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Analysis and Design of an Efficient NoSQL Data Storage Schema for an Interlinked Recent Activity Feed

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Preface

The idea for this research originally comes from the Open Weblides project and its many little side activities in development. One of the things that has always fascinated me was how the platform would handle a massive influx of users, and specifically how it would relate to the non-critical data storage in the Recent Activity feed. I wanted to find out how a NoSQL data store would be integrated into the flow of information, and what kind of data store would be the most efficient, scalable solution for this problem. My interest was also piqued by the usage of a graph database (Neo4j) in a personal project, and how the data of the Open Weblides project would fit into the graph theoretical model. Digging into this subject while still maintaining my vision on the Ruby on Rails implementation in the platform allowed me to let the question bloom into this research thesis.

This thesis was in part achieved by the support of my promotor, who has given me many tips and tricks, and provided a framework for conducting a proper research. My co-promotor also had an important influence on certain decisions taken in the research and development phase, being a person who is immersed in the relational and non-relational academic database world. Finally, my friends and family also deserve recognition for helping me accomplish this paper, which is the culmination of three years higher education in a fast-moving and innovative field.

Samenvatting

Abstract

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List of abbreviations

BASE Basically Available, Soft state, Eventual consistency. 38

CAP Consistency, Availability, Partition tolerance. 38

RDBMS Relational Database Management System. 25

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1. Introduction

1.1 Context

In this age of computers and smartphones, older and deprecated methods of teaching are quickly being replaced by their digital equivalents. Course content has shifted from being printed in full-text on paper, to static slides on an overhead projector, to the digital canvas present in every modern classroom. However, there is a lot more potential to gain from the modernization of educational content. Education is still too often a one-way street, where students are obligated to process the course content without being provided much challenge or activity. This does not allow for any dialogue to take place between students and teachers concerning feedback and improvement of the course material itself.

Technological constraints in current iterations of educational software do not allow this co-creation discourse easily. Material being locked to specific versions of proprietary software is just one of the many problems teachers might encounter when trying to apply this concept in real life.

By utilizing interactive tools and applications, the teacher can engage the students more directly.

By building on modern, open standards, the Open WebSlides project (Open WebSlides, 2017a) aims to provide a platform that solves these problems. It creates a user-friendly environment where teachers can create courses based on open source technologies and standards, and it allows teachers and students to apply the co-creation narrative easily. This also enables users to share their material not just with their immediate environment but with a much broader educational audience.

The concrete implementation of the Open Weblides platform is built to cater to higher education institutions. This entails that the traffic on the platform will be relatively high yet predictable, and seasonally bound. One of the consequences of this idiosyncratically shaped traffic is the requirement for the platform and infrastructure to efficiently handle these workload peaks.

1.2 Problem statement

The Open Weblides platform incorporates several ways to stimulate spontaneous co-creation between teachers and students. One of the most prominent elements is the *Recent Activity* feed. This reverse chronologically ordered list enumerates the most recent user interactions with the platform and with other users. Feed items range from simple actions such as a user having created or modified course content, to more complex social interactions like a discussion facilitated by comments or a student's changes being incorporated in a teacher's courses.

The size of the activity feed data set is directly correlated to the size and activity of the users. It has the potential to grow explosively, especially in timespans critical to teachers and students such as examination periods. In order for the infrastructure to be able to handle the deluge of the queried information, designing a system that allows for efficient querying and easy scalability is of paramount importance. Further decoupling of this subsystem from the business-critical processes is also important to ensure that downtime of this subsystem has no impact on round the clock availability of course content to the users.

This research thesis will provide a comparative and analytical study of existing NoSQL database solutions for the Open Weblides project to implement an efficient, scalable data storage system in the context of the *Recent Activity* feed. A basic solution is already implemented within the platform, however the proposed data model allows for more flexibility and accommodates any future functionality expansion.

1.3 Research questions

The research in this thesis is focused on finding an answer to three main research questions:

1. What frameworks and software packages currently exist in the industry to store structured non-relational graph or document data?
2. How is the social graph as introduced by the Open Weblides' Recent Activity feed conceptually and logically structured and how is this data consumed?
3. What NoSQL data store is the most appropriate and efficient data store to store this social graph?

Determining and exploring the NoSQL database landscape will provide us with a general idea of the current state of affairs. This knowledge will then be utilized to answer the

second research question in form of concrete data models and queries. Finally, The first two questions will then lead into an analysis to answer the last research question. It will also provide a practical approach for the Open Weblides development team to take into account this research paper when developing the platform in the future.

1.4 Research goal and objectives

Since this research contains both a general, more theoretical and a practical, use-case oriented panel, the goal of this thesis is dual. First, a comparative study of the NoSQL horizon, its applications and vendors, which might be of interest for a broader audience. Second, a concrete recommendation of a NoSQL data store for the Open Weblides project.

1.5 Expected results and conclusions

We expect to find a comprehensive answer to all of the research questions proposed in this chapter. Primarily, this involves finding the best NoSQL data model for the studied use case. We expect to find that graph databases are the most fitting data store in this context. This results from the train of thought that the Open Weblides Recent Activity feed data is structured as a directed graph, and behaves as such. In addition, we also expect to find that RDF triple stores may be an interesting option to entertain, however these are out of the scope of this research. Finally, we predict that relational, NewSQL and other databases that provide ACID guarantees will not be among the most efficient solutions for this use case, due to strong consistency not being a requirement.

In chapter 2 an overview will be presented of current research, applied solutions and other related work in the research domain based on a literature study. This literary review will then be used to give a broad and global overview of the research domain, introducing various relational and NoSQL concepts and explaining the ideas behind NoSQL and polyglot persistence in chapter 3.

