

High Performance Visual Analysis

- the Challenges and Opportunities

面向科学数据的高性能可视分析

袁晓如 北京大学

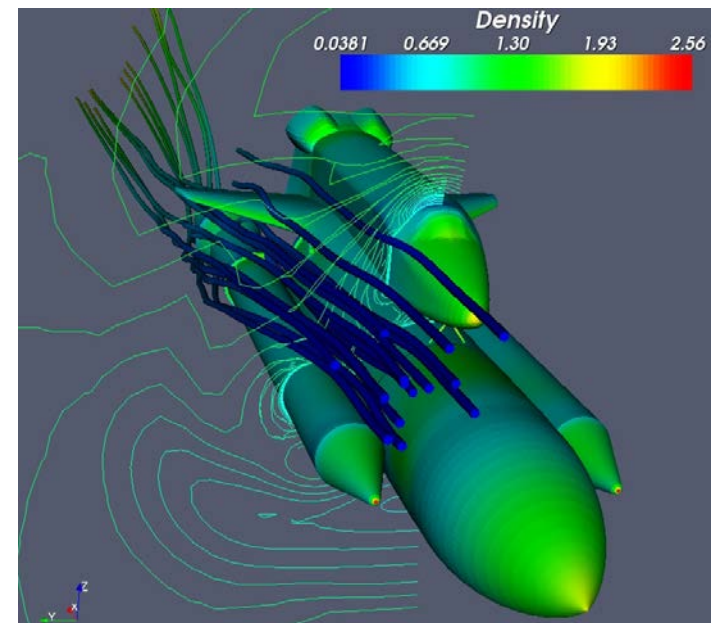
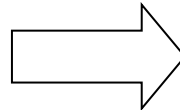
2013.11.14

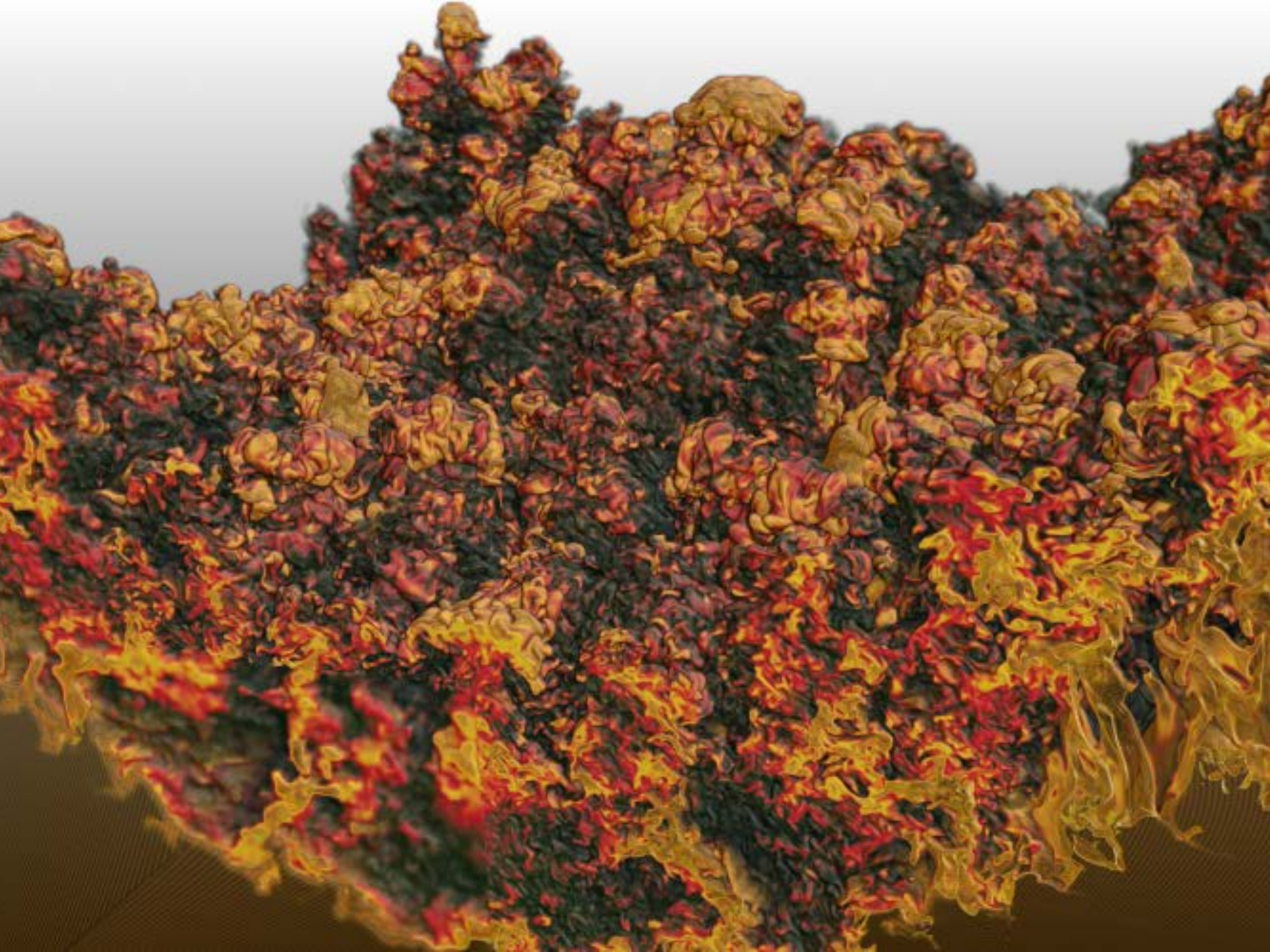
From Data to Visualization

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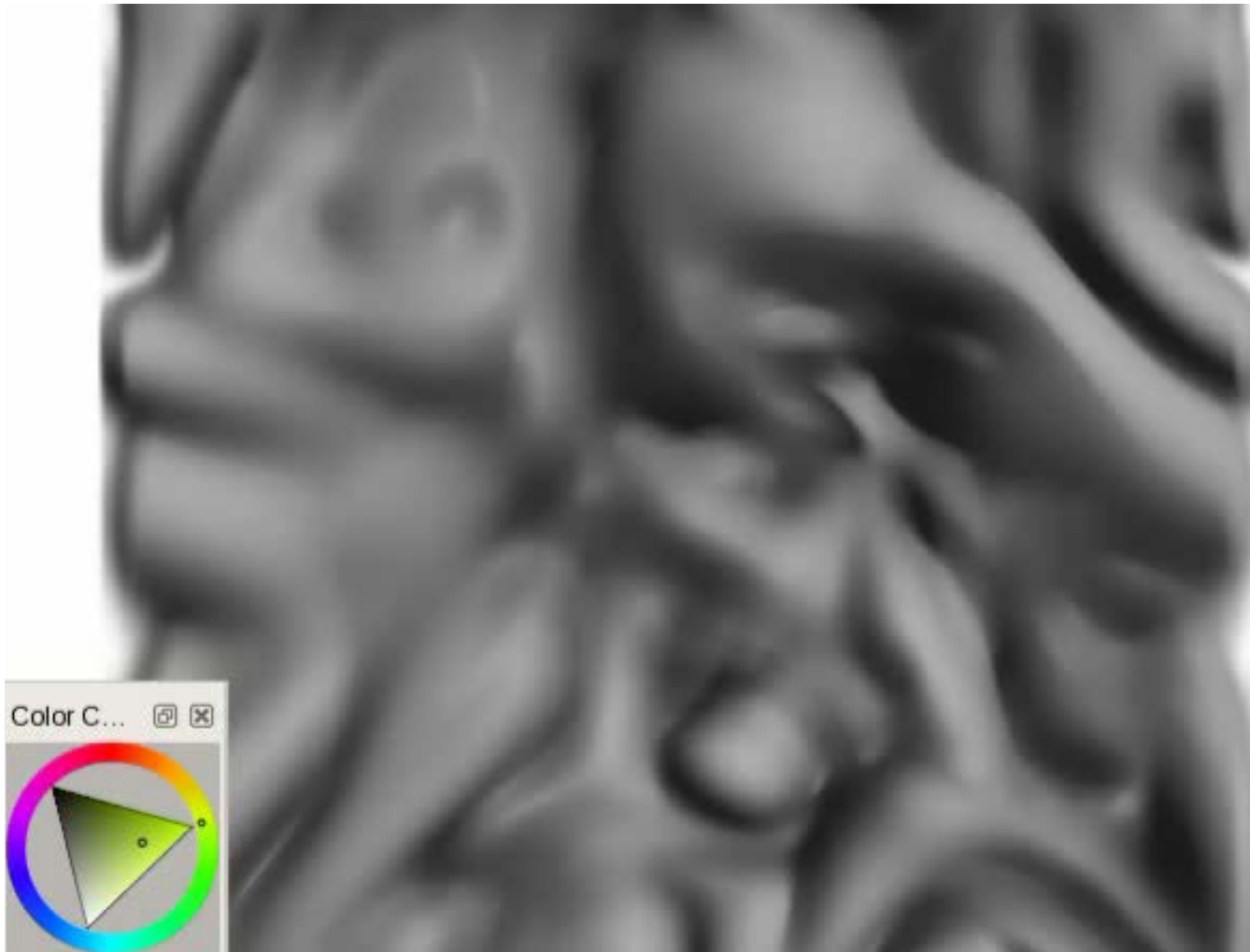




Large Volume Data Visualization



Interaction in Vis





Collaborative Vis with Advanced UI



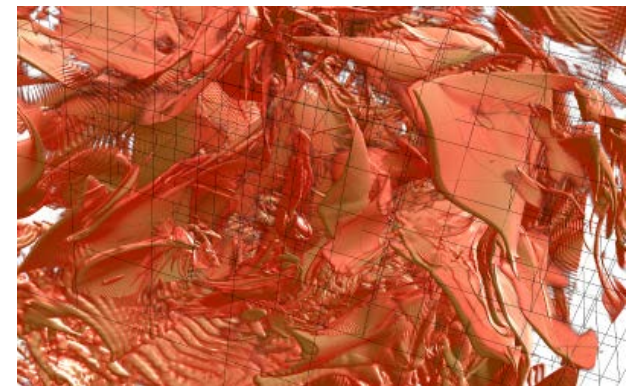
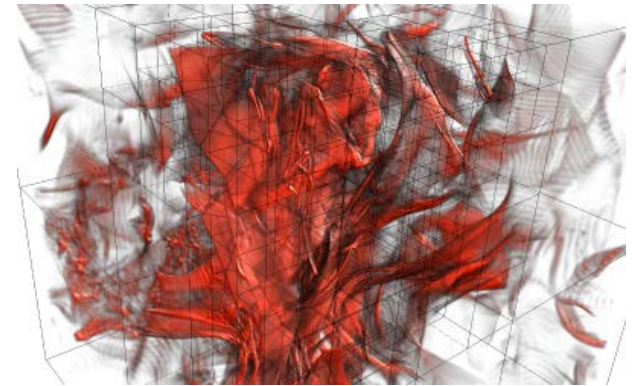
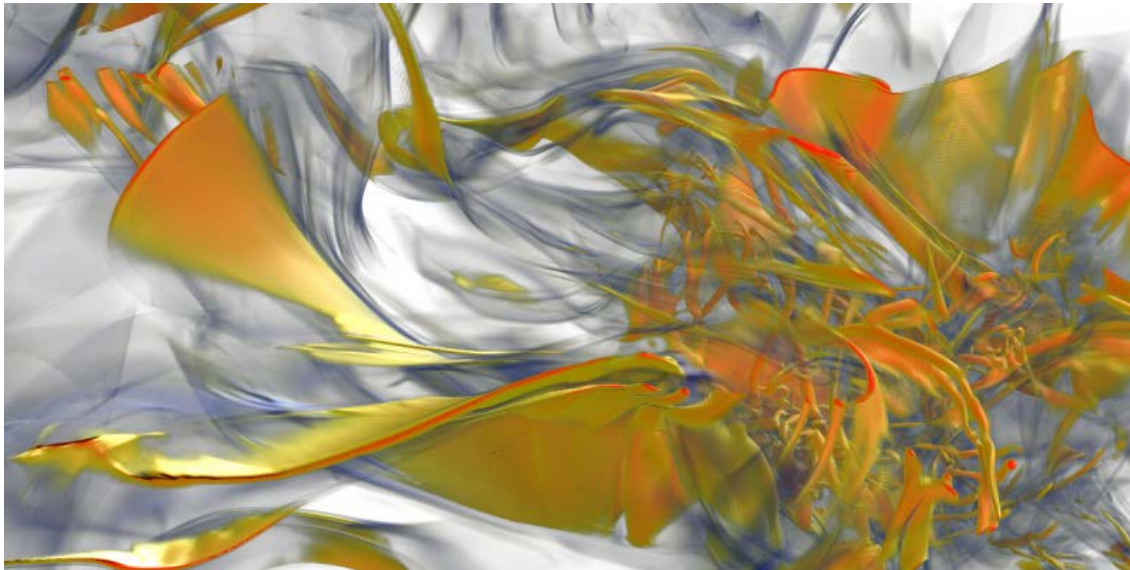
Work in Progress



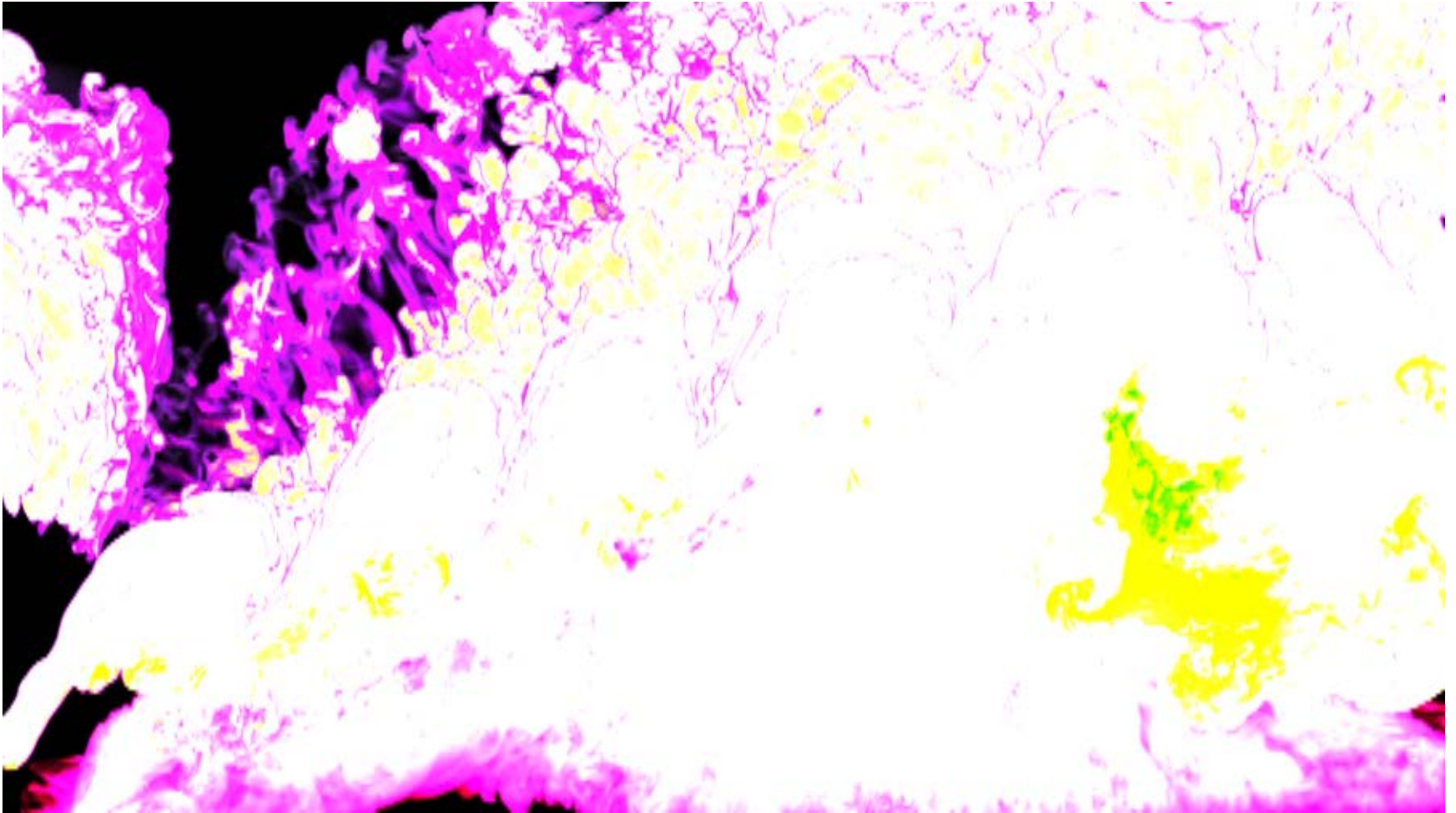
Work in Progress

Large Volume Visualization

- Level of Details
- Out of Core
- Parallel Visualization

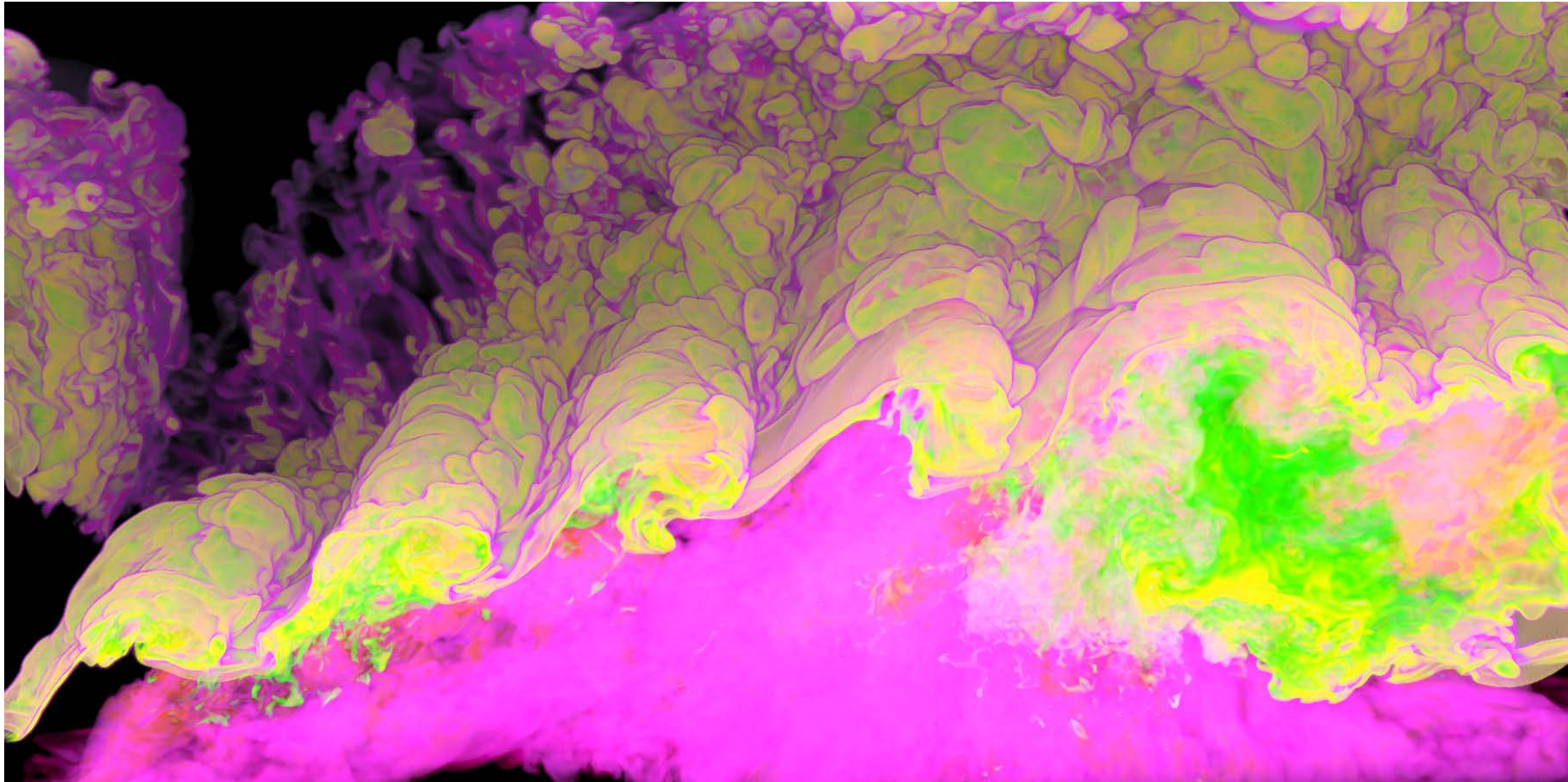


HDR VolVis – Turbulent Mixing of Air/SF₆



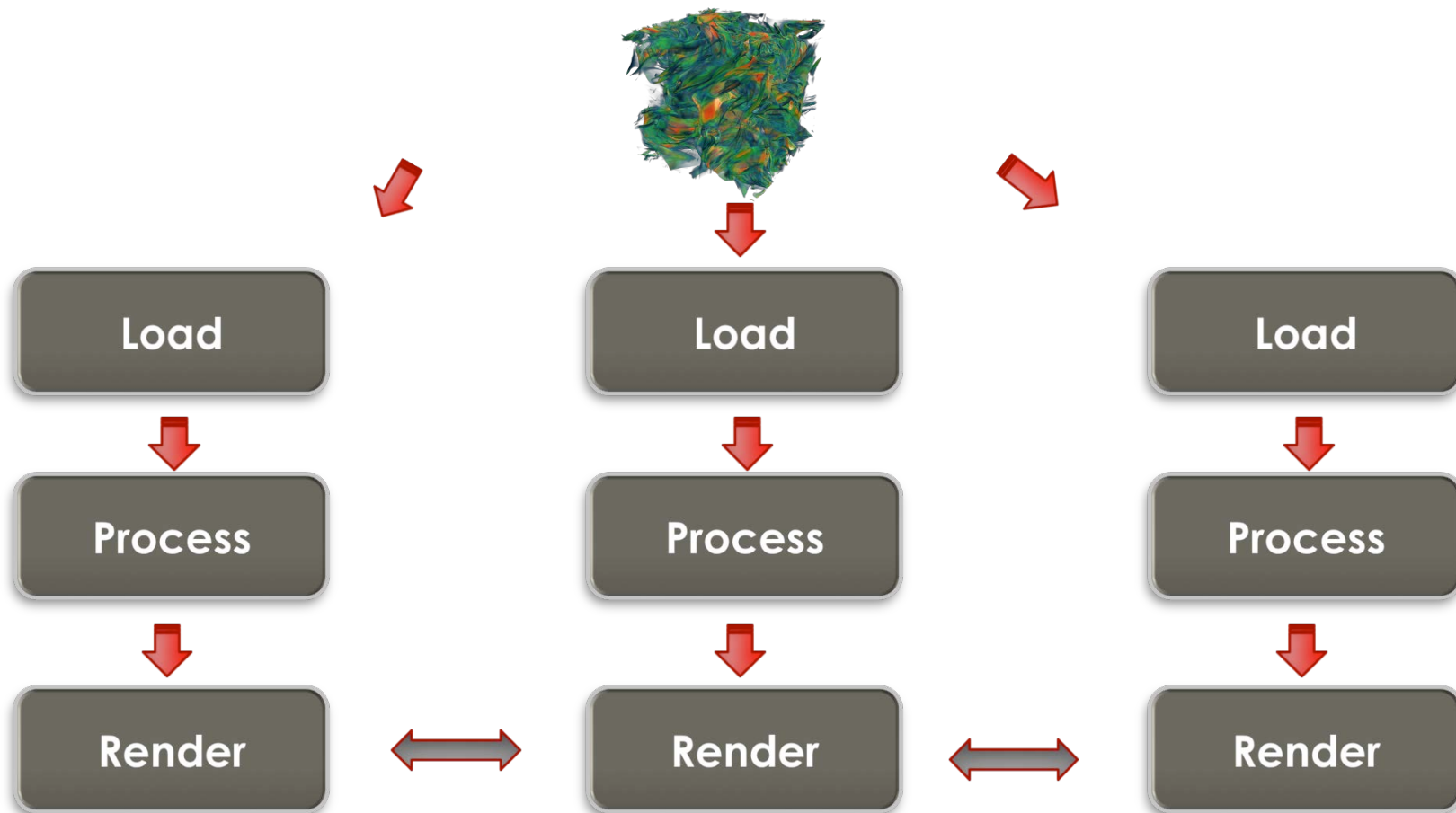
Without High dynamic range rendering results

