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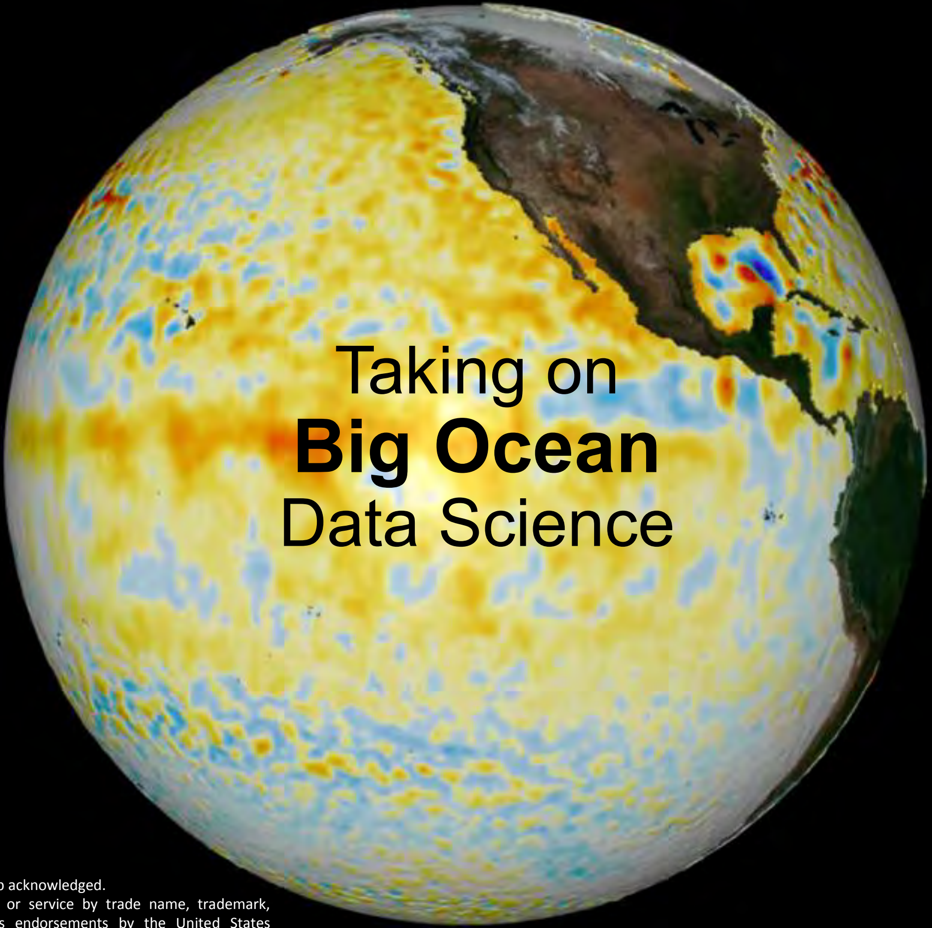

Taking on **Big Ocean** Data Science

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Data Scientist @NASA/JPL

- **Project Technologist** for the NASA's Physical Oceanography Distributed Active Archive Center (PO.DAAC) – http://podaac.jpl.nasa.gov
- **Architect** for the NASA Sea Level Change Portal – https://sealevel.nasa.gov
- **Principal Investigator / Co-Investigator** in several NASA-funded Big Data Analytic Projects
    - OceanXtremes: Oceanographic Data-Intensive Anomaly Detection and Analysis Portal – https://oceanxtremes.jpl.nasa.gov
    - Distributed Oceanographic Matchup Service (DOMS) – https://doms.jpl.nasa.gov
    - Mining and Utilizing Dataset Relevancy from Oceanographic Datasets (MUDROD)
    - Enhanced Quality Screening for Earth Science Data – https://vqss.jpl.nasa.gov
    - NEXUS - Big Data Analytic on the Cloud
- **Architect** for Tactical Data Science Framework for Naval Research
- **Ontologist** for the Semantic Web for Earth and Environmental Terminology (SWEET) Ontologies – http://sweet.jpl.nasa.gov
- **Chair** for The Federation of Earth Science Information Partners (ESIP) Cloud Computing Cluster
- **Chair/Co-Chair** for the NASA Earth Science Data System Working Groups

NASA'S PO.DAAC

- The **NASA Physical Oceanography Distributed Active Archive Center (PO.DAAC)** at Jet Propulsion Laboratory is an element of the **Earth Observing System Data and Information System (EOSDIS)**. The EOSDIS provides science data to a wide communities of user for NASA's Science Mission Directorate.

- Archives and distributes data relevant to the physical state of the ocean

- The mission of the PO.DAAC is to **PRESERVE** NASA's ocean and climate data and make these universally **ACCESSIBLE** and **MEANINGFUL**.

http://podaac.jpl.nasa.gov

Big Data Reality

Reality

- With large amount of observational and modeling data, downloading to local machine is becoming inefficient
- Data centers are starting to provide additional services
    - Better searches – faceted, spatial, keyword, relevancy, etc.
    - Data subsetting – data reduction
    - Visualization – visual discovery

2015 NASA ESTO/AIST Big Data Study Roadmap: Moving from Data Archiving to Data Analytics

Increasing “big data” era is driving needs to

- Scale computational and data infrastructures
- Support new methods for deriving scientific inferences
- Shift towards integrated data analytics
- Apply computational and data science across the lifecycle

Scalable Data Management

- Capturing well-architected and curated data repositories based on well-defined data/information architectures
- Architecting automated pipelines for data capture

Scalable Data Analytics

- Access and integration of highly distributed, heterogeneous data
- Novel statistical approaches for data integration and fusion
- Computation applied at the data sources
- Algorithms for identifying and extracting interesting features and patterns

Facts

Moving/copying science data (and managing copies) is more expensive than computation.

Hardware & software do not yet make science data analysis easy at terabyte scales.

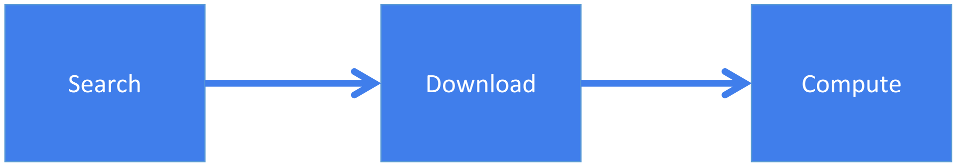
Current analytics are mostly I/O bound.

Next generation - "advanced" analytics will be compute bound (simulations, distributed linear algebra). Efficiency matters.

