

Using the Semantic Web in Digital Humanities: Shift from Data Publishing to Data-analysis and Serendipitous Knowledge Discovery

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Editors: Pascal Hitzler, Kansas State University, Manhattan, KS, USA; Krzysztof Janowicz, University of California, Santa Barbara, USA
Solicited reviews: Rafael Goncalves, Stanford University, CA, USA; Peter Haase, metaphacts GmbH, Walldorf, Germany; One anonymous reviewer

Abstract. This paper discusses a shift of focus in research on Cultural Heritage semantic portals, based on Linked Data, and envisions and proposes new directions of research. Three generations of portals are identified: Ten years ago the research focus in semantic portal development was on data harmonization, aggregation, search, and browsing (“first generation systems”). At the moment, the rise of Digital Humanities research has started to shift the focus to providing the user with integrated tools for solving research problems in interactive ways (“second generation systems”). This paper envisions and argues that the next step ahead to “third generation systems” is based on Artificial Intelligence: future portals not only provide tools for the human to solve problems but are used for finding research problems in the first place, for addressing them, and even for solving them automatically under the constraints set by the human researcher. Such systems should preferably be able to explain their reasoning, which is an important aspect in the source critical humanities research tradition. The second and third generation systems set new challenges for both computer scientists and humanities researchers.

Keywords: Digital Humanities, Linked Data, Semantic portals, Data analysis, Knowledge discovery

1. Introduction

Cultural Heritage (CH) has become a most active area of application of Linked Data and Semantic Web (SW) technologies [1]. Large amounts of CH content and metadata about it are available openly for research and public use based on collections in museums, libraries, archives, and media organizations. For example, data has been aggregated in large national and international repositories, web services, and portals such

as Europeana¹ and Digital Public Library of America², and forms a substantial part of DBpedia³ and Wikidata⁴.

The availability of Big Data has boosted the rapidly emerging new research area of Digital Humanities (DH) [2, 3] where computational methods are developed and applied to solving problems in humanities

¹<http://europeana.eu>

²<https://dp.la/>

³<http://dbpedia.org>

⁴<http://wikidata.org>

and social sciences. In this context Big Data means data that is too big or complex to be analyzed manually by close reading [4].

From a SW research point of view, CH data provide interesting challenges for DH research: First, the data is syntactically heterogeneous (text, images, sound, videos, and structured data in different formats, such as XML, JSON, CSV, and RDF) and written in different languages. Second, the data is semantically rich covering all aspects of life in different times and places. Third, the data are often incomplete, imprecise, uncertain, or fuzzy due to the nature of history. Fourth, the data is interlinked across different data sources, distributed in different countries and databases. Helping the humanities researcher to deal with such data in semantically complex problems addressed in humanities sets for computer scientists interesting methodological problems.

This paper analyses and discusses this line of research and development at the crossroads of Semantic Web research, humanities, and social sciences, from the early days of the Semantic Web to next steps in the future. Three conceptual generations of semantic portals on the Semantic Web are first identified. After this the ideas are made more concrete by an example case study system exhibiting features of the three generations.

2. First Generation: Portals for Search and Browsing

Due to the challenges in CH data, SW research in CH has been initially focused on issues related to syntactic and semantic interoperability and data aggregation. A great deal of work has been devoted in developing metadata standards and data models for harmonizing data, including application agnostic W3C standards⁵ (RDF, OWL, SKOS, etc.), document centric models, such as Dublin Core and its dumb down principle⁶, and event-centric models for data harmonization on a more fundamental level, such as CIDOC CRM⁷ [5] for museums and its extensions⁸, and FRBRoo [6] and IFLA Library Reference Model (LRM)⁹ in libraries. In document-centric metadata models the

⁵<http://www.w3.org/standards/semanticweb>

⁶<https://www.dublincore.org/>

⁷<http://cidoc-crm.org>

⁸<http://www.cidoc-crm.org/collaborations>

⁹<https://www.ifla.org/publications/node/11412>

idea is to agree upon a shared way of describing the properties of documents, and how different models can be mapped on each other for interoperability. The event-centric approach focuses on developing more fundamental ontological models of the real world onto which different data and metadata can be transformed for interoperability. Once the data is harmonized in one way or another, it can be published in a SPARQL endpoint, and semantic portals can be created on top of it via APIs.

