

Analysis on the Graph Techniques for Data-mining and Visualization of Heterogeneous Biodiversity Data Sets

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Keywords: Biodiversity Data Mining, Ontology Engineering, Biodiversity Metadata Visualization, Graph.

Abstract: Existing biodiversity databases contain an abundance of information. To turn such information into knowledge, it is necessary to address several information-model issues. Biodiversity data are collected for various scientific objectives, often even without clear preliminary objectives, may follow different taxonomy standards and organization logic, and be held in multiple file formats and utilising a variety of database technologies. This paper presents a graph catalogue model for the metadata management of biodiversity databases. It explores the possible operation of data mining and visualization to guide the analysis of heterogeneous biodiversity data. In particular, we would propose contributions to the problems of (1) the analysis of heterogeneous distributed data found across different databases, (2) the identification of matches and approximations between data sets, and (3) the identification of relationships between various databases. This paper describes a proof of concept of an infrastructure testbed and its basic operations, presenting an evaluation of the resulting system in comparison with the ideal expectations of the ecologist.

1 INTRODUCTION

Accurate and publicly available information on biodiversity observations can contribute to scientific knowledge, foster multidisciplinary studies, and provide new perspectives to environmental and societal responses including decision-making (Lausch et al., 2015). To this end, several biodiversity metadata projects have been established which describe and characterize the information hosted in a range of distributed databases (David et al., 2016; Dodge et al., 2013).

As these metadata projects grow horizontally, with more databases and types of data sets, as well as vertically, with more documents, the ecologist will need information management tools to enable the following common tasks:

1. To discover existing but heterogeneous, dispersed data sets of different origins and scales of observation;
2. To discover relationships between documents of

potential interest to the scientist;

3. To interpret the semantic meaning of relations without the need to know the meta-model;
4. To enable the scientist to understand the data context and collection methods of multiple fields and topics;
5. To determine the quality associated with the data, including the data sets inter-calibration; and
6. To be aware of the conditions of access and use.

This proof of concept presents a case study of the ECOSCOPE metadata catalogue (Taffoureau et al., 2016; Eco,) which provides data mining and visualization capabilities. ECOSCOPE is a metadata collection service for databases of different fields of ecology *in lato sensu*.

We follow a behaviour-driven development (BDD)(Solis and Wang, 2011) of a minimal operational set and the assessment of the ecologist at each operation. The resulting evaluation is used to

propose a new graph catalogue service architecture. The new graph catalogue can be used for metadata discovery and visualization, integrated with the existing and future data management service. The current ECOSCOPE web catalogue is used to collect metadata in a standardized way, using an authorization service which provides the ecologist accredited access to various storage systems.

2 RELATED WORK AND MOTIVATION

The consortium IndexMed (renamed recently “IndexMeed - Indexing for Mining Ecological and Environmental Data” to build international projects) was created by the axis “Management of biodiversity and natural spaces” of the IMBE (Mediterranean Institute of freshwater and marine Biodiversity and Ecology) (David et al., 2015). Its main goal is to develop awareness of databases and their effective use in the ecological research community. This consortium is particularly useful as a bridge between existing networks and initiatives at national and international levels. The aim of the consortium is to index biodiversity data (and to provide an index of qualified existing open datasets) and to make it possible to build graphs to assist in the analysis and the development of new ways to mine data. Standards and specific protocols can be applied to interconnect databases. Semantic approaches greatly increase data interoperability. The project should develop new transdisciplinary methods of data analysis, focusing on open data, open source and free methods and development tools.

ECOSCOPE is an infrastructure funded and managed by French research organisations through the Foundation for Research on Biodiversity (FRB) which ensures its coordination. The scientific aim is to document the state and trends of biodiversity and ecosystem services, enabling scenarios for the future to be built. In this framework, ECOSCOPE promotes the complementarity of observations and links between research observation systems that vary across spatial and temporal scales, variables, studied ecosystems and kingdoms, levels of organization and data sources. In cooperation with existing initiatives, ECOSCOPE provides an entry point for the discovery of observations and datasets for research on biodiversity across the entire data life cycle, facilitating links between data producers and users.

Note - for deletion on final version: Scholes does not refer to ECOSCOPE. hence it cannot be used in this manner as a reference. You could say “These aims are consistent with Scholes et al.(2012).”

