Databases for the New World

Parallel Functional Programming

John Hughes

New Demands on Databases

- Very high throughput
 - Requires large clusters to process all the operations
- Low and predictable latency
 - Good customer experience for (almost) all customers
 - Not average latency, but 99,9th percentile
- Always available
 - Think Amazon, Twitter, Facebook
 - Failed operations==lost business

"The Network is Reliable" (aphyr.com)

 "During a planned network reconfiguration to improve reliability, Fog Creek suddenly lost access to their network. A network loop had formed...it resulted in two hours of total service unavailability."

 "Mystery RabbitMQ partitions...upping the partition detection timeout to 2 minutes reduced the frequency of partitions, but didn't prevent them altogether."

"The Network is Reliable" (aphyr.com)

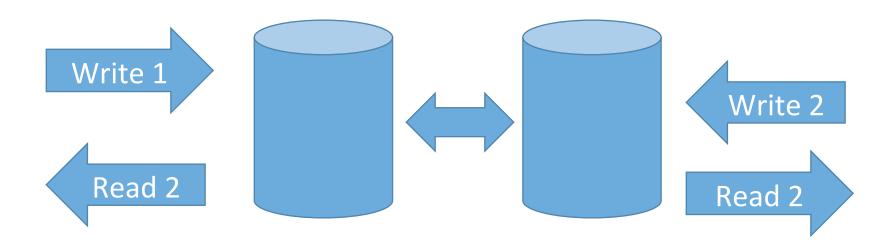
 "DRBD split-brain...both nodes can remain online and accept writes...the only realistic option is to discard all writes not made to a selected component."

 "Github...a 90 second network partition caused file servers to send "Shoot the other node in the head" messages to each other...when the network recovered, both nodes shot each other at the same time...recovering took five hours."

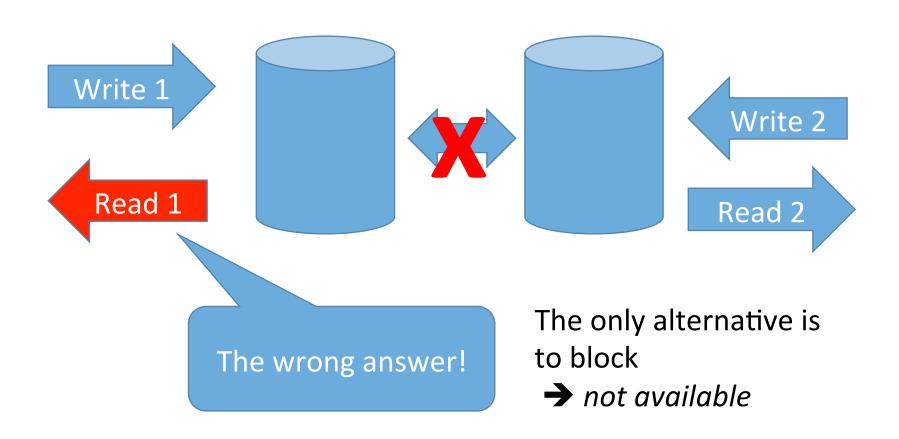
"The Network is Reliable" (aphyr.com)

 "MongoDB...partition caused two hours of write loss...network events causing failover on EC2 are common...simultaneous primaries accepting writes for multiple days are not unknown."

Network Partitions in a Database Cluster



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Consistency **Pick** two! A vailability P artition-tolerance theorem

Conjecture: Eric Brewer, 2000 Proof: Gilbert and Lynch, 2002

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Luckily...

- For many applications, consistency is not essential
 - E.g. Facebook posts
- "Eventual" consistency is good enough
 - Eventually we get the right answer
 - Mechanisms to discover and tolerate inconsistencies
- Often, simple queries are all that is needed
 - Primary key only, no relational joins

Amazon Dynamo (2007)

Dynamo: Amazon's Highly Available Key-value Store

Giuseppe DeCandia, Deniz Hastorun, Madan Jampani, Gunavardhan Kakulapati, Avinash Lakshman, Alex Pilchin, Swaminathan Sivasubramanian, Peter Vosshall and Werner Vogels

Amazon.com

ABSTRACT

Reliability at massive scale is face at Amazon.com, one of the the world; even the slightest consequences and impacts of platform, which provides service is implemented on top of an infi

servers and network components located in many datacenters around the world. At this scale, small and large components fail continuously and the way persistent state is managed in the face

of these failures drives the software systems.

This paper presents the design highly available key-value stora core services use to provide achieve this level of availabilit under certain failure scenarios, versioning and application-assis

that provides a novel interface for developers to use.

