

Distributed Data Management Data Models and Query Languages Thorsten Papenbrock Felix Naumann

F-2.03/F-2.04, Campus II

Hasso Plattner Institut

Introduction Layering Data Models

1. Conceptual layer

- Data structures, objects, modules, ...
 - Application code

2. Logical layer

- Relational tables, JSON, XML, graphs, ...
 - Database management system (DBMS) or storage engine

3. Representation layer

- Bytes in memory, on disk, on network, ...
 - > Database management system (DBMS) or storage engine

4. Physical layer

- Electrical currents, pulses of light, magnetic fields, ...
 - Operating system and hardware drivers



Distributed Data Management

our focus now

Data Models and Query Languages

Overview Relational and Non-Relational Data Models





Overview Relational and Non-Relational DBMSs



"No SQL" or rather "not only SQL" systems because most support some SQL dialect.

A class of relational DBMSs that seek to provide the same scalable performance of NoSQL systems for OLTP workloads while still maintaining all ACID guarantees.



© 2012 by The 451 Group. All rights reserved

Overview
Data Models

A data model consists of three parts:

1. Structure

- physical and conceptual data layout
- 2. Constraints
- inherent limitations and rules
- 3. Operations
- possible query and modification methods

Distributed Data Management

Data Models and Query Languages

ThorstenPapenbrock Slide **5**

[Hector Garcia-Molina, Jeffrey D. Ullman, and Jennifer Widom. *Database Systems: The Complete Book*. Prentice Hall Press, Upper Saddle River, NJ, USA, 2 edition, 2008. ISBN 9780131873254]



Overview Relational and Non-Relational Data Models





The Relational Data Model Natural Relational Data



Transactional Data



Statistical Data

Governance	Strategic	Balanced Scorecar	d Busines	s planning Knov	ledge Risk and c	ompliance	Business continuit	y Partners
	Tactical	Legal Audi	t Monitoring	Quality assurance	Change Co	Pr Project M	anagement	
Solutions	Operations	People	Finance	Sales	Marketing	Projects	Computing	
Requirements Acquisition Development Implementation Maintenance	Value chain Operations plan Supply Production Distribution Stock control TOM IT Production	Policy People plan Org structure Safety and health Culture Development Incentive Recruitment Appraisal Reorganisation Termination	Financial plan Budget Tax AP AR Payroll Financial contro Investment Assets	Customer (CRM) Strategy Lead generation Lead tracking Campaign Forecasting Retail Export	Analysis Strategy Products Demand Channels	Portfolio Resources Planning Execution	Computing Plan Assessment Governance Architecture Security Release Provisioning IT Support Information	
Service Levels Service Desk / Inc Problem Availability CSIP User satisfaction Capacity								

Business Data

The Relational Data Model Popular relational DBMS



HPI Hasso Plattner Institut

https://db-engines.com/en/ranking

Distributed Data Management

Data Models and Query Languages

[E. F. Codd. A Relational Model of Data for Large Shared Data Banks. Communications of the ACM, 13(6):377–387, 1970]

The Relational Data Model **Definition**

- 1. Structure
- Schemata: named, non-empty, typed, and unordered sets of attributes
 - Example: Person(ID,Surname,Name,Gender,Address)
- Instances: sets of records, i.e., functions that assign values to attributes
 - Example: (275437,`Miller´,`Frank´,`male´,`Millstr. 5´)
- 2. Constraints
- Integrity constraints: data types, keys, foreign-keys, ...
- 3. Operations
- Relational algebra (and relational calculus)
- Usually implemented as Structured Query Language (SQL)

Distributed Data Management

Data Models and Query Languages



The Relational Data Model Querying: SQL



SELECT<attribute list>FROM<relation list>WHERE<conditions>GROUP BY<grouping attributes>HAVING<grouping conditions>ORDER BY<attribute list>;

Declarative query languages specify the result of a query and not how it should be obtained:

- Easier to understand
- Transparently optimizable
- > Implementation independent

Further keywords:

DISTINCT, AS, JOIN AND, OR MIN, MAX, AVG, SUM, COUNT NOT, IN, LIKE, ANY, ALL, EXISTS UNION, EXCEPT, INTERSECT

DDL\DML:

CREATE TABLE DROP TABLE ALTER TABLE INSERT INTO ... VALUES DELETE FROM ... WHERE UPDATE ... SET ... WHERE

Distributed Data Management

Data Models and Query Languages

The Relational Data Model Querying: SQL

Grundbausteine:

SELECT <Attributliste> FROM <Relationenliste> WHERE <Bedingungen>; (= Schema der Ergebnisrelation; * für alle Attribute) (Relationen aus denen die <u>Tupe</u>] stammen) (Bedingungen an die Daten; Verknüpfung über Schlüsselwort **AND**)

SELECT:

- Entspricht Projektion π der relationalen Algebra
- Umbenennung: SELECT Titel AS Filmtitel
- Arithmetik: SELECT Länge * 3.1415 AS Länge MalPi
- Konstanten: SELECT ,Herr' AS Titel
- Duplikateliminierungδ: SELECT DISTINCT Titel

FROM:

- Entspricht Kreuzprodukt × der relationalen Algebra (falls mehrere Relationen gewählt)

WHERE:

- Entspricht Selektion σ der relationalen Algebra
- WHERE-Teil der Anfrage ist optional
- Operatoren: =, <>, >, <, <=, >=, LIKE, NOT, ANY, ALL, EXISTS, IN,
- Kann Kreuzprodukt des FROM-Teils zum Join machen:
 - WHERE Person.ID = Mensch.ID
 (Natürlicher Join ⋈)

 WHERE Person.Name = Mensch.Vorname
 (Theta Join ⋈_n)

Komplexe Anfragen:

- SELECT <Attributliste>
- FROM < Relationenliste >
- WHERE < Bedingungen>
- GROUP BY <Gruppierungsattribute>
- HAVING <Bedingungen auf Gruppierungsattribute>
- ORDER BY <Attributliste>;

GROUP BY ... HAVING:

- Entspricht Gruppierungγder relationalen Algebra
- Aggregationsoperatoren für SELECT-Statement: AVG(<Attribut>)

COUNT(<Attribut>) SUM(<Attribut>)

- HAVING entspricht einer Selektion nach der Gruppierung

ORDER BY:

- Entspricht Sortierung τ der relationalen Algebra
- Sortiert das Ergebnis der Anfrage entsprechend der Attributliste
- Aufsteigend: ORDER BY Vorname, Nachname ASC
- Absteigend: ORDER BY Vorname, Nachname DESC

Datendefinition: Data Definition Language (DDL)

CREATE TABLE <Tabellenname>(<Attributliste mit Datentypen>);

- Aufgabe: neue Tabelle erstellen
- Datentypen mit Länge n bzw. m: CHAR(n), VARCHAR(n), BIT(n), DECIMAL(n,m)

CLOB, BLOB

- Datentypen mit impliziter Länge: INT, FLOAT
- Datentypen f
 ür Objekte:
- Datentypen für Zeiten: TIME, DATE, TIMESTAMP
- Bsp.: CREATE TABLE Schauspieler (
 - Name CHAR(30),
 - Adresse VARCHAR(255)
 - Geschlecht CHAR(1),
 - Geburtstag DATE);
- Nebenbedingungen:
 - Primärschlüssel: PRIMARY KEY
 - Eindeutigkeit: UNIQUE
 - Default-Werte: DEFAULT<<u>Defaultwert</u>>
 - Nicht-Null: NOT NULL
 - Fremdschlüssel: FOREIGN KEY (<Attributliste>) REFERENCES <Tabellenname>(<Attributliste>)
 - Weitere: CHECK, CREATE TRIGGER, CREATE ASSERTION, ...
- Bsp.: CREATE TABLE Schauspieler (
 - SchauspielerNummer INT PRIMARY KEY, Name CHAR(30) NOT NULL, Adresse VARCHAR(255) NOT NULL UNIQUE, Geschlecht CHAR(1), Geburtstag DATE DEFAULT DATE, 0000-00-00', FOREIGN KEY (Adresse) REFERENCES Haus(Adresse));

