



## Introducing Kafka Streams

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Michael Noll

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 miguno

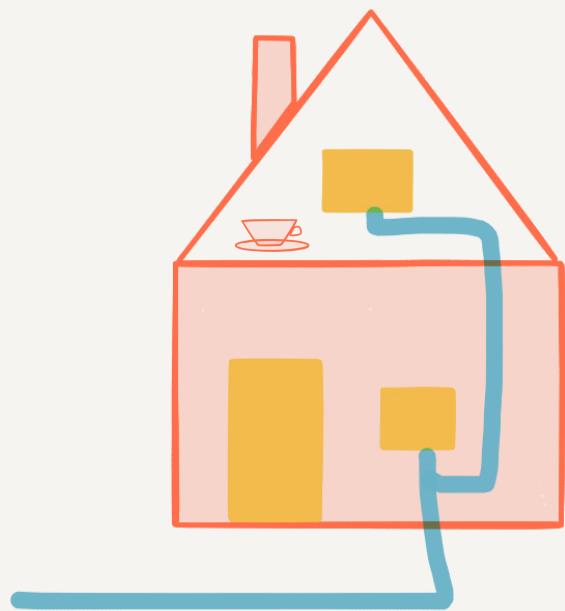
Berlin Buzzwords, June 06, 2016

COFFEE = 





DATA = WATER  
KAFKA = PIPES





KAFKA STREAMS  
= COFFEE MACHINE

HAPPY  
USER  
↓

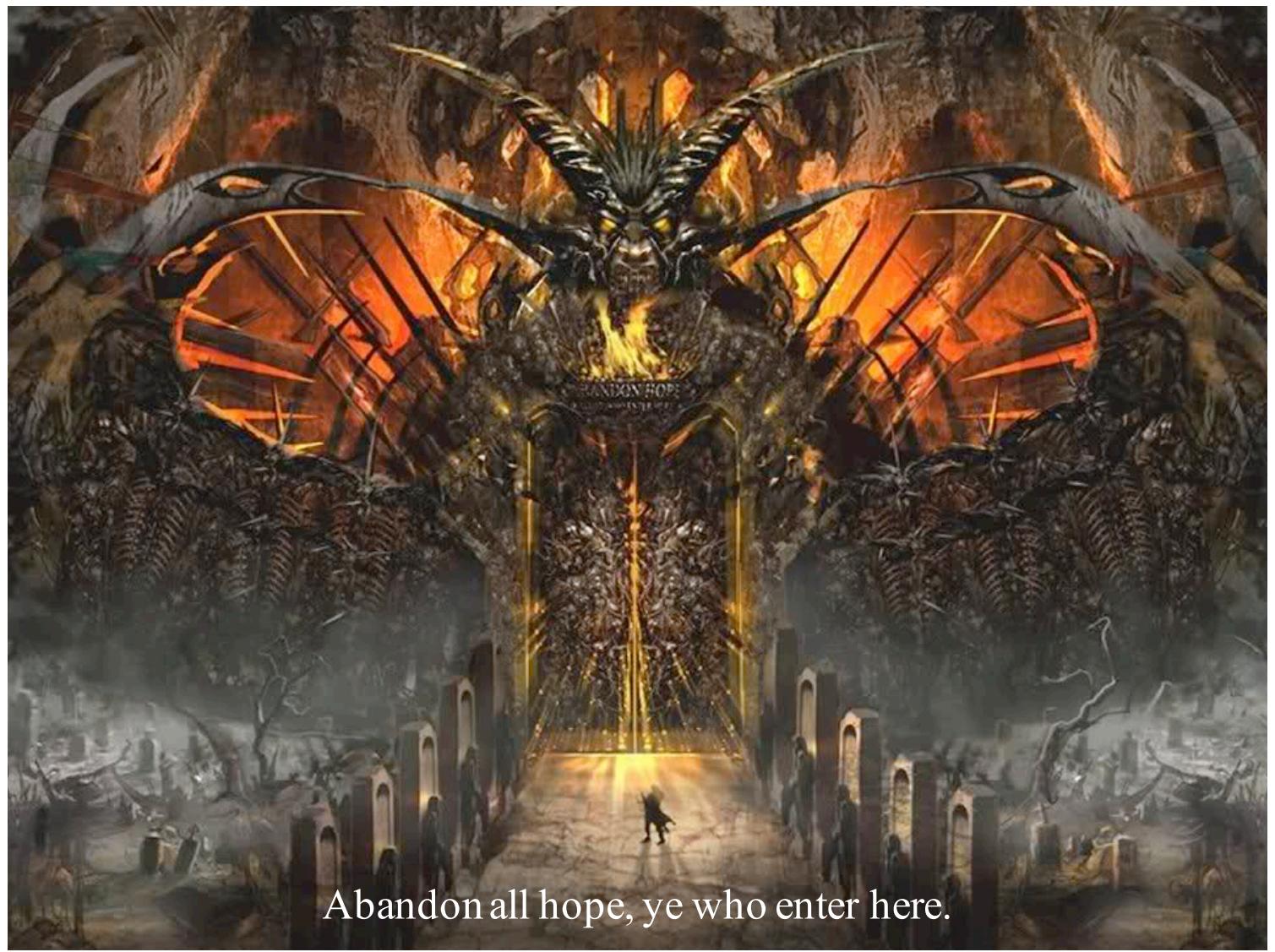


# Stream processing in Real Life™

...before Kafka Streams

...somewhat exaggerated

...but perhaps not that much



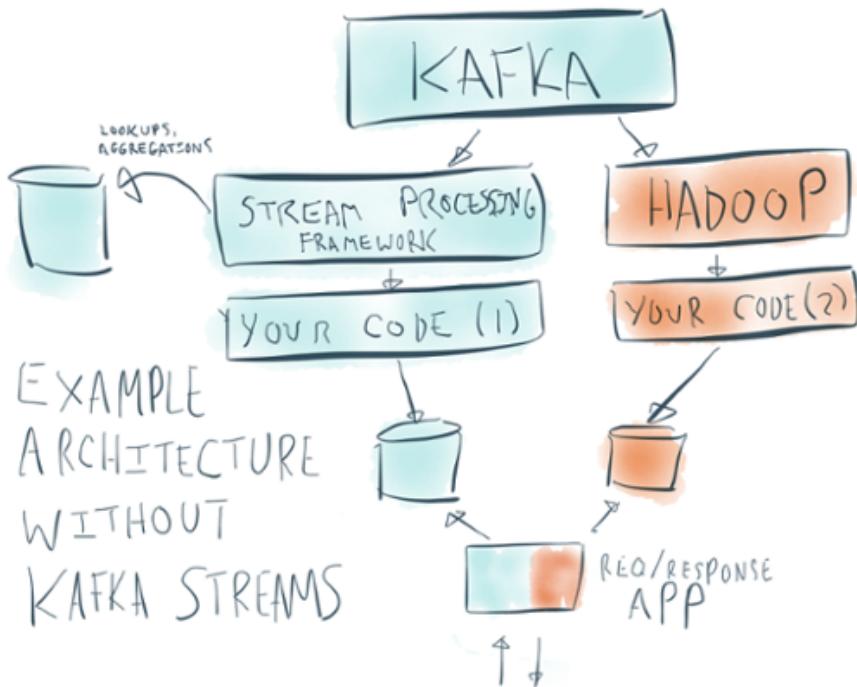
Abandon all hope, ye who enter here.

## How did this... (#machines == 1)

---

```
scala> val input = (1 to 6).toSeq  
  
// Stateless computation  
scala> val doubled = input.map(_ * 2)  
Vector(2, 4, 6, 8, 10, 12)  
  
// Stateful computation  
scala> val sumOfOdds = input.filter(_ % 2 != 0).reduceLeft(_ + _)  
res2: Int = 9
```

...turn into stuff like this (#machines > 1)





MAX(VALUE) && MIN(DISTRACTION)

MAKE COMPLEX THINGS

SIMPLE EASY FUN

"DEVELOPER EFFICIENCY"

## SCALING HUMANS

(OUR BRAIN CAPACITY DOES NOT  
DOUBLE EVERY 18 MONTHS 😞)



"Who is using Kafka?"

Taken at a session at ApacheCon: Big Data, Hungary, September 2015

# Kafka Streams

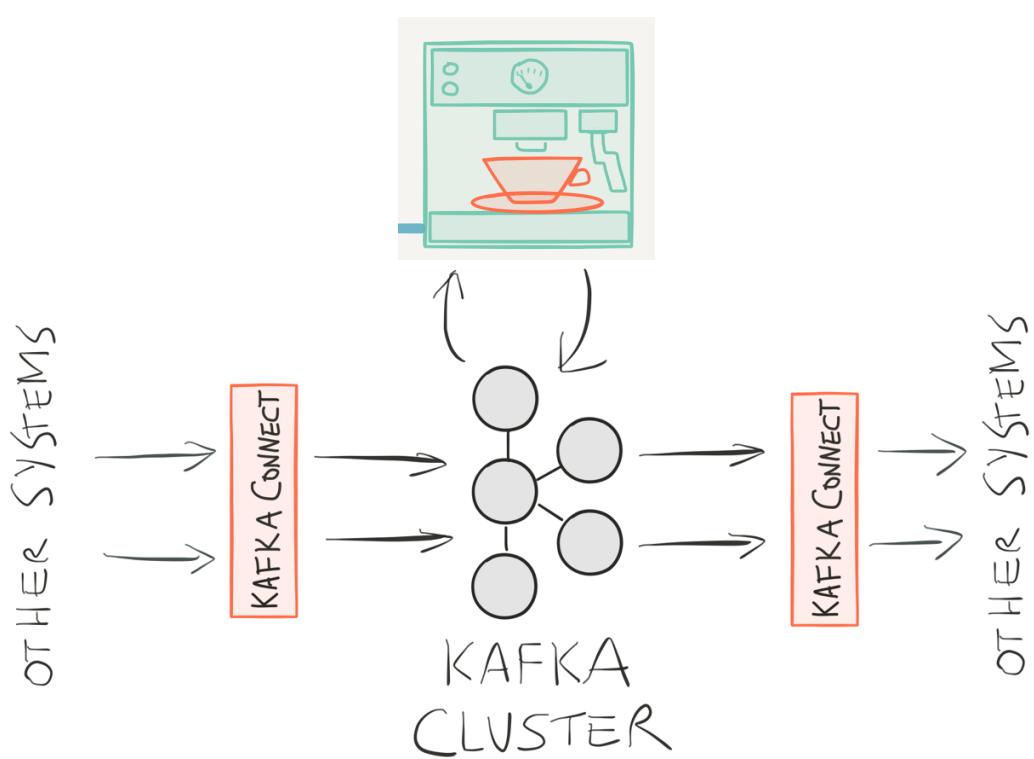
## stream processing made simple

# Kafka Streams

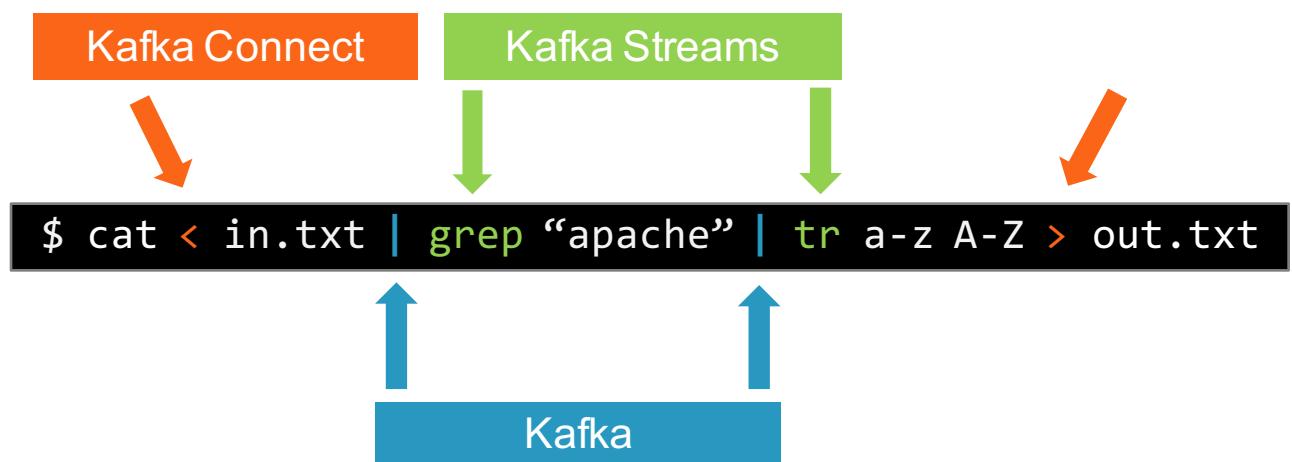
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- **Powerful yet easy-to use Java library**
- Part of open source Apache Kafka, introduced in v0.10, May 2016
- Source code: <https://github.com/apache/kafka/tree/trunk/streams>
- Build your own stream processing applications that are
  - highly scalable
  - fault-tolerant
  - distributed
  - stateful
  - able to handle late-arriving, out-of-order data
  - <more on this later>

