



Achieving prod-ready streaming pipelines with Kafka Streams



KSUG meetup • 28.05.19

Agenda

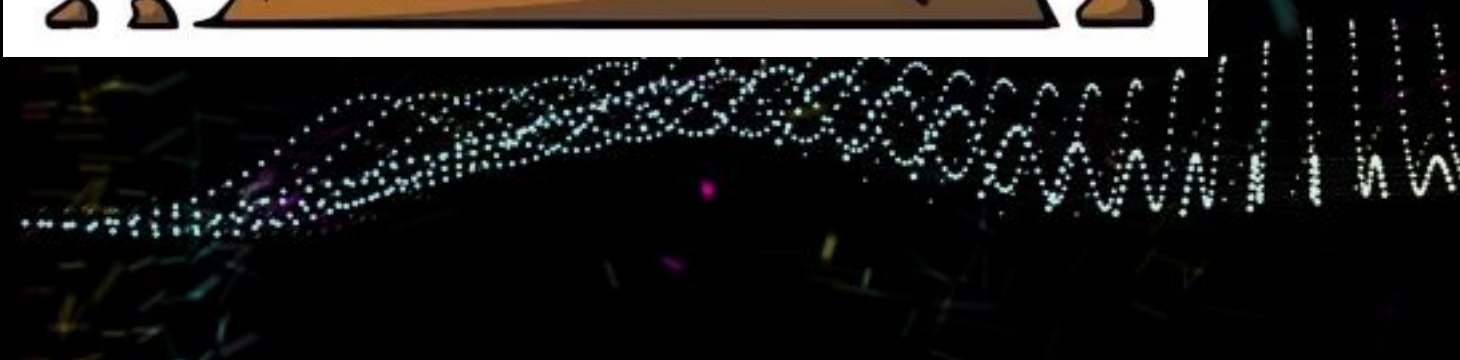
- Kafka introduction
- Kafka Streams API
- Live coding



Disclaimer



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In case of emergency



Let's talk about



Apache Kafka

- Initially conceived as a messaging queue
- Distributed commit log
- Distributed event-streaming platform



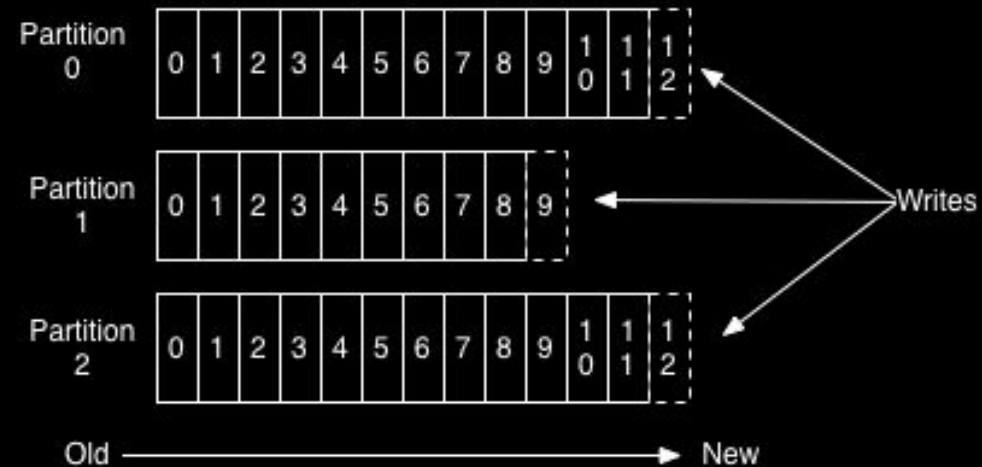
So... what is it?

