



Scalable Cloud Data Management

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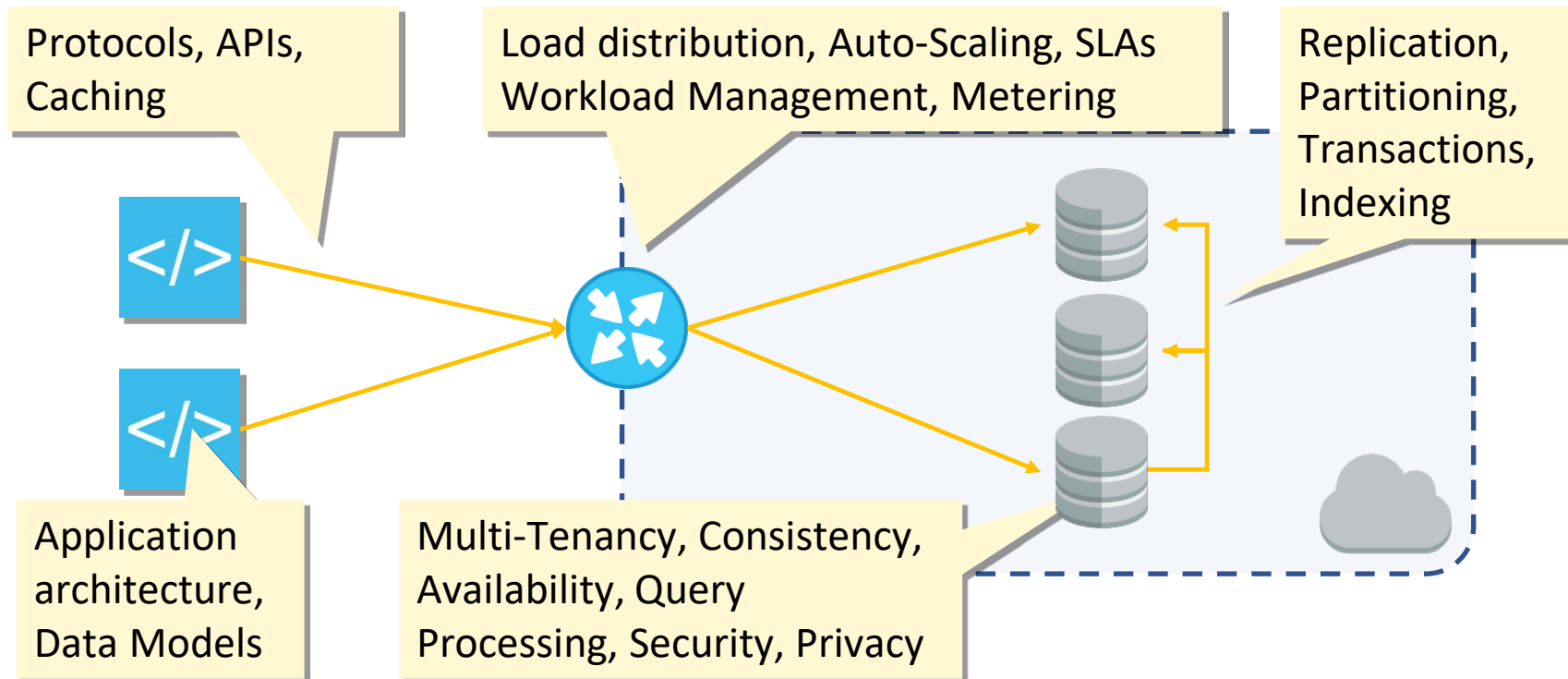
Universität Hamburg



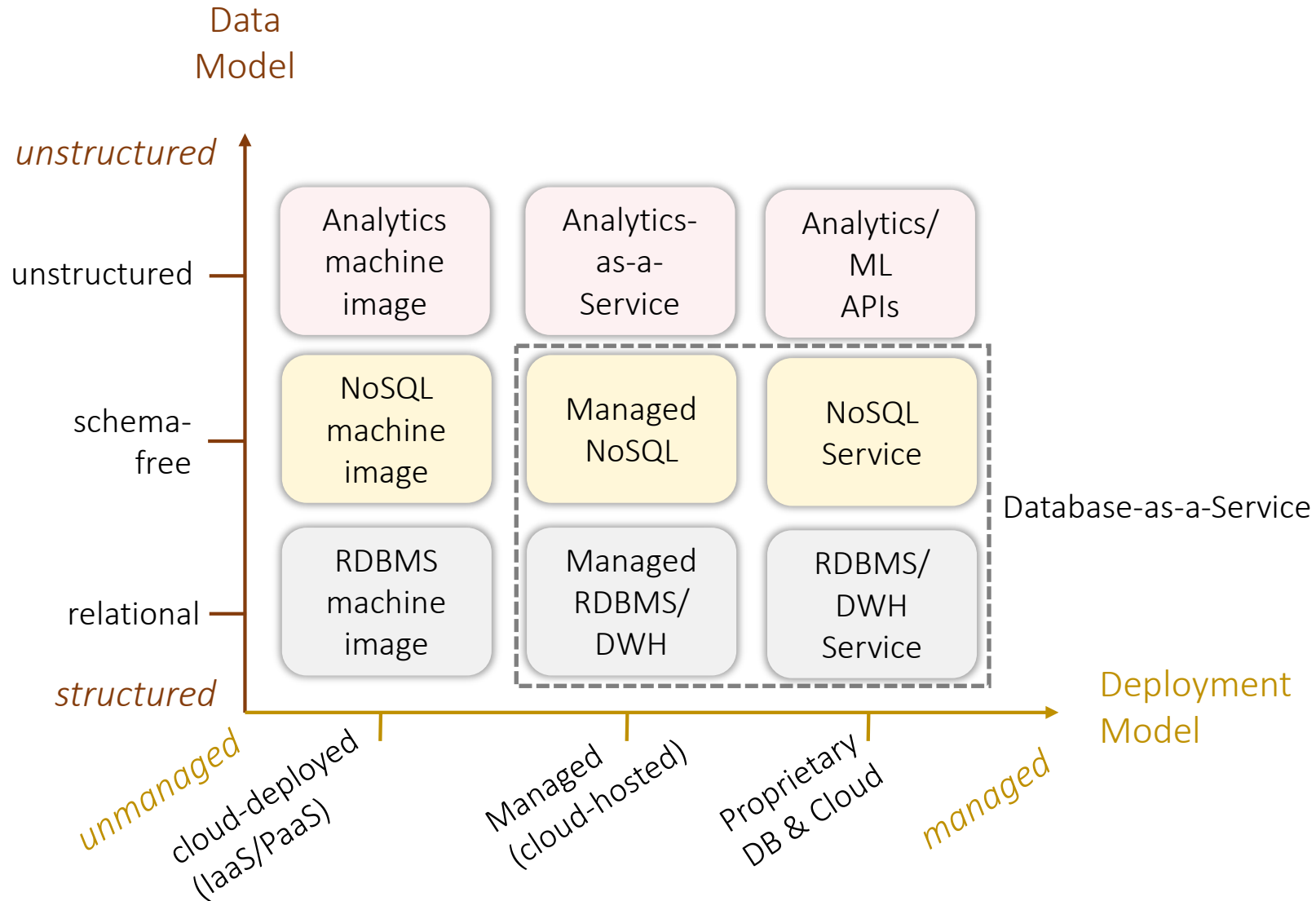
www.baqend.com

Cloud Data Management

- New field tackling the *design, implementation, evaluation* and *application implications* of **database systems** in cloud environments:

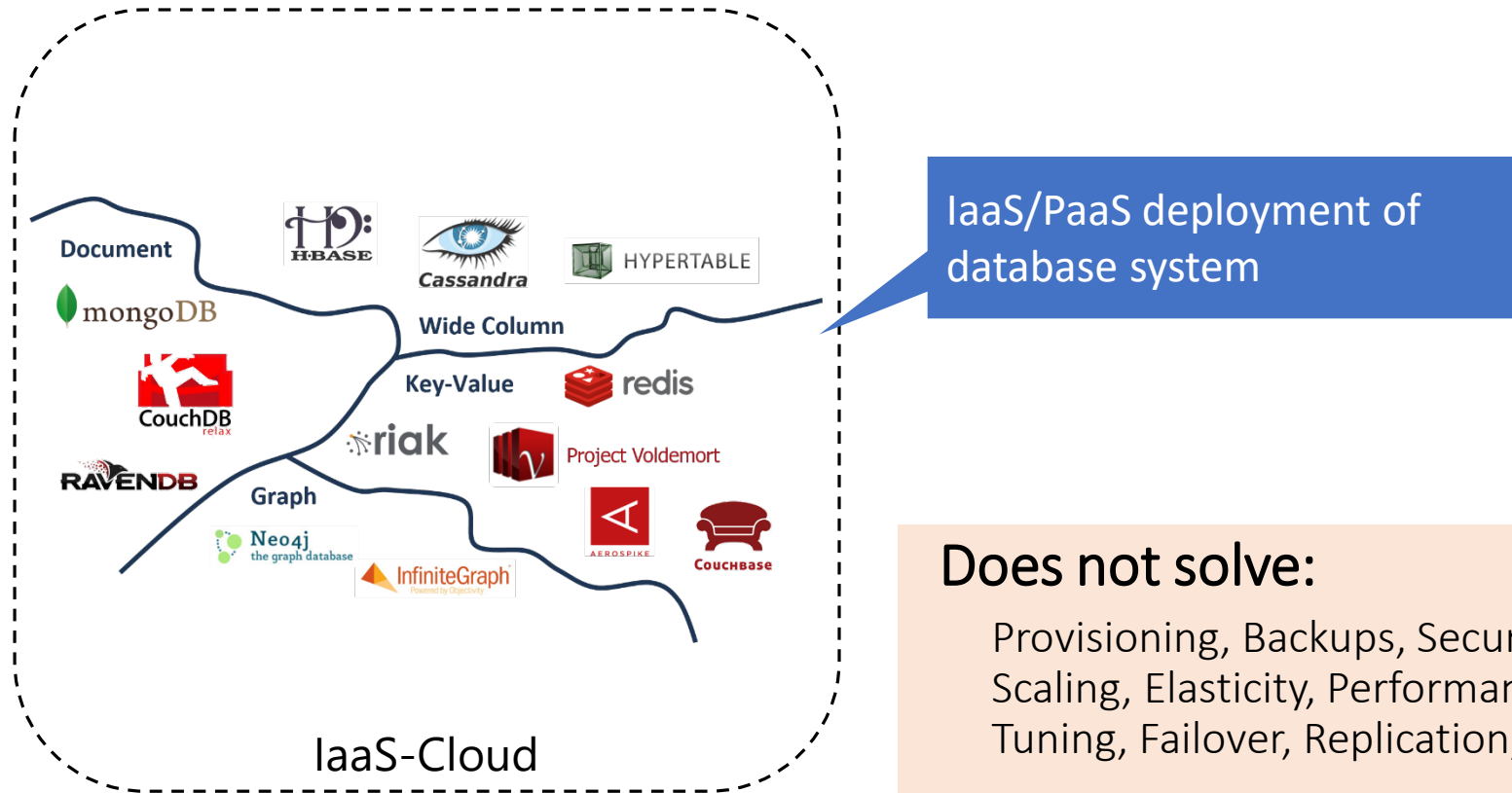


Cloud-Database Models



Cloud-Deployed Database

Database-image provisioned in IaaS/PaaS-cloud

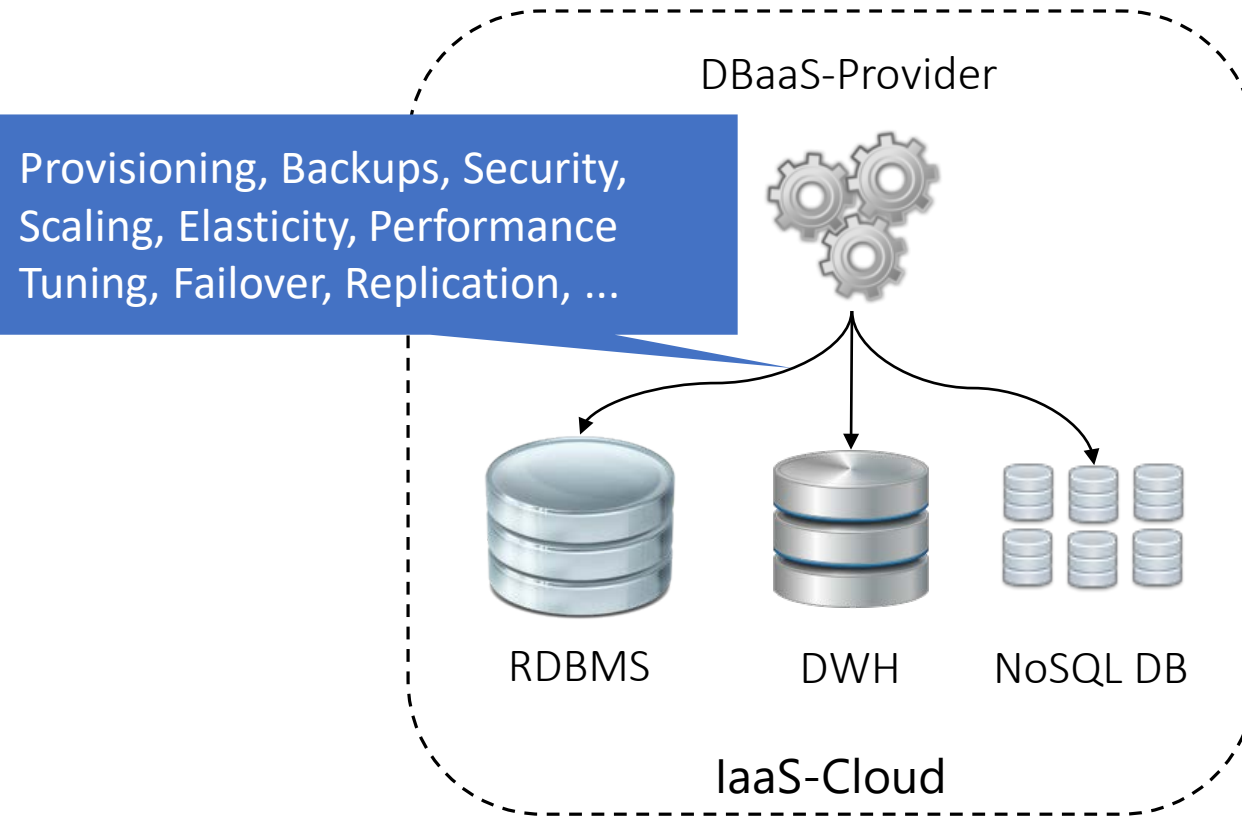

















Does not solve:

Provisioning, Backups, Security,
Scaling, Elasticity, Performance
Tuning, Failover, Replication, ...