HDR VolVis – Turbulent Mixing of Air/SF₆

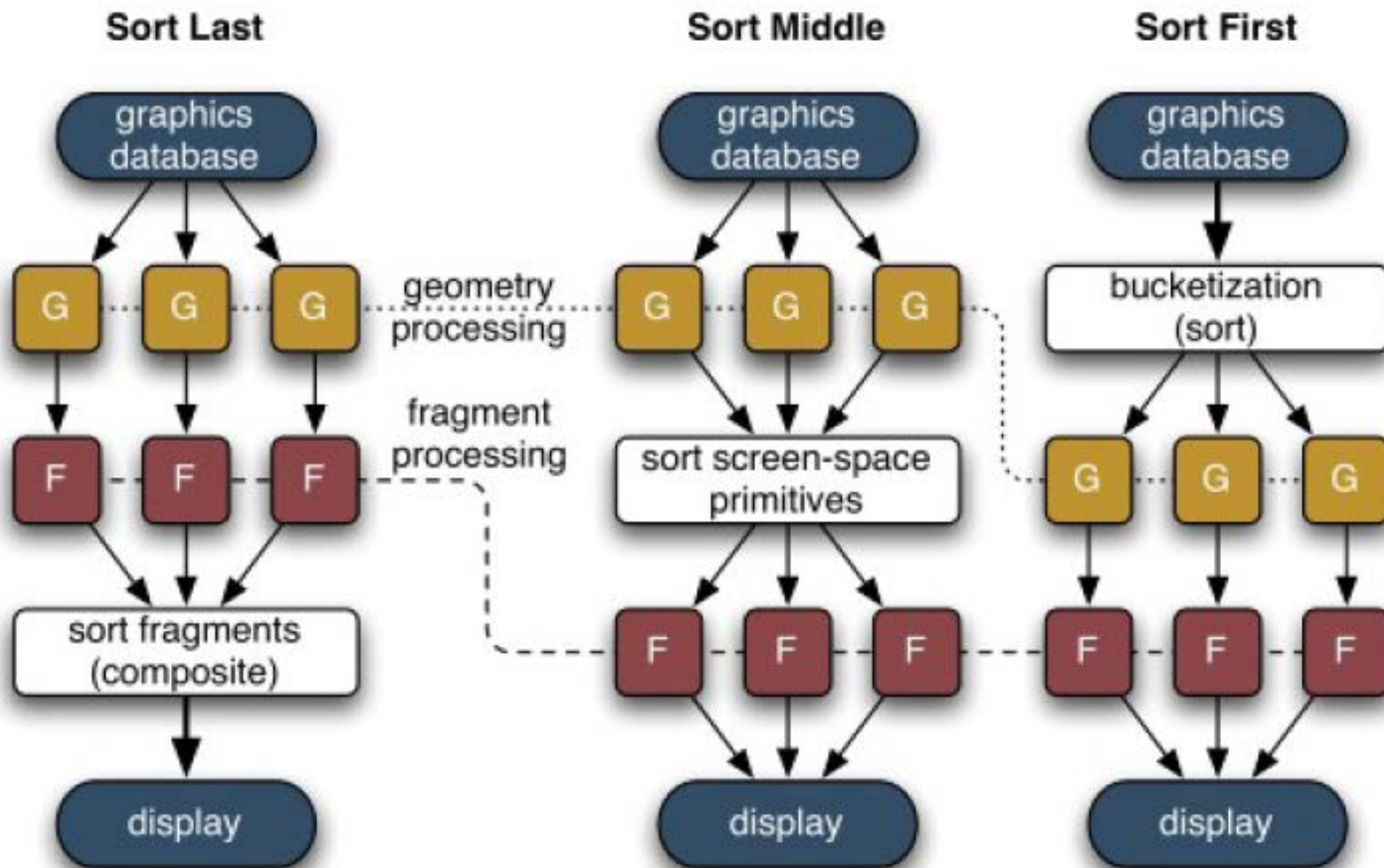


High dynamic range rendering results

Parallel Data (Volume) Vis.



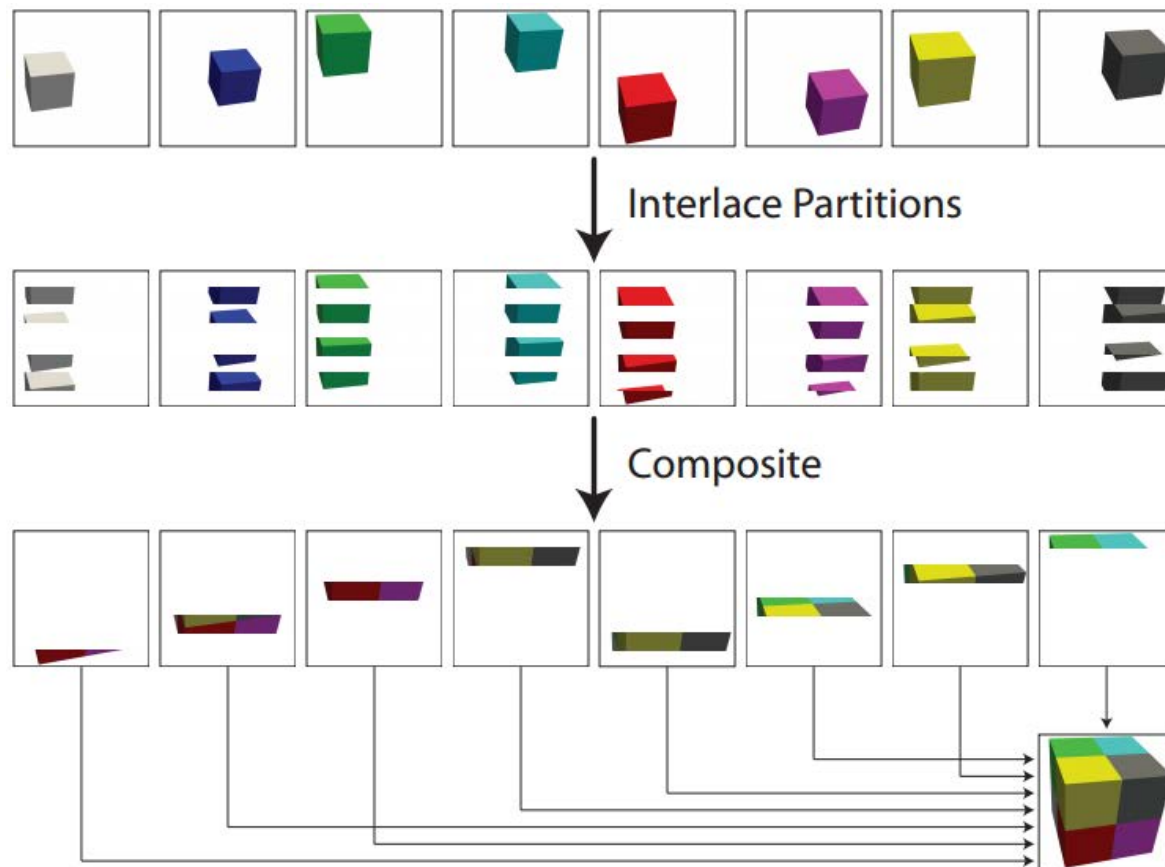
Parallel Image Compositing



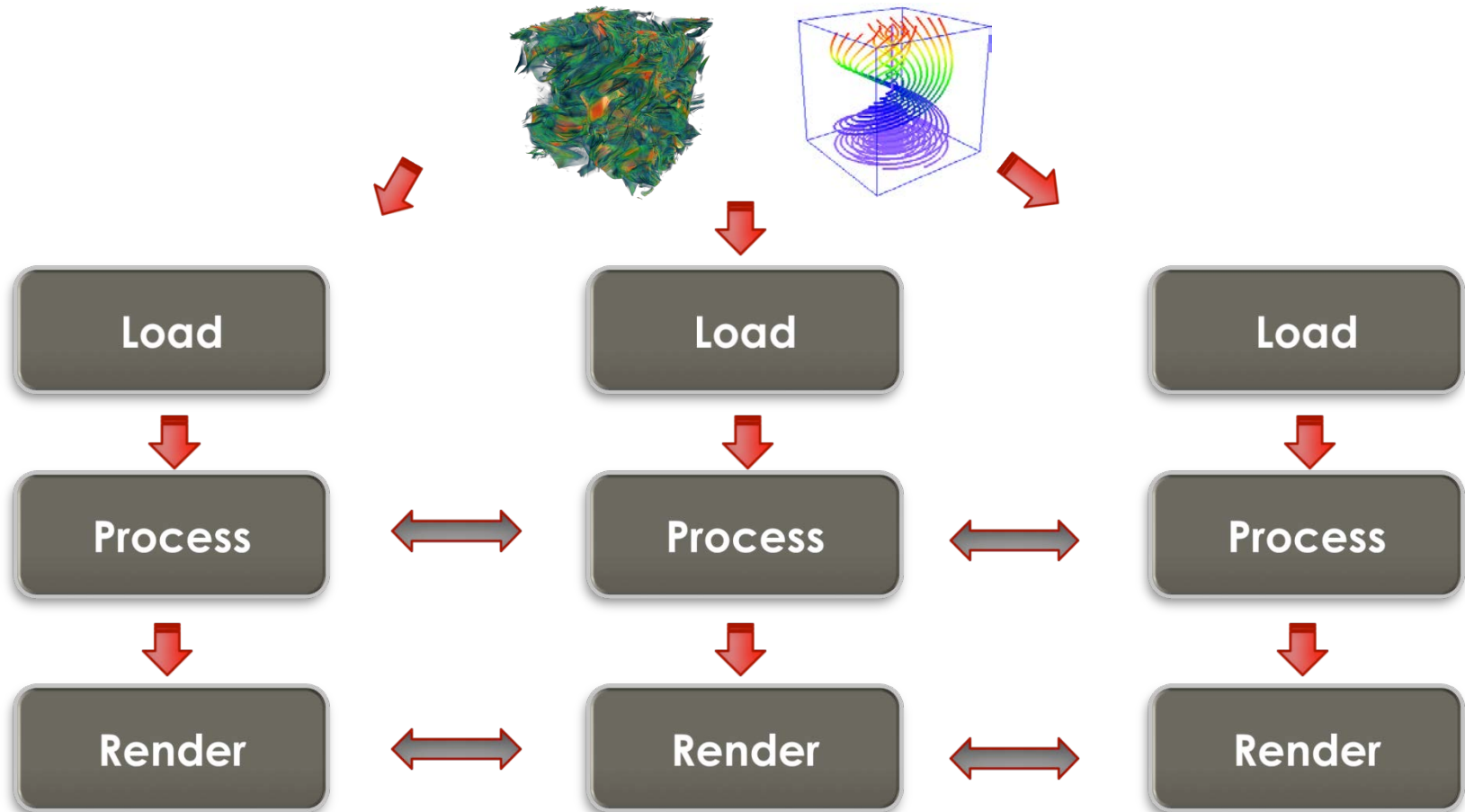
Sort-Last Image Compositing

■ Stages in final parallel visualization algorithms

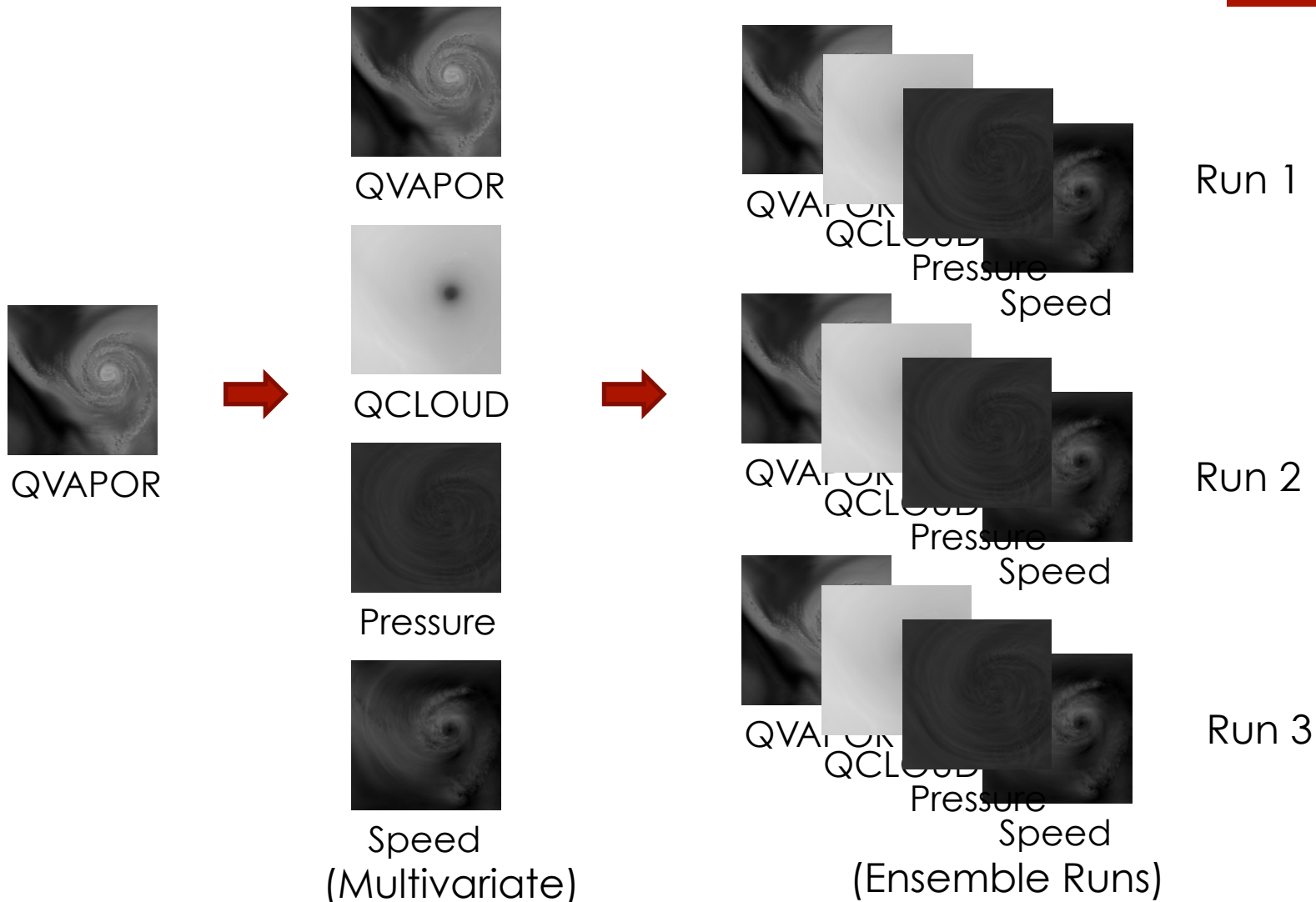
- Data partitioning
- Local data rendering
- Result image compositing



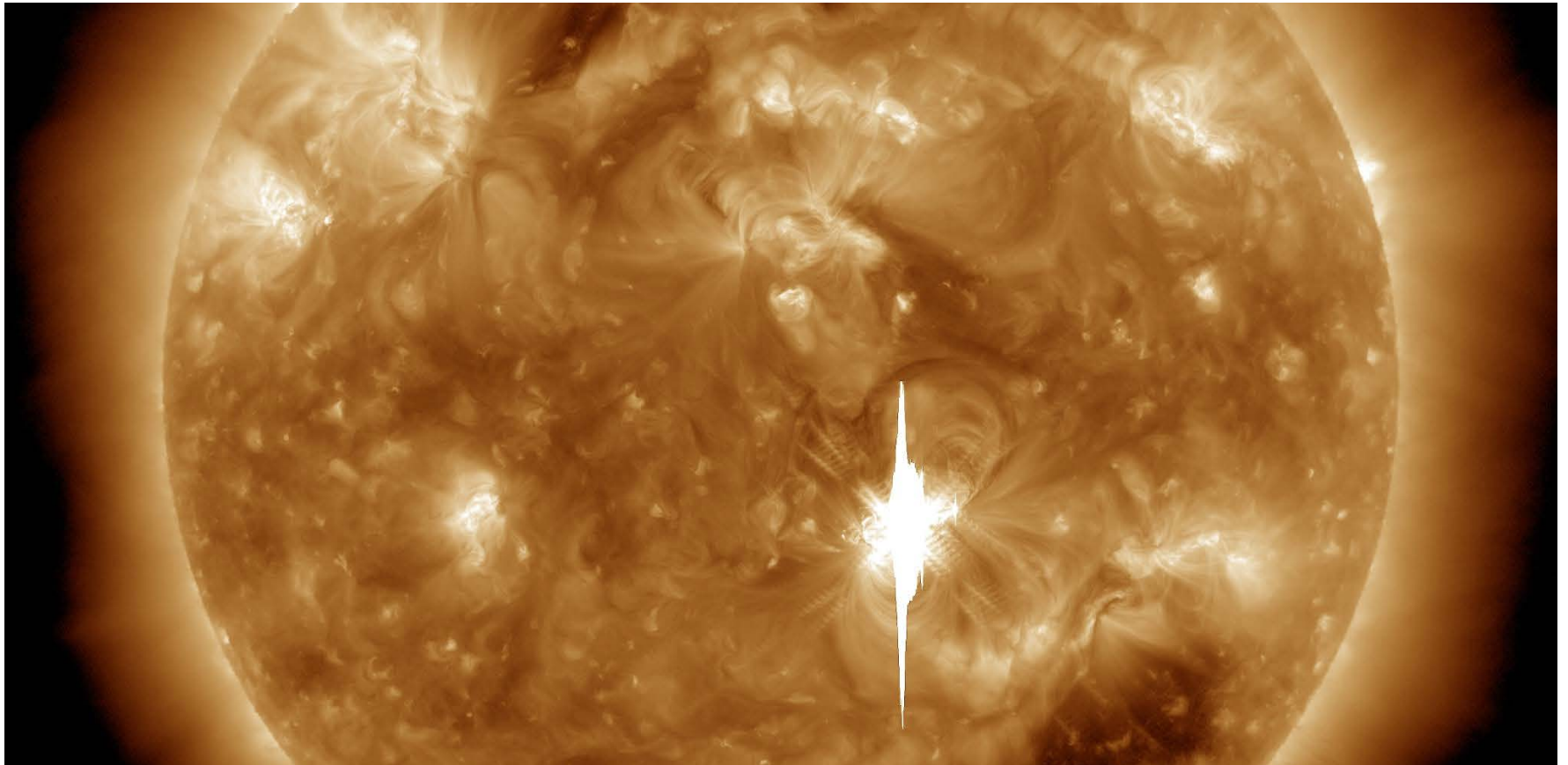
Parallel Data Vis.



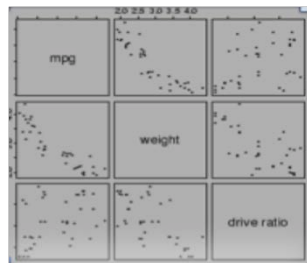
Multivariate to Multi-Run Visual Analysis



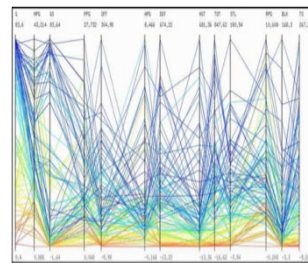
Multivariate Data



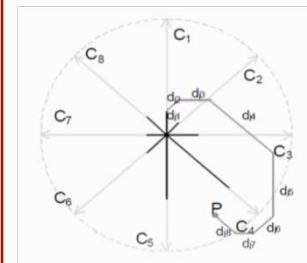
Multi-Dimensional Data Vis



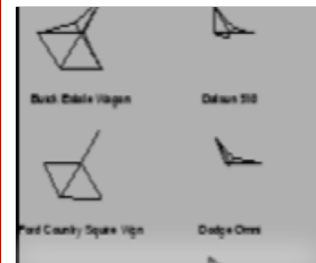
Scatter Plot Matrix



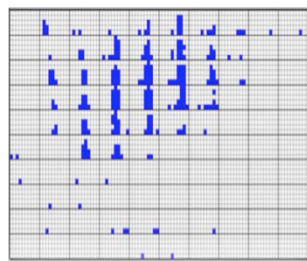
Parallel Coordinates



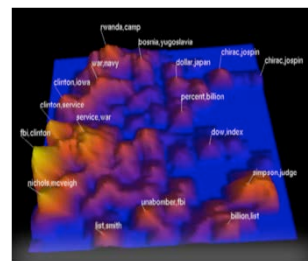
Star Coordinates



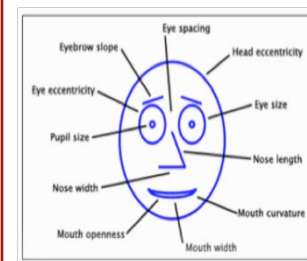
Star Glyphs



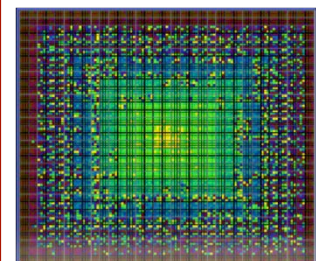
Dimension Stacking



Landscapes



Chernoff faces



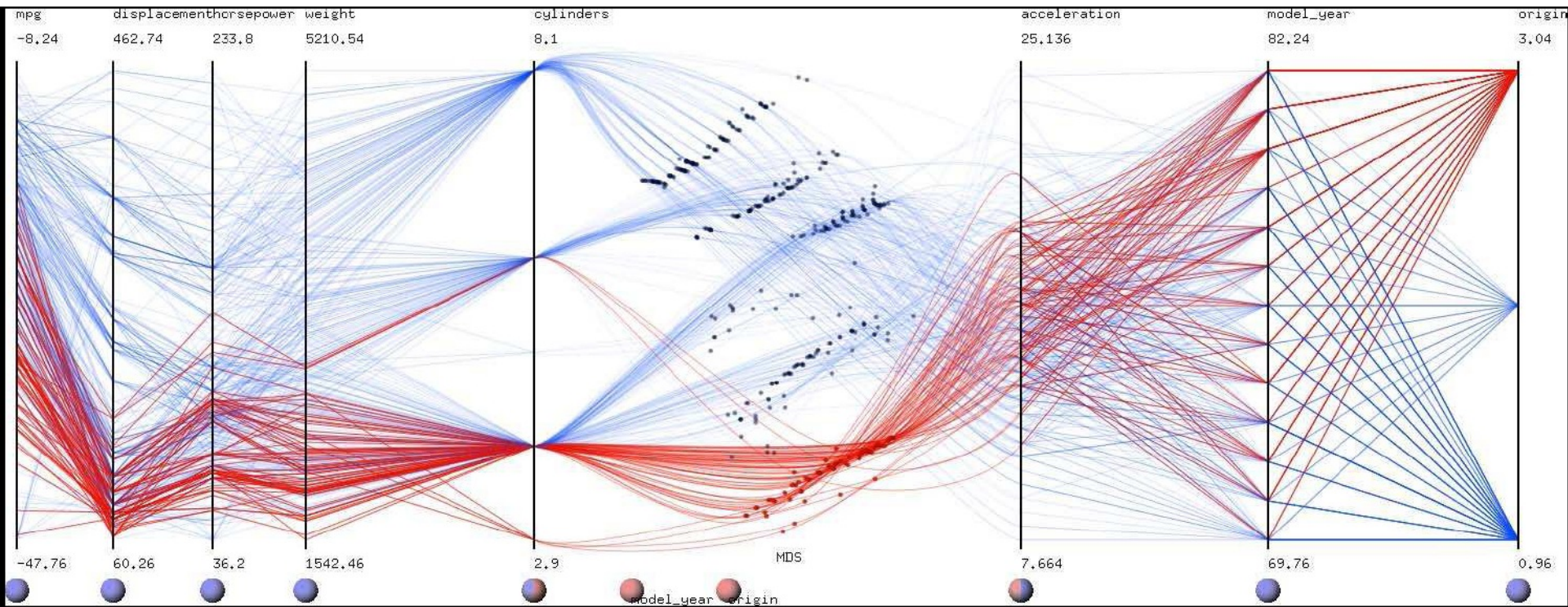
Pixel-based

Parallel Coordinates Basics

- To represent N dimensional data
 - Set N vertical axes in parallel
 - Put data to intersects on corresponding axes
 - Connect intersects

