

SciSpark: Interactive and Highly Scalable Climate Model Analytics

Chris Mattmann

*Chief Architect, Instrument and Science Data Systems Section, NASA JPL
Adjunct Associate Professor, USC
Director, Apache Software Foundation*

Agenda

- NASA + Big Data
- Motivation for SciSpark
- Apache Software Foundation
- Spark
- Prototype Effort
- Where we're going

And you are?

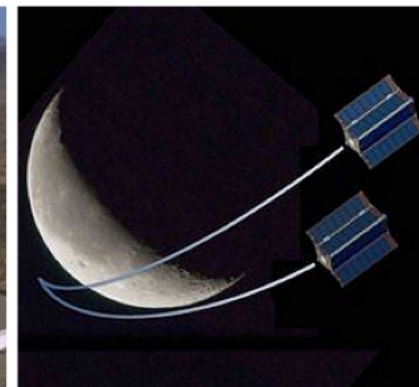
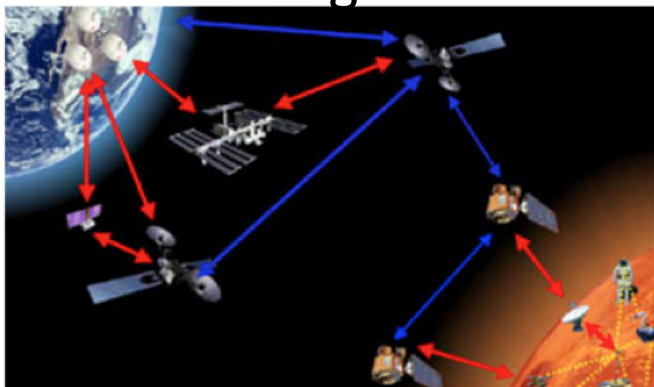
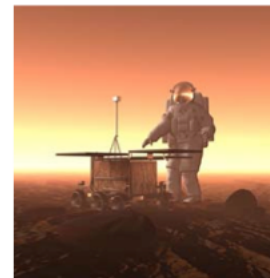
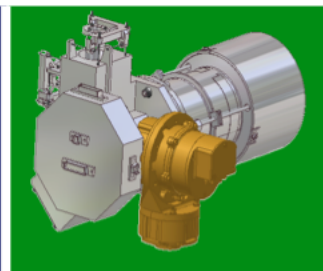
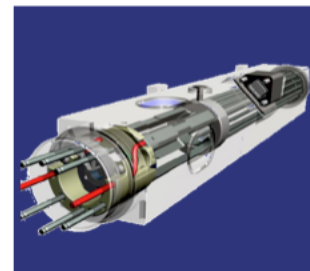
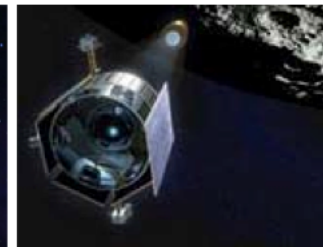
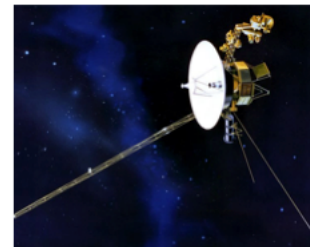


- Chief Architect, Instrument and Science Data Systems Section at NASA JPL in Pasadena, CA USA
- Software Architecture/ Engineering Prof at Univ. of Southern California
- Apache Board of Directors involved in
 - OODT (VP, PMC), Tika (PMC), Nutch (PMC), Incubator (PMC), SIS (PMC), Gora (PMC), Airavata (PMC)

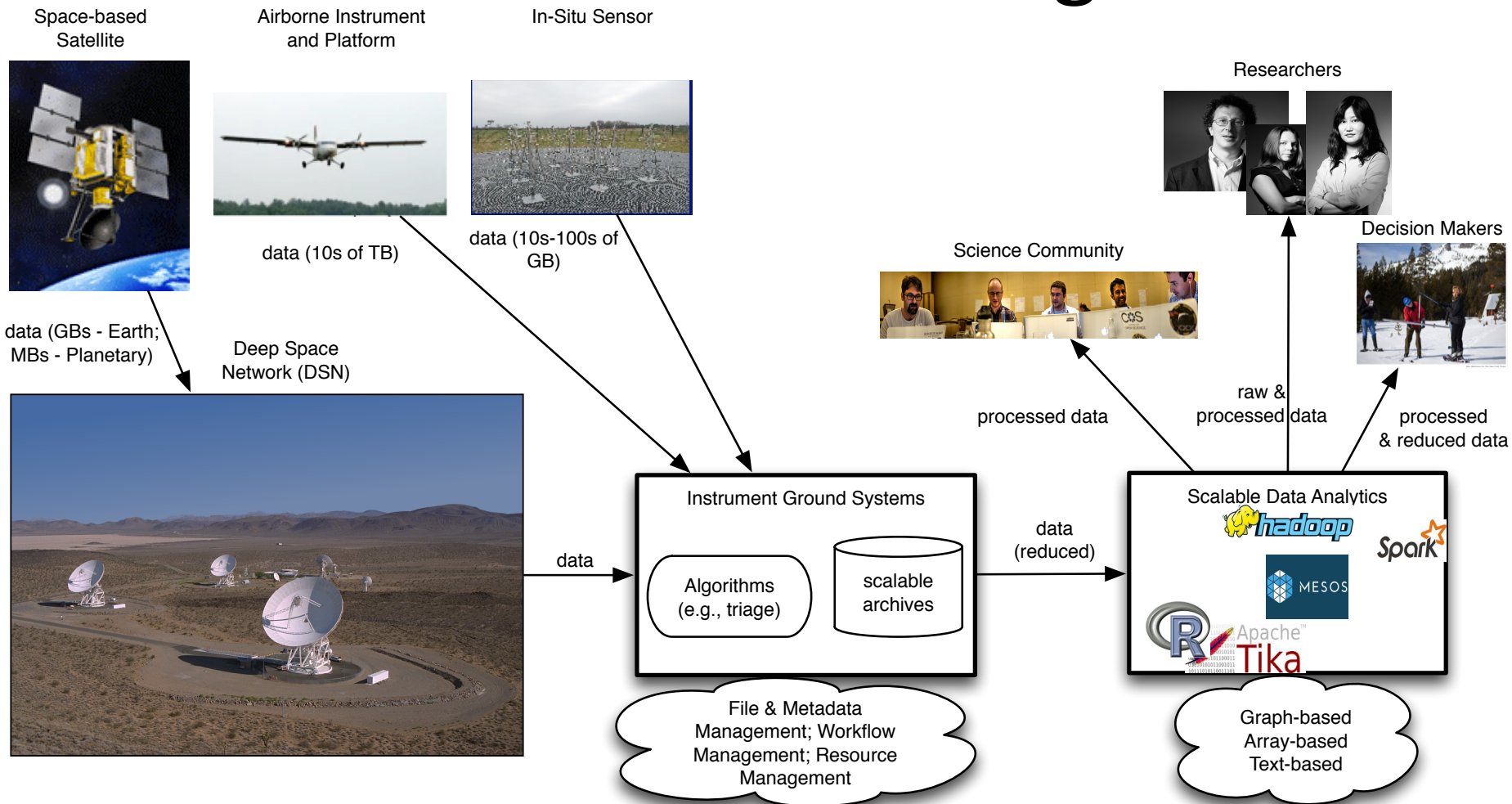
Instrument & Ground Data Systems

(Section 398)

