Big Data (NoSQL Databases)

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Smart Software System Laboratory

"Big data is at the foundation of all the megatrends that are happening today, from social to mobile to cloud to gaming."

- Chris Lynch, Vertica Systems





- Relational databases mainstay of business
- Web-based applications caused spikes
- Explosion of social media sites (Facebook, Twitter) with large data needs
- rise of cloud-based solutions such as Amazon S3 (simple storage solution)
- Hooking RDBMS to web-based application becomes trouble



Issues with Scaling up

- Best way to provide ACID and Rich Query Model is to have the dataset on a single machine
- Limits to scaling up (or vertical scaling: make a "single" machine more powerful) dataset is just too big!
- Scaling out (or horizontal scaling: adding more smaller / cheaper servers)
 is a better choice
- Different approaches for horizontal scaling (multi-node database):
 - Master / Slave
 - Sharding (partitioning)





Master / Slave

- All writes are written to the master.
- All reads performed against the replicated slave databases
- Critical reads may be incorrect as writes may not have been propagated down
- Large datasets can pose problems as master needs to duplicate data to slaves



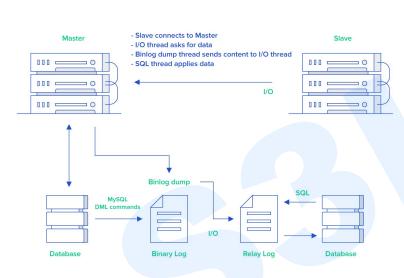
Sharding (Partitioning)

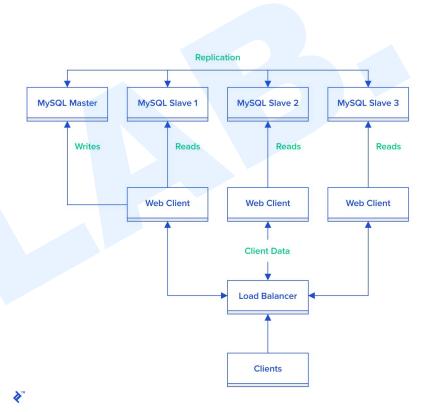
- Scales well for both reads and writes
- Not transparent, application needs to be partition-aware
- Can no longer have relationships/joins across partitions
- Loss of referential integrity across shards



Scaling out RDBMS

Master / Slave







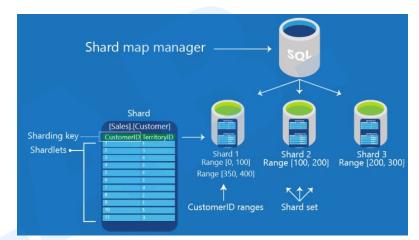


Scaling out RDBMS

Sharding (Partitioning)

Original Table

CUSTOMER ID	FIRST NAME	LAST NAME	FAVORITE COLOR
1	TAEKO	OHNUKI	BLUE
2	O.V.	WRIGHT	GREEN
3	SELDA	BAĞCAN	PURPLE
4	JIM	PEPPER	AUBERGINE



Vertical Partitions

VP1			
CUSTOMER ID	FIRST NAME	LAST NAME	
1	TAEKO	OHNUKI	
2	O.V.	WRIGHT	
3	SELDA	BAĞCAN	
4	JIM	PEPPER	

VP2			
CUSTOMER ID	FAVORITE COLOR		
1	BLUE		
2	GREEN		
3	PURPLE		
4	AUBERGINE		

Horizontal Partitions

HP1

CUSTOMER ID	FIRST NAME	LAST NAME	FAVORITE COLOR
1	TAEKO	OHNUKI	BLUE
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HP2

CUSTOMER ID	FIRST NAME	LAST NAME	FAVORITE COLOR
3	SELDA	BAĞCAN	PURPLE
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The other ways

- Multi-Master replication
- INSERT only, not UPDATES/DELETES
- No JOINs, thereby reducing query time
 - This involves de-normalizing data
- In-memory databases







- This name stands for Not Only SQL
- The term NOSQL was introduced by Carl Strozzi in 1998 to name his file-based database
- It was again re-introduced by Eric Evans when an event was organized to discuss open source distributed databases
 - Eric states that "... but the whole point of seeking alternatives is that you need to solve a
 problem that relational databases are a bad fit for. ..."





