TDDD43 Advanced Data Models and Databases

Topic: NoSQL

Olaf Hartig



HOW TO WRITE A CV



"NoSQL"?





http://gook.and.noka.com/gookandnoka/2011/1/27/good htt



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``NoSQL"

- · Some interpretations (without precise definition):
 - "no to SQL"
 - "not only SQL"
 - "not relational"
- · 1998: first used for an RDBMS* that omitted usage of SQL
- 2009: picked up again to name a conference on "open-source, distributed, non-relational databases"
- Since then, "NoSQL database" loosely specifies a class of non-relational DBMSs
 - Relax some requirements of RDBMSs to gain efficiency and scalability for use cases in which RDBMSs are a bad fit

*RDBMS = relational database management system



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Goal of the lecture

What are key characteristics of such systems?

What do databases supported by these systems look like?

What can you do with these databases? (in comparison to the databases supported by RDBMSs)

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Relational Database Management Systems

• Well-defined formal foundations (relational data model)

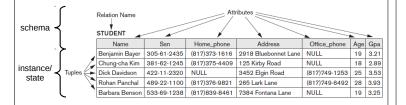


figure from "Fundamentals of Database Systems" by Elmasri and Navathe, Addison Wesl

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Relational Database Management Systems

- Well-defined formal foundations (relational data model)
- SQL powerful declarative language
 - querying
 - data manipulation
 - database definition
- Support of transactions with ACID properties (Atomicity, Consistency preservation, Isolation, Durability)
- Established technology (developed since the 1970s)
 - many vendors
 - highly mature systems
 - experienced users and administrators

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Business world has evolved

- Organizations and companies (whole industries) shift to the digital economy powered by the Internet
- Central aspect: new IT applications that allow companies to run their business and to interact with costumers
 - Web applications
 - Mobile applications
 - Connected devices ("Internet of Things")



mage source: https://pixabay.com/en/technology-information-digital-2082642

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New Challenges for Database Systems

- Increasing numbers of concurrent users/clients
 - tens of thousands, perhaps millions
 - globally distributed
 - expectations: consistently high performance and 24/7 availability (no downtime)
- Different types of data
 - huge amounts (generated by users and devices)
 - data from different sources together
 - frequent schema changes or no schema at all
 - semi-structured and unstructured data
- Usage may change rapidly and unpredictably



Image source: https://www.flickr.com/photos/groucho/552336927

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Scalability

- Data scalability: system can handle growing amounts of data without losing performance
- Read scalability: system can handle increasing numbers of read operations without losing performance
- Write scalability: system can handle

increasing numbers of write operations without losing performance

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Vertical Scalability vs. Horizontal Scalability

· Vertical scalability ("scale up")

 Add resources to a server (e.g., more CPUs, more memory, more or bigger disks)



- Horizontal scalability ("scale out")
 - Add nodes (more computers) to a distributed system

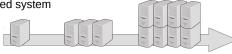


Image source: https://pixabay.com/en/server-web-network-computer-567943

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NoSQL: BASE rather than ACID

- Idea: by giving up ACID guarantees, one can achieve much higher performance and scalability
- · Basically Available
 - system available whenever accessed, even if parts of it unavailable
- Soft state
 - the distributed data does not need to be in a consistent state at all times
- **E**ventually consistent
 - state will become consistent after a certain period of time
- BASE properties suitable for applications for which some inconsistency may be acceptable



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Typical* Characteristics of NoSQL Systems

- Ability to scale horizontally over many commodity servers with high performance, availability, and fault tolerance
 - achieved by giving up ACID guarantees
 - and by partitioning and replication of data
- Non-relational data model, no requirements for schemas

*Attention, there is a *broad variety* of such systems and not all of them have these characteristics to the same degree



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Data Models

- · Key-value model
- · Document model
- · Wide-column models
- · Graph database models

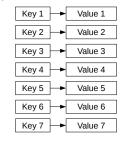
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Key-Value Stores: Data Model

- · Database is simply a set of key-value pairs

 - keys are uniquevalues of arbitrary data types
- · Values are opaque to the system



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Example

· Assume a relational database consisting of a single table:

User	<u>login</u>	name	website	twitter
	alice12	Alice	http://alice.name/	NULL
	bob_in_se	Bob	NULL	@TheBob
	charlie	Charlie	NULL	NULL

• How can we capture this data in the key-value model?



