### **NoSQL Database Systems:** A Survey and Decision Guidance

<u>Felix Gessert</u>, Wolfram Wingerath, Steffen Friedrich, <u>Norbert Ritter</u> gessert@informatik.uni-hamburg.de June 28, SummerSOC 2016

## 2016 tutorial): slideshare.net/felixgessert

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### Outline



NoSQL Foundations and Motivation

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The NoSQL Toolbox: Common Techniques



NoSQL Systems



- NoSQL: Motivation and Origins
- The 4 Classes of NoSQL Databases:
  - Key-Value Stores
  - Wide-Column Stores
  - Document Stores
  - Graph Databases
- CAP Theorem

# Introduction: Diversity of NoSQL data stores

### How to choose a database system?

#### Many Potential Candidates



### How to choose a database system?

#### Many Potential Candidates



### **NoSQL** Databases

- "NoSQL" term coined in 2009
- Interpretation: "Not Only SQL"
- Typical properties:
  - Non-relational
  - Open-Source
  - Schema-less (schema-free)
  - Optimized for distribution (clusters)
  - Tunable consistency

NoSQL-Databases.org:

Current list has over 150 NoSQL systems Wide Column Store / Column Families

Hadoop / HBasc AFt: Java / any writer, Protocol: any write call, Quey Mehod: MapReduce Java / any coxec, Replication: Mitten in: Java, Concurrence, 1, Mise: Links: 3 Books (J. 2, 3)

Cassandramashcip vasilobi, parilihone ros idor. matoricia siricotar: linear solar obromanoe, na single points of failure, readwite subport across multiple data contras s (logi valiability) sonta APT (Joury) Monod. CRL and Thriffs (robleation) pace-to-pacer without in: Java Construction, Machedica estable factor, lines both cons construction, Machedica estable factor, lines both cons construction, Machedica estable, lines (Journal et al. 1997) indexes, security features. Lines <u>Decumentation</u>, Flancer-<u>Construction</u>, Machedica estable, lines (Journal et al. 1997) indexes, security features. Lines <u>Decumentation</u>, Flancer-

Hypertable API: Thrift (Java, PHP, Perl, Python, Ruby, ctt), Prosocia Thrift (Java, PHP, Perl, Python, Ruby, API, Reglication, HDPS Replication, Chourency, MVCC, Consistency Model: Fully consistent Misc Hip portomance G-- implementation of Google's Bigtable. <u>2</u> Commercial Juscent

Accumulo Accumulo is based on BigErabic and is built on too of Hadro<u>on</u> Tookcorp, and Thirling It features improvements on the Bafasic access in the form of cellbased access control improved compression and a sorveide pregramming mechanism that can modify registrate pairs at various points in the case management process.

Amazon SimpleDB Mise: not open source / part of AVS, <u>Book</u> (will be outperformed by DynamoDB ?!) <u>Cloudata</u> Google's Big table clone like HBase, <u>a Article</u>

Clouders Professional Software & Services based on Haddoo. Haddoo. HPCC from <u>Lexisticuis, info, anticle</u> Stratosphere (research system) massive parallel & flexible.

Stratosphere (research system) massive parallel a flexible execution, MR generalization and extendion (paper, poster). (Openheptune, (base, KDI) Document Store

#### MongoDB APE BSON, Protocol: C, Quey Method: dynamic object-based language & MapReduce, Replication: Master Slave & Auto-Sharding Writen

Rolleaton Master Slave a Auto-Sharefing Witen in C-4, concurrency Update in Place Mis-Intervention, Grieffs, Freenare + Commercial Uncerse Unit - Risk Index - Company Biosociecaerch AP, REST and many linearisation Surging automatic and configurable uniterin Javas Misc scheming and Suborgi.

