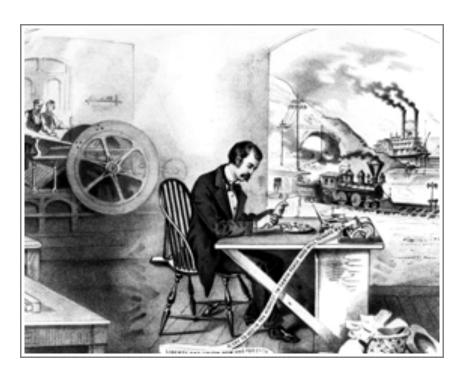


# Data and Computing in the Astronomy Community

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Information technology revolution is historically unprecedented - in its impact it is like the industrial revolution and the invention of printing combined

Science and scholarship are slowly adopting the new tools and technologies and there are great scientific and leadership opportunities in this arena

We are effectively developing a new methodology of science and scholarship for the 21st century



Exponential
Growth of
Data
Volumes
on Moore's law time scales

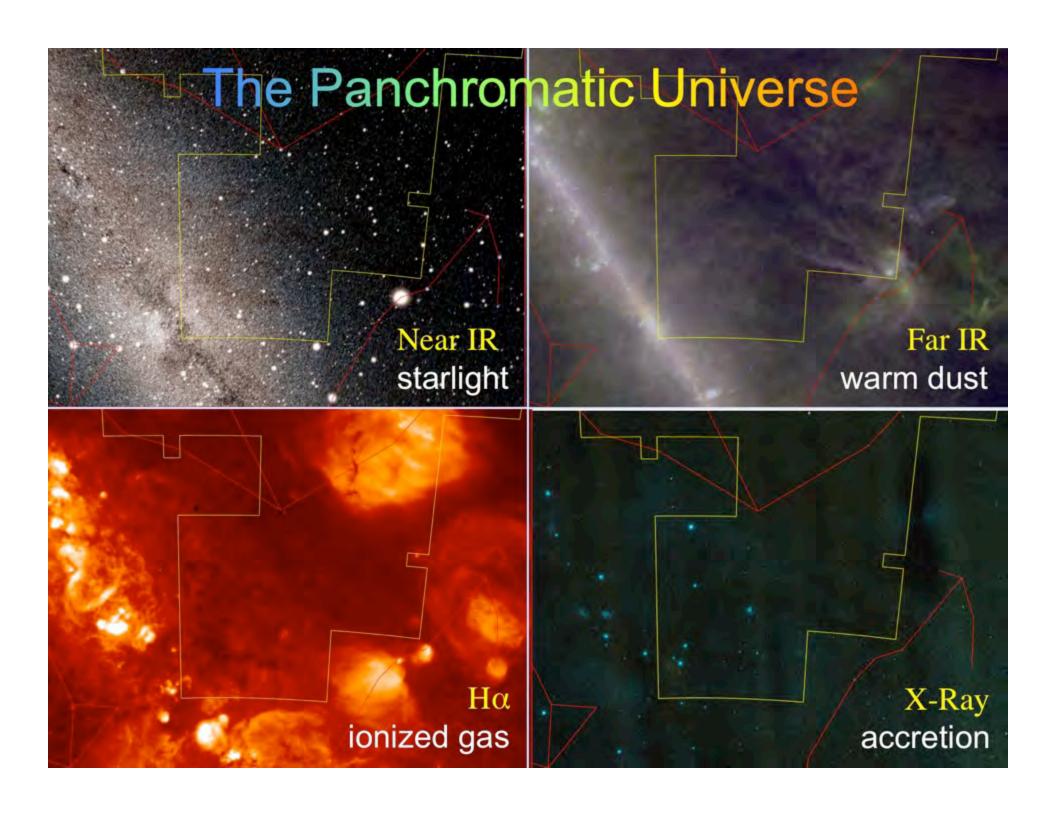
Complexity

From data poverty to data glut complex phenomena
From data sets to data streams requires complex data!
Theory expressed as data
From static to dynamic, evolving data
From anytime to real-time analysis and discovery
From centralized to distributed resources
From ownership of data to ownership of expertise

Understanding of

There Are Lots Of Stars In The Sky...

