# Introduction

One of the benefits of using NoSQL databases has usually been performance over traditional SQL databases. With an appropriate use case, NoSQL databases might offer significant performance benefits. In this study, the focus is on comparing traditional relational model and NoSQL graph model. A graph model is one of the four NoSQL types. As the graph model consists of nodes and edges, the model has its own benefits when handling relationship rich data. While in SQL database multiple tables have to be joined for a relational query, in graph databases relational information can be queried by navigating through the graph.

There are already several studies where the performance graph databases, especially Neo4J, have been compared with the traditional SQL databases. Often the results show better performance in favor of graph databases. This study is about handling simulated enterprise data. In enterprise use, performance is often crucial. Thus, it is very important take into consideration when choosing database. Due to the importance of performance considerations, the main question is to find out how an SQL database and a graph database compare when it comes to the performance.

This study will show how the databases perform when query complexity grows. Recursive query performance and the effect of indexing on the performance is also studied. The databases compared in this study are MySQL 5.1.41, MariaDB 10.5.6 and Neo4J 4.1.3. Instead of using existing benchmarks such as [1] or [2], a dedicated test bench was implemented for this study. The test bench is called Invoicing Database Test Bench and its source code is available from GitHub [3]. The program generates given amount of data into the test invoicing database schema and performs various query tests. This approach was taken in order to simulate a program that queries the data.

The rest of the paper is organized as follows. Section 2 is the related study where previous performance related study related to Neo4J and MariaDB is reviewed. Section 3 presents the schema that is used for the test data. Section 4 presents the implemented benchmarking program. Section 5 presents the test queries. Section 6 presents the query tests. Section 7 is an analysis of the results and section 8 is the conclusions.

# Related study

SQL database and Neo4J have been compared in several studies including [4], [5], [6], [7] and [8]. The study [4] compares tuned Oracle 11g and Neo4J 3.03 Community Edition. Healthcare data is used including data about patients, medicines and medical staff. Performance of the databases is evaluated with ten different count(\*) queries. Many of the queries also perform some table joins. Physical database tuning technique called tablespaces is used for Oracle. The same databases were compared without physical database tuning in [6]. The physical database tuning technique decreased the overall average query time of Oracle from 4.34 to 2.78 seconds. However, the overall average query time for Neo4J in query tests was only 0.67 seconds. Thus, Neo4J performed better compared to Oracle.

In [5] Neo4J version 1.8 performance is tested with different backend solutions. Neo4J is benchmarked as embedded with native object access, as a dedicated server through RESTful Web Services, with embedded Cypher queries, with Cypher through REST optimized for remote execution and with Gremlin queries through REST. MySQL version 5.5.27 is also included with Java Persistence API based backend. Queries are done using Cypher, Gremlin and SQL query languages. The test data consist of people related data. Relational test queries are executed such as friends of friends. As the database gets larger, the advantages of Neo4J over MySQL become more prevalent. Neo4J performance stays nearly constant when MySQL performance drops by factors 5 and 7-9. Both Neo4J query languages Gremlin and Cypher have performance benefits over MySQL with JPA.

Article [7] is from the year 2010 and compares MySQL Community Server version 5.1.42 and Neo4J version 1.0-b11. In the article the graph database is stored into a relational database as nodes and edges. Three types of structural and three types of data queries are made. First structural query finds all the orphan nodes and the two other ones traverse the graph in the depths of 4 and 128. The data queries are count(\*) queries counting nodes with certain payload. Neo4J performed better in structural queries. However, with the integer-based queries MySQL was more efficient due to the fact that the tested Neo4J used Lucene indexing. As it treated the data as text by default, conversions had to be made and thus they impacted the performance. The article [7] is referenced in articles [4], [5] and [6].

Article [8] is from the year 2012 and compares MySQL version 5.1.41 and Neo4J Community version 1.6. The article relates closely to this study. A schema with tables user, friends, fav\_movies and actors is used for testing. The databases are tested with three queries: “Find all friends of Esha”, “Find all favourite movies of Esha’s friends” and “Find the lead actors of Esha’s friends favourite movies”. Queries are done with 100 and 500 objects. Neo4J has 2-5 times faster query times with 100 objects data set and 15-30 times faster in 500 objects data set. The article [8] is referenced in article [5].