The ECOSCOPE metadata catalogue delivers freely available online information about who, where, what, when, why and how the research observation data were collected. It is build on the EBV concept, developed by GEO BON (Group on Earth Observation Biodiversity Observation Network), which is designed to serve as the foundation for interoperable sub-national, national, regional and global monitoring initiatives.

Precise and public information on biodiversity observation datasets contribute to data openness and reuse, in full conformity with data producers and owners. Metadata formats the description, characterisation and specification of data hosted in datasets, allowing the discovery of data, whether heterogeneous or dispersed and across locational and observational scales. Metadata permits the understanding of the context of the dataset, collection methods and data quality. It gives information on access and use of data and other resource conditions and the contact persons (Michener 2006).

Michener W.M. (2006) Meta-information concepts for ecological data management. *Ecological Informatics* 1:3-7

The ECOSCOPE metadata catalogue answers to this need thanks to providers and exchanges with other information systems. As it is based on standards in use, the metadata profile can be exported into other information systems, and metadata files (such as Ecological Metadata Language: EML) can be imported into the ECOSCOPE metadata portal. It contributes to global efforts to make research on biodiversity data more available for scientific projects, synthesis and indicators.

In this context of a prototype for collecting ecological metadata of various fields and topics, our motivation is to explore the possibilities of graph techniques for visualization and data mining supported in graph databases. The graph databases have been a proven feasible backend to provide semantic services (Riesen and Bunke, 2008; Angles and Gutierrez, 2008). Furthermore, the indexing capabilities of graph databases ensures the scalability of the system response (Williams et al., 2007), which is a critical factor in our needs to increase various databases, data sets, and document integration.

In a graph catalogue the database mapping model is isomorphic with the represented structure. The resulting model enables the evolution of applications with linear complexity in the data mining operation, which is critical for scale-up in data volume and variety.

The overall architecture of our vision is shown in Figure 1. There are increasing number of database

browse of a graph focused in the core information with temporal and spatial information.

Test:

```
MATCH (n)
WHERE NOT n:Attribute
AND NOT n:Keyword
AND NOT n:TaxonomicClassification
AND NOT n:TaxonomicCoverage
RETURN n
```

Assets: Here a clear segmentation of the graph in 5 categories is obtained (Figure 2). This general spatial temporal graph shows two types of node segments. On one hand, some segments are irrelevant because a single data set (in blue) is related to its core information: the three segments in the bottom right. In the other hand, node segments with several data sets are related with temporal spatial information. For example, the segment on the top shows all the data sets (nodes in blue) are related to a single geographical area (node in yellow), even when they have been tagged with different geographical names in the database (nodes in pink).

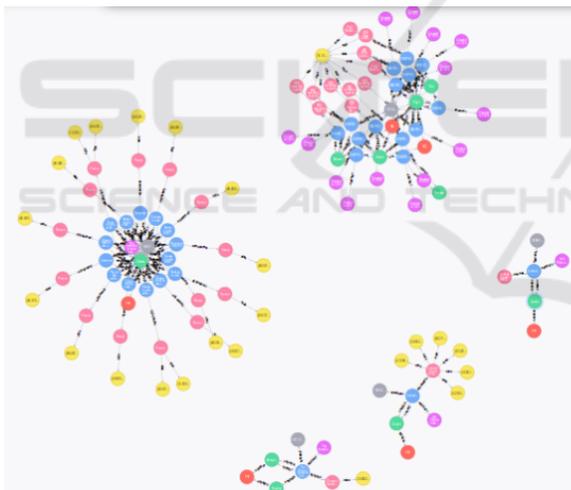


Figure 2: A general view of the spatial temporal graph.

Weakness: The resulting segmentation does not show relations between data sets of different databases. Eventually, a more precise matching in geographical area is needed, for example by area proximity or overlap. Another approach could be to draw such areas in the map to give to the ecologist a visual map of the data sets.

3.1.3 Operation: Show Graph for Taxonomy and Organizational Relations

Given the node classification above.

When it is needed for analysis of data set categories,

Then it can be of interest to the visualization and browse of a graph focused in core information with the organizational logic metadata.

Test:

```
MATCH (n)
WHERE NOT n:Attribute
AND NOT n:TemporalCoverage
AND NOT n:GeographicCoverage
AND NOT n:GPolygonOuterRing
RETURN n
```

Assets: Figure 3 shows a clear segmentation of data sets (nodes in blue) by the organization logic metadata of Keyword (in red), TaxonomicClassification (in green) and TaxonomicCoverage (in pink), but with all the nodes connected, which is of high interest to enable multidisciplinary relationship discovery. Our results show some metadata fields which are relation-hubs between data sets of different databases, particularly a few generalist TaxonomicClassification values and Keywords.