Current files formats are good for data archival, NOT for data analysis

> **"The scientific file-formats of HDF, NetCDF, and FITS can represent tabular data but they provide minimal tools for searching and analyzing tabular data… Performing this filter-then-analyze, data analysis on large datasets with conventional procedural tools runs slower and slower as data volumes increase."**
>
> **-- Jim Gray, Scientific Data Management in the Coming Decade**

Traditional Data Analysis



- Depending on the data volume (size and number of files)
- It could take many hours of download – (e.g. 10yr of observational data could yield thousands of files)
- It could take many hours of computation
- It requires expensive local computing resource (CPU + RAM + Storage)
- After result is produced, purge downloaded files

Observation

- Traditional methods for data analysis (time-series, distribution, climatology generation) can't scale to handle large volume, high-resolution data. They perform poorly
- Performance suffers when involve large files and/or large collection of files
- A high-performance data analysis solution must be free from file I/O bottleneck

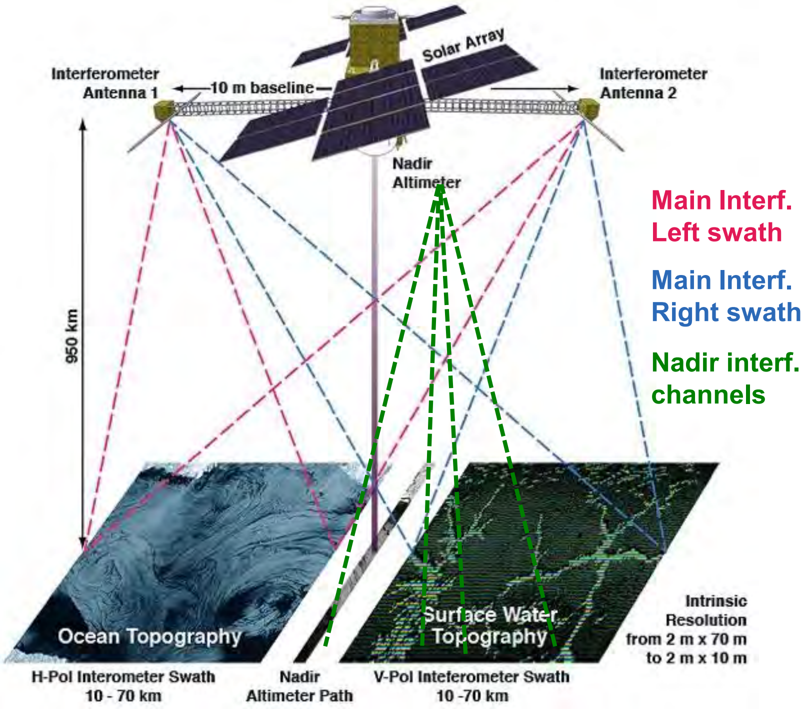
NASA's Upcoming Big Data Mission: Surface Water and Ocean Topography (SWOT)

Oceanography: Characterize the ocean mesoscale and sub-mesoscale circulation at spatial resolutions of 10 km and greater.

Hydrology: To provide a global inventory of all terrestrial water bodies whose surface area exceeds $(250m)^2$ (lakes, reservoirs, wetlands) and rivers whose width exceeds 100 m (requirement) (50 m goal) (rivers).

- To measure the global storage change in fresh water bodies at sub-monthly, seasonal, and annual time scales.
- To estimate the global change in river discharge at sub-monthly, seasonal, and annual time scales.

- **Data Volume:**
  - 17PB of original data
  - 6 PB of reprocessed data
- **Total of about 23PB for a nominal 3-year mission**
- **Add roughly 450TB/month for any mission extension**



Launches April of 2021
https://swot.jpl.nasa.gov

The Silver Bullet?

Moore's Law is coming to an end due to physical limits of CMOS
"the number of transistors can put on a microchip doubles every year or so."

Cloud Computing provides an elastic infrastructural approach to Big Data

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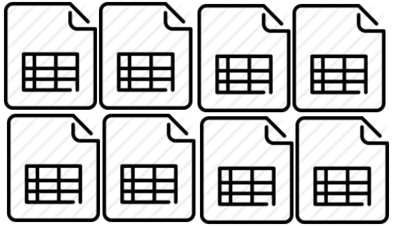
But… How Do We Get There?

Here → **There**



NEXUS Deep Data Analytics: One-Minute Summary

NEXUS is an emerging technology developed at JPL

- Open Source: https://github.com/dataplumber/nexus
- A Cloud-based/Cluster-based data platform that performs scalable handling of observational parameters analysis designed to scale horizontally by
- Leveraging high-performance indexed, temporal, and geospatial search solution
- Breaks data products into small chunks and stores them in a Cloud-based data store

Data Volumes Exploding

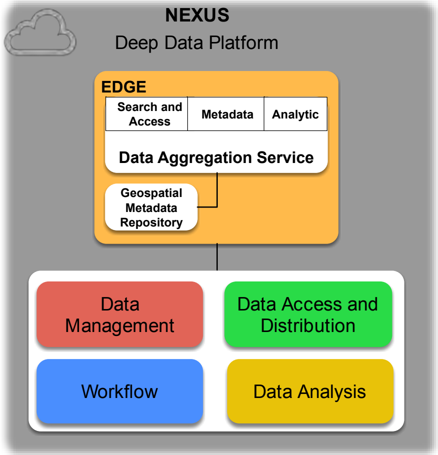
- SWOT mission is coming
- File I/O is slow

Scalable Store & Compute is Available

- NoSQL cluster databases
- Parallel compute, in-memory map-reduce
- Bring Compute to Highly-Accessible Data (using Hybrid Cloud)

Pre-Chunk and Summarize Key Variables

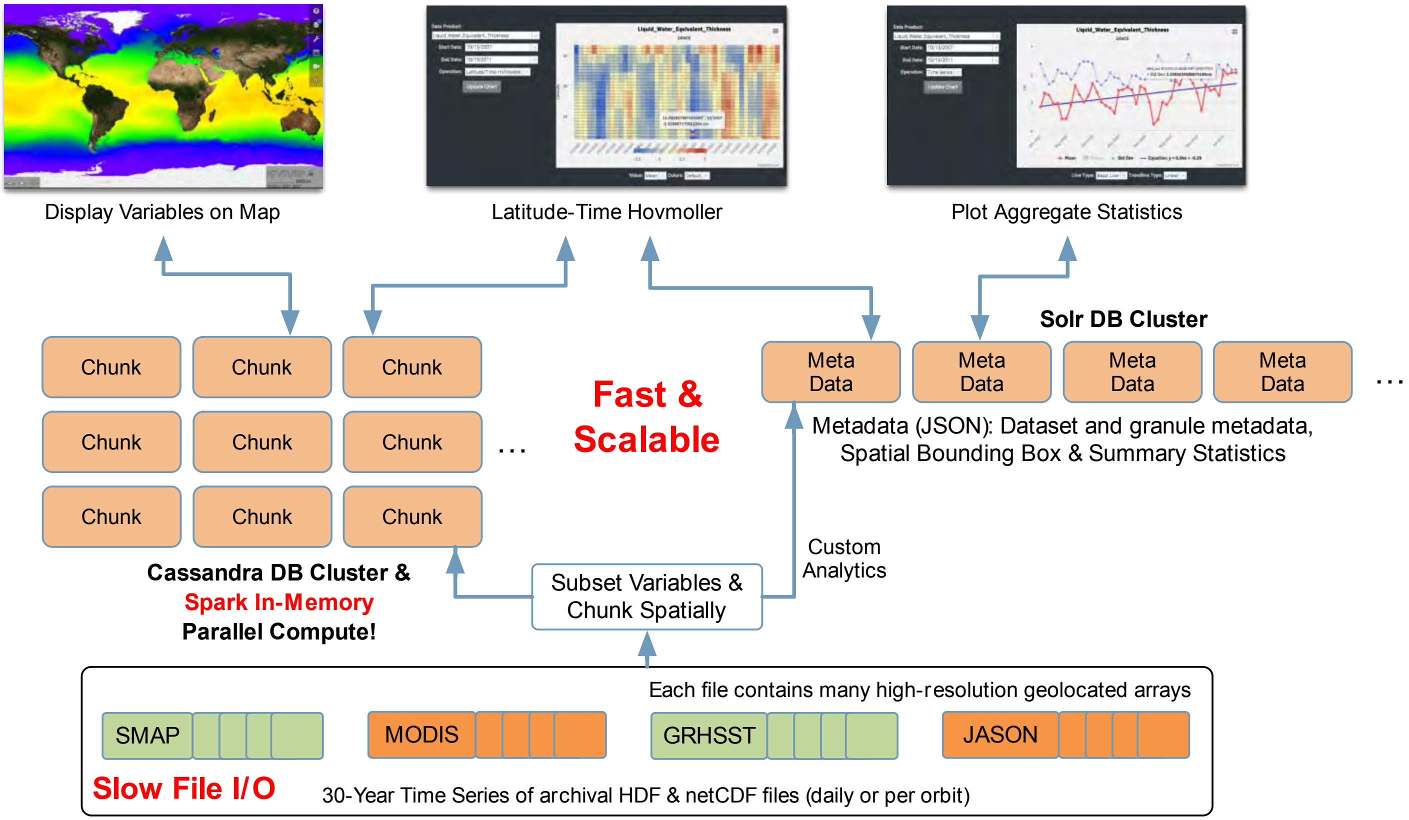
- Easy statistics instantly (milliseconds)
- Harder statistics on-demand (in seconds)
- Visualize original data (layers) on a map quickly