There also exist previous performance studies where MariaDB is involved. In [9] the performance of MariaDB 10.0.21 and MySQL 5.6 is compared using Sysbench and OLTP [2] software. OLTP-Simple and OLTP-Seats workloads are used. Both databases consumed the same number of resources. However, when an increasing number of threads in OLTP-Simple and a number of workers in OLTP-Seats MySQL is clearly more effective outperforming MariaDB. The Common Table Expression capabilities of MariaDB are studied in [10] along with PostgresSQL. This study showed that Postgres has better results when only a few steps of recursion is needed. However, MariaDB is a better choice for a long recursive process on a huge amount of data.

# Invoicing database

The test database is a general example of an invoicing database. One of the most important use cases is the calculation of the price for a customer invoice. This is done by calculating the used time for work of different work types and the price of the items used when working. Invoices might also have relations to other invoices if several invoices are sent to the customer.

The database has 10 tables. The basic tables are customer, invoice, target, work, worktype and item. These tables contain the customer information, customer’s invoices, the target where the work is done, a listing of each work, a listing of different worktypes with different prices and information about the items used for each work. Relational data between the tables is stored into M:N tables worktarget, workinvoice, useditem and workhours.

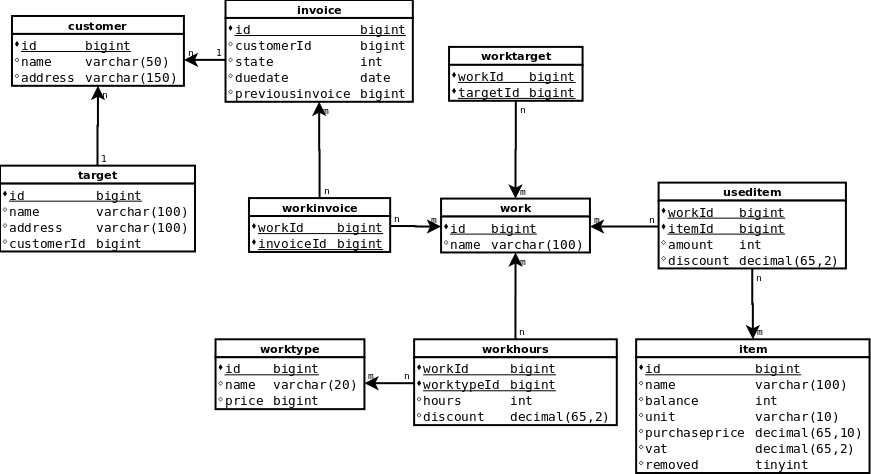
.**

Figure 1: Invoicing database in relational format.

On a graph format edges are used to represent the relationships. For N:M relationships, bidirectional edges are used. The tables customer, invoice, target, work, worktype are represented as nodes. The edges between the nodes are PAYS between the customer and invoice, CUSTOMER\_TARGET between the customer and target, WORK\_TARGET between work and target, WORK\_INVOICE between work and invoice, WORKHOURS between the work and worktype and USED\_ITEM between work and item.



Figure 2: Invoicing database in graph format.

# Test program

The test data are generated using a Java program. The program uses sample data that are based on openly available name and address data sets [13], [12]. The sample data are used when generating customer and target names. The generation process is divided into three parts. Items and work types should be generated first, then work and customer data. The Java program has threaded classes for each part. Multiple threads can be used to insert data as well as factors for related data. In data generation, seed numbers are used every time when there is something defined by random. This way the test data is repeatable.

When generating the work data, amount of the related worktypes and items can be defined by setting the worktype factor and item factor. When generating customer data, the number of the related invoices, targets and work can be defined by setting the invoice factor, target factor and work factor. When generating work and customers, the given amount of work and customers are generated. For each work, the given factor of relations for worktypes and invoices are generated. For each customer, the given number of invoices and targets are generated. The generator will also generate workinvoice and worktarget relationships based on the given factor.

The program has a class called QueryTester that used to perform the query tests. Query tests are repeated the selected number of times. The test program collects the results into a list structure. The program removes the biggest and the smallest number from the list and calculates an average and a standard deviation from the rest of the results.

# Test queries

The query tests contain different relational queries for calculating the price of work and invoice and a recursive query. These queries test different capabilities of the databases. As both relational and graph databases are tested, the queries are in SQL and Cypher form. The query tests are ordered from simple to complex starting from the work price and the work price with items ending in the work prices and work prices for a given customer. There is also a recursive query test.