Key features (Advantages)

- non-relational
- don't require schema
- data are replicated to multiple nodes (so, identical & fault-tolerant)
 and can be partitioned:

 Redis
 - down nodes easily replaced
 - o no single point of failure
- horizontal scalable





Key features (Advantages)

- cheap, easy to implement (open-source)
- massive write performance
- fast key-value access









Disadvantages

- Don't fully support relational features
 - o no join, group by, order by operations (except within partitions)
 - no referential integrity constraints across partitions
- No declarative query language (e.g., SQL) more programming
- Relaxed ACID (see CAP theorem) fewer guarantees
- No easy integration with other applications that support SQL





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- Three major papers were the "seeds" of the NOSQL movement:
 - BigTable (Google)
 - DynamoDB (Amazon)
 - Ring partition and replication
 - Gossip protocol (discovery and error detection)
 - Distributed key-value data stores
 - Eventual consistency
 - CAP Theorem





- Large datasets, acceptance of alternatives, and dynamically-typed data has come together in a "perfect storm"
- Not a backlash against RDBMS
- SQL is a rich query language that cannot be rivaled by the current list of NOSQL offerings

CAP Theorem

Suppose three properties of a distributed system (sharing data)



- <u>C</u>onsistency:
 - Reads and writes are always executed atomically and are strictly consistent (linearizable). Put differently, all clients have the same view on the data at all times.
- Availability:
 - Every non-failing node in the system can always accept read and write requests by clients and will eventually return with a meaningful response, i.e. not with an error message.
- Partition-tolerance:
 - system properties (consistency and/or availability) hold even when network failures prevent some machines from communicating with others. A system can continue to operate in the presence of a network partitions

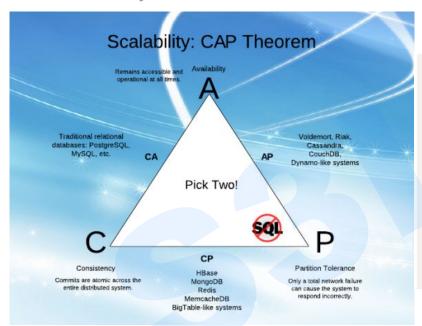


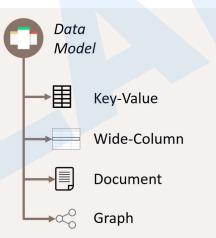


- Brewer's CAP Theorem:
 - For any system sharing data, it is "impossible" to guarantee simultaneously all of these three properties
 - You can have at most two of these three properties for any shared-data system
- Very large systems will "partition" at some point:
 - That leaves either C or A to choose from (traditional DBMS prefers C over A and P)
 - In almost all cases, you would choose A over C (except in specific applications such as order processing)



Consistency





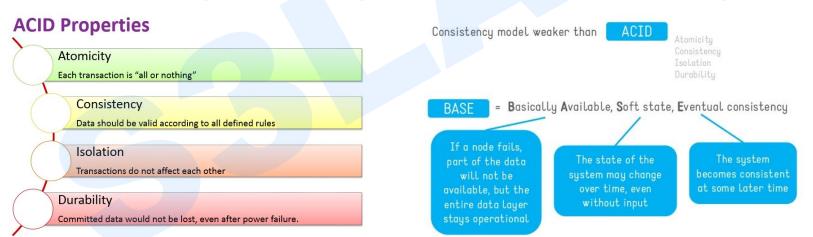






Consistency

- Have 2 types of consistency:
 - Strong consistency ACID (Atomicity, Consistency, Isolation, Durability)
 - Weak consistency BASE (<u>Basically Available</u> <u>Soft-state</u> <u>Eventual consistency</u>)