DROP TABLE <Tabellenname>;

- Aufgabe: bestehende Tabelle löschen
- Bsp.: DROP TABLE Schauspieler;

ALTER TABLE <Tabellenname> <Aktion>;

- Aufgabe: bestehende Tabelle ändern
- ADD: Attribut hinzufügen
 - Bsp.: ALTER TABLE Schauspieler ADD Telefon CHAR(16)
- DROP: Attribut löschen
 - Bsp.: ALTER TABLE Schauspieler DROP Geburtstag

CREATE INDEX <Indexname> ON <Tabellenname>{<Attributliste des neuen Index>};

- Aufgabe: Index erstellen

DROP INDEX <Indexname>;

Aufgabe: Index löschen

Datenbearbeitung: Data Modelling Language (DML)

INSERT INTO <Tabellenname>{<Attributliste>} VALUES (<Attributliste>);

- Aufgabe: Tupel einfügen
- Bsp.: INSERT INTO Studio(Name, Nummer) VALUES ('Pixa', 34);
- Ergebnis einer Anfrage für Einfügen nutzen:
 - Bsp.: INSERT INTO Studios(Name)
 - SELECT DISTINCT StudioName
 - FROM Film
 - WHERE StudioName NOTIN
 - (SELECT Name
 - FROM Studios);

DELETE FROM <Tabellenname>WHERE <Bedingung>;

- Aufgabe: <u>Tupel</u> löschen
- Bsp.: DELETE FROM Studio WHERE Name='Pixa';

UPDATE <Tabellenname> SET <Zuweisung> WHERE <Bedingung>;

- Aufgabe: Attributwerte ändern
- Bsp.: UPDATE Studio SET Name='Pixa' WHERE Name='Pi';

Bulk insert: IMPORT, LOAD, ... (→ DBMS spezifisch)

Mengenoperationen

(<Anfrage>) UNION (<Anfrage>) (Liefert Vereinigung "∪" der beiden Ergebnismengen)

HPI

Hasso Plattner

Institut

- (<Anfrage>) EXCEPT (<Anfrage>) (Liefert Differenz ,- "der beiden Ergebnismengen)
- (<Anfrage>) INTERSECT (<Anfrage>) (Liefert Schnittmenge "

 "der beiden Ergebnismengen)
- UNION, EXCEPT und INTERSECT nutzen Mengensemantik (→eliminieren Duplikate)
- UNION ALL, EXCEPT ALL und INTERSECT ALL nutzen Multimengensemantik (→erhalten Duplikate)

Join-Varianten

1.) Kreuzprodukt mit Bedingung:

CREATE VIEW <Sichtname > AS <Anfrage >:

- SELECT *
- FROM <Join-Relation1>, <Join-Relation2> WHERE <Join-Attribut1> = <Join-Attribut2>;

- Aufgabe: Erstelle eine Sicht für die gegebene SQL-Anfrage

2.) Schlüsselwort:

<Tabellenname> CROSS JOIN <Tabellenname> <Tabellenname> NATURAL IOIN <Tabellenname> <Tabellenname> NATURAL INNER JOIN <Tabellenname> <Tabellenname> NATURAL LEFT OUTER JOIN <Tabellenname> <Tabellenname> NATURAL RIGHT OUTER JOIN <Tabellenname> <Tabellenname> NATURAL FULL OUTER JOIN <Tabellenname>

Sichten

The Relational Data Model Querying: SQL – Examples



Schemata:

- Product(maker, model, type)
- PC(model, speed, ram, hd, rd)
- Laptop (<u>model</u>, speed, ram, hd, screen)

(SELECT DISTINCT maker FROM Product, Laptop WHERE Product.model = Laptop.model) EXCEPT (SELECT DISTINCT maker FROM Product, PC WHERE Product.model = PC.model);

"Find all makers that produce Laptops but no PCs."

SELECT * FROM PC PC1, PC PC2 WHERE PC1.speed = PC2.speed AND PC1.ram = PC2.ram AND PC1.model < PC2.model;

"Find all pairs of PCs with same speed and ram sizes." SELECT COUNT(hd) FROM PC GROUP BY hd HAVING COUNT(model) > 2;

"How many hard disk sizes are built into more than two PCs?"

Distributed Data Management

Data Models and Query Languages

The Relational Data Model Strengths and Weaknesses



Strengths

- Strict schemata good for point queries, error prevention, compression, ...
- Universal data model serving linked and unconnected data, all data types, ...
- Consistency checking (ACID) with support for different consistency levels
 Weaknesses
- Schemata need to be altered globally if certain records require additional attributes
- Impedance Mismatch:
 - Objects, structs, pointers vs. relations, records, attributes
 - Object-relational mapping (ORM) frameworks like ActiveRecord or Hibernate to the rescue
 - Complicates and slows data access; source for errors

Distributed Data Management

Data Models and Query Languages

The Relational Data Model Storage Variations

Row-Based

Store rows continuously

Row Oriented Database

 See "Database Systems II" course

Column-Based

- Store columns continuously
- See "Trends and Concepts in Software Industry" course

<u>date</u>	price	<u>size</u>
2011-01-20	10.1	10
2011-01-21	10.3	20
2011-01-22	10.5	40
2011-01-22	10.5	40
2011-01-23	10.4	5
2011-01-24	11.2	55
2011-01-25	11.4	66
2013-03-31	17.3	100

Column Oriented Database

date	price	size
2011-01-20	10.1	10
2011-01-21	10.3	20
2011-01-22	10.5	40
2011-01-23	10.4	5
2011-01-24	11.2	55
2011-01-25	11.4	66
2013-03-31	17.3	100

Distributed Data Management

Data Models and Query Languages

ThorstenPapenbrock Slide **14**

© http://www.timestored.com/time-series-data/what-is-a-column-oriented-database

Table of Data

> 55 66

100

HPI Hasso Plattner Institut

The Relational Data Model Storage Variations



Row-Based

Store rows continuously

Column-Based

Store columns continuously

Operation	Row-Based	Column-Based
Single column aggregation	Slow (full table scan)	Fast (single column scan)
Compression	Only NULL compression	Run length encoding
Column scans	Slow (skip irrelevant data)	Fast (one continuous read)
Insert/update of records	Fast (simply append)	Slow (many inserts; move data)
Single record point queries	Fast (one continuous read)	Slow (many seeks and reads)

ThorstenPapenbrock Slide **15**

Better **OLTP** performance

Better **OLAP** performance

The Relational Data Model Storage Variations



Row-Based

- Store rows continuously
- Examples by popularity:
 - Oracle
 - MySQL (open source)
 - Microsoft SQL Server
 - PostgreSQL (open source)
 - DB2

. . .

Microsoft Access



Column-Based

- Store columns continuously
- Examples by popularity:
 - Teradata
 - SAP HANA
 - SAP Sybase IQ
 - Vertica
 - MonetDB (open source)
 - C-Store (open source)

...

Distributed Data Management

Data Models and Query Languages

Definition

- Relational structure but no constraints (no key-enforcement, data types, consistency checking, ...)
- Operations: linear read and appending insert

Properties

- Encoding (ASCII, UTF-8, UTF-16, ...)
- Value separator (usually semicolon `;', comma `,', or tab ` ')
- Quote character (usually double-quotes `"')
- Escape character (usually slash `\')

Uses

- Data archiving and data exchange between heterogeneous systems
- File system storage engines (HDFS, NTFS, Ext3, ...)
- Data dumping: sensor data, measurement data, scientific data, ...