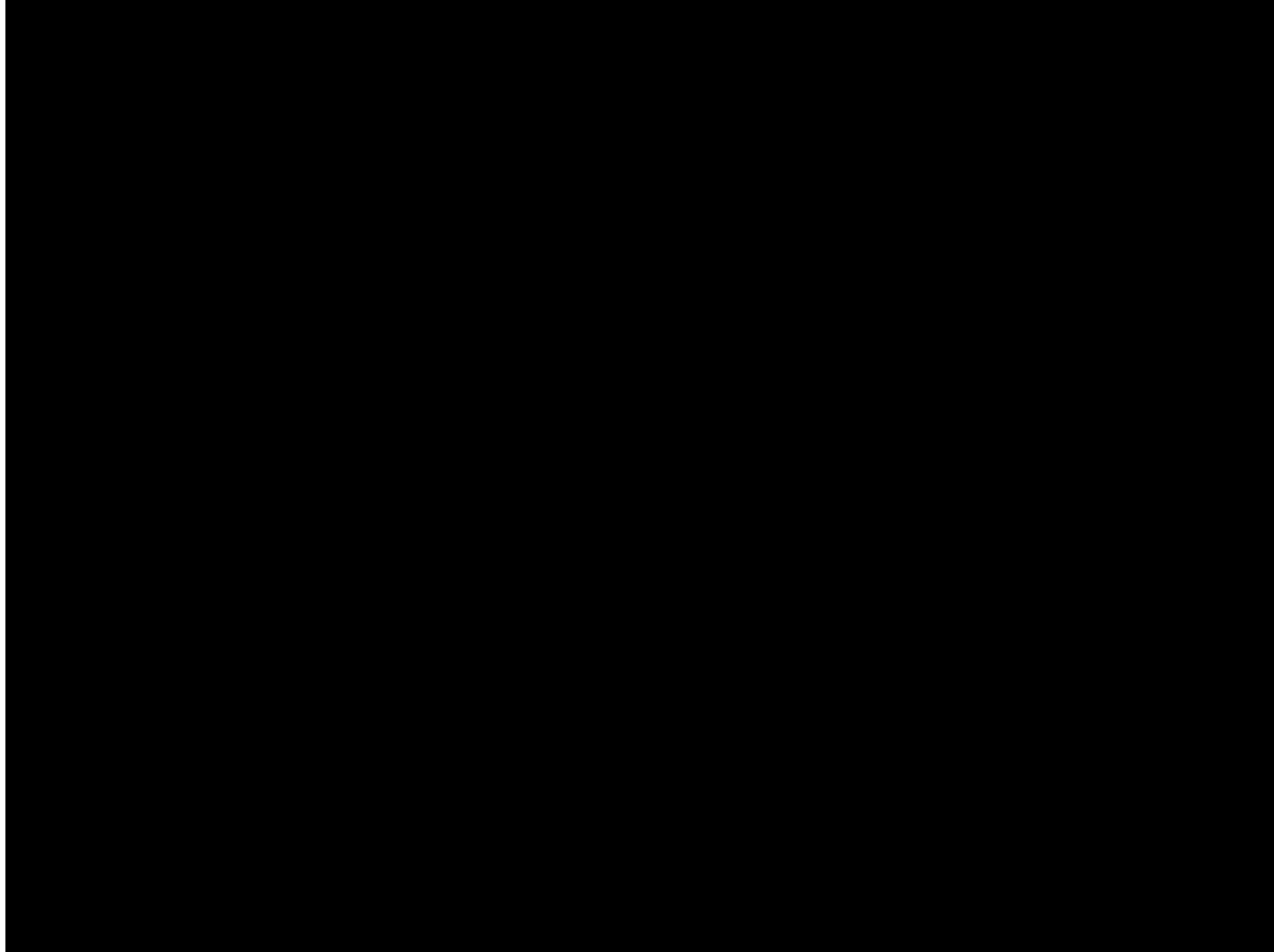


Data Exploration with SPPC

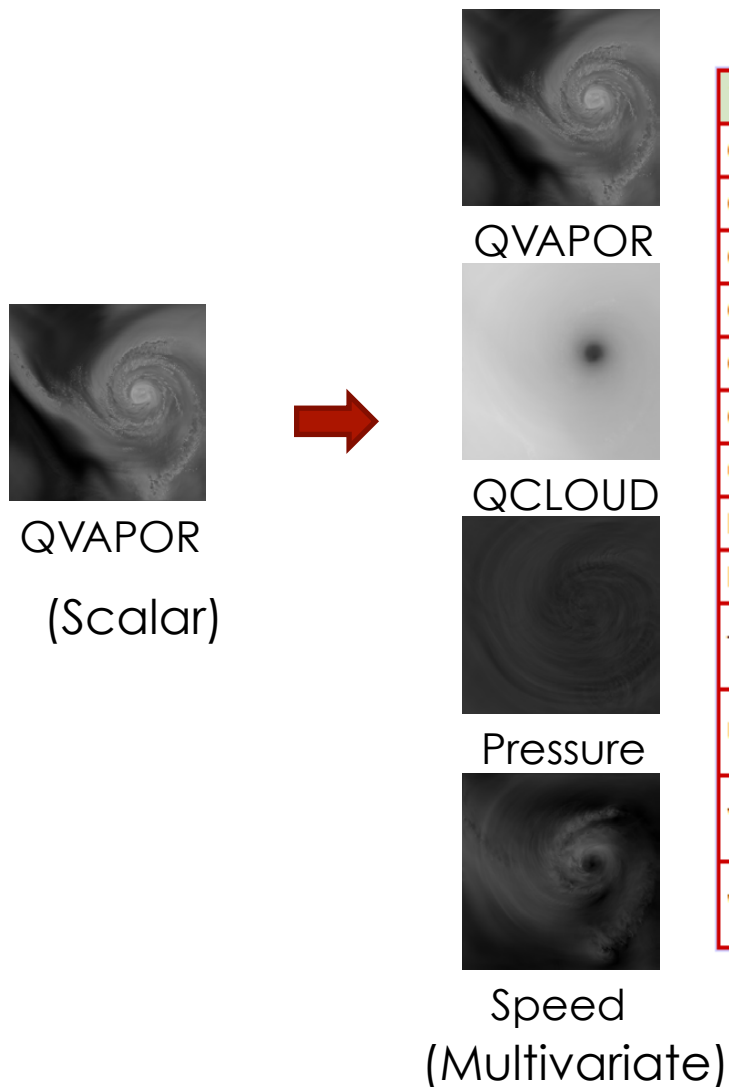


[Yuan et al. TVCG 2009 (InfoVis)]

Data Exploration with SPPC



Multivariate Visual Analysis

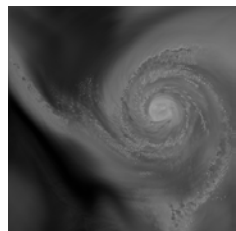


Variable	Description	Min / Max
QCLOUD	Cloud water	0.00000 / 0.00332
QGRAUP	Graupel	0.00000 / 0.01638
QICE	Cloud ice	0.00000 / 0.00099
QRAIN	Rain	0.00000 / 0.01132
QSNOW	Snow	0.00000 / 0.00135
QVAPOR	Water vapor	0.00000 / 0.02368
CLOUD	Total cloud (QICE + QCLOUD)	0.00000 / 0.00332
PRECIP	Total precipitation (QGRAUP+QRAIN+QSNOW)	0.00000 / 0.01672
P	Pressure: weight of the atmosphere above a grid point	-5471.85791 / 3225.42578
TC	Temperature	-83.00402/31.51576
U	X wind component: west-east wind component in model coordinate; positive means winds blow from west to east	-79.47297/85.17703
V	Y wind component: south-north wind component in model coordinate; positive means winds blow from south to north	-76.03391/82.95293
W	Z wind component: vertical wind component in model coordinate; positive means upward motion	-9.06026/28.61434

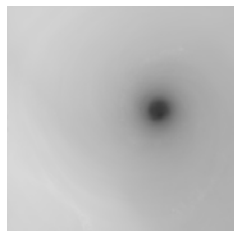
Isabel Hurricane

Transfer Functions

- Multivariate TFs for multi-modal data



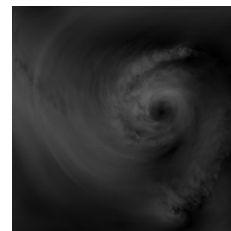
QVAPOR



QCLOUD

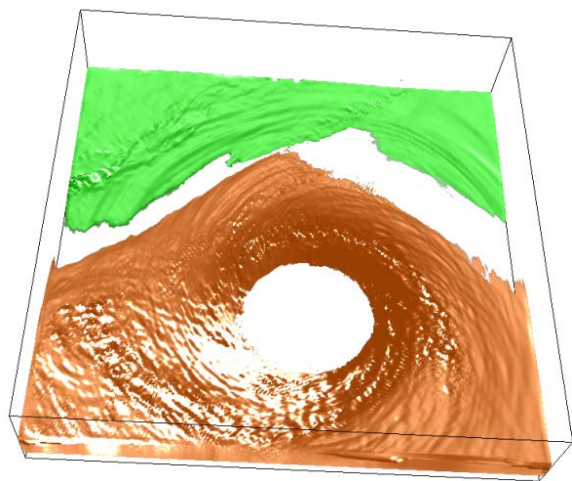


Pressure

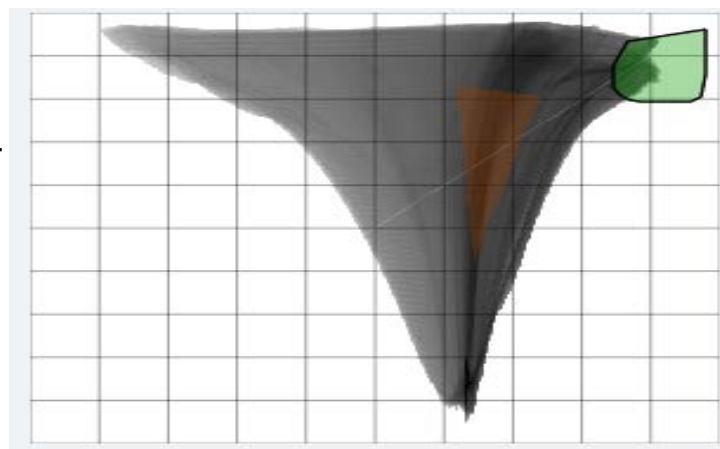


Speed

...



Temperature

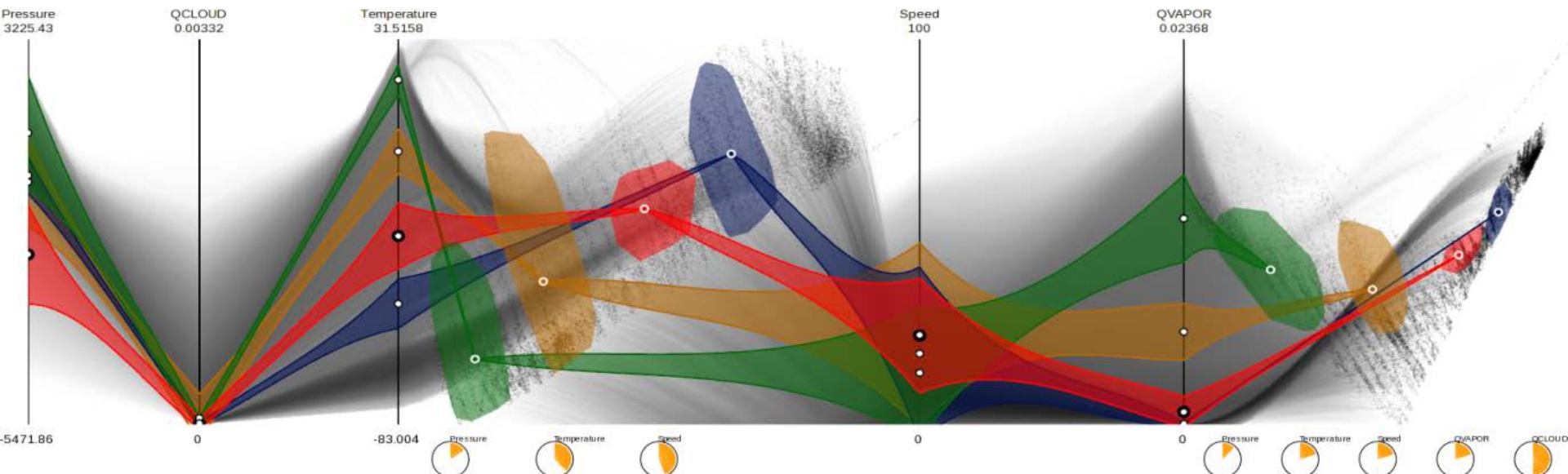


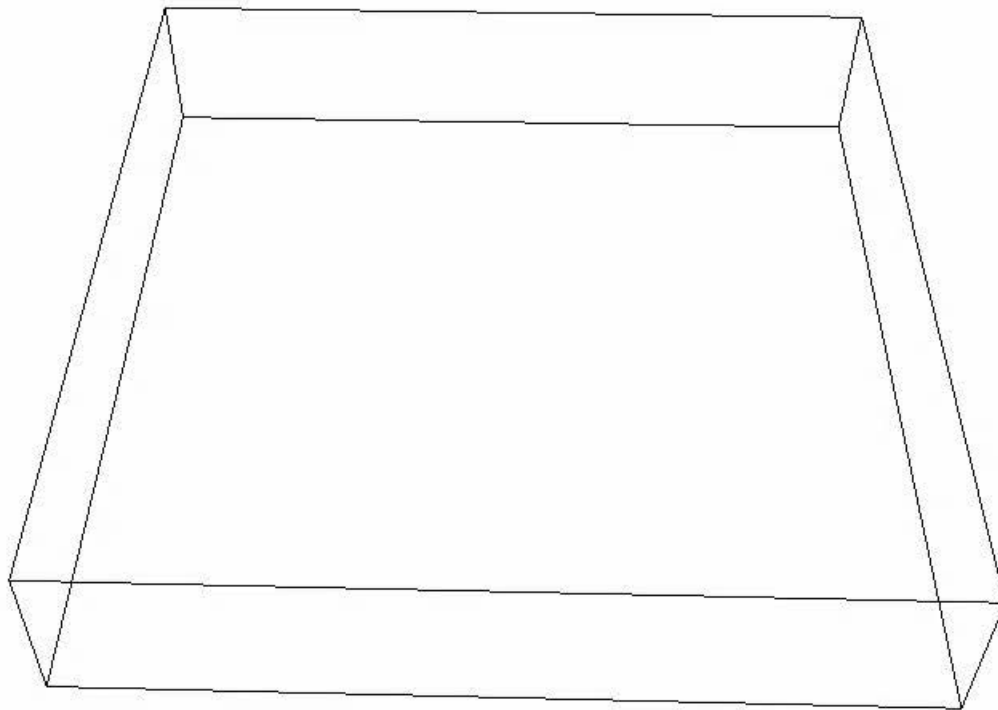
Pressure

High Dimensional Transfer Function Design

- Integrate MDS into PCP to facilitate multivariate TF design.
 - Avoids context jumps between polyline and point regions when exploring data clusters
 - Provides multiple perspective views upon the data, supporting linked queries

[Guo, Xiao, and Yuan. PacificVis 2011]





Pressure
3225.43

QCLOUD
0.00332

Temperature
31.5158

Speed
100

QVAPOR
0.02368

-5471.86

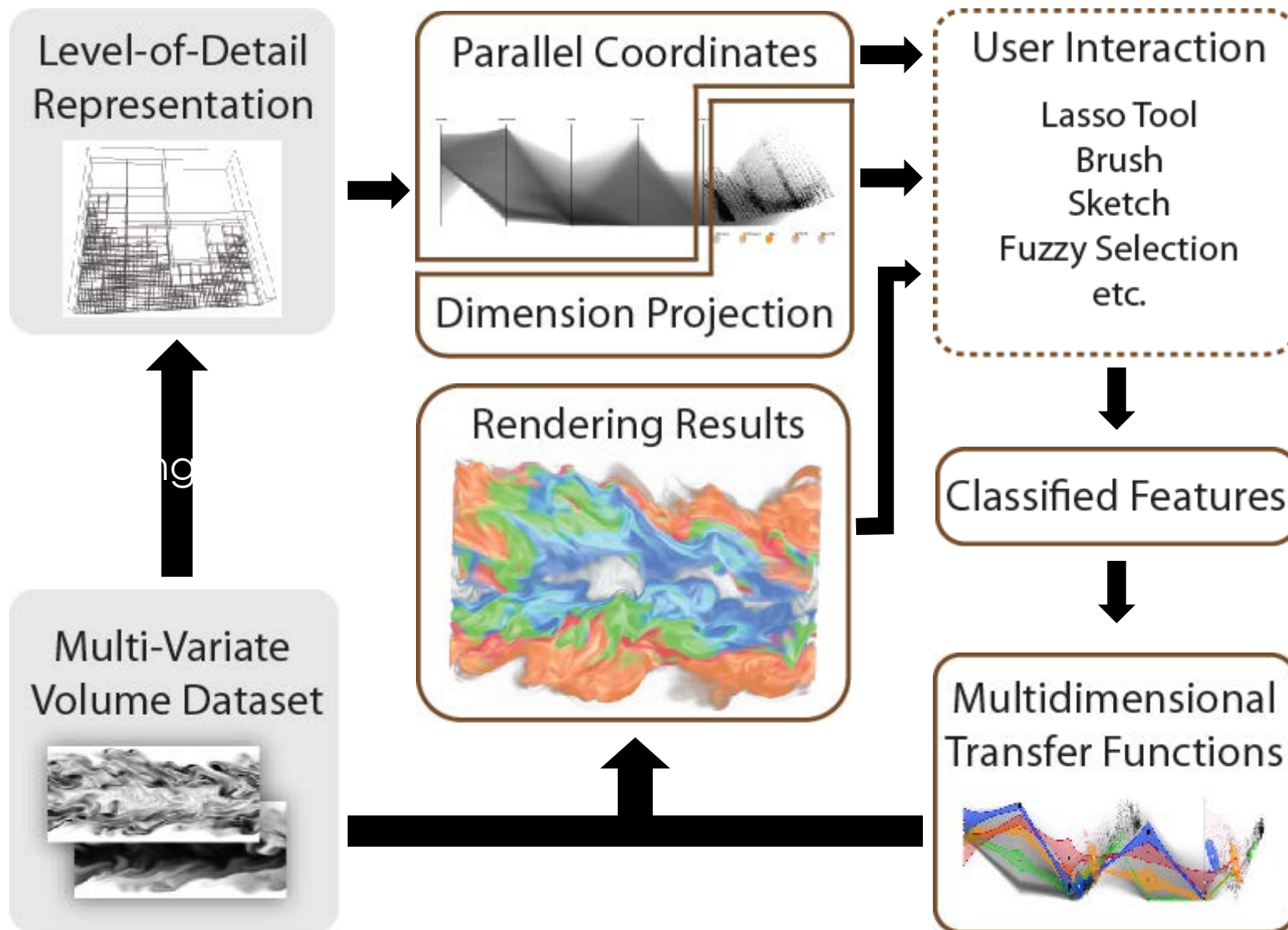
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-83.004

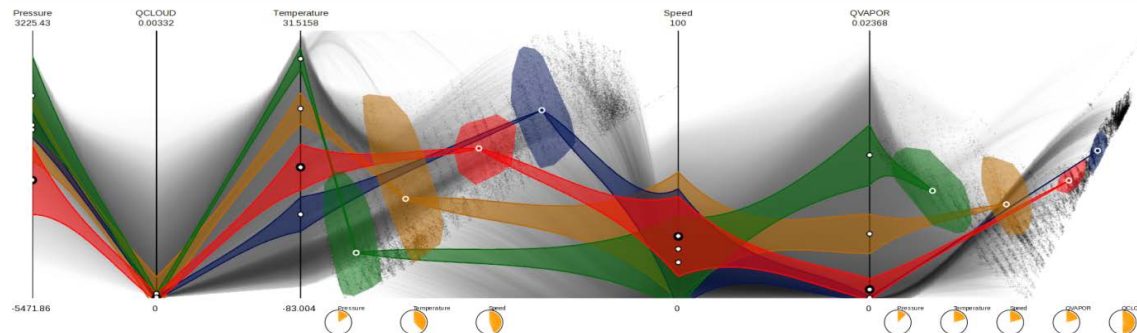
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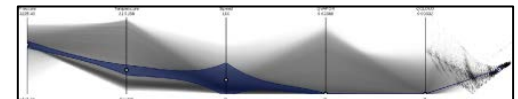
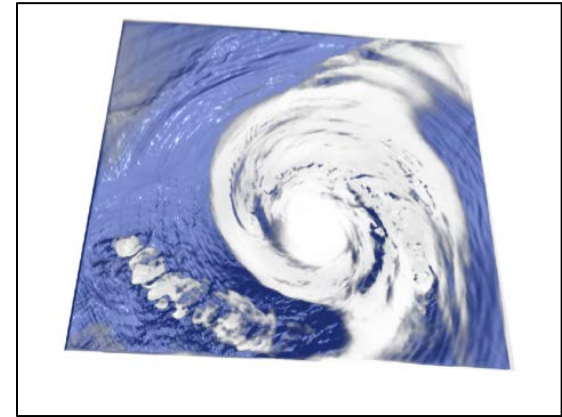
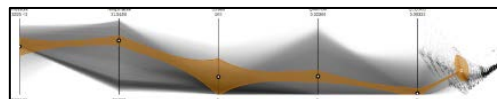
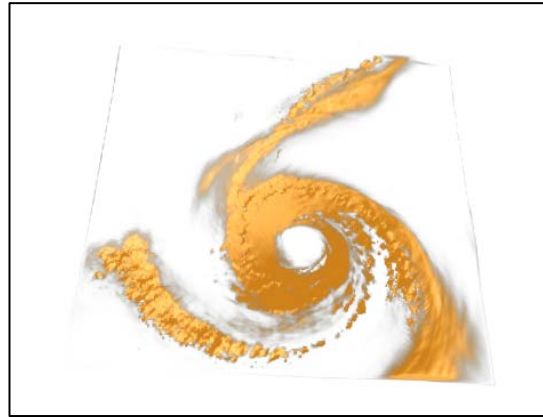
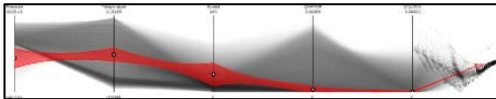
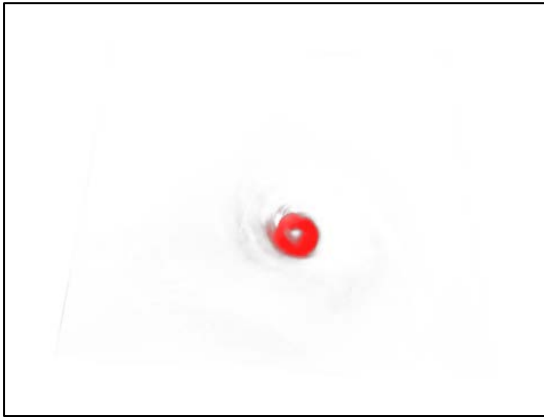