Both document-centric and event-centric approaches have been successful. Dublin Core and its extensions have become the metadata norm for representing documents on the Web, and a lot of use cases and applications of CIDOC CRM¹⁰ and other event-centric systems have been published.

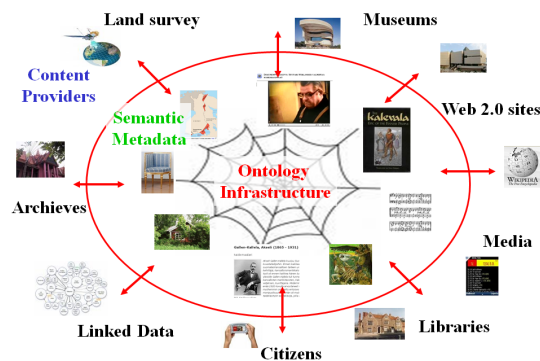


Fig. 1. A model for distributed Linked Data publishing. The data publishers around the circle, i.e., a joint publishing system, provide data using the vocabularies of a shared ontology infrastructure in the middle. The data are automatically interlinked and enrich each other.

The ideas of the Semantic Web and Linked Data can be applied to address the problems of semantic data interoperability and distributed content creation at the same time, as depicted in Fig. 1. Here the publication system is illustrated by a circle. A shared semantic ontology infrastructure is situated in the middle. It includes shared domain ontologies, modeled using SW standards. If content providers outside of the circle provide the system with metadata about CH based on the same ontologies, the data are automatically linked through shared URIs, enrich each other, and form a joint knowledge graph.

For example, if metadata about a painting created by Picasso comes from an art museum, it can be enriched

¹⁰<http://www.cidoc-crm.org/useCasesPage>

1 by data links to, e.g., biographies from Wikipedia and
 2 other sources, photos taken of Picasso, information
 3 about his wives, books in a library describing his works
 4 of art, related exhibitions open in museums, and so on.
 5 At the same time, the contents of any organization in
 6 the portal having Picasso-related material get enriched
 7 by the metadata of the new artwork entered in the sys-
 8 tem. This is a win-win business model for everybody
 9 to join such a system; collaboration pays off.

10 Combining the infrastructure with the idea of decou-
 11 pling the data services for machines from the applica-
 12 tions for the human user creates a model for building
 13 collaborative Semantic Web applications. This model
 14 has been developed and tested in practice, e.g., in the
 15 “Sampo” series of semantic portals¹¹ [7, 8]. The idea
 16 of collaborative content creation using Linked Data
 17 has been developed also in other settings, e.g., in Re-
 18 searchSpace¹².

19 The main use case in CH portals has been providing
 20 the user with enhanced information retrieval (IR) facil-
 21 ities [9], such as faceted search [10], semantic search,
 22 entity search, and semantic recommendation systems
 23 [11] for exploring the data in intelligent ways. Such
 24 CH search and browsing systems based on harmonized
 25 aggregated linked data will be called *first generation*
 26 *systems*.

