Decentralizing the Persistence and Querying of RDF Datasets Through Browser-Based Technologies

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Abstract. Even though linked open data are inherently capable of persisting in a decentralized fashion, the Semantic Web community is currently lacking effective workflows for seamlessly sharing and querying RDF datasets beyond relatively small scales. The most accessible methods for cloning and subsequently querying an RDF dataset are quite involved, often requiring users to locate and download a data dump, install and run a local SPARQL engine, load data into the triplestore, and then query their endpoint using some interface. While these actions may qualify as the first step towards decentralization, the incentive for such a user to then make their local endpoint available to the World Wide Web and thus act as an auxiliary host or 'mirror' is hard to surmise. In this work, we answer the intelligent client challenge by presenting a browser-based RDF query engine coupled with web-based peer-to-peer networks to illustrate the decentralized persistence and use of RDF datasets via intelligent peers.

1 Introduction and Motivation

It is no surprise that in recent years, web services and standalone web applications have been introduced as viable competitors to desktop native applications thanks to innovative Web technologies [7]. When it comes to hosting and querying RDF datasets however, native applications are almost always involved, typically on the server side. While this technique is trivial for centralized architectures, it does not promote the decentralization of persisting and querying RDF datasets. [10] demonstrates the value that decentralization has on the Semantic Web by thoroughly examining the methods and applications found at the intersection of the Semantic Web and Peer-to-Peer. Additionally, as several previous works have demonstrated [12,13], *intelligent clients* play an important role when it comes to improving both the accessibility to RDF datasets as well as the reliability of obtaining query results. In this work, we strive towards the decentralization of persisting and querying RDF datasets by creating *intelligent peers* that may simultaneously act as data publishers, data hosts and data consumers.

2 Peer-to-Peer Communication

Using peer-to-peer (p2p) networks to disseminate large RDF datasets has two main advantages for the Semantic Web community: (1) data publishers who operate with limited resources may circumvent hosting costs once their dataset reaches a sufficient demand, i.e., a substantial number of *seeds*¹ and (2) data consumers should see an in-

¹A peer that is uploading content for other peers

creased availability of datasets correlate with larger *swarm*² sizes. From the data publisher's perspective, a p2p content delivery network (CDN) operates at a fraction of the cost compared to the equivalent centralized approach because a p2p CDN only requires the cost of initial seeds. From the data consumer's perspective, as the number of seeds increases, so does the probability of faster content delivery for any new peer. While other works in the literature have focused on distributed RDF repositories [2,3,1,4], our approach is to deliver the entire dataset, or a complete selection of the dataset, to each peer so that applications can continue to function offline. We have also yet to see a p2p implementation of RDF persistence or querying in the browser.

Web Real-Time Communication (WebRTC)³ is a set of protocols and APIs that is currently being standardized by W3C to enable communication between clients, within the browser, over p2p connections. A demonstration of this technology, already implemented in most major browsers, is made available by the open-source JavaScript library WebTorrent⁴, which offers many of the same features as traditional torrent programs. For example, incoming torrent data are available to use on-arrival, meaning that applications can start operating on fragments of a dataset before the full file is downloaded. Although our prototype does not yet take advantage of this capability, a use case is demonstrated by the streaming video playback on the WebTorrent home page. We employ the WebTorrent library to demonstrate how WebRTC can be used to deliver RDF dataset content to clients within the browser.

3 Querying RDF Datasets in the Browser

While there are several existing browser-based triplestores and query engines [8,11,6,13], the major shortcomings of these JavaScript libraries have to do with the size of datasets they can reliably handle, query performance, or the overhead time of acquiring a dataset prior to querying. The cause for these shortcomings can be narrowed down to the fact that no client-side implementation is making use of a binary serialization format that applies *succinct data structures*. Such serializations are intended to create compact storage footprints while also enabling high-performance query operations without the need for decompression. HDT [5,9] is the pioneer binary serialization format for encoding RDF datasets using compact data structures that can be queried *in-place*, i.e., without transformations. In order to drive forward real-time Semantic Web applications that take place in the browser, intelligent clients should be able to quickly and efficiently transmit, receive and query RDF datasets with little overhead.

