Tutorial 18: Going for Speed

	09:00	Introduction
Up next!	09:15	Part 1: Efficient Frontend Design
	09:40	Q&A
	09:50	Coffee Break
	10:00	Part 2: High-Performance Networking
	10:40	Q&A
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	11:00	Part 3: Scalable Backend Architectures
	11:40	Q&A
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	12:00	Part 4: Performance Tracking & Analysis
	12:00	The Core Web Vitals (<u>Google Guest Speaker!</u>)
	12:30	Measuring Web Performance
	12:50	Q&A





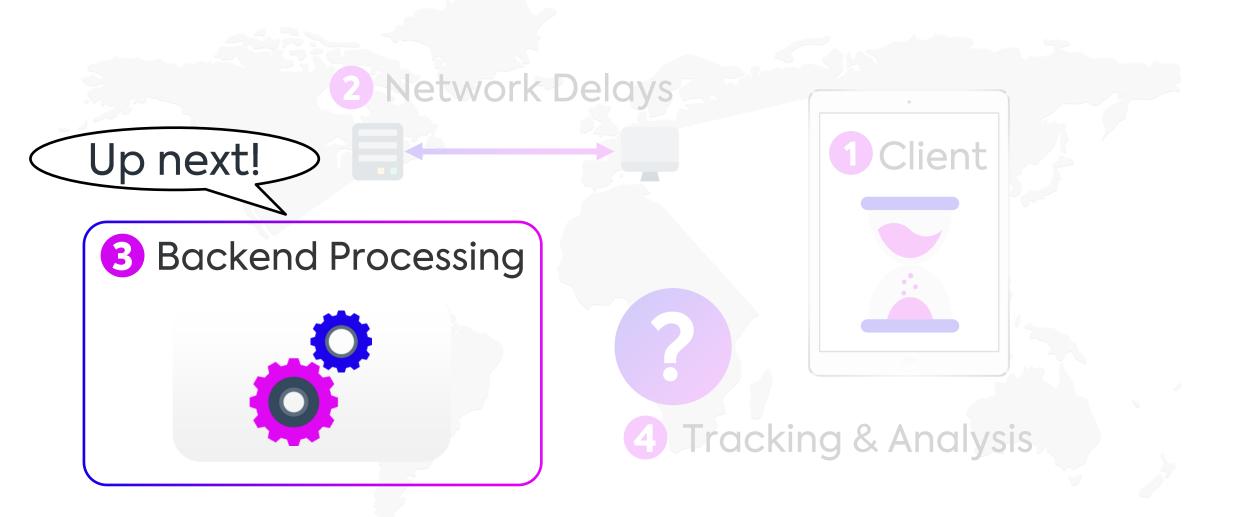
Part 3:

Scalable Backend Architectures

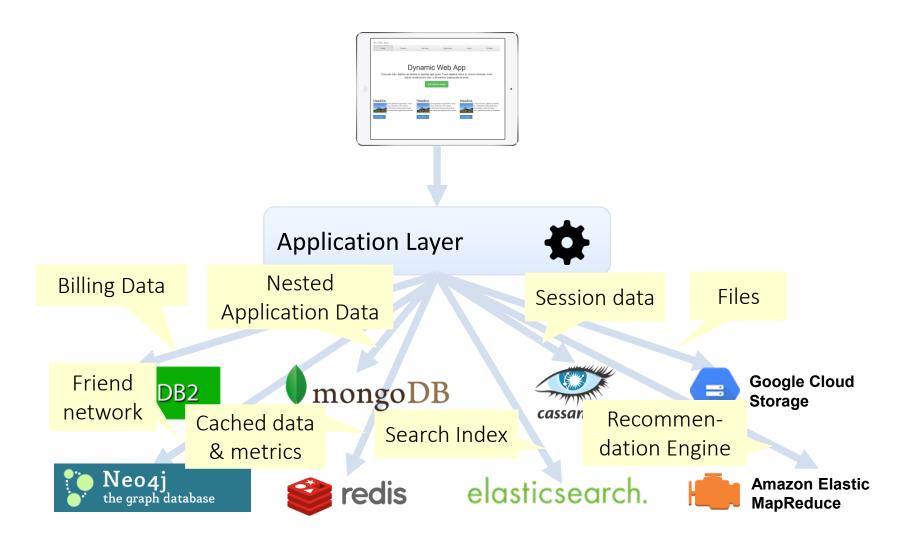
The Web Conference April 15, 2021 Tutorial

Wolfram Wingerath, Benjamin Wollmer, Felix Gessert, Stephan Succo, Norbert Ritter

The 4 Challenges of Web Performance



How to Choose a Database System



How to Choose a Database System





- "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 200 NoSQL systems

Wide Column Store / Column Families Hadoop / HBase API: Java / any writer, Protocol: any write call, Quey Method: MapReduce Java / any exce. Replication: HDPS Replication. Writen in Java, Concurrency: ?, Misc: Links: 3 Books [], 2, 3] Cassandra massively scalable, partitioned row stor nasteriess architecture, linear scale performance, no single points of failure, read/write support across multiple data conters & cloud availability zones. API / Ouery Method: COL and Thrift, replication: pccr-to-pccr, written in: Java, Concurrency: tunable consistency, Misc: built-in data compression, HapReduce support, primary/secondary indexes, security features. Links: Documentation, PlanetC* Leneral: <u>Hypertablic</u> API: Thrift (Java, PAP, Pol, Python, Ruby, dt.), Protocol: Thrift (Jury Michod: HQL, native Thrift API, Replication: HDPS Replication, Concurrency: HVCC, Consistency Model: Fully consistent Misc High porformance C++ implementation of Geogle's Bigable. <u>B</u> Commercial support Accumule Accumule is based on BigTable and is built on top of Hadoop. Zookcoper, and Thrift. It features improvements on the BigTable design in the form of cell-based access control, improved compression, and a sover-side programming mechanism that can modify key/value pairs at various points in the data management DEDCESS. Amazon SimpleDB Mise: not open source / part of AWS, Book (will be outperformed by DynamoDB ?!) Cloudata Google's Big table clone like HBase, » Article Cloudera Professional Software & Services based on Hadron HPCC from Loxisticxis, info, article Stratosphere (research system) massive parallel & flexible execution, WR generalization and extention (paper, poster). (OpenNeptune, Obase, KDI) Document Store MongoDB AFE BSON, Protocol: C, Query Method: dynamic object-based language & MapReduce, Scolication: Master Slave & Auto-Sharding Written n: C++, Concurrency: Update in Place. Mise: Indexing, GridFS, Freeware + Commercial License Links: » Talk, » Notes, » Company Elasticscarch API: REST and many languages, Protocol: REST, Query Method: via JSON, Application + Sharding: automatic and configurable, written in: Java, Misc: schema mapping, multi tenancy with arbitrary indexes, Company and Support 2 Couchbase Server API: Memorached API+protoco (binary and 4500), most languages. Protocol: Memorached REST interface for cluster conf + management. Whiten in: C/C+++ Erlang(clustering), Banagement, Witch in: CrC+++ Eriang (outcome) Replication: Peer to Peer, fully consistent, Mic: Transparent topology changes during operation, provides memcached-compatible caching buckets, commercially supported version available, Links <u>> Wiki, > Article</u> CouchDB API: JSON, Protocol: REST, Query Method: MapReduccR of JavaScript Funcs, Replication: laster Master, Written in: Erlang, Concurrency: MVCC Links: » 3 Couch06 books, » Couch Lounge (partitioning clusering), » Dr. Dobbs RethinkDB API: protobuf-based, Quey Method: unified chainable query language (incl. JOINs, sub-queries, MapReduce, GroupedMapReduce); Replication: Sync and Async Master Slave with per-table acknowledgements, Sharding guided range-based, Written in: C++, Concurrency, MVCC. Mise los-structured storage engine with concurrent incrementa parbage compactor RavenDB Jict solution. Provides HTTP/JSON access. LING queries & Sharding supported. <u>• Misc</u> Hantlogic Scruer/Honesconnectal| AP: ISON, XIII, Java Protocols HTTP, RESTQuey Nichod: Pull Text Scarch, XPath, XQuery, Range, Geospatial Writen in: C++ Concerneny: Shared-nothing cluster, MVCC Misc: Petabyt-scalable, cloudable, ACID transactions, autosharding, failover, master slave replication, secure with ACLs Developer Community $\underline{\mathbf{z}}$ Clusterpoint Server (Incolar-commodal) APE XML, PHP, Java, MET Protocols: HTTP, REST, native TCP/IP Quey Nethod: full text search, XML, range and Xpath queries: Witten in C++ Concurrency. ACID-compliant, transactional, multi-master cluster Mis: Petabyte-scalable document store and full text search engine. Information ranking. Replication. Cloudable ThruDB (please help provide more facts!) Uses Apache Thrif to integrate multiple backend databases as BerkeleyOB, Disk, MySQL, S3. Terrastore API: Java a http: Protocol: http: Language: Java, Queying: Range queries, Predicates, Replication: Partitioned with consistent hashing. Consistency: Per-record strict consistency, Mise Based on Terracotta

