Real-world consistency explained

or the challenges of modern persistence

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@ufried

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Some kudos first …

A lot of this talk was inspired by the great posts of Adrian Colyer

especially by his blog series "Out of the fire swamp”

see [Col], [Col2015a-c]


Past

RDBMS

ACID

RDBMS

- "One database to rule them all"
- Good all-rounder
  - Rich schema
  - Rich access patterns
- Designed for scarce resources
  - Storage, CPU, Backup are expensive
  - Network is slow
- Shared database
  - Replication was expensive
  - Licenses were expensive
  - Operations were expensive
  - Easy integration model
  - "Strange attractor"
  - Hard to change schemas
  - Data spaghetti

ACID

- Atomicity
- Consistency
- Isolation
- Durability

- Great programming model
  - No temporal inconsistencies
  - No anomalies
  - Easy to reason about

- But reality often is different!
  - ACID does not necessarily mean "serializability"
  - Databases often run at lower consistency levels
  - Anomalies happen
  - Most developers are not aware of it

ANSI SQL

Anomalies

- Dirty write (P0): w1[x]…w2[x]…(c1 or a1)
- Dirty read (P1): w1[x]…r2[x]…(c1 or a1)
- Fuzzy read (P2): r1[x]…w2[x]…(c1 or a1)
- Phantom read (P3): r1[P]…w2[y in P]…(c1 or a1)

Isolation levels

	Dirty write	*Dirty read*	*Fuzzy read*	*Phantom read*
Read uncommitted	Not possible	Possible	Possible	Possible
Read committed	Not possible	Not possible	Possible	Possible
Repeatable read	Not possible	Not possible	Not possible	Possible
Serializable	Not possible	Not possible	Not possible	Not possible

See [Ber+1995]

Extended anomaly model

- Dirty write (P0): w1[x]...w2[x]...(c1 or a1)
- Dirty read (P1): w1[x]...r2[x]...(c1 or a1)
- Lost update (P4): r1[x]...w2[x]...w1[x]...c1
- Lost cursor u. (P4C): rc1[x]...w2[x]...wc1[x]...c1.
- Fuzzy read (P2): r1[x]...w2[x]...(c1 or a1)
- Phantom read (P3): r1[P]...w2[y in P]...(c1 or a1)
- Read skew (A5A): r1[x]...w2[x]...w2[y]...c2...r1[y]...(c1 or a1)
- Write skew (A5B): r1[x]...r2[y]...w1[y]...w2[x]...(c1 and c2 occur)

see [Ber+1995]

Extended isolation level model

Isolation level	*Dirty write*	*Dirty read*	*Cursor lost update*	*Lost update*	*Fuzzy read*	*Phantom read*	*Read skew*	*Write skew*
Read uncommitted	Not possible	Possible	Possible	Possible	Possible	Possible	Possible	Possible
Read committed	Not possible	Not possible	Possible	Possible	Possible	Possible	Possible	Possible
Cursor stability	Not possible	Not possible	Not possible	Sometimes possible	Sometimes possible	Possible	Possible	Sometimes possible
Repeatable read	Not possible	Not possible	Not possible	Not possible	Not possible	Possible	Not possible	Not possible
Snapshot	Not possible	Not possible	Not possible	Not possible	Not possible	Sometimes possible	Not possible	Possible
Serializable	Not possible	Not possible	Not possible	Not possible	Not possible	Not possible	Not possible	Not possible

See [Ber+1995]

Default & maximum isolation levels

Database	Default	Maximum
Actian Ingres 10.0/10S [1]	S	S
Aerospike [2]	RC	RC
Akiban Persistit [3]	SI	SI
Clustrix CLX 4100 [4]	RR	RR
Greenplum 4.1 [8]	RC	S
IBM DB2 10 for z/OS [5]	CS	S
IBM Informix 11.50 [9]	Depends	S
MySQL 5.6 [12]	RR	S
MemSQL 1b [10]	RC	RC
MS SQL Server 2012 [11]	RC	S
NuoDB [13]	CR	CR
Oracle 11g [14]	RC	SI
Oracle Berkeley DB [7]	S	S
Oracle Berkeley DB JE [6]	RR	S
Postgres 9.2.2 [15]	RC	S
SAP HANA [16]	RC	SI
ScaleDB 1.02 [17]	RC	RC
VoltDB [18]	S	S
RC: read committed, RR: repeatable read, SI: snapshot isolation, S: serializability, CS: cursor stability, CR: consistent read		

Table 1: Default and maximum isolation levels for ACID and NewSQL databases as of January 2013.

See [Bai+2013a]

Wrap-up – Past

- The relational model is a good tradeoff
- ACID makes a developer's life easy
- Yet, we often live (unknowingly) with less than serializability


Present

Cloud

μService

NoSQL

BASE

Cloud

- Self Service
- Elasticity
- Pay per use

- Great resource provisioning model

- Improves
  - Autonomy
  - Response time (lead time)
  - Elasticity
  - Cost efficiency (if done right)

- Trade-offs
  - Scale out (“distributed hell”)
  - Reduced availability of individual resources

- "Microservices are the mapping of organizational autonomy to software architecture"
  - Limited in scope
  - Self-dependent
  - Loosely coupled
- Improves
  - Autonomy
  - Response time (if done right)
  - Elasticity
- Trade-offs
  - Higher design effort
  - Harder to operate
  - Distributed by default
- Shared nothing
  - No shared data
  - No cross-service coordination

