B4M36DS2, BE4M36DS2: Database Systems 2

http://www.ksi.mff.cuni.cz/~svoboda/courses/191-B4M36DS2/

Lecture 1 Introduction

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23. 9. 2019

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Lecture Outline

Big Data

- Characteristics
- Current trends

NoSQL databases

- Motivation
- Features

Overview of NoSQL database types

• Key-value, wide column, document, graph, ...

What is Big Data?

Buzzword? Bubble? Gold rush? Revolution?



Dan Ariely: 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.

What is Big Data?

No standard definition

• Gartner (*research and advisory company*): High Performance Computing

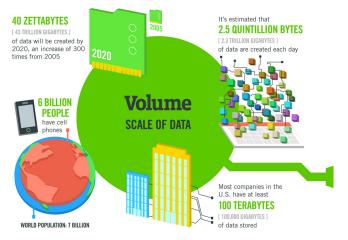
Big Data is **high volume**, **high velocity**, and/or **high variety** information assets that require **new forms of processing** to enable enhanced decision making, insight discovery and process optimization.

Where is Big Data?

Sources of Big Data

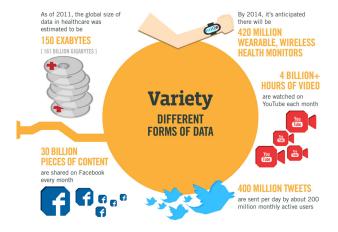
- Social media and networks
 - ...all of us are generating data
- Scientific instruments
 - ...collecting all sorts of data
- Mobile devices
 - ...tracking all objects all the time
- Sensor technology and networks
 - ...measuring all kinds of data

Volume (Scale)



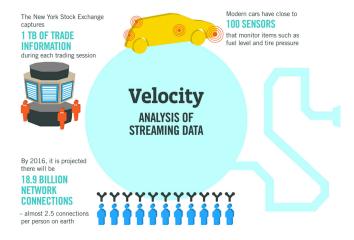
Source: http://www.ibmbigdatahub.com/

Variety (Complexity)



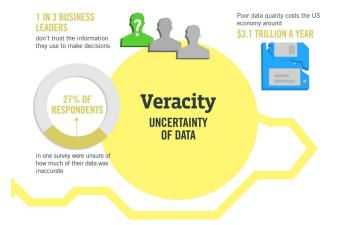
Source: http://www.ibmbigdatahub.com/

Velocity (Speed)



Source: http://www.ibmbigdatahub.com/

Veracity (Uncertainty)



Source: http://www.ibmbigdatahub.com/

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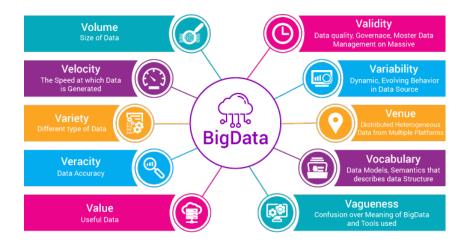
Basic 4V

- <u>Volume</u> (Scale)
 - Data volume is increasing exponentially, not linearly
 - Even large amounts of small data can result into Big Data
- Variety (Complexity)
 - Various formats, types, and structures (from semi-structured XML to unstructured multimedia)
- <u>Velocity</u> (Speed)
 - Data is being generated fast and needs to be processed fast
- Veracity (Uncertainty)
 - Uncertainty due to inconsistency, incompleteness, latency, ambiguities, or approximations

Additional V and C

- Value
 - Business value of the data (needs to be revealed)
- Validity
 - Data correctness and accuracy with respect to the intended use
- <u>V</u>olatility
 - Period of time the data is valid and should be maintained
- <u>Cardinality</u>
- <u>C</u>ontinuity
- <u>Complexity</u>

Additional V



Source: https://www.xenonstack.com/blog/big-data-engineering/ingestion-processing-big-data-iot-stream/

Data model

 $\mathsf{Instance} \to \textbf{database} \to \textbf{table} \to \textbf{row}$

Query languages

- Real-world: SQL (Structured Query Language)
- Formal: Relational algebra, relational calculi (domain, tuple)

Query patterns

 Selection based on complex conditions, projection, joins, aggregation, derivation of new values, recursive queries, ...

