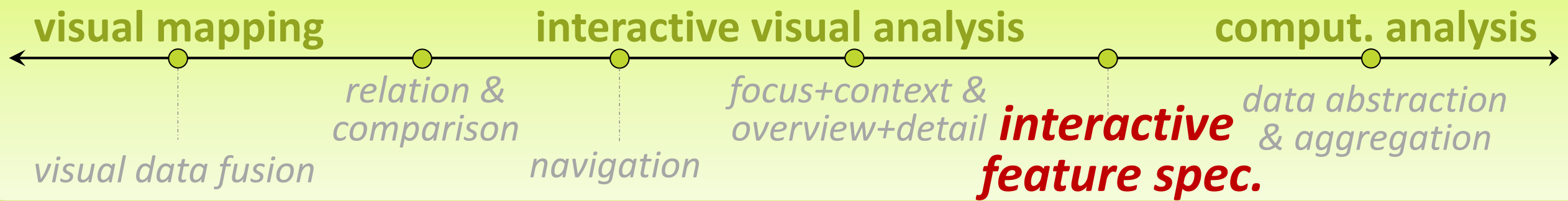




Select function graphs based on similarity [Muigg et al. 08]

- pattern sketched by user
- similarity evaluated on gradients (1st derivative)

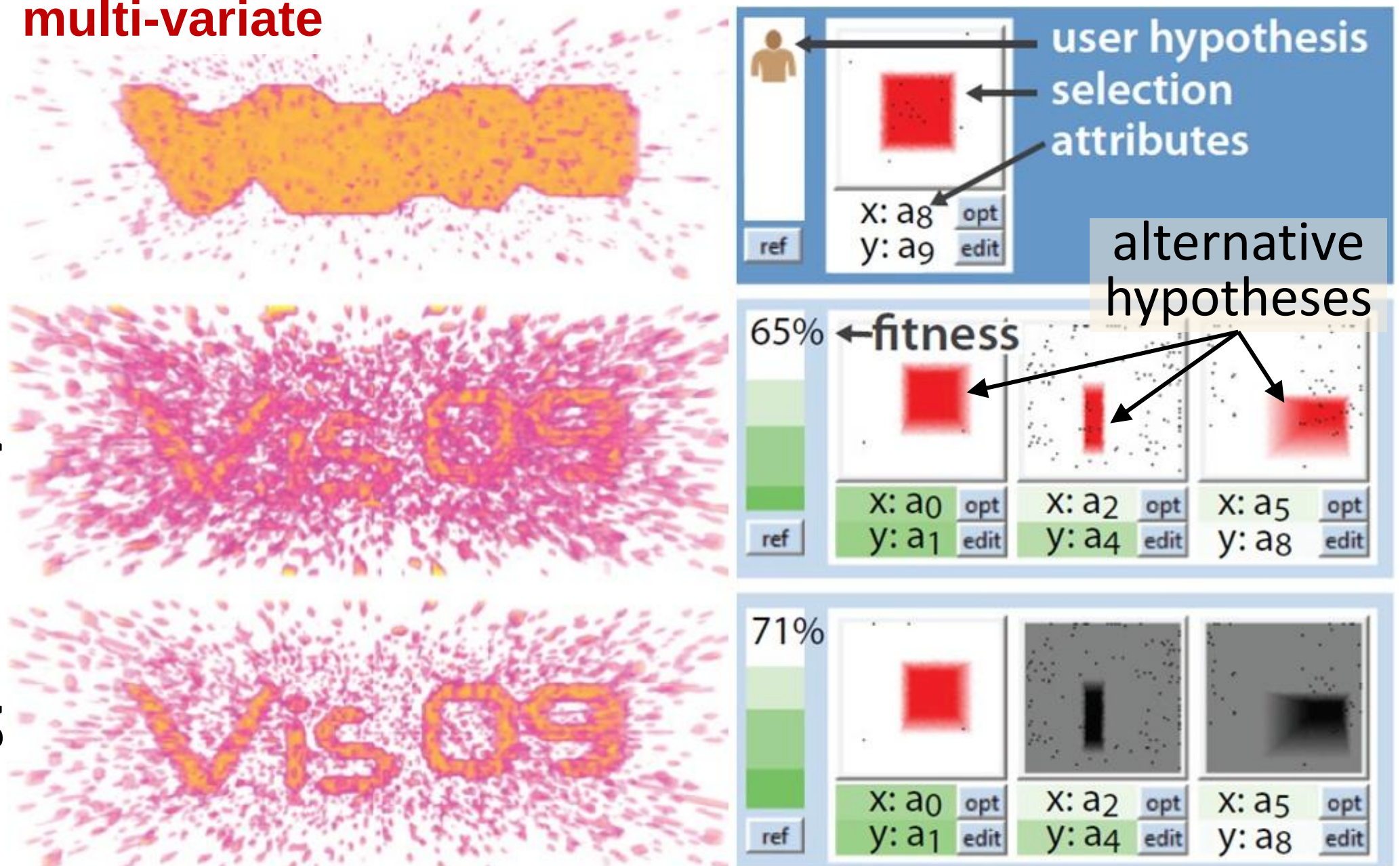