CAP Theorem



Consistency

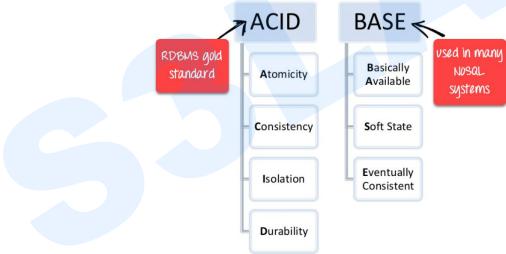
- A consistency model determines rules for visibility and apparent order of updates
- Example:
 - Row X is replicated on nodes M and N
 - Client A writes row X to node N
 - Some period of time t elapses
 - Client B reads row X from node M
 - Open client B see the write from client A?
 - Consistency is a continuum with tradeoffs
 - For NOSQL, the answer would be: "maybe"
 - CAP theorem states: "strong consistency can't be achieved at the same time as availability and partition-tolerance"





Consistency

- Cloud computing
 - ACID is hard to achieve, moreover, it is not always required, e.g. for blogs, status updates, product listings, etc.



NoSQL



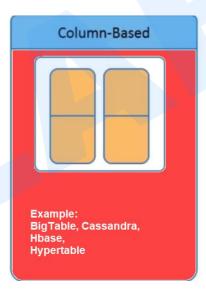
- "No-schema" is a common characteristics of most NOSQL storage systems
- Provide "flexible" data types
- Other or additional query languages than SQL
- Distributed horizontal scaling
- Less structured data
- Supports big data

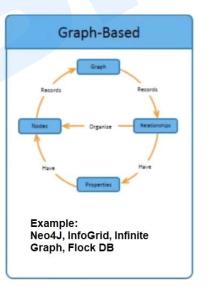






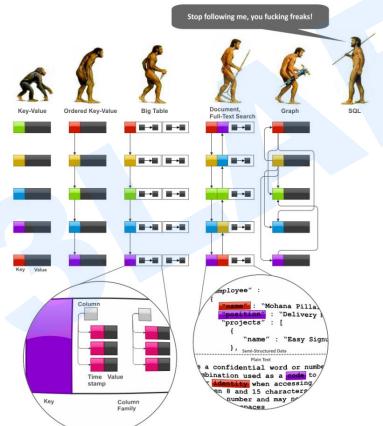








NoSQL Categories







- Focus on scaling to huge amounts of data
- Designed to handle massive load
- Based on Amazon's dynamo paper
- Data model: (global) collection of Key-value pairs
- Dynamo ring partitioning and replication
- Example: (DynamoDB)
 - o items having one or more attributes (name, value)
 - An attribute can be single-valued or multivalued like set.
 - o items are combined into a table









- Basic API access:
 - o get(key): extract the value given a key
 - o put(key, value): create or update the value given its key
 - delete(key): remove the key and its associated value
 - execute(key, operation, parameters): invoke an operation to the value (given its key) which is a special data structure (e.g. List, Set, Map etc)





• Pros:

- very fast
- very scalable (horizontally distributed to nodes based on key)
- simple data model
- eventual consistency
- fault-tolerance

Cons:

Can't model more complex data structure such as objects

NoSQL Categories



Key-value

Name	Producer	Data model	Querying
SimpleDB	Amazon	set of couples (key, {attribute}), where attribute is a couple (name, value)	restricted SQL; select, delete, GetAttributes, and PutAttributes operations
Redis	Salvatore Sanfilippo	set of couples (key, value), where value is simple typed value, list, ordered (according to ranking) or unordered set, hash value	primitive operations for each value type
Dynamo	Amazon	like SimpleDB	simple get operation and put in a context
Voldemort	LinkedIn	like SimpleDB	similar to Dynamo



employee_id	first_name	last_name	address
1	John	Doe	New York
2	Benjamin	Button	Chicago
3	Mycroft	Holmes	London

employee:\$employee id:\$attribute name = \$value

```
employee:1:first_name = "John"
employee:1:last_name = "Doe"
employee:1:address = "New York"

employee:2:first_name = "Benjamin"
employee:2:last_name = "Button"
employee:2:address = "Chicago"

employee:3:first_name = "Mycroft"
employee:3:last_name = "Holmes"
employee:3:address = "London"
```





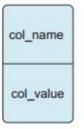
Column-based

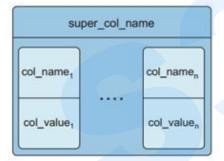
- Based on Google's BigTable paper
- Like column oriented relational databases (store data in column order) but with a twist
- Tables similarly to RDBMS, but handle semi-structured
- Data model:
 - Collection of Column Families
 - Column family = (key, value) where value = set of related columns (standard, super)
 - o indexed by row key, column key and timestamp