Calculating the invoice prices is one of the most important queries. The schema does not store invoice prices explicitly. The price has to be calculated based on the amount of the workhours and the items used. The “price of work” and the “price of work with items” are the subqueries for calculating this price. The query calculating invoice prices for a given customer add customer information into this query. The recursive query queries all the interrelated invoices. One practical example is that customer has not paid the invoice and there will be additional invoices based on the same invoice.

## Query optimization

In addition to just testing databases with the queries, it was interesting to find out how to make the query performance better. Indexing is one of the ways making queries faster. It has to be taken into consideration when comparing MySQL, MariaDB and Neo4J that both MySQL and MariaDB index primary key and foreign key by default. Neo4J does not create indexes for properties by default. The effect of indexing is also different when comparing an SQL database with a graph database. When querying relations, in SQL databases the relations are formed by joining the tables based on primary key and foreign key information. Thus, SQL databases usually benefit from the indexing of primary key and foreign key. In a graph database, we are traversing the graph when querying data. As such, it does not benefit indexing the properties the way SQL databases do.

In order to study the effects of indexing, certain columns and properties that were used in queries were indexed in all the databases. Table 1 shows the extra indexes created. As ids in customer and invoice tables are indexed by default in MySQL and MariaDB, an extra index was not needed.

|  |  |  |
| --- | --- | --- |
| Table/Node | SQL | Neo4J |
| Customer | - | customerId |
| Invoice | previousinvoice | invoiceId, previousinvoice |
| Item | purchaseprice | purchaseprice |
| Workhours | hours, discount | hours, discount |
| Worktype | price | price |
| Useditem | amount, discount | amount, discount |

Table 1: Indexed columns/properties in SQL and Neo4J.

Besides indexing, in Neo4J 4.1.3 the queries can be optimized using CALL subqueries [13]. The CALL clause makes it possible to execute subqueries in other queries. It is like a function that gets input parameters from the main query and returns some values. The subquery is executed for each incoming input row from calling the query from the main query. CALL has been supported from Neo4J onwards. In this study Cypher queries with and without CALL are used in order for backward compatibility. This way it is also possible to see how much CALL subquery improves the query performance.

## Short query, price of work

The first query to be tested calculates the price of work. One work can have different work types with different prices. The price of one work is defined by the number of hours done of the work type. There can also be a discount on the prices and the discount is included in calculation. This query shows how databases perform with a fairly simple query. Table 2 shows the queries to query the price of work in SQL and Cypher.

|  |  |
| --- | --- |
| Type | Query |
| SQL | SELECT work.id AS workId, SUM( (worktype.price \* workhours.hours \* workhours.discount) ) AS price  FROM work  INNER JOIN workhours ON work.id = workhours.workId  INNER JOIN worktype ON worktype.id = workhours.worktypeId  GROUP BY work.id |
|  |
|  |
|  |
|  |
|  |
| Cypher | MATCH (wt:worktype)-[h:WORKHOURS]->(w:work)  WITH SUM(h.hours\*h.discount\*wt.price) as price, w  RETURN w.workId as workId, price |  |
|  |
|  |
| Cypher with CALL | MATCH (w:work) CALL {  WITH w  MATCH (wt:worktype)-[h:WORKHOURS]->(w)  RETURN SUM((h.hours\*h.discount\*wt.price)) as price  } RETURN w.workId as workId, price |  |
|  |
|  |
|  |
|  |
|  |

Table 2: Price of work query in SQL and Cypher

## Long query, price of work with items

The query for the price of work with items is an extended version of the query for the price of work. This query adds item prices into work prices. As items are also included, the longer relational query is needed. With this query it is possible to see how databases perform when more relations and calculations are included in the query. Item purchase price is a floating-point number so this will add more challenges to the calculations. Table 3 shows the queries for the price of work with items in SQL and Cypher.

|  |  |
| --- | --- |
| Type | Query |
| SQL | SELECT work.id AS workId, SUM((worktype.price \* workhours.hours \* workhours.discount) + (item.purchaseprice \* useditem.amount \* useditem.discount) ) AS price  FROM work  INNER JOIN workhours ON work.id = workhours.workId  INNER JOIN worktype ON worktype.id = workhours.worktypeId  INNER JOIN useditem ON work.id = useditem.workId  INNER JOIN item ON useditem.itemId = item.id  GROUP BY work.id |
|  |
|  |
|  |
|  |
|  |
|  |
| Cypher | MATCH (wt:worktype)-[h:WORKHOURS]->(w:work)-[u:USED\_ITEM]->(i:item)  WITH SUM((h.hours\*h.discount\*wt.price)+(u.amount\*u.discount\*i.purchaseprice)) as price, w  RETURN w.workId as workId, price |  |
|  |
|  |
| Cypher with CALL | MATCH (w:work)  CALL {  WITH w  MATCH (wt:worktype)-[h:WORKHOURS]->(w)-[u:USED\_ITEM]->(i:item) RETURN SUM((h.hours\*h.discount\*wt.price)+(u.amount\*u.discount\*i.purchaseprice)) as price }  RETURN w.workId as workId, price |  |
|  |
|  |
|  |
|  |
|  |

Table 3: The price of work with items query in SQL and Cypher.