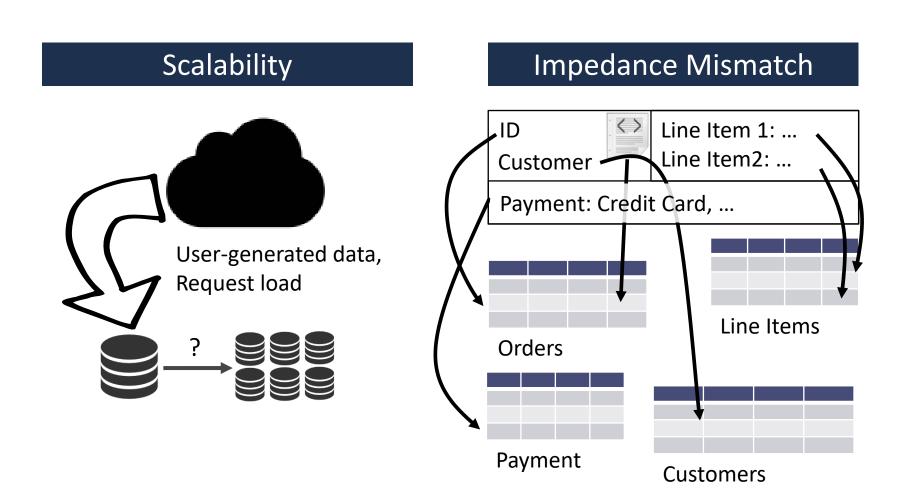
JaxDB Liphwight open source document database written in Jax for nep performance, runs in-memory, supports Androide. AFL [304], Java Quory Minotes, REST Obasta Skylic Query language, Java fluent Query API Concurnery: Atografic document writes indexes: eventually consistent indexes.

RaptorDB (SON based, Document store database with complice .nct map functions and automatic hybrid bitmap indexing and LINQ query filters

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

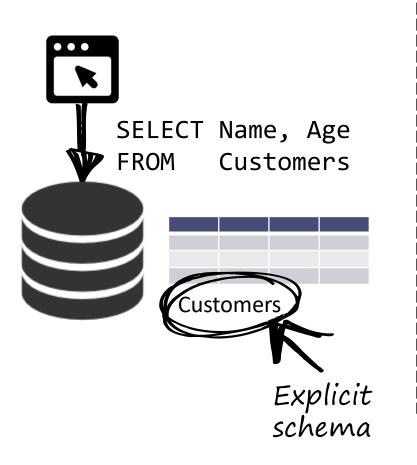
cjondb djon06 API: BSON, Protocol: C++, Quary Mathad: dynamic quarties and map/reduce; Drives: Java, C++, PHP Miss: ACID compliant, Pull shall console over google v8 angine; djondb requirements are submitted by users, are more thereast. Bit and compared the submitted by users.

NoSQL: Two Main Motivations

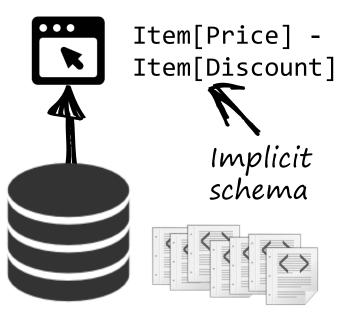


Schema-Free Data Remodeling

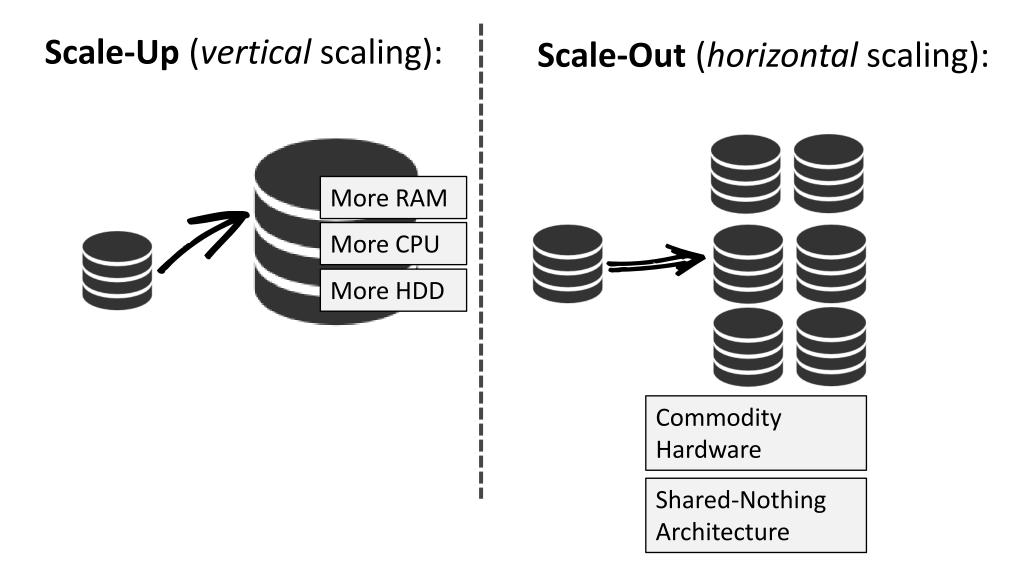
RDBMS:



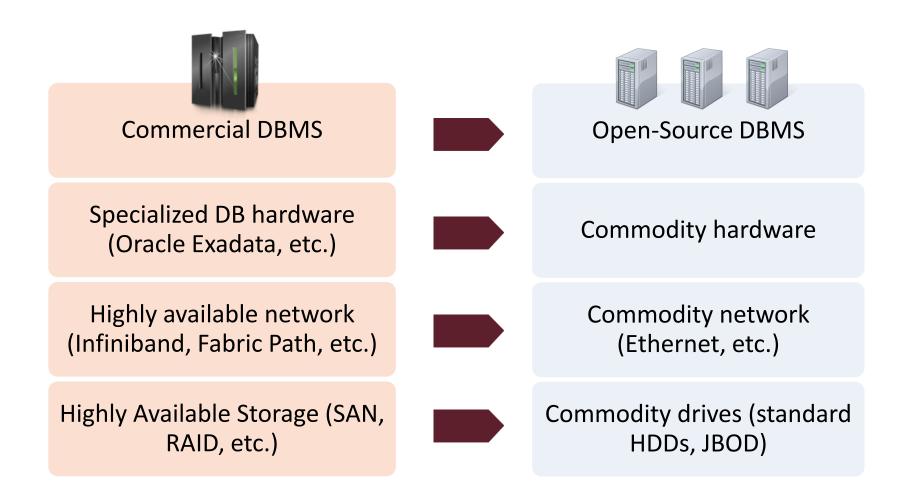
NoSQL DB:



Scale-Up vs. Scale-Out

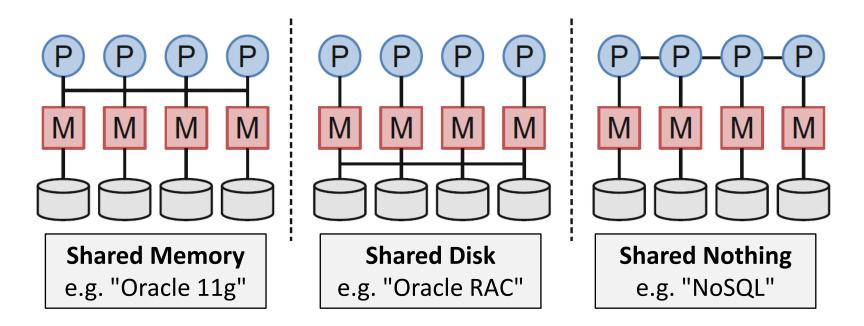


Paradigm Shift: Open-Source & Commodity



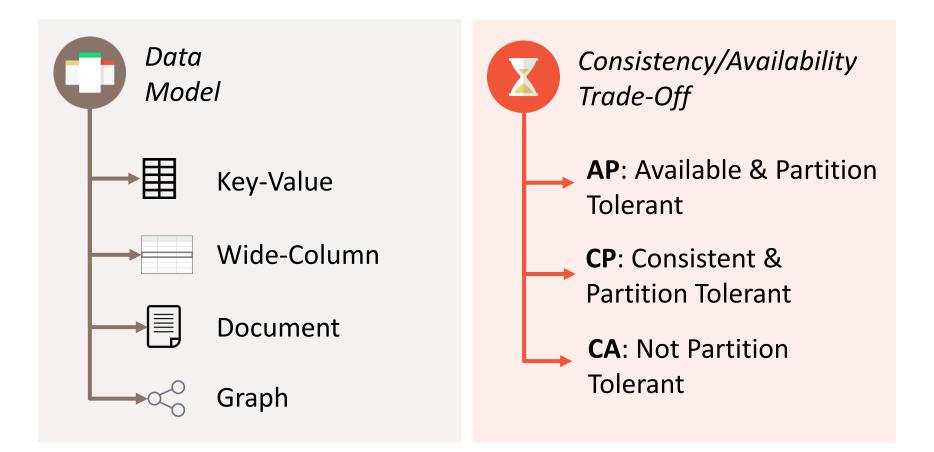
Paradigm Shift: Shared Nothing

Shift towards higher distribution & less coordination:

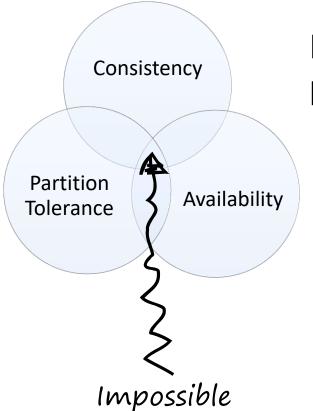


Typical Classification Schemes

Two common criteria:



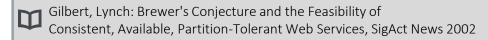




In a distributed system, only 2 out of the following 3 properties are achievable at the same time:

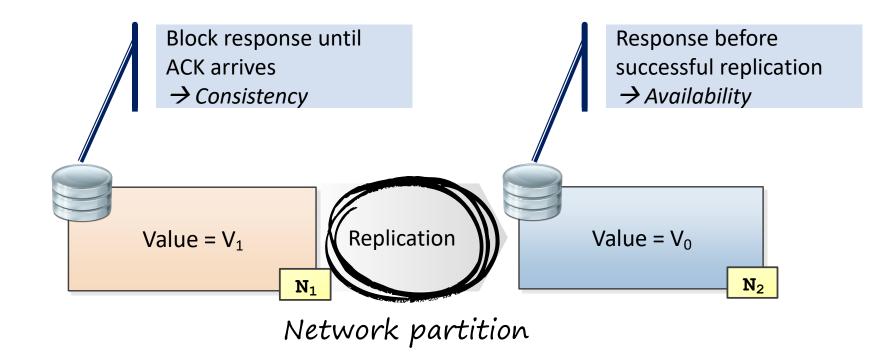
- **Consistency**: all clients have the same view on the data
- Availability: every request to a non-failed node most result in correct response
- Partition tolerance: the system has to continue working, even under arbitrary network partitions

Eric Brewer, ACM-PODC Keynote, Juli 2000

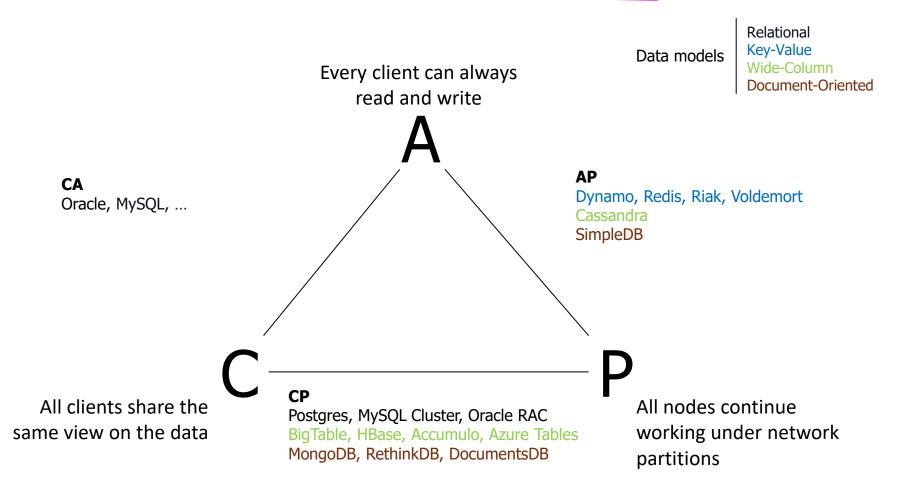


CAP Theorem: Intuitive Explanation

Problem: when a network partition occurs, either consistency or availability have to be given up

