



Akka Streams, Kafka, Kinesis

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#StreamProcessingBe



whoami : Peter Vandenabeele

@peter_v

@All_Things_Data (my consultancy)

current client:

Real Impact Analytics @RIAnalytics

Telecom Analytics (emerging markets)



Agenda

5' Intro (Peter)

40' Akka Streams, Kafka, Kinesis (Peter)

45' Spark Streaming and Kafka Demo (Gerard)

15' Open discussion (all)

30' beers (doors close at 21:30)



Many thanks to

@All_Things_Data (beer)

@maasg (Gerard Maas)

you !

=> Note: always looking for locations

Agenda



Why ?

Akka + Akka Streams

Demo

Kafka + Kinesis

Demo

Why !

Why ?

distributed state

distributed failure

slow consumers





Akka



Why (for iHeartRadio 📻)

Why we picked Akka cluster as the core architecture for our microservice

- Out-of-the-box clustering infrastructure
- Loose coupling without the cost of JSON parsing.
- Transparent programming model within and across microservices.
- Strong community and commercial support
- High resiliency, performance and scalability.

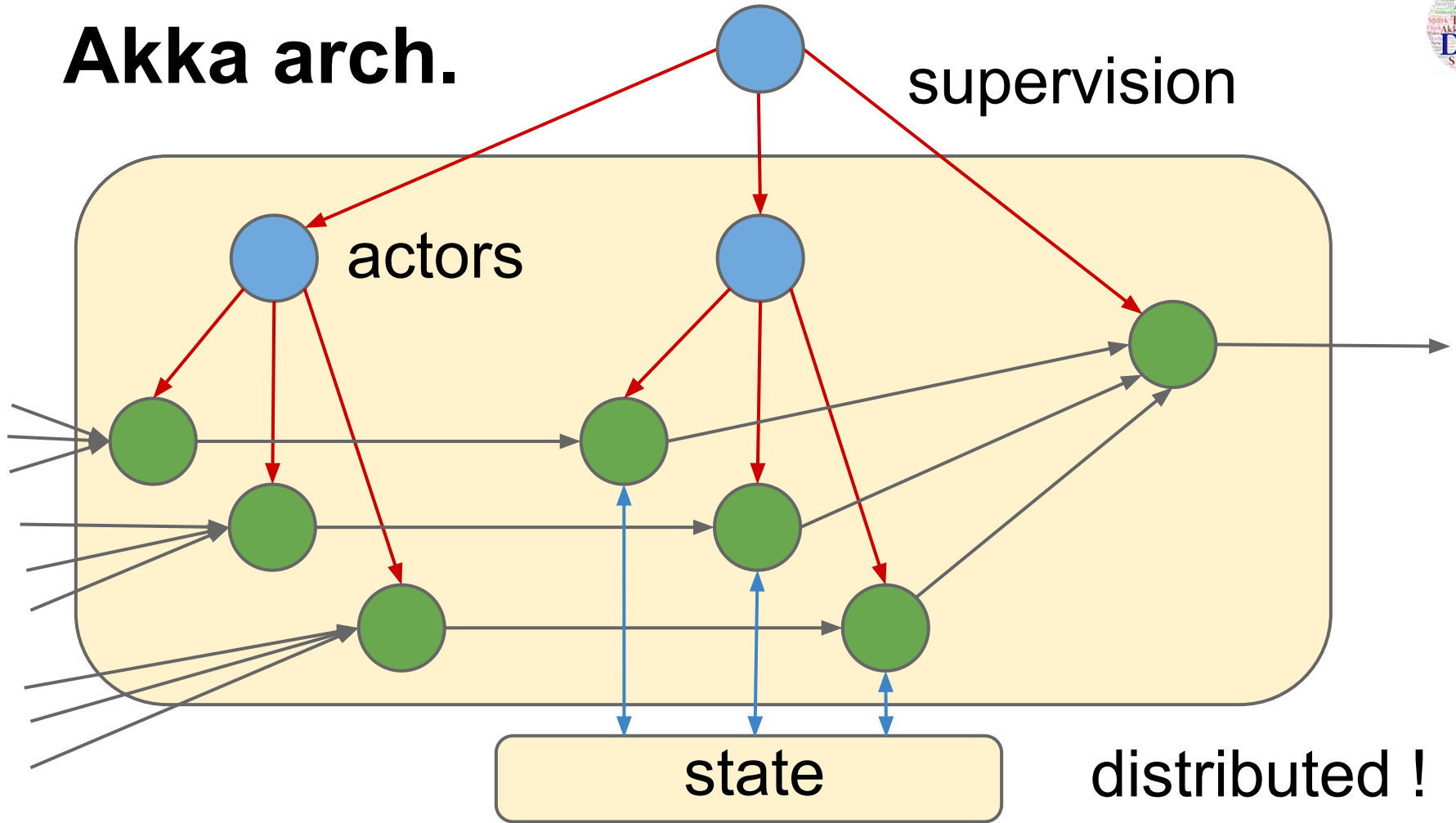
Akka design



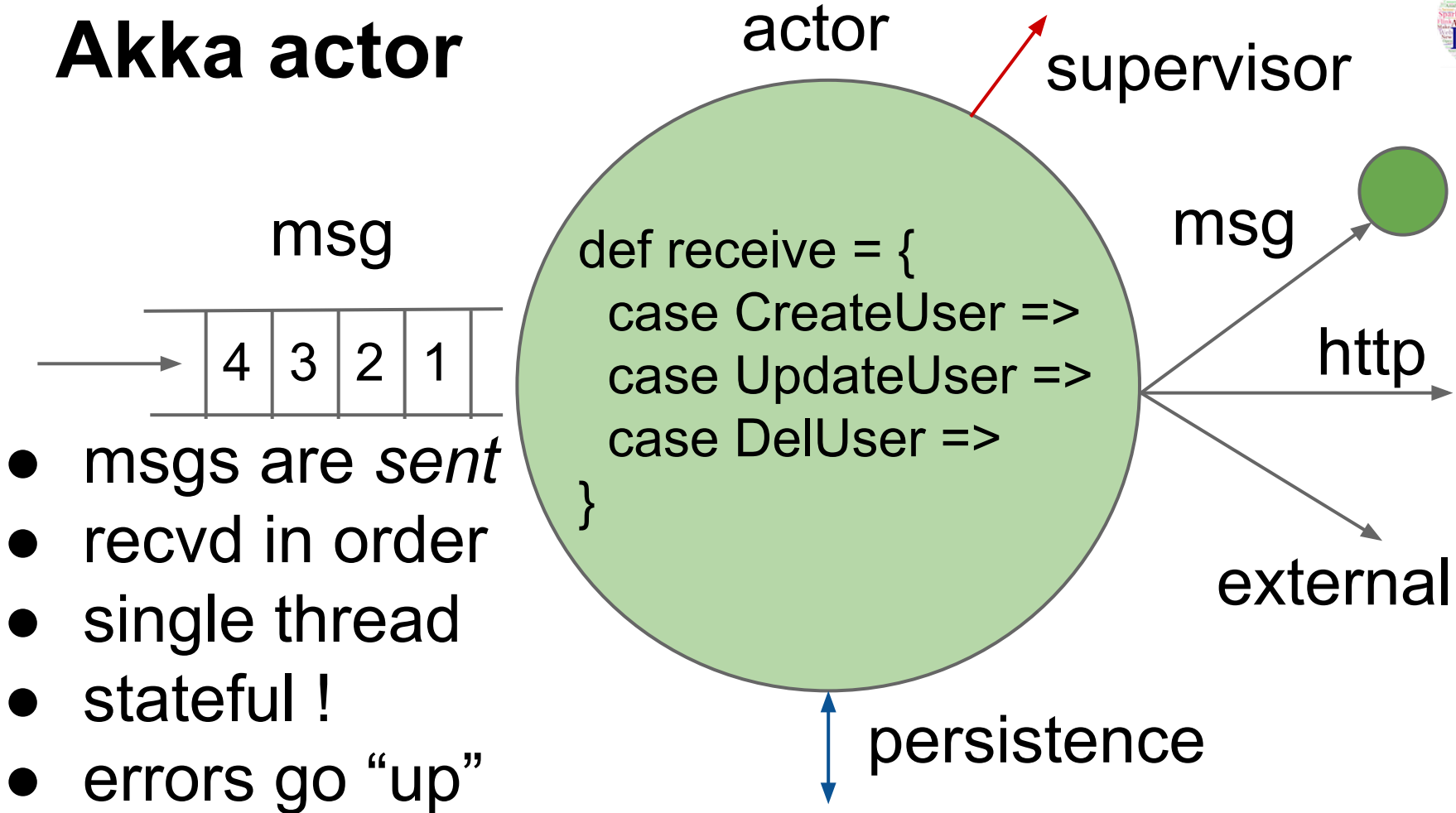
Building

- concurrent \leq many (slow) CPU's
- distributed \leq distributed state
- resilient \leq distributed failure
- applications \leq platform
- on JVM \leq Erlang OTP

Akka arch.