Distributed Data Management

Data Models and Query Languages





represented as CSV File

Format Example

Name	Туре	Equatorial diameter	Mass	Orbital radius	Orbital period	Rotation period	Confirmed moons	Rings	Atmosphere
Mercury	Terrestrial	0.382	0.06	0.47	0.24	58.64	0	no	minimal
Venus	Terrestrial	0.949	0.82	0.72	0.62	-243.02	0	no	CO ₂ , N ₂
Earth	Terrestrial	1.000	1.00	1.00	1.00	1.00	1	no	N ₂ , O ₂ , Ar
Mars	Terrestrial	0.532	0.11	1.52	1.88	1.03	2	no	CO ₂ , N ₂ , Ar
Jupiter	Giant	11.209	317.8	5.20	11.86	0.41	67	yes	H ₂ , He
Saturn	Giant	9.449	95.2	9.54	29.46	0.43	62	yes	H ₂ , He
Uranus	Giant	4.007	14.6	19.22	84.01	-0.72	27	yes	H ₂ , He
Neptune	Giant	3.883	17.2	30.06	164.8	0.67	14	yes	H ₂ , He

WDC_planets.csv

"Name", "Type", "EquatorialDiameter", "Mass", "OrbitalRadius", "OrbitalPeriod", "RotationPeriod", "ConfirmedMoons", "Rings", "Atmosphere"
"Mercury", "Terrestrial", "0.382", "0.06", "0.47", "0.24", "58.64", "0", "no", "minimal"
"Venus", "Terrestrial", "0.949", "0.82", "0.72", "0.62", "-243.02", "0", "no", "CO2 N2"
"Earth", "Terrestrial", "1.000", "1.00", "1.00", "1.00", "1", "no", "N2 02 Ar"
"Mars", "Terrestrial", "0.532", "0.11", "1.52", "1.88", "1.03", "2", "no", "CO2 N2 Ar"
"Jupiter", "Giant", "11.209", "317.8", "5.20", "11.86", "0.41", "67", "yes", "H2 He"
"Saturn", "Giant", "4.007", "14.6", "19.22", "84.01", "-0.72", "27", "yes", "H2 He"
"Name", "Giant", "3.883", "17.2", "30.06", "164.8", "0.67", "14", "yes", "H2 He"



Java 1.7 using *au.com.bytecode.opencsv*

```
Access Example
CSVWriter writer = null;
try {
 writer = new CSVWriter(
   new OutputStreamWriter(new FileOutputStream(dataFile, true), StandardCharsets.UTF_8), ',', '\"', '\\';
  for (String[] record : records) {
   writer.writeNext(record);
  writer.close();
                                                                                                       write
CSVReader reader = null;
try {
 reader = new CSVReader(
   new InputStreamReader(new FileInputStream(dataFile), StandardCharsets.UTF 8), ',', '\"');
 String[] record = null;
 while ((record = reader.readNext()) != null) {
   this.process(record);
 reader.close();
                                                                                                        read
```

Distributed Data Management

Data Models and Query Languages



Java 1.8 using *au.com.bytecode.opencsv*

try (CSVWriter writer = new CSVWriter(

Access Example

new OutputStreamWriter(new FileOutputStream(dataFile, true), StandardCharsets.UTF_8), ',', '\"', '\\')) { Arrays.stream(records).forEach(record -> writer.writeNext(record));

write

try (CSVReader reader = new CSVReader(new InputStreamReader(new FileInputStream(dataFile), StandardCharsets.UTF_8), ',', '\"')) { reader.forEach(record -> this.process(record)); }

Distributed Data Management

Data Models and Query Languages

ThorstenPapenbrock Slide **20**

read

Overview Relational and Non-Relational Data Models





The Key-Value Data Model Popular Key-Value Stores



https://db-engines.com/en/ranking

Distributed Data Management

Data Models and Query Languages



The Key-Value Data Model **Definition**



Some implementations do support this and some don't.

- 1. Structure
- Index (e.g. hash map): (large, distributed) key-value data structure
- 2. Constraints
- Each value is associated with a unique key.
- 3. Operations
- Store a key-value pair.
- Retrieve a value by key.
- Remove a key-value mapping.

Distributed Data Management

Data Models and Query Languages

The Key-Value Data Model **Example**





Distributed Data Management

Data Models and Query Languages

ThorstenPapenbrock Slide **24**

©Jorge Stolfi (https://commons.wikimedia.org/wiki/File:Hash_table_3_1_1_0_1_0_SP.svg)

The Key-Value Data Model Querying: Redis API

Redis

- In-memory key-value store with file persistence on disk
- Supports five data structures for values:
 - Strings: byte arrays that may represent actual strings or integers, binary serialized objects, ...
 - Hashes: dictionaries that map secondary keys to strings
 - Lists: sequences of strings that support insert, append, pop, push, trim, and many further operations
 - Sets: duplicate free collections of strings that support set operations such as diff, union, intersect, ...
 - Ordered sets: duplicate free, sorted collections of strings that use explicitly defined scores for sorting and support range operations



Distributed Data Management

Data Models and Query Languages



The Key-Value Data Model Querying: Redis API – Examples

Redis API

Strings: SET hello "hello world" GET hello → "hello world" SET users:goku {race: 'sayan', power: 9001} GET users:goku → {race: 'sayan', power: 9001}

Hashes:

HSET users:goku race 'sayan' HSET users:goku power 9001 HGET users:goku power → 9001

"<group>:<entity>" is a naming convention.

Lists:	
LPUSH mylist a	// [a]
LPUSH mylist b	// [b,a]
RPUSH mylist c	// [b,a,c]
LRANGE mylist 0	1
→ b, a	
RPOP mylist	
\rightarrow c	

Sets: SADD friends:lisa paul SADD friends:lisa duncan SADD friends:paul duncan SADD friends:paul gurney SINTER friends:lisa friends:paul

 \rightarrow duncan

Ordered sets: ZADD lisa 8 paul ZADD lisa 7 duncan ZADD lisa 2 faradin ZRANGEBYSCORE lisa 5 8 → duncan → paul

Distributed Data Management

Data Models and Query Languages



The Key-Value Data Model Strengths and Weaknesses

Strengths

- Efficient storage: fast inserts of key-value pairs
- Efficient retrieval: fast point queries, i.e., value look-ups
- Key-value pairs are easy to distribute across multiple machines
- Key-value pairs can be replicated for fault-tolerance and load balancing
 Weaknesses
- No filtering, aggregation, or joining of values/entries
 - Must be done by the application (or cluster computing framework!)
- (Usually) no parsing of complex values; must be done by the application
 - Must be done by the application (or cluster computing framework!)

Distributed Data Management

Data Models and Query Languages



Overview Relational and Non-Relational Data Models





The Column-Family Data Model Popular Column-Family Stores



Aug 2017	Rank Jul 2017	Aug 2016	DBMS	Database Model	Score Aug Jul Aug 2017 2017 2016
1.	1.	1.	Cassandra 🖶	Wide column store	126.72 +2.60 -3.52
2.	2.	2.	HBase	Wide column store	63.52 -0.10 +8.01
3.	3.	个 4.	Microsoft Azure Cosmos DB 🗄	Multi-model 🚺	9.42 +1.72 +6.87
4.	4.	4 3.	Accumulo	Wide column store	3.66 +0.17 +0.31
5.	5.		Microsoft Azure Table Storage	Wide column store	2.96 -0.13
6.	6.	6.	Google Cloud Bigtable	Wide column store	0.70 -0.11 +0.41
7.	个 8.		MapR-DB	Multi-model 🚺	0.51 -0.04
8.	个 9.	4 7.	Sqrrl	Multi-model 🚺	0.50 +0.01 +0.24
9.	1 0.	4 8.	ScyllaDB	Wide column store	0.39 -0.00 +0.35
10.			Alibaba Cloud Table Store	Wide column store	0.01

Distributed Data Management

Data Models and Query Languages

https://db-engines.com/en/ranking

The Column-Family Data Model **Definition**



1. Structure

- Multi-dimensional index (e.g. multi-dimensional hash map)
 - (large, distributed) key-value data structure that uses a hierarchy of up to three keys for one typed value
- Conceptually equivalent to sparse relational tables, i.e., each row supports arbitrary subsets of attributes.

