SPARKLIS: An Expressive Query Builder for SPARQL Endpoints with Guidance in Natural Language

Sébastien Ferré *,**

IRISA, Université de Rennes 1, campus de Beaulieu, 35042 Rennes cedex, France

Abstract. SPARKLIS is a Semantic Web tool that helps users explore and query SPARQL endpoints by guiding them in the interactive building of questions and answers, from simple ones to complex ones. It combines the fine-grained guidance of faceted search, most of the expressivity of SPARQL, and the readability of (controlled) natural languages. No knowledge of the vocabulary and schema are required for users. Many SPARQL features are covered: multidimensional queries, union, negation, optional, filters, aggregations, ordering. Queries are verbalized in either English or French, so that no knowledge of SPARQL is ever necessary. All of this is implemented in a portable Web application, SPARKLIS, and has been evaluated on many endpoints and questions. No endpoint-specific configuration is necessary as the data schema is discovered on the fly by the tool. Online since April 2014, thousands of queries have been formed by hundreds of users over hundreds of endpoints.

Keywords: semantic search, SPARQL endpoint, query builder, faceted search, natural language

1. Introduction

A wealth of semantic data is accessible through SPARQL endpoints. DBpedia alone contains several billions of triples covering all sorts of topics (e.g., people, places, buildings, species, films, books). Although different endpoints may use different vocabularies and ontologies, they all share a common interface to access and retrieve semantic data: the SPARQL query language [22]. In addition to being a widely-adopted W3C standard, the advantages of SPARQL are its *expressivity*, especially since version 1.1, and its *scalability* for large RDF stores thanks to highly optimized SPARQL engines (e.g., Virtuoso, Jena TDB). Its main drawback is that writing SPARQL queries is a tedious and error-prone task, and is largely unaccessible to most potential users of semantic data.

Our motivation in developing SPARKLIS¹, shared by many other developers of Semantic Web tools and

applications, is to unleash access to semantic data by making it easier to define and send SPARQL queries to endpoints. The novelty of SPARKLIS is to combine in an integrated fashion different search paradigms: Faceted Search (FS), Query Builders (QB), and Natural Language Interfaces (NLI). That integration is the key to reconcile properties for which there is generally a trade-off in existing systems: user guidance, expressivity, readability of queries, scalability, and portability to different endpoints.

Section 2 discusses related work on making semantic search more usable. Section 3 presents the principles and architecture of SPARKLIS, and Section 4 illustrates them on a concrete navigation scenario on DB-pedia. Section 5 precisely states the capabilities and limitations of SPARKLIS, and Section 6 reports experimental evaluation and impact on real usage worldwide. Finally, Section 7 concludes and draws perspectives. This paper focuses on the functional and nonfunctional properties of SPARKLIS, as perceived by users. Technical details about its internal working can be found in previous papers [6,5].

^{*}Corresponding author. E-mail: ferre@irisa.fr.

^{**}This work is supported by ANR project IDFRAud.

¹Online at http://www.irisa.fr/LIS/ferre/sparklis/

2. Related work

There are mainly two approaches to make semantic search more usable: user interaction (UI) and natural language (NL). UI-based systems reuse and adapt UI paradigms to semantic data: hypertext browsing (e.g., Fluidops Information Workbench²), query builders (e.g., SemanticCrystal [13]), faceted search (FS) [21], or OLAP [3]. Query builders generally offer more expressivity, but lack readability because they are based on formal languages. Moreover, their guidance is mostly based on syntax, and sometimes on a data schema, but not on actual data, like in FS. Most FS-based systems do not claim for a contribution in term of expressivity, and contribute either to the design of better interfaces and visualizations, or to methods for the rapid or user-centric configuration of faceted views: e.g., Ontogator [16], mSpace³, Longwell⁴. Similarly, OLAP-based systems emphasize visualization, and require substantial amount of configuration to extract cubic views over RDF graphs: e.g., Cubix [19], Linked Data Query Wizard [11]. Therefore, their contributions are somewhat orthogonal to ours, and could certainly complement them. A few FS-based systems extend faceted search expressivity: e.g., Slash-Facet [10], BrowseRDF [20], gFacet [9], VisiNav [8], SemFacet [1], OpenLink FS⁵, Vinge Query&Explore⁶. While more expressive than classical FS, those systems are still much less expressive than SPARQL 1.1, and approximately cover basic graph patterns. None of them support union, negation, or aggregation. All except Vinge Query&Explore present only lists of results, rather than tables. That expressivity is reflected by the frequent choice to use trees and graphs to represent the query. Those representations have a good match with SPARQL graph patterns, but do not scale well to express union, negation, or aggregations, unlike natural language.