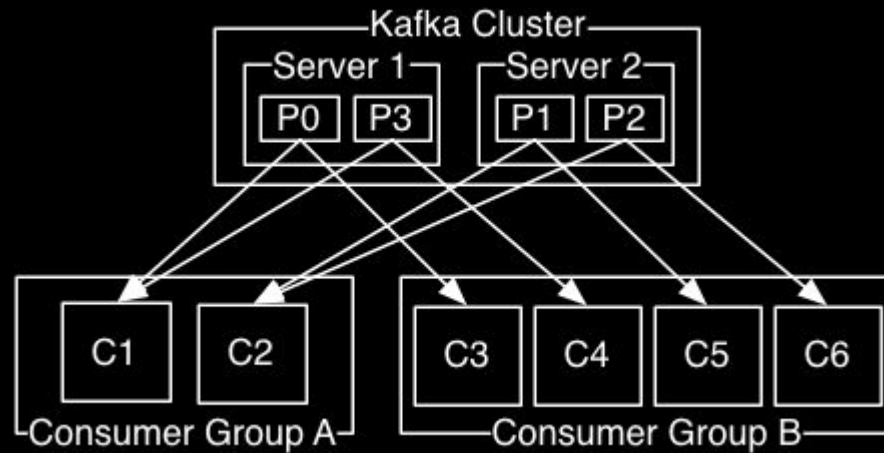
- A messaging system (queueing, pub-sub)
- A storage system (high-perf, low-latency, replication)
- A streaming platform



Kafka basics

Anatomy of a Topic





- Any partition is consumed by one and only one member of the same consumer group
- Any partition is consumed by all of the present consumer groups

Implications


- Every record published to a topic is delivered to all of the subscribed consumer groups.
- Each consumer **in a single** consumer group => effective **load balancing** over all consumers.
- Each consumer **in its own** consumer group => effective **broadcast** over all consumers.



Let's talk about **Kafka Streams API**



Kafka Streams

- Lightweight Java library
 - Enables transforming and enriching data from the Kafka topics in a streaming fashion
 - No separate cluster required
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- Code-centred (Scala API)
- Elastic, highly scalable, fault-tolerant
- Deploy to containers, VMs, bare metal, cloud
- Equally viable for small, medium, & large use cases
- Comes with all the Kafka goodies (partition-scalability , Kafka Security, exactly-once semantics).

Enter the DSL

Transformations:

- Stateless (`filter`, `map`, `flatMap`, `foreach`, `branch`, etc...)
- Stateful (`count`, `reduce`, `aggregate`, `join`, `windowed ops`)



Stateless operators

Transform 1 message into 0 or more messages:

- 1:1 - map, branch
- 1:[0,1] - filter
- 1:[0,...,n] - flatMap



An example (Word Count)

```
builder.stream[String, String]("streams-file-input")  
  .flatMapValues { value => value.toLowerCase().split(" ") }  
  .map { (_, value) => (value, value) }  
  .groupByKey  
  .count("Counts")  
  .to("streams-wordcount-output")
```

```
val streams = new KafkaStreams(builder, props)  
streams.start()
```