Representatives

- Oracle Database, Microsoft SQL Server, IBM DB2
- MySQL, PostgreSQL

Representatives











Features: Normal Forms

Model

- Functional dependencies
- 1NF, 2NF, 3NF, BCNF (Boyce-Codd normal form)

Objective

- Normalization of database schema to BCNF or 3NF
- Algorithms: decomposition or synthesis

Motivation

- Diminish data redundancy, prevent update anomalies
- However:
 - Data is scattered into small pieces (high granularity), and so
 - these pieces have to be joined back together when querying!

Features: Transactions

Model

• **Transaction** = flat sequence of database operations (READ, WRITE, COMMIT, ABORT)

Objectives

- Enforcement of ACID properties
- Efficient parallel / concurrent execution (slow hard drives, ...)

ACID properties

- <u>Atomicity</u> partial execution is not allowed (all or nothing)
- <u>Consistency</u> transactions turn one valid database state into another
- Isolation uncommitted effects are concealed among transactions
- <u>Durability</u> effects of committed transactions are permanent

Big Data

- Volume: terabytes → zettabytes
- Variety: structured \rightarrow structured and unstructured data
- Velocity: batch processing → streaming data

• ...

Big users

- Population online, hours spent online, devices online, ...
- Rapidly growing companies / web applications
 - Even millions of users within a few months

Everything is in cloud

- SaaS: Software as a Service
- PaaS: Platform as a Service
- IaaS: Infrastructure as a Service

Processing paradigms

- OLTP: Online Transaction Processing
- OLAP: Online Analytical Processing
- ...but also...
- RTAP: Real-Time Analytical Processing

Data assumptions

- Data format is becoming unknown or inconsistent
- Linear growth \rightarrow unpredictable exponential growth
- Read requests often prevail write requests
- Data updates are no longer frequent
- Data is expected to be replaced
- Strong consistency is no longer mission-critical

\Rightarrow New approach is required

Relational databases simply do not follow the current trends

Key technologies

- Distributed file systems
- MapReduce and other programming models
- Grid computing, cloud computing
- NoSQL databases
- Data warehouses
- Large scale machine learning

NoSQL Databases

What does NoSQL actually mean?

A bit of history ...

- 1998
 - First used for a relational database that omitted usage of SQL
- 2009
 - First used during a conference to advocate non-relational databases

So?

- Not: no to SQL
- Not: not only SQL
- NoSQL is an accidental term with no precise definition

NoSQL Databases

What does NoSQL actually mean?

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 databases = 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**, a **huge data amount**, and more.

Source: http://nosql-database.org/

Types of NoSQL Databases

Core types

- Key-value stores
- Wide column (column family, column oriented, ...) stores
- Document stores
- Graph databases

Non-core types

- Object databases
- Native XML databases
- RDF stores
- ...

Key-Value Stores

Data model

- The most simple NoSQL database type
 - Works as a simple hash table (mapping)
- Key-value pairs
 - Key (id, identifier, primary key)
 - Value: binary object, black box for the database system

Query patterns

- Create, update or remove value for a given key
- Get value for a given key

Characteristics

- Simple model ⇒ great performance, easily scaled, ...
- Simple model \Rightarrow **not for complex queries nor complex data**

Key-Value Stores

Suitable use cases

- Session data, user profiles, user preferences, shopping carts, ...
 - I.e. when values are only accessed via keys

When not to use

- Relationships among entities
- Queries requiring access to the content of the value part
- Set operations involving multiple key-value pairs