Managed RDBMS/DWH/NoSQL DB

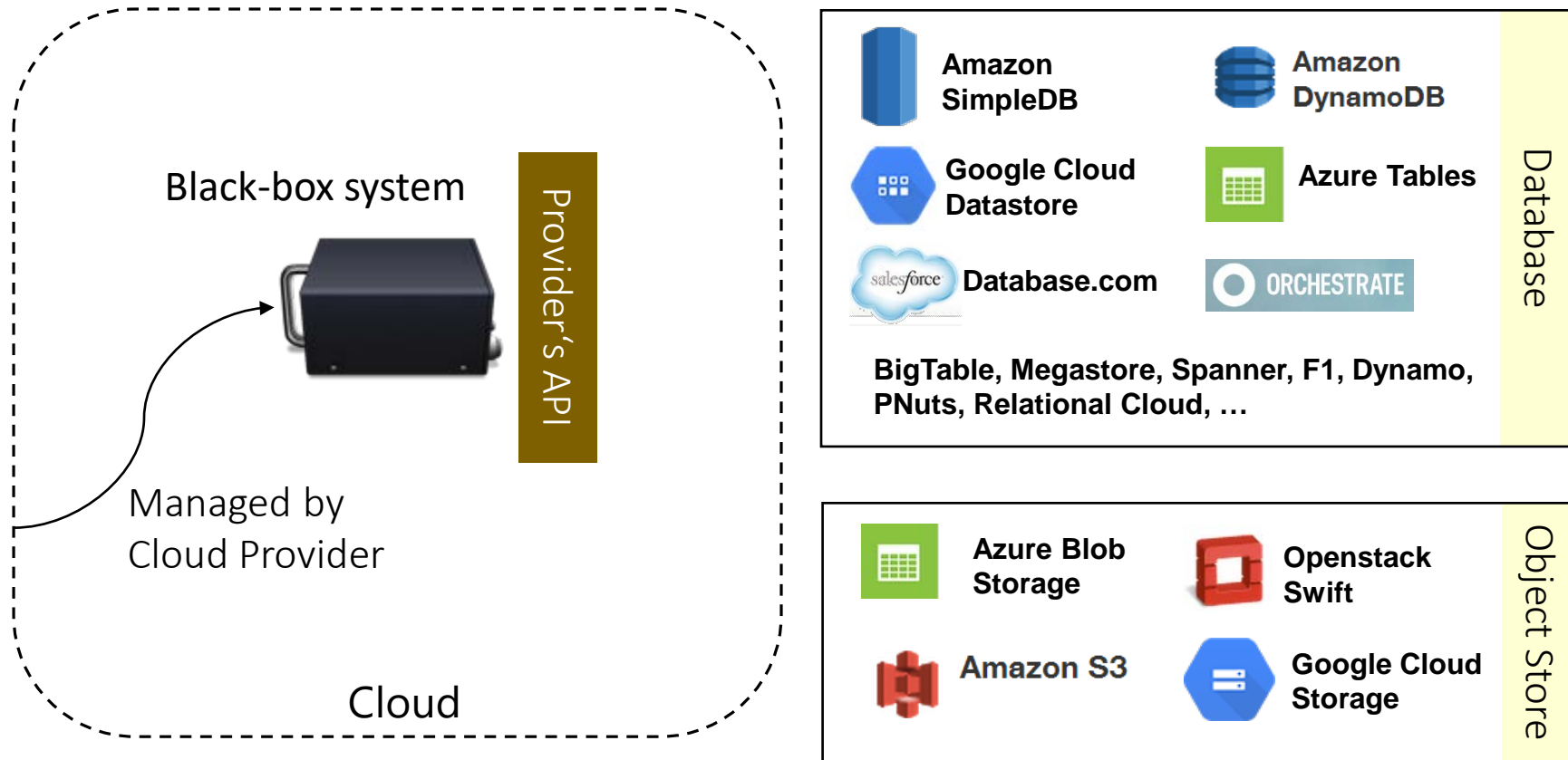
Cloud-hosted database



 Amazon RDS	 SQL Azure	RDBMS
 Clustrix	 EDB ENTERPRISEDB	
 Google Cloud SQL	 Heroku Postgres	
 Amazon ElastiCache	 mongoHQ	NOSQL DB
 mongolab	 Cloudant an IBM® Company	
 redis	 instaclustr	
 bonsai	 Iris Couch	
 Amazon Redshift	DWH	

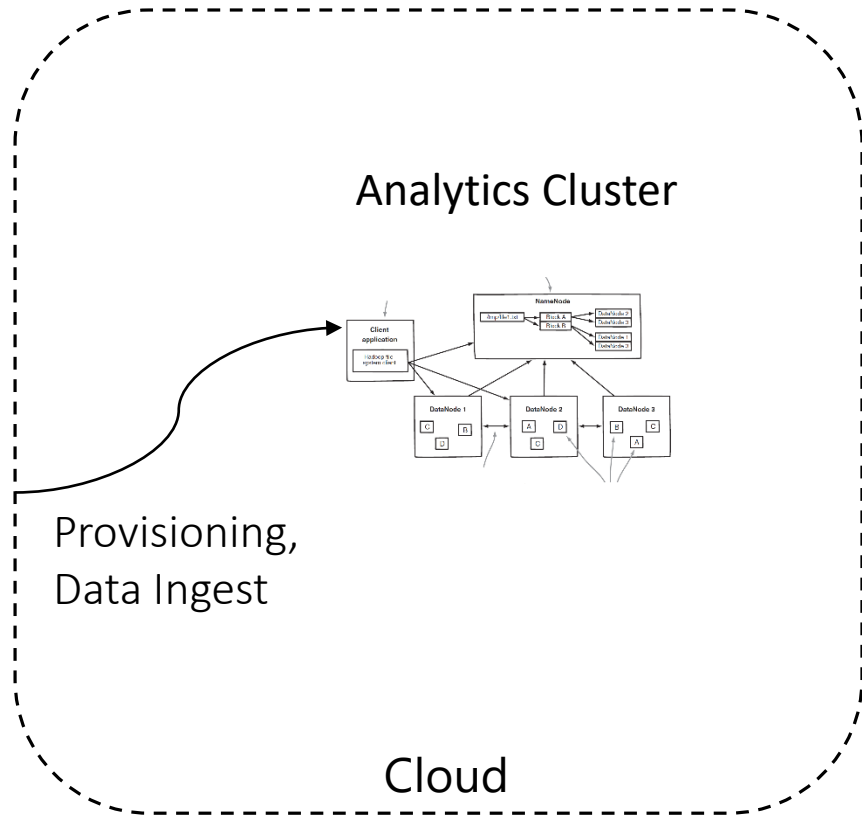
Proprietary Cloud Database

Designed for and deployed in vendor-specific cloud environment



Analytics-as-a-Service

Analytic frameworks and machine learning with service APIs



**Amazon Elastic
MapReduce**



**Azure
HDInsight**

Analytics



**Google
BigQuery**

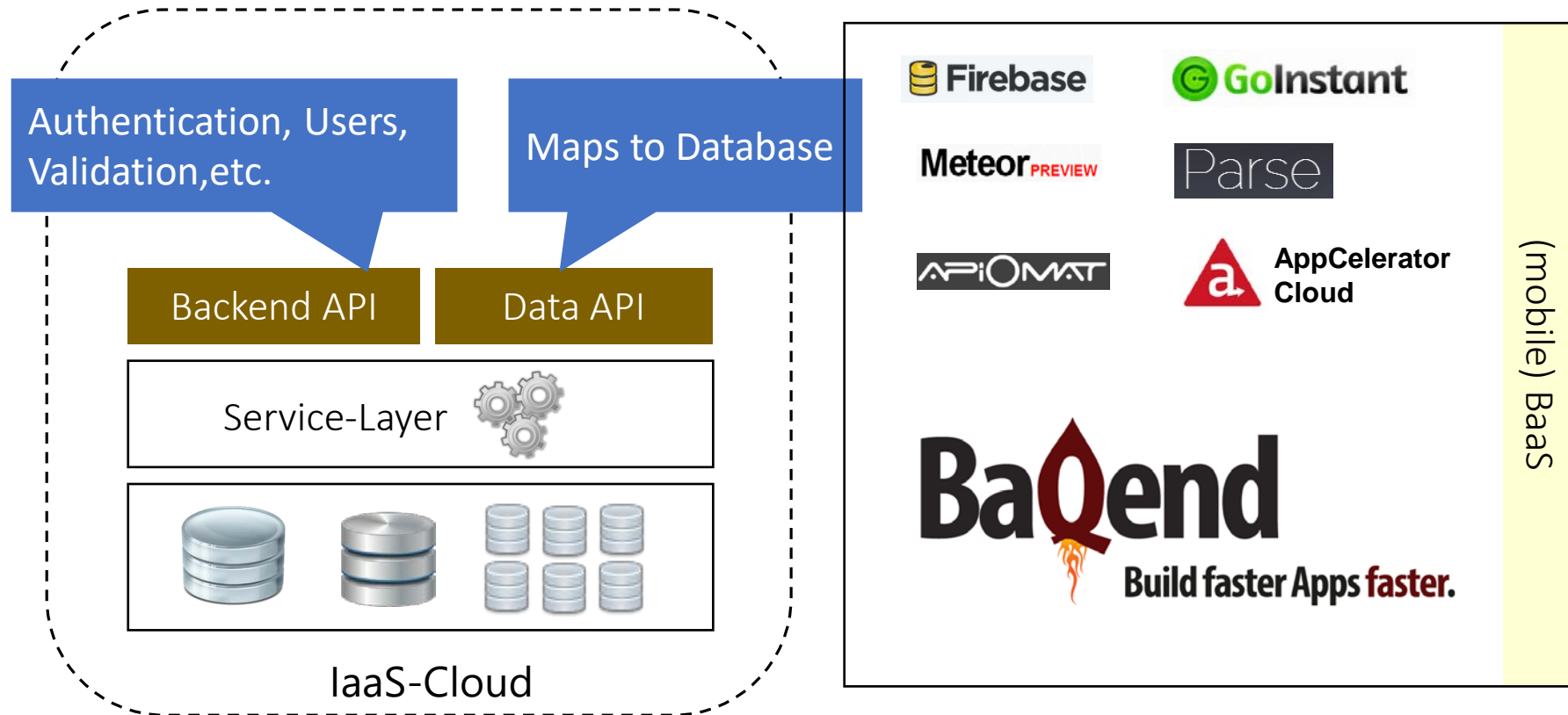


**Google
Prediction API**

ML

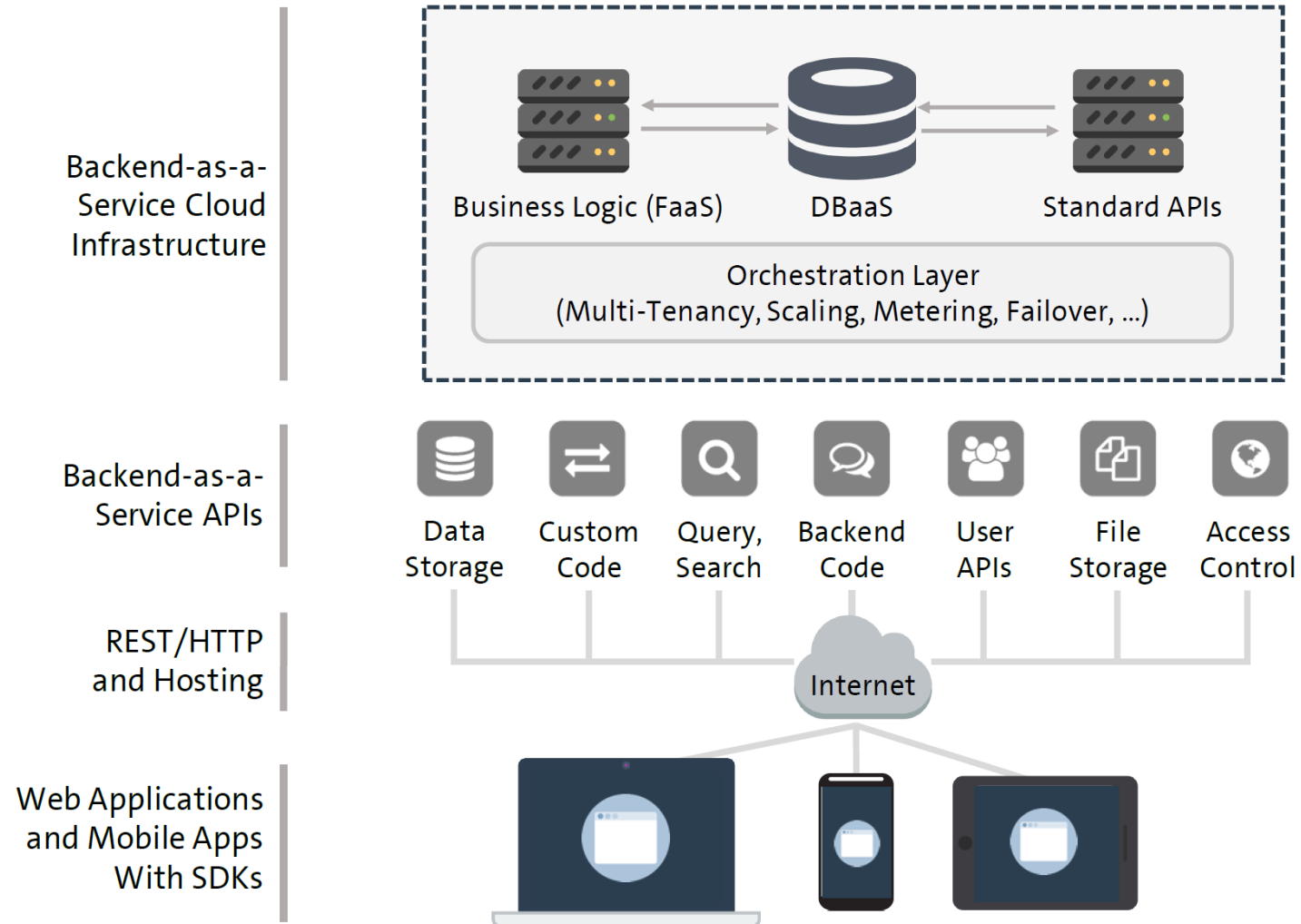
Backend-as-a-Service

DBaaS with embedded custom and predefined application logic



Backend-as-a-Service

DBaaS with embedded custom and predefined application logic



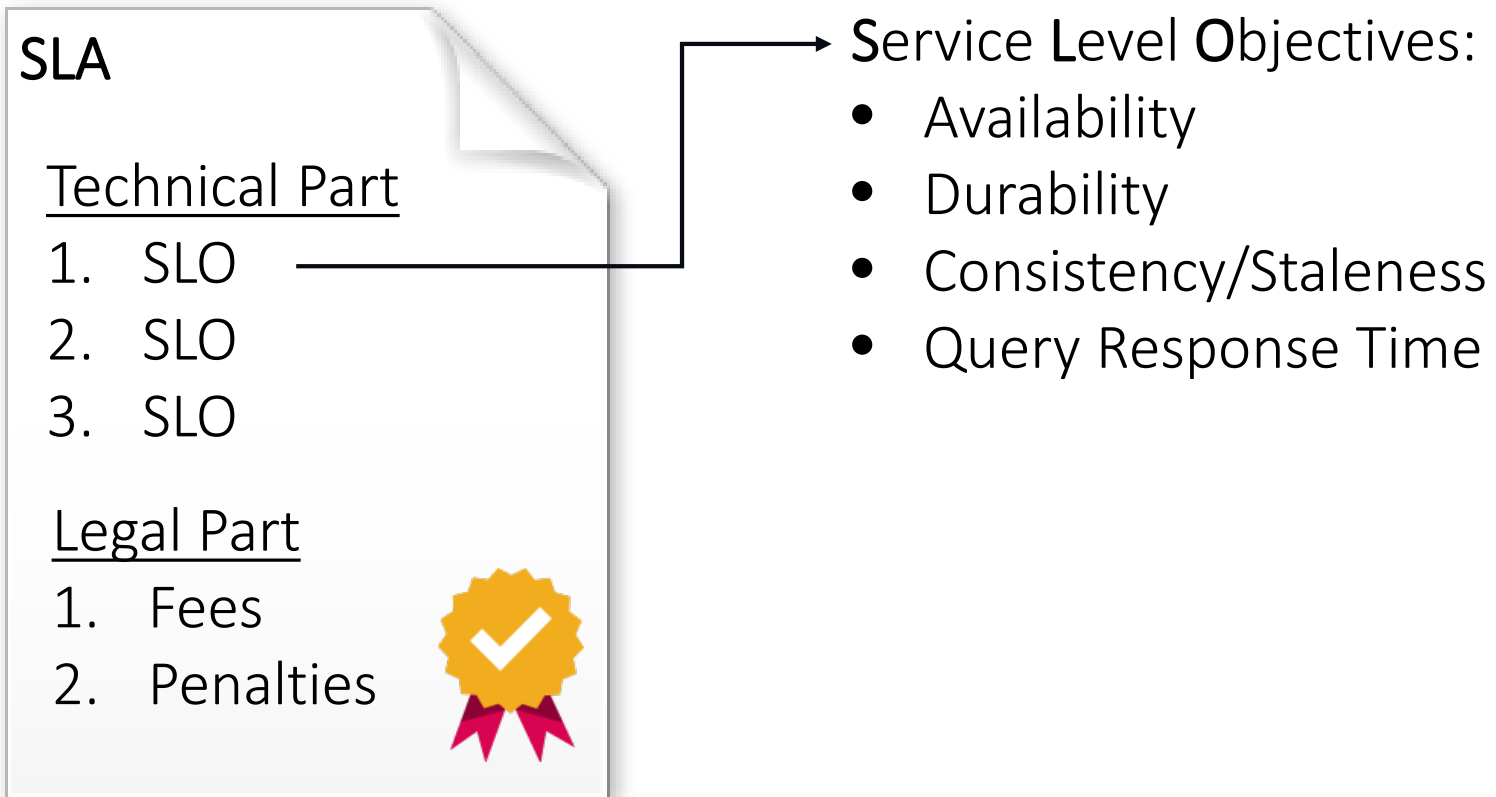
Backend-as-a-Service

DBaaS with embedded custom and predefined application logic

- rather recent trend
 - progress currently driven by industry projects
(similarly to early cloud computing and big data processing)
 - structured research yet to be established
- most comfortable approach for applications
- but many unsolved problems
 - latency challenge: all clients access the service via high-latency WAN
 - persistence on top of one single database technology
 - service/NoSQL-DBS selection problem
 - usually, tenants colocated on a shared database cluster
→ database system configuration (e.g., the replication protocol)
prescribes the guarantees for each tenant