Chapter 4 will clarify the analytical research approaches used in this thesis to attempt to formulate an answer to the research questions. The research is split into three distinct parts. The first part, Chapter 5 will evaluate existing NoSQL data stores using a comparative study in the context of the Open Weblides use case. Certain NoSQL categories will also be ruled out in this chapter, and a concrete selection of NoSQL database management systems will be made for further comparison.

The second part in chapter 6 describes the domain model that is used in the Open Weblides project, and proposes the conceptual and logical data models for every included NoSQL database. Example queries that cater to the needs of the domain will also be drawn up and implemented. Chapter 7 combines the proposed data models and reference queries in order to perform and discuss a quantitative study on the data stores selected in chapter 5.

Chapter 8 will conclude and reflect upon the opportunities and inadequacies encountered in the research, providing an entrypoint for future studies.

Finally, chapter 9 will present a general conclusion and summarize the research, formulating answers on all of the research questions.

2. State of the Art

Ever since the rise of the NoSQL databases in 2009 (Sadalage & Fowler, 2012) it has been a subject for vigorous academic and professional research. The contrast with relational databases, optimal use cases, performance and scalability are only some of the aspects that have been analyzed with great regularity. This chapter will summarize previous publications relevant to this thesis.

The book by Sadalage and Fowler (2012) provides an excellent entry into the world of NoSQL. It explains the motivation behind the use of NoSQL techniques, and how this differs from relational data storage. Furthermore it introduces the segregation of NoSQL data stores into four main categories: key-value, document, column and graph databases. Second, Sadalage and Fowler touch the concept of polyglot persistence. This describes the concept of an application using multiple types of data stores to store heterogeneously structured data. This technique is relevant in particular to this thesis, as the describe data schema only relates to one of the database management systems integrated in the Open Webslides platform. The second part of the book provides a more practical approach to using polyglot persistence in an enterprise application. The authors have written down many pointers and guides in order to pick the right database for a particular use case.

There have already been numerous studies to differentiate the different types of NoSQL databases and comparative studies between NoSQL database systems. Nayak, Poriya, and Poojary (2013a) present a fifth NoSQL category: object-oriented databases. Other studies such as Moniruzzaman and Hossain (2013) and Maroo (2013) provide a feature-based comparison for various NoSQL vendors and database systems. Finally, the similarity and resemblance of relational and NoSQL data stores is also a well-researched topic in current literature. Studies and surveys such as Mohamed and Ismail (2014) and Cattell (2010) tackle this subject in great detail.

Grolinger, Higashino, Tiwari, and Capretz (2013) present a use-case based approach to comparing different NoSQL and NewSQL data stores. The survey incorporates a feature-based comparison over different aspects such as querying, scalability and security, and analyzes these concepts in the context of a select number of NoSQL data stores.

Hecht and Jablonski (2011) provides a feature-based comparison of different NoSQL database types and vendors. The researchers compare the data model, querying access, concurrency, partitioning and replication. The paper uses a duality-based approach, where a minus indicates that the feature is not supported by the database system, and a plus if the feature is supported. The paper also presents the problem of a lack of unified querying interface for NoSQL databases. Furthermore, the importance of choosing the right NoSQL database type for the use case is emphasized, however Hecht and Jablonski do not present a specific case study.

The proceedings of the 2013 IEEE International Conference on Big data by Kaur and Rani (2013) describe the theoretical modeling and querying of SQL and NoSQL data stores. The paper then proceeds with a case study of a social networking site similar to Slashdot (Malda & Bates, 1997). Starting from an entity-relationship diagram (ERD), the researchers then proceed by modeling the entities in both a document and a graph database. Finally, a set of seven queries related to the use case is then drawn up and compared for the PostgreSQL, MongoDB and Neo4j data stores.

Zhao (2015) explores the use of NoSQL data stores to store huge amounts of observational data generated by astronomical research. It briefly discusses using filesystems and relational data stores, before comparing NoSQL alternatives. A concrete data model to store the astronomical data in a MongoDB data store is then presented, together with eight scenarios and queries that may be used in a production system. Furthermore, performance measurements of MongoDB are also analyzed. Data insertion, querying and deletion using the aforementioned data scheme and real observational data are used in this section.

The proceedings of the AGILE 2015 conference by Schmid, Galicz, and Reinhardt (2015) present an overview of selected SQL and NoSQL databases, focusing on the geo-functionalities of the systems. It uses performance tests between two document-based NoSQL data stores (MongoDB and CouchBase). The researchers conclude that geospatial calculations in NoSQL database systems are still only supported for basic queries. Relational databases still perform superior to NoSQL databases in small to larger data sets for queries with geo-functions. However the NoSQL response time only increases slightly relative to data set size.

The technical report by Barahmand, Ghandeharizadeh, and Li (2015) quantifies the scalability of MongoDB and HBase for processing simple operations using the social networking benchmark BG (Barahmand & Ghandeharizadeh, 2013). It considers both horizontal and vertical scalability of the data stores using the Social Action Rating (SoAR) introduced by the benchmarking tool. In order to perform these benchmarks, two logical data models for the database design are presented. The report concludes that while both data stores scale superlinearly, their speedup is limited by the resources of a few nodes out of many becoming fully utilized.

Another performance-based study written by Abramova, Bernardino, and Furtado (2014) compares five popular NoSQL databases (Cassandra, HBase, MongoDB, OrientDB and Redis) using the Yahoo! Cloud Serving Benchmark (Cooper, Silberstein, Tam, Ramakrishnan, & Sears, 2010). The study compares read and write query performance. It concludes that over the five compared data stores, MongoDB, Redis and OrientDB are more read-optimized, and Cassandra and HBase are more update-optimized.

