

Amazon Kinesis and Amazon MSK overview Analyzing real-time data streams on AWS

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Agenda

- Amazon Kinesis overview
 - Introduction for Kinesis family
 - Key concept of Kinesis data stream
 - Lab
- Amazon MSK overview
 - Key concept of MSK
 - Lab
- Amazon Kinesis vs. Amazon MSK
- Online testing



Amazon Kinesis overview



Enabling real-time analytics

Data streaming technology enables a customer to ingest, process and analyze high volumes of high velocity data from a variety of sources in real time





Streaming data with AWS

Easily collect, process, and analyze data streams in real time









Capture and store data streams

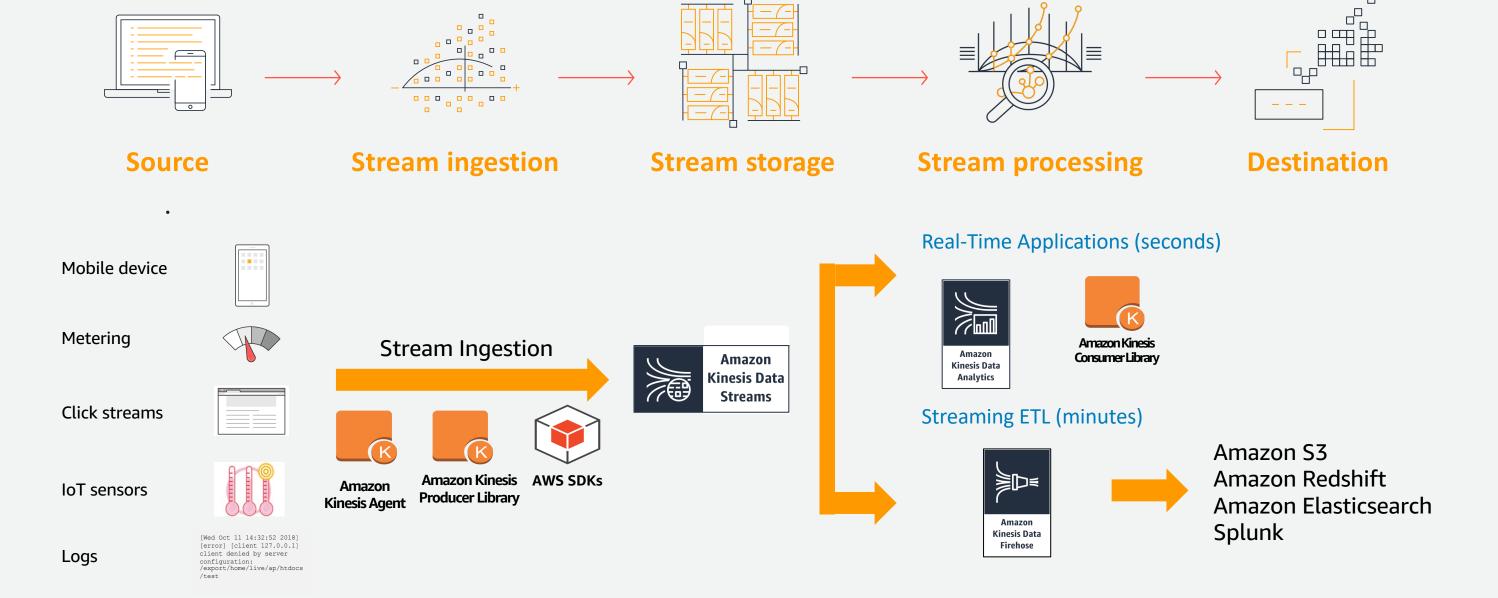
Analyze data streams in real time streams, data lakes