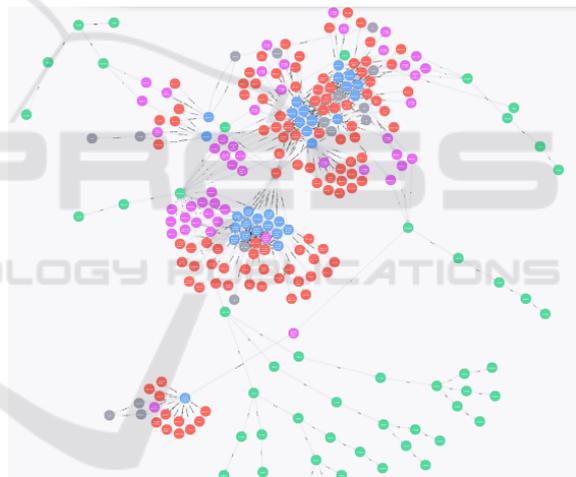


Figure 3: A general view of the organizational logic graph.

Weakness: Even when the visualization tool is able to do a zoom of Figure 3, this is not enough to ensure a systematic discovery.

3.2 Common Data Mining Operations

This sub-section presents operations in the metadata graph database to provide a subset of the graph according to the behaviours required by the ecologist, which have been described in the enumeration of the introduction section.

3.2.1 Operation: Common 1

Given the general graph visualizations above,

When ecologist want to discover existing but hetero-

geneous, dispersed data sets of different origins and scales of observation;

Then restrict the graph of visualization operation 3.1.2, which contains geographical origins and temporal scales, to match a single hub node and related data sets. The hub node is taken from previous general view of operation 3.1.3

Test:

```
START keyword=node(*)
MATCH (n)->[]-(d)-[r]->(keyword)
WHERE keyword.word = "AGROVOC"
AND NOT n:Attribute
AND NOT n:Keyword
AND NOT n:TaxonomicClassification
AND NOT n:TaxonomicCoverage
RETURN n,d,r,keyword
```

Assets: In Figure 4 a zoom view of all the data sets related to the hub metadata. In red the hub node (*Keyword=AGROVOC*). The cluster on the top is a segment of DIMPIE data sets, of the same database, with a temporal coverage in green (1975) and to the left there is a blue data set (*Collection moisissures IFV*), which temporal coverage of 2008 in green. So both databases and datasets are related by the hub node.

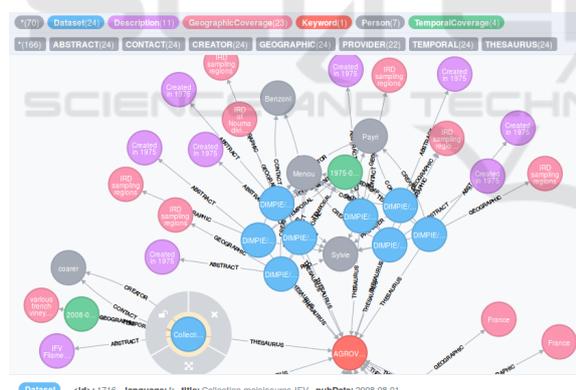


Figure 4: A zoom of different origins and scale matching with a hub node.

Weakness: There is no systematic way of filtering origins and scales.

3.2.2 Operation: Common 2

Given the metadata catalogue,
When the ecologist wants to discover relationships in a document of potential interest;

Then starting from operations to match data sets, filter the desired documents.

Weakness: In our case study the metadata source has not the details of each document. It is necessary

to collect the metadata information of the document in the meta catalogue.

3.2.3 Operation: Common 3

Given a data set,

When the ecologist wants to interpret the semantic meaning of relationships without the need to know the meta-model;

Then to dig in the relationships without an explicit relationship label

Test:

```
START d1=node(*)
MATCH (d1)-[*1..5]->(n)
WHERE d1.title =~ "Donkey.*"
AND NOT n:Attribute
AND NOT n:Keyword
AND NOT n:TaxonomicClassification
AND NOT n:TaxonomicCoverage
RETURN d1,n
```

Assets: Figure 5 illustrates the great possibilities of the graph approach to describe meta-model semantics, without explicit knowledge of the model. The clause *MATCH (d1)-[*1..5]->(n)* gets a maximum depth of 5 levels, and the result shows a maximum of only two levels of relations from the given data set (node in blue). The rest of the MATCH clause is restricting the results to the basic information and the spatial temporal information of the visualization operation in 3.1.2. Another interesting filter would be to show the taxonomy and organizational relations of the visualization operation in 3.1.3.

