NIF4OGGD - NLP Interchange Format for Open German Governmental Data

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Abstract

In the last couple of years the amount of structured open government data has increased significantly. Already now, citizens are able to leverage the advantages of open data through increased transparency and better opportunities to take part in governmental decision making processes. Our approach increases the interoperability of existing but distributed open governmental datasets by converting them to the RDF-based NLP Interchange Format (NIF). Furthermore, we integrate the converted data into a geodata store and present a user interface for querying this data via a keyword-based search. The language resource generated in this project is publicly available for download and via a dedicated SPARQL endpoint.

Keywords: NLP, NIF, Open Data

1. Motivation

The open data movement has become increasingly important as a key driver for economical success. For instance, the German government has passed an E-Government-Law (Bundesgesetzesblatt, 2013), which emphasises the importance of machine-readable data provided by official agencies and other German government organizations. To use this potential we present our multi-dataset mashup NIF4OGGD.[⁴]

Many open data platforms, however, still provide data in (often proprietary, non-standard) formats that lack machine-readability. There is a growing number of data repositories. For example, the city of Berlin (see Section 2.) provides more than 200 datasets from kindergarten locations across city districts to ozone pollution distributions. While this diversity allows to create a wide range of apps and mashups, the integration of several sources remains a challenging problem. In order to provide a standardized solution to gather and correlate open data documents, we propose to use formats and tools that achieve interoperability between Natural Language Processing (NLP) tools, language resources and annotations. A format that fits these requirements is the Linked Data-based Natural Language Processing Interchange Format (NIF) (Hellmann et al., 2012, Hellmann et al., 2013). We employ NIF to connect three different government data repositories and interlink it with spatial information in the Web of Data. The resulting language resource is published along with a user interface for browsing it.

The contributions of this paper are: (1) we describe a process for creating a novel language resource (LR) comprising several open datasets across Germany, which we (2) made publicly available. Furthermore, we (3) use Linked Data via NIF as multi-lingual interchange format to allow queries across datasets. Additionally, we (4) offer a simple search engine interface for end users. Finally, we (5) provide use cases that show the potential impact of NIF4OGGD.

2. State of the Art of Open German Governmental Data

In general, in Germany it is not allowed to publish person-centered data without the consent of the described persons. Taking this restriction into account, most data portals upload statistical or textual data from public hearings. This data is difficult to understand without an extra layer of structural information and most often proprietary, unstructured, not standardized and thus not machine-readable. We aim to overcome these problems via NIF (NLP interchange format) which is based on the principles of Linked Data. Tim-Berners Lee postulated the 5 Star principle (Berners-Lee, 2011) for sharing open data as Linked Data. The first step is to make data available on the web via an open licence and put it as a second step online in an structured format like Excel. The data will get a third star if it is in a non-proprietary format like CSV and a fourth star if URIs are used to denote resources. Last, it is 5 Star data if it links to other datasets creating a richer context.

We analyzed three local portals providing open data with respect to their 5 Star quality. Moreover, we look for the availability of geodata, SPARQL (Prud’hommeaux and Seaborne, 2008) endpoint availability and whether the portal has a visual interface for analysing the data on-site, e.g., a heat map function. The results can be seen in Table 1. Berlin² is the pioneer portal for open data in Germany. Although it does not provide 5 Star data, it comes up with an API and several well structured and non-proprietary data files about, e.g., public wireless LAN locations, events or a list of all memorials. Overall, there are 289 datasets in 21 categories.

Bonn² is by now not able to deliver content via download or API. The city of Bonn is discussing how and what data should be delivered. A part of the administrative data is available via an administration management system, cf. Section 3.²
Table 1: Different data portals, their 5 Star classification and further features. (✓) means that not all data is available at this particular star level.

