

LDBC Social Network Benchmark (SNB) – 0.3.0

Coordinator: LDBC Social Network Benchmark Task Force

ABSTRACT

LDBC's Social Network Benchmark (LDBC SNB) is an effort intended to test various functionalities of systems used for graph-like data management. For this, LDBC SNB uses the recognizable scenario of operating a social network, characterized by its graph-shaped data.

LDBC SNB consists of two workloads that focus on different functionalities: the interactive workload (interactive transactional queries) and the business intelligence workload (analytical queries).

This document contains the definition of the Interactive Workload and the first draft of the Business Intelligence Workload. This includes a detailed explanation of the data used in the LDBC SNB benchmark, a detailed description for all queries, and instructions on how to generate the data and run the benchmark with the provided software.

EXECUTIVE SUMMARY

The new data economy era, based on complexly structured, distributed and large datasets, has brought on new demands on data management and analytics. As a consequence, new industry actors have appeared, offering technologies specially built for the management of graph-like data. Also, traditional database technologies, such as relational databases, are being adapted to the new demands to remain competitive.

LDBC's Social Network Benchmark (LDBC SNB) is an industrial and academic initiative, formed by principal actors in the field of graph-like data management. Its goal is to define a framework where different graph based technologies can be fairly tested and compared, that can drive the identification of systems' bottlenecks and required functionalities, and can help researchers to open new research frontiers.

The philosophy around which LDBC SNB is designed is to be easy to understand, flexible and cheap to adopt. For all these reasons, LDBC SNB will propose different workloads representing all the usage scenarios of graph-like database technologies, hence, targeting systems of different nature and characteristics. In order increase its adoption by industry and research institutions, LDBC SNB provides all necessary software, which are designed to be easy to use and deploy at a small cost.

This document contains:

- A detailed specification of the data used in the whole LDBC SNB benchmark.
- A detailed specification of the workloads.
- A detailed specification of the execution rules of the benchmark.
- A detailed specification of the auditing rules and the full disclosure report's required contents.

Table of Contents

| Ex | ECUT | ive Summary | 3 |
|----|------|---|----|
| Do | CUMI | ENT INFORMATION | 4 |
| 1 | Intr | ODUCTION | 8 |
| | 1.1 | Motivation for the Benchmark | 8 |
| | 1.2 | Relevance to Industry | 8 |
| | 1.3 | General Benchmark Overview | 9 |
| | 1.4 | Related Projects | 10 |
| | 1.5 | Participation of Industry and Academia | 10 |
| 2 | Ben | CHMARK SPECIFICATION | 12 |
| | 2.1 | Requirements | 12 |
| | 2.2 | Software and Useful Links | 12 |
| | 2.3 | Data | 12 |
| | | 2.3.1 Data Types | 12 |
| | | 2.3.2 Data Schema | 13 |
| | | 2.3.3 Output Data | 17 |
| | | 2.3.4 Scale Factors | 19 |
| | | 2.5.4 Scale Lactors | 1) |
| 3 | Wor | RKLOADS | 20 |
| | 3.1 | Query Description Format | 20 |
| | 3.2 | Query Definitions | 20 |
| | | 3.2.1 Notation | 20 |
| | | 3.2.2 Morphisms | 20 |
| | | 3.2.3 Graph Patterns | 21 |
| | 3.3 | Substitution Parameters | 21 |
| | 3.4 | Load Definition | 22 |
| 4 | Inte | RACTIVE WORKLOAD | 24 |
| | 4.1 | Complex Reads Query Descriptions | 24 |
| | 4.2 | Short Reads Query Descriptions | 39 |
| | 4.3 | Update Query Descriptions | 43 |
| | | | |
| 5 | Busi | iness Intelligence Workload | 48 |
| | 5.1 | Read Query Descriptions | 48 |
| | 5.2 | Update Query Descriptions | 72 |
| 6 | Aud | ITING RULES | 73 |
| | 6.1 | Preparation | 73 |
| | | 6.1.1 Collect System Details | 73 |
| | | 6.1.2 Setup the Benchmark Environment | 73 |
| | | 6.1.3 Load Data | 74 |
| | 6.2 | Running the Benchmark | 74 |
| | 6.3 | Recovery | 74 |
| | 6.4 | Serializability | 75 |
| | U.T | Dominization of the contraction | 1 |

| 7 | Related Work | 76 |
|---|---------------------------------|--------|
| | 7.1 Graph Database Benchmarks | 76 |
| | 7.2 Scalable Graph Generators | |
| | 7.3 LDBC Publications | |
| A | Choke Points | 78 |
| | A.1 Aggregation Performance | 78 |
| | A.2 Join Performance | 80 |
| | A.3 Data Access Locality | 82 |
| | A.4 Expression Calculation | 83 |
| | A.5 Correlated Sub-queries | 83 |
| | A.6 Parallelism and Concurrency | 84 |
| | A.7 RDF and Graph Specifics | 84 |
| В | Scale Factor Statistics | 86 |
| | B.1 Scale Factor Statistics | 86 |

ACKNOWLEDGMENTS

Special thanks to all the people that have contributed to the development of this benchmark:

- Renzo Angles (Universidad de Talca)
- Alex Averbuch (Neo Technologies)
- Peter Boncz (Vrije Universiteit Amsterdam and CWI)
- Orri Erling (Google)
- Andrey Gubichev (Google)
- Moritz Kaufmann (Tableau)
- Josep Lluís Larriba Pey (Universitat Politècnica de Catalunya)
- Minh-Duc Pham (Altran)
- Marcus Paradies (SAP)
- Arnau Prat (Sparsity Technologies)
- Mirko Spasić (Openlink Software)
- Norbert Martínez (Huawei Technologies)
- Gábor Szárnyas (MTA-BME Lendület Research Group on Cyber-Physical Systems)

DEFINITIONS

DATAGEN: Is the data generator provided by the LDBC SNB, which is responsible for generating the data needed to run the benchmark.

DBMS: A DataBase Management System.

LDBC SNB: Linked Data Benchmark Council Social Network Benchmark.

Query Mix: Refers to the ratio between read and update queries of a workload, and the frequency at which they are issued.

SF (**Scale Factor**): The LDBC SNB is designed to target systems of different size and scale. The scale factor determines the size of the data used to run the benchmark, measured in Gigabytes.

SUT: The System Under Test is defined to be the database system where the benchmark is executed.

Test Driver: A program provided by the LDBC SNB, which is responsible for executing the different workloads and gathering the results.

Full Disclosure Report (FDR): The FDR is a document which allows reproduction of any benchmark result by a third party. This contains complete description of the SUT and the circumstances of the benchmark run, e.g. configuration of SUT, dataset and test driver, etc.

Test Sponsor: The Test Sponsor is the company officially submitting the Result with the FDR and will be charged the filing fee. Although multiple companies may sponsor a Result together, for the purposes of the LDBC processes the Test Sponsor must be a single company. A Test Sponsor need not be a LDBC member. The Test Sponsor is responsible for maintaining the FDR with any necessary updates or corrections. The Test Sponsor is also the name used to identify the Result.

Workload: A workload refers to a set of queries of a given nature (i.e. interactive, analytical, business), how they are issued and at which rate.

1 Introduction

1.1 Motivation for the Benchmark

The new era of data economy, based on large, distributed and complexly structured data sets, has brought on new and complex challenges in the field of data management and analytics. These data sets, usually modeled as large graphs, have attracted both industry and academia, due to new opportunities in research and innovation they offer. This situation has also opened the door for new companies to emerge, offering new non-relational and graph-like technologies that are called to play a significant role in upcoming years.

The change in the data paradigm calls for new benchmarks to test new emerging technologies, as they set a framework where different systems can compete and be compared in a fair way, they let technology providers identify the bottlenecks and gaps of their systems and, in general, drive the research and development of new information technology solutions. Without them, the uptake of these technologies is at risk by not providing the industry with clear, user-driven targets for performance and functionality.

The LDBC Social Network Benchmark (LDBC SNB) aims at being comprehensive benchmark setting the rules for the evaluation of graph-like data management technologies. LDBC SNB is designed to be a plausible look-alike of all the aspects of operating a social network site, as one of the most representative and relevant use case of modern graph-like applications.

LDBC SNB includes the Interactive Workload [4], which consists of user-centric transactional-like interactive queries, and the Business Intelligence Workload, which includes analytical queries to respond to business critic questions. Initially, a graph analytics workload was also included in the roadmap of LDBC SNB, but this was finally delegated to the Graphalytics benchmark project [6], which was adopted as an official LDBC graph analytics benchmark. LDBC SNB and Graphalytics combined target a broad range of systems with different nature and characteristics. LDBC SNB and Graphalytics aims at capturing the essential features of these usage scenarios while abstracting away details of specific business deployments.

This document contains the definition of the Interactive Workload and the first draft of the Business Intelligence Workload. This includes a detailed explanation of the data used in the LDBC SNB benchmark, a detailed description for all queries, and instructions on how to generate the data and run the benchmark with the provided software.

1.2 Relevance to Industry

LDBC SNB is intended to provide the following value to different stakeholders:

- For **end users** facing graph processing tasks, LDBC SNB provides a recognizable scenario against which it is possible to compare merits of different products and technologies. By covering a wide variety of scales and price points, LDBC SNB can serve as an aid to technology selection.
- For **vendors** of graph database technology, LDBC SNB provides a checklist of features and performance characteristics that helps in product positioning and can serve to guide new development.
- For **researchers**, both industrial and academic, the LDBC SNB dataset and workload provide interesting challenges in multiple choke-point areas, such as query optimization, (distributed) graph analysis, transactional throughput, and provides a way to objectively compare the effectiveness and efficiency of new and existing technology in these areas.

The technological scope of LDBC SNB comprises all systems that one might conceivably use to perform social network data management tasks:

• Graph database systems (e.g. Neo4j, InfiniteGraph, Sparksee, Titan) are novel technologies aimed at storing directed and labeled graphs. They support graph traverals, typically by means of APIs, though some of them also support some sort of graph oriented query language (e.g. Neo4j's Cypher). These systems' internal structures are typically designed to store dynamic graphs that change over time. They ofter support transactional queries with some degree of consistency, and value-based indexes to quickly

locate nodes and edges. Finally, their architecture is typically single-machine (non-cluster). These systems can potentially implement all three workloads, though Interactive and Business Intelligence workloads are where they will presumably be more competitive.

- Graph programming frameworks (e.g. Giraph, Signal/Collect, GraphLab, Green Marl) are designed to perform global graph queries computations, executed in parallel or lockstep. These computations are typically long latency, involving many nodes and edges and often consist of approximation answers to NP-complete problems. These systems expose an API, sometimes following a vertex-centric paradigm, and their architecture targets both single-machine and cluster systems. Though these systems will likely implement the Graph Analytics workload.
- RDF database systems (e.g. OWLIM, Virtuoso, BigData, Jena TDB, Stardog, Allegrograph) are systems that implement the SPARQL 1.1 query language, similar in complexity to SQL-92, which allows for structured queries, and simple traversals. RDF database system often come with additional support for simple reasoning (sameAs, subClass), text search and geospatial predicates. RDF database systems generally support transactions, but not always with full concurrency and serializability and their supposed strength is integrating multiple data sources (e.g. DBpedia). Their architecture is both single-machine and clustered, and they will likely target Interactive and Business Intelligence workloads.
- Relational database systems (e.g. Postgres, MySQL, Oracle, DB2, SQLserver, Virtuoso, MonetDB, Vectorwise, Vertica, but also Hive and Impala) treat data as relational, and queries are formulated in SQL and/or PL/SQL. Both single-machine and cluster systems exist. They do not normally support recursion, or stateful recursive algorithms, which makes them not at home in the Graph Analytics workloads
- NoSQL database systems (e.g. key-value stores such as HBase, REDIS, MongoDB, CouchDB, or even MapReduce systems like Hadoop and Pig) are cluster-based and scalable. Key-value stores could possibly implement the Interactive Workload, though its navigational aspects would pose some problems as potentially many key-value lookups are needed. MapReduce systems could be suited for the Graph Analytics workload, but their query latency would presumably be so high that the Business Intelligence workload would not make sense, though we note that some of the key-value stores (e.g. MongoDB) provide a MapReduce query functionality on the data that it stores which could make it suited for the Business Intelligence workload.

1.3 General Benchmark Overview

LDBC SNB aims at being a complete benchmark, designed with the following goals in mind:

- **Rich coverage**. LDBC SNB is intended to cover most demands encountered in the management of complexly structured data.
- **Modularity**. LDBC SNB is broken into parts that can be individually addressed. In this manner LDBC SNB stimulates innovation without imposing an overly high threshold for participation.
- **Reasonable implementation cost**. For a product offering relevant functionality, the effort for obtaining initial results with SNB should be small, on the order of days.
- **Relevant selection of challenges**. Benchmarks are known to direct product development in certain directions. LDBC SNB is informed by the state of the art in database research so as to offer optimization challenges for years to come while not having a prohibitively high threshold for entry.
- Reproducibility and documentation of results. LDBC SNB will specify the rules for full disclosure of benchmark execution and for auditing of benchmark runs. The workloads may be run on any equipment but the exact configuration and price of the hardware and software must be disclosed.

LDBC SNB benchmark is modeled around the operation of a real social network site. A social network site represents a relevant use case for the following reasons:

• It is simple to understand for a large audience, as it is arguably present to our every-day life in different shapes and forms.

- It allows testing a complete range of interesting challenges, by means of different workloads targeting systems of different nature and characteristics.
- A social network can be scaled, allowing the design of a scalable benchmark targeting systems of different sizes and budgets.

In Section 2.3, LDBC SNB defines the schema of the data used in the benchmark. The schema, represents a realistic social network, including people and their activity in the social network during a period of time. Personal information of each person, such as name, birthday, interests or places where people work or study, is included. Persons' activity is represented in the form of friendhisp relationships and content sharing (i.e. messages and pictures). LDBC SNB provides a scalable synthetic data generator based on the MapReduce parallel paradigm, that produces networks with the described schema with distributions and correlations similar to those expected in a real social network. Furthermore, the data generator is designed to be user-friendly. The proposed data schema is shared by all the different proposed workloads, those we currently have, and those that will be proposed in the future.

In chapter 3, the Interactive Workload and the first draft of the Business Intelligence workload are proposed. Workloads are designed to mimic the different usage scenarios found in operating a real social network site, and each of them targets one or more types of systems. Each workload defines a set of queries and query mixes, designed to stress the SUTs in different choke-point areas, while being credible and realistic. Interactive workload reproduces the interaction between the users of the social network by including lookups and transactions that update small portions of the data base. These queries are designed to be interactive and target systems capable of responding such queries with low latency for multiple concurrent users. Business Intelligence workload, represents those business intelligence analytics a social network company would like to perform in the social network, in order to take advantage of the data to discover new business opportunities. This workload explores moderate to large portions of data from different entities, and performing more resource intensive operations.

LDBC SNB provides an execution test driver, which is responsible for executing the workloads and gathering the results. The driver is designed with simplicity and portability in mind, to ease the implementation on systems with different nature and characteristics, at a low implementation cost. Furthermore, it automatically handles the validation of the queries by means of a validation dataset provided by LDBC. The overall philosophy of LDBC SNB is to provide all the necessary software tools to run the benchmark, and therefore to reduce the benchmark's entry point as much as possible.

1.4 Related Projects

Along the Social Network Benchmark, LDBC [1] provides other benchmarks as well:

- The Semantic Publishing Benchmark (SPB)¹ measures the performance of RDF engines operating on semantic data sets.
- The Graphalytics benchmark [7] measures the performance of graph analysis operations (e.g. PageRank, local clustering coefficient).

1.5 Participation of Industry and Academia

The list of institutions that take part in the definition and development of LDBC SNB is formed by relevant actors from both the industry and academia in the field of linked data management. All the participants have contributed with their experience and expertise in the field, making a credible and relevant benchmark that meets all the desired needs. The list of participants is the following:

- FOUNDATION FOR RESEARCH AND TECHNOLOGY HELLAS
- MTA-BME LENDUELET RESEARCH GROUP ON CYBER-PHYSICAL SYSTEMS
- NEO4J

¹http://ldbcouncil.org/benchmarks/spb

- ONTOTEXT
- OPENLINK
- TECHNISCHE UNIVERSITAET MUENCHEN
- UNIVERSITAET INNSBRUCK
- UNIVERSITAT POLITECNICA DE CATALUNYA
- VRIJE UNIVERSITEIT AMSTERDAM

Besides the aforementioned institutions, during the development of the benchmark several meetings with the technical and users community have been conducted, receiving an invaluable feedback that has contributed to the whole development of the benchmark in every of its aspects.

