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A Linked Science Investigation: Enhancing Climate Change Data Discovery with Semantic Technologies

Line C. Pouchard,

Oak Ridge National Laboratory, 1 Bethel Valley Road, Oak Ridge, TN 37831

Marcia L. Branstetter,

Oak Ridge National Laboratory, 1 Bethel Valley Road, Oak Ridge, TN 37831

Robert B. Cook,

Oak Ridge National Laboratory, 1 Bethel Valley Road, Oak Ridge, TN 37831

Ranjeet Devarakonda,

Oak Ridge National Laboratory, 1 Bethel Valley Road, Oak Ridge, TN 37831

Jim Green,

Oak Ridge National Laboratory, 1 Bethel Valley Road, Oak Ridge, TN 37831

Giri Palanisamy,

Oak Ridge National Laboratory, 1 Bethel Valley Road, Oak Ridge, TN 37831

Paul Alexander, and

Stanford Center for Biomedical Informatics Research, Stanford University, Stanford, CA 94305

Natalya F. Noy

Stanford Center for Biomedical Informatics Research, Stanford University, Stanford, CA 94305

Abstract

Linked Science is the practice of inter-connecting scientific assets by publishing, sharing and linking scientific data and processes in end-to-end loosely coupled workflows that allow the sharing and re-use of scientific data. Much of this data does not live in the cloud or on the Web, but rather in multi-institutional data centers that provide tools and add value through quality assurance, validation, curation, dissemination, and analysis of the data. In this paper, we make the case for the use of scientific scenarios in Linked Science. We propose a scenario in river-channel transport that requires biogeochemical experimental data and global climate-simulation model data from many sources. We focus on the use of ontologies—formal machine-readable descriptions of the domain—to facilitate search and discovery of this data. Mercury, developed at Oak Ridge National Laboratory, is a tool for distributed metadata harvesting, search and retrieval. Mercury currently provides uniform access to more than 100,000 metadata records; 30,000 scientists use it each month. We augmented search in Mercury with ontologies, such as the ontologies in the Semantic Web for Earth and Environmental Terminology (SWEET) collection by prototyping a component that provides access to the ontology terms from Mercury. We evaluate the coverage of SWEET for the ORNL Distributed Active Archive Center (ORNL DAAC).

Corresponding author: Line C. Pouchard, pouchardlc@ornl.gov, Telephone: 865-574-6125, Fax: 865-241-4811.

Availability and Requirements

NCBO BioPortal software is available at: <http://www.github.com/ncbo>

Authors refer to (Devarakonda et al., 2010) for Mercury.

Keywords

Linked Science; ontologies; BioPortal; semantic search; climate change; data discovery

1. Linked Science: requirements and examples

The ways in which scientists conduct research in earth sciences, chemistry, biology, geography, ecology, sociology, and other scientific fields is changing significantly. Often, the most challenging research questions require them to understand and use practices, data, methods and software from many scientific disciplines. *Linked Science* is the practice of inter-connecting scientific assets by publishing, sharing and linking scientific data and processes in end-to-end loosely coupled workflows that allow the sharing and re-use of scientific data (Kauppinen and de Espindola, 2011, Kauppinen et al., 2011, Kauppinen et al., 2012). Linked Science requires new ways of integrating and aggregating structured and unstructured data and information derived from physical, chemical, biological, sociological, and other traditional fields of scientific study. Linked Science is grounded in interdisciplinary research and highlights the reasons why such research is arduous: a scientist who is already an expert in a domain must become fluent in the language and practices of another domain in order to start addressing a scientific question. Some examples of Linked Science, such as DataONE, put a special emphasis on working from scientific scenarios.

Data Observation Network for Earth (DataONE) is the foundation of new innovative environmental science through a distributed framework and sustainable cyber-infrastructure that meets the needs of science and society for open, persistent, robust, and secure access to well-described and easily discovered Earth observational data (Michener et al., 2012). The goal of DataONE is to ensure the preservation, access, use and reuse of multi-scale, multi-discipline, and multi-national science data via cyber-infrastructure elements and a broad education and outreach program. DataONE researchers have developed a number of scientific scenarios that require such multi-disciplinary integration of data. In DataONE's initial efforts at using data to address scientific research projects, scientists used bird observations and a variety of environmental data layers to estimate changes in the occurrence of bird species seasonally in the conterminous U.S. (Fink et al., 2010, Kelling et al., In press).

Gil and colleagues (Gil et al., 2012) propose another Linked Science example and developed a scenario focused on understanding the carbon cycle in water that requires integrating data and analyses by scientists studying river, lake, ocean, and coastal ecosystems. Here, a semantic framework allows collective metadata editing and acknowledgement of scientists' contributions around the scenarios. The Semantic Water Quality Portal (Patton et al., 2011), also an example of Linked Science, integrates domain data related to water quality from several agencies, including the US Geological Survey (USGS), the Environmental Protection Agency (EPA) and multiple regulation ontologies. Ontologies enable detection of pollution events and communities can monitor pollution results according to a regulation of their choice.

