Training and hackathon on building biodiversity knowledge graphs

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Abstract

Knowledge graphs have the potential to unite disconnected digitized biodiversity data, and there are a number of efforts underway to build biodiversity knowledge graphs. More generally, the recent popularity of knowledge graphs, driven in part by the advent and success of the Google Knowledge Graph, has breathed life into the ongoing development of semantic web infrastructure and prototypes in the biodiversity informatics community. We describe a one week training event and hackathon that focused on applying three specific knowledge graph technologies – the Neptune graph database; Metaphactory; and Wikidata - to a diverse set of biodiversity use cases.

We give an overview of the training, the projects that were advanced throughout the week, and the critical discussions that emerged. We believe that the main barriers towards adoption of biodiversity knowledge graphs are the lack of understanding of knowledge graphs and the lack of adoption of shared unique identifiers. Furthermore, we believe an important advancement in the outlook of knowledge graph development is the emergence of Wikidata as an identifier broker and as a scoping tool. To remedy the current barriers towards biodiversity knowledge graph development, we recommend continued discussions...
at workshops and at conferences, which we expect to increase awareness and adoption of knowledge graph technologies.

**Keywords**

Biodiversity knowledge graph; linked open data; biodiversity informatics; Neptune; Metaphactory; Wikidata; graph database; RDF; SPARQL

**Date and place**

The workshop was held at Agriculture and Agri-Food Canada's Experimental Farm in Ottawa, Ontario, Canada from Oct. 15 to 19, 2018.

**List of participants**

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Introduction

Natural Science collection digitization, genomic sample sequencing, and field survey data collection have accelerated in pace in recent years, thanks to focused investments and increasing efficiency of workflows, among other factors (Hobern et al. 2013). While this is an auspicious development for the advancement of knowledge of biodiversity on earth, several problems have emerged. First, generation of all of this data has led to a “data deluge,” with more data being generated than can be managed or assessed and not enough bioinformaticians to handle these massive datasets (Bell et al. 2009, Hey and Trefethen 2003). Further, these complementary datasets are often held in private data stores or in disconnected public repositories (so-called “data silos”). The knowledge that could be gleaned by organizing and uniting these resources is immense.

The need for dataset integration has led biodiversity scientists in search of solutions. Here, we describe a workshop that brought together biodiversity data professionals of diverse backgrounds to explore the utility of biodiversity knowledge graphs. Knowledge graphs unite data silos using linked data technologies. On the semantic web, knowledge graphs are constructed using the Resource Description Framework (RDF) - a family of W3C specifications in which graphs are specified by collections of \(<subject, predicate, object>\) triples. The subjects and objects represent nodes in the graph, and the predicates represent directed arcs from subjects into objects. Objects can be either literals or URIs, but subjects and predicates must be URIs. It is this final requirement that glues RDF graphs to other RDF graphs, and which puts the “web” in semantic web.

The purpose of our workshop was to explore the integration of linked, open datasets toward the development of biodiversity knowledge graphs, the feasibility and status of building a single biodiversity knowledge graph resource, and to explore several use cases for the development and application of these graphs. First, we held hands-on sessions where participants learned how to access and utilize the AWS Neptune graph database within the Metaphactory suite of tools (Haase et al. 2019). We did this using the neXtProt dataset from UniProt, a human proteome database (Lane et al. 2011). Next, we learned how to interact with Wikidata using Metaphactory (e.g., how to write SPARQL queries). Peter Haase from Metaphactory was available via conference call to answer all questions the participants had. Finally, workshop participants used what they had learned about Neptune, Wikidata, Metaphactory, SPARQL, and other technologies to advance their own biodiversity informatics projects and explore several use cases for knowledge graph application.
Specific workshop aims

Knowledge graphs

One observation that emerged from our workshop was that there is a lack of understanding of knowledge graph utility for biodiversity science. This knowledge gap was addressed through training sessions before we began exploring tools and our case studies. Additionally, participants of the workshop sought to better understand what problems or use cases are well suited to the use of triplestores. Thus, a central goal of the workshop became to learn about knowledge graphs. We wanted to understand what opportunities building knowledge graphs could bring and how to use modern technologies, like Neptune, Blazegraph and Metaphactory to set one up.