- Largest Section on Lab
- 250+ people
- Data Science, Machine Learning, Visualization, Operations groups
- OCO-2, NPP Sounder PEATE, SMAP, MER, MSL, Mars 2020, Image Processing



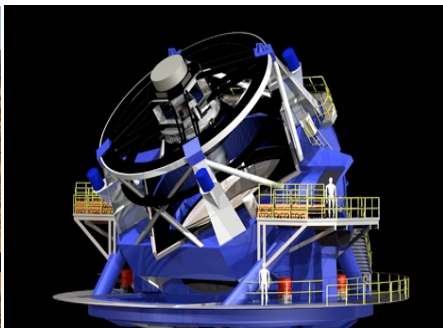
Instrument and Ground Systems: Earth Monitoring



Some “Big Data” Grand Challenges I’m interested in

- *How do we handle 700 TB/sec of data coming off the wire when we actually have to keep it around?*
 - Required by the Square Kilometre Array
- *Joe scientist says I’ve got an IDL or Matlab algorithm that I will not change and I need to run it on 10 years of data from the Colorado River Basin and store and disseminate the output products*
 - Required by the Western Snow Hydrology project
- *How do we compare petabytes of climate model output data in a variety of formats (HDF, NetCDF, Grib, etc.) with petabytes of remote sensing data to improve climate models for the next IPCC assessment?*
 - Required by the 5th IPCC assessment and the Earth System Grid and NASA
- *How do we catalog all of NASA’s current planetary science data?*
 - Required by the NASA Planetary Data System

Big Data Strategic Initiative



Future Opportunities: Mission and instrument competitions, data-intensive industries, LSST, future radio observatories.

JPL Concept: Big data technology for data triage, archiving, etc.

Key Challenges this work enables: Broaden JPL business base (relevant to 1X, 3X, 4X, 7X, 8X, 9X Directorates)

Initiative Long Term Objectives

- Apply lower-efficient digital architectures to future JPL flight instrument developments and proposals.
- Expand and promote JPL expertise with machine learning algorithm development for real-time triage.
- Utilize intelligent anomaly classification algorithms in other fields, including data-intensive industry.
- Build on JPL investments in large data archive systems to capture role in future science facilities.
- Enhance the efficiency and impact of JPL's data visualization and knowledge extraction programs.

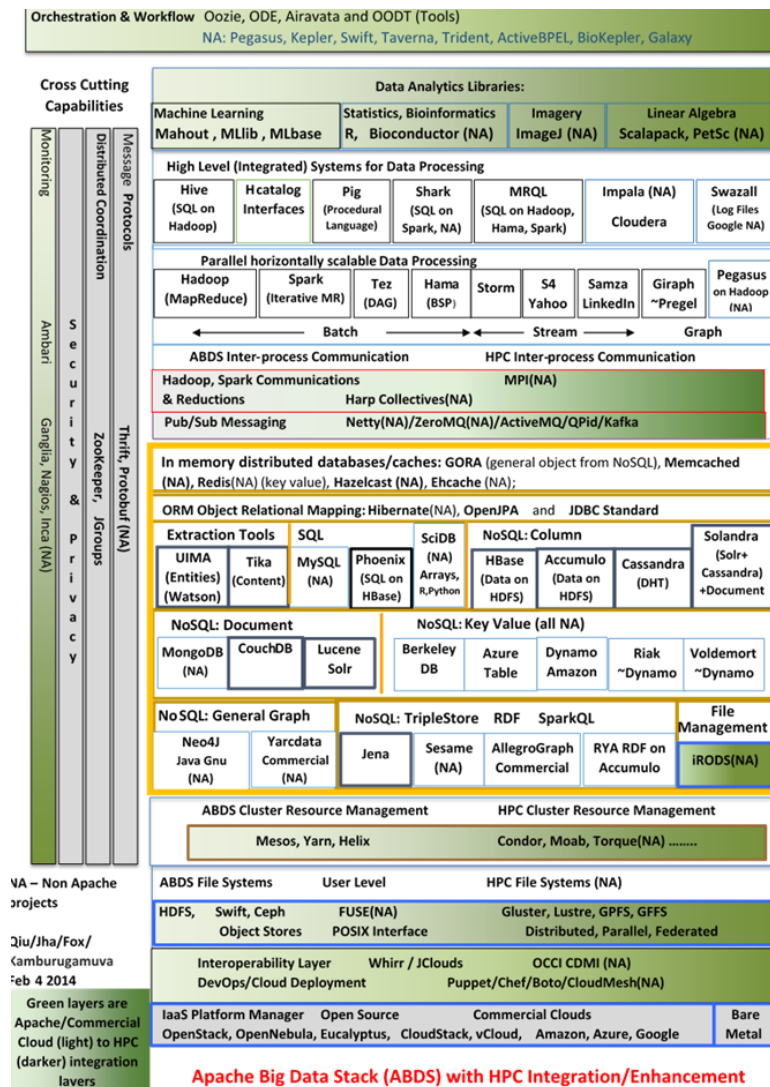
Initiative Leader: Chris Mattmann
Steering Committee Leader: Joseph Lazio

Task Title	PI	Org
1 Power Minimization in Signal Processing for Data-Intensive Science	Larry D'Addario	335
2 Machine Learning for Smart Triage of Big Data	Umaa Rebbapragada	398
3 Archiving, Processing and Dissemination for the Big Date Era	Paul Ramirez	398
4 Knowledge driven Automated Movie Production Environment distribution and Display (AMPED) Pipeline	Shigeru Suzuki	398

Initial Major Milestones for FY13	Date
Report on end-to-end power optimization of instruments	Jun 2013
Hierarchical classification method for VAST and ChemCam	Jan 2013
Demonstrate smart compression for Hyperion and CRISM	Mar 2013
Cloud computing research and scalability experiments	Feb 2013
Data formats and text, metadata extraction in big data sys.	Aug 2013
Develop AMPED pipeline and install in VIP Center	Dec 2012

NIST Big Data Architecture

- What is this on the right?
- Effort by NIST to identify Big Data Architectural Components
- Apache Software Foundation is home to *many* of these components
- Including Spark!



