Big Data Management and NoSQL Databases

Lecture 1. Introduction

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What is Big Data?

- buzzword?
- bubble?
- gold rush?
- revolution?

“Big data is like teenage sex: everyone talks about it, nobody really knows how to do it, everyone thinks everyone else is doing it, so everyone claims they are doing it.”

Dan Ariely
What is Big Data?

- No standard definition
- First occurrence of the term: High Performance Computing (HPC)

Gartner: “Big Data” is high volume, high velocity, and/or high variety information assets that require new forms of processing to enable enhanced decision making, insight discovery and process optimization.
**What is Big Data?**

*IBM: Depending on the industry and organization, Big Data encompasses information from internal and external sources such as transactions, social media, enterprise content, sensors, and mobile devices. Companies can leverage data to adapt their products and services to better meet customer needs, optimize operations and infrastructure, and find new sources of revenue.*

- **Social media and networks** (all of us are generating data)
- **Scientific instruments** (collecting all sorts of data)
- **Sensor technology and networks** (measuring all kinds of data)
- **Mobile devices** (tracking all objects all the time)
Big Data Characteristics:
Volume (Scale)

40 ZETTABYTES
[43 TRILLION GIGABYTES]
of data will be created by 2020, an increase of 300 times from 2005

6 BILLION PEOPLE
have cell phones

WORLD POPULATION: 7 BILLION

Data volume is increasing exponentially, not linearly

Volume
SCALE OF DATA

10^{21}

10^{18}

10^{12}

Most companies in the U.S. have at least 100 TERABYTES
[100,000 GIGABYTES] of data stored

http://www.ibmbigdatahub.com/

It's estimated that 2.5 QUINTILLION BYTES
[2.3 TRILLION GIGABYTES] of data are created each day
Big Data Characteristics: 

**Variety (Complexity)**

As of 2011, the global size of data in healthcare was estimated to be 150 exabytes (161 billion gigabytes).

By 2014, it's anticipated there will be:
- 420 million wearable, wireless health monitors
- 4 billion+ hours of video are watched on YouTube each month
- 30 billion pieces of content are shared on Facebook every month
- 400 million tweets are sent per day by about 200 million monthly active users

Various formats, types, and structures (from semi-structured XML to unstructured multimedia)

Static data vs. streaming data
Big Data Characteristics: Velocity (Speed)

The New York Stock Exchange captures 1 TB OF TRADE INFORMATION during each trading session.

Modern cars have close to 100 SENSORS that monitor items such as fuel level and tire pressure.

By 2016, it is projected there will be 18.9 BILLION NETWORK CONNECTIONS – almost 2.5 connections per person on earth.

Data is being generated fast and need to be processed fast.

Online Data Analytics

http://www.ibmbigdatahub.com/
Big Data Characteristics: **Veracity** (Uncertainty)

**Veracity**

**Uncertainty of Data**

Poor data quality costs the US economy around $3.1 trillion a year.

1 in 3 business leaders don’t trust the information they use to make decisions.

27% of respondents in one survey were unsure of how much of their data was inaccurate.

Uncertainty due to inconsistency, incompleteness, latency, ambiguities, or approximations.
Processing Big Data

- **OLTP**: Online Transaction Processing (DBMSs)
  - Database applications
  - Storing, querying, multiuser access
- **OLAP**: Online Analytical Processing (Data Warehousing)
  - Answer multi-dimensional analytical queries
  - Financial/marketing reporting, budgeting, forecasting, …
- **RTAP**: Real-Time Analytical Processing (Big Data Architecture & Technology)
  - Data gathered & processed in a real-time
    - Streaming fashion
  - Real-time data queried and presented in an online fashion
  - Real-time and history data combined and mined interactively
Key Big Data-Related Technologies

- Distributed file systems
- **NoSQL databases**
- Grid computing, cloud computing
- MapReduce and other new paradigms
- Large scale machine learning

http://e-theses.imtlucca.it/34/
Relational Database Management Systems (RDMBSs)