Service Level Agreements (SLAs)

Specification of Application/Tenant Requirements



Service Level Agreements

Expressing application requirements

Functional Service Level Objectives

- Guarantee a „feature“
- Determined by database system
- *Examples:* transactions, join



Non-Functional Service Level Objectives

- Guarantee a certain *quality of service* (QoS)
- Determined by database system and service provider
- *Examples:*
 - **Continuous:** response time (latency), throughput
 - **Binary:** Elasticity, Read-your-writes

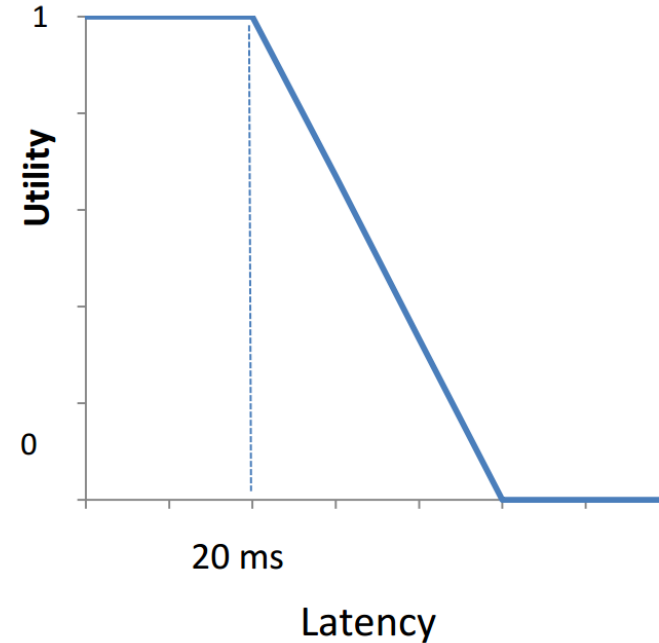
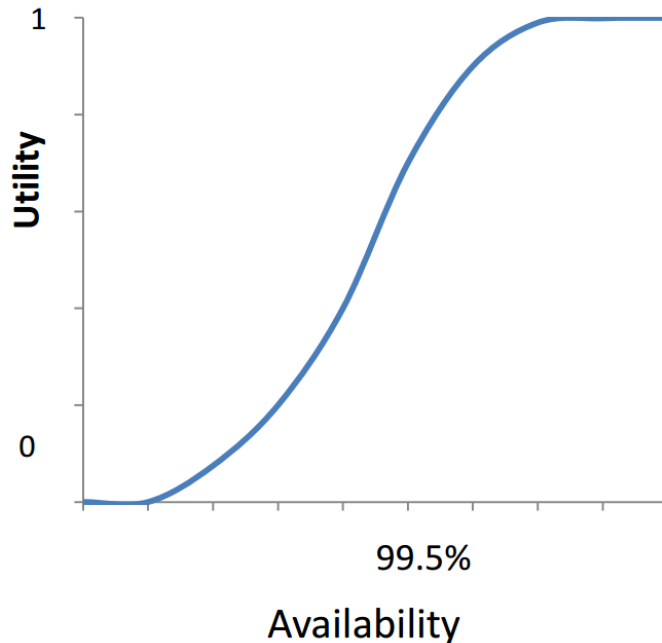


Service Level Objectives

Making SLOs measurable through utilities

Utility expresses „value“ of a continuous non-functional requirement:

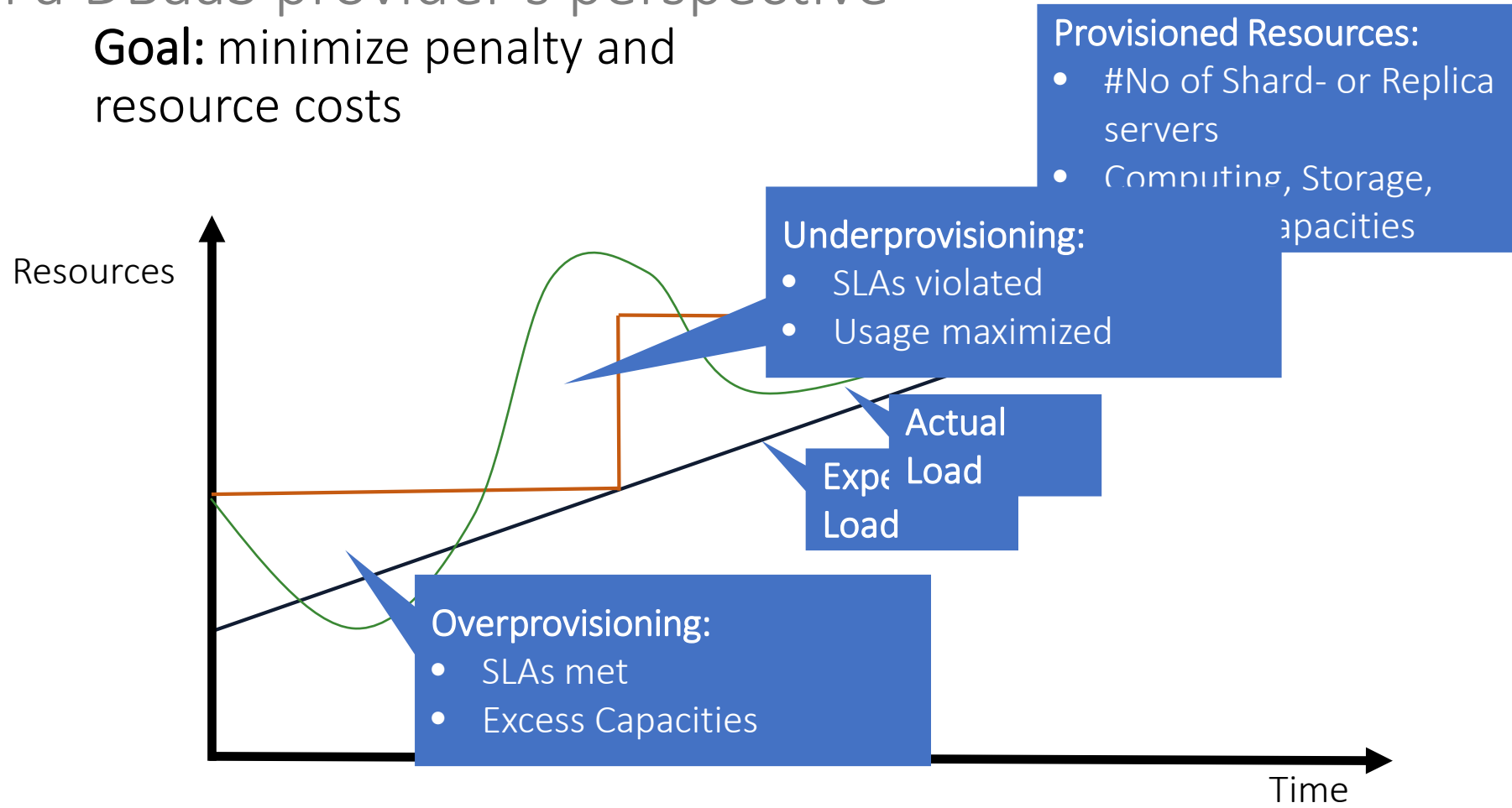
$$f_{utility}(metric) \rightarrow [0,1]$$



Resource & Capacity Planning




From a DBaaS provider's perspective

Goal: minimize penalty and resource costs












SLAs in the wild









Most DBaaS systems offer no SLAs, or only a simple uptime guarantee

	Model	CAP	SLAs
SimpleDB	Table-Store (NoSQL Service)	CP	
Dynamo-DB	Table-Store (NoSQL Service)	CP	
Azure Tables	Table-Store (NoSQL Service)	CP	99.9% uptime
AE, Cloud DataStore	Entity-Group Store (NoSQL Service)	CP	
S3, Az. Blob, GCS	Object-Store (NoSQL Service)	AP	99.9% uptime (S3)

Managed NoSQL Services

	Model	CAP	Scans	Sec. Indices	Largest Cluster	Learning	Lic.	DBaaS
HBase	Wide-Column	CP	Over Row Key		~700	1/4	Apache	 (EMR)
MongoDB	Document	CP	yes		>100 <500	4/4	GPL	
Riak	Key-Value	AP			~60	3/4	Apache	 (Softlayer)
Cassandra	Wide-Column	AP	With Comp. Index		>300 <1000	2/4	Apache	
Redis	Key-Value	CA	Through Lists, etc.	manual	N/A	4/4	BSD	

Proprietary Database Services

	Model	CAP	Scans	Sec. Indices	Queries	API	Scale-out	SLA
SimpleDB	Table-Store	CP	Yes (as queries)	Auto-matic	SQL-like (no joins, groups, ...)	REST + SDKs		
Dynamo-DB	Table-Store	CP	By range key / index	Local Sec. Global Sec.	Key+Cond. On Range Key(s)	REST + SDKs	Automatic over Prim. Key	
Azure Tables	Table-Store	CP	By range key		Key+Cond. On Range Key	REST + SDKs	Automatic over Part. Key	99.9% uptime
AE/Cloud DataStore	Entity-Group	CP	Yes (as queries)	Auto-matic	Conjunct. of Eq. Predicates	REST/ SDK, JDO,JPA	Automatic over Entity Groups	
S3, Az. Blob, GCS	Blob-Store	AP				REST + SDKs	Automatic over key	99.9% uptime (S3)

A person with long hair is seen from behind, sitting on a dark pier or ledge. They are looking out over a body of water towards a port area. In the background, several large port cranes are visible, their lights glowing against a sunset sky with warm orange and yellow hues. The water reflects the lights from the cranes and the sky. A semi-transparent white rectangular box is overlaid on the left side of the image, containing the text 'Our SCDM Approach'.

Our SCDM Approach

NoSQL Database Systems: A Survey and Decision Guidance

Felix Gessert, Wolfram Wingerath, Steffen Friedrich, and Norbert Ritter

Universität Hamburg, Germany

{gessert, wingerath, friedrich, ritter}@informatik.uni-hamburg.de

Abstract. Today, data is generated and consumed at unprecedented scale. This has lead to novel approaches for scalable data management subsumed under the term “NoSQL” database systems to handle the over-

1. Aim at fully managed Backend (BaaS)

of contrasting the implementation specifics of individual representatives, we propose a comparative classification model that relates functional and non-functional requirements to techniques and algorithms employed in NoSQL databases. This NoSQL Toolbox allows us to derive a simple decision tree to help practitioners and researchers filter potential system candidates based on central application requirements.

1 Introduction

Traditional relational database management systems (RDBMSs) provide

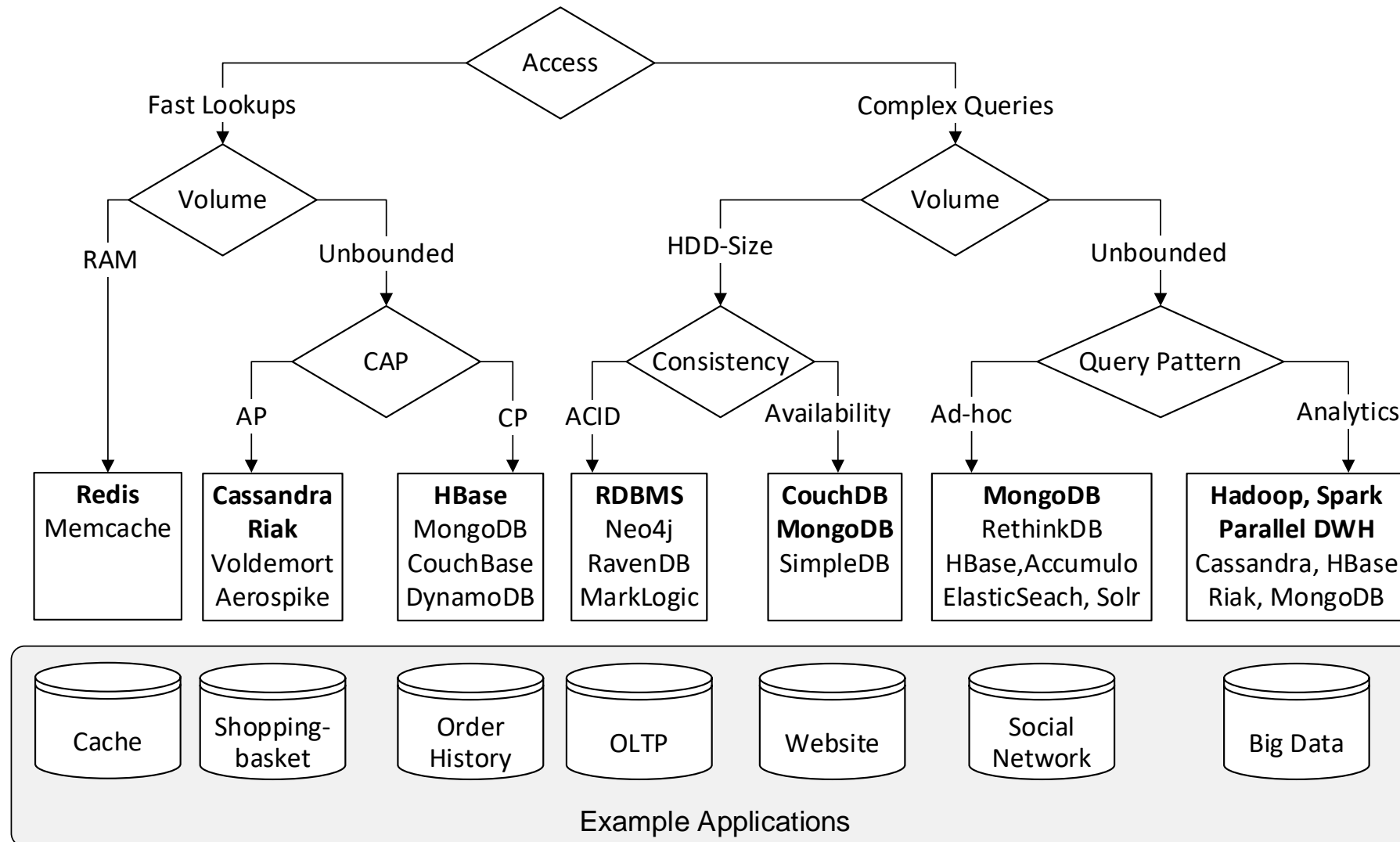
2. Exploit modern (NoSQL) Database Technology

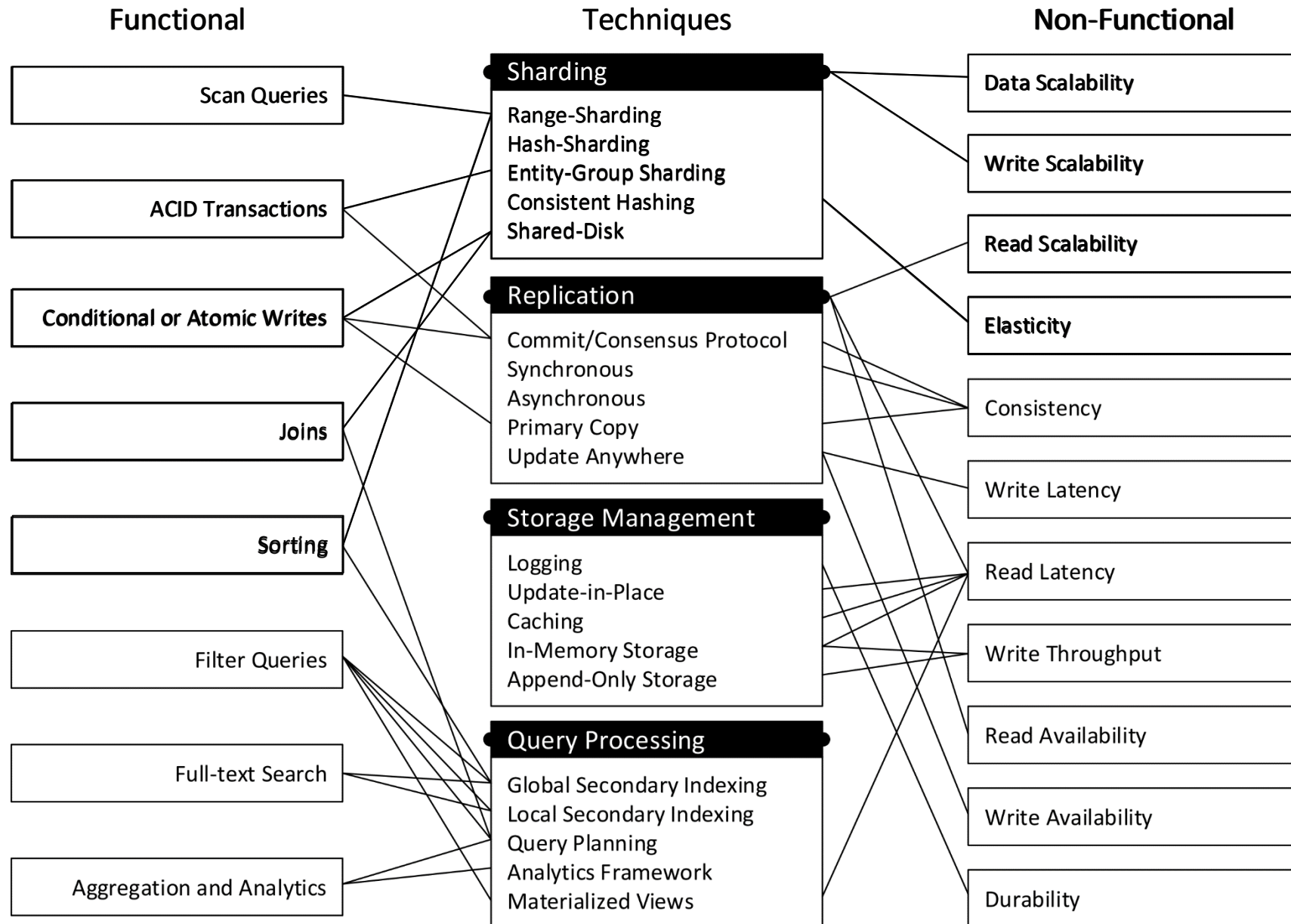
so vast that it cannot be stored or processed by traditional database solutions. User-generated content in social networks or data retrieved from large sensor networks are only two examples of this phenomenon commonly referred to as **Big Data** [35]. A class of novel data storage systems able to cope with Big Data are subsumed under the term **NoSQL databases**, many of which offer horizontal scalability and higher availability than relational databases by sacrificing querying capabilities and consistency guarantees. These trade-offs are pivotal for service-oriented computing and as-a-service models, since any stateful service can only be as scalable and fault-tolerant as its underlying data store.