Spark Introduction

Apache Spark is an in-memory map-reduce platform.

<http://spark.apache.org/>

Spark features include : (1000x faster than Hadoop)



- Stream processing
- SQL Query Syntax
- Integration with Apache Mesos cluster manager
- Spark grew out of the Berkeley AMP Lab (Mattmann is on steering com)
 - Algorithms, Machines and People, investment from 80+ industry partners, DARPA XDATA and NSF CISE Expeditions in Computing

Credit: Mike Starch

Spark Community / Growth

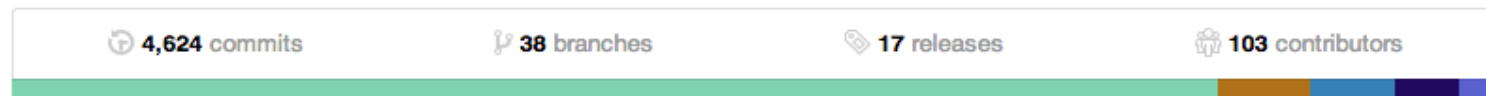
Development Community

With over 100 developers and 25 companies, one of the most active communities in big data



Mirror of Apache Spark

Credit: Matei Zaharia, Spark Summit 2013



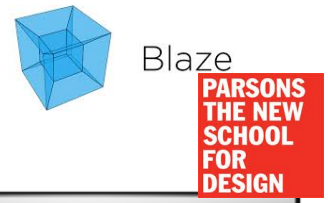
Comparison: Storm (48), Giraph (52), Drill (18), Tez (12)

Past 6 months: more active devs than Hadoop MapReduce!

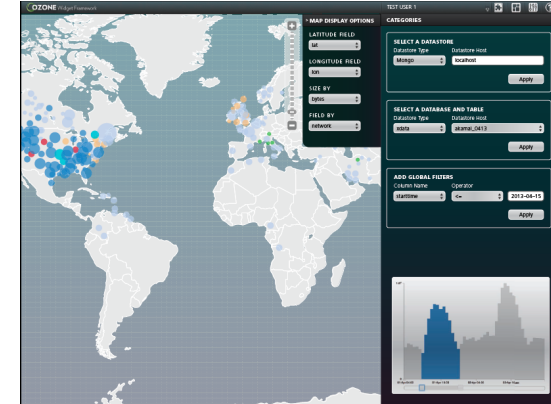
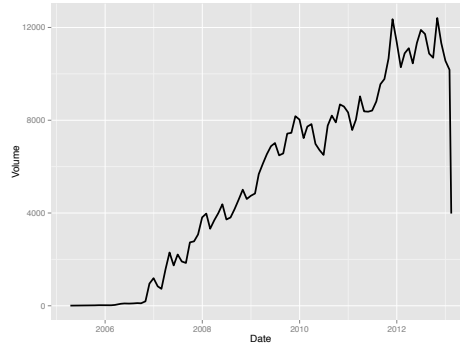
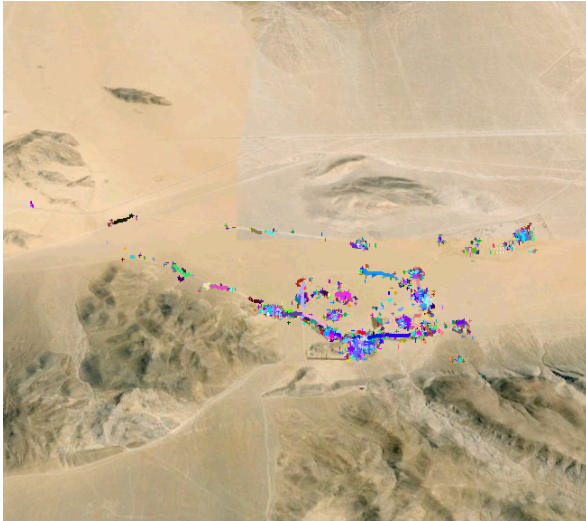
How did I get involved with
Spark?

DARPA XDATA

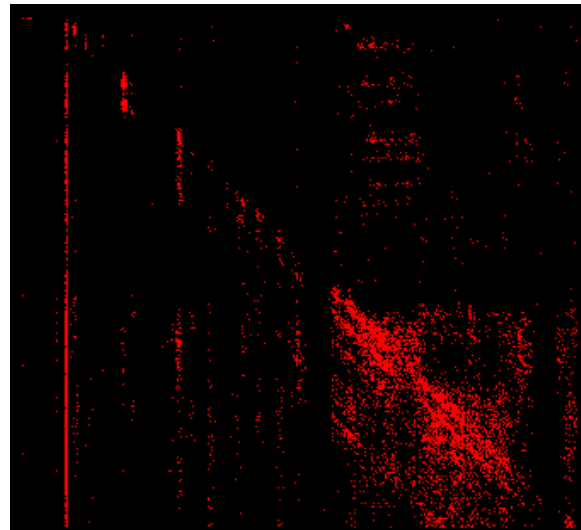
- Grab the principals behind the leading infrastructure/viz technologies
 - Shove them in a tight space
 - Provide beer coffee and snacks
 - Provide awesome data and challenges
 - Provide infrastructure and connectivity
- Check in every day and 1x a week
- Wall of Shame/Fame
- New Challenges Each Week
- Midterm Presentations
 - Peanut Gallery
- Make people talk/socialize
- Put that all together