A more query-oriented study was performed by Zhou, He, Sheng, and Wang (2013). The paper considers both the academic and the industry definition and description of data models and system architectures. The researchers identify two kinds of searching approaches: the MapReduce-oriented and the SQL-like querying.

The work of Atzeni, Bugiotti, Cabibbo, and Torlone (2016) dives deeper into the world of non-relational data modeling. The paper investigates how traditional data modeling can be used in the context of schemaless and heterogeneous data stores. Atzeni et al. propose NoAM (NoSQL Abstract Modeling), an abstract data model to describe NoSQL databases based on the common surfaces of the various data store types. This technique can be used to describe system-independent application data and later to implement this in the specific data stores, taking advantage of the various target system idiosyncrasies. Further articles by the same authors (Bugiotti, Cabibbo, Atzeni, & Torlone, 2014) expand upon this abstract data model to present a database design methodology for NoSQL systems.

The main differences between this study and the previous studies are:

1. Many studies have been conducted to understand the motivation between the NoSQL principles and the shift from relational data stores. The division of NoSQL data store types into four most commonly recognized categories and elaboration upon this is usually also a topic in these studies. This research paper builds upon that knowledge, providing only a brief introduction in the world of NoSQL and NewSQL concepts.
2. Some of the previously mentioned research papers also discuss a case study applied to a specific use case. This is mostly related to business critical systems that store and process large volumes of data. The use case described in this thesis is very specific in that it's a complementary subsystem that does not affect critical data. Consequently, certain comparative attributes such as security and availability are not considered in this research.

This thesis aims to provide a case study of data storage the Open WebSlides (2017a) platform. Several use case based surveys and studies already exist, however they aim at replacing a relational database in an application with a NoSQL database without bringing polyglot persistence into account. Sadalage and Fowler (2012) is one notable exception in this aspect. In the case of Open WebSlides, the NoSQL data store only complements the relational database and does not fulfill a critical function. Therefore, several constraints such as security and availability differ in interpretation from existing studies.

3. Overview

3.1 Relational data stores

Relational Database Management System (RDBMS) have been the de facto standard for over two decades to efficiently store and query information in a wide variety of native and web applications. These data stores are based on the relational model devised by Edgar Codd. The data in the system is presented as relations: a collection of data consisting of rows and columns. The user of the database can query and manipulate the data using relational operators.

Codd presented thirteen rules for a database in order for it to be considered a *relational database*. However, many of the modern database systems do not adhere strictly to all of these rules. More commonly, a relational database is defined as a database that exposes its information using a collection of rows and columns.

Relational database management systems provide certain guarantees during database transactions. Reuter and Härder introduced the concept of ACID: an acronym referring to a set of transactional properties.

1. **Atomicity**

Each transaction is “all or nothing”, either the transaction completes successfully and the data is mutated in an atomic way, or the transaction fails entirely and none of the data in the database is modified.

2. **Consistency**

Each transaction, when successful, only commits legal results. This entails that the database is always in a consistent state.

3. **Isolation**

Actions within a single transaction are not visible to other, concurrently running transactions.

4. Durability

Once a transaction has completed successfully and been committed to the database, the system must guarantee that these results survive any subsequent malfunction.

3.2 NoSQL data stores

In the early 2000s, a totally different concept of data storage was popularized. NoSQL data stores provide a system of storage that is “non SQL” or “non relational”. More recently, the NoSQL denomination has also been explained as “not only SQL”, pointing on the fact that some data stores provide a different interface besides SQL. Many databases expose a REST API as primary execution interface. Other data stores have devised their own binary communication protocol, such as the Bolt protocol (Neo Technology, 2007a) for the Neo4j graph data store.

The use of non-relational data stores was motivated by the needs of Web 2.0 companies such as Facebook and Google. NoSQL provides a way to store massive amounts of data in a flexible way. The architecture of such systems is usually more simple, and focuses more on horizontal as opposed to vertical scalability. Data is not stored in rows and columns as is the case in the relational model, but rather in a different data structure. Nayak et al. (2013a) divides the data models used by NoSQL data stores into five categories.

3.2.1 Key-value

Key-value data stores are simple in design, yet powerful and efficient when used correctly. A key-value data store allows the user to store schemaless data using a plain, unique key, creating a *key* and *value* pair. Similar to hash tables, the values are stored and queried using the provided key which is used as an index. Modern key-value data stores prefer high scalability over consistency, entailing that more advanced ad-hoc querying and analytical operations on the data such as joins are not available.

3.2.2 Column-oriented

Column-oriented data stores are designed to store data by column rather than by row. This kind of NoSQL data store is more similar to relational database systems than the other categories. In column-oriented data stores, each key is associated with a set of attributes, stored in columns. Concretely this means that the data is indexed by column value, rather than by row.

3.2.3 Document

Document databases store their data in documents, indexed by a unique key. This is technically a subclass of key-value stores, however the difference lies in the interpretation of the data. Since documents are schemaless, each document may have a comparable structure, or a completely different one. In contrast with key-value data stores, the value (document) is not opaque to the database management system, but is interpreted and used for query optimization.

3.2.4 Graph

As the name suggests, graph data stores keep their data in the form of a graph. The graph is made up of nodes and edges, the former being the entities containing the data and the latter being the relations between these entities. Graph data stores use a technique called *index-free adjacency*, where every node contains a logical pointer to the adjacent node. This makes graph traversal a very fast operation. Graph databases are ACID compliant.