We explored the state of interoperability amongst biodiversity knowledge graphs. If we imagine the hypothetical federation of all biodiversity knowledge graphs, how well connected is the resulting unified graph? A graph is connected if there exists a path between every two nodes. The more paths that exist, the higher the connectivity, and the less siloed the data (see the Glossary for a precise definition of graph connectivity). A key interest of workshop participants was contributing to general purpose biodiversity knowledge graph infrastructure.

Use cases

We sought to accomplish our learning goals by exploring and advancing three separate products:

1. Expanding the scope of Ozymandias, a biodiversity knowledge graph of Australian fauna (Page 2018)
2. Investigating the use of Wikidata as a component of the Integrated Flora of Canada (Pender et al. 2018)
3. Exploring the use of knowledge graphs for heuristic food webs (Compson et al., unpublished data)

This structure allowed us to solicit feedback on our work and advance our projects in an exploratory and collaborative fashion (i.e., through instantaneously implementing various ideas that arose).

Ozymandias

Rod Page has assembled a knowledge graph that incorporates a classification of the Australian fauna (via the Atlas of Living Australia), associated taxonomic names and publications (from the Australian Faunal Directory), augmented by data from BioStor (Page 2011) and the Biodiversity Literature Repository (Egloff et al. 2017). For details of the construction of this knowledge graph ("Ozymandias") see Page 2018. Additionally, a web interface is available at https://ozymandias-demo.herokuapp.com. Ozymandias contains only a subset of the elements included in the biodiversity knowledge graph illustrated in
Page 2016, so for the workshop Page investigated adding additional entities such as GBIF specimen data, GenBank records, and DNA barcodes. Specimens were of particular interest, given that Steve Baskauf has done extensive work on representing Darwin Core occurrence records in RDF (Baskauf and Webb 2016), and having specimen data would expand the set of questions that could be asked using the knowledge graph. For example, Fontaine et al. 2012 reported that the average lag time between the discovery of a specimen representing a new species and the description of that species is 21 years. Having specimen data in Ozymandias would enable us to see whether this pattern also applied to the Australian fauna.

**Integrated Flora of Canada**

The hosts of the workshop at Agriculture and Agri-Food Canada (Joel Sachs, James Macklin, Jocelyn Pender, Beatriz Lujan-Toro) are working on an Integrated Flora of Canada resource, based on their Semantic MediaWiki representation of Flora of North America (Pender et al. 2018, [floranorthamerica.org](http://floranorthamerica.org)). Although desperately needed, a Flora of Canada does not exist. The Integrated Flora of Canada project seeks to replace the laborious process of composing a de novo Canadian Flora by mashing together pre-existing data sources and curating these to produce up-to-date, ontology-driven, dynamic treatments. To achieve this, the team is parsing botanical treatments into machine readable qualitative and quantitative properties that can be queried and exported en masse within the Integrated Flora (Cui 2010). The goals at the workshop for the Integrated Flora of Canada were to explore the use of Wikidata or Wikibase as a technology for the data store and to investigate the integration of existing biodiversity data (i.e., sequence, soil microbiome, habitat data).

**Food webs**

Zacchaeus Compson is leading a project that aims to create trait-based food webs (or heuristic food webs) by combining taxa lists generated from DNA metabarcoding with traits that are found in literature using text mining. Visualizing food webs requires knowledge of taxa and their trophic habits, but sampling limitations and a lack of readily available prior knowledge make this challenging. Recent advances in genomics and computational analysis show promise for overcoming some of these challenges. Using DNA to characterize the composition of communities can be effective because it is non-invasive, sensitive, and provides a standardized detection method that circumvents the need for trained taxonomists. Meanwhile, text-mining provides a method for targeted data-mining, allowing trait data to be gathered for specific taxa across large PDF databases (e.g., Google Scholar, Web of Science). We used a scalable engine for semantic automatic discovery and integration to query online databases for knowledge on benthic macroinvertebrate traits (e.g., trophic links, body size). Previously, we have generated heuristic food webs using taxon lists from matched DNA and morphology samples for the Peace and Athabasca River deltas in the Mackenzie Basin, Alberta, Canada, and have explored how metrics of these food webs vary through space and time. In Compson et al. (2018), we describe both text mining and trait-based food web pipelines that can be used
for rapid construction of heuristic food webs. For this workshop, we explored ways of integrating the pairwise contingency matrix generated from our food web pipeline with information from other existing databases in order to make knowledge graphs for these trait-based food webs.