ETL streaming into and warehouses

Capture and store data streams



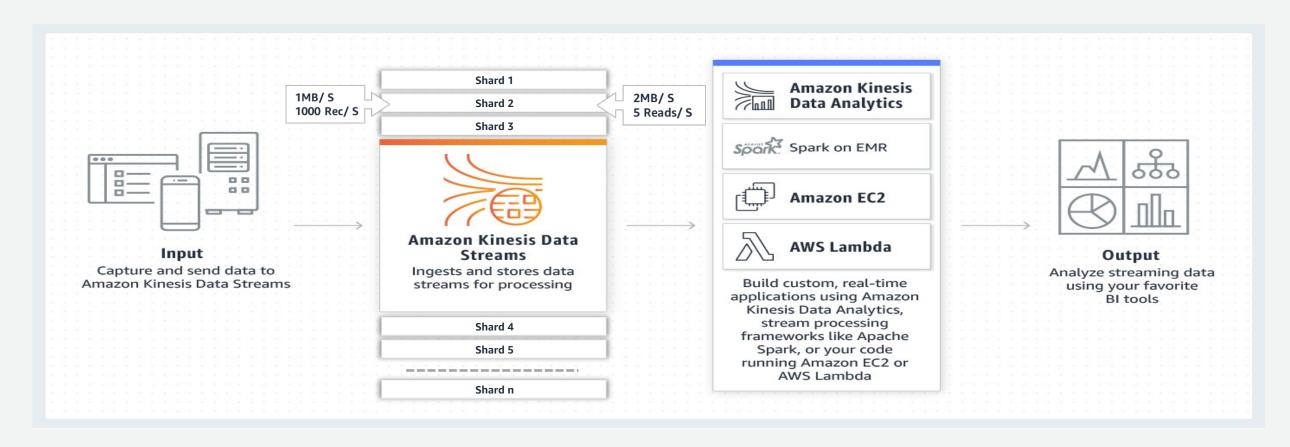
An Example Architecture





Amazon Kinesis Data Streams



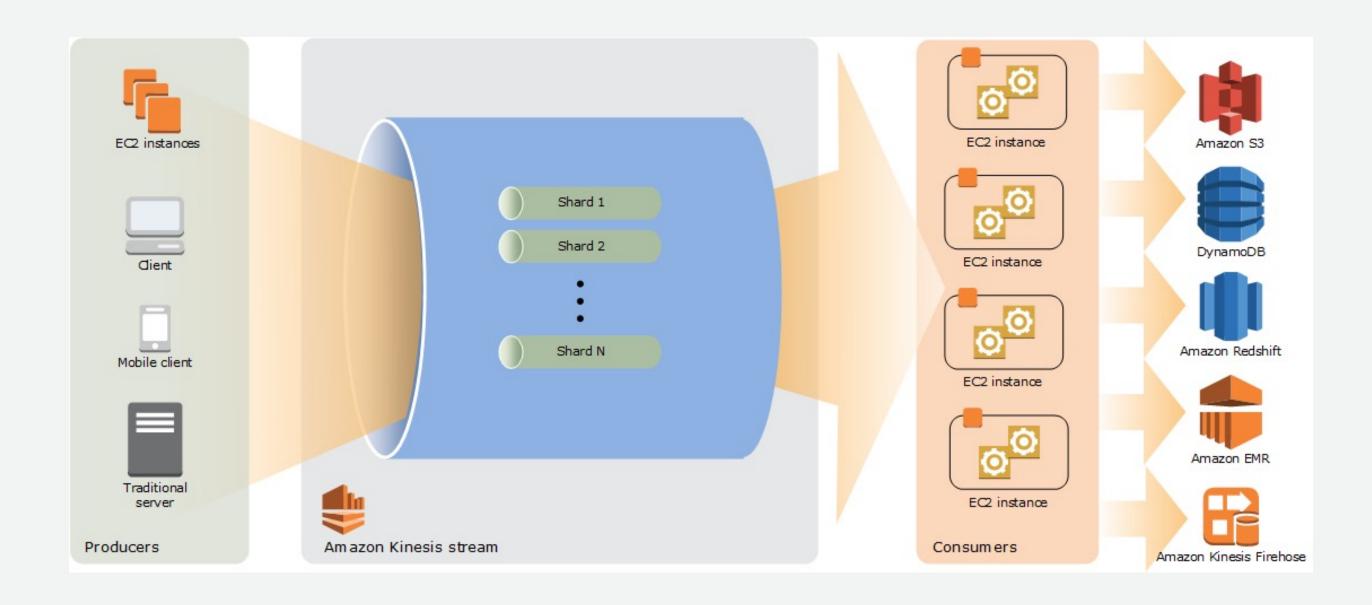


- Easy administration and low cost
- Real-time, elastic performance
- Secure, durable storage
- Available to multiple real-time analytics applications

- Average latency of 200ms with one standard consumer
- Enhanced Fan Out with SubscribeToShard API offers typical average latency of 70 ms

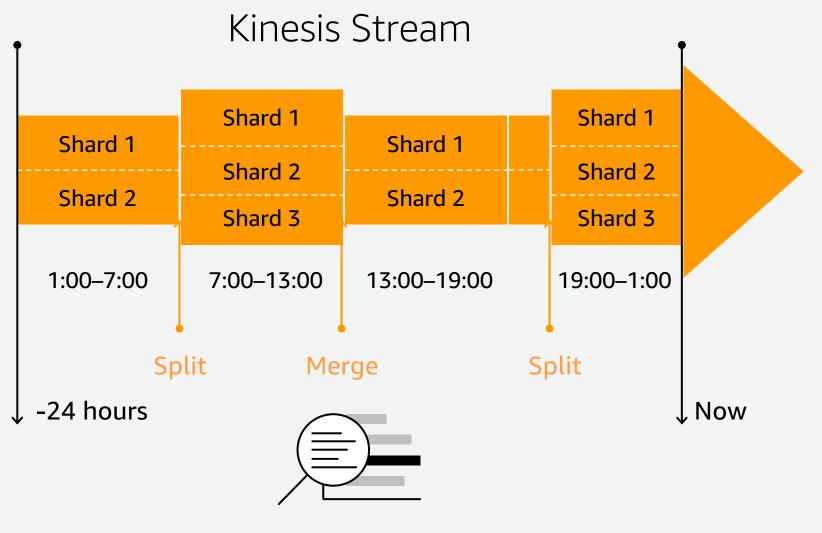


Kinesis Data Streams High-Level Architecture





Managed ability to capture & store data



Time-based seek

- Data streams are made of Shards
- Each shard ingests data up to 1MB/sec, and up to 1000 TPS
- Each Shard emits up to 2 MB/sec
- All data is stored by default for 24 hours and can be extended up to 365 days
- Scale Kinesis data streams by splitting or merging Shards
- Replay data inside of
 24 hour 365 day window