Repartitioning

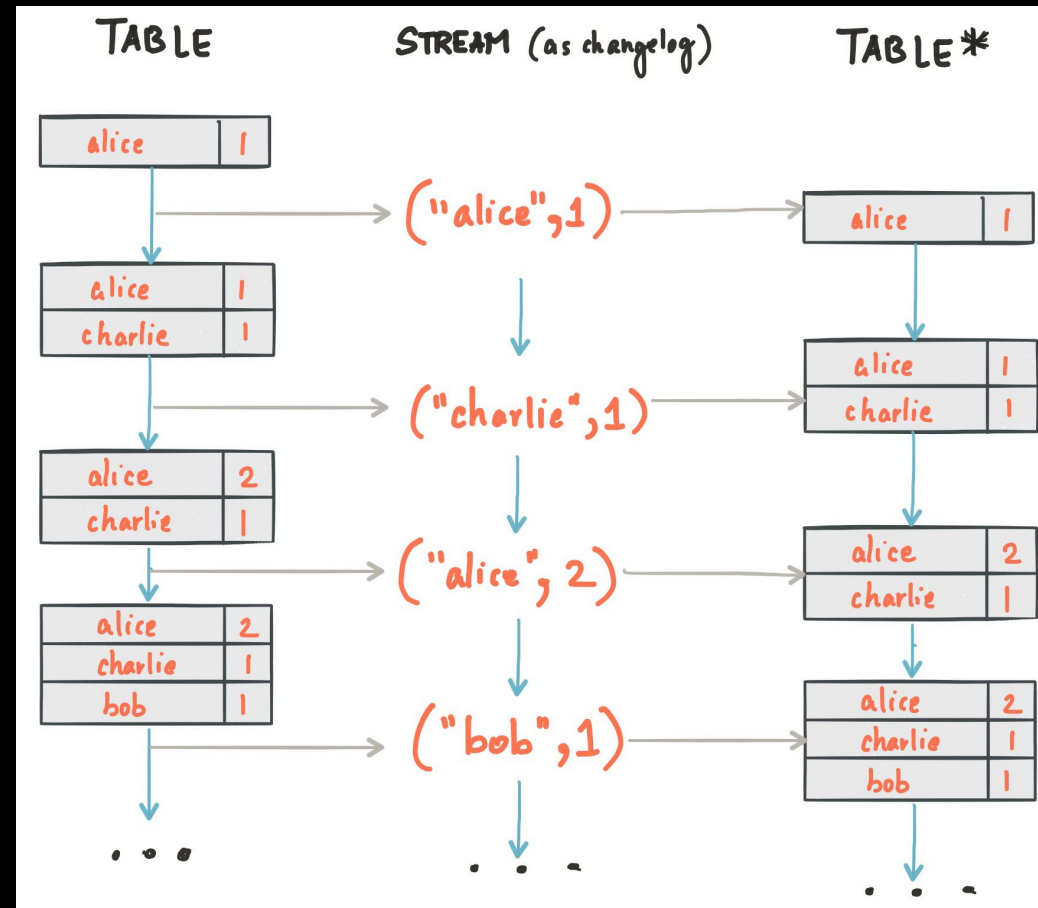
- Some of the operators may change the key of the transformed message(s). That may cause the creation of an intermediate repartition-topic.
- Operators indicating no key-change:
 - `filter`, `mapValues`, `flatMapValues`

Duality of Streams and Tables

- Stream processing usually needs both streams and databases
- A stream may be seen as a changelog of a table
- Table may be seen as a snapshot of the latest value for each key in a stream.



Duality of Streams and Tables



KStreams and KTables

- **KStream** is an abstraction of a **record** stream
 - log compaction may remove important data
- **KTable** is an abstraction of a **changelog** stream
 - only latest value is significant
 - log compaction highly encouraged
 - table-lookup via joins



Joins

Problem:

Process records only with the specific key.

What if the key is dynamically defined?

What if the key is defined on another topic?

Example:

Stream (main) record structure:
(key: *PersonID*, value: *ImportantData*)

Other (configuration) topic:
(key: *PersonID*, value: *Boolean*)



Solution:

- KStream-to-KTable join (inner join or left join)

Different types of joins:

- KTable-to-KTable (inner join, left join, outer join)
- KStream-to-KStream (same, but only windowed)