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· How can we capture this data in the key-value model?



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Example

· Let's add another table:

Fav	user	favorite
	alice12	bob_in_se
	alice12	charlie
website		twitter

User	<u>login</u>	name	website	twitter
	alice12	Alice	http://alice.name/	NULL
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alice12

bob_in_se

· How can we capture this data in the key-value model?



Key-Value Stores: Querying

- · Only CRUD operations in terms of keys
 - CRUD: create, read, update, delete
 - put(key, value); get(key); delete(key)
- · No support for value-related queries
 - Recall that values are opaque to the system (i.e., no secondary index over values)
- · Accessing multiple items requires separate requests
 - Beware: often no transactional capabilities



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- Accessing multiple items requires separate requests
 - Beware: often no transactional capabilities
- Advantage of these limitations: partition the data based on keys ("horizontal partitioning", also called "sharding") and distributed processing can be very efficient



Example (cont'd)

- Assume we try to find all users for whom Bob is a favorite
- · It is possible (how?), but very inefficient
- · What can we do to make it more efficient?



Example (cont'd)

- · Assume we try to find all users for whom Bob is a favorite
- It is possible (how?), but very inefficient
- What can we do to make it more efficient?
 - Add redundancy (downsides: more space needed, updating becomes less trivial and less efficient)



Examples of Key-Value Stores

- In-memory key-value stores
 - Memcached
 - Redis



- · Persistent key-value stores
 - Berkeley DB
 - Voldemort
 - RiakDB







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Data Models

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- Document model



- · Wide-column models
- · Graph database models

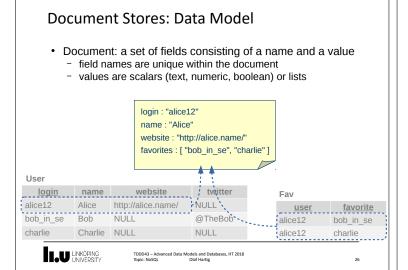


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Document Stores: Data Model

- · Document: a set of fields consisting of a name and a value
 - field names are unique within the document
 - values are scalars (text, numeric, boolean) or lists
 - in some systems, values may also be other documents



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Document Stores: Data Model

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- · Database is a set of documents (or multiple such sets)
 - each document additionally associated with a unique identifier (typically system-generated)
 - schema free: different documents may have different fields



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Collection: Users

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 - grouping of documents into separate sets (called "domains" or "collections")
- · Partitioning based on collections and/or on document IDs
- Secondary indexes over fields in the documents possible
 - different indexes per domain/collection of documents

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Document Stores: Querying

- · Querying in terms of conditions on document content
- Queries expressed in terms of program code using an API or in a system-specific query language

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Document Stores: Querying

- Querying in terms of conditions on document content
- · Queries expressed in terms of program code using an API or in a system-specific query language
- Examples (using MongoDB's query language):
 - Find all docs in collection *Users* whose *name* field is "Alice" db.Users.find({name: "Alice"})
 - Find all docs in collection *Users* whose *age* is greater than 23 db.Users.find({age: {\$qt: 23}})
 - Find all docs about Users who favorite Bob

db.Users.find({favorites: {\$in: ["bob_in_se"]}})

login: "alice12" name : "Alice" website: "http://alice.name/" TDDD43 - Advanced Data Models and Databases, HT 2018 FAVORITIES: ["bob_in_se", "charlie" Topic: NOSQ.