As a proof of concept, we provide here⁵ a working prototype that combines our browser-based RDF query engine, a connection to a p2p network, and the use of RDF datasets that have been encoded using a custom binary serialization format based on HDT. The web application initially downloads an RDF dataset by connecting to a p2p network via WebTorrent. We don't go into details about our custom binary serialization format here, but the encoding software is written entirely in JavaScript and is therefore

²All active peers in a p2p network sharing the same torrent

³https://webrtc.org

⁴https://webtorrent.io

⁵http://phuzzy.link/query/

also capable of being used by intelligent peers to create new RDF datasets from within the browser. Result times will vary depending on the machine used to trial our web application demo, but we believe that its performance illustrates the potential that this type of approach has.

In its current state, our query engine works exclusively on basic graph patterns, i.e., sets of triple patterns and filters. In order to generate graph patterns, we designed a JavaScript API that follows a graph traversal query paradigm, much like the Gremlin Graph Traversal Language⁶. This low-level approach of using an API to build queries offers several compelling advantages to us over the more user-friendly alternative of a query language such as SPARQL. First of all, the data structures we are using to store RDF terms and RDF triples, as well as their complementary search algorithms, make the API-built queries straightforward to construct and evaluate. More importantly however, supporting a high-level query language such as SPARQL requires that some library will handle parsing, query planning and query optimization. With a graph traversal query paradigm, the costs associated with these processes are either voided or significantly mitigated by virtue of the query paradigm itself, e.g., parsing is handled by the runtime system directly and query planning is essentially performed by the query author. Reducing costs at this stage of query execution is pivotal to our goal of competing with the time it takes distant endpoints to return query results. In the broader scheme of providing a usable query engine, we wish to start by offering the bare essentials for querying in the form of an API. Support for a more advanced query language such as SPARQL is therefore supplemental and appropriate for future work. A sample query is shown in Listing 1, and its equivalent SPARQL query shown in Listing 2. Documentation about our query API is available online⁷.

```
(graph) => graph.pattern()
    .object('dbo:City').edgeIn('rdf:type')
    .subjects().mark('city')
    .forkOut({
        'dbo:isPartOf': e => e.object('dbr:California'),
        'rdfs:label': e => e.literals()
        .data('language', k => k.in('en')).mark('name'),
        'dbo:populationTotal': e => e.literals()
        .data('number', k => k.is('>=',1e5).is('<',2e5).save('population')),
})
.exit();</pre>
```

Listing 1 A sample query that shows how our low-level JavaScript API is used to create graph patterns. This query selects all cities in California that have a population between 100,000 and 200,000 from a selection of the DBpedia dataset.

```
select * {
    ?city a dbo:City;
    dbo:isPartOf dbr:California;
    rdfs:label ?name;
    dbo:populationTotal ?population .

filter(isLiteral(?name))
    filter(langMatches(lang(?name), "en"))
    filter(isNumeric(?population))
    filter(?population >= 100000 && ?population < 200000)
}</pre>
```

 $^{^6}$ http://tinkerpop.apache.org/gremlin.html

⁷http://phuzzy.link/query/documentation

Listing 2 The equivalent SPARQL query to the graph pattern shown in Listing 1.

To illustrate how such an RDF query engine compares to a traditional triplestore, the query shown in Listing 1 yielded 45 results from a dataset consisting of 4.15 million triples (all things of type $\mathtt{dbo:City}$) in 329ms on a modern laptop while consuming 421MiB in memory, shown in Figure 1, compared to the 45 results obtained in 365ms on a cold, local Fuseki instance with the exact same dataset while consuming 2.43GiB in memory. The most likely explanation for why we see comparable performance and a smaller memory footprint out of our browser-based query engine, and the reason that such comparisons are not fair to draw conclusions from, could be due to the fact that our system operates on a read-only snapshot of an efficiently packed dataset file with pre-built indexes. Nonetheless, the comparable query result times are a good indication that browser-based query engines may indeed have the strength to compete with distant endpoints where clients would normally be confronted by network latency, network throughput, and server-side resource limitations.