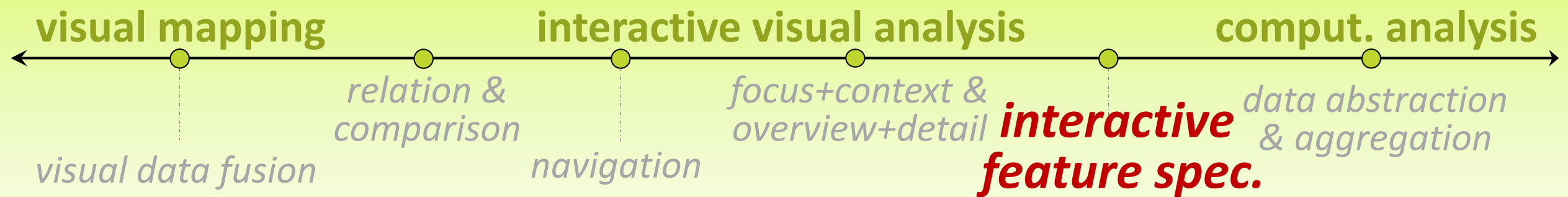


■ Tight integration with supervised machine learning

multi-variate

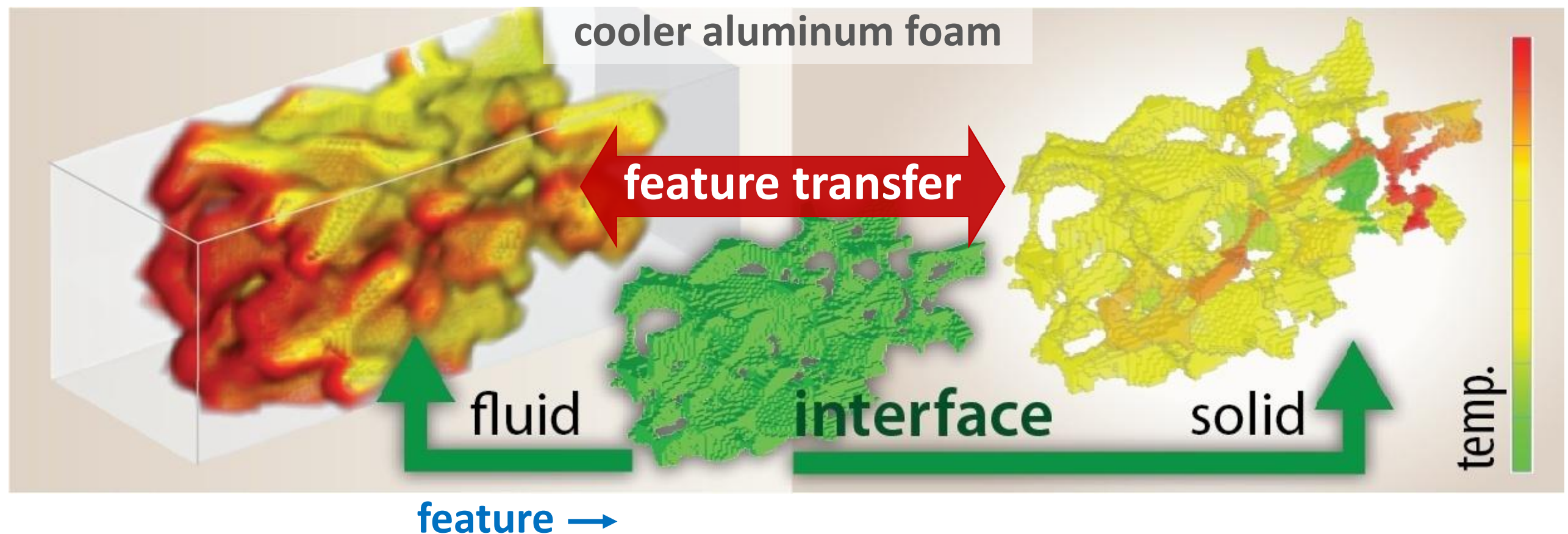
Visual human+machine learning [Fuchs et al. 09]