# Kafka Streams



## What is Kafka Streams: Unix analogy



## What is Kafka Streams: Java analogy

---

1996

1 core

java.lang

---

2004

multi-core

java.util.concurrent

---

2016

multi-machine

java.distributed

org.apache.kafka.streams

# When to use Kafka Streams (as of Kafka 0.10)

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## Recommended use cases

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- **Application Development**
- “Fast Data” apps (small or big data)
- Reactive and stateful applications
- Linear streams
- Event-driven systems
- Continuous transformations
- Continuous queries
- Microservices

## Questionable use cases

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- **Data Science / Data Engineering**
- “Heavy lifting”
- Data mining
- Non-linear, branching streams (graphs)
- Machine learning, number crunching
- What you’d do in a data warehouse

## Alright, can you show me some code now? 😊

---

- API option 1: Kafka Streams DSL (declarative)

```
KStream<Integer, Integer> input = builder.stream("numbers-topic");

// Stateless computation
KStream<Integer, Integer> doubled = input.mapValues(v -> v * 2);

// Stateful computation
KTable<Integer, Integer> sumOfOdds = input
    .filter((k,v) -> v % 2 != 0)
    .selectKey((k, v) -> 1)
    .reduceByKey((v1, v2) -> v1 + v2, "sum-of-odds");
```

## Alright, can you show me some code now? 😊

- API option 2: low-level Processor API (imperative)

```
public PrintToConsoleProcessor implements Processor<K, V> {  
  
    @Override  
    public void init(ProcessorContext context) {  
        // No initialization needed in this case.  
    }  
  
    @Override  
    public void process(K key, V value) {  
        System.out.println("Received data record with " +  
            "key=" + key + ", value=" + value);  
    }  
  
    @Override  
    public void punctuate(long timestamp) {  
        // No periodic actions needed in this case.  
    }  
  
    @Override  
    public void close() {  
        // No shutdown logic needed in this case.  
    }  
}
```

The diagram illustrates the mapping of the Processor API methods to Kafka lifecycle events:

- Startup**: Corresponds to the `init` method.
- Process a record**: Corresponds to the `process` method.
- Periodic action**: Corresponds to the `punctuate` method.
- Shutdown**: Corresponds to the `close` method.

A photograph of a large iceberg floating in the ocean. The visible portion above the water's surface is small compared to the massive submerged portion below. An orange arrow points upwards from the submerged part to a white rectangular overlay containing the text.

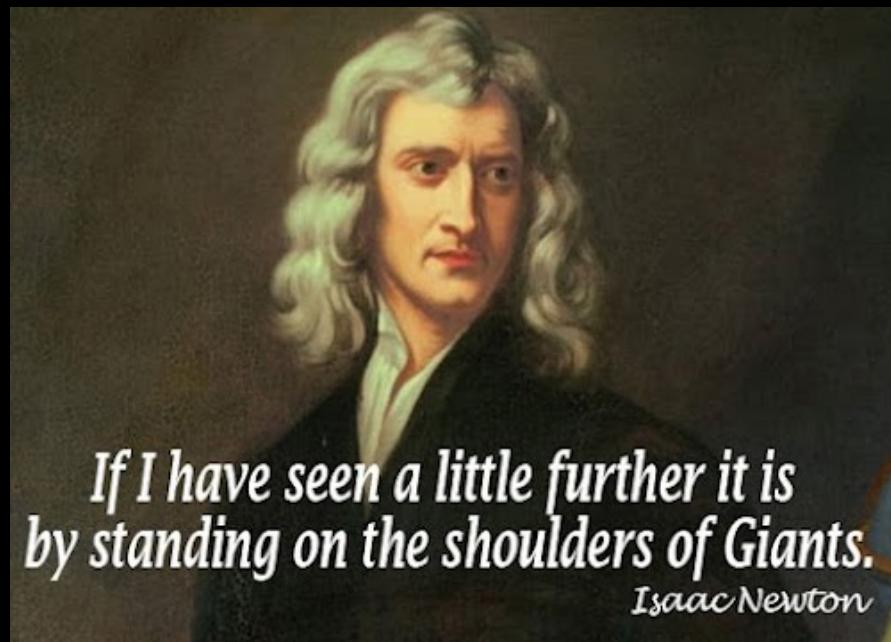
API, coding

MAKE COMPLEX THINGS  
SIMPLE EASY FUN

An orange arrow points downwards from the white rectangular overlay to the submerged portion of the iceberg.

Operations, debugging, ...

Kafka Streams outsources hard problems to Kafka



## How do I install Kafka Streams?

---

- There is and there should be no “install”.
- It’s a library. Add it to your app like any other library.

```
<dependency>
  <groupId>org.apache.kafka</groupId>
  <artifactId>kafka-streams</artifactId>
  <version>0.10.0.0</version>
</dependency>
```

**ONE DOES NOT SIMPLY PROCESS DATA  
WITHOUT A CLUSTER**

## Do I need to install a CLUSTER to run my apps?

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- No, you don't. Kafka Streams allows you to stay lean and lightweight.
- Unlearn bad habits: “do cool stuff with data != must have cluster”

Ok.



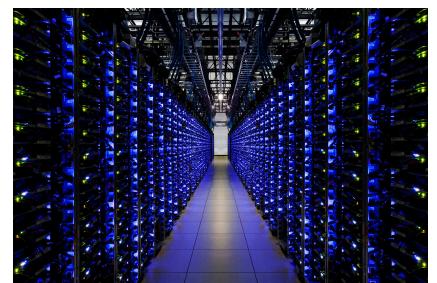
Ok.



Ok.



Ok.



# How do I package and deploy my apps? How do I ...?

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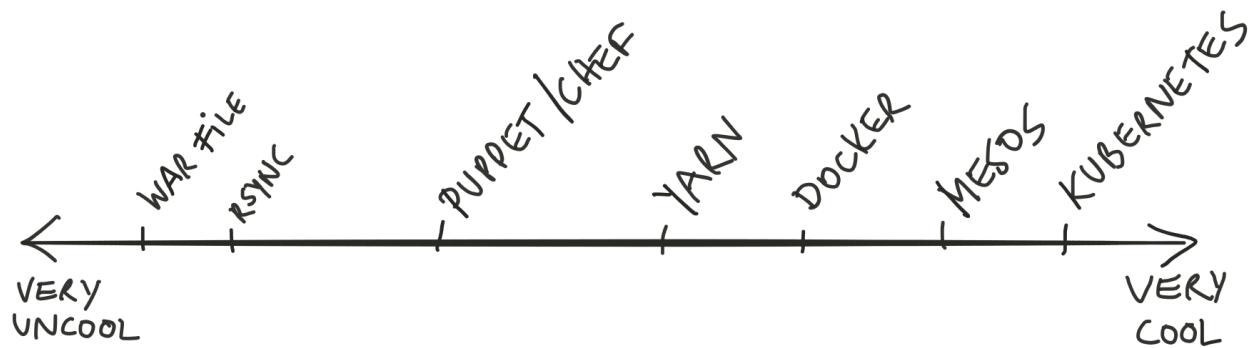
Introducing Kafka Streams, Michael G. Noll, Berlin Buzzwords, June 2016 | 28



imgflip.com

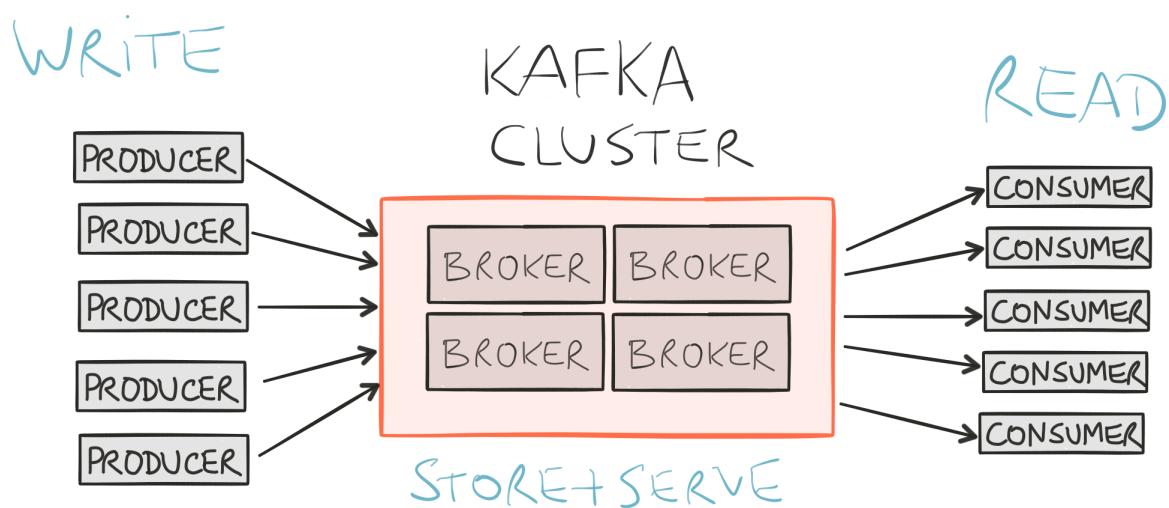
## How do I package and deploy my apps? How do I ...?

- Whatever works for you. Stick to what you/your company think is the best way.
  - Why? Because an app that uses Kafka Streams is...a normal Java app.
- Your Ops/SRE/InfoSec teams may finally start to **love** not hate you.