**Workshop outcomes**

**Knowledge graphs and linked open data**

During the training part of the workshop, we explored knowledge graphs and the graph data model. Graphs differ from the traditional relational data model in several keys ways, which enable knowledge representation and integration. First, graph databases place an emphasis on relationships between entities, which may be real world objects or abstract concepts, stored as structured or unstructured data. This is useful for representing biological heterogeneous and integrated datasets (Yoon et al. 2017), where relationships between nodes may be complex and difficult to query using relational databases. Second, data integration is trivial (in principal) with a graph data model. Relationships between entities are explicitly represented in the “triples” (i.e., RDF data model) rather than in a database schema, and as a result new entities can be added to the graph without need for schema revision. Third, the graph data model enables intuitive visualization of data relationships, which is particularly useful when combining datasets. Metaphactory showcased this with its bubble visualization builder. Fourth, through the use of semantic descriptions and axiomatic knowledge, a graph data model can support automated reasoning. In particular, property chains in SPARQL allow for inferencing when logic is encoded in a database directly. Neptune and Blazegraph are graph database technologies that can be used to build a knowledge graph, and Metaphactory is a platform for knowledge graph management, using either Neptune or Blazegraph as a backend database.

Linked open data (LOD) was also explored at the training. LOD principles are a set of standards, principles for publishing, sharing, and interrelating structured knowledge that help break down data silos and create a “Web of Data”. Following these principles extends the utility of data in a knowledge graph: data can be integrated and reused via semantic queries. Through a series of demos using tools on the Metaphactory platform, Peter Haase demonstrated how to use identifier matching (most ideal scenario) or string matching (least ideal scenario, due to disambiguation problems) to unite existing datasets (e.g., the neXtProt database) with other freely available datasets (e.g., Wikidata). Queries or visualizations could then be constructed and developed that would be difficult or impossible to build with the individual datasets alone. There are an increasing number of externally available datasets to connect to, especially in the life sciences, as can be seen in the LOD cloud (https://lod-cloud.net/).

Following the demonstration of neXtProt and Wikidata integration, our goal was to integrate our use case datasets (i.e., Ozymandias, Integrated Flora of Canada, trait-based food webs) and external datasets (e.g., Wikidata and metagenomic data). Before this could
occur, however, we needed to convert the use case datasets into a graph by conversion to RDF. This required using tools to convert CSV and XML files into RDF so that they could be loaded into the Metaphactory triplestore. Important obstacles to overcome included (1) converting data into RDF, (2) establishing links between our data and other datasets to enhance content, and, importantly, (3) determining shared identifiers. As a result of our experiences, at the workshop and otherwise, we believe that the absence of shared identifiers to make links is the most problematic barrier to adoption of LOD today.

An obvious driver towards realizing a biodiversity knowledge graph is its potential to generate metrics demonstrating the efforts and value of collections and biodiversity scientists. Often overlooked and undervalued, the work of biodiversity scientists is not adequately represented by existing metrics (McDade et al. 2011). Metrics such as the number of specimen citations or the use of collections could be quantified by a biodiversity knowledge graph, and we believe generation of these metrics could motivate investment in biodiversity knowledge graph infrastructure (see Page 2018 for a more thorough discussion).

**Wikidata and Wikibase**

We also explored the utility of Wikidata. The central goal of Wikidata is to allow users to query all knowledge contained within Wikipedia, a significant knowledge base and the sixth most popular website globally (Auer et al. 2007). Wikidata is collaborative, like Wikipedia, and has grown substantially since its launch in 2012. We learned how to interact with Wikidata using the Wikidata Query Service, the Metaphactory Wikidata platform that employs the Wikidata Label Service, and Quick Statements.

Wikidata discussions focused on its utility as an “identity broker”. Wikidata provides a comprehensive list of item identifiers collected collaboratively from across the web. This is a significant development in the world of linked data, because mapping identifiers to identifiers is a laborious task, and the absence of shared identifiers is a major barrier to the adoption of LOD technology. Due to its collaborative nature, Wikidata is also powerful for "property scoping", or reaching consensus on properties for a certain type of object. The limitations of Wikidata for biodiversity data were also discussed, including its inability to distinguish between a name and a "taxon concept".

