



Using Erlang, Riak and the ORSWOT CRDT at bet365 for Scalability and Performance

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Background



bet365 in stats

- Founded in 2000
- Located in Stoke-on-Trent
- The largest online sports betting company
- Over 19 million customers
- One of the largest private companies in the UK
- Employs more than 2,000 people
- 2013-2014: Over £26 billion was staked
 - Last year is likely to be around 25% up
 - Business growing very rapidly!
- Very technology focused company

bet365 technology stats

- Over 500 employees within technology
- £60 million per year IT budget
- Fifteen datacentres in seven countries worldwide
- 100Gb capable private dark fibre network
- 9 upstream ISPs
- 150 Gigabits of aggregated edge bandwidth
 - 25 Gbps and 6M HTTP requests/sec at peak
- Around 1 to 1.5 million markets on site at any time
- 18 languages supported
- Push systems burst to 100,000 changes per second
 - Almost all this change generated via automated models
- Database systems running at > 500K TPS at peak
- Over 2.5 million concurrent users of our push systems
- We stream more live sport than anyone else in Europe



Production systems using Erlang and Riak

• Cash-out

A system used by customers to close out bets early.

Stronger

An online transaction processing (OLTP) data layer.



Why Erlang and Riak?



Our historical technology stack

- Very pragmatic
- What would deliver a quality product to market in record time
- Mostly .NET with some Java middleware
- Lot and lots of SQL Server



But we needed to change

- Complexity of code and systems
- Needed to make better use of multi-core CPUs
- Needed to scale out
 - Could no longer scale our SQL infrastructure
 - Had scaled up and out as far as we could
 - Lack of scalability caused undue stress on the infrastructure
 - Lead to loss of availability



Erlang Adoption

Erlang – Key learnings

- You can get a lot done in a short space of time. A plus and a minus!
- Tooling is limited
- Hot code upgrades with state can be hard
- Dependency management could be better
- Get as much visibility into the system as possible e.g. stats, data etc
- Use OTP / reuse proven code
- Keep to standards (e.g. code layout etc)

bet₃₆₅

Erlang – Key learnings

- Check your supervision tree
- Message passing is a double edged sword
- Keep state small (e.g. gen_server state)
- Binaries and GC: Heap vs reference-counted
- Explore whether you need the Transparent Huge Pages (THP) feature in the Linux kernel
- Don't use error_logger as your main logger
- Validate all data coming into the Erlang system at the edge



Riak Adoption

Riak – Brief overview

- Key value store
- Inspired by the Dynamo paper
 - http://www.read.seas.harvard.edu/~kohler/class/cs239-w08/decandia07dynamo.pdf
- Traditionally, an eventually consistent system (AP from CAP)
 - Riak 2.0+: Introduction of a strongly consistent option (CP from CAP)
- A Riak cluster is a 160-bit integer space the Riak Ring
- Split into partitions a virtual node (vnode) is responsible for each one
- A vnode lives on one of the Riak nodes
- Data is stored in a number of partitions (n_val setting default 3)
 - Consistent hashing technique helps with identifying the partitions for putting and getting the data



Riak – Why?

- Open source aspect
- Uses Erlang
- Based on solid ideas
- Horizontally scalable
- Highly available
- Masterless: No global locks performance is predictable
- Designed to be operationally simple
- Support and community exists

Riak – Key learnings

- Eventually consistent: Eventually the data will converge to the consistent value
 - Keep in mind:
 - A get/read may return an old value
 - A put/write may be accepted for a key at the same time as another concurrent put for the same key in the cluster (i.e. no global locking)
 - Bend your problem! E.g. Look at it from another side
- Data model for your use case i.e. normalisation isn't key: Trade off puts vs gets for your use case
- No bigger object sizes than 1MB
- Store data in a structure which helps with version upgrades
- Riak Enterprise
 - Multi-Datacenter Replication + Support

Riak – Key learnings

- Consult Riak's System Performance Tuning documentation
- Different internode network vs inbound
- Monitor network and disk usage
- Use the "riak-admin diag" command
- Use Basho Bench to load test your cluster
- For bitcask backend: load test merging and tune for your use case
 - Setting log_needs_merge to true will help with this tuning
- Allow siblings (i.e. allow_mult = true)
 - With resolving siblings asap

Riak – What are siblings?