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CouchDB API: JSON, Protocol: REST, Query Michael: MapReduceR of JavaScript Funes, Replication: Master Master, Written in: Erlang, Concurrency, MVCC, Mise:

Links: <u>> 3 CouchOB books</u>, <u>> Couch Lounge</u> (partitioning / clusering), <u>> Dr. Dobbs</u>

Rethinking AFF protobul-based, Quey Nened: unified charable duery language (inc.). (Ditk, sub-dueries, MapReduce, GroupedkapReduce) Realization Syne, and Asyne Master Slave with portable acknowledgements. Sharing pulled range-based, When in C-A., Concurrent, MYCCL Mice legandured storage regine with concurrent incremental grasge consistor

RavenDB Net solution. Provides HTTP/JSON access. LING queries a Sharding supported. <u>a Mise</u>

MarkLogic Server (Itenare-commercial AFE (SON, Xall, Java Fotocols: HTTP, RESTQuey lichos: Fuil Text Scarch, XPath, XQuey, Range, Goospatial Million in: C+- Concurreny: Shared-nothing cluster, MVCC Mise: Petaplosciable, (double), ACD sharedons, subsharing, failour, masto slave replication, server with ACLs. Developer Community

Custore of the second for a second se

ThruDB (picasc help provide more facts!) Uses Apache <u>Thrift</u> to internate multiple backend databases as BerkeleyOB, Olsk, MySQL, 53.

Terrastore APE Java & http://rotocol.http://anguage Java, Gorying: Range queries, Predicates, Replication Partitioned with consistent hashing Consistency Per-record strict consistency Misc Based on Toracota

JasoB Liphoncipit open source document database written in Java Ton high performance, rung innemory, supports Android. API: JSON, Java Query Nichold: REST Obstas Stylic Query Language, Java fluent Query API Concurrency: Atomice document writtes Indoces:

eventually consistent indexes <u>RaptorDB</u> JS0N based, Document store database with compiled .net map functions and automatic hybrid bimap indexing and LUNQ query filters

SisoDB A Document Store on top of SQL Server. SDB For small online databases, PHP / JSON interface, implemented in PHP.

Gioneb gioneb AF: BSDM, Protocol: C++, Query Method: dynamic queries and map/reduce, Driver: Java, C++, PHP Misc. ACIC compliant, Full static console over pogle: v8 engine, djoneb requirements are submitted by users, and more (Linear, GB) and compared.

### **NoSQL System Classification**

#### Two common criteria:

![](_page_7_Figure_2.jpeg)

### **Key-Value Stores**

- Data model: (key) -> value
- Interface: CRUD (Create, Read, Update, Delete)

![](_page_8_Figure_3.jpeg)

Examples: Amazon Dynamo (AP), Riak (AP), Redis (CP)

### Wide-Column Stores

- Data model: (rowkey, column, timestamp) -> value
- Interface: CRUD, Scan

![](_page_9_Figure_3.jpeg)

 Examples: Cassandra (AP), Google BigTable (CP), HBase (CP)

### **Document Stores**

- Data model: (collection, key) -> document
- Interface: CRUD, Querys, Map-Reduce

![](_page_10_Figure_3.jpeg)

 Examples: CouchDB (AP), Amazon SimpleDB (AP), MongoDB (CP)

### **Graph Databases**

- Data model: G = (V, E): Graph-Property Modell
- Interface: Traversal algorithms, querys, transactions

![](_page_11_Figure_3.jpeg)

 Examples: Neo4j (CA), InfiniteGraph (CA), OrientDB (CA)

### **Graph Databases**

Data model: G = (V, E): Graph-Property Modell

![](_page_12_Figure_2.jpeg)

### Soft NoSQL Systems Not Covered Here

![](_page_13_Picture_1.jpeg)

#### Search Platforms (Full Text Search):

- No persistence and consistency guarantees for OLTP
- *Examples*: ElasticSearch (AP), Solr (AP)

#### **Object-Oriented Databases:**

- Strong coupling of programming language and DB
- *Examples*: Versant (CA), db4o (CA), Objectivity (CA)

![](_page_13_Picture_8.jpeg)

#### XML-Databases, RDF-Stores:

- Not scalable, data models not widely used in industry
- Examples: MarkLogic (CA), AllegroGraph (CA)

### CAP-Theorem

![](_page_14_Figure_1.jpeg)

Only 2 out of 3 properties are achievable at a time:

- Consistency: all clients have the same view on the data
- Availability: every request to a nonfailed node most result in correct response
- Partition tolerance: the system has to continue working, even under arbitrary network partitions

Eric Brewer, ACM-PODC Keynote, Juli 2000

![](_page_14_Picture_7.jpeg)

Gilbert, Lynch: Brewer's Conjecture and the Feasibility of Consistent, Available, Partition-Tolerant Web Services, SigAct News 2002

![](_page_15_Picture_0.jpeg)

Data Models and CAP provide high-level classification.

