

Ontology Use for Semantic e-Science

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Abstract. Ontologies are being widely used in online science activities, or e-Science, mostly in roles related to managing and integrating data resources and workflows. We suggest this use has focused on enabling e-science infrastructures to operate more efficiently, but has had less emphasis on scientific knowledge innovation. A greater focus on online innovation can be achieved through more explicit representation of scientific artifacts such as theories and models, and more online tools to enable scientists to directly generate and test such representations. This should lead to regular use of ontologies by scientists as part of their routine online activity.

Keywords: ontology use, e-Science, semantic web

1. Introduction

Before the onset of scientific computing, the data, methods and theory used for science were often kept close together, in the head and notebook of the researcher. Development of computational infrastructure over the last 50 years has allowed first data and next methods to move far from their scientific creators. Many research communities are now congregating around online infrastructures that contain shared repositories of primarily data and methods. Such infrastructures are being used for discovery, retrieval, and integration of online scientific resources, mainly scientific databases, and increasingly also to capture and describe scientific instruments, software, workflows, and experiments. The infrastructures and associated activities collectively comprise e-Science [6]. The number of e-Science initiatives is vast. Some examples are:

- The Geosciences Network: *GEON* (www.geongrid.org) [10]
- Cancer Biomedical Informatics Grid: *caBIG* (<https://cabig.nci.nih.gov/>) [13]
- Global Ocean Observing Initiative: *GOOS* (<http://www.ioc-goos.org/>) [2]

These and similar efforts are realizing important scientific benefits, which can be largely attributed to three factors: improvements in resource quantity and efficiency, improvements in representation, and improvements in communication:

(1) **Improvements in resource quantity and efficiency** are realized by leveraging and integrating greater numbers of relevant online resources. New results ensue when more and bigger online assets are brought to bear on a problem, for example, such as when distributed computing is running remote applications, often automated and in parallel, over networks of massive databases or sensors. Data that is often expensive to capture or create is then more likely to see secondary use. The same goes for methods and other e-resources.

(2) **Improvements in representation** are realized by recording more complete and complex expressions of scientific knowledge and research activities. More reliable results then accrue from far better levels of repeatability and explanation, because online environments host machine-processable representations of many (ideally all) aspects of scientific investigation, and these can be accessed by greater numbers of scientists.

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(3) **Improvements in communication** are realized by facilitating greater online collaboration between scientists. The enhanced connectivity of online science environments then increases the exchange of ideas.

Ontologies are already playing a pivotal role in these areas. For example, in virtual observatories [8] ontologies are (1) being used to annotate the structure and content of scientific databases and workflows to make them interoperable, (2) helping guide the structure and content of scientific workflow provenance to illuminate scientific reasoning, and more generally [14] they are (3) facilitating scientific discourse by providing content and context for online dialog in virtual communities.

However, these improvements are largely impacting the online use of scientific data and methods, while the surrounding knowledge, the theory, assumptions, reasoning and other context, have largely been left behind. This is highlighted by the position of ontologies in the infrastructures, where they are frequently shuffled to the background. Indeed, ontologies are rarely used directly by scientists, despite the potential for them to help represent the missing knowledge, but they are more often directly used by machines to enable automated components of the infrastructure to work properly. This raises outstanding questions about how effectively ontologies are being used to innovate knowledge from their background position in the infrastructure.

We suggest that ontologies are underutilized in the development of new scientific knowledge in each of the three aspects above. This is largely due to the fact that—for the most part—ontologies are being treated as engineering artifacts required to execute tasks more efficiently, rather than knowledge artifacts that, for example, help to describe some gap in scientific theory or flaw in the reasoning. Indeed, we claim e-Science ontology use is at present largely motivated by operational efficiency, with downstream impacts on scientific knowledge development minimized at present, and significantly below their potential. A contrasting vision prioritizes knowledge innovation in which scientists use ontologies both to express hypotheses, theories and models, and also to generate and test them [4, 17]. In this aspirational vision, scientists use ontologies directly as part of routine scientific investigation because the e-Science environments are designed to do so. Such direct scientist interaction with the ontology-enabled knowledge, i.e. ‘in-silico’ semantic science, should then help revitalize online scientific methodology by helping generate

richer insights, and improving our ability to repeat, report, and validate scientific findings.

2. Quantity and efficiency

The focus on operational efficiency is best exemplified by the quantity and efficiency aspect (described in 1 above), in which significantly more online resources can be marshaled and then applied to some task. This usually involves ontology-enabled semantic interoperability to connect greater volumes of data, software, instruments, and computing resources. The associated ontologies typically consist of application ontologies that describe particular resources, or a slightly more general domain ontology that spans the application ontologies and serves as a unifying conceptualization for the system [15, 20]. However, neither of these ontology types typically encapsulates broad domain knowledge, as each tends to include only those concepts needed to enable the interoperability of specific resources. These ontologies are seldom even seen by scientists and they mainly remain part of black-box components that allow the system to automatically handle greater volumes of resources than could perhaps be handled manually. Knowledge innovation is thus tied to insights gleaned by scientists from greater resource integration—rather than deeper understanding of these resources. Arguably, this often does not involve the application of new online scientific methods, but rather the mirroring of manual methods within the online environment, such that well-established lines of reasoning are carried out by scientists. While this is certainly leading to new scientific results, there remains the possibility that dramatic new insights might be achieved with complementary lines of investigation involved increased use of machine techniques related to forms of learning, reasoning, data mining and so on.