Akka actor





Akka usage + courses

- concurrent programming *not* easy ...
- but *without* Akka ... would be much harder
- Spark (see log extract next slide)
- Flink (version 0.9 of 24 June)
- local projects (e.g. "Wegen en verkeer")
- BeScala Meetup now runs Akka intro course
- commercial courses (Cronos, Scala World...)



Spark heavily based on Akka

log extract from Spark:

```
java -cp ...
```

```
"org.apache.spark.executor.CoarseGrainedExecutorBackend"
```

```
"--driver-url" "akka.tcp://sparkDriver@docker-master:  
51810/user/CoarseGrainedScheduler"
```

```
"--executor-id" "23" "--hostname" "docker-slave2" "--cores" "8"
```

```
"--worker-url" "akka.tcp://sparkWorker@docker-slave2:  
50268/user/Worker"
```





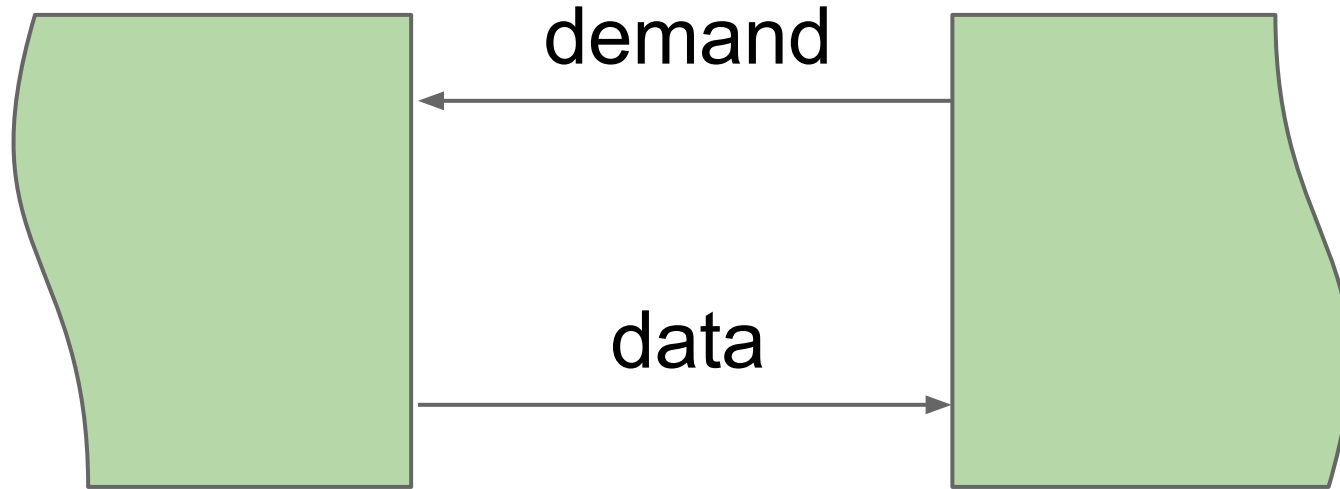
Reactive Streams

- <http://reactive-streams.org>
- exchange of
stream data across
asynchronous boundary in
bounded fashion
- building and industry standard (open IP)

Demand based

producer

consumer



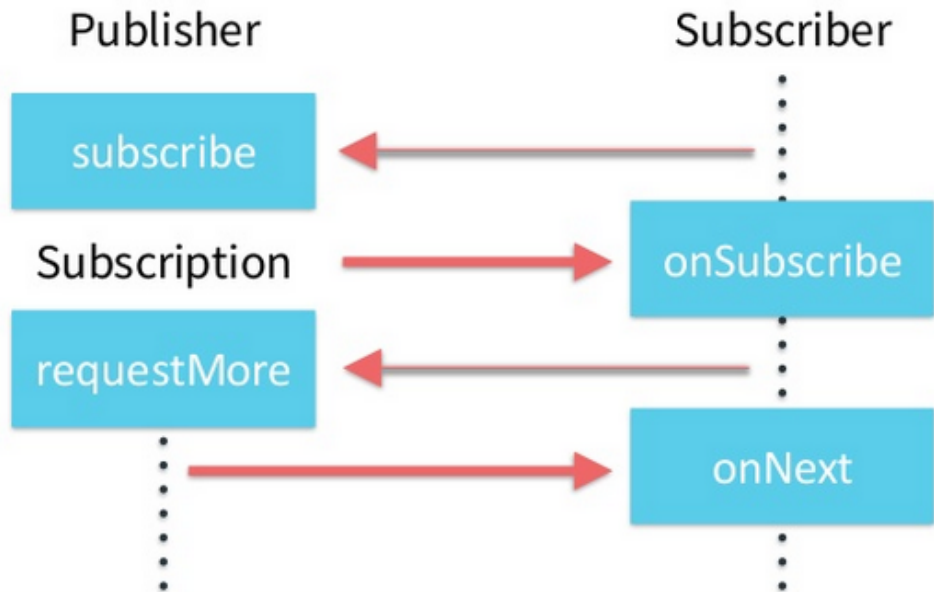
“sending 2, 5, 10, ...”