## Complex query, invoice price

This query calculates the sum of a work price for each invoice. The query contains two subqueries. The first one finds the relation of the invoices and work. The second query is the previously presented “price of work with items”. The results of these queries are joined and the sums of prices are aggregated based on the id of the invoice. This is one of the heaviest queries and as such it is useful to see the performance differences when executing a complex query. Table 4 present the queries for calculating the invoice price in SQL and Cypher.

|  |  |
| --- | --- |
| Type | Query |
| SQL | SELECT q1.invoiceId, SUM(q2.price) AS invoicePrice  FROM (  SELECT workinvoice.invoiceId, workinvoice.workId  FROM workinvoice  INNER JOIN invoice ON workinvoice.invoiceId = invoice.id  ) AS q1 INNER JOIN (  SELECT workhours.workid AS workId, SUM( (worktype.price \* workhours.hours \* workhours.discount) + (item.purchaseprice \* useditem.amount \* useditem.discount) ) AS price  FROM workhours  INNER JOIN worktype ON workhours.worktypeid = worktype.id  INNER JOIN useditem ON workhours.workid = useditem.workid  INNER JOIN item ON useditem.itemid = item.id  GROUP BY workhours.workid ) AS q2 USING (workId)  GROUP BY q1.invoiceId |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
| Cypher | MATCH (inv:invoice)-[:WORK\_INVOICE]->(w:work)  WITH inv, w  OPTIONAL MATCH (wt:worktype)-[h:WORKHOURS]->(w:work)-[u:USED\_ITEM]->(i:item)  WITH inv, w, SUM((h.hours\*h.discount\*wt.price)+(u.amount\*u.discount\*i.purchaseprice)) as workPrice RETURN inv, SUM(workPrice) as invoicePrice |  |
|  |
|  |
|  |
| Cypher with CALL | CALL {  WITH inv  MATCH (inv)-[:WORK\_INVOICE]->(w:work)  RETURN w }  CALL {  WITH w  MATCH (wt:worktype)-[h:WORKHOURS]->(w)-[u:USED\_ITEM]->(i:item)  RETURN SUM((h.hours\*h.discount\*wt.price)+(u.amount\*u.discount\*i.purchaseprice)) as workPrice  }  RETURN inv, SUM(workPrice) as invoicePrice |  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |

Table 4:The query for invoice prices in SQL and Cypher.

## Query with defined key, invoice prices for customer with id 0

It is often needed to find out all the invoice prices for a given customer. The query that calculates invoice prices for a given customer is an extended query from the query that calculates invoice prices. A subquery to get customer’s relation to invoices is included. This query is the most complex of the tested queries. From the technical point of view this query shows how databases perform when there is a certain key defined for which the data should be related to. Table 5 presents the queries for calculating invoice prices for a given customer.