► Fetch all rows at once	45 results in 329.785ms		
<pre>(graph) > graph.pattern() c) ject('dbo:City').edgeIn('rdf:type') subjects().mark('city') '.forkOut('dbo:isPartOf': e > e.object('dbr:Californi 'rdfs:label': e > e.literals() data('language', k > k.in('en')).mark('dbo:populationTotal': e > e.literals() data('number', k > k.is('>=', 100e3).i) .exit();</pre>	city	name	population
	dbr:Antioch,_California	"Antioch, California"@en	110542
	dbr:Berkeley, California	"Berkeley, California"@en	112580
	dbr:Burbank, California	"Burbank, California"@en	103340
	dhe Carlahad California	"Carlsbad, California"@en	105328
	dbr:Concord, California	"Concord, California"@en	122067
		"Corona, California"@en	160000
	dbr:Costa_Mesa,_California	"Costa Mesa, California"@en	109960
	dbr:Daly_City,_California	"Daly City, California"@en	101123
	dbr:Downey, California	"Downey, California"@en	111772
	dbr:El_Monte,_California	"El Monte, California"@en	113475
	dbr:Elk Grove, California	"Elk Grove, California"@en	153015
	dbr:Escondido, California	"Escondido, California"@en	143911
	dbr:Fairfield,_California	"Fairfield, California"@en	105321
<pre>select * { ?city a dbo:City; dbo:isPartOf dbr:California; rdfs:label ?label; dbo:populationTotal ?population . filter(isLiteral(?label)) filter(langmatches(lang(?label), "en")) filter(shumeric(?population))</pre>	dbr:Fullerton,_California	"Fullerton, California"@en	135161
	dbr:Garden Grove, California	"Garden Grove, California"@en	170883
	dbr:Glendale, California	"Glendale, California"@en	191719
	dbr:Hayward, California	"Hayward, California"@en	149392
	dbr:Huntington_Beach,_California	"Huntington Beach, California"@en	189992
	dbr:Inglewood,_California	"Inglewood, California"@en	109673
	dbr:Lancaster, California	"Lancaster, California"@en	156633
filter(?population >= 100000 && ?population < 2000	dbr:Moreno Valley, California	"Moreno Valley, California"@en	193365
filter(?population >= 100000 && population < 2000.	dbr:Murrieta,_California	"Murrieta, California"@en	103466
	dbr:Norwalk, California	"Norwalk, California"@en	105549
	dbr:Oceanside, California	"Oceanside, California"@en	167086
	dbr:Ontario, California	"Ontario, California"@en	166866
	dbr:Orange,_California	"Orange, California"@en	136416
	dbr:Palmdale,_California	"Palmdale, California"@en	152750
	dbr:Pasadena,_California	"Pasadena, California"@en	137122
	dbr:Pomona,_California	"Pomona, California"@en	149058
	dbr:Rancho_Cucamonga,_California	"Rancho Cucamonga, California"@en	165269
	dbr:Richmond,_California	"Richmond, California"@en	103701
	dbr:Roseville,_California	"Roseville, California"@en	118788
	dbr:Salinas,_California	"Salinas, California"@en	150441
	dbr:Santa_Clara,_California	"Santa Clara, California"@en	116468
	dbr:Santa_Rosa,_California	"Santa Rosa, California"@en	167815
	dbr:Simi_Valley,_California	"Simi Valley, California"@en	124237
	dbr:Sunnyvale,_California	"Sunnyvale, California"@en	140081
	dbr:Temecula,_California	"Temecula, California"@en	100097
	dbr:Thousand_Oaks,_California	"Thousand Oaks, California"@en	126683
	dbr:Torrance,_California	"Torrance, California"@en	145438
	dbr:Vallejo,_California	"Vallejo, California"@en	115942
	dbr:Ventura,_California	"Ventura, California"@en	108961
	dbr:Victorville,_California	"Victorville, California"@en	115903
	dbr:Visalia,_California	"Visalia, California"@en	130753
	dbr:West Covina, California	"West Covina, California"@en	106098

Fig. 1 A snapshot of the demo interface showing the sample query from Listing 1 on the left evaluated on a selection of the DBpedia dataset, consisting of 4.15 million triples, and the results shown on the right.

As these tools are still under active research and development, we are not yet ready to publish an evaluation of this work. However, we hope to demonstrate the potential implications that the associated methods have. The source code is made available under the development branch on GitHub⁸.

4 Closing Remarks

In this work, we took steps towards decentralizing the Semantic Web from entirely within the browser. We combined the concepts of intelligent clients with peer-to-peer communication to create *intelligent peers*. We introduced our RDF query engine that operates on binary serializations of RDF in an effort to compete with the round trip times it takes distant endpoints to process and return query results to the client. We hope to continue our research in this direction so that datasets persisting in a decentralized architecture can come to fruition within the browser and so that they may be used and queried by data consumers offline.

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⁸https://github.com/blake-regalia/graphy.js/tree/development

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