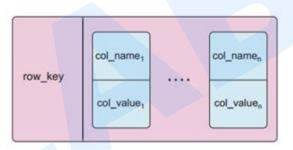


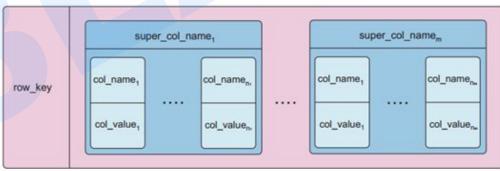


Column-based





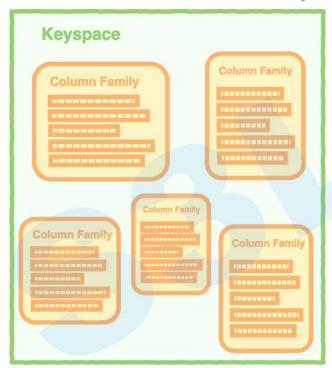


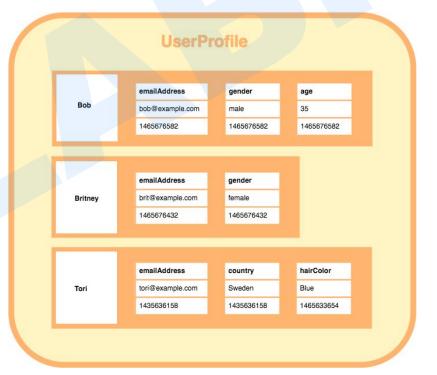




NoSQL Categories

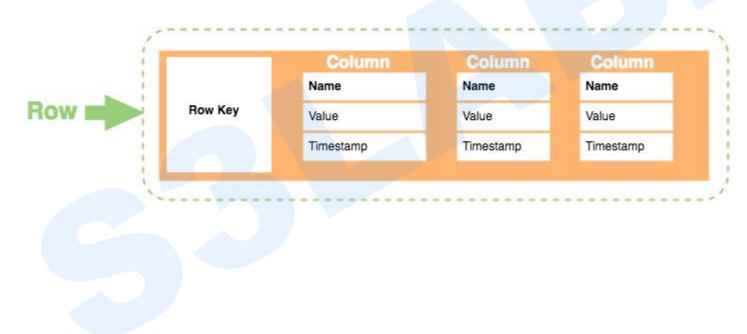
Keyspace ~ Schema, Column Family ~ Table







Row structure





Column-based

- One column family can have variable numbers of columns
- Cells within a column family are sorted "physically"
- Very sparse, most cells have null values
- Comparison: RDBMS vs column-based NOSQL
 - Query on multiple tables
 - RDBMS: must fetch data from several places on disk and glue together
 - Column-based NOSQL: only fetch column families of those columns that are required by a query (all columns in a column family are stored together on the disk, so multiple rows can be retrieved in one read operation data locality)



NoSQL Categories

Column-based

Operation	Column-Oriented Database	Row-Oriented Database
Aggregate Calulation of Single Column e.g. sum(price)	✓ fast	slow
Compression	✓ Higher. As stores similar data together	
Retrieval of a few columns from a table with many columns	✓ Faster	has to skip over unnecessary data
Insertion/Updating of single new record	Slow	✓ Fast
Retrieval of a single record	Slow	✓ Fast



Column-based

Example: (Cassandra column family--timestamps removed for simplicity)

```
UserProfile = {
     Cassandra
                 = {
                          emailAddress:"casandra@apache.org", age:"20"
     TerryCho
                          emailAddress:"terry.cho@apache.org", gender:"male"
     Cath
                          emailAddress:"cath@apache.org",
                          age:"20",gender:"female",address:"Seoul"
```



