

RDF Triple Stores for Client-side Applications

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1 Introduction

Client driven web applications are becoming increasingly popular [8]. At the same time REpresentation State Transfer (REST) protocols for datasets have become more standardized, allowing for Javascript frameworks to implement standardized create, read, update and delete (CRUD) functions. This presents some interesting problems when working with current RDF triple stores [7].

2 Material and Methods

2.1 RDF Triple Stores

We have decided to focus on Apache Jena [1], Blazegraph [2], and Sesame [4] RDF stores for our testing due to their support for RDFS reasoning, open source licenses, ability to handle large datasets and REST endpoints for interaction. [10]

Testing was performed on VirtualBox VM [6] with 2x 2.4 GHz 6-core Xeon processors and 8GB of ram. The operating system was CentOS 7 [3] and all of the stores were run through the Java application server Tomcat [5].

Apache Jena uses a Java based front-end called Fuseki these are distributed under the Apache License 2.0. Fuseki can either be run standalone or under a Java application server. For the testing the most recent version Fuseki 2.0.0 under Tomcat was used with RDFS level reasoning.

Systrap's Blazegraph (formerly known as Bigdata) is distributed under either the GPLv2 or a commercial license. It can also either be run as a standalone or in a Java application server. Blazegraph 1.5.1 under Tomcat was used for testing with a triple + inference graph.

Much like Jena, Sesame is a framework that also includes a web front-end and native triple store. Sesame version 2.8.4 was used for testing with a native triple store with RDFS inference.

2.2 Testing Method

We used a modified version of the Lehigh University Benchmark (LUBM) [9] to test performance and capability of the triple stores. We did not use the default testing queries with this dataset as they were more designed to measure the performance of OWL inference models, however this benchmark gave us an arbitrarily large set of data with RDFS classes that we could use for load testing.

We also modified the ontology provided to allow for RDFS inference in cases where OWL reasoning was required.

To service a high performing client-side application the triple store must be able to both return queries quickly and in large volume. We built a testing framework to benchmark the triple stores we were testing. This framework measured three areas, data loading time, large query runtime, and throughput of small queries.

Data load time was measured by converting the LUBM dataset into groups of 10 n3 formatted triples and then measuring both the time it took to load one million of those groups and the average load time per group of triples. These were loaded through the HTTP REST interfaces for each of the stores and the times were measured from the python script that was loading them.

Large queries were defined as a query that would be used to generate lists and as a starting point for more queries that would pull details out. To simulate this we ask for the URIs of all students, professors, classes, departments and universities. Once we have these we produce 1,000 smaller queries to ask for details about the individuals using increasingly complex queries.

There were three small queries that were run 1,000 times initially asking for the details of a single student, then asking for the details for all professors in one university, and finally asking for the details for all students taking classes under a specific professor. The average of these was used to produce the results.

3 Results

The results of the loading test shows a large difference between the materialized inference triple stores Sesame and Blazegraph and the query time inference model of Jena.

Store	Total	Mean	Standard Deviation	Median
Jena	1776	0.013	0.101	0.005
Blazegraph	39341	0.309	0.429	0.012
Sesame	8414	0.066	0.205	0.035
Sesame2	267644*	21.037	20.964	10.883

The second row for sesame is a setting with a higher inference level which greatly slowed down the load time was stopped only 20% through.

The results of the small query performance shows that the inference at query time system Jena has much worse query performance over this large dataset.

Store	Total	Mean	Standard Deviation	Median
Jena	4250	4.257	0.157	4.235
Blazegraph	738	0.033	?	?
Sesame	?	?	?	?
Sesame2	?	?	?	?

4 Discussion

In client-side web applications response time is important for keeping users happy. Understanding why triple stores perform the way they do is helpful in

choosing the correct one for any given project. We have demonstrated the large performance differences that can appear even in relatively small datasets between the different inference models.

References

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