Categories and Subject Descriptors

D.4.2 [Operating Systems]: Storage Management; D.4.5 [Operating Systems]: Reliability; D.4.2 [Operating Systems]: Performance;

More than 2,000 citations

pple, customers should be able shopping cart even if disks are failing, network routes are flapping, or data centers are being destroyed by tornados. Therefore, the service responsible for

> that it can always write to and its data needs to be available

n has learned from operating

liability and scalability of a

application state is managed.

zed, loosely coupled, service

hundreds of services. In this

need for storage technologies

Many, many systems follow this design

performance.

To meet the reliability and scaling needs, Amazon has developed a number of storage technologies, of which the Amazon Simple Storage Service (also available outside of Amazon and known as Amazon S3), is probably the best known. This paper presents the

cture comprised of millions of f operation; there are always a rver and network components As such Amazon's software a manner that treats failure thout impacting availability or

What has this to do with Erlang?

Erlang excels at scalable services (e.g. WhatsApp),
 which often need a scalable database

- Erlang is good at *implementing* a scalable distributed database
 - CouchDB, Couchbase, Riak, Scalaris, Dynomite...
- Basho's Riak is one of the big noSQL databases
 - (Rovio, Danish medical card, UK National Health Service)

API

• Dynamo is a Distributed Key-Value Store

```
get :: Key -> ** Kitalk@th** [conttext]
```

put :: (Klæk/Ologetute) >> ok

RiakObject

Riak splits a key into a *key* and a *bucket* (like a table name)

Key Context Value

Cluster

- Dynamo is designed for clusters of up to a few hundred machines
 - Largest Riak clusters are 60-80 machines

- Each machine handles a share of the load
 - Stores a part of the data (in a local back-end, such as Google LevelDB)
- Data is replicated N times for durability/availability
 - At Amazon, replicas are in *different data-centres*

Consistent hashing

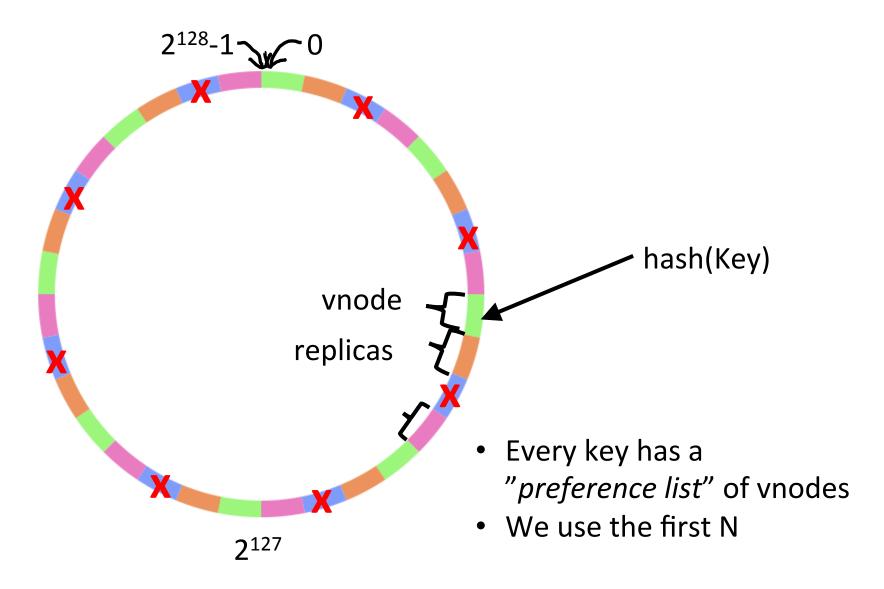
2128-1-0

My datas

vnode replicas hash(Key)

- Hash determines vnode
- Vnode determines physical node
- Any node can determine where a key is stored

What if a node is unavailable?



- Put: send writes to N nodes
 - And wait for N acknowledgements before acknowledging the user

Read: send read requests to N nodes

And wait for N replies before replying to the user



What if a node doesn't reply?

Mark it unavailable, use the next from the preference list.

- Put: send writes to N nodes
 - And wait for N acknowledgements before acknowledging the user

Read: send read requests to N nodes

And wait for N replies before replying to the user



Why not reply to a read when we get the first reply?
It might be a stale value!

- Put: send writes to N nodes
 - And wait for N acknowledgements before acknowledging the user

Read: send read requests to N nodes

And wait for N replies before replying to the user



Why not reply to a read when we get the Thefitte Candina latency vs consistency!

- Put: send writes to N nodes
 - And wait for N acknowledgements before acknowledging the user

Read: send read requests to N nodes

And wait for R replies before replying to the user



Why not acknowledge

a write after one
Because that node might
crash and the other writes

might fail! We'd lose data.