27 28 29 **3. Second Generation: Portals with Tools for** 30 **Distant Reading**

31
32 As more and more harmonized aggregated linked
 33 datasets are available, the time has come to take a next
 34 step forward to *second generation* of CH semantic por-
 35 tals. The novelty of such systems is to provide the
 36 user with tools for solving Digital Humanities (DH)
 37 research problems, not only tools for searching and
 38 browsing the data. For example, the researcher may be
 39 interested in finding out, how historical persons, ships,
 40 or manuscripts have been moving around geographi-
 41 cally, what topics have appeared and when in par-
 42 liamentary discussions, newspapers, or other corpora,
 43 what kind of social networks or correspondences there
 44 have been between members of a society, and so on.
 45 In DH, a key goal is to use computational methods for
 46 solving humanities and social science problems using
 47 large datasets that have become available. A variety of

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49 ¹¹[https://www.europenowjournal.org/2019/09/09/
 50 linked-data-in-use-sampo-portals-on-the-semantic-web/](https://www.europenowjournal.org/2019/09/09/linked-data-in-use-sampo-portals-on-the-semantic-web/)

51 ¹²<https://www.researchspace.org/>

1 technologies have been developed and applied for such
 2 tasks, such as sentiment analysis [12], topic modeling
 3 [13], network analysis [14, 15], and visualizations [16]
 4 in addition to traditional and novel statistical methods,
 5 such as word embeddings and neural networks [17–
 6 19].

7 Many of the methods and tools above are domain
 8 independent, and there are a lot of software packages
 9 available for using them, such as Gephi¹³, R [20], and
 10 various Python libraries¹⁴. However, each of them have
 11 their own input formats and user interfaces. Further-
 12 more, visualizations are crafted case by case; tools for
 13 formulating, adjusting, and comparing them in gener-
 14 alizable ways would be helpful for the user. A major
 15 problem here is that using the tools typically requires
 16 technical expertise and skills not common among the
 17 humanities researchers. Furthermore, the tools usually
 18 do not support Linked Data formats and data services,
 19 and there is the burden of transforming and transport-
 20 ing linked data into formats required by the different
 21 data analysis tools. A challenge therefore is how to cre-
 22 ate the tools in a generalizable way so that the end user
 23 can adapt them for her own particular research prob-
 24 lems.

25
26 At the moment, many portals include tools but they
 27 are mostly aimed for visualizing and exploring the
 28 data. Showing data on maps and timelines are common
 29 examples of this. The same applies to some systems
 30 for network analysis, such as Six Degrees of Francis
 31 Bacon¹⁵, where one can search for a person or a group
 32 whose networks are then shown for exploration using
 33 interactive graphs. It is also possible to show the group
 34 on a timeline and, e.g., filter the connections in the
 35 network based on parameters. To move on to second gen-
 36 eration systems with a clearer focus on data analytic
 37 tooling one could, e.g., compute various connectivity
 38 parameters and statistics of the networks, such as most
 39 connected nodes, hubs, and connections in the data.
 40 Such tool-oriented systems are largely still missing in
 41 semantic portals; data analysis in Digital Humanities is
 42 usually done by downloading data and by transforming
 43 it locally to be used in specific off-line tools. Integra-
 44 tion of data analytic tools with online semantic portals
 45 is a promising future direction of work [21, 22].
 46

47
48 ¹³<https://gephi.org/>

49 ¹⁴[https://bigdata-madesimple.com/
 50 top-20-python-libraries-for-data-science/](https://bigdata-madesimple.com/top-20-python-libraries-for-data-science/)

51 ¹⁵<http://www.sixdegreesoffrancisbacon.com/>

4. Third Generation: Portals for Serendipitous Knowledge Discovery

Current DH systems have focused on semantic data aggregation, enrichment, validation, search, exploration, visualization, and in some cases even data analysis. The idea has been to search and present the data to the DH researcher using statistical charts, maps, timelines, graphs, and other means so that the researcher can more easily analyze the data related to her/his research problem. What is still largely missing in the DH methodology and tools is the next conceptual level of Artificial Intelligence where the DH tool is able not only to present the data to the human researcher in useful ways but also to 1) find, address, or solve the DH research problems *automatically by itself* and 2) also explain its reasoning or solution to the researcher. This is a grand challenge for research in the future.