“give me max 20”

“give me max 10 more”

How does it work?

Don't try this at home



```

trait Publisher[T] {
  def subscribe(sub: Subscriber[T])
}
trait Subscription {
  def requestMore(n: Int): Unit
  def cancel(): Unit
}
trait Subscriber[T] {
  def onSubscribe(s: Subscription):
  def onNext(elem: T): Unit
  def onError(thr: Throwable): Unit
  def onComplete(): Unit
}
  
```



- ```

graph LR
 in((in)) -- f1 --> bcast((bcast))
 bcast -- f2 --> merge((merge))
 bcast -- f4 --> merge
 merge -- f3 --> out((out))

```



# Akka Streams : advantages

- Types (stream of T)
- makes it trivially simple :-)
- Many examples online (fast and simple)
- demo of simplistic case

# simplistic Akka Streams demo



```
import akka.actor.ActorSystem
import akka.stream.ActorFlowMaterializer
import akka.stream.scaladsl._

object Words {

 implicit val system = ActorSystem("System")
 implicit val materializer = ActorFlowMaterializer()
 import system.dispatcher

 def processWords(sinkAction: (String => Unit)): Unit = {

 val text =
 """|Lorem Ipsum is simply text
 |Lorem Ipsum has been the industry standard dummy text ever since the 1500s
 |when an unknown printer took a galley of type and scrambled it to make a type
 |specimen book""".stripMargin

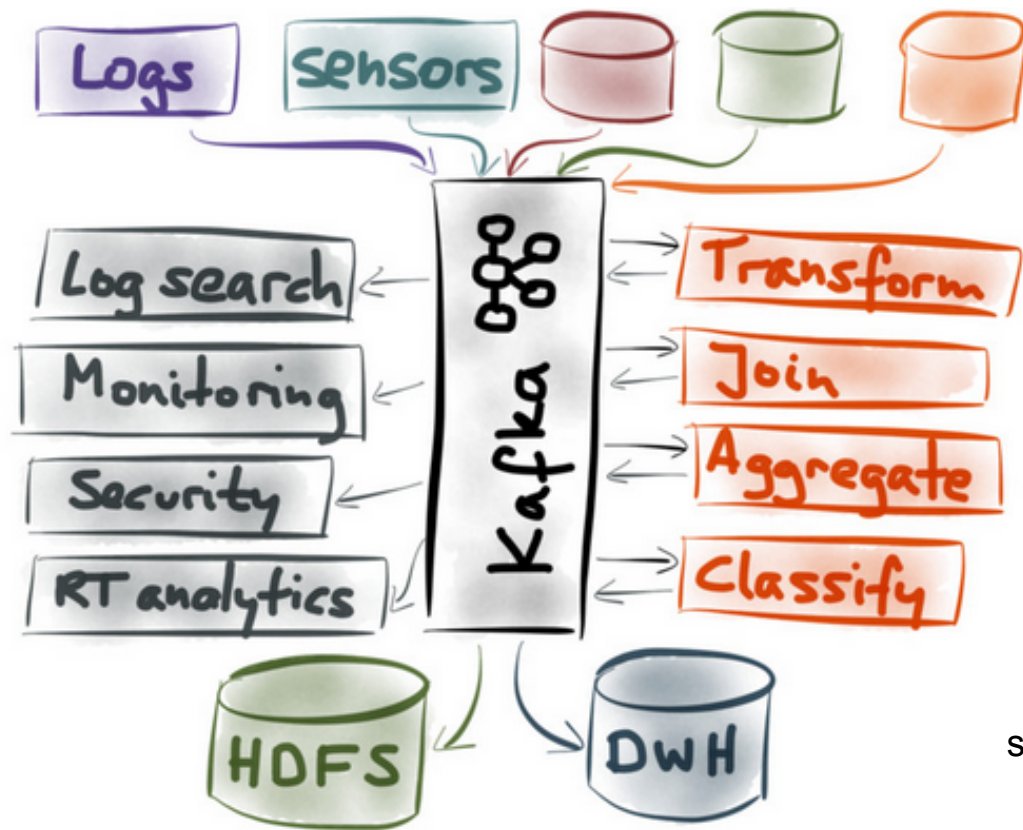
 Source(() => text.split("\\s").iterator)
 .map(_.>toUpperCase)
 .filter(line => line.length > 3)
 .runForeach(sinkAction)
 .onComplete(_ => system.shutdown())

 // in ~> f1 ~> bcast ~> f2 ~> merge ~> f3 ~> out
 // bcast ~> f4 ~> merge
 }
}
```