Modern sky surveys obtain  $\sim 10^{12} - 10^{15}$  bytes of images, catalog  $\sim 10^8 - 10^9$  objects (stars, galaxies, etc.), and measure  $\sim 10^2 - 10^3$  numbers for each



#### **Numerical Simulations:**

A qualitatively different and necessary way of doing theory, beyond the analytical approach

#### Theory

is expressed as data, an output of a numerical simulation, not as a set of equations

... and then must be matched against complex measurements

#### Astronomy Has Become Very Data-Rich

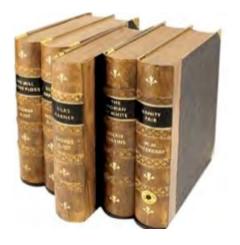
- Typical digital sky surveys generate ~ 10 100 TB each, plus a comparable amount of derived data products
  - PB-scale data sets are imminent
- Astronomy today has ~ a few PB of archived data, and generates ~ 10 TB/day
  - Both data volumes and data rates grow exponentially, with a doubling time ~ 1.5 years
  - Even more important is the growth of data complexity
- For comparison:

Human Genome < 1 GB

Human Memory < 1 GB (?)

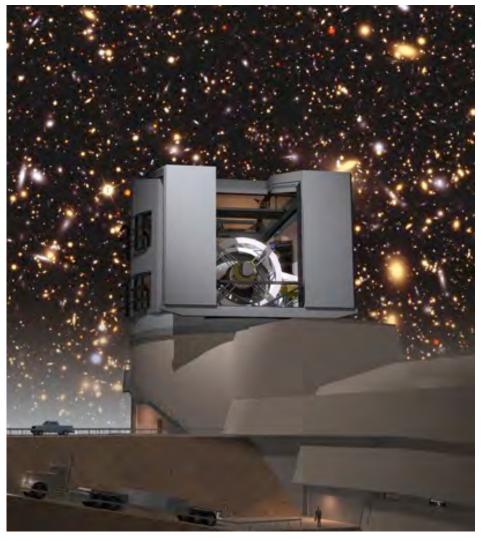
1 TB ~ 2 million books

Human Bandwidth ~ 1 TB / year (±)

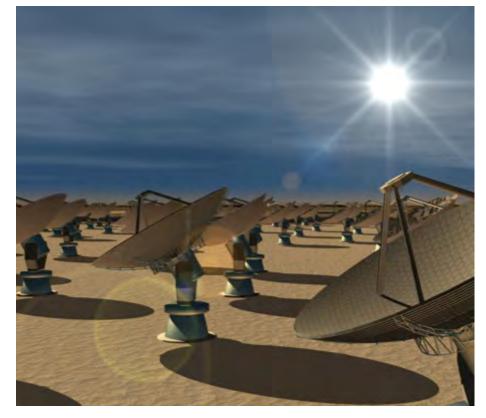


#### ... And It Will Get Much More So

Large Synoptic Survey Telescope (LSST) ~ 30 TB / night



Square Kilometer Array (SKA)
~ 1 EB / second (raw data)
(EB = 1,000,000 TB)



#### The Era of Cosmic Cinematography

- Synoptic digital sky surveys are now becoming the dominant data producers in astronomy
  - From Terascale to Petascale data streams
- A major new growth area of astrophysics
  - Driven by the new generation of large digital synoptic sky surveys (CRTS, PTF, PS1, ... Fermi), leading to LSST, SKA, etc.
- All the challenges of traditional sky surveys, plus the time dimension and time-critical analysis requirements
- A broader significance for an automated, real-time knowledge discovery in massive data streams











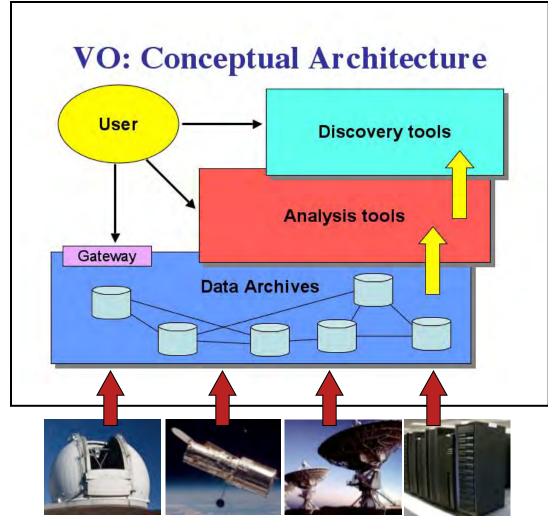
# The Rise of Virtual Scientific Organizations Compute Resources Analysis Tools