Linked Science relies upon the collection, organization, classification, storage, discovery, access, transport, distribution, sub-setting, aggregation, dissemination, and visualization of large, diverse types of data. Scientists store and disseminate data through archives and data centers, supported by organizations in government, academia, and industry. These data centers guarantee data quality and reproducible transformation of data through processes so that the credibility of scientific results is preserved. This is essential in the study of climate

change, where results influence national and international policy. Data centers include experimental, observational, and computer-generated data. They provide tools and add value through quality assurance, validation, curation, dissemination, and analysis of the data. These data typically cannot be consumed by a browser, an audio or video reader, and usually require specialized applications that these data centers also provide.

Problem Statement

Discovery and access to this data poses a major challenge, one that we describe below. On the Web, with Linked Open Data, every resource has a unique identifier. By contrast, with Linked Science, uniform access to datasets provided by unique identifiers is not always available because the data does not live in the cloud or on the Web, but in multi-institutional data centers. Uniform data access would be advantageous to searches and for discovering new links between datasets and/or scientists. Globally unique identifiers, a unique reference number used as an identifier in software and on the web, like a Uniform Resource Identifier, are gaining some traction but not universally used. Common schemes like the Digital Object Identifiers system, a unique reference ISO standard, are often linked to publications rather than datasets. Critically, each dataset must be accompanied by discovery metadata to enable access. Metadata must specify which services or software can consume the data and where they are offered to allow automation of processes. In addition, metadata must describe the way that the data was generated, potential errors, and uncertainty or variability in the calculations and measurements.

Thus, not only must we have a vocabulary to describe this extensive metadata, but also this vocabulary needs to be shared among multiple data providers and we must be able to perform automated reasoning in order to discover, access, and integrate data described by this metadata. We investigate the use of *formal ontologies* as such metadata vocabularies. Ontologies help with data access as they can provide additional keywords for a search. The Semantic Web for Earth and Environmental Terminology (SWEET) (Raskin and Pan, 2005) is a mature foundational ontology that can provide such terms. For example a search on “carbon” in SWEET returns “Carbon sequestration,” “carbon footprint,” “Dissolved organic carbon concentration protocol” and “kartz,” that all can be used for new searches. Ontologies define the terms in a domain of discourse (shared metadata terms in our case), provide constraints on the values and define formal semantics that enable automated reasoning. Furthermore, the World-Wide Web Consortium (W3C) has defined OWL, a formal language for representing and sharing ontologies on the Web, enabling scientists to publish and integrate metadata using standard Web protocols. One can think of an ontology as a taxonomy of terms with added rules and relationships that can be used by computer algorithms. Ontologies and semantic descriptions of the scientific data and processes provide the necessary objects supporting the production of new knowledge by allowing interoperability of the processes, shared annotations and integration of the data.

Organization

This paper makes the following contributions:

- We propose a scientific scenario in climate change that illustrates Linked Science.
- We describe the motivation for semantic annotations within this scenario.
- We present a tool that uses ontologies to improve the quality of search across heterogeneous big-data resources.
- We describe our semantic work in the context of Linked Science

- We analyze the domain coverage that SWEET provides for a large collection of datasets.

The paper is organized as follows. The next section describes related work. In Section 3, we describe the domain application and provide a scientific scenario where heterogeneous data sources must be collected to perform a scientific investigation of climate change for river water transport. We characterize the various types of datasets available in this domain. In section 4 we focus on the design and implementation of our system and the integration of several open-source components to improve data discovery using semantics. We also describe a prototype tool highlighting the use of ontologies developed for this scenario. In Section 5, we present the results that we obtained with the prototype and we evaluate the ontology coverage. In the final section we analyze our results and then present our conclusion and future directions.

2. Related work

Researchers in the Semantic Web community have studied *semantic search* and a variety of approaches to it. A recent survey (Tran et al., 2011) presented a general model for semantic search and identified different types of semantic search. In general, there are two key approaches. In one, the (linked) data is represented in RDF or OWL and the search engine provides access to a collection of such data, either through keyword search or through SPARQL (e.g., SWSE (Hogan et al., 2011), or Sindice (Tummarello et al., 2007)). Uren and colleagues provide a survey of this type of semantic-search engines (Uren et al., 2007).

The second class of semantic-search applications are document-retrieval applications that use semantics to expand or constrain the user query (e.g., (Chu-Carroll et al., 2006, Castells et al., 2007)). The query expansion method (Navigli and Velardi, 2003) uses ontology terms related to terms of the original query as additional search terms. Related ontology terms can specify synonyms, sub- or super-classes for query terms, thus providing additional search terms that are not linguistically related to query terms. Semantic technologies, such as ontologies, help improve search results by adding these additional search terms and thus potentially increasing the number of returned results.

The application that we describe here is closer to the second category as we use ontologies for query expansion using superclasses and sub-classes. It provides access to heterogeneous collections of structured data, but this data is not represented in Semantic Web formats, thus does not fall into the first category. At the same time, it uses semantics on the “front-end,” augmenting the user query, but we use this query expansion to access structured data and not a set of documents. Thus, to the best of our knowledge, the application that we describe is unlike many semantic-search applications because it uses semantics on the query side, provides access to structured data, but not in RDF and OWL format.