Case Study – Hurricane Isabel



Case Study – Hurricane Isabel

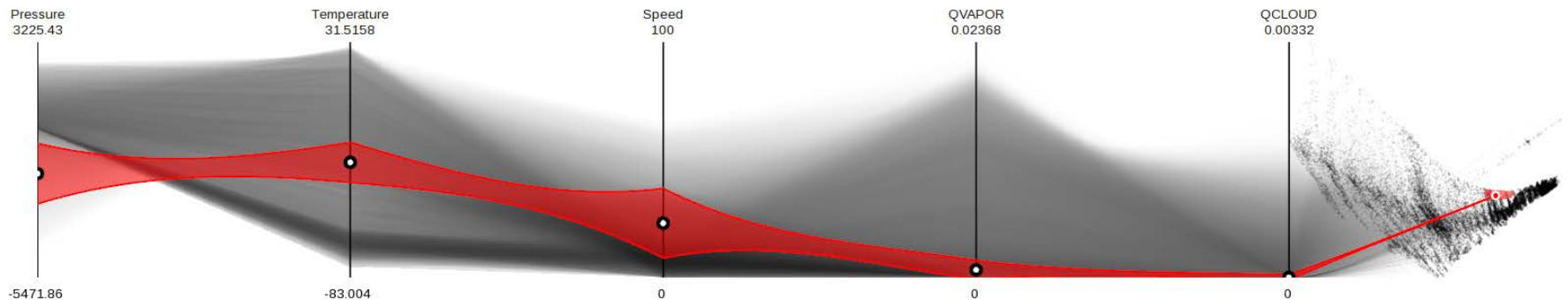
- Different components



Case Study – Hurricane Isabel



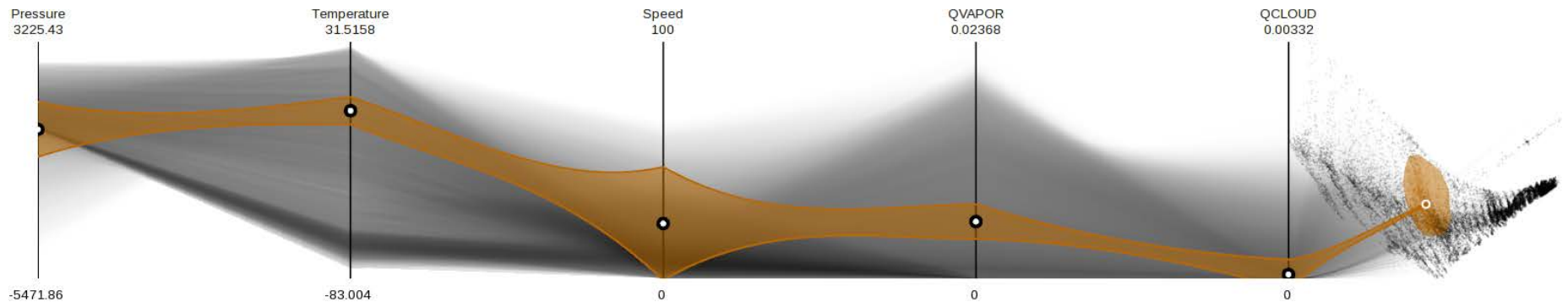
Red part (hurricane eye):
low pressure,
medium temperature,
lower QCLOUD,
high wind speed



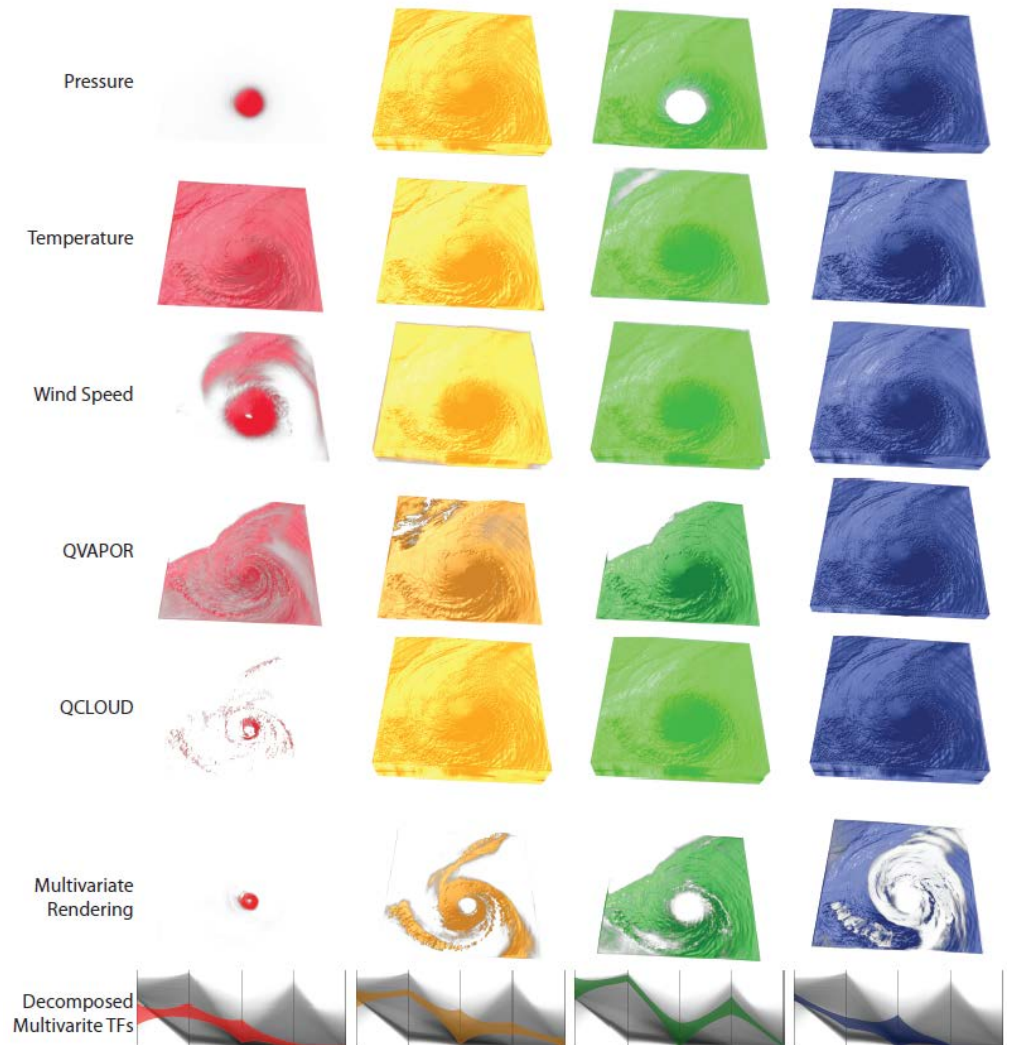
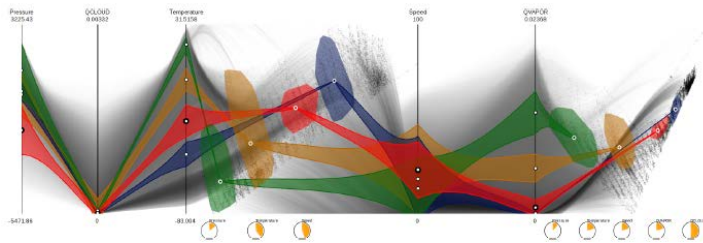
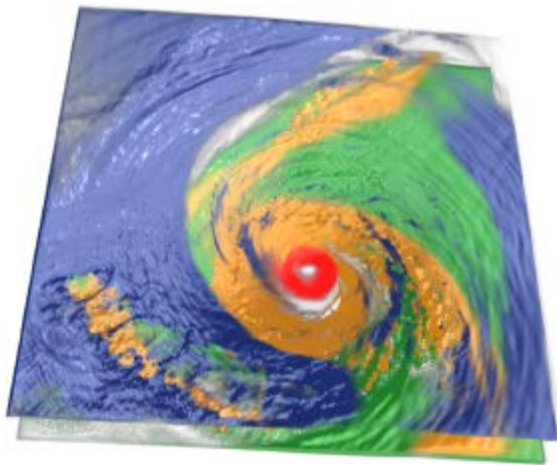
Case Study – Hurricane Isabel



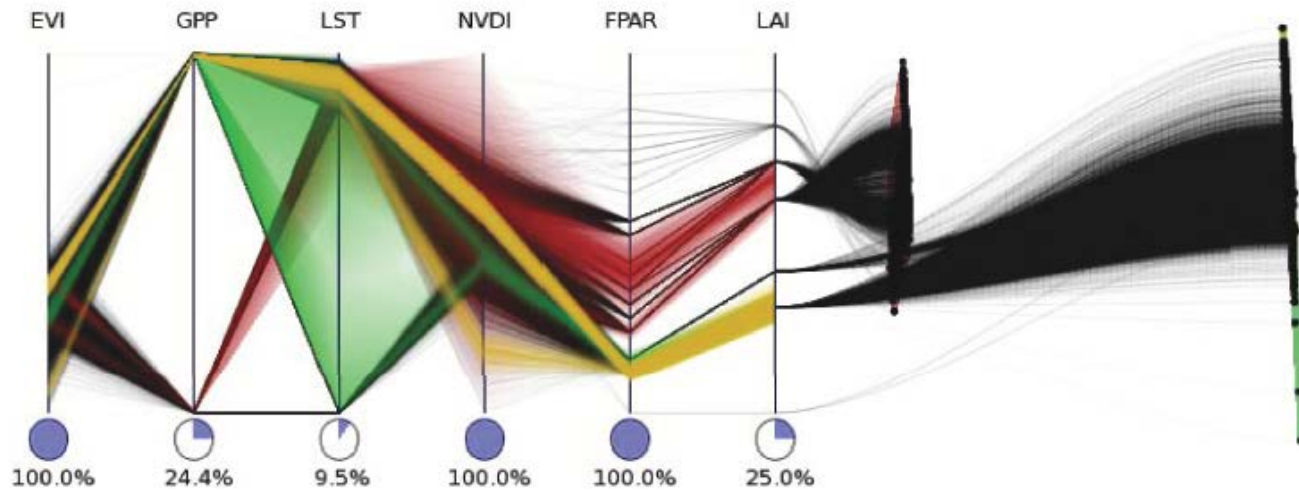
Yellow part:
higher pressure
lower wind speed



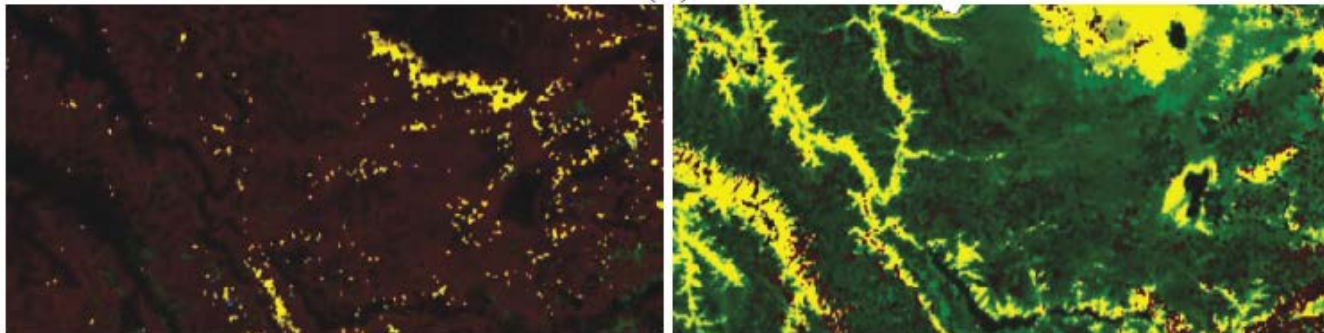
Hurricane



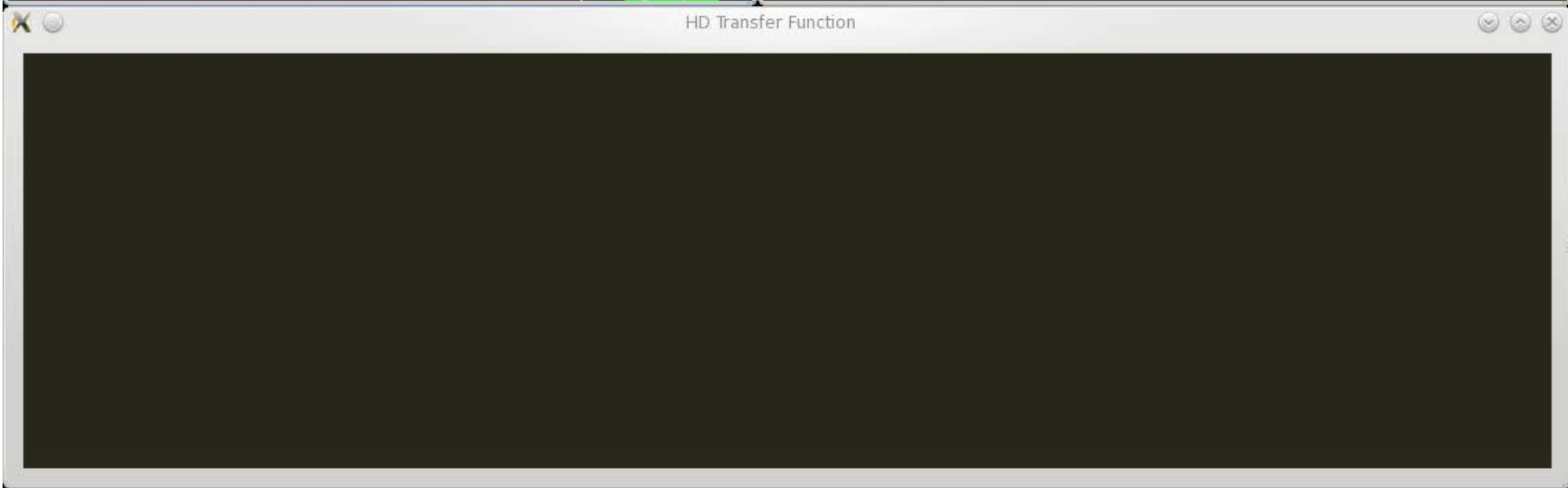
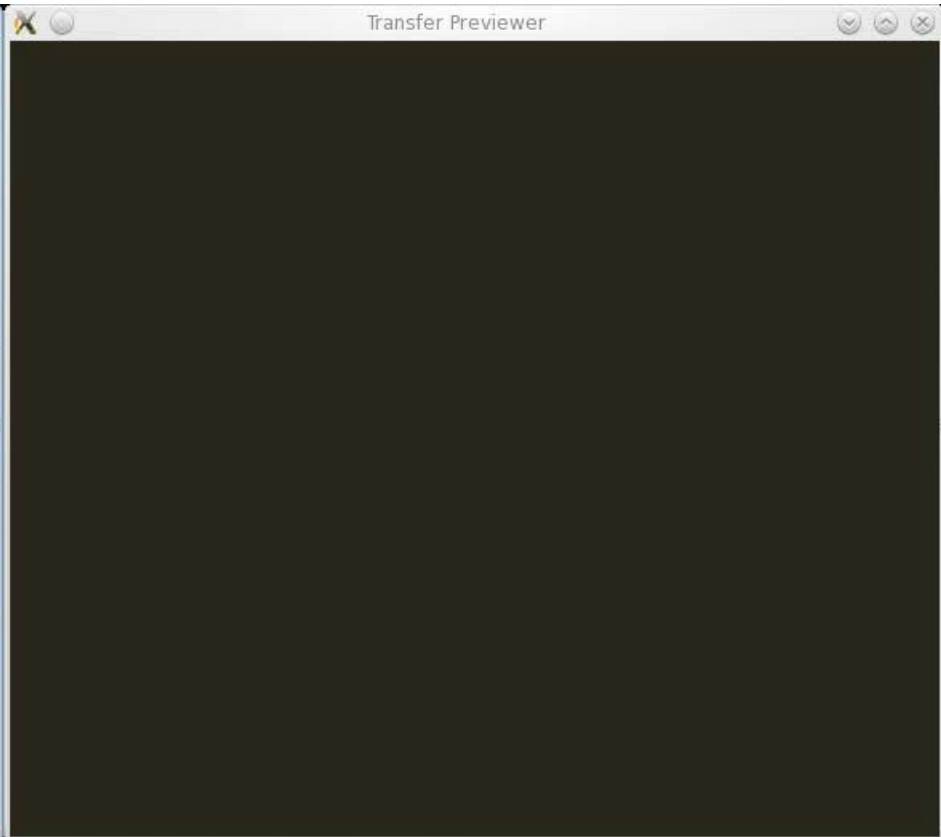
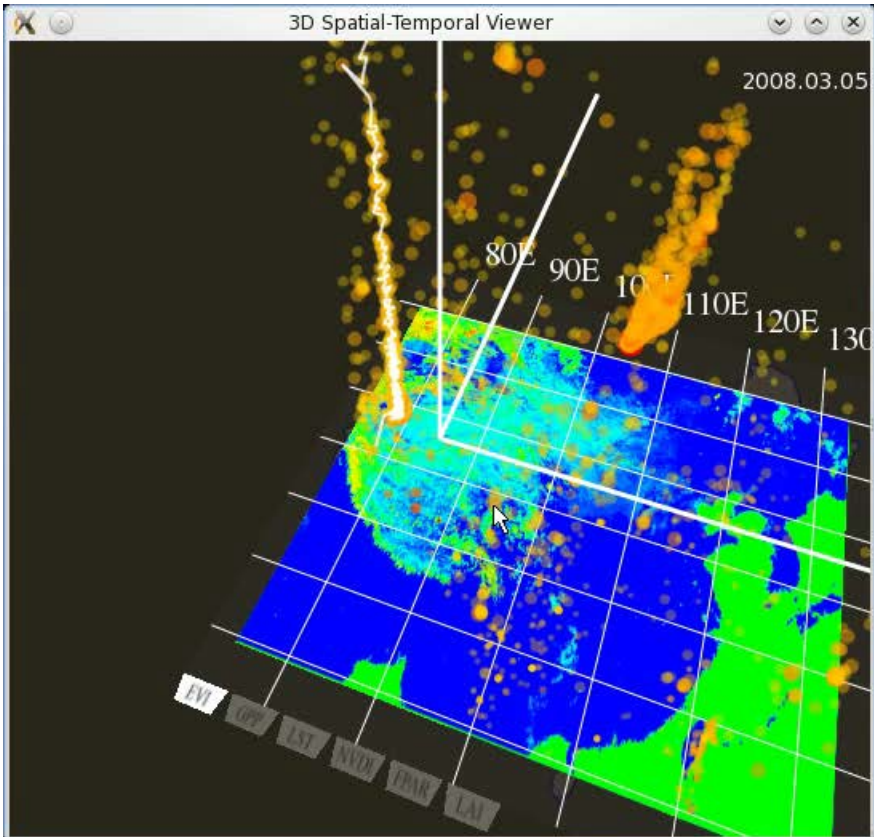
Multivariate Analytics for Images



(c)



[Yuan et al. TVCG 2010 (Vis)]



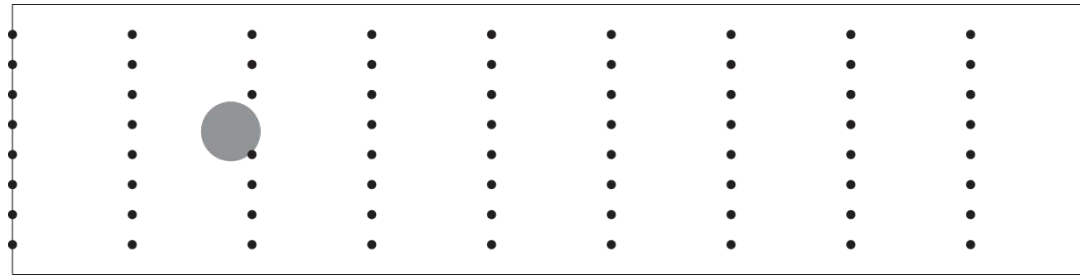
Scalable Multivariate Analytics of Seismic and Satellite-based Observational Data



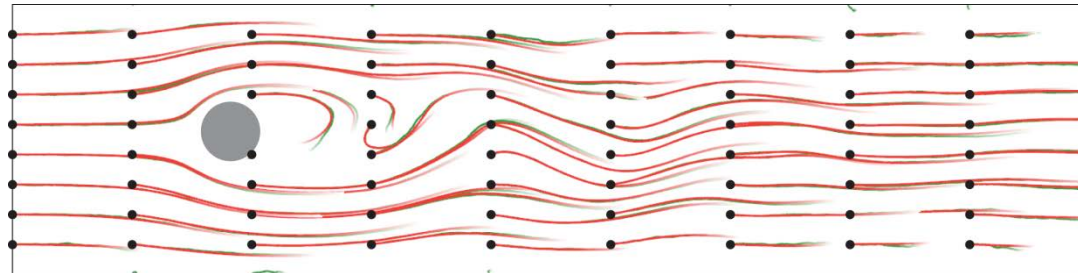
[Yuan et al. TVCG 2010 (Vis)]

Eulerian and Lagrangian Specifications

- Eulerian: $\mathbf{v} = \mathbf{v}(\mathbf{x}, t), p = p(\mathbf{x}, t), T = T(\mathbf{x}, t),$



- Lagrangian: $\mathbf{X} = \mathbf{X}(\mathbf{a}, t), p = p(\mathbf{a}, t), T = T(\mathbf{a}, t),$

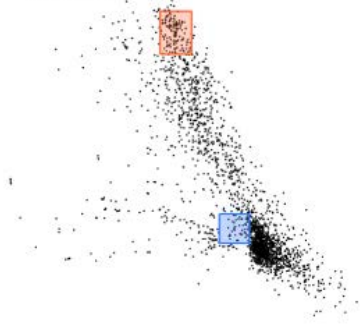


- Relationships between two specifications (flow map):