- A grassroots response of scientific communities to the challenges and opportunities brought by the data glut
- Domain-specific, not institution-based; inherently distributed
  - The human, data, and compute resources are distributed
  - A new type of a scientific organization, needing new management models
- Should VO's have a finite lifetime, as they fulfill their role?

#### The Virtual Observatory Concept

 A complete, dynamical, distributed, open research environment for the new astronomy with massive and complex data sets

- Provide and federate
   content (data, metadata)
   services, standards, and
   analysis/compute services
- Develop and provide data exploration and discovery tools
- Harness the IT revolution in the service of astronomy
- A part of the broader e-Science /Cyber-Infrastructure



#### Scientific Roles and Benefits of a VO

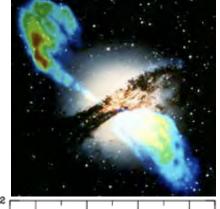
- Facilitate science with massive data sets (observations and theory/simulations) efficiency amplifier
- Provide an added value from federated data sets (e.g., multi-wavelength, multi-scale, multi-epoch ...)
  - Discover the knowledge which is present in the data,
     but can be uncovered *only* through data fusion
- Enable and stimulate some qualitatively new science with massive data sets (not just old-but-bigger)
- Optimize the use of expensive resources (e.g., space missions, large ground-based telescopes, computing ...)
- Provide R&D drivers, application testbeds, and stimulus to the partnering disciplines (CS/IT, statistics ...)

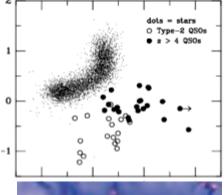
#### Virtual Observatory Science Examples

 Combine the data from multi-TB, billion-object surveys in the optical, IR, radio, X-ray, etc., for:

- Precision large scale structure in the universe
- Precision structure of our Galaxy
- Discover rare and unusual (one-in-a-million or one-in-a-billion) types of sources
  - E.g., extremely distant or unusual quasars, brown dwarfs, new types, etc.
- Probe the evolution of quasars, galaxies, or clusters discovered using different techniques over the cosmic time
- Match Peta-scale numerical simulations of star or galaxy formation with equally large and complex observations

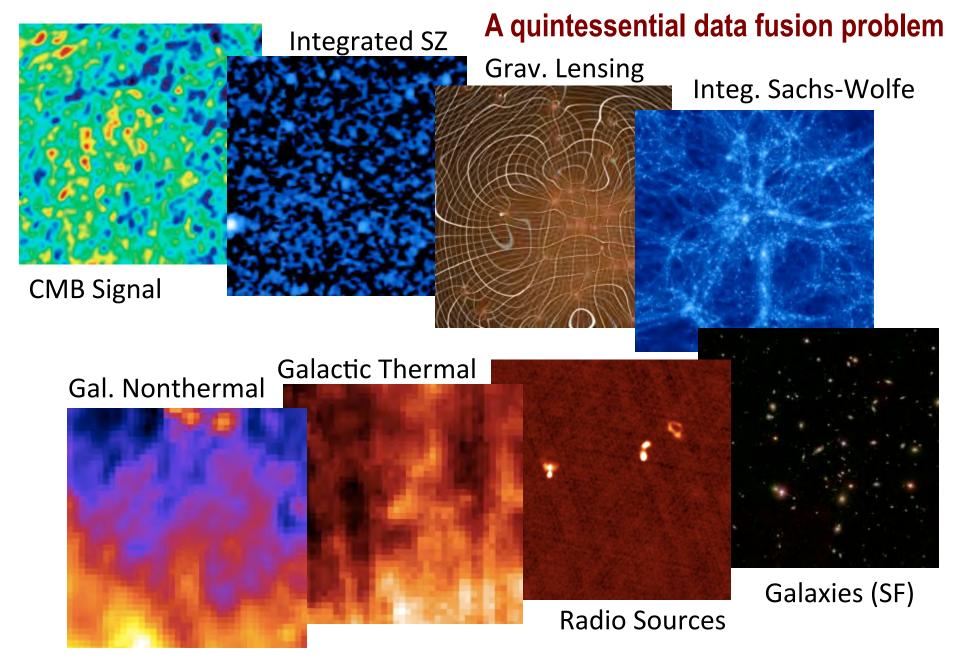
... etc., etc.