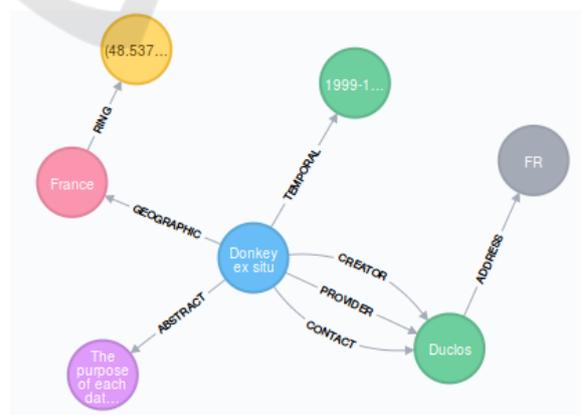


Figure 5: Semantic visualization of relations of a given data set.

Weakness: The test command shows only outgoing relations from the given data set. Eventually, a more generalist operation shall ask the ecologist

whether it is requested outbound, inbound or both relations to or from a given data set.

3.2.4 Operation: Common 4

Given a data set,

When To guide the ecologist to understand the data context and collection methods;

Then to dig in the collection and context information of the data set.

Test:

```
START d1=node(*)
MATCH (d1)-[*1..8]->(n)
WHERE d1.title =~ "Donkey.*"
AND NOT n:Address
AND NOT n:Person
AND NOT n:TemporalCoverage
AND NOT n:GeographicCoverage
AND NOT n:GPolygonOuterRing
RETURN d1,n
```

Assets: Given a data set in blue, Figure 6 shows on one hand the information on collection methods about taxonomic coverage (in yellow) and the corresponding sub-graph of the taxonomic classification in green. The names displayed are the category of the classification, while by clicking in a particular green node will give the corresponding value for the data set. On the other hand, the context information is shown in the attributes (in grey) of the documents in the data set, as well as the keywords (in red) of the data set. Spatial and temporal information would be other interesting data context information.

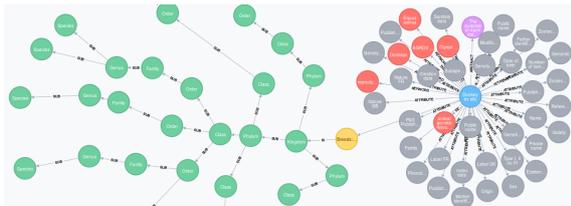


Figure 6: Data context and collection methods.

Weakness: Even when we have the attribute list of the documents, we don't have a document catalogue, so this information is of little value. Eventually we should include the document catalogue in the graph catalogue.

3.2.5 Operation: Common 5

Given a data set,

When the ecologist would like to estimate the quality associated with the data, including the data sets

inter-calibration;

Then Show detailed information about the data quality of the corresponding nodes and inter-calibration.

Assets: Figure 7 shows the content of the Description node of a data set, which gives information about the associated data quality and a few calibration details.

Description

 <Id>: 1618

purpose:
The purpose of each dataset is to give a core of information related to the biological material which is stored.

abstract:
Core data include species/breed or strain/Type of reproductive material / Genetic type (I, II, III) / animal individual identification. When possible pedigree information and/or phenotype data are also available, however for most species (i.e. equids, ruminants and pigs) a national data center exists that gathers all this information. The only information needed to get a correspondence between the national data center and the French Cryobank database is the animal national identification.

Figure 7: Data quality and callibration details.

Weakness: There is more quality and calibration information in some of the Attribute nodes. However, the name of the Attribute with valuable information is dependent on the particular data set. Therefore, it will be necessary to include a label in those Attribute nodes which are related to calibration and data quality to display such nodes for a given data set so the ecologist could browse the details.

3.2.6 Operation: Common 6

Given the metadata catalogue,

When the ecologist wants to be aware of the conditions of access and use of data sets and more documents;

Then provide access policy to sets and objects

Weakness: In our case study the metadata source only provides a secondary way to obtain the data, by giving the contact person and web information for a data set. So the ecologist can manually manage their access to the data, and there is no automation in this behaviour. This is a critical point to overcome the current collection scope by tools and methods to enable the access policy to the existing database objects.