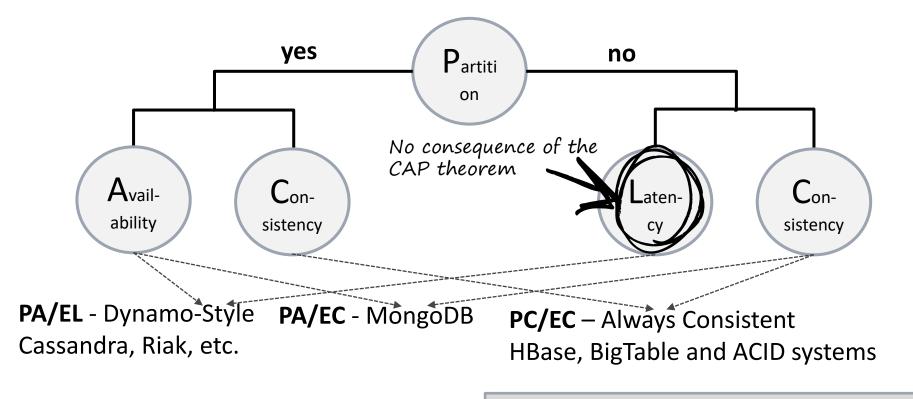




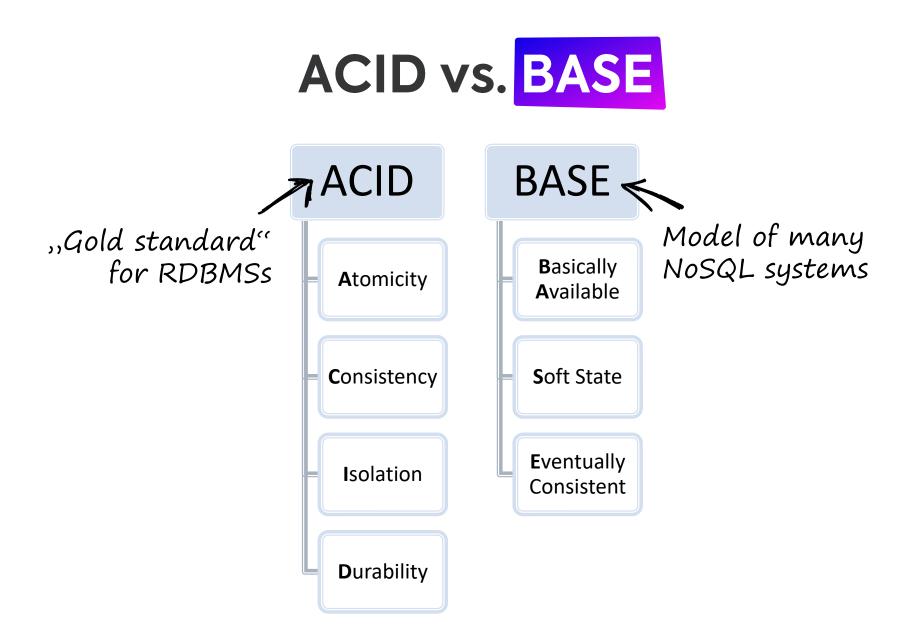


PACELC: What About Normal Operation?

Idea: Classify systems by behavior during network partitions and normal operation



Abadi, Daniel. "Consistency tradeoffs in modern distributed database system design: CAP is only part of the story."

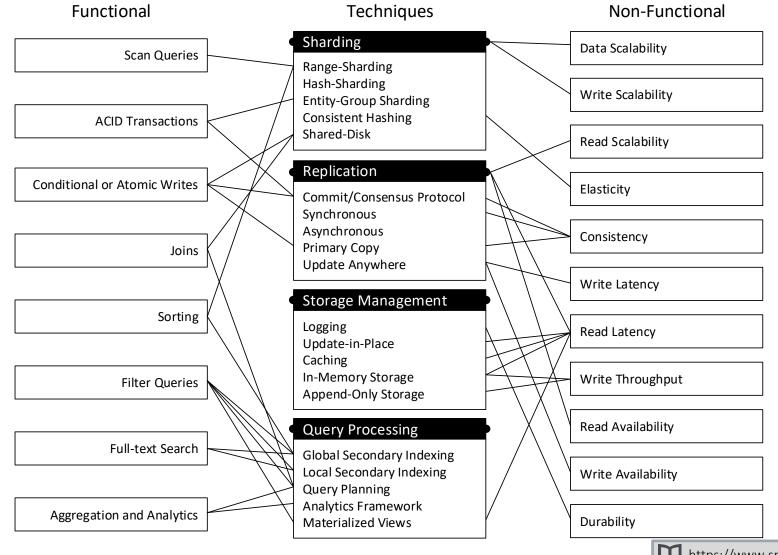






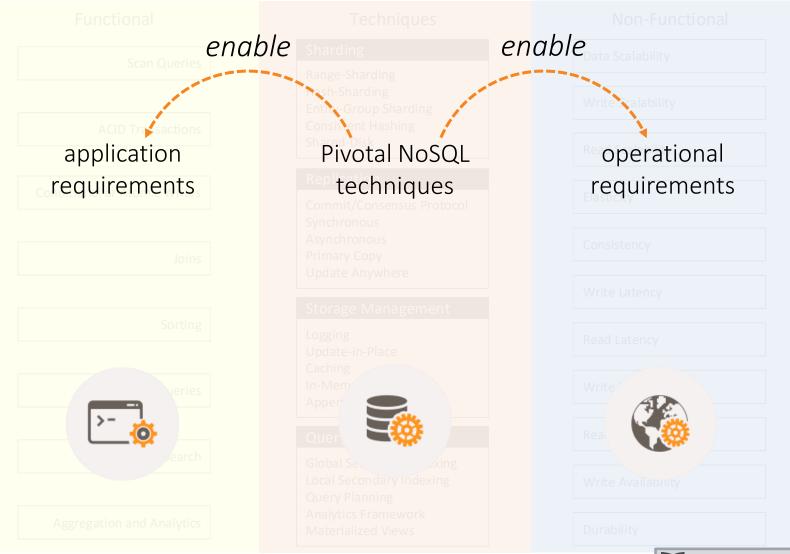


Requirements vs. Techniques

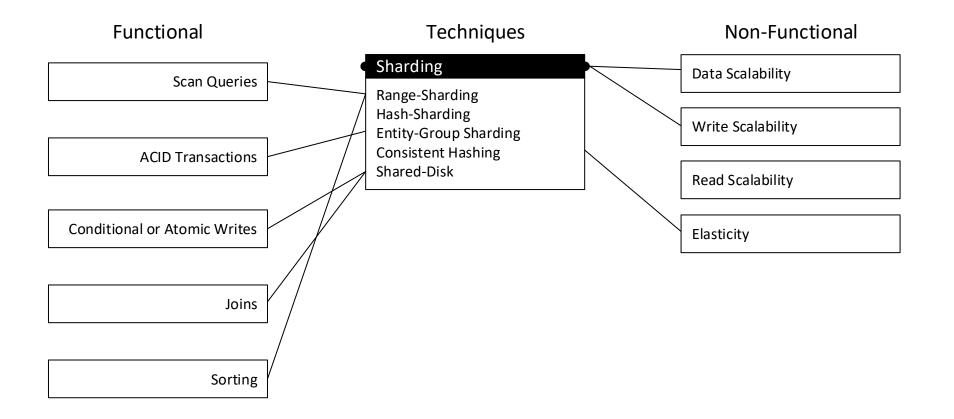


https://www.springer.com/gp/book/9783030435059

Requirements vs. Techniques

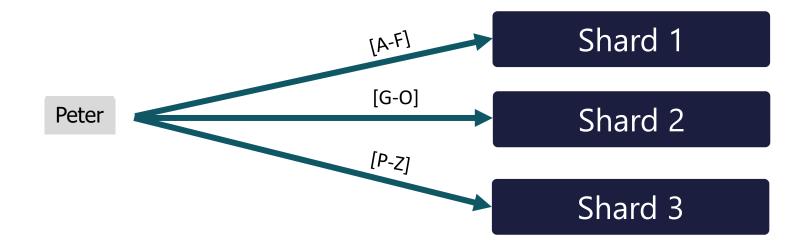


https://www.springer.com/gp/book/9783030435059



Sharding: **Scaling Storage & Throughput**

Horizontal distribution of data over nodes



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

Sharding: Approaches

Hash-based Sharding

- Hash of data values (e.g. key) determines partition (shard)
- **Pro**: Even distribution
- Contra: No data locality