2 BENCHMARK SPECIFICATION

2.1 Requirements

LDBC SNB is designed to be flexible and to have an affordable entry point. From small single node and in memory systems to large distributed multi-node clusters have its own place in LDBC SNB. Therefore, the requirements to fulfill for executing LDBC SNB are limited to pure software requirements to be able to run the tools. All the software provided by LDBC SNB have been developed and tested under Linux-based operating systems.

LDBC SNB does not impose the usage of any specific type of system, as it targets systems of different nature and characteristics, from graph databases, graph processing frameworks and RDF systems, to traditional relational database management systems. Consequently, any language or API capable of expressing the proposed queries can be used. Similarly, data can be stored in the most convenient manner the test sponsor may decide.

2.2 Software and Useful Links

- LDBC Driver 0.3 https://github.com/ldbc/ldbc_driver: The driver responsible for executing the LDBC SNB workload.
- DATAGEN 0.2.5 https://github.com/ldbc/ldbc_snb_datagen: The data generator used to generate the datasets of the benchmark.

2.3 Data

This section introduces the data used by LDBC SNB. This includes the different data types, the data schema, how it is generated and the different scale factors.

2.3.1 Data Types

Table 2.1 describes the different types used in the whole benchmark.

| Type | Description | | | |
|----------------|---|--|--|--|
| ID | integer type with 64-bit precision. All IDs within a single entity are unique | | | |
| 32-bit Integer | integer type with 32-bit precision | | | |
| 64-bit Integer | integer type with 64-bit precision | | | |
| String | variable length text of size 40 Unicode characters | | | |
| Long String | variable length text of size 256 Unicode characters | | | |
| Text | variable length text of size 2000 Unicode characters | | | |
| Date | date with a precision of a day, encoded as a string with the following format: yyyy-mm- | | | |
| | dd, where yyyy is a four-digit integer representing the year, the year, mm is a two-digit | | | |
| | integer representing the month and dd is a two-digit integer representing the day. | | | |
| DateTime | date with a precision of milliseconds, encoded as a string with the following format: | | | |
| | yyyy-mm-ddTHH:MM:ss.sss+0000, where yyyy is a four-digit integer representing the | | | |
| | year, the year, mm is a two-digit integer representing the month and dd is a two-digit | | | |
| | integer representing the day, HH is a two-digit integer representing the hour, MM is | | | |
| | a two digit integer representing the minute and ss.sss is a five digit fixed point real | | | |
| | number representing the seconds up to millisecond precision. Finally, the $+0000$ of | | | |
| | the end represents the timezone, which in this case is always GMT. | | | |

Table 2.1: Description of the data types.

2.3.2 Data Schema

Figure 2.1 shows the data schema in UML. The schema defines the structure of the data used in the benchmark in terms of entities and their relations. Data represents a snapshot of the activity of a social network during a period of time. Data includes entities such as Persons, Organisations, and Places. The schema also models the way persons interact, by means of the friendship relations established with other persons, and the sharing of content such as messages (both textual and images), replies to messages and likes to messages. People form groups to talk about specific topics, which are represented as tags.

LDBC SNB has been designed to be flexible and to target systems of different nature and characteristics. As such, it does not force any particular internal representation of the schema. The DATAGEN component supports multiple output data formats to fit the needs of different types of systems, including RDF, relational DBMS and graph DBMS.

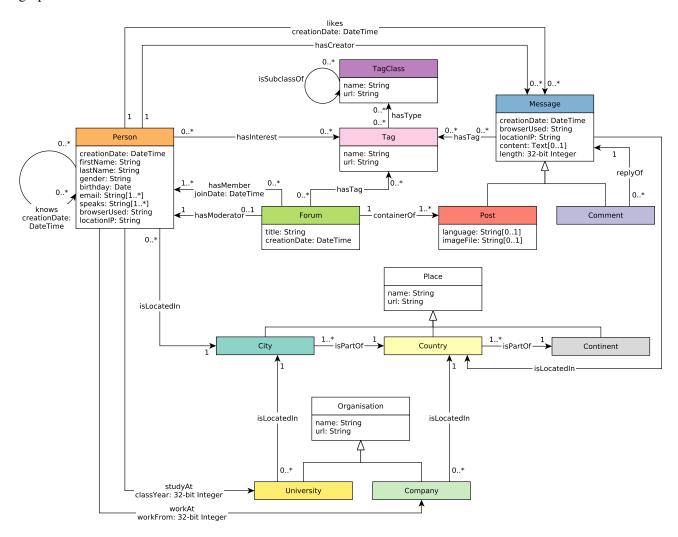


Figure 2.1: The LDBC SNB data schema

The schema specifies different entities, their attributes and their relations. All of them are described in the following sections.

Textual Restrictions and Notes

- Posts have content or imageFile. They have one of them but not both. The one they do not have is an empty string.
- Posts in a forum can be created by a non-member person if and only if that person is a modeartor.

Entities

City: a sub-class of a Place, and represents a city of the real world. City entities are used to specify where persons live, as well as where universities operate.

Comment: a sub-class of a Message, and represents a comment made by a person to an existing message (either a Post or a Comment).

Company: a sub-class of an Organisation, and represents a company where persons work.

Country: a sub-class of a Place, and represents a continent of the real world.

Forum: a meeting point where people post messages. Forums are characterized by the topics (represented as tags) people in the forum are talking about. Although from the schema's perspective it is not evident, there exist three different types of forums: persons' personal walls, image albums, and groups. They are distinguished by their titles. Table 2.2 shows the attributes of Forum entity.

| Attribute | Type | Description | | |
|--------------|-------------|---------------------------------|--|--|
| id ID | | The identifier of the forum. | | |
| title | Long String | The title of the forum. | | |
| creationDate | DateTime | The date the forum was created. | | |

Table 2.2: Attributes of Forum entity.

Message: an abstract entity that represents a message created by a person. Table 2.3 shows the attributes of Message abstract entity.

| Attribute | Type | Description | | |
|---------------------------------|----------------|--|--|--|
| id ID The identifier of the mea | | The identifier of the message. | | |
| browserUsed | String | The browser used by the Person to create the message. | | |
| creationDate | DateTime | The date the message was created. | | |
| locationIP | String | The IP of the location from which the message was created. | | |
| content | Text[01] | The content of the message. | | |
| length | 32-bit Integer | The length of the content. | | |

Table 2.3: Attributes of Message interface.

Organisation: an institution of the real world. Table 2.4 shows the attributes of Organisation entity.

| Attribute | Type | Description | | |
|---|-------------|-------------------------------------|--|--|
| id ID The identifier of the organisation. | | The identifier of the organisation. | | |
| name | Long String | The name of the organisation. | | |
| url | Long String | The URL of the organisation. | | |
| url | Long String | The URL of the organization. | | |

Table 2.4: Attributes of Organisation entity.

Person: the avatar a real world person creates when he/she joins the network, and contains various information about the person as well as network related information. Table 2.5 shows the attributes of Person entity.

| Attribute | Type | Description | | |
|--------------|-----------------|--|--|--|
| id | ID | The identifier of the person. | | |
| firstName | String | The first name of the person. | | |
| lastName | String | The last name of the person. | | |
| gender | String | The gender of the person. | | |
| birthDay | Date | The birthday of the person. | | |
| email | Long String[1*] | The set of emails the person has. | | |
| speaks | String[1*] | The set of languages the person speaks. | | |
| browserUser | String | The browser used by the person when he/she registered to the social network. | | |
| locationIp | String | The IP of the location from which the person was registered to the social network. | | |
| creationDate | DateTime | The date the person joined the social network. | | |

Table 2.5: Attributes of Person entity.

Place: a place in the world. Table 2.6 shows the attributes of Place entity.

| Attribute | Type | Description | |
|-------------------|-------------|------------------------------|--|
| id ID The identif | | The identifier of the place. | |
| name | Long String | The name of the place. | |
| url | Long String | The URL of the place. | |

Table 2.6: Attributes of Place entity.

Post: a sub-class of Message, that is posted in a forum. Posts are created by persons into the forums where they belong. Posts contain either content or imageFile, always one of them but never both. The one they do not have is an empty string. Table 2.7 shows the attributes of Post entity.

| Attribute | Type | Description |
|---------------------|------------|----------------------------|
| language String[01] | | The language of the post. |
| imageFile | String[01] | The image file of the post |

Table 2.7: Attributes of Post entity.

Tag: a topic or a concept. Tags are used to specify the topics of forums and posts, as well as the topics a person is interested in. Table 2.8 shows the attributes of Tag entity.

| Attribute | Type | Description | |
|----------------------------------|-------------|----------------------------|--|
| id ID The identifier of the tag. | | The identifier of the tag. | |
| name | Long String | The name of the tag. | |
| url | Long String | The URL of the tag. | |

Table 2.8: Attributes of Tag entity.

TagClass: a class or a category used to build a hierarchy of tags. Table 2.9 shows the attributes of TagClass entity.

| Attribute | Туре | Description | |
|---------------------------------------|-------------|---------------------------------|--|
| id ID The identifier of the tagclass. | | The identifier of the tagclass. | |
| name | Long String | The name of the tagclass. | |
| url Long String | | The URL of the tagclass. | |

Table 2.9: Attributes of TagClass entity.

University: a sub-class of Organisation, and represents an institution where persons study.

Relations

Relations connect entities of different types. Entities are defined by their "id" attribute.

| Name | Tail | Head | Type | Description | |
|--------------|----------------|--------------|------|---|------------------------------|
| containerOf | Forum[1] | Post[1*] | D | A Forum and a Post contained in it | |
| hasCreator | Message[0*] | Person[1] | D | A Message and its creator (Person) | |
| hasInterest | Person[0*] | Tag[0*] | D | A Person and a | Tag representing a topic the |
| | | | | person is interested in | |
| hasMember | Forum[0*] | Person[1*] | D | A Forum and a member (Person) of the fo- | |
| | | | | rum | |
| | | | | Attribute | joinDate |
| | | | | Type | DateTime |
| | | | | Description | The Date the person |
| | | | | Description | joined the forum |
| hasModerator | Forum[0*] | Person[1] | D | A Forum and it | ts moderator (Person) |
| hasTag | Message[0*] | Tag[0*] | D | A Message and | a Tag representing the mes- |
| | | | | sage's topic | |
| hasTag | Forum[0*] | Tag[0*] | D | A Forum and | a Tag representing the fo- |
| | | | | rum's topic | |
| hasType | Tag[0*] | TagClass[0*] | D | | gClass the tag belongs to |
| isLocatedIn | Company[0*] | Country[1] | D | | d its home Country |
| isLocatedIn | Message[0*] | Country[1] | D | A Message and the Country from which it | |
| | | | | was issued | |
| isLocatedIn | Person[0*] | City[1] | D | A Person and the | - |
| isLocatedIn | University[0*] | City[1] | D | A University and the City where the uni- | |
| | | | | versity is | |
| isPartOf | City[1*] | Country[1] | D | • | Country it is part of |
| isPartOf | Country[1*] | Continent[1] | D | | the Continent it is part of |
| isSubclassOf | TagClass[0*] | TagClass[0*] | D | A TagClass its parent TagClass Two Persons that know each other | |
| knows | Person[0*] | Person[0*] | U | | |
| | | | | Attribute | creationDate |
| | | | | Type | DateTime |
| | | | | Description | The date the knows |
| | | | | | relation was established |
| likes | Person[0*] | Message[0*] | D | A Person that likes a Message | |
| | | | | Attribute | creationDate |
| | | | | Type | DateTime |
| | | | | Description | The date the like was |
| | | | | Description | issued |

| replyOf | Comment[0*] | Message[1] | D | A Comment and the Message it replies | | |
|---------|-------------|----------------|---|--|-------------------------------|--|
| studyAt | Person[0*] | University[0*] | D | A Person and a University it has studied | | |
| | | | | Attribute | classYear | |
| | | | | Type | 32-bit Integer | |
| | | | | Description | The year the person | |
| | | | | Description | graduated | |
| workAt | Person[0*] | Company[0*] | D | A Person and a | Person and a Company it works | |
| | | | | Attribute | workFrom | |
| | | | | Type | 32-bit Integer | |
| | | | | | The year the person | |
| | | | | Description | started to work at that | |
| | | | | | company | |

Table 2.10: Description of the data relations.

Domain Concepts

A thread consist of Messages, starting with a single Post and Comments that transitively reply to that Post.

2.3.3 Output Data

DATAGEN produces outputs three different items:

- **Dataset**: The dataset to be bulk loaded by the SUT. It corresponds to roughly the 90% of the total generated network.
- **Update Streams**: A set of update streams containing update queries, which are used by the driver to generate the update queries of the workloads. This update streams correspond to the remaining 10% of the generated dataset.
- **Substitution Parameters**: A set of files containing the different parameter bindings that will be used by the driver to generate the read queries of the workloads.

The SUT have to take care only of the generated Dataset to be bulk loaded. Three different formats are supported by DATAGEN:

- **CSV:** Data output in CSV format, one file per different entity and on file per different relation. Also, there is a file por those attributes whose cardinality is larger than one (i.e. Person.email, Person.speaks, etc.).
- **CSVMergeForeign:** Similar to CSV format, but in this case, those relations of the form 1 to 1 and 1 to N, are stored in the tail entity file as a foreign keys.
- Turtle: Dataset in Turtle format for RDF systems.

CSV

This is a comma separated format. Each entity, relation and properties with a cardinality larger than one, are output in a separate file. Generated files are summarized at Table 2.11. Depending on the number of threads used for generating the dataset, the number of files varies, since there is a file generated per thread. The * in the file names indicates a number between 0 and NumberOfThreads -1.

CSV_MERGE_FOREIGN

This is a comma separated format. It is similar to CSV, but those relations connecting two entities A and B, where an entity A has a cardinality of one, A is output as a column of entity B. Generated files are summarized at Table 2.12. Depending on the number of threads used for generating the dataset, the number of files

| File | Content |
|--------------------------------------|---|
| comment_*.csv | id creationDate locationIP browserUsed content length |
| comment_hasCreator_person_*.csv | Comment.id Person.id |
| comment_hasTag_tag_*.csv | Comment.id Tag.id |
| comment_isLocatedIn_place_*.csv | Comment.id Place.id |
| comment_replyOf_comment_*.csv | Comment.id Comment.id |
| comment_replyOf_post_*.csv | Comment.id Post.id |
| forum_*.csv | id title creationDate |
| forum_containerOf_post_*.csv | Forum.id Post.id |
| forum_hasMember_person_*.csv | Forum.id Person.id joinDate |
| forum_hasModerator_person_*.csv | Forum.id Person.id |
| forum_hasTag_tag_*.csv | Forum.id Tag.id |
| organisation_*.csv | id type("university", "company") name url |
| organisation_isLocatedIn_place_*.csv | Organisation.id Place.id |
| person_*.csv | id firstName lastName gender birthday creationDate locationIP browserUsed |
| person_email_emailaddress_*.csv | Person.id email |
| person_hasInterest_tag_*.csv | Person.id Tag.id |
| person_isLocatedIn_place_*.csv | Person.id Place.id |
| person_knows_person_*.csv | Person.id Person.id creationDate |
| person_likes_comment_*.csv | Person.id Post.id creationDate |
| person_likes_post_*.csv | Person.id Post.id creationDate |
| person_speaks_language_*.csv | Person.id language |
| person_studyAt_organisation_*.csv | Person.id Organisation.id class Year |
| person_workAt_organisation_*.csv | Person.id Organisation.id workFrom |
| place_*.csv | id name url type("city", "country", "continent") |
| place_isPartOf_place_*.csv | Place.id Place.id |
| post_*.csv | id imageFile creationDate locationIP browserUsed language content length |
| post_hasCreator_person_*.csv | Post.id Person.id |
| post_hasTag_tag_*.csv | Post.id Tag.id |
| post_isLocatedIn_place.csv | Post.id Place.id |
| tag_*.csv | id name url |
| tag_hasType_tagclass_*.csv | Tag.id TagClass.id |
| tagclass_*.csv | id name url |
| tagclass_isSubclassOf_tagclass_*.csv | TagClass.id TagClass.id |

Table 2.11: Files output by the CSV serializer (33 in total)

varies, since there is a file generated per thread. The * in the file names indicates a number between 0 and NumberOfThreads -1.

| File | Content |
|--------------------------------------|---|
| comment_*.csv | id creationDate locationIP browserUsed content length creator place replyOfPost replyOfComment |
| comment_hasTag_tag_*.csv | Comment.id Tag.id |
| forum_*.csv | id title creationDate moderator |
| forum_hasMember_person_*.csv | Forum.id Person.id joinDate |
| forum_hasTag_tag_*.csv | Forum.id Tag.id |
| organisation_*.csv | id type("university", "company") name url |
| person_*.csv | id firstName lastName gender birthday creationDate locationIP browserUsed place |
| person_email_emailaddress_*.csv | Person.id email |
| person_hasInterest_tag_*.csv | Person.id Tag.id |
| person_knows_person_*.csv | Person.id Person.id creationDate |
| person_likes_comment_*.csv | Person.id Post.id creationDate |
| person_likes_post_*.csv | Person.id Post.id creationDate |
| person_speaks_language_*.csv | Person.id language |
| person_studyAt_organisation_*.csv | Person.id Organisation.id classYear |
| person_workAt_organisation_*.csv | Person.id Organisation.id workFrom |
| place_*.csv | id name url type("city", "country", "continent") |
| post_*.csv | id imageFile creationDate locationIP browserUsed language content length creator Forum.id place |
| post_hasTag_tag_*.csv | Post.id Tag.id |
| tag_*.csv | id name url |
| tag_hasType_tagclass_*.csv | Tag.id TagClass.id |
| tagclass_*.csv | id name url |
| tagclass_isSubclassOf_tagclass_*.csv | TagClass.id TagClass.id |

Table 2.12: Files output by the CSV_MERGE_FOREIGN serializer (22 in total)

Turtle

This is the standard Turtle¹ format. DATAGEN outputs two files: <code>0_ldbc_socialnet_static_dbp.ttl</code> and <code>0_ldbc_socialnet.ttl</code>.