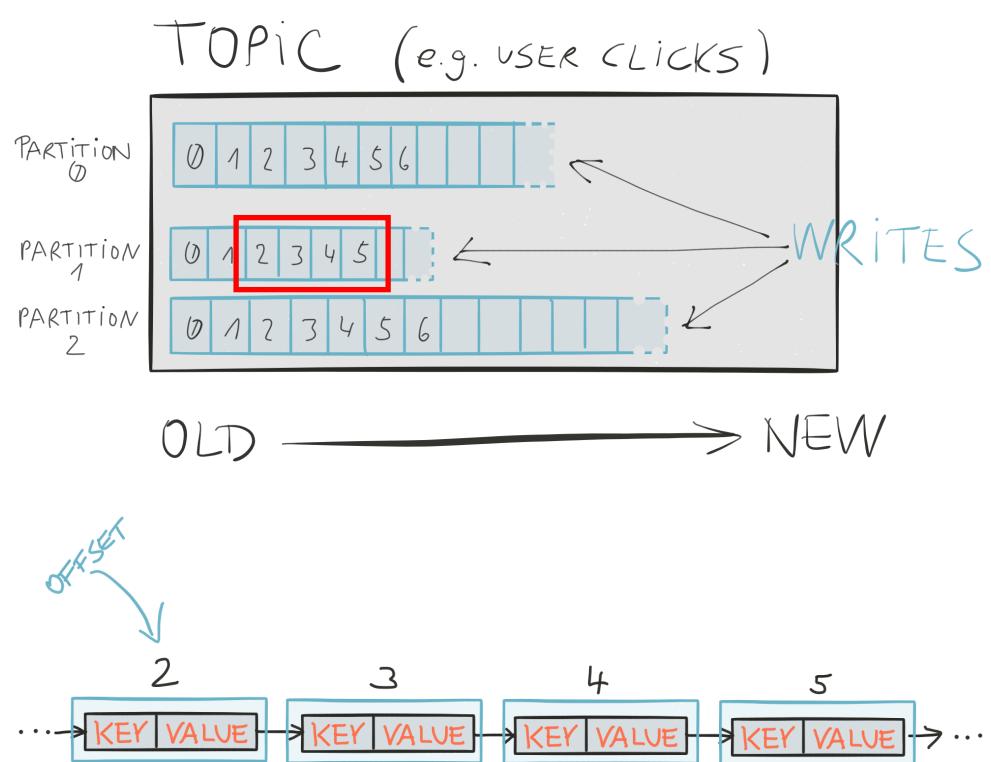


# Kafka concepts

## Kafka concepts



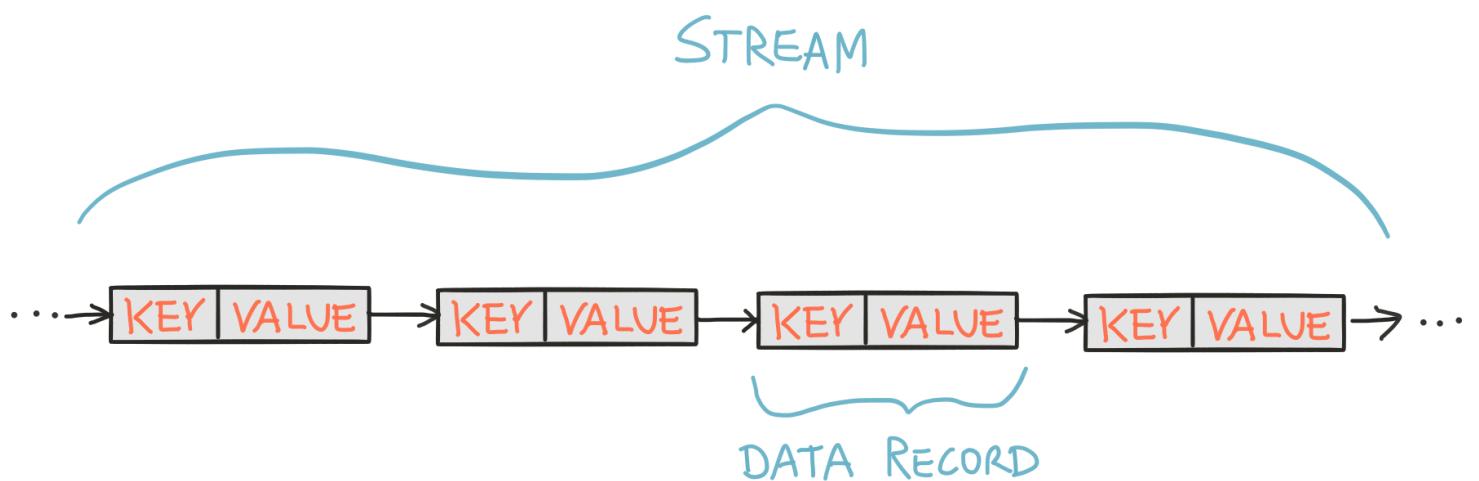
# Kafka concepts



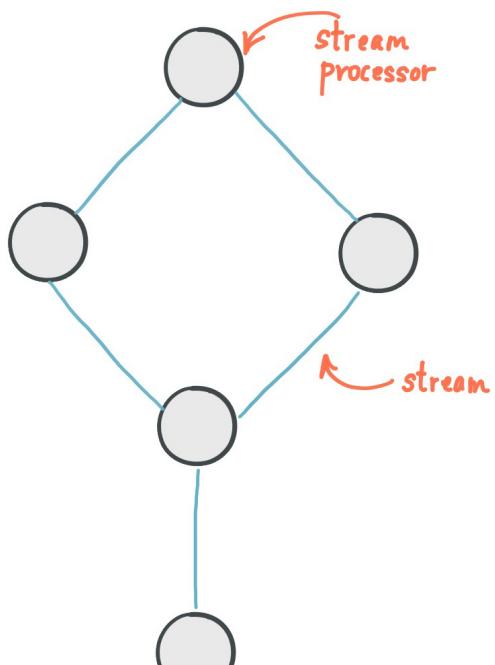
# Kafka Streams

## concepts

# Stream

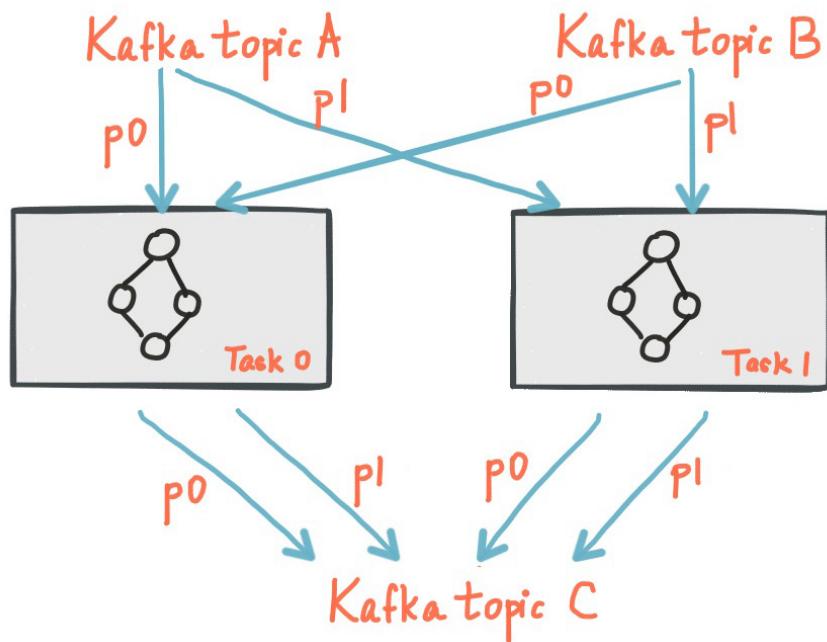


## Processor topology

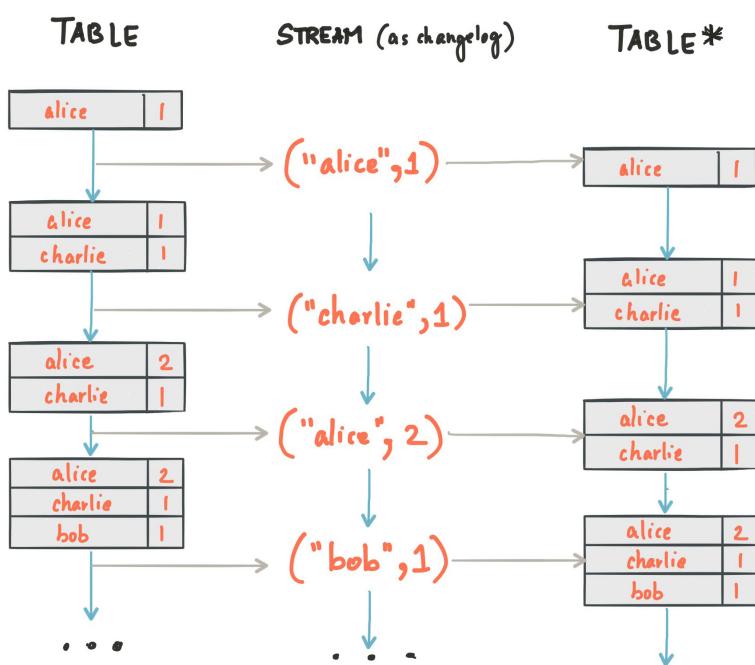


PROCESSOR TOPOLOGY

## Stream partitions and stream tasks

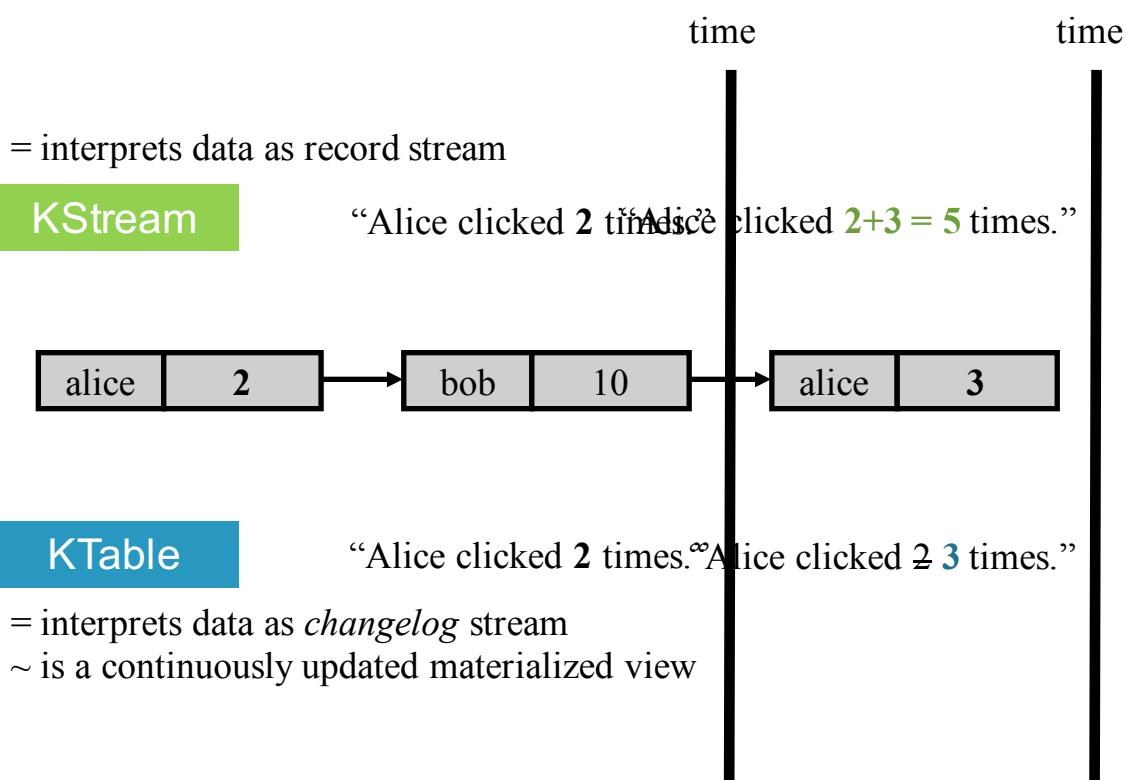


# Streams meet Tables

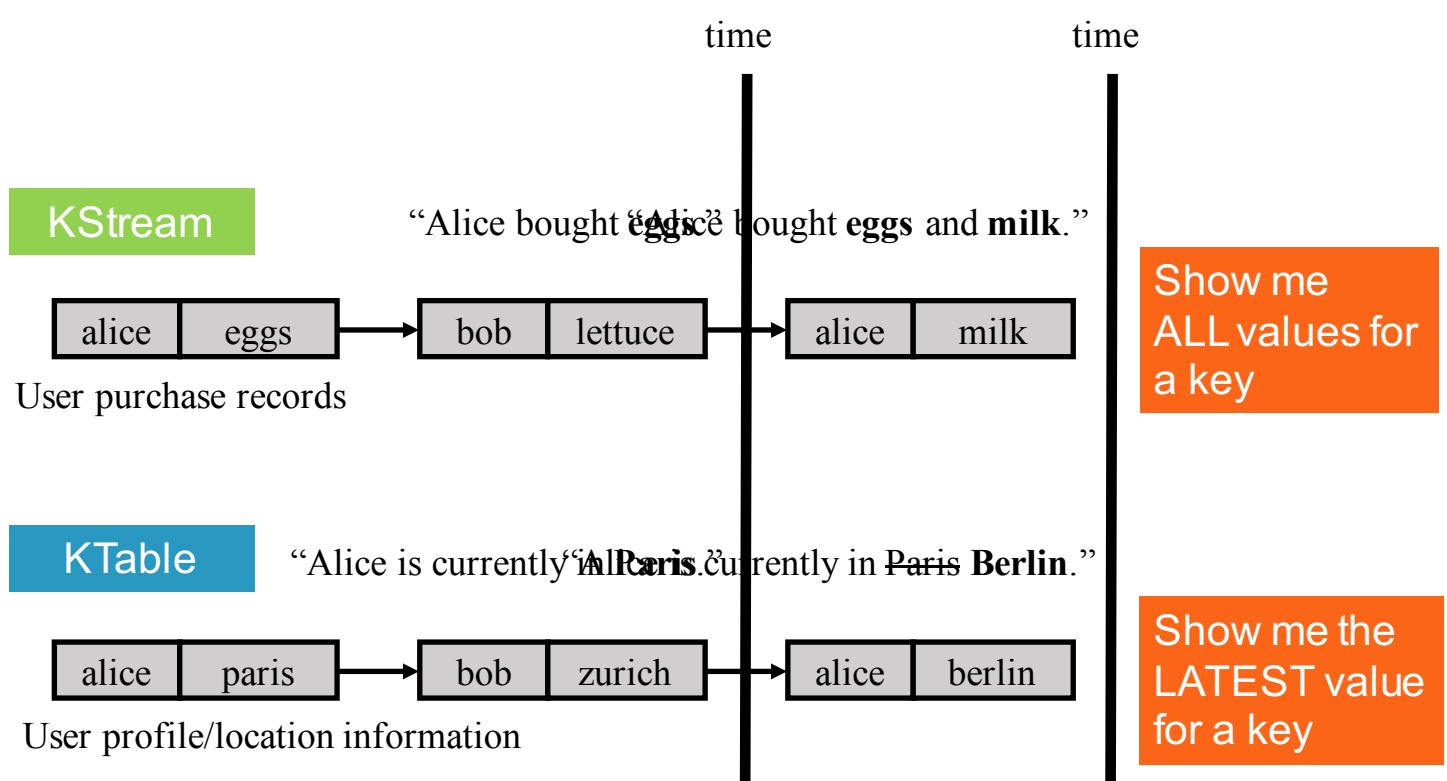


<http://www.confluent.io/blog/introducing-kafka-streams-stream-processing-made-simple>  
<http://docs.confluent.io/3.0.0streams/concepts.html#duality-of-streams-and-tables>