Range-based Sharding

- Assigns ranges defined over fields (shard keys) to partitions
- Pro: Enables Range Scans and Sorting
- Contra: Repartitioning/balancing required

Entity-Group Sharding

- Explicit data co-location for single-node-transactions
- **Pro**: Enables ACID Transactions
- **Contra**: Partitioning not easily changable

Implemented in

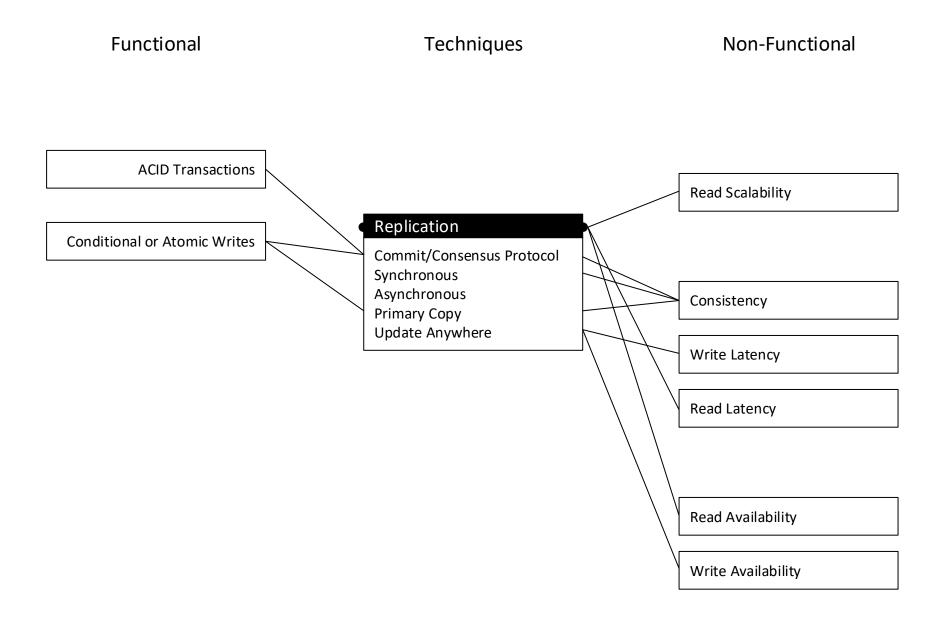
Dynamo, Cassandra, MongoDB, Riak, Redis, Azure Table,

Implemented in

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

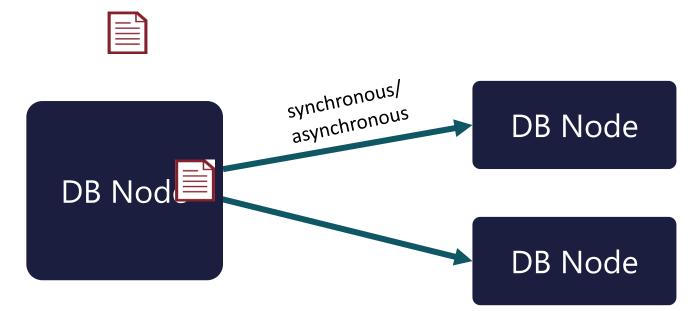
Implemented in

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



Replication: Read Scalability & Fault Tolerance

Stores N copies of each data item



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

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

Implemented in

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

Implemented in

BigTable, HBase, Accumulo, CouchBase, MongoDB, RethinkDB

Replication: Where

Master-Slave (Primary Copy)

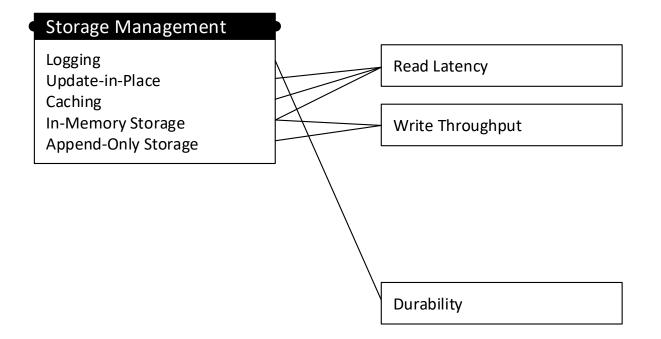
- Only a dedicated master is allowed to accept writes, slaves are read-replicas
- Pro: reads from the master are consistent
- **Contra**: master is a bottleneck and SPOF

Multi-Master (Update anywhere)

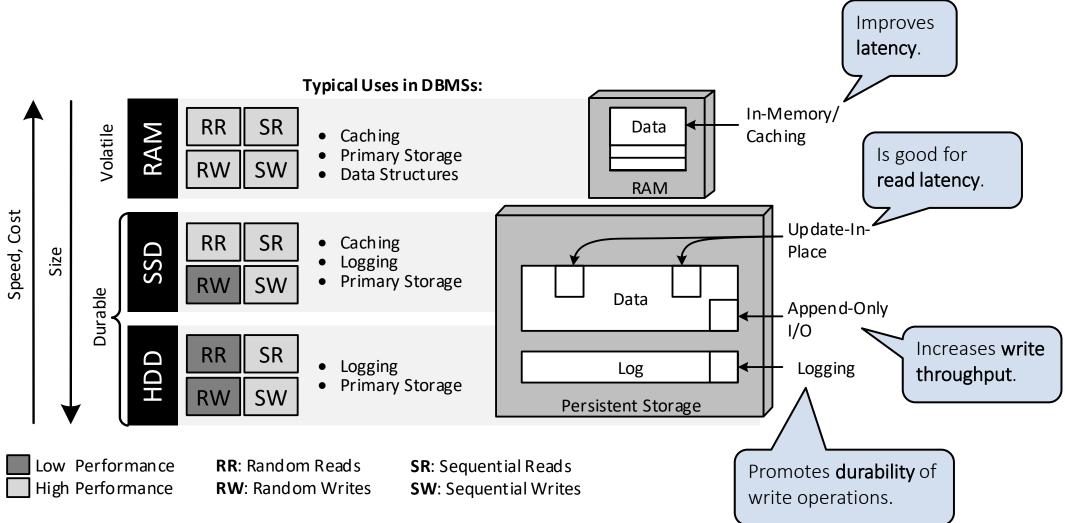
- The server node accepting the writes synchronously propagates the update or transaction before acknowledging
- **Pro**: fast and highly-available
- **Contra**: either needs coordination protocols (e.g. Paxos) or is inconsistent