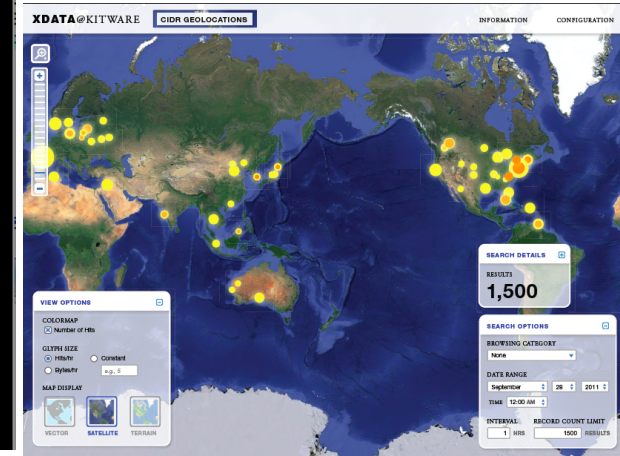
DARPA XDATA: Analytics + Viz



23-Jun-15



ESTF 2015



13

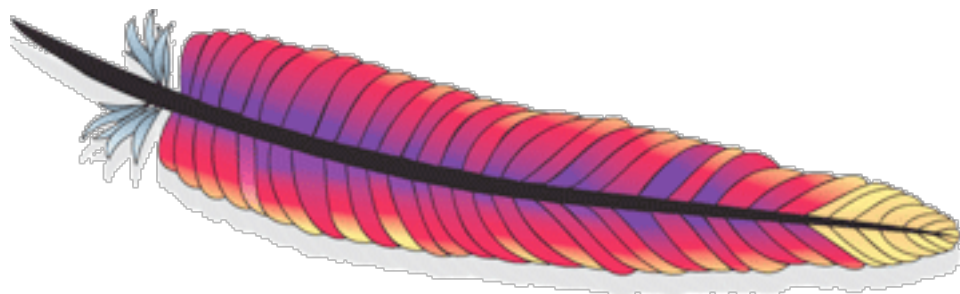
Met these fine people

- Ion Stoica
CEO, DataBricks
Co-Director,
AMP Lab
- Matt Massie
Dev Manager,
AMP Lab
- Dr. Chris White, DARPA
XDATA PM



The Apache Software Foundation

- Largest open source software development entity in the world
 - Over 2600+ committers
 - Over 4200+ contributors
 - Over 400+ members
- 100+ Top Level Projects
 - 57 Incubating
 - 32 Lab Projects
- 12 retired projects in the “Attic”
- Over 1.2 *million* revisions
- 501(c)3 non-profit organization incorporated in Delaware

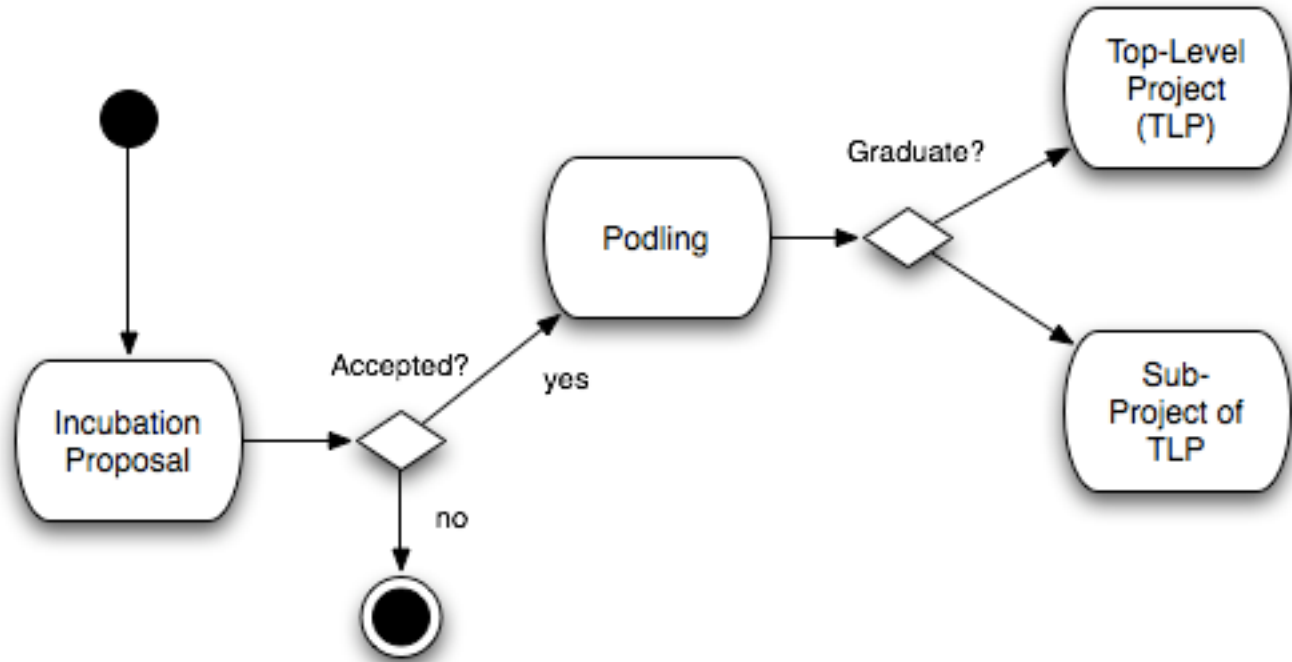


-Over 10M successful requests served a day across the world

-HTTPD web server used on 100+ million web sites (52+% of the market)

Apache Maturity Model

- Start out with Incubation
- Grow community
- Make releases
- Gain interest
- Diversify



- When the project is ready, graduate into
 - Top-Level Project (TLP)
 - Sub-project of TLP
- Increasingly, Sub-projects are discouraged compared to TLPs

Apache is a well recognized brand



GOVERNOR ARNOLD SCHWARZENEGGER

November 5, 2009

Apache Software Foundation

It is a great pleasure to extend my greetings to all those attending ApacheCon and congratulations on your tenth anniversary.

I applaud your incredible work over the past decade and appreciate you choosing California as the place to celebrate this fantastic milestone. Our state is a land of innovation, and you have likewise fostered great technological advancements that have touched the lives of millions of people around the world.

Whether managing financial systems, positioning satellites or powering websites through the Apache HTTP Server, your open source projects play key roles in making our information age possible. Thank you for your extraordinary accomplishments and commitment to discovery.

On behalf of all Californians, I send my gratitude to everyone in attendance for your participation, and I offer my best wishes for a rewarding conference and continued success.

Sincerely,

Arnold Schwarzenegger

Why Spark and NASA?

Where does Spark fit into science?



U.S. National Climate Assessment
(pic credit: Dr. Tom Painter)



SKA South Africa: Square Kilometre Array
(pic credit: Dr. Jasper Horrell, Simon Ratcliffe)

NASA Science & Architecture



Science Data File Formats

- Hierarchical Data Format (HDF)
 - <http://www.hdfgroup.org>
 - Versions 4 and 5
 - Lots of NASA data is in 4, newer NASA data in 5
 - Encapsulates
 - Observation (Scalars, Vectors, Matrices, NxMxZ...)
 - Metadata (Summary info, date/time ranges, spatial ranges)
 - Custom readers/writers/APIs in many languages
 - C/C++, Python, Java