2.3.4 Scale Factors

LDBC SNB defines a set of scale factors (SFs), targeting systems of different sizes and budgets. SFs are computed based on the ASCII size in Gigabytes of the generated output files using the CSV serializer. For example, SF 1 weights roughly 1 GB in CSV format, SF 3 weights roughly 3 GB and so on and so forth. The proposed SFs are the following: 1, 3, 10, 30, 100, 300, 1000. The Test Sponsor may select the SF that better fits their needs, by properly configuring the DATAGEN.

The size of the resulting dataset, is mainly affected by the following configuration parameters: the number of persons and the number of years simulated. Different SFs are computed by scaling the number of Persons in the network, while fixing the number of years simulated. Table 2.13 shows the parameters used in each of the SFs.

| Scale Factor | 1 | 3 | 10 | 30 | 100 | 300 | 1000 |
|--------------|------|------|------|------|------|-------|------|
| # of Persons | 11K | 27K | 73K | 182K | 499K | 1.25M | 3.6M |
| # of Years | 3 | 3 | 3 | 3 | 3 | 3 | 3 |
| Start Year | 2010 | 2010 | 2010 | 2010 | 2010 | 2010 | 2010 |

Table 2.13: Parameters of each scale factor.

For example, SF 30 consists of the activity of a social network of 182K users during a period of three years, starting from 2010.

¹Description of the Turtle RDF format http://www.w3.org/TR/turtle/

3 Workloads

3.1 Query Description Format

Queries are described in natural language using a well-defined structure that consists of three sections: *description*, a concise textual description of the query; *parameters*, a list of input parameters and their types; and *results*, a list of expected results and their types. The syntax used in *parameters* and *results* sections is as follows:

- Entity: entity type in the dataset.
 - One word, possibly constructed by appending multiple words together, starting with uppercase character and following the camel case notation, e.g. TagClass represents an entity of type "TagClass".
- **Relationship**: relationship type in the dataset.
 - One word, possibly constructed by appending multiple words together, starting with lowercase character and following the camel case notation, and surrounded by arrow to communicate direction, e.g. –worksAt-> represents a directed relationship of type "worksAt".
- Attribute: attribute of an entity or relationship in the dataset.
 - One word, possibly constructed by appending multiple words together, starting with lowercase character and following the camel case notation, and prefixed by a "." to dereference the entity/relationship, e.g. Person.firstName refers to "firstName" attribute on the "Person" entity, and -studyAt->.classYear refers to "classYear" attribute on the "studyAt" relationship.
- Unordered Set: an unordered collection of distinct elements.
 - Surrounded by { and } braces, with the element type between them, e.g. {String} refers to a set of strings.
- Ordered List: an ordered collection where duplicate elements are allowed.
 - Surrounded by [and] braces, with the element type between them, e.g. [String] refers to a list of strings.
- Ordered Tuple: a fixed length, fixed order list of elements, where elements at each position of the tuple have predefined, possibly different, types.
 - Surrounded by < and > braces, with the element types between them in a specific order e.g. <String, Boolean> refers to a 2-tuple containing a string value in the first element and a boolean value in the second, and [<String, Boolean>] is an ordered list of those 2-tuples.

here comes the new stuff - Gabor

Results are categorized according to their source of origin:

- raw (R), if the result is returned with an unmodified value and type.
- calculated (C), if the result is calculated from other values and conditions.
- aggregated (A), if the result is an aggregated value, e.g. a count or a sum of another value. If a result is both calculated and aggregated (e.g. count(x) + count(y) or avg(x + y)), it is considered an aggregated result.

3.2 Query Definitions

3.2.1 Notation

Intervals. Closed interval boundaries are denoted with [and], while open interval boundaries are denoted with (and). For example, [0, 1) denoted an interval between 0 and 1, closed on the left and open on the right.

3.2.2 Morphisms

Discuss homomorphism vs. isomorphism

3.2.3 Graph Patterns

To illustrate queries, we use graph patterns such as Figure 3.1 with the following notation:

- Filtering conditions are typeset in *italic*.
- Properties that should be returned are denoted in normal (book) font.
- Negative conditions, i.e., edges that are now allowed in the graph are denoted with dashed red lines.
- Aggregations are shown in dashed boxes.

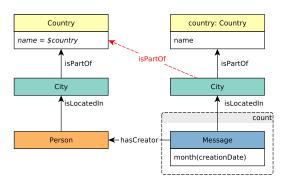


Figure 3.1: Example graph pattern.

Keywords. The notation uses a small set of keywords:

- UNWIND unnests a list, i.e. produces a set of one-tuples. For example, UNWIND [1,2,3] results in $\{\langle 1 \rangle, \langle 2 \rangle, \langle 3 \rangle\}$)
- Aggregation operations: count, average.
- Functions: floor, year (extracts the year from a given date), month (extracts the month from a given date).

Resolving ambiguity. Note that if the textual description and the graph pattern are different for a particular query (either due to an error or the lack of sophistication in the graphical syntax), *the textual description takes precedence*.

3.3 Substitution Parameters

Together with the dataset, DATAGEN produces a set of parameters per query type. Parameter generation is designed in such a way that for each query type, all of the generated parameters yield similar runtime behaviour of that query.

Specifically, the selection of parameters for a query template guarantees the following properties of the resulting queries:

- P1: the query runtime has a bounded variance: the average runtime corresponds to the behavior of the majority of the queries
- P2: the runtime distribution is stable: different samples of (e.g. 10) parameter bindings used in different query streams result in an identical runtime distribution across streams
- P3: the optimal logical plan (optimal operator order) of the queries is the same: this ensures that a specific query template tests the system's behavior under the well-chosen technical difficulty (e.g. handling voluminous joins or proper cardinality estimation for subqueries, etc.)

As a result, the amount of data that the query touches is roughly the same for every parameter binding, assuming that the query optimizer figures out a reasonable execution plan for the query. This is done to avoid bindings that cause unexpectedly long or short runtimes of queries, or even result in a completely different

optimal execution plan. Such effects could arise due to the data skew and correlations between values in the generated dataset.

In order to get the parameter bindings for each of the queries, we have designed a *Parameter Curation* procedure that works in two stages:

- 1. for each query template for all possible parameter bindings, we determine the size of intermediate results in the *intended* query plan. Intermediate result size heavily influences the runtime of a query, so two queries with the same operator tree and similar intermediate result sizes at every level of this operator tree are expected to have similar runtimes. This analysis is effectively a side effect of data generation, that is we keep all the necessary counts (number of friends per user, number of posts of friends etc.) as we create the dataset.
- 2. then, a greedy algorithm selects ("curates") those parameters with similar intermediate result counts from the domain of all the parameters.

Parameter bindings are stored in the substitution_parameters folder inside the data generator directory. Each query gets its bindings in a separate file. Every line of a parameter file is a JSON-formatted collection of key-value pairs (name of the parameter and its value). For example, the Query 1 parameter bindings are stored in file query_1_param.txt, and one of its lines may look like this:

```
{"PersonID": 1, "Name": "Lei", "PersonURI": "http://www.ldbc.eu/ldbc_socialnet/1.0/data/pers1"}
```

Depending on implementation, the SUT may refer to persons either by IDs (relational and graph databases) or URIs (RDF systems), so we provide both values for the Person parameter. Finally, parameters for short reads are taken from those in complex reads and updates.

3.4 Load Definition

LDBC SNB Test Driver is in charge of the execution of the Interactive Workload. At the beginning of the execution, the Test Driver creates a query mix by assigning to each query instance, a query issue time and a set of parameters taken from the generated substitution parameter set described above.

Query issue times have to be carefully assigned. Although substitution parameters are chosen in such a way that queries of the same type take similar time, not all query types have the same complexity and touch the same amount of data, which causes them to scale differently for the different scale factors. Therefore, if all query instances, regardless of their type, are issued at the same rate, those more complex queries will dominate the execution's result, making faster query types purposeless. To avoid this situation, each query type is executed at a different rate. The way the execution rate is decided, also depends on the nature of the query: complex read, short read or update.

Update queries' issue times are taken from the update streams generated by the data generator. These are the times where the actual event happened during the simulation of the social network. Complex reads' times are expressed in terms of update operations. For each complex read query type, a frequency value is assigned which specifies the relation between the number of updates performed per complex read. Table 3.1 shows the frequencies assigned to each query type for SF1. The frequencies of the different scale factors can be found in Section B.1.

| Query Type | freq | Query Type | freq |
|------------|------|------------|------|
| Query 1 26 | | Query 8 | 45 |
| Query 2 | 37 | Query 9 | 157 |
| Query 3 | 69 | Query 10 | 30 |
| Query 4 | 36 | Query 11 | 16 |
| Query 5 | 57 | Query 12 | 44 |
| Query 6 | 129 | Query 13 | 19 |
| Query 7 | 87 | Query 14 | 49 |

Table 3.1: Frequencies for each query type for SF1.

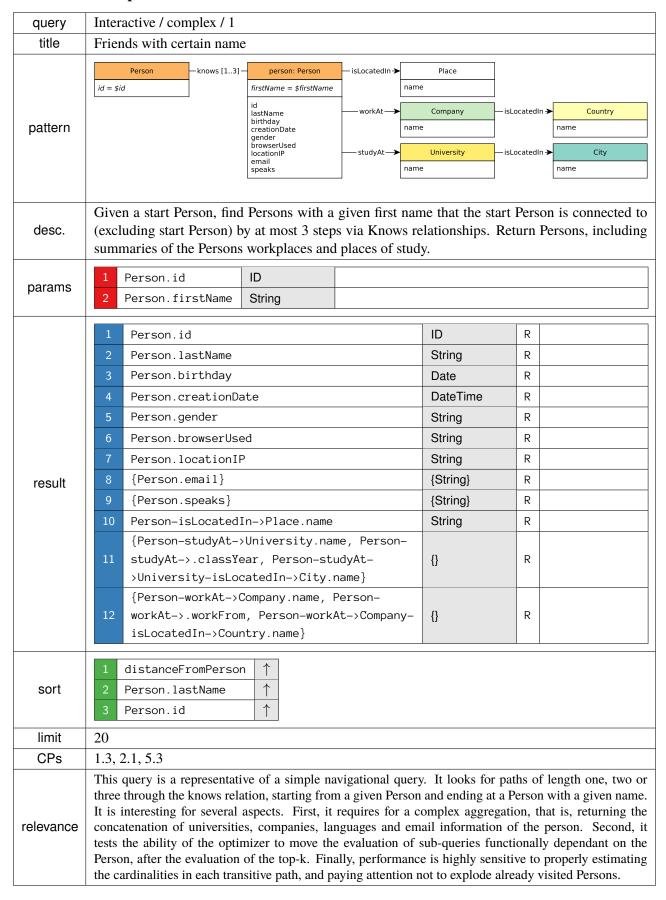
Finally, short reads are inserted in order to balance the ratio between reads and writes, and to simulate the behavior of a real user of the social network. For each complex read instance, a sequence of short reads is planned. There are two types of short read sequences: Person centric and Message centric. Depending on the type of the complex read, one of them is chosen. Each sequence consists of a set of short reads which are issued in a row. The issue time assigned to each short read in the sequence is determined at run time, and is based on the completion time of the complex read it depends on. The substitution parameters for short reads are taken from the results of previously executed complex reads and short reads. Once a short read sequence is issued (and provided that sufficient substitution parameters exist), there is a probability that another short read sequence is issued. This probability decreases for each new sequence issued. Since the same random number generator seed is used across executions, the workload is deterministic.

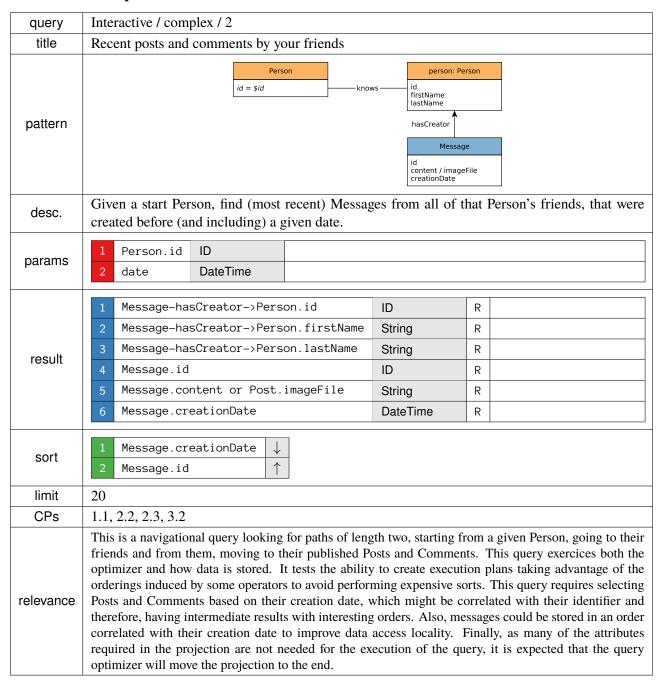
The specified frequencies, implicitly define the query ratios between queries of different types, as well as a default target throughput. However the Test Sponsor may specify a different target throughput to test, by "squeezing" together or "stretching" apart the queries of the workload. This is achieved by means of the "Time Compression Ratio" that is multiplied by the frequencies (see Table 3.1). Therefore, different throughputs can be tested while maintaining the relative ratios between the different query types.

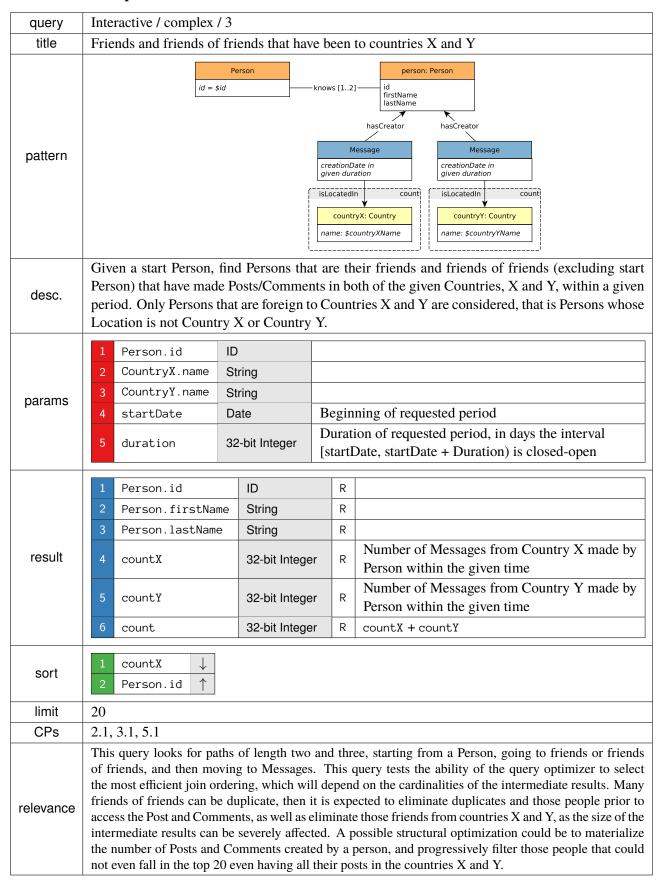
4 Interactive Workload

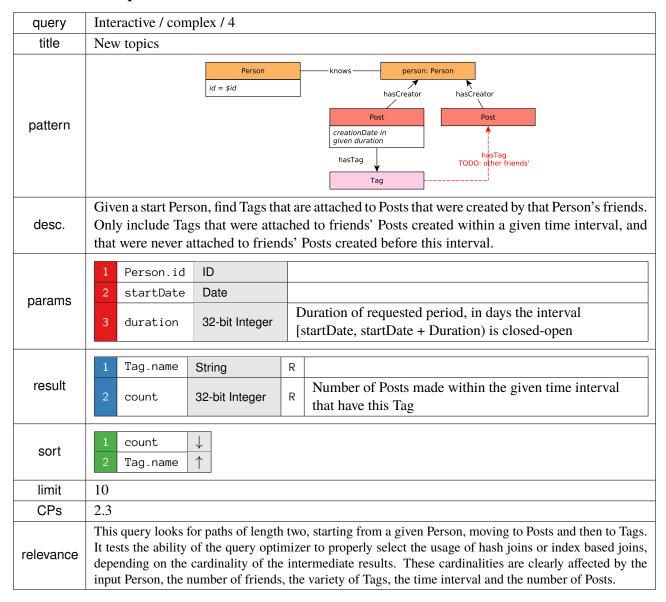
4.1 Complex Reads Query Descriptions

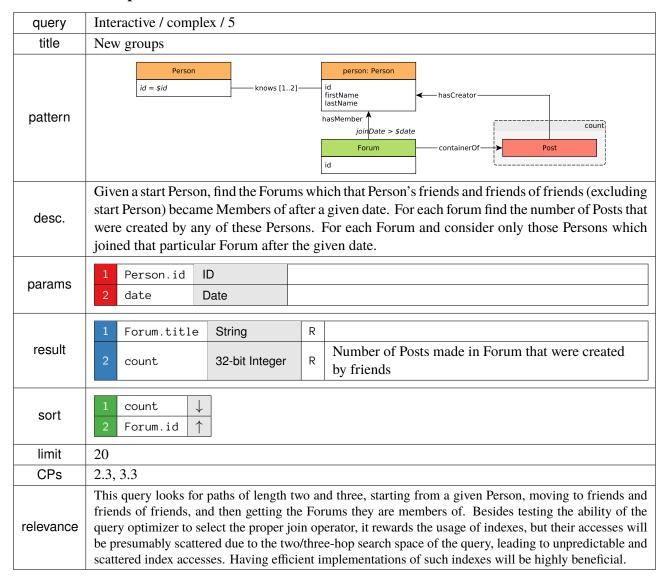
Query cards for complex read follow.