2. Constraints

- Each value is associated with a unique key.
- Hierarchy of keys is a tree.
- Integrity constraints: keys, foreign-keys, cluster-keys (for distribution), ...
- 3. Operations
- At least: store key-value pair; retrieve value by key; remove key-value pair
- Usually: relational algebra support without joins (with own SQL dialect)

For this reason, they are also called "Wide Column Stores".

Distributed Data Management

Data Models and Query Languages

The Column-Family Data Model **Example**





The Column-Family Data Model Example 1





Hierarchy of keys enables:

- Flexible schemata (column names model attributes and row keys records)
- Value groupings (by super column names and row keys)

Management

Data Models and Query Languages

The Column-Family Data Model Example 2





Analogy:

Relational Model	Cassandra Model
Database	Keyspace
Table	Column Family (CF)
Primary key	Row key
Column name	Column name/key
Column value	Column value

Distributed Data Management

Data Models and Query Languages

Hierarchy of keys enables:

- Flexible schemata (column names model attributes and row keys records)
- Value groupings (by super column names and row keys)

The Column-Family Data Model Querying: CQL

Cassandra Query Language CQL ...

- is an SQL dialect (same syntax).
- supports all DML and DDL functionalities.
- does not support:
 - joins, group by, triggers, cursors, transactions, or (stored) procedures
 - OR and NOT logical operators (only AND)
 - subqueries
- makes, inter alia, the following restrictions:
 - WHERE conditions *should* be applied only on columns with an index
 - timestamps are comparable only with the equal operator (not <,>,<>)
 - UPDATE statements only work with a primary key (they do not work based on other columns or as mass update)
 - INSERT overrides existing records, UPDATE creates non-existing ones



Data Models and Query Languages







The Column-Family Data Model Querying: CQL – Examples



SQL:	CQL:	
CREATE DATABASE myDatabase;	CREATE KEYSPACE myDatabase WITH replication = { 'class': 'SimpleStrategy', 'replication_factor': 1};	
SELECT * FROM myTable WHERE myField > 5000 AND myField < 100000;	SELECT * FROM myTable WHERE myField > 5000 AND myField < 100000 ALLOW FILTERING;	Distributed Data Management
	Otherwise: Bad Request: Cannot execute this query as it might involve data filtering and thus may have unpredictable performance. If you want to execute it despite the performance unpredictability, use ALLOW FILTERING.	Data Models and Query Languages ThorstenPapenbrock Slide 36
The Column-Family Data Model Strengths and Weaknesses

Strengths

- Efficient storage: fast inserts of data items
- Efficient retrieval: fast point queries, i.e., value look-ups
- Data structure is easy to distribute across multiple machines
- Data structure can be replicated for fault-tolerance and load balancing
- Flexible schemata

Weaknesses

- No join and limited filtering support (filtering might also be super slow)
 - Must be done by the application (or cluster computing framework!)
- Multi-key structure groups values to entities but general groupings and aggregations are not supported
- Non-point queries, i.e., those that read more than one mapping, are costly

Distributed Data Management

Data Models and Query Languages



The Column-Family Data Model
Strengths and Weaknesses

"Writes are cheap. Write everything the way you want to read it."
If you have people and addresses and you need to read people and their addresses, then store people and addresses additionally(!) in one column family.
"Not just de-normalize, forget about normalization all together."

Alex Meng https://medium.com/@alexbmeng/cassandra-query-language-cql-vs-sql-7f6ed7706b4c

Weaknesses

- No join and limited filtering support (filtering might also be super slow)
 - Must be done by the application (or cluster computing framework!)
- Multi-key structure groups values to entities but general groupings and aggregations are not supported
- Non-point queries, i.e., those that read more than one mapping, are costly

Distributed Data Management

Hasso Plattner

Institut

Data Models and Query Languages

Overview Relational and Non-Relational Data Models





The Document Data Model Natural Document Data



Digital Documents

##fileformat=VCFv4.0
##fileDate=20090805
##source=myImputationProgramV3.1
##reference=1000GenomesPilot-NCBI36
##phasing=partial
##INFO= <id=ns,number=1,type=integer,description="number data"="" of="" samples="" with=""></id=ns,number=1,type=integer,description="number>
##INFO= <id=dp,number=1,type=integer,description="total depth"=""></id=dp,number=1,type=integer,description="total>
##INFO= <id=af,number=.,type=float,description="allele frequency"=""></id=af,number=.,type=float,description="allele>
##INFO= <id=aa,number=1,type=string,description="ancestral allele"=""></id=aa,number=1,type=string,description="ancestral>
##INFO= <id=db,number=0,type=flag,description="dbsnp 129"="" build="" membership,=""></id=db,number=0,type=flag,description="dbsnp>
##INFO= <id=h2,number=0,type=flag,description="hapmap2 membership"=""></id=h2,number=0,type=flag,description="hapmap2>
##FILTER= <id=q10,description="quality 10"="" below=""></id=q10,description="quality>
##FILTER= <id=s50,description="less 50%="" data"="" have="" of="" samples="" than=""></id=s50,description="less>
##FORMAT= <id=gt,number=1,type=string,description="genotype"></id=gt,number=1,type=string,description="genotype">
##FORMAT= <id=gq,number=1,type=integer,description="genotype quality"=""></id=gq,number=1,type=integer,description="genotype>
##FORMAT= <id=dp,number=1,type=integer,description="read depth"=""></id=dp,number=1,type=integer,description="read>
##FORMAT= <id=hq,number=2,type=integer,description="haplotype quality"=""></id=hq,number=2,type=integer,description="haplotype>
#CHROM POS ID REFALT QUAL FILTER INFO FORMAT N
A00001 NA00002 NA00003
20 14370 rs6054257 G A 29 PASS NS=3;DP=14;AF=0.5;DB;H2 GT:GQ:DP:
HQ 0 0:48:1:51,51 1 0:48:8:51,51 1/1:43:5:.,.
20 17330 . T A 3 q10 NS=3;DP=11;AF=0.017 GT:GQ:DP:
HQ 0 0:49:3:58,50 0 1:3:5:65,3 0/0:41:3
20 1110696 rs6040355 A G,T 67 PASS NS=2;DP=10;AF=0.333,0.667;AA=T;DB GT:GQ:DP:
HQ 1 2:21:6:23,27 2 1:2:0:18,2 2/2:35:4
20 1230237. T . 47 PASS NS=3;DP=13;AA=T GT:GQ:DP:
HQ 0 0:54:7:56,60 0 0:48:4:51,51 0/0:61:2
20 1234567 microsat1 GTCT G,GTACT 50 PASS NS=3;DF=9;AA=G GT:GQ:DP
0/1:35:4 0/2:17:2 1/1:40:3