Amazon Kinesis Data Streams

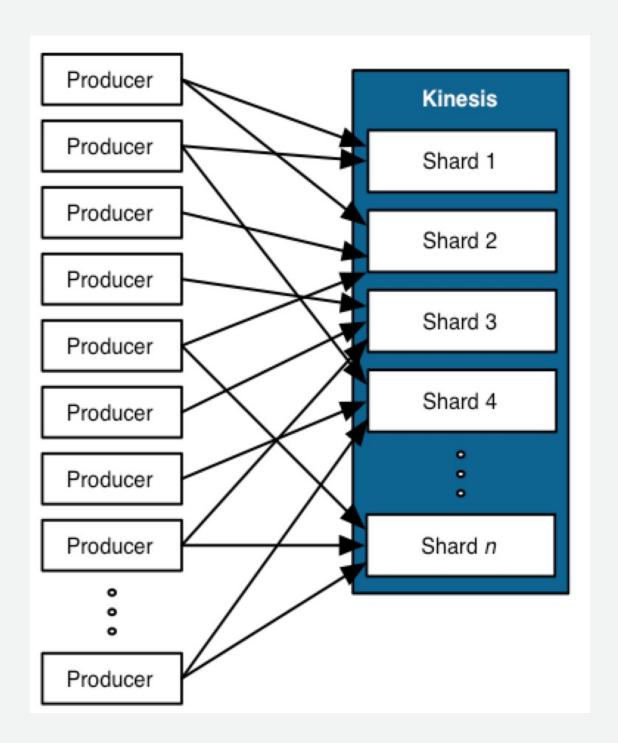
Core Concepts

Partition Keys

- Supplied by producer with each message
- Used to distribute data across shards

Sequence Numbers

- Returned to producer for each successfully written message
- Not an index to data in shard
- Can be used to enforce strict ordering





Amazon Kinesis Data Streams

Core Concepts

Message Ordering

Unordered processing

 Randomize partition key to distribute events over many shards and use multiple workers

Exact order processing

 Control partition key to ensure events are grouped into the same shard and read by the same worker

Need both? Use global sequence number





Amazon Kinesis Data Streams Data Producers

SDKs

Publish directly from application code via PutRecord and PutRecords APIs

Kinesis Agent

Tail log files and forward lines as messages to Kinesis Data Streams.

Kinesis Producer Library (KPL)

- Background process aggregates and batches messages.
- Producer application calls addUserRecord method.

3rd-party and open source

- Log4j appender
- Flume, fluentd source libraries



Kinesis Data Streams Standard Consumer



No additional cost

5 transactions/sec shard limit means 5 consumers can run concurrently

All consumers share the 2MB/sec shard limit

Lowest latency with a single consumer is about 200 millisecs

Uses the polling model to get records from a shard

To enable more than 5 consumers, use the fan-out pattern



Kinesis Data Streams Enhanced Fan-Out



Enhanced fan-out allows customers to scale the number of functions reading from a stream in parallel while maintaining performance

HTTP/2 data retrieval API improves data delivery speed between data producers and Lambda functions by more than 65%

Provides dedicated 2MB/s bandwidth per consumer – no bandwidth sharing with other consumers