4 SERVICE ARCHITECTURE

These tests have demonstrated the feasibility of graph techniques to provide semantic features in visualization and data mining of ecological metadata. However, to facilitate the ecologist’s discovery and visualization, it also is necessary to provide high level applications alongside the existing graph database. The weakness analysis on the common expected operations, points to the need for more generic operations and systematic approaches, adapted to the characteristic multidisciplinary database of the ecologist.

The required behaviour on several of the common operations needs the inclusion of the metadata of the documents, not only as generic information of the data set.

- *Common 2* and *Common 4* need the integration metadata of the documents in the catalogue.
- *Common 6* is a critical operation to provide automated access policies to the documents.

For these reasons the present paper proposes a model-view-controller (MVC) service architecture (Deacon, 2009) as show in Figure 8

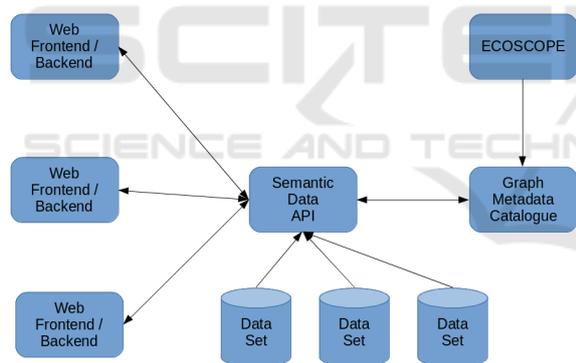


Figure 8: Proposed MVC Service Architecture.

- **View-Controller**
 - **Web Frontend/Backend** a common web framework for all the specific webs of the multidisciplinary studies, including basic frontend forms, backend handlers and driver connectivity to the common API. It supports the identity management and it is the user entry point to the data.
- **Model**
 - **Semantic Data API** Including all the high level methods for visualization, data mining and the gateway for data access policies to various storages.

- **Graph Metadata Catalogue** With the existing meta model for data sets, but also including the metadata of the documents, as well as the access policies between identities and documents.
- **Data Set** with secured access to the documents
- **ECOSCOPE** the metadata collection and standardization portal.

The developing, releasing and deployment of the service components can be enabled by a container compose (Mulfari et al., 2015) or virtual machine infrastructure manager (Caballer et al., 2015).

5 CONCLUSIONS

This paper presents preliminary studies on metadata semantics. Indeed, it demonstrates improved metadata qualification needs using tools, standards and recommendations at both national (SINP [National Information System on Biodiversity], RBDD [Network of Research Databases]) and international levels (MedOBIS [Mediterranean Ocean Biogeographic Information System], OBIS, GBIF (Cryer et al., 2009), Life-Watch, GEO-BON, etc.) or shared by other research entities (i.e. IRD [Institute of Research for the Development] or MNHN [National Museum of Natural History, Paris])

The proof of concept demonstrates the potential of graph databases to enable metadata visualization and common operations in a scalable way through the graph database capabilities in the horizontal relations. Furthermore, it has identified the commonalities for a high level semantic data API, into a service architecture for several specific web front-ends, contributing to economies of scale in the development and exploitation of the information system.

The promising results encourage future work following the proposed service architecture, to facilitate ecological studies in heterogeneous fields and topics with their increasingly complex requirements.

ACKNOWLEDGEMENTS

The authors would like to thanks Alison Specht, director of CESAB (FRB) and to Robin Goffaux from FRB for they advisory and support to this paper. This work is co-funded by the EGI-Engage project (Horizon 2020) under Grant number 654142 and by the Spanish MICINN project number TIN2014-53234-C2-1-R. IndexMed consortium is funded by the CNRS défi “VIGI-GEEK (VISualisation of Graph

In transdisciplinary Global Ecology, Economy and Sociology data-Kernel”, CNRS INEE through the “CHARLIEE” project in 2015 and CNRS “Mission pour l’Interdisciplinarité” in 2016. Data used for this article were obtained through ECOSCOPE metadata tools. The authors acknowledge the support of France Grilles for providing computing resources on the French National Grid Infrastructure. Supplementary acknowledgement to organisers of the EGI workshop “design your e-infrastructure” which started this work.

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