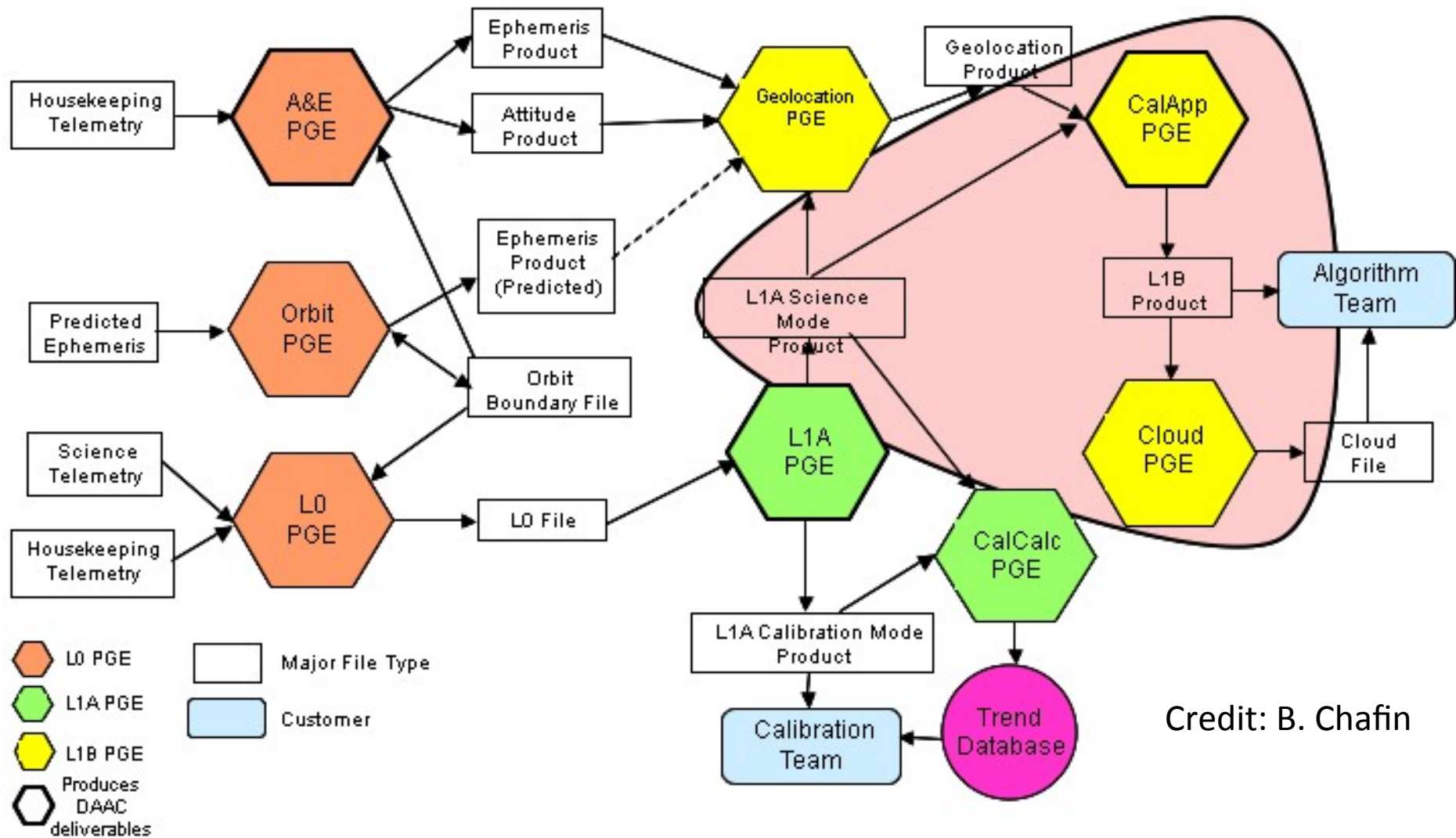


Science Data File Formats

- network Common Data Form (netCDF)
 - www.unidata.ucar.edu/software/netcdf/
 - Versions 3 and 4
 - Heavily used in DOE, NOAA, etc.
 - Encapsulates
 - Observation (Scalars, Vectors, Matrices, NxMxZ...)
 - Metadata (Summary info, date/time ranges, spatial ranges)
 - Custom readers/writers/APIs in many languages
 - C/C++, Python, Java
 - Not Hierarchical representation: all flat