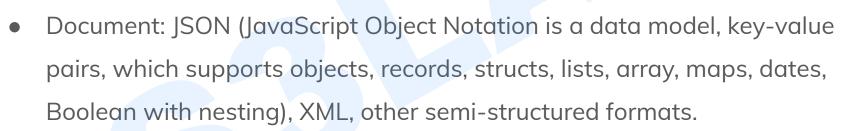
Column-based

Name	Producer	Data model	Querying				
BigTable	Google	set of couples (key, {value})	selection (by combination of row, column, and time stamp ranges)				
HBase	Apache	groups of columns (a BigTable clone)	JRUBY IRB-based shell (similar to SQL)				
Hypertable	Hypertable	like BigTable	HQL (Hypertext Query Language)				
CASSANDRA	Apache (originally Facebook)	columns, groups of columns corresponding to a key (supercolumns)	simple selections on key, range queries, column or columns ranges				
PNUTS	IUTS Yahoo (hashed or or arrays, flexibl		selection and projection from a single table (retrieve an arbitrary single record by primary key, range queries, complex predicates, ordering, top-k)				



Document-based

- Can model more complex objects
- Inspired by Lotus Notes
- Data model: collection of documents







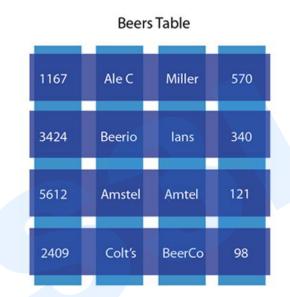
Document-based

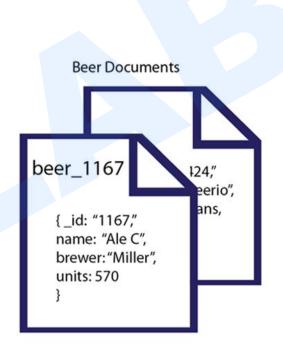
Example: (MongoDB) document

```
Name:"Jaroslav",
Address: "Malostranske nám. 25, 118 00 Praha 1",
Grandchildren: {Claire: "7", Barbara: "6", "Magda: "3", "Kirsten: "1", "Otis: "3", Richard: "1"}
Phones: [ "123-456-7890", "234-567-8963" ]
             object
                                                             number
                                                             object
                                                              true
```



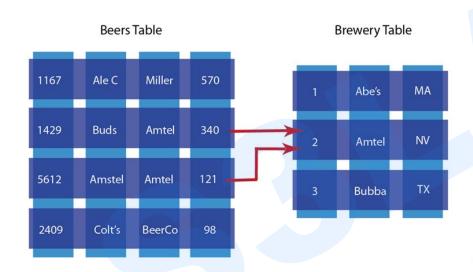
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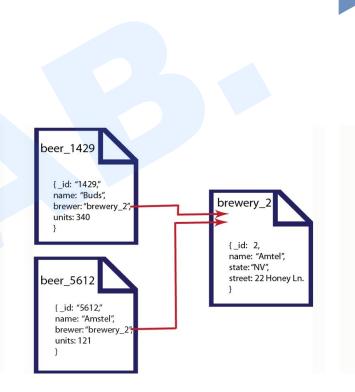






Document-based





Big Data

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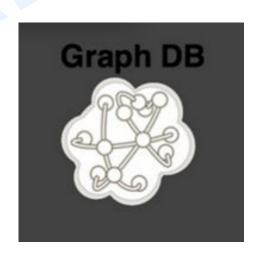


Document-based

Name	Producer	Data model	Querying		
MongoDB	10gen	object-structured documents stored in collections; each object has a primary key called ObjectId	manipulations with objects in collections (find object or objects via simple selections and logical expressions, delete, update,)		
Couchbase	Couchbase ¹	document as a list of named (structured) items (JSON document)	by key and key range, views via Javascript and MapReduce		



- Focus on modeling the structure of data (interconnectivity)
- A graph is composed of two elements: a node and a relationship.
- Scales to the complexity of data
- Inspired by mathematical Graph Theory (G=(E,V))
- Data model:
 - (Property Graph) nodes and edges
 - Nodes may have properties (including ID)
 - Edges may have labels or roles
 - Key-value pairs on both

