

Stream Processing: Using Kafka and KSQL for Twitter data

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## Agenda



Problem statement



**Project Overview** 



Kafka



KSQL



Stream Processing with Twitter data

#### Problem Statement

With today's ever-increasing demand for real-time analytics, traditional batch-oriented data processing doesn't suffice.

#### Approach:

- Identify how stream processing allows us to harness the full value of the data; extracting valuable information.
- Use stream processing to automatically filter, aggregate, and analyze
   Twitter data

Project Overview Read tweets from Twitter's API by using KSQL's ability to look at Kafka event streams

Implement continuous stream processing from Python environments by writing simple SQL-like statements

Identify how Kafka and KSQL will help us analyze Twitter data and implement desirable machine learning methods to understand underlying information in Twitter data.



Stream Processing allows users to harness the full value of data as it is produced; extracting valuable information in motion

Requires different tools from those used in traditional batch processing architecture. We will use Apache **Kafka** and **KSQL** 

**Kafka** takes data published by 'producers', which maybe apps, files, or databases, and makes it available for 'consumers' subscribed to streams of different 'topics.'

**KSQL**, built on top of Kafka's Streams API, supports stream processing operations like filtering, transformations, aggregations, joins and windowing by using SQL statements instead of writing a lot of codes

## KAFKA capabilities

Distributed Streaming
Platform that can
publish and subscribe
to streams of records

Fault tolerant storage

Replicates Topic Log Partitions to multiple servers

Process records as they occur

Fast, efficient IO, batching, compression, and more

Used to decouple data streams

#### Kafka Use Cases

- Stream Processing
- Website Activity Tracking
- Metrics Collection and Monitoring
- Log Aggregation
- Real time analytics
- Capture and ingest data into Spark / Hadoop
- CRQS, replay, error recovery
- Guaranteed distributed commit log for in-memory computing



Lucidworks

DATADOG

#### Kafka Fundamentals

Records have a key (optional), value and timestamp; Immutable

Topic a stream of records ("/orders", "/user-signups"), feed name

Log topic storage on disk

Partition / Segments (parts of Topic Log)

Producer API to produce a streams or records

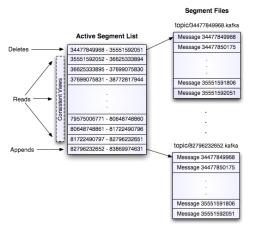
Consumer API to consume a stream of records

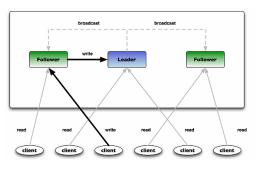
Broker: Kafka server that runs in a Kafka Cluster. Brokers form a cluster. Cluster consists on many Kafka Brokers on many servers.

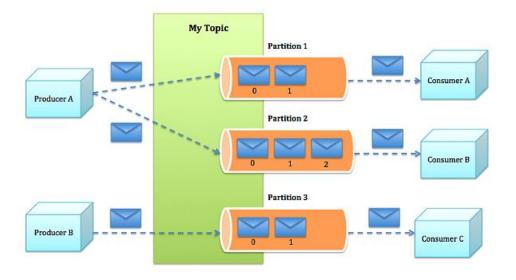
ZooKeeper: Does coordination of brokers/cluster topology. Consistent file system for configuration information and leadership election for Broker Topic Partition Leaders

### KAFKA Architecture

- Each partition of a topic corresponds to a logical log
- Physically, a log is implemented as a set of segment files of equal sizes
- Every time a producer publishes a message to a partition, the broker simply appends the message to the last segment file
- Segment file is flushed to disk after configurable numbers of messages have been published or after a certain amount of time elapsed
- Messages are exposed to consumer after it gets flushed.
- Consumer always consumes messages from a particular partition sequentially and if the consumer acknowledges particular message offset, it implies that the consumer has consumed all prior messages.