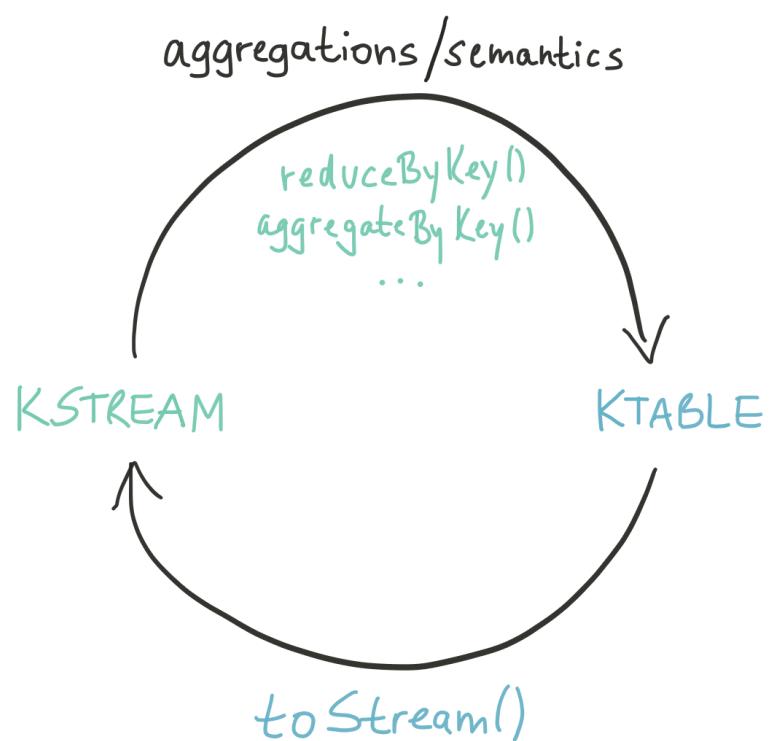
## Streams meet Tables – in the Kafka Streams DSL



## Streams meet Tables – in the Kafka Streams DSL



## Streams meet Tables – in the Kafka Streams DSL



## Streams meet Tables – in the Kafka Streams DSL

---

- JOIN example: compute user clicks by region via KStream.leftJoin(KTable)

```
// e.g. "alice" -> 13L
KStream<String, Long> userClicksStream = ...;

// e.g. "alice" -> "europe"
KTable<String, String> userRegionsTable = ...;
```

# Streams meet Tables – in the Kafka Streams DSL

- JOIN example: compute user clicks by region via KStream.leftJoin(KTable)

```
// e.g. "alice" -> 13L
KStream<String, Long> userClicksStream = ...;

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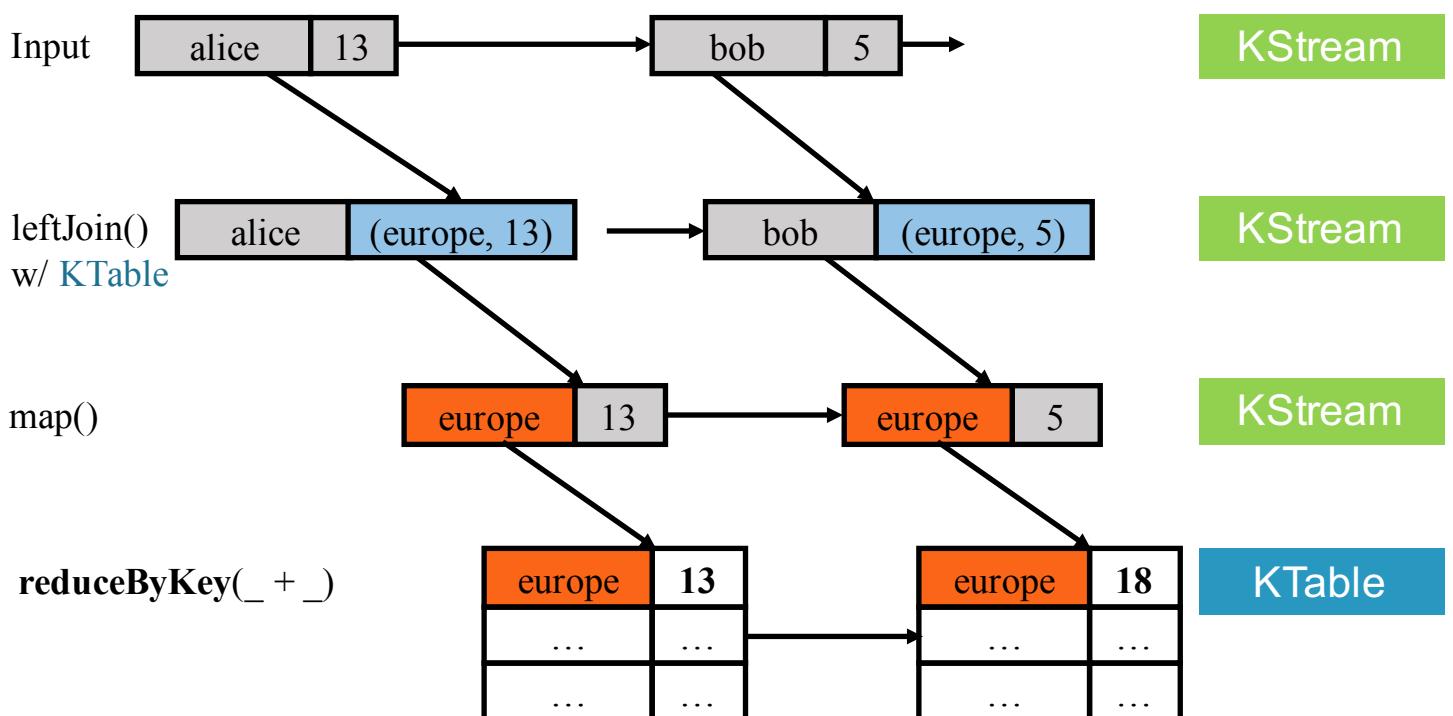
// Compute the number of user clicks per region, e.g. "europe" -> 13L
KTable<String, Long> clicksPerRegion = userClicksStream
    .leftJoin(userRegionsTable, (clicks, region) -> new RegionWithClicks(region == null ? "UNKNOWN" : region, clicks))
    .map((user, regionWithClicks) -> new KeyValue<>(regionWithClicks.region(), regionWithClicks.clicks()))
    .reduceByKey(
        (firstClicks, secondClicks) -> firstClicks + secondClicks,
        stringSerde, longSerde, "ClicksPerRegion");
```

Even simpler in Scala because, unlike Java, it natively supports tuples:

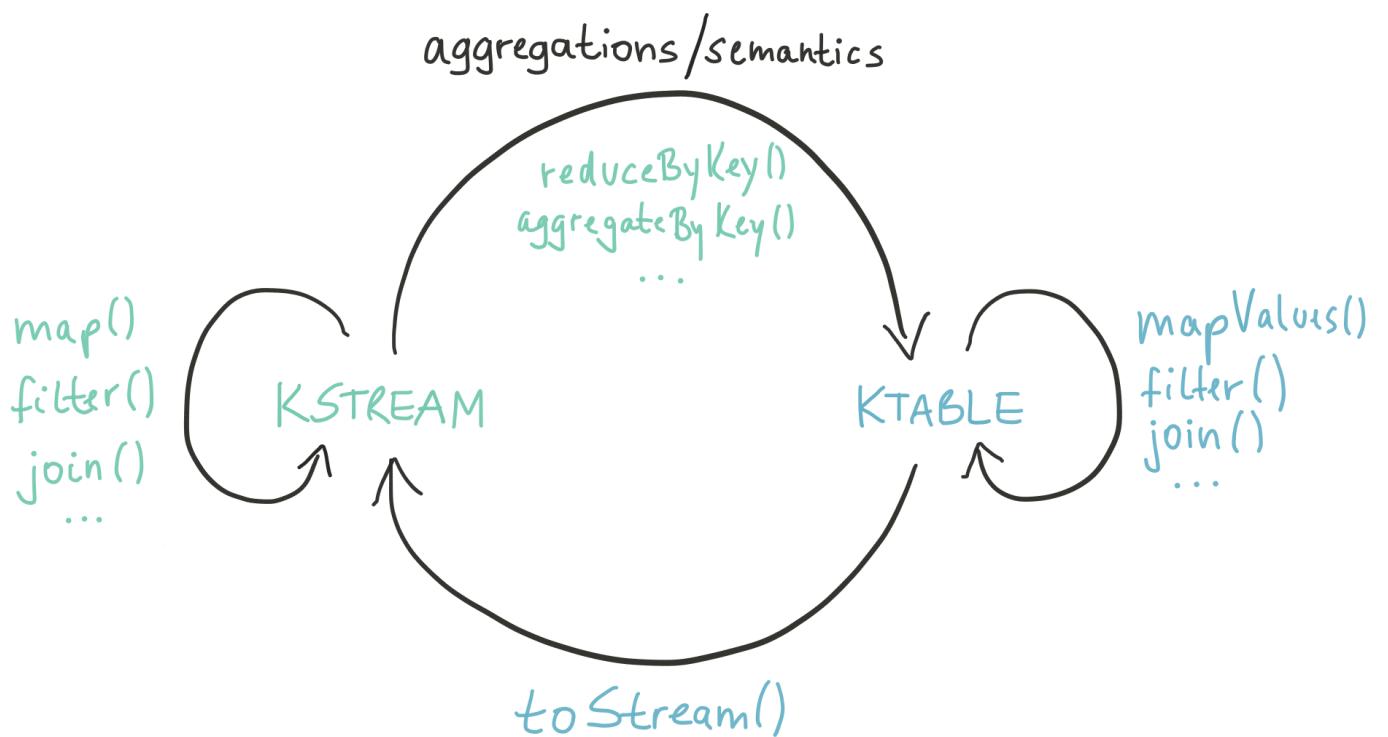
```
.leftJoin(userRegionsTable, (clicks: Long, region: String) => (if (region == null) "UNKNOWN" else region, clicks))
.map((user: String, regionWithClicks: (String, Long)) => new KeyValue(regionWithClicks._1, regionWithClicks._2))
```