Kauppinen and colleagues frame the challenges of Linked Science in the form of an “executable paper” (Kauppinen and de Espindola, 2011), with publication of validated and well-sourced data as one of the key requirements. Contributions to the First and Second Linked Science workshops¹ (Kauppinen et al., 2011) investigated several issues related to Linked Science and Linked Data but did not focus on semantic searches for structured datasets in dedicated archives. Researchers discussed: the requirements for Linked Science in the geo-physical sciences (Mäs et al., 2011); the use of rules for interactively mapping data sources in databases to ontology and generating RDF triples (Knoblock et al., 2011); the need for trust in the data sources with an emphasis on formally describing the

¹ <http://linkedsience.org/events/lisc2011/>
<http://linkedsience.org/events/lisc2012/>

relationship between data and sources in bibliographic resources (McCusker et al., 2011); and challenges in the bioinformatics (Vision et al., 2011) and astronomy domains. Thus, researchers are actively addressing the trends in Linked Science. Our effort describes our semantic work in the context of Linked Science and is complementary to the approaches described in these papers.

3. Use case scenario and datasets

Consider the following scenario (Fig. 1). A hydrologist focuses on validating model simulation multi-decadal trends for nutrient transport in a river channel within a watershed. In this case we are considering the Community Climate System Model (Gent et al., 2011). This climate model simulation of the earth system used to investigate climate change has four components: land, sea ice, ocean and atmosphere. The land component currently includes simulations of river flow. Future models of the earth system will contain biogeochemical species such as nitrogen, carbon, and phosphorus compounds (e.g., those contained in fertilizers). Changes in the chemistry of rivers from two different scenarios are particularly relevant to climate change. First, biogeochemical species resulting from fertilizer use are washed from the soil, carried from water streams into larger rivers, and eventually end up in coastal oceans. Second, deforestation from biomass burning also causes changes to the chemical composition of the water that flows into rivers. The transport of biogeochemical species, particularly riverine nitrogen, may have an even larger effect: these species cause hypoxia (reduction in the oxygen concentration in water) and fish mortality in the coastal oceans (Doney, 2010). In order to characterize these effects realistically, the hydrologist will need access to two types of data, which are generally available to earth scientists: (1) *computational* data that record the results of computer modeling and simulation; and (2) *observational* data that contain results of specific measurements.

Getting the data

In our use case, the computational data will include models of river flow and transport of biogeochemical species; the observational data will describe stream flow, water quality, precipitation, air and water temperature, sediment data, biogeochemical species, and soil moisture. For computational-model data, our hydrologist can turn to the Earth System Grid Federation (ESGF²) gateway at the National Center for Atmospheric Research (Bernholdt et al., 2005). At the time of this writing, it contains 3,384 datasets of computational data totaling about 1.3 Petabytes of data and representing 368 variables. She will need to know, however, that file names in this source attempt to reflect variable name, such as “qchanr” (river flow), or “qchocnr” (river discharge into the ocean).

For observational data, the hydrologist can get data from the Gravity Recovery and Climate Mission (Tapley et al., 2004) and the Tropical Rainfall Measuring Mission (Theon, 1993) from the National Aeronautics and Space Administration (NASA) to validate the outputs of the climate model simulation. These datasets contain remote sensing imagery for tropical precipitation and storage. Ground stream flow data is available from the USGS. Fertilizer input and water-quality measurements may come from the EPA and the US Department of Agriculture.

Biogeochemical data is available to the hydrologist from the NASA-sponsored Distributed Active Archive Center at the Oak Ridge National Laboratory³ (ORNL DAAC). This center holds about 1,000 datasets amounting to 2 Terabytes relevant to biogeochemical dynamics, ecological data, and environmental processes, as well as 60 TB of land product data subsets

² <http://www.earthsystemgrid.org/>

³ <http://daac.ornl.gov/>

(measurements of surface radiance, reflectance, emissivity, and temperature) from the Moderate Resolution Imaging Spectroradiometer (MODIS) Instrument aboard the Terra and Aqua satellites.

A scientific user may typically be familiar with computational climate datasets, such as those found in ESGF, or with observational earth and ecological science datasets such as those found in the ORNL DAAC, but not both. Both types of sources currently present their data in faceted searches along attributes such as Project, Model, Experiment, Product, Variable Name, and Ensemble for ESG, and Parameter, Sensor, Topic, Project, Keywords in the ORNL DAAC. A faceted search exposes different views of a search result set along attributes contained in structured metadata. Note that in computational data the facet “Experiment” denotes experiments “in silico.” In the observational data, one also finds “Models,” a term typically reserved for simulations, where datasets are used in assessments and policy studies and simulate ecological systems: observational data can also be the result of simulations.

Thus, data solutions to the scientific question require the use of heterogeneous data. The hydrologist will need to search for datasets from different data centers to discover useful data. Each data domain has its own portal, its own metadata formats, and its own query-building methods for obtaining datasets. The exact definition of variables and observational parameters may require substantial searches for unfamiliar topics. In order to advance investigation of climate change, scientists need access to formal descriptions of the multiple objects present in each activity and to tools that permit seamless searches across all types of data. The next section presents tools that enable heterogeneous data access and improve searches.

4. Design and implementation

To enable data access and to facilitate searches, we integrated several existing technologies and developed an added module for query expansion using Earth Sciences domain ontologies. Our system is composed of the following components, which we describe in detail in the rest of this section:

- The National Center for Biomedical Ontology (NCBO) BioPortal ontology repository.
- The Mercury search engine
- Our added module providing programmatic access from Mercury to ontology terms stored in our BioPortal instance.

A key factor in integrating these components was the existence of open source APIs.