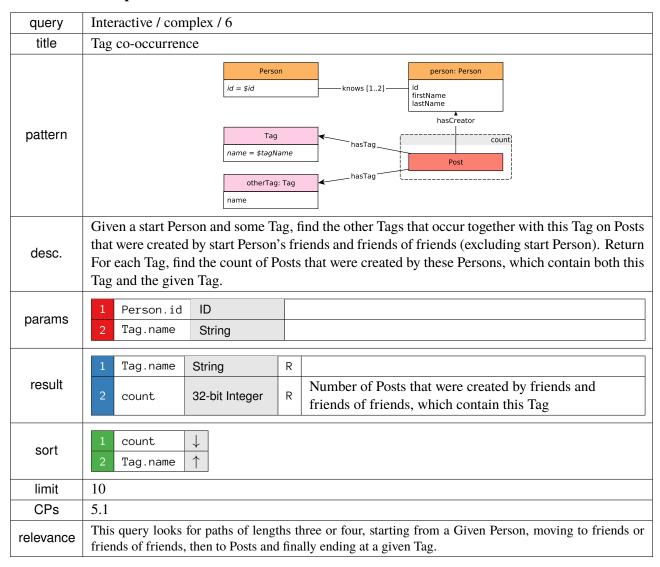


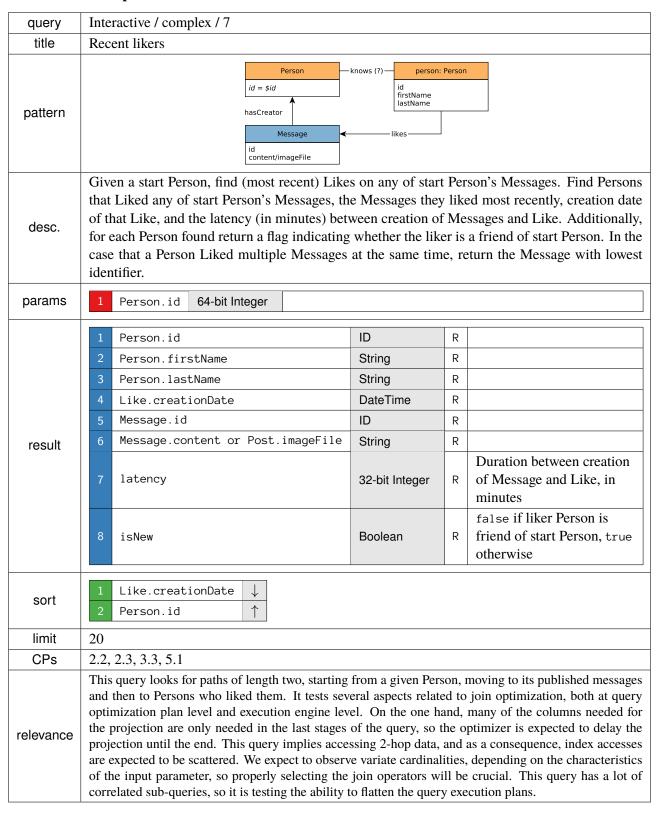


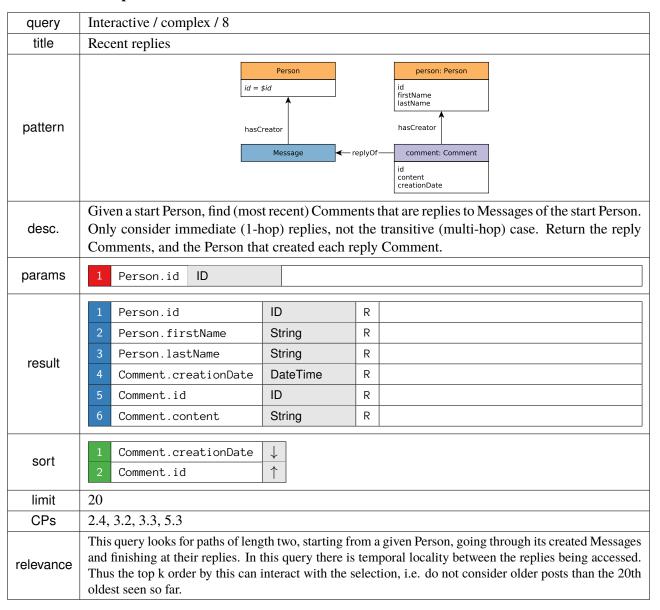


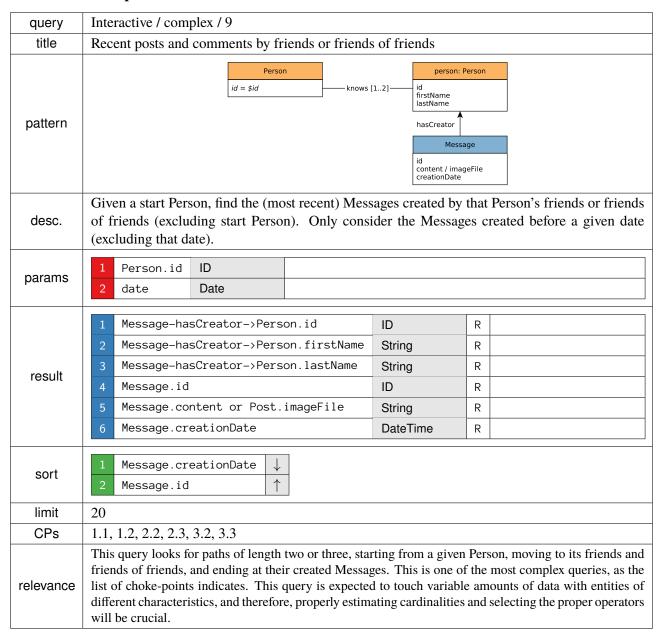




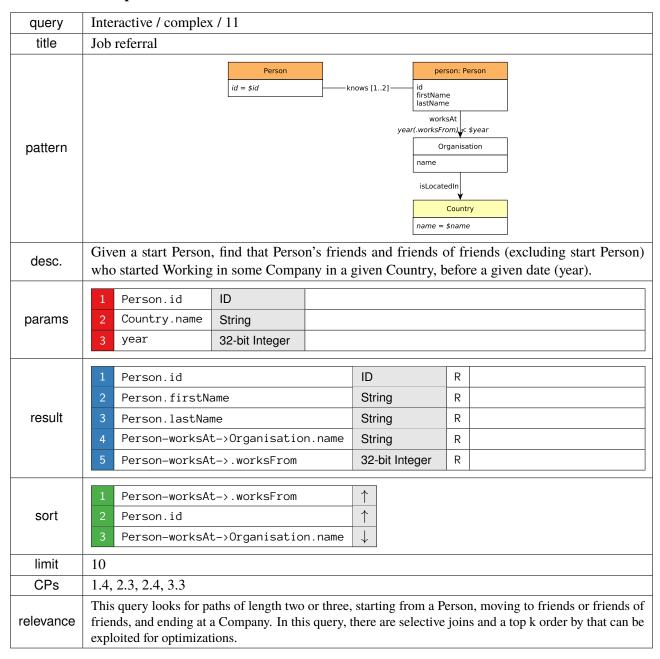


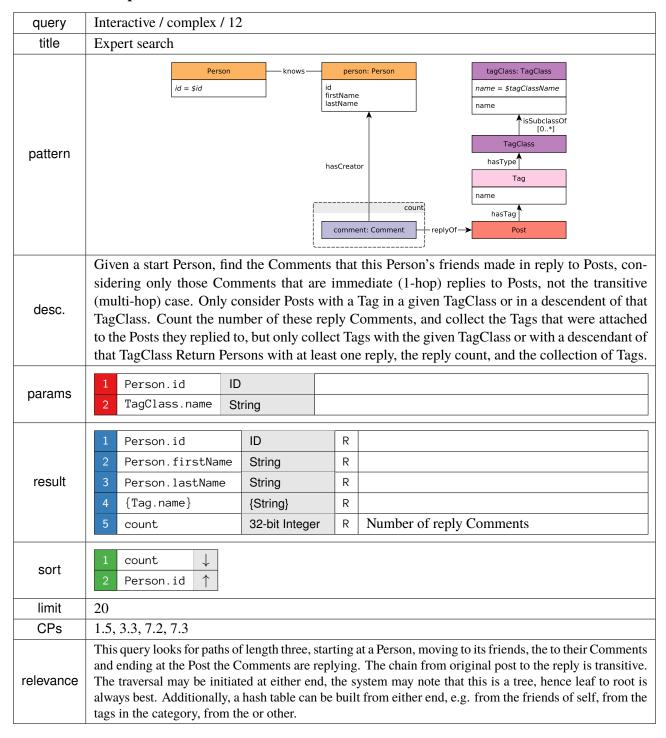






| query | Interactive / complex / 10 | | | | | |
|-----------|---|--|--|--|--|--|
| title | Friend recommendation | | | | | |
| pattern | Person id = \$id knows [22] birthDay cond's id firstName lastName gender Person Person hasCreator rag hasInterest Tag | | | | | |
| desc. | Given a start Person, find that Person's friends of friends (excluding start Person, and immediate friends), who were born on or after the 21st of a given month (in any year) and before the 22nd of the following month. Calculate the similarity between each of these Persons and start Person, where similarity for any Person is defined as follows: • common = number of Posts created by that Person, such that the Post has a Tag that start Person is Interested in • uncommon = number of Posts created by that Person, such that the Post has no Tag that start Person is Interested in • similarity = common - uncommon | | | | | |
| params | 1 Person.id ID 2 month 32-bit Integer Between 1-12 | | | | | |
| result | 1 Person.id ID R 2 Person.firstName String R 3 Person.lastName String R 4 similarity 32-bit Integer R 5 Person.gender String R 6 Person-isLocatedIn->Place.name String R | | | | | |
| sort | 1 similarity ↓ 2 Person.id ↑ | | | | | |
| limit | 10 | | | | | |
| CPs | 2.3, 3.3, 4.1, 4.2, 5.1, 5.2, 6.1, 7.1 | | | | | |
| relevance | This query looks for paths of length two, starting from a Person and ending at the friends of their friends. It does widely scattered graph traversal, and one expects no locality of in friends of friends, as these have been acquired over a long time and have widely scattered identifiers. The join order is simple but one must see that the anti-join for "not in my friends" is better with hash. Also the last pattern in the scalar sub-queries joining or anti-joining the tags of the candidate's posts to interests of self should be by hash. | | | | | |





Interactive / complex / 13

| query | Interactive / complex / 13 | | |
|-----------|---|--|--|
| title | Single shortest path | | |
| pattern | | | |
| desc. | Given two Persons, find the shortest path between these two Persons in the subgraph induced by the Knows relationships. Return the length of this path: • -1: no path found • 0: start person = end person • > 0: regular case | | |
| params | 1 person1.id ID 2 person2.id ID | | |
| result | 1 length 32-bit Integer R | | |
| CPs | 3.3, 7.2, 7.3 | | |
| relevance | This query looks for a variable length path, starting at a given Person and finishing at an another given Person. Proper cardinality estimation and search space prunning, will be crucial. This query also allows for possible parallel implementations. | | |

Interactive / complex / 14

| query | Interactive / complex / 14 | | |
|-----------|---|--|--|
| title | Weighted/unweighted paths | | |
| pattern | Person Person id = \$person1ld id = \$person2ld | | |
| desc. | Given two Persons, find all (unweighted) shortest paths between these two Persons, in the subgraph induced by the Knows relationship. Then, for each path calculate a weight. The nodes in the path are Persons, and the weight of a path is the sum of weights between every pair of consecutive Person nodes in the path. The weight for a pair of Persons is calculated such that every reply (by one of the Persons) to a Post (by the other Person) contributes 1.0, and every reply (by ones of the Persons) to a Comment (by the other Person) contributes 0.5. Return all the paths with shortest length, and their weights. Do not return any rows if there is now path between the two Persons. | | |
| params | 1 person1.id ID 2 person2.id ID | | |
| result | 1 [Person.id] [ID] R Identifiers representing an ordered sequence of the Persons in the path 2 weight 64-bit Float R | | |
| sort | 1 weight \ | | |
| CPs | 3.3, 7.2, 7.3 | | |
| relevance | This query looks for a variable length path, starting at a given Person and finishing at an another given Person. This is a more complex query as not only requires computing the path length, but returning it and computing a weight. To compute this weight one must look for smaller sub-queries with paths of length three, formed by the two Persons at each step, a Post and a Comment. | | |

4.2 Short Reads Query Descriptions

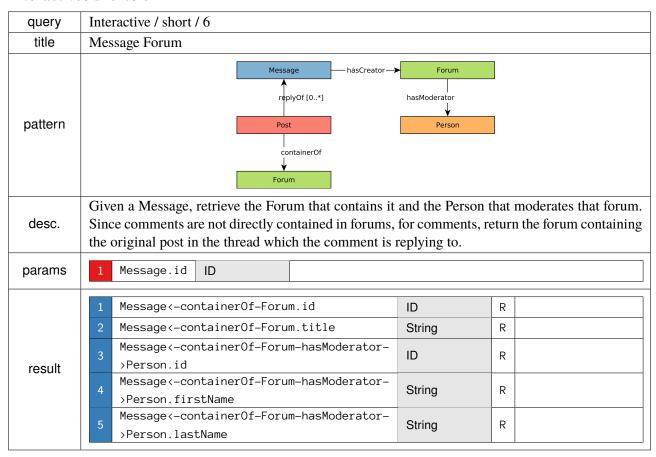
| query | Inte | Interactive / short / 1 | | | | |
|---------|------|--------------------------------------|--|----------------|-------|---------------------------------------|
| title | Pers | rson Profile | | | | |
| pattern | | | person: Person id lastName birthday ipAddress browser | isLocatedIn> | | city: City |
| desc. | | en a start Person, retr esidence. | ieve their first | name, last nam | e, bi | rthday, IP address, browser, and city |
| params | 1 | Person.id ID | | | | |
| | 1 | Person.firstName | | String | R | |
| | 2 | Person.lastName | | String | R | |
| | 3 | Person.birthDay | | Date | R | |
| result | 4 | Person.locationIP | | String | R | |
| resuit | 5 | Person.browserUse | d | String | R | |
| | 6 | Person-isLocatedI | n->Place.id | 32-bit Integer | R | |
| | 7 | Person.gender | | String | R | |
| | 8 | Person.creationDa | te | DateTime | R | |