To address this challenge one has to study serendipitous¹⁶ knowledge discovery (KD) [23, 24] in the context of historical Cultural Heritage Big Data. Another direction of research to draw ideas from is Computational Creativity in Artificial Intelligence [25]. Serendipitous knowledge discovery is one of the original promises of the Semantic Web [26]. However, there is surprisingly little research about it. A reason for this may be a shortage of high quality densely interlinked datasets needed for studying serendipity. Also the notion of serendipity is conceptually complicated. Better understanding of the notion of serendipity and how insightful knowledge discovery can be implemented and utilized is needed. This could lead to new insights of scientific discovery in humanities and to a paradigm change where the role of the computer is changing from a passive tool to a proactive intelligent agent.

For this challenge the research agenda for the future should seek answers to, e.g., the following fundamental research questions:

1. *How can one formalize the notion of serendipity in terms of 'interestingness' [27] in a generalizable way?* It does not make sense to hard code serendipity in a system using specific ad hoc rules, otherwise reasoning would not be serendipitous.
2. *How can serendipitous phenomena and their explanations be extracted from the data?*

¹⁶Serendipity means 'happy accident' or 'pleasant surprise', even 'fortunate mistake'.

3. *How can the notion of serendipity (1) and the methods for discovering it (2) be used in practice for finding, addressing, and solving humanities research problems?*

4. *How can semantically rich-enough linked datasets for (1–3) be created, based on combining both structured and non-structured data?* An important research topic here is Natural Language Understanding, since the primary data is typically available in textual forms.

In previous sections, semantic portals have been categorized conceptually into three generations. However, in practice the later generation systems have to address the challenges of the former generations, too: a requisite for both second and third generation systems is availability of harmonized linked data, as in first generation systems, and third generation systems also focus on tools in a way similar to second generation systems.

In order to make the ideas presented above more concrete by an example, a semantic portal, BiographySampo, is presented next. This system was created with the goal of making a paradigm shift in its field from state-of-the-art first generation systems to a second generation systems. However, the system also includes a third generation tool for serendipitous knowledge discovery.

5. A Case Study: BiographySampo – Biographies on the Semantic Web

Biography is a research area in humanities that studies life stories of particular people of significance, with the aim of getting a better understanding of their personality and actions, e.g., to understand their motives [28]. An important resource in this research field are biographical dictionaries [29] that may contain tens of thousands of short biographies of historical persons of importance¹⁷. Traditionally, such dictionaries have been published as printed book series but nowadays major biographical dictionaries have opened their editions on the Web with search engines for finding and (close) reading biographies of interest.

¹⁷On-line national biographical collections include, e.g., USA's American National Biography [30], Germany's Neue Deutsche Biographie [31], Biography Portal of the Netherlands [32], Dictionary of Swedish National Biography [33], and National Biography of Finland [34].

1 In BiographySampo¹⁸ [22], linked data and natural
2 language technology was used for creating a knowl-
3 edge graph encompassing the data related to 13 100
4 biographies, including the National Biography of Fin-
5 land [34]. The data was harmonized using an extension
6 of CIDOC DRM and was linked to 16 external datasets
7 for enriching the contents. The data was published in
8 a SPARQL endpoint, and faceted search was imple-
9 mented on top of the data service for finding biogra-
10 phies and exploring them by browsing. These features
11 make BiographySampo a state-of-the-art first genera-
12 tion system.

13 In contrast to biography, the focus of *prosopogra-*
14 *phy* research is to study life histories of groups of peo-
15 ple in order to find out some kind of commonness or
16 average in them [35]. For example, the research ques-
17 tion may be to find out what happened to the students
18 of a school in terms of social ranking and employ-
19 ment after their graduation. The prosopographical re-
20 search method [35, p. 47] has two steps: First, a tar-
21 get group of entities in the data is selected that share
22 desired characteristics for solving the research ques-
23 tion at hand. Second, the target group is analyzed, and
24 possibly compared with other groups, in order to solve
25 the research question. The analysis may involve, e.g.,
26 creating pie charts, histograms or other statistics of the
27 target group, mapping the target group geographically,
28 network analysis, etc.