<table>
<thead>
<tr>
<th>Portal</th>
<th>✓</th>
<th>✓</th>
<th>✓</th>
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<th>✓</th>
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<td>✓</td>
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<td></td>
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<tr>
<td>Bonn</td>
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<td></td>
<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Cologne</td>
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<td>✓</td>
<td>✓</td>
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<tr>
<td>NIF4OGGD</td>
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<td>✓</td>
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</tr>
</tbody>
</table>

Listing 1: Select all streets of Berlin along with latitude and longitude.

```sparql
PREFIX lgd: <http://linkedgeodata.org/ontology/>
PREFIX geovocab: <http://geovocab.org/geometry#>
PREFIX geo: <http://www.w3.org/2003/01/geo/wgs84_pos#>

SELECT DISTINCT ?s ?streetLabel ?lat ?long
FROM <http://thedatahub.org/dataset/lgd-berlin>
WHERE{
  ?s a lgd:HighwayThing;
  rdfs:label ?streetLabel;
  geovocab:geometry ?geometry.
  ?posSeq ?posSeqP ?posSeqO.
  ?s2 geovocab:geometry ?posSeqO;
  geo:lat ?lat;
  geo:long ?long.
}
```

Cologne offers 172 datasets from 9 categories for download. Although, the portal provides several open license datasets, no dataset exists following the Linked Data paradigm. Like in the case of Bonn, administrative data is available via an administration management system that already has been scraped by the Cologne Open Data Portal.

3. Dataset

In this section, we briefly present LinkedGeoData (Stadler et al., 2012) and our data extraction from E-Government data portals. Moreover, we point out current problems and how we overcome each of them by using NIF and the Linked Data paradigm.

3.1. LinkedGeoData

The OpenStreetMap (OSM) (Haklay and Weber, 2008) project offers a freely available and rich source of spatial data. OSM consists of more than 1 billion nodes and 100 million ways stored in a relational database. LinkedGeoData (LGD) provides a transformation of OSM data into RDF (Auer et al., 2009), which comprises approximately 20 billion triples. LGD is available according to the Linked Data principles and interlinked with DBpedia (Auer et al., 2008) and GeoNames. LGD provides its RDF data not only in form of free dump files, but the data can also be queried via a SPARQL endpoint. LGD provides an ontology for structuring the information in OpenStreetMap. For instance, it contains more than forty subclasses of HighWay. As an example, Listing 1 is a SPARQL query, which retrieves all streets of the city of Berlin, along with latitude and longitude information.

For obtaining the relevant datasets, we downloaded OSM dumps for Berlin and North Rhine-Westphalia and applied the LGD conversion to them. The resulted datasets are stored in the project endpoint.

3.2. Data Extraction

To enrich the spatial data, government data was retrieved from administration management systems of Bonn and Cologne, see Section 2. All of the data is document based, containing PDF documents with administrative decisions and documented enquiries of citizens. Each document serves as a resource in the administration management systems and features a title and minor metadata. To obtain the data, the portals were queried by custom web scrapers. In the case of Bonn, the PDF documents were downloaded via curl (Steinberg, 2012) and converted to textual data via Linux command pdf2text. Because of this procedure, the original formatting of the documents and any information contained in it (like tabular data) was lost. The negative effect of this is limited in our case, since we only perform text search over the documents. However,
keeping this information and performing advanced extraction methods is one of our steps in a larger research agenda. In case of Cologne, the above mentioned open data portal already performed this step and allowed us to use the textual data of the documents and the titles of the resources.

4. Architecture

The NIF4OGGD architecture shown in Figure 1 has three main modules: Conversion of documents to NIF, Enrichment and Visualization & Search. The modules provide a flexible solution to integrate multiple web data sources using (semantic) web standards and NIF.

4.1. Conversion of Documents to NIF

For the conversion to NIF, two resources where established for every document: First, the textual content of a document was added to a resource of the type nif:Context with the nif:isString predicate as a literal. The URL of the source document was also included in the metadata to ensure full traceability of the original data. Second, the title string of the document was in turn added to a resource of the type nif:Title with the nif:anchorOf predicate and linked to the context resource via nif:referenceContext. Listing 2 provides an example of a NIF conversion of a document presented in Figure 3.