## Streams meet Tables – in the Kafka Streams DSL

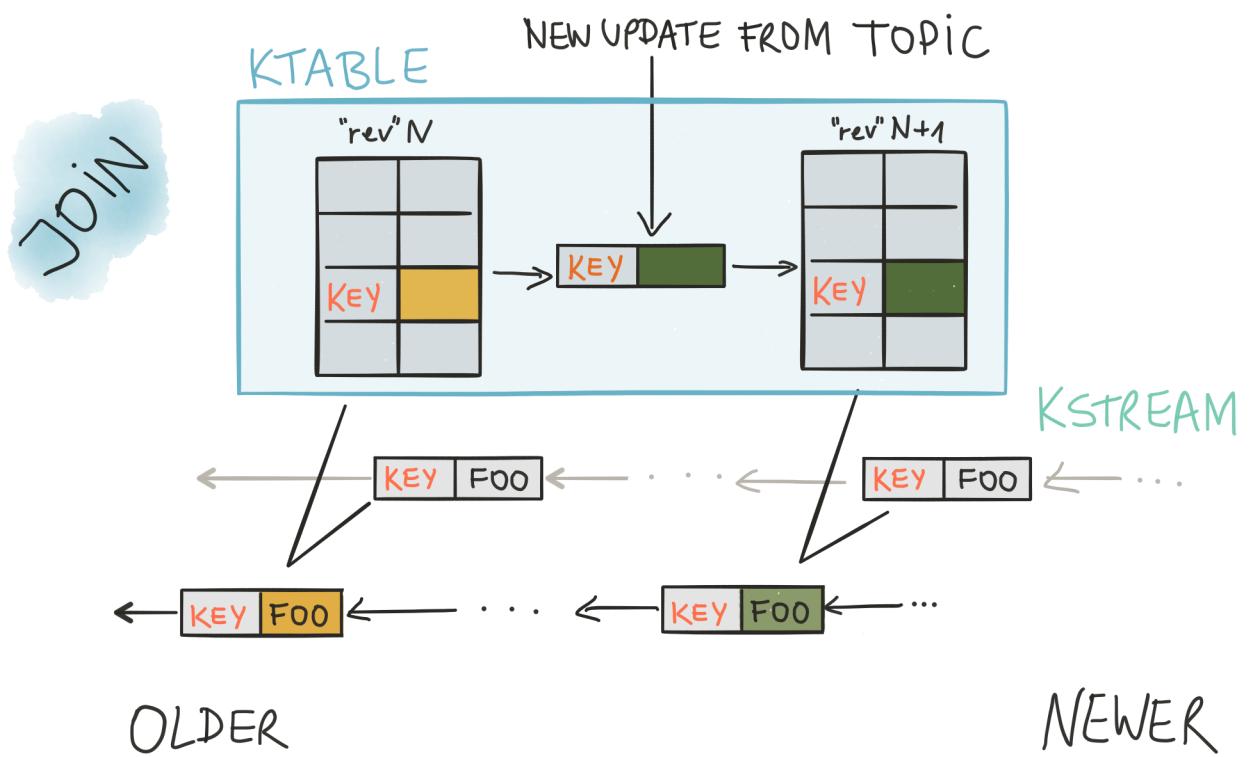
- JOIN example: compute user clicks by region via KStream.leftJoin(KTable)



## Streams meet Tables – in the Kafka Streams DSL



## Streams meet Tables – in the Kafka Streams DSL



# Kafka Streams key features

## Key features in 0.10

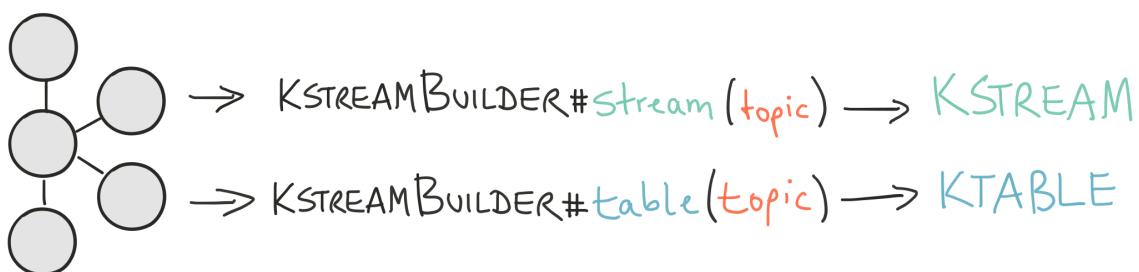
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- **Native, 100%-compatible Kafka integration**
  - Also inherits Kafka's security model, e.g. to encrypt data-in-transit
  - Uses Kafka as its internal messaging layer, too

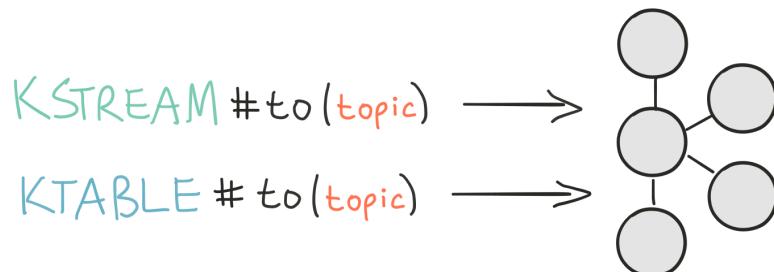
## Native Kafka integration

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- Reading data from Kafka

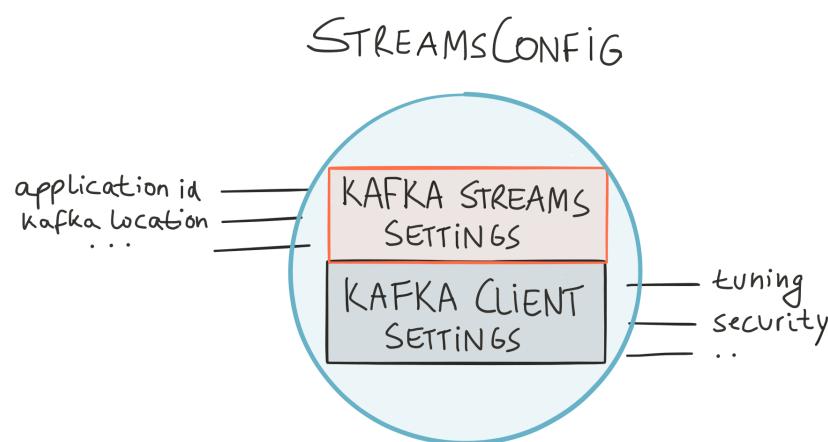


- Writing data to Kafka



## Native Kafka integration

- You can configure both Kafka Streams plus the underlying Kafka clients



```
Properties cfg = new Properties();
cfg.put(StreamsConfig.APPLICATION_ID_CONFIG, "berlin-buzzwords-demo-app");
cfg.put(StreamsConfig.BOOTSTRAP_SERVERS_CONFIG, "kafka-broker1:9092,kafka-broker2:9092");
cfg.put(ConsumerConfig.AUTO_OFFSET_RESET_CONFIG, "earliest");
// ...and so on...
StreamsConfig streamsConfig = new StreamsConfig(cfg);
```

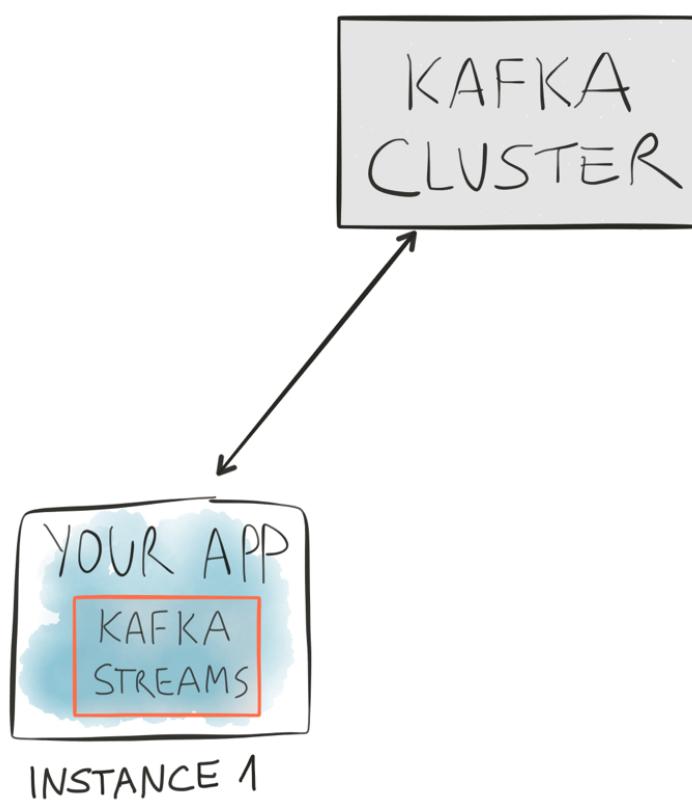
## Key features in 0.10

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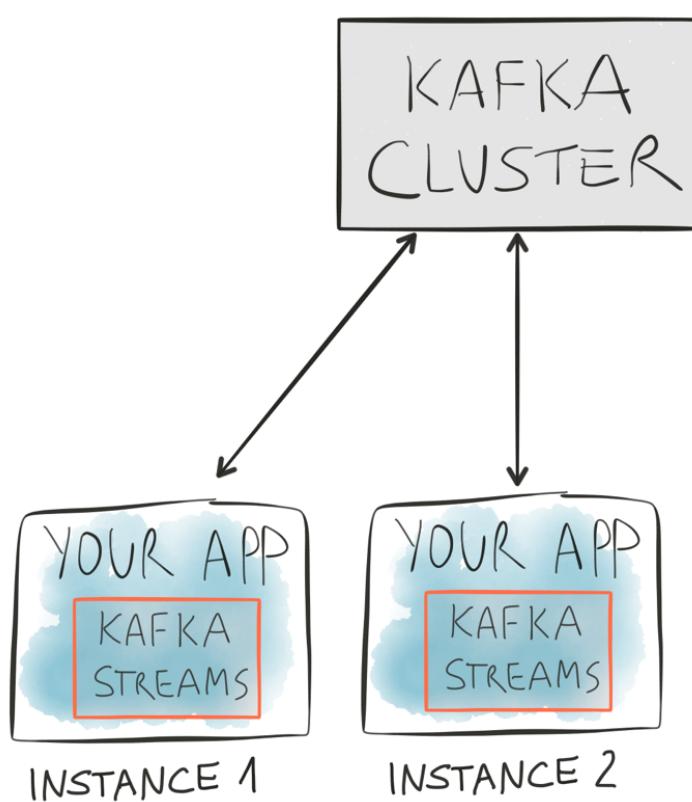
- Native, 100%-compatible Kafka integration
  - Also inherits Kafka's security model, e.g. to encrypt data-in-transit
  - Uses Kafka as its internal messaging layer, too
- **Highly scalable**
- **Fault-tolerant**
- **Elastic**

