# Saving Large Semantic Data in Cloud: A Survey of the Main DBaaS Solutions

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In the last decades, the evolution of ICT has been spectacular, having a major impact on all the other sectors of activity. New technologies have emerged, coming up with solutions to existing problems and opening up new opportunities. This article discusses solutions that combine big data, semantic web and cloud computing technologies. The authors analyze various possibilities of storing large volumes of data in triplestore databases, which are currently the matter of choice for storing semantic web data. The paper first presents the existing solutions for installing triplestores on the premises and then focuses on triplestores as DBaaS (in cloud). Comparative analyzes are made between the various identified solutions. This paper provides useful means for choosing the most appropriate database solution for semantic web data representation, both on premises or as DBaaS.

Keywords: Big Data, Cloud computing, Semantic Web

# **1** Introduction

L One of the main limitations of the World Wide Web (abbreviated WWW) is that it was not designed to be machine-readable, but only human-understandable. In 1994, at the very first International WWW Conference, five years after he invented the World Wide Web, Tim Berners-Lee introduced the idea of a semantic web that can be understood by machines [1]. WWW Consortium (abbreviated W3C) is an international organization which aims to develop standards for the WWW. W3C was founded by Tim Berners-Lee at the same conference where he announced the need for semantic web [1]. Ever since, one of the objectives of W3C is to improve the WWW, by upgrading it from a web of documents to a web of data. [2]

Until the end of the 20<sup>th</sup> century, the work on semantic web was mostly theoretical; no practical application emerged as an international standard. The first important practical approach consisted in microformats. Instead of using HTML tags for their usual purpose, microformats can be sent as metadata. information annexing understandable by machines. In the meantime, new standards have been developed (e.g. RDF, RDFS, OWL, SPRAQL and JSON-LD) as well as new approaches. [3] In this paper, the authors describe the current state of semantic web, present the main approaches used for semantic web and discuss the most important semantic web solutions for cloud computing. The aim of this article is to present and compare the main solutions for saving semantic data in cloud.

The main semantic web representations are currently based on the semantic triple. A triple is a statement that links two objects, and follows the rule Subject-Predicate-Object. Every part of the triple has a uniform resource identifier (URI) associated with it. Semantic data can be stored as a large graph, where the subject and object are represented as nodes connected by a predicate, in the form of an edge. Any information can be represented by this simplistic model. The maximum potential of this approach can be reached if all the resources on the WWW have an associated URI and are connected with as many other resources as possible by triples. In this scenario, the WWW becomes a large unified database where information from more websites can be automatically extracted and correlated by simple queries. Table 1 shows the format of an URI and describes its components.

| Table 1. UKI Schema   |                                     |  |  |  |  |  |  |
|---|-------------------------------------|--|--|--|--|--|--|
| schema"://[user "@"]host[:port][/path][ "?"query][ "#fragment"] |                                     |  |  |  |  |  |  |
| The descrip   | tion of an URI's components         |  |  |  |  |  |  |
| Components  | Description                         |  |  |  |  |  |  |
| schema  | Access type (HTTP, HTTPS,           |  |  |  |  |  |  |
| schema  | etc.)                               |  |  |  |  |  |  |
| NEOR  | Credential used for accessing the   |  |  |  |  |  |  |
| user  | resource                            |  |  |  |  |  |  |
| host  | Domain name                         |  |  |  |  |  |  |
| port  | Communication channel               |  |  |  |  |  |  |
| path  | Can contain parameters              |  |  |  |  |  |  |
| query   | Specifies the parameters            |  |  |  |  |  |  |
| fuggin out  | Specifies the required parts of the |  |  |  |  |  |  |
| fragment  | document                            |  |  |  |  |  |  |