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 - Find all docs about Users who favorite Bob db.Users.find({favorites: {\$in: ["bob_in_se"]}})
- · However, no cross-document queries (like joins)
 - have to be implemented in the application logic



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Examples of Document Stores

· Amazon's SimpleDB



CouchDB



Couchbase



MongoDB





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Data Models

- · Key-value model
- Document model
- Wide-column models
- Graph database models

also called column-family models

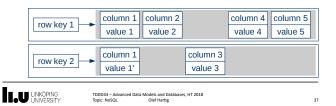
extensible-record models

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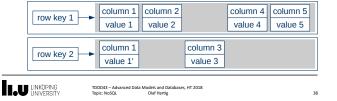
Wide-Column Stores: Data Model (Basic)

- Database is a set of "rows" each of which ...
 - ... has a unique key, and
 - ... a set of key-value pairs (called "columns")
- Schema free: different rows may contain different columns



Wide-Column Stores: Data Model

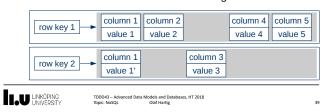
- · Database is a set of "rows" each of which ...
 - ... has a unique key, and
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- Schema free: different rows may contain different columns
- · Like a single, very wide relation (SQL table) that is a) extensible, b) schema-free, and c) potentially sparse

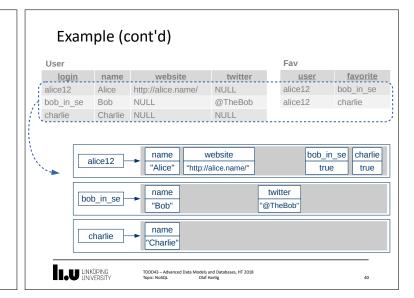




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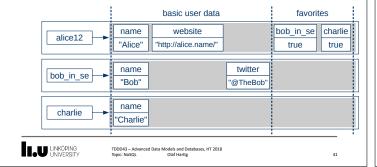
 a) extensible, b) schema-free, and c) potentially sparse
- · Like the document model without nesting





Wide-Column Stores: Data Model (cont'd)

- Columns may be grouped into so called "column families"
 - Hence, values are addressed by (row key, column family, column key)



Wide-Column Stores: Data Model (cont'd)

- Columns may be grouped into so called "column families"
 - Hence, values are addressed by (row key, column family, column key)
- Data may be partitioned ...
 - ... based on row keys (horizontal partitioning),
 - ... but also based on column families (vertical partitioning),
 - ... or even on both

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Wide-Column Stores: Data Model (cont'd)

- Columns may be grouped into so called "column families"
 Hence, values are addressed by
 - (row key, column family, column key)
- Data may be partitioned ...
 - ... based on row keys (horizontal partitioning),
 - ... but also based on column families (vertical partitioning),
 - ... or even on both
- Secondary indexes can be created over arbitrary columns

Wide-Column Stores: Querying

- · Querying in terms of keys or conditions on column values
- Queries expressed in a system-specific query language or in terms of program code using an API
 - Conceptually similar to queries in document stores
- No joins
 - Again, must be implemented in the application logic

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Examples of Wide-Column Stores

- · Basic form (no column families):
 - Amazon SimpleDB



- Amazon DynamoDB
- With column families:Google's BigTable



- Hadoop HBase



- Apache Cassandra





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 NoSQL systems support non-relational data models (key-value, document, wide-column, graph)

support for semi-structured and unstructured data

· NoSQL systems provide high (horizontal) scalability

with high performance, availability, and fault tolerance

• data partitioning (effective due to data model limitations)

limited query capabilities (no joins!)

See next lecture

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Summary

schema free

- achieved by:

· data replication

Data Models

· Key-value model

Document modelWide-column models

Graph database models

Data Models

- · Key-value model
- · Document model
- · Wide-column models
- · Graph database models

There are also multi-model NoSQL stores

Examples:

• OrientDB (key-value, documents, graph)



• ArangoDB (key-value, documents, graph)



• Cosmos DB (key-value, documents, wide-column, graph)



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• giving up consistency requirements

Reading Material

- NoSQL Databases: a Survey and Decision Guidance. F. Gessert. *Blog post*, August 2016.
- Data Management in Cloud Environments: NoSQL and NewSQL Data Stores. K. Grolinger et al. Journal of Cloud Computing 2:22, 2013
 - Considers not only NoSQL but also NewSQL systems
 - Includes comprehensive comparison of various systems over a large number of dimension
- Scalable SQL and NoSQL Data Stores. R. Cattell. ACM SIGMOD Record 39(4), 2011
 - More detailed overview of several example systems
- · NoSQL Databases. C. Strauch. Lecture Notes, 2012
 - Comprehensive discussion of several example systems



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