### But what about **fine-grained requirements**, e.g. query capabilites?

![](_page_15_Picture_3.jpeg)

### Outline

![](_page_16_Figure_1.jpeg)

NoSQL Foundations and Motivation

	-	
-		

The NoSQL Toolbox: Common Techniques

![](_page_16_Picture_5.jpeg)

NoSQL Systems

![](_page_16_Picture_7.jpeg)

- Techniques for Functional and Non-functional Requirements
  - Sharding
  - Replication
  - Storage Management
  - Query Processing

![](_page_17_Figure_0.jpeg)

![](_page_18_Figure_0.jpeg)

![](_page_19_Figure_0.jpeg)

Sharding (aka Partitioning, Fragmentation)

Horizontal distribution of data over nodes

![](_page_20_Figure_2.jpeg)

Partitioning strategies: Hash-based vs. Range-based
Difficulty: Multi-Shard-Operations (join, aggregation)

### Sharding

#### Hash-based Sharding

- Hash of data values (e.g. key) d MongoDB, Riak,
- **Pro**: Even distribution
- Contra: No data locality

#### Range-based Sharding

- Assigns ranges defined over field
- Pro: Enables Range Scans and \$
- Contra: Repartitioning/balancir

#### **Entity-Group Sharding**

- Explicit data co-location for sin
- Pro: Enables ACID Transactions
- Contra: Partitioning not easily d

#### Implemented in

MongoDB, Riak, Redis, Cassandra, Azure Table, Dvnamo

Implemented in

BigTable, HBase, DocumentDB Hypertable, MongoDB, RethinkDB, Espresso

#### Implemented in

G-Store, MegaStore, Relation Cloud, Cloud SQL Server

David J DeWitt and Jim N Gray: "Parallel database systems: The future of high performance database systems," Communications of the ACM, volume 35, number 6, pages 85–98, June 1992.

![](_page_22_Figure_0.jpeg)

### Replication

#### Stores N copies of each data item

![](_page_23_Figure_2.jpeg)

Consistency model: synchronous vs asynchronous
Coordination: Multi-Master, Master-Slave

![](_page_23_Picture_4.jpeg)

Özsu, M.T., Valduriez, P.: Principles of distributed database systems. Springer Science & Business Media (2011)

### **Replication: When**

### Asynchronous (lazy)

- Writes are acknowledged immdediately
- Performed through *log shipping* or *update propagation*
- Pro: Fast writes, no coordination needed
- Contra: Replica data potentially stale (*inconsistent*)

#### Synchronous (eager)

- The node accepting writes synchronously propagates updates/transactions before acknowledging
- **Pro**: Consistent
- Contra: needs a commit protocol (more roundtrips), unavaialable under certain network partitions

![](_page_24_Picture_10.jpeg)

### **Replication: When**

#### Asynchronous (lazy)

- Writes are acknowledged imn
- Performed through *log shippi*.
- Pro: Fast writes, no coordinati
- Contra: Replica data potential

#### Synchronous (eager)

- The node accepting writes synd Implemented in updates/transactions before a
- **Pro**: Consistent
- **Contra**: needs a commit proto **RethinkDB** unavaialable under certain networк partitions

#### Implemented in

Dynamo , Riak, CouchDB, Redis, Cassandra, Voldemort, MongoDB, RethinkDB

BigTable, HBase, Accumulo,

CouchBase, MongoDB,

Charron-Bost, B., Pedone, F., Schiper, A. (eds.): Replication: Theory and Practice, Lecture Notes in Computer Science, vol. 5959. Springer (2010)

toc

### **Replication: Where**

#### Master-Slave (Primary Copy)

- Only a dedicated master is allowed to accept writes, slaves are read-replicas
- Pro: reads from the master are consistent
- Contra: master is a bottleneck and SPOF

#### Multi-Master (Update anywhere)

- The server node accepting the writes synchronously propagates the update or transaction before acknowledging
- Pro: fast and highly-available
- Contra: either needs coordination protocols (e.g. Paxos) or is inconsistent

![](_page_27_Figure_1.jpeg)

Viotti, Paolo, and Marko Vukolić. "Consistency in Non-Transactional Distributed Storage Systems." arXiv (2015).

![](_page_28_Figure_1.jpeg)

Viotti, Paolo, and Marko Vukolić. "Consistency in Non-Transactional Distributed Storage Systems." arXiv (2015).