| Table 1. URI schema |
|---------------------|
|                     |

The standard for representing a semantic triple is the Resource Description Framework (abbreviated RDF). Through RDF, a triple can be represented as a succession of three URIs. There are more ways of representing triples with RDF, called serialization formats. The description language varies from one serialization format to another. The main serialization formats are RDF/XML and JSON-LD. However, Turtle (abbreviation of Terse RDF Triple Language) and N-Triples are worth mentioning, as they can be easily understood by humans. Both RDF/XML and JSON-LD serialization are based on popular standards used to represent and transfer data between applications. XML (abbreviation from Extensible Markup Language) has the advantage of being easy to interpret by applications, but in general it is considered a difficult writing format. JSON (abbreviation of JavaScript Object Notation) is used to represent data structures and transmit them between applications. All-important programming languages support JSON formats. JSON-LD (abbreviation from JSON - Linked Data) is designed to facilitate the representation of RDF relationships. [4]

| Components   | Description   |
|--------------|---|
| Classes      | Collections, sets, types of objects representing an entity in an ontology |
| Individuals  | Class instances   |
| Relations    | Describe how classes and individuals are related                          |
| Attributes   | Properties or parameters of a class or individual                         |
| Restrictions | Dependencies between classes, restricting the set of valid arguments      |
| Rules        | "If-then" statements  |
| Axioms       | Logical assertions  |
| Events       | Changes of values of attributes or relationships                          |

**Table 2.** The components of an ontology

RDF triples are implemented with associated models named ontologies. Ontologies are designed with sets of rules, terms and vocabularies. These sets provide definitions of the entities found in reality. Ontologies are used to develop a large number of applications in different areas, such as knowledge management, intelligent information integration, information retrieval, natural language processing, database design and integration, e-commerce, bioinformatics and education [5]. The main components of an ontology are illustrated in Table 2.

RDFS (RDF Schema) and OWL (abbreviation of Web Ontology Language) are semantic tools used to represent ontologies. They are also called description languages, and define classes and attributes of URIs and their relationships.

#### 2 Semantic Web Databases - Triplestores

A triplestore is a database built to store semantic web data in the form of semantic triples. As RDF is the standard used for Semantic Web, triplestores are also called RDF stores.

This section presents the main issues that should be considered when choosing a database for representing RDF triples. The authors describe the main triplestores available nowadays and present a comparative analysis of the existing solutions. Survey [6] presents the state of art of triplestores in 2014 and performs a qualitative analysis of the main RDF stores available at the time of writing. A similar approach is carried on in this paper, but with an update of criteria used and inclusion of the approaches and solutions developed in the meantime.

To be taken into account for the analysis performed in this paper, RDF stores should include the following components: repository and middleware. A repository is a database constructed or adapted for storing and managing RDF triples. As RDF triples are simple data structures and their XML and JSON serializations are compatible with many programming languages, almost any database can be remodelled to store RDF triples. Therefore, to be regarded as a triplestore and to be able to easily communicate with the repository, the RDF database has to include middleware components (e.g. query engine, RDF parser, APIs for connecting with other applications and storage provider). Triplestores can be implemented both in traditional (SQL) or NoSQL database. NoSQL paradigm seems more suitable for storing and managing RDF triples, since it offers more flexibility. The main NoSQL types of models are: document, key-value, column and graph. RDF stores can be implemented in all these models, but the most reliable one is the graph. [6]

The main technical characteristics to be considered when choosing a triplestore are:

- Semantic web standards supported (e.g. RDF, RDFS, OWL, SPARQL);
- Programming languages supported for connecting the triplestore with other applications;
- Support to reasoning the ability to make logical inferences based on a set of facts and axioms, thus acquiring new knowledge;
- Licenses both commercial and open source solutions are considered;
- Last release date is very important, as old triplestore may not be compatible with more recent applications;
- Operating systems which are compatible with the triplestore.

Table 3 illustrates the main triplestores available in the present together with their technical characteristics.