We also investigated the use of Wikibase, the software that powers Wikidata, for the Integrated Flora of Canada project. In particular, we saw potential in the data attribution capabilities of Wikibase through the use of “qualifiers”, as well as the collaborative functionality. During the “hacking” phase of the workshop, we discovered a blog post outlining the steps to getting a local Wikibase instance up and running (https://medium.com/@thisismattmiller/wikibase-for-research-infrastructure-part-1-d3f640dfad34). This provided us with a starting point for continuing our post-workshop investigations of Wikibase.
Associated technology

This was a hands-on workshop, and a focus was placed on learning new technologies. We learned how to set up a Metaphactory instance using AWS directly, how to write SPARQL queries, how to write and interpret Turtle syntax, and how to construct federated queries out of Neptune using Ephedra. Additionally, we learned how to use Metaphactory's powerful visualization tool, which allows users to build bubble graphs, maps, and various types of charts from SPARQL queries on loaded datasets directly. In parallel, we shared useful tools (e.g., Kitematic for fast Docker container deployment, sloppy.io for Docker hosting, Fuseki as a SPARQL server, W3C validator for RDF XML, BaseX for XML database management, d3sparql.js for building visualizations directly from SPARQL queries).

We discussed the uses of different database technologies (relational, non-relational, graph, etc.). Importantly, we discussed the challenges of data curation, which requires a majority of the effort to build a dataset, and no technological solution will change this. Instead, we asked how we can share data curation efforts so that they never need to be duplicated.

Use cases

Ozymandias

The biggest challenge in adding specimen data ("occurrences") to Ozymandias is deciding how to model occurrences. One approach is to treat occurrence records as "flat" data structures, in much the same way that occurrence data is currently served by GBIF. Existing RDF providers, such as members of the Consortium of European Taxonomic Facilities (CETAF), adopt this approach (Güntsch et al. 2017). In contrast, describe a more elaborate model of an occurrence that separates entities, such as events, localities, identifications, and specimens. This is the model that Page adopted during the workshop, and a small number of GBIF records for type specimens were added to Ozymandias. A SPARQL query was devised to compute the difference in dates between the type specimen of a species being collected and the date the species description was published. For example, the species Amplirhagada cambridgensis Solem was described in 1988 (Solem 1988) and there is a paratype specimen in GBIF that was collected in 1984 (https://www.gbif.org/occurrence/1065115500); hence, the lag between species collection and description was four years (Fig. 1).

Integrated Flora of Canada

At the workshop, the Integrated Flora of Canada team developed a wish list of data integrations for the future platform, explicitly outlining identifier and data providers. In addition, we worked on developing federated SPARQL queries that used Wikidata and Uniprot SPARQL endpoints to discover and collect sequence and reference data. Using these queries as a starting point, discussions around the added capabilities of a flora with up-to-date sequence and reference information ensued. For example, with updated
references, one could construct a reference timeline. This would allow us to ask questions such as, “is research on a particular taxon still ongoing?”, and if so, “who are the current experts?” This has applications for directing treatment revisions to the most appropriate botanists.

Additionally, one of the outcomes for the Integrated Flora of Canada project was clarifying its value proposition. There are many services available online that aggregate data from elsewhere. What the Integrated Flora of Canada provides, beyond aggregating relevant resources, is parsed morphological data from flora documents, which we are currently working to prepare. Once complete, the user will be able to query quantitative and qualitative properties (e.g., leaf coloration), across all taxa represented in the flora.

Stimulating discussions about the future of floras were also part of the workshop. For example, how far away are we from automatically generated floras using existing data sources? Can we generate treatments on the fly as a database query?
Lastly, we hacked at our existing Flora of North America data to convert it into RDF and to experiment with Metaphactory as a platform for hosting our data. The source data was an XML file (Suppl. material 1) used to generate a text version of a floral treatment through the Semantic MediaWiki platform. Because the data were in XML format, it was relatively easy to use an XQuery script (Suppl. material 2) to convert the data to RDF/XML (Suppl. material 3). Once the RDF file was loaded into Metaphactory, it could be queried using the SPARQL interface. For example, the following query could be used to determine which taxa had yellow petals.

```
PREFIX bios: <http://www.github.com/biosemantics/>
PREFIX rdfs: <http://www.w3.org/2000/01/rdf-schema#>

SELECT DISTINCT ?name ?description
WHERE {
  ?treatment bios:hasMajorPart ?major.
  FILTER(CONTAINS(?description, "yellow"))
  OPTIONAL {
    ?major bios:majorDescription ?description.
    ?major bios:majorName "petals".
  }
  OPTIONAL {
    ?major bios:hasMinorPart ?minor.
    ?minor bios:minorName "petals".
  }
}
```