The contrasting vision would thus leverage ontologies and more automated methods, such as knowledge discovery, analogical and abductive reasoning, which are already being applied to scientific discovery [5, 13], to facilitate the proposal of new hypotheses as well as offer mechanisms to test their validity. The role of the scientist is not diminished but the system plays a greater and more direct role in knowledge innovation, as do the ontologies which are regularly updated and extended as part of this process. The ontologies would then constitute a far richer knowledge repository for a domain, and con-

sist of theories, models, methods, and other aspects of scientific work. This is a significant advance on present ontology contents, which primarily contain scientific categories such as ‘granite’, ‘mass’, ‘temperature’, and ‘cancer’.

3. Representation

The representation aspect is best exemplified by the role of ontologies in online scientific provenance [11, 18]. Scientific provenance refers to the historical context surrounding some scientific activity or result, and typically involves a description of the methods and applications used, the processes and reasoning steps carried out by a scientist for some purpose, and the old as well as new states of knowledge and data [19]. It is most widely encountered in established scientific workflow environments, such as myExperiment [7], which orchestrate scientific processing and from which the provenance elements can be readily captured. While traditional best practices would necessarily have such process information recorded manually, online environments allow this to happen transparently by recording each operation as it occurs, and also recording it more finely so that each step can be captured, repeated and questioned.

Ontologies are widely used in provenance systems. They serve as common conceptualizations in the query interface for viewing and querying provenance, and for semantic interoperability across various data and provenance stores [11]. They are also used to annotate metadata associated with components in a scientific workflow [10, 11], and underpin trust systems that evaluate the quality and reliability of a scientific resource [2]. However, as with the quantity factor, such ontology use primarily has an efficiency imperative: more often than not the ontologies are used to describe low-level system resources such as a web service interface or a specific data product, rather than scientific objectives such as the hypothesis being tested or the scientific method used. The ontologies are thus primarily used to make the provenance system work, but how this affects knowledge generation is left to the scientist to determine. Our vision of provenance extends these notions to include ontologies of scientific method, such that online processing steps can be understood in terms of scientific objectives—such as to verify a result or to evaluate a hypothesis. A particular workflow could thus be described in terms of system operations as well as scientific reasoning steps, so that scientists

could interact with the workflow in terms of scientific goals as well as system mechanics. This necessarily involves a conceptualization of the science knowledge cycle and effective interfaces and functions to operate over it.

4. Communication

The communication factor is best exemplified by online scientific collaboratories in which scientists utilize multi-media and social networking resources to work together on common tasks [16]. The general intent is scientific progress through increased scientific interaction, with a particular focus on augmented and clearer online discourse. Tools to represent discourse are usually coupled to literature repositories or other resources, which provide subject matter for the discourse. Ontologies are used to represent concepts inherent in the discourse and are often realized as annotations to variously structured text, such as online research papers. The emphasis, though, is on the nature of the rhetoric surrounding some knowledge [3, 14] and on the validity of a given line of reasoning typically within a descriptive logic, with less focus on the representation and evolution of higher-order scientific concepts such as theories and models. Again, this can be largely viewed as a gain in efficiency in that more scientists are able to collaborate more often, as well as a gain in knowledge interpretation as scientific statements are critically evaluated and inconsistencies in the reasoning are identified and conceptual gaps are highlighted. Critical evaluation might include the proposal of new hypotheses, but discourse systems on their own do not enable those hypotheses to be tested in a scientific sense, with data and methods; at least not without being coupled to additional resources such as workflows, databases, instruments, and so forth. Our vision would see that coupling take place, such that dynamic hypothesis generation and testing could occur during online scientific discourse, where it could be tracked as well as evaluated for trust and reuse.

5. Conclusions

Our conception of semantic e-science amalgamates the enhanced visions discussed above. It includes semantic repositories of knowledge in which ontologies are a base representation for scientific

concepts, theories, models, methods, and other science knowledge elements. These are coupled to workflow operations driven by scientific objectives and methods, and to scientific provenance described in terms of scientific reasoning steps. Finally, scientific laboratories enable community discourse to occur over any of the previously mentioned components, to evaluate them for quality, trust, veracity and re-usability. In such an online environment scientists would focus on knowledge innovation, in as transparent a way as possible, harnessing both efficiency and innovation objectives for next generation science.

Acknowledgements

We gratefully acknowledge of the support of the Geological Survey of Canada and the Ministry of Research, Science and Technology, New Zealand.

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