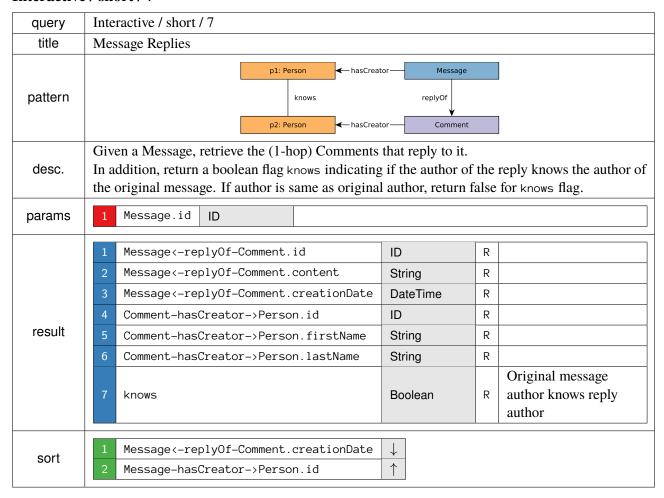
| query | Interactive / short / 2 | | |
|---------|---|--|--|
| title | Person Recent Messages | | |
| pattern | person: Person id lastName birthday ipAddress browser hasCreator op: Person | id lastName birthday ipAddress browser | |
| desc. | Given a start Person, retrieve the last 10 Messages of that message, the original post in its conversation Messages is a Post, then the original Post will be the twice in that result. | , and the author of that post. If any of the | |
| params | 1 Person.id ID | | |
| result | 1 Message.id 64-bit Integer R 2 Message.content or Post.imageFile String R 3 Message.creationDate DateTime R 4 Post.id or Comment-replyOf*->Post.id ID R 5 Post-hasCreator->Person.id or Comment-replyOf*->Porson.id Post-hasCreator->Person.id Post-hasCreator->Person.id Post-hasCreator->Person.firstName or Comment-replyOf*->Post-hasCreator-Serson.id Post-hasCreator->Post-hasCreator-Serson.id R 7 Post-hasCreator->Person.lastName or String R 7 Comment-replyOf*->Post-hasCreator-String R 7 >Person.lastName String R 7 >Person.lastName | | |
| sort | 1 Message.creationDate ↓ 2 Message.id ↓ | | |

| query | Interactive / short / 3 | | |
|---------|---|--|--|
| title | Person Friends | | |
| pattern | | | |
| desc. | Given a start Person, retrieve all of their friends, and the date at which they became friends. | | |
| params | 1 Person.id ID | | |
| result | 1 Person.id ID R 2 Person.firstName String R 3 Person.lastName String R 4 Knows.creationDate DateTime R | | |
| sort | 1 Knows.creationDate ↓ 2 Person.id ↑ | | |

| query | Interactive / short / 4 | |
|---------|--|--|
| title | Message Content | |
| pattern | Message content or Post.imageFile creationDate | |
| desc. | Given a Message, retrieve its content and creation date. | |
| params | 1 Message.id ID | |
| result | 1 Message.creationDate ID R 2 Message.content or Post.imageFile String R | |

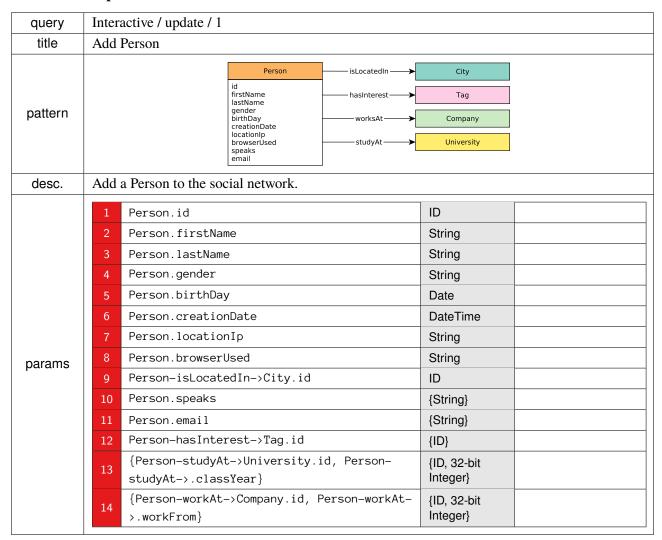
| query | Interactive / short / 5 | | |
|---------|---|--|--|
| title | Message Creator | | |
| pattern | hasCreator person: Person id lastName birthday ipAddress browser | | |
| desc. | Given a Message, retrieve its author. | | |
| params | 1 Message.id ID | | |
| result | 1 Message-hasCreator->Person.id ID R 2 Message-hasCreator->Person.firstName String R 3 Message-hasCreator->Person.lastName String R | | |





4.3 Update Query Descriptions

Each update query inserts either (1) a single vertex of a certain type, along with its edges to other existing vertices or (2) a single edge of a certain type between two existing vertices.



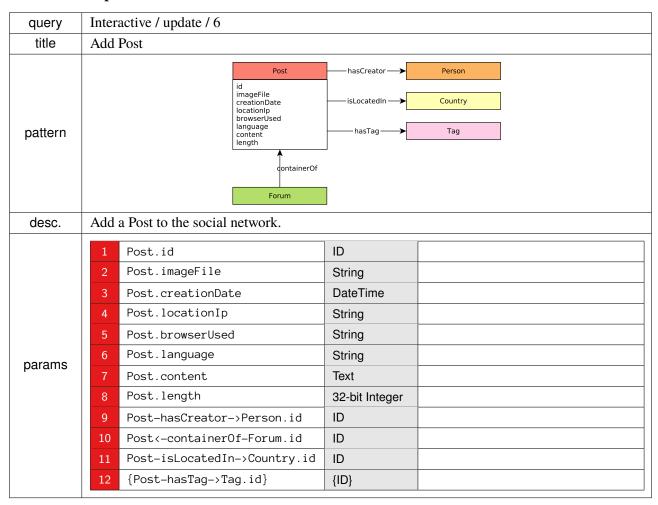
| query | Interactive / update / 2 | | |
|---------|---|--|--|
| title | Add Post Like | | |
| pattern | Person likes Post | | |
| desc. | Add a Like to a Post of the social network. | | |
| | 1 Person.id ID | | |
| params | 2 Post.id ID | | |
| | 3 Person-likes->.creationDate | | |

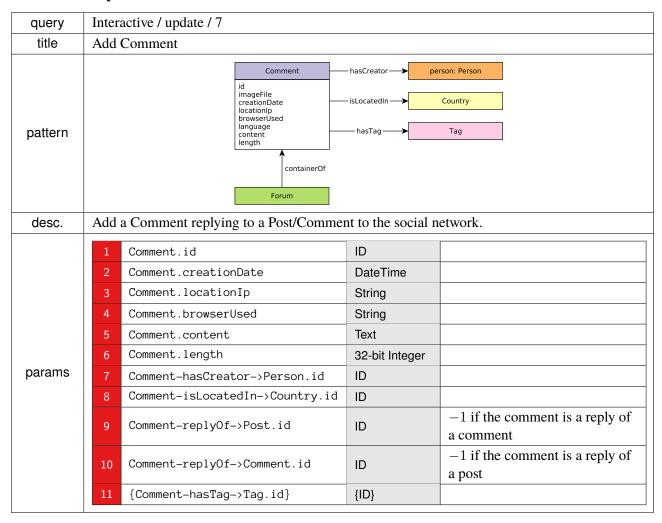
| query | Interactive / update / 3 | | |
|---------|--|--|--|
| title | Add Comment Like | | |
| pattern | Person likes — Comment | | |
| desc. | Add a Like to a Comment of the social network. | | |
| | 1 Person.id ID | | |
| params | 2 Comment.id ID | | |
| | 3 Person-likes->.creationDate | | |

Interactive / update / 4

| query | Interactive / update / 4 | | |
|---------|------------------------------------|------------------------|--|
| title | Add Forum | | |
| pattern | Forum I hasTag Tag | —hasModerator → Person | |
| desc. | Add a Forum to the social network. | | |
| | 1 Forum.id | ID | |
| | 2 Forum.title | String | |
| params | 3 Forum.creationDate | DateTime | |
| | 4 Forum-hasModerator->Person.id | ID | |
| | 5 Forum-hasTag->Tag.id | {ID} | |

| query | Interactive / update / 5 | |
|---------|---|--|
| title | Add Forum Membership | |
| pattern | person: Person hasMember → Forum | |
| desc. | Add a Forum membership to the social network. | |
| | 1 Person.id ID | |
| params | 2 Forum-hasMember->Person.id ID | |
| | 3 Forum-hasMember->.joinDate | |





| query | Interactive / update / 8 | | |
|---------|---|------------------------|--|
| title | Add Friendship | | |
| pattern | person1: Person | knows——person2: Person | |
| desc. | Add a friendship relation to the social network | | |
| | 1 Person.id II | Person 1 | |
| params | 2 Person.id II | Person 2 | |
| | 3 Person-knows->.creationDate D | PateTime | |

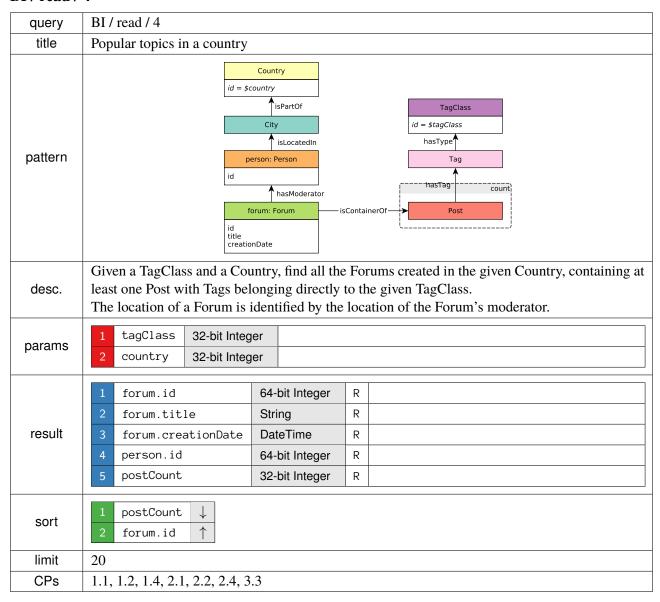
5 Business Intelligence Workload

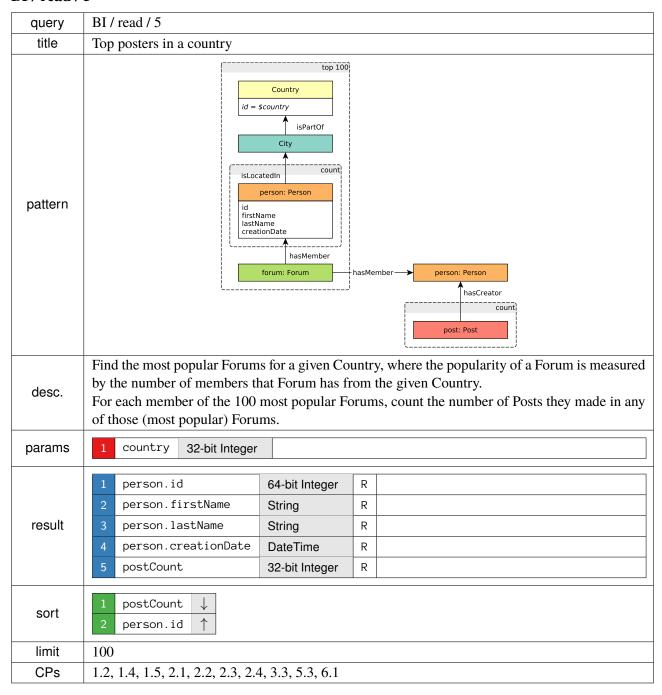
5.1 Read Query Descriptions

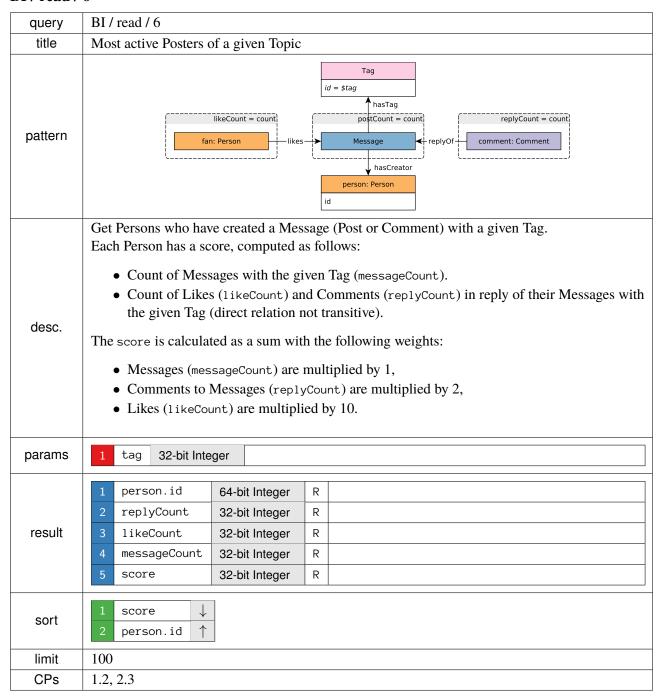
| query | BI / read / 1 | | | | | |
|---------|--|---|--|--|--|--|
| title | Posting summary | | | | | |
| pattern | message: Message creationDate < \$date length year(creationDate) | | | | | |
| desc. | Given a date, find all Messages created before that date. Group them by a 3-level grouping: 1. by year of creation 2. for each year, group into message types (Posts or Comments) 3. for each year-type group, split into four groups based on length of their content • 0 <= length < 40: short • 40 <= length < 80: one liner • 80 <= length < 160: tweet • 160 <= length: long | | | | | |
| params | 1 date Date | | | | | |
| | 1 messageYear 32-bit Integer R Year of the message 2 messageType String R post/comment (in lowercase) 3 lengthCategory String R short/one-liner/tweet/long (in lowercase) 4 messageCount 32-bit Integer R Total number of Messages (Posts/Comments) in that group | | | | | |
| result | averageMessageLength 32-bit Integer R Average length of the Message contract that group | | | | | |
| | 5 sumMessageLength 32-bit Integer R Sum of all message content length Number of messages in group as percentageOfMessages 32-bit Float R Percentage of all messages create the given date | a | | | | |
| sort | <pre>1 year</pre> | | | | | |
| CPs | 1.2, 3.2, 4.1 | | | | | |

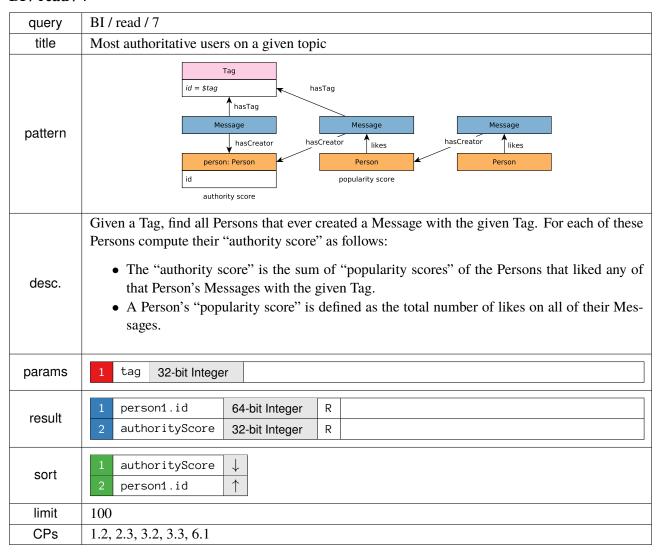
| query | BI / read / 2 | | | | |
|---------|---|--|--|--|--|
| title | Top tags for country, age, gender, time | | | | |
| pattern | country: Country name = \$country1 or name = \$country2 name hasTag hasTag message: Message \$date1 <= creationDate and creationDate and creationDate <= \$date2 month(creationDate) | | | | |
| desc. | Select all Messages (Posts & Comments) created between date1-date2 (inclusive) by persons located in country1 or country2. Select the creators (Persons) and the Tags of these Messages. Split these Persons, Tags and Messages into a 5-level grouping: 1. name of country of person, 2. month message was created, 3. gender of person, 4. age group of person, defined as years between person's birthday and end of simulation (2013-01-01), divided by 5, rounded down, 5. name of tag attached to message. Consider only those groups where number of messages is greater than 100. | | | | |
| params | 1 date1 Date 2 date2 Date 3 country1 String 4 country2 String | | | | |
| result | 1 country.name String R 2 messageMonth 32-bit Integer R 1-12 3 person.gender String R male/female 4 ageGroup 32-bit Integer R 5 tag.name String R 6 messageCount 64-bit Integer R 7 The number of messages in the group | | | | |
| sort | <pre>1 messageCount</pre> | | | | |
| limit | 100 | | | | |
| CPs | 1.1, 1.2, 1.4, 2.1, 2.3, 3.1, 3.2 | | | | |