Natural Language Interfaces (NLI) [14] use NL in various forms, going from full natural language (e.g., PowerAqua [15]) to mere keywords (e.g., NLP-Reduce [13]). In between, there are also controlled natural languages (e.g., Ginseng [13], SQUALL [4]). Systems based on full NL or keywords devote most of their effort to bridging the gap between lexical forms

and ontology triples (mapping and disambiguation), and process only the simplest questions, i.e., they generate SPARQL queries with only one or two triples. Most of them support none of aggregations (e.g., counting), comparatives, or superlatives, even though those features are relatively frequent (see QALD-3 challenge [2]).

Some systems integrate the UI approach in NLIs to alleviate the *habitability problem* [13], in which users have not a precise knowledge about what can be understood by the NLI system, and therefore can be frustrated by syntax errors or empty results. Ginseng [13] uses auto-completion based on the grammar and an ontology. Atomate [23] uses a Controlled Natural Language (CNL) and dynamic forms to guide users in the definition of reactive rules. Those systems can be seen as query builders based on a controlled natural language. They improve the former with readability, and the latter with guidance, but they still lack the finegrained guidance of FS that is necessary to fully solve the habitability problem.

In our objective to reconcile guidance and expressivity, we wanted to avoid some unsatisfactory effects that generally result from the loose integration of two existing approaches. First, we wanted guidance from the beginning, rather than asking users to start with a category (e.g., BrowseRDF), an entity (e.g., Vinge Query&Explore), or a keyword search (e.g., OpenLink FS). Guidance from the beginning avoids the writer's block (blank field with no suggestion), and initial suggestions provide an overview over the whole dataset to users, which is especially valuable when the dataset is unknown. Second, we wanted to directly explore RDF graphs, rather than first extracting a dataset on which standard techniques can then be applied, e.g. faceted search on extracted facets over a selected collection of items (e.g.. Ontogator), OLAP on an extracted data cube (e.g., Linked Data Query Wizard, Cubix). The latter approaches generally require an expert for the extraction part, and strongly limit expressivity for end users. Finally, we want to avoid the habitability problem [13], found in NLIs, in which users have not a precise knowledge about what can be understood and answered by the system, and therefore can be frustrated by syntax errors or empty results.

3. Principles and architecture

This section explains how SPARKLIS integrates the principles of Faceted Search (FS), Query Builders

²http://iwb.fluidops.com/

³http://mspace.fm/

⁴http://simile.mit.edu/wiki/Longwell

⁵http://dbpedia.org/fct/facet.vsp

⁶http://www.vingefree.com/querybyexplore/

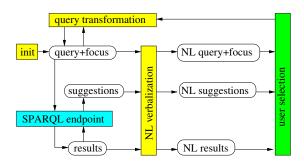


Fig. 1. The system architecture of SPARKLIS

(QB), and Natural Language Interfaces (NLI). That integration is summarized by Figure 1, which shows the system architecture of SPARKLIS. The state of the system is determined by the query and focus, where the focus refers to an entity in the query, and is used as an insertion position for applying query transformations. The SPARQL endpoint is used to compute results from the query, and to compute a set of suggestions from the query and results. Each suggestion is a query element that can be inserted at the focus to refine the query: i.e. an entity, a class, a property, or an operator (e.g., an aggregation). The query, suggestions, and results are verbalized in NL for rendering to the user in a readable form. The user can select a suggestion, or activate a control in the query or results, to trigger a query transformation, and reach a new system state.

SPARKLIS re-uses and generalizes the interaction model of FS [21], where users are guided step-bystep in the selection of items. In FS, at each step, the system gives a set of suggestions to refine the current selection, and users only have to pick a suggestion according to their preferences. The suggestions are specific to the selection, and therefore support exploratory search [17] by providing overview and feedback during the search process. To overcome expressivity limitations of FS, and of existing FS extensions for the Semantic Web (e.g., gFacet [9], VisiNav [8], SemFacet [1]), we have generalized it to Query-based Faceted Search (QFS) [6], where the selection of items is replaced by a structured query. The latter is built step-by-step through the successive choices of the user. This makes SPARKLIS a kind of Query Builder (QB), like SemanticCrystal [13]. QBs have the advantage to allow for a high expressivity while assisting users about syntax by listing eligible constructs at each step. In particular, this enables to completely avoid syntax errors. However, the FS-based guidance of SPARKLIS is more fine-grained than in QBs. In fact, SPARKLIS only allows the building of queries that do return results, preventing users to fall on empty results. Note that non-empty results necessarily imply no syntax error, and no vocabulary error as a consequence. That is because system suggestions are computed for the individual results, not for their common class. In fact, SPARKLIS is as much about building answers as about building questions.