| Table 3. Technical Characteristics of Triplestores |  |  |                      |                     |                         |                        |  |  |
|--|--|--|----------------------|---------------------|-------------------------|------------------------|--|--|
| Name   | Semantic Web<br>standards  | Programming<br>Languages   | Support to reasoning | License             | Last<br>Release<br>Date | Operating<br>system    |  |  |
| 3Store   | RDF, SPARQL  | C, Lisp  | No                   | Open source         | July 2006               | Linux, Mac,<br>Windows |  |  |
| 4Store   | RDF  | -  | No                   | Open source         | September<br>2009       | Linux/Unix             |  |  |
| AllegroGraph<br>RDF Store                          | RDF, SPARQ,<br>RDFS, OWL,<br>RDFa  | Java, LISP, Pyt<br>hon, Prolog, C#,<br>Ruby, Perl  | Yes                  | Commercial          | March 2018              | Linux                  |  |  |
| Blazegraph<br>(Formerly<br>Bigdata®)               | RDF, SPARQ,<br>RDFS, OWL   | Java   | Yes                  | Commercial,<br>Free | August 2016             | Linux, Mac,<br>Windows |  |  |
| Dydra  | RDF, SPARQL  | Java, LISP, Pyt<br>hon, C++, Ruby  | Yes                  | Free                | September<br>2017       | Linux                  |  |  |
| GraphDB<br>(former<br>OWLIM)                       | RDF, RDFS,<br>SPARQL,OWL   | Java, C#   | Yes                  | Commercial,<br>Free | 2018                    | Linux, Mac,<br>Windows |  |  |
| Semantics.Serv<br>er/Intellidimen<br>sion          | RDF, RDFS,<br>OWL,SPARQL   | ASP.NET,<br>Microsoft<br>Silverlight   | Yes                  | Commercial,<br>Free | June 2010               | Linux, Mac,<br>Windows |  |  |
| MarkLogic  | RDF, SPARQL  | Java,<br>JavaScript,<br>XQuery   | Yes                  | Commercial,<br>Free | 2018                    | Linux, Mac,<br>Windows |  |  |
| Marmotta<br>(Kiwi<br>Triplestore)                  | RDF, SPARQ,<br>SKOS  | Java   | Yes                  | Open source         | December<br>2014        | Linux, Mac,<br>Windows |  |  |
| Intelligent<br>Topic Manager                       | SPARQL, O<br>WL, SKOS  | Java   | Yes                  | Commercial          | 2018                    | Linux, Mac,<br>Windows |  |  |
| Mulgara  | RDF, SPARQ,<br>SKOS  | Java   | Yes                  | Open Source         | January<br>2017         | Linux,<br>Windows      |  |  |
| Open Anzo  | RDF, SPARQ,<br>OWL   | Javascript,<br>Java, .Net  | Yes                  | Open Source         | 2011                    | Linux,<br>Windows      |  |  |
| OpenLink<br>Virtuoso                               | RDF, RDFS, S<br>PARQL,OWL,<br>GRDDL,RDFa,<br>RDB2RDF, R<br>2RML, Direct<br>Mapping | C, C++, Python,<br>PHP, Java,<br>Javascript, C#,<br>ActionScript,<br>Tcl, Perl, Ruby,<br>Obj-C | Yes                  | Commercial,<br>free | April 2016              | Linux, Mac,<br>Windows |  |  |
| StrixDB  | RDF, SPARQL  | C, C++, Lua  | Yes                  | Commercial,<br>free | November<br>2010        | Linux,<br>Windows      |  |  |
| Parliament   | RDF, SPARQ,<br>OWL, RDFS   | Java, C  | Yes                  | Open source         | January<br>2016         | Linux, Mac,<br>Windows |  |  |
| Profium Sense                                      | RDF, RDFS, S<br>PARQL,OWL  | Java, Javascript   | Yes                  | Commercial          | 2018                    | Linux, Mac,<br>Windows |  |  |
| RDFox  | RDF, OWL, S<br>PARQL   | C++, Java  | Yes                  | Open Source         | 2015                    | Linux, Mac,<br>Windows |  |  |
| RDFD Store   | RDF, SPARQL  | Perl, C  | No                   | Open Source         | June 2006               | Linux                  |  |  |

