Towards a Solution to the Red Wedding Problem

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Abstract

Edge computing promises lower latency interactions for clients operating at the edge by shifting computation away from Data Centers to Points of Presence which are more abundant and located geographically closer to end users. However, most commercially available infrastructure for edge computing focuses on applications without shared state. In this paper, we present the Red Wedding Problem, a real-world scenario motivating the need for stateful computations at the edge. We sketch the design and implementation of a prototype database for operation at the edge that addresses the issues presented in the Red Wedding Problem and present issues around implementing our solution on commercial edge infrastructure due to limitations in these offerings.

1 Edge Computing

Edge computing promises lower latency interactions for clients operating at the edge by shifting computation away from Data Centers (DCs) to Points of Presence (PoPs) which are more abundant and located geographically closer to end users. Not only do PoPs reduce latency, but they also serve to alleviate load on origin servers enabling applications to scale to sizes previously unseen.

Recently, large-scale cloud providers have started providing commercial access to this edge infrastructure through the use of “serverless” architectures, giving application developers the ability to run arbitrary code at these PoPs. To support this, client code is normally executed inside of a transient container where the application developer must resort to using storage that’s typically provided only at the DC. Therefore, to take most advantage of the edge, application developers are incentivized to use these architectures for applications where there is no shared state, thereby ruling out one of the most common types of applications developed today: collaborative applications which rely on replicated, shared state.

The first generation of edge computing services provided by Content Delivery Networks (CDNs) let customers supply code that will be run at PoPs during the HTTP request-response cycle. Akamai [1], for instance, allows customers to author arbitrary JavaScript code that operates in a sandboxed environment and is permitted to modify responses before the content is returned to the user. Fastly [7] gives users the ability to extend their CDN with similar rewrite functionality by authoring code in the Varnish Configuration Language (VCL), which is also executed in a restricted environment. In both cases, customers are not allowed to inspect or modify content operating at the PoP and can only modify requests to and from the origin and to and from the end user.

Second generation services, such as Amazon’s Lambda [2] and its extension to the edge, Amazon Lambda@Edge, enables the customer to write stateless Lambda functions that either run inside one of Amazon’s region availability zones or at any number of the hundreds of PoPs it has through its CloudFront caching service. While Lambda provides black-box binary execution in a restricted containerized environment, Lambda@Edge is limited to Node.js applications. Both Google Cloud’s Cloud Functions and Microsoft Azure’s Functions at IoT Edge provide similar services to Amazon’s Lambda and Lambda@Edge.

This paper defines a real-world scenario for motivating stateful computations at the edge, titled the Red Wedding Problem. We sketch the design and implementation of a prototype potentially consistent, peer-to-peer replicated database operating at the edge using Amazon Lambda that is aimed at addressing the issues presented by the Red Wedding Problem. In realizing our prototype on real-world edge infrastructure, we ran into numerous limitations imposed by Amazon’s infrastructure that are presented here.
2 The Red Wedding Problem

We present the Red Wedding Problem\(^2\): an industry use case presented to us by a large commercial CDN in the United States around the handling of traffic spikes at the edge.

Game of Thrones [9] is a popular serial fantasy drama that airs every Sunday night on HBO in the United States. During the hours leading up to the show, throughout the hour-long episode, and into the hours following the airing, the Game of Thrones Wiki [6] experiences a whirlwind of traffic spikes. During this occurrence, articles related to characters that appear in that episode, the page for the details of the episode, and associated pages to varying concepts referenced in the episode undergo traffic spikes—read operations upon viewing related pages—and write spikes while updating related pages as events in the episode unfold.

While handling read spikes is what CDNs were designed for, the additional challenge of handling write spikes is what makes the Red Wedding Problem interesting. More specifically, the Red Wedding Problem requires that the programmer be made aware of and optimize for the following constraints:

- **Low latency writes.** By accepting and servicing writes at the PoP, user’s experience lower latency requests when compared to an approach that must route writes to the origin;
- **Increased global throughput.** By accepting writes at the PoP, and avoiding routing writes through to the origin DC, write operations can be periodically sent to the origin in batches, removing the origin DC as a global throughput bottleneck.

However, several design considerations of the Red Wedding Problem make the problem difficult to solve.

- **Storing state.** How should state be stored at the edge, especially when leveraging “serverless” infrastructures at the edge which are provided by most cloud providers today;
- **Arbitrating concurrent writes.** How should concurrent writes be arbitrated when accepting writes at the edge to minimize conflicts and maximize batching?
- **Application logic.** As clients do not communicate with the database directly in most applications, how should application logic be loaded and leveraged at the edge.