To overcome the lack of readability of SPARQL queries for most users, SPARKLIS queries, suggestions, and results are verbalized in natural language so that SPARQL queries and variables never need to be shown to users (see [5] for more details). This makes SPARKLIS a kind of Natural Language Interface (NLI), like PowerAqua [15]. The important difference is that questions are built through successive user choices in SPARKLIS instead of being freely input in NLIs. SPARKLIS interaction makes question formulation more constrained, slower, and less spontaneous, but it provides guidance and safeness with intermediate answers and suggestions at each step. Moreover, it avoids the hard problem of NL understanding: i.e., ambiguities, out-of-scope questions. A few NLI systems, like Ginseng [13], are based on a controlled NL and auto-completion to suggest the next words in a question. However, their suggestions are not fine-grained like with FS, and less flexible because they only apply to the end of the question. In SPARKLIS, questions form complete sentences at any step of the search; and suggestions are not words but meaningful phrases (e.g., that has a director), and can be inserted at any position in the current question. That current insertion position is called focus, and suggestions are specific to each insertion position.

In order to address scalability issues, only a limited number of results are retrieved from the query, and only a limited number of suggestions are computed from the partial results. The negative consequence of partial results and suggestions is that suggestions may be incomplete, making some queries unreachable. To restore some form of completeness, we have designed an intelligent auto-completion. Autocompletion is a well-known user interface mechanism that provides guidance and feedback, and has already been adapted to semantic contexts [12,7]. SPARKLIS auto-completion is directly available at the top of each suggestion list, and dynamically filters suggestion lists at each keystroke for immediate feedback. It is intelligent in that it uses a cascade of three stages to ensure completeness relative to user input. At stage 1, the partial list of suggestions is filtered on the client side, which can be done efficiently. At stage 2, if the filtered

Table 1
Navigation scenario in SPARKLIS over DBpedia

Step Query

- 1 Give me something
- 2 Give me a writer
- 3 Give me a writer that has a nationality
- 4 Give me a writer that has nationality Russians
- 5 Give me a writer that has nationality Russians and that has a birth date
- 6 Give me a writer that has nationality Russians and whose birth date is after 1800
- 7 Give me a writer that has nationality Russians and whose birth date is after 1800 and that is the author of something
- 8 Give me a writer that has nationality Russians and whose birth date is after 1800 and that is the author of a book
- 9 Give me a writer that has nationality Russians and whose birth date is after 1800 and that is the author of a number of book
- 10 Give me a writer that has nationality <u>Russians</u> and whose birth date is after 1800 and that is the author of **the highest-to-lowest** number of book
- 11 Give me a writer that has nationality Russians or <u>something</u> and whose birth date is after 1800 and that is the author of the highest-to-lowest number of book
- 12 Give me a writer that has nationality Russians or **Russia** and whose birth date is after 1800 and that is the author of the highest-to-lowest number of book

list gets empty, the list of suggestions is re-computed by sending to the SPARQL endpoint a new query that includes the user filter. This means that the same partial query results are used, but a constraint is put on the expected classes and properties. At stage 3, when the filtered list is still empty, new queries are again sent to the SPARQL endpoint, using the full SPARQL query instead of the partial results, in addition to the user filter. This ensures that all query results are used in the computation of suggestions.

To promote portability, SPARKLIS is entirely based on Web standards. It uses SPAROL endpoints for RDF storage and querying, HTTP requests to query them, JavaScript (JS) for the application code, and HTML5/CSS3 for the user interface. SPARKLIS only needs the URL of an endpoint to explore it, without any further required configuration. Users can however customize the number of results to be retrieved, or the labelling properties to verbalize URIs. Queries to SPARQL endpoints are sent directly from the client browser, using AJAX requests. It makes SPARKLIS independent from a server, hence trivial to deploy, and efficient because all application code runs on the client. When a SPARQL endpoint does not enable cross-domain AJAX requests, another endpoint can be used as a proxy, based on the SERVICE feature of SPARQL. For code safety and development speed, the JS code is compiled from a high-level strong-typed language (OCaml using js_of_ocaml⁷). The source code counts about 5000 lines of code, and the minimized JS code weights about 340k.

4. Navigation scenario on Core DBpedia

We detail in this section a navigation scenario using SPARKLIS on Core English DBpedia, the subset of DBpedia limited to classes and properties of the DBpedia ontology, and to English labels. The same scenario can be played on the full DBpedia but with more noisy suggestions because it includes all data from Wikipedia infoboxes. When SPARKLIS is loaded, the Core English DBpedia is the default endpoint, and users can switch to DBpedia, Live DBpedia, or a few other predefined endpoints. Users can also enter the URL of the endpoint of their choice (see at the top of Figure 2).