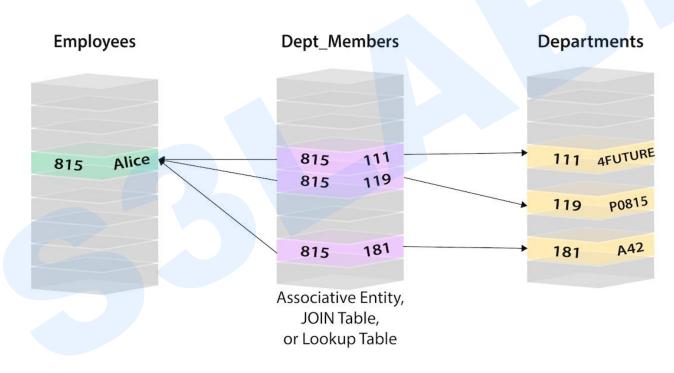




- Interfaces and query languages vary
- Single-step vs path expressions vs full recursion
- Example:
 - Neo4j, FlockDB, Pregel, InfoGrid ...

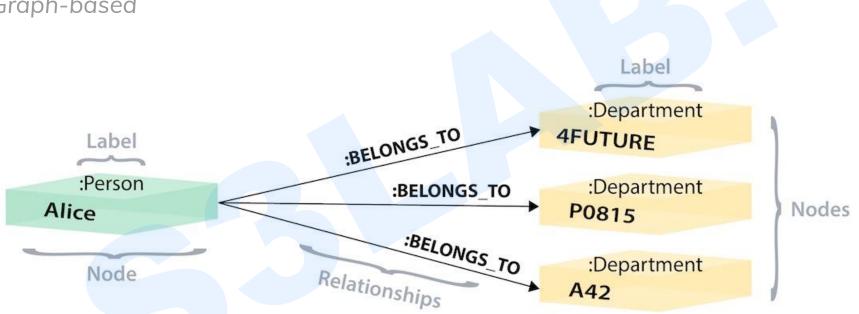












Comparison



Attributes NoSQL Databases										
Database model		Document-Stored		Wide-Column Stored			Key-Value Stored		Graph- oriente d	
	Features	MongoDB	CouchDB	DynamoBD	HBase	Cassandra	Accumulo	Redis	Riak	Neo4j
Design & Features	Data storage	Volatile memory File System	Volatile memory File System	SSD	HDFS		Hadoop	Volatile memory File System	Bitcask LevelDB Volatile memory	File System Volatile memory
	Query language	Volatile memory File System	JavaScript Memcached- protocol	API calls	API calls REST XML Thrift	API calls CQL Thrift		API calls	HTTP JavaScript REST Erlang	API calls REST SparQL Cypher Tinkerpo p Gremlin
	Protocol	Custom, binary (BSON)	HTTP, REST		HTTP/REST Thrift	Thrift & custom binary CQL3	Thrift	Telnet-like	HTTP, REST	HTTP/RES Tembedd ing in Java
	Conditional entry updates	Yes		Yes		No		No		
	MapReduce	Yes	Yes	Yes	Yes	Yes	Yes	No		No
	Unicode	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
	TTL for Entries	Yes	Yes	No	Yes	Yes	Yes	Yes	Yes	
	Compression	Yes	Yes		Yes	Yes	Yes	Yes	Yes	
	Integrity model	BASE	MVCC	ASID	Log Replicati on	BASE	MVCC			ASID
	Atomicity	Conditional	Yes	Yes		Yes	Condition al	Yes	No	Yes
.≧	Consistency	Yes	Yes	Yes	Yes	Yes	Yes	Yes	No	Yes
teg	Isolation	No	Yes	Yes	No	No	4	Yes	Yes	Yes
Int	Durability (data storage)	Yes	Yes	Yes		Yes	Yes	Yes	*	Yes
	Transactions	No	No	No	Yes	No	Yes	Yes	No	Yes
	Referential integrity	No		No	No	No		Yes		Yes
	Revision control	No	Yes	Yes	Yes	No	Yes	No	Yes	No
	Secondary Indexes	Yes		No	Yes	Yes	Yes			
ing	Composite keys	Yes		Yes	Yes	Yes	Yes		Yes	=
Indexing	Full text search	No	No	No	No	No	Yes	No	Yes	Yes
l	Geospatial Indexes			No	No	No	Yes			Yes
	Graph support	No	No	No	No	No	Yes	No	Yes	Yes
	Horizontal scalable	Yes	Yes	Yes	Yes	Yes	Yes		Yes	No
_	Replication	Yes	Yes	Yes	Yes	Yes	Yes		Yes	Yes
Distribution	Replication mode	Master- Slave- Replica Replication	Master- Slave Replicatio n		Master- Slave Replicati on	Master- Slave Replicatio n	-	Master- Slave Replicati on	Multi- master replicati on	-
	Sharding	Yes	Yes	Yes	Yes	Yes	Yes	No	Yes	Yes
	Shared nothing architecture	Yes		Yes	Yes	Yes	-			-
	Value size max.	16MB	20MB	64KB	2TB	2GB	1EB		64MB	
System	Operating system	Cross- platform	Ubuntu Red Hat Windows Mac OS X	Cross- platform	Cross- platform	Cross- platform	NIX 32 entries Operating system	Linux *NIX Mac OS X Window s	Cross- platform	Cross- platfor m
	Programming language	C++	Erlang C++ C Python	Java	Java	Java	Java	C C++	Erlang	Java