Built with open source technologies

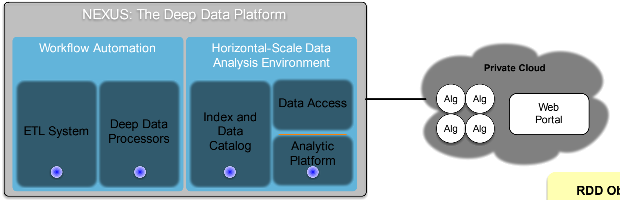
- Apache Solr
- Apache Cassandra
- Apache Spark/PySpark
- Apache Mesos/YARN
- Apache Kafka
- Apache Zookeeper
- Tornado
- Spring XD
- EDGE

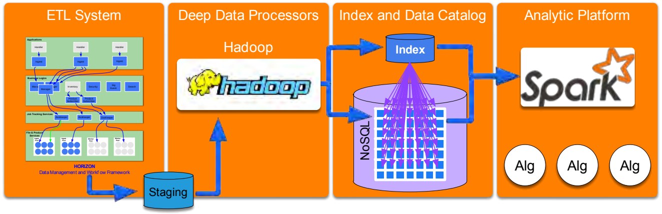
Analytics & Summarization Stack Using Two Scalable Database Technologies

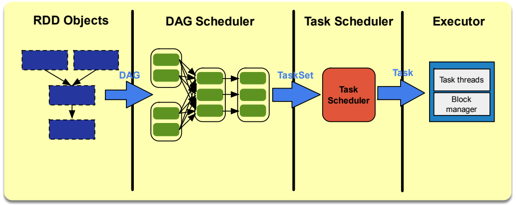


Deep Data Computing Environment (DDCE)

- ETL System – Ingest and stage data
- Deep Data Processors – metadata, statistics, and tiles
- Index and Data Catalog – horizontal-scale geospatial search and tile retrieval
- Analytic Platform – Spark-based domain-specific analytics
- Data Access – tile and collection-based data access
- Cloud Platform – portal and custom VMs