One appealing feature of the Metaphactory platform is that it can generate graph diagrams from data based on screening criteria specified by graph patterns in SPARQL CONSTRUCT queries. To generate the diagram, a page is created using Metaphactory's semantic-map page template by enclosing the query within a `<semantic-graph>` XML element:

```
<semantic-graph height="1000" query="[insert query here]">
"</semantic-graph>
```

In doing this, we were able to generate a diagram visualizing the paths that one could take to arrive at a determination using the Cucurbitaceae key present in the treatment (Fig. 2).

The primary limitation of modelling our data in RDF was the inability to order nodes (i.e., we were unable to designate the 1st discussion node, 2nd discussion node, and 3rd discussion node in order). However, we learned what Metaphactory could afford us in terms of semantic property searches, including semantic auto suggestion, faceted search and exploration of item collections, and translation of keyword queries into SPARQL queries.
In order to use the Metaphactory tools to visualize food web data from the Peace-Athabasca Delta, it was necessary to convert data from several spreadsheets into RDF. Within the spreadsheets, some columns contained identifiers linking the described consumer to known resources. Other columns contained metadata about the consumers and resources. Whenever possible, the metadata columns were mapped to well-known vocabulary terms in the Darwin Core, Dublin Core, or RDF Schema (RDFS) vocabularies. After establishing the mappings, an RDF graph was created for each spreadsheet using CSV to RDF conversion software (Baskauf 2016a).

The specific data converted to RDF is shown in Figs. 3 through 6. The spreadsheet containing basic data about genera included in the study included columns whose headers are shown in the table at the lower left of Fig. 3. That table shows how each column was mapped to standard terms that served as predicates of the RDF triples. The Turtle serialization at the lower right shows how data for one row of the table (for the genus Anax) was converted to RDF. In order to create valid RDF, fake URIs were created by appending the genus name to the made-up namespace "http://unb.ca/taxa/". Fig. 3 shows part of the graph in diagram form, with most of the literal-value properties omitted. Another spreadsheet contained columns of notes about the genera. Since all columns were essentially equivalent comments about the genus, all of the columns were mapped to rdfs:comment properties (Fig. 4).

The food web data were collected during sampling events (i.e., collected at a particular location on a particular date; see Compson et al. 2018 for details). A CSV table with two columns related genera to the event at which they were observed. The link from event to genus present was made with a made-up predicate "unb:hasGenus", and a fake URI for
the event was made by appending the event code to the namespace "http://unb.ca/sample/". The graph resulting from this spreadsheet is illustrated in Fig. 5.

![Graph generated from CSV spreadsheet containing data about genera.](image1)

![Graph generated from CSV spreadsheet containing notes about genera.](image2)

The actual food web relations were contained in a fourth CSV table. Each row of the table documented a known predation relationship, with columns stating the participants in the relationship (consumer and resource), a code for the source of the information, and a locally unique identifier for the relationship. The predation relationship was defined in RDF by linking the predation event to the consumer via the made-up predicate "unb:consumer"
and to the resource it consumed by the made-up predicate "unb:resource". An example of one predation relationship is shown in Fig. 6, along with the RDF/Turtle serialization of the portion of the graph that was diagrammed.

After each of the CSV spreadsheets was converted to RDF/Turtle, the four graphs were loaded into the Metaphactory triplestore to create a single graph that was the union of the four graphs. A small portion of the merged graph is shown in Fig. 7. In the figure, each of
the diagrams in Figs 3, 4, 5, 6 are highlighted in different colors to show how the four individual spreadsheets contributed to the overall knowledge graph.

Once the four RDF/Turtle files generated from the spreadsheets were loaded into the Metaphactory triplestore, it was possible to construct SPARQL queries to ask questions that could not be answered by examining a single spreadsheet. For example, the query in Fig. 8 asks what genera of food resources known to be utilized by the genus *Anax* were observed at sampling events with *Anax*.