3.2.5 Object-oriented

Object-oriented data stores represent the data as an object, closely resembling the concept of an object in object-oriented programming languages. This puts object-oriented data stores much closer to the programming environment than other database systems. It provides all features inherent to object-oriented languages, such as data encapsulation, inheritance and polymorphism. The concepts of class, class attributes and an object can be mapped onto the relational concepts of a table, columns and a tuple. This concept of data storage makes software development more flexible.

3.2.6 Multi-model

Data stores are generally built and optimized around one data model. However, there exist multi-model data stores as well. These database systems allow storing data using any of the different data models mentioned before, while integrating these into the same server package. One example of such a data store is OrientDB (OrientDB Ltd, 2010). Multi-model data stores support and stimulate the principle of polyglot persistence, where an application utilizes more than one type of data store, while reducing its operation complexity.

3.2.7 NewSQL

3.2.8 Triple store

Most NoSQL data stores are built around the concept of *eventual consistency*. This is a consistency model that dictates that all accesses to a particular piece of data will eventually return the last updated value. This principle is broadly implemented in distributed computing systems. Systems providing this property are also classified as BASE: Basically Available, Soft state, Eventual consistency. In contrast to the ACID properties, systems built around the BASE principles prefer availability over consistency.

In 2000, Eric Brewer presented a conjecture known as the CAP theorem (Brewer, 2000). This conjecture, later formally proven (Gilbert & Lynch, 2002a, 2), asserts that it is impossible for a distributed data store to exhibit more than two out of three of the following properties:

1. **Consistency:** Every read operation receives the most recent write result
2. **Availability:** Every request receives a non-error result
3. **Partition tolerance:** The system continues to work despite failure to communicate between nodes

The CAP theorem states that when a network partition is present, the database developer has to choose between providing consistency or availability. Note that *consistency* as defined by the CAP theorem is not the same concept as consistency as described by the ACID properties. Database systems respecting the ACID guarantees choose consistency over availability, while database systems built on the BASE principle generally choose availability over consistency.

NewSQL is a type of relational database management system providing scalability similar to NoSQL data stores, while still maintaining the ACID guarantees. Since this a blanket term, many different architectures exist in NewSQL databases. However, one common feature is their relational model inheritance and their ability to utilize SQL as query interface.

4. Methodology

Analyzing and comparing every NoSQL data storage solution is not feasible due to the sheer number of competing products. Therefore we have made a selection of the most popular NoSQL data stores in the different NoSQL categories that are included in the research. This selection is mainly motivated by a list of the most popular databases kept up to date by Solid IT (2018). For every category, we select the most popular databases by reported score, which is in turn based on several other criteria. DB-Engine Ranking attempts to estimate the overall popularity of the data store on the Web.

Next, the selected data stores are committed to a general comparison. In this comparative study we analyze every solution based on various aspects. The data model, which is the main driving force behind many other aspects, is one of the main focus points. We also focus on the querying capabilities, scaling, partitioning and replication, consistency and concurrency control. Certain aspects are not discussed in detail due to their irrelevance to the presented use case. Examples of these criteria include security, authentication and auditing.

In the following chapter, the conceptual and logical data model of the Open Webslides project is presented. Building upon this, a physical data model is discussed and implemented for the various categories of NoSQL data stores that were selected in the first part of the research. Finally, the characteristic data access flow within the application is examined, and a few reference queries are introduced. These queries are examples of queries that could typically be used to retrieve data from the data store as part of the normal operating procedures of the Open Webslides application.

Finally, the implementation of the data model and the reference queries are used in the last part of the research as a base for qualitative benchmarks. Since the Open Webslides

application is built on Ruby and Ruby on Rails, the benchmarks will be implemented in a Ruby on Rails application, making use of the available Ruby language bindings to the various data stores examined. For every data store used in the benchmark, the most popular ORM gem is used as determined by the RubyGems repository. Related work and previous NoSQL database benchmarks are also considered in this part, however it is referenced only as a baseline due to the fact that previous work does not cover the specific use case and NoSQL data stores studied in this thesis.

5. Data stores

5.1 Selected data stores

Due to the large number of active and maintained NoSQL data stores it is not feasible to consider all of them for this research. We will consider every NoSQL category, discuss the use of the respective categories applied to the use case, and decide whether or not it will be included in the research. In the applicable NoSQL categories the most popular data stores are selected, where popularity is based on certain predetermined parameters.

Solid IT (2018) maintains a list of database systems ranked by popularity, based on parameters such as number of mentions on websites, Google Trends and relevance in social networks. These parameters are also cross-referenced with professional networks such as LinkedIn and Upwork using the number of available job offers and professional profiles. This ranking is called the DB-Engine Ranking. The list includes not only NoSQL databases but also other types of data storage systems such as relational database systems.

The information technology division of Gartner maintains a yearly report on emerging relational and non-relational database technologies.

The DB-Engine Ranking is used primarily to determine which data stores are the most interesting to consider in this research. The Magic Quadrant for Operational Database Management Systems was used as a secondary resource. Due to licensing concerns regarding the Open Weblides project, database systems that do not fall under a free license as classified by the Free Software Foundation (1985) will not be considered.

Certain categories of NoSQL data stores will not be considered. Key-value stores will not be accounted for due to the relative simplicity of this type of data store. The main use case

of key/value stores is not storing more complex information, rather the emphasis lies more on scalability and consistency. Processing complex queries that consist of the NoSQL equivalent of relational joins is not efficient in key-value data stores. Implementing the proposed data model and queries in such a system is a more ambitious task that does not lie within the scope of the research.

