





Business Requirements

- Near real-time activity processing
- Billons activities per customer per day
- Improve cost efficiency of operations while scaling up
- Global enterprise grade security and governance



SAAS Requirements

- Customers are added and removed
- Fairness and throttling per customer
- Strict sequential event processing for some applications
- Temporarily suspend a customer, when errors occur





Use Cases

React to activities

- Send an email when someone visits a web page
- Change the score when someone fills a form

Replicate data

- Build Solr Indexes, near real-time
- Update DataXChange an internal lead cache
- Syncto/from CRM Systems

Analytics

- Incrementally update email reports
- Enrich activities and feed to Druid for advanced email/web reports



Why Spark Streaming?

- Micro-batching provides sink-side efficiencies
- Great integration with Kafka
- No strict real time processing requirements
- Great community, industry adoption



Challenges with Spark + Kafka

- No way to add/remove topics on the fly
- No out of the box support for sequencing RDDs
- No support for turning off topics under errors
- Does not play well with scaling Kafka partitions up/down, when ordering is required



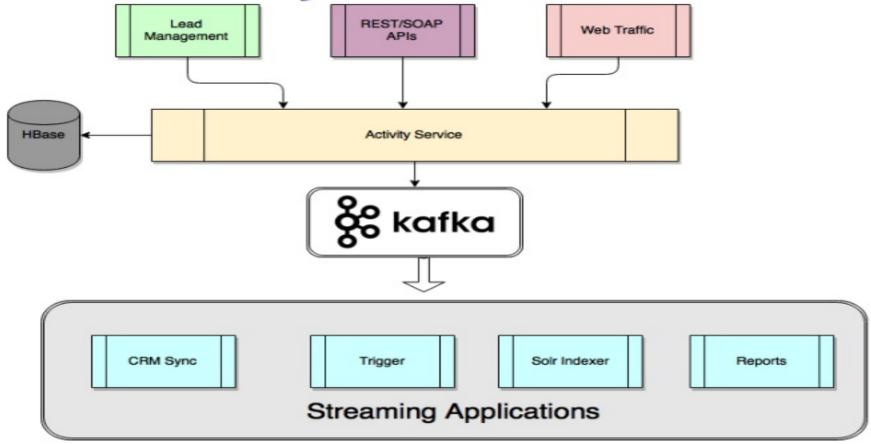
Challenges - Stragglers

- A batch can't complete until the slowest operation finishes
- Many of our batches include slow operations
 - Sometimes don't complete within the batch time
- Batches are multitenant
 - one customers operation can delay processing for other customers in the same batch
- Severe impact on utilization & batch delay





Marketo Activity Architecture





Kafka Topics Organization

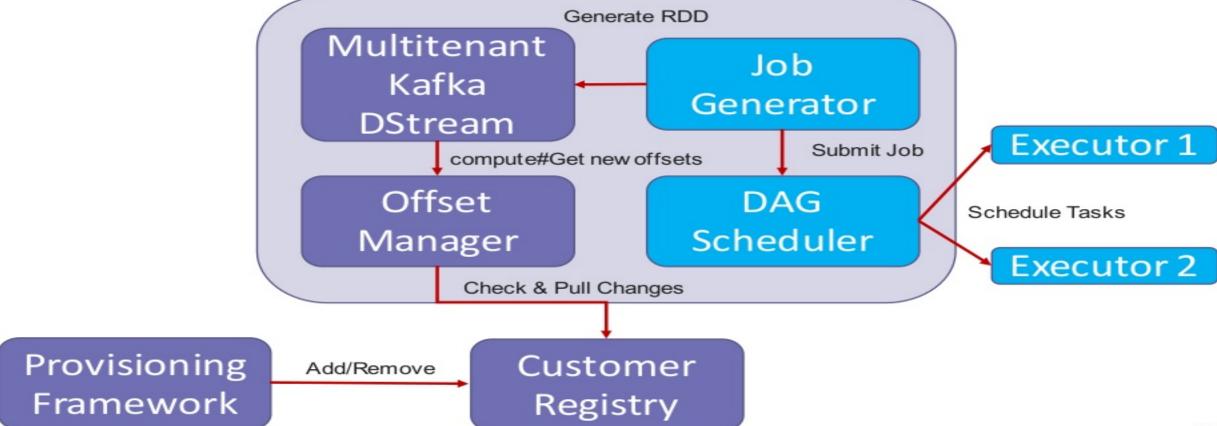
- One topic per use case, data from all customers
 - Easy to manage
 - A single customer can create backlogs for others during activity storms
 - Fairness/throttling is hard to implement
- One topic per use case, per customer
 - Storms are isolated to the customer
 - Fairness/throttling is easy to control, by tweaking the topic
 - Pressure on Kafka ZK so far not a problem