There are dozens of NoSQL database systems and it is hard to keep track of where they excel, where they fail or even where they differ, as implementation details change quickly and feature sets evolve over time. In this article, we therefore aim to provide an overview of the NoSQL landscape by discussing employed concepts rather than system specificities and explore the requirements typically posed to NoSQL database systems, the techniques used to fulfil these requirements and the trade-offs that have to be made in the process. Our focus lies on key-value, document and wide-column stores, since these NoSQL categories

<http://www.baqend.com/files/nosql-survey.pdf>

NoSQL Decision Tree





System Properties

According to the NoSQL Toolbox

- ▶ For fine-grained system selection:

Functional Requirements								
	Scan Queries	ACID Transactions	Conditional Writes	Joins	Sorting	Filter Query	Full-Text Search	Analytics
Mongo	x		x		x	x	x	x
Redis	x	x	x					
HBase	x		x		x			x
Riak							x	x
Cassandra	x		x		x		x	x
MySQL	x	x	x	x	x	x	x	x

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	X	X	X		X	X	X		X		X
Redis			X		X	X	X	X	X		X
HBase	X	X	X	X	X	X		X			X
Riak	X	X	X	X		X	X	X	X	X	X
Cassandra	X	X	X	X		X		X	X	X	X
MySQL			X		X						X

System Properties

According to the NoSQL Toolbox

- ▶ For fine-grained system selection:

	Techniques																			
	Range-Sharding	Hash-Sharding	Entity-Group Sharding	Consistent Hashing	Shared-Disk	Transaction Protocol	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	Materialized Views
Mongo	x	x					x	x	x		x		x	x	x		x	x	x	
Redis								x	x		x		x							
HBase	x						x		x		x		x		x					
Riak		x		x				x		x	x	x	x			x	x		x	
Cassandra		x		x				x		x	x		x		x	x	x			x
MySQL					x			x	x		x	x	x				x	x		



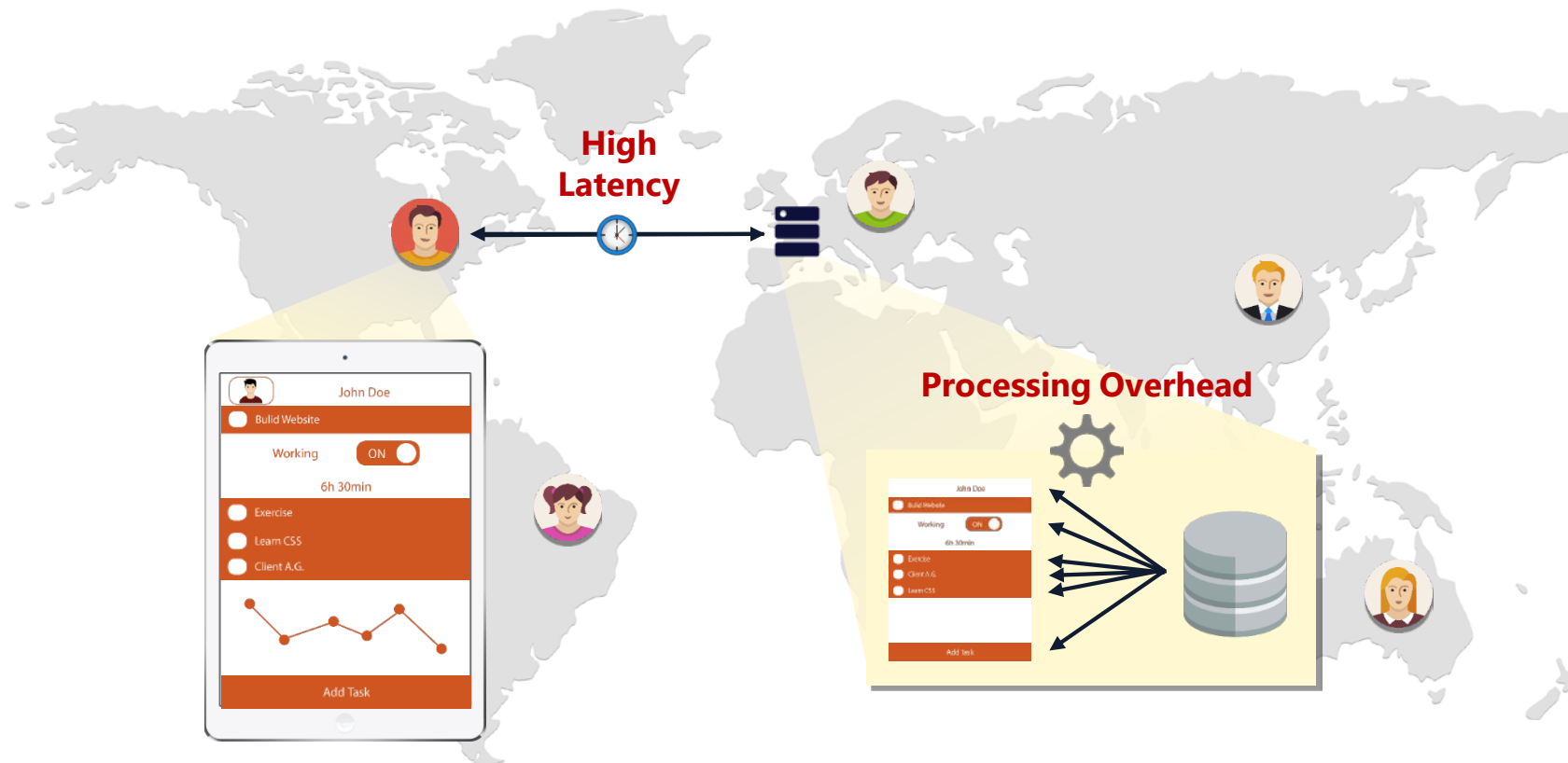
NoSQL DBS support applications in achieving horizontal scalability and backend performance through differentiated trade-offs in functionality and consistency!