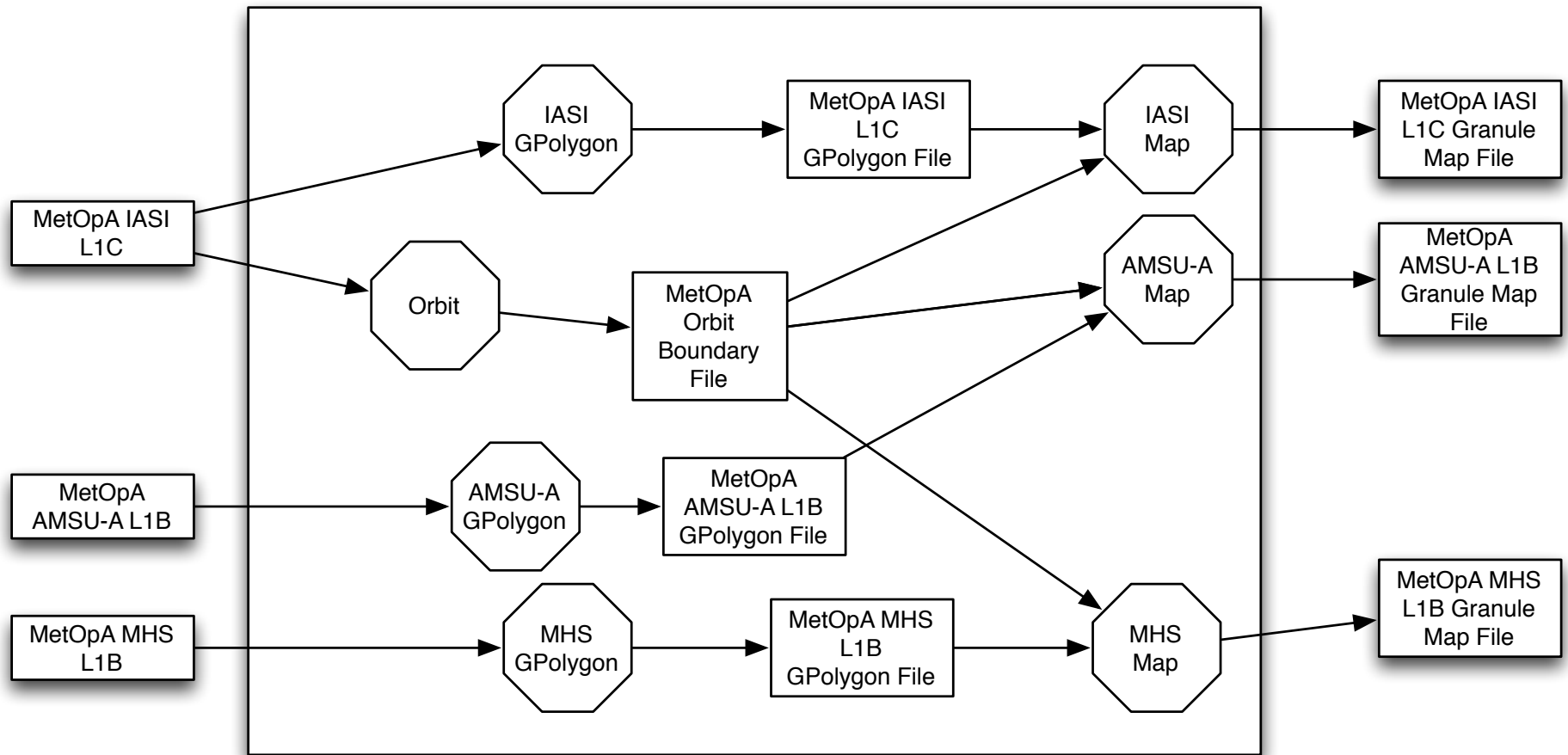


OCO-1 Workflow



Credit: B. Chafin

NPP Sounder PEATE



Credit: B. Foster

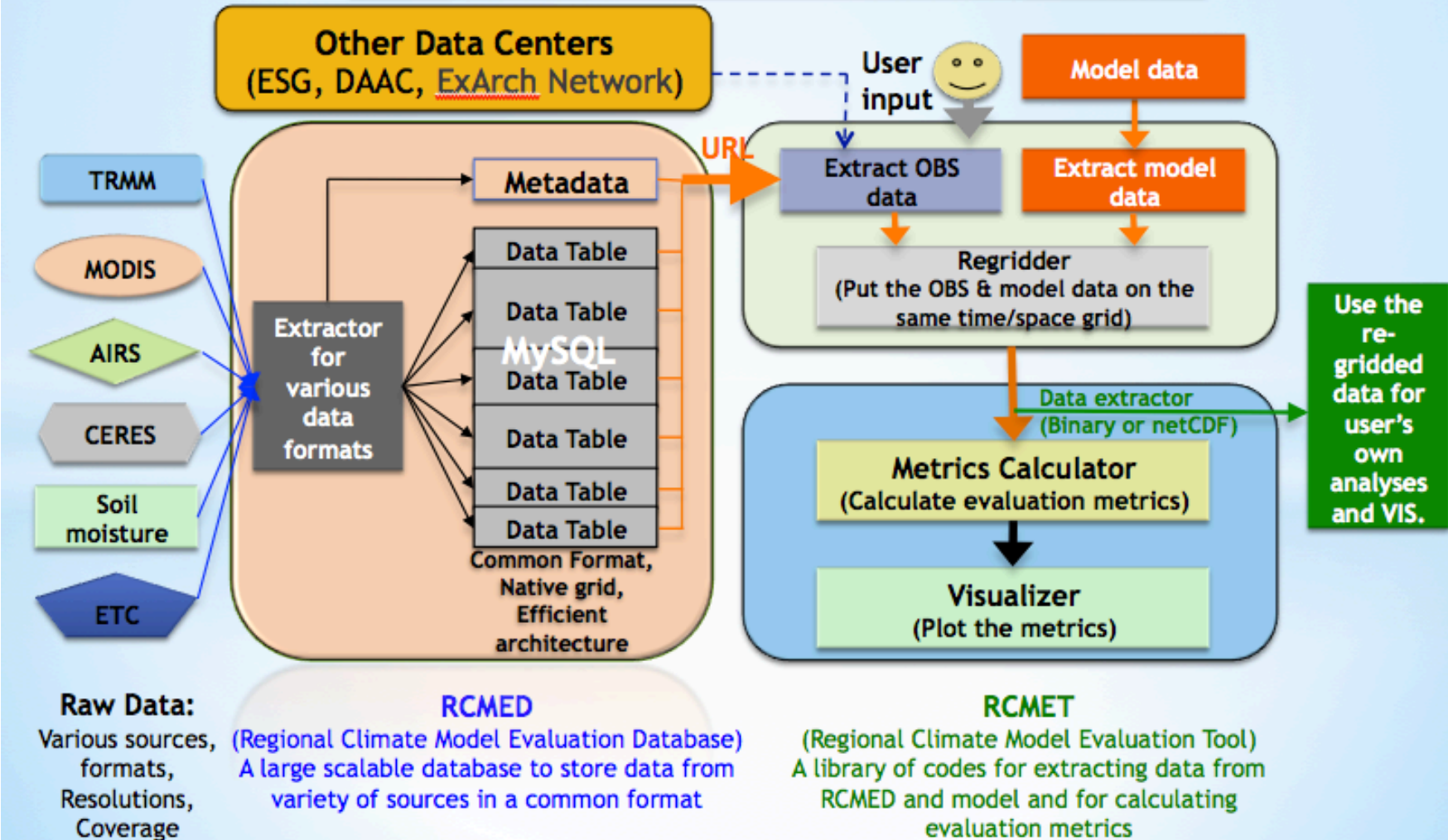
Data-Reuse Between Stages

- All of these science data pipelines
 - Read/Write NetCDF, HDF files
 - Write to distributed file systems (only recently HDFS, GlusterFS, etc.)
- Have timing constraints
- Include jobs with varying timing
 - Some early completing jobs (<1ms)
 - Some long running jobs
- What does this sound like? SPARK