![](_page_29_Figure_1.jpeg)

Viotti, Paolo, and Marko Vukolić. "Consistency in Non-Transactional Distributed Storage Systems." arXiv (2015).

![](_page_30_Figure_1.jpeg)

Viotti, Paolo, and Marko Vukolić. "Consistency in Non-Transactional Distributed Storage Systems." arXiv (2015).

![](_page_31_Figure_1.jpeg)

Viotti, Paolo, and Marko Vukolić. "Consistency in Non-Transactional Distributed Storage Systems." arXiv (2015).

![](_page_32_Figure_1.jpeg)

Viotti, Paolo, and Marko Vukolić. "Consistency in Non-Transactional Distributed Storage Systems." arXiv (2015).

![](_page_33_Figure_1.jpeg)

Viotti, Paolo, and Marko Vukolić. "Consistency in Non-Transactional Distributed Storage Systems." arXiv (2015).

![](_page_34_Figure_1.jpeg)

Viotti, Paolo, and Marko Vukolić. "Consistency in Non-Transactional Distributed Storage Systems." arXiv (2015).

![](_page_35_Figure_3.jpeg)
#### NoSQL Storage Management In a Nutshell



Low Performance High Performance **RR**: Random Reads **RW**: Random Writes **SR**: Sequential Reads **SW**: Sequential Writes

#### NoSQL Storage Management In a Nutshell







## Local Secondary Indexing

Partitioning By Document

Partition I			
	Кеу	Color	
Data	12	Red	
	56	Blue	
	77	Red	
×	Term	Match	
apu	Red	[12,77]	
<u> </u>	Blue	[56]	

	Partition II				
	Кеу	Color			
Data	104	Yellow			
	188	Blue			
	192	Blue			
×	Term	Match			
apr	Yellow	[104]			
<u> </u>	Blue	[188,192]			



#### Local Secondary Indexing

#### Partitioning By Document



#### Local Secondary Indexing

#### Partitioning By Document



# **Global Secondary Indexing**

#### Partitioning By Term

Partition I					
	Кеу	Color			
Ita	12	Red			
Da	56	Blue			
	77	Red			
×	Term	Match			
apu	Yellow	[104]			
<u> </u>	Blue	[56, 188, 192]			
			ノ		

	Partition II				
	Кеу	Color			
Ita	104	Yellow			
Da	188	Blue			
	192	Blue			
ех	Term	Match			
Ind	Red	[12,77]			



# Global Secondary Indexing

#### Partitioning By Term



# **Global Secondary Indexing**

#### Partitioning By Term



#### **Query Processing Techniques**

Summary

- Local Secondary Indexing: Fast writes, scatter-gather queries
- Global Secondary Indexing: Slow or inconsistent writes, fast queries
- (Distributed) Query Planning: scarce in NoSQL systems but increasing (e.g. left-outer equi-joins in MongoDB and θ-joins in RethinkDB)
- Analytics Frameworks: fallback for missing query capabilities
- Materialized Views: similar to global indexing



# How are the techniques from the NoSQL toolbox used in actual data stores?

# Outline



NoSQL Foundations and Motivation



The NoSQL Toolbox: Common Techniques

- Overview & Popularity
- Dynamo & Riak
- HBase
- Cassandra
- Redis
- MongoDB



NoSQL Systems



#### NoSQL Landscape HYPERTABLE webservices HBASE **Document** Amazon DynamoDB Google Cassandra Datastore mongoDB Wide Column redis **Key-Value** CouchDB *mriak* amazon webservices™ S3 RAVENDB Graph **Project Voldemort** Neo4j