**3. Consider the entire path
from the (mobile) application
through the net
to the data backend!**

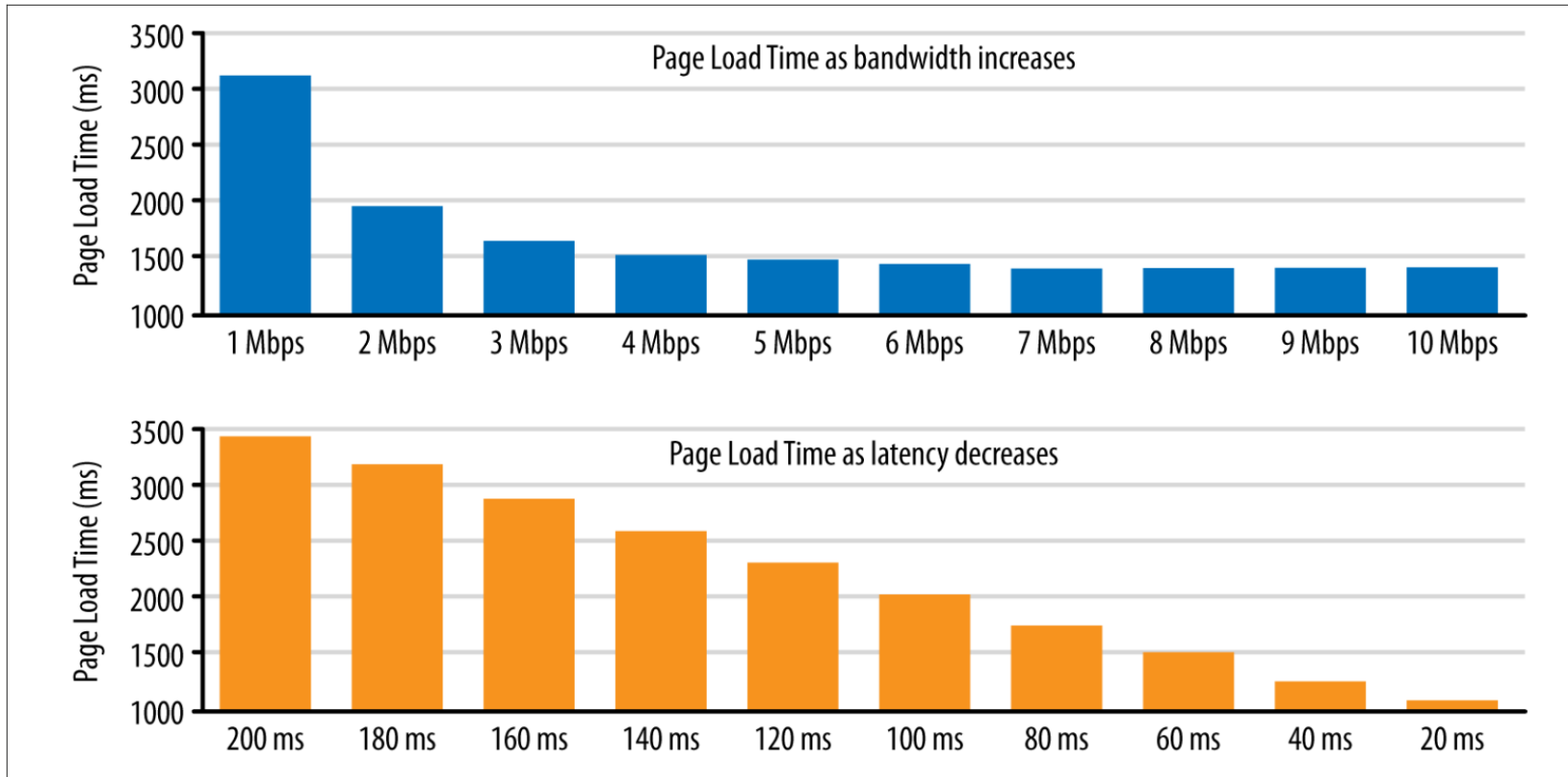


Challenge: Slow Websites / mobile Apps

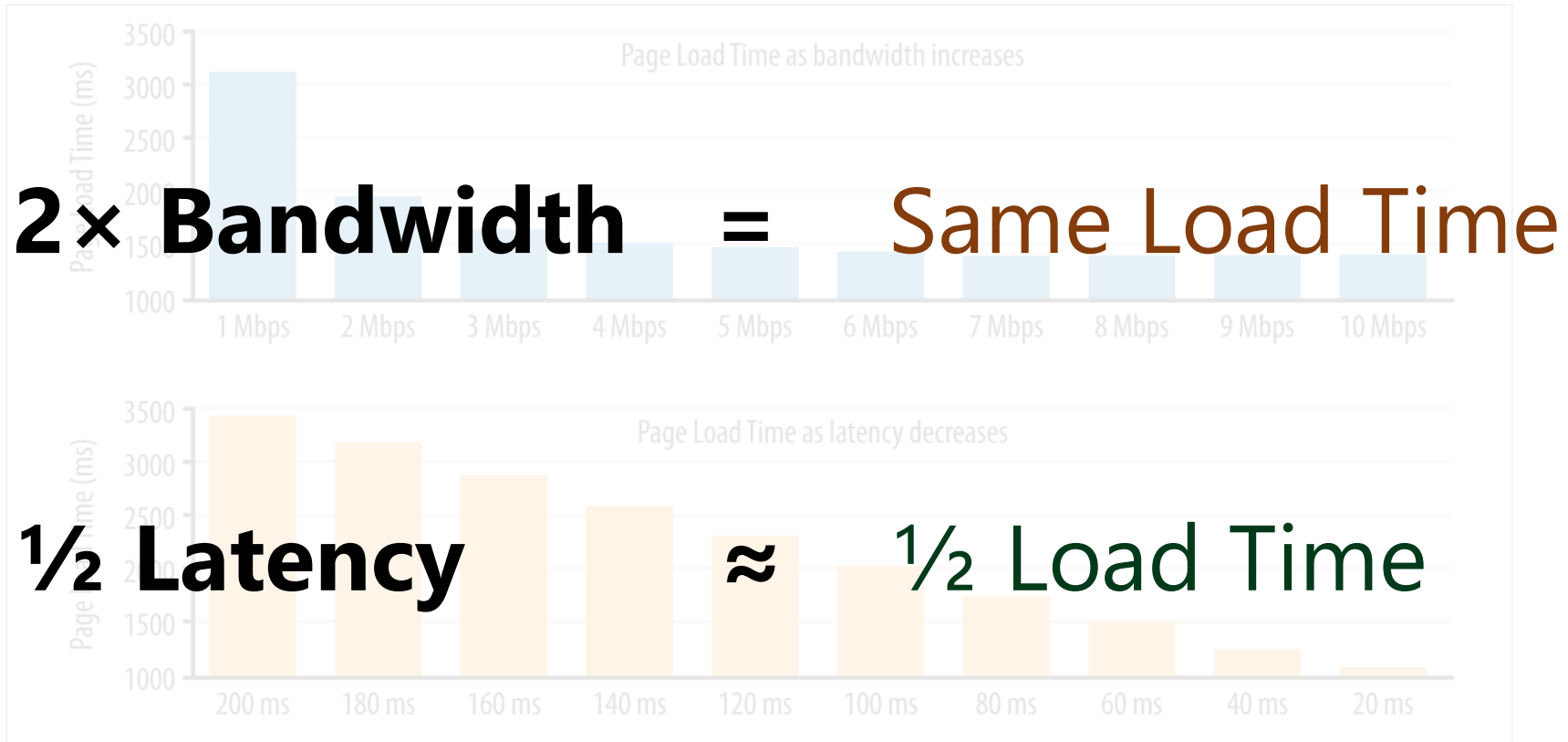
Two Bottlenecks: Latency and Processing



Network Latency: Impact



Network Latency: Impact



Orestes Architecture

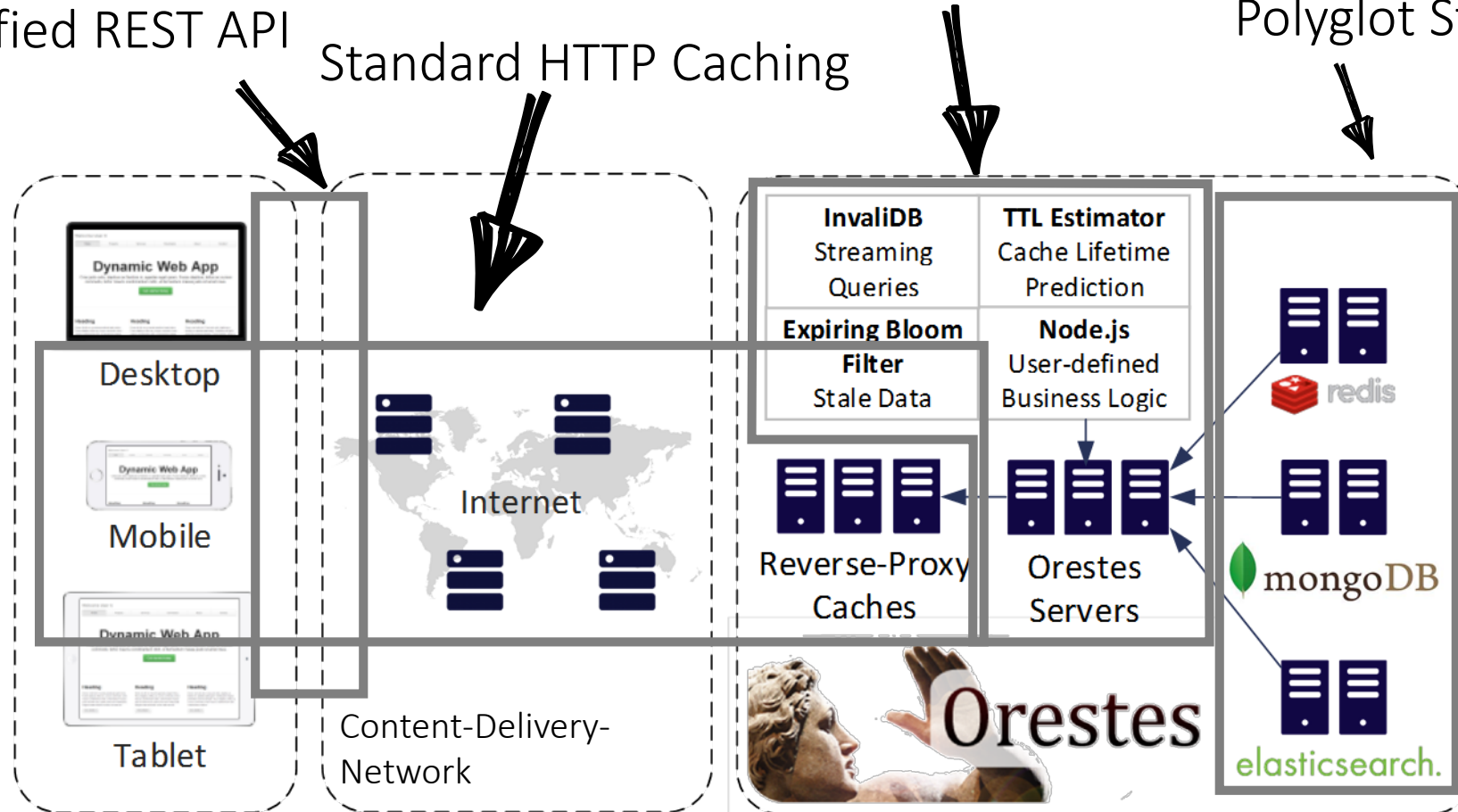
Infrastructure

Backend-as-a-Service Middleware:
Caching, Transactions, Schemas,
Invalidation Detection, ...

Unified REST API

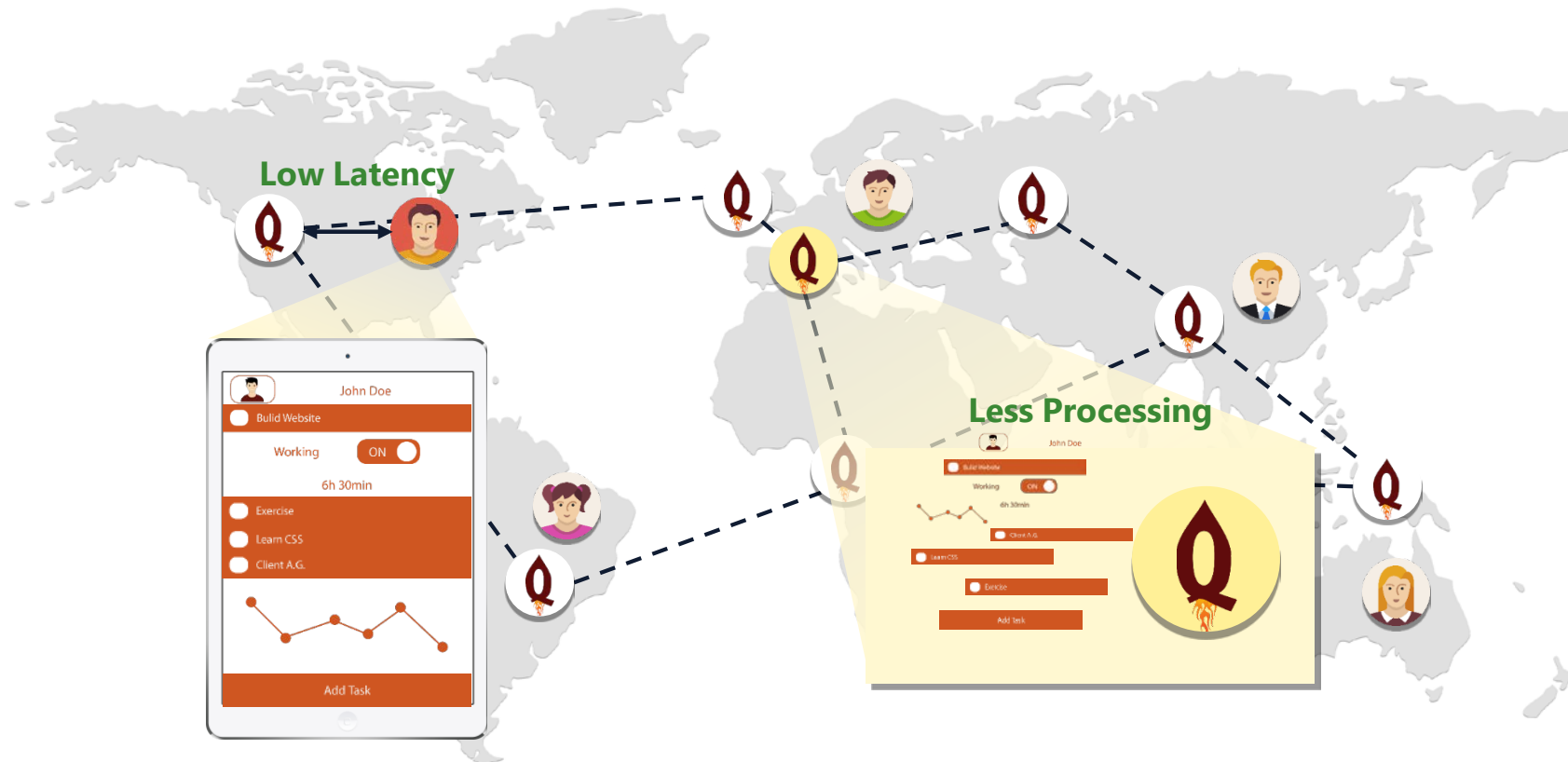
Standard HTTP Caching

Polyglot Storage



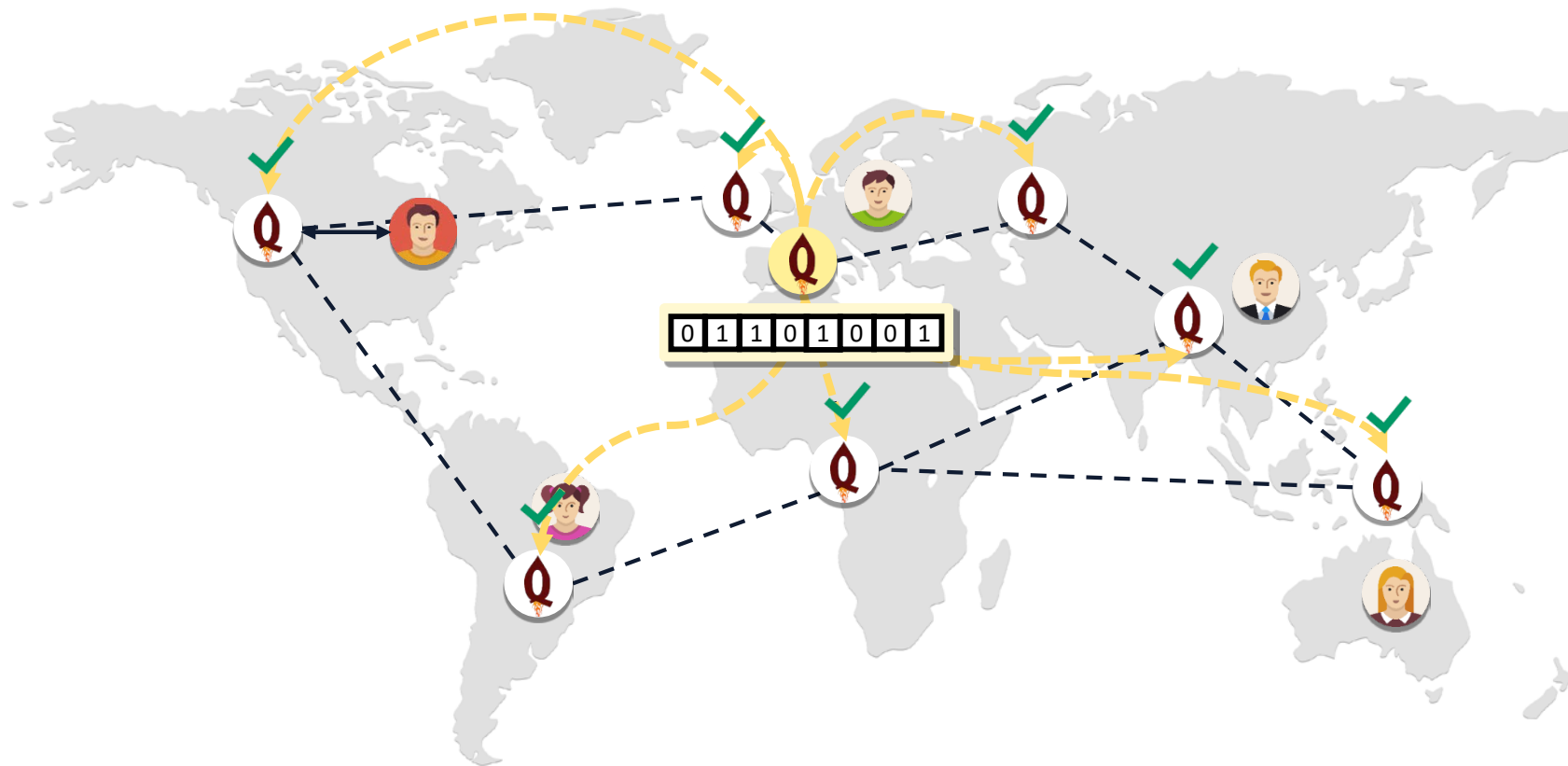
Solution: Global Caching

Fresh Data From Distributed Web Caches



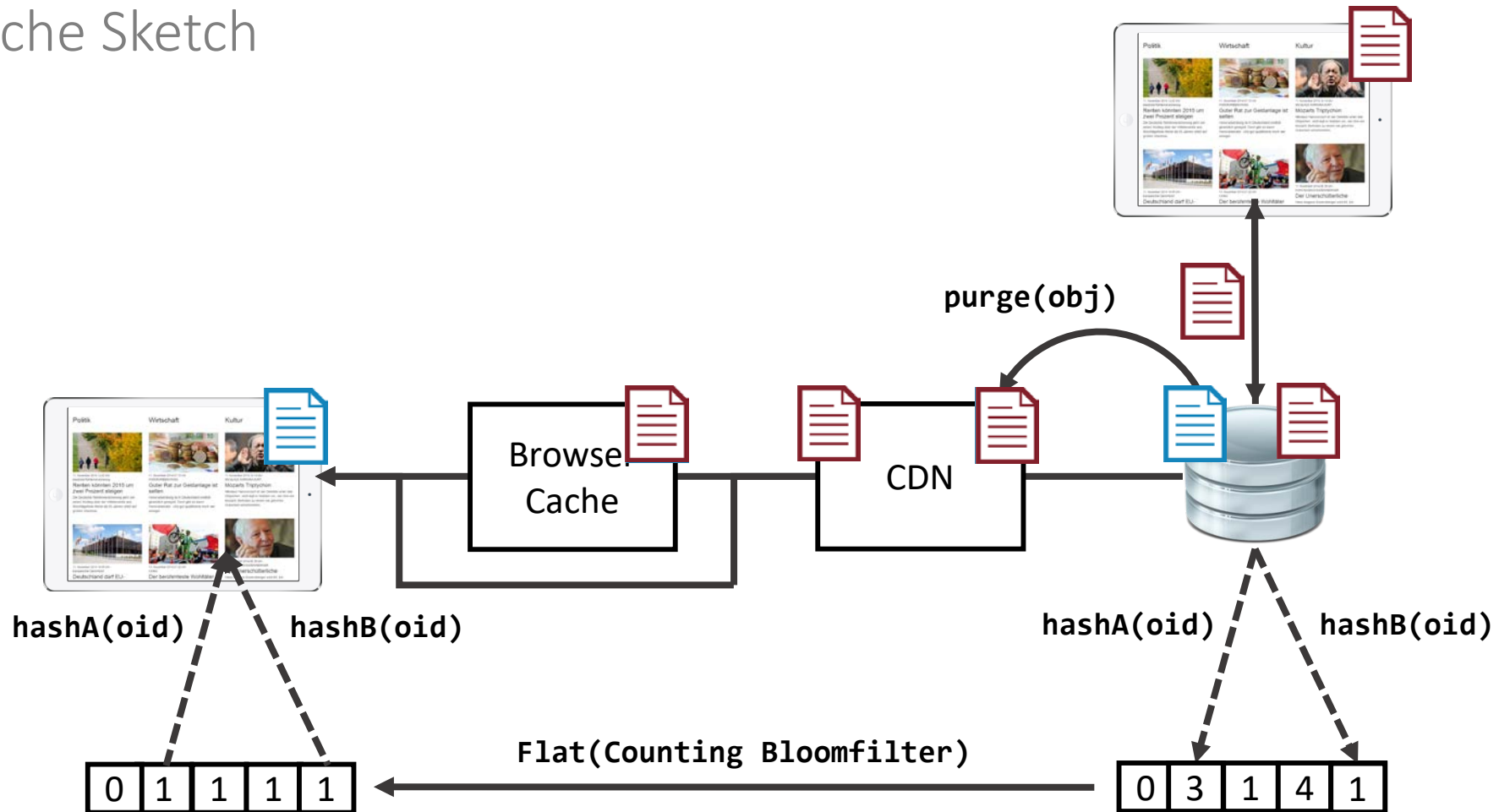
New Caching Algorithms

Solve Consistency Problem



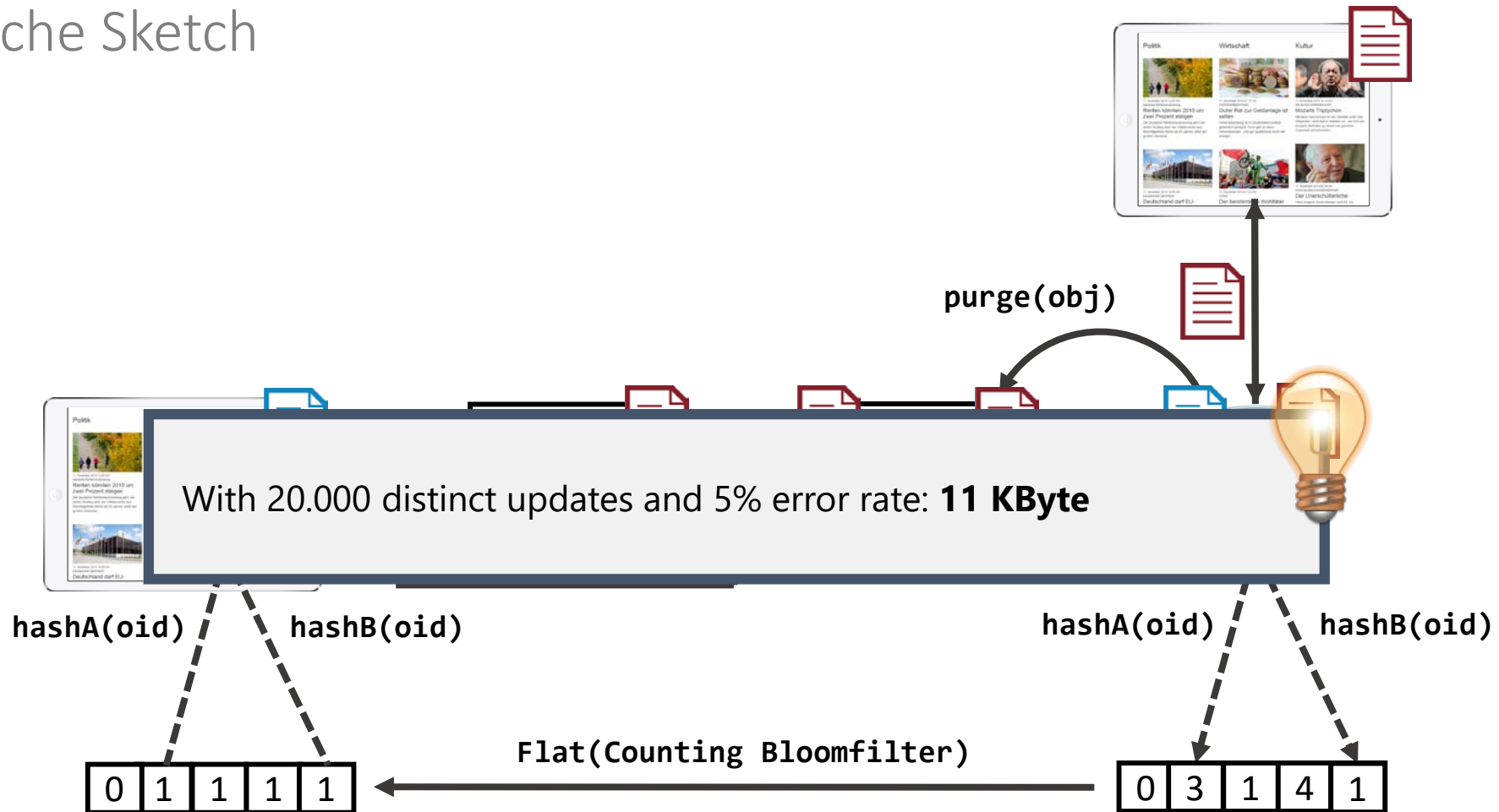
Consistent Web Caching

The Cache Sketch



Consistent Web Caching

The Cache Sketch





4. Make the backend push-based,
additionally (real-time queries)!

Challenge: Real-Time Queries



C₁: Scalability:

- ▶ Handle additional queries
- ▶ Handle increasing throughput



C₂: Expressiveness:

- ▶ Content search? Composite filters?
- ▶ Ordering? Limit? Offset?