Column-oriented data stores are a category of NoSQL data stores that is questionably useful in this case study. These data stores are commonly seen as inverse relational database systems, where the storage of attributes per entity is more flexible regarding nullable values and unstructured information. Column-oriented data stores are very efficient when retrieving a subset of columns for a certain record. Since the data store will be deployed as additional data store next to the authoritative relational database, it will not contain any data that will not be used when querying the database. This workflow cancels out the efficiency and usefulness of column-oriented data stores, and subsequently we will not consider this NoSQL category as a viable candidate.

Furthermore, comparison and application of NewSQL data stores will not be included in this paper either. As described in chapter 3, NewSQL aims to provide scalable performance similar to that of NoSQL while still guaranteeing ACID properties. Since ACID is not a major concern for this use case, NewSQL does not provide significant advantage over NoSQL, and will therefore be omitted from the comparison. Similarly, object-oriented and multi-model database such as OrientDB (OrientDB Ltd, 2010) are not within scope of this research.

In conclusion, document and graph data stores will provide the most beneficial data storage and are the main focus of this research.

The most popular document database systems according to the DB-Engine Ranking are MongoDB (MongoDB Inc., 2009), CouchDB (Apache Software Foundation, 2005a) and Couchbase (Couchbase, Inc., 2010). While Couchbase is technically a multi-model data store, it is being ranked as a document database. However, Couchbase is a conceptual merge between CouchDB and Membase, and mainly adds improved scalability, clustering and auto-sharding. Couchbase will not be investigated upon in a practical capacity, however it is considered as an option to increase multi-host scalability. This leads to MongoDB and CouchDB being the contestants in the document database system category.

Finally, as sole graph database the Neo4j system (Neo Technology, 2007b) stands out greatly over competitors in the ranking. This database system will be analyzed as only graph database in this research.

In conclusion, the data stores included in the study will be MongoDB and CouchDB as document databases, and Neo4j as graph databases.

5.2 Comparative study

Data stores and databases can be compared and analyzed using many quantitative and qualitative aspects. In this section we propose a selection of criteria based on the usefulness applied to the studied use case. Since the landscape and feature-set of NoSQL data stores is changing on a weekly base, the impact of these technologies must be carefully considered in order to reach a durable conclusion in this research.

The features of a data store that are taken into account in this comparative study are:

- Querying capabilities
 - Language
 - Protocol
 - MapReduce
- Programming language
- Language bindings
- Integrity model
- Atomicity
- Revision control
- Consistency
- Persistence
- Partitioning
- High Availability
- Concurrency
- Replication
- License
- Commercial support

Certain aspects are not relevant to the presented use case because of various reasons. Aspects omitted from the study are:

- Security
- Compression
- Full-text search
- Geospatial functionality
- Cloud hosting

The specific use case described in this thesis does not focus on security, since the data store is not public facing and clients do not interact directly with it. Security measures include but are not limited to authentication, authorization, encryption and auditing. None of these are features that are required or useful for the comparison. Authentication and authorization is functionality that is present in all databases. It introduces the concept of multiple clients or roles connecting to a database, and assigning permissions to these clients in order to enforce permissions-based access. However since there will only be one client connecting to the database - the platform itself - deeper integration with LDAP, ActiveDirectory or similar is superfluous and omitted from the comparison.

Encryption refers to the mechanism where data is encrypted and unreadable for unauthorized third parties. Encryption in databases is threefold: encryption of data at rest, client-to-server communication and server-to-server communication (Grolinger et al., 2013). However, since data protection is a comprehensive topic, it does not fall within the scope of this research. Consequently, encryption functionality is not considered in the comparative study.

Database auditing is a facility offered by the database management system that keeps track of the usage of database resources and authorization. Operations on the database leave a trail of events, called an *audit log*. Similarly to authentication, the usefulness of this functionality is somewhat lost when there is only one client operating on the database. However, many security standards such as PCI-DSS and HIPAA require the existence of an audit log.

Compression of data in the database is not included in the comparison. Builtin compression may provide additional storage space but as it is a disk space-CPU usage tradeoff we have chosen to only consider CPU usage. Akin to compression, we leave the choice of cloud hosting up to the database administrator.

Full-text search and geospatial functions are provided by certain data stores, however these features are not used by the platform and subsequently are not relevant to this comparative study.

| | | Querying | | Language | | |
|-----------------|---------|---|--|-----------|----------|--|
| | | Language | Protocols | MapReduce | Language | Language bindings |
| Document stores | MongoDB | JavaScript | MongoDB Wire Protocol REST ¹ | Yes | C++ | C/C++, C#, Java, Node.js, Perl, PHP, Python, Ruby, Scala Erlang ¹ , Go ¹ |
| | CouchDB | REST | HTTP | Yes | Erlang | C/C++ ¹ , Dart ¹ , Go ¹ , Java ¹ , Lua ¹ , Node.js ¹ , Python ¹ , R ¹ , Ruby ¹ , Scala ¹ |
| Graph stores | Neo4j | Cypher SparQL ¹ Gremlin ¹ | HTTP Bolt | No | Java | C#, Java, JavaScript, Python, C/C++ ¹ , Clojure ¹ , Erlang ¹ , Go ¹ , Haskell ¹ , Perl ¹ , PHP ¹ , R ¹ , Ruby ¹ |

Table 5.1: NoSQL data stores: Querying capabilities and language support

¹3rd party community supported

One of the most important factors when deciding on a NoSQL data store is the capability to communicate with the data store, since NoSQL does not have a unified interface like SQL signifies for relational databases. Choosing a certain data store may allow the developer to integrate data storage easier into the application. Every NoSQL data store has its own standard for composing queries, related but not depending on the protocol used for communication with the server. MongoDB allows querying using a JavaScript API, natively over MongoDB's binary protocol or over HTTP (REST) using a third party plugin. Subsequently, MongoDB is popular among server-side applications written in JavaScript and Node.js. CouchDB takes another approach and provides a RESTful interface over HTTP to query, modify and manage the database.