Uses a push model to push records to consumers instead of a polling model

Not subject to the 5 transactions/sec/shard limit of the standard consumer

Achieves an average latency of 70 millisec

Does have an additional cost



Amazon MSK overview

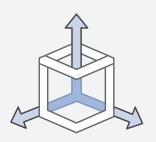


Challenges operating Apache Kafka

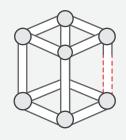
Difficult to setup



Tricky to scale



Hard to achieve high availability



AWS integrations = development



No console, no visible metrics

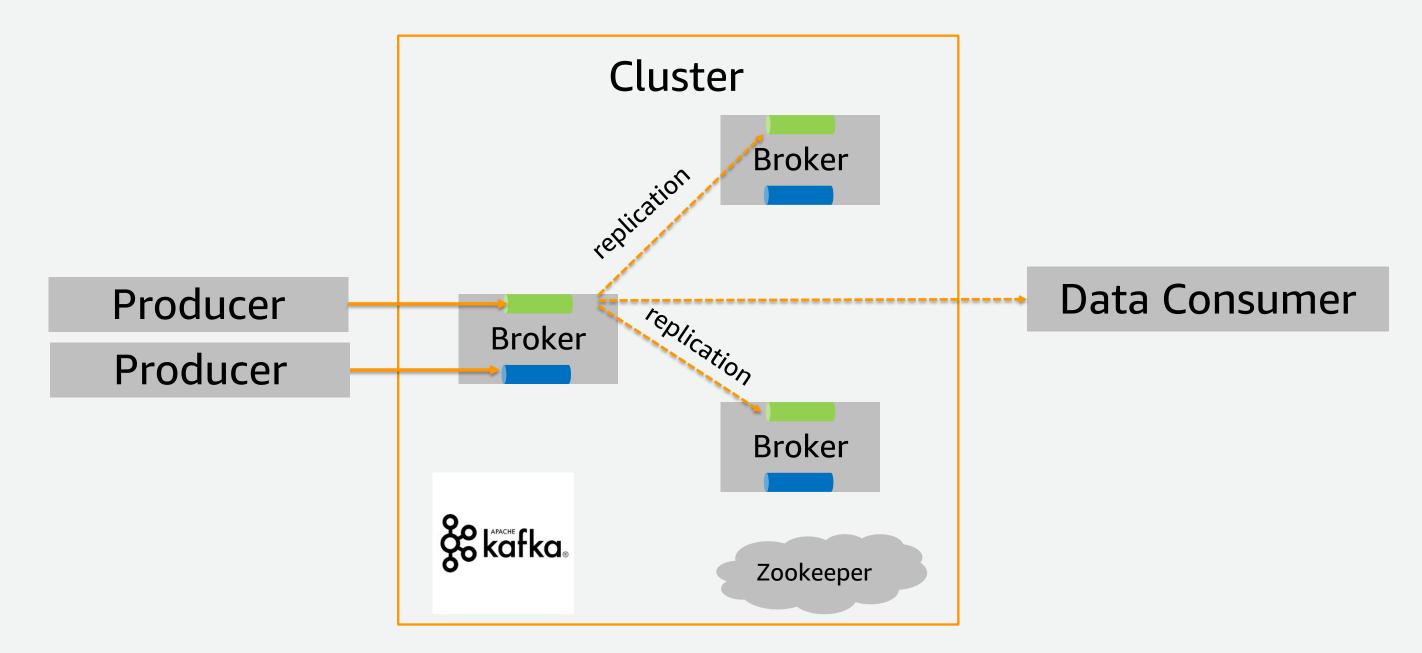


$$f(kafka_{usage}) = \sum_{n=1}^{\infty} (SRE)$$





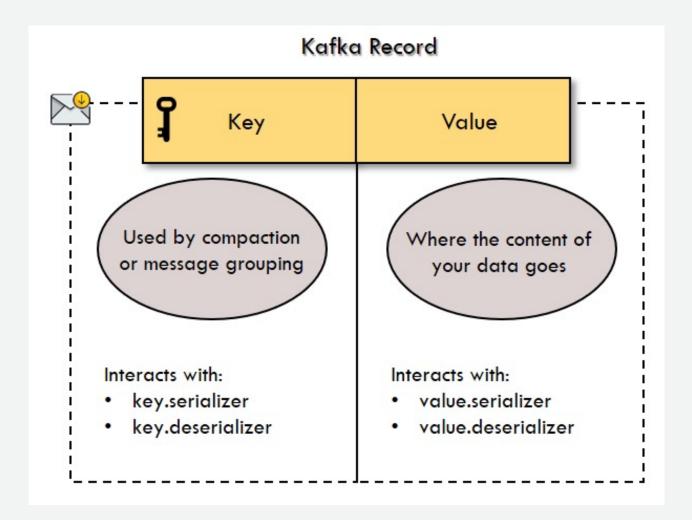
Apache Kafka Anatomy 101





Kafka Record

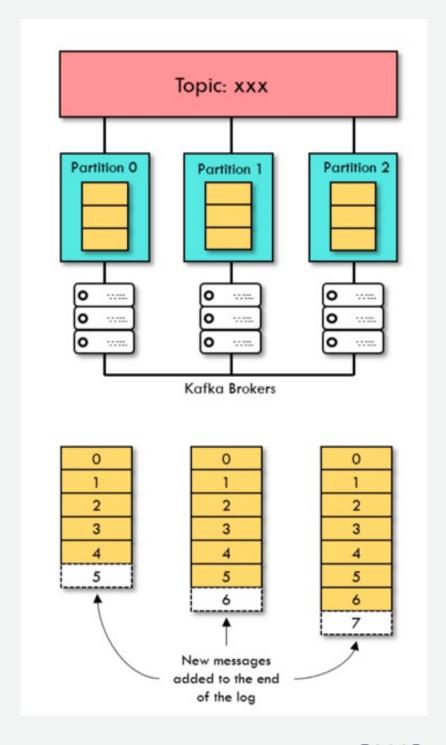
- Every message publish to Kafka called "Record"
- Record contain two parts:
 - Key: Used by compaction or for message grouping
 - Value: The content of data goes