Research Question: „How can expressive push-based real-time queries be implemented on top of an existing pull-based database in a scalable and generic manner?“



C₃: Legacy Support:

- ▶ Real-time queries for *existing databases*
- ▶ *Decouple* OLTP from real-time workloads



C₄: Abstract API
















- ▶ Data independence
- ▶ Self-maintaining queries

Overview Real-time DBSs



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

Overview Stream-Processors

	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)			
Processing Model	one-at-a-time	micro-batch	one-at-a-time	micro-batch	one-at-a-time
Backpressure			no (buffering)		
Ordering		between batches	within partitions	between batches	within partitions
Elasticity					

Overview Stream-Processors

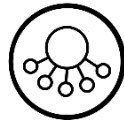
	Storm	Trident	Samza	Spark Streaming	Flink (streaming)
Strictest Guarantee	at-least-once	exactly-once	at-least-once	exactly-once	exactly-once
Achievable Latency	<<100 ms	<100 ms	<100 ms	<100 ms	<100 ms
State Management	no	no	no	no	no
Batching	time	time	time	micro-batch	one-at-a-time
Buffering	no	✓	no (buffering)	✓	✓
Batching Scope	between	between	within	between batches	within partitions
Elasticity	✓	✓	✗	✓	✗

“Storm [...] and Flink [...] show sub-second latencies at relatively high throughputs with Storm having the lowest 99th percentile latency. Spark streaming [...] supports high throughputs, but at a relatively higher latency.”

From <https://yahooeng.tumblr.com/post/135321837876/benchmarking-streaming-computation-engines-at>

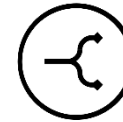
InvaliDB: A Scalable Real-Time Database Design

System Model & Overview



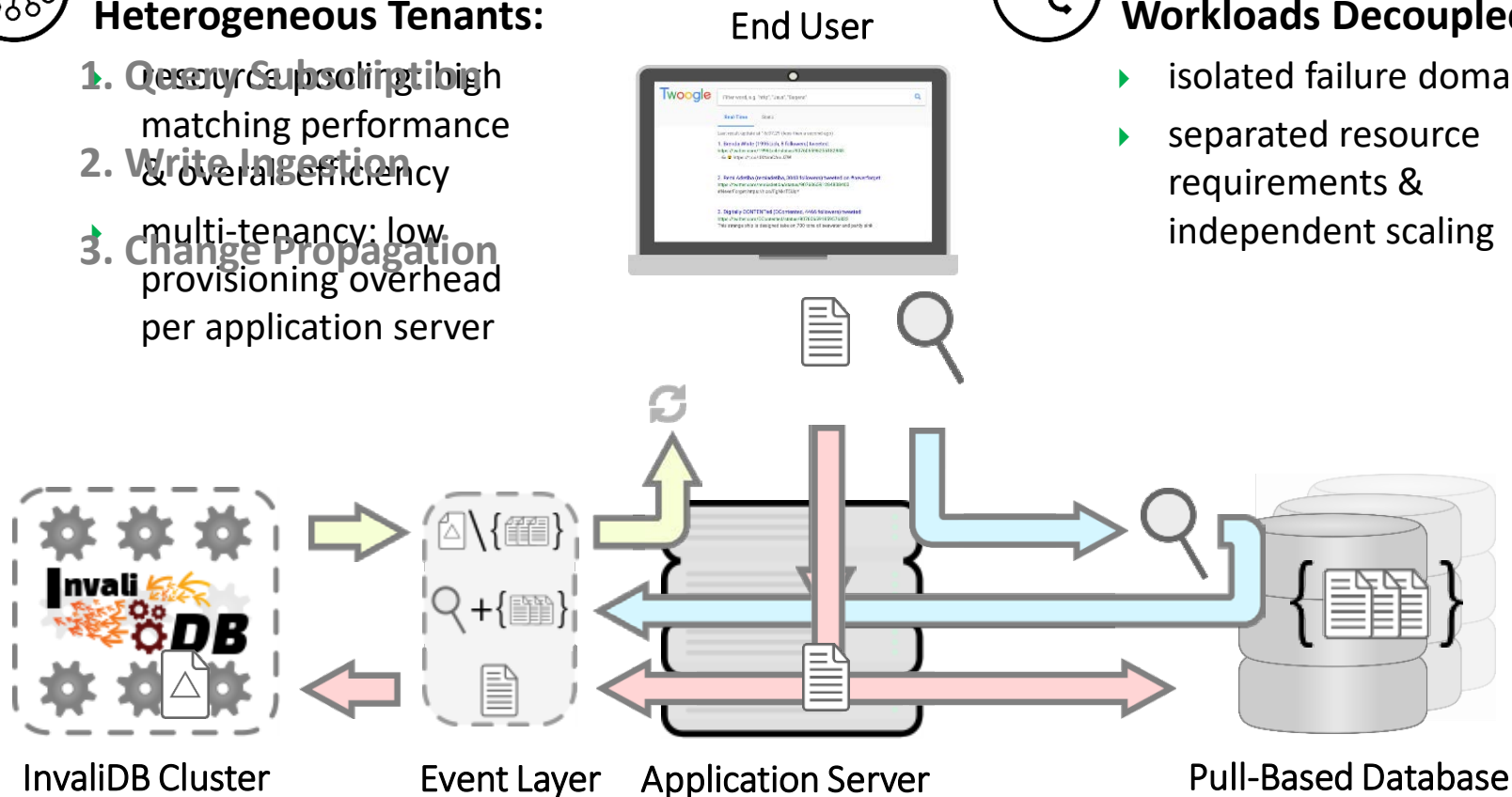
Realtime-as-a-Service For Heterogeneous Tenants:

1. Query Subpooling: high matching performance
2. Write Ingestion & overall efficiency
3. Change Propagation: multi-tenancy: low provisioning overhead per application server



Real-Time & OLTP Workloads Decoupled:

- ▶ isolated failure domains
- ▶ separated resource requirements & independent scaling



InvaliDB: A Scalable Real-Time Database Design

Two-Dimensional Workload Partitioning



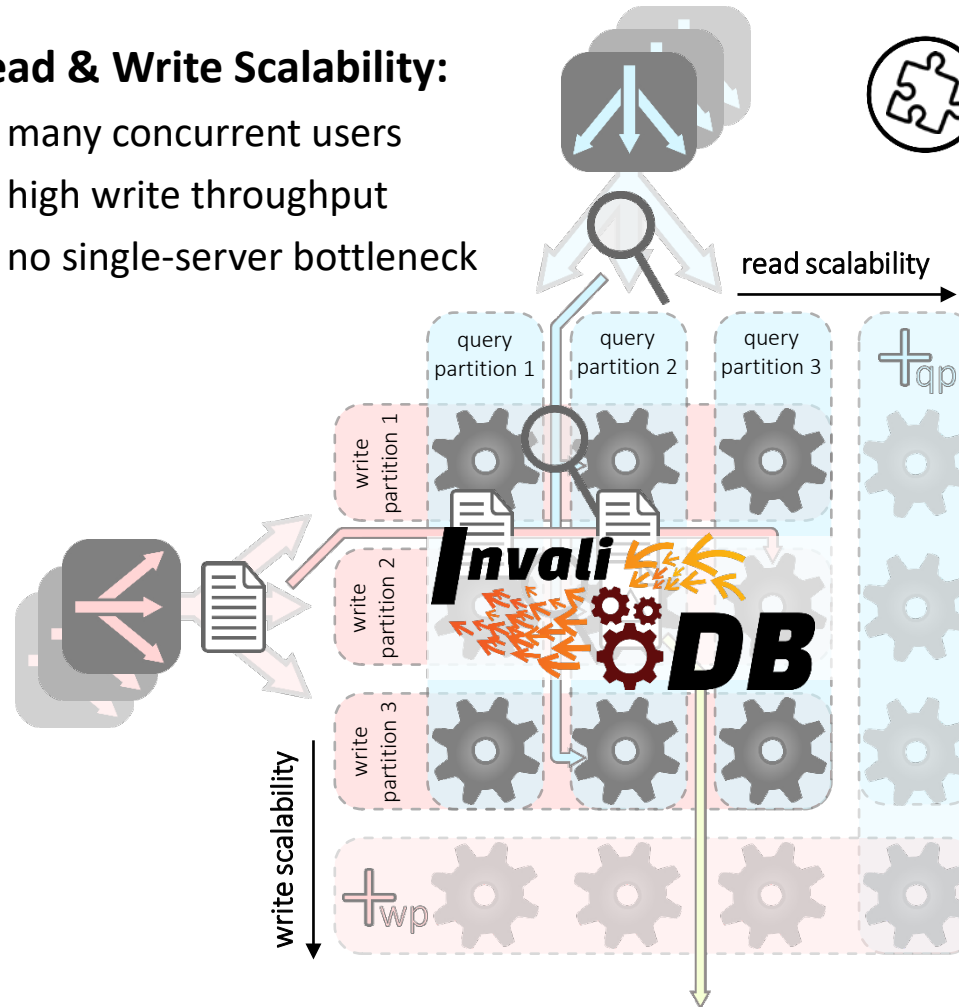
Read & Write Scalability:

- ▶ many concurrent users
- ▶ high write throughput
- ▶ no single-server bottleneck



Pluggable Query Engine:

- ▶ legacy-compatibility
- ▶ multi-tenancy across databases



W. Wingerath, Scalable Push-based Real-time Queries on top of Pull-based Databases. PhD thesis, University of Hamburg, 2019.

realises 1. to 4. as commercially available service

Platform



- Platform for building (Progressive) **Web Apps**
- **15x** Performance Edge
- Faster **Development**

Speed Kit

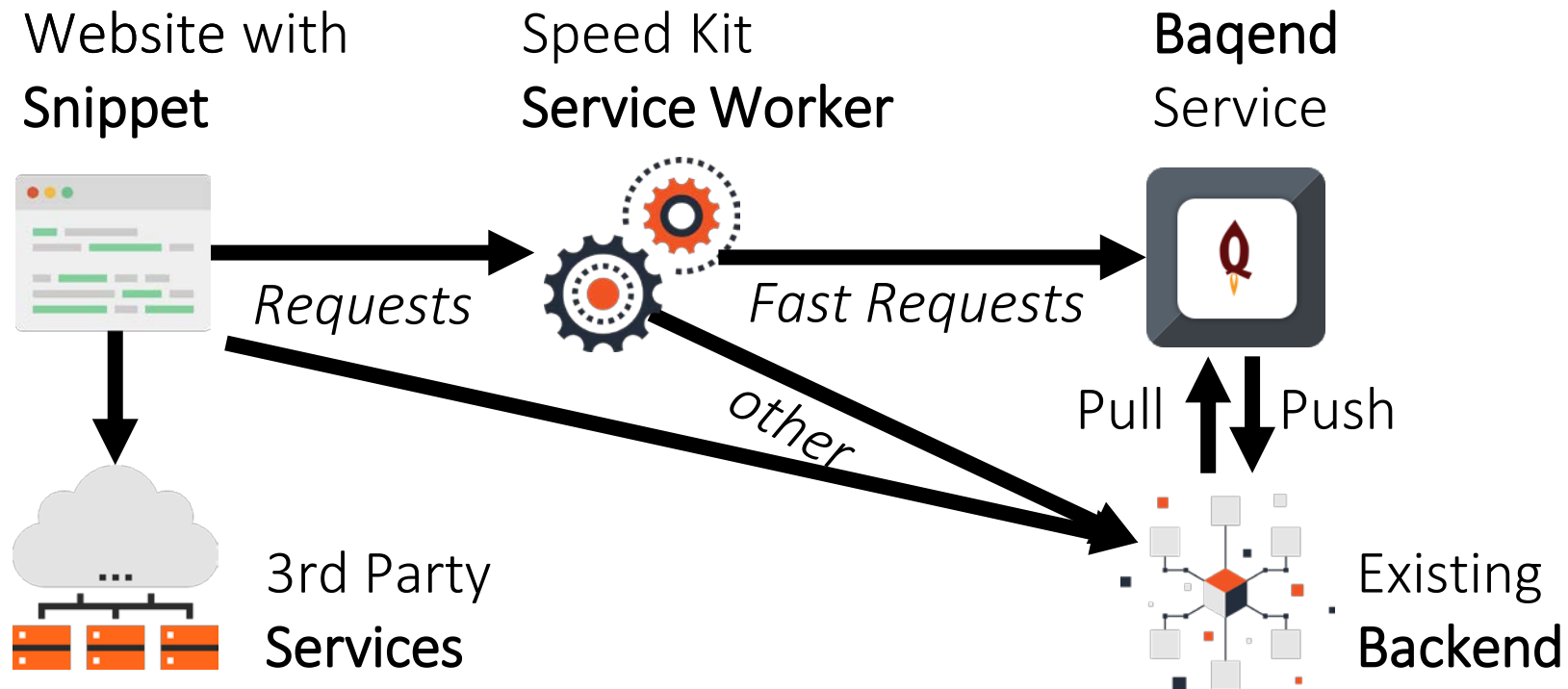


- Turns Existing Sites into **PWAs**
- **50-300% Faster** Loads
- **Offline** Mode

Try It Out!

Speed Kit

Baqend Caching for Legacy Websites



Speed Kit

Measure Your Page Speed!