- A sibling happens when Riak does not know which value is the causally recent (E.g. because of concurrent puts/writes)
 - Uses version vectors to know this
 - · Version vector A is Concurrent to version vector B (as opposed to Descends or Dominates)
 - Explained later
 - Referenced as vector clocks in the Riak documentation should have been named version vectors
 - Talk: A Brief History of Time in Riak. Sean Cribbs (Basho). RICON 2014
 - https://haslab.wordpress.com/2011/07/08/version-vectors-are-not-vector-clocks/
 - Similar logic, however:
 - Vector clocks is about tracking events to a computation
 - Version vectors is about tracking updates to data replicas
 - Riak 2.0 introduced the option of dotted version vectors instead
 - Similar idea to the ORSWOT CRDT dot functionality (explained later)
 - Reduces potential number of siblings (i.e. causality tracking is more accurate) -> limits sibling explosion
- All sibling values are returned (i.e. more than one)
 - Big difference to the normal experience with SQL type data stores

Riak – allow_mult=false

- You can set allow_mult to false (i.e. no siblings to the client) with:
 - last_write_wins set to false
 - Uses version vectors. In conflict, the sibling with the highest timestamp wins.
 - last_write_wins set to true
 - Doesn't use version vectors new value overwrites current value
- However, not recommended* because of potential data loss
 - Network problems
 - Reading: Fallacies of Distributed Computing Explained, Arnon Rotem-Gal-Oz
 - Complexity of time synchronisation across servers (speed of light, machines fail etc)
 - Reading: There is No Now Problems with simultaneity in distributed systems, Justin Sheehy. 2015

* Perhaps for immutable data with separate unique keys



Riak – Dealing with siblings

- Sibling values are returned on a get request
- Need to have a merge function to produce the correct value
- The merge function should be deterministic by having the following properties:
 - Associativity
 - Order of the merge function being applied to the data doesn't matter as long as the sequence of the data items is not changed
 - Commutativity
 - Order into the merge function does not matter
 - Idempotence
 - · Merge function applied twice to the same value results in the same value

• Can be hard to get right

- Can lead to possible data loss and incorrect data



CRDTs

CRDTs – What are they?

- Conflict-Free Replicated Data Types
- Can be:
 - Operation based: Commutative Replicated Data Types
 - State based: Convergent Replicated Data Types
- Reduces complexity by having no client side siblings
 - But still having no data loss
- Readings:
 - A comprehensive study of Convergent and Commutative Replicated Data Types. Marc Shapiro, Nuno Preguiça, Carlos Baquero, Marek Zawirski. 2011.
 - Conflict-free replicated data types. Marc Shapiro, Nuno Preguiça, Carlos Baquero, Marek Zawirski. 2011.
 - CRDTs: Consistency without concurrency control. Mihai Letia, Nuno Preguiça, Marc Shapiro. 2009.
 - http://christophermeiklejohn.com/crdt/2014/07/22/readings-in-crdts.html



CRDTs – Operation based

- Commutative Replicated Data Types
- All replicas of the data are sent operational updates
- Relies more on a good network and reliably delivering updates
- Knowing the current true membership is more important

CRDTs – State based

- Convergent Replicated Data Types
- Data is locally updated, sent to replicas and merged
- Update function must be monotonically increasing
- Generally easier to understand than operation based
- Easier to have an elastic membership of replicas
- However, more data is sent around the network



CRDTs – Types

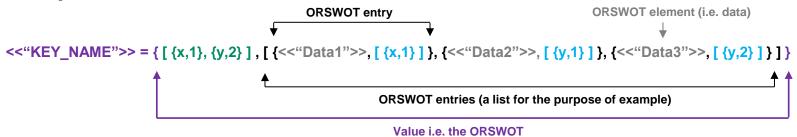
- Different data type implementations exist for:
 - Counters
 - Sets
 - Maps
 - ...
- For our core use case the Sets data type made sense
- However, we were and are using Riak 1.4+ and a Sets CRDT isn't available
 - Introduced in Riak 2.0+
 - At the time, Riak 2.0+ wasn't even a release candidate

CRDTs - riak_dt

- We decided to use the riak_dt dependency and integrate it ourselves into our system using Riak 1.4+
 - <u>https://github.com/basho/riak_dt</u>
 - Apache License Version 2.0 (http://www.apache.org/licenses/LICENSE-2.0)
- Different set based implementations exist (all state based CRDT's):
 - G-Set: Grow only set
 - i.e. no remove
 - OR-Set: Observe Remove Set
 - Able to add and remove
 - · However, when an element is removed a tombstone still exists i.e. Size problem
 - ORSWOT: Observe Remove Set Without Tombstones
 - Able to add and remove, but doesn't have tombstones
 - In a concurrent add and remove of the same element -> add-wins
 - Reading: An optimized conflict-free replicated set. Annette Bieniusa, Marek Zawirski, Nuno Preguiça, Marc Shapiro, Carlos Baquero, Valter Balegas, Sérgio Duarte. 2012.
- For our core use case we chose to use the ORSWOT