SciSpark

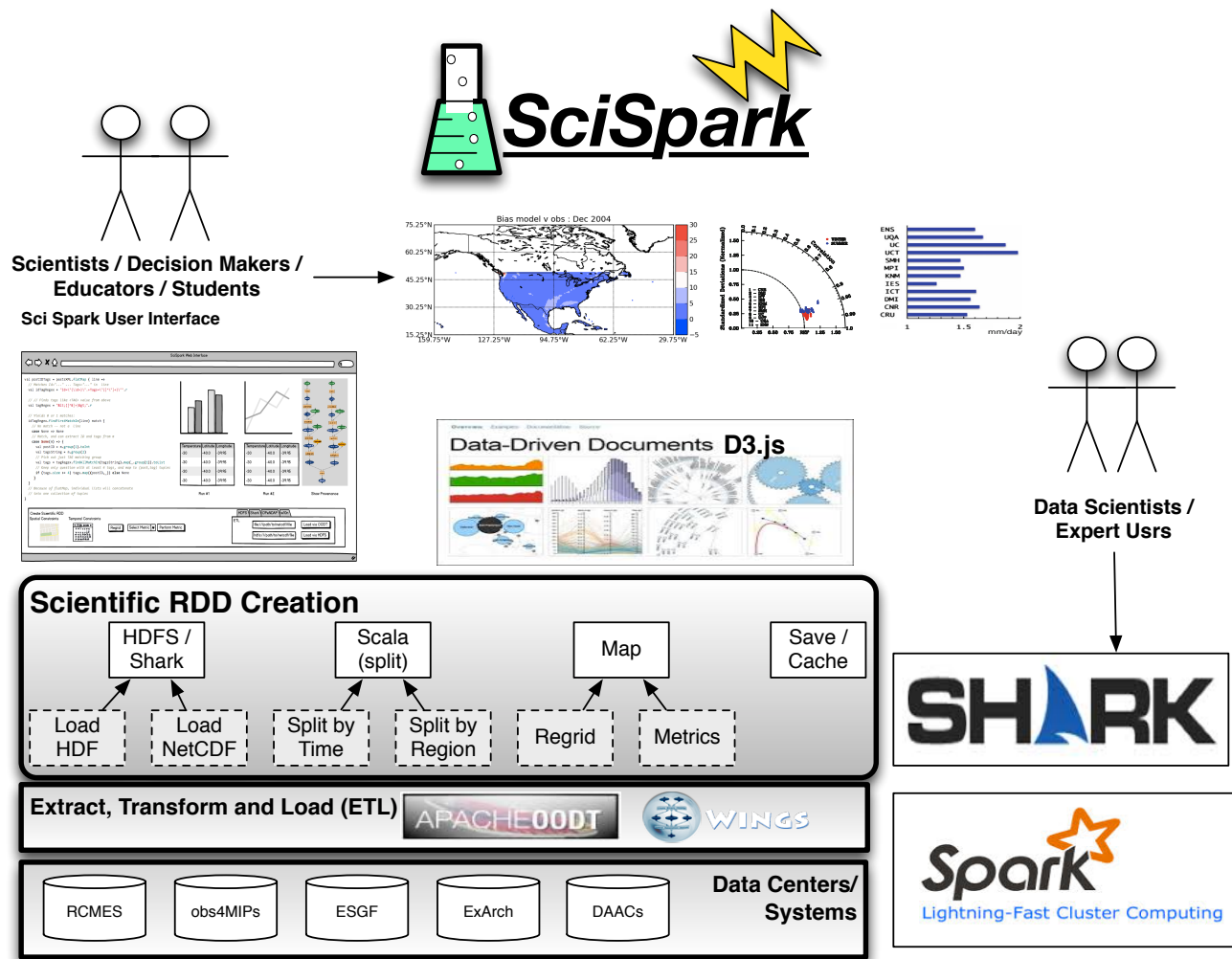
RCMES v2.0 - High-Level Architecture



Motivation for SciSpark

- Experiment with *in memory and frequent data reuse* operations
 - Regridding, Interactive Analytics such as MCC search, and variable clustering (min/max) over decadal datasets could benefit from in-memory testing (rather than frequent disk I/O)
 - Data Ingestion (preparation, formatting)

Architecture of SciSpark



Sci Spark – Visualization (D3 and friends)

- Tika, Nutch, Blaze, Bokeh, Solr, Tangelo

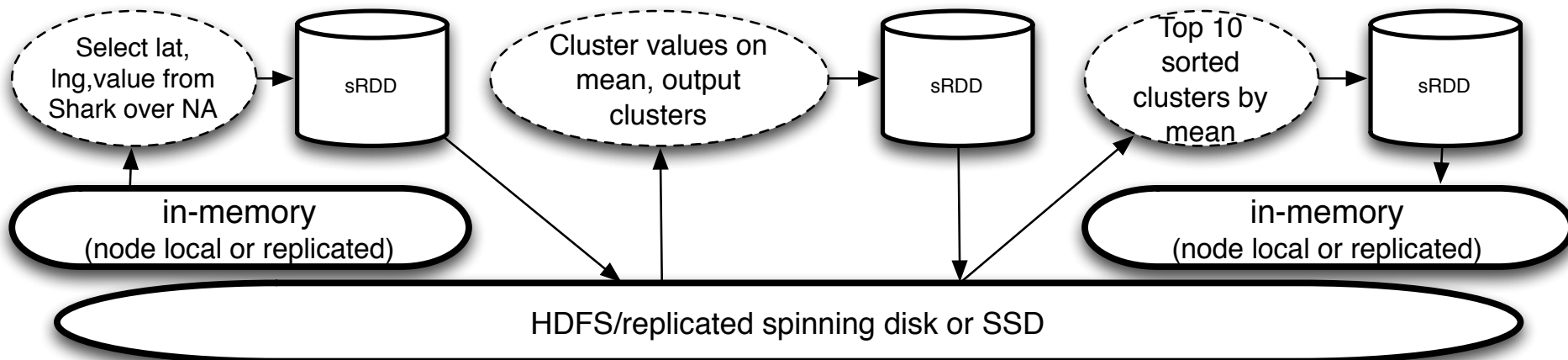
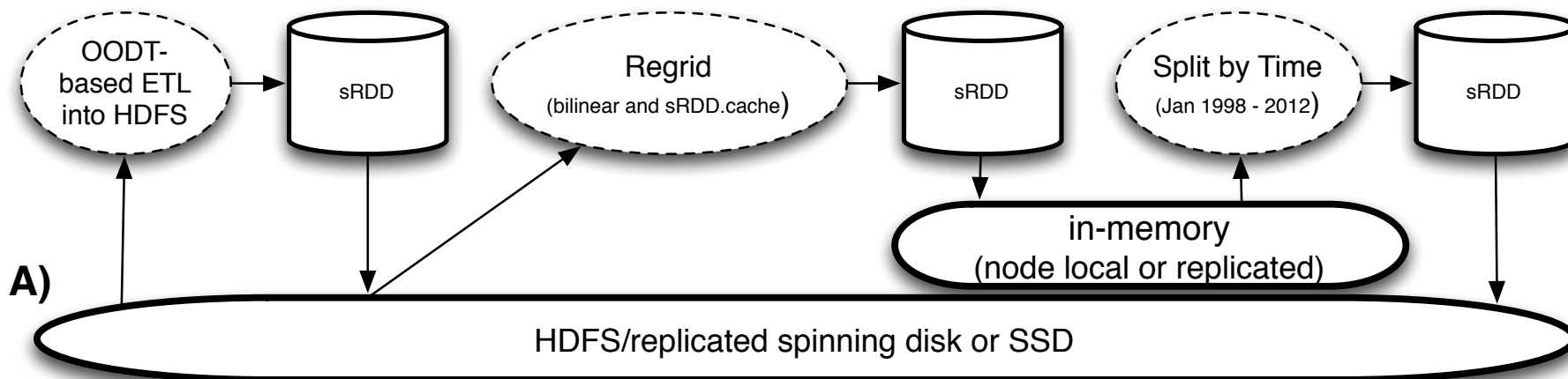
The central diagram illustrates the Sci Spark architecture. It features a 'Protocol Layer' containing 'Tika (parsing framework)', 'Gora (storage framework)', and 'HBase/Hadoop'. This layer connects to 'web pages' and an 'index' (powered by 'Solr (indexing framework)'). The data flow is shown as 'Storage' leading to 'Query' and then 'Filtering'. The background is a collage of various scientific visualizations, including a map with green dots, a line graph, a heatmap, a box plot, a scatter plot, a pie chart, and a table of chemical elements.