µService

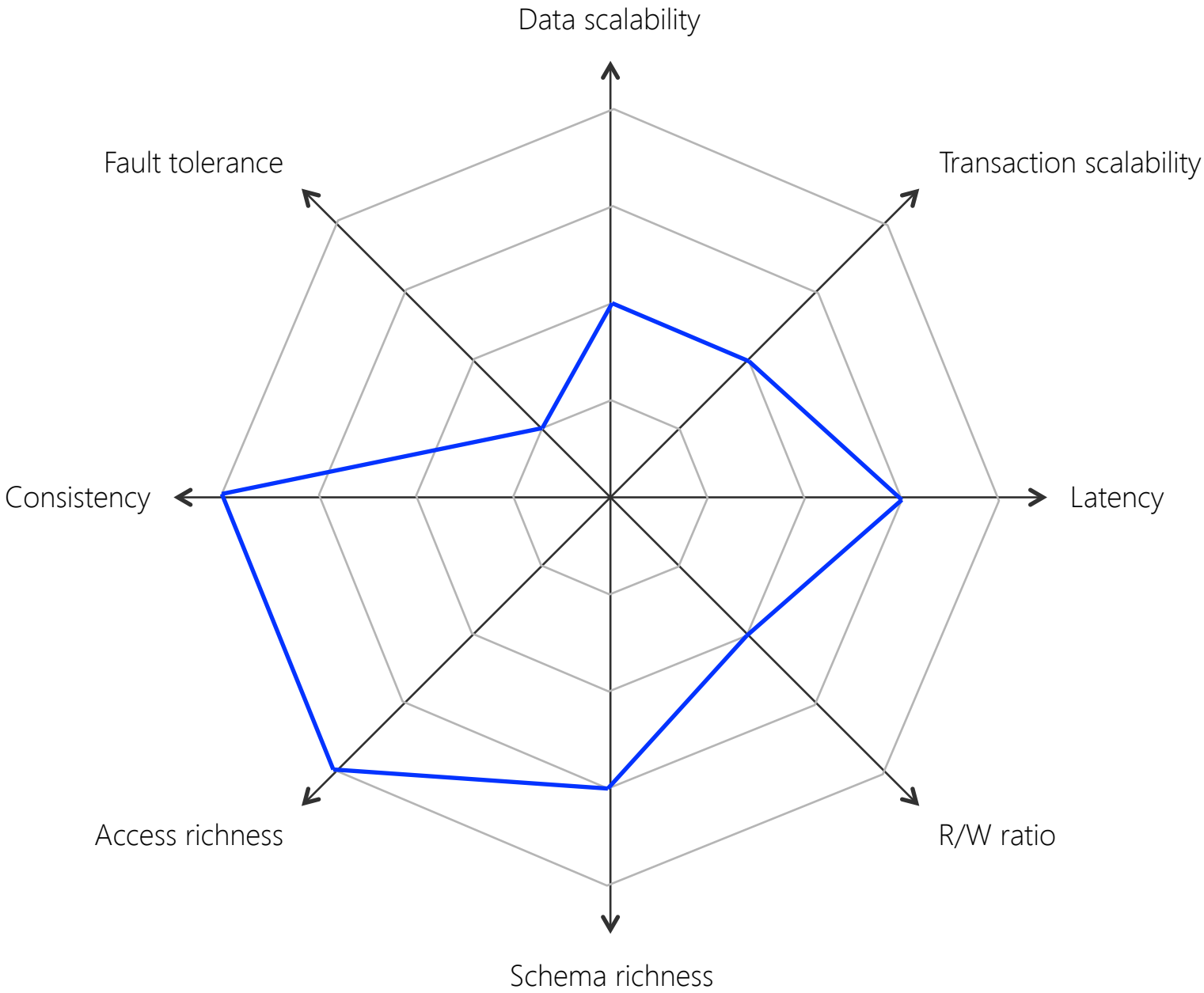
NoSQL

- Extension of the storage solution space
  - Before NoSQL RDBMS and file system were predominant solutions
  - NoSQL tries to fill the gaps
- New options
  - Scalability (Volume & Velocity)
  - Relaxed schema
  - Availability in cloud environments
- Trade-offs
  - CAP Theorem
  - Capabilities and limitations often poorly understood