 Table 3. Technical Characteristics of Triplestores

| Rdfstore-js                    | RDF, SPARQL   | Javascript                     | No  | Open Source         | September<br>2016 | NA                     |
|--------------------------------|---|--------------------------------|-----|---------------------|-------------------|------------------------|
| RDF4 (former<br>Sesame)        | RDF, RDFS, S<br>PARQL                                   | Java, Python, P<br>HP          | Yes | Open Source         | December<br>2017  | Linux, Mac,<br>Windows |
| Stardog                        | RDF, SPARQ,<br>OWL                                      | Java, Groovy,<br>Clojure, .Net | Yes | Commercial,<br>free | 2018              | Linux, Mac,<br>Windows |
| Strabon                        | RDF, SPARQL,<br>Geospatial, s<br>tSPARQL, G<br>eoSPARQL | Java                           | No  | Open source         | March 2018        | Linux,<br>Windows      |
| Apache Jena -<br>TDB           | RDF, RDFS, SP<br>ARQL, OWL,<br>GRDDL                    | Java                           | Yes | Open source         | 2018              | Linux, Mac,<br>Windows |
| Oracle<br>Database 12 <i>c</i> | RDF, SPARQL,<br>RDFS, OWL                               | Java                           | Yes | Commercial,<br>Free | 2018              | Linux,<br>Windows      |

## **3** Choosing the best triplestore solutions

The authors compared the data gathered for each triplestore in order to decide which solution is the best to use for storing semantic web data at the moment. Each criterion selected above has to be considered, but obviously not all of them are equally important.

Taking into account that IT technologies evolve very fast nowadays, the authors agree that last release date is can be the starting point for filtering the data. If the release date is not recent there are two issues to consider:

- The possibility that some solutions will not work together due to versions incompatibilities.
- It is less likely that the solution still has support.

The authors consider that the minimum requirement for a medium to big project should be to have the release date in 2017 or more recent and for a small to medium project from 2015 until present. Given these assumptions, there are 10 triplestores to be considered for a medium to big project and 14 for a small to medium one.

The type of license is of course important when choosing a solution. While triplestores with open source licenses offer the advantages of flexibility and the possibility of lower costs (or none), the ones with commercial licenses are more likely to have good support. To offer good flexibility in implementation, the authors consider that the minimum requirements regarding semantic web standards supported by the triplestores should be: RDF, OWL and SPARQL.

The reasoner is an essential component in semantic web. As can be observed in Table 3, most of the triplestores have this type of tool included.

Having more programming languages available can be useful when accessing and managing store contents from local or semantic client application. On the other hand, all modern triplestores are compatible with at least one major programming language, which can be sufficient if proper specialists are engaged in the project.

The authors consider the operating system as the least important criterion chosen. All triplestores are Linux compatible, which should be good enough considering that most of the servers run on Linux distributions.

Considering the above, the authors consider the top triplestores available nowadays to be: AllegroGraph RDF Store, GraphDB (former OWLIM), MarkLogic, Mulgara, Profium Sense, RDF4 (former Sesame), Stardog, Apache Jena–TDB and Oracle Database 12c, in no specific order. Other studies came to similar conclusions, such as [6] or [7]. Future tests for evaluating reasoning capabilities, security and loading speed can further reduce the list of best triplestore. Such tests to compare performances between triplestores have been performed on some of the solutions identified [8] [9].

### 4 Short presentation of the main DBaaS

All the major cloud computing providers (the authors are talking here about the big four: Amazon, Microsoft, Google and IBM - see Figure 1 for details) have in their portfolios different DBMSs available in cloud as Databases as a Service. The advantages of these systems are that they can be much safer than a small-medium company could keep them and that they are scalable. This means that the customer is paying just for what he uses and once the business or the stored data grows, the customer will pay more and more without worrying about data migration, switching technologies or buying new servers. All the major providers have in their portfolios both SQL and NoSQL databases, but because this paper focuses mainly on the storage of semantic data, which is known to be unstructured, the authors will present in detail just the NoSQL databases available in cloud for each one of the providers.