Aggregation

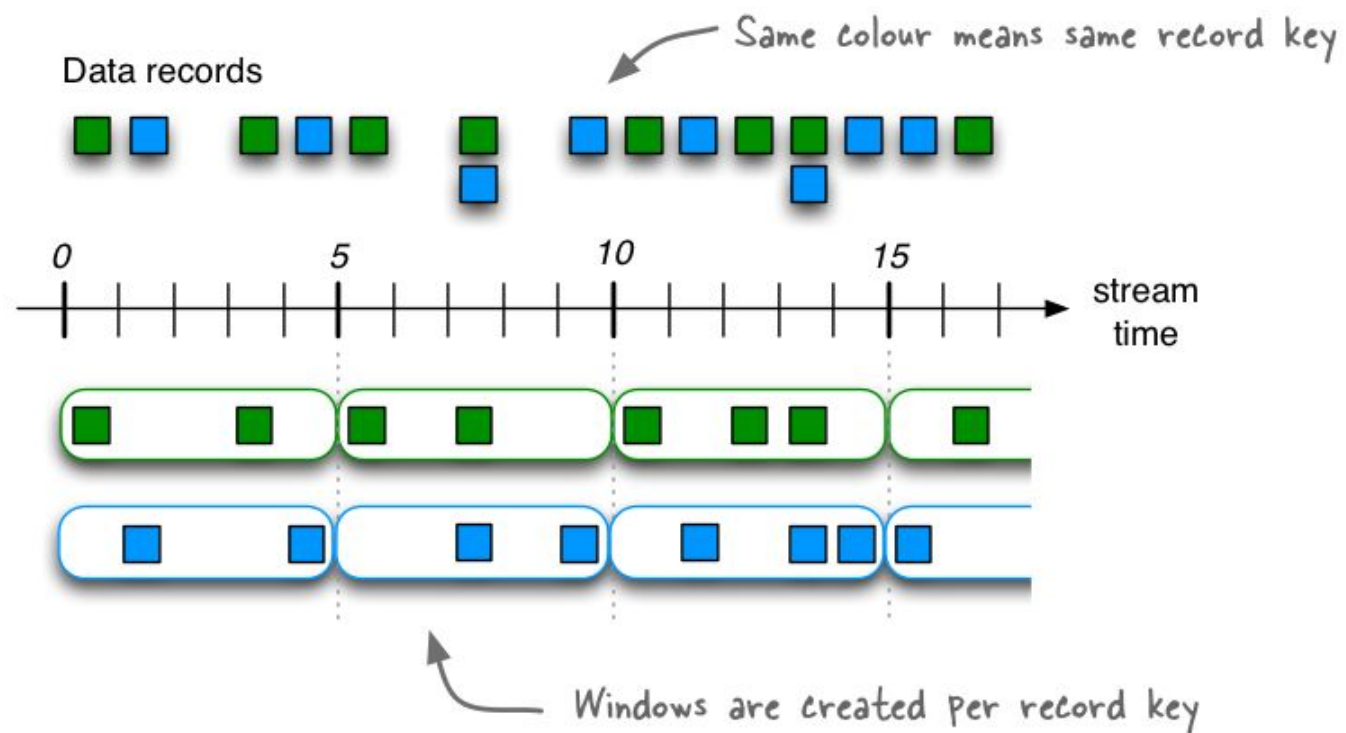
- Transforms N (grouped) messages into 1 message
- Key-based operation, records need to be grouped
 - only grouping by key causes no repartitioning
- May be both windowed and non-windowed
- 3 types of aggregate functions:
 - count, reduce, aggregate