- Predominant technology for storing structured data
  - Web and business applications
- Relational calculus, SQL
- Often thought of as the only alternative for data storage
  - Persistence, concurrency control, integration mechanism, …
- Alternatives: Object databases or XML stores
  - Never gained the same adoption and market share
“NoSQL”

- 1998 first used for a relational database that omitted the use of SQL
  - Carlo Strozzi
- 2009 used for conferences of advocates of non-relational databases
  - Eric Evans
  - Blogger, developer at Rackspace

NoSQL movement = “the whole point of seeking alternatives is that you need to solve a problem that relational databases are a bad fit for”
„NoSQL“

- Not „no to SQL“
  - Another option, not the only one
- Not „not only SQL“
  - Oracle DB or PostgreSQL would fit the definition
- „Next Generation Databases mostly addressing some of the points: being non-relational, distributed, open-source and horizontally scalable. The original intention has been modern web-scale databases. Often more characteristics apply as: schema-free, easy replication support, simple API, eventually consistent (BASE, not ACID), a huge data amount, and more“
The End of Relational Databases?

- Relational databases are not going away
- Compelling arguments for most projects
  - Familiarity, stability, feature set, and available support
- We should see relational databases as one option for data storage
  - Polyglot persistence – using different data stores in different circumstances
  - Search for optimal storage for a particular application
Motivation for NoSQL Databases

- Huge amounts of data are now handled in real-time
- Both data and use cases are getting more and more dynamic
- Social networks (relying on graph data) have gained impressive momentum
  - Special type of NoSQL databases: graph databases
- Full-texts have always been treated shabbily by RDBMS
Example: FaceBook
Statistics from 2010

- 500 million users
- 570 billion page views per month
- 3 billion photos uploaded per month
- 1.2 million photos served per second
- 25 billion pieces of content (updates, comments) shared every month
- 50 million server-side operations per second

2008: 10,000 servers
2009: 30,000 servers
...

→ One RDBMS may not be enough to keep this going on!

And even bewer numbers:
https://research.facebook.com/blog/1522692927972019/facebook-s-top-open-data-problems/
Example: FaceBook
Architecture from 2010

Cassandra
- NoSQL distributed storage system with no single point of failure
- For inbox searching

Hadoop/Hive
- An open source MapReduce implementation
- Enables to perform calculations on massive amounts of data
- Hive enables to use SQL queries against Hadoop
Example: FaceBook
Architecture from 2010 and later

**Memcached**
- Distributed memory caching system
- Caching layer between the web servers and MySQL servers
  - Since database access is relatively slow

**HBase**
- Hadoop database, used for e-mails, instant messaging and SMS
- Has recently replaced MySQL, Cassandra and few others
- Built on Google’s BigTable model
NoSQL Databases

Five Advantages

1. Elastic scaling
   - “Classical” database administrators **scale up** – buy bigger servers as database load increases
   - **Scaling out** – distributing the database across multiple hosts as load increases

2. Big Data
   - Volumes of data that are being stored have increased massively
   - Opens new dimensions that cannot be handled with RDBMS

http://www.techrepublic.com/blog/10things/10-things-you-should-know-about-nosql-databases/1772
NoSQL Databases

Five Advantages

3. Goodbye DBAs (see you later?)
   - Automatic repair, distribution, tuning, … vs. expensive, highly trained DBAs of RDBMS

4. Economics
   - Based on cheap commodity servers → less costs per transaction/second

5. Flexible Data Models
   - Non-existing/relaxed data schema → structural changes cause no overhead
NoSQL Databases

Five Challenges

1. Maturity
   - Still in pre-production phase
   - Key features yet to be implemented

2. Support
   - Mostly open source, result from start-ups
     - Enables fact development
   - Limited resources or credibility

3. Administration
   - Require lot of skill to install and effort to maintain
NoSQL Databases
Five Challenges

4. Analytics and Business Intelligence

- Focused on web apps scenarios
  - Modern Web 2.0 applications
  - Insert-read-update-delete