Representatives

- <u>Redis</u>, MemcachedDB, Riak KV, Hazelcast, Ehcache, Amazon SimpleDB, Berkeley DB, Oracle NoSQL, Infinispan, LevelDB, Ignite, Project Voldemort
- Multi-model: OrientDB, ArangoDB

redis

Key-Value Stores

Representatives



Μ

MEMERCHED





BERKELEY DB



*riakky

Data model

- Documents
 - Self-describing
 - Hierarchical tree structures (JSON, XML, ...)
 - Scalar values, maps, lists, sets, nested documents, ...
 - Identified by a unique identifier (key, ...)
- Documents are organized into collections

Query patterns

- Create, update or remove a document
- Retrieve documents according to complex query conditions Observation
 - Extended key-value stores where the value part is <u>examinable</u>!

Suitable use cases

- Event logging, content management systems, blogs, web analytics, e-commerce applications, ...
 - I.e. for structured documents with similar schema

When not to use

- Set operations involving multiple documents
- Design of document structure is constantly changing
 - I.e. when the required level of granularity would outbalance the advantages of aggregates

Representatives

- <u>MongoDB</u>, Couchbase, Amazon DynamoDB, CouchDB, RethinkDB, RavenDB, Terrastore
- *Multi-model*: **MarkLogic**, **OrientDB**, OpenLink Virtuoso, ArangoDB

Representatives







Couchbase



amazon





terrastore

Data model

- Column family (table)
 - Table is a collection of similar rows (not necessarily identical)
- Row
 - Row is a collection of columns
 - Should encompass a group of data that is accessed together
 - Associated with a unique row key
- Column
 - Column consists of a column name and column value (and possibly other metadata records)
 - Scalar values, but also flat sets, lists or maps may be allowed

Query patterns

- Create, update or remove a row within a given column family
- Select rows according to a row key or <u>simple</u> conditions

Warning

• Wide column stores are <u>not just a special kind of RDBMSs</u> with a variable set of columns!

Suitable use cases

- Event logging, content management systems, blogs, ...
 - I.e. for structured flat data with similar schema

When not to use

- ACID transactions are required
- Complex queries: aggregation (SUM, AVG, ...), joining, ...
- Early prototypes: i.e. when **database design may change** Representatives
 - <u>Apache Cassandra</u>, Apache HBase, Apache Accumulo, Hypertable, Google Bigtable

Representatives











Graph Databases

Data model

- Property graphs
 - Directed / undirected graphs, i.e. collections of ...
 - nodes (vertices) for real-world entities, and
 - relationships (edges) between these nodes
 - Both the nodes and relationships can be associated with additional properties

Types of databases

- Non-transactional = small number of very large graphs
- Transactional = large number of small graphs

Graph Databases

Query patterns

- Create, update or remove a node / relationship in a graph
- Graph algorithms (shortest paths, spanning trees, ...)
- General graph traversals
- Sub-graph queries or super-graph queries
- Similarity based queries (approximate matching)

Representatives

- Neo4j, Titan, Apache Giraph, InfiniteGraph, FlockDB
- Multi-model: OrientDB, OpenLink Virtuoso, ArangoDB

Graph Databases

Suitable use cases

- Social networks, routing, dispatch, and location-based services, recommendation engines, chemical compounds, biological pathways, linguistic trees, ...
 - I.e. simply for graph structures

When not to use

- Extensive batch operations are required
 - Multiple nodes / relationships are to be affected
- Only too large graphs to be stored
 - Graph distribution is difficult or impossible at all





Native XML Databases

Data model

- XML documents
 - Tree structure with nested elements, attributes, and text values (beside other less important constructs)
 - Documents are organized into collections

Query languages

- XPath: XML Path Language (navigation)
- XQuery: XML Query Language (querying)
- XSLT: XSL Transformations (transformation)

- Sedna, Tamino, BaseX, eXist-db
- Multi-model: <u>MarkLogic</u>, OpenLink Virtuoso

Native XML Databases

Representatives







Native XML Database System





RDF Stores

Data model

- RDF triples
 - Components: subject, predicate, and object
 - Each triple represents a statement about a real-world entity
- Triples can be viewed as graphs
 - Vertices for subjects and objects
 - Edges directly correspond to individual statements