#### Understanding the CMBR Foregrounds



## Virtual Observatory

http://www.euro-vo.org



http://usvao.org

Is Real!

Discover, retrieve, and analyze astronomical data from archives and data centers around the world.



Hungarian Virtual Observatory

#### A Modern Scientific Discovery Process

Pata Gathering (e.g., from sensor networks, telescopes...)



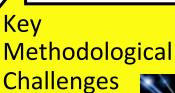
Storage/Archiving Indexing, Searchability Data Fusion, Interoperability

Database Technologies



Data Mining (or Knowledge Discovery in Databases):

Key Technical Challenges Pattern or correlation search Clustering analysis, classification Outlier / anomaly searches Hyperdimensional visualization



**Data Understanding** 

**New Knowledge** 



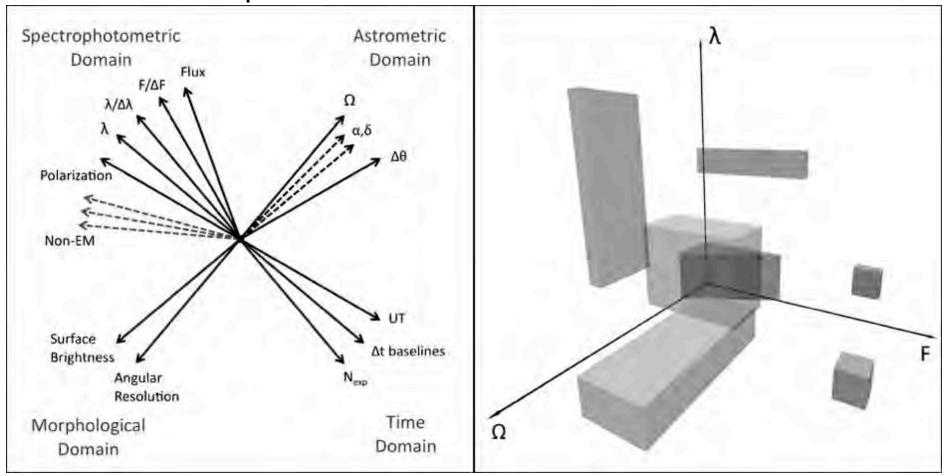


## Systematic Exploration of the Observable

Parameter Space (OPS)

Its axes are defined by the observable quantities

Every observation, surveys included, carves out a hypervolume in the OPS



Technology opens new domains of the OPS New discoveries

#### Astronomy in the Time Domain

- Rich phenomenology, from the Solar system to cosmology and extreme relativistic physics
  - Touches essentially every field of astronomy
- For some phenomena, time domain information is a key to the physical understanding
- A qualitative change:

Static ⇒ Dynamic sky
Sources ⇒ Events

 Real-time discovery/reaction requirements pose new challenges for knowledge discovery