Table 1 shows the successive queries, as verbalized in SPARKLIS, of a navigation scenario that leads the user in 12 steps to a list of "Russian writers born since 1800, and ordered by decreasing number of written books". That scenario is only one of several possible scenarios leading to the same results: e.g., the birth date could have been constrained before the nationality. At each step, the bold part represents the newly inserted query element, chosen by the user at the previous step among system suggestions, and the underlined part represents the query focus that is used for the next query transformation. The query focus is moved simply by clicking on different parts of the query. The query elements that are suggested for insertion at query

⁷http://ocsigen.org/js_of_ocaml/

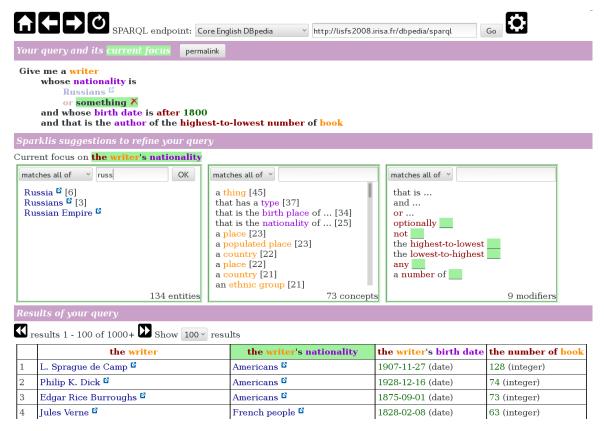


Fig. 2. SPARKLIS screenshot at step 11 of scenario in Table 1

focus can be entities (e.g., Russians), classes (e.g., a book), properties in both directions (e.g., is the author of, has birth date), filters (e.g., after 1800), and various modifiers (e.g., number of, or).

Figure 2 is a SPARKLIS screenshot at step 11, during the specification of an alternative nationality for the writer (Russia as a synonym nationality of Russians). The user interface is made of three parts: top, middle, bottom. The top part shows the current query, and highlights the current focus, here something. The first branch of disjunction is transparent to reflect the fact that it is ignored during the construction of the second branch. The bottom part is the result table of the current query, with a column for each entity/value in the query (here: writer, nationality, birth date, and number of books). The focus column, here the nationality, is highlighted. The middle part contains relevant query elements for insertion at the query focus. It is split in three lists. The first list contains entities (URIs) and values (literals) found in the focus column. It also enables the construction of filters over values. The second list contains *concepts* (classes and properties) that apply to entities/values found in the focus column. The third list contains modifiers that are applicable to the query focus, such as Boolean connectors, aggregation operators, and ordering. Each list provides filtering for quickly locating a query element, and auto-completion for retrieving more matching elements from the endpoint. To save resources, auto-completion is not triggered at each character input, but only for trailing spaces, i.e. at the end of input words. In addition to selecting a query element to insert it at query focus, the query part under focus can be deleted by clicking the red cross in the query.

Figure 3 is another SPARKLIS screenshot at final step 12. The current focus and suggestions have been hidden by clicking on the query head "Give me", in order to emphasize the query results. Figure 4 shows the SPARQL translation of the current query, which is sent to the endpoint, and is displayed at all times at the bottom of the page for the curious users or for reuse in other Semantic Web tools. Note that the layout, URI abbreviations, and variable names have been designed so as to make those SPARQL queries more

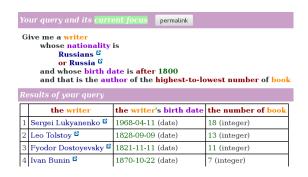


Fig. 3. SPARKLIS screenshot at final step 12, hidding suggestions

Fig. 4. The SPARQL translation of the query at step 12

human-readable. When users want to come back frequently to the current query, they can get a *permalink* to reach the current query and results in one click (see button above the query in Figure 3). Entities and values in the table of answers can be inserted in the query, at the focus corresponding to their column. For example, selecting the birth date 1968-04-11 in Figure 3 would replace the constraint after 1800 by the value 1968-04-11. It also works for aggregated values, so that retrieving all writers having written a given number of books is possible.

5. Capabilities and limitations

We here detail the capabilities and limitations of SPARKLIS w.r.t. five important properties for user experience: expressivity, guidance, readability, scalability, and portability.

5.1. Expressivity: large subset of SPARQL 1.1

SPARKLIS is concerned with the SELECT and ASK forms of SPARQL queries. It covers many features of SPARQL: basic graph patterns, includ-

ing cycles; simple filters on strings, numbers, dates and times, language tags, and datatypes; UNION patterns; OPTIONAL patterns; NOT EXISTS constraints; ORDER BY clauses; multiple aggregations with GROUP BY and HAVING clauses. Triple patterns with a variable in predicate position are also covered through the suggestions that has a relation to ... and that has a relation from ..., where a relation stands for a property placeholder. The focus can then be put on it, and all kinds of constraints can be applied to it through SPARKLIS navigation. Blank nodes are correctly handled. They are shown in results, and suggestions are given about them, but they cannot be inserted in the query.

From a linguistic point of view, SPARKLIS verbalized queries include: noun phrases, verb phrases, and relative clauses for graph patterns; coordinations for relational algebra (UNION, NOT EXISTS, OPTIONAL); superlatives for ordering; determiners for aggregations; and prepositions for some filters. All 100 training questions over DBpedia of the QALD-3 challenge [2] fall in the expressivity scope of SPARKLIS. In practice, 9 questions could not be built because they are "out of scope", i.e. have no answers in DBpedia.

