B4M36DS2: Database Systems 2

http://www.ksi.mff.cuni.cz/~svoboda/courses/2016-1-B4M36DS2/

Lecture 1

# Introduction

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# **Lecture Outline**

## **Big Data**

- Characteristics
- Current trends

### **NoSQL** databases

- Motivation
- Features

Overview of NoSQL database types

Key-value, document, wide column, 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

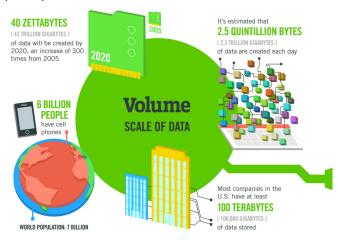
Big Data is high volume, high velocity, and/or high variety information assets that require new forms of processing to enable enhanced decision making, insight discovery and process optimization.

# Where is Big Data?

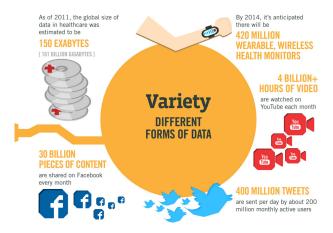
## Sources of Big Data

- 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)



## Variety (Complexity)



## 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



**Velocity** 

ANALYSIS OF STREAMING DATA

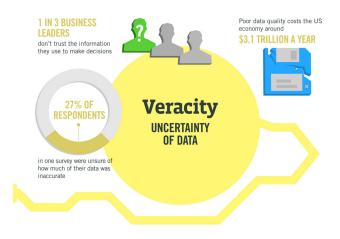


#### 18 9 BILLION NETWORK CONNECTIONS

- almost 2.5 connections per person on earth



### **Veracity** (Uncertainty)



#### **Basic 4V**

- Volume (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)
- Velocity (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

- Value
  - Business value of the data (needs to be revealed)
- Validity
  - Data correctness and accuracy with respect to the intended use
- Volatility
  - Period of time the data is valid and should be maintained

### Data model

```
Instance 	o database 	o table 	o row
```

## Query patterns

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

## Query languages

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

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











**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

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

## **Big Data**

- Volume: terabytes → zettabytes
- Variety: structured → 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
- laaS: Infrastructure as a Service

### Processing paradigms

- OLTP: Online Transaction Processing
- OLAP: Online Analytical Processing
- ...but also...
- RTAP: <u>Real-Time</u> Analytic Processing

## Data assumptions

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

- ⇒ New approach is required
  - Relational databases simply do not follow the current trends

## Key technologies

- Distributed file systems
- NoSQL databases
- MapReduce and other programming models
- Data warehouses
- Grid computing, cloud computing
- 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
- Document stores
- Wide column (column family, column oriented, ...) 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 ⇒ 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

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

# **Key-Value Stores**



















#### 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

- MongoDB, Couchbase, Amazon DynamoDB, CouchDB, RethinkDB, RavenDB, Terrastore
- Multi-model: MarkLogic, OrientDB, OpenLink Virtuoso, ArangoDB



















#### 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 simple 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

 Apache Cassandra, Apache HBase, Apache Accumulo, Hypertable, Google Bigtable











# **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)

- <u>Neo4j</u>, Titan, Apache Giraph, InfiniteGraph, FlockDB
- Multi-model: OrientDB, OpenLink Virtuoso, ArangoDB

# **Graph Databases**

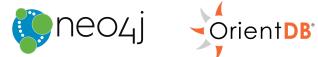
#### Suitable use cases

- 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

# **Graph Databases**













### **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: MarkLogic, 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

### **RDF Stores**









#### 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?

- Certainly no
  - 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

## **Course Overview**

**Conclusion + Outline + Objectives** 

### Basic **principles**

- Distribution, scaling, sharding, replication, consistency
- Parallelism, transactions, visualization, processing of graphs
- Data formats: JSON, YAML, XML, RDF, ...

NoSQL technologies: principles, models, interfaces, languages, ...

- MapReduce: Apache Hadoop
- Core databases: Riak, Redis, MongoDB, Cassandra, Neo4j
- Non-core databases: XML, RDF