29 To support prosopography, a second generation CH
30 application with tooling is needed. Filtering out the tar-
31 get group is not enough but tools and visualizations
32 are needed for analyzing it, too. In developing Biog-
33 raphySampo, a major goal has been in providing the
34 DH researchers with generic tools for data visualiza-
35 tion and analysis. Moreover, the tools can be applied
36 not only to one target group but also to two parallel
37 groups in order to compare them. For example, Fig. 2
38 compares the life charts of Finnish generals and admiral-
39 s in the Russian armed forces in 1809–1917 when
40 Finland was an autonomous Grand Duchy within the
41 Russian Empire (on the left) with the members of the
42 Finnish clergy (1800–1920) (on the right). With a few
43 selections from the facets the user can filter out the
44 two target groups and see that, for some reason, quite a
45 few officers moved to Southern Europe when they re-
46 tired (like retirees today) while the Lutheran ministers
47 tended to stay in Finland.

50 ¹⁸The portal is online at <http://biografiasampo.fi> and has had tens
51 of thousands of users

1 In the same way, the statistical application perspec-
2 tive in the system includes histograms showing vari-
3 ous numeric value distributions of the members of the
4 target groups, e.g., their ages, number of spouses and
5 children, and pie charts visualizing proportional dis-
6 tributions of professions, societal domains, and work-
7 ing organizations. There is also a network perspective
8 based on the idea of visualizing and studying networks
9 among target groups filtered out using facets. The net-
10 works are based on the reference links between the bi-
11 ographies, either handmade or based on automatically
12 detected mentions. The depth of the networks can be
13 controlled by limiting the number of links, and color-
14 ing of the nodes can be based on the gender or societal
15 domain of the person (e.g., military, medical, business,
16 music, etc.).

17 The biographies can also be analyzed as a collection
18 of artefacts by using linguistic analysis. For example,
19 it turns out that the biographies of female Members of
20 the Parliament (MP) frequently contain words "fam-
21 ily" and "child", but these words are seldom used in
22 the biographies of male MPs. The analyses are based
23 on a linguistic knowledge graph of the texts.

24 These tools and functionalities make BiographySam-
25 po a second generation system. To study and explore
26 the possibilities and challenges of third generation
27 systems, yet another application perspective was cre-
28 ated in BiographySampo for finding interesting ser-
29 endipitous connections in the biographical knowledge
30 graph. This application idea is related to relational
31 search [36, 37]. In our case a new knowledge-based ap-
32 proach was developed to find out in what ways (groups
33 of) people are related to places and areas. Such con-
34 nections can reveal hidden indirect relations that are
35 new and surprising to the user. This method, described
36 in more detail in [38], rules out non-sense relations ef-
37 fectively and is able to create natural language expla-
38 nations for the connections.

39 The question to be solved is formulated by making
40 selections on facets about people, professions, places,
41 and generic relation types. For example, the question
42 "How are Finnish artists related to Italy?" is solved
43 by selecting "Italy" from the place facet and "artist"
44 from the profession facet. The results include connec-
45 tions between people and places constrained by the
46 facet selections, e.g., that "Elin Danielson-Gambogi
47 received in 1899 the Florence City Art Award" and
48 "Robert Ekman created in 1844 the painting 'Land-
49 scape in Subiaco' depicting a place in Italy". Finding
50 out hidden "new" semantic associations and their ex-
51 planations like these in a large knowledge graph (over