SPARKLIS does not address SPARQL queries that return graphs (DESCRIBE and CONSTRUCT), nor updates. The SPARQL features that are not yet covered are: GRAPH patterns to select named graphs; nested queries, and in particular nested aggregations; expressions to perform computations; and transitive closures of property paths.

5.2. Guidance: safeness and completeness

There are two important properties for guidance in query building: safeness and completeness. A safe guidance avoids dead-ends (i.e., empty results) by providing only relevant suggestions, i.e. suggestions that match actual data. A complete guidance fulfills the expressivity potential by providing all relevant suggestions. In a previous work [6], we formally proved the theoretical safeness and completeness of Querybased Faceted Search (QFS), on which SPARKLIS is founded. For scalability reasons, only a subset of results and suggestions might be computed and displayed in SPARKLIS. Guidance completeness is restored through the intelligent auto-completion mechanism (see Section 3). The query focus plays a crucial role in combining high expressivity (see Section 5.1) and guidance completeness. It determines an insertion position, and the suggestions that are relevant to it. It also provides flexibility in the query construction process, allowing query elements to be inserted in many orderings. When the focus is in the scope of an *opaque* operator (e.g., negation, aggregation), that operator is temporarily ignored in the translation of the query in SPARQL so that results and suggestions can be computed.

The main limitation in guidance is that, for a given query, it gives the same suggestions to everybody, ignoring any user preferences or past usage. This can be a difficulty when the vocabulary is large, like in DBpedia. One could imagine that suggestions that have been selected by the user (or another user of his community) in similar circumstances were ranked higher in the list of suggestions.

5.3. Readability: verbalization in natural language

All content elements of the user interface of SPARK-LIS are verbalized in natural language for readability by users unaware of RDF and SPARQL: the query, the suggestions, the column headers, and the entities in the table of answers. By default, URIs are verbalized by uncamelizing their local name (e.g., dbo:birthDate \leadsto birth date), or by replacing underscores by spaces (e.g., dbr:Barack_Obama \leadsto Barack Obama). Alternately, users can specify in the configuration panel a property, typically rdfs:label, and a language tag to retrieve URI verbalizations from the endpoint. Class and property local names or labels are assumed to be nouns, but a few common patterns are taken into account: e.g., hasParent \leadsto parent, partOf \leadsto (inverse of) part.

To verbalize queries, each query construct is first verbalized locally using a fixed syntactic pattern, and syntactic transformations are then applied globally to make the verbalization look more natural. For example, the first verbalization that has a birth date that is after 1800 is transformed into whose birth date is after 1800. In case of cyclic graph patterns, it is necessary for one part of the query to refer to another part of the query. In order to completely avoid the use of unnatural variables, we verbalize variables as anaphoras. For example, in the SPARKLIS query Give me a film whose director is an actor of the film, the noun phrase the film is implicitly the SPARQL variable used for a film. The same variable verbalizations are used to label column headers. SPARKLIS so far offers verbalization in both English and French, and it should be relatively easy to extend it to other languages that have the same syntactic categories, and no declensions (e.g., Spanish, Italian).

It is certainly possible to improve verbalization at the syntactic level, but the main room for improvement lies at the lexical level. For a real improvement, it is necessary to know for each URI all its possible lexical forms, along with their syntactic category (e.g., noun, verb), number (singular/plural), and gender when relevant. The LEMON vocabulary [18] has been defined to that purpose, but lexicons are rarely available. We think that SPARQL endpoint should include their ontology and lexicon so that lexical information can be retrieved along with data.

5.4. Scalability: billions of triples

SPARKLIS is responsive on the largest well-known endpoint, DBpedia, which has a few billions of triples. For a language as expressive as SPARQL, short response times cannot be guaranteed, but from our experience, time-outs are rarely encountered. Among the 91 QALD-3 questions having answers, half can be answered in less than 30s, and the worst one required 109s (wall-clock time including user interaction and system computations but not user thinking). Only one question could not be answered because of a time-out when searching for a YAGO class by auto-completion.

In fact, the scalability of SPARKLIS is mostly limited by the scalability of the endpoint, because SPARQL requests account for most of the computation time. In particular, the lack of a standard way to perform state-of-the-art full-text search over resources, rather than using FILTER, is an important limitation for efficient auto-completion. Note that it may also be possible to optimize SPARQL engines for the particular kind of queries used by SPARKLIS, especially in the case where many users use it on a same endpoint. So far, federated search is not supported by SPARK-LIS, i.e. a single endpoint can be queried at any given time. Distributed evaluation of queries over several endpoints, although supported by the SERVICE feature of SPARQL, raises the scalability issue one order higher.