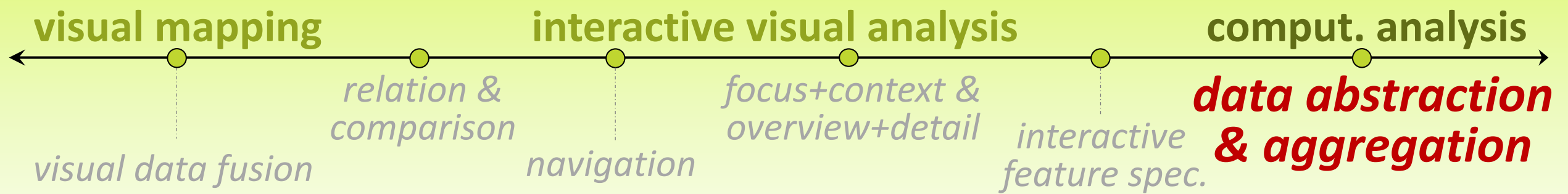




Fluid-structure interactions (multi-model data)

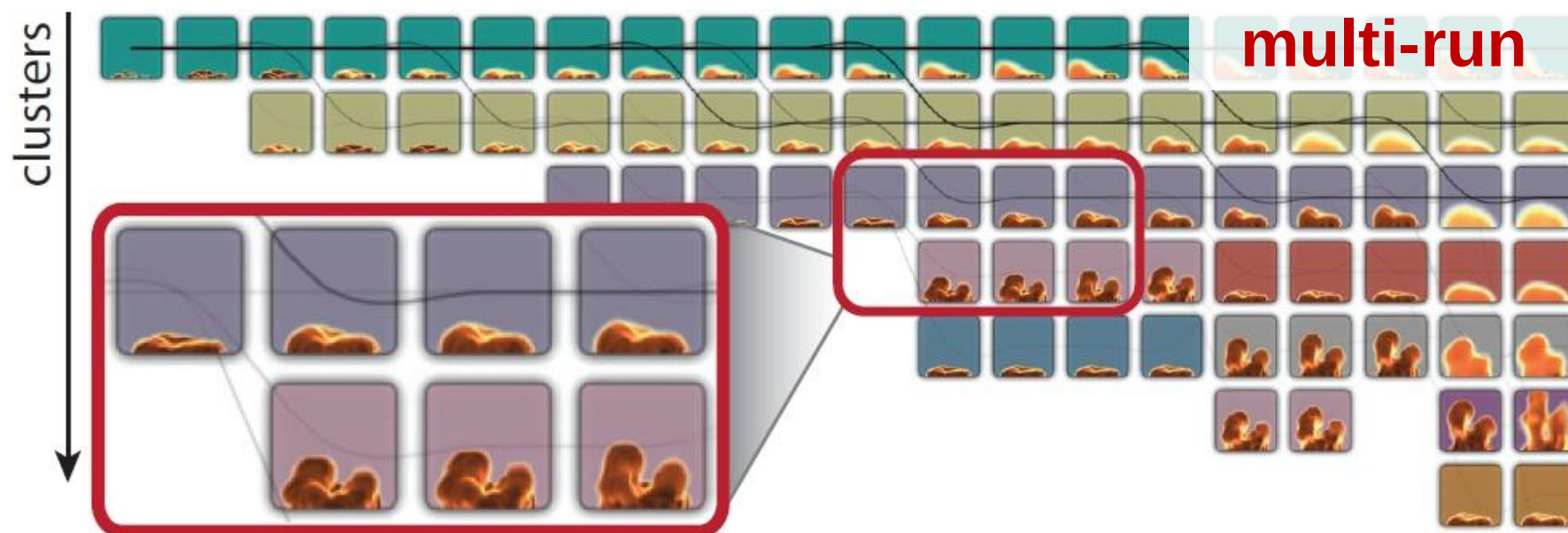
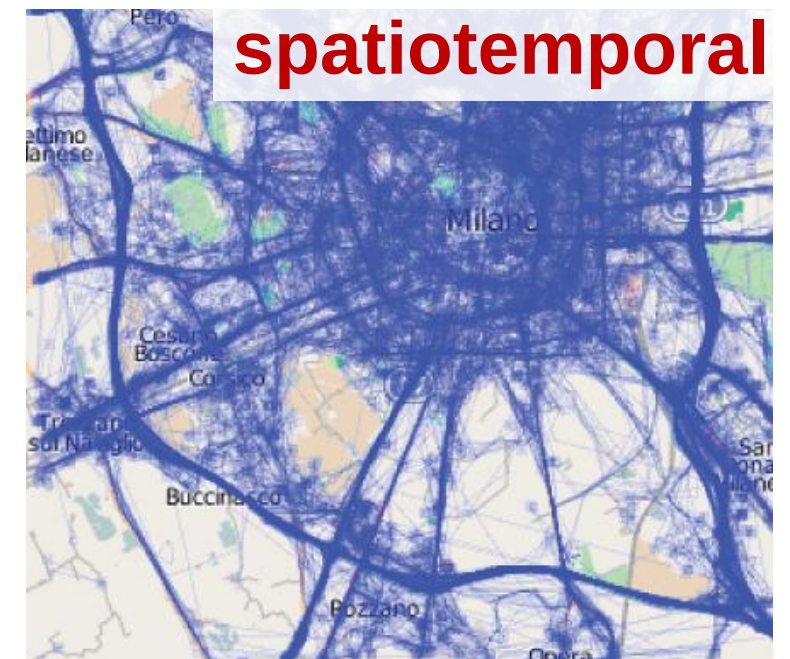
- heat exchange between fluid $\Leftrightarrow$ structure
- feature specification/transfer across data parts [Kehrer et al. 11]