1 <! DOCTYPE html> <html> <head> <title>Canvas-Rotation</title> <meta charset="UTF-8" /> <style> #square { border: 1px solid black; transform: scale(10) rotate(3deg) translateX(0px); -moz-transform: scale(10) rotate(3deg) translateX(0px) .box { transition-duration: 2s; transition-property: transform; transition-timing-function: linear; </style> </head> 20 <body> <canvas id="square" width="200" height="200"></canvas> <script> var canvas = document.createElement('canvas'); canvas.width = 200; canvas.height = 200; var image = new Image(); 28 image.src = 'images/card.png'; 29 image.width = 114; 30 image.height = 158; image.onload = window.setInterval(function() { 32 rotation(); 33 }, 1000/60); </script> 35 </body> 36 </html>

Web Pages



Structured Data

👷 Problems @ Javadoc 😡 Declaration 🛷 Search 📮 Console 🕱	
🔳 🗶 💥 🗟 🚮 🔛 🗲	🚇 📑 🖘 📑 🕶
<terminated> HyFD [Maven Build] /usr/lib/jvm/java-8-oracle/bin/java (Aug 24, 2017, 1:20:15 PM)</terminated>	
[INF0] org/xml/sax/SAXNotSupportedException.class already added, skipping	
[INFO] org/xml/sax/SAXParseException.class already added, skipping	
[INF0] org/xml/sax/XMLFilter.class already added, skipping	
[INF0] org/xml/sax/XMLReader.class already added, skipping	
[INFO] META-INF/maven/ already added, skipping	
[INF0] META-INF/ already added, skipping	
[INFO] META-INF/MANIFEST.MF already added, skipping	
[INFO] de/ already added, skipping	
[INFO] de/metanome/ already added, skipping	
[INFO] META-INF/Maven/ atready added, Skipping	fior! is missing
[waxwined of stacking the accombly file: /home/thersten/Data/Doublement/worksha	o /papophrock /HyED /1
NOTE: If multiple descriptors or descriptor-formats are provided for this project	t, the value of th
[WARNING] Replacing pre-existing project main-artifact file: /home/thorsten/Data	/Development/works
with assembly file: /home/thorsten/Data/Development/workspace/papenbrock/HyFD/ta	rget/HvFD-1.1-SNAPS
[INF0]	5 . 5
[INFO] maven-install-plugin:2.4:install (default-install) @ HyFD	
[INF0] Installing /home/thorsten/Data/Development/workspace/papenbrock/HyFD/targ	et/HyFD-1.1-SNAPSH(
[INF0] Installing /home/thorsten/Data/Development/workspace/papenbrock/HyFD/pom.	xml to /home/thors
[INF0]	
[INFO] BUILD SUCCESS	
LINFO]	
[INFO] Total time: 32.567 s	
[INF0] Finished at: 2017-08-24113:20:49+02:00	
INFO; Final Hemory, 304/642H	
[10.0]	

Distributed Data Management

Data Models and Query Languages

ThorstenPapenbrock Slide **40**

Scientific Data Formats

Log Data

The Document Data Model Popular Document Stores

Aug 2017	Rank Jul 2017	Aug 2016	DBMS	Database Model	Score Aug Jul 2017 2017	Aug 2016
1.	1.	1.	MongoDB 🔁 👾	Document store	330.50 <mark>-2.2</mark> 7	+12.01
2.	2.	个 3.	Amazon DynamoDB 담	Document store	37.62 +1.16	+11.02
3.	3.	4 2.	Couchbase 軠	Document store	32.97 <mark>-0.05</mark>	+5.57
4.	4.	4.	CouchDB	Document store	21.34 -0.81	+0.28
5.	5.	5.	MarkLogic	Multi-model 🚺	12.50 +0.07	+2.46
6.	6.	个 10.	Microsoft Azure Cosmos DB 🞛	Multi-model 🚺	9.42 +1.72	+6.87
7.	7.	4 6.	OrientDB 🕂	Multi-model 🚺	5.67 +0.10	-0.30
8.	8.	4 7.	RethinkDB	Document store	4.88 -0.07	+0.12
9.	个 10.		Firebase Realtime Database	Document store	4.51 +0.44	
10.	4 9.	4 8.	Cloudant	Document store	4.31 -0.33	-0.20
11.	11.	† 17.	Google Cloud Datastore	Document store	3.49 -0.40	+2.41
12.	12.	4 9.	RavenDB 🖶	Document store	3.27 -0.24	-1.17
13.	13.	4 12.	Apache Drill	Multi-model 🚺	3.21 -0.01	+0.83
14.	14.	4 13.	PouchDB	Document store	2.99 <mark>-0.07</mark>	+0.64
15.	15.	4 14.	ArangoDB	Multi-model 🚺	2.92 -0.04	+0.99
16.	16.	4 15.	CloudKit	Document store	2.20 -0.21	+0.37
17.	17.	4 11.	Virtuoso	Multi-model 🚺	1.98 <mark>-0.01</mark>	-0.42
18.	18.	4 16.	Datameer	Document store	1.73 +0.11	-0.01
19.	19.	4 18.	Mnesia	Document store	1.17 -0.21	+0.12



https://db-engines.com/en/ranking

Distributed Data Management

Data Models and Query Languages

The Document Data Model **Definition**

1. Structure

- Index: (large, distributed) key-value data structure
- Documents: values are documents or collections of documents that (usually) contain hierarchical data.
 - > XML, JSON, RDF, HTML, ...
- 2. Constraints
- Each value/document is associated with a unique key.

3. Operations

- Store a key-value pair.
- Retrieve a value by key.
- Remove a key-value mapping.
- Update a value of a key.

Document stores are often considered to be schemaless, but since the applications usually assume some kind of structure they are rather schema-on-read in contrast to schema-on-write.

Distributed Data Management

Data Models and Query Languages



The Document Data Model **Definition**





Relational data model

Highly-structured table organization with rigidly-defined data formats and record structure.



Document data model

Collection of complex documents with arbitrary, nested data formats and varying "record" format.

Distributed Data Management

Data Models and Query Languages



The Document Data Model Hasso HPI Plattner Example 2 Institut " id": 1, document ID "username": "ben", "password": "ughiwuv", Benno87 "contact": { "phone": 0331-1781471, embedded "email": "ben87@gmx.de", subdocument "skype": "benno.miller" **}**, "access": { embedded "level": 3, AnnaMT "group": "user" subdocument }, **Distributed Data** "supervisor": { Management "\$ref": "AnnaMT", document "\$id": 2, Data Models and reference "\$db": "users" Query Languages } } ThorstenPapenbrock Slide 45

The Document Data Model



<_id>1</_id> <username>ben</username> <password>ughiwuv</password> <contact> <phone>0331-1781254</phone> <email>ben87@gmx.de</email> <skype>benno.miller</skype> </contact> <access> <level>3</level> <group>user</group> </access> <supervisor> <ref>AnnaMT</ref> <id>2</id>

<db>users</db>

</supervisor>

HPI Hasso Plattner Institut

Note that relational databases also support hierarchical data types (e.g. XML and JSON) in their attributes.

Distributed Data Management

Data Models and Query Languages

The Document Data Model Strengths and Weaknesses



Strengths

- Efficient storage: fast inserts of key-value pairs
- Efficient retrieval: fast point queries, i.e., document (collection) look-ups
- Document (collections) are easy to distribute across multiple machines
- Document (collections) can be replicated for fault-tolerance and load balancing
- Flexible document formats: self-describing documents that may use different formats
 Weaknesses
- (Usually) developers need to explicitly/manually plan for distribution of data across instances (key-value and column-family stores do this automatically)
- Updates to documents are expensive if they alter encoding or size

Distributed Data Management

Data Models and Query Languages

The Document Data Model Querying: MongoDB API

MongoDB ...

- is a free and open-source document-oriented DBMS.
- uses JSON-like documents with schemata and integrity constraints (keys).