5.5. Portability: no required configuration

SPARKLIS conforms to the SPARQL standard, and requires no preprocessing or configuration to explore an endpoint. It entirely relies on the endpoint to discover data and its schema. However, a few settings can be configured to adapt to differences between end-

points. When an endpoint does not allow cross-domain requests from Web browsers (same-origin policy), a default endpoint located at our institution IRISA is used as a proxy. The use of a proxy can be deactivated, and another proxy endpoint can be specified. By default, results to SPARQL queries are cached for improved performance, but caching can be deactivated, for example in case of frequently changing data. The default limit for query results is 200 and can be changed to adapt to the responsiveness of the endpoint. Similarly for the number of suggested classes and properties. The default language of the interface and verbalization is English, and can be changed to French. By default, URIs are verbalized from their local names. For each of entities, classes, and properties, a label property and language tag can be specified for the verbalization of URIs. The latter is more demanding on the endpoint, and requires it to support BIND patterns.

The limits that we have encountered by trying SPARKLIS on many endpoints come either from the implementation of the endpoint, or from its contents. Some endpoints do not support HTTP requests with the POST method or do not support some essential SPARQL features, such as UNION or BIND. A number of endpoints only contain terminological knowledge, and no instances. We recall that SPARKLIS heavily relies on instances for guidance. Another limitation is when the endpoint does not apply RDFS or OWL inference, so that some suggestions are missing. From its client side, SPARKLIS has no easy way to compensate for this lack of inference.

6. Evaluation and impact

In previous papers [6,5], we have presented controlled experiments about the expressivity, scalability, and portability of SPARKLIS, and the guidance of query-based faceted search. The main results are recalled in Section 5. A user study in [6] has shown that students without any prior knowledge of semantic technologies and data were able to answer questions involving several classes and properties, disjunction and negation, tree-shape patterns, and cycles in patterns.

In this paper, we analyze real usage of SPARKLIS worldwide. SPARKLIS has been developed since December 2013, and was put online and publicized since April 2014. It continues to evolve – the last version was released in December 2014 – but its user interface

has remained very stable over time. Several people became regular users, and provided valuable feedback and suggestions. Important improvement came from needs expressed by those people, e.g., support of blank nodes in answers, use of labelling properties for verbalizing URIs. We collected an anonymous usage log from 19/06/2014 to 26/01/2015, where entries are navigation steps. Each step is described by the timestamp, the endpoint URL, the current query, and the user IP. Because IPs are not reliable to identify users and sessions, especially from wireless connection, we also include a random session ID since 29/10/2014. For privacy sake, the user can deactivate the logging (checkbox in the configuration panel), so that the following analysis may underestimate real usage.

As an additional proof of the capabilities of SPARK-LIS, we want to emphasize that SPARKLIS itself was used to explore its own log, and to produce all the results below. The log was translated into RDF, and loaded in a local SPARQL endpoint using the Fuseki server. The most useful features were filters, ordering, and aggregation. What we found missing sometimes are nested aggregations (e.g., average number of steps per session), and the direct visualization of results by charts rather than numeric tables.

Global statistics. Over the 221 days of the log period, SPARKLIS has been loaded 2776 times, hence 12.6 times per day on average. This is consistent with the number of sessions per day on the more recent period with session IDs. The total number of steps (not including the initial loadings) is 11942, hence 54 steps per day on average. Although only seven endpoints are suggested in SPARKLIS, 260 different endpoints have been explored with SPARKLIS. About 6000 different queries have been built overall, which means that queries are visited twice on average. Finally, we have counted 354 different user IPs. This is a rough estimate of unique users because several people may hide behind a same IP (e.g., wireless networks), but it is consistent with the number of views of the YouTube demo (374). It is noticeable that the latter was seen from 38 countries: France (122), United States (38), Italy (26), Germany (26), Japan (19), Russia (17), Australia (15), etc.

Endpoints. Table 2 shows a shortlist of the endpoints that attracted the most users, as counted by the number of user IPs. The most popular endpoint is the default endpoint, the Core English DBpedia hosted at IRISA. It is closely followed by the reference DBpedia end-

Table 2

Most popular endpoints (number of user IPs)

endpoint URL	#user
http://lisfs2008.irisa.fr/dbpedia/sparql	160
http://dbpedia.org/sparql	152
http://data.nobelprize.org/sparql	27
http://live.dbpedia.org/sparql	20
http://drugbank.bio2rdf.org/sparql	15
http://rdf.insee.fr/sparql	11
http://db.lodc.jp/sparql	10
http://lod.euscreen.eu/sparql	9
http://datos.bcn.cl/sparql	8

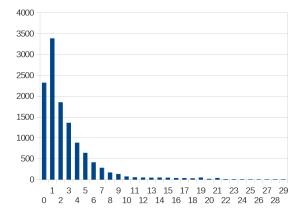


Fig. 5. Distribution of query sizes (number of steps)

point. Other popular endpoints cover various topics, and are hosted in countries worldwide.