- Put: send writes to N nodes
 - And wait for N acknowledgements before acknowledging the user

Read: send read requests to N nodes

And wait for R replies before replying to the user



Why not acknowledge a write after W

Then Weblish?tune latency against durability!

- Put: send writes to N nodes
 - And wait for W acknowledged acknowledging the user

In practice, a "sloppy quorum", because we use the first N available nodes

- Read: send read requests to N noues
 - And wait for R replies before replying to user



guarantees a *quorum*—each read sees the latest write

Handoff

- What happens if data is written to the wrong node, and then the correct node comes back up?
 - We know where the data should be stored—we record that on the replacement node.
 - When the correct node recovers, the replacement node "hands off" the data to the correct one.
 - But before handoff is complete, stale data may be read.

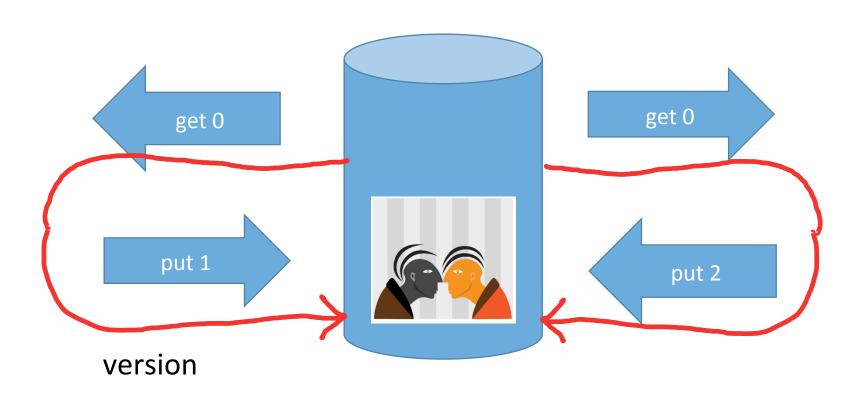
Read repair

- What happens if we get a key, and one of the replicas returns stale data?
 - We can *update* that replica with fresh data from another replica... "read repair".
 - This is why the node co-ordinating a get request waits for all N replies, even if only R are needed to respond to the client.

Recognising stale data

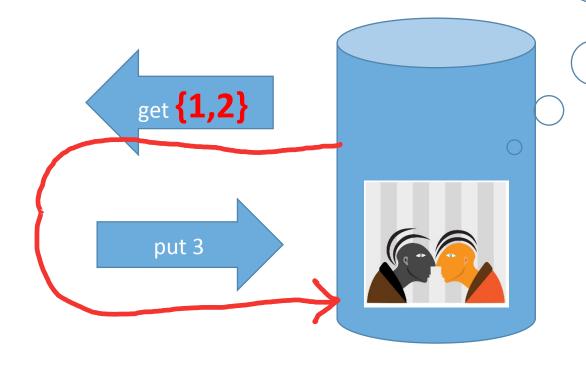
- How do we know when data is out-of-date?
 - Time stamps are not reliable!
- Remember that "context" information...?
 - It's version information... a vector clock.

Versioning



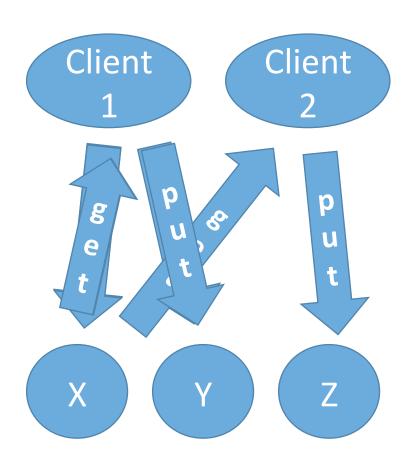
Versioning

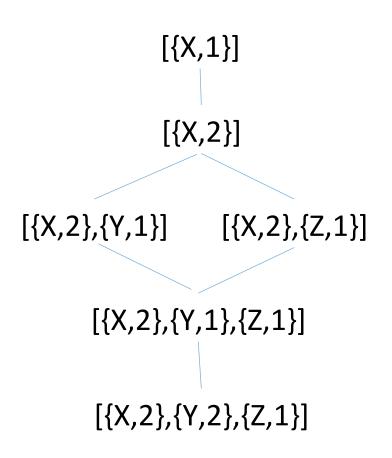
Application specific conflict resolution



ExampleAmazon shopping basket

Vector Clocks





Vector Clocks

Consist of a list of node ids and counts

- c1 descends from c2 if: √ n ∈ Nodes. count(n,c1) >= count(n,c2)
- A value v1 supercedes v2 if its vector clock descends from the clock of v2.

How often do conflicts arise?

Number of versions	%ge of read requests
1	99,94
2	0,00057
3	0,00047
4	0,00009

Conflict resolution is only needed occasionally Source: Amazon

Deletion

How do we delete a key?