The Neo4j graph store is a different story. The native querying language of Neo4j is called Cypher, and is much closer to SQL than MongoDB or CouchDB's way of querying. Cypher can be used natively both over HTTP and over Neo4j's binary Bolt protocol. Third party plugins can extend Neo4j to provide interfaces for SparQL and Gremlin querying languages.

Another factor that might come into play when choosing a data store is the ability to support MapReduce, explicitly or under the hood. MapReduce is a conceptual framework and implementation for processing large data sets using multiple processing entities. It is composed of a *map* and a *reduce* procedure. The former filters and sorts the data, while the latter performs an aggregating or summarizing operation as a result to the query.

The chosen document data stores both support MapReduce as normal operating procedure. The graph store does not support Neo4j, instead building upon *index-free adjacency*.

Luckily, all major vendors prove to be adequately supported by first- or third-party efforts.

| | | Integrity | | | |
|-----------------|---------|-----------|-------------------|------------------|---|
| | | Model | Atomicity | Revision control | Consistency |
| Document stores | MongoDB | BASE | Document level | No | Configurable: Eventual or strong consistency |
| | CouchDB | BASE | Document level | Yes | Eventual consistency |
| Graph stores | Neo4j | ACID | Transaction level | No | Eventual consistency |

Table 5.2: NoSQL data stores: Data integrity

Even though all of the compared data stores are classified as NoSQL and based on the principle of *eventual consistency*, the integrity model is different for each. MongoDB is designed with the Basically Available, Soft state, Eventual consistency (BASE) integrity model in mind, which prioritizes availability over consistency over data in the Consistency, Availability, Partition tolerance (CAP) theorem. CouchDB is designed around the BASE model as well, however the data store solved the problem in a different way. Documents in CouchDB are stored using a technique called Multi-Version Concurrency Control (MVCC). Neo4j however, enforces ACID guarantees as only data store included in the comparison.

Data stores designed around the BASE model typically do not provide any strong consistency guarantees. This is reflected in the fact that MongoDB and CouchDB provide atomic operations only on document level. This means that operations on documents – and embedded child documents – are atomic, however operations on multiple documents are not guaranteed to be completely atomic. Neo4j is ACID-compliant and does provide atomic transactions for multiple operations akin to the database transactions found in relational data stores.

CouchDB's concurrency control method is based on the technique of MVCC. Subsequently it is the only data store that provides native version control as a result of the concurrency control method used in the database engine. Neither MongoDB or Neo4j provide any similar feature natively.

NoSQL data stores are in general based on the principle of eventual consistency, as tradeoff versus the availability according to the CAP theorem. However, since the start of the development on MongoDB a lot of research and implementation has been done, and the software will soon provide both the option to configure the database to enable strong consistency, and atomic operations on multiple documents – similar to transactions. Both MongoDB and Neo4j provide eventual consistency as dictated by the principles of the CAP theorem.

| | | Scalability | | | |
|-----------------|---------|----------------|--------------------|--------------|---|
| | | Persistence | Partitioning | Replication | Concurrency control |
| Document stores | MongoDB | Memory Disk | Shard key | Master-slave | MVCC (document) Locks (global, database, collection) |
| | CouchDB | Memory Disk | Consistent hashing | Multi-master | MVCC |
| Graph stores | Neo4j | Memory Disk | Cache sharding | Master-slave | Locks |

Table 5.3: NoSQL data stores: Scalability

| | | Features | | |
|-----------------|---------|---|--------------------|---|
| | | License | Commercial support | Cloud environment |
| Document stores | MongoDB | GNU AGPLv3 (database) Apache (drivers) | Yes | Amazon EC2 Google Cloud Platform Microsoft Azure Digital Ocean Cloud hosting partners |
| | CouchDB | Apache | No | No |
| Graph stores | Neo4j | GNU GPLv3 (Community Edition), AGPLv3 (Enterprise Edition) | Yes | Amazon EC2 Google Cloud Platform Microsoft Azure Digital Ocean Cloud hosting partners |

Table 5.4: NoSQL data stores: Hosting concerns

Table 5.4 provides insight into the aspects related to hosting one or more database instances. First, the type of license is important when hosting the data store. Since we have restricted our research to licenses classified as free by the Free Software Foundation (1985), none of these data stores prevent hosting instances without incurring additional licensing costs. However, for two out of three data stores commercial support is available as well. Commercial support includes licensing the product for enterprise use, which usually includes additional functionality – commonly related to scalability – and customer support.

MongoDB is dual-licensed. The database management system itself is licensed under the GNU Affero General Public License version 3. The language bindings – called drivers – while officially supported are licensed under a different license, the Apache license

The GNU Affero General Public License version 3 is a modified version of the GNU General Public License version 3. The GNU General Public License requires developers that modify software licensed under the GPL to distribute their modifications as well. However, the license does not cover the case of service providers: if a developer modifies the source code and runs the software on a server, allowing other users to interact with it, distribution of the modified source code is not required. The AGPL adds a provision to prevent this loophole. Whenever a modified version of software licensed under the AGPL runs on a server, the modified source code must be available to download. In the case of data stores, this makes sure that the software remains free and cannot be commercially exploited.

CouchDB, being an Apache project, is licensed under the Apache license.