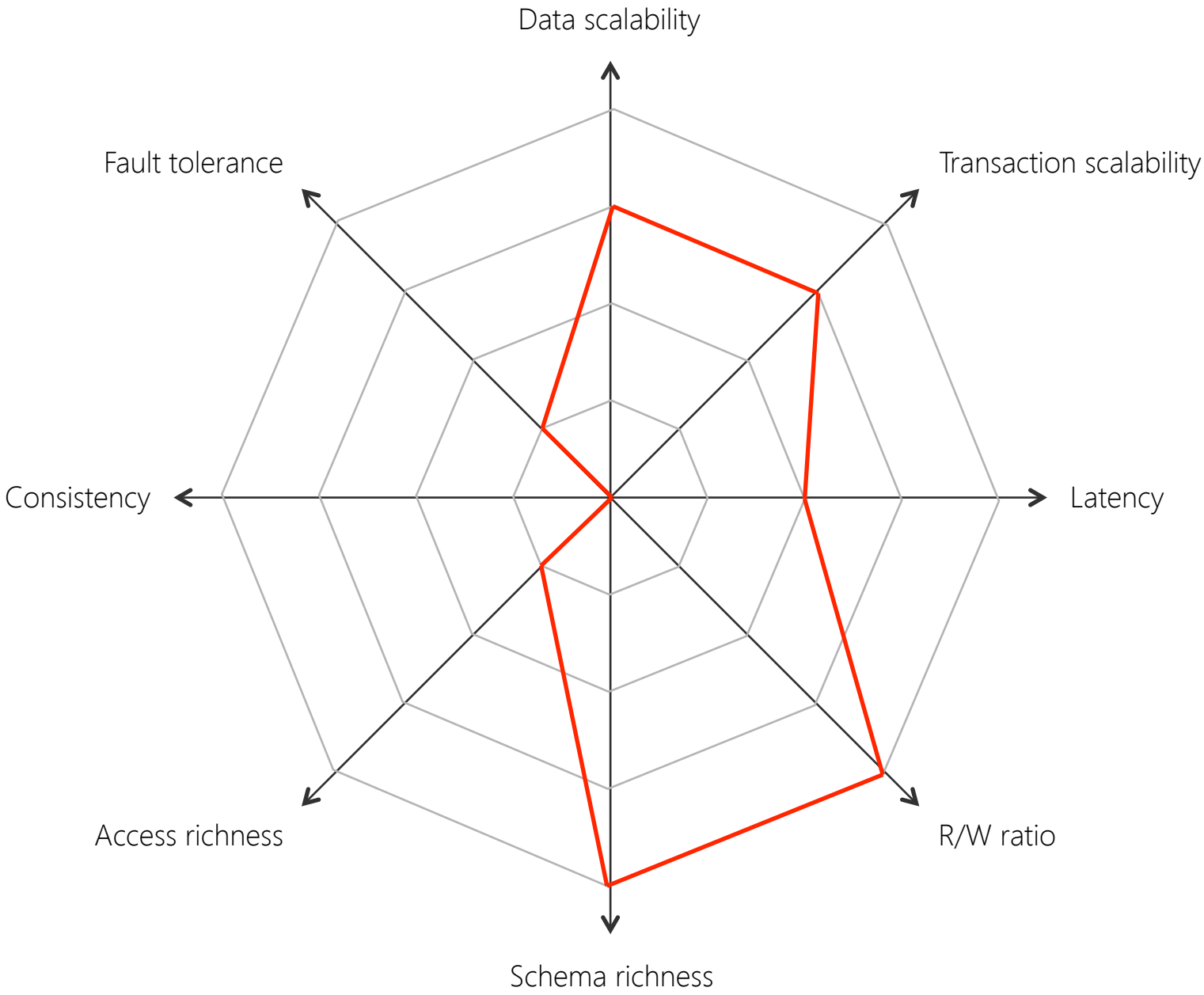
The 8 dimensions of storage

- Data Scalability (amount of data)
- Transaction Scalability (access rate)
- Latency (response time considering scalability)
- Read/Write Ratio (variability of r/w mix considering scalability)
- Schema Richness (variability of data model)
- Access Richness (variability of access patterns)
- Consistency (data consistency guarantees)
- Fault Tolerance (ability to handle failures gracefully)