4.2. Enrichment

In order to link geographical and governmental data, the governmental data is enriched by geodata via LingPipe Exact Dictionary-Based Chunker (LingPipe, 2013). LingPipe relies on the Aho-Corasick string matching algorithm (Aho and Corasick, 1975). The output of this extracting is stored as standardized NIF files. The collection of LGD locations and government data is additionally stored as a set of documents in a Lucene (Hatcher et al., 2004) index.

Furthermore, we built an in-memory dictionary for the data provided by LGD. This dictionary is used for performing an analysis in all NIF files that contain government data. Specifically in the data stored in nif:isString property, when a location name occurs in a document, all the data is stored in the Lucene document corresponding to that location. Figure 2 shows all fields stored for an indexed document.

4.3. Visualization & Search

Aiming to allow an easy integration of NIF4OGGD into external web processes, we implemented RESTful and SOAP web services for the search process. The web service interface allows access to query a Lucene index and returns JSON and NIF format. NIF4OGGD is deployed as a web service and has a very simple user interface for demonstration. The source code is available at our project repository [13]. The user interface was built using the Google Maps API, allowing to search locations by name. Once the location is selected by the user, the map shows it on a map and displays related government documents stored in the index. It is also possible to annotate the texts using DBpedia Spotlight (Mendes et al., 2011) and use the annotated resources to discover more information in a graph database. Figure 3 shows the NIF4OGGD user interface.

5. Use Cases

In this section, we outline application scenarios and use cases for our system.

Data Retrieval. An example use case for data retrieval are citizens searching for events in their neighbourhood. NIF4OGGD provides data about what governmental events happen in a specific area. For instance, Figure 3 shows a user query for governmental documents mentioning Neumarkt in Cologne. As shown, there are many documents about the search topic, in which the user is free to browse.

Interoperability using NIF. The aligned governmental documents using NIF representation enable searching for the same entity across different authorities’ documents. NIF’s interoperability capabilities make it easy to query all

occurrences of a certain text segment in all available documents without the need of any additional indexing. Especially, sophisticated SPARQL queries can posed to interlinked NIF documents, e.g., to find all places across Berlin where demonstrations were declared about some particular political issue.

Listing 3 introduces an example of using NIF to retrieve all documents mentioning Baubeschluss.

Information Aggregation. Using DBpedia Spotlight, NIF4OGGD annotates the governmental documents using the cross-domain dataset of DBpedia, which provides added value to the data. As an example, in Figure 3 user can use the annotated entities of the presented document to get detailed data about Köln Innenstadt.

6. Conclusion and Future Work

NIF4OGGD is a novel language resource providing geographical data that is interlinked with governmental information. We presented an extraction process for creating this language resource and made it freely available. In addition, our project constitutes a new central sharing point for Open German Governmental Data which is published following the 5 Star principles. In the future, we plan to extend our datasets and enrich it with more data from the LOD-cloud. Furthermore, we plan to employ additional NLP-algorithms, e.g., to just show documents with a valid time range, to provide more trustful documents via NIF4OGGD.

7. References


Sebastian Hellmann, Jens Lehmann, and Sören Auer. 2012. Linked-data aware uri schemes for referencing text frag-
Figure 3: Searching for governmental documents mentioning Neumarkt in Cologne

Listing 3: List of all occurrences of “Baubeschluss” using NIF

```
PREFIX str: <http://nlp2rdf.lod2.eu/schema/string/>

SELECT ?document { 
?textSegment str:referenceContext ?s; 
str:anchorOf "Baubeschluss".
}
```


Daniel Steinberg. 2012. libcurl: The multiprotocol file transfer library.

Acknowledgments

Parts of this work were supported by the ESF, the Free State of Saxony and the EU project GeoKnow (GA no. 318159).