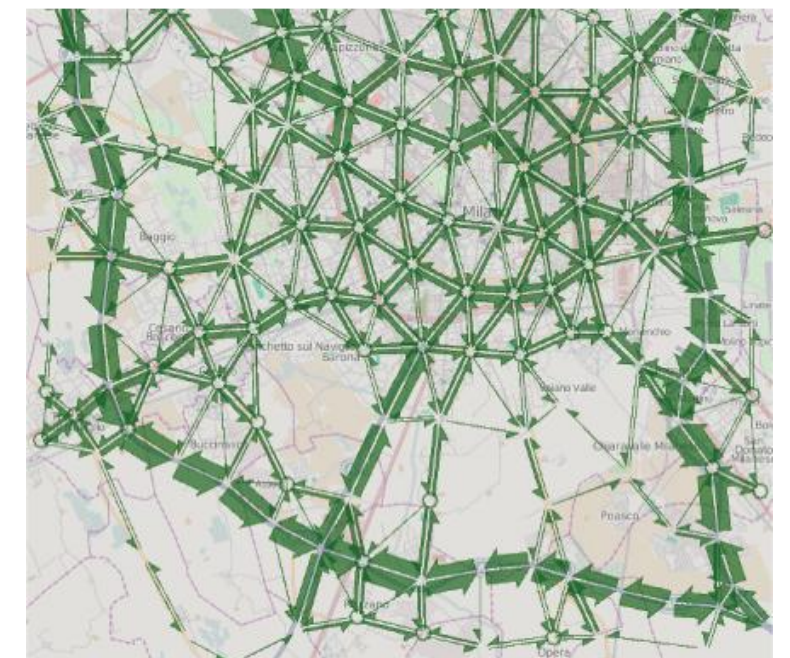


Algorithmically extract values & patterns

- dimensionality reduction (PCA, SOM, MDS)
- aggregation, summary statistics
- clustering, outliers, etc.



clustering of multi-run simulations [Bruckner & Möller 10]



[Andrienko & Andrienko 11]