Topics and partitions

- Topics: a particular stream of data
 - Similar to a table in a database (without all the constraints)
 - You can have as many topics as you want
 - A topic is identified by its name
- Topics are split in partitions
 - Each partition is ordered
 - Each message within a partition gets an incremental id, called offset





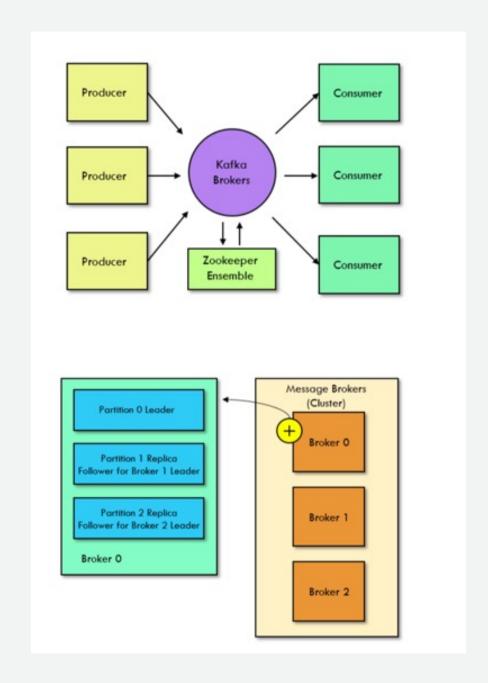
Topics and partitions

- Offset only have a meaning for a specific partition.
 - E.g. offset 3 in partition doesn't represent the same data as offset 3 in another partition
- Order is guaranteed only within a partition (not across partitions)
- Data is kept only for a limited time (default is one weeks)
- Once the data is written to a partition, it can't be changed (immutability)
- Data is assigned randomly to a partition unless a key is provided (more on this later)
- You can have as many partitions per topics as you want)



Brokers

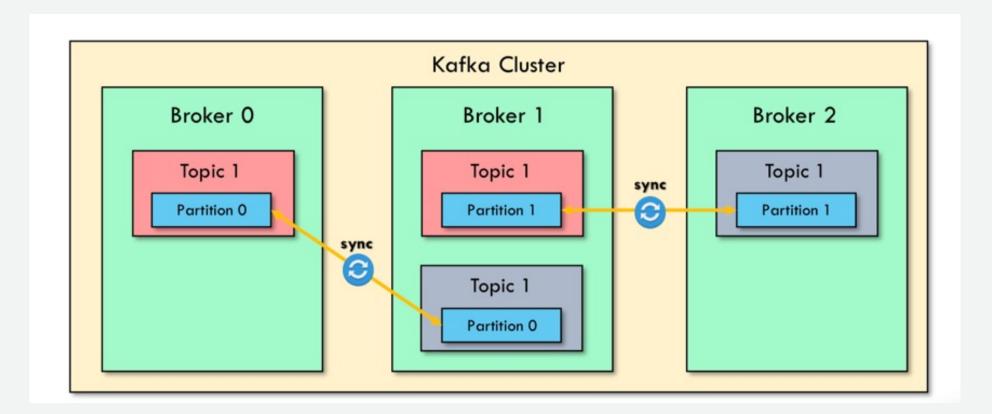
- A Kafka cluster is composed of multiple brokers (servers)
- Each broker is identified with its ID (integer)
- Each broker contains certain topic partitions
- After connecting to any broker (called a bootstrap broker), you will be connected to the entire cluster
- A good number to get started is 3 brokers, but some big clusters have over 100 brokers





Topic replication factor

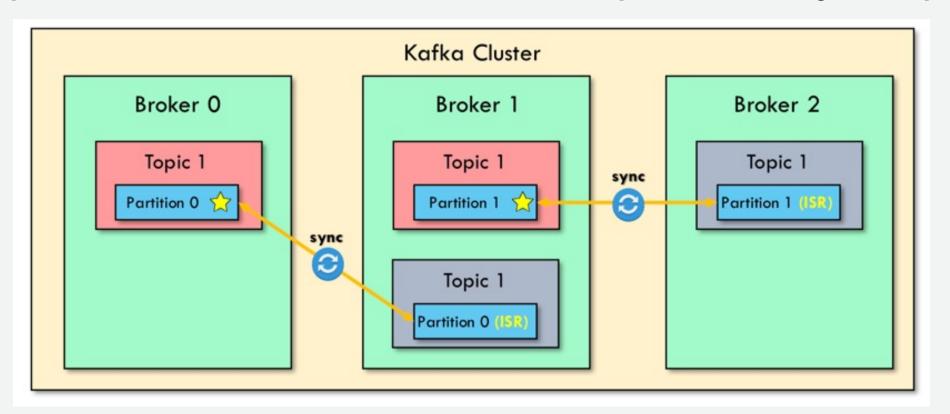
- Topics should have a replication factor > 1 (usually between 2 and 3)
- This way if a broker is down, another broker can serve the data
- Example: 1 topic with 2 partitions and replication factor of 2