SQL Terms/Concepts	MongoDB Terms/Concepts
database	database
table	collection
row/record	document
column/attribute	field
index	index
table join	\$lookup, embedded document
primary key (any column)	primary key (always the _id filed)
aggregation (group by)	aggregation pipeline





Distributed Data Management

Data Models and Query Languages



SOL MongoDB Create/Drop **CREATE TABLE** people (db.people.insertOne({ id MEDIUMINT NOT NULL user_id: "abc123", Document: AUTO_INCREMENT, age: 55, user_id Varchar(30), **{** status: "A" _id: 1, age Number, }) user_id: "abc123", status char(1), age: 55, **PRIMARY KEY** (id) status: 'A' **Distributed Data** Management Data Models and Query Languages **DROP TABLE** people db.people.drop() ThorstenPapenbrock Slide 49 First insert automatically creates the document collection "people" but no schema! https://docs.mongodb.com/manual/





Insert, Update	SQL	MongoDB
and Delete	INSERT INTO people(user_id,	<pre>db.people.insertOne(</pre>
Document: <pre>{ _id: 1, user_id: "abc123",</pre>	status) VALUES ("bcd001", 45, "A"))
age: 55, status: 'A' }	UPDATE people SET status = "C" WHERE age > 25	<pre>db.people.updateMany({ age: { \$gt: 25 } }, { \$set: { status: "C" } })</pre>
	DELETE FROM people	<pre>db.people.deleteMany({ status: "D" })</pre>

WHERE status = "D"

AND

age <= 50



Select	SQL SELECT * FROM people	<pre>MongoDB db.people.find()</pre>	Always selected if
Document: {id: 1,	SELECT user_id, status FROM people WHERE status = "A"	db.people.find({ status: "A" }, { user_id: 1, stat)	<pre>not deselected. tus: 1, _id: 0 }</pre>
user_id: "abc123", age: 55, status: 'A' }	SELECT * FROM people WHERE status = "A" OR age = 50	db.people.find({ \$or: [{ status: { age: 50)	: "A" } , 0 }] }
	SELECT * FROM people WHERE age > 25	<pre>db.people.find({ age: { \$gt: 25, 5)</pre>	\$lte: 50 } }



Aggregate	SQL	MongoDB		
	SELECT COUNT(*)	<pre>db.people.count({</pre>	<pre>age: { \$gt: 30 } })</pre>	
	FROM people			
Document:	WHERE age > 30	MongoDB's aggre	egation pipeline:	
{	db.sales.aggregate(after the \$group to fu	rther refine the result.	
_id: 1,	[{ \$group : {	[{ \$group : { id : { month: { \$month: "\$date" },		
user_id: "abc123",	_id : { month: { \$			
age: 55,	year: { \$y	year: { \$year: "\$date" } }, totalPrice: { \$sum: { \$multiply: ["\$price", "\$quantity"] } }		
status: 'A'	totalPrice: { \$sum			
}	averageQuantity: count: { \$sum: 1	averageQuantity: { \$avg: "\$quantity" }, count: { \$sum: 1 } }])		
			1	
https://docs.mongodb.com/ma	Group the documents by calculate the total price , t and the count of docu	month and year and he average quantity, Iments per group.	ThorstenPapenbrock Slide 53	

The Document Data Model Querying: MongoDB API –	<pre>{ "_id" : 1, inventory_docs "item" : "abc", "price" : 12, "quantity" : 2, "inventory docs" : [</pre>
<pre>Join db.orders.aggregate([{ \$lookup: { from: "inventory", localField: "item", foreignField: "sku", as: "inventory_docs" } }]) Orders { "_id" : 1 "item": "abc", "price" : 12, "quantity" : 2 } { "_id" : 2 "item": "jkl", "price" : 20, "quantity" : 1 } t"_id" : 3 } inventory { "_id" : 1, "sku": "abc", description: "product 1", "instock" : 120 } { "_id" : 2 "sku" : "def", description: "product 2", "instock" : 120 } { "_id" : 3 "sku" : "ijk", description: "product 2", "instock" : 60 } { "_id" : 4 "sku" : "jkl", description: "product 4", "instock" : 70 } { "_id" : 5, "sku": null, description: "Incomplete" } </pre>	<pre>{ "_id" : 1, "sku" : "abc", description: "product 1", "instock" : 120 }] } { "_id" : 2, "item" : "jkl", "price" : 20, "quantity" : 1, "inventory_docs" : [{ "_id" : 4, "sku" : "jkl", "description" : "product 4", "instock" : 70 }] } { [{ "_id" : 5, "sku" : null, "description" : "Incomplete" }, { "_id" : 6 }] }</pre>

ThorstenPapenbrock Slide **54**



Index



Document:

{
_id: 1,
user_id: ``abc123",
age: 55,
status: 'A'
}



For Indexes, the DBMS maintains the document offsets in collections so that indexes work similar to indexes in relational databases.

ThorstenPapenbrock Slide 55

Very rich API

Document:



Visit the manual!

- db.collection.aggregate()
- db.collection.bulkWrite()
- db.collection.copyTo()
- db.collection.count()
- db.collection.createIndex()
- db.collection.dataSize()
- db.collection.deleteOne()
- db.collection.deleteMany()
- db.collection.distinct()
- db.collection.drop()
- db.collection.dropIndex()
- db.collection.dropIndexes()
- db.collection.ensureIndex()
- db.collection.explain()
- db.collection.find()
- db.collection.findAndModify()
- db.collection.findOne()
- db.collection.findOneAndDelete
- db.collection.findOneAndReplac
- db.collection.findOneAndUpdat

- cursor.batchSize()
- cursor.close()
- cursor.collation()
- cursor.comment()
- cursor.count()
- cursor.explain()
- cursor.forEach()
- cursor.hasNext()
- 💿 cursor.hint()
- cursor.itcount()
- cursor.limit()
- cursor.max()
- cursor.maxScan()
- cursor.maxTimeMS()
- cursor.min()
- cursor.next()
- cursor.noCursorTimeout()
- cursor.objsLeftInBatch()
- cursor.pretty()

- db.adminCommand()
- db.cloneCollection()
- db.cloneDatabase()
- db.commandHelp()
- db.copyDatabase()
- db.createCollection()
- db.createView()
- db.currentOp()
- db.dropDatabase()
- db.eval()
- db.fsyncLock()
- db.fsyncUnlock()
- db.getCollection()
- db.getCollectionInfos()
- db.getCollectionNames()
- db.getLastError()
- db.getLastErrorObj()
- db.getLogComponents()
- db.getMongo()
- db.getName()



- sh.addShard()
- sh.addShardTag()
- sh.addShardToZone()
- sh.addTagRange()
- sh.disableBalancing()
- sh.enableBalancing()
- sh.enableSharding()
- sh.getBalancerHost()
- sh.getBalancerState()
- sh.removeTagRange()
- sh.removeRangeFromZone()
- sh.help()
- sh.isBalancerRunning()
- sh.moveChunk()
- sh.removeShardTag()
- sh.removeShardFromZone()
- sh.setBalancerState()
- sh.shardCollection()
- sh.splitAt()
- sh.splitFind()

Overview Relational and Non-Relational Data Models





The Graph Data Model Natural Graph Data



Social Graphs



Road and Rail Maps



Linked Open Data

.....