visual mapping

interactive visual analysis

comput. analysis

visual data fusion

relation & comparison

navigation

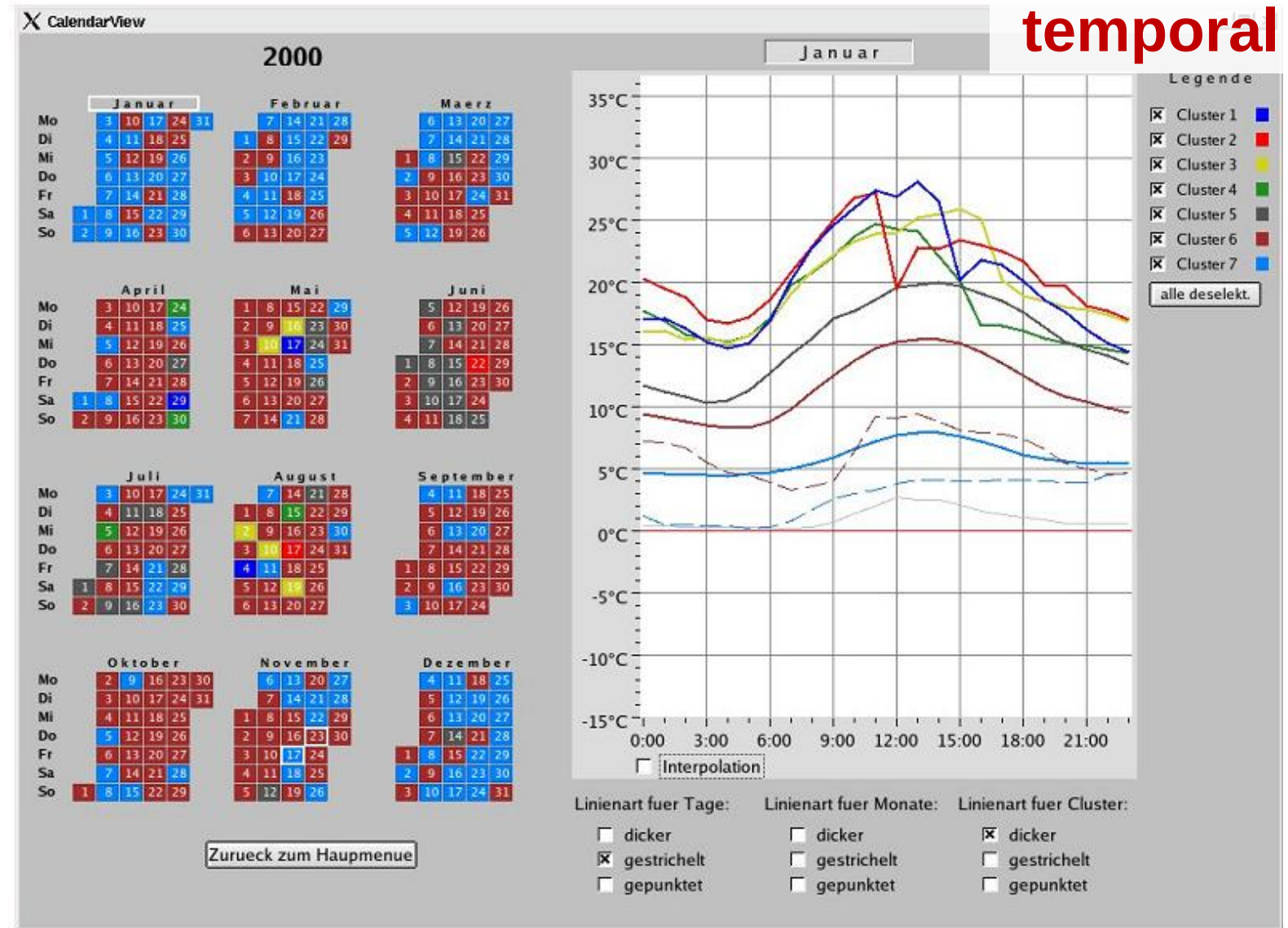
focus+context & overview+detail

interactive feature spec.

data abstraction & aggregation

Cluster Calendar View [vanWijk & van Selow '99]

- Time series clustered by similarity (K-means)