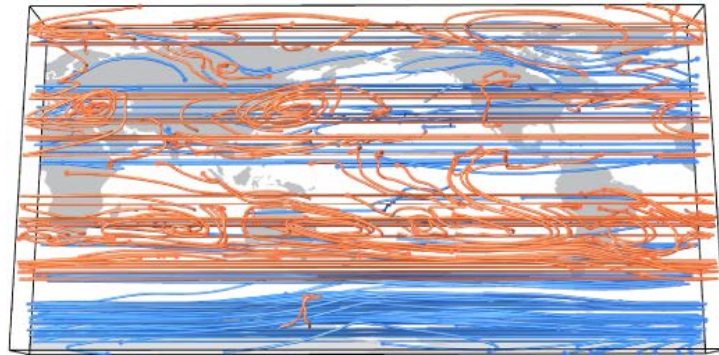
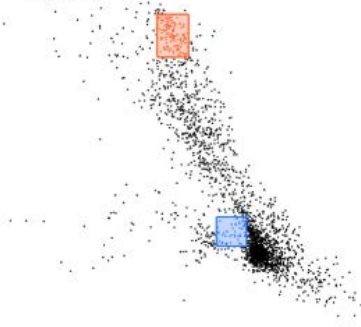
$$\mathbf{v}(\mathbf{X}(\mathbf{a}, t), t) = \frac{\partial \mathbf{X}(\mathbf{a}, t)}{\partial t}. \quad \Phi : \mathbf{x} \mapsto \Phi_{t_0}^{t_0+t}(\mathbf{x})$$

GEOS-5 Simulations

Run 0

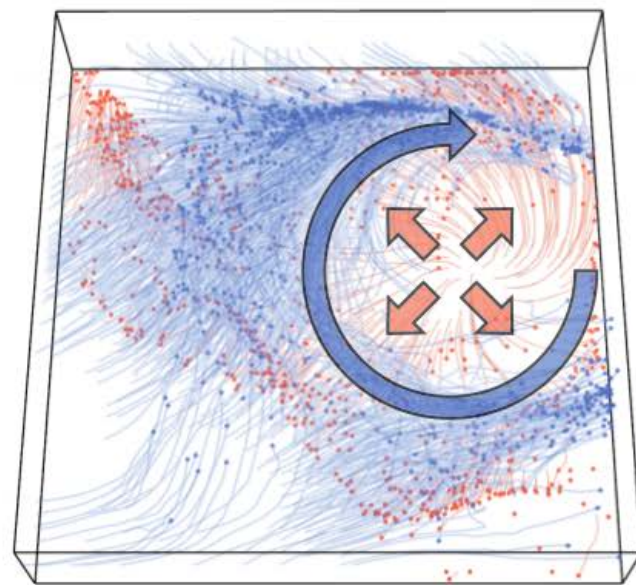
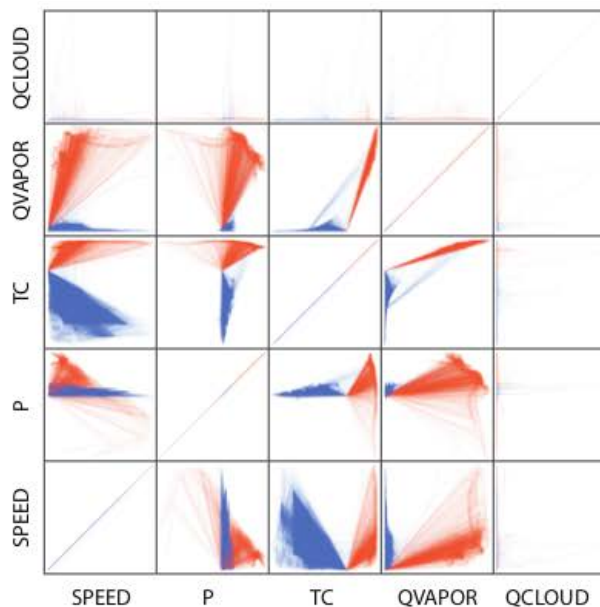
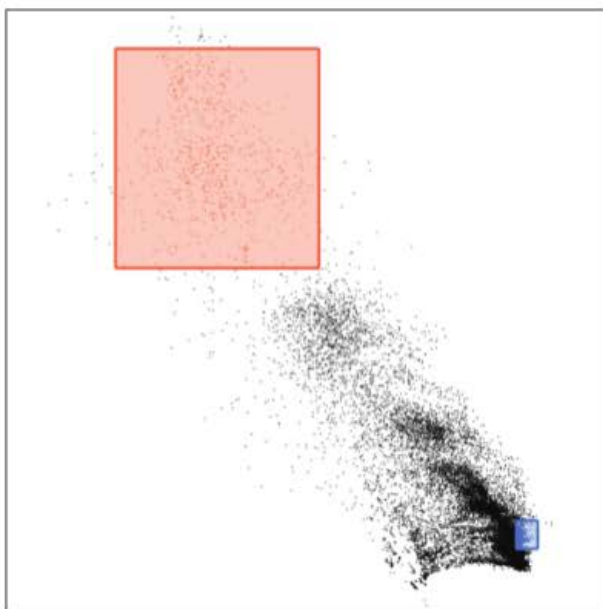


Run 1



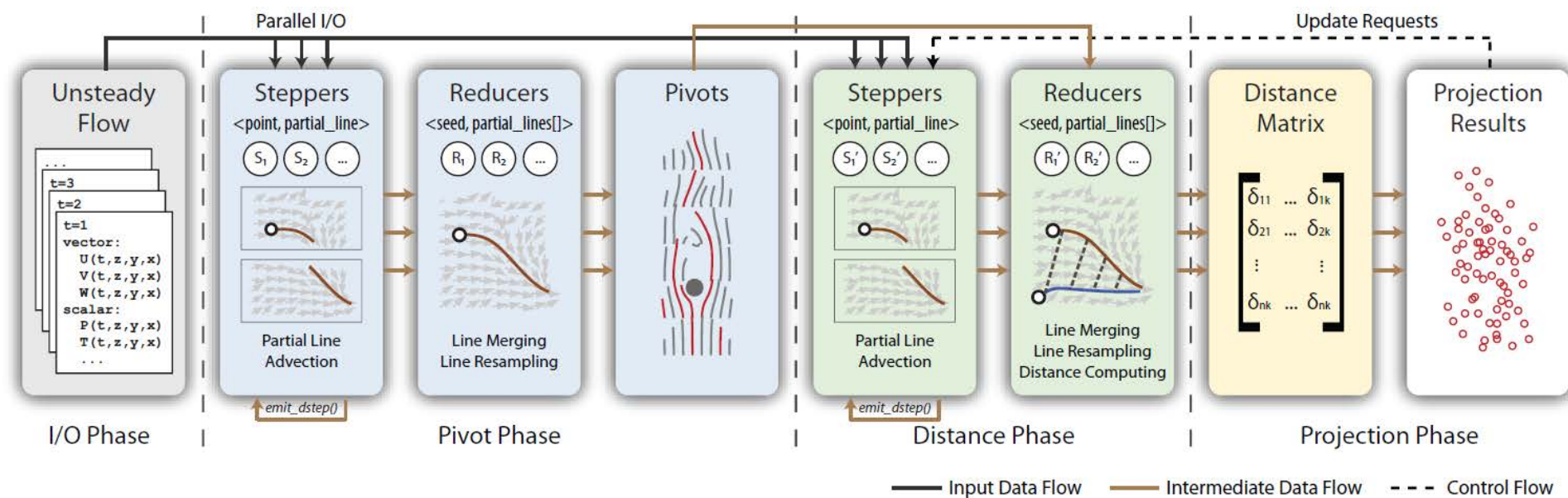
Hurricane Isabel Simulation

- Attributes along pathlines are significantly different in the two group of selected features



The Workflow of the Parallel System LASP

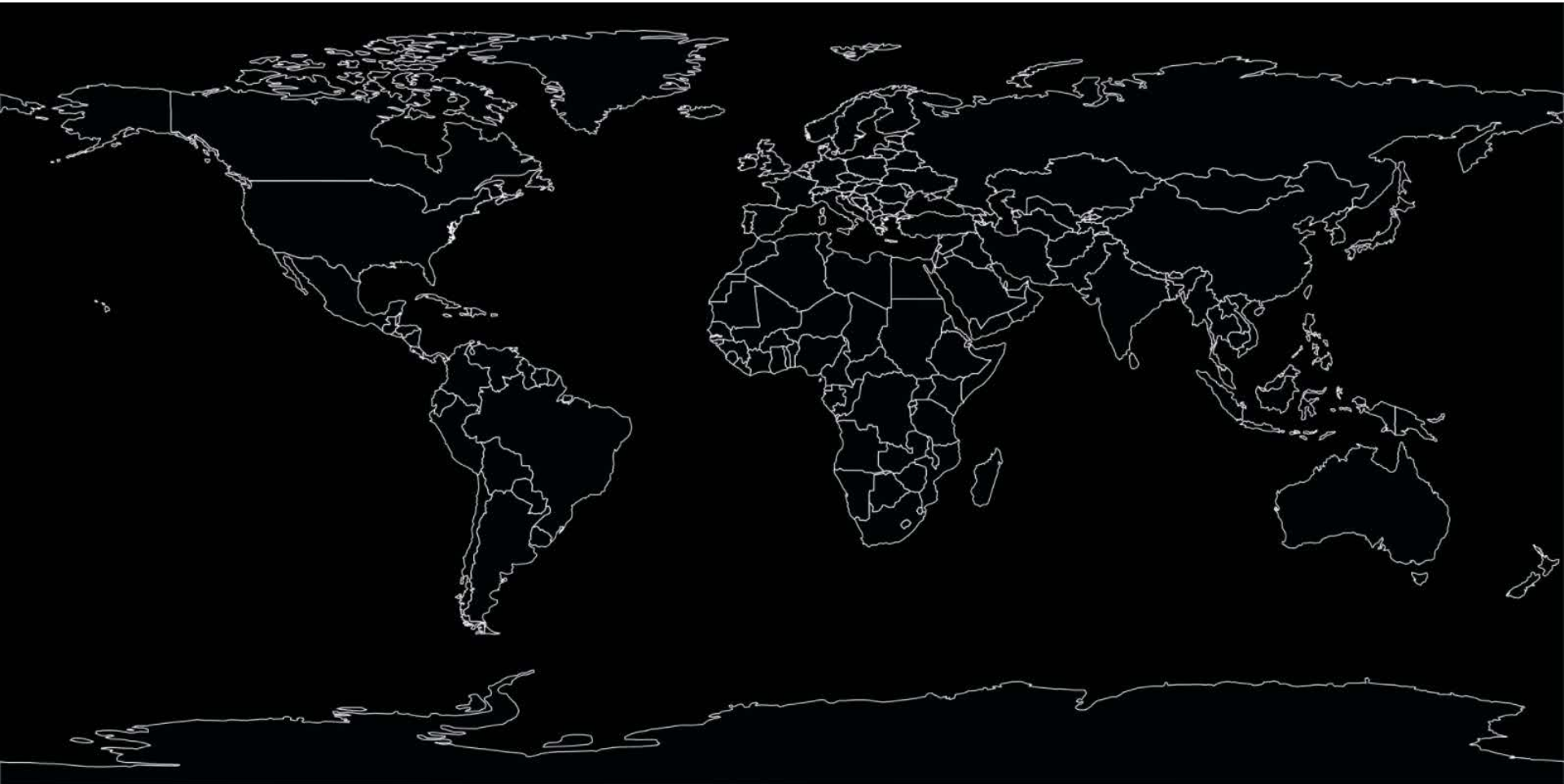
- I/O Phase: loading blocked data in parallel
- Pivot Phase: computing a small number of pivots
- Distance Phase: computing huge number of pathlines, and their distances to pivots
- Projection Phase: projecting with SPMDs in reducers



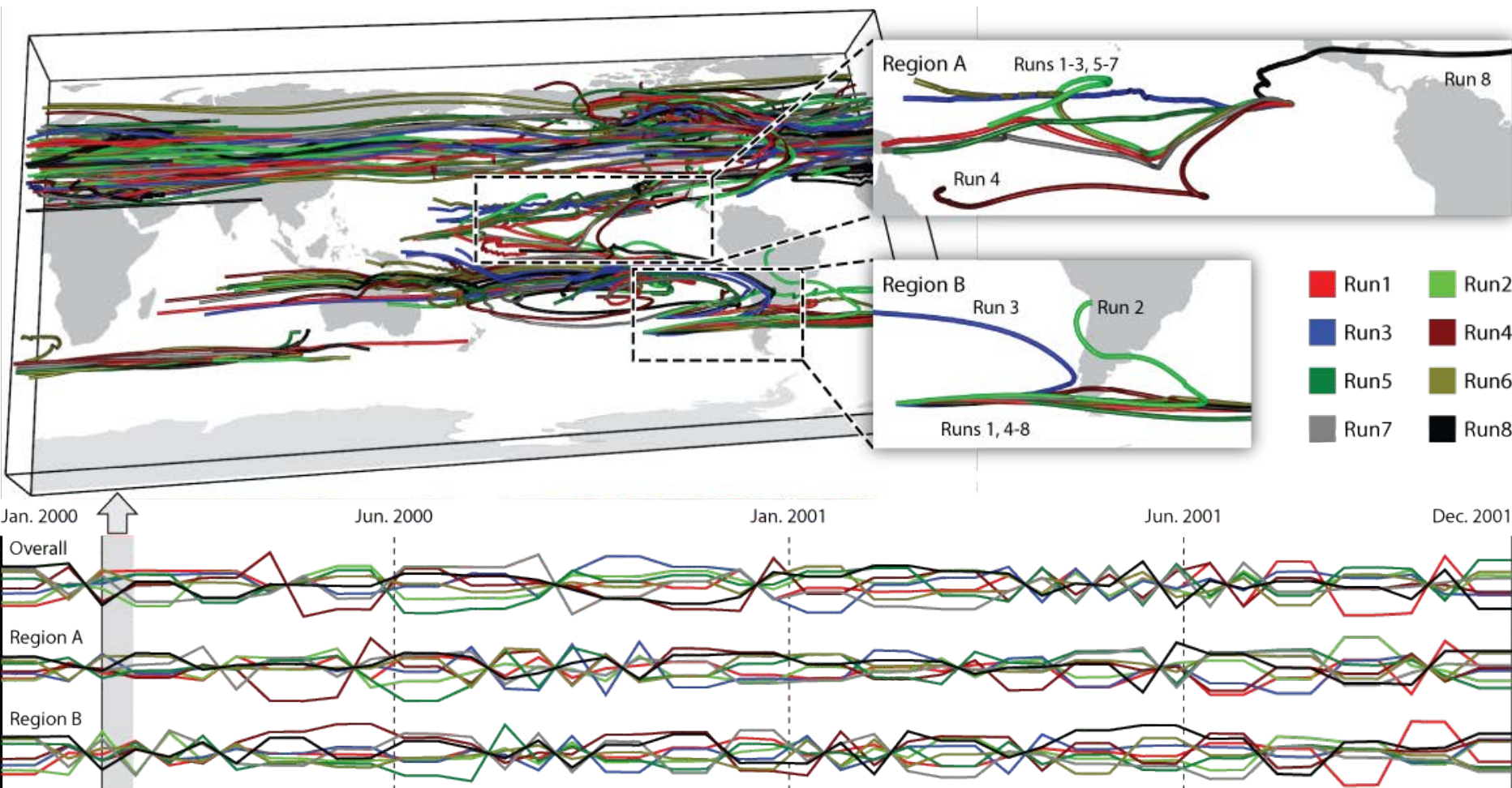
Pollution From China



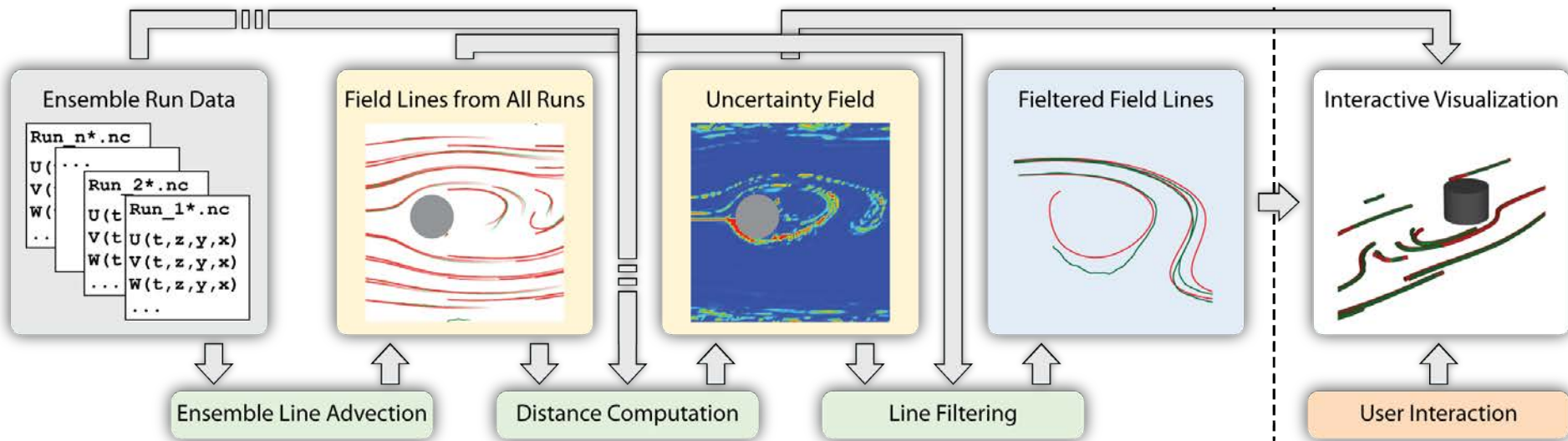
Pollution from USA



Case: GEOS-5 Simulation



Conceptual Flow Chart



- Ensemble data (large)
- Field line data (much larger than ensemble data)
- Variation field (small)
- Filtered lines (even smaller)

Benchmark Platform: NCSSJN

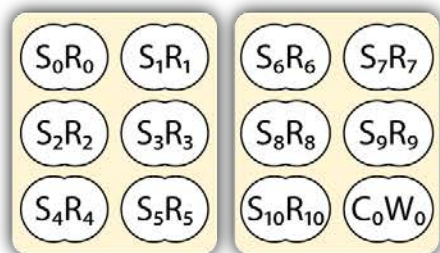
- ShenWei-based supercomputer
 - SW1600 processor, 1.0~1.1GHz
 - 1GB memory for each core
 - 40Gbps high-speed interconnection

- x86-based supercomputer
 - Intel Xeon E5675 hexa-core processor, 3.06GHz
 - 4GB memory for each core
 - QDR Infiniband interconnection

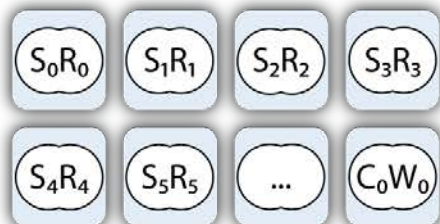
- Shared global filesystem: SWGFS

Scalability

- Strong scalability test in National Super Computer Center in Jinan (ShenWei and x86 architectures)

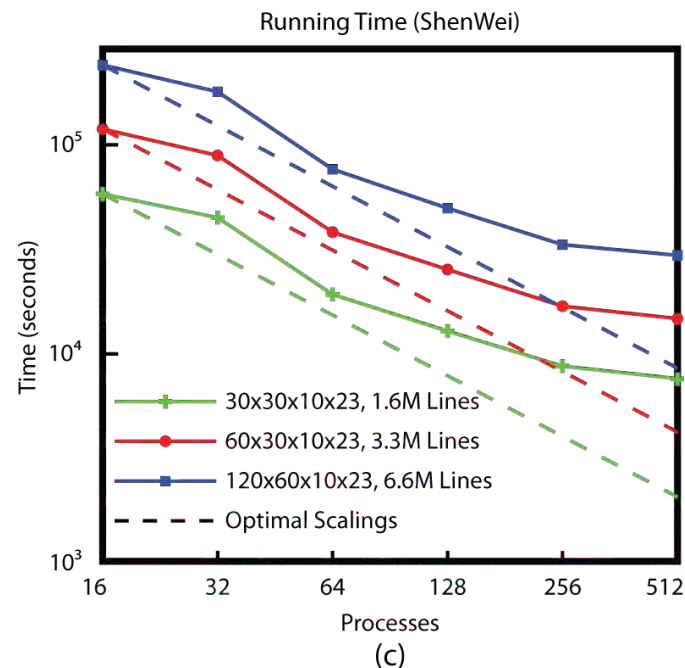
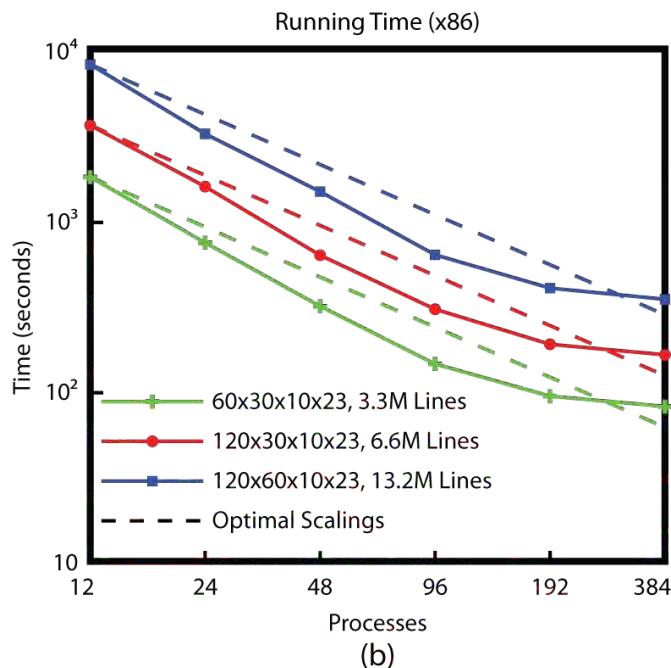


Intel Xeon E5675 (Hexa-Core)



ShenWei Virtual Node (4 Cores)

(a)