Credit JPL: T. Huang, B. Wilson, G. Chang, E. Armstrong, T. Chin

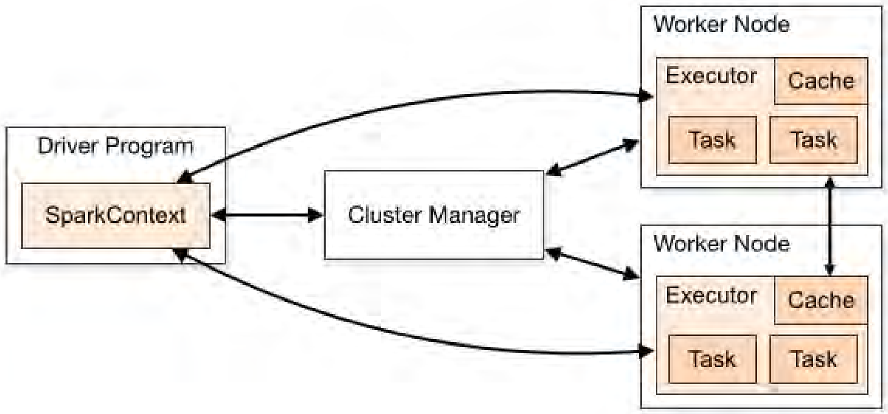
AIST-14: OceanXtremes

openstack CLOUD SOFTWARE

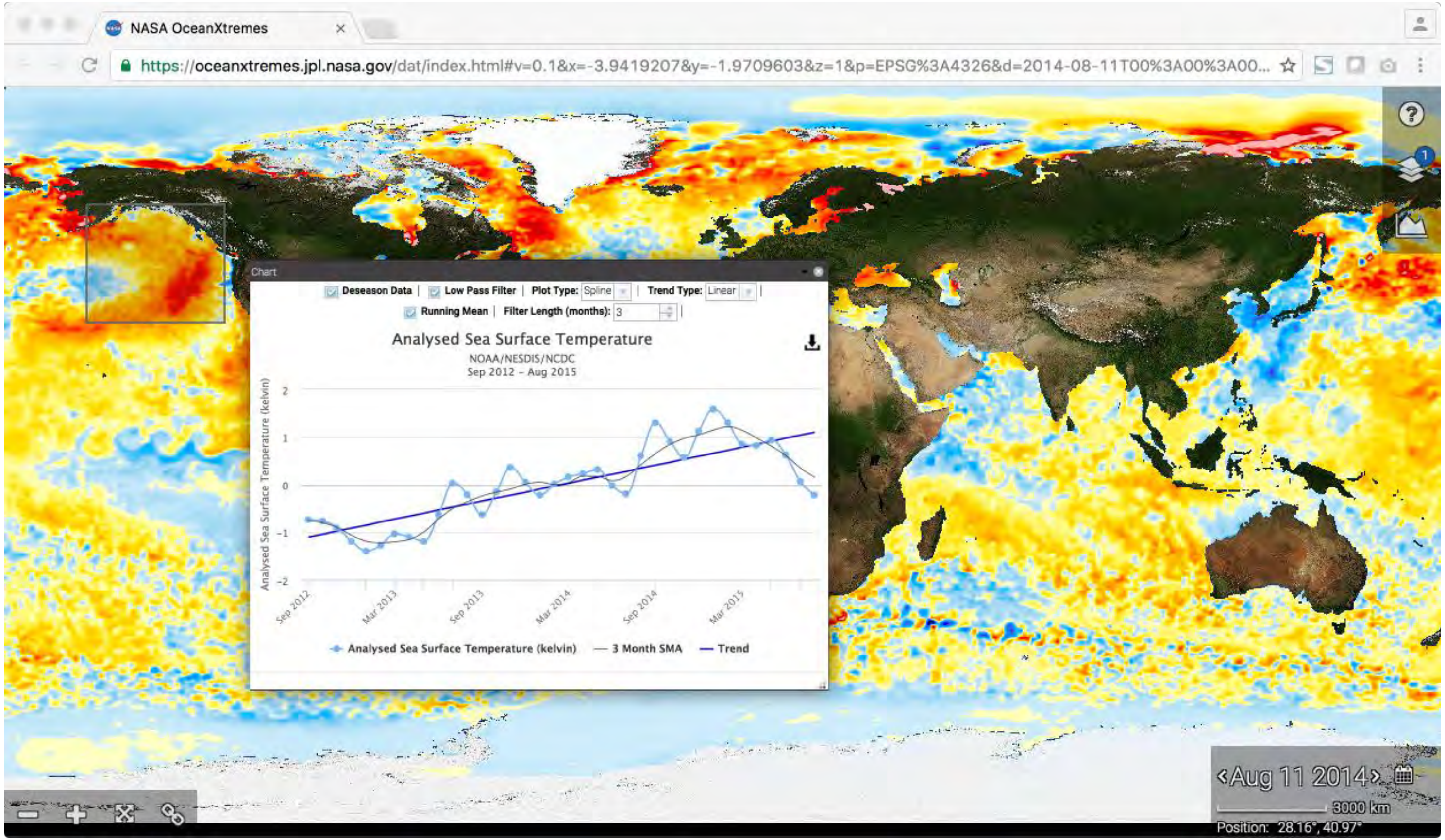
amazon web services

Using Apache Spark and Cassandra

- **Apache Spark**
  - In-Memory Map-Reduce framework
  - Datasets partitioned across a compute cluster by key
  - Resilient Distributed Dataset (RDD)
    - Fault-tolerant, parallel data structures
    - Intermediate results persisted in memory
    - User controls the partitioning to optimize data placement
  - Rich set of operators on RDD's: `Map, Filter, GroupByKey, ReduceByKey`, etc.
  - Computation is implicit (Lazy) until answers needed
  - Uses YARN/Mesos
- **Apache Cassandra**
  - Horizontal-scale NoSQL database
  - Constant-time writes regardless of the size of data set grows
  - No-single-point of failure architecture



NEXUS Real Time Analysis



The Notebook

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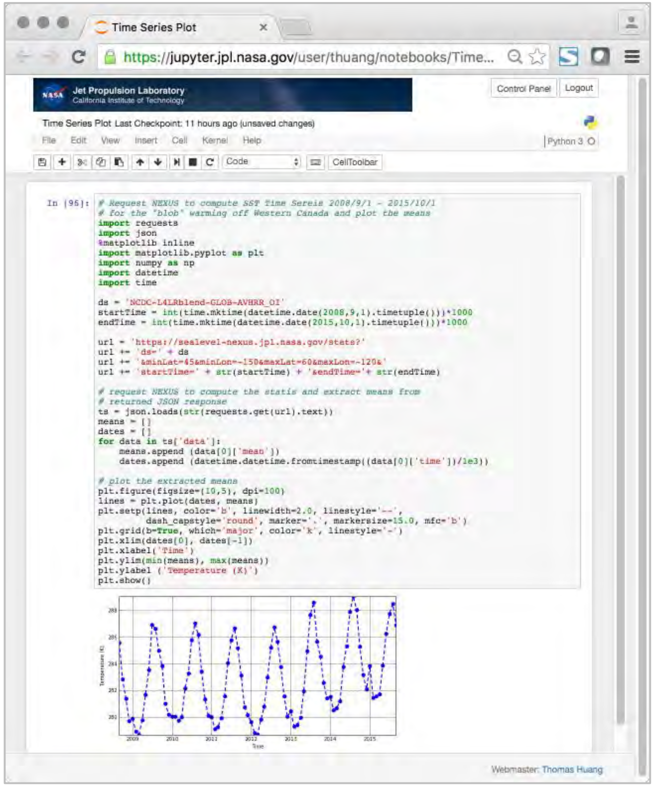
Interact with NEXUS using Jupyter Notebook



- **`/capabilities`**: list of capabilities
- **`/chunks`**: list data chunks by location, time, and datasets
- **`/correlationMap`**: Correlation Map
- **`/datainbounds`**: Matchup operation to fetch values from dataset within geographic bounds
- **`/datapoint`**: Matchup operation to fetch value at lat/lon point
- **`/dailydifferenceaverage`**: Daily difference average
- **`/latitudeTimeHofMoeller`**: Latitude Time Hovmoeller
- **`/list`**: list available datasets
- **`/longitudeLatitudeMap`**: Longitude Latitude Map
- **`/longitudeTimeHofMoeller`**: Longitude Time Hovmoeller
- **`/stats`**: Statistics (standard deviation, count, min/max, time, mean)

NEXUS 2.0 Performance Challenge

The Challenge: Show that NEXUS performs 2X or greater speed improvement compare to Giovanni – Sponsored by NASA/ESDIS

- **Dataset**: TRMM Daily Precipitation (TRMM 3B42 Daily V7), 18 years, 6574 granule files, ~26GB
- **Algorithms**
  - Area Averaged Time Series
  - Global Time Averaged Map
  - Correlation Map
- **Giovanni** – web application for researchers to analyze NASA's gridded data.
  - Backed by the popular NCO (NetCDF Operator) library, highly optimized C/C++ library
- **NEXUS**
  - Apache Solr for spatial searches, metadata, and pre-computed statistics
  - Apache Cassandra for clustered data storage where granule data is partitioned into tiles
  - Apache Spark for data analytic platform

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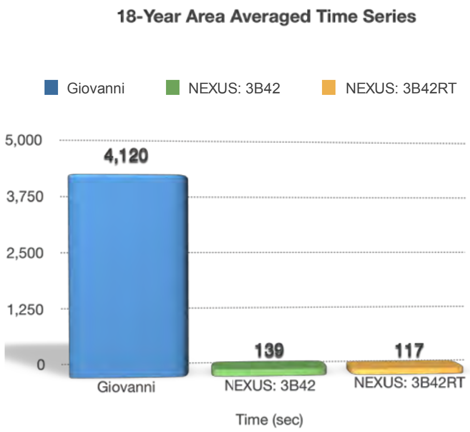
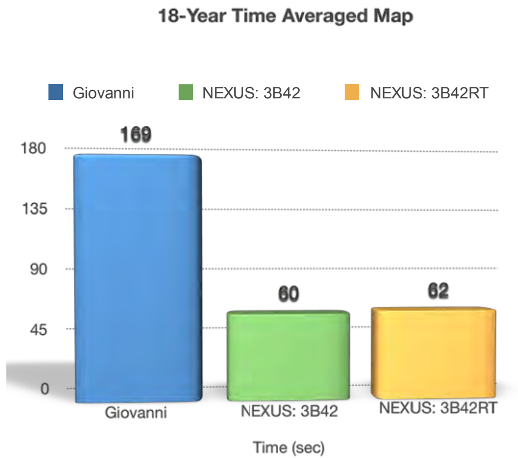
Performance Statistics

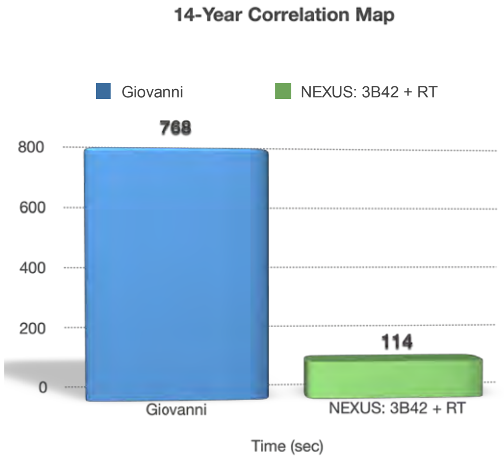
All performed on Apache Spark cluster with 16-way parallel