The semantic-graph feature of Metaphactory can be used to create a visualization of all of the predation relationships that might be present among genera that were present in a particular sampling event. For example, a diagram for the Aug2012-11A event can be generated using the following query:

```
PREFIX unb: <http://unb.ca/vocab/>

CONSTRUCT {
}

WHERE {
}
```

where :eats is a made-up predicate that connects a consumer directly to a resource.
The resulting visualization is shown in Fig. 9. Although Metaphactory had no problem creating the diagram, there was clearly too much information when the entire food web was displayed, as indicated by the clutter of the resultant figure. However, the visualization was interactive and click-and-drag could be used to drag individual nodes from the mass of nodes to examine individual links in the food web (e.g. *Arrenurus* eats *Callibaetis*, *Notodromas*, and *Caenis*).
We created a more meaningful visualization by restricting the display to parts of the web that met particular criteria. For example, consider the case where genus1 eats genus2, which eats genus3, which eats genus4. For genera present in the Aug2012-11A sampling event, we can ask “in what cases does genus1 also eat genus3 and genus4?” In other words, are there cases where a genus eats broadly down the food chain? To answer this question, we set up the following page template:

```semantic-graph height="1000" query="
PREFIX unb: <http://unb.ca/vocab/>
CONSTRUCT {
}
```

Figure 9. doi
Food web visualization showing all species present at one sampling event