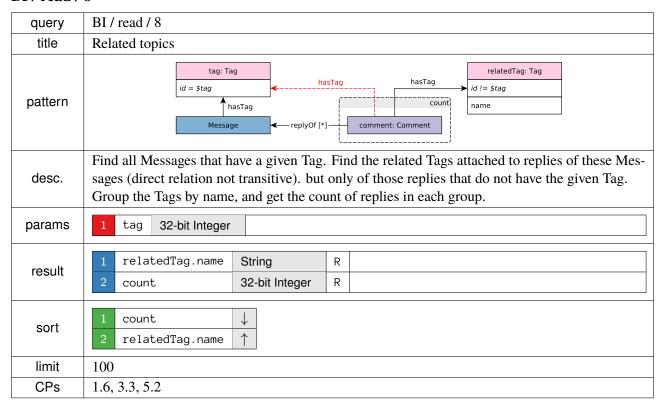
| query | BI / read / 3 | | | | |
|---------|--|--|--|--|--|
| title | Tag evolution | | | | |
| pattern | CountMonth1 = count Message year(creationDate) = \$year and month(creationDate) = \$month hasTag tag: Tag name CountMonth2 = count countMonth2 = count year(creationDate) = \$year + \$month/12 and month(creationDate) = \$month + \$month%12 tag: Tag name | | | | |
| desc. | Given a year and a month, find the Tags that were used in Messages during the given month of the given year, and the Tags that were used during the month after the given month of the given year. For both months, compute the count of Messages that used each of the Tags. | | | | |
| params | 1 year 32-bit Integer | | | | |
| | 2 month 32-bit Integer | | | | |
| | 1 tag.name String R | | | | |
| | 2 countMonth1 32-bit Integer R Occurrences of the tag during year-month 1 | | | | |
| result | 3 countMonth2 32-bit Integer R Occurrences of the tag during year-month 2 | | | | |
| | diff 32-bit Integer R Difference between occurrences of this Tag in month 1 and month 2 | | | | |
| sort | 1 diff ↓ 2 tag.name ↑ | | | | |
| limit | 100 | | | | |
| CPs | 2.4, 3.1, 3.2, 4.1, 4.3, 5.3, 6.1 | | | | |



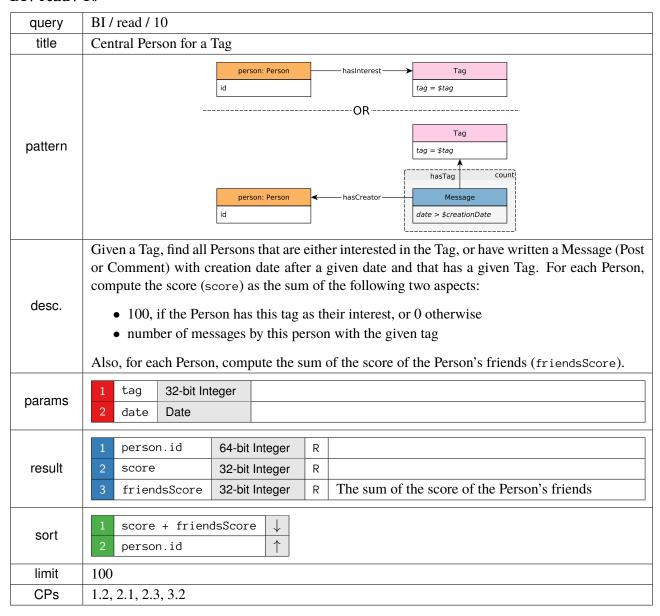








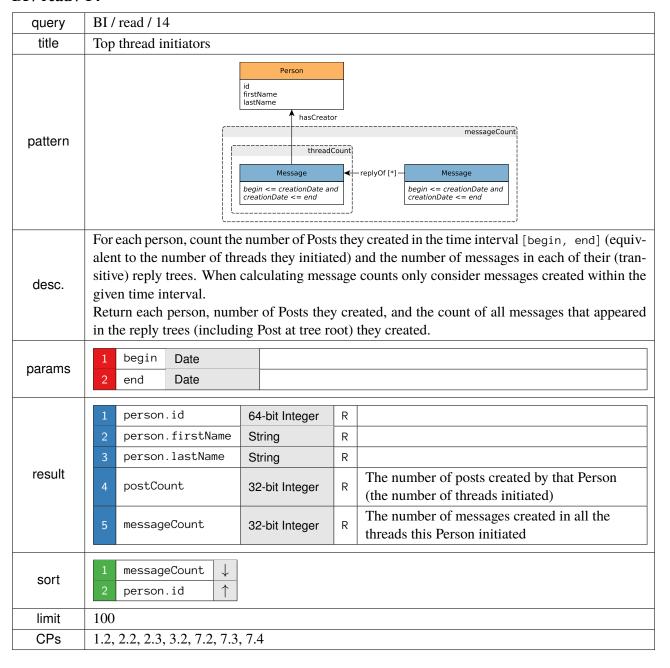
| query | BI / read / 9 | | | | | |
|---------|--|--|--|--|--|--|
| title | Forum with related Tags | | | | | |
| pattern | Forum with related Tags TagClass id = \$tagClass1 hasType Tag hasTag count1 = count Post Post Post Forum hasMember members = count \$threshold < members | | | | | |
| desc. | Given two TagClasses (tagClass1 and tagClass2), find Forums that contain at least one Post with a Tag from tagClass1 and at least one Post with a Tag from tagClass2 (direct children not transitive) – this may be the same Post. Consider the Forums with a number of members greater than a given threshold. For every such forum, count the number of Posts that have a Tag from TagClass1 (count1), and the number of posts that have a tag from TagClass2 (count2). | | | | | |
| params | 1 tagClass1 32-bit Integer 2 tagClass2 32-bit Integer 3 threshold 32-bit Integer | | | | | |
| result | 1 forum.id 64-bit Integer R 2 count1 32-bit Integer R Number of Posts with at least one tag belonging to tagClass1 3 count2 32-bit Integer R Number of Posts with at least one tag belonging to tagClass2 | | | | | |
| sort | 1 count2 - count1 ↓ 2 forum.id ↑ | | | | | |
| limit | 100 | | | | | |
| CPs | 1.2, 1.4, 2.1, 2.3, 2.4 | | | | | |

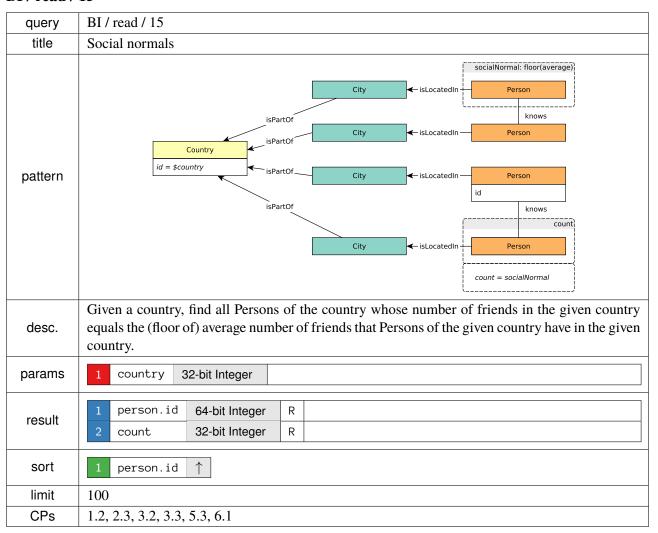


| query | BI / read / 11 | | | |
|--------------|---|--|--|--|
| title | Unrelated replies | | | |
| pattern | Country country = \$country isPartOf City Tag hasTag hasTag person HasCreator Message replyTo Comment content does not contain blacklisted words | | | |
| desc. | Find those Persons of a given country that replied to any Message, such that the reply does not have any Tag in common with the Message (only direct replies are considered, transitive ones are not). Consider only those replies not containing any word from a given blacklist. For each Person and valid reply, retrieve the Tags associated with the reply, and retrieve the number of likes on the reply. | | | |
| params | 1 country 32-bit Integer 2 blacklist String[] | | | |
| result | 1 person.id 64-bit Integer R 2 tag.name String R 3 countLikes 32-bit Integer R The count of Likes to replies with that Tag. 4 countReplies 32-bit Integer R The count of replies with that Tag. | | | |
| sort | 1 countLikes ↓ 2 person.id ↑ 3 tag.name ↑ | | | |
| limit CPs | 100 1.1, 2.1, 2.2, 2.3, 3.1, 3.2, 6.1 | | | |

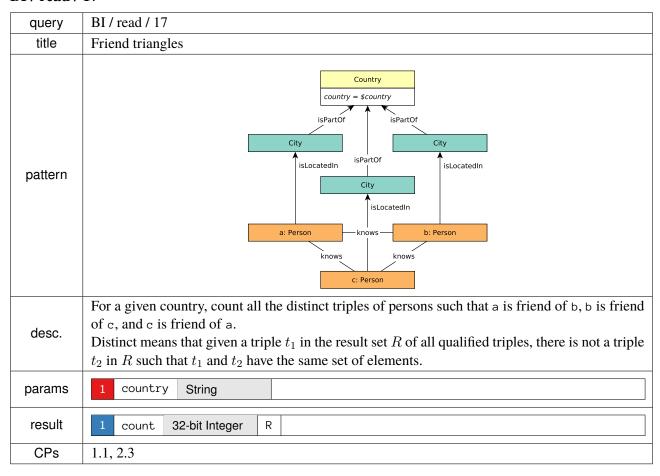
| query | BI / read / 12 | | | | |
|---------|--|--|--|--|--|
| title | Trending Posts | | | | |
| pattern | message: Message creationDate > \$creationDate id likes counti | | | | |
| desc. | Find all Messages created after a given date, that received more than a given number of likes. | | | | |
| params | 1 creationDate Date 2 likeThreshold 32-bit Integer | | | | |
| result | 1 message.id 64-bit Integer R 2 message.creationDate DateTime R 3 creator.firstName String R The first name of the post's creator 4 creator.lastName String R The last name of the post's creator 5 likeCount 32-bit Integer R The number of Likes the Post received | | | | |
| sort | 1 likeCount ↓ 2 message.id ↑ | | | | |
| limit | 100 | | | | |
| CPs | 1.2, 2.2, 3.1, 6.1 | | | | |

| query | BI / read / 13 | | | |
|---------|--|--|--|--|
| title | Popular Tags per month in a country | | | |
| pattern | Country name = \$country isLocatedIn Message year(creationDate) month(creationDate) hasTag tag: Tag | | | |
| desc. | Find all Messages in a given Country, as well as their Tags. Group messages by creation year and month. For each group, find the 5 most popular Tags, where popularity is the number of Messages (from within the same group) where the Tag appears. | | | |
| params | 1 country String | | | |
| | 1 year 32-bit Integer R year(message.creationDate) | | | |
| | 2 month 32-bit Integer R month(message.creationDate) | | | |
| result | popularTags TagPairs R (tag.name - String, popularity - 32-bit Integer), sorted descending by popularity, then ascending by tag name | | | |
| sort | 1 year ↓ 2 month ↑ | | | |
| limit | 100 | | | |
| CPs | 1.2, 2.2, 2.3, 3.2, 6.1 | | | |

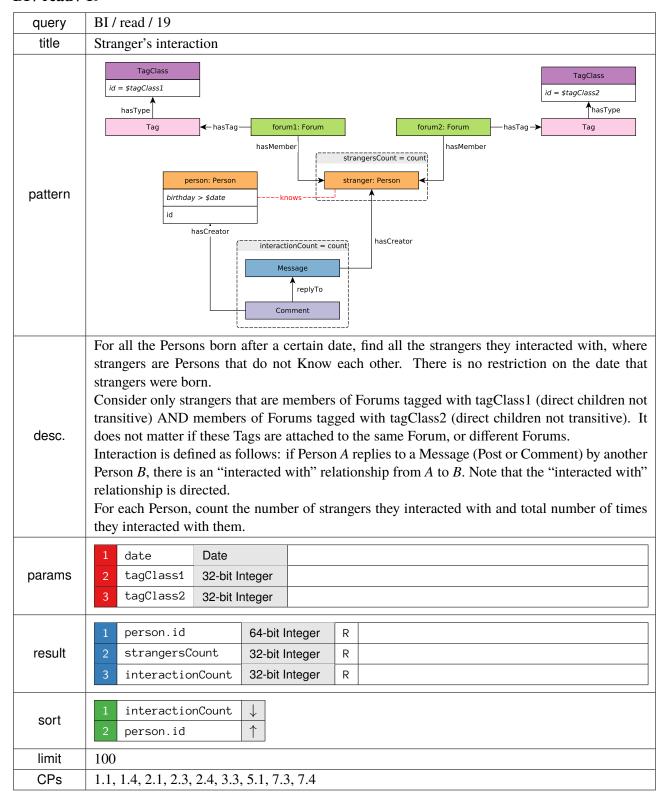




| query | BI / read / 16 | | | |
|---------|--|--|--|--|
| title | Experts in social circle | | | |
| pattern | Country name = \$country isPartOf City isLocatedIn person: Person id = \$personId Aname = \$tagClass name = \$tagClass Tag hasTag Message hasTag tag: Tag name | | | |
| desc. | Given a Person, find all other Persons that live in a given country and are connected to given person by a transitive path with length in range [min, max] through the knows relation. In the path, an edge can be only traversed once while nodes can be traversed multiple times. For each of these Persons, retrieve all of their Messages (Posts & Comments) that contain at least one Tag belonging to a given TagClass (direct relation not transitive). For each Message, retrieve all of its Tags. Group the results by Persons and Tags, then count the Messages by a certain Person having a certain Tag. | | | |
| params | 1 personId 64-bit Integer 2 country String 3 tagClass String 4 minPathDistance 32-bit Integer 5 maxPathDistance 32-bit Integer | | | |
| result | 1 person.id 64-bit Integer R 2 tag.name String R 3 messageCount 32-bit Integer R Number of Messages created by that Person containing that Tag | | | |
| sort | <pre>1 messageCount ↓ 2 tag.name ↑ 3 person.id ↑</pre> | | | |
| limit | 100 | | | |
| CPs | 1.2, 1.4, 2.3, 2.4, 3.3, 7.1, 7.2, 7.3 | | | |



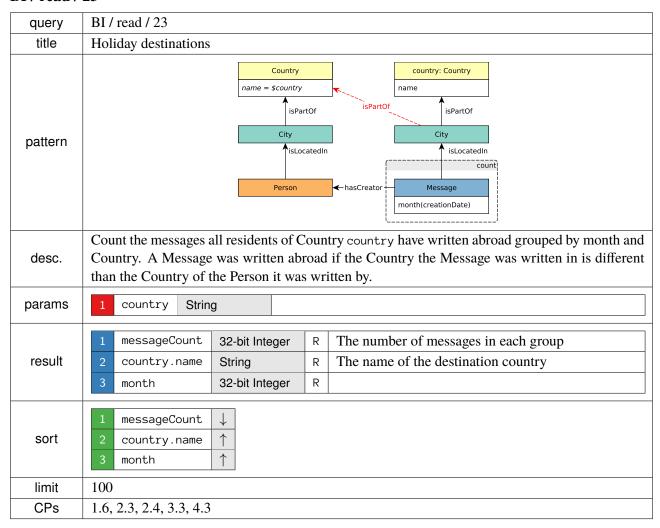
| query | BI / read / 18 | | | |
|---------|--|------------------------------------|--------|--|
| title | How many persons have a given number of posts | | | |
| pattern | 2. personCount = count Nessage Person AsCreator Message | | | |
| desc. | For each Person, count the number of Messages (Posts & Comments) they made. Only consider messages with: • length below the lengthThreshold • creationDate after date (exclusive, equality is not allowed) • any of the given languages The language of a Comment is that of the Post that initiates the thread where Comment belongs to. | | | |
| params | 2 lengthThreshold | Date 32-bit Integer String[] | | |
| result | | | R R | Number of messages created The number of Persons with messageCount messages |
| sort | 1 personCount \ \ | | | |
| CPs | 1.1, 1.2, 1.6, 3.2, 4.2, 4.3 | | | |