- Limited ad-hoc querying
  - Even a simple query requires significant programming expertise

5. Expertise

- Few number of NoSQL experts available in the market
## Data Assumptions

<table>
<thead>
<tr>
<th>RDBMS</th>
<th>NoSQL</th>
</tr>
</thead>
<tbody>
<tr>
<td>integrity is mission-critical</td>
<td>OK as long as most data is correct</td>
</tr>
<tr>
<td>data format consistent, well-defined</td>
<td>data format unknown or inconsistent</td>
</tr>
<tr>
<td>data is of long-term value</td>
<td>data are expected to be replaced</td>
</tr>
<tr>
<td>data updates are frequent</td>
<td>write-once, read multiple (no updates, or at least not often)</td>
</tr>
<tr>
<td>predictable, linear growth</td>
<td>unpredictable growth (exponential)</td>
</tr>
<tr>
<td>non-programmers writing queries</td>
<td>only programmers writing queries</td>
</tr>
<tr>
<td>regular backup</td>
<td>replication</td>
</tr>
<tr>
<td>access through master server</td>
<td>sharding across multiple nodes</td>
</tr>
</tbody>
</table>
NoSQL Data Model

Aggregates

- Data model = the model by which the database organizes data
- Each NoSQL solution has a different model
  - Key-value, document, column-family, graph
  - First three orient on aggregates

Aggregate

- A data unit with a complex structure
  - Not just a set of tuples like in RDBMS
- Domain-Driven Design: “an aggregate is a collection of related objects that we wish to treat as a unit"
  - A unit for data manipulation and management of consistency
<table>
<thead>
<tr>
<th>Customer</th>
<th>Orders</th>
</tr>
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<tbody>
<tr>
<td>Id</td>
<td>Id</td>
</tr>
<tr>
<td></td>
<td>CustomerId</td>
</tr>
<tr>
<td></td>
<td>ShippingAddressIdId</td>
</tr>
<tr>
<td>1</td>
<td>99</td>
</tr>
<tr>
<td>Martin</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>77</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Product</th>
<th>BillingAddress</th>
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<tbody>
<tr>
<td>Id</td>
<td>Id</td>
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<tr>
<td></td>
<td>CustomerId</td>
</tr>
<tr>
<td></td>
<td>AddressId</td>
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<td>27</td>
<td>55</td>
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<tr>
<td>NoSQL Distilled</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>77</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>OrderItem</th>
<th>Address</th>
</tr>
</thead>
<tbody>
<tr>
<td>Id</td>
<td>Id</td>
</tr>
<tr>
<td></td>
<td>City</td>
</tr>
<tr>
<td></td>
<td>77</td>
</tr>
<tr>
<td></td>
<td>Chicago</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>OrderPayment</th>
<th>Id</th>
<th>OrderId</th>
<th>CardNumber</th>
<th>BillingAddressId</th>
<th>txnId</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>33</td>
<td>99</td>
<td>1000-1000</td>
<td>55</td>
<td>abelif879rft</td>
</tr>
</tbody>
</table>
// in customers
{
    "customer": {
        "id": 1,
        "name": "Martin",
        "billingAddress": [{"city": "Chicago"}],
        "orders": [{
            "id": 99,
            "customerId": 1,
            "orderItems": [
                {
                    "productId": 27,
                    "price": 32.45,
                    "productName": "NoSQL Distilled"
                }
            ],
            "shippingAddress": [{"city": "Chicago"}]
        }]
    }
}
NoSQL Data Model

Aggregates – aggregate-ignorant

- There is no universal strategy how to draw aggregate boundaries
  - Depends on how we manipulate the data
- RDBMS and graph databases are aggregate-ignorant
  - It is not a bad thing, it is a feature
  - Allows to easily look at the data in different ways
  - Better choice when we do not have a primary structure for manipulating data
NoSQL Data Model