3 Solving the Red Wedding Problem

Our proposed solution is presented in Figure 1. In solving the Red Wedding Problem, we had the following design considerations:

- **Application logic at the edge.** Application logic for authentication and authorization, as well as for mapping user requests into database reads and writes, must also live at the PoP;
- **Elastic replica scalability at the PoP.** To avoid moving the bottleneck from the origin to the PoP, there must be elastic scalability at the PoP that allows the system to instantiate more data replicas on demand;
- **Inter-replica communication.** Replicas located at the PoP should be able to communicate with one another to allow for data sharing between instances at the edge without having to communicate through the origin server;
- **Convergent data structures.** As modifications will be effecting data items concurrently, the data model needs to support objects that have well-defined merge semantics to ensure eventual convergence;
- **Origin batching.** To alleviate load on the origin server, updates should be able to be batched while removing redundancy and periodically sent back to the origin server.

Our solution assumes that application code is containerized and can be deployed to the PoP to interpose on requests to the origin: enabling the application to scale independently at the edge. As demand from end users ramps up, more instances of the application are spawned to handle user requests. These instances of the application generate read and write traffic that is routed to the database. In traditional architectures, these application instances would normally communicate with a database operating at the DC.

We use a containerized database to enable data storage at the PoP. Each of these database instances would be scaled in the same manner as the application code at the PoP and its state would be bootstrapped from the data center upon initialization. These instances of the database would be instantiated on demand and interpose on read and write operations from the application code inside of the PoP.

Each replica running at the PoP should be able to bootstrap from other replicas, if still running, at the PoP. This enables faster bootstrapping and communication between replicas that can serve to speed up the anti-entropy process within the PoP, increasing the probability that data items are the freshest amongst the different replicas within the same PoP.

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1 Private communication, Fastly.
2 Many examples of the Red Wedding Problem exist: live European football game commentary posted on Reddit is one such example.
Finally, the use of concurrent data structures can be used at the edge to avoid conflicting updates that cannot be merged. For instance, several previous works [5, 12, 3] have identified how to build abstract data types where all operations commute or have predefined merge functions to avoid issues where changes do not commute.

4 Implementation

We have developed a prototype that enables us to store content inside of transient serverless invocations using Lambda. To achieve this, we have taken an open source distributed peer-to-peer key-value database [10, 11] written in Erlang and have embedded it inside of a Node.JS 6.10 application that is deployed to Lambda in several AWS regions.

4.1 Lambda

Amazon’s Lambda is a stateless computation service for end users, allowing them to upload functions that will be invoked in response to triggered events. With Lambda, users specify the events that uploaded code should respond to and users’ application code is guaranteed to automatically scale, when necessary, by creating additional invocations to deal with the increasing demand.

Invocations in Lambda are designed to be opaque. Users upload a compressed file containing their application code and upon the first event, the application code will be decompressed and executed inside of a container. When the invocation completes, the container is paused until the next invocation; however, after a period of somewhere between 5 to 60 minutes, the container will be terminated. Containers will be reused for invocations when possible, but concurrent invocation will cause additional containers to be created on demand. This mechanism is transparent to the end user, as the user only interacts on a per-event level.

As a result of this, invocations are incentivized to be completely stateless, or authorized and able to store and fetch required state from an external service, such as S3 or DynamoDB, as Amazon recommends.

4.2 Prototype

We built an eventually-consistent replicated peer-to-peer database that runs in Lambda. The interface provided to the user is Redis-like; read and update operations are identified by a unique key and a data structure type that are issued against the database.

Data structures provided by the data store are Conflict-Free Replicated Data Types [12] (CRDT), distributed
data structures with a merge function to ensure that state
remains convergent in the event of concurrent modifications
at multiple locations. CRDTs come in a variety of
different flavors providing distributed versions of many
common abstract data types: ie. registers, sets, dictionaries,
graphs.

Operations are always performed at a local replica and
asynchronously propagated to other nodes in the system
using an anti-entropy [4] protocol: this prevents blocking
the system for write acknowledgements.

When the Lambda function is invoked, an instance of
the database starts up. There can be multiple instances of
the database running in each Amazon region. Every time
da database instance is invoked, it kicks off a compulsory
anti-entropy session. The databases bootstrap one an-
other by running anti-entropy sessions over AMQP. All data
structures in the database are CRDTs in order to
avoid conflicting updates.

4.3 Limitations of Lambda

In building our prototype on Lambda, we ran into a num-
ber of complications. We discuss those here.