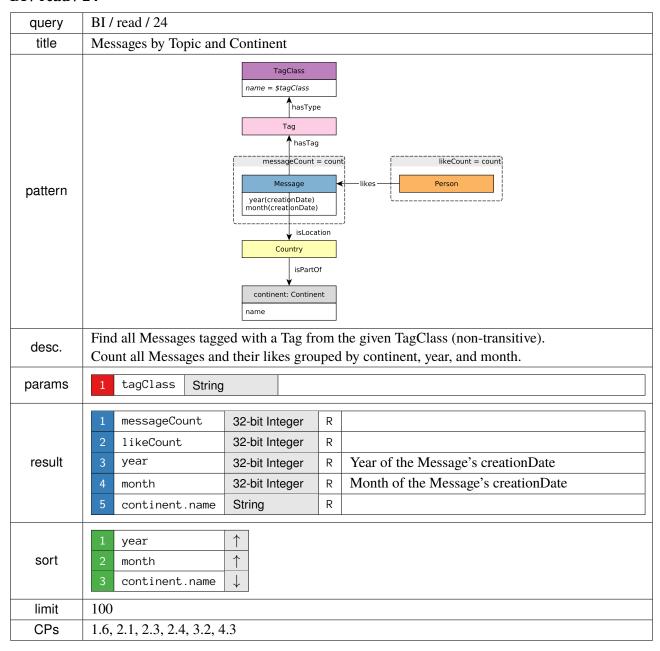


| query | BI / read / 20 | | | |
|---------|--|--|--|--|
| title | High-level topics | | | |
| pattern | tagClass: TagClass name = \$tagClassName name hasType Tag isSubclassOf [0.*] Tag hasTag Message | | | |
| desc. | For all given TagClasses, count number of Messages that have a Tag that belongs to that TagClass or any of its children (all descendants through a transitive relation). | | | |
| params | 1 tagClasses String[] | | | |
| result | 1 tagClass.name String R The TagClass of the root 2 postCount 32-bit Integer R | | | |
| sort | <pre>1 postCount ↓ 2 tagClass.name ↑</pre> | | | |
| limit | 100 | | | |
| CPs | 1.6, 2.1, 6.1 | | | |

| query | BI / read / 21 | | | | | |
|---------|---|--|--|--|--|--|
| title | Zombies in a country | | | | | |
| pattern | Country country = \$country isPartOf City isLocatedIn person: Person creationDate < \$endDate hasCreator realLikeCount = count hasCreator creationDate < \$endDate | | | | | |
| desc. | Find zombies within the given country, and return their zombie scores. A zombie is a Person created before the given endDate and that has created between [0, 1) Messages per month, on average, during the time range between profile creation date and the given endDate. The number of months spans the time range from creation date of the profile to the endDate and also includes partial months. For each person, calculate the following: • zombieLikeCount: the number of likes received from zombies • totalLikeCount: the total number of likes received on that Person's messages • zombieScore: zombieLikeCount / totalLikeCount | | | | | |
| | In all cases, only count likes received from profiles that were created before the given endDate. | | | | | |
| params | 1 country String 2 endDate Date | | | | | |
| result | 1 person.id 64-bit Integer R 2 zombieLikeCount 32-bit Integer R 3 totalLikeCount 32-bit Integer R 4 zombieScore 32-bit Float R The ratio of zombieLikeCount and totalLikeCount | | | | | |
| sort | <pre>1 zombieScore ↓ 2 person.id ↑</pre> | | | | | |
| limit | 100 | | | | | |
| CPs | 1.2, 2.1, 2.3, 2.4, 3.2, 3.3, 5.1, 5.3 | | | | | |

| query | BI / read / 22 | | | | |
|---------|--|--|--|--|--|
| title | International dialog | | | | |
| pattern | Country country = \$countryX isPartOf cityX: City name isLocatedIn p1: Person id | | | | |
| desc. | Consider all pairs of people (p1, p2) such that one is located in a city of Country countryX and the other is located in a city of Country countryY. For each city of Country countryX, return the highest scoring pair. The score of a pair is defined as the sum of the scores of the following kinds of interaction: • p1 has created a reply Comment to at least one Comment or Post by p2: Score = 4 • p1 has created at least one Post or Comment that p2 has created a reply Comment to: Score = 1 • p1 and p2 Know each other: Score = 15 • p1 liked at least one Post or Comment by p2: Score = 10 • p1 has created at least one Post or Comment that was liked by p2: Score = 1 Consequently, the maximum score a pair can obtain is: 4 + 1 + 15 + 10 + 1 = 31 | | | | |
| params | 1 countryX String 2 countryY String | | | | |
| result | 1 p1.id 64-bit Integer R 2 p2.id 64-bit Integer R 3 cityX.name String R 4 score 32-bit Integer R | | | | |
| sort | 1 score ↓ 2 p1.id ↑ 3 p2.id ↑ | | | | |
| CPs | 1.4, 1.6, 2.1, 3.1, 3.3, 5.1, 5.2, 5.3 | | | | |





| query | BI / read / 25 |
|---------|--|
| title | Weighted paths |
| | person1: Person id = \$person2Id Person knows [*] — person2: Person id = \$person2Id |
| pattern | Post Comment replyOf Comment hasContainer Forum begin <= creationDate & creationDate & creationDate <= end |
| desc. | Given two Persons, find all (unweighted) shortest paths between these two Persons, in the subgraph induced by the Knows relationship. Then, for each path calculate a weight. The nodes in the path are Persons, and the weight of a path is the sum of weights between every pair of consecutive Person nodes in the path. The weight for a pair of Persons is calculated such that every reply (by one of the Persons) to a Post (by the other Person) contributes 1.0, and every reply (by ones of the Persons) to a Comment (by the other Person) contributes 0.5. Only consider messages that were created in a forum that was created within the timeframe [startDate, endDate]. Return all the paths with shortest length, and their weights. |
| params | 1 person1Id 64-bit Integer 2 person2Id 64-bit Integer 3 startDate Date 4 endDate Date |
| result | person.id 64-bit Integer[] R Identifiers representing an ordered sequence of the Persons in the path weight |
| sort | 1 weight \ |
| CPs | 1.2, 2.1, 2.2, 2.4, 3.3, 5.1, 5.3, 7.2, 7.3 |

5.2 Update Query Descriptions

TODO

6 Auditing Rules

This chapter describes the rules to audit benchmark runs, that is, what techniques are allowed and what are not, what must be provided to the auditor and guidelines for the auditors to perform the audit.

6.1 Preparation

The first step when doing an audit is to determine the versions of the following items that will be used for the benchmark:

- The benchmark specification
- The data generator
- The driver

These must be reported in the full disclosure report to guarantee that the benchmark run can be reproduced exactly in the future. Similarly, the test sponsor will inform the auditor the scale factor to test. Finally, a clean test system with enough space to store the scale factor must be provided, including the update streams and substitution parameters.

6.1.1 Collect System Details

The next step is to collect the technical and pricing details of the system under test. This includes the following items:

- Common name of the system, e.g. Dell PowerEdge xxxx.
- Type and number of CPUs, cores/threads per CPU, clock frequency and cache hierarchy characteristics (levels, size per level, etc.).
- The amount of the system's memory, type and frequency.
- The disk controller or motherboard type if disk controller is on the motherboard.
- For each distinct type of secondary storage device, the number and characteristics of the device.
- The number and type of network controllers.
- The number and type of network switches. Wiring must be disclosed.
- Date of availability of the system.

Only the network switches and interfaces that participate in the run need to be reported. If the benchmark execution is entirely contained on a single machine, no network need be reported. The price of the hardware in question must be disclosed and should reflect the single quantity list price that any buyer could expect when purchasing one system with the given specification. The price may be either an item by item price or a package price if the system is sold as a package

Besides hardware characteristics, also software details must be collected:

- The DBMS and operating system name and versions.
- Installation and configuration information of both the DBMS and operating system, which must be provided by the test sponsor.
- Price of the software license used, which can be tied to the number of concurrent users or size of data.
- Date of availability of the software.

Also, the test sponsor must provide all the source code relevant to the benchmark.

6.1.2 Setup the Benchmark Environment

Once all the information has been collected, the auditor will setup the environment to perform the benchmark run. This setup includes configuring the following items:

- Setup the LDBC Data generator in the test machine if datasets are not available from a trusted source.
- Setup the LDBC driver with the connectors provided by the test sponsor. The test sponsor must provide the configuration parameters to configure the driver (tcr, number of threads, etc.).

 The ldbc sph interactive update interleave driver parameter must come from the

The ldbc.snb.interactive.update_interleave driver parameter must come from the updateStream.properties file, which is created by the data generator. That parameter should never be set manually. Also, make sure that the -rl/--results_log is enabled. Make sure that all operations are enabled and the frequencies are those for the selected scale factor. These can found in Section B.1. If the driver will be executed on a separate machine, gather the characteristics of that machine in the same way as specified above.

6.1.3 Load Data

The test sponsor must provide all the necessary documentation and scripts to load the dataset into the database to test. The system under test must support the different data types needed by the benchmark for each of the attributes at their specified precision. No data can be filtered out, everything must be loaded. The test sponsor must provide a tool to perform arbitrary checks of the data or a shell to issue queries in a declarative language if the system supports it. The auditor will measure the time to load the data, which will be disclosed.

6.2 Running the Benchmark

Running the benchmark consists of three separate parts: (1) validating the query implementations, (2) warming the database and (3) performing the benchmark run. The queries are validated by means of the official validation datasets provided by LDBC consortium in their official software repositories. The auditor must load the provided dataset and run the driver in validation mode, which will test that the queries provide the official results.

The warmup can be performed either using the LDBC driver or externally, and the way it is performed must be disclosed.

A valid benchmark run must last at least 2 hours of simulation time (datagen time). Also, in order to be valid, a benchmark run needs to meet the following requirements. The results_log.csv file contains the actual_start_time and the scheduled_start_time of each of the issued queries. In order to have a valid run, 95% of the queries must meet the following condition:

```
{\sf actual\_start\_time} - {\sf scheduled\_start\_time} < 1\ {\sf second}
```

If the execution of the benchmark is valid, the auditor must retrieve all the files from directory specified by -rd/--results_dir which includes configuration settings used, results log, results summary, which will be disclosed.

6.3 Recovery

Once an official run has been validated, the recovery capabilities of the system must be tested. The system and the driver must be configured in the same way as in during the benchmark execution. After a warmup period, an execution of the benchmark will be performed under the same terms as in the previous measured run.

At an arbitrary point close to 2 hours of simulation execution time, the machine will be disconnected. Then, the auditor will restart the database system and will check that the last committed update (in the driver log file) is actually in the database. The auditor will measure the time taken by the system to recover from the failure. Also, all the information about how durability is ensured must be disclosed. If checkpoints are used, these must be performed with a period of 10 minutes at most.

6.4 Serializability

Optionally, the test sponsor can execute update queries atomically. The auditor will verify that serializability is guaranteed.

7 Related Work

7.1 Graph Database Benchmarks

Train Benchmark: model validation [12].
Graph database comparison [8].
An "interactive social network benchmark": [3].

7.2 Scalable Graph Generators

The DATAGEN component is a fork of the generator described in [11].

7.3 LDBC Publications

LDBC publications are listed at http://ldbcouncil.org/publications.

References References

REFERENCES

[1] R. Angles, P. A. Boncz, J. Larriba-Pey, I. Fundulaki, T. Neumann, O. Erling, P. Neubauer, N. Martínez-Bazan, V. Kotsev, and I. Toma. The Linked Data Benchmark Council: a graph and RDF industry benchmarking effort. *SIGMOD Record*, 43(1):27–31, 2014.

- [2] A. Averbuch and A. Prat-Pérez. Benchmark design for navigational pattern matching benchmarking. Technical report, Linked Data Benchmark Council, 2014. http://ldbcouncil.org/sites/default/files/LDBC_D3.3.34.pdf.
- [3] S. Barahmand and S. Ghandeharizadeh. BG: A benchmark to evaluate interactive social networking actions. In CIDR 2013, Sixth Biennial Conference on Innovative Data Systems Research, Asilomar, CA, USA, January 6-9, 2013, Online Proceedings, 2013.
- [4] O. Erling, A. Averbuch, J. Larriba-Pey, H. Chafi, A. Gubichev, A. Prat-Pérez, M. Pham, and P. A. Boncz. The LDBC Social Network Benchmark: Interactive workload. In *SIGMOD*, pages 619–630, 2015.
- [5] G. Graefe. Query evaluation techniques for large databases. ACM Comput. Surv., 25(2):73–170, 1993.
- [6] A. Iosup, T. Hegeman, W. L. Ngai, S. Heldens, A. Prat-Pérez, T. Manhardt, H. Chafi, M. Capota, N. Sundaram, M. J. Anderson, I. G. Tanase, Y. Xia, L. Nai, and P. A. Boncz. LDBC graphalytics: A benchmark for large-scale graph analysis on parallel and distributed platforms. *PVLDB*, 9(13):1317–1328, 2016.
- [7] A. Iosup, T. Hegeman, W. L. Ngai, S. Heldens, A. Prat-Pérez, T. Manhardt, H. Chafi, M. Capota, N. Sundaram, M. J. Anderson, I. G. Tanase, Y. Xia, L. Nai, and P. A. Boncz. LDBC Graphalytics: A benchmark for large-scale graph analysis on parallel and distributed platforms. *PVLDB*, 9(13):1317–1328, 2016.
- [8] M. Lissandrini, M. Brugnara, and Y. Velegrakis. An evaluation methodology and experimental comparison of graph databases. Technical report, University of Trento, 2017. https://disi.unitn.it/~lissandrini/pdf/lissandrini-techreport-gdb.pdf.
- [9] G. Moerkotte. Small materialized aggregates: A light weight index structure for data warehousing. In *PVLDB*, pages 476–487, 1998.
- [10] T. Neumann and G. Moerkotte. A framework for reasoning about share equivalence and its integration into a plan generator. In *BTW*, pages 7–26, 2009.
- [11] M. Pham, P. A. Boncz, and O. Erling. S3G2: A scalable structure-correlated social graph generator. In Selected Topics in Performance Evaluation and Benchmarking 4th TPC Technology Conference, TPCTC 2012, Istanbul, Turkey, August 27, 2012, Revised Selected Papers, pages 156–172, 2012.
- [12] G. Szárnyas, B. Izsó, I. Ráth, and D. Varró. The Train Benchmark: Cross-technology performance evaluation of continuous model validation. *Softw Syst Model*, 2017.

A CHOKE POINTS

Introduction

Choke points are a superset of [2] with the exception of CP 7.1, which was removed and replaced with a new choke point.

Choke points X queries are displayed in Table A.1.

A.1 Aggregation Performance

CP-1.1: [QOPT] Interesting orders

TPC-H 1.2

This choke-point tests the ability of the query optimizer to exploit the interesting orders induced by some operators. Apart from clustered indexes providing key order, other operators also preserve or even induce tuple orderings. Sort-based operators create new orderings, typically the probe-side of a hash join conserves its order, etc.

Queries. BI 2 BI 4 BI 11 BI 17 BI 18 BI 19 Interactive 2 Interactive 9

CP-1.2: [QEXE] High Cardinality group-by performance

TPC-H 1.1

This choke-point tests the ability of the execution engine to parallelize group-by's with a large number of groups. Some queries require performing large group-by's. In such a case, if an aggregation produces a significant number of groups, intra query parallelization can be exploited as each thread may make its own partial aggregation. Then, to produce the result, these have to be re-aggregated. In order to avoid this, the tuples entering the aggregation operator may be partitioned by a hash of the grouping key and be sent to the appropriate partition. Each partition would have its own thread so that only that thread would write the aggregation, hence avoiding costly critical sections as well. A high cardinality distinct modifier in a query is a special case of this choke point. It is amenable to the same solution with intra query parallelization and partitioning as the group-by. We further note that scale-out systems have an extra incentive for partitioning since this will distribute the CPU and memory pressure over multiple machines, yielding better platform utilization and scalability.

Queries. BI 1 BI 2 BI 4 BI 5 BI 6 BI 7 BI 9 BI 10 BI 12 BI 13 BI 14 BI 15 BI 16 BI 18 BI 21 BI 25 Interactive 9

CP-1.3: [QEXE] Complex aggregate performance

This choke-point test the performance of the execution engine to perform complex aggregates. Many databases offer user defined aggregates and more complex aggregation operations than the basic count, sum, max and min, for example string concatenation aggregation operator. These types of aggregates are expected to benefit from the same optimizations as the basic built-in ones, for example partitioning.

Queries. Interactive 1

CP-1.4: [QOPT] Top-k push down

Top-k push down. This choke-point tests the ability of the query optimizer to perform optimizations based on top-k selections. Many times queries demand for returning the top-k elements. Once k results are obtained, extra restrictions in a selection can be added based on the properties of the kth element currently in the top-k, being more restrictive as the query advances, instead of sorting all elements and picking the highest k.