Queries. The keywords corresponding to conjunctive queries (i.e., thing, that, has, is, and) are present in thousands of steps each. The keywords corresponding to relational algebra (i.e., or, optionally, not), ordering (i.e., highest-to-lowest, lowest-to-highest), and string matching filters are less frequent but still present in hundreds of steps each. Other filters and aggregations are more rare, and occur in tens of steps.

Figure 5 shows the distribution of query sizes in number of steps. The size of the query is the number of query elements that must be selected to build the query, and is therefore the minimal number of navigation steps to reach it from the empty query (size 0). The query size ranges from 0 to 29, and most queries have a size under 10. Here is an example of query with size 9, built on the bio2rdf endpoint by 9 different users: Give me a drug that has a title and is the drug of a target-relation that has action inhibitor and whose target has a title and has a specific-function. It returns not onlys drugs, but also the related targets and the specific-functions of each tar-

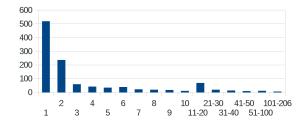


Fig. 6. Distribution of session lengths (number of sessions)

get. Another simpler example with size 3 on DBpedia is: Give me a celestial body that is a galaxy and that has a ngc. It retrieves galaxies along with their NGC number.

Sessions. Figure 6 shows the distribution of session lengths in number of sessions. It reveals that those lengths range from 1 to 206 steps, but that 518 out of 1121 sessions have only one step. This amounts to 46% of sessions, but only 4% of steps. However, 1042 sessions out of 1121 have at least one non-empty query. From our examination of a number of sessions, there are several explanations to the high number of very short sessions. First, about half endpoints do not allow cross-domain requests from browsers, even though they are public, and other endpoints are not responsive enough or do not support essential SPARQL features or have no instances. The cross-domain issue has been solved with a proxy endpoint, but only mid-December 2014. Second, derived projects, at least a customized interface for bio2rdf and a student project to extend SPARKLIS, have generated a lot of testing, i.e. many sessions with 1 or 2 steps.

Table 3 details an example session about cars on Core English DBpedia. It shows an untrained user first learning by trial and error, and finally reaching a useful query of size 6 after 14 steps and less than 5 minutes. At step 1, he tries to filter individual entities with keyword "car", which does not work as expected. At step 3, he finds the class that interests him, an automobile. After exploring tentatively from property class, he seems to really get the trick at step 10, from which he reaches the final query in a straight line. He gets the cars he is looking for, along with a picture of them.

7. Conclusion and perspectives

SPARKLIS provides a stable and working answer to a frequent question in the Semantic Web: "How to explore and query a large unknown endpoint beyond the most simple queries without reading or writing any SPARQL, and without preprocessing or configura-

Table 3

An example session about cars on Core English DBpedia

Step	Time	Query
1	17:05:58	Give me something that matches all of car
2	17:06:07	Give me something
3	17:06:40	Give me an automobile
4	17:07:09	Give me an automobile that has a class
5	17:07:51	Give me an automobile whose class has a thumbnail
6	17:08:18	Give me an automobile that has a class
7	17:08:30	Give me an automobile whose class has a manufacturer
8	17:08:54	Give me an automobile that has a class
9	17:08:56	Give me an automobile
10	17:09:10	Give me an automobile that has a manufacturer
11	17:09:18	Give me an automobile whose manufacturer is Toyota
12	17:09:48	Give me an automobile whose manufacturer is Toyota and that has a transmission
13	17:10:23	Give me an automobile whose manufacturer is Toyota and whose transmission is Super ECT 4-Speed automatic
14	17:10:33	Give me an automobile whose manufacturer is Toyota and whose transmission is Super ECT 4-Speed automatic
		and that has a thumbnail

tion?". Because of a novel interaction paradigm combining faceted search and query builders, user studies have shown that users need to learn how to use it, but that after a short training, they can answer complex queries. Complex queries such as analytical queries involving multiple dimensions, aggregations, and filters, are only a few clicks away, under safe and complete guidance. Moreover, guidance is entirely mediated through natural language.

In the future, we will continue to maintain and improve SPARKLIS, especially w.r.t. expressivity and readability. For expressivity, we plan to cover almost all of SPARQL 1.1, i.e. to add expressions for computations, nested aggregations for rich analytics, and graphs as results for CONSTRUCT queries and updates. For readability, we plan to improve verbalization with lexicons, to extend multi-lingual support, and to offer graphical visualizations of results.