Concept of Leader for a partition

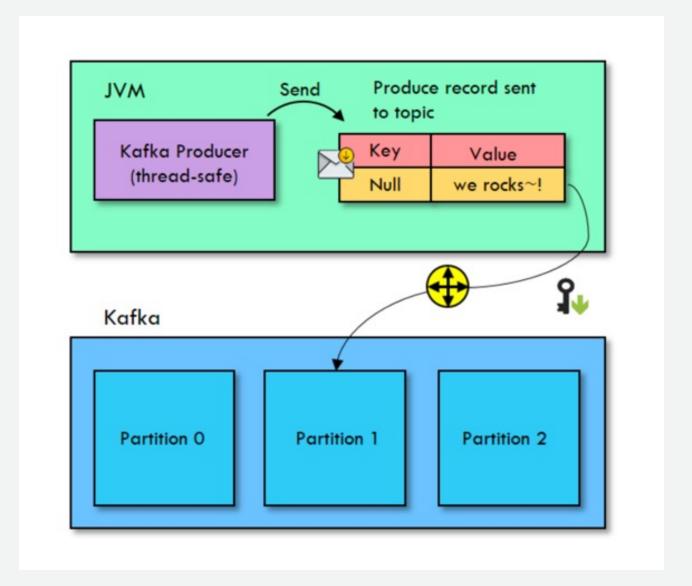
- At any time only 1 broker can be a leader for a given partition
- Only that leader can retrieve and serve data for a partition
- The other brokers will synchronize the data
- There each partition has: one leader, and multiple ISR (in-sync replica)





Producers

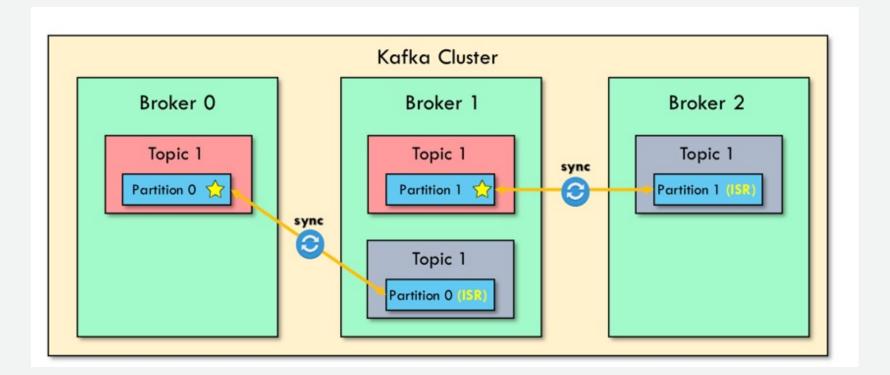
- Producers write data to topics.
- They only have to specify the topic name and one broker to connect to, and Kafka will automatically take care of routing the data to the right brokers.





Producers: Broker acknowledgement

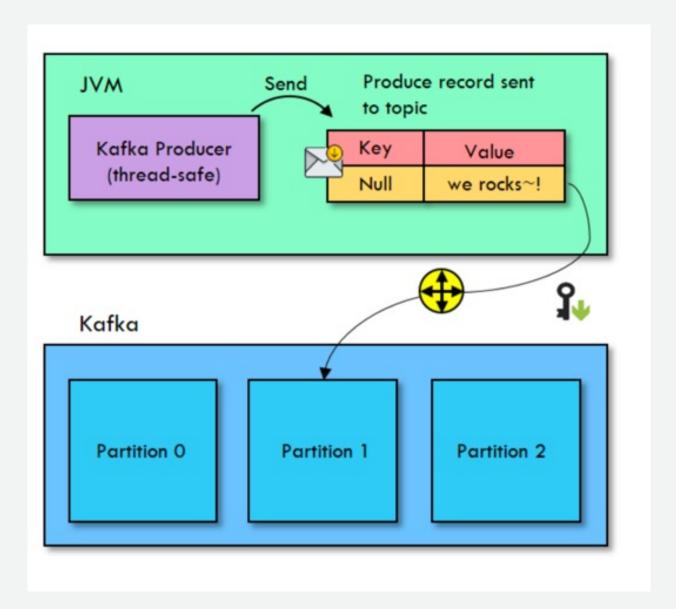
- Producer can choose to receive acknowledgement of data writes:
- acks = 0 : Producer won't wait for acknowledgement (possible data loss)
- acks = 1 : Producer will wait for leader acknowledgement (limited data loss)
- acks = all (-1): Leader + replicas acknowledgment (no data loss)