Fig. 1. Public cloud platforms market share (source: https://www.skyhighnetworks.com/cloud-security-blog/microsoft-azure-closes-iaas-adoptiongap-with-amazon-aws/)

## 4.1 Amazon

The cloud computing platform from Amazon, called Amazon Web Services (AWS) gives the user five different DBMSs to choose from: Amazon RDS, Amazon Redshift, Amazon DynamoDB, Amazon ElastiCache and Amazon Neptune. Even though. nonrelational database systems just are DynamoDB, ElastiCache and Neptune (see Figure 2).

|            |            |            |                 | ₿                  | E                       |                   |
|------------|------------|------------|-----------------|--------------------|-------------------------|-------------------|
|            | Amazon RDS |            | Amazon Redshift | Amazon<br>DynamoDB | Amazon<br>ElastiCache   | Amazon<br>Neptune |
| Aurora     | Commercial | Community  | Data Warehouse  | Key Value          | In-Memory<br>Data Store | Graph             |
|            |            | MysqL      |                 | Document           | 📚 redis 🛛 🕅             |                   |
| PostgreSQL | SQL Server | PostgreSQL |                 | Coconcil           | MEMCREHED               |                   |

**Fig. 2.** Databases available in Azure Web Services (source: https://aws.amazon.com/products/databases/?nc2=h\_l3\_db)

DynamoDB is a fast document based or keyvalue store (the user can switch between those two) that uses SSD technologies, auto-scaling and auto-management. It is used by big names in the industry like Airbnb, Lyft and Netflix and accessible directly from code as a local database [10].

ElastiCache is Amazon's in memory data store service. It is fully compatible with Redis and Memcached and provides real-time data processing. It is used by companies like Airbnb and McDonald's [11].

Neptune, on the other side, is a graph database at its core and the only DBaaS that natively supports W3C standards like RDF or SPARQL. Like the other ones from Amazon, it is auto-scalable, fully-managed and ACID compliant [12].

# 4.2 Microsoft

Microsoft's cloud computing platform, called Azure, offers right now the following database solutions: Azure SQL Database, Azure Database for MySQL, Azure Database for PostgreSQL, SQL Data Warehouse, Azure Cosmos DB, Table Storage, Redis Cache and DataFactory. Of these, the authors focused just on Cosmos DB and Table Storage because the other ones are dedicated only to structured, hybrid data or caching.

Cosmos DB is a NoSQL database dedicated for low latency systems. It can be used as a key-value, graph, column family or document store, all-in-one. It is also compatible with APIs like SQL, JavaScript, Gremlin, MongoDB, Apache Cassandra and Azure Table Storage. [13]

Table Storage is a key-value store that uses semi-structured datasets to store large amounts of data. Client libraries for .NET, Java, Android, C++, Node.js, PHP, Ruby and Python are provided by Microsoft, but any other language that accepts HTTP requests is compatible due to the RESTful API. [14]

# 4.3 Google

The cloud computing platform from Google, called generically Google Cloud, provides the user with a variety of six different storage technologies: Cloud Storage, Cloud SQL, Cloud Bigtable, Cloud Spanner, Cloud Datastore and Persistent Disk. From those just Bigtable and Datastore are NoSQL databases, the other ones provide just SQL-based engines or simple file storage.

Cloud Bigtable is Google's NoSQL Big Data database service. According to them it is the same database that is used for the Search, Analytics, Maps, and Gmail products [15]. Its features include high performance for big data, security (the content of the database is encrypted), auto-scaling and automatic rebalancing of the data when cluster nodes are deleted. It is also fully compatible with Hadoop and HBase API.

Cloud Datastore is a highly-scalable NoSQL database for applications, that don't save necessary very large amounts of data [16]. It

is very easy to use and provides also a RESTful API for data access. As for Bigtable, Google is in charge with the auto-scaling and management of the database. Another important feature of Datastore is that it is ACID compliant.

# 4.4 IBM

IBM's cloud platform called simple Cloud (formerly Bluemix) offers four main database engines [17]: Cloudant, Compose, Db2 and Informix. If Cloudant is a simple NoSQL JSON document store optimized for heavy I/O operations, Compose is more like a cloud database engine that can work with different DBMSs like Elasticsearch, etcd, JanusGraph, MongoDB, MySQL, PostgreSQL, Redis, RethinkDB and ScyllaDB. On the other hand, Db2 and Informix are dedicated specially to structured data (the case of the first one) or mixed data altogether – the case of the second one). Cloudant is a JSON document store, that is optimized for handling heavy workloads of concurrent reads and writes in the cloud. It can be a good choice for mobile apps that use features that need offline sync. It is accessible as a RESTful API and fully compatible with Apache CouchDB.