<https://test.speed-kit.com/>



Speed Kit

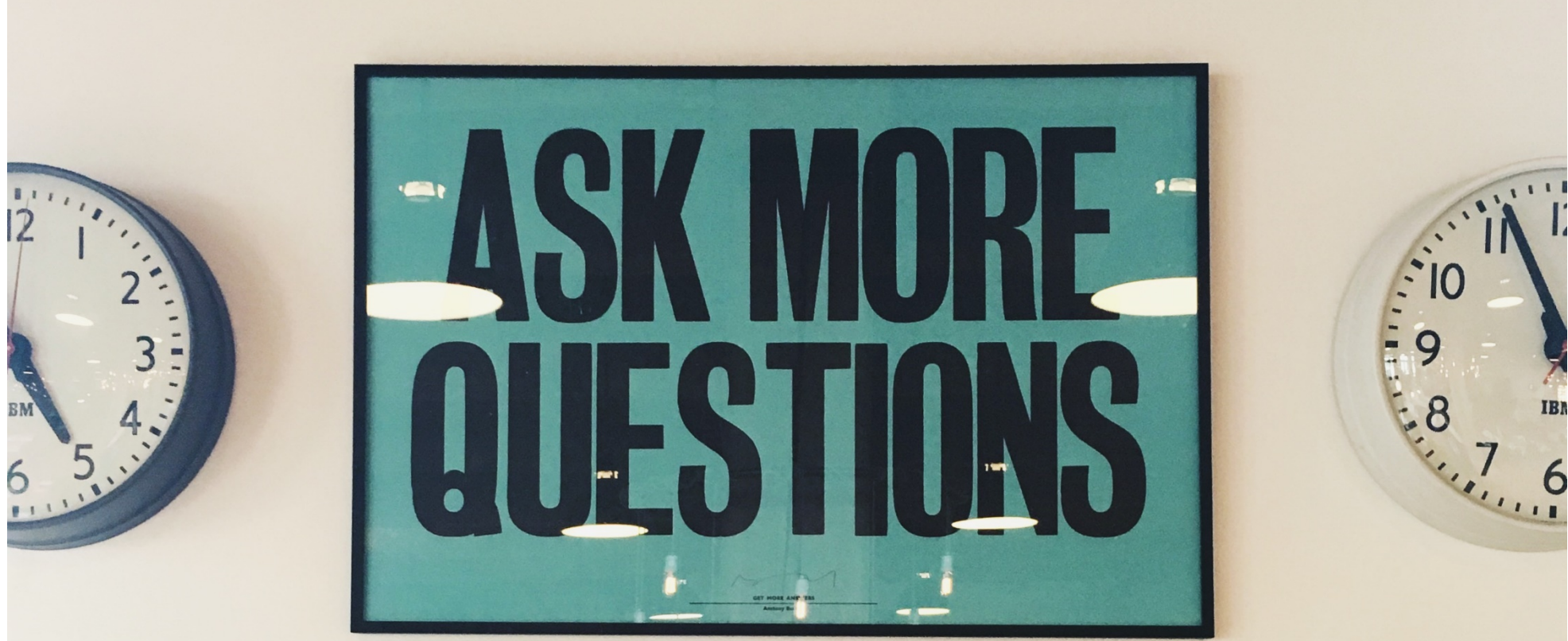
Built for Market Leaders

For a large e-commerce company like Baur, supreme performance and a snappy user experience are vital. **Speed Kit** helps Baur.de stay ahead of the competition by accelerating page loads through **cutting-edge technology**. Finally, there is a German player in the web performance market that does not only pioneer a **superior approach**, but also shines through competent onboarding and immediate support.

Revenue: 1 000 000 000 € for 2018
Traffic: 70 000 000 PIs per month

BAUR 

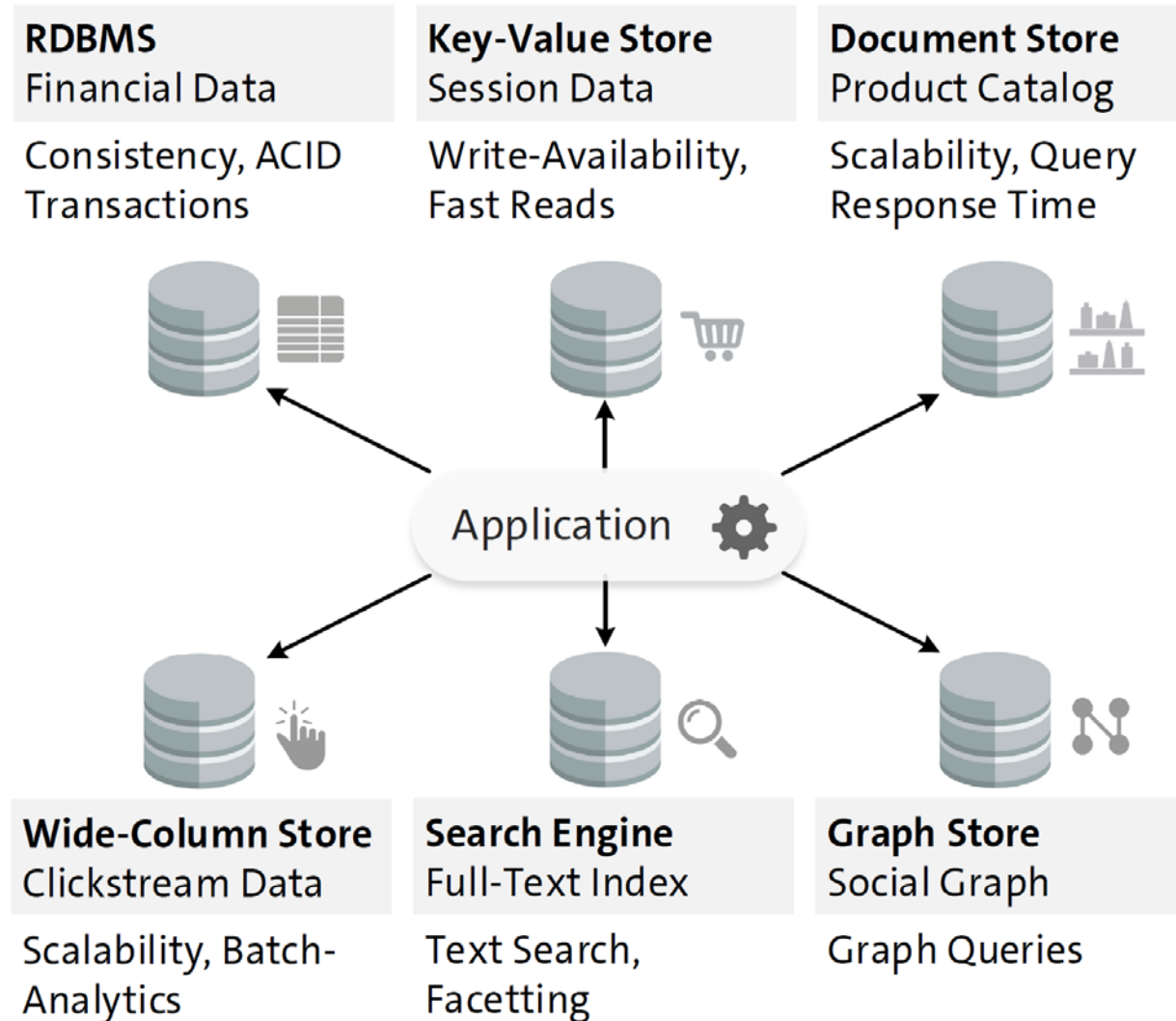
A member of the otto group



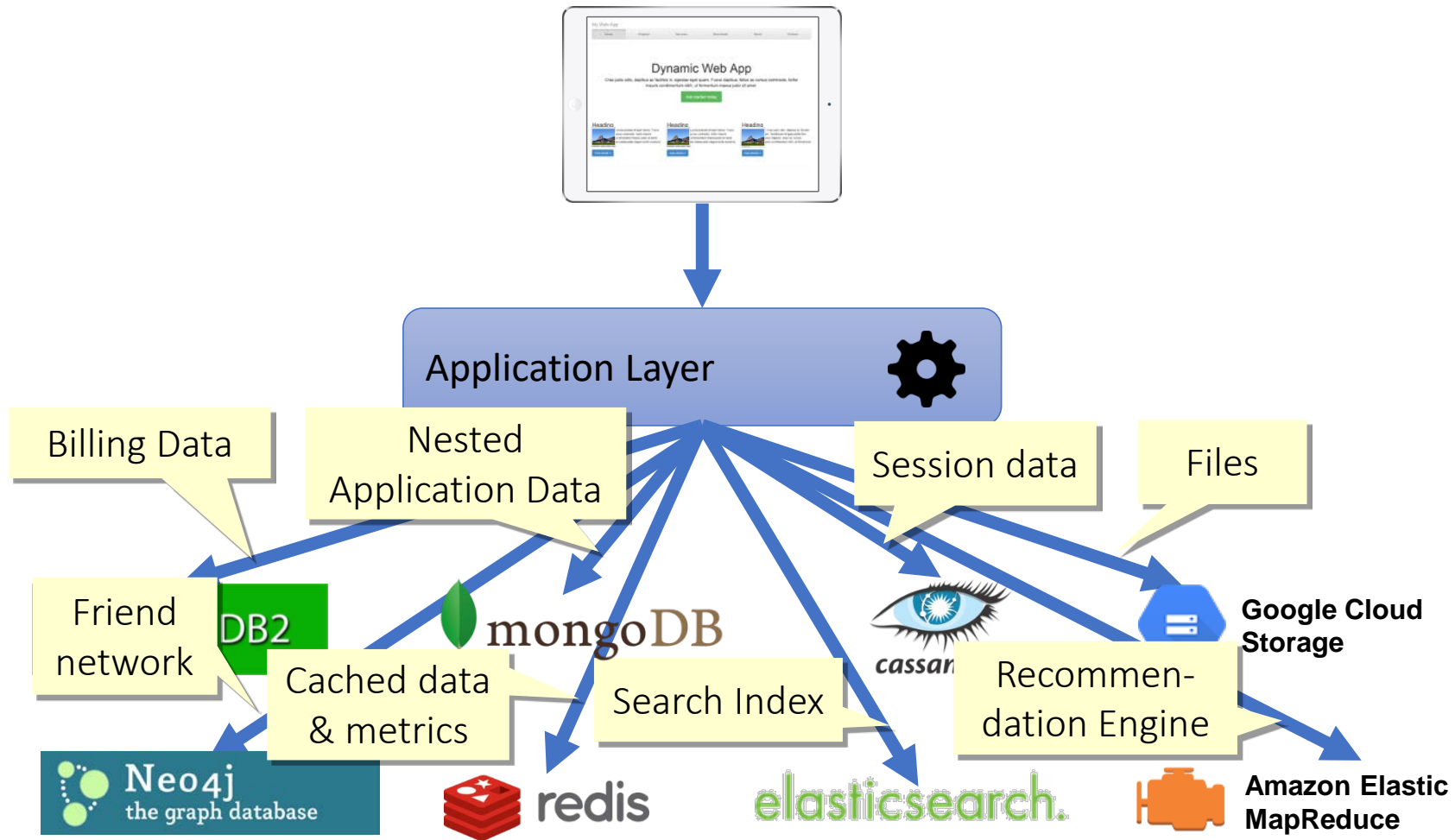
Vision



5. Provide Polyglot Persistence!



Challenge: ,automated' mediation



Challenge: ,automated' mediation

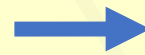


Research Question:

Can we automate the

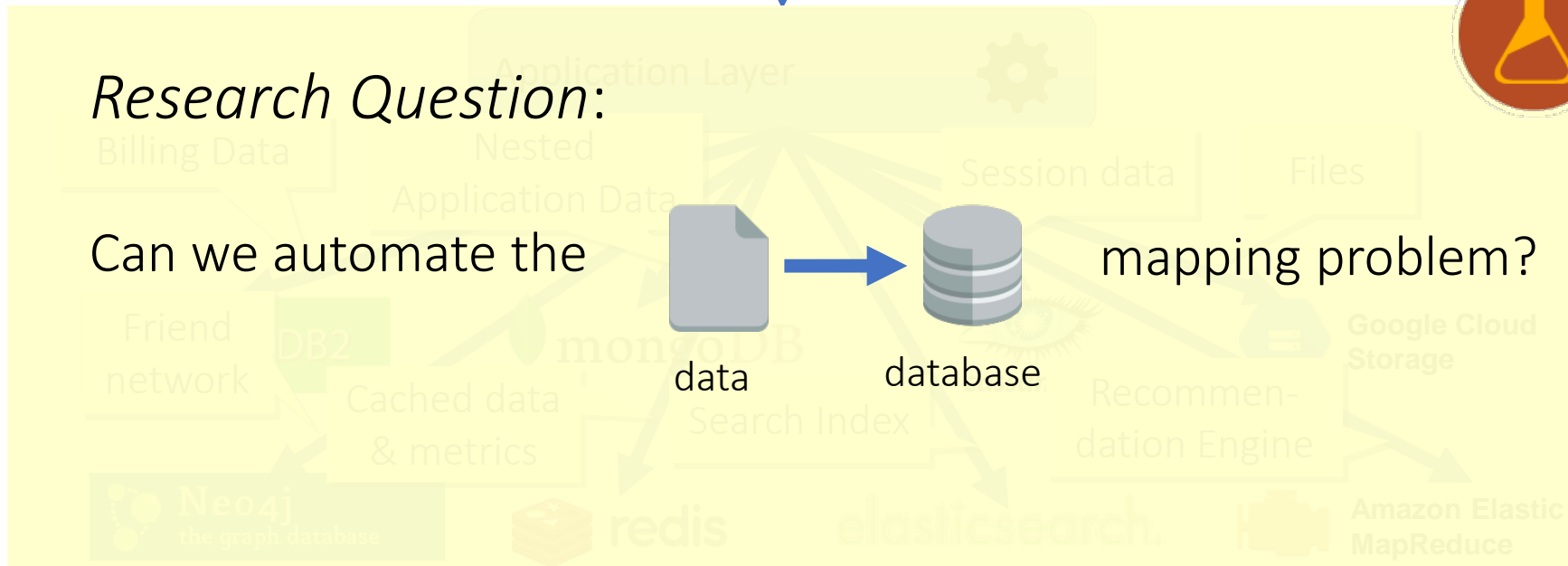


data



database

mapping problem?

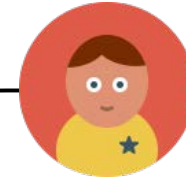


Vision

Schemas can be annotated with requirements



- Write Throughput > 10,000 RPS
- Read Availability > 99.9999%
- Scans = **true**
- Full-Text-Search = **true**
- Monotonic Read = **true**