Aggregates – aggregate-oriented

- Aggregate orientation
  - Aggregates give the database information about which bits of data will be manipulated together
    - Which should live on the same node
  - Helps greatly with running on a cluster
    - We need to minimize the number of nodes we need to query when we are gathering data

- Consequence for transactions
  - NoSQL databases support atomic manipulation of a single aggregate at a time
NoSQL Databases

Materialized Views

- Disadvantage: the aggregated structure is given, other types of aggregations cannot be done easily
  - RDBMSs lack of aggregate structure → support for accessing data in different ways (using views)

- Solution: materialized views
  - Pre-computed and cached queries

- Strategies:
  - Update materialized view when we update the base data
    - For more frequent reads of the view than writes
  - Run batch jobs to update the materialized views at regular intervals
NoSQL Databases

Schemalessness

- When we want to store data in a RDBMS, we need to define a schema
- Advocates of schemalessness rejoice in freedom and flexibility
  - Allows to easily change your data storage as we learn more about the project
  - Easier to deal with non-uniform data
- Fact: there is usually an implicit schema present
  - The program working with the data must know its structure
Types of NoSQL Databases

Core:
- Key-value databases
- Document databases
- Column-family (column-oriented/columnar) stores
- Graph databases

Non-core:
- Object databases
- XML databases
- …

http://nosql-database.org/
Key-value store
Basic characteristics

- The simplest NoSQL data stores
- A simple hash table (map), primarily used when all access to the database is via primary key
- A table in RDBMS with two columns, such as ID and NAME
  - ID column being the key
  - NAME column storing the value
    - A BLOB that the data store just stores
- Basic operations:
  - Get the value for the key
  - Put a value for a key
  - Delete a key from the data store
- Simple → great performance, easily scaled
- Simple → not for complex queries, aggregation needs
Key-value store
Representatives

- riak
- redis
- MemcachedDB

Not open-source:
- Amazon DynamoDB

Open-source version:
- ORACLE BERKELEY DB
- Project Voldemort
Key-value store

Suitable Use Cases

Storing Session Information
- Every web session is assigned a unique session_id value
- Everything about the session can be stored by a single PUT request or retrieved using a single GET
- Fast, everything is stored in a single object

User Profiles, Preferences
- Every user has a unique user_id, user_name + preferences such as language, colour, time zone, which products the user has access to, ...
- As in the previous case:
  - Fast, single object, single GET/PUT

Shopping Cart Data
- Similar to the previous cases
Key-value store

When Not to Use

Relationships among Data
- Relationships between different sets of data
- Some key-value stores provide link-walking features
  - Not usual

Multioperation Transactions
- Saving multiple keys
  - Failure to save any one of them → revert or roll back the rest of the operations

Query by Data
- Search the keys based on something found in the value part

Operations by Sets
- Operations are limited to one key at a time
- No way to operate upon multiple keys at the same time
Column-Family Stores
Basic Characteristics

- Also “columnar” or “column-oriented”
- **Column families** = rows that have **many** columns associated with a **row key**
- Column families are groups of related data that is often accessed together
  - e.g., for a customer we access all profile information at the same time, but not orders
Column-Family Stores
Representatives

Google’s BigTable
H·BASE
Cassandra
HYPERTABLE
SimpleDB
Column-Family Stores

Suitable Use Cases

- **Event Logging**
  - Ability to store any data structures → good choice to store event information

- **Content Management Systems, Blogging Platforms**
  - We can store blog entries with tags, categories, links, and trackbacks in different columns
  - Comments can be either stored in the same row or moved to a different keyspace
  - Blog users and the actual blogs can be put into different column families
Column-Family Stores

When Not to Use

Systems that Require ACID Transactions
- Column-family stores are not just a special kind of RDBMSs with variable set of columns!

Aggregation of the Data Using Queries
- (such as SUM or AVG)
- Have to be done on the client side

For Early Prototypes
- We are not sure how the query patterns may change
- As the query patterns change, we have to change the column family design
Document Databases

Basic Characteristics

- Documents are the main concept
  - Stored and retrieved
  - XML, JSON, ...