Producers: Message keys

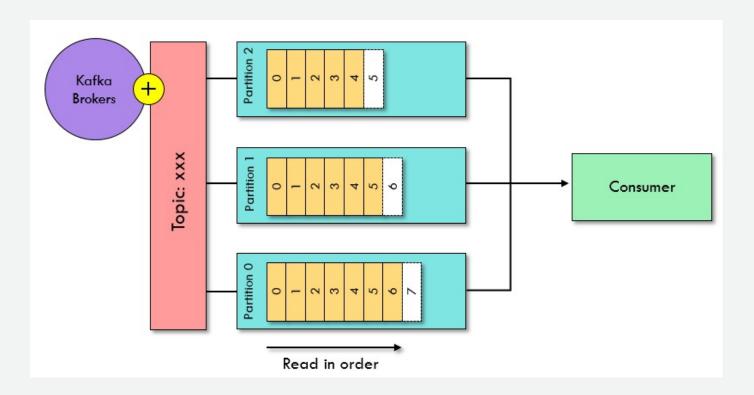
- Producers can choose to send a key with the message
- If a key is sent, then the producer has the guarantee that all messages for that key will always go to the same partition
- This enables to guarantee ordering for a specific key





Consumers

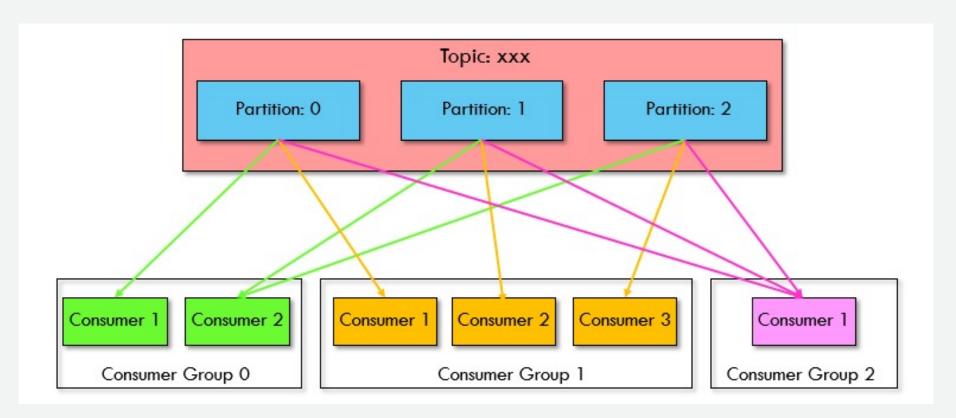
- Consumers read data from a topic
- They only have to specify the topic name and one broker to connect to, and Kafka will automatically take care of pulling the data from the right brokers
- Data is read in order for each partitions





Consumer Groups

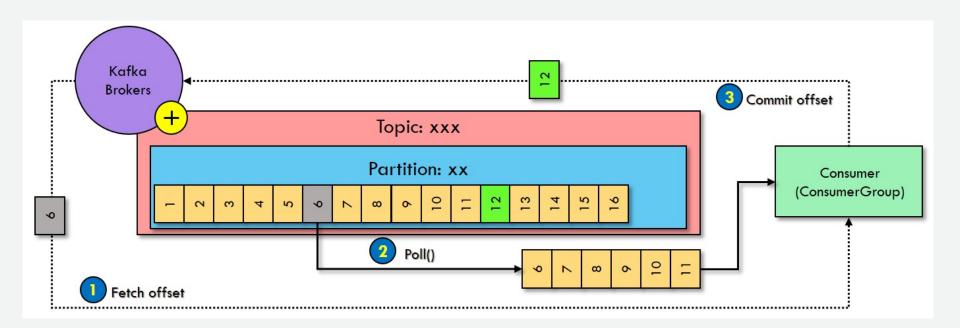
- Consumers read data in consumer groups
- Each consumer within a group reads from exclusive partitions
- You should control consumers instances less or equal than partitions (otherwise some will be inactive)





Consumer Offsets

- Kafka stores the offsets at which a consumer group has been reading
- The offsets commit live in a Kafka topic named "__consumer_offsets"
 - Key = [group, topic, partition], Value=offset
- When a consumer has processed data received from Kafka, it should be committing the offsets
- If a consumer process dies, it will be able to read back from where it left off





Kafka Guarantees

- Messages are appended to a topic-partition in the order they are sent
- Consumers read messages in the order stored in a topic-partition
- With a replication factor of N, producers and consumers can tolerate up to N-1 brokers being down
- This is why a replication factor of 3 is a good idea:
 - Allows for one broker to be taken down for maintenance
 - Allows for another broker to be taken down unexpectedly
- As long as the number of partitions remains constant for a topic (no new partitions), the same key will always go to the same partition



Log Cleanup Policies



- Many Kafka clusters make data expire, according to a policy
- That concept is called "log cleanup".
 - Policy 1: log.cleanup.policy=delete (Kafka default for all user topics)
 - Delete based on age of data (default is a week)
 - Delete based on max size of log (default is -1 == infinite)
 - Policy 2: log.cleanup.policy=compact (Kafka default for topic __consume_offsets)
 - Delete based on keys of your messages
 - Will delete old duplicate keys after the active segment is committed
 - Infinite time and space retention