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```semantic-graph height="1000" query="
PREFIX unb: <http://unb.ca/vocab/>
CONSTRUCT {
}
```

Figure 9. doi
Food web visualization showing all species present at one sampling event
WHERE {
}
">
</semantic-graph>

Fig. 10 shows that this rarely occurred. By dragging the nodes around, we could identify the cases. *Lethocerus* eats *Blethisa*, which eats *Chaetogaster*, which eats *Limnesia*, while *Lethocerus* also eats *Chaetogaster* and *Limnesia* lower down on the chain. There were several more cases and we could see that there were also a few cases where this pattern occurred though five trophic levels (e.g. *Lethocerus, Sisyra, Blethisa, Chaetogaster, Piona*).
These exercises and visualizations illustrate that the utility of Metaphactory as a data exploration tool is only limited by our ability to imagine SPARQL queries that reflect interesting ecological questions that we might want to ask about the food webs. Additionally, Metaphactory’s use as a food web visualization tool is restricted when the amount of information displayed is too great.

**Discussion questions**

While the focus of the workshop was training and advancing individual projects, key discussion questions arose. Some of these originated from the relative inexperience of participants, and others from the considerable experience and cynicism of other participants. Most of these questions are unresolved, and the authors invite readers to engage in an ongoing discussion via post-publication review. Here we present our consensus views.

*How can we do a better job at showcasing the utility of knowledge graphs?*

We believe tools like Metaphactory can help to showcase the utility of knowledge graphs.

*Is unfamiliarity with the SPARQL query language a barrier for the biodiversity informatics community towards adopting biodiversity knowledge graphs?*

Undoubtedly, SPARQL is a much more niche query language than SQL. Therefore, we believe unfamiliarity with SPARQL is a barrier. The cultivation of communities of practice within scientific disciplines, achieved through workshops and breakouts at conferences, would help to mitigate this issue. As previously stated, more user friendly front-end interfaces are needed, such as those provided by Metaphactory.

*Does Wikidata change the outlook of building a united biodiversity knowledge graph?*

Wikidata acts as identifier broker. Wikidata has the potential to contribute to consensus-based term definitions, via its collaborative property pages.

*Is the lack of a single access point to the “biodiversity knowledge graph” a limiting factor in demonstrating the utility of integrated data? If so, is this a technological barrier or a social barrier (i.e., who will build the access point)?*

A participant of the hackathon and active member of the biodiversity informatics community, Steve Baskauf, argues that it is hard to do anything meaningful when data are scattered across multiple endpoints (Baskauf 2016b describes difficulties running federated SPARQL queries). However, a fellow workshop participant, Rod Page, suggests that the Linked Data Fragments project (Verborgh et al. 2016) addresses some of the difficulty that Baskauf reports. But what about eliminating the need for federated queries altogether (at least for now), by creating a great big biodiversity knowledge collider (i.e., a single point of access to the biodiversity knowledge graph)? The advantages are that it eliminates our problems with federated queries, and allows us to focus our attention on other
technological barriers. But what data will go into such an access service? Who will pay? Who maintains the access point, and how is it sustained economically?

How can we advance our cause of developing shared identifiers?

The success of shared identifier adoption has varied across domains. Some domains are easy, for example ORCIDs have been widely adopted for the identification of people (authors and contributors). Some are slightly harder, for example DOIs are often used to identify publications. Some really seem to resist consensus. Notably, agreement on taxon concept identifiers has not been reached, and there are multiple schemes in existence (e.g., LSIDs, NCBI taxonomy identifiers, Catalogue of Life identifiers). This is compounded by the lack of clarity on the difference between taxa and taxon concepts (i.e., some services conflate the two, others differentiate them with separate identifiers). Further, taxonomy, which is continually being refined and updated, can make make the adoption of universal identifiers for a taxon concept difficult. This is especially true in the microbial sciences, where species concepts are obfuscated by horizontal gene transfer and rapid evolution. Wikidata helps as an identifier broker, but only on straightforward cases. Identifiers for traits are similarly lacking. A trait like “body size” can have many different names and units for the same entity.

What other data sources should be linked to the biodiversity knowledge graph but can't easily be linked?

This depends on what we want our knowledge graphs to do. If we are interested in building heuristic food webs, we would want to link to traits databases, particularly those that report trophic linkages, such as the Database of Trophic Interactions, the USGS Traits Database, or the Global Biotic Interactions database. However, linking to these and other traits databases that have added trait information (e.g., body size and other life history information) could radically expand the utility of knowledge graphs beyond ecological network applications, making these tools much more appealing to people in disparate fields of the natural sciences. Some of the key traits and trophic linkage databases are below:

1. Database of Trophic Interactions (DTI)
2. USGS Traits Database
3. Global Biotic Interactions (GLOBI)
4. Encyclopedia of Life
5. IsoScape (natural abundance stable isotope repository)

Would the knowledge graph scale?

This question prompted further questions: (1) Can consensus around identifier schemes be achieved amongst all producers of biodiversity data? and (2) Can we develop consensus around semantics and definitions of terms? Computing resources (for federation over the internet; reasoning; graph traversal, etc.), alongside consensus development, may be a limiting factor in the feasibility of a knowledge graph at scale.
Are the pieces that we are trying to merge into a knowledge graph incompatible because they are based on different graph models? Do we need agreement on data models to build a biodiversity knowledge graph?