# Kafka

# Kafka (LinkedIn) : Martin Kleppmann



source : Martin Kleppmann  
at strata Hadoop London

# consumers

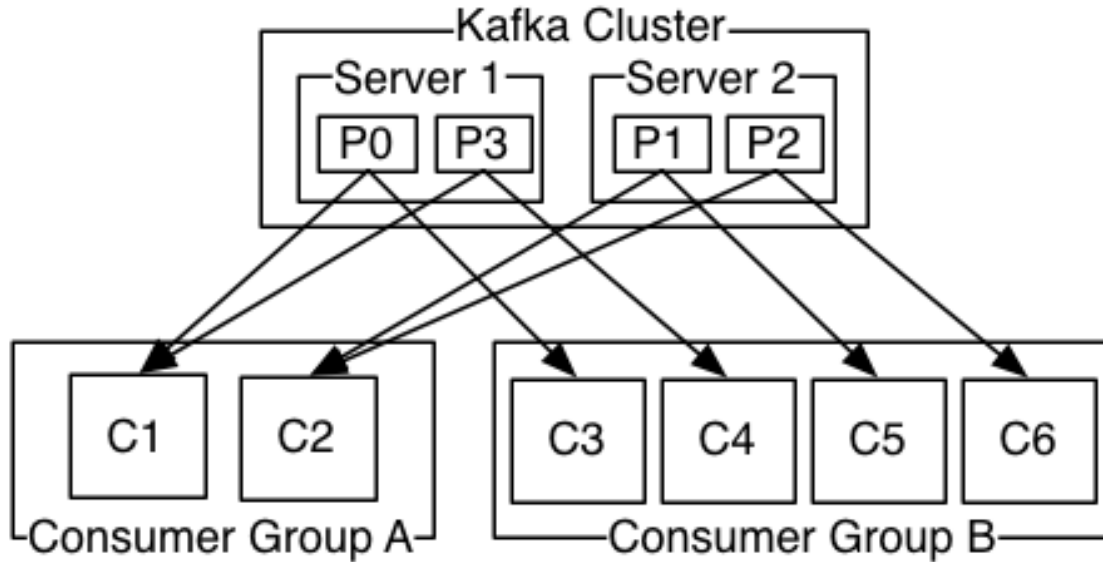
The diagram illustrates the data lifecycle and partitioning for a streaming workload. It shows a sequence of partitions (48 to 123) categorized into 'new', '1 week', and 'del' (deleted) states. Arrows indicate data flow from 'new' partitions to 'real-time', 'batch', 'replay', and 'ad-hoc' processing stages.

| State | Partition Range | Values                            |
|-------|-----------------|-----------------------------------|
| new   | 48 - 54         | 48, 47, 46, 45, 44, 43, 42        |
|       | 129 - 135       | 129, 128, 127, 126, 125, 124, 123 |
|       | del             | (Dashed lines)                    |

Processing Stages:

- real-time
- batch
- replay
- ad-hoc

# Kafka partitions



**A two server Kafka cluster hosting four partitions (P0-P3) with two consumer groups. Consumer group A has two consumer instances and group B has four.**





# Kafka (LinkedIn) : Jay Kreps

- 175 TB of in-flight log data per colo
- Low-latency: ~1.5 ms
- Replicated to each datacenter
- Tens of thousands of data producers
- Thousands of consumers
- 7 million messages written/sec
- 35 million messages read/sec
- Hadoop integration