Inter-node communication. Inter-node communication
is required for the compulsory anti-entropy sessions
amongst database replicas. Lambda does not allow pro-
cesses to bind ports inside of the container they are ex-
cuting in. Therefore, our key-value store was unable
to open sockets and receive incoming connections from
other nodes in the system. To work around this limi-
tation, we used an external message queue service that
provided AMQP and sent out all of the inter-node com-
munication on this transport layer. Given nodes could
not accept incoming connections, we established connec-
tions out to an external AMQP broker and reused those
connections to receive traffic from other nodes.

Concurrent invocations. Utilizing multiple invoca-
tions at the same time is key to elastic scalability and the
operation of multiple replicas. Lambda’s unique design
does not provide the developer any insight to the num-
ber of containers that are currently being invoked, and
scale-out is transparent as demand calls for it. Therefore,
we needed a mechanism for all nodes to identify one an-
other in order to route messages to each other and partic-
ipate in an anti-entropy session. To achieve this, we set a
single topic on the AMQP broker for membership com-
munication. On this specific channel, a single CRDT set
containing all of the members of the cluster gets period-
ically broadcast to all nodes in the system. Upon receipt
of this membership information, each node updates its
local view of membership.

Per-invocation work. Keeping the instances alive long
does not provide the developer any insight to the num-

the throughput of the system and ensuring no updates
are lost. Each Lambda has a maximum invocation of
300 seconds. Therefore, we ensure that a Lambda is in-
voked per-region faster than that interval, and after ser-
vicing a request, each invocation performs a compulsory
anti-entropy session with its peers to populate incoming
replica state and establish that existing replicas dissemi-
nated their state before termination. Scaling out the num-
ber of concurrent invocations only required increasing
this interval per-region.

4.4 Lambda@Edge Challenges

Operating our key-value store on Lambda was the neces-
sary first step towards moving to Lambda@Edge, Amaz-
on’s extension of Lambda to CloudFront PoPs. Once
running on Lambda@Edge, we should be able to inter-
pose on writes to and from the origin servers through
CloudFront, and, therefore, provide low-latency writes to
clients, serviced from their local and closest PoP, rather
than from their origin server.

In planning our migration to Lambda@Edge, we ran
into a number of other complications. We discuss those
here.

Message brokering at the edge. Inter-node communi-
cation at the PoP is important for keeping the anti-
entropy sessions as efficient as possible and increasing
system scalability. One notable drawback to Amazon’s
infrastructure is that there is no mechanism for colocat-
ing a message broker at a CloudFront cache location.
Since Lambda invocations cannot communicate with one
another directly, this requires that Lambdas must com-
municate through their closest Amazon region and avail-
ability zone, inflating latency costs to the local availabil-
ity zone for anti-entropy between instances. As nodes
will communicate solely with their local broker, and
therefore may not observe messages in the same order as
other nodes in the system, our convergent CRDT-based
model is essential.

Application code. Operation of application code at the
edge enables our system to keep most of the execution
inside of the PoP and not rely on the DC for request pro-
cessing. Therefore, our design assumes that users will be
able to run a component of their application code at the
edge location as well.

5 Related Work

ExCamera [8] is a system for parallel video encoding de-
veloped on top of Lambda to parallelize video encoding.
ExCamera partitions the job of video encoding into a se-
ries of chunks and uses a work-queue pattern leveraging
the maximum amount of Lambda instances available to reduce video encoding time as much as possible.

Our system differs from the ExCamera work in a number of ways.

- ExCamera relies on the use of a central rendezvous server for passing messages between different Lambda invocations. Our system relies on the use of AMQP, which supports geographically distributed exchanges and brokers to be leveraged for message delivery, allowing for our nodes to communicate with one another through their closest geographically located exchange;

- State in the ExCamera system is stored in Amazon S3. In our system, state is stored in the transient instances themselves, which are bootstrapped by other nodes as they come up online, and that handoff their state to other nodes before termination;

- Finally, ExCamera was primarily designed for work dispatch and as fast as possible concurrent processing by a large number of Lambda instances. Our system is designed for transient state management at the edge, mainly to reduce real-time latency costs during traffic spikes, as demonstrated by The Red Wedding Problem.

6 Conclusion

Recent innovations in publicly available edge computing services, such as Amazon’s Lambda@Edge, enable developers to run application code closer to end users, taking advantage of proximity resulting in both lower latency interactions and increased scalability. However, most of these services available only enable stateless computation at the edge or require collaborative applications that operate with shared state use a data store that’s located at one of the provider’s data centers, thereby reducing the full potential for edge computing. This paper presents an alternative design for a system that enables these interactions at the edge.

References