## Execution model

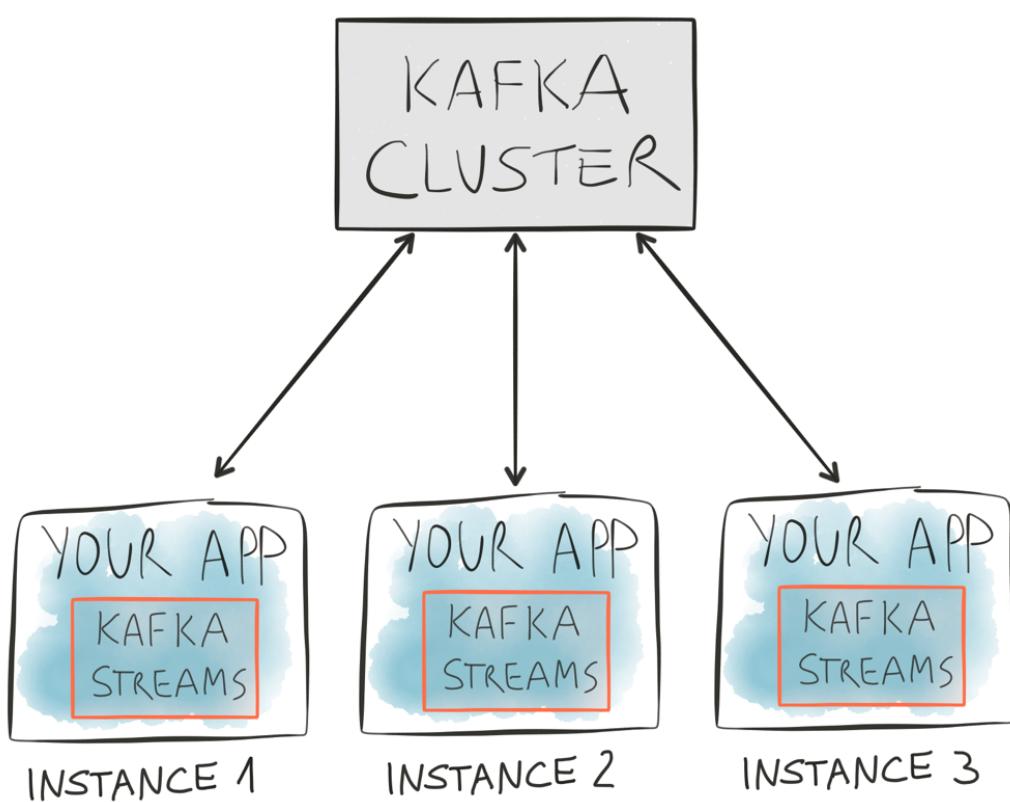
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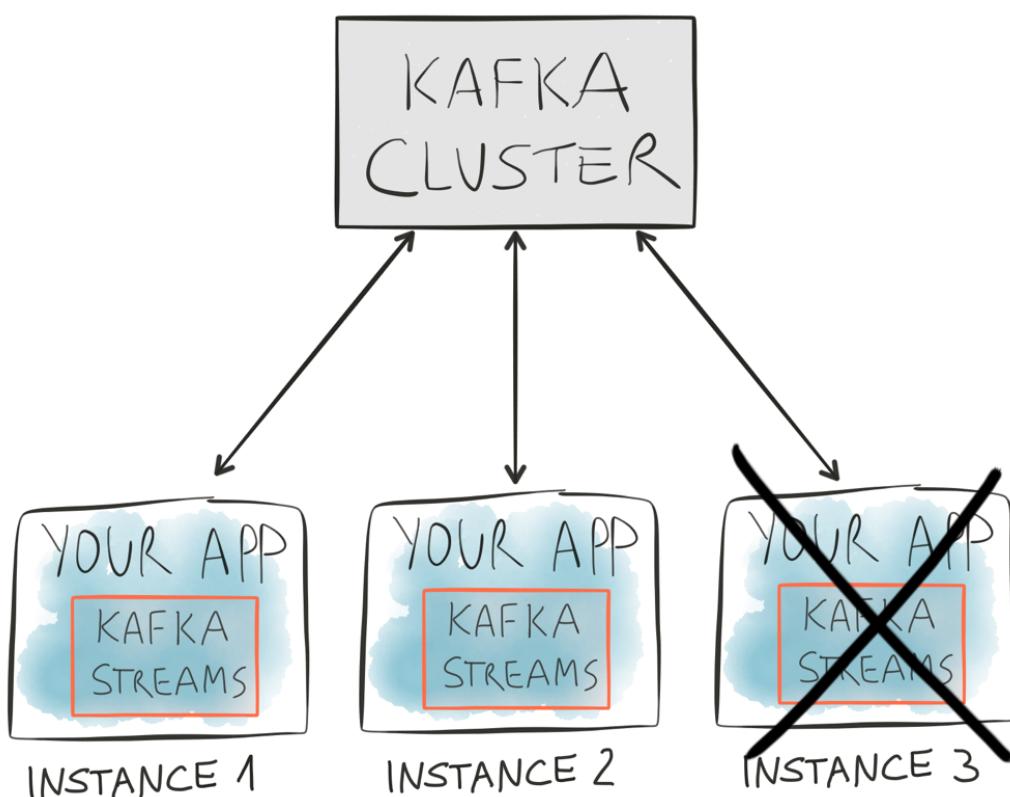
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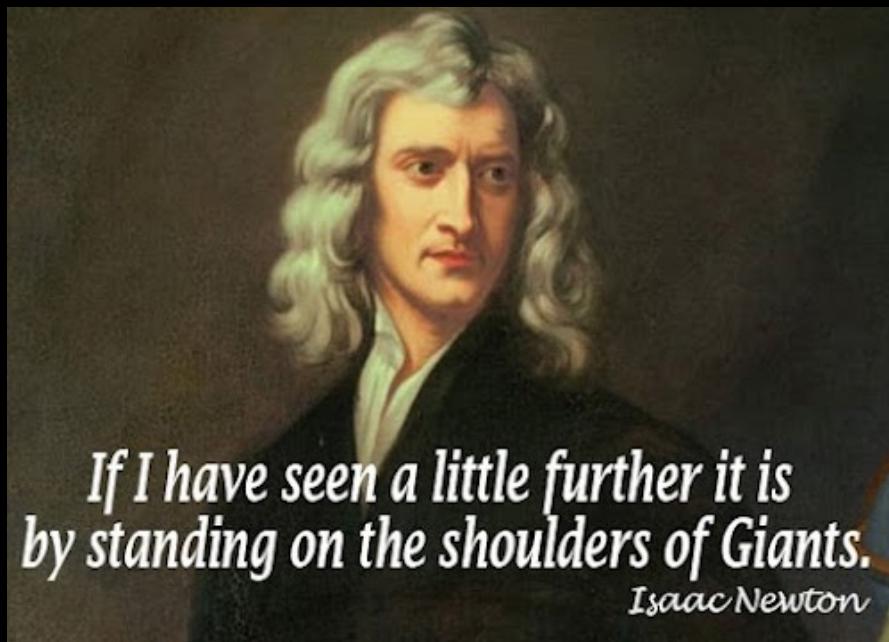
## Execution model



## Execution model



Kafka Streams outsources hard problems to Kafka



## Key features in 0.10

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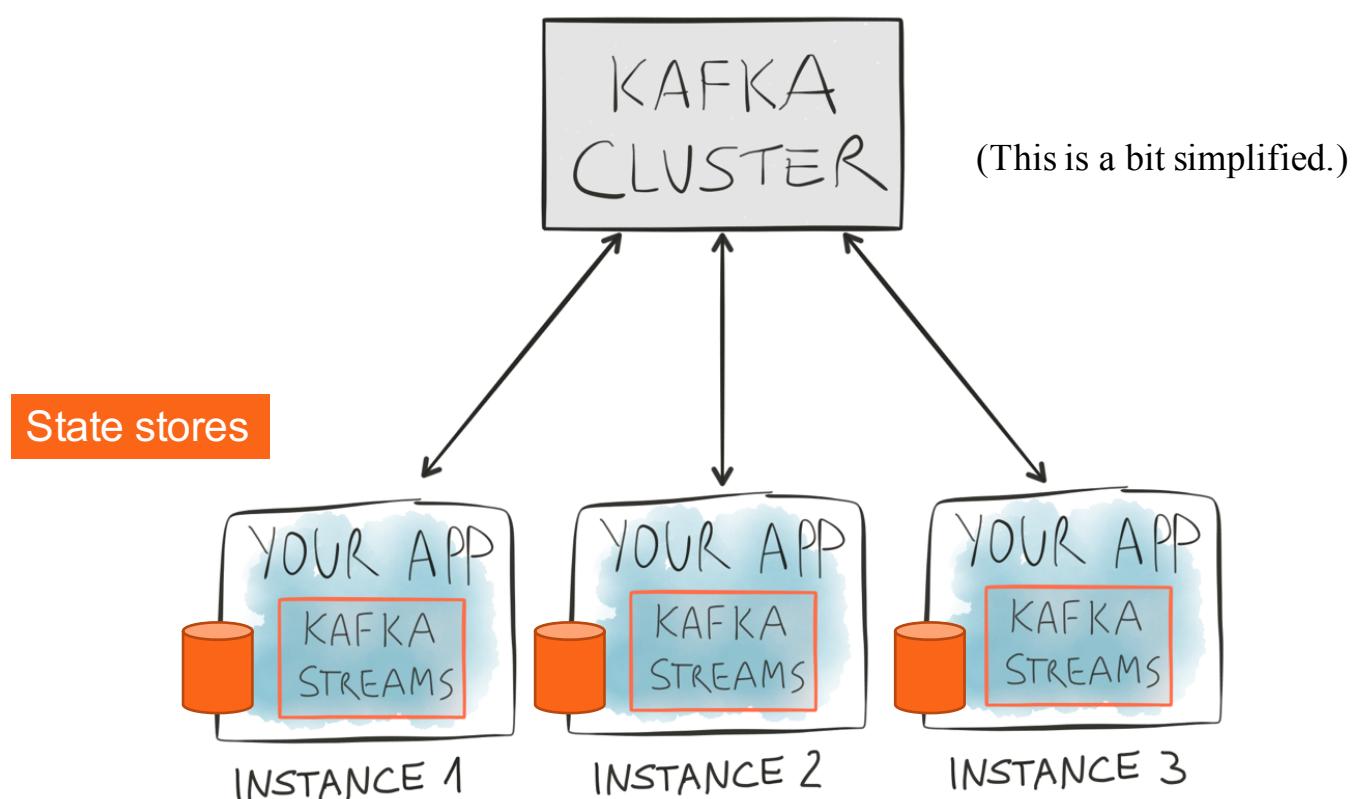
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  - Uses Kafka as its internal messaging layer, too
- Highly scalable
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- Elastic
- **Stateful and stateless computations (e.g. joins, aggregations)**

# Stateful computations

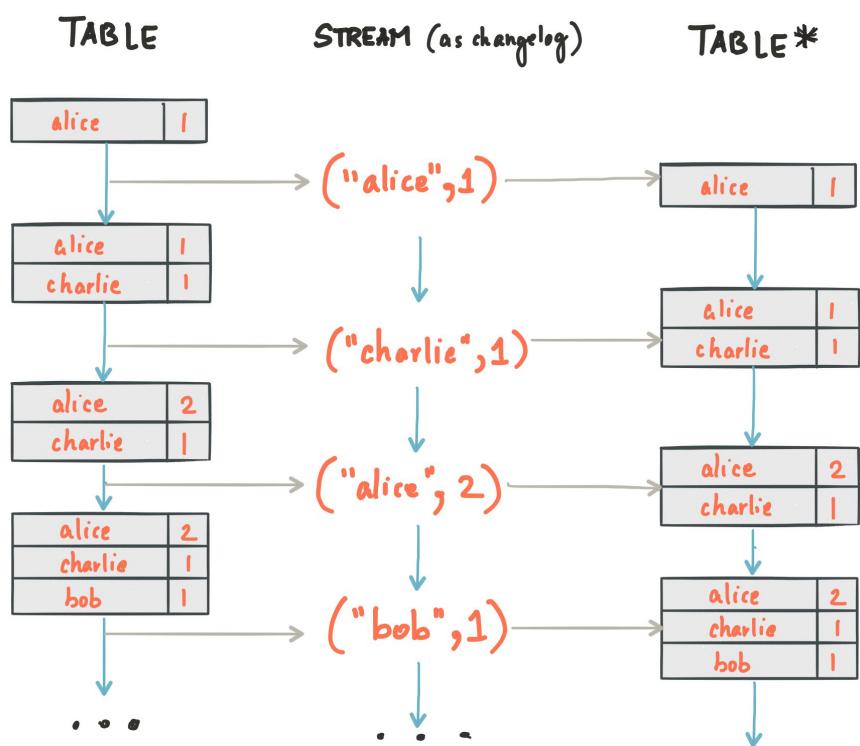
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- Stateful computations like **aggregations** or **joins** require state
  - We already showed a join example in the previous slides.
  - Windowing a stream is stateful, too, but let's ignore this for now.
- **State stores** in Kafka Streams
  - Typically: key-value stores
  - Pluggable implementation: RocksDB (default), in-memory, your own ...
- State stores are **per stream task** for isolation (think: share-nothing)
- State stores are **local** for best performance
- State stores are **replicated to Kafka** for elasticity and for fault-tolerance