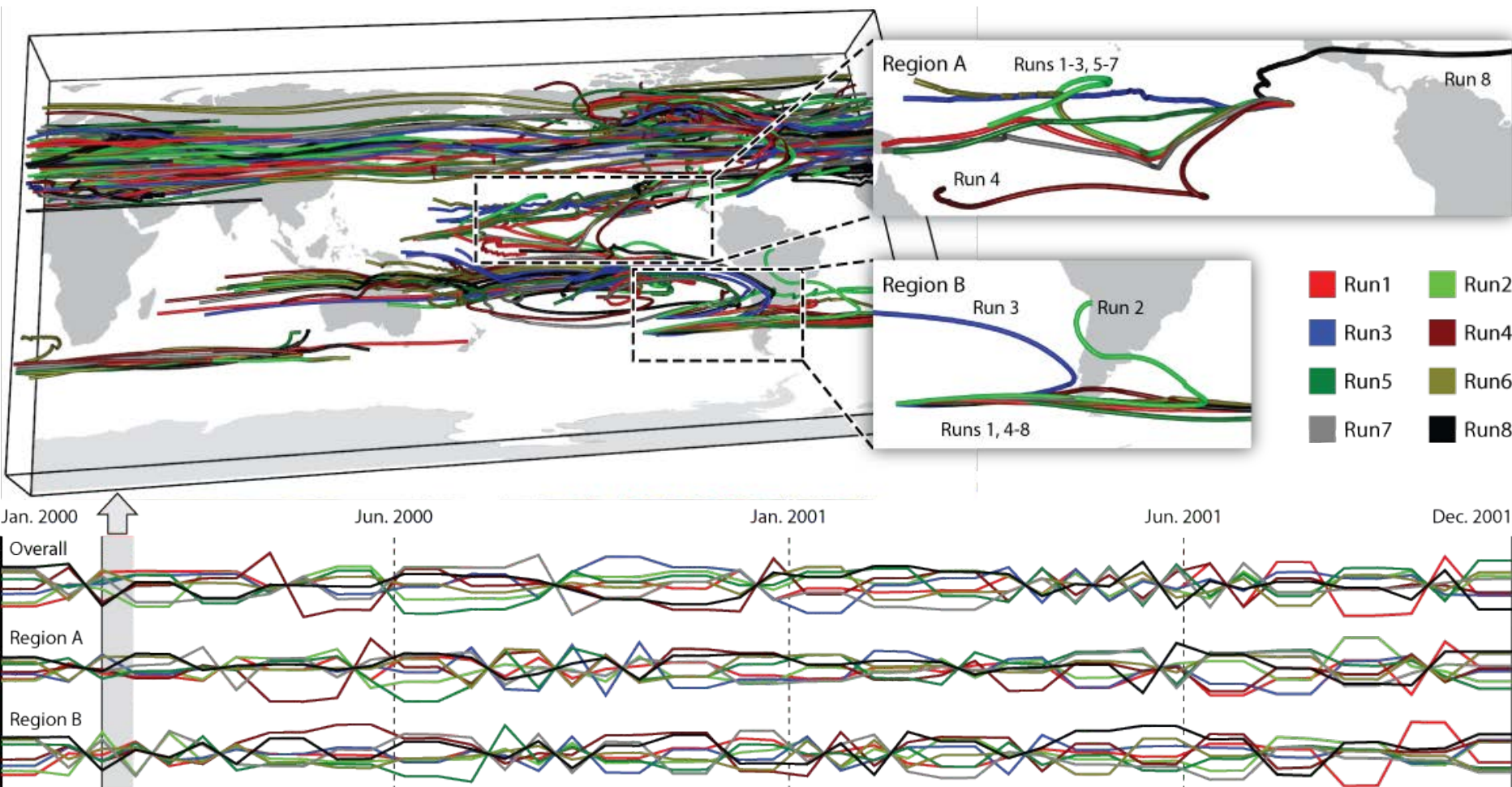


Applications

- GEOS-5 global climate model from NASA Goddard Space Flight Center
 - 8 run ensemble simulation
 - 1°x1.5° resolution, with 72 pressure levels
 - From Jan 2000 to Dec 2011

- WRF ARW model
 - **base** and **no_urban** runs to investigate the impact of urbanization
 - 100x100x27 resolution, East China
 - From 2012-7-1 00:00:00 UTC to 2012-7-10 18:00:00 UTC

GEOS-5 Simulation

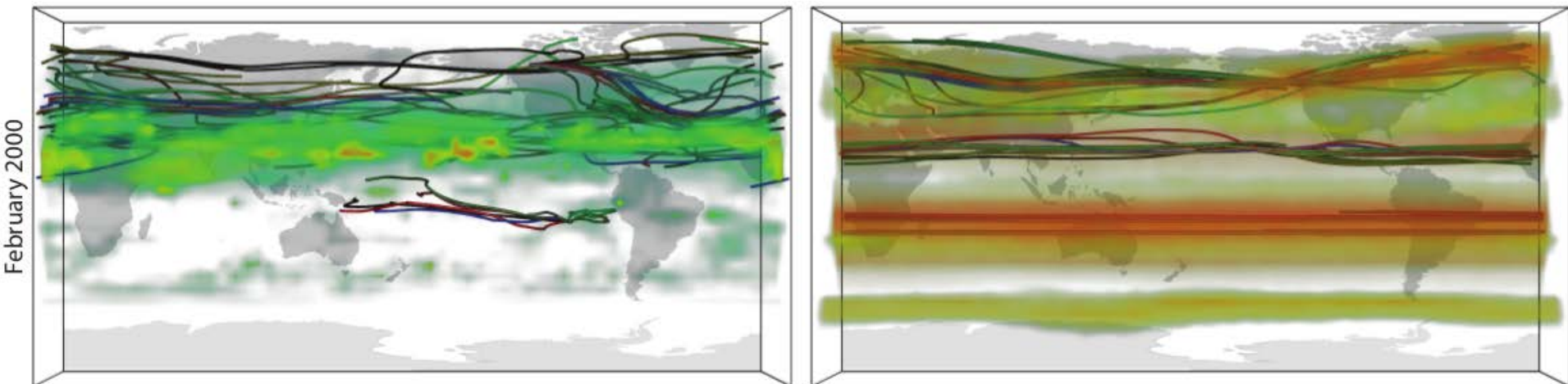


GEOS-5 Simulation

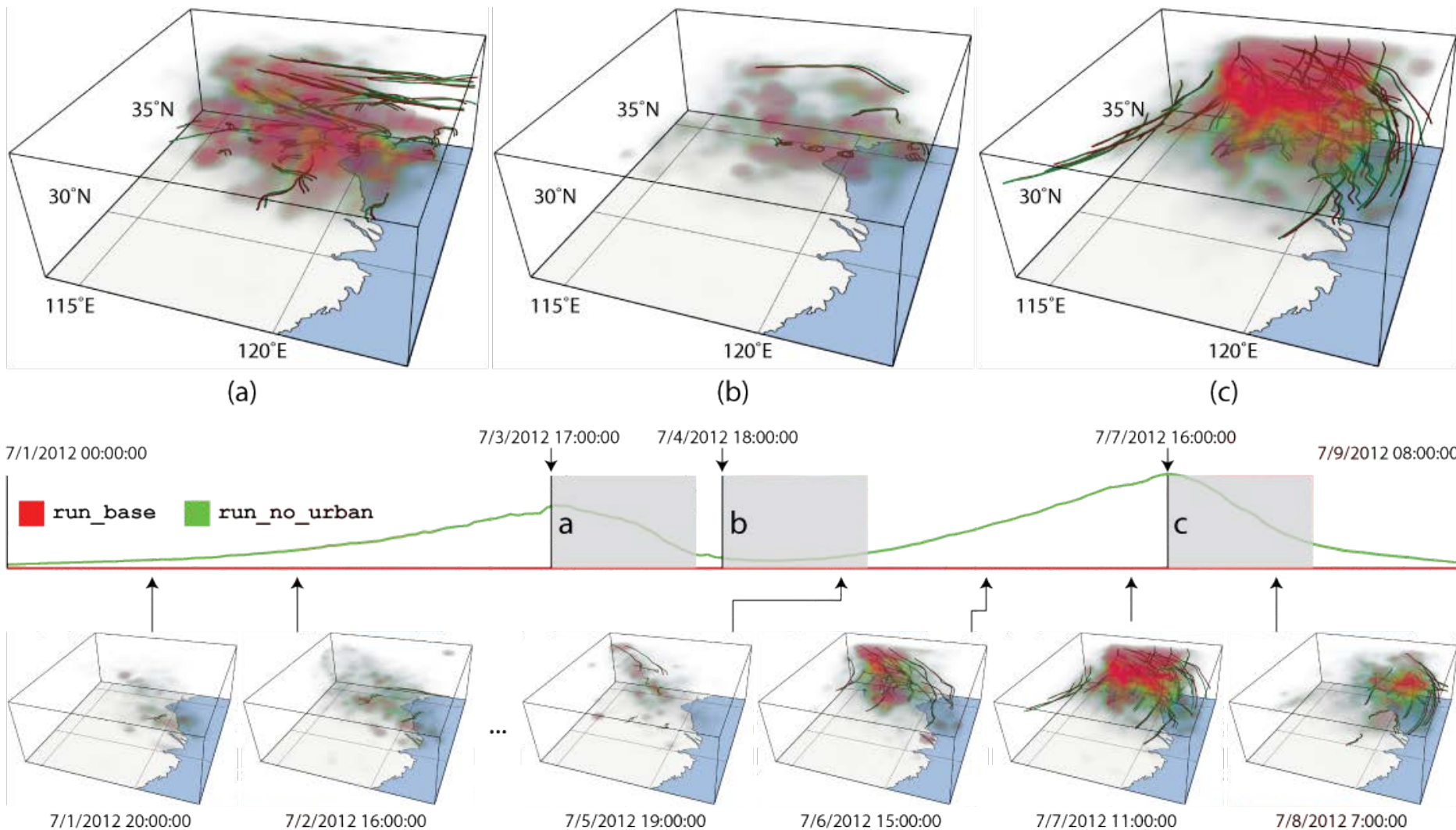
8-Run GEOS-5 Simulation Data
March, 2003

GEOS-5 Simulation: CO₂-based Metric

- The metric: the differences of locations / CO₂ concentration along the pathline
- Findings
 - The variation of the wind field is high in the north hemisphere
 - However, The CO₂ difference is higher in south hemisphere and some places in the north
 - CO₂ concentration is not sensitive to wind in above regions

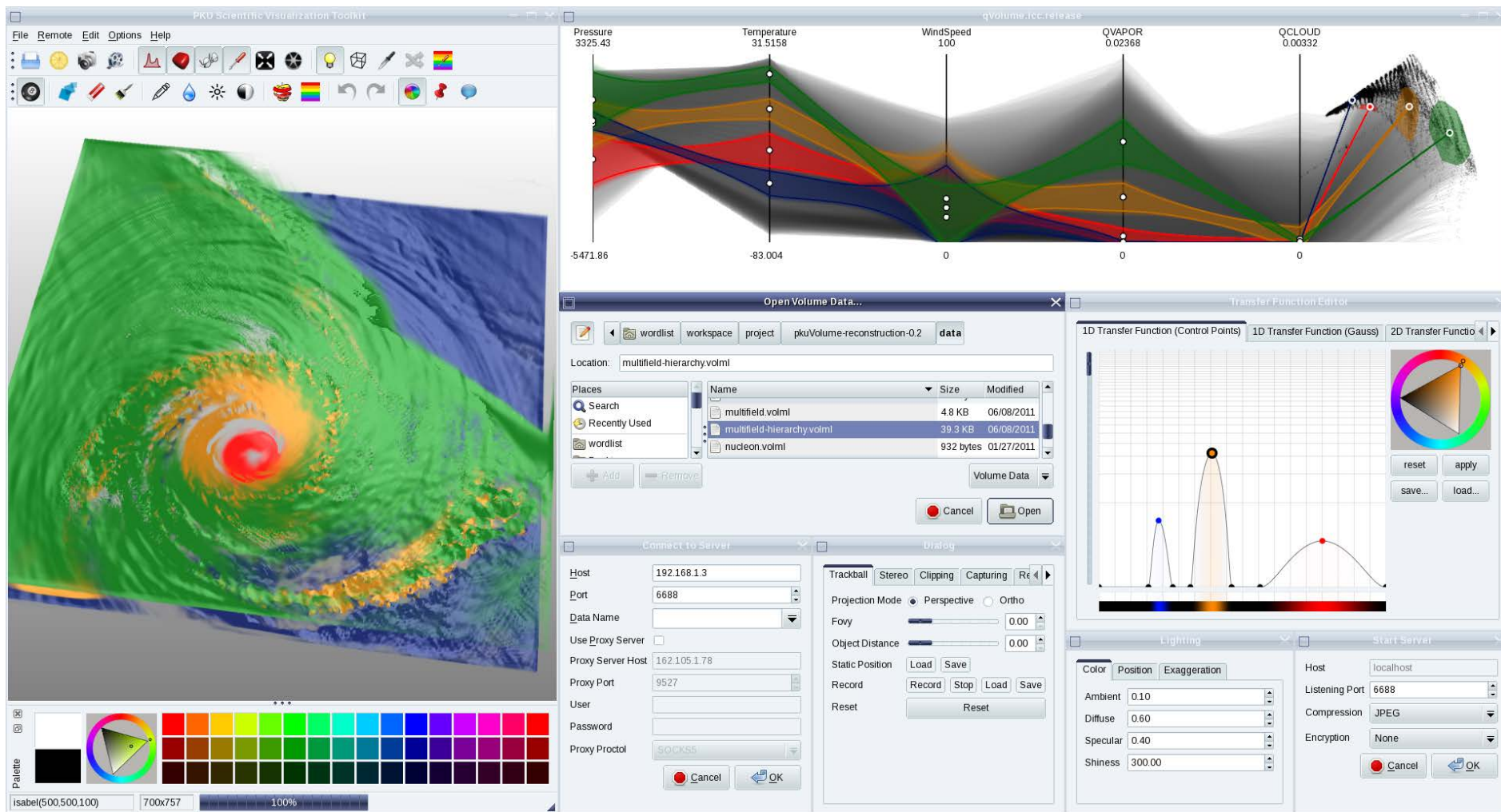


WRF Simulation



WRF Simulation

2-Run WRF Simulation Data



Challenges in VIS and VA

- Scalability in Data/Task complexity
 - Science imposes more computational challenges methods for visualization and visual analysis on large scientific data
- Scalability in User
 - Collaborative Visualization and Analysis on large data
 - Can scientist create novel visualization without programming

Double Gulf



**Visualization
User**

Evaluation

**Conceptual
Model**

Execution

Visualization

**Visualization
Designer**

Representation



Data

Manipulation

Double Gulf



**Visualization
User**

Evaluation

**Conceptual
Model**

Execution

Visualization

**Visualization
Designer**

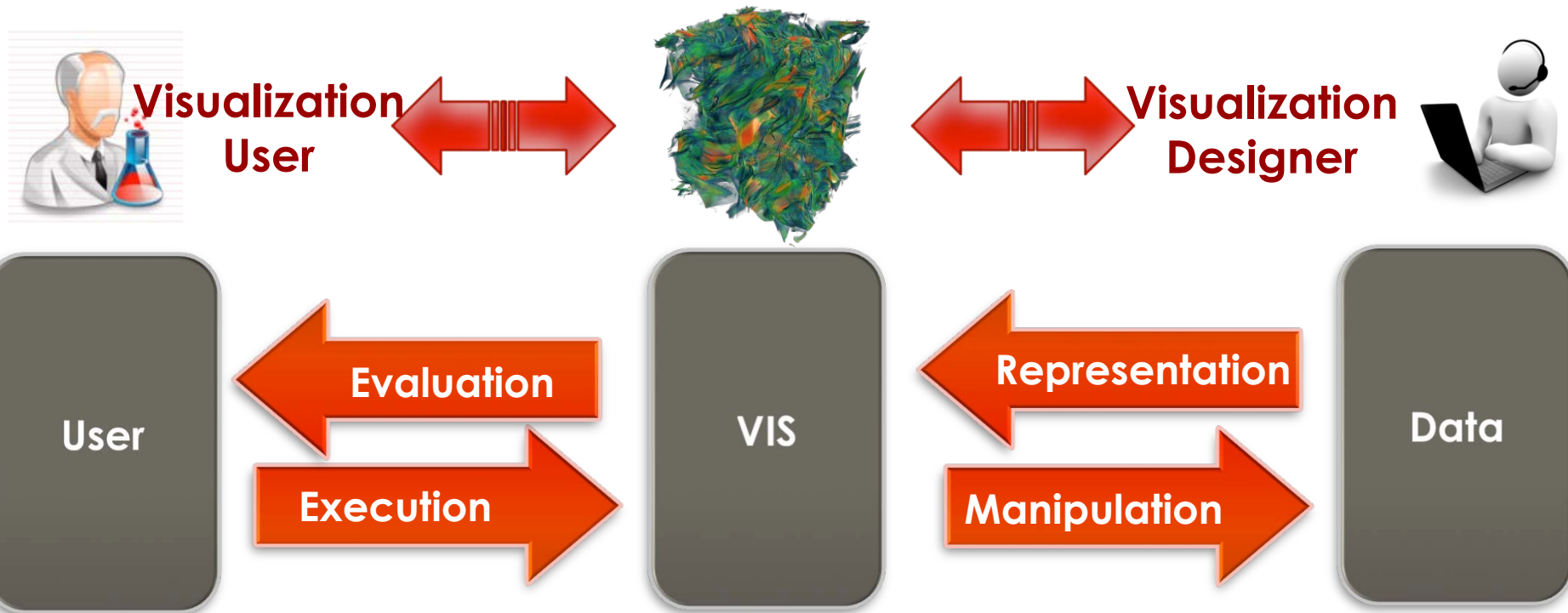
Representation



Data

Manipulation

From Data to User



Scalability In Users



**Visualization
User**



**Conceptual
Model**



Visualization

**Visualization
Designer**



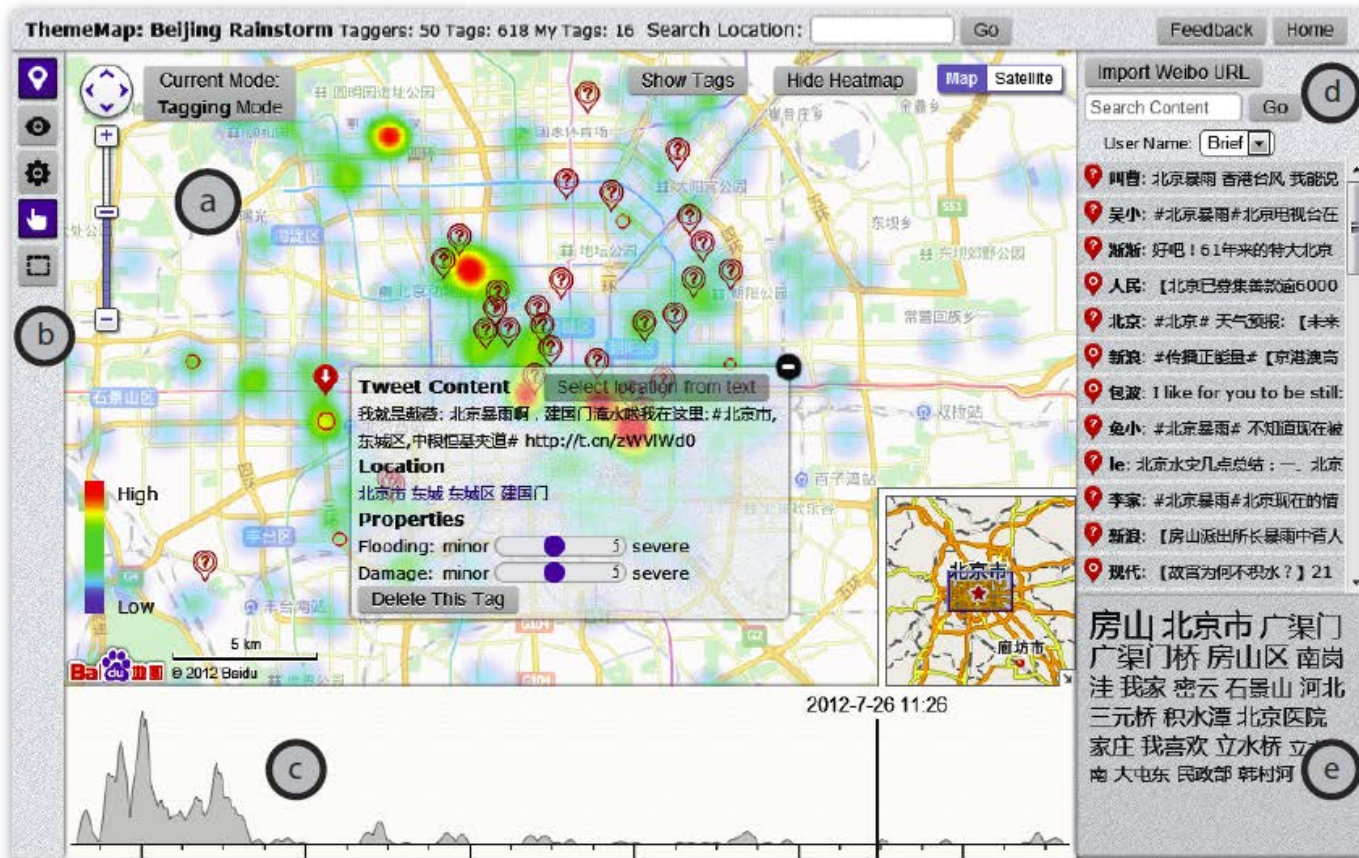
Data



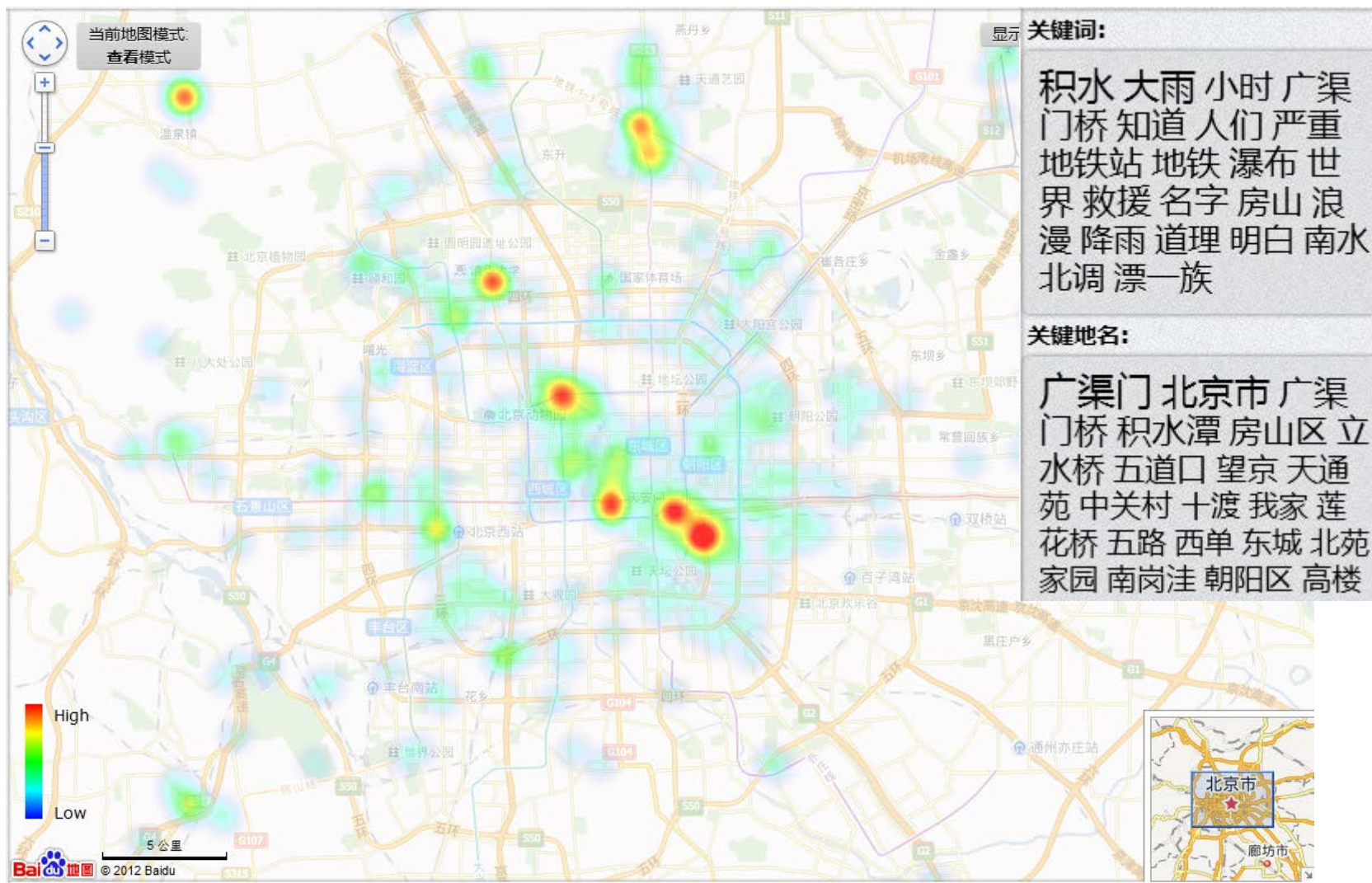
Scalability In Users – Collaborative Visualization