Giovanni: over an hour
NEXUS: a little over 2min
~30X faster



Giovanni: about 3min
NEXUS: 1min
~3X faster



Giovanni: about 13min
NEXUS: 2min
~7X faster

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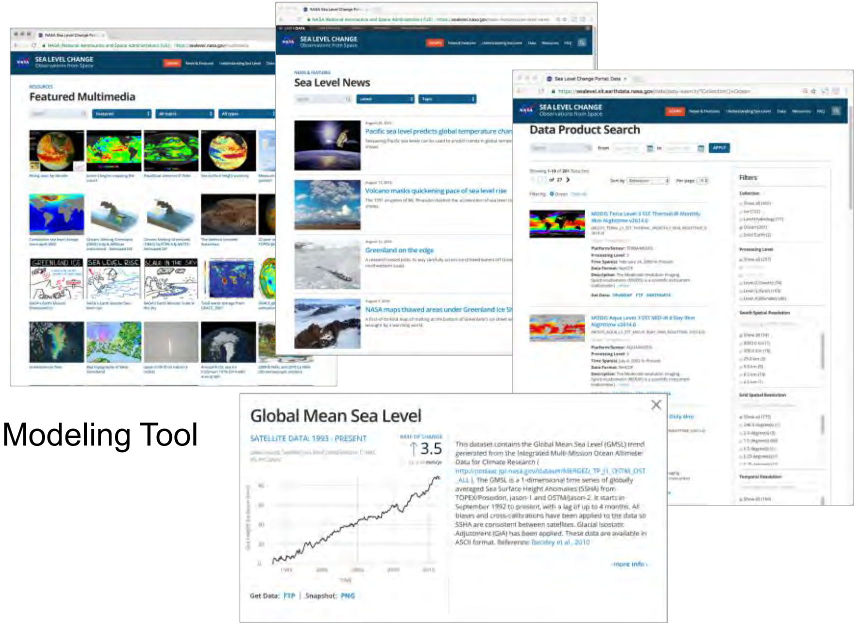
NASA Sea Level Change Portal

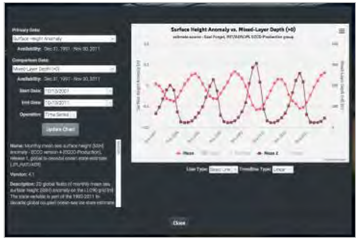
Goals

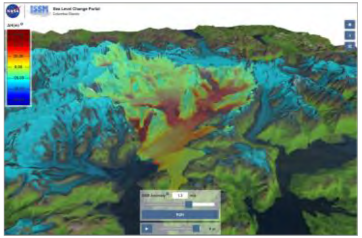
- Provide scientists and the general public with a “one-stop” source for current sea level change information and data
- Provide interactive tools for accessing and viewing regional data
- Provide virtual dashboard for sea level indicators
- Provide latest news, quarterly report, and publications
- Provide ongoing updates through a suite of editorial products
  - Content articles
  - Multimedia

Features

- Featured news
- Sea level indicators
- Understanding sea level
  - Causes
  - Observations
  - Projections
  - Adaptation
- Data search
- Data Analysis Tool
- Ice Sheet Simulation and Modeling Tool
- Multimedia
- Sea level news
- Scientist interviews
- Publications
- Commentary
- Featured multimedia
- Subscription for newsletter

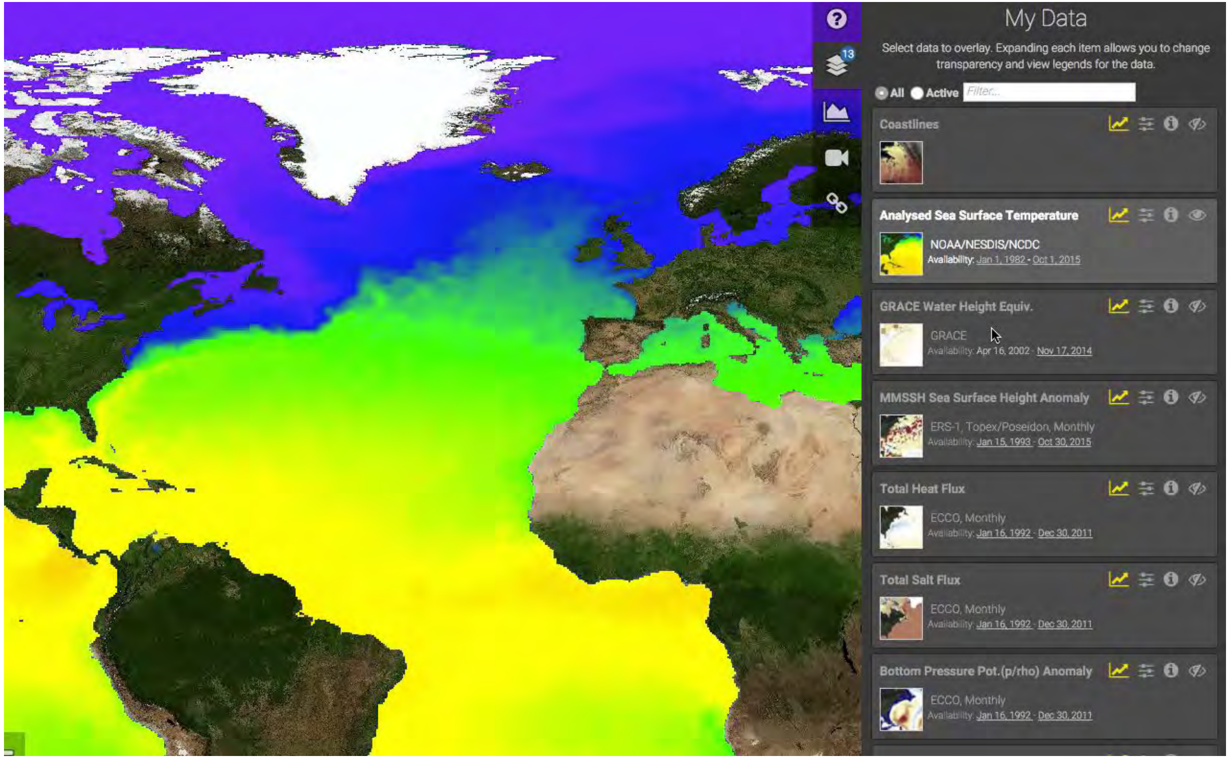






https://sealevel.nasa.gov

On-The-Fly Analysis for Sea Level Rise Research



Sea Level Change - Data Analysis Tool

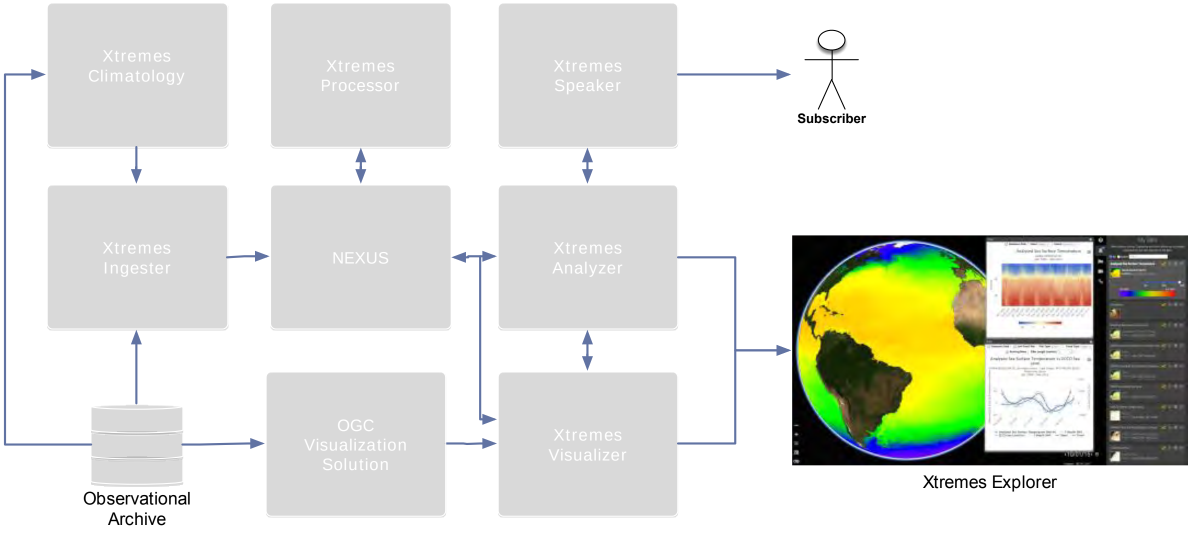
- Visualizations – WMTS – tiled imagery webserivce
- Time Series
- Data Comparison
- Latitude/Time Hofmoeller
- Etc.