#### **KSQL**

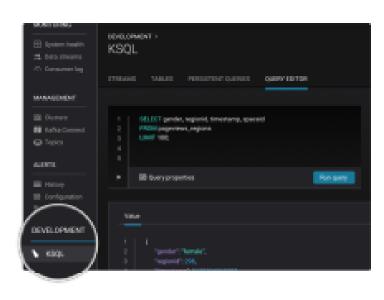
- Open Source
- · Zero Programming in Java, Scala
- Elastic, Scalable, Fault-Tolerant, Distributed
- Powerful Processing incl. Filters, Transforms, Joins, Aggregations, Windowing
- Runs Everywhere
- Supports Streams and Tables
- Exactly-Once Processing
- Event-Time Processing
- Kafka Security Integration





## Interactive KSQL usage

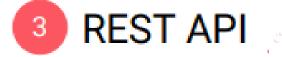
ksql>



POST /query







## KSQL for Data Exploration

- An easy way to inspect data in Kafka
- Join data from a variety of sources to see the full picture

```
SHOW TOPICS;

PRINT 'my-topic' FROM BEGINNING;

SELECT page, user_id, status, bytes
FROM clickstream
WHERE user_agent LIKE 'Mozilla/5.0%';
```

```
CREATE STREAM enriched_payments AS

SELECT payment_id, u.country, total

FROM payments_stream p

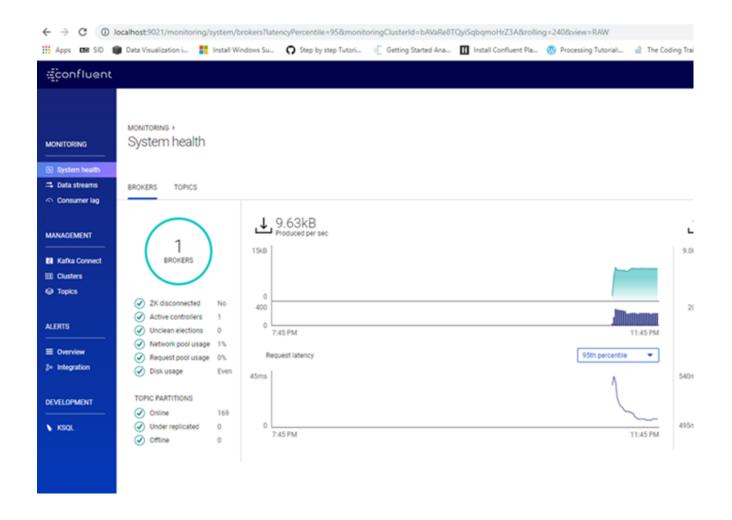
LEFT JOIN users_table u

ON p.user_id = u.user_id;

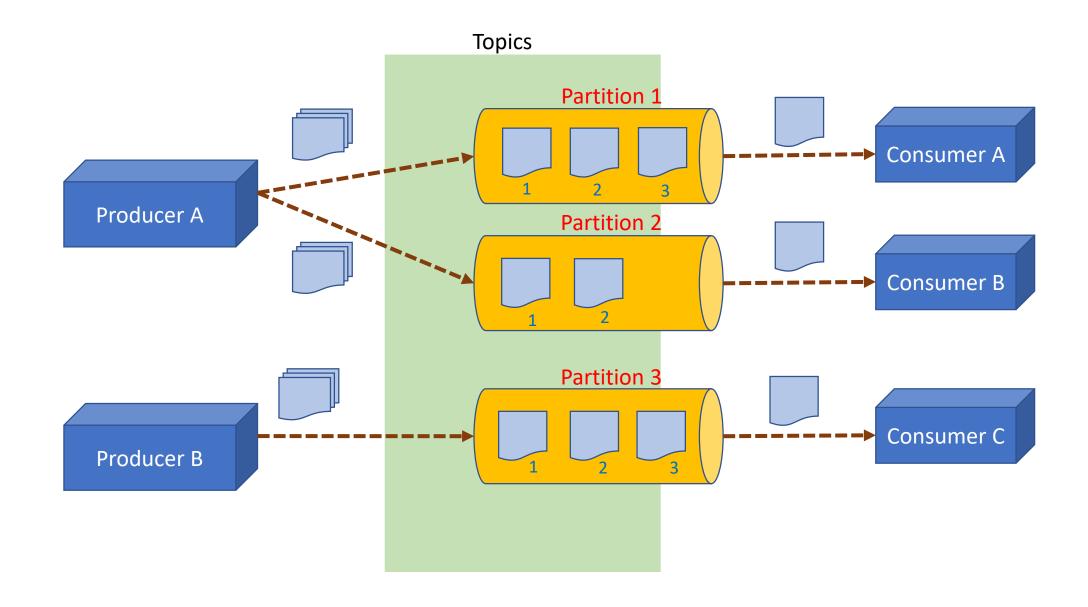
1 Stream-table join
```