Network Topologies



Hasso

Plattner

Institut

HPI

Circuit Diagrams

The Graph Data Model Popular Graph DBMS

	Rank	Σ.			S	core	
Aug 2017	Jul 2017	Aug 2016	DBMS	Database Model	Aug 2017	Jul 2017	Aug 2016
1.	1.	1.	Neo4j 🖶	Graph DBMS	38.00	-0.52	+2.43
2.	2.	个 4.	Microsoft Azure Cosmos DB 🗄	Multi-model 🚺	9.42	+1.72	+6.87
3.	3.	4 2.	OrientDB 🚦	Multi-model 🚺	5.67	+0.10	-0.30
4.	4.	4 3.	Titan	Graph DBMS	5.21	+0.29	+0.33
5.	5.	个 6.	ArangoDB	Multi-model 🚺	2.92	-0.04	+0.99
6.	6.	4 5.	Virtuoso	Multi-model 🚺	1.98	-0.01	-0.42
7.	7.	7.	Giraph	Graph DBMS	1.05	-0.01	+0.10
8.	8.	个 9.	AllegroGraph 🞛	Multi-model 🚺	0.63	+0.02	+0.17
9.	9.	4 8.	Stardog	Multi-model 🚺	0.57	+0.02	+0.03
10.	10.	个 12.	GraphDB 🗄	Multi-model 🚺	0.57	+0.04	+0.40
11.	11.	4 10.	Sqrrl	Multi-model 🚺	0.50	+0.01	+0.24
12.	12.		Graph Engine	Multi-model 🚺	0.33	-0.03	
13.	13.	4 11.	InfiniteGraph	Graph DBMS	0.30	+0.00	+0.11
14.	1 5.	14.	Dgraph	Graph DBMS	0.27	-0.01	+0.12
15.	4 14.	个 17.	Blazegraph	Multi-model 🚺	0.26	-0.03	+0.18
16.	16.		JanusGraph	Graph DBMS	0.23	-0.00	
17.	17.	个 19.	Sparksee	Graph DBMS	0.19	+0.03	+0.13
18.	18.	4 15.	FlockDB	Graph DBMS	0.17	+0.01	+0.05
19.	19.	4 18.	HyperGraphDB	Graph DBMS	0.15	-0.00	+0.08

HPI Hasso Plattner Institut

https://db-engines.com/en/ranking

Distributed Data Management

Data Models and Query Languages





1. Structure

- Nodes: entities equivalent to records in the relational model
- Edges: (un)directed connections between nodes; represent relationships
- Properties: information relating to nodes (and edges); equivalent to attributevalue or key-value pairs

2. Constraints

- Nodes consist of a unique identifier, a set of outgoing edges, a set of incoming edges, and a collection of properties.
- Edges consist of a unique identifier, the end- and start-nodes, a label, and a collection of properties.

3. Operations

- Insert/query/update/delete notes, edges, and properties (CRUD)
- Traverse edges; aggregate queries (avg, min, max, count, sum, ...)
- Most popular query language: Cypher (declarative; uses pattern matching)



Distributed Data Management

Data Models and Query Languages

The Graph Data Model **Definition**







The Graph Data Model Storage Variations



Native Graph Storage (e.g. Neo4j)

- Stores graph in a specialized graph format that points nodes directly to their adjacent nodes.
- Graph processing engines can traverse the graph by simply following links between nodes.



Non-Native Graph Storage (e.g. Titan)

- Stores graph in relational or objectoriented format and uses indexes or join-tables to find adjacent nodes.
- Graph processing engine needs to look-up links in a global index or join records/entities.



Distributed Data Management

Data Models and Query Languages

The Graph Data Model Storage Variations



Native Graph Storage (e.g. Neo4j)

- Stores graph in a specialized graph format that points nodes directly to their adjacent nodes.
- Graph processing engines can traverse the graph by simply following links between nodes.



Non-Native Graph Storage (e.g. Titan)

• Example for relational model:

CREATE TABLE vertices (

id integer **PRIMARY KEY**,

```
properties json
```

```
);
```

CREATE TABLE edges (

id	integer PRIMARY KEY ,
tail_vertex	integer REFERENCES vertices(id),
head_vertex	integer REFERENCES vertices(id),
label	text,
properties	json
);	

CREATE INDEX edges_tails **ON** edges (tail_vertex); **CREATE INDEX** edges_heads **ON** edges (head_vertex);

Cypher ...

- is a declarative query language for graphs.
- formulates queries as **patterns** to match them against the graph.
- uses an ascii-art syntax:
 - Nodes: statements in parentheses, e.g. (node)
 - Relationships: statements in arrows, e.g. -[connects]->
 - Properties: statements in curly brackets, e.g. {name:"Peter"}
- is designed for Neo4j but intended as a standard (like SQL).
- is shortened CQL (Cypher Query Language), which is not to be confused with CQL (Cassandra Query Language)!



Data Models and Query Languages





General structure for patterns



The Graph Data Model Querying: Cypher Example: Basic graph queries SQL	Nodes and edges listed with labels PART OF Product Product Product Order Cypher
SELECT p.*	MATCH (p:Product)
FROM products as p;	Return p;
SELECT p.ProductName, p.UnitPrice	MATCH (p:Product)
FROM products as p	RETURN p.productName, p.unitPrice
ORDER BY p.UnitPrice DESC	ORDER BY p.unitPrice DESC
LIMIT 10;	LIMIT 10;
SELECT p.ProductName, p.UnitPrice	MATCH (p:Product)
FROM products AS p	WHERE p.productName = "Chocolade"
WHERE p.ProductName = 'Chocolade';	RETURN p.productName, p.unitPrice;
	MATCH (p:Product {productName:"Chocolade"})

Example: Edge traversal queries

Nodes and edges listed with labels Employee Product Product Product Order Supplier



SELECT DISTINCT c.Name FROM customers c, orders o, order_details od, products p WHERE c.CustomerID = o.CustomerID AND o.OrderID = od.OrderID AND od.ProductID = p.ProductID AND p.ProductName = 'Chocolade';

Cypher

SOL

```
MATCH (c:Customer)-[:PURCHASED]->(:Order)-[:PRODUCT]->(p:Product)
WHERE p.productName = "Chocolade"
RETURN distinct c.name;
```

Note that indexing is also possible on graphs: CREATE INDEX ON :Product(productName); Slide 68

Example: Aggregation queries

SQL



SELECT e.name, count(o.OrderID) AS Count FROM Employee e JOIN Order o ON (o.EmployeeID = e.EmployeeID) GROUP BY e.EmployeeID, e.name ORDER BY Count DESC LIMIT 10;

Cypher

MATCH (:Order)<-[:SOLD]-(e:Employee) RETURN e.name, count(o.id) AS Count ORDER BY Count DESC LIMIT 10; Grouping for aggregation is implicit: The first aggregation function causes all non-aggregated columns to automatically become grouping keys. → group by employee ID



Example: Creating a graph

```
CREATE (you:Person {name:"You"})
RETURN you
```

MATCH (you:Person {name:"You"})
CREATE (you)-[like:LIKE]->(neo:Database {name:"Neo4j" })
RETURN you,like,neo

```
MATCH (you:Person {name:"You"})
FOREACH (name in ["Johan","Rajesh","Anna","Julia","Andrew"] |
CREATE (you)-[:FRIEND]->(:Person {name:name}))
```

```
MATCH (neo:Database {name:"Neo4j"})
MATCH (anna:Person {name:"Anna"})
CREATE (anna)-[:FRIEND]->(:Person:Expert {name:"Amanda"})-[:WORKED_WITH]->(neo)
```

Two labels!

https://neo4j.com/developer/cypher-query-language/

Slide 70

Andre

You

mand

Johan

Anna

Hasso Plattner

Institut



Ways to Model Properties/Relationships

Model "Nodes that have an address", which should be used for filtering.

a) Using a property and then filtering by property (node {address: "address"})

The Graph Data Model

- b) Using a specific relationship type and then filtering by relationship type (node)-[:HAS_ADDRESS]->(address)
- C) Using a generic relationship type and then filtering by end node label (node)-[:HAS]->(address:Address)
- d) Using a generic relationship type and then filtering by relationship property (node)-[:HAS {type: "address"}]->(address)
- e) Using a generic relationship type and then filtering by end node property (node)-[:HAS]->(address {type: "address"})
 - Best way depends on query performance (for filtering probably b)), semantic fit (maybe c)), and extensibility (maybe a) or d))

https://neo4j.com/developer/cypher-query-language/ — For further reading on Cypher



Data Models and Query Languages


The Graph Data Model Triple-Stores

Definition

 Same graph definition as property graphs, but graph is stored in simple three-part, sentence-like statements of the form

(subject, predicate, object) instead of nodes with collections of direct links.