4.1. The NCBO BioPortal and its Virtual Appliances

The NCBO BioPortal is a community-based ontology repository (Musen et al., 2011, Whetzel et al., 2011). BioPortal allows users to browse ontologies and to search for specific ontologies that have terms that are relevant for their work. The mappings between ontologies not only allow users to compare the use of related terms in different ontologies, but also allow analysis of how whole ontologies compare with one another. BioPortal provides access to the ontologies through a REST interface, thus enabling easy integration with Mercury. While the instance of BioPortal that runs at NCBO is a repository of biomedical ontologies—with more than 300 of them at the time of this writing—the BioPortal software is domain-independent.

For the communities that want to run their own ontology repositories using the BioPortal code base, the NCBO team generates a *Virtual Appliance (VA)*—a packaged copy of the web-server software that other communities can install and maintain. These communities use the repository to share and access ontologies that are relevant to their domain. The Earth Science Information Partners Federation (ESIP) Semantic Portal⁴ deployed such a VA on the Amazon EC2 cloud node procured by ESIP (Pouchard et al 2012). Fig. 2 shows ontologies uploaded in this repository. The NCBO BioPortal at ORNL DAAC is another such installation behind a firewall. It contains the same version of the SWEET ontology as the ESIP ontology portal. We used the ORNL DAAC instance in our added module for ease of implementation and evaluation.

4.2. The Mercury tool: aggregating metadata

Mercury is a tool for distributed metadata harvesting, search, and retrieval originally developed for NASA. Mercury is currently used by projects funded by NASA, USGS, and U.S. Department of Energy (DOE) (Devarakonda et al., 2010). More than 30,000 scientists use Mercury each month. Mercury provides a single portal to search quickly for data and information contained in disparate data-management systems. It collects metadata and key data from contributing project servers distributed around the world and builds a centralized index. Mercury allows data providers to advertise the availability of their data and maintain complete control and ownership of that data. Fig. 3 shows a diagram of the Mercury architecture including our added BioPortal module.

Mercury currently provides access to over 100,000 metadata records. It supports several widely used metadata standards and protocols such as the Federal Geographic Data Committee, Dublin Core, Darwin Core, the Ecological Metadata Language, the International Standards Organization's ISO-19115, XML, Library of Congress protocols Z39.50⁵ and Search/Retrieve via URL⁶, and Amazon subsidiary A9's OpenSearch⁷.

The Mercury architecture includes a harvester, an indexing tool, and a user interface. Mercury's harvester typically harvests metadata records from publicly available external servers. Data providers and principal investigators create metadata for their datasets and place these metadata in a publicly accessible place such as a web directory or FTP directory. Mercury then harvests these metadata, builds the centralized index, and makes it available for the Mercury search user interface. Mercury also harvests metadata records from external catalogs using the Open Archives Initiative Protocol for Metadata Harvest (OAI-PMH) (Devarakonda et al., 2011) and other web-based harvesting techniques.

The Mercury search interfaces allow users to perform simple, attribute-based, spatial and temporal searches across these metadata sources. The Mercury repository of metadata for distributed data sources provides low latency search results to the user. For instance a full-text search of 70,000 XML documents returned 48 records in 90 milliseconds; a fielded search of the same collection returned 7 documents in 122 milliseconds (Devarakonda et al., 2010).

Mercury's query engine is built using a service-oriented architecture, which includes a rich user interface. This interface allows users to perform various types of search capabilities, including 1) simple search, which performs a full text search, 2) advanced search, which allows users to search against controlled-vocabulary keywords, time period, spatial extent

⁴ <http://semanticportal.esipfed.org/>

⁵ <http://old.cni.org/pub/NISO/docs/z39.50-brochure/>

⁶ <http://www.loc.gov/standards/sru/>

⁷ <http://www.opensearch.org/>

and data provider information, and 3) web browser tree search, which enables a drill-down through the metadata facets using a hierarchical keyword tree.

4.3. Adding semantics and integrating components

With the breadth of sciences represented within the Mercury metadata records, scientists can address some key interdisciplinary scientific challenges related to climate change and its environmental and ecological impacts, such as carbon sequestration and mitigation. However, the wealth of data and metadata also makes it difficult to pinpoint the datasets that are relevant to particular scientific inquiries.

We have applied semantic technologies—ontologies, in particular—to improve search results. There are several reasons for using this approach. First, simply using popularity determined by pointing links to provide a high ranking to a search result, as with Google, typically is not useful in the case of scientific data queries. Each scientific inquiry tends to be unique, and datasets are not directly indexed so that result ranking may not be useful. Thus, we must be able to improve search results based on the *meaning* of the data descriptions. Ontologies represent such *meaning* in a machine-readable way. Second, scientific queries are unlike everyday queries because they return specific datasets, which themselves have numerous parameters that may or may not be exposed to a general search engine. For example, the Earth System Grid Federation (ESGF) gateway exposes 368 variables to search. Third, each domain science has its own terminology, more or less curated and consensual, and with various degrees of standardization. The same term may refer to different linguistic or scientific objects across domains (semantic plurality), and different terms mean the same thing (synonymy). For all these reasons, we decided to use scientific ontologies because they can provide a context for search results, in a way that keywords never will.