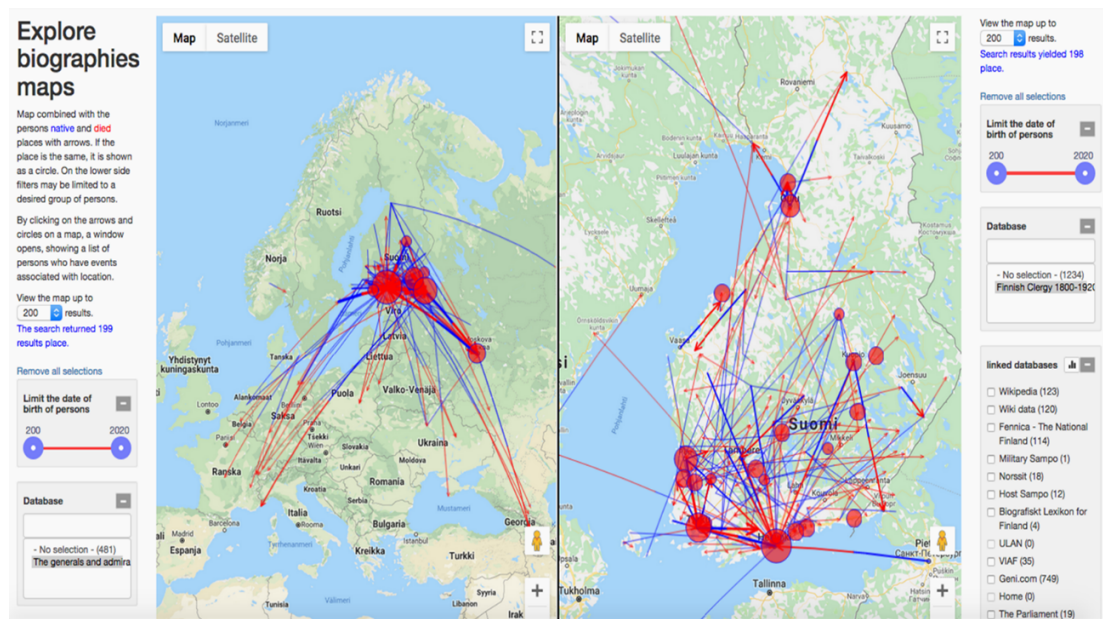


Fig. 2. Comparing the life charts of two target groups, admirals and generals (left) and clergy (right) of the historical Grand Duchy of Finland (1809–1917).

10 million triples), created using the model of Fig.1, can arguably be considered serendipitous knowledge discovery. This makes BiographySampo an example of a third generation semantic portal. Knowledge discovery in this application is performed by transforming the knowledge graph into instances of serendipitous connections and their explanations in a preprocessing phase using rule-based reasoning. After this, relational search can be reduced into faceted search on the connection instances.

6. Conclusions

This paper discussed how focus in developing semantic portals for Cultural Heritage has been evolving during the last 10 years, and proposes and envisions next steps ahead. A three generation model was presented for characterising the process: The first generation systems provided the end user with search and browsing facilities on top of a data service of harmonized linked data (SPARQL endpoint). The second generation systems provide the user also with data-analytic tools that help the Digital Humanities researcher in addressing and solving research problems. In the envisioned third generation systems a step on a new conceptual level towards Artificial Intelligence is taken: the role of the portal is not only to provide tools

for the human researcher to use but also actively and automatically find interesting serendipitous patterns in the data and even solve problems by itself, preferably with explicit explanations. In addition to knowing that the meaning of life is “42”, as suggested by the computer in the novel *Hitchhiker’s Guide to the Galaxy* by Douglas Adams, we also need to know why so.

This shift of research focus from data publishing to data analysis and tooling and finally to Artificial Intelligence brings in novel research challenges in, e.g., knowledge extraction, data visualization, machine learning, knowledge discovery, and computational creativity. Interpreting the results of a tool typically requires a great deal of domain knowledge and understanding the underlying algorithms and the characteristics of the data, such as modeling principles used and completeness, uncertainty, and fuzziness of the data. Using advanced computational tools in Digital Humanities raises the demand for source criticism on a new, higher level.

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