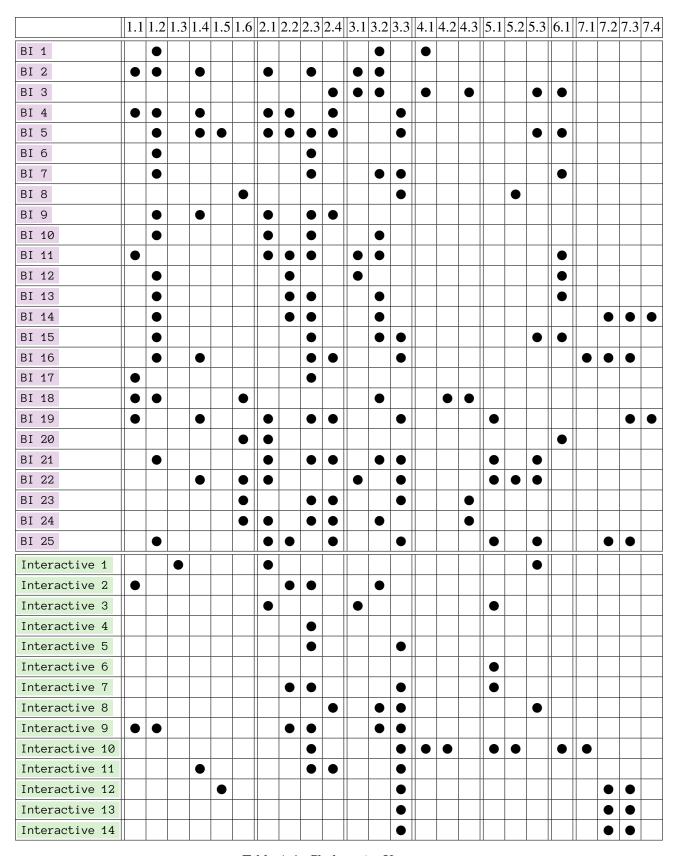


Table A.1: Choke point X query

Queries. BI 2 BI 4 BI 5 BI 9 BI 16 BI 19 BI 22 Interactive 11

CP-1.5: [QEXE] Dependent group-by keys

TPC-H 1.4

This choke-point tests the ability of the query optimizer to exclude those functionally dependent group-bys. Sometimes queries require for group-by's on a set of columns and a key, where the value of the key determines the columns. In this situation, the aggregation operator should be able to exclude certain group-by attributes from the key matching.

Queries. BI 5 Interactive 12

CP-1.6: [QEXE] Low Cardinality group-by performance

TPC-H 1.3

This choke-point tests the ability to efficiently perform group by evaluation when only a very limited set of groups is available. This can require special strategies for parallelization, e.g. pre-aggregation when possible. This case also allows using special strategies for grouping like using array lookup if the domain of keys is small.

Queries. BI 8 BI 18 BI 20 BI 22 BI 23 BI 24

A.2 Join Performance

CP-2.1: [QOPT] Rich join order optimization

TPC-H 2.3

This choke-point tests the ability of the query optimizer to find optimal join orders. A graph can be traversed in different ways. In the relational model, this is equivalent as different join orders. The execution time of these orders may differ by orders of magnitude. Therefore, finding an efficient join (traversal) order is important, which in general, requires enumeration of all the possibilities. The enumeration is complicated by operators that are not freely re-orderable like semi-, anti-, and outer-joins. Because of this difficulty most join enumeration algorithms do not enumerate all possible plans, and therefore can miss the optimal join order. Therefore, these chokepoint tests the ability of the query optimizer to find optimal join (traversal) orders.

Queries. BI 2 BI 4 BI 5 BI 9 BI 10 BI 11 BI 19 BI 20 BI 21 BI 22 BI 24 BI 25 Interactive 1 Interactive 3

CP-2.2: [QOPT] Late projection

TPC-H 2.4

This choke-point tests the ability of the query optimizer to delay the projection of unneeded attributes until late in the execution. Queries where certain columns are only needed late in the query. In such a situation, it is better to omit them from initial table scans, as fetching them later by row-id with a separate scan operator, which is joined to the intermediate query result, can save temporal space, and therefore I/O. Late projection does have a trade-off involving locality, since late in the plan the tuples may be in a different order, and scattered I/O in terms of tuples/second is much more expensive than sequential I/O. Late projection specifically makes sense in queries where the late use of these columns happens at a moment where the amount of tuples involved has been considerably reduced; for example after an aggregation with only few unique group-by keys, or a top-k operator.

Queries. BI 4 BI 5 BI 11 BI 12 BI 13 BI 14 BI 25 Interactive 2 Interactive 7 Interactive 9

CP-2.3: [QOPT] Join type selection

This choke-point tests the ability of the query optimizer to select the proper join operator type, which implies accurate estimates of cardinalities. Depending on the cardinalities of both sides of a join, a hash or an index index based join operator is more appropriate. This is especially important with column stores, where one usually has an index on everything. Deciding to use a hash join requires a good estimation of cardinalities on both the probe and build sides. In TPC-H, the use of hash join is almost a foregone conclusion in many cases, since an implementation will usually not even define an index on foreign key columns. There is a break even point between index and hash based plans, depending on the cardinality on the probe and build sides

Queries. BI 2 BI 5 BI 6 BI 7 BI 9 BI 10 BI 11 BI 13 BI 14 BI 15 BI 16 BI 17 BI 19 BI 21 BI 23 BI 24 Interactive 2 Interactive 4 Interactive 5 Interactive 7 Interactive 9 Interactive 10 Interactive 11

CP-2.4: [QOPT] Sparse foreign key joins

TPC-H 2.2

This choke-point tests the performance of join operators when the join is sparse. Sometimes joins involve relations where only a small percentage of rows in one of the tables is required to satisfy a join. When tables are larger, typical join methods can be sub-optimal. Partitioning the sparse table, using Hash Clustered indexes or implementing Bloom filter tests inside the join are techniques to improve the performance in such situations [5].

Queries. BI 3 BI 4 BI 5 BI 9 BI 16 BI 19 BI 21 BI 23 BI 24 BI 25 Interactive 8 Interactive 11

CP-2.5: [QOPT] Index choice - Decide scan order taking into account order of results for down-stream (parent) operators

Add description

this is not covered by the queries - at least according to the docs

CP-2.6: [QEXE] Unpredictable, widely scattered indexed access pattern (graph walk)

The efficiency of index lookup is very different depending on the locality of keys coming to the indexed access. Techniques like vectoring non-local index access by simply missing the cache in parallel on multiple lookups vectored on the same thread may have high impact. Also detecting absence of locality should turn off any locality dependent optimizations if these have overhead when there is no locality. A graph neighborhood traversal is an example of an operation with random access without predictable locality.

this is not convered by the queries - at least according to the docs

CP-2.7: [QOPT] Join type - correct choice of hash vs. index based on cardinality on either side

Specially with stores, where one usually has an index on everything, deciding to use a hash join requires a good estimation of cardinalities on both the probe and build sides. In TPC-H, the use of hash join is almost a foregone conclusion in many cases, since an implementation will usually not even define an index on foreign key columns. There is a break even point between index and hash based plans, depending on the cardinality on the probe and build sides.sides. This choke point tests whether this break-even point is correctly modeled and whether the cost model produces accurate estimates as input to this choice.

this is not convered by the queries - at least according to the docs

A.3 Data Access Locality

CP-3.1: [QOPT] Detecting correlation

TPC-H 3.3

This choke-point tests the ability of the query optimizer to detect data correlations and exploiting them. If a schema rewards creating clustered indexes, the question then is which of the date or data columns to use as key. In fact it should not matter which column is used, as range- propagation between correlated attributes of the same table is relatively easy. One way is through the creation of multi-attribute histograms after detection of attribute correlation. With MinMax indexes, range-predicates on any column can be translated into qualifying tuple position ranges. If an attribute value is correlated with tuple position, this reduces the area to scan roughly equally to predicate selectivity.

Queries. BI 2 BI 3 BI 11 BI 12 BI 22 Interactive 3

CP-3.2: [STORAGE] Dimensional clustering

This chokepoint tests suitability of the identifiers assigned to entities by the storage system to better exploit data locality. A data model where each entity has a unique synthetic identifier, e.g. RDF or graph models, has some choice in assigning a value to this identifier. The properties of the entity being identified may affect this, e.g. type (label), other dependent properties, e.g. geographic location, date, position in a hierarchy etc, depending on the application. Such identifier choice may create locality which in turn improves efficiency of compression or index access.

Queries. BI 1 BI 2 BI 3 BI 7 BI 10 BI 11 BI 13 BI 14 BI 15 BI 18 BI 21 BI 24 Interactive 2 Interactive 8 Interactive 9

CP-3.3: [QEXE] Scattered index access patterns

This choke-point tests the performance of indexes when scattered accesses are performed. The efficiency of index lookup is very different depending on the locality of keys coming to the indexed access. Techniques like vectoring non-local index accesses by simply missing the cache in parallel on multiple lookups vectored on the same thread may have high impact. Also detecting absence of locality should turn off any locality dependent optimizations if these are costly when there is no locality. A graph neighborhood traversal is an example of an operation with random access without predictable locality.

this is not convered by the queries - at least according to the docs

CP-3.4: [QEXE] Use of zone maps for accelerating scans

Add description

this is not convered by the queries - at least according to the docs

CP-3.5: [STORAGE] Dimensional clustering - assigning pk/fk/uri/vertex ids in function of attributes of the entity concerned

A data model where each entity has a unique synthetic identifier, e.g. RDF or graph models, has some choice in assigning a value to this identifier. The properties of the entity being identified may affect this, e.g. type (label), other dependent properties, e.g. geographic location, date, position in a hierarchy etc, depending on the application. Such identifier choice may create locality which in turn improves efficiency of compression or index access.

this is not convered by the queries - at least according to the docs

A.4 Expression Calculation

CP-4.1: [QOPT] Common subexpression elimination

TPC-H 4.2a

This choke-point tests the ability of the query optimizer to detect common sub-expressions and reuse their results. A basic technique helpful in multiple queries is common subexpression elimination (CSE). CSE should recognize also that average aggregates can be derived afterwards by dividing a SUM by the COUNT when those have been computed.

Queries. BI 1 BI 3 Interactive 10

CP-4.2: [QOPT] Complex boolean expression joins and selections

TPC-H 4.2d

This choke-point tests the ability of the query optimizer to reorder the execution of boolean expressions to improve the performance. Some boolean expressions are complex, with possibilities for alternative optimal evaluation orders. For instance, the optimizer may reorder conjunctions to test first those conditions with larger selectivity [9].

Queries. BI 18 Interactive 10

CP-4.3: [QEXE] Low overhead expressions interpretation

This choke-point tests the ability to efficiently evaluate simple expressions on a large number of values. A typical example could be simple arithmetic expressions, mathematical functions like floor and absolute or date functions like extracting a year.

Queries. BI 3 BI 18 BI 23 BI 24

CP-4.4: [QEXE] String matching performance

this is not covered by the queries - at least according to the docs

A.5 Correlated Sub-queries

CP-5.1: [QOPT] Flattening sub-queries

TPC-H 5.1

This choke-point tests the ability of the query optimizer to flatten execution plans when there are correlated sub-queries. Many queries have correlated sub-queries and their query plans can be flattened, such that the correlated sub-query is handled using an equi-join, outer-join or anti-join. To execute queries well, systems need to flatten both sub-queries, the first into an equi-join plan, the second into an anti-join plan. Therefore, the execution layer of the database system will benefit from implementing these extended join variants. The ill effects of repetitive tuple-at-a-time sub-query execution can also be mitigated if execution systems by using vectorized, or block-wise query execution, allowing to run sub-queries with thousands of input parameters instead of one. The ability to look up many keys in an index in one API call creates the opportunity to benefit from physical locality, if lookup keys exhibit some clustering.

Queries. BI 19 BI 21 BI 22 BI 25 Interactive 3 Interactive 6 Interactive 7 Interactive 10

CP-5.2: [QEXE] Overlap between outer and sub-query

TPC-H 5.3

This choke-point tests the ability of the execution engine to reuse results when there is an overlap between the outer query and the sub-query. In some queries, the correlated sub-query and the outer query have the same joins and selections. In this case, a non-tree, rather DAG-shaped [10] query plan would allow to execute the common parts just once, providing the intermediate result stream to both the outer query and correlated sub-query, which higher up in the query plan are joined together (using normal query decorrelation rewrites). As such, the benchmark rewards systems where the optimizer can detect this and the execution engine supports an operator that can buffer intermediate results and provide them to multiple parent operators.

Queries. BI 8 BI 22 Interactive 10

CP-5.3: [QEXE] Intra-query result reuse

TPC-H 5.2

This choke-point tests the ability of the execution engine to reuse sub-query results when two sub-queries are mostly identical. Some queries have almost identical sub-queries, where some of their internal results can be reused in both sides of the execution plan, thus avoiding to repeat computations.

Queries. BI 3 BI 5 BI 15 BI 21 BI 22 BI 25 Interactive 1 Interactive 8

A.6 Parallelism and Concurrency

CP-6.1: [QEXE] Inter-query result reuse

TPC-H 6.3

This choke-point tests the ability of the query execution engine to reuse results from different queries. Sometimes with a high number of streams a significant amount of identical queries emerge in the resulting workload. The reason is that certain parameters, as generated by the workload generator, have only a limited amount of parameters bindings. This weakness opens up the possibility of using a query result cache, to eliminate the repetitive part of the workload. A further opportunity that detects even more overlap is the work on recycling, which does not only cache final query results, but also intermediate query results of a "high worth". Here, worth is a combination of partial-query result size, partial-query evaluation cost, and observed (or estimated) frequency of the partial-query in the workload.

Queries. BI 3 BI 5 BI 7 BI 11 BI 12 BI 13 BI 15 BI 20 Interactive 10

A.7 RDF and Graph Specifics

CP-7.1: [QEXE] Path pattern reuse

This choke-point tests the ability of the execution engine to reuse work across graph traversals

complete in more detail

For example, when computing paths within a range of distances, it is often possible to incrementally compute longer paths by reusing paths of shorter distances that were already computed

Queries. BI 16 Interactive 10

CP-7.2: [QOPT] Cardinality estimation of transitive paths

This choke-point tests the ability of the query optimizer to properly estimate the cardinality of intermediate results when executing transitive paths. A transitive path may occur in a "fact table" or a "dimension table" position. A transitive path may cover a tree or a graph, e.g. descendants in a geographical hierarchy vs. graph neighborhood or transitive closure in a many-to-many connected social network. In order to decide proper join order and type, the cardinality of the expansion of the transitive path needs to be correctly estimated. This could for example take the form of executing on a sample of the data in the cost model or of gathering special statistics, e.g. the depth and fan-out of a tree. In the case of hierarchical dimensions, e.g. geographic locations or other hierarchical classifications, detecting the cardinality of the transitive path will allow one to go to a star schema plan with scan of a fact table with a selective hash join. Such a plan will be on the other hand very bad for example if the hash table is much larger than the "fact table" being scanned.

Queries. BI 14 BI 16 BI 25 Interactive 12 Interactive 13 Interactive 14

CP-7.3: [QEXE] Execution of a transitive step

This choke-point tests the ability of the query execution engine to efficiently execute transitive steps. Graph workloads may have transitive operations, for example finding a shortest path between vertices. This involves repeated execution of a short lookup, often on many values at the same time, while usually having an end condition, e.g. the target vertex being reached or having reached the border of a search going in the opposite direction. For the best efficiency, these operations can be merged or tightly coupled to the index operations themselves. Also parallelization may be possible but may need to deal with a global state, e.g. set of visited vertices. There are many possible tradeoffs between generality and performance

Queries. BI 14 BI 16 BI 19 BI 25 Interactive 12 Interactive 13 Interactive 14

CP-7.4: [QEXE] Efficient evaluation of termination criteria for transitive queries

This tests the ability of a system to express termination criteria for transitive queries so that not the whole transitive relation has to be evaluated as well as efficient testing for termination.

Queries. BI 14 BI 19

B Scale Factor Statistics

B.1 Scale Factor Statistics

| Query Type | SF1 | SF3 | SF10 | SF30 | SF100 | SF300 | SF1000 |
|-------------------|-----|-----|------|------|-------|-------|--------|
| Query 1 | 26 | 26 | 26 | 26 | 26 | 26 | 26 |
| Query 2 | 37 | 37 | 37 | 37 | 37 | 37 | 37 |
| Query 3 | 69 | 79 | 92 | 106 | 123 | 142 | 165 |
| Query 4 | 36 | 36 | 36 | 36 | 36 | 36 | 36 |
| Query 5 | 57 | 61 | 66 | 72 | 78 | 84 | 91 |
| Query 6 | 129 | 172 | 236 | 316 | 434 | 580 | 796 |
| Query 7 | 87 | 72 | 54 | 48 | 38 | 32 | 25 |
| Query 8 | 45 | 27 | 15 | 9 | 5 | 3 | 1 |
| Query 9 | 157 | 209 | 287 | 384 | 527 | 705 | 967 |
| Query 10 | 30 | 32 | 35 | 37 | 40 | 44 | 47 |
| Query 11 | 16 | 17 | 19 | 20 | 22 | 24 | 26 |
| Query 12 | 44 | 44 | 44 | 44 | 44 | 44 | 44 |
| Query 13 | 19 | 19 | 19 | 19 | 19 | 19 | 19 |
| Query 14 | 49 | 49 | 49 | 49 | 49 | 49 | 49 |

Table B.1: Frequencies for each query and SF.