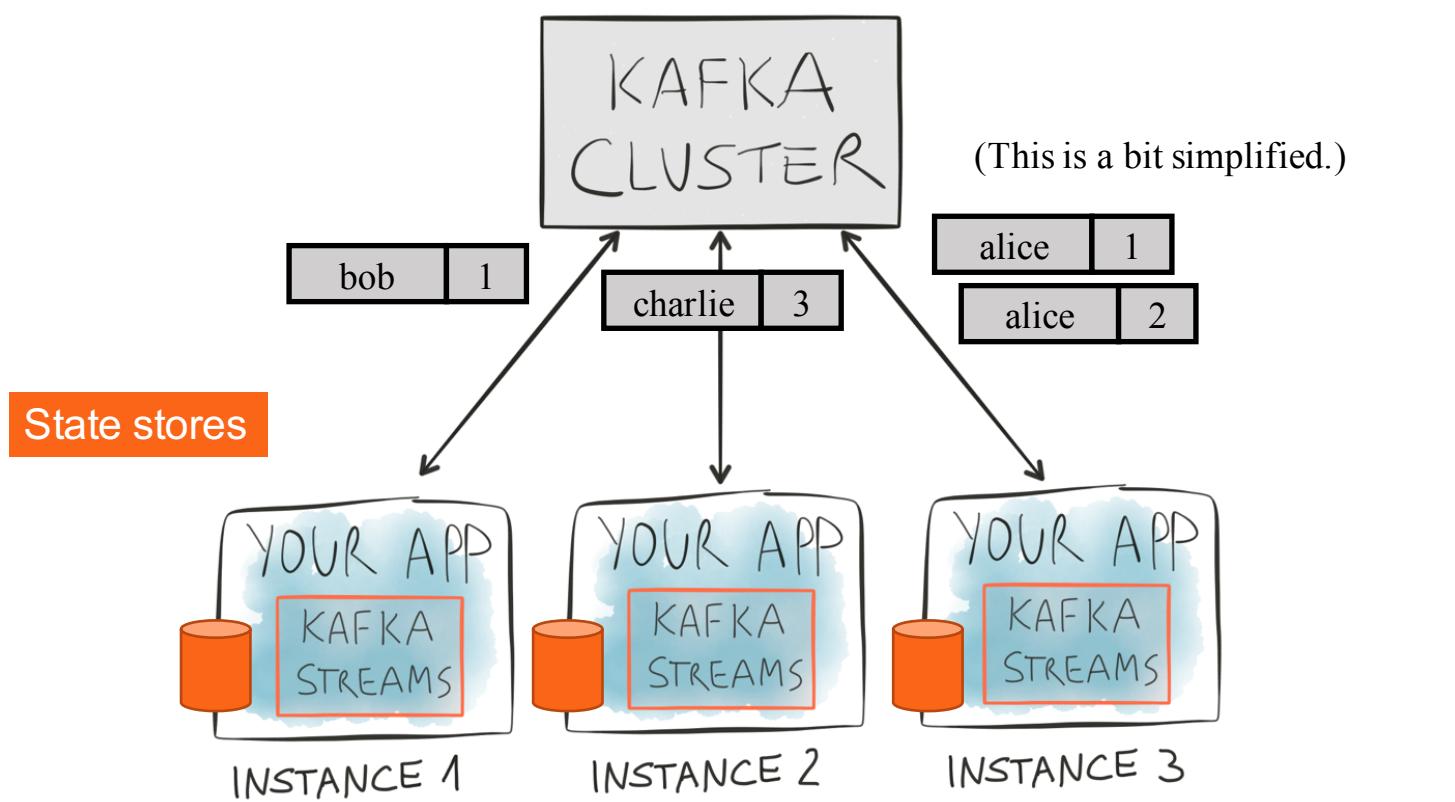
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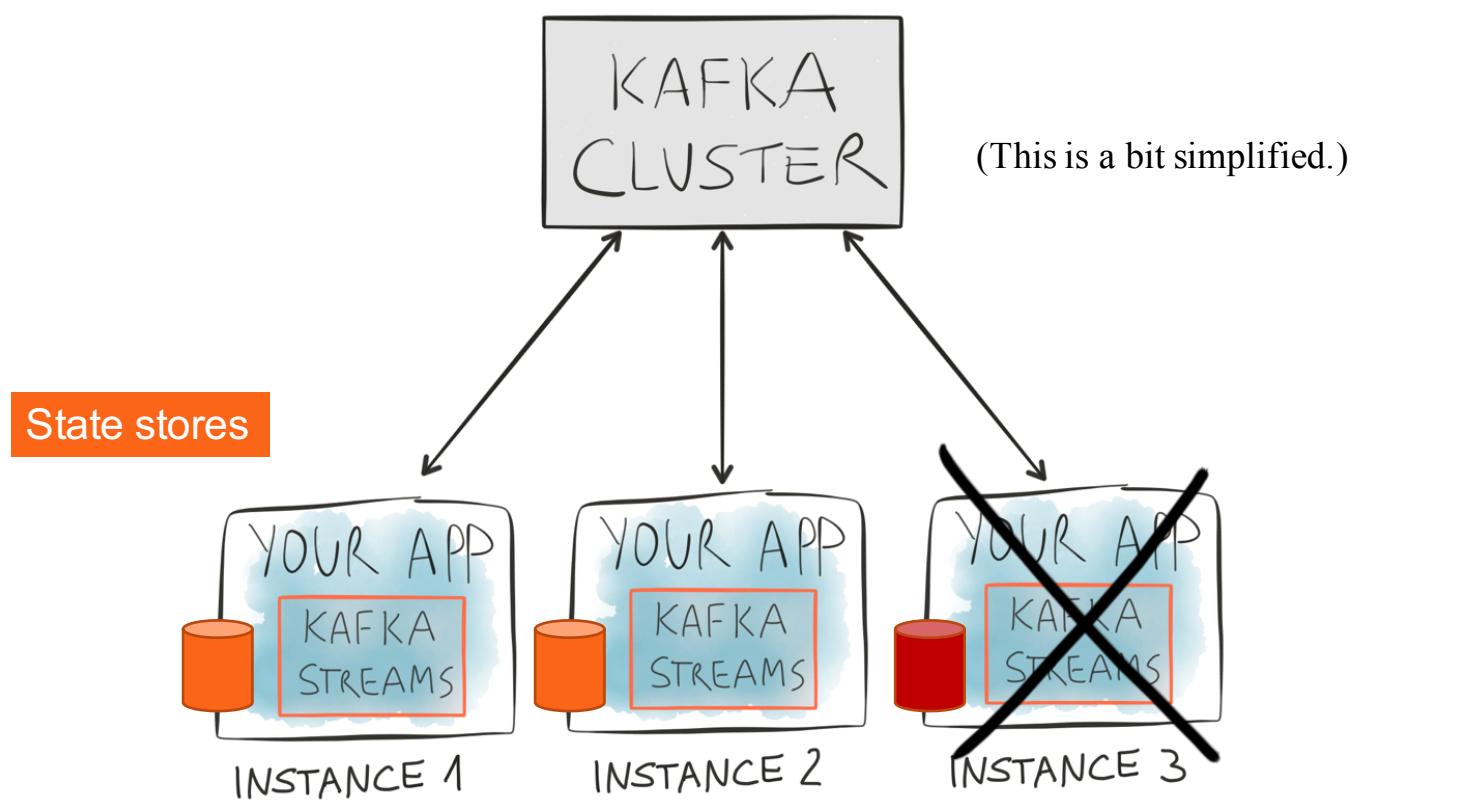
## Remember?



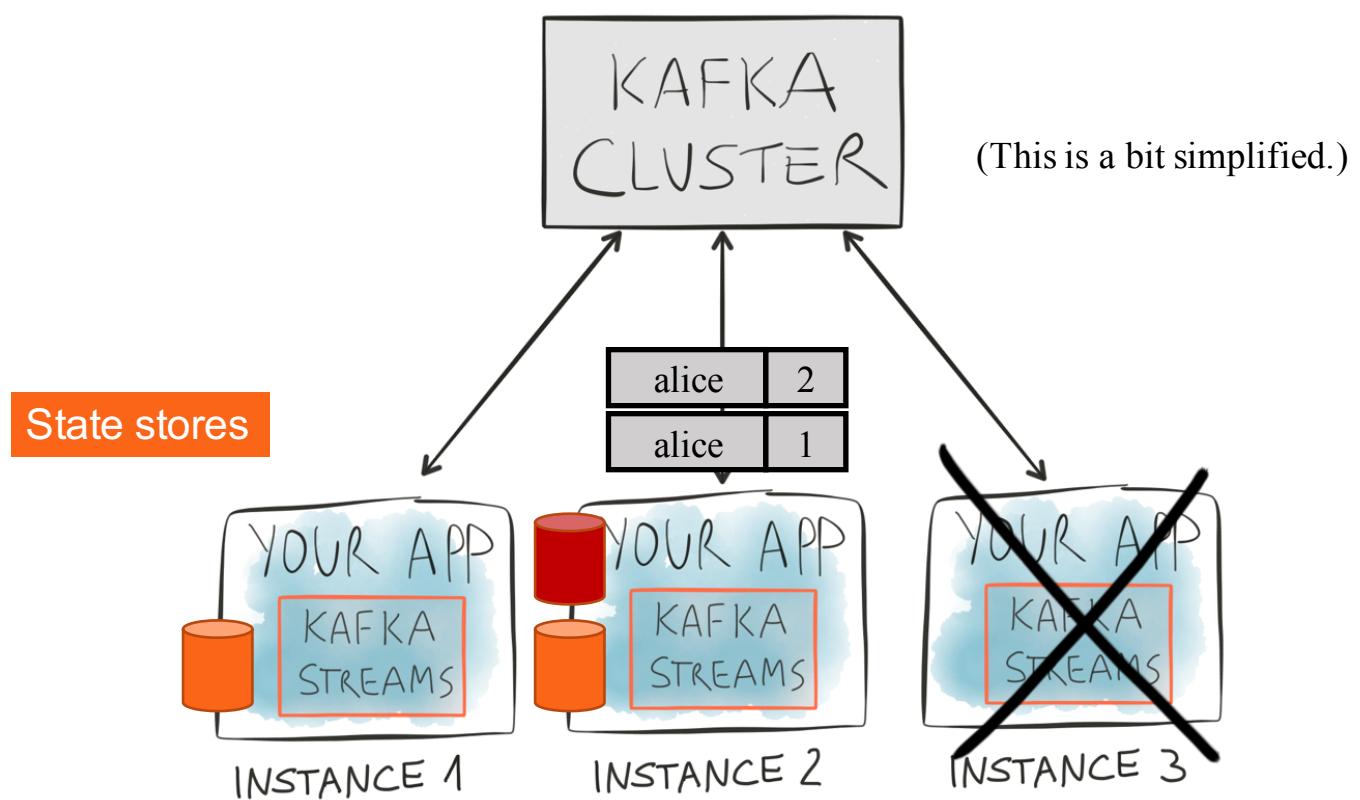
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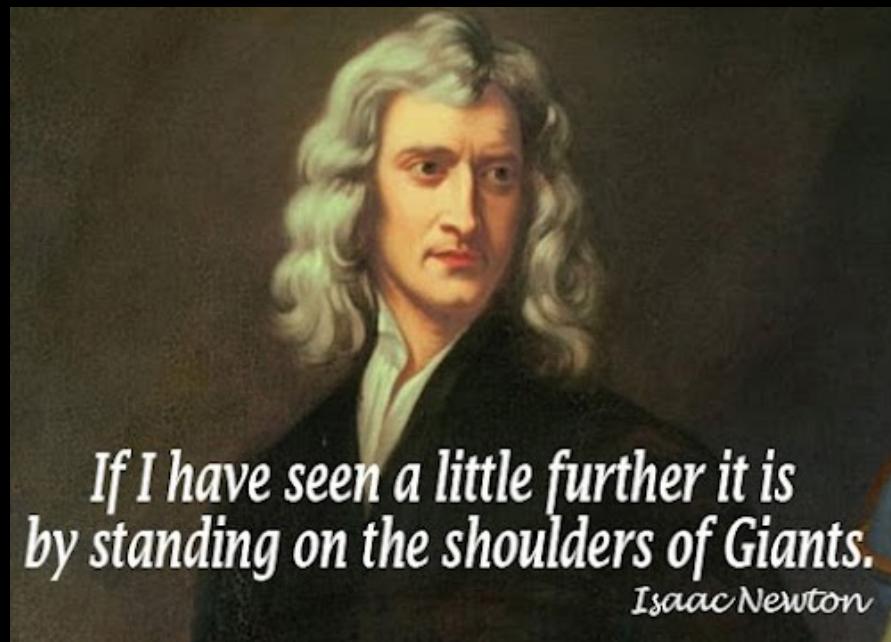
## Execution model



## Execution model



Kafka Streams outsources hard problems to Kafka



## Stateful computations

---

- **Kafka Streams DSL:** abstracts state stores away from you
  - Stateful operations include
    - count(), reduceByKey(), aggregate(), ...
- **Low-level Processor API:** direct access to state stores
  - Very flexible but more manual work for you

# Stateful computations

- Use the low-level Processor API to interact directly with state stores

```
public class WordCountProcessor extends Processor<byte[], String> {  
  
    private KeyValueStore<String, Long> stateStore;  
  
    @Override  
    public void init(ProcessorContext context) {  
        stateStore = (KeyValueStore) context.getStateStore("WordCounts");  
    }  
  
    @Override  
    public void process(byte[] key, String word) {  
        Integer oldValue = stateStore.get(word);  
        if (oldValue == null) {  
            stateStore.put(word, 1L);  
        } else {  
            stateStore.put(word, oldValue + 1L);  
        }  
    }  
  
    // rest omitted  
}
```