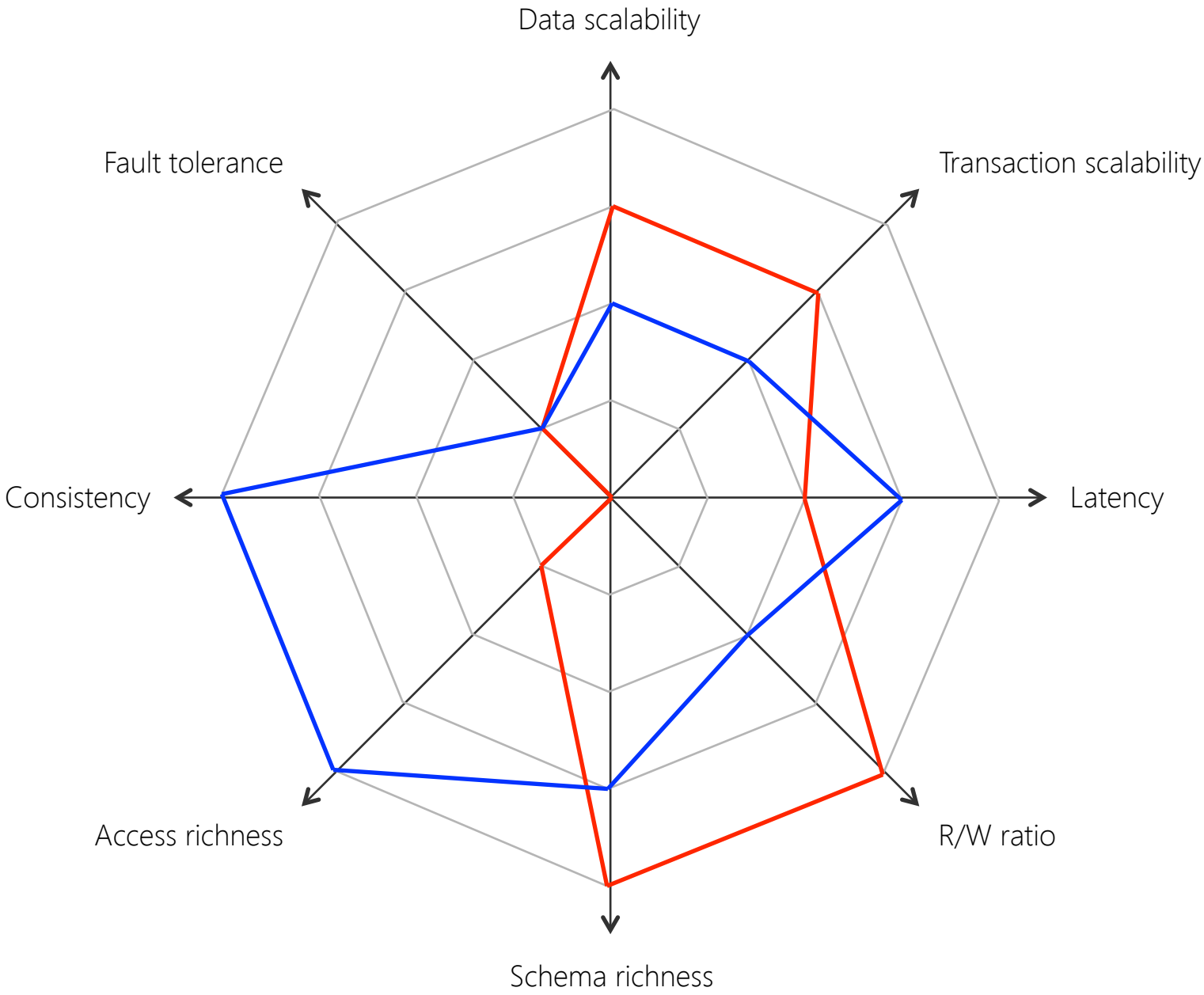
Relational database



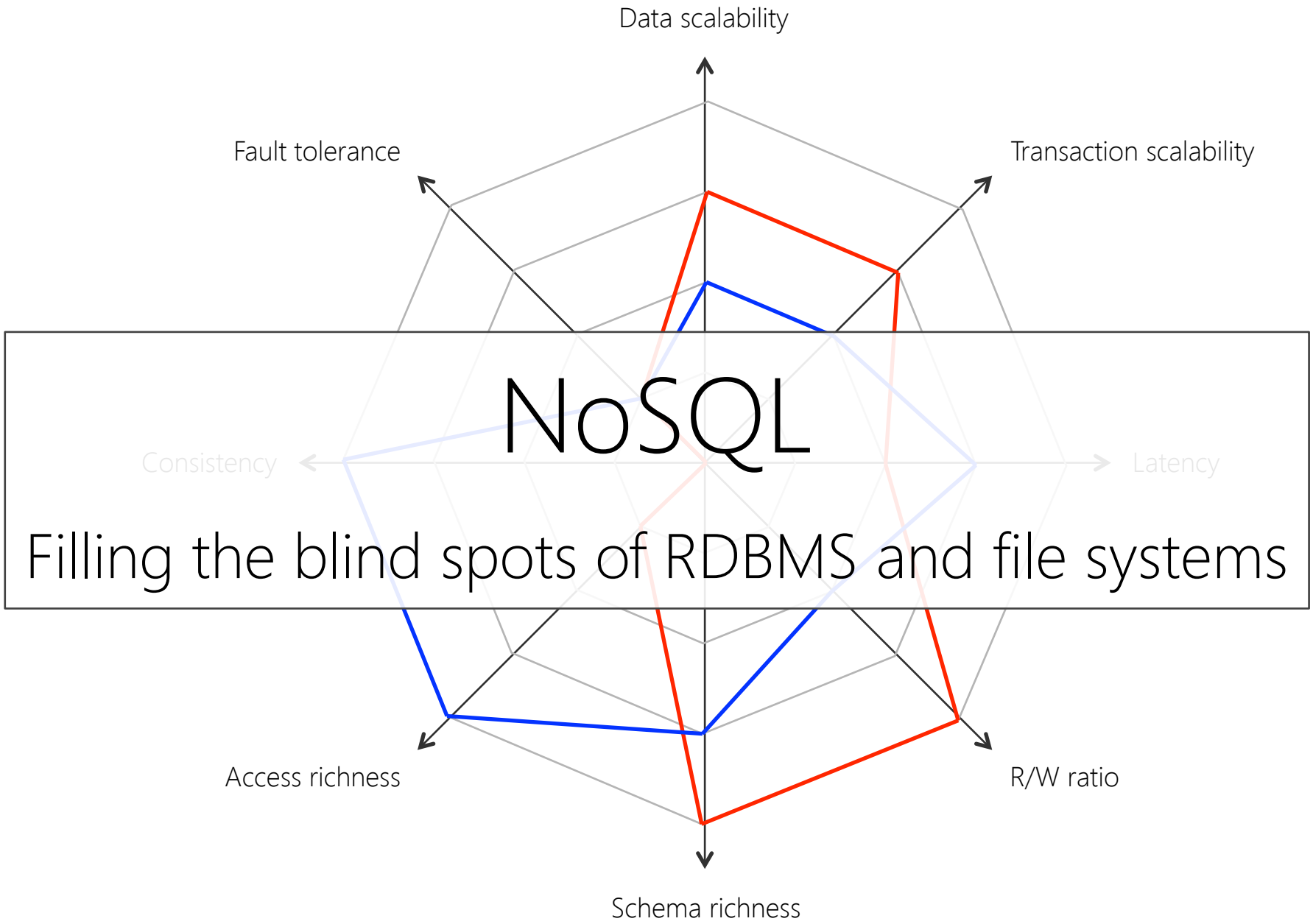
File system


Data scalability
Transaction scalability
Latency
R/W ratio
Schema richness
Access richness
Consistency
Fault tolerance