URL	Count	Rank	Relevance
1 http://endcomp.com/index-page-4.htm	13404	0.14	0.25
2 http://www.backpage.com	13404	0.92	0.18
3 http://endcomp.com/Pine-Tree-Spa-Pineas-Park-Florida.htm	13404	0.31	0.59
4 http://endcomp.com/Eastern-Cozy-Spa-Wayne-New-Jersey.htm	13404	0.03	0.49
5 http://endcomp.com/index-page-8.htm	13404	0.66	0.38
6 http://www.xdating.com/	13404	0.11	0.74
7 http://endcomp.com/index-page-2.htm	13404	0.61	0.09
8 http://endcomp.com/index-page-3.htm	13404	0.07	0.05
9 http://www.theeurovision.com/	13404	0.94	0.98
10 http://endcomp.com/Asian-Pure-Spa-Henderson-Nevada.htm	13404	0.33	0.50

Use Cases

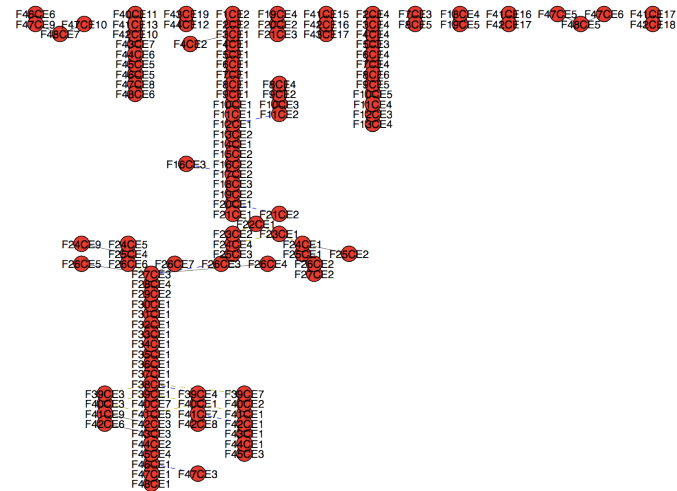
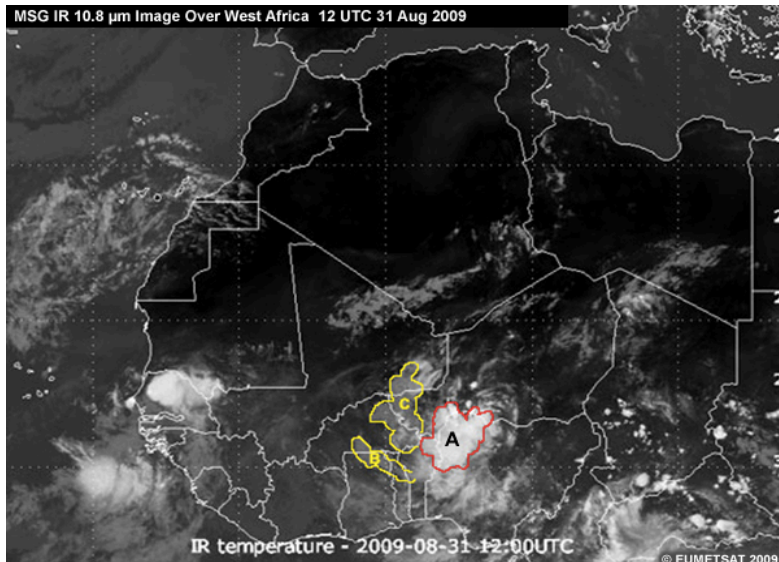
- (A) Multi-stage generation to generate time-split data
- (B) Multi-stage operation to select data from Shark, and cluster by deviation from mean

Climate Metrics on SciSpark



SciSpark – just getting started

- Funded NASA AIST14 award to construct
- SciSpark climate scenarios
 - Climate extremes / impact analysis and clustering
 - Mesoscale Convective Complex Search



Proposed SciSpark Interface

SciSpark Web Interface

```

val postIDTags = postsXML.flatMap { line =>
  // Matches Id="..." ... Tags="..." in Line
  val idTagRegex = "Id=\"(\\d+)\".+Tags=\"([^\"]+)\".r

  // Finds tags like <TAG> value from above
  val tagRegex = "&lt;([^\"]+)&gt;".r

  // Yields 0 or 1 matches:
  idTagRegex.findFirstMatchIn(line) match {
    // No match -- not a Line
    case None => None
    // Match, and can extract ID and tags from m
    case Some(m) => {
      val postID = m.group(1).toInt
      val tagsString = m.group(2)
      // Pick out just TAG matching group
      // Pick out just TAG matching group
      val tags = tagRegex.findAllMatchIn(tagsString).map(_.group(1)).toList
      // Keep only question with at least 4 tags, and map to (post,tag) tuples
      if (tags.size >= 4) tags.map((postID, _)) else None
    }
  }
  // Because of flatMap, individual lists will concatenate
  // into one collection of tuples
}
        
```

Temperature	Latitude	Longitude
-30	-40.0	-39.95
-30	-40.0	-39.95
-30	-40.0	-39.95
-30	-40.0	-39.95

Run #1

Temperature	Latitude	Longitude
-30	-40.0	-39.95
-30	-40.0	-39.95
-30	-40.0	-39.95
-30	-40.0	-39.95
-30	-40.0	-39.95

Run #2

Show Provenance

Create Scientific RDD

Spatial Constraints

Temporal Constraints

← FEB 2006 →
 S M T W T F S
 3 4 5 6 7 8
 10 11 12 13 14 15
 17 18 19 20 21 22
 24 25 26 27 28 29

HDFS Shark OPeNDAP w10n

ETL

file:///path/to/netcdf/file

Load via OOD

hdfs:///path/to/netcdf/file

Load via HDFS

23-Jun-15

ESTF 2015

35

Conclusions

- Lots of places in science and NASA for Spark
- Great connections already
- Existing Apache projects to integrate upstream
- Downstream use cases
- Come chat with me today!

EVERY SINGLE SATELLITE ORBITING THE EARTH

Credit: Vala Afshar, Extreme Networks



Thank you!

chris.a.mattmann@nasa.gov
@chrismattmann/Twitter

[http://sunset.usc.edu/
~mattmann/](http://sunset.usc.edu/~mattmann/)