Compose for etcd is a key-value store that uses auto-scaling and has a great focus on data consistency. Similar, Compose for MongoDB brings all the advantages of the BSONs in the cloud.

Compose for JanusGraph is a scalable graph database optimized for storing and querying highly-interconnected data. It is not built-in to be compatible with W3C's semantic data formats, but it is with Apache TinkerPop framework.

Table 4 summarizes all the features of the presented databases with a focus on the accepted semantic web technologies.

| DBaaS                      | Туре                                       | Querying<br>languages                    | <b>RESTful</b><br>access | Other type of access                           | W3C's<br>semantic<br>data<br>formats<br>compatible |
|----------------------------|--|--|--------------------------|--|--|
| Amazon<br>DynamoDB         | Key-Value or<br>Document<br>Store<br>NoSQL | -  | Yes                      | Libraries for main<br>programming<br>languages | No   |
| Amazon<br>ElastiCache      | In-memory<br>NoSQL                         | -  | No                       | Libraries for main<br>programming<br>languages | No   |
| Amazon<br>Neptune          | Graph NoSQL                                | Gremlin, SPARQL                          | No                       | Libraries for main<br>programming<br>languages | Yes  |
| Microsoft<br>Cosmos DB     | Multi-model<br>NoSQL                       | Gremlin,<br>DocumentDB API,<br>Table API | Yes                      | Dependent of the chosen model                  | No   |
| Microsoft Table<br>Storage | Wide column<br>NoSQL                       | -  | Yes                      | Libraries for main<br>programming<br>languages | No   |
| Google Cloud<br>Bigtable   | Wide column<br>NoSQL                       | BigQuery                                 | Yes                      | HBase API                                      | No   |
| Google Cloud<br>Datastore  | Document-<br>Store NoSQL                   | GQL                                      | Yes                      | Libraries for main<br>programming<br>languages | No   |

Table 4. DBaaS systems comparison

| IBM Cloudant                  | Document-<br>Store NoSQL | -       | Yes | Libraries for some<br>programming<br>languages | No |
|-------------------------------|--------------------------|---------|-----|--|----|
| IBM Compose<br>for etcd       | Key-Value<br>NoSQL       | -       | No  | Libraries for main<br>programming<br>languages | No |
| IBM Compose<br>for MongoDB    | Document-<br>Store NoSQL | -       | No  | Libraries for main<br>programming<br>languages | No |
| IBM Compose<br>for JanusGraph | Graph NoSQL              | Gremlin | No  | Libraries for some<br>programming<br>languages | No |

## 5 Picking the winners

The previous chapter showed that every one of the four big cloud computing providers has the capability to offer to the client a solution for saving semantic data. Now the question that can be raised is what to choose, depending on the cloud provider that one is already using. In our opinion, Amazon has a clear winner in the name of Amazon Neptune. As stated in [18] a graph database usually surpasses in terms of performance a key-value or document-store for large amounts of data, especially if the non-graph database doesn't use MapReduce techniques. Additionally, Neptune supports both W3C standards (RDF 1.1 and SPARQL 1.1) and the Apache TinkerPop framework and is ranked 4<sup>th</sup> by the DB-Engines Ranking in the category of RDF stores based on its popularity [7]. It is also placed 8<sup>th</sup> in the graph databases category (Figure 3).

From Microsoft, the winner is **Microsoft Cosmos DB**. Even if it is a hybrid type of NoSQL, it can be configured to work as a graph database that supports the Apache TinkerPop framework. It is ranked 2<sup>nd</sup> in the DB-Engines Ranking in the category of graph databases [19] and not present in the RDF stores categories because it isn't directly compatible with this format.

Google Cloud doesn't provide yet any type of graph database, so in the case of this platform it is necessary to choose between the existing NoSQL databases. Our choice is **Google Cloud Bigtable** because it is designed for large amounts of data and also supports MapReduce techniques that can significantly reduce the querying time [18]. Because Bigtable is neither a graph database nor a RDF store, it is possible to find just in the overall ranking of DB-Engines where it is placed 125<sup>th</sup> [19].