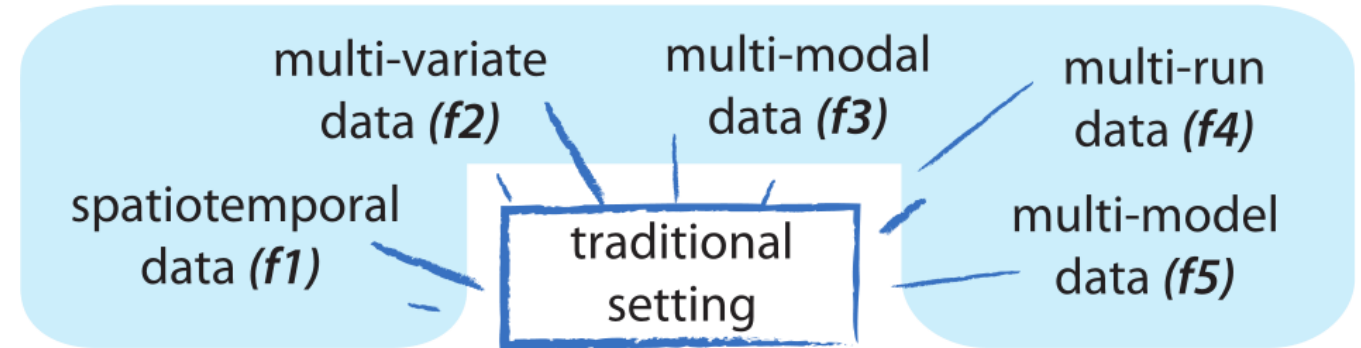


temporal

Categorization of approaches

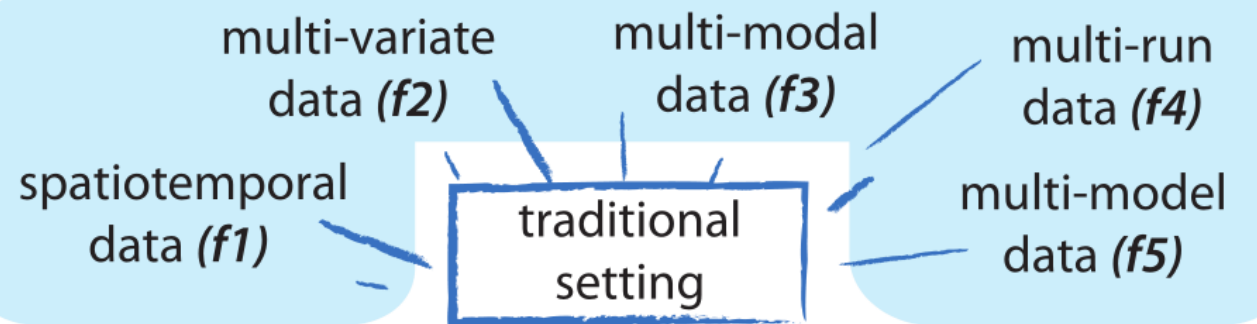
	visual mapping	interactive visual analysis				computational analysis
	visual data fusion	relation & comparison	navigation	focus+context & overview+detail	interactive feature specification	data abstraction & aggregation
multi-dimensional	maps [13], [14], [92]; Helix glyphs [93]; flow maps [105]; function graphs [70], [71], [72]; Time Histograms [94], [110], [111]; chrono volumes [98]; illustrative techniques [99]; texture-based flow vis. [100]	2-tone coloring [20]; Helix glyphs [93]; juxtaposed views [19], [110]; difference views [107]	search, zooming and panning [40], [54]	2-tone coloring [20]; multi-level focus+context [71]; pixel-based multi-resolution techn. [104]	brushing [21], [70], [71], [95], [113]; transfer functions [110], [111]	aggregation [15], [103], [105]; trends [21]; flow features [82]; clustering [83], [84], [110]; PCA [17], [78], [85]; SOM [89], [90]; KDE [106], [107]; information theory [108]; wavelet analysis [109], [110]
multi-variate	attribute views [22], [50], [67]; color & texture [119]; layering [115], [124], [126]; 2-level volume rendering [127], [128]; glyphs [120], [121], [122], [123], [124], [125]	correlation fields [133]; operators [134]; multiple linked views [9], [26], [29], [73], [74], [76]	grand tour [47]; ScatterDice [46]; ranking & quality metrics [48], [130], [131], [132]	illustrative vis. [115], [116]; outlier-preserving methods [69]; smooth brushing [80]	brushing [9], [50], [74], [75], [112]; multi-dim. transfer func. [114], [115]; machine learning [91], [135], [136]	clustering [68], [130]; data binning [69]; PCA [78]; MDS [86], [87]; SOM [88], [89]; projections [47], [48], [130], [132]; point clouds [129]