|  |  |
| --- | --- |
| Type | Query |
| SQL | SELECT q1.customerId, q2.invoiceId, SUM(q3.price) AS invoicePrice  FROM (  SELECT customer.id AS customerId, invoice.id AS invoiceId  FROM invoice  INNER JOIN customer ON invoice.customerId=customer.id  ) AS q1  INNER JOIN ( SELECT workinvoice.invoiceId, workinvoice.workId  FROM workinvoice  INNER JOIN invoice ON workinvoice.invoiceId = invoice.id ) AS q2 USING (invoiceId) INNER JOIN (  SELECT workhours.workid AS workId, SUM( (worktype.price \* workhours.hours \* workhours.discount) + (item.purchaseprice \* useditem.amount \* useditem.discount) ) AS price  FROM workhours  INNER JOIN worktype ON workhours.worktypeid = worktype.id  INNER JOIN useditem ON workhours.workid = useditem.workid  INNER JOIN item ON useditem.itemid = item.id GROUP BY workhours.workid ) AS q3 USING (workId) WHERE q1.customerId=0 GROUP BY q2.invoiceId |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
| Cypher | MATCH (c:customer)-[:PAYS]->(inv:invoice) WHERE c.customerId=0  WITH c, inv  OPTIONAL MATCH (inv)-[:WORK\_INVOICE]->(w:work)  WITH c, inv, w  OPTIONAL MATCH (wt:worktype)-[h:WORKHOURS]->(w:work)-[u:USED\_ITEM]->(i:item)  WITH c, inv, w, SUM((h.hours\*h.discount\*wt.price)+(u.amount\*u.discount\*i.purchaseprice)) as workPrice RETURN c, inv, SUM(workPrice) as invoicePrice |  |
|  |
|  |
|  |
|  |
|  |
| Cypher with CALL | MATCH (inv:invoice) WHERE inv.customerId=0  CALL {  WITH inv  MATCH (c:customer)-[:PAYS]->(inv)  RETURN c  } CALL {  WITH c, inv  MATCH (inv)-[:WORK\_INVOICE]->(w:work)  RETURN w  }  CALL {  WITH w  MATCH (wt:worktype)-[h:WORKHOURS]->(w)-[u:USED\_ITEM]->(i:item)  RETURN SUM((h.hours\*h.discount\*wt.price)+(u.amount\*u.discount\*i.purchaseprice)) as workPrice  }  RETURN c, inv, SUM(workPrice) as invoicePrice |  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |

Table 5: The query for the Invoice prices of a defined customer in SQL and Cypher.

## Recursive query, invoices related to invoice id 100000

The recursive query gets all the sequential invoices related to given invoice id. The query is useful to test the recursive query capabilities of the databases. In SQL Common Table Expressions is used to make the query. In Cypher there is a way to optimize the recursive query by negating irrelevant relationships. The optimized query does not return exactly the same result as the basic query. While the basic query returns a set of individual nodes, the optimized query returns a list structure containing nodes. However, it still returns similar results and as such it is a relevant query. Table 6 presents the queries for finding sequential invoices for a given invoice.

|  |  |
| --- | --- |
| Type | Query |
| SQL | WITH RECURSIVE sequential\_invoices AS ( SELECT id, customerId, state, duedate, previousinvoice  FROM invoice  WHERE id=10000  UNION ALL  SELECT i.id, i.customerId, i.state, i.duedate, i.previousinvoice  FROM invoice AS i  INNER JOIN sequential\_invoices AS j ON i.previousinvoice = j.id  WHERE i.previousinvoice <> i.id)  SELECT \* FROM sequential\_invoices |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
|  |
| Cypher | MATCH (i:invoice { invoiceId:10000 })-[p:PREVIOUS\_INVOICE \*0..]->(j:invoice) RETURN \* |  |
| Cypher optimized | MATCH inv=(i:invoice { invoiceId:10000})-[p:PREVIOUS\_INVOICE \*0..]->(j:invoice) WHERE NOT (j)-[:PREVIOUS\_INVOICE]->() RETURN nodes(inv) |  |

Table 6: The Recursive query to get the sequential invoices related to defined invoice in SQL and Cypher.

# Query tests

## Test settings

The query tests were performed with MacBook Pro Laptop with following specifications:

* macOS Catalina version 10.15.5
* 1,4 GHz quad core Intel Core i5
* 8 GB 2133 MHz LPDDR3
* Intel Iris Plus Graphics 645, 1536 MB

MySQL version 5.1.41, MariaDB version 10.5.6 and Neo4J community edition version 4.1.3 were installed on this computer. MariaDB and Neo4J were the latest when making this study. MySQL version 5.1.41 is already considered as "end-of-life" when making this study. However, it that was used in a previous study [8] and version 5.1.42 in [7] as shown in related studies. MariaDB driver version 2.7 and Neo4J driver version 4.1.1 were used.

There was interest to see how the old MySQL compares with the new versions of MariaDB and Neo4J and whether it was possible to repeat the previous results. MariaDB was chosen because MySQL is nowadays often replaced with MariaDB. There are various reasons for this including more open development compared with modern MySQL. There are also not yet many studies about MariaDB yet. Especially such studies where it would be compared with a graph database.

When making this study, DB-Engines site ranks MariaDB as 8th out of 138 of relational databases [14]. Neo4J ranks the first out of 32 databases on the same sites. As both of the databases are quite popular, they are often considered to be used in many enterprises. A dataset was generated using the test program. Table 7 presents the number of rows/objects generated for the dataset. A double the amount of useditem, workhours, workinvoice and worktarget relationships were generated into Neo4J as in graph form. A many-to-many relationship is expressed as a bidirectional relationship.