InfiniteGraph

E (

the graph database



Соиснваѕе

# Popularity

http://db-engines.com/de/ranking

#	System	Model	Score	11.	Elasticsearch	Search engine	86.31
1.	Oracle	Relational DBMS	1462 02	12.	Teradata	Relational DBMS	73.74
			1102.02	13.	SAP Adaptive Server	Relational DBMS	71.48
2.	MySQL	Relational DBMS	1371.83	14.	Solr	Search engine	65.62
3.	MS SQL Server	Relational DBMS	1142.82	15.	HBase	Wide column store	51.84
Δ	MongoDB	Document store	220.22	16.	Hive	Relational DBMS	47.51
7.	INIONGODD	Document store	520.22	17.	FileMaker	Relational DBMS	46.71
5.	PostgreSQL	Relational DBMS	307.61	18.	Splunk	Search engine	44.31
6.	DB2	Relational DBMS	185.96	19.	SAP HANA	<b>Relational DBMS</b>	41.37
-	0		424 50	20.	MariaDB	<b>Relational DBMS</b>	33.97
7.	Cassandra	wide column store	134.50	21.	Neo4j	Graph DBMS	32.61
8.	Microsoft Access	Relational DBMS	131.58	22.	Informix	Relational DBMS	30.58
9.	Redis	Key-value store	108.24	23.	Memcached	Key-value store	27.90
				24.	Couchbase	Document store	24.29
10.	SQLite	Relational DBMS	107.26	25.	Amazon DynamoDB	Multi-model	23.60

**Scoring**: Google/Bing results, Google Trends, Stackoverflow, job offers, LinkedIn

#### Dynamo (AP)

- Developed at Amazon (2007)
- Sharding of data over a ring of nodes
- Each node holds multiple partitions
- Each partition replicated N times







# **Reading and Writing**

- > An arbitrary node acts as a coordinator
- ▶ N: number of replicas
- **R**: number of nodes that need to confirm a read
- **W**: number of nodes that need to confirm a write



# Riak (AP)

- Open-Source Dynamo-Implementation
- Extends Dynamo:
  - Keys are grouped to Buckets
  - KV-pairs may have metadata and links
  - Map-Reduce support
  - Secondary Indices, Update Hooks, Solr Integration
  - Riak CS: S3-like file storage, Riak TS: time-series database



<b>in riak</b>	
Riak	
Model:	
Key-Value	
License:	
Apache 2	
Written in:	
Erlang und C	

#### Dynamo and Riak Classification



# Redis (CA)

- Remote Dictionary Server
- Rich Key-Value model
- Asynchronous Master-Slave Replication
- Tunable persistence: logging and snapshots
- Optimistic batch transactions (Multi blocks)
- Very high performance: >100k ops/sec per node
- Redis Cluster (sharding) still in the early stages



#### **Redis Data structures**

String, List, Set, Hash, Sorted Set



#### Classification: Redis Techniques



# Google BigTable (CP)

- Published by Google in 2006
- Original purpose: storing the Google search index

A Bigtable is a sparse, distributed, persistent multidimensional sorted map.

Data model also used in: HBase, Cassandra, HyperTable, Accumulo



### Wide-Column Data Modelling

Storage of crawled web-sites ("Webtable"):



## Architecture



## Storage: Sorted-String Tables

- **Goal**: Append-Only IO when writing (no disk seeks)
- Achieved through: Log-Structured Merge Trees
- Writes go to an in-memory memtable that is periodically persisted as an SSTable as well as a commit log
- Reads query memtable and all SSTables



#### Storage

r5

#### Logical to physical mapping:



r1:cf2:c1:t1:<value>
r2:cf2:c2:t1:<value>
r3:cf2:c2:t2:<value>
r3:cf2:c2:t1:<value>
r5:cf2:c1:t1:<value>
File cf2

r1:cf1:c1:t1:<value>
r2:cf1:c2:t1:<value>
r3:cf1:c2:t1:<value>
r3:cf1:c1:t2:<value>
r5:cf1:c1:t1:<value>
File cf1





## Apache HBase (CP)

- Open-Source Implementation of BigTable
- Hadoop-Integration
  - Data source for Map-Reduce
  - Uses Zookeeper and HDFS
- Data modelling challenge: key design, tall vs wide
  - **Row Key**: only access key (no indices)  $\rightarrow$  key design important
  - Tall: good for scans
  - Wide: good for gets, consistent (*single-row atomicity*)
- Interface: REST, Avro, Thrift

HBASE
HBase
Model:
Wide-Column
License:
Apache 2
Written in:
Java

#### Classification: HBase Techniques



# Apache Cassandra (AP)

- Published 2007 by Facebook
- Idea:
  - BigTable's wide-column data model
  - Dynamo ring for replication and sharding
- Cassandra Query Language (CQL): SQL-like query- and DDL-language
- ► Compound indices: partition key (shard key) + clustering key (ordered per partition key) → Limited range queries
- Secondary indices: hidden table with mapping  $\rightarrow$  queries with simple equality condition