RDBMS & file system



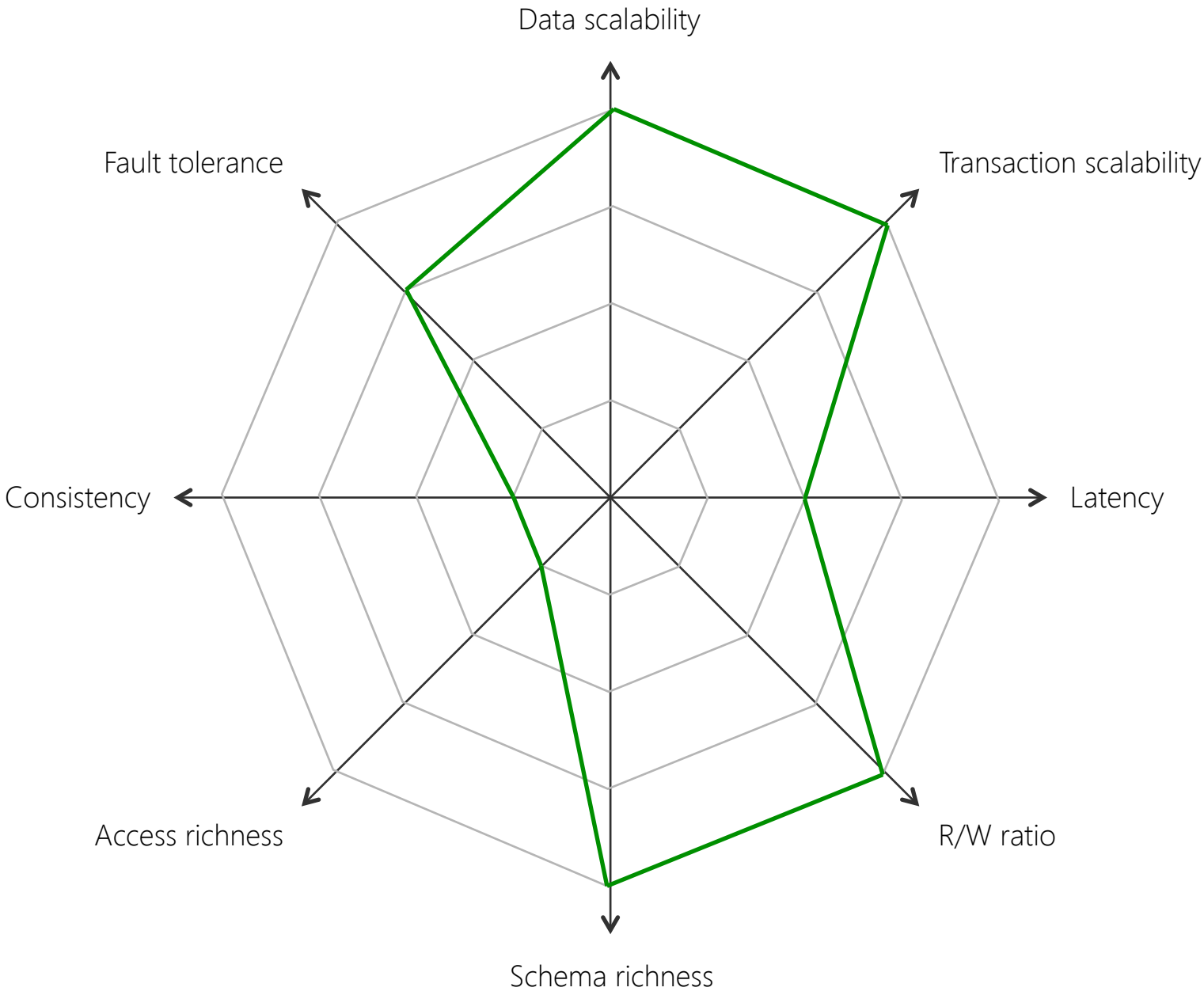
RDBMS & file system



NoSQL

Filling the blind spots of RDBMS and file systems

Cassandra


Data scalability
Transaction scalability
Latency
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Schema richness
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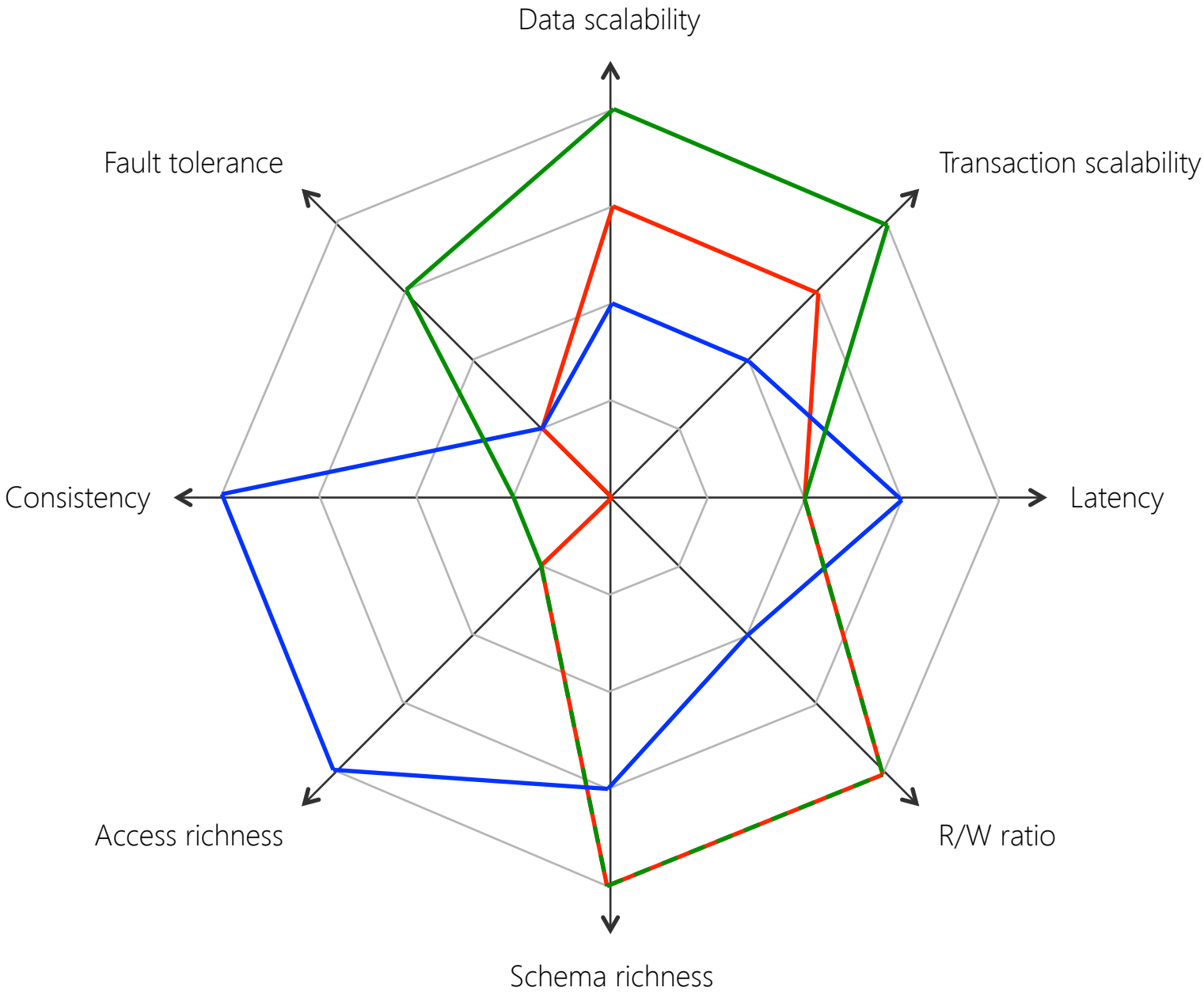
Consistency
Fault tolerance