Table 7: The numbers of the generated rows/objects in SQL and Neo4J.

## Test results

Each query test was executed with 12 iterations. Each query result contains an average time for the query in the milliseconds and coefficient of variation (CV) of the result list. As Neo4J has outperformed SQL databases in many previous studies, Neo4J was chosen as a reference database where others are compared with. Overall, the inclusion of CALL into a query made queries faster so Cypher queries with CALL were chosen as the reference queries. If Neo4J version above 4.1 would be used, CALL would be preferred for more performance. Non-indexed tests show percentage slower related to non-indexed Neo4J and indexed tests show percentage slower related to indexed Neo4J.

* + 1. Short query, price of work

Results for the query that queries the price of work can be found in table 8. From the generated dataset, the query returned 10000 rows/objects. With this query, Neo4J outperforms SQL databases as in several previous studies. MySQL is the slowest and MariaDB the second. Inclusion of CALL does not seem to bring benefits to Neo4J with this query. Indexing does either not seem to bring much benefits either. CV values do not have a remarkable difference in these results.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Database | Avg | CV | Slower than Neo4J 4.1.3 CALL | Avg indexed | CV indexed | Slower than Neo4J 4.1.3 CALL indexed |
| MySQL 5.1.41 | 405 | 2,56 % | 64 % | 413 | 0,71 % | 65 % |
|  |
| MariaDB 10.5.6 | 209 | 8,90 % | 31 % | 204 | 7,13 % | 28 % |  |
|  |
| Neo4J 4.1.3 | 144 | 1,15 % | -1 % | 148 | 4,97 % | 1 % |  |
|  |
| Neo4J 4.1.3 CALL | 145 | 1,07 % | - | 146 | 1,63 % | - |  |
|  |

Table 8: Results for the query for the price of work.

* + 1. Long query, price of work with items

Results for the query that queries the price of work can be found in table 8. From the generated dataset, the query returned 10000 rows/objects. With this query, Neo4J outperforms SQL databases as in several previous studies. MySQL is the slowest and MariaDB the second. Inclusion of CALL does not seem to bring benefits to Neo4J with this query. Indexing does not seem to bring benefits either. The CV value of MySQL and MariaDB results is very low Neo4J CV values being significantly higher.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Database | Avg | CV | Slower than Neo4J 4.1.3 CALL | Avg indexed | CV indexed | Slower than Neo4J 4.1.3 CALL indexed |
| MySQL 5.1.41 | 4952 | 0,63 % | 62 % | 4901 | 0,34 % | 40 % |
|  |
| MariaDB 10.5.6 | 2197 | 0,43 % | 15 % | 2193 | 0,23 % | -34 % |  |
|  |
| Neo4J 4.1.3 | 6880 | 18,82 % | 73 % | 8250 | 21,79 % | 64 % |  |
|  |
| Neo4J 4.1.3 CALL | 1859 | 8,67 % | - | 2931 | 13,53 % | - |  |
|  |

Table 9: Results for price of work with items query.

* + 1. Complex query, invoice price

Table 10 shows results for the query that calculates invoice prices. From the generated dataset, the query returns 100000 rows/objects. Without CALL included in query Neo4J performs again even worse than old MySQL 5.1.41. Inclusion of CALL gives significant performance benefit. However, MariaDB still radically faster. Indexing gives also some performance benefits in MariaDB. However, with other databases it does not seem to improve performance. CV value for MariaDB and indexed Neo4J with CALL. The CV values of MariaDB and indexed Neo4J with CALL are significantly higher than others, which are quite low.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Database | Avg | CV | Slower than Neo4J 4.1.3 CALL | Avg indexed | CV indexed | Slower than Neo4J 4.1.3 CALL indexed |
| MySQL 5.1.41 | 235817 | 1,03 % | 3 % | 236875 | 0,76 % | -28 % |
|  |
| MariaDB 10.5.6 | 4282 | 24,45 % | -5254 % | 3840 | 23,06 % | -7800 % |  |
|  |
| Neo4J 4.1.3 | 753587 | 0,73 % | 70 % | 957325 | 1,35 % | 68 % |  |
|  |
| Neo4J 4.1.3 CALL | 229256 | 1,66 % | - | 303365 | 19,61 % | - |  |
|  |

Table 10: Results for the query for the invoice prices.