Log Cleanup Policy: log.cleanup.policy=delete

log.retention.hours:

- Number of hours to keep data for (default is 168 one week)
- Higher number means more disk space
- Lower number means that less data is retained (your consumers may need to replay more data than less)

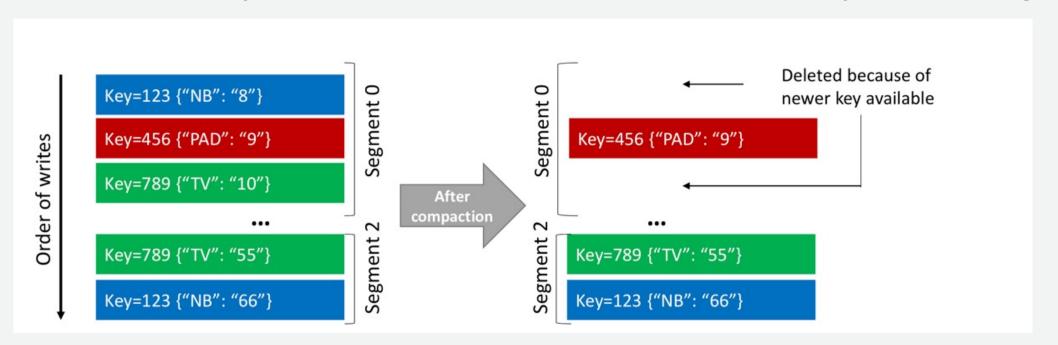
log.retention.bytes:

- Max size in Bytes for each partition (default is -1 == infinite)
- Useful to keep the size of a log under a threshold



Log Cleanup Policy: log.cleanup.policy=compact

- Log compaction ensures that your log contains at least the last know value for a specific key within a partition
- Very useful if we just require a SNAPSHOT instead of full history (such as for a data table in a database)
- The idea is that we only keep the latest "update" for a key in our log





Other advanced configurations

- max.messages.bytes (default is 1MB): if your messages get bigger than 1MB, increase this parameter on the topic and your consumers buffer
- min.isync.replicas (default is 1): if using acks=all, specify how many brokers need to acknowledge the write
- unclean.leader.election (danger zone! default false): if set to true, it will allow replicas which are not in sync to become leader as a last resort if all ISRs are offline. This can lead to data loss. If set to false, the topic will go offline until the ISRs come back up



Further thinking

- Topics are made of partitions, and the partitions are made of ...??
- How kafka knows where to find data in a constant time or offset?
- Producer Synchronous and Asynchronous send.
- What is the consumer group rebalance? Describe the mechanism of rebalance process.







Comparing Amazon MSK with Amazon Kinesis Data Streams



Comparing Amazon Kinesis Data Streams to MSK



Amazon Kinesis Data Streams

- Streams and shards
- AWS API experience
- Throughput provisioning model
- Seamless scaling
- Typically lower costs
- Deep AWS integrations



Amazon MSK

- Topics and partitions
- Open-source compatibility
- Strong third-party tooling
- Cluster provisioning model
- Apache Kafka scaling isn't seamless to clients
- Raw performance

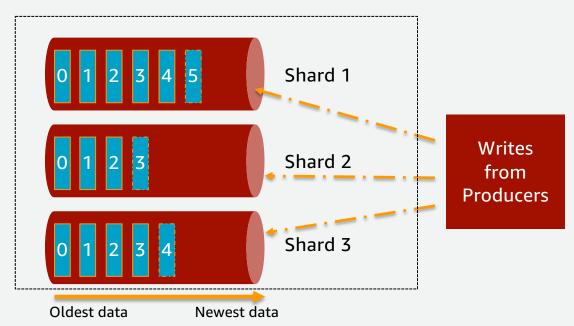


Comparing Amazon Kinesis Data Streams to MSK



Amazon Kinesis Data Streams

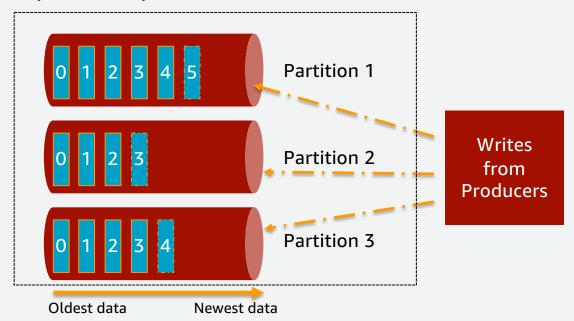
Stream with 3 shards





Amazon MSK

Topic with 3 partitions





Online-testing

• https://github.com/zhwenhao-amzn/ELS-MSK-Kinesis/tree/main/streaming-quiz

Mail the result to zhwenhao@amazon.com



Thank you!