Query language

• SPARQL: SPARQL Protocol and RDF Query Language

- Apache Jena, rdf4j (Sesame), Algebraix
- Multi-model: MarkLogic, OpenLink Virtuoso











Data model

- Traditional approach: relational model
- (New) possibilities:
 - Key-value, document, wide column, graph
 - Object, XML, RDF, ...
- Goal
 - Respect the real-world nature of data (i.e. data structure and mutual relationships)

Aggregate structure

- Aggregate definition
 - Data unit with a complex structure
 - Collection of related data pieces we wish to treat as a unit (with respect to data manipulation and data consistency)
- Examples
 - Value part of key-value pairs in key-value stores
 - Document in document stores
 - Row of a column family in wide column stores

Aggregate structure

- Types of systems
 - Aggregate-ignorant: relational, graph
 - It is not a bad thing, it is a feature
 - Aggregate-oriented: key-value, document, wide column
- Design notes
 - No universal strategy how to draw aggregate boundaries
 - Atomicity of database operations:

just a single aggregate at a time

Elastic scaling

- Traditional approach: scaling-up
 - Buying bigger servers as database load increases
- New approach: scaling-out
 - Distributing database data across multiple hosts
 - Graph databases (unfortunately): difficult or impossible at all

Data distribution

- Sharding
 - Particular ways how database data is split into separate groups
- Replication
 - Maintaining several data copies (performance, recovery)

Automated processes

- Traditional approach
 - Expensive and highly trained database administrators
- New approach: automatic recovery, distribution, tuning, ...

Relaxed consistency

- Traditional approach
 - Strong consistency (ACID properties and transactions)
- New approach
 - Eventual consistency only (BASE properties)
 - I.e. we have to make trade-offs because of the data distribution

Schemalessness

- Relational databases
 - Database schema present and strictly enforced
- NoSQL databases
 - Relaxed schema or completely missing
 - Consequences: higher flexibility
 - Dealing with non-uniform data
 - Structural changes cause no overhead
 - However: there is (usually) an implicit schema
 - We must know the data structure at the application level anyway

Open source

• Often community and enterprise versions (with extended features or extent of support)

Simple APIs

Often state-less application interfaces (HTTP)

Current State: Five advantages

- Scaling
 - Horizontal distribution of data among hosts
- Volume
 - High volumes of data that cannot be handled by RDBMS
- Administrators
 - No longer needed because of the automated maintenance
- Economics
 - Usage of cheap commodity servers, lower overall costs
- Flexibility
 - Relaxed or missing data schema, easier design changes

Current State: Five challenges

- Maturity
 - Often still in pre-production phase with key features missing
- Support
 - Mostly open source, limited sources of credibility
- Administration
 - Sometimes relatively difficult to install and maintain
- Analytics
 - Missing support for business intelligence and ad-hoc querying
- Expertise
 - Still low number of NoSQL experts available in the market

Conclusion

The end of relational databases?

- <u>Certainly no</u>
 - They are still suitable for most projects
 - Familiarity, stability, feature set, available support, ...
- However, we should also consider different database models and systems
 - Polyglot persistence = usage of different data stores in different circumstances

Lecture Conclusion

Big Data

• 4V characteristics: volume, variety, velocity, veracity

NoSQL databases

- (New) logical models
 - Core: key-value, wide column, document, graph
 - Non-core: XML, RDF, ...
- (New) principles and features
 - Horizontal scaling, data sharding and replication, eventual consistency, ...

Course Overview

Outline and Objectives

Principles

- Scaling, distribution, consistency
- Transactions, visualization, ...

Technologies

- MapReduce programming model
 - Apache Hadoop
- Data formats
 - XML, JSON, RDF, ...
- NoSQL databases
 - Core: RiakKV, Redis, MongoDB, Cassandra, Neo4j
 - Non-core: XML, RDF
 - Data models, query languages, ...