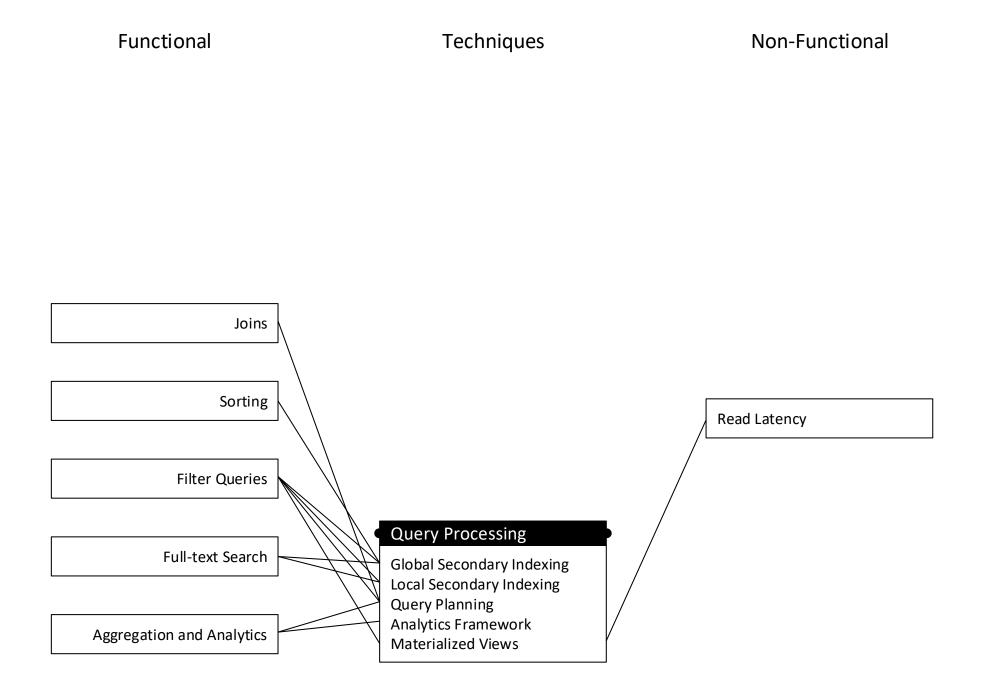
Functional

Techniques

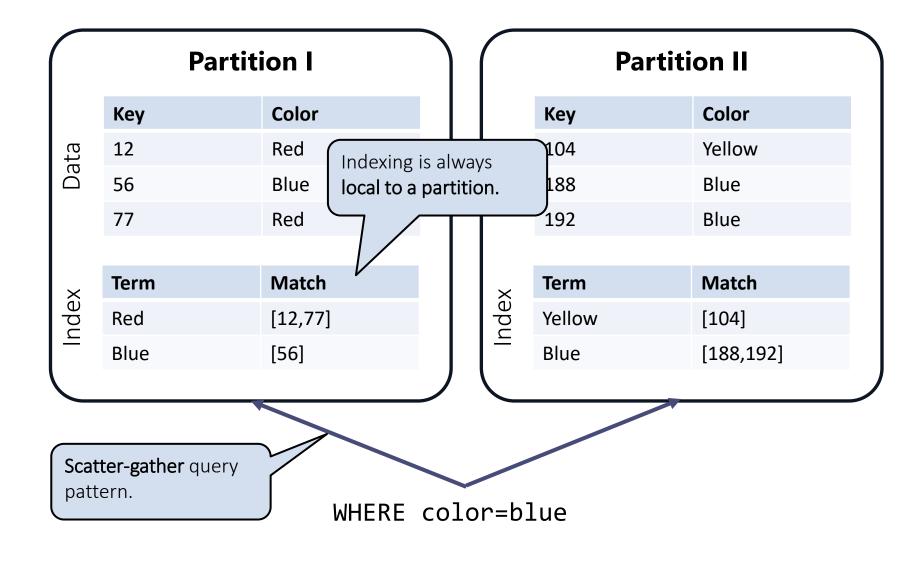


Storage Management





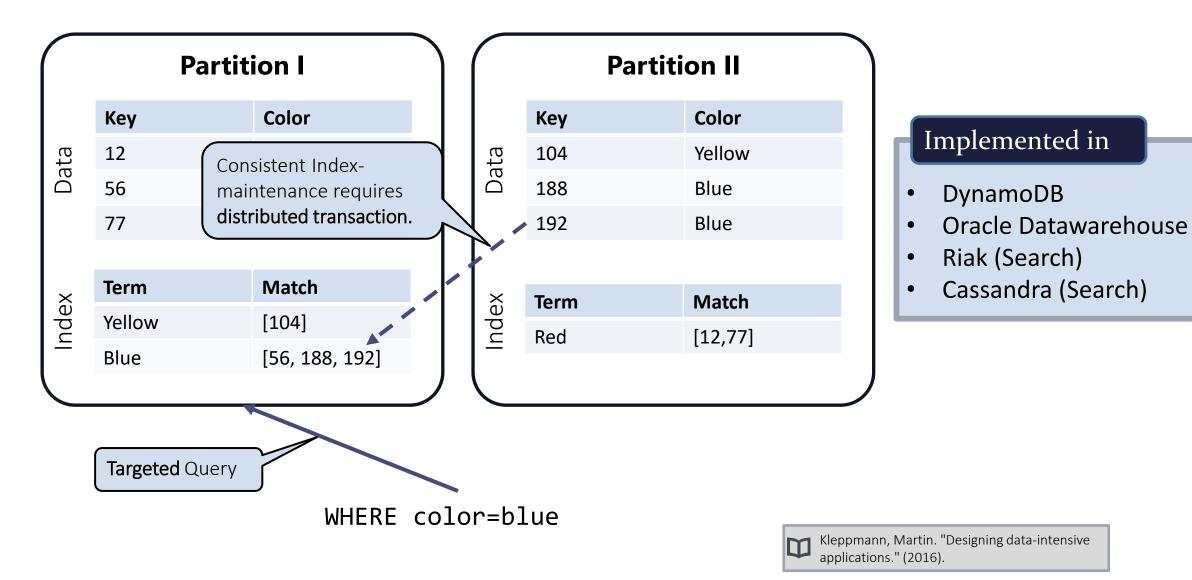
Local Secondary Indexing



Implemented in

- MongoDB
- Riak
- Cassandra
- Elasticsearch
- SolrCloud
- VoltDB

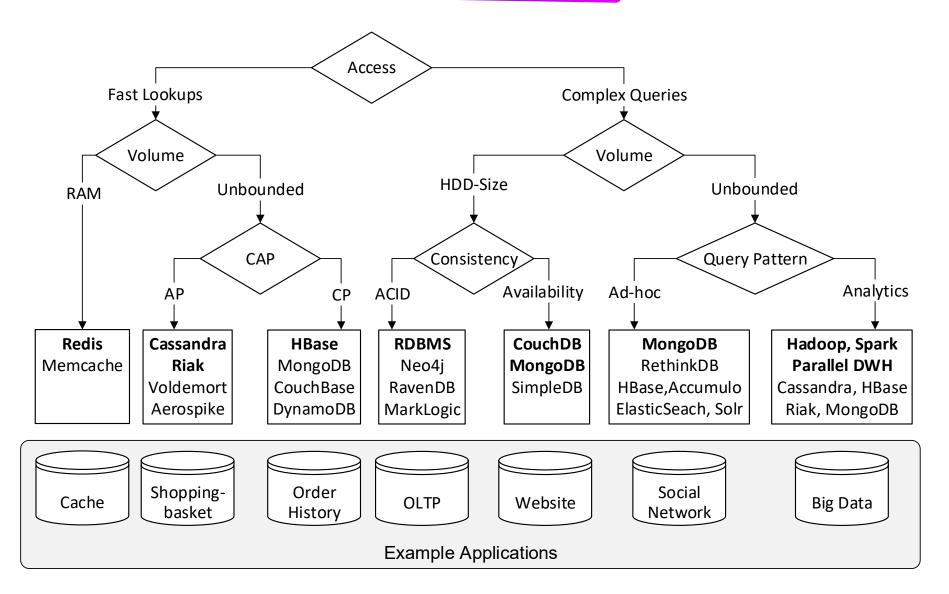
Global Secondary Indexing



Query Processing Techniques: Summary

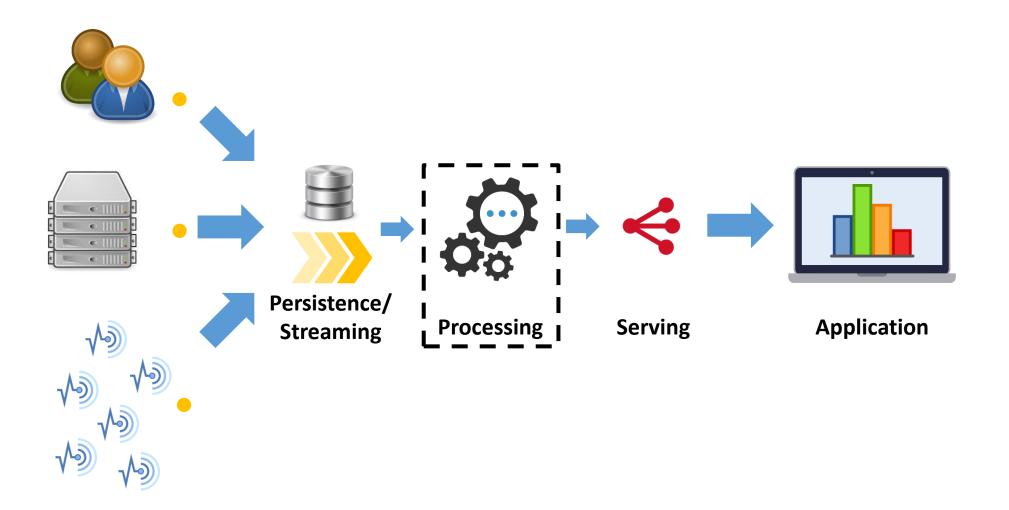
- Local Secondary Indexing: Fast writes, scatter-gather queries
- Global Secondary Indexing: Slow or inconsistent writes, fast queries
- (Distributed) Query Planning: scarce in NoSQL systems but increasing (e.g. left-outer equi-joins in MongoDB and θ-joins in RethinkDB)