The Neo4j graph store is dual-licensed as well: the Community Edition is licensed under the GPL version 3, and the Enterprise edition is licensed under the AGPL version 3. Deciding on data store hosting is a difficult topic, which is not within the scope of this research. We merely enumerate the options available for every data store. Since all products are licensed under a free license, it is possible for individuals and companies to host instances themselves. MongoDB and Neo4j provide commercial support for this use case. However, running a data store instance can also be outsourced to a plethora of companies and hosting providers. Both MongoDB and Neo4j integrate with various cloud providers such as Amazon EC2, Google Cloud Platform or the Digital Ocean hosting platform.

6. Data model

In this section we describe the conceptual domain of this research, and provide a logical data model that fits the domain. For every NoSQL data model selected, we also propose a physical data model. This data model will then be implemented in a Ruby on Rails application using the available language bindings. Finally, a set of reference queries will be proposed, queries that may typically be used in the context of the Recent Activity feed.

6.1 Domain description

On the Open Weblides platform, there is technically no distinction between a teacher's and a student's account. Both of them are represented by the User entity in the database. This entity contains information pertaining to the user, such as email address, first name and last name. The data models described in the next sections will closely follow the existing data model of the platform. However, certain irrelevant attributes are omitted from the model in order to improve readability and simplify abstraction.

A user can create or modify course content in an interactive online editor. The actual course content, formally called a topic, is stored inside a git repository on the filesystem. However, the platform also maintains a record of topic metadata in the relational database. This metadata includes title and description, but also permissions and contributors on the course content. Per illustration, the `title` and `description` attributes are included in the data models.

From a technological perspective, the user has three distinct paths of action for (co-)creation on course content. Since the permission model in the platform is not relevant to this research, we will not go into detail on it. First, the user can directly modify the course

content if the user has permission to do so. The second option is creating annotations on the topic. This allows the user to attach private or public notes to specific content on the topic. Annotations are stored in the relational database, mapped to the `Annotation` entity. The entity contains a logical pointer to the annotated content, which is omitted from the data models. The final possibility to integrate user content into a topic is by adding comments. In contrast to annotations, comments have a typical structure. They can take the form of questions, notes, suggestions, and can also be nested - which allows simple interaction and conversation between multiple users, and effectively enables dialogue between students and teachers.

The intention of this thesis is to use the NoSQL data storage as storage mechanism for the Recent Activity feed. This entails that the authoritative information will not be stored in that data store, but rather be extracted from the relational database whenever an activity event is generated. Accordingly, some information may be omitted from this data store, while other information is copied.

Consider the following domain description structured as activity events in the Recent Activity feed.

“ **John** created **Topic A**. ”

“ **Jane** commented on **Topic B**: *This is not a good example. Try and find a better one.* ”

“ **John** commented on **Jane**’s comment on **Topic B**: *I agree.* ”

“ **Jane** annotated **Topic A**. ”

“ **Bob** reacted to **John**’s comment on **Topic B**. ”

Using this description of the domain, we can start to derive logical and physical data models for the selected NoSQL data stores.

6.2 Physical data model

6.2.1 Document-oriented data model

Document data stores provide a number of different approaches to storing linked data. We attempt to cover the different methods, discussing common approaches and trade offs for each option. In general, data can be linked in two different ways.

1. Embedded collections (denormalized data). Embedding of data stores information in a single document. This technique is commonly used when the entity *contains* the embedded entity: for example storing contact details of a user. Another use for this approach is storing one-to-many relationships, where the child documents are always queried within the context of the parent document.
2. References (normalized data): Storing a reference to another document, similar to

storing keys to other tables in the relational model.

6.2.2 Column-oriented data model

6.2.3 Graph-oriented data model

Using the examples of activity events in section 6.1, we can identify the components used in database graph model. The first entity that is identified are the *nodes*

From the domain description, we can derive the labels, nodes and relationships to build the graph data model. The first entities that are identified are the nodes. In the domain, we have four main entities.

- User
- Topic
- Comment
- Annotation

In order for the Recent Activity feed to string all this together, there is also an entity to represent an activity event called Event. Events have relationships to all other entities according to the (subject, predicate, object)-triple explained in ??.

Neoj data modeling supports something called *Labels*, a graph construct that groups nodes into sets. A set contains nodes with the same label. A node can have any number of labels.

In the domain model, we can distinguish one case where labels are very useful. The *object* of the event triple can be one of multiple entities, in this thesis the scope is restricted to Topic, Comment and Annotation. Using this label abstraction allows future expansion with more object entity types. Consequently, we propose the label Object which aggregates nodes of these types. To keep the naming in the data model consistent, the label Subject will also be applied to nodes of the type User.

We can identify the following interactions between the entities:

- A User is the author of many Topics
- A Topic has many Comments attached to it
- A Topic has many Annotations attached to it
- Both Comments and Annotations are created by a User
- A Comment optionally has many children comments

Additionally, the Event entity introduces the following interactions:

- An Event relates to a User (the subject)
- An Event relates to an object

This leads us to the graph model in figure ??.

6.3 Reference queries

7. Comparative study

8. Opportunities

9. Conclusion

A. Research proposal

The subject of this thesis is based upon a research proposal that was graded by the promotor in advance. This proposal has been included as an attachment.

A.1 Introduction

The Open Weblides (2017b) project provides a user-friendly platform to collaborate on weblides - slides made with modern web technologies such as HTML, CSS and JavaScript. One of the core features this application provides is *co-creation*. The co-creation aspect manifests itself in several forms within the application; annotations on slides and a change suggesting system resembling GitHub's pull request feature are the main mechanisms. Because of the inherent social nature of co-creation, a basic notifications feed was also implemented. This feed is tailored to the user, and reflects the most recent changes, additions and comments relevant to the slide decks the user is interested in.