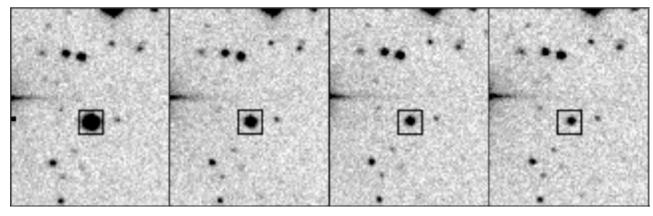




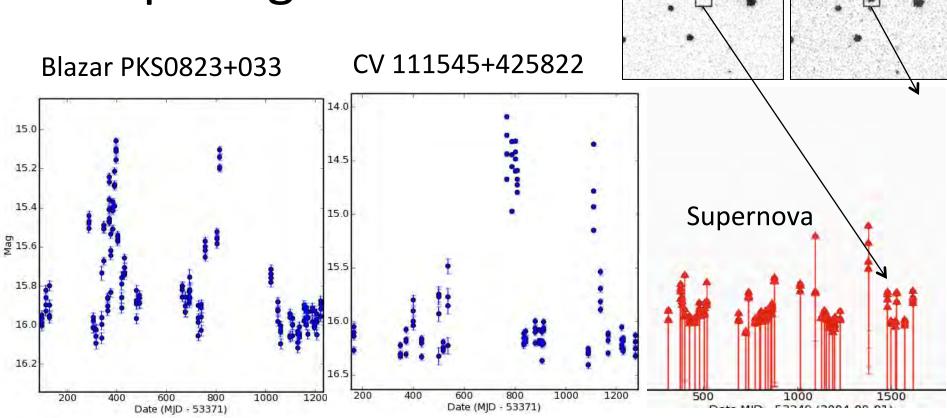
Synoptic, panoramic surveys  $\rightarrow$  event discovery Rapid follow-up and multi- $\lambda \rightarrow$  keys to understanding

# The Catalina Real-Time Transient Survey (CRTS)

- Collaboration with UAz/LPL search for NEA/PHA asteroids
- 3 small telescopes up to 2,500 deg²/night with 4 exposures/ pointing, limiting mags ~ 19 – 21, several tens of passes per year, total area coverage ~ 33,000 deg², time baselines from 10 min to years, ~ 7+ years coverage
- Real time processing and event discovery and publication
- Open data policy: all data are made public immediately
- ~ 6,500 unique transients so far, a number of discoveries made