RDBMS-FS-Cassandra

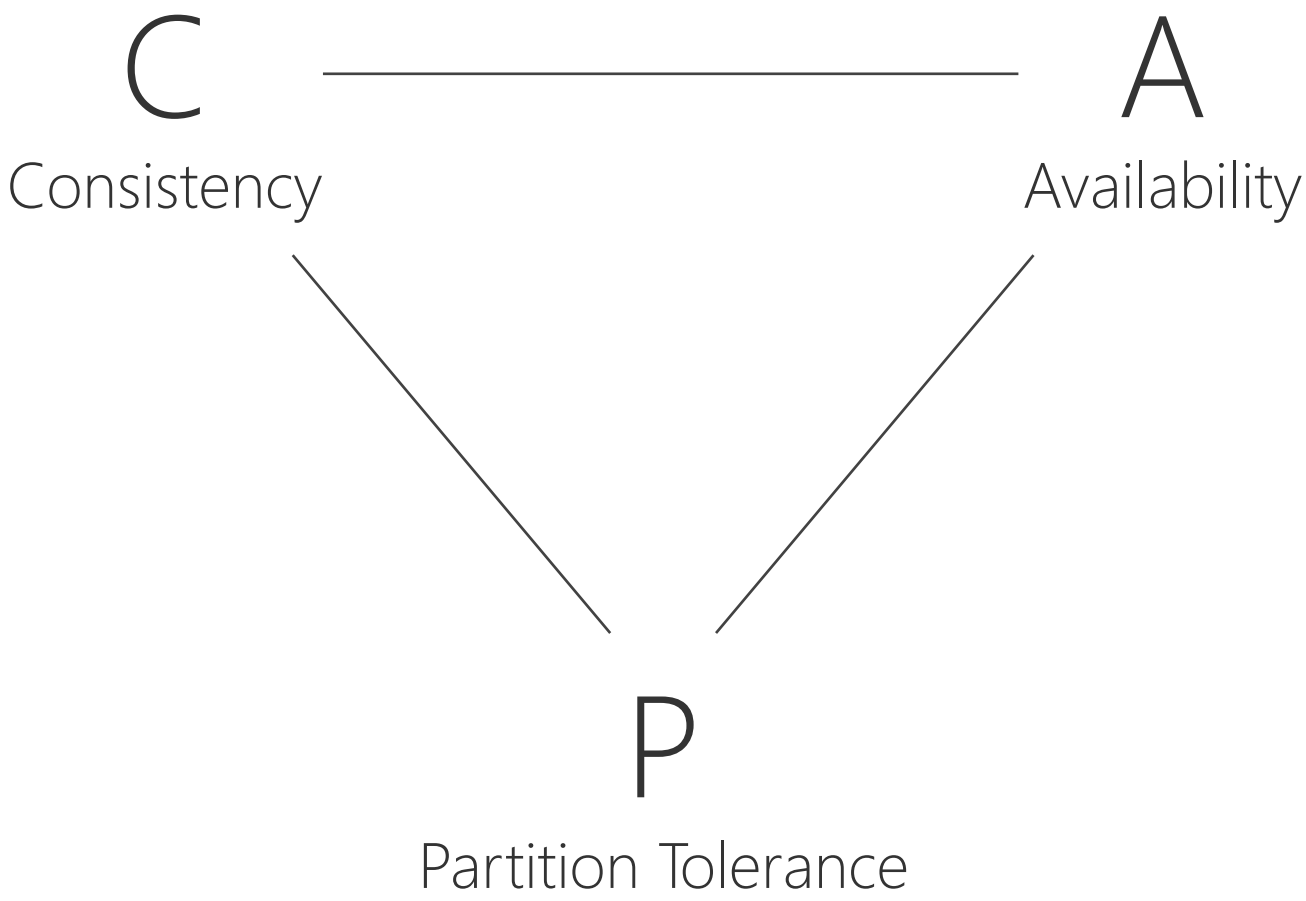


- Response to CAP theorem
  - "Relax temporal constraints in exchange for better availability"
- Improves
  - Scalability
  - Availability
- Trade-offs
  - Temporal inconsistencies and all kinds of anomalies become visible
  - Very hard programming model
- Key readings
  - [Bre2000] introduced CAP and BASE
  - [Hel2007] "defined" boundaries of eventual consistency for years
  - [Sha+2011] introduced CRDTs, improving convergence guarantees in eventually consistent systems

BASE

But those failures rarely happen, do they?

Do we really need to care about partition tolerance?


C
Consistency
A
Availability
P
Partition Tolerance

Failure types

- Crash failure
- Omission failure
- Timing failure
- Response failure
- Byzantine failure

"A 2011 study of several Microsoft datacenters observed over 13,300 network failures with end-user impact, with an estimated median 59,000 packets lost per failure. The study found a mean of 40.8 network link failures per day (95th percentile: 136), with a median time to repair of around five minutes (and up to one week)."

see [Bai+2014a]

	H2	H3
H1	0.55	0.56
H2		0.50

(a) Within us-east-b AZ

	C	D
B	1.08	3.12
C		3.57

(b) Across us-east AZs

	OR	VA	TO	IR	SY	SP	SI
CA	22.5	84.5	143.7	169.8	179.1	185.9	186.9
OR		82.9	135.1	170.6	200.6	207.8	234.4
VA			202.4	107.9	265.6	163.4	253.5
TO				278.3	144.2	301.4	90.6
IR					346.2	239.8	234.1
SY						333.6	243.1
SP							362.8

(c) Cross-region (CA: California, OR: Oregon, VA: Virginia, TO: Tokyo, IR: Ireland, SY: Sydney, SP: São Paulo, SI: Singapore)

Table 1: Mean RTT times on EC2 (min and max highlighted)

see [Bai+2014a]

… and that's only the network.

We haven't talked yet about failing nodes, faulty processes, etc., which also materialize as partitioning in distributed system

Partitions in distributed systems happen
way more often than most people expect it

Ignoring partition tolerance is not an option in distributed systems

You need to balance availability and consistency
if you want to scale out your data store