* + 1. Query with a defined key, invoice prices for a given customer

Results for the query that gets invoice prices for a given customer can be found in table 11. From the generated dataset, the query returns 10 rows/objects. MySQL 5.1.41 was left out as the performance was so poor. The query took over one hour on average. In practical use, it would be unusable. With Neo4J, the inclusion of CALL does not give performance benefits. However, indexing seems to bring improvements with basic Cypher query. With indexing Neo4J finds the customer 0 from the graph faster. Although Neo4J performs well, MariaDB outperforms it by a small margin. The CV values of these results are overall high compared to previous results ones non-indexed MariaDB and Neo4J being quite low.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Database | Avg | CV | Slower than Neo4J 4.1.3 CALL | Avg indexed | CV indexed | Slower than Neo4J 4.1.3 CALL indexed |
| MariaDB 10.5.6 | 42 | 1,17 % | -38 % | 30 | 18,02 % | -97 % |
|  |
| Neo4J 4.1.3 | 56 | 25,17 % | -4 % | 50 | 17,91 % | -18 % |  |
|  |
| Neo4J 4.1.3 CALL | 58 | 1,63 % | - | 59 | 12,09 % | - |  |
|  |

Table 11: Results for the query for the invoice prices for a defined customer.

6.2.5 Recursive query, invoices related to invoice id 100000

The recursive query lists all the sequential invoices related to the invoice with given id. The tests were performed with 100 and 1000 invoices. With 10000 invoices Neo4J performance without optimization was so poor that it would have taken too long to complete. Table 12 presents the results when querying 100 sequential invoices and table 13 presents results when querying 1000 invoices. The CV value of MariaDB is significantly higher because of its radically different result. Average of 1 divided by 0.4 standard deviation yields this result. Without the index, MariaDB has quite low CV others having notably higher value.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Database | Avg | CV | Slower than Neo4J 4.1.3 Optimized | Avg indexed | CV indexed | Slower than Neo4J 4.1.3 Optimized indexed |
| MariaDB 10.5.6 | 4473 | 0,33 % | 81 % | 1 | 40,00 % | -86800 % |
|  |
| Neo4J 4.1.3 | 150 | 12,00 % | -468 % | 153 | 4,33 % | -468 % |  |
|  |
| Neo4J 4.1.3 Optimized | 852 | 9,63 % | - | 869 | 12,44 % | - |  |
|  |

Table 12: Results for the recursive query with 100 rows/objects.

With 100 invoices, Neo4J seems to have the best performance without query optimization. The optimized query does not seem to improve performance. However, when indexes are used MariaDB benefits dramatically from indexing. The query takes just 1ms average clearly making MariaDB the best performer. Indexing does not improve performance for Neo4J. There is variation in CV values non-indexed MariaDB having the lowest value. Most variation was in the results of optimized Neo4J query results.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Database | Avg | CV | Slower than Neo4J 4.1.3 Optimized | Avg indexed | CV indexed | Slower than Neo4J 4.1.3 Optimized indexed |
| MariaDB 10.5.6 | 45004 | 0,46 % | 92 % | 10 | 6,40 % | -51070 % |
|  |
| Neo4J 4.1.3 | 1239446 | 12,09 % | 100 % | 1637223 | 16,71 % | 100 % |  |
|  |
| Neo4J 4.1.3 Optimized | 3662 | 27,24 % | - | 5117 | 5,71 % | - |  |
|  |

Table 13: Results for the recursive query with 1000 rows/objects.

As the number of invoices was increased to 1000, Neo4J performance drops dramatically. With the optimized query though Neo4J becomes much faster outperforming MariaDB. However, as MariaDB benefits from indexing, with indexing the query takes only ten milliseconds on average making it yet again the best performer. The performance is radically faster in comparison with Neo4J. MariaDB is 51070 % faster. Indexing does not improve performance for Neo4J.

# Analysis

With the query tests performed Neo4J is often outperformed by MariaDB in many cases. In some tests, Neo4J performs even worse than old MySQL 5.1.41. When comparing Neo4J with MySQL and MariaDB we are comparing a Java program with C/C++ program. Obviously, the latter can be optimized better. It has to be also taken into consideration that MariaDB indexes primary keys and foreign keys by default. This gives benefits in every query where the table joins are done. Neo4J does not seem to benefit from indexing in many cases. One such case where indexing had benefits was when Neo4J needs to find the starting point from the graph.