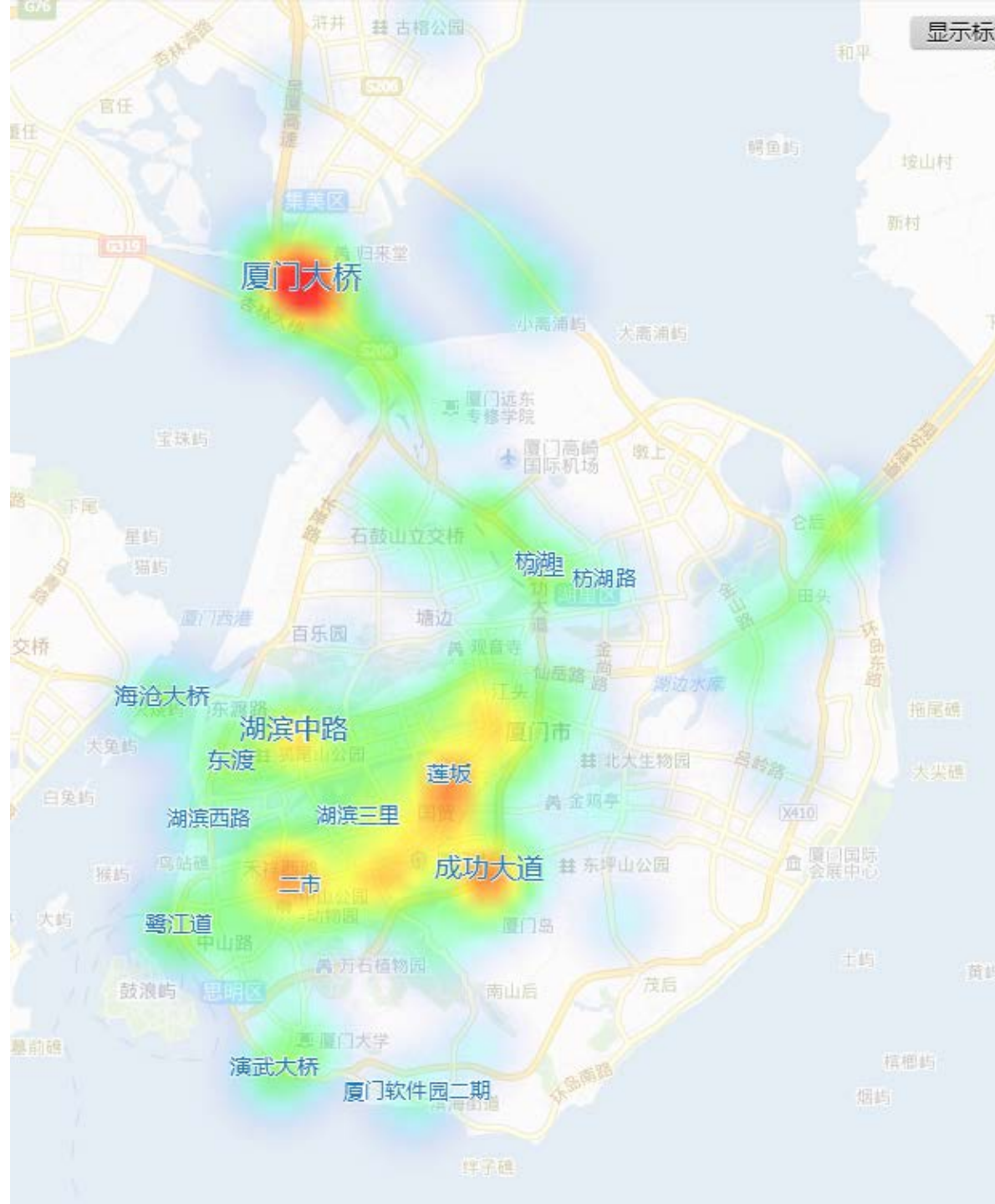


ThemeMap – Crowd Sourcing



ThemeMap – Crowd Sourcing





关键词:

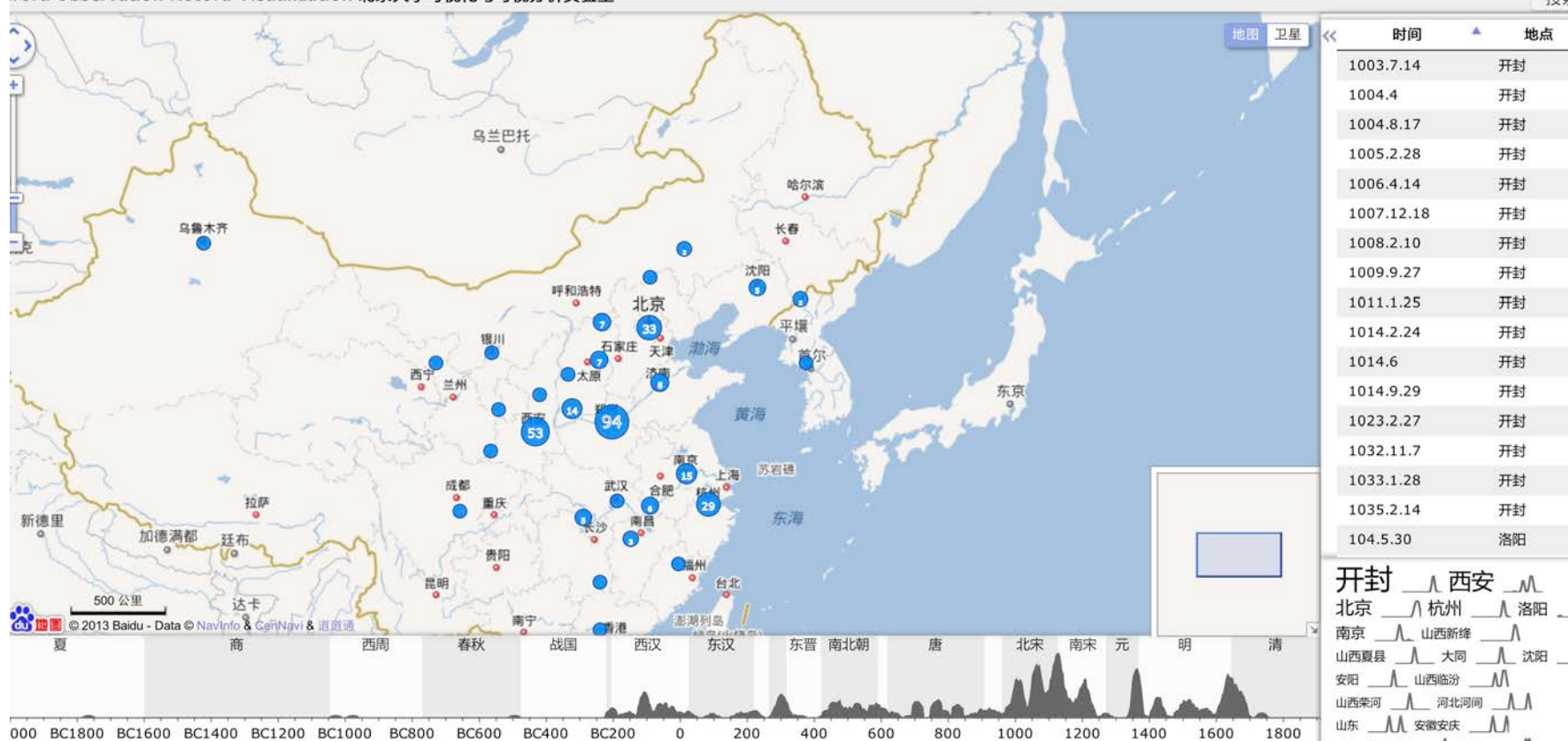
大道 严重 隧道 红绿灯
 仙岳 路口 火车站 司机 路
 段 海沧 时间 上班 知道 轮
 渡 东渡 方向 中路 开车 公
 园 车道

关键地名:

厦门大桥 成功大道 湖滨
 中路 桥上 地下停车场 东渡
 海沧大桥 演武大桥 莲坂 鹭江
 道 集美 南湖公园 湖滨三里 湖里
 厦门软件园二期 二楼 二市 枋湖
 湖滨西路 枋湖路

Aurora Observation Vis

Aurora Observation Record Visualization 北京大学可视化与可视分析实验室



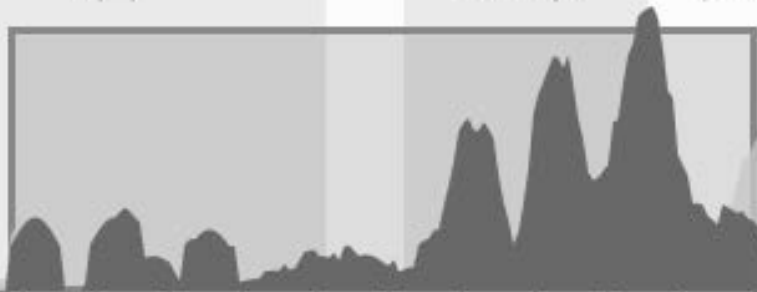
Aurora Observation Vis



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北宋

南宋



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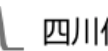
文献: 宋真宗景德四年十一月己巳, 中天有赤气如扫, 长七尺, 在舆鬼南。

开封  西安 

杭州  北京 

内蒙巴林右旗  湖北江陵 

陕西潼关  湖北西北部 

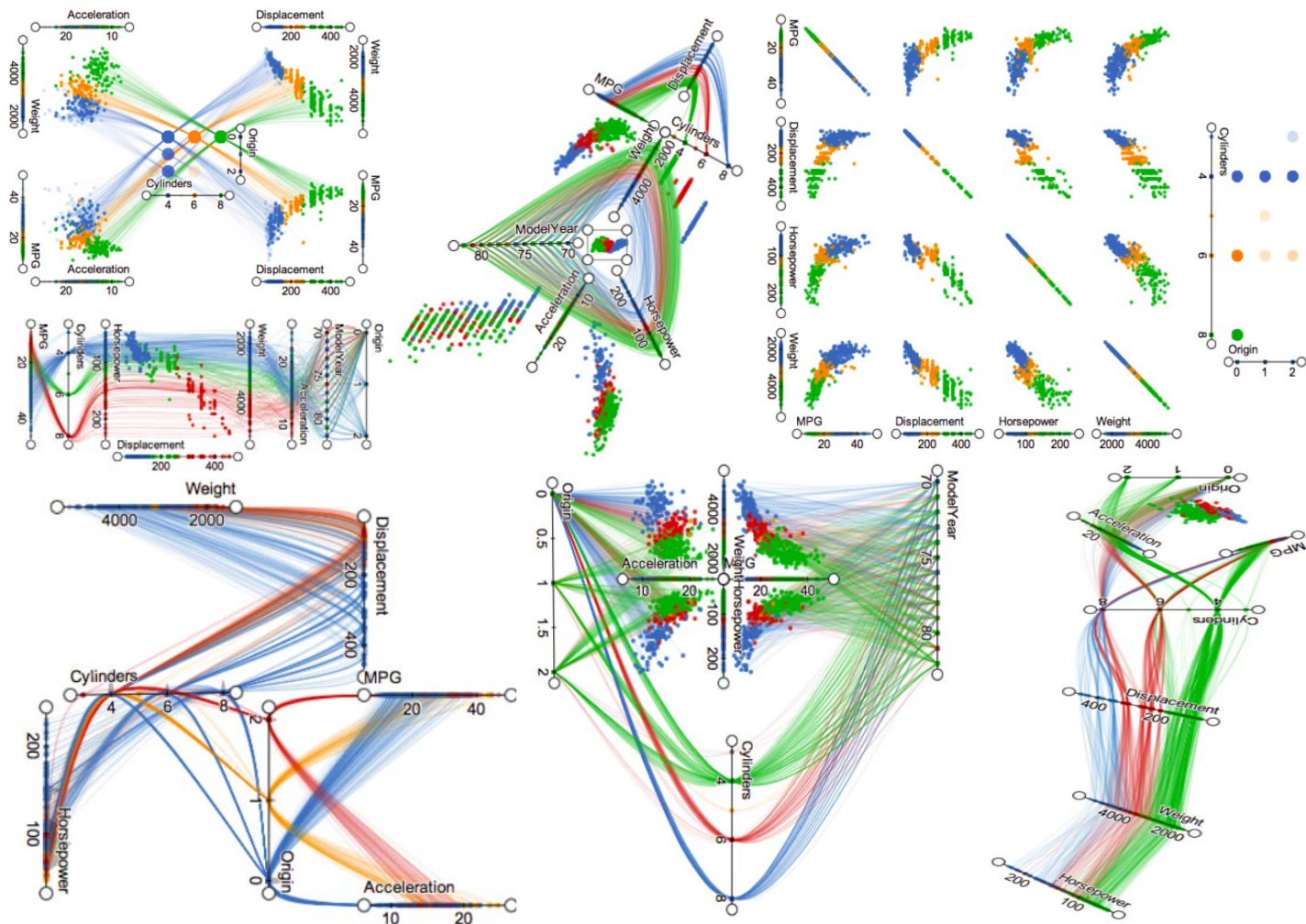
洛阳  四川仁寿井研 

嘉兴  洞庭湖 

北京空气质量可视化 mapview periodview



Scalability In Users – User - Visualization Expert

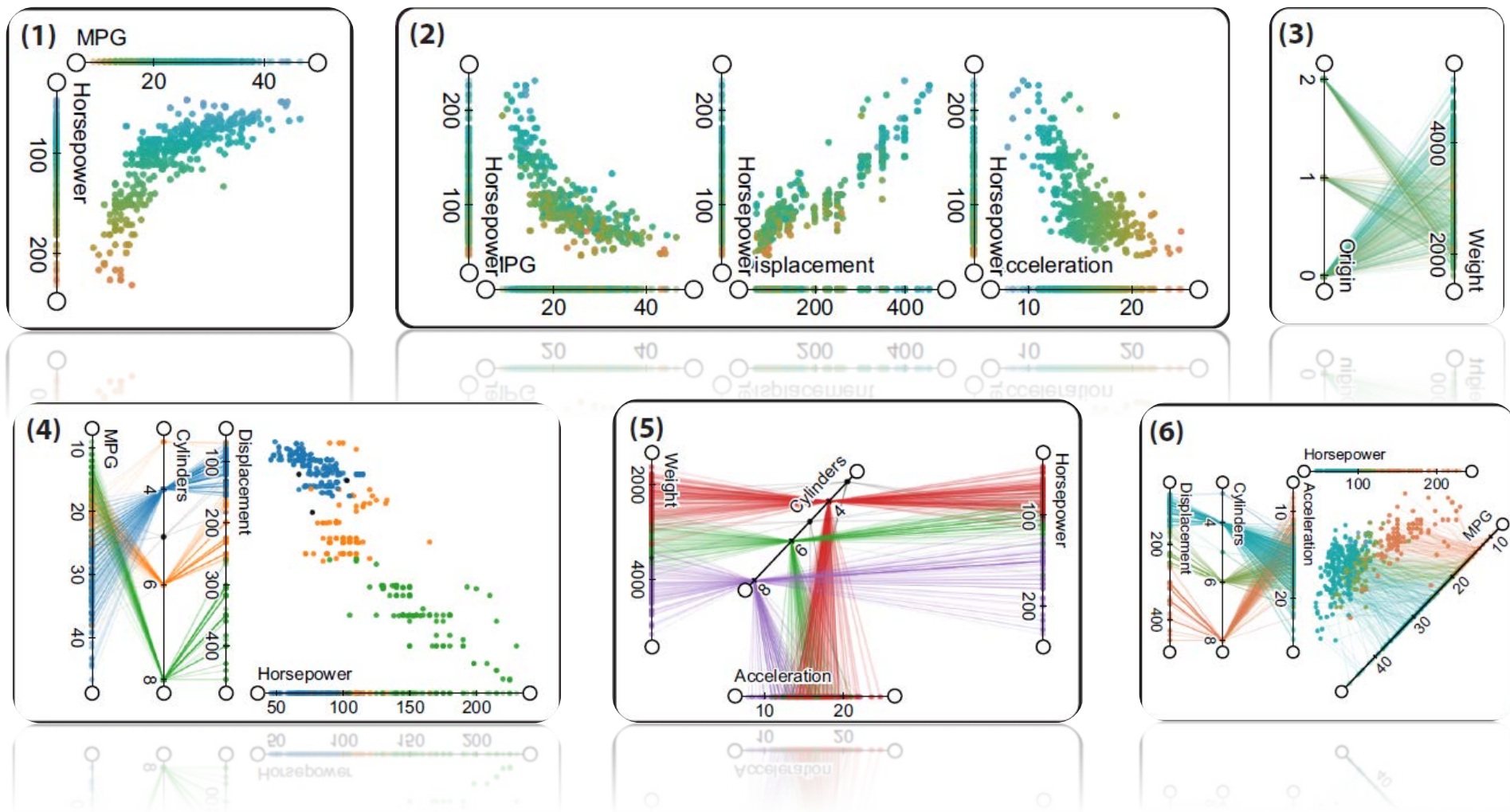


Visualization Assembly Line

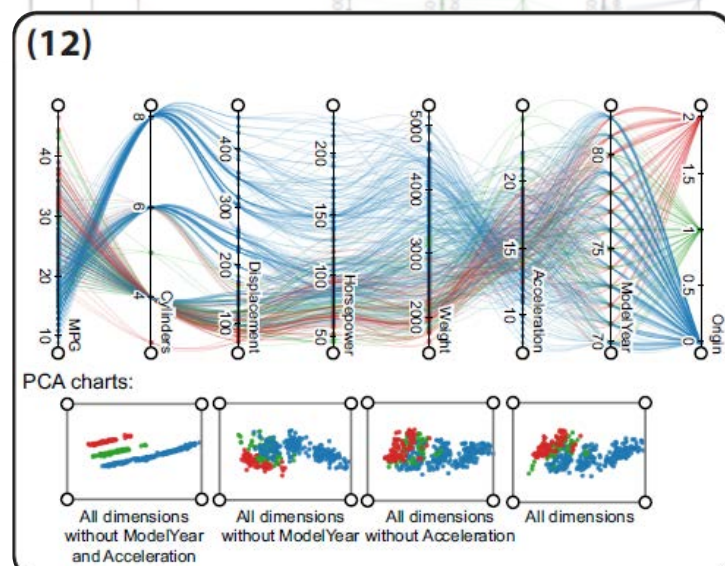
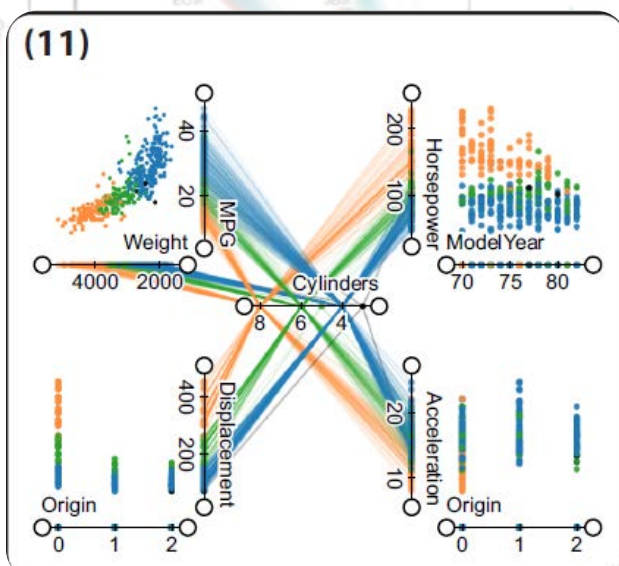
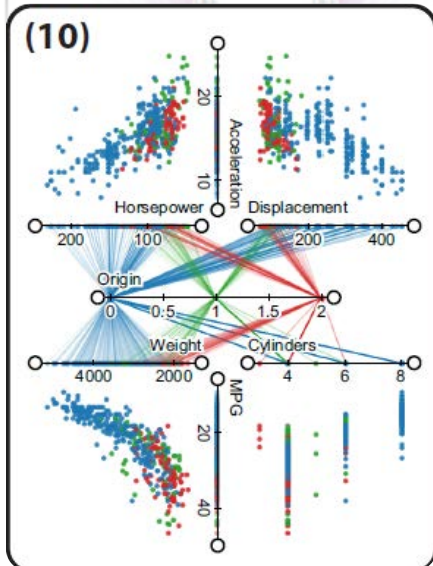
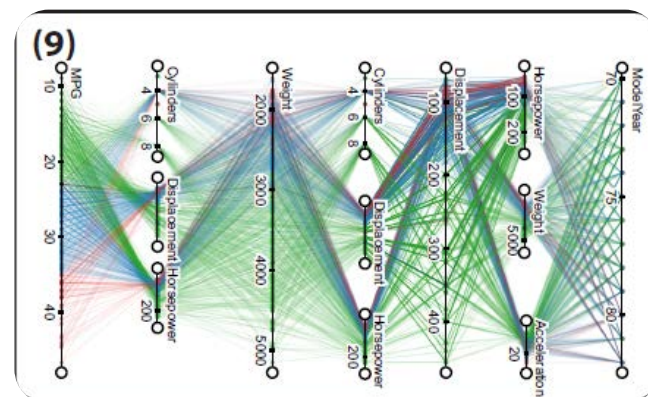
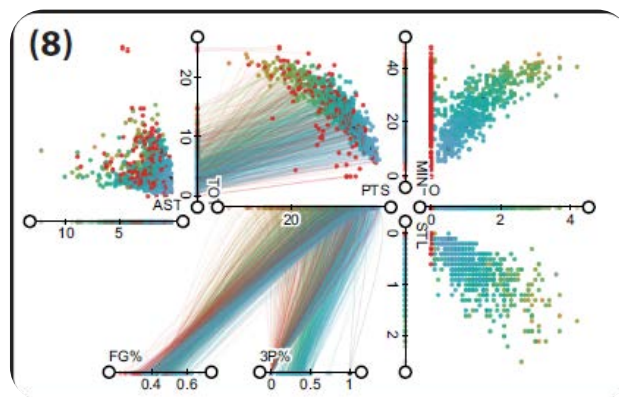
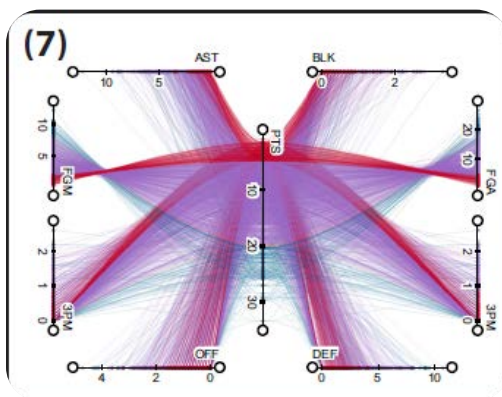