The SWEET ontologies are currently under the ESIP governance and can be accessed from the ESIP Semantic Portal. SWEET 2.3 currently contains over 4,500 terms organized in 200 OWL ontologies classifying 9 top-level classes. For SWEET 2.3 these top-level classes are:

- Representation (math, space, science, time, data)
- Realm (ocean, land surface, terrestrial hydrosphere, atmosphere, heliosphere, cryosphere, geosphere)
- Phenomena (macro-scale ecological and physical)
- Processes (micro-scale physical, biological, chemical, and mathematical)
- Matter (living thing, material thing, material thing)
- Human Activities (decision, commerce, jurisdiction, environmental, research),
- Property (binary, categorical, ordinal, quantity)
- Role (biological, chemical, geographic, impact, representative, trust),
- Relation (human, physical, space, time, chemical).

We used the SWEET ontologies to improve the results of the Mercury search interface. The ontologies provide context by linking individual keywords to a scientific realm and suggest additional keywords for searches. We designed an ontology service that allows integration of ontology terms into search results. The Mercury search system passes its search terms to the VA, which returns one or several matching terms through the REST interface. The user can choose any of these as additional search term for Mercury, or directly display the results indicated by the ontology sub-class terms. For example, an ontology-based search on

biomass returns the keywords “biomass” and “litter” because litter is a sub-class of biomass in SWEET.

5. Results

The ontology service provides domain context, parameter attribute, and entity annotations to the Mercury search system. Mercury user interface uses a faceted search approach; we present the ontology results to the user in the same user interface (Fig. 4). The search term used in this figure is “biomass.” The five top boxes (“Filter by”) show the faceted search results without semantic search. The bottom three boxes (“Ontology”) present the results of the semantic search.

Unlike a faceted search that highlights attributes within a set of results but cannot enlarge the set, the semantic solution can implement both restrictions and expansions of the initial set of results. In our semantic search, there are four new dimensions enabled by ontologies: Ontology Concepts, Ontology Super-classes, Ontology Sub-classes and Filter by keywords and all sub-classes. *Ontology concepts* present each search term within the ontological hierarchy. *Ontology Super-classes* shows the hierarchical level one level up and *Ontology sub-classes*—one level down. To display the facet “Filter by keywords and all sub-classes” the ontology service sends the subclass terms to Solr, which returns links to datasets of interest (not shown in the figure).

5.1. Using ontologies to improve context

Recall the scenario that we described in Section 2. Our hydrologist will need to search for datasets annotated with “biomass” because she wants to analyze the transport of biochemical species in the river flow. She will search for datasets containing the term “biomass.” A Mercury search using controlled vocabulary keywords returns 35 datasets, a full-text search returns 187 datasets. A search for “biomass OR humus” (a type of biomass) returns 192 datasets, indicating that 5 potentially relevant datasets are not included in the search on biomass. Querying the SWEET ontologies through BioPortal’s REST API, the ontology service exposes “humus” as an additional search term for Mercury in the first discovery session about “biomass.” Humus is a sub-class of biomass in SWEET. Thus, the semantic search returns the five additional datasets without the user having to know about specific types of biomass. “Biomass” also acquires scientific context when the ontology service exposes that it can be a form of Energy Storage and a Living Entity. These examples demonstrate how ontologies help expand the search and provide scientific context for the search terms.

5.2. Using ontologies to reduce the number of search results

“Carbon” is another popular search term in Mercury, since the increase in the concentration of carbon dioxide in the atmosphere is considered a potential factor of climate change. A Mercury search for “carbon” returns 264 datasets from the ORNL DAAC. With the ontology service integrating the results of an ontology search into the faceted search, “carbon” acquires a scientific context and additional query terms that can be used to reduce the scope of the original search. For example, the individual in one of the ontologies, “stateTimeGeologic2:Carboniferous,” links results to datasets relevant to geological times (paleo-environmental science), while the subclass “carbon offset” links to datasets relevant to “human environmental control” and “human activity.” In addition, “offset” is not a facet offered by the Mercury search system but the ontology search suggests this sub-class to reduce the result set further. Limiting the search to both “carbon” and “offset” produces only two results.

While ontologies provide additional search terms, we also use the ontology structure to enable the user to filter the results in a meaningful way.

5.3. Analyzing the coverage of ontologies

In addition to SWEET 2.3, the ESIP ontology collection includes the Plant Ontology, which describes structure and developmental stages of a plant (Bruskiewich et al., 2002), and the Extensible Observational Ontology (OBOE) for representing scientific observations and measurements (Madin et al., 2007), and its extension to represent ecological and environmental data.

Using simple term matching, we evaluated how well the terms in the SWEET ontologies cover the top 100 controlled-vocabulary keywords that were used for indexing datasets in Mercury. “Biomass” is the top keyword currently indexing 138 datasets. Fig. 5 shows the results of this evaluation. 21 of the top 100 keywords do not appear in the ontologies. Thus, 79% of the top 100 keywords in Mercury have at least one match in the selected ontologies. At the long tail of the distribution one keyword (water) has 38 matches, and two (air, carbon) have 28 matches. A fifth of the Mercury keywords do not appear in the SWEET ontologies. Thus, the scientific community needs to develop additional ontologies to enhance the keyword collection adequately.