Windowing

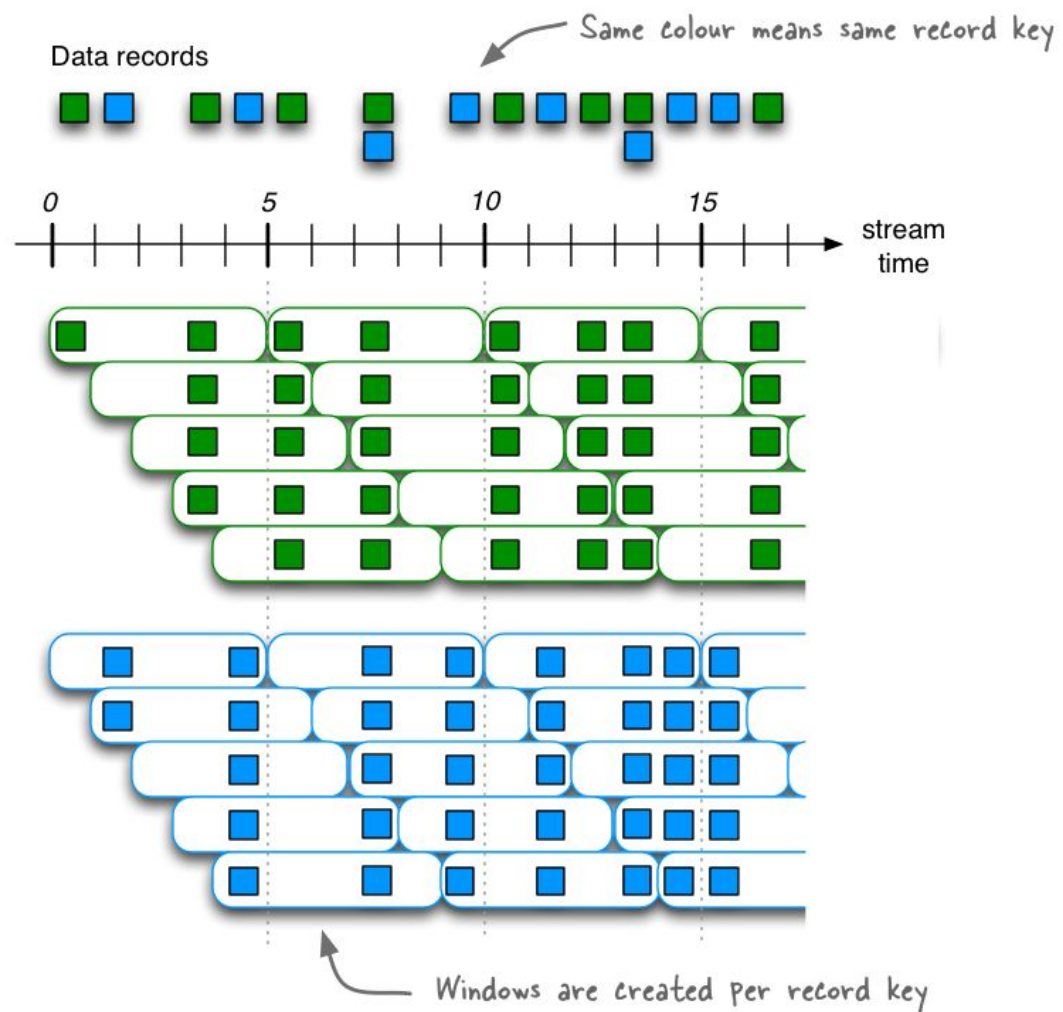
- Different types of windows:
 - *tumbling window, hopping window* (windowed aggregates)
 - *sliding window* (windowed joins)
 - *session window* (session aggregates, e.g. user activities)