<http://vis.pku.edu.cn/mddv/val/>

Visualization Assembly Line



Visualization Assembly Line



Challenges in VIS and VA

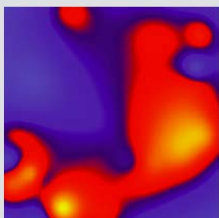
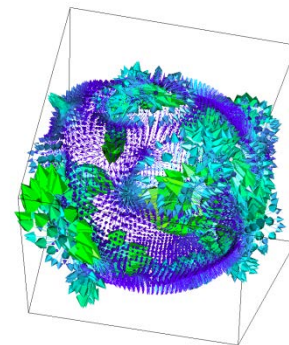
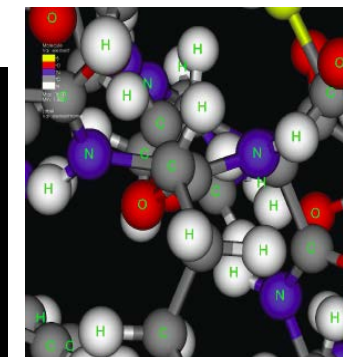
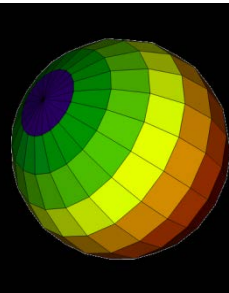
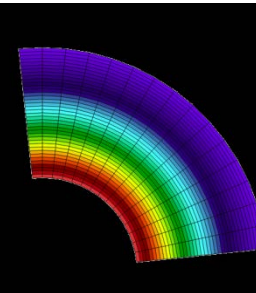
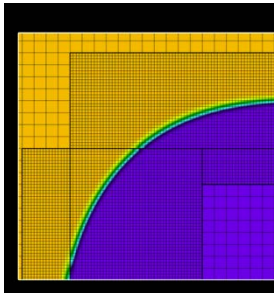
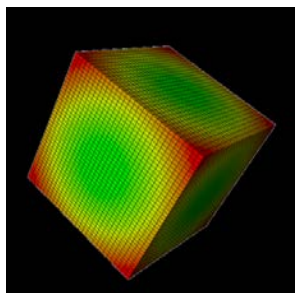
- Scalability in Data/Task complexity
 - Science imposes more computational challenges methods for visualization and visual analysis on large scientific data
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 - Collaborative Visualization and Analysis on large data
 - Can scientist create novel visualization without programming
- System Development
 - *Domain and Development Libraries, Frameworks, and Tools*
 - *Social, Community, and Government Engagements*

SCIVIS Visualization Systems

- VisIt - LLNL
<https://wci.llnl.gov/codes/visit>
- ParaView- Kitware/SNL/LANL
<http://www.paraview.org>
- IceT (Image Composition Engine for Tiles) - Sandia
<http://icet.sandia.gov>
- Daxtoolkit - Data Analysis at Extreme
<http://www.daxtoolkit.org>
- PISTON - Portable Data-Parallel Visualization and Analysis Library - LANL
<http://viz.lanl.gov/projects/PISTON.html>

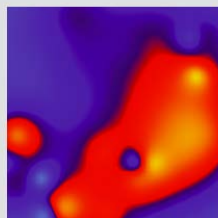
VisIt

- Production end-user tool supporting scientific and engineering applications.
- Parallel post-processing that scales from desktops to massive HPC

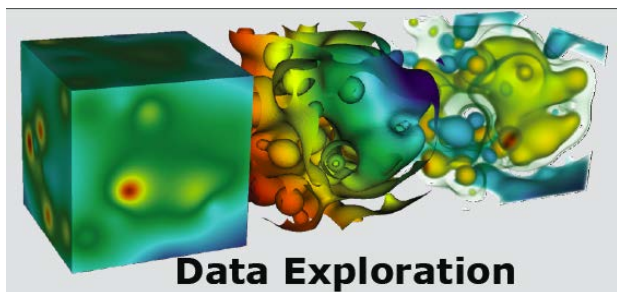


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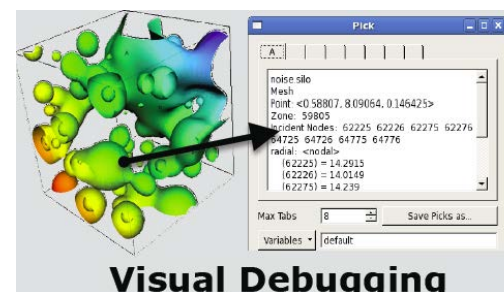
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Comparative Analysis



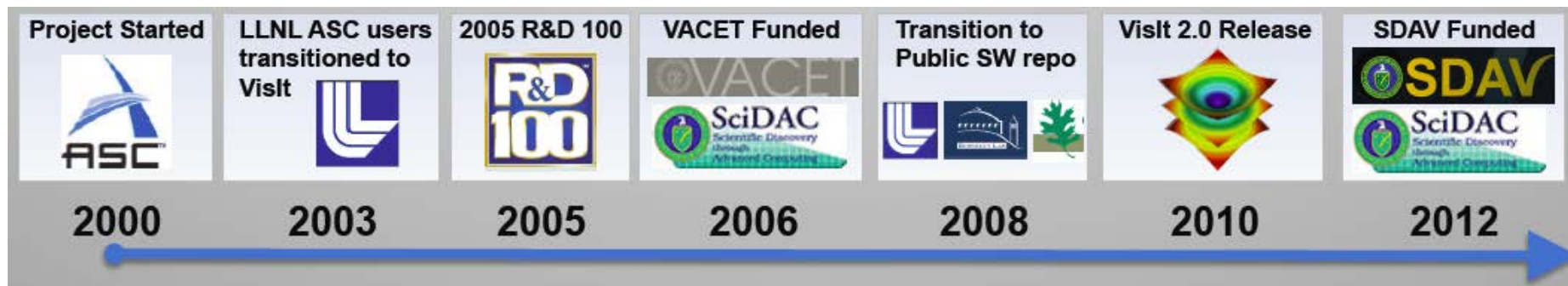
Data Exploration



Visual Debugging

Development of VisIt

- The VisIt project started in 2000 to support LLNL's large scale ASC physics codes.
- Supported by multiple organizations: LLNL, LBNL, ORNL, UC Davis, Univ. of Utah, ...
- Over 75 person years effort.
- 1.5+ million lines of code.



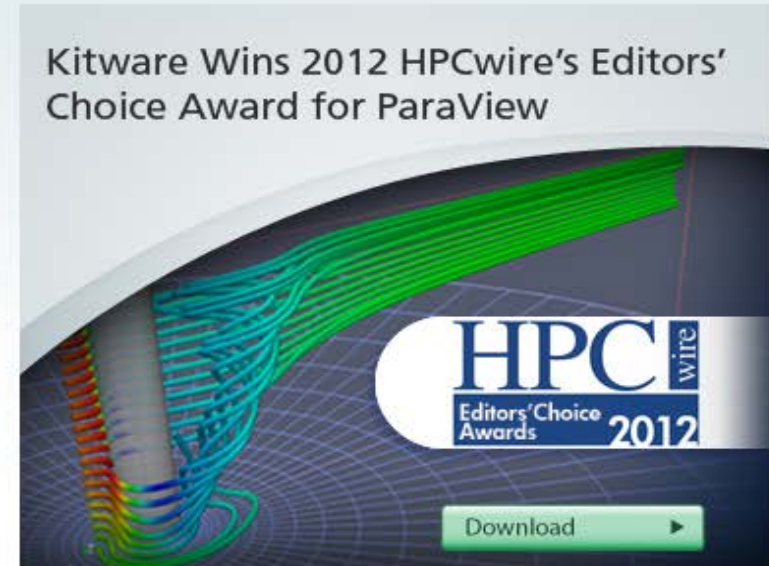
ParaView is an open-source, multi-platform data analysis and visualization application. ParaView users can quickly build visualizations to analyze their data using qualitative and quantitative techniques. The data exploration can be done interactively in 3D or programmatically using ParaView's batch processing capabilities.

ParaView was developed to analyze extremely large datasets using distributed memory computing resources. It can be run on supercomputers to analyze datasets of terascale as well as on laptops for smaller data.

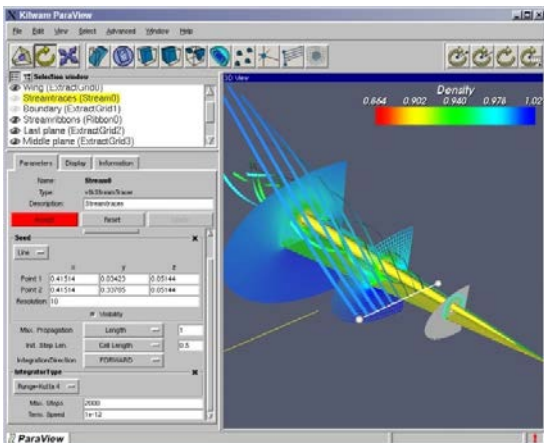
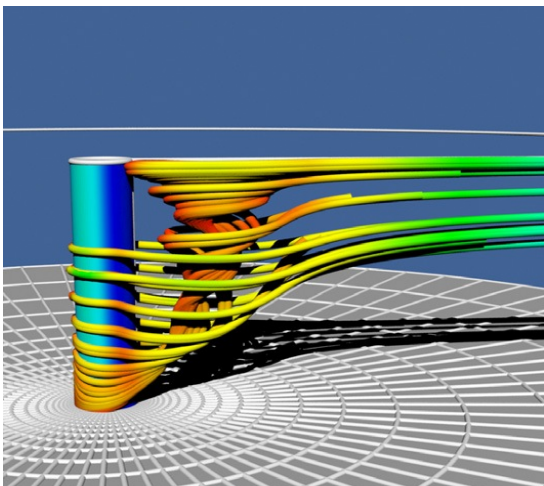
News

[More News >](#)

- 05.07.2013** Kitware to Enhance the Visualization Toolkit (VTK) to Accelerate ...
- 03.06.2013** KiwiViewer 2.0 Released and Available for Download in iTunes
- 02.12.2013** ParaView 3.98.1 Now Available
- 12.03.2012** ParaView 3.98.0 Now Available
- 11.14.2012** Kitware Receives Honors in 2012 HPCwire Readers' and Editors' Cho...



VTK



1993年, 3名来自GE Corporate R&D 中心的可视化研究员开始开发一个开源的可视化系统 Visualization Toolkit (VTK)。

开始被创作的意图是用于教学和研究的（因此它是开源的）。这个软件很快的被接受，不光因为它的面向对象设计和软件过程的理念，还因为它得到使用者们的改善。

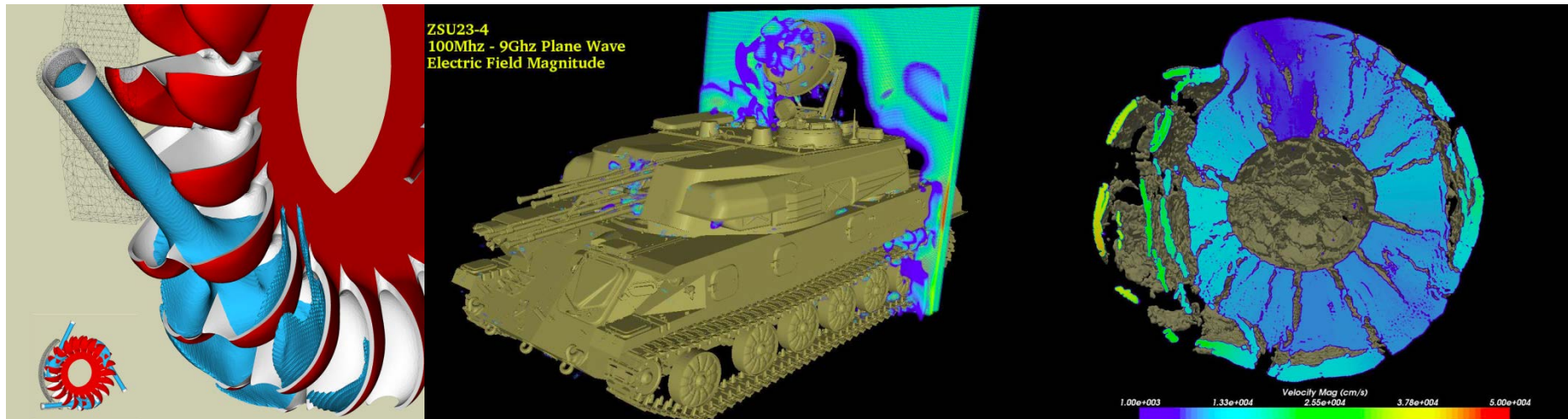
VTK现在被广泛的使用，许多公司和相关产品由于使用它而诞生，比如Kitware公司于1998年建立后编写了基于VTK的一款开源可视化系统 ParaView和一款商业的体绘制应用程序VolView。VTK持续被学术人员，美国国家实验室和商家完善，并且用于许多商业软件中。

W.J. Schroeder, K. Martin, and W. Lorensen, *The Visualization Toolkit: An Object Oriented Approach to Computer Graphics, Third Edition*, Kitware, Inc., ISBN-1-930934-12-2 (2004).

S. E. Rogers, D. Kwak, and U. K. Kaul, A numerical study of three-dimensional incompressible flow around multiple post. In *Proceedings of AIAA Aerospace Sciences Conference*. AIAA Paper 86-0353. Reno, Nevada, 1986.

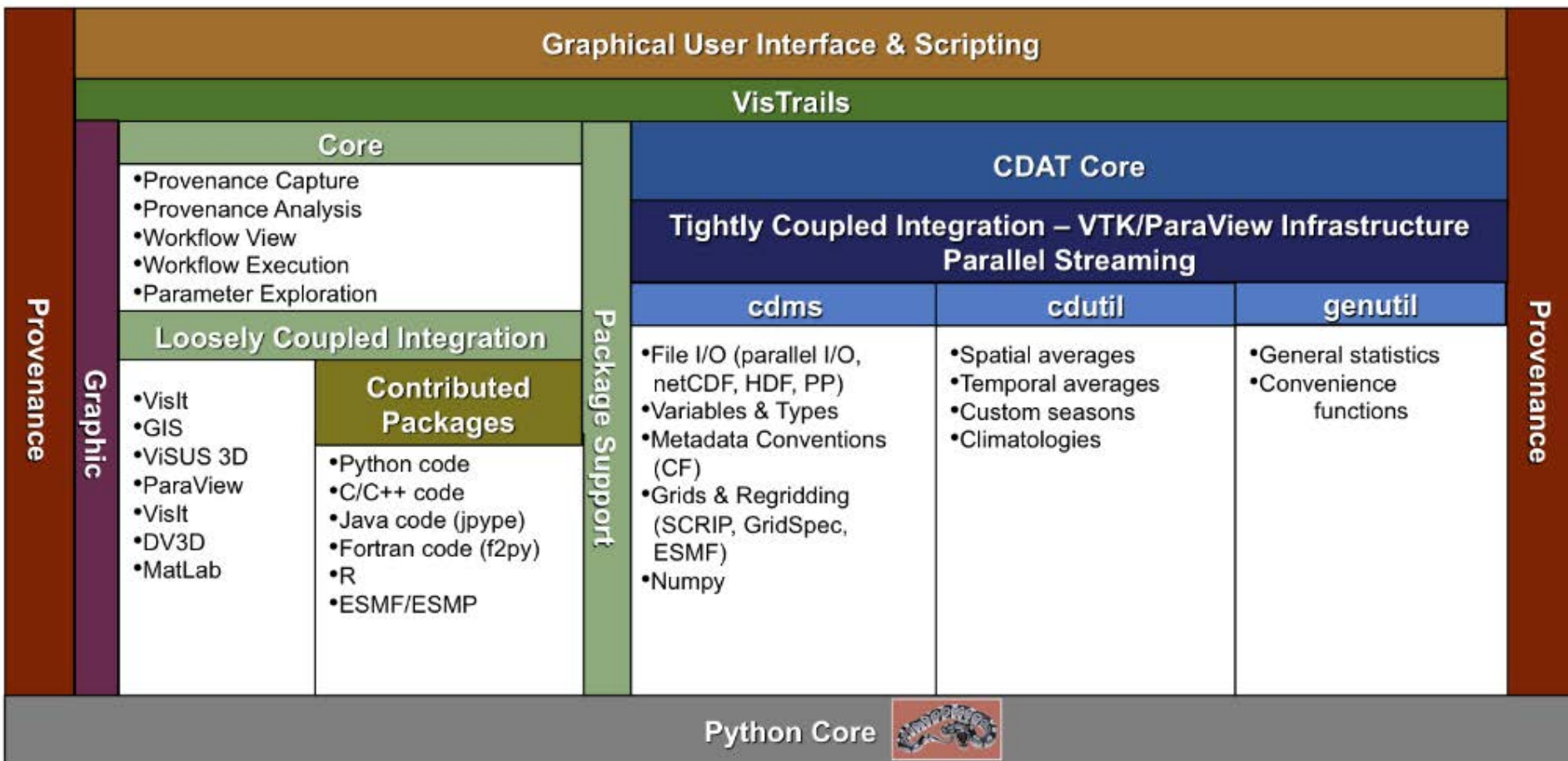
ParaView

- 2000 Los Alamos National Laboratories and Kitware Inc.
- 2005 Sandia National Laboratories and Kitware Inc.
- Used by academic, government, and commercial institutions worldwide.
- Downloaded ~100K times per year.



UV-CDAT Project

Ultra-scale Visualization Climate Data Analysis Tools (UV-CDAT) Architectural Layers



Build a successful vis system

- System Design
- Domain User – Visualization Scientist “Co-design”
- Stable Development Team
- Funding Mechanism

Social, Community, and Government Engagements

■ Universities

- University of Tennessee in Knoxville
- Ohio State University
- SCI Institute, University of Utah
- University of California, Davis
- University of California, San Diego
- University of Nebraska-Lincoln
- Michigan Technological University
- Drexel University

■ Supercomputer centers

- San Diego Supercomputer Center (SDSC)
- Texas Advanced Computing Center (TACC)
- National Center for Supercomputing Applications at the University of Illinois (NCSA)

■ DoE Labs

- Argonne National Laboratory (ANL)
- Lawrence Berkeley National Laboratory (LBNL)
- Lawrence Livermore National Laboratory (LLNL)
- Los Alamos National Laboratory (LANL)
- Pacific Northwest National Laboratory (PNNL)
- Oak Ridge National Laboratory (ORNL)
- Sandia National Laboratories (SNL)
- National Renewable Energy Laboratory (NREL)

■ Companies

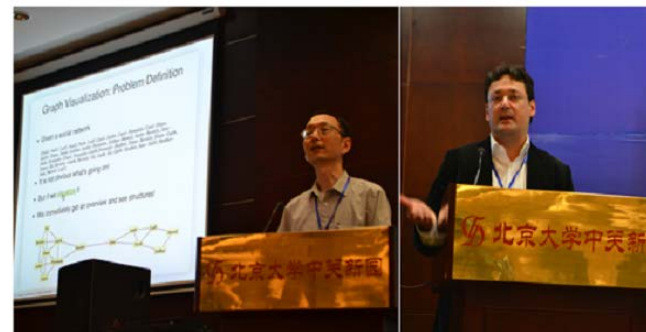
- Kitware

Vis Workshop 2013 @ PKU

■ 2013.7.12-13



会场



Top 10 Challenges in Extreme-Scale Data Visual Analytics



Visualization Viewpoints

Editor:
Theresa-Marie Rhynes

The Top 10 Challenges in Extreme-Scale Visual Analytics

Pak Chung Wong
Pacific Northwest National Laboratory

Han-Wei Shen
Ohio State University

Christopher R. Johnson
University of Utah

Chaomei Chen
Drexel University

Robert B. Ross
Argonne National Laboratory

In this issue of CG&A, researchers share their R&D findings and results on applying visual analytics (VA) to extreme-scale data. Having surveyed these articles and other R&D in this field, we've identified what we consider the top challenges of extreme-scale VA. To cater to the magazine's diverse readership, our discussion evaluates challenges in all areas of the field, including algorithms, hardware, software, engineering, and social issues.

Background

The September/October 2004 issue of CG&A introduced the term visual analytics to the computer science literature.¹ In 2005, an international team of multidisciplinary panelists consensually and collectively defined the then newly established area as "the science of analytical reasoning facilitated by interactive visual interfaces."² The means and targets of VA have since evolved and expanded significantly, covering both scientific and nonscientific data of different types, shapes, sizes, domains, and applications. As extreme-scale datasets began revolutionizing our daily working life,³ researchers looked to VA for solutions to their big-data problems.

Today's extreme-scale VA applications often combine high-performance computers for computation, high-performance database appliances and/or

cloud servers for data storage and management, and desktop computers for human-computer interaction. Sources of extreme-scale data often come from models or observations, arising from different scientific, engineering, social, and cyber applications. Although many petabyte (10^{15}) or even terabyte (10^{12}) data analytics problems remain unsolved, scientists have begun analyzing exabyte (10^{18}) data.

The Top 10 Challenges

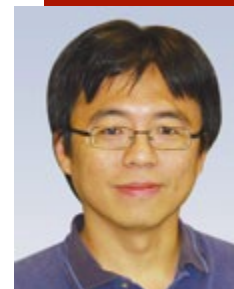
Addressing the top challenges has profound, far-reaching implications, for not only fulfilling the critical science and technical needs but also facilitating the transfer of solutions to a wider community. We thus evaluate the problems from both technical and social perspectives. The order of the following challenges doesn't reflect their relative importance but rather the content correlation among individual challenges.

1. In Situ Analysis

The traditional postmortem approach of storing data on disk and then analyzing the data later might not be feasible beyond petascale in the near future. Instead, in situ VA tries to perform as much analysis as possible while the data are still in memory. This approach can greatly reduce I/O



Pak Chung Wong (PNNL)



Han-Wei Shen (OSU)



Chris Johnson (Utah)



Chaomei Chen (Drexel)



Robert Ross (Argonne)

Top 10 Challenges in Extreme-Scale Data Visual Analytics

■ *In Situ Analysis*

- Perform as much analysis as possible while the data are still in memory

■ *Interaction and User Interfaces*

- Machine-based automated systems vs. Human Cognition

■ *Large-Data Visualization*

- Data projection and dimension Reduction, display technology

■ *Databases and Storage*

- A cloud-based solution might not meet the needs

■ *Algorithms*

- Address both data-size and visual-efficiency issues

Top 10 Challenges in Extreme-Scale Data Visual Analytics

- ***Data Movement/Transport, & Network Infrastructure***
 - Efficiently use networking resources and provide convenient abstractions
- ***Uncertainty Quantification***
 - Cope with incomplete data
- ***Parallelism***
- ***Domain and Development Libraries, Frameworks, and Tools***
 - Affordable resource libraries, frameworks, and tools
- ***Social, Community, and Government Engagements***

Acknowledgement

- Students
- Funds
 - NSFC
 - 863
 - PKU
 - Beijing NSF
- Collaborators
 - Jian Huang University of Tennessee
 - Zhu Xiaoming, SDSCC
 - Yongxian Zhang, China Earthquake Network Center
 - Xiaoguang Ma, CAS IAP
 - More ...



<http://vis.pku.edu.cn/wiki>
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