• We don't! Write a "tombstone" over it.



• This means dead keys can come back to life as a result of conflicts...

How do nodes know the ring

Joining and leaving the ring is done explicitly

- Nodes "gossip" the ring to each other
 - Periodically send the ring to random other nodes
 - All nodes quickly become aware of changes
- Basho is working on optimizations to this

How do nodes join or leave?

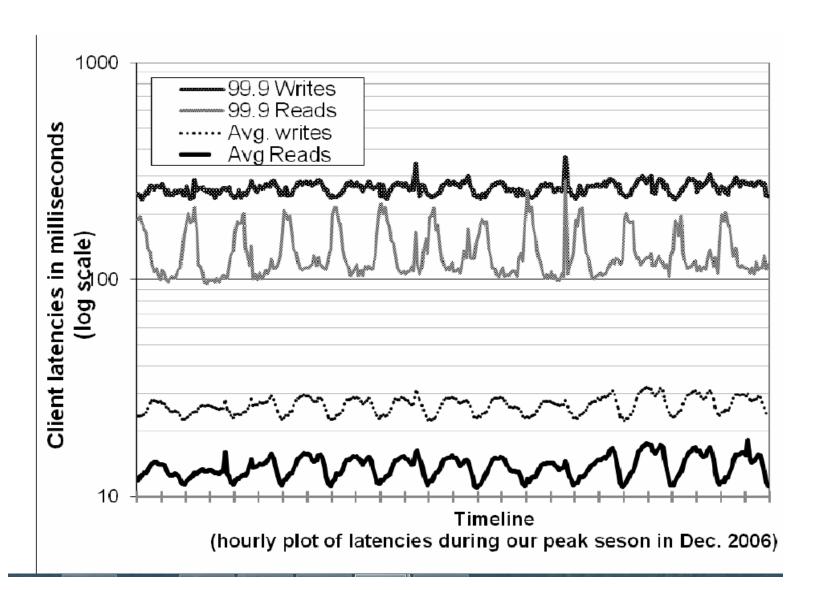
 A new node takes over its share of vnodes from other nodes

 For balanced load, it should take roughly the same number of vnodes from each other node

Requires many more vnodes than nodes!

Adding a node can take a day...

How well does it work?



Riak

- Riak Core
 - Distribution framework: vnodes, consistent hashing, the ring, vector clocks, etc
- Riak KV
 - The key-value store
- Riak Map-Reduce
 - For aggregating data
- Riak Pipe
 - A generalisation of map-reduce
- ...

Borrowing from Sean Cribbs...

 Magical Horizontally-Scaling Unicorns



- Magical Horizontally-Scaling Unicorns
- Application-Friendly Rainbows in the Cloud



- Magical Horizontally-Scaling Unicorns
- Application-Friendly Rainbows in the Cloud
- Data-Shredding
 MapReduce Ninjas

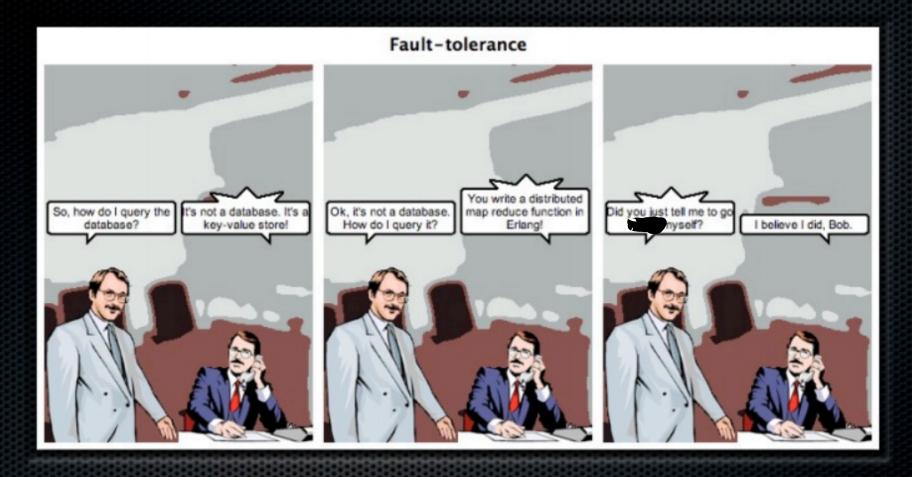


This Thursday



Guest lecture: Russell Brown from Basho in the UK

- CRDTs: Convergent Replicated Datatypes (a recent Riak extension for higher performance)
 - What are they?
 - Equivalents in Haskell
 - Turning research into a product
 - QuickCheck's rôle in all of this



Riak Map-Reduce

I believe I did, Bob.