- Subject: start node label
- Predicate: edge/property label

then: triple = property

Object: end node label or static value with primitive data type

then: triple = edge

Examples

- (Jim, likes, Bananas)
- (Jim, age, 28)
- (Leon, is_a, Lion)
- (Leon, lives_in, Africa)
- (Africa, is_a, Continent)



Data Models and Query Languages



The Graph Data Model Triple-Stores



Triple-Stores

- Examples:
 - Datomic
 - AllegroGraph
 - Virtuoso
- Query languages:
 - SPARQL
 - Datalog

Property Graph DBMSs

- Examples:
 - Neo4j
 - Titan
 - InfiniteGraph
- Query languages:
 - Cypher
 - Gremlin

Distributed Data Management

Data Models and Query Languages

The Graph Data Model **Triple-Stores**

Semantic Web

- Initiative of the World Wide Web Consortium (W3C) to extend the Web through standards for data formats and exchange protocols
- Most popular use case for triple stores
- Idea: Store entities/relations AND their semantic meaning in machine readable format!
- Approach: "Resource Description Framework" (RDF)
 - Subject, predicate and object in triples are represented as URIs
 - Example:

<http://www.hpi.de/#TPapenbrock>

<http://www.w3.org/1999/02/22-rdf-syntax-ns#type> <http://www.w3.org/2000/10/swap/pim/contact#Person> .

Ensures that datasets can be combined without semantic conflicts: <http://www.hpi.de/#HS1> ≠ <http://www.uni-potsdam.de/#HS1>

Distributed Data URIs don't need to Management resolve to web pages

Data Models and Query Languages

ThorstenPapenbrock Slide 75





Which tie their meaning to an ID



- "Resource Description Framework Schema" (RDFS)
 - A set of well defined RDF classes and properties to describe ontologies (=formal description of "real" entities in some domain)
 - Example for RDFS classes:

rdfs:Class (declares a node as a class for other nodes)
foaf:Person rdf:type rdfs:Class .

→ ex:Lisa rdf:type foaf:Person .

• Example for RDFS properties:

rdfs:domain (declares the subject type for a predicate)
rdfs:range (declares the object type for a predicate)
ex:student rdfs:domain foaf:Person .
ex:student rdfs:range foaf:University .

 \rightarrow ex:Lisa ex:student ex:UniversityPotsdam .

If RDFS is insufficient to build your ontology, use its extension **OWL** ("Web Ontology Language")



The Graph Data Model Querying: SPARQL

SPARQL ...

- is a declarative query language for triple-store graphs in RDF format.
- formulates queries in RDF syntax.
- is an acronym for "SPARQL Protocol and RDF Query Language".
- Example:

SELECT	?locationName
WHERE	{
?hpi	:name "HPI gGmbH" .
?hpi	:location ?locationName .
}	

MATCH (hpi {name: "HPI gGmbH"})-[:location]->(loc) RETURN loc.name Cypher



Data Models and Query Languages

ThorstenPapenbrock Slide **78**



SPARQL

The Graph Data Model Querying: SPARQL SPARQL ...

- is a declarative query language for triple-store graphs in RDF format
- formulates queries in RDF syntax
- is an acronym for "SPARQL Protocol and RDF Query Language"
- Example:

```
      SELECT ?personName
      SPARQL

      WHERE {
      ?person :name ?personName .
      Distributed Data Management

      ?person :bornIn / :within* / :name "Europe" .
      Distributed Data Management

      ?url>:label
      <variable>:label
      Data Models and Query Languages

      MATCH (person)-[:bornIn]->()-[:within*0..]->(location {name: "Europe"}})
      Slide 79
```



SPARQL and Cipher

are quite similar.

The Graph Data Model Strengths and Weaknesses

Strengths

- Many-to-many relationships (other data models heavily prefer one-to-many)
- Efficient traversal of relationships between entities (relationship queries)
 - Traversal costs proportional to the average out-degree of nodes (and not proportional to the overall number of relationships)
 - > Join performance scales naturally with the size of the data
- Natural support for graph queries: shortest path, community detection, ...
- Flexible schemata due to flexible edge and property definitions
- Direct mapping of nodes/edges to data structures of object-oriented applications
 Weaknesses
- OLTP and CRUD operations on many nodes are comparatively slow
- Data Distribution is hard, because workload is based on data locality
- Querying difficult due to unknown schema (flexibility leads to misuse)

Distributed Data Management

Data Models and Query Languages



The Graph Data Model Graph DBMSs and Distribution

Replication/Clustering

- Supported by most graph DBMSs
- Same techniques for consistency management as other DBMSs
- Queries can be routed to any replica and then be served from it

Partitioning/Sharding

- Performance-wise problematic, because graph queries have join character rather than point query character and often cross partition boundaries.
 - Most systems offer rudimentary partitioning support, but try to avoid it and go for replication (e.g. Neo4j).
- Challenge: Find a graph partitioning with ...
 - a) possibly few inter-partition links;
 - b) possibly balanced partition sizes;
 - c) a certain number of partitions that matches physical nodes.



Hasso Plattner

nstitut

Data Models and Query Languages

Subject to

research!

The Graph Data Model Further Reading on Graph Databases





NEW OPPORTUNITIES FOR CONNECTED DATA

lan Robinson, Jim Webber & Emil Eifrem

Graph Databases

- Free to download as pdf at:
 - <u>http://graphdatabases.com/</u>

Distributed Data Management

Data Models and Query Languages

Overview Relational and Non-Relational Data Models







Data Models and Query Languages Summary



Discrete Data Minimally Connected Data	Connected Data Focused on Data Relationships		
Other NoSQL	Relational Databases	Graph Databases	
 inhomogeneous data 	 homogeneous data 	 highly linked data 	
 frequent schema changes 	 relatively reliable schemata 	 frequent schema changes 	
 fast growth 	 moderate growth 	 moderate growth 	
 little/no relationship support 	 full relationship support 	 specialized on relationships 	
 usually sacrifice ACID 	 usually comply with ACID 	 usually comply with ACID 	Distributed Data Management
 usually horizontal scaling 	 usually vertical scaling 	 usually vertical scaling 	Data Models and
 data distribution 	 data compression 	 data optimization 	Query Languages
 throughput 	 transactions and security 	 relationship traversal 	ThorstenPapenbrock
 OLTP focus 	 OLTP and OLAP 	 OLAP focus 	Slide 85
Image: © https://neo4j.com/blog/	'aggregate-stores-tour/		

Data Models and Query Languages Check yourself

- Train your query skills with the following exercises:
 - MongoDB
 - <u>https://www.w3resource.com/mongodb-exercises/</u>
 - (includes solutions)
 - Neo4j / Cypher
 - <u>https://www.uio.no/studier/emner/matnat/ifi/INF3100/v17/undervisni</u> <u>ngsmateriale/graph-dbs---neo4j.pdf</u>
- It helps if you really set up a database and try the queries yourself. If you face any problems in doing so, please do not hesitate to ask us or the mailing list for help.

Distributed Data Management

Data Models and Query Languages