- > Analytics Frameworks: fallback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

NoSQL Decision Tree



What About Push-Based Systems?

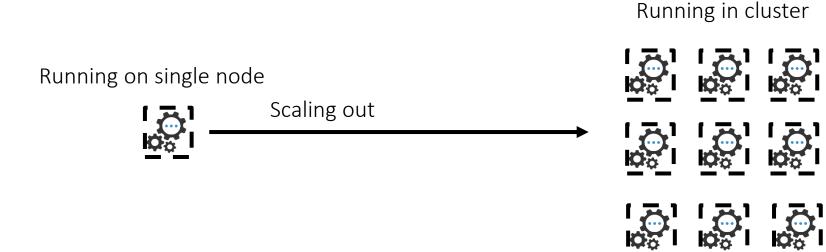
A Data Processing Pipeline



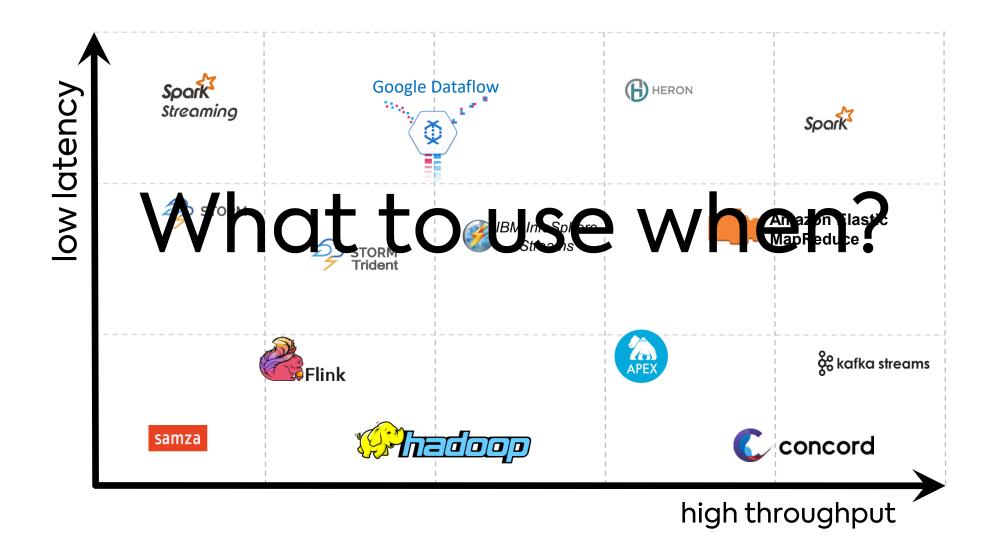
Scale-Out Made Feasible

Data processing frameworks hide complexities of scaling, e.g.:

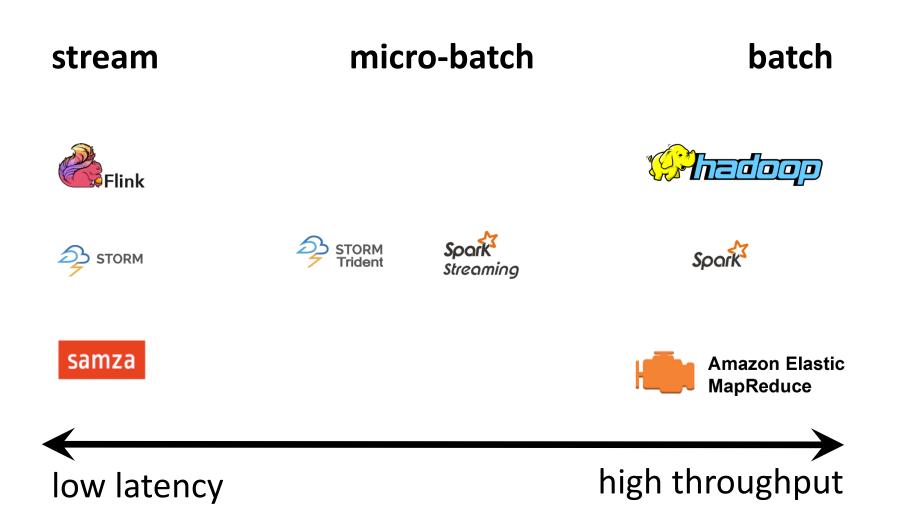
- **Deployment -** code distribution, starting/stopping work
- Monitoring health checks, application stats
- Scheduling assigning work, rebalancing
- Fault-tolerance restarting workers, rescheduling failed work