However, at the moment the functionality implemented in the system contains only the bare necessities. The module will be expanded in the future, and doing so requires a structural and conceptual rethinking of how the notifications are generated, stored and queried. The size of the dataset is also expected to grow linearly with user activity, therefore scalability is a requirement as well.

This paper has two research questions.

1. What frameworks and software packages currently exist in the industry to store structured non-relational graph or document data?
2. How is the social graph provided by the Open Weblides' notification feed structured

and how is this data consumed?

Answering these questions is paramount for the final section of the paper, which describes a data storage mechanism that performs well given the functional requirements of the project's data flow.

A.2 Use case

This thesis is a study of NoSQL data storage techniques applied to the Open Weblides (2017b) project. The online, interactive platform that the project provides includes a list of notifications in reverse chronological order, tailored to the user. This feature is called the *social feed*. It enumerates the most recent social activity on the platform. For example, if a user updates a slide deck, the user's friends would be able to find a change notification in their respective social feeds.

From the perspective of the application code that generates the social feed, the user, slide deck and notification should be considered separate entities. The notification itself has relations to the other entities: the author (the *subject*), the slide deck (the *object*), along with the *predicate* property that provides information on what operation was performed (for example creating or updating a slide deck).

This data model is also characterized by the *write-once read-many* nature of the information; once the notification has been generated, it does not need to be modified again. It will also only be queried in a very specific way: the application server will always attempt to retrieve the most recent notifications starting from the user entity. This principle is an important aspect to take into consideration in the choice of data storage mechanism.

A.3 State-of-the-art

In current literature, studies such as Moniruzzaman and Hossain (2013), Nayak, Poriya, and Poojary (2013b) and Dayne Hammes and Mitchell (2014) have already analyzed the disparity between traditional relational database systems and NoSQL stores. However, the conceptual and technical difference between these data storage models will not be scrutinized any further, since this paper presents a data storage solution applied to the social notification feed of the Open Weblides (2017b) project.

There are already many existing free and commercial products for the storage of NoSQL data, such as Redis (Sanfilippo, 2009), CouchDB (Apache Software Foundation, 2005b), MongoDB (Inc., 2009) and Neo4j (Technology, 2007). Finding the right database model for this use case (section A.2) is one of the hurdles this paper intends to handle. Zhao (2015) describes the development of a messaging system for astrophysical transient event notifications. Part of this paper is a qualitative comparison between document-based NoSQL storage solutions fit for this particular use case. We expect this paper to provide

a solid base of reasoning in order to find a scalable and efficient solution for resolving similar computational challenges.

The goal of this paper is to provide a performant, scalable and maintainable data storage schema for the Open Weblides (2017b) platform, regarding the linked social graph that powers the *Social Feed* functionality present in the platform.

A.4 Methodology

First, a range of industry-standard NoSQL database management systems such as MongoDB (Inc., 2009), HBase (Apache Software Foundation, 2005c) and Neo4j (Technology, 2007) will be qualitatively analyzed. Three of the five types of NoSQL database types (Nayak et al., 2013b) will be included in the study: column-oriented, document based and graph databases. Criteria for comparison include how the database management system concretely stores its data on disk, the query format and specific programming language bindings. Another important aspect is the distributed nature of many NoSQL databases. Using Brewer's conjecture (Gilbert & Lynch, 2002b, 2) – often called the CAP theorem – the existing types of data storage systems will be examined and summarized. There is also a practical factor present in the research; this includes the license of the project, its active maintainability and future prospects. Common types of NoSQL databases include key-value store, column-oriented, document store and graph databases (Nayak et al., 2013b). This paper will provide a short introduction to these types, before proceeding with the type that fits our use case the most.

Second, the data model specific to the Open Weblides project will be examined. We will start from the data model that is already implemented in the current iteration of the platform. At the time of writing, the existing base implementation of the social notification feed only contains two types of notifications. This paper will try to extrapolate this concept into a more generalized, abstract system in which developers can easily plug additional notification types. The physical properties of the data model will also be taken into account: the data will be written to the data storage only once, but read many times. It is also highly interlinked information, as a notification will always relate to one or more users as a subject, and a target object as well – most likely a slide deck or collection of slide decks. These links need to be maintained, and efficiently reconstructed when queried.

Finally, a sample dataset will be constructed using the aforementioned detailed analysis. Empirical testing will be conducted against multiple database management systems, and the results will be summarized and interpreted. Various information flows will be tested; however, the most important process remains efficiently querying the stored data.

Using the comparative study of storage engines, data model analysis and the empirical results an implementation plan will be constructed. This plan will serve as a recommendation for future development.

A.5 Expected results and conclusions

The NoSQL ecosystem, unlike relational databases, is headed towards specialization, so different solutions are headed in different directions (Maroo, 2013). In this paper, we expect to find one type of NoSQL database that is a better fit for the Open Weblides use case, in clear contrast with the other types of storage engines. Due to the inherently highly interlinked nature of the stored data, we suspect a graph-based database management system to provide most advantages, and generally the most performant experience.

This expectation is amplified by the availability and good community support of Ruby bindings to the most popular graph database management systems.

Since the platform being discussed only caters to a small to medium user base, we do not expect the need to scale horizontally beyond one instance. However, the vertical scalability is still a topic for discussion, and we expect to determine the computational order of magnitude in order to efficiently query the given dataset during this study.

Finally, the implementation plan should describe a concrete roadmap, stretching over a development period with a baseline expectation of one to three months. Roll-out of this mechanism should also be included in this plan.

We also expect that this thesis will provide a good reference to a further stable, scalable and extendable implementation of the *social feed* feature in the Open Weblides (2017b) project as outlined in section A.2.

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