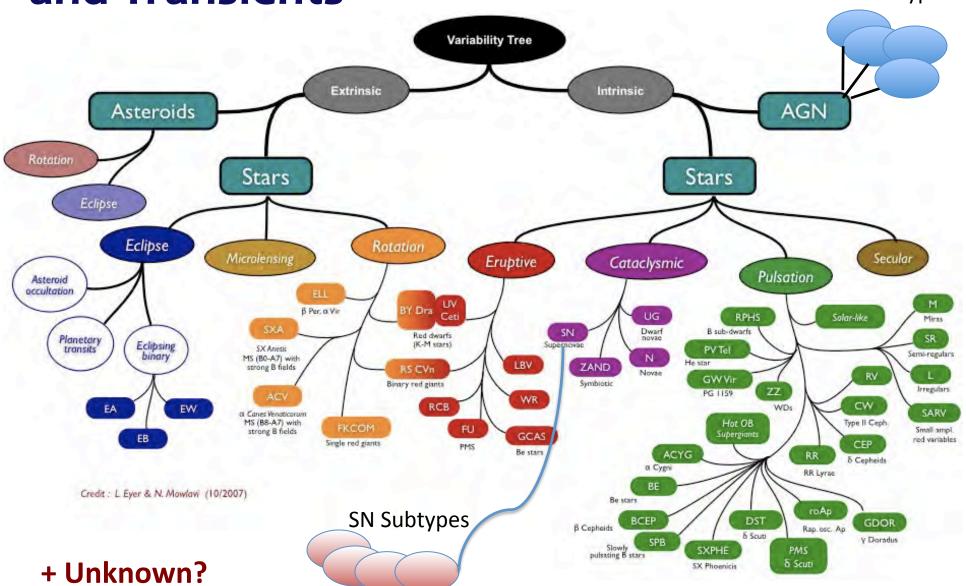
#### Sample Light Curves



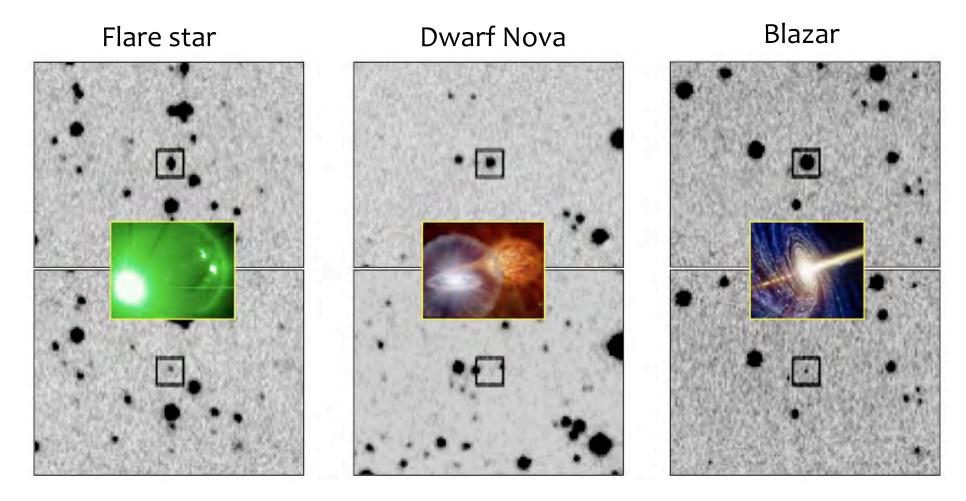
- Large-amplitude transients published immediately, light curves accumulated for every source (~ 500 million)
- Transients are perishable must be followed rapidly in order to get the science, but the follow-up is very limited

Semantic Tree of Astronomical Variables and Transients

AGN Subtypes



#### **Automated Classification of Transients**



Vastly different physical phenomena, yet they look the same! Which ones are the most interesting and worthy of follow-up?



Rapid, automated transient classification is a critical need!

#### This is a Critical Problem

(and it will get a lot worse)



- Now: data streams of ~ 0.1 TB / night, ~ 10<sup>2</sup> transients / night (CRTS, PTF, various SN surveys, microlensing, etc.)
  - ♦ We are already in the regime where we cannot follow them all
  - ♦ Spectroscopy is the key bottleneck now, and it will get worse
- Forthcoming on a time scale ~ 1 5 years:
   ~ 1 TB / night, ~ 10<sup>3</sup> 10<sup>4</sup> transients / night
   (PanSTARRS, Skymapper, VISTA, VST, SKA precursors...)
- Forthcoming in ~ 8 10 (?) years: LSST, ~ 30
   TB / night, ~ 10<sup>5</sup> 10<sup>7</sup> transients / night, SKA
- So... which ones will you follow up?
- Follow-up resources will likely remain limited



Transient classification is essential

#### **Event Classification is a Hard Problem**

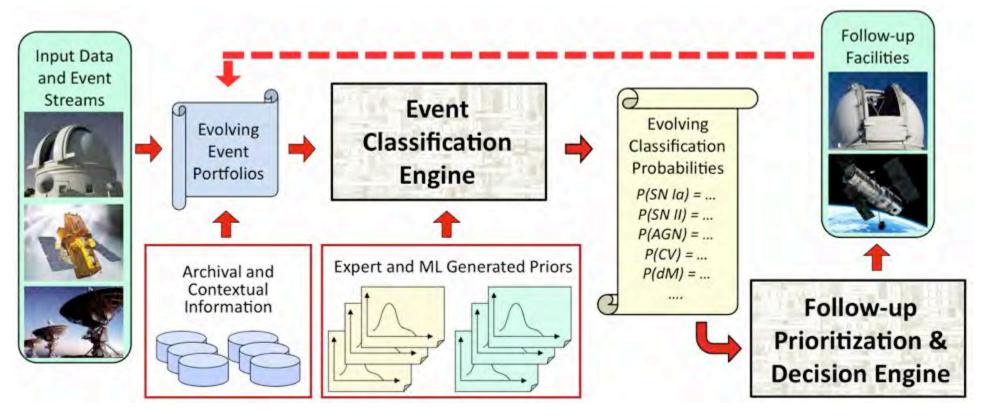
- Classification of transient events is essential for their astrophysical interpretation and uses
  - Must be done in real time and iterated dynamically
- Human classification is already unsustainable, and will not scale to the Petascale data streams
- This is hard:
  - Data are sparse and heterogeneous: feature vector approaches do not work; using Bayesian approach
  - Completeness vs. contamination •
  - Follow-up resources are expensive and/or limited: only the most interesting events
  - Iterate classifications dynamically as new data come in
- Traditional DP pipelines do not capture a lot of the relevant contextual information, prior/expert knowledge, etc.