• Example:



Version vector exists as part of the ORSWOT:

- List of tuples
- Each tuple being {UniqueActorName, Counter}
- ORSWOT operations happen through an unique actor
- Similar but a different instance of version vector to what will also exist for the overall key/value

Dots exist for each element in the ORSWOT set:

- Just a minimal version vector
- Each dot pair represents a tuple in the ORSWOT version vector at a particular point
- Usually a list of one
 - Can be a list of more than one when, for example, two ORSWOTs for two concurrent adds of the same element (i.e. using two different unique actors) are merged



Adding an element:

- Version vector as part of the ORSWOT incremented
 - Counter for unique actor being used is incremented (or set as 1 if not currently in the version vector)
- The updated {UniqueActorName, Counter} pair is stored with the element as its dots
 - · If an entry already exists in the ORSWOT for the element, this is replaced

Example:

Adding <<"Data2">> using unique actor y to the existing ORSWOT:

{ [{x,1}] , [{<<"Data1">>, [{x,1}] }] }

Results in the new ORSWOT:

```
{ [ {x,1}, {y,1} ] , [ {<<"Data1">>, [ {x,1} ] }, {<<"Data2">>, [ {y,1} ] } ]
```

and ORSWOT value (i.e. ignoring metadata / what a client would be interested in) of:

[<<"Data1">>, <<"Data2">>]



Removing an element:

- Version vector as part of the ORSWOT does not change
 - Of course when the put happens to store the ORSWOT's updated value, the version vector for the overall key/value is incremented
- The elements entry is simply removed from the ORSWOT
 - i.e. No tombstones

Example:

Removing <<"Data1">> from the existing ORSWOT:

```
{ [ {x,1}, {y,1} ] , [ {<<"Data1">>, [ {x,1} ] }, {<<"Data2">>, [ {y,1} ] } ]
```

Results in the new ORSWOT:

{ [{x,1}, {y,1}] , [{<<"Data2">>, [{y,1}] }] }

* Further options do exist, such as being able to delay removes

- E.g. the ORSWOT object doing a remove might have not seen the original add for the element yet (e.g. because of being a replica not merged with the add yet)
- Would also add further logic to the merge operation

CRDTs – ORSWOT overview

Merging ORSWOT's: ORSWOT A and ORSWOT B

- E.g. because of siblings detected by version vectors for the overall key/value
- Version vectors for ORSWOT A and ORSWOT B merged
 - i.e. the least possible common descendant of both (*)
- Elements merged:
 - Common elements only kept if there exists a non-empty dots for them from merging (*):
 - Common dot pairs for the element in ORSWOT A and ORSWOT B
 - Dot pairs for the element only in ORSWOT A where the dot pair count is greater than any count* for the same actor in ORSWOT B's version vector
 - Dot pairs for the element only in ORSWOT B where the dot pair count is greater than any count* for the same actor in ORSWOT A's version vector
 - Elements only in ORSWOT A only kept if there exists a non-empty dots for them after:
 - Keeping only dot pairs for the element where the dot pair count is greater than any count* for the same actor in ORSWOT B's version vector
 - Elements only in ORSWOT B only kept if there exists a non-empty dots for them after:
 - Keeping only dot pairs for the element where the dot pair count is greater than any count* for the same actor in ORSWOT A's version vector

* As you might think, if there doesn't exist an appropriate actor pair/count in the version vector then the dot pair is merged/kept



Merging example:

ORSWOT A:	{ [{x,1}, {y,2}] , [{<<"Data1">>, [{x,1}] }, {<<"Data2">>, [{y,1}] }, {<<"Data3">>, [{y,2}] }] }	
Seen:	 Adding element <<"Data1">> via actor x Adding element <<"Data2">> via actor y Adding element <<"Data3">> via actor y 	
ORSWOT B:	{ [{x,1}, {y,1}, {z,2}] , [{<<"Data2">>, [{y,1}] }, {<<"Data3">>, [{z,1}] }, {<<"Data4">>, [{z,2}] }] }	
Seen:	 Adding element <<"Data1">> via actor x Adding element <<"Data2">> via actor y Adding element <<"Data3">> via actor z Adding element <<"Data4">> via actor z Removing element <<"Data1">> via actor z 	
Merged ORSWOT:	{ [{x,1}, {y,2}, {z,2}] , [{<<"Data2">>, [{y,1}] }, {<<"Data3">>, [{y,2}, {z,1}] }, {<<"Data4">>, [{z,2}] }] }	