multi-modal	resampling [138]; data model [142]; illumination model [143]; multi-volume rendering [128], [139], [143], [144], [145], [146]	difference views [107]; multi-image view [153]; nested surfaces [31], [154], [156]; features [44], [155]	viewpoint selection [49]	cutaway views [147], [139], [49]	transfer functions [143], [144]	registration [27]; mutual information [28]; comparison metrics [148], [151], [152], [133]
multi-run	glyphs & box plots [37], [43], [162], [163], [164]; shape descriptors [164]; families of surfaces [41]; spaghetti plots [35], [42], [165]	aggregated & multi-run data [36], [37], [41], [174]; HyperMoVal [51], [52]	aggregated & multi-run data [36], [37], [41]; parameter space nav. [51], [52]	aggregated & multi-run data [36], [37], [41]; simulation process vis. [173], [174]	trends & outliers [36], [37], [41]; visual steering [172]	overview statistics [31], [35], [36]; projections [41], [51], [170], [171]; operators [33]; PCA [169]; clustering [167], [168], [169]
multi-model	feature fusion across multiple data parts [37]	feature relation across data parts [37]	x	x	feature spec. across data parts [37]	x

Open Issues

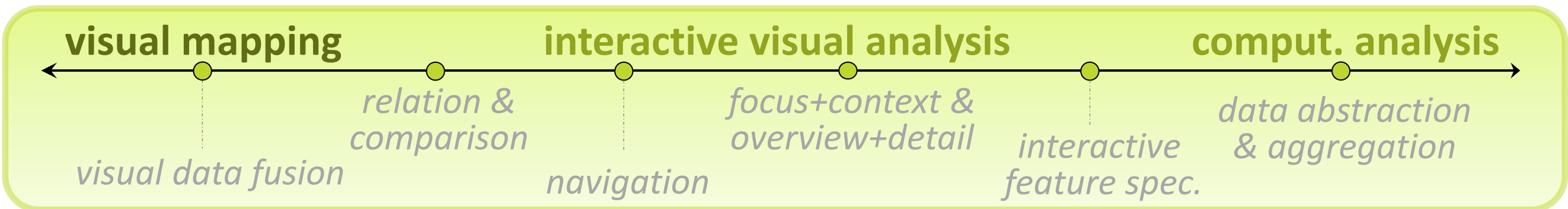


- **How to deal with data heterogeneity?**
 - most approaches only address one or two data facet
 - coordinated multiple views with linking & brushing
 - investigation of features across views, data facets, levels of abstraction, and data sets
 - fusion of heterogeneous data at feature/semantic level
- **Combination of vis., interaction, and comput. analysis**
 - analytical methods can controll steps in visualization pipeline (e.g., visualization mapping or quality metrics)
 - interactive feature specification + machine learning