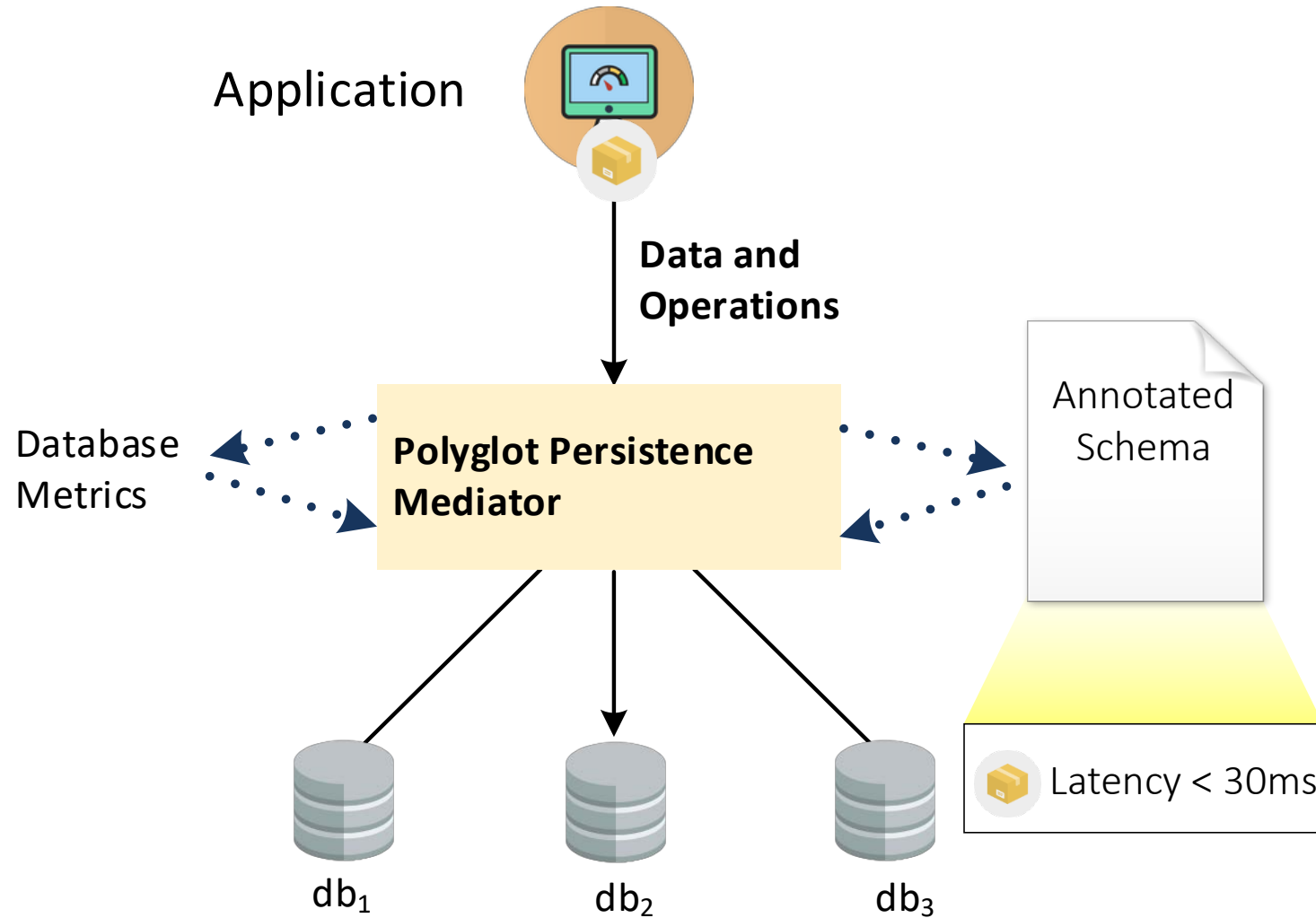
Schema

DBs
Tables
Fields



Vision

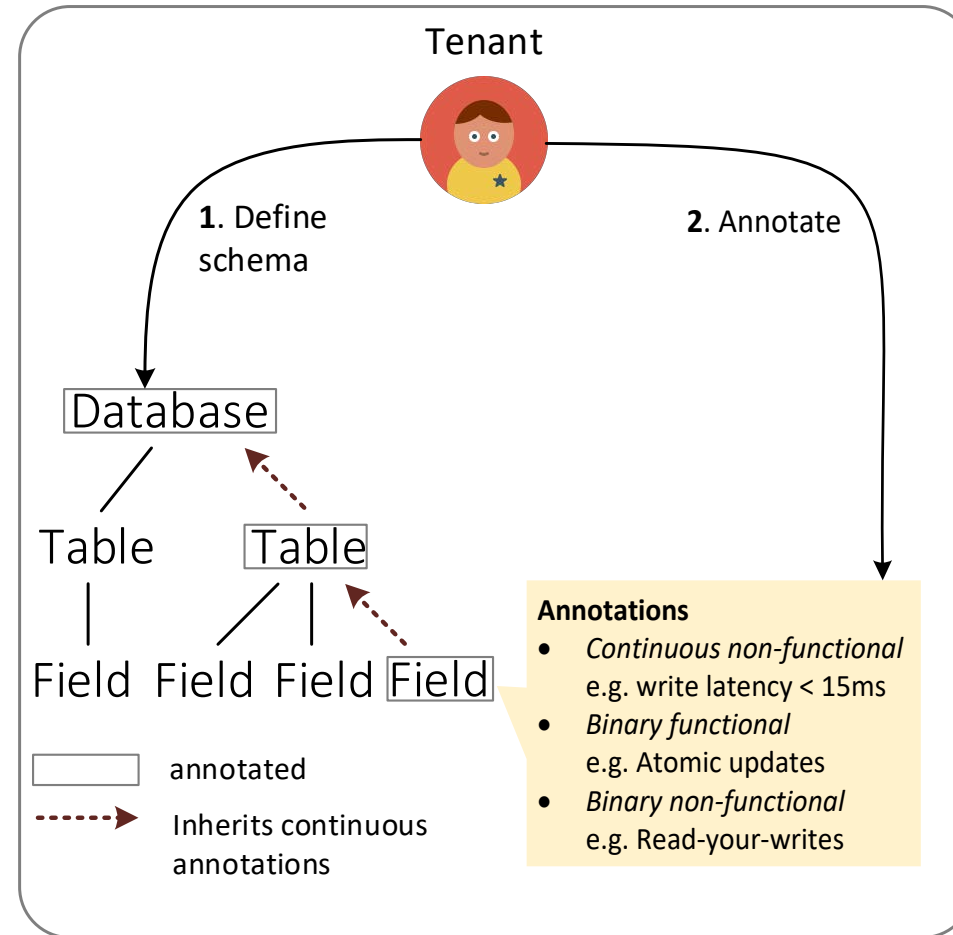
The Polyglot Persistence Mediator chooses the database



Step I - Requirements

Expressing the application's needs

Annotation	Type	Annotated at
Read Availability	Continuous	*
Write Availability	Continuous	*
Read Latency	Continuous	*
Write Latency	Continuous	*
Write Throughput	Continuous	*
Data Vol. Scalability	Non-Functional	Field/Class/DB
Write Scalability	Non-Functional	Field/Class/DB
Read Scalability	Non-Functional	Field/Class/DB
Elasticity	Non-Functional	Field/Class/DB
Durability	Non-Functional	Field/Class/DB
Replicated	Non-Functional	Field/Class/DB
Linearizability	Non-Functional	Field/Class
Read-your-Writes	Non-Functional	Field/Class
Causal Consistency	Non-Functional	Field/Class
Writes follow reads	Non-Functional	Field/Class
Monotonic Read	Non-Functional	Field/Class
Monotonic Write	Non-Functional	Field/Class
Scans	Functional	Field
Sorting	Functional	Field
Range Queries	Functional	Field
Point Lookups	Functional	Field
ACID Transactions	Functional	Class/DB
Conditional Updates	Functional	Field
Joins	Functional	Class/DB
Analytics Integration	Functional	Field/Class/DB
Fulltext Search	Functional	Field
Atomic Updates	Functional	Field/Class

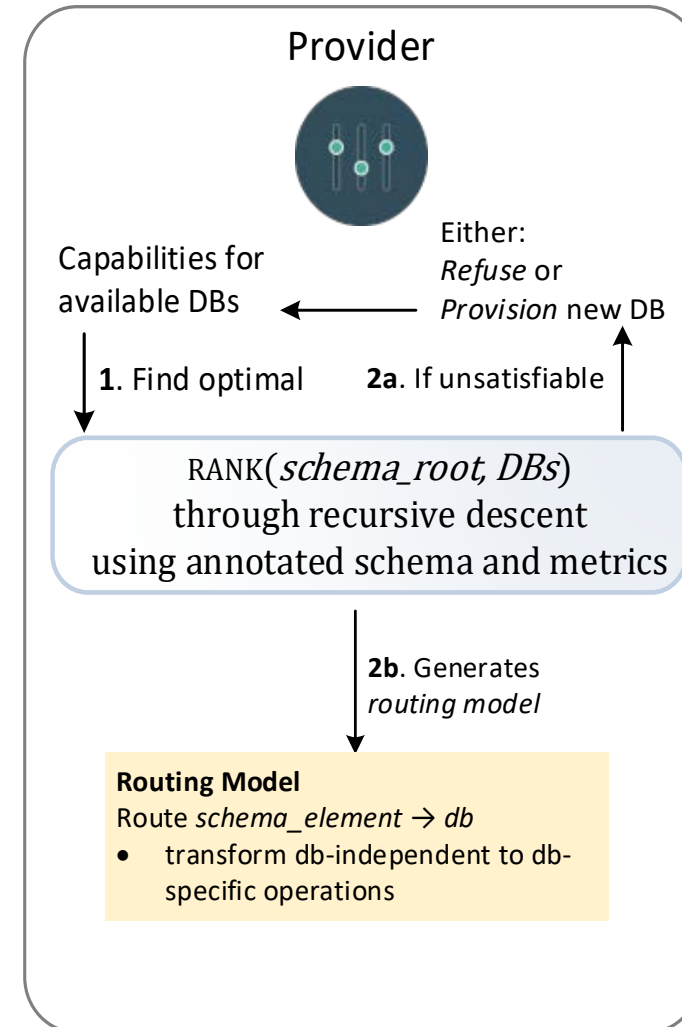


1 Requirements

Step II - Resolution

Finding the best database

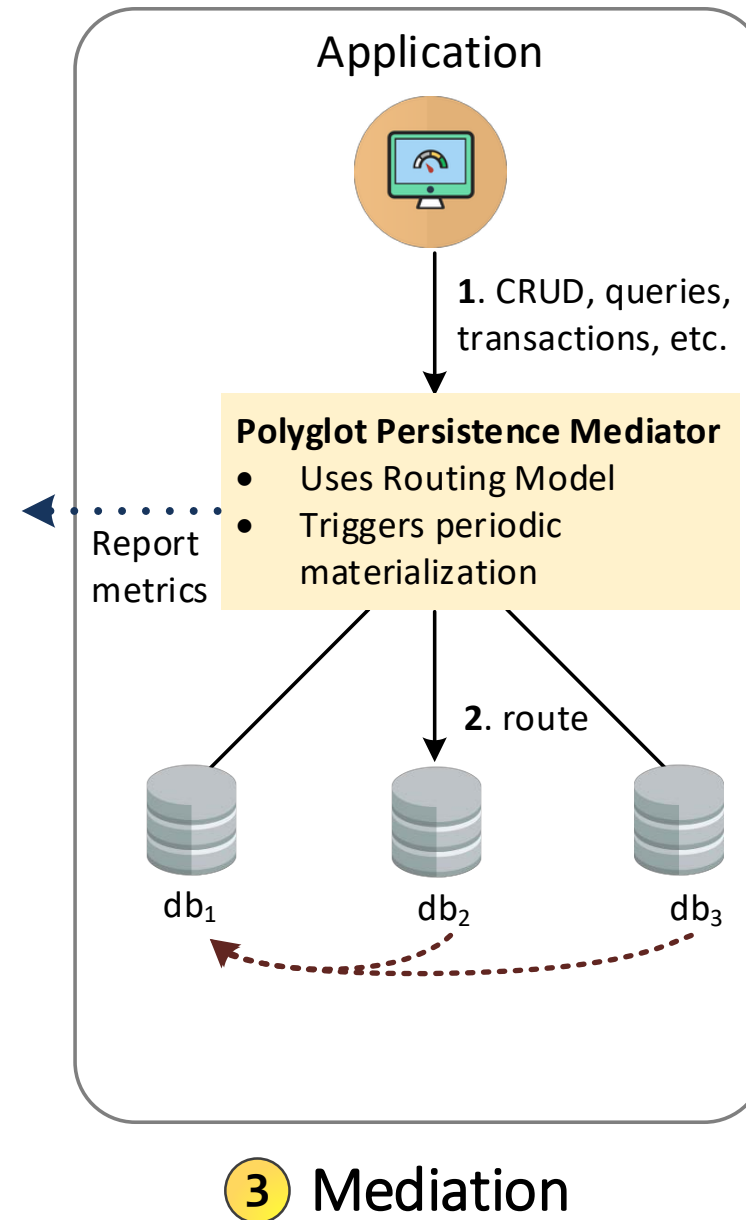
- The Provider resolves the requirements
- **RANK**: scores available database systems
- **Routing Model**: defines the optimal mapping from schema elements to databases



Step III - Mediation

Routing data and operations

- The PPM routes data
- **Operation Rewriting:** translates from abstract to database-specific operations
- **Runtime Metrics:** Latency, availability, etc. are reported to the resolver
- **Primary Database Option:** All data periodically gets materialized to designated database



Evaluation: News Article

Prototype of Polyglot Persistence Mediator in ORESTES

Scenario: news articles with impression counts

Objectives: low-latency top-k queries, high-throughput counts, article-queries

Article



The image shows a screenshot of a Hacker News article snippet. It features an orange header bar with the 'Y' logo, 'Hacker News' text, and navigation links: 'new | threads | comments | show | ask | jobs | submit'. Below this is a list item: '1. * NoSQL Databases: A Survey and Decision Guidance (medium.com)' followed by '297 points by DivineTraube 9 days ago | past | web | 73 comments | in pocket speichern'. At the bottom right of the snippet, it says 'read by 53,222'. Two arrows point to the snippet: one from the word 'Article' on the left pointing to the article title, and another from the word 'Counter' on the right pointing to the 'read by 53,222' text.

Y **Hacker News** new | threads | comments | show | ask | jobs | submit
submissions

1. * NoSQL Databases: A Survey and Decision Guidance (medium.com)
297 points by DivineTraube 9 days ago | past | web | 73 comments | in pocket speichern

read by 53,222

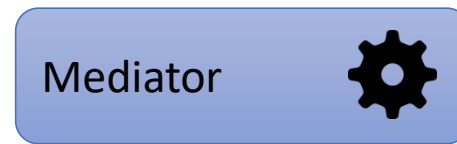
Counter

Evaluation: News Article

Prototype built on ORESTES

Scenario: news articles with impression counts

Objectives: low-latency top-k queries, high-throughput counts, article-queries



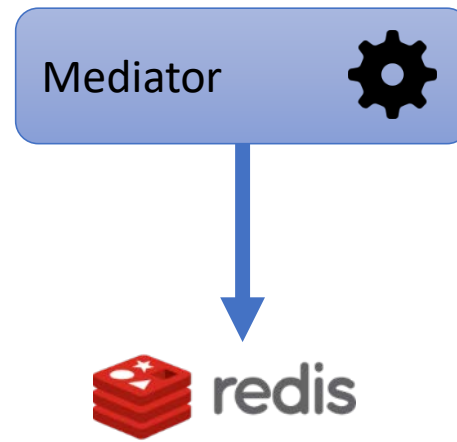
Counter updates kill performance

Evaluation: News Article

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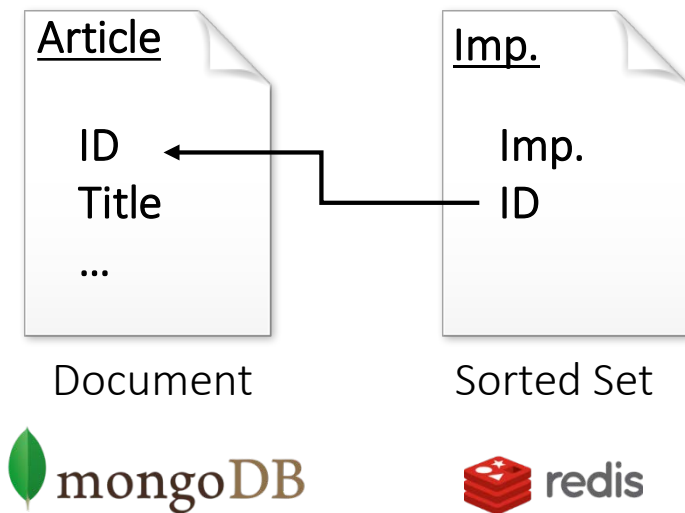
No powerful queries

Evaluation: News Article

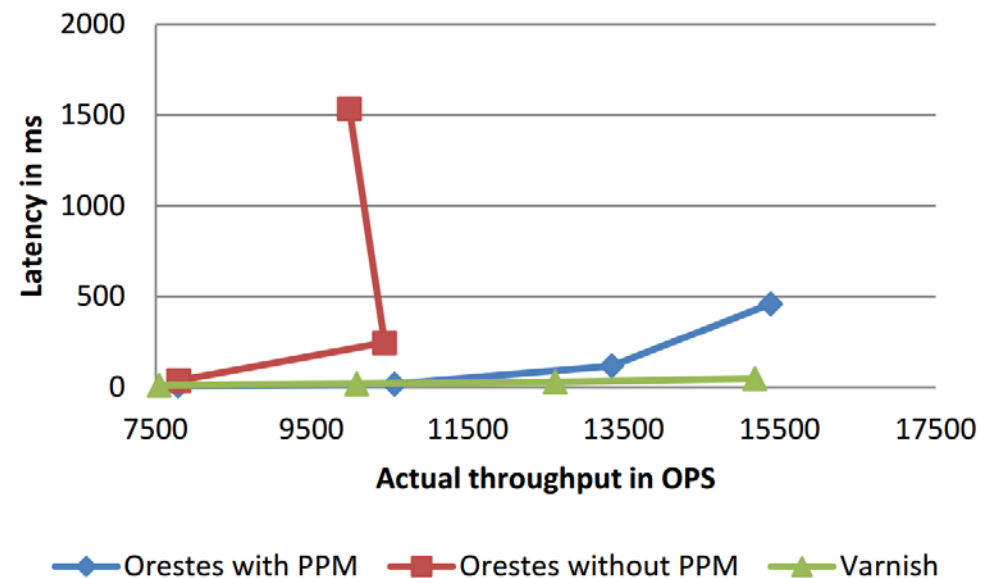
Prototype built on ORESTES

Scenario: news articles with impression counts

Objectives: low-latency top-k queries, high-throughput counts, article-queries



Found Resolution



Polyglot Persistence: Challenges



Database Selection: Actively minimize SLA violations



Utility Functions/SLAs: Capture trade-offs comprehensively



Meta-DBaaS: Mediate over DBaaS-Systems, unify SLAs



Live Migration: adapt to changing requirements




Workload Management: Adaptive Runtime Scheduling



Transaction Management: Alignment of ACID with NoSQL and scalability



Multi-Tenancy/Privacy: Dream: full homomorphic encryption

A glowing incandescent light bulb hangs from a dark background. The bulb is illuminated, casting a warm glow. The background is dark with several out-of-focus blue and white light sources, creating a bokeh effect. The light bulb is suspended by a dark cord.

CLOSING TIME

Summary

SCDM

1. Aim at fully managed Backend (BaaS)

2. Exploit modern (NoSQL) DB Technology

3. Consider the entire path from the (mobile)
Application through the Net to the Data Backend!

4. Make the backend push-based, additionally
(real-time queries)!

5. Provide Polyglot Persistence!

6. Other Problems? ... certainly!