Dynamic provisioning capacity

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Marketo Offset Manager

- Tracks multitenancy
- Streaming Jobs process data for many customers
 - Accessing multiple Kafka topics and partitions
- Adds new topics
- Remove/Deactivate/Suspend topics



Multitenant DStream

- Enables efficient multitenant RDDs
- Controlled sequencing of RDDs
- Coalesce Kafka partitions
 - Bin-packing for efficiency
- Maintains partition lineage for offset management



Provisioning

- Manages allocating customers to a spark streaming application
- round robin + resource affinity
- Enables rebalancing of customers across spark streaming jobs
- Oozie based framework



Dynamic Resource Allocation

- SPARK-12133
 - Goal "make processing time infinitely close to duration"
 - Assumes tasks are roughly similar
- Stragglers throw this goal off
- What we really want :
 - DRA + Safe concurrent job execution



Results so far

- ~ 10 different use cases
- > 100 Spark Executors
- >1000 Kafka Partitions
- Processing latencies < 5s (99th %)
- Rolled out to ~20% customers





Application Scheduling

- Scheduling within an application to handle stragglers
- spark.streaming.concurrentJobs
- Exploring scheduler pools
- Changes to Streaming Job Scheduler, to execute multiple RDDs safely



Scaling Up Kafka Partitions

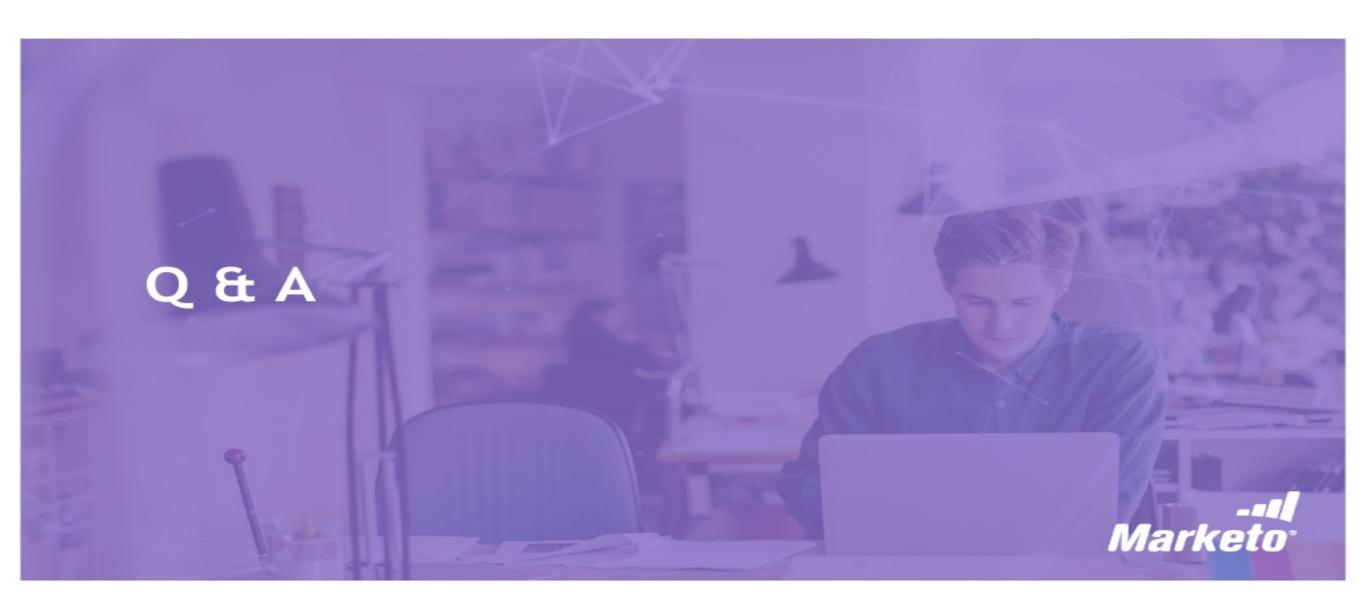
- Our customers grow in size over a period of time
- Ordering requirements mean we cannot alter topic on the fly
- Coordination required on both producer & consumer fronts
- Enhance provisioner to manage partition up/down scaling



Move to 2.x and Open Source!







Architecture Requirements

- Maximize utilization of hardware
- Multitenancy support with fairness
- Encryption, Authorization & Authentication
- Applications must scale horizontally