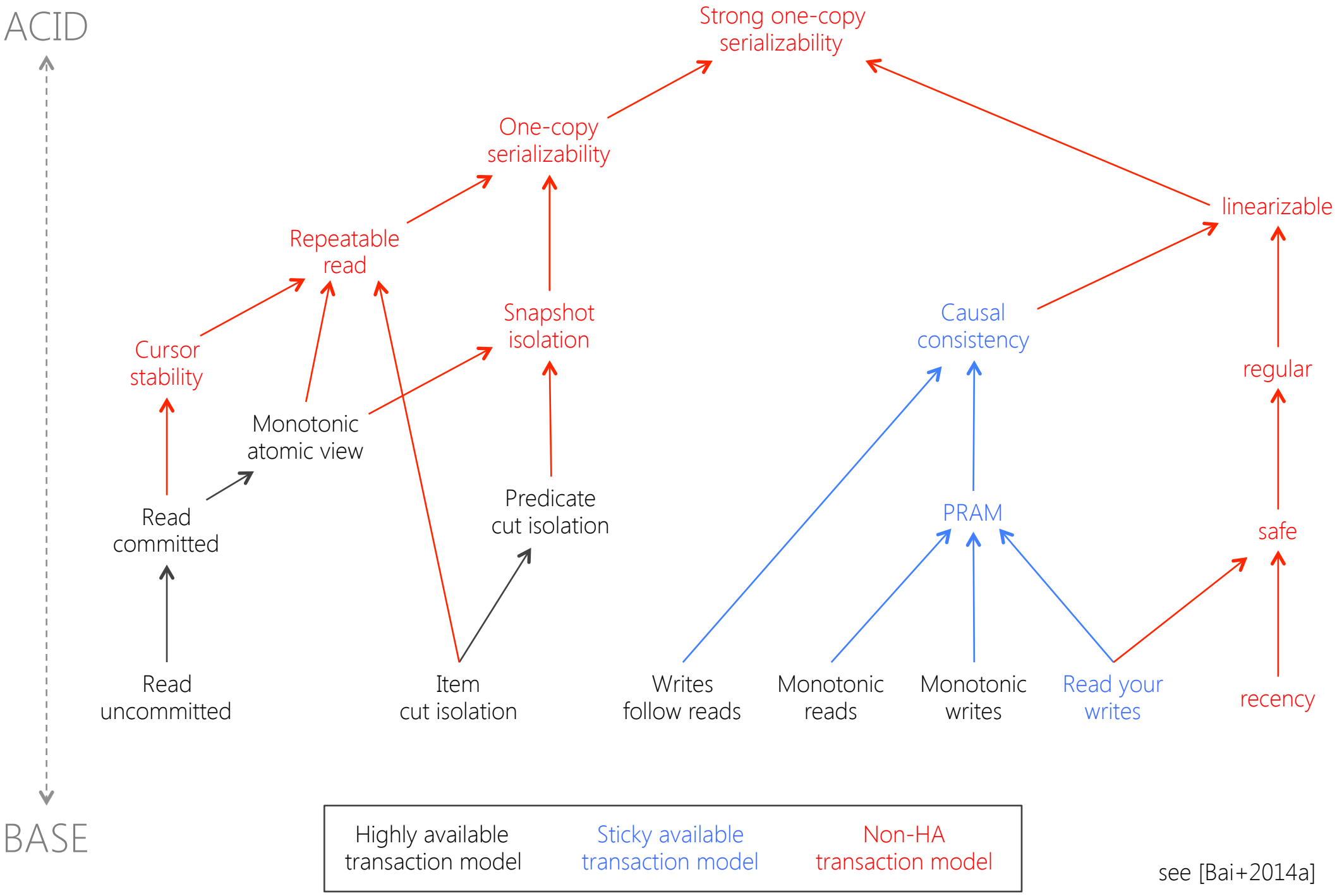
Wrap-up – Present

- IT goes distributed (“scale out”)
- NoSQL fills the empty spots in the storage solution space
- BASE transactions imply a very hard programming model


Future

So, can I only choose between “ACID” and “BASE”?

Remember, that “ACID” does not necessarily mean serializability?



see [Bai+2014a]

There are a lot of consistency options to choose from between “ACID” and “BASE”

But ...

The old model assumed the work would be processed in exactly one order of execution. There was a default “single system of record” form of isolation provided by the classic database system running at the primary. This single history allows for a low-level READ and WRITE semantic that depends on “replaying history”.

In this new [distributed] world, history cannot be exactly replayed and we must count on the ability to reorder the work. This means that we cannot completely know the accurate state of the system. It also means we must move the correctness and reordering semantics up from being based on system properties (i.e. READ and WRITE) to application based business operations.

see [Hel+2009]

It is no longer sufficient to understand and reason about distributed application consistency at the level of data store access.

Instead you need to understand and reason about distributed application consistency at the level of application operations.

1. You need to analyze the consistency requirements of the application carefully based on the business requirements.
2. The application (usually) needs to contribute explicitly to the implementation of the required consistency model.

Current research explores the “frontiers”

- "HAT, not CAP: Towards Highly Available Transactions” [Bai+2013b]
  HA causal consistency in distributed systems
- "Scalable Atomic Visibility with RAMP Transactions" [Bai+2014b]
  New isolation level “atomic read” (between MAV and SI), implemented in HA fashion
- "Building Consistent Transactions with Inconsistent Replication" [Zha+2015]
  Low-latency distributed transactions by building a transactional application protocol on top of inconsistent replication
- "Putting Consistency Back into Eventual Consistency" [Bal+2015]
  Explicit application-level consistency using application-defined invariants on top of eventual consistent data stores
- "Implementing Linearizability at Large Scale and Low Latency" [Lee+2015]
  Implementing fast and scalable exactly-once semantics, enabling linearizable operations on top of it

- "Spanner: Google’s Globally-Distributed Database" [Cor+2012]
  “Case study” for a very different approach to scalable serializability by using hardware to solve the “time problem”
- "High-Performance ACID via Modular Concurrency Control" [Xie+2015]
  Speedup of traditional ACID transactions by grouping transactions into independent sets that can be handled concurrently

And there is more to come …

Current memory and storage trends

- Terabyte memory computers

  Full in-memory computing of large data sets

- Storage Class Memory (SCM) / Non-Volatile RAM (NVRAM)

  Many technologies under development filling the gap between RAM and SSD

- Remote Direct Memory Access (RDMA)

  Accessing a remote machine's RAM bypassing the CPU, allowing very low-latency remote memory access (around 2 order of magnitude faster than SSD access)

Keeping the CPU busy is no longer the core challenge

New system architectures and programming models will emerge

New consistency options not available today may also emerge

see [Coc2016], [Col2016]