Even when we agree on the data model, things are very fragile. For example, although data providers may agree to use RDF, vocabularies and shared identifiers may be used inconsistently (Page 2011). Regarding the need for a common data model, mixed sentiments emerged. While we believe that agreement on RDF technology is required, partial agreement on ontology is satisfactory. The workshop participants believe that the problems we face in building a distributed knowledge graph are not solved by ontologies. However, disagreement on the role of ontologies exists, within the broader biodiversity informatics community and at the workshop itself.

Conclusions

To unite our data silos in biodiversity science, we need agreement and adoption of a data modelling framework. A knowledge graph built using RDF, supported by an identity broker such as Wikidata, has the potential to link data and change the way biodiversity science is conducted. For example, in the future floral treatments may be created on the fly as biodiversity knowledge graph queries, and may replace existing botany workflows.

Biodiversity scientists are only beginning to understand the value and utility of constructing a common knowledge graph. While workshop members have pursued independent projects working towards prototype systems (e.g., Ozymandias, the Integrated Flora of Canada), there is still a lack of adoption of shared identifiers and RDF-formatted data. The value of a biodiversity knowledge graph should continue to be demonstrated by use cases, and discussions should continue as the data available grows. Questions that remain include understanding how knowledge graphs scale and how to maintain updated and accurate datasets that feed knowledge graphs. Moving forward, we aim to continue conversations at the Biodiversity_Next conference in 2019 in Leiden, the Netherlands, or at an alternate venue and date. We will also investigate setting up a Wikidata challenge or workshop at the conference, to broaden participation in our discussions of biodiversity knowledge graphs.

Glossary

AWS: Amazon Web Services, a cloud computing service that provides hosting of computational resources

A Biodiversity Knowledge Graph: biodiversity knowledge organized in a graph form, including data related to specimens, images, people and publications, and their relationships. Represents data using common identifiers and open standards such as the W3C’s RDF and TDWG’s Darwin Core to facilitate data integration.
The Biodiversity Knowledge Graph: The hypothetical federation of all biodiversity knowledge graphs. In theory, this is what everyone working on individual biodiversity knowledge graphs is working towards, and contributing to.

Blazegraph: a triplestore and graph database, used in the Wikidata SPARQL endpoint

Connected graph: A graph in which a path exists between any two nodes.

Connectivity: The minimum number of vertices that need to be deleted to render a graph disconnected. An unconnected graph (connectivity=0) represents data that is still siloed. Intuitively, the higher the connectivity of a knowledge graph, the less siloed the data is.

CETAF: the Consortium of European Taxonomic Facilities

CSV: Comma Separated Values delimited text file

Darwin Core: a standard maintained by the Biodiversity Information Standards (TDWG) group, that includes a glossary of terms facilitating the sharing of biodiversity data

Dublin Core: a glossary of terms that can be used to describe digital and physical resources maintained by the Dublin Core Metadata Initiative (dublincore.org)

Ephedra: a SPARQL federation engine aimed at processing hybrid queries that forms a part of the Metaphactory platform (metaphacts.com/product)

Federated query: A SPARQL query whose graph pattern includes triple patterns present at different SPARQL endpoints.

GBIF: Global Biodiversity Information Facility (gbif.org)

Graph Database: a database that models data in graph form, with objects represented by vertices, and relationships between objects represented by edges.

Knowledge Graph: knowledge stored in graph form, with a focus on the unification of knowledge silos through the use of common identifiers and open standards. In our case, a knowledge graph uses URIs to identify objects (vertices) and properties (edges), and the RDF as the data definition language.

Linked Open Data: open structured data that utilizes standards technologies (e.g., RDF, URIs), so that it can be interlinked to other data sources and can be queried

Metaphactory: a knowledge graph management platform

Neptune: a graph database product, offered as part of Amazon Web Services

RDF: the Resource Description Framework, a set of World Wide Web Consortium (W3C) specifications for describing and modelling metadata, often used in knowledge graphs and to model linked open data
**SPARQL**: an RDF query language, a key technology of the semantic web

**Triplestore**: a database for storage and queries of RDF data or triples. A triple is a data entity composed of subject-predicate-object

**Triple**: data modelled in a subject-predicate-object format

**Turtle**: a syntax and file format for representing RDF data

**URI**: Uniform Resource Identifier, a string that unambiguously identifies an entity using a defined naming scheme. Used with the RDF specifications

**W3C**: World Wide Web Consortium

**Wikibase**: a freely available set of MediaWiki extensions that can be used to manage versioned data and allow for data collaboration in a central data repository

**Wikidata**: a collaborative knowledge base offering data provenance and other features, hosted by the Wikimedia foundation and supporting Wikipedia

**XML**: eXtensible Markup Language, a data storage file format recommended by the W3C

### References

- Egloff W, Agosti D, Kishor P, Patterson D, Miller J (2017) Copyright and the use of images as biodiversity data. Research Ideas and Outcomes 3: e12502. [https://doi.org/10.3897/ri.o.3.e12502](https://doi.org/10.3897/ri.o.3.e12502)


Supplementary materials

Suppl. material 1: Source XML data from Flora of North America for the family Cucurbitaceae

Authors: Flora of North America Editorial Committee, eds.
Data type: Taxon description
Brief description: The data in this XML file was used to generate the textual treatment of Cucurbitaceae at http://beta.floranorthamerica.org/wiki/Cucurbitaceae using the Semantic MediaWiki platform.
Filename: V6_1.xml - Download file (24.70 kb)

Suppl. material 2: Script to transform idiosyncratic XML data into RDF/XML

Authors: Baskauf, Steven
Data type: code
Brief description: This script was written in XQuery and processed using the BaseX XQuery processor
Filename: make-rdf-xml-family.xq - Download file (10.52 kb)

Suppl. material 3: RDF/XML serialization of Cucurbitaceae taxonomic treatment

Authors: Nesom, Guy L. and Steven J. Baskauf
Data type: RDF
Brief description: The taxonomic treatment was written by Guy L. Nesom with the conversion to RDF by Steven J. Baskauf
Filename: cucurbitaceae.rdf - Download file (33.76 kb)