- Documents are
  - Self-describing
  - Hierarchical tree data structures
  - Can consist of maps, collections (lists, sets, ...), scalar values, nested documents, ...

- Documents in a collection are expected to be similar
  - Their schema can differ

- Document databases store documents in the value part of the key-value store
  - Key-value stores where the value is examinable
Document Databases
Representatives

mongoDB
CouchDB
OrientDB
RAVENDB
Lotus Notes
Storage Facility
Document Databases

Suitable Use Cases

Event Logging
- Many different applications want to log events
  - Type of data being captured keeps changing
- Events can be sharded (i.e. divided) by the name of the application or type of event

Content Management Systems, Blogging Platforms
- Managing user comments, user registrations, profiles, web-facing documents, …

Web Analytics or Real-Time Analytics
- Parts of the document can be updated
- New metrics can be easily added without schema changes
  - E.g. adding a member of a list, set,…

E-Commerce Applications
- Flexible schema for products and orders
- Evolving data models without expensive data migration
Document Databases
When Not to Use

Complex Transactions Spanning Different Operations
- Atomic cross-document operations
  - Some document databases do support (e.g., RavenDB)

Queries against Varying Aggregate Structure
- Design of aggregate is constantly changing → we need to save the aggregates at the lowest level of granularity
  - i.e. to normalize the data
Graph Databases
Basic Characteristics

- To store entities and relationships between these entities
  - Node is an instance of an object
  - Nodes have properties
    - e.g., name
  - Edges have directional significance
  - Edges have types
    - e.g., likes, friend, …

- Nodes are organized by relationships
  - Allow to find interesting patterns
  - e.g., “Get all nodes employed by Big Co that like NoSQL Distilled”
Graph Databases

RDBMS vs. Graph Databases

- When we store a graph-like structure in RDBMS, it is for a single type of relationship
  - “Who is my manager”
- Adding another relationship usually means a lot of schema changes
- In RDBMS we model the graph beforehand based on the Traversal we want
  - If the Traversal changes, the data will have to change
  - In graph databases the relationship is not calculated at query time but persisted
Graph Databases
Representatives

- Neo4j
- InfiniteGraph
- OrientDB
- FlockDB
Graph Databases

Suitable Use Cases

Connected Data
- Social networks
- Any link-rich domain is well suited for graph databases

Routing, Dispatch, and Location-Based Services
- Node = location or address that has a delivery
- Graph = nodes where a delivery has to be made
- Relationships = distance

Recommendation Engines
- “your friends also bought this product”
- “when invoicing this item, these other items are usually invoiced”
Graph Databases
When Not to Use

- When we want to update all or a subset of entities
  - Changing a property on all the nodes is not a straightforward operation
  - e.g., analytics solution where all entities may need to be updated with a changed property

- Some graph databases may be unable to handle lots of data
  - Distribution of a graph is difficult or impossible
NoSQL Data Model
Aggregates and NoSQL databases

Key-value database
- Aggregate = some big blob of mostly meaningless bits
  - But we can store anything
- We can only access an aggregate by lookup based on its key

Document database
- Enables to see a structure in the aggregate
  - But we are limited by the structure when storing (similarity)
- We can submit queries to the database based on the fields in the aggregate
NoSQL Data Model
Aggregates and NoSQL databases

Column-family stores

- A two-level aggregate structure
  - The first key is a row identifier, picking up the aggregate of interest
  - The second-level values are referred to as columns
- Ways to think about how the data is structured:
  - **Row-oriented**: each row is an aggregate with column families representing useful chunks of data (profile, order history)
  - **Column-oriented**: each column family defines a record type (e.g., customer profiles) with rows for each of the records; a row is the join of records in all column families
References

- Pramod J. Sadalage – Martin Fowler: *NoSQL Distilled: A Brief Guide to the Emerging World of Polyglot Persistence*
- Sherif Sakr – Eric Pardede: *Graph Data Management: Techniques and Applications*
- Shashank Tiwari: *Professional NoSQL*