Ocean Anomalies

Identifying observations which do not conform to an expected pattern in a dataset or time series.

OceanXtremes – Data-Intensive Anomaly Detection System

- Funded by the NASA Advanced Information System Technology program
- Current and future oceanographic missions and our research communities present us with challenges to rapidly identify features and anomalies in increasingly complex and voluminous observations
- Goals
    - Provide one-stop portal registry of ocean anomalies
    - Provide on-the-fly analysis and mining on observational data
- Typically this is a two-stage procedure
    1. Determine a long-term/periodic mean ("climatology")
    2. Deviations from the mean are searched. Step 1 could be omitted in cases where a climatology data set already exists.



OceamXtremes System Architecture

Credit: T. Huang, E. Armstrong, G. Chang, T. Chin, B. Wilson

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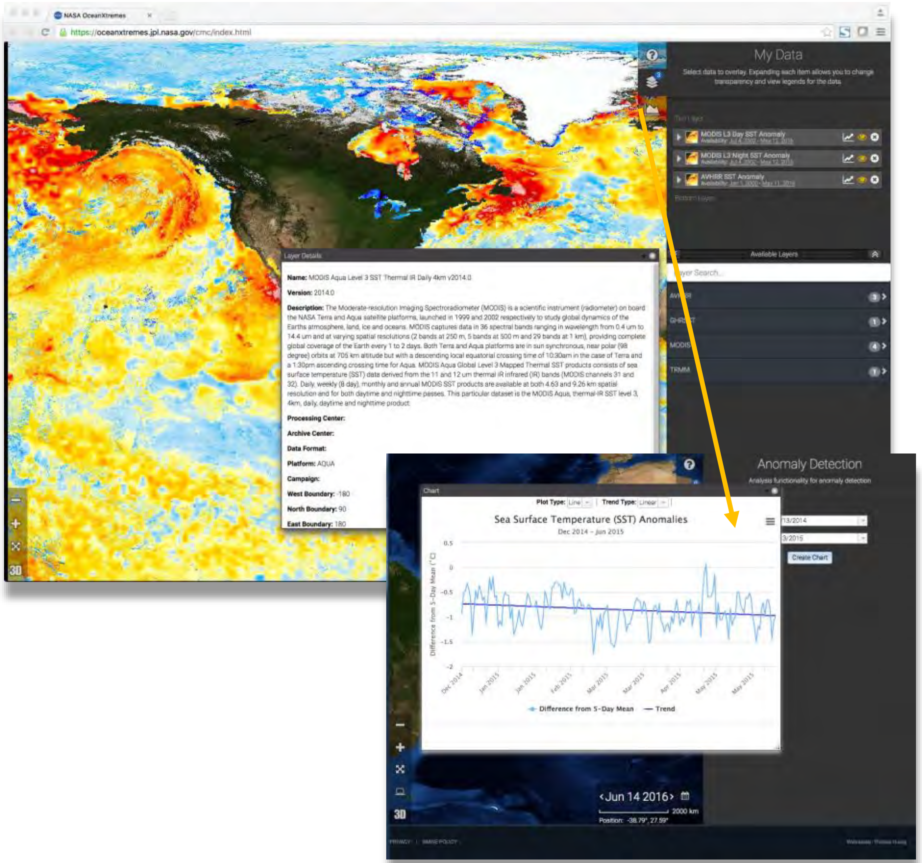
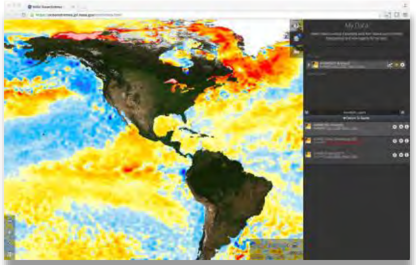
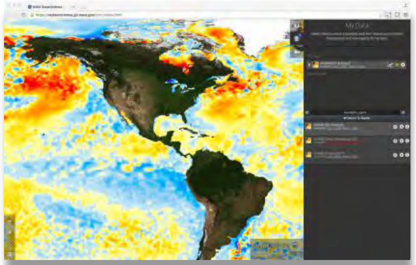
Pasadena, California

Xtremes Explorer: Daily Anomaly

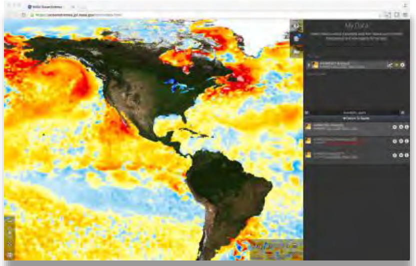




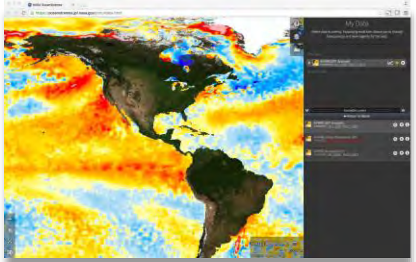
Aug 02, 2012



Aug 02, 2013



Aug 02, 2014



Aug 02, 2015

Matchup

Reconciliation of satellite and in-situ datasets

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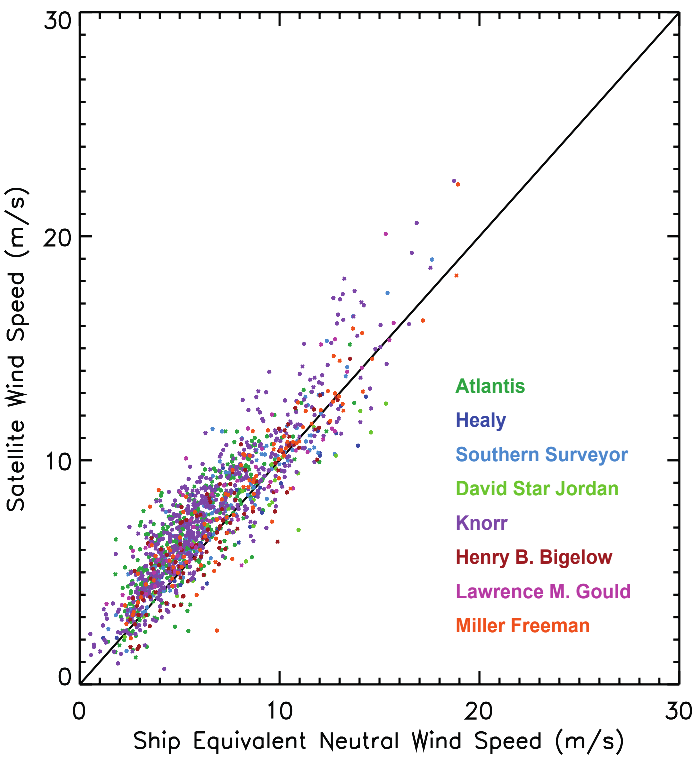
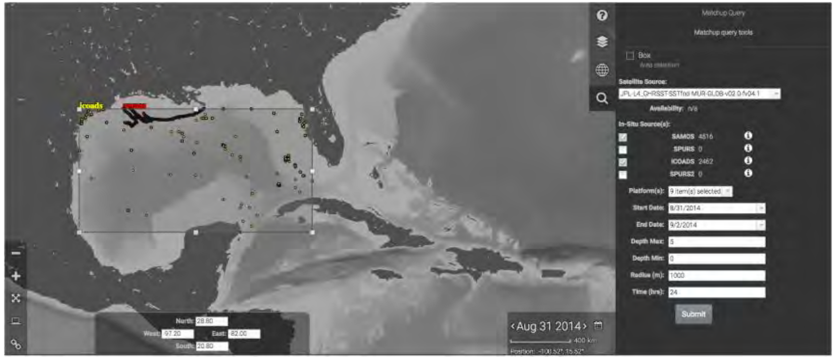
Distributed Oceanographic Matchup Service (DOMS)