Big Data Processing Frameworks





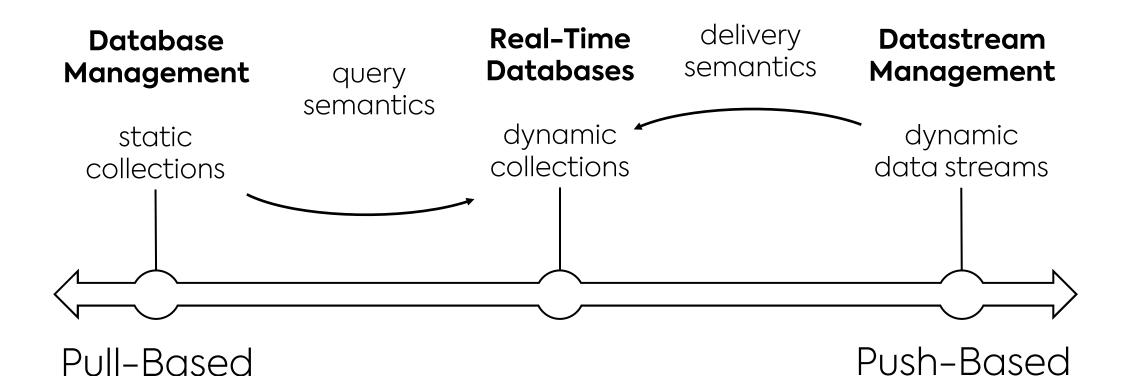


Stream Processing System Comparison

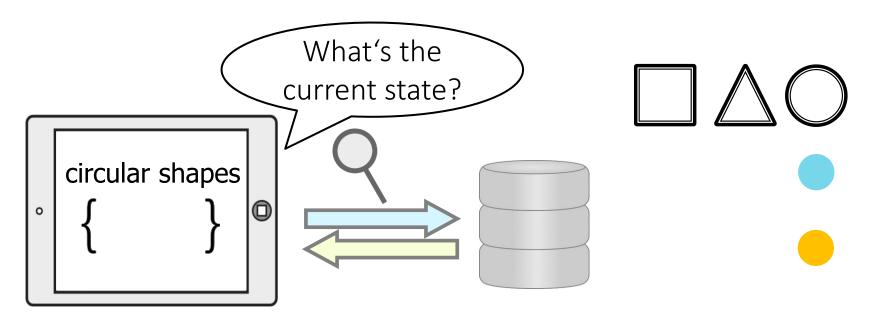
	Storm	Trident	Samza	Spark Streaming	Flink (streaming)
Strictest Guarantee	at-least- once	exactly- once	at-least- once	exactly-once	exactly-once
Achievable Latency	≪100 ms	<100 ms	<100 ms	<1 second	<100 ms
State Management	(small state)	(small state)	\checkmark	\checkmark	\checkmark
Processing Model	one-at-a- time	micro-batch	one-at-a- time	micro-batch	one-at-a- time
Backpressure	\checkmark	\checkmark	no (buffering)	\checkmark	\checkmark
Ordering	×	between batches	within partitions	between batches	within partitions
Elasticity	\checkmark	\checkmark	×	\checkmark	×

Real-Time Databases: Combining Push & Pull

Push vs. Pull: Trade-Offs in Data Management

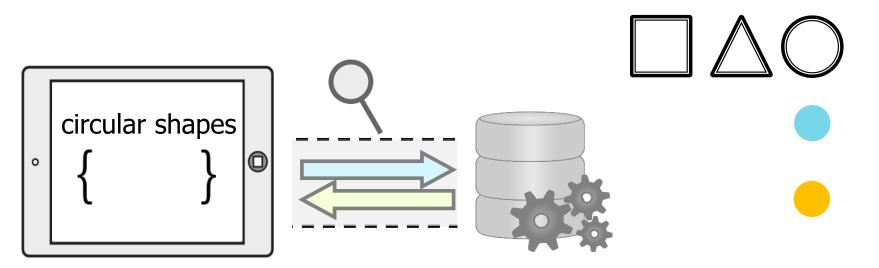


Traditional Databases: No Request, No Data!



- Periodic Polling for query result maintenance:
 → inefficient
- \rightarrow slow

Real-Time Databases: Always Up-to-Date



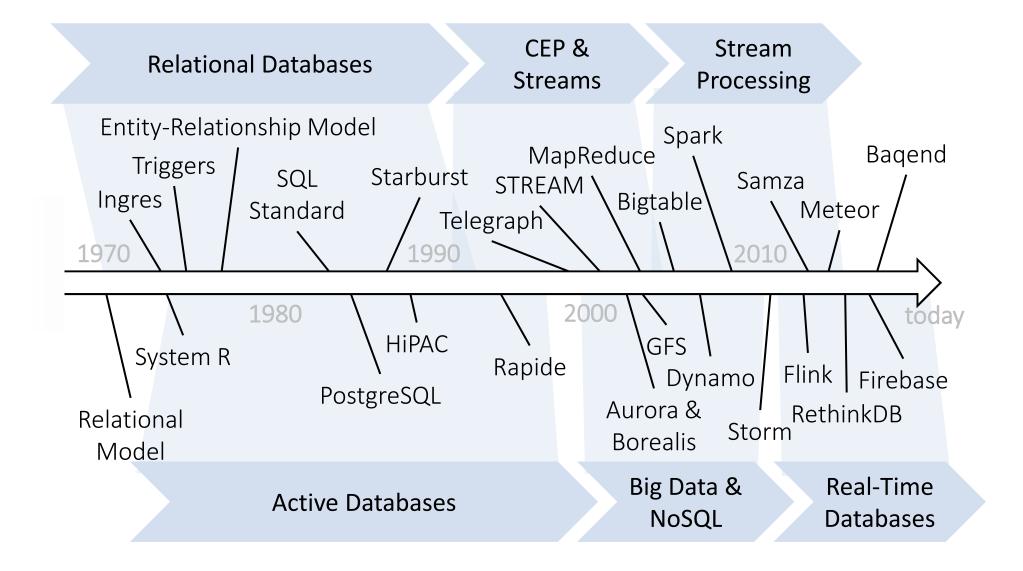
Real-Time Queries for query result maintenance:
→ efficient
→ fast

Real-Time Database Comparison

	METE		() RethinkDB	Parse	😕 Firebase
	Poll-and-Diff		Change Log Tailing		Unknown
Write Scalability	\checkmark	×	×	×	×
Read Scalability	×	\checkmark	\checkmark	\checkmark	? (100k connections)
Composite Filters (AND/OR)	\checkmark	\checkmark	\checkmark	\checkmark	(AND In Firestore)
Sorted Queries	\checkmark	\checkmark	\checkmark	×	(single attribute)
Limit	\checkmark	\checkmark	\checkmark	×	\checkmark
Offset	\checkmark	\checkmark	×	×	(value-based)
Self-Maintaining Queries	\checkmark	\checkmark	×	×	×
Event Stream Queries	\checkmark	\checkmark	\checkmark	✓ _	\checkmark

W. Wingerath, F. Gessert, N. Ritter: InvaliDB: *Scalable Push-Based Real-Time Queries on Top of Pull-Based Databases (Extended)*, VLDB 2020

Wrapup: A Short History of Data Management



Felix Gessert Wolfram Wingerath Norbert Ritter

Fast and

Scalable

Cloud Data

Management

Deringer









Wolfram Wingerath Norbert Ritter **Felix Gessert Real-Time &** Stream Data Management Push-Based Data in Research & Practice Deringer

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Up next!