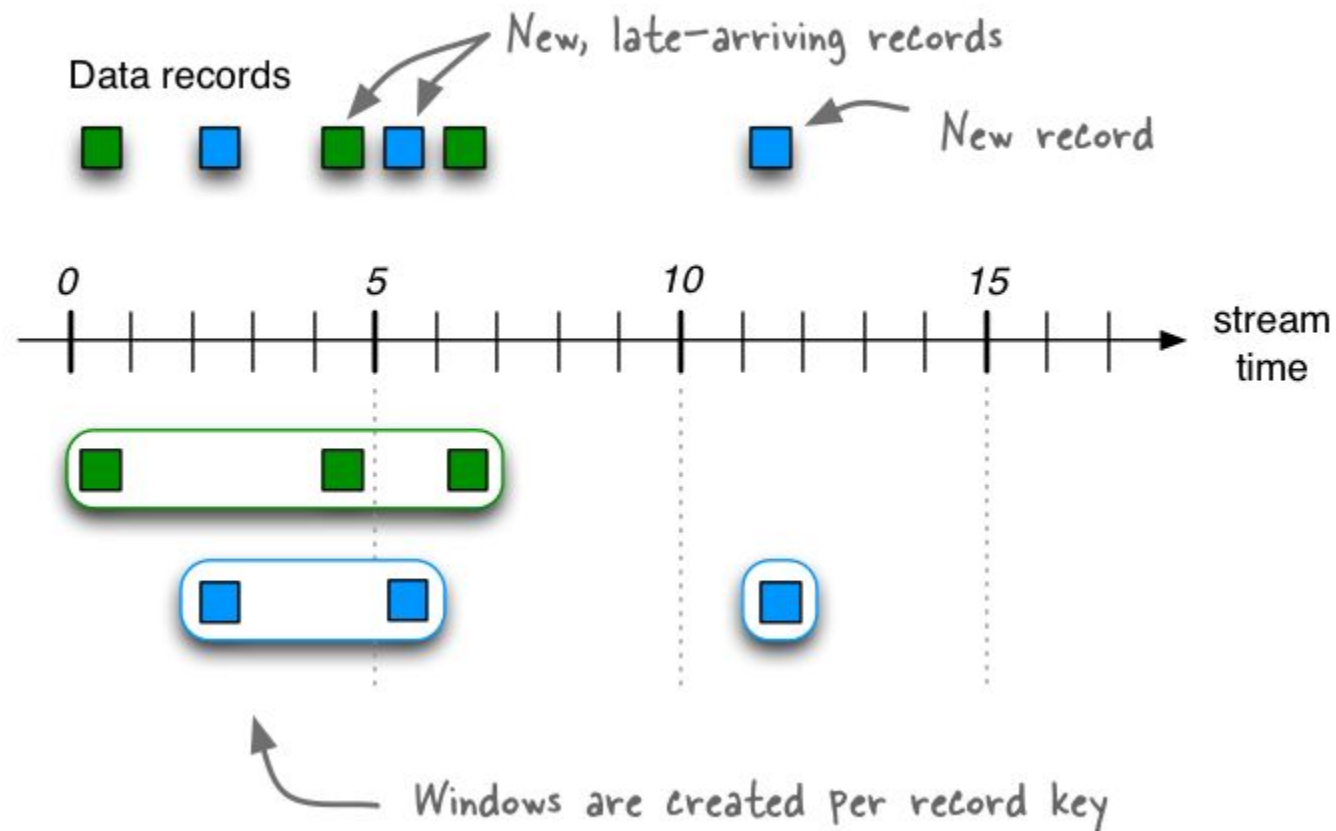
A 5-min Tumbling Window



A 5-min Hopping Window with a 1-min "hop"



A Session Window with a 5-min inactivity gap



Time semantics

3 types of time in terms of stream processing:

- Event time (real event timestamp at “source”)
- Ingestion-time (when event was produced to Kafka)
- Processing-time (when event was processed)



Windowed time semantics

3 types of window-time configuration:

- **Window duration** (size of the window)
- **Window interval** (frequency of subsequent windows)
- **Grace period** (actual size of the window, when late records are still accepted)



When exactly are the windowing results produced?

- After every update? (yes, when the cache is disabled)
- Whenever the earliest of *commit.interval.ms* or *cache.max.bytes.buffering* hits.



Input records



Output records



(a) No cache - commit interval has no impact on forwarding.



(b) With cache



Suppression (Kafka 2.1.0)

- Final windowing results produced after the grace period (finally).

<https://www.confluent.io/blog/kafka-streams-take-on-watermarks-and-triggers>



```
events

    .groupByKey()

    .windowedBy(

        TimeWindows.of(Duration.ofMinutes(2)).withGrace(Duration.ofMinutes(2))

    )

    .count(Materialized.as("count-metric"))

    .suppress(Suppressed.untilWindowClose(BufferConfig.unbounded()))
```