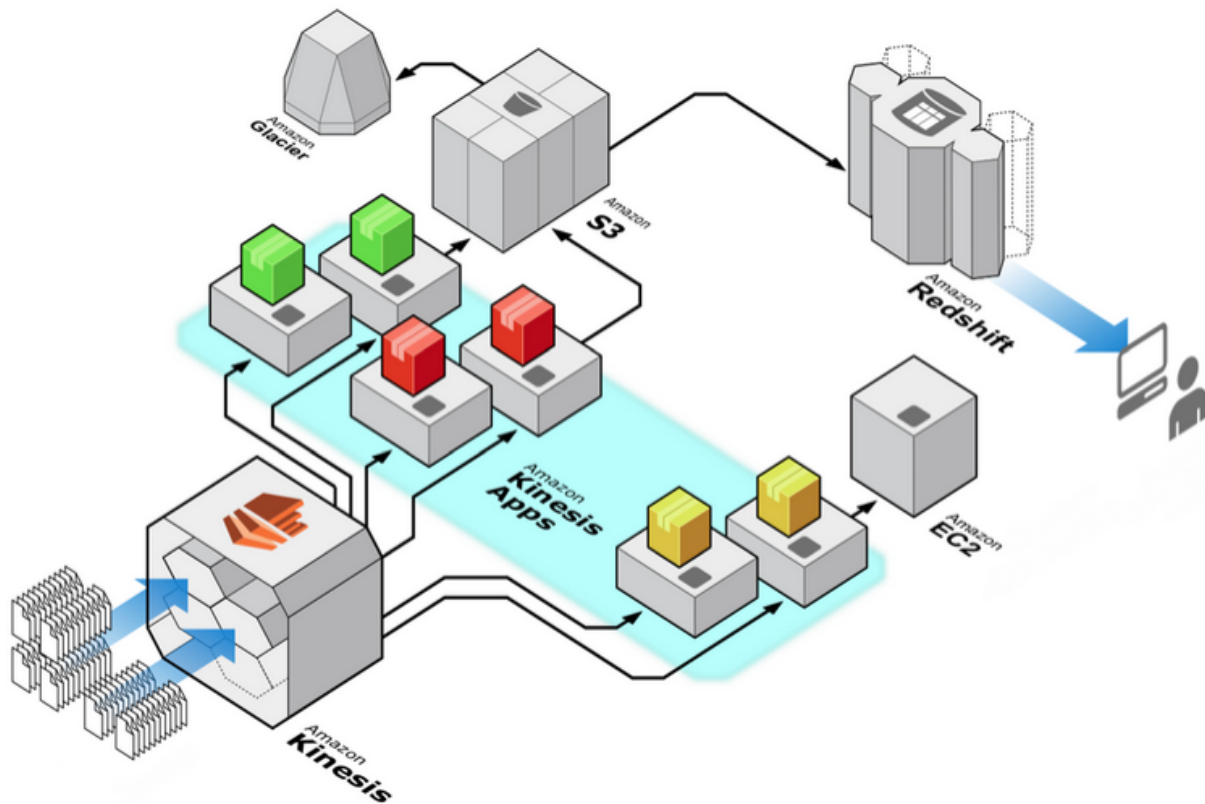
source: Jay Kreps  
on slideshare  
"I ♥ Log"

Real-time Data and Apache Kafka



# Kinesis

# Kinesis : Kafka as a Service



source: <http://aws.amazon.com/kinesis/details/>



# Kinesis design

- Fully (auto-)managed
- Strong durability guarantees
- Stream (= topic)
- Shard (= partition)
- “fast” writers (but ... round-trip 20 ms ?)
- “slow” readers (max 5/s per shard ??)
- Kinesis Client Library (java)



# Kinesis limitations ...

- writing latency (20 ms per entry - replicated)
- 24 hours data retention
- 5 reads per second

<https://brandur.org/kinesis-in-production>

- “vanishing history” after shard split
- “if I’d understood the consequences ... earlier, I probably would have pushed harder for Kafka”



Kinesis consumer with Amazon DynamoDB :: reused from <http://docs.aws.amazon.com/kinesis/latest/dev/kinesis-sample-application.html>





# Problem 1: Distributed state

## Akka

=> state encapsulated in Actors

=> exchange self-contained messages

## Kafka

=> immutable, ordered update queue (Kappa)





# Problem 2: Distributed failure

## Akka

=> explicit failure management (supervisor)

## Kafka

=> partitions are replicated over brokers

=> consumers can replay from log



# Problem 3: Slow consumers

## Akka Streams