- Funded by the NASA Advanced Information System Technology program
- A distributed data service (a.k.a DOMS) to match satellite and in situ marine observations to support platform comparisons, cross-calibration, validation, and quality control
- Use Cases
    - Satellite Cal/Val and algorithm development
    - Decision support
        - Planning field campaigns
        - Real-time operational activities
    - Scientific investigation
        - Process studies
        - Model assimilation services
        - User friendly interface to support student research
    - Alternate matching
        - Satellite to satellite
        - Satellite/in situ to model



DOMS Data



- **In-Situ**
  - **Shipboard Automated Meteorological and Oceanographic System (SAMOS)** initiative provides high-quality underway data from research vessels.
    - Hosted at Florida State University's Center for Ocean-Atmospheric Prediction Studies (COAPS), Tallahassee, Florida
    - Example OpenSearch query: `"http://doms.coaps.fsu.edu/ws/search/samos?startTime=2012-08-01T00:00:00Z&endTime=2013-10-31T23:59:59Z&bbox=-45,15,-30,30"`
  - **International Comprehensive Ocean-Atmosphere Data Set (ICOADS)** is a global ocean marine meteorological and surface ocean dataset.
    - Hosted at the National Center for Atmospheric Research (NCAR), Boulder, Colorado. Over 500 million measurements since year 1662.
    - Example OpenSearch query: `"http://rda-data.ucar.edu:8890/ws/search/icoads?startTime=2012-08-01T00:00:00Z&endTime=2013-10-31T23:59:59Z&bbox=-45,15,-30,30"`
  - **Salinity Processes in Upper Ocean Regional Study (SPURS)**
    - Hosted at the NASA Physical Oceanography Distributed Active Archive Center (PO.DAAC), Pasadena, California.
    - Example OpenSearch query: `"https://doms.jpl.nasa.gov/ws/search/spurs?startTime=2012-08-01T00:00:00Z&endTime=2013-10-31T23:59:59Z&bbox=-45,15,-30,30"`
- **Satellite** – All managed by **NASA's Physical Oceanography Distributed Active Archive Center (PO.DAAC)**
  - Group for High Resolution Sea Surface Temperature (GHRSST) Level 4 Multiscale Ultrahigh Resolution (MUR)
  - Advanced Scatterometer (ASCAT) Level 2 coastal ocean surface wind vector
  - Soil Moisture Active Passive (SMAP) Level 2 Sea Surface Salinity (SSS)

DOMS Architecture

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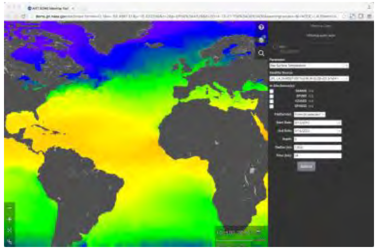
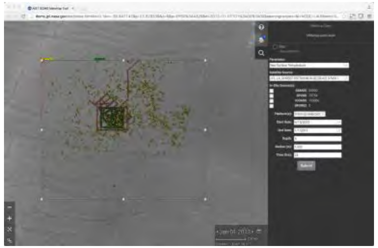
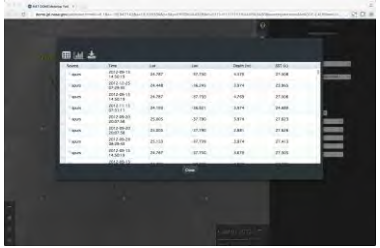


High resolution data visualization



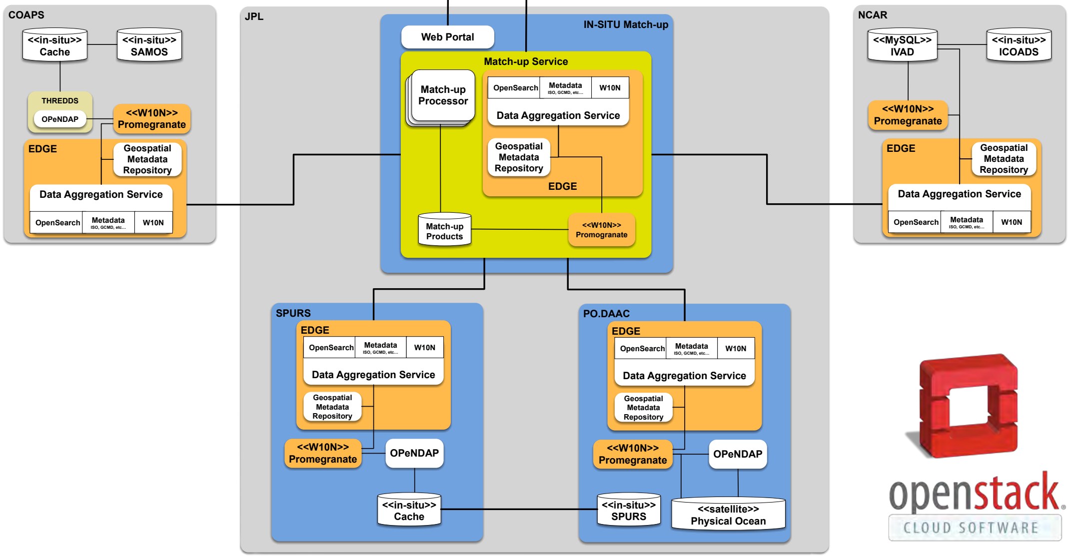
Initial match according to user initial selection



SST matchup with SPURS

Dynamic Match-up

Credit: S. Smith/FSU, S.Worley/NCAR, T. Huang/JPL, V. Tsontos/JPL, B. Holt/JPL



openstack CLOUD SOFTWARE

```
% curl -X GET
"https://doms.jpl.nasa.gov/nexus/match_spark?primary=JPL-L4_GHRSST-SSTfnd-MUR-GLOB-v02.0-fv04.1&matchup=spurs&startTime=2012-09-25T00:00:00Z&endTime=2012-09-30T23:59:59Z&b=-40,25,-45,30&platforms=1,2,3,4,5,6,7,8,9&depthMin=0.0&depthMax=5.0&tt=86400&rt=1000.0&parameter=sst"
```