Conclusion



- NOSQL database cover only a part of data-intensive cloud applications (mainly Web applications)
- Problems with cloud computing:
 - SaaS (Software as a Service or on-demand software) applications require enterprise-level functionality, including ACID transactions, security, and other features associated with commercial RDBMS technology, i.e. NOSQL should not be the only option in the cloud
 - Hybrid solutions:
 - Voldemort with MySQL as one of storage backend
 - deal with NOSQL data as semi-structured data
 - -> integrating RDBMS and NOSQL via SQL/XML

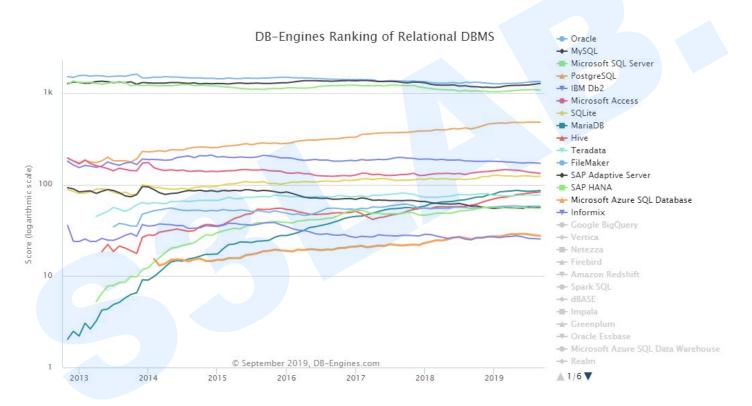




- next generation of highly scalable and elastic RDBMS: NewSQL databases (from April 2011)
 - they are designed to scale out horizontally on shared nothing machines,
 - still provide ACID guarantees,
 - o applications interact with the database primarily using SQL,
 - o the system employs a lock-free concurrency control scheme to avoid user shut down,
 - the system provides higher performance than available from the traditional systems.
- Examples: MySQL Cluster (most mature solution), VoltDB, Clustrix,
 ScalArc, etc.

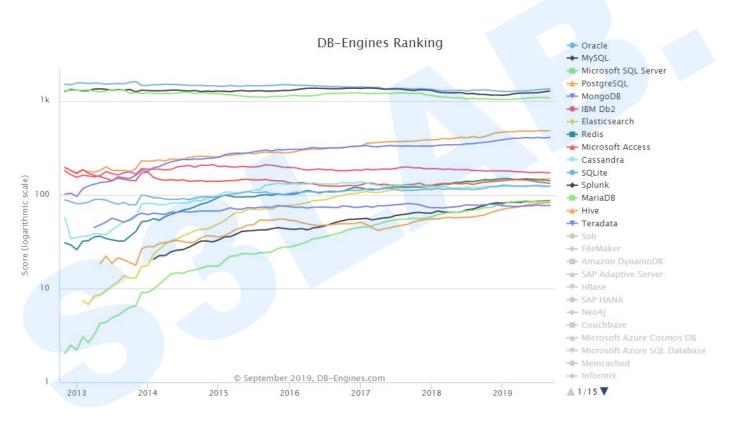






Conclusion





Q & A





Cảm ơn đã theo dõi

Chúng tôi hy vọng cùng nhau đi đến thành công.