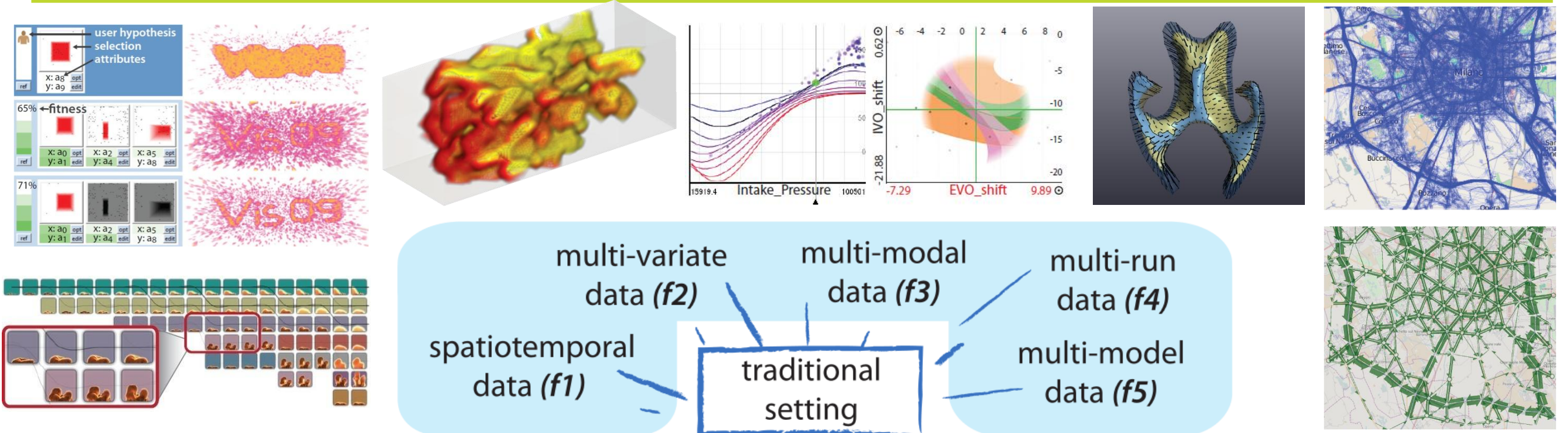
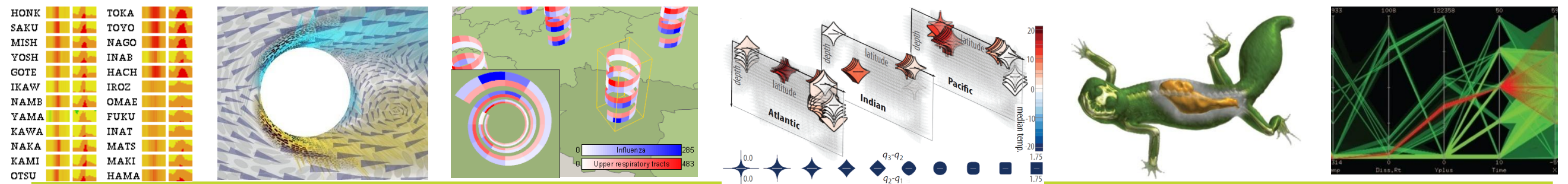
Conclusions



- Scientific data are becoming multi-faceted
- Categorization based on common visualization, interaction, and comput. analysis methods



- Promising data facets, e.g., multi-run & multi-model data



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