#### Towards an Automated Event Classification



- Incorporation of the contextual information (archival, and from the data themselves) is essential
- Automated prioritization of follow-up observations, given the available resources and their cost
- A dynamical, iterative system

#### The Key Challenge: Data Complexity

#### Or: The Curse of Hyper-Dimensionality

#### 1. Data mining algorithms scale very poorly:

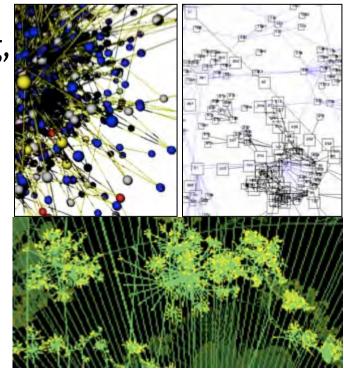
N = data vectors,  $\sim 10^8 - 10^9$ , D = dimension,  $\sim 10^2 - 10^3$ 

- Clustering  $\sim$  N log N  $\rightarrow$  N<sup>2</sup>,  $\sim$  D<sup>2</sup>
- Correlations  $\sim$  N log N  $\rightarrow$  N<sup>2</sup>,  $\sim$  D<sup>k</sup> (k  $\geq$  1)
- Likelihood, Bayesian  $\sim N^m (m \ge 3), \sim D^k (k \ge 1)$



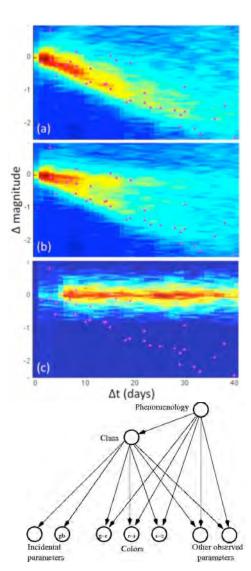
#### 2. Visualization in >> 3 dimensions

- The complexity of data sets and interesting, meaningful constructs in them is exceeding the cognitive capacity of the human brain
- We are biologically limited to perceiving D ~ 3 - 10(?) dimensions
- Visualization must be a component of the data mining / exploration process
- It is the bridge between the quantitative content of data and human understanding



#### Look to new techniques (for astronomy)

- Data are sparse and heterogeneous
- Light curve characterization/feature extraction
  - Thiel-San estimator
  - AR(1) time series
  - Bayesian blocks
  - Local regression
- Classification
  - Bayesian networks
  - Symbolic regression
  - Probabilistic structure functions
  - Knowledge-based (semantic)
  - Hierarchical approaches
  - Fusion modules



#### **VO Functionality Today**

#### What we did so far:

- Progress on interoperability, standards, etc.
- An incipient data grid of astronomy
- Some useful web services
- Community training, EPO

#### What we did not do (yet):

- Significant data exploration and mining tools
   That is where the science will come from!
   Thus, little VO-enabled science so far
   Thus, a slow community buy-in
- → Development of powerful knowledge discovery tools should be a key priority



#### Some Thoughts About e-Science

- Comput*ational* science ≠ Comput*er* science
- Data-driven science is not about data, it is about knowledge extraction (the data are incidental to our real mission)
- Information and data are (relatively) cheap, but the expertise is expensive
  - Just like the hardware/software situation
- Computer science as the "new mathematics"
  - It plays the role in relation to other sciences which mathematics did in ~ 17<sup>th</sup> - 20<sup>th</sup> century
- Computation: an interdisciplinary glue/lubricant
  - Many important problems (e.g., climate change) are inherently inter/multi-disciplinary







# The quantitative change in the information volume and complexity will enable the

#### Science of a Qualitatively Different Nature:

- Statistical astronomy done right
  - Precision cosmology, Galactic structure, stellar astrophysics ...
  - Discovery of significant patterns and multivariate correlations
  - Poissonian errors unimportant
- Systematic exploration of the observable parameter spaces
   (NB: Energy content ≠ Information content)
  - Searches for rare or unknown types of objects and phenomena
  - Low surface brightness universe, the time domain ...
- Confronting massive numerical simulations with massive data sets
  - + things we have not thought of yet ...

### **Beyond Virtual Scientific Organizations:**

The Rise of X-Informatics (X = Astro, Bio, Geo, ..)

- Domain-specific amalgam fields (science + CS + ICT)
- A mechanism for a broader community inclusion (both as contributors and as consumers)
- A mechanism for interdisciplinary e-Science methodological sharing