In the case of IBM Cloud, the authors picked **IBM Compose for JanusGraph** as the winner because it is the only NoSQL database from IBM that is a graph-based and because it is compatible with the Apache TinkerPop framework. JanusGraph is ranked 14<sup>th</sup> in the graph databases category by the DB-Engines Ranking (Figure 3).

| Mar<br>2018 2 | Rank<br>Feb<br>2018 | Mar          | DRMG                        |                |             | Score       |             |
|---------------|---------------------|--------------|-----------------------------|----------------|-------------|-------------|-------------|
| 2018 2        |                     | Mar          |                             |                |             | score       |             |
|               |                     | 2017         | DBMS                        | Database Model | Mar<br>2018 | Feb<br>2018 | Mar<br>2017 |
| 1.            | 1.                  | 1.           | Neo4j 🖶                     | Graph DBMS     | 40.90       | +1.08       | +6.58       |
| 2.            | 2.                  | <b>1</b> 4.  | Microsoft Azure Cosmos DB 🗄 | Multi-model 🚺  | 16.76       | +0.57       | +12.81      |
| 3.            | 3.                  |              | Datastax Enterprise 🔂       | Multi-model 🚺  | 8.00        | +2.04       |             |
| 4.            | 4.                  | <b>4</b> 2.  | OrientDB 🔁                  | Multi-model 🚺  | 6.46        | +0.52       | +1.12       |
| 5.            | 5.                  | 5.           | ArangoDB                    | Multi-model 🚺  | 4.12        | +0.65       | +1.74       |
| 6.            | 6.                  | 6.           | Virtuoso                    | Multi-model 🚺  | 1.82        | -0.01       | -0.15       |
| 7.            | 7.                  | 7.           | Giraph                      | Graph DBMS     | 1.08        | +0.06       | +0.02       |
| 8.            | 8.                  |              | Amazon Neptune              | Multi-model 🚺  | 0.70        | -0.05       |             |
| 9.            | 9.                  | <b>4</b> 8.  | AllegroGraph 🖶              | Multi-model 🚺  | 0.64        | +0.00       | +0.16       |
| 10.           | 10.                 | 10.          | Stardog                     | Multi-model 🚺  | 0.53        | +0.01       | +0.16       |
| 11. 1         | 13.                 | 11.          | Sqrrl                       | Multi-model 🚺  | 0.47        | +0.05       | +0.14       |
| 12.           | 12.                 | <b>4</b> 9.  | GraphDB 🖶                   | Multi-model 🚺  | 0.45        | +0.03       | +0.05       |
| 13. 🤘         | 11.                 | <b>†</b> 17. | Graph Engine                | Multi-model 🚺  | 0.43        | -0.01       | +0.33       |
| 14. 1         | 15.                 |              | JanusGraph                  | Graph DBMS     | 0.30        | +0.09       |             |
| 15. 🤘         | 14.                 | <b>†</b> 21. | Sparksee                    | Graph DBMS     | 0.23        | -0.01       | +0.17       |

29 systems in ranking, March 2018

Fig. 3. DB-Engines Ranking for Graph Databases in March 2018

Based on the DB-Engines Rankings and on the supported technologies it can be considered that the overall winner could be Amazon Neptune, but in the absence of a benchmark test and of a monthly average price survey of the whole group of analyzed database, it can be a subjective choice.

### 6 Conclusions and future work

In this article, the authors aimed to identify what DBaaS solution from the large cloud computing providers (Amazon, Microsoft, Google and IBM) can be considered when it comes to save large amounts of semantic data. After the semantic web technologies and domain standards were presented, the authors have made a comparative analysis of the existing triplestores (databases specially created for saving semantic data) available on premises. Unfortunately, none of them is present in the portfolios of the large cloud providers and many of them are not up-todate. For this reason, the NoSQL databases available as DBaaS were analysed and the authors tried to figure out which one of them could be the best match when it comes to save large quantities of semantic data. Even though Amazon Neptune seems to be the overall winner, the authors will focus their future work on doing a benchmark testing on all the identified databases together with a cost analysis.

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