Wrap-up – Future

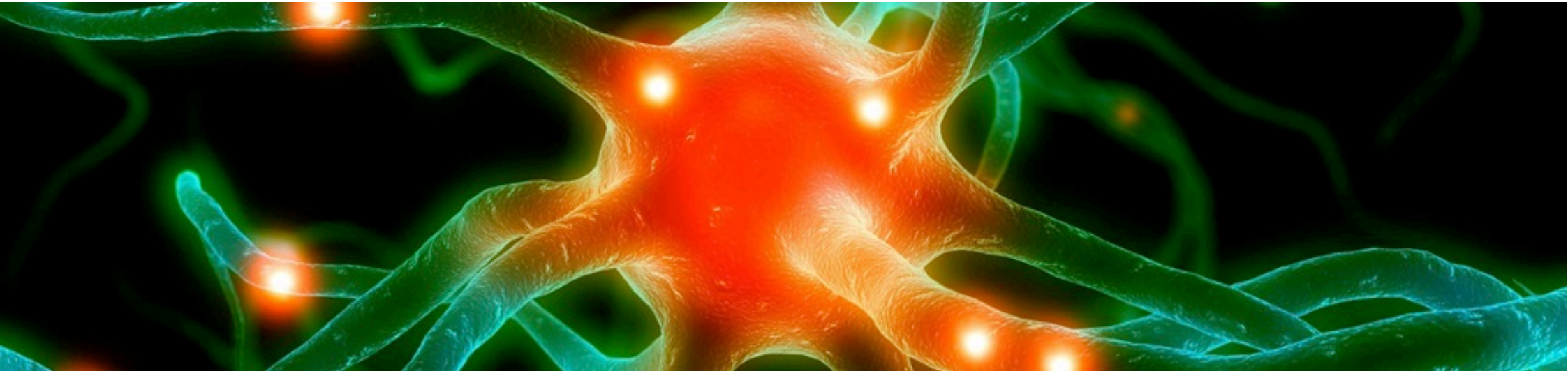
- Lots of options to balance consistency constraints and intricacy of the programming model
- Higher consistency guarantees than “plain BASE” in distributed HA databases might come soon
- Most of them will require effort on the application level

Recommendations

Across service/storage boundaries

- Don't coordinate writes across service/storage boundaries
  - Usually your design is wrong (entity-driven instead of behavior-driven)
  - Remember: DDD is about *domains*, not entities! Domains include behavior
  - The activation path of a use case should be as short as possible

- Use a relaxed temporal constraint model plus reconciliation
  - Consider applying the concepts of promise theory [Bur2005] and memory, guesses and apologies [Hel+2009]

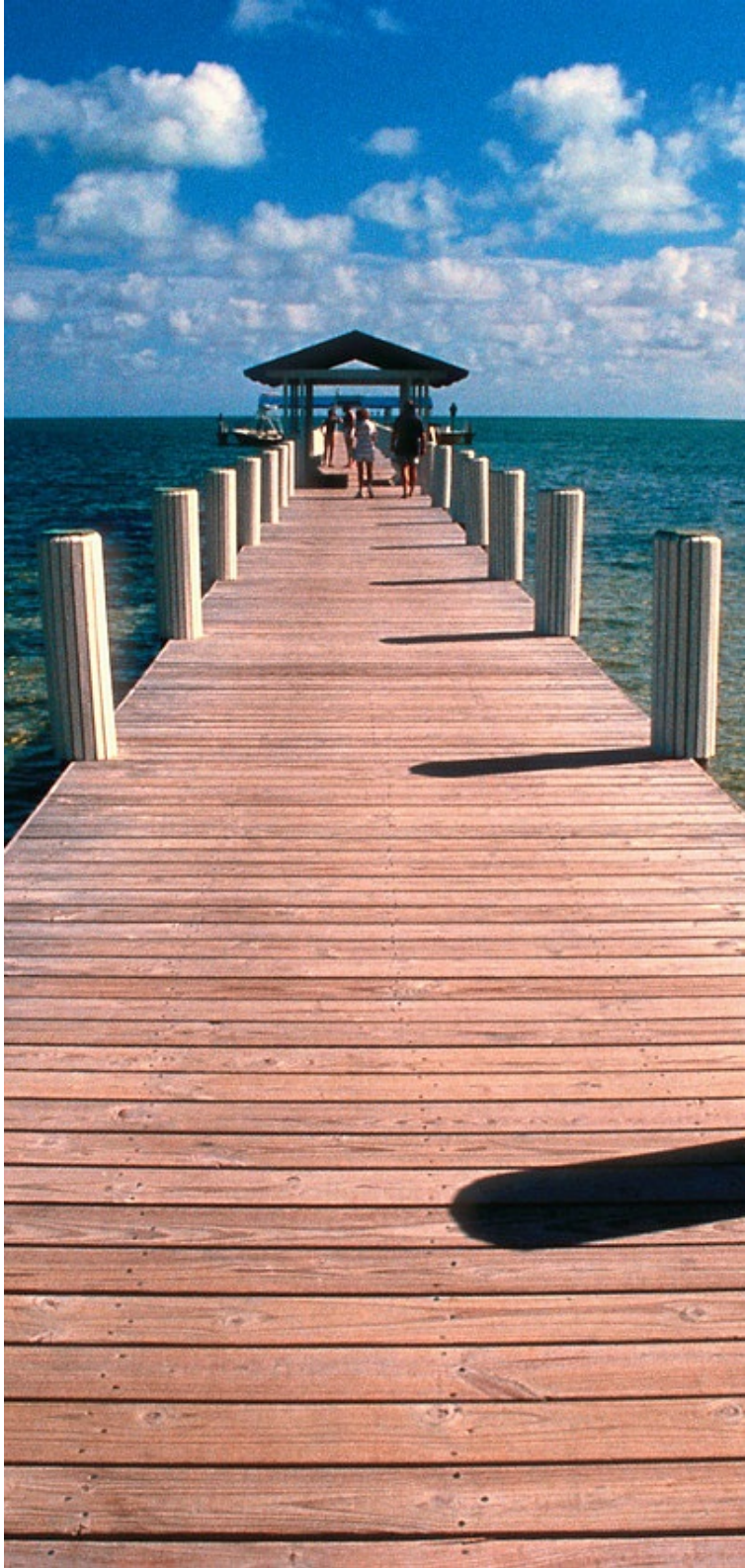


Within service/storage boundaries

- Thoroughly analyze your storage requirements (8 dimensions)
- Thoroughly analyze your consistency requirements
- Be wary of serializability (database reality is different)
- Don't think consistency is solely a data store issue
- Don't distribute data or relax consistency without an explicit need
- Choose wisely – and have your developers in mind

Wrap-up

- Past: RDBMS and ACID
  - Great programming model
  - ACID does not necessarily mean serializability
- Presence: Cloud, µServices, NoSQL & BASE
  - More options, more challenges
  - Very hard programming model
- Future: Exploring the boundaries
  - Many options between ACID and BASE
  - Often requires awareness on the application level
- Know your options!



There is no “one-size-fits-all” solution

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