Get the store

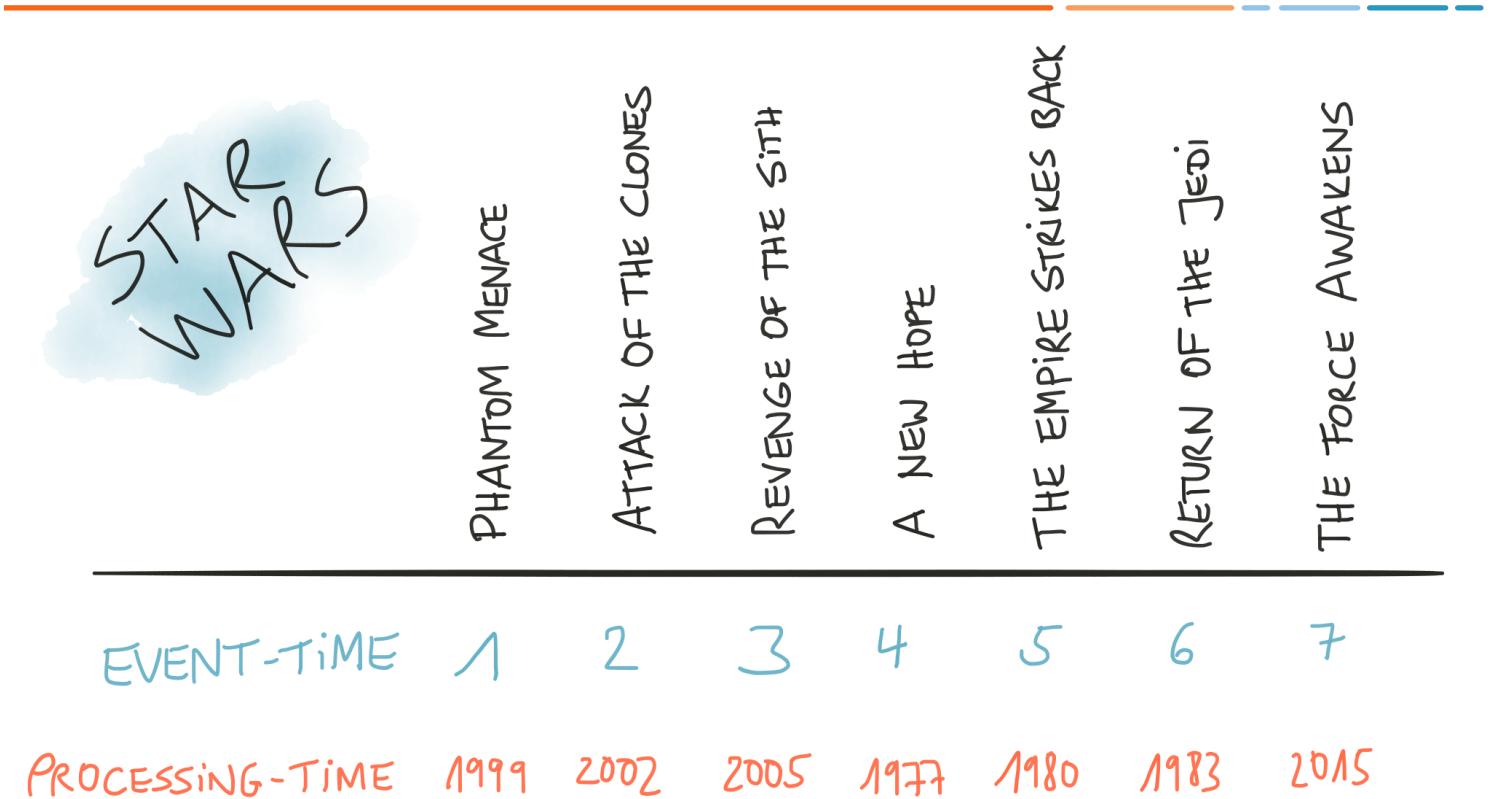
Use the store

## Key features in 0.10

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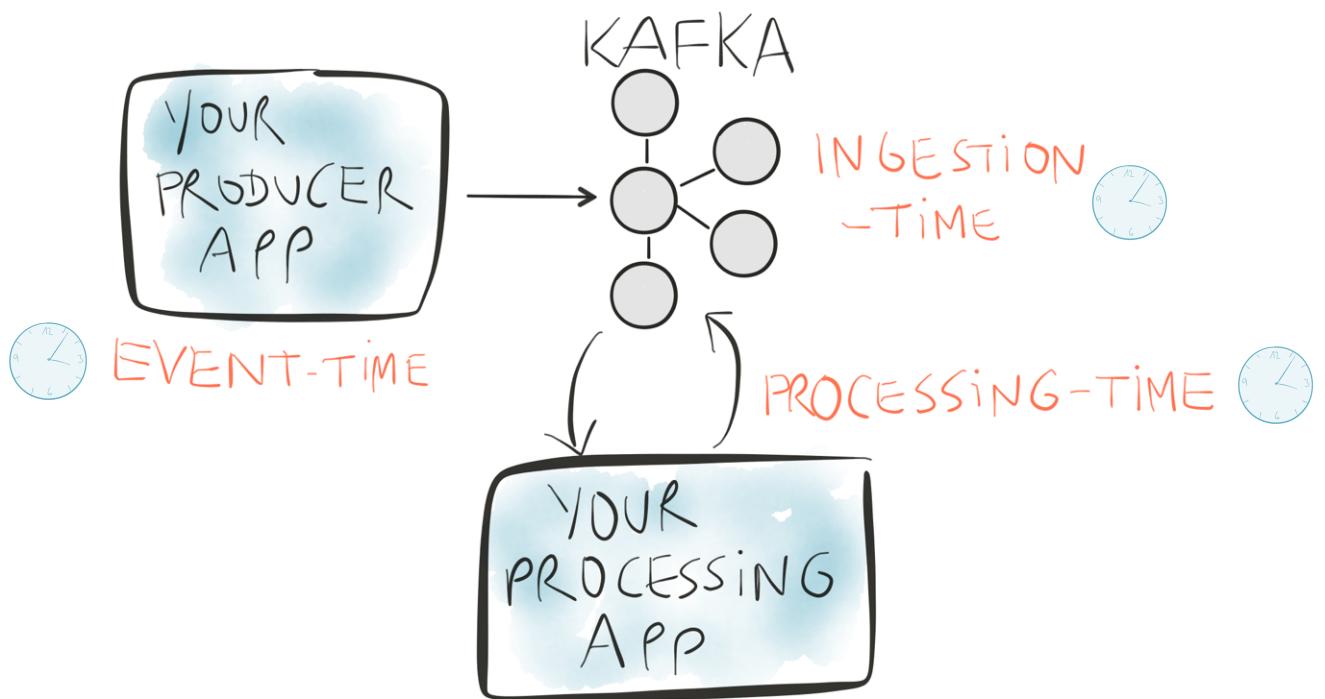
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- Highly scalable
- Fault-tolerant
- Elastic
- Stateful and stateless computations
- **Time model**

Time



## Time

---



# Time

---

- You configure the desired time semantics through **timestamp extractors**
- Default extractor yields **event-time** semantics
  - Extracts embedded timestamps of Kafka messages (introduced in v0.10)

```
// Event-time (default timestamp extractor in 0.10)
public class ConsumerRecordTimestampExtractor implements TimestampExtractor {
    @Override
    public long extract(ConsumerRecord<Object, Object> record) {
        return record.timestamp();
    }
}
```

```
// Processing-time
public class WallclockTimestampExtractor implements TimestampExtractor {
    @Override
    public long extract(ConsumerRecord<Object, Object> record) {
        return System.currentTimeMillis();
    }
}
```

## Key features in 0.10

---

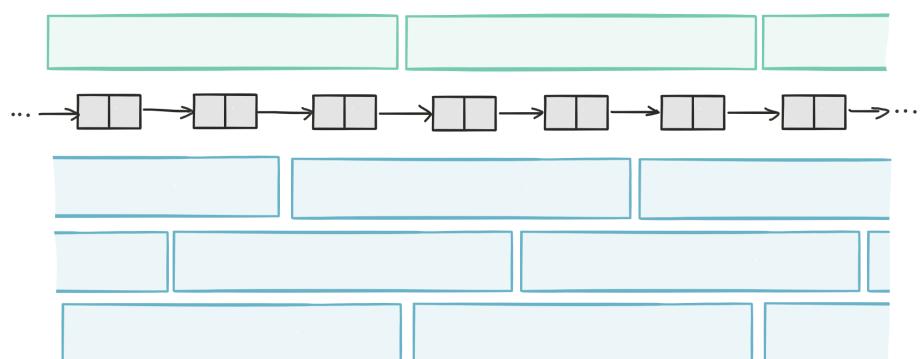
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- **Windowing**

# Windowing

`TimeWindows.of(3000)`

"aggregate for 3 secs, tell me every 3 sec"

TUMBLING WINDOWS

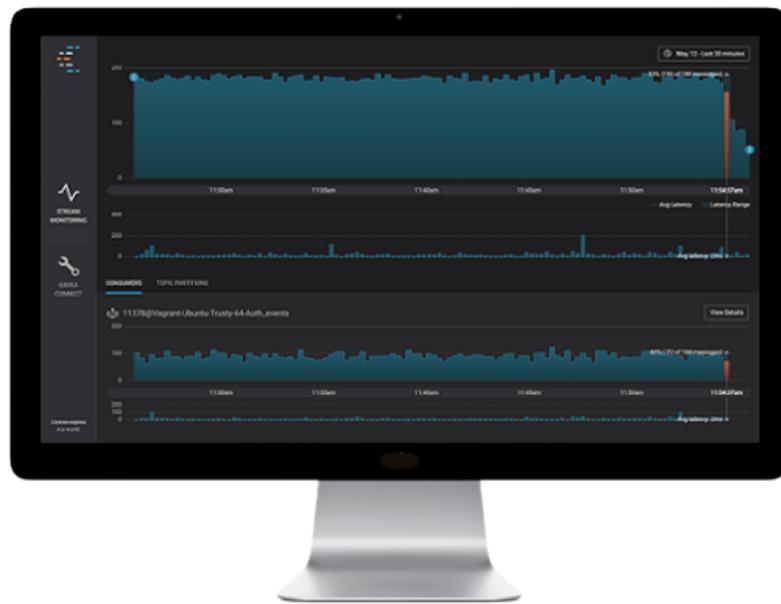


HOPPING WINDOWS

"aggregate for 3 secs, tell me every 1 sec"

`TimeWindows.of(3000).advanceBy(1000)`

## Windowing use case: monitoring (1m/5m/15m averages)



Confluent Control Center for Kafka

## Key features in 0.10

---

- Native, 100%-compatible Kafka integration
  - Also inherits Kafka's security model, e.g. to encrypt data-in-transit
  - Uses Kafka as its internal messaging layer, too
- Highly scalable
- Fault-tolerant
- Elastic
- Stateful and stateless computations
- Time model
- Windowing
- **Supports late-arriving and out-of-order data**
- **Millisecond processing latency, no micro-batching**
- **At-least-once processing guarantees (exactly-once is in the works)**

# Wrapping up

## Where to go from here?

---

- Kafka Streams is available in Apache Kafka 0.10 and Confluent Platform 3.0
  - <http://kafka.apache.org/>
  - <http://www.confluent.io/download> (free + enterprise versions, tar/zip/deb/rpm)
- Kafka Streams demos at <https://github.com/confluentinc/examples>
  - Java 7, Java 8+ with lambdas, and Scala
  - WordCount, Joins, Avro integration, Top-N computation, Windowing, ...
- Apache Kafka documentation: <http://kafka.apache.org/documentation.html>
- Confluent documentation: <http://docs.confluent.io/3.0.0/streams/>
  - Quickstart, Concepts, Architecture, Developer Guide, FAQ
- Join our bi-weekly *Ask Me Anything* sessions on Kafka Streams
  - Contact me at [michael@confluent.io](mailto:michael@confluent.io) for details

## Some of the things to come

---

- Exactly-once semantics
- Queriable state – tap into the state of your applications
- SQL interface
- Listen to and collaborate with the developer community
  - Your feedback counts a lot! Share it via [users@kafka.apache.org](mailto:users@kafka.apache.org)



**Tomorrow's keynote (09:30 AM) by Neha Narkhede,**  
co-founder and CTO of Confluent

“Application development and data in the emerging world  
of stream processing”

Want to contribute to Kafka and open source?

---

**Join the Kafka community**  
**<http://kafka.apache.org/>**

...in a great team with the creators of Kafka?

**Confluent is hiring 😊**  
**<http://confluent.io/>**

Questions, comments? Tweet with [#bbuzz](#) and /cc to [@ConfluentInc](#)



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