Merging example:

ORSWOT A:	{ [{x,1}, {y,2}] ,	<pre>{<<"Data1">>, [{x,1}] }, {<<"Data2">>, [{y,1}] }, {<<"Data3">>, [{y,2}] }] }</pre>
ORSWOT B:	{ [{x,1}, {y,1}, {z,2}] ,	<<"Data2">>, [{y,1}] }, {<<"Data3">>, [{z,1}] }, {<<"Data4">>, [{z,2}] }] }
Merged =	{ [{x,1}, {y,2}, {z,2}] ,	{<<"Data2">>, [{y,1}] }, {<<"Data3">>, [{y,2}, {z,1}] }, {<<"Data4">>, [{z,2}] }] }

- [{x,1}, {y,2}, {z,2}] is the least possible common descendant for the version vectors of ORSWOT A and ORSWOT B
- Common elements:
 - <<"Data2">> : ORSWOT's have common dot pair {y,1} no other dot pairs to consider/merge with it => element in
 - <<"Data3">> : ORSWOT A has {y,2} which has a greater count than {y,1} in ORSWOT B's version vector [{x,1}, {y,1}, {z,2}] ORSWOT B has {z,1} which is included because ORSWOT A's version vector [{x,1}, {y,2}] doesn't have a count for z. Therefore {y,2} and {z,1} are merged to give [{y,2}, {z,1}] => element in
- Elements only in ORSWOT A:
 - <<"Data1">> : From ORSWOT A, no dots exist from [{x,1}] which are greater than ORSWOT B's version vector
 [{x,1}, {y,1}, {z,2}] => element not in
- Elements only in ORSWOT B:
 - <<"Data4">> : From ORSWOT B, dot pair {z,2} for the element is kept because ORSWOT A's version vector [{x,1}, {y,2}] doesn't have a count for z => element in



CRDTs – riak_dt integration

- Erlang middle layer between clients and Riak
- Middle layer needed to have unique actors (gen_server's) as part of using the ORSWOT implementation
 - With getting a balance on the number of them e.g. due to impacting ORSWOT version vector size
- Unique actor names pre-defined as part of server setup configuration
 - With making them small e.g. due to impacting ORSWOT version vector size
- Clients don't have to deal with siblings
- Middle layer (i.e. client to Riak) does need to bring back siblings
 - Not as good as a Riak server side CRDT (i.e. Riak 2.0+)
- Still using version vectors on the overall key/value's
 - i.e. to know to do the ORSWOT merge operation



Getting It Live



Tooling

- From day 1 we ate our own dog food
- Monitoring
- Performance counters
- Error reporting (including correlation between systems)
- Built custom adhoc query tool
 - On replicated Riak cluster
- Custom reconciliation between new system and old system
- Created build and release scripts for automation



Released in phases

- Able to build on stable ground
- Able to get data to impact future decisions
- Built confidence
- Move functionality to using Erlang and Riak
- Not all phases were immediately business impacting
- However, overall able to get business impacting functionality out sooner



Replay testing

- Captured logs asynchronously in one phase
- Common interface between old and new systems
- Able to do reconciliation between the systems
- Big range of test data / realistic load profile
- Used for functional and performance testing
- Different logs for long weekend run vs quick run
- Complemented other testing such as specific unit/integration testing, fuzz testing and formal UAT

Performance testing

- Done early and repeated
- Built custom client using replay logs
- Able to increase load profile easily
- Used our custom tooling / monitoring
- Identified bottlenecks / tested horizontal scalability of system
- Adhoc changes and fed back into development
- Had a specific profile which could give a fair test between changes
- Basho Bench used to sanity test Riak cluster setup



Failure testing

- Did failure testing
- Built confidence and understanding
- Trying to make sure a failure seen in production isn't the first time it's experienced
- Tested common procedures e.g. taking nodes in and out of service
- Failures will happen embrace them

Today with Stronger

- The project was a success!
- We have a system which is:
 - Performant and able to deal with many times our peak load
 - Reliable and deals with failure using minimal human intervention
- We have introduced the business to a number of new technologies
- We have grown the capabilities of the business and our people



Questions?