Cassandra
Cassandra
Model:
Wide-Column
License:
Apache 2
Written in:
Java

1111/11/



#### Classification: Cassandra Techniques



## MongoDB (CP)

- ▶ From hu**mongo**us ≅ gigantic
- Tunable consistency
- Schema-free document database
- Allows complex queries and indexing
- Sharding (either range- or hash-based)
- Replication (either synchronous or asynchronous)
- Storage Management:
  - Write-ahead logging for redos (*journaling*)
  - Storage Engines: memory-mapped files, in-memory, Logstructured merge trees (WiredTiger)





## Sharding und Replication



#### Classification: MongoDB Techniques





How can the choices for an appropriate system be narrowed down?
## Outline



NoSQL Foundations and Motivation

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-		-

The NoSQL Toolbox: Common Techniques



NoSQL Systems



Decision Guidance: NoSQL Decision Tree

- Decision Tree
- Classification Summary
- Literature Reommendations

## **NoSQL** Decision Tree



## **NoSQL** Decision Tree



#### System Properties According to the NoSQL Toolbox

For fine-grained system selection:

	Functional Requirements									
	Scan Queries	ACID Transactions	<b>Conditional Writes</b>	Joins	Sorting	Filter Query	Full-Text Search	Analytics		
Mongo	х		х		Х	х	Х	х		
Redis	х	х	х							
HBase	x		х		X			Х		
Riak							x	х		
Cassandra	x		х		х		х	х		
MySQL	х	х	х	х	х	х	х	х		

#### System Properties According to the NoSQL Toolbox

For fine-grained system selection:

	Non-functional Requirements										
	Data Scalability	Write Scalability	Read Scalability	Elasticity	Consistency	Write Latency	Read Latency	Write Throughput	Read Availability	Write Availability	Durability
Mongo	Х	Х	Х		Х	X	X		X		Х
Redis			Х		Х	X	X	Х	Х		Х
HBase	Х	Х	X	Х	Х	X		Х			Х
Riak	Х	Х	Х	Х		X	X	Х	Х	Х	Х
Cassandra	х	Х	Х	Х		Х		Х	Х	Х	Х
MySQL			Х		Х						Х

#### System Properties According to the NoSQL Toolbox

For fine-grained system selection:

	Techniques																			
	Range-Sharding	Hash-Sharding	Entity-Group Sharding	<b>Consistent Hashing</b>	Shared-Disk	<b>Transaction Protocol</b>	Sync. Replication	Async. Replication	Primary Copy	Update Anywhere	Logging	Update-in-Place	Caching	In-Memory	Append-Only Storage	Global Indexing	Local Indexing	Query Planning	Analytics Framework	<b>Materialized Views</b>
Mongo	х	Х					Х	Х	Х		Х		Х	Х	Х		Х	Х	Х	
Redis								х	Х		х		х							
HBase	х						х		Х		х		х		х					
Riak		х		х				х		Х	х	х	х			х	х		Х	
Cassandra		х		х				х		х	х		х		х	х	х			х
MySQL					х			х	х		х	х	х				х	х		



Select Requirements in Web GUI:



System makes suggestions based on data from practitioners, vendors and automated benchmarks:



# Future Work Polyglot Persistence Mediator





# Bacend Build faster Apps faster.

www.baqend.com

#### Approach: API as a Superset For Web-Apps and Mobile





#### Approach: API as a Superset For Web-Apps and Mobile

#### Standard HTTP Caching



#### Approach: API as a Superset For Web-Apps and Mobile



# Summary



- High-Level NoSQL Categories:
  - Key-Value, Wide-Column, Docuement, Graph
  - Two out of {Consistent, Available, Partition Tolerant}
- The NoSQL Toolbox: systems use similar techniques that promote certain capabilities







#### 4<sup>th</sup> Workshop on Scalable Cloud Data Management

Co-located with the <u>IEEE BigData Conference</u>. Washington D.C., December 5th 2016.

Submit Paper

June 6, 2016



#### SCDM 2016 announced

The fourth Scalable Cloud Data Management Workshop (SCDM 2016) will again be held in conjunction with the IEEE BigData 2016 - this year in Washington D.C.



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*		
Q	SCDM 2016 announced	Paper Deadline
Q		

# Thank you!

gessert@informatik.uni-hamburg.de

www.baqend.com www.scdm2016.com vsis-www.informatik.uni-hamburg.de