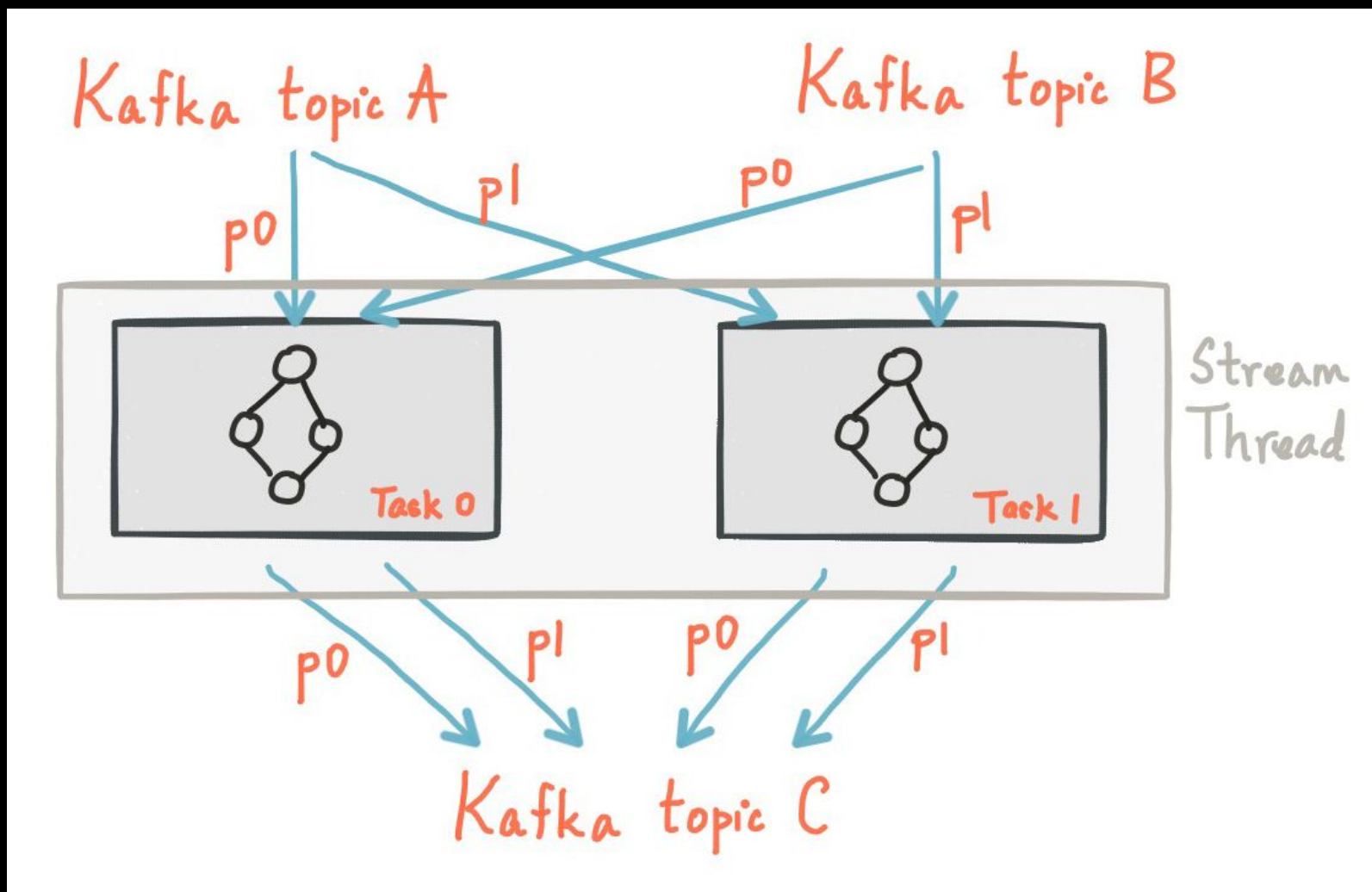
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Threading model

- **Stream partition** is a totally ordered sequence of data records and maps to a Kafka topic partition.
- **Stream task** is a logical unit of parallelism model - an application's topology is scaled by breaking it into multiple stream tasks.
- **Streaming thread** is just a thread and may execute many stream tasks.





Stores

- Used to store and query data (for stateful operations)
- Automatically recoverable from changelog topics
- Implementations to choose:
 - RocksDB (by default), in-memory hashmap, or anything you implement



Testing

Two options:

- [TopologyTestDriver.java](#) (comes with Kafka dependencies, constantly updated, rich API, Java :<)
- [MockedStreams.scala](#) (really lightweight Scala wrapper for the TopologyTestDriver)


```
import com.madewithtea.mockedstreams.MockedStreams

val input = Seq(("x", "v1"), ("y", "v2"))
val exp = Seq(("x", "V1"), ("y", "V2"))
val strings = Serdes.String()

MockedStreams()

  .topology { builder => builder.stream(...) [...] } // Scala DSL
  .input("topic-in", strings, strings, input)
  .output("topic-out", strings, strings, exp.size) shouldEqual exp
```

Live coding / demo time

DEMO



More?

Repo: <https://github.com/mowczare/kafka-streams-scala>

Blog: <https://www.avsystem.com/blog/large-scale-data-monitoring-with-kafka-streams/>

Careers: <https://www.workwiththebest.pl>



Thanks for the attention!