Search and Discovery

Finding the right data and uncover related data and services

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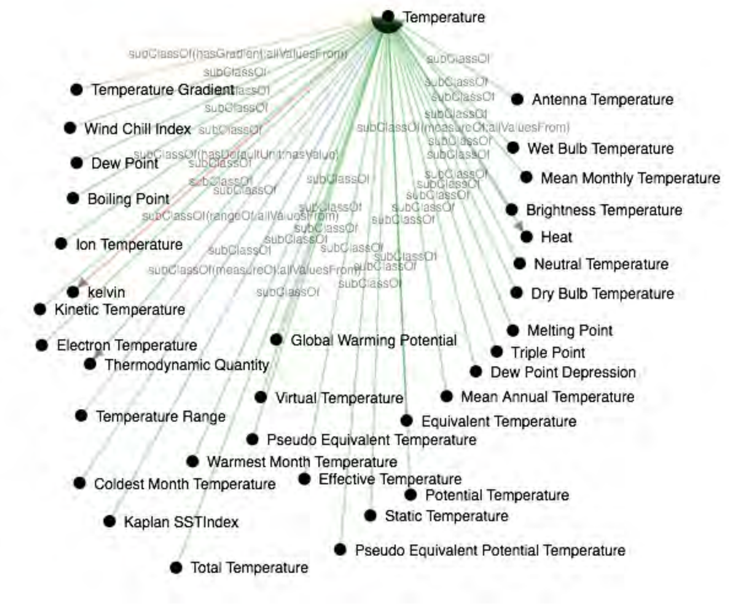
Search Relevancy

- Traditional keyword/key-phase search will not be adequate when dealing v petabyte-scale data
- What happen when a keyword/key-phase search returns thousands/millior hits? Which one should the user look at or download?
- **Search** – look for something you expect to exist
  - Information tagging
  - Indexed search technologies like Apache Solr or ElasticSearch
  - The solution is pretty straightforward
- **Discovery** – find something new, or in a new way
  - This is non-trivial
  - Traditional ontological method doesn't quite add up
  - The strength of semantic web is in inference
  - What happen when we have a lot of **`subClassOf`**, **`equivalentClassOf`**, **`sameAs`**?
  - How wide and deep should we go?
- **Relevancy**
  - It is domain-specific
  - It is personal
  - It is temporal
  - It is dynamic

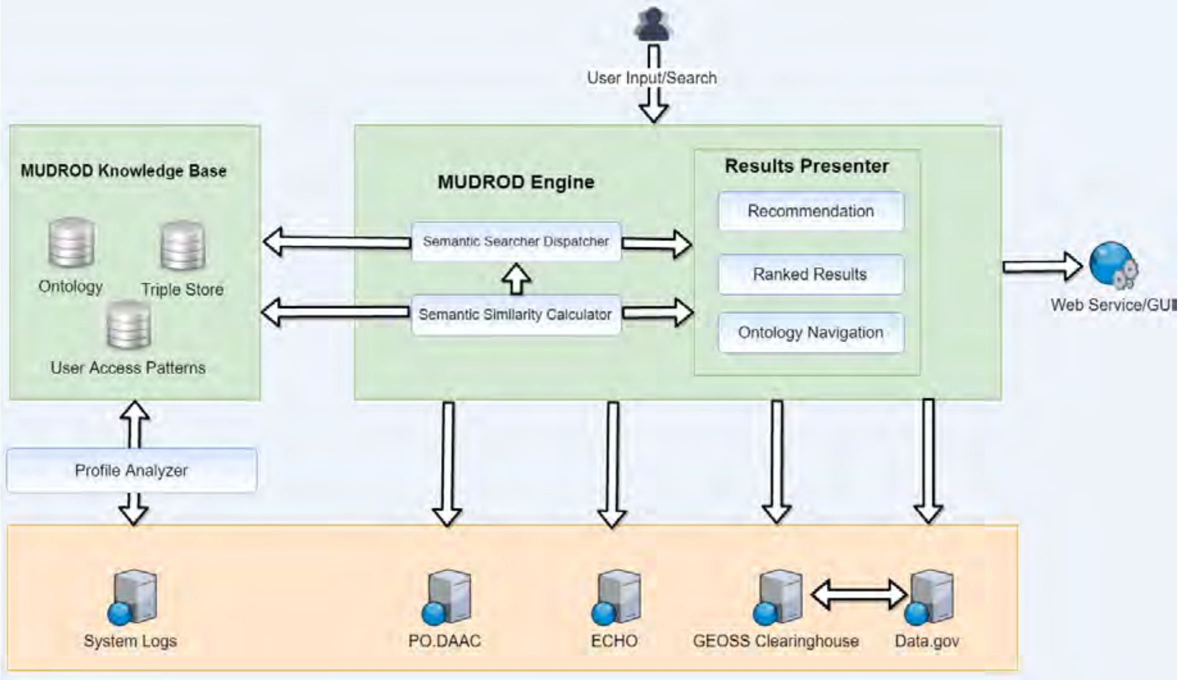


SWEET Ontologies' Temperature Concept
http://sweet.jpl.nasa.gov

Mining and Utilizing Dataset Relevancy from Oceanographic Datasets (MUDROD)

- Funded by the NASA Advanced Information System Technology program
- Analyze web logs to discover user knowledge (the connections between datasets and keyword)
- Construct knowledge base by combining semantics and profile analyzer
- Improve data discovery by
  - Better ranked results
  - Recommendation
  - Ontology navigation

✓ Web log processing
✓ Session reconstruction
✓ Vocabulary semantic relationship extraction
✓ Search ranking
✓ Recommendation



Key Takeaways

- OceanXtremes, DOMS, and MUDROD just finished their 1st year of development. Year 2 focuses on more performance updates and increase TRL level.
- More parallelism ≠ Faster performance
  - For global analytics, a lot of smaller tiles actually yields slower performance, because of scheduling, data transport, data queries, etc.
- Big Data ≠ Cloud Computing
- Big Data is not a new computing problem. Cloud Computing opens up new approaches in tackling Big Data
- While Cloud Computing has many benefits, it only plays a part in the overall Big Data architecture
  - Process, Information Model, Technologies, etc.
- Apply Cloud Computing where it make senses
  - Data-Intensive Science, Cost reduction, Service reliability, etc.
- Part of the architectural design involves modernizing existing software solutions in order to
  - Truly leverage the elasticity of the Cloud
- It makes sense to bring the computing close to the data - On-premise Cloud (currently)
  - Need local experts
  - Governance
- Use automation deployment – Puppet, Chef, Salt
- Many technologies are mature in their standalone context
  - It doesn’t mean they are high TRL when integrated into our domain-specific architecture
- Look into Open Source Solutions before build your own

Further Information

- **NEXUS is available through open source – Apache License 2.0**
  - https://github.com/dataplumber/nexus

- **NEXUS at AGU 2016 Fall Meeting**
  - NASA Sea Level Change Portal – Boening, C. – Demonstration at NASA Booth
  - Session IN12A/B: Big Data Analytics – Huang, T., Lynnes, C., Vance, T., and Yang, C.
  - NEXUS-released Abstracts:
    - Gill, K., et.al, 2016: "Analysis of Sea Level Rise in Action"
    - Greguska, F, et.al, 2016: "Tackling the Four V's with NEXUS"
    - Jacob, J., et.al, 2016: "Performance Comparison of Big Data Analytics With NEXUS and Giovanni"
    - Lynnes, C., et.al, 2016: "Benchmark Comparison of Cloud Analytics Methods Applied to Earth Observations"
    - Quach, N., et.al, 2016: "Sea Level Rise Data Discovery"
    - Wilson, B., et.al, 2016: "OceanXtremes: Scalable Anomaly Detection in Oceanographic Time-Series"

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