#### Software Setup Confluent Kafka

- Installed Confluent Kafka on pure Linux environment
  - Issue with the Windows
     10 Linux install
- Ubuntu 18 was chosen because of its ease of installation and graphical environment
  - Ubuntu 18 is a open source Debian-based Linux distribution

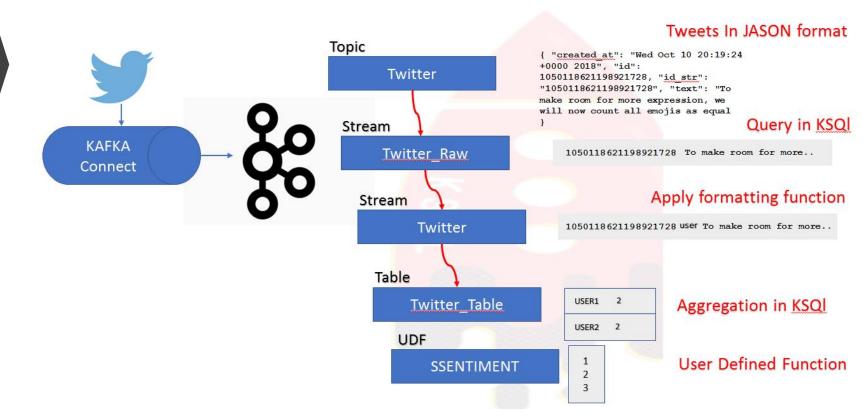


KSQL is the querying side of Kafka. It can be ran in both command line and a graphical environment and run from a web page on the local computer



# Analyzing Twitter Data in Apache Kafka through KSQL

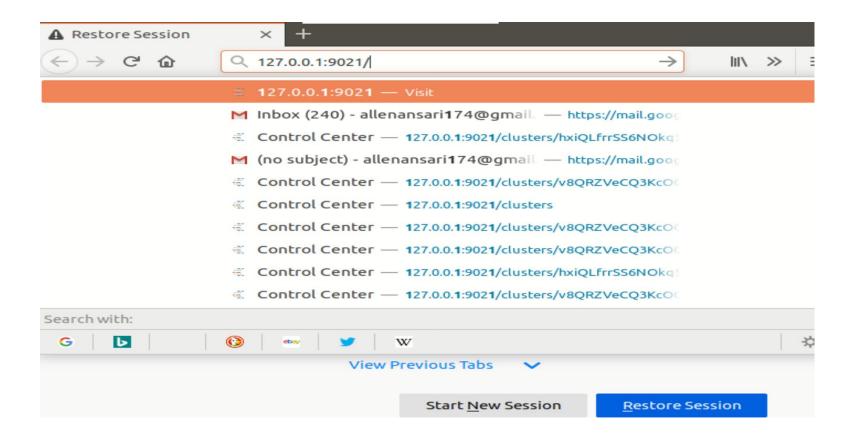
- Used Kafka Connect to pull the data from Twitter
- Used Twitter connector that uses the twitter streaming API to listen for status update messages and convert them to a Kafka Connect struct on the fly.
  - The goal is to match as much of the Twitter Status object as possible



## KSQL CLI

```
allen@allen:~$
```

## KSQL UI



## Twitter Sentiment Analysis

```
File Edit View Search Terminal Help
ksql> select text from twitter;
Press CTRL-C to interrupt
                                           allen@allen: ~
File Edit View Search Terminal Help
<sql> select ssentiment(text) from twitter;
Press CTRL-C to interrupt
```

1 = Negative

2 = Neutral

3 = Positive