6. Discussion

Our approach to the investigation of climate change has led to the integration of search capabilities and the development of a semantic service for discovering multi-disciplinary datasets in Earth and Environmental sciences. Scientists can use our semantic service to discover new datasets that were not included in the original search results, thus expanding the original queries.

We used a BioPortal instance as a source for ontologies rather than a triple-store or an OWL API to process the ontologies for several reasons. First, the REST service interface that BioPortal provided was easy to integrate into the Mercury architecture. Second, ontology authors sometimes use idiosyncratic approaches to representing some features of their ontologies, such as preferred names or synonyms for terms. These lexical features are key to user searches but ontologies use different properties to represent them. BioPortal uses ontology metadata to extract these properties and provides its users with a single service call to access this information across all ontologies in a repository. Finally, BioPortal enables scientists to submit new ontologies through its web interface and these ontologies become available to the semantic search in Mercury. Thus, if a scientist discovers a new ontology that covers her domain of interest, she can add it to her set of ontologies to expand the meaningful results from her semantic search.

We set up the ESIP instance of BioPortal because this user community needs a stable ontology repository that covers the Earth and Environmental Sciences domains. This instance of BioPortal is accessible to users with all the functionality provided by BioPortal, including annotations, ontology extensions, and term mappings. New community additions to the ontologies made through this instance are directly accessible to the semantic service. ESIP is currently discussing curation mechanisms for ontologies in the ESIP portal.

However, our approach has several limitations. First, the faceted display becomes crowded very quickly and a more dynamic presentation of search results may be beneficial. Another, more serious, limitation is that the quality of the newly discovered metadata is contingent on the quality of the ontologies used in our implementation. BioPortal curates the ontologies by enforcing compliance with ontology language standards and resolving relationships and

axioms to detect potential conflicts, but it cannot check for coverage or correctness in terms of domain expertise. Search terms and thesaurus keywords in Mercury may be absent from current ontologies, or the ontology classification may not bring additional information that is not already presented by the faceted terms. However, as semantic technologies mature, more substantial ontologies become available in many scientific domains.

7. Conclusion and Future Directions

We have made several initial steps in order to address the limitation on coverage and quality of the ontologies. We will use the features that are currently available in BioPortal to solicit feedback and to provide additional information about the ontologies. Specifically, BioPortal includes a comment field for each ontology term that users can edit. ESIP members can take advantage of this feature to resolve conflicts and to propose new terms. Second, the ESIP Semantic Web portal team is currently working on manual evaluation of the coverage of ontologies. The team plans to submit proposals and annotations of terms to the ESIP community for approval. Finally we plan to use the mapping function of BioPortal for creating mappings between terms in the ontologies, thus helping to extend coverage.

The solution that we presented in this paper leverages the federated search capabilities in Mercury that collect metadata records from several scientific domains, and the storage, access and curation functionality of BioPortal. Our solution provides guidance on how to leverage semantic capabilities for improving search results. Use of ontologies—even lightweight ones—provides a path for helping domain experts find the information that they need from heterogeneous datasets. We demonstrated that the tools that are already available today enabled us, with minimal additional effort, to build on two mature systems and to find relevant datasets for interdisciplinary inquiries. The paper thus indicates a direction for linking environmental, ecological and biological sciences.

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List of Abbreviations

API	Application Programming Interface
EC2	Amazon's Elastic Compute Cloud
EPA	Environmental Protection Agency
ESGF	Earth System Grid Federation
ESIP	Earth Science Information Partners
FTP	File Transfer Protocol
ISO	International Standards Organization
MODIS	Moderate Resolution Imaging Spectroradiometer
NASA	National Aeronautics and Space Administration
NCBO	National Center for Bio-medical Ontologies
OAI-PMH	Open Archives Initiatives-Protocol for Metadata Harvesting
ORNL	Oak Ridge National Laboratory
ORNL DAAC	ORNL Distributed Active Archive Center

OWL	(W3C) Web Ontology Language
RDF	(W3C) Resource Description Format
SKOS	(W3C) Simple Knowledge Organization System
SPARQL	W3C Query Language for RDF
SWEET	Semantic Web Earth and Environmental Terminology
SWSE	Semantic Web Search Engine
TB	Terabyte
URL	Uniform Resource Locator
USGS	US Geological Survey
VA	Virtual Appliance
W3C	World Wide Web Consortium
XML	(W3C) Extensible Markup Language