The benefit of indexing in MariaDB can also be seen as the benefit of the traditional relational database model. As the relations with the tables are created when executing the SQL query, indexing the keys becomes beneficial. The graph model does not benefit from such indexing as there are no tables that are joined by keys. Querying a graph database is done by traversing the graph. One of the benefits of the graph model can be seen in recursive query tests. By optimizing the query, performance becomes clearly better and, in this case, even better than SQL database with CTE query. However, with recursive queries, indexing still brings dramatical benefits for SQL database.

With the invoicing database schema used in this study, calculating the price is done with complex queries. If this database was used in some real case, the usage of table views would probably be preferred to simplify the queries. When it comes to using views, it is also a benefit of SQL databases over Neo4J as the time writing this article Neo4J does not have an exact equivalent of such feature as views in SQL databases.

# Conclusions

This study compared MySQL 5.1.41, MariaDB 10.5.6 and Neo4J 4.1.3 with various query tests related to the invoicing database. A simulated enterprise dataset was used as test data. The query tests were performed using a Java program developed for this study. The query tests included relational queries with increasing complexity and recursive queries. With the simpler relational queries, Neo4J had the best performance which correlates with previous results shown in [7] and [10]. However, with more complex queries MariaDB outperformed Neo4J.

The fact that SQL database outperformed a graph database is a new finding that has not been presented in previous studies. As presented in previous studies, Neo4J often outperformed SQL databases. In study [4] for example, Neo4J outperformed Oracle with various tests using count(\*) queries. In this study aggregation queries were also used but the result was different. This study also indicated the benefit of indexing in SQL database in many of the tests. SQL databases seemed to benefit from indexing and in some cases very dramatically. However, the tested database Neo4J often seemed not to benefit from indexing. Neo4J benefited from indexing when a starting point in the graph was indexed.

Overall MariaDB is the clear winner in this study when it comes to performance with more complex queries. Especially the query test for querying the invoice prices, MariaDB shows its performance. Indexing also gives clear benefits in MariaDB. Especially with recursive queries the improvement was dramatical. The results in this study show how a relational database is still a strong alternative when it comes to performance compared with a NoSQL graph database.

# References

[1] Cooper, Brian F., et al. "Benchmarking cloud serving systems with YCSB." Proceedings of the 1st ACM symposium on Cloud computing. 2010.

[2] Difallah, Djellel Eddine, et al. "Oltp-bench: An extensible testbed for benchmarking relational databases." *Proceedings of the VLDB Endowment* 7.4 (2013): 277-288.

[3] GitHub. InvoicingDBTestBench repository. https://github.com/homebeach/InvoicingDBTestBench, Accessed 13.12.2020

[4] Khan, Wisal, et al. "SQL Database with physical database tuning technique and NoSQL graph database comparisons." *2019 IEEE 3rd Information Technology, Networking, Electronic and Automation Control Conference (ITNEC)*. IEEE, 2019.

[5] Holzschuher, Florian, and René Peinl. "Performance of graph query languages: comparison of cypher, gremlin and native access in Neo4j." *Proceedings of the Joint EDBT/ICDT 2013 Workshops*. 2013.

[6] Khan, Wisal, and Waseem Shahzad. "Predictive Performance Comparison Analysis of Relational & NoSQL Graph Databases." *Int. J. Adv. Comput. Sci. Appl* 8 (2017): 523-530.

[7] Vicknair, Chad, et al. "A comparison of a graph database and a relational database: a data provenance perspective." *Proceedings of the 48th annual Southeast regional conference*. 2010.

[8] Batra, Shalini, and Charu Tyagi. "Comparative analysis of relational and graph databases." *International Journal of Soft Computing and Engineering (IJSCE)* 2.2 (2012): 509-512.

[9] Tongkaw, Sasalak, and Aumnat Tongkaw. "A comparison of database performance of MariaDB and MySQL with OLTP workload." 2016 IEEE conference on open systems (ICOS). IEEE, 2016.

[10] Shalygina, Galina, and Boris Novikov. "Implementing common table expressions for MariaDB." Second Conference on Software Engineering and Information Management (SEIM-2017) (full papers). 2017.

[11] Data.world. Names datasets. https://data.world/datasets/names, 2020

[12] OpenAddresses. A summary view of OpenAddresses data. http://results.openaddresses.io, 2020

[13] Neo4J, Inc. The Neo4J Cypher Manual v4.2. CALL {subquery}. https://neo4j.com/docs/cypher-manual/current/clauses/call-subquery, 2020

[14] DB-Engines. https://db-engines.com/, Accessed 13.12.2020