References

- M. Arenas, B.C. Grau, E. Kharlamov, Š. Marciuška,
 D. Zheleznyakov, and E. Jimenez-Ruiz. SemFacet: Semantic faceted search over YAGO. In World Wide Web Conf. Companion, pages 123–126. WWW Steering Committee, 2014.
- [2] P. Cimiano, V. Lopez, C. Unger, E. Cabrio, A.-C. Ngonga Ngomo, and S. Walter. Multilingual question answering over linked data (QALD-3): Lab overview. In P. Forner, H. Müller, R. Paredes, P. Rosso, and B. Stein, editors, *Information Ac*cess Evaluation. Multilinguality, Multimodality, and Visualization - Int. Conf. CLEF Initiative, LNCS 8138, pages 321–332. Springer, 2013.

- [3] E.F. Codd, S.B. Codd, and C.T. Salley. Providing OLAP (Online Analytical Processing) to User-Analysts: An IT Mandate. Codd & Date, Inc, San Jose, 1993.
- [4] S. Ferré. SQUALL: a controlled natural language for querying and updating RDF graphs. In T. Kuhn and N.E. Fuchs, editors, *Controlled Natural Languages*, LNCS 7427, pages 11–25. Springer, 2012.
- [5] S. Ferré. Expressive and scalable query-based faceted search over SPARQL endpoints. In P. Mika and T. Tudorache, editors, *Int. Semantic Web Conf.* Springer, 2014.
- [6] S. Ferré and A. Hermann. Reconciling faceted search and query languages for the Semantic Web. *Int. J. Metadata, Se*mantics and Ontologies, 7(1):37–54, 2012.
- [7] H. Haller. QuiKey an efficient semantic command line. In Knowledge Engineering and Management by the Masses (EKAW), pages 473–482. Springer, 2010.
- [8] A. Harth. VisiNav: A system for visual search and navigation on web data. *J. Web Semantics*, 8(4):348–354, 2010.
- [9] P. Heim, T. Ertl, and J. Ziegler. Facet graphs: Complex semantic querying made easy. In L. Aroyo et al., editor, *Extended Semantic Web Conference*, LNCS 6088, pages 288–302. Springer, 2010.
- [10] M. Hildebrand, J. van Ossenbruggen, and L. Hardman. /facet: A browser for heterogeneous semantic web repositories. In I. Cruz et al, editor, Int. Semantic Web Conf., LNCS 4273, pages 272–285. Springer, 2006.
- [11] P. Hoefler, M. Granitzer, V. Sabol, and S. Lindstaedt. Linked data query wizard: A tabular interface for the semantic web. In *The Semantic Web: ESWC 2013 Satellite Events*, pages 173– 177. Springer, 2013.
- [12] E. Hyvönen and E. Mäkelä. Semantic autocompletion. In *The Semantic Web (ASWC)*, pages 739–751. Springer, 2006.
- [13] E. Kaufmann and A. Bernstein. Evaluating the usability of natural language query languages and interfaces to semantic web knowledge bases. J. Web Semantics, 8(4):377–393, 2010.
- [14] V. Lopez, V. S. Uren, M. Sabou, and E. Motta. Is question answering fit for the semantic web?: A survey. *Semantic Web*, 2(2):125–155, 2011.

- [15] V. Lopez, M. Fernández, E. Motta, and N. Stieler. PowerAqua: Supporting users in querying and exploring the semantic web. Semantic Web, 3(3):249–265, 2012.
- [16] E. Mäkelä, E. Hyvönen, and S. Saarela. Ontogator a semantic view-based search engine service for web applications. In I. F. Cruz et al., editor, Int. Semantic Web Conf., LNCS 4273, pages 847–860. Springer, 2006.
- [17] G. Marchionini. Exploratory search: from finding to understanding. Communications of the ACM, 49(4):41–46, 2006.
- [18] J. McCrae, D. Spohr, and P. Cimiano. Linking lexical resources and ontologies on the semantic web with lemon. In *Extended Semantic Web Conference (ESWC)*, LNCS 6643, pages 245–259. Springer, 2011.
- [19] C. Melo, A. Mikheev, B. Le Grand, and M.-A. Aufaure. Cubix: A visual analytics tool for conceptual and semantic data. In *Int. Conf. Data Mining Workshops*, pages 894–897. IEEE

- computer society, 2012.
- [20] E. Oren, R. Delbru, and S. Decker. Extending faceted navigation to RDF data. In I. Cruz et al, editor, Int. Semantic Web Conf., LNCS 4273, pages 559–572. Springer, 2006.
- [21] G. M. Sacco and Y. Tzitzikas, editors. *Dynamic taxonomies and faceted search*. The information retrieval series. Springer, 2009
- [22] SPARQL11. SPARQL 1.1 query language, 2012. URL http://www.w3.org/TR/sparq111-query/. http://www.w3.org/TR/sparq111-query/, W3C Recommendation.
- [23] M. Van Kleek, B. Moore, D.R. Karger, P. André, and M.C. Schraefel. Atomate it! end-user context-sensitive automation using heterogeneous information sources on the web. In *Int. Conf. World Wide Web*, pages 951–960. ACM, 2010.