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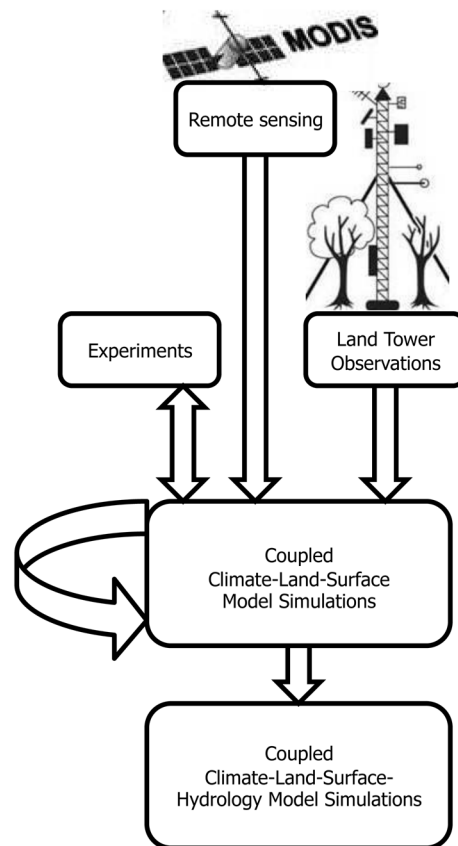
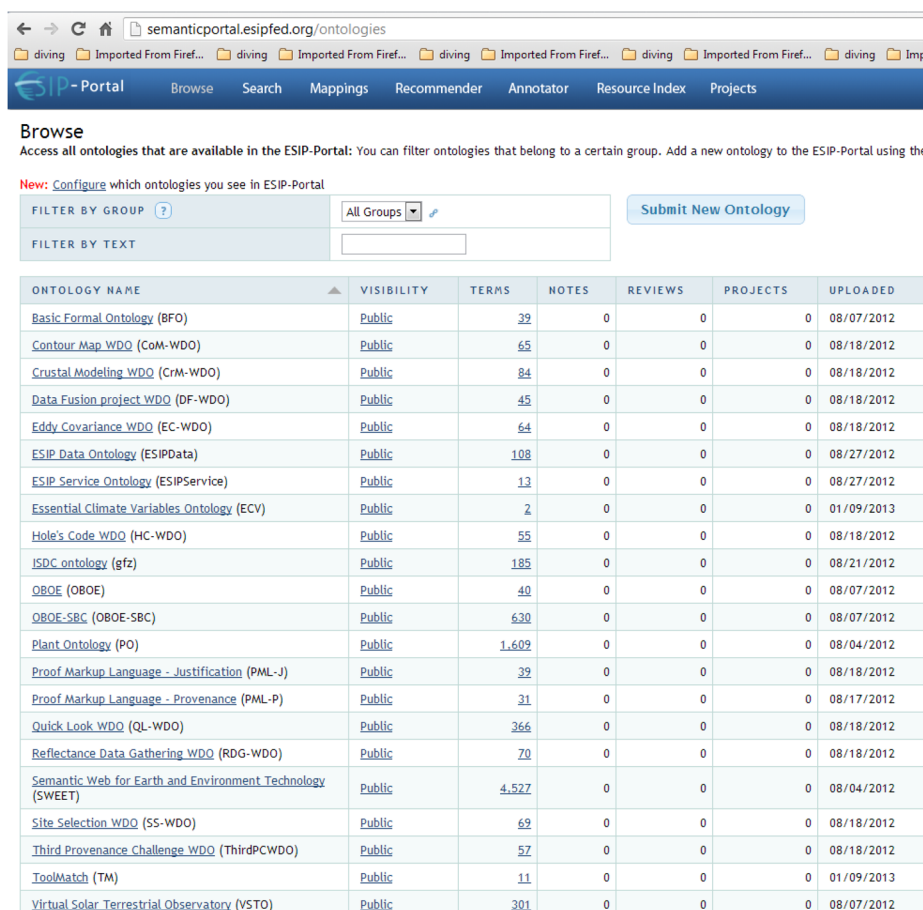


Fig. 1.
Different types of data in our river channel transport use case scenario



ONTOLOGY NAME	VISIBILITY	TERMS	NOTES	REVIEWS	PROJECTS	UPLOADED
Basic Formal Ontology (BFO)	Public	39	0	0	0	08/07/2012
Contour Map WDO (CoM-WDO)	Public	65	0	0	0	08/18/2012
Crustal Modeling WDO (CrM-WDO)	Public	84	0	0	0	08/18/2012
Data Fusion project WDO (DF-WDO)	Public	45	0	0	0	08/18/2012
Eddy Covariance WDO (EC-WDO)	Public	64	0	0	0	08/18/2012
ESIP Data Ontology (ESIPData)	Public	108	0	0	0	08/27/2012
ESIP Service Ontology (ESIPService)	Public	13	0	0	0	08/27/2012
Essential Climate Variables Ontology (ECV)	Public	2	0	0	0	01/09/2013
Hole's Code WDO (HC-WDO)	Public	55	0	0	0	08/18/2012
JSDC ontology (sfz)	Public	185	0	0	0	08/21/2012
OBOE (OBOE)	Public	40	0	0	0	08/07/2012
OBOE-SBC (OBOE-SBC)	Public	630	0	0	0	08/07/2012
Plant Ontology (PO)	Public	1,609	0	0	0	08/04/2012
Proof Markup Language - Justification (PML-J)	Public	39	0	0	0	08/18/2012
Proof Markup Language - Provenance (PML-P)	Public	31	0	0	0	08/17/2012
Quick Look WDO (QL-WDO)	Public	366	0	0	0	08/18/2012
Reflectance Data Gathering WDO (RDG-WDO)	Public	70	0	0	0	08/18/2012
Semantic Web for Earth and Environment Technology (SWEET)	Public	4,527	0	0	0	08/04/2012
Site Selection WDO (SS-WDO)	Public	69	0	0	0	08/18/2012
Third Provenance Challenge WDO (ThirdPCWDO)	Public	57	0	0	0	08/18/2012
ToolMatch (TM)	Public	11	0	0	0	01/09/2013
Virtual Solar Terrestrial Observatory (VSTO)	Public	301	0	0	0	08/07/2012

Fig. 2.

The ESIP Semantic Portal lists ontologies relevant to Earth sciences. It contains the SWEET ontology and many others. Users can browse each ontology hierarchy or search for terms of interest across all ontologies.

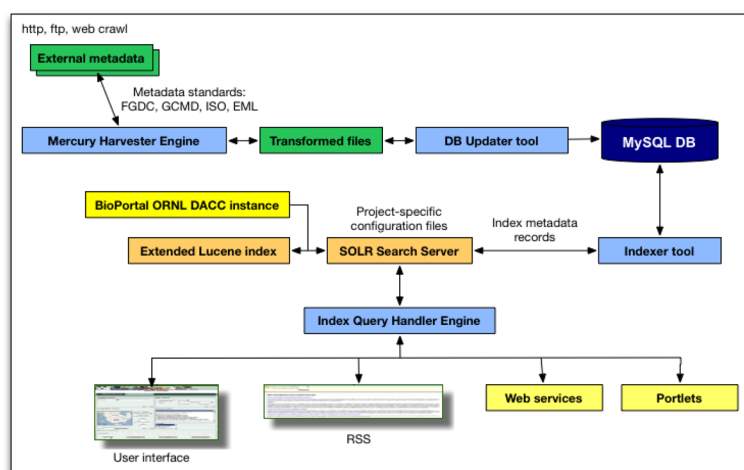


Fig. 3. Architecture of the Mercury Search Engine and its integration with BioPortal ORNL DACC instance. Blue boxes indicate reusable software components. Green boxes are metadata files. Yellow boxes are external services. The Mercury Search service calls BioPortal REST services to add ontology knowledge to the queries

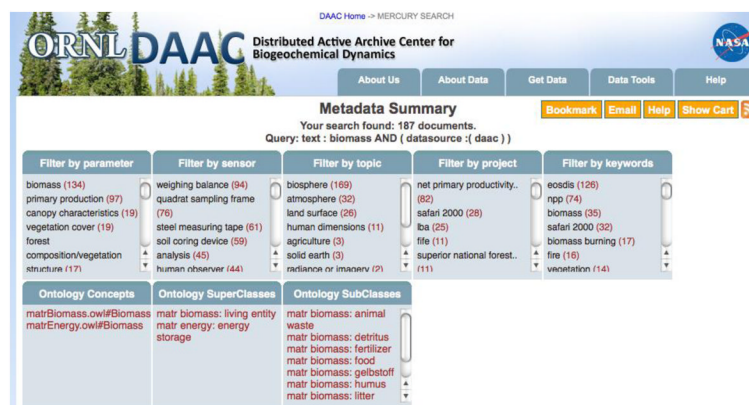


Fig. 4. User interface for the semantic search in ORNL DAAC. The user has searched for “biomass” and the interface suggest additional related terms based on the ontology search. The five tabs named “Filter” display Mercury search results. The three tabs named “Ontology” display the results of the ontology-based search obtained with our prototype

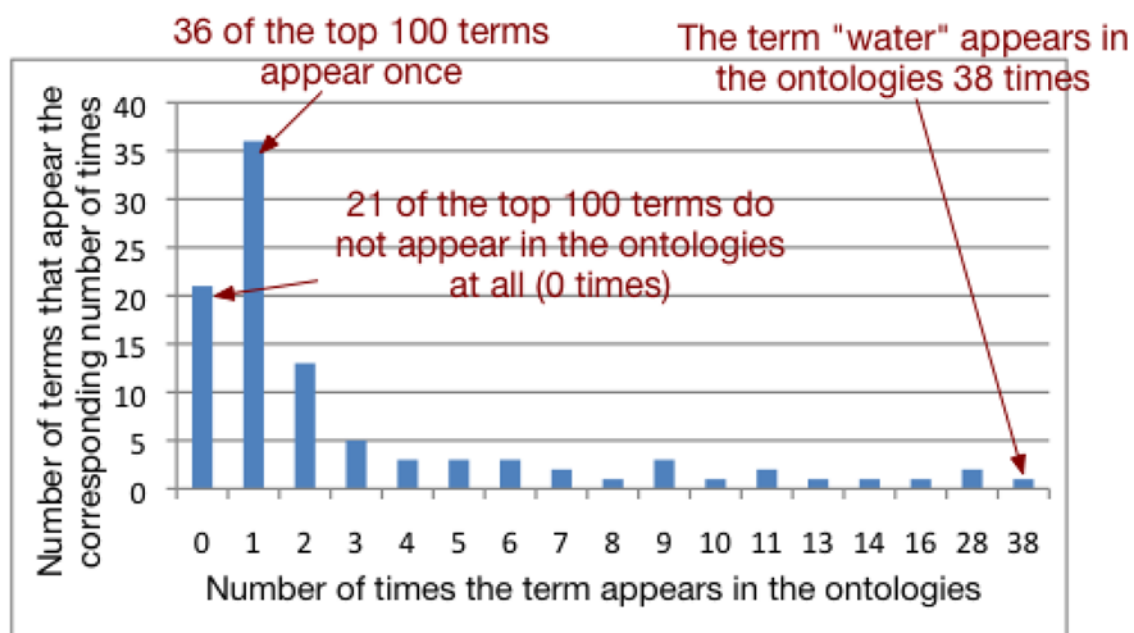


Fig. 5.
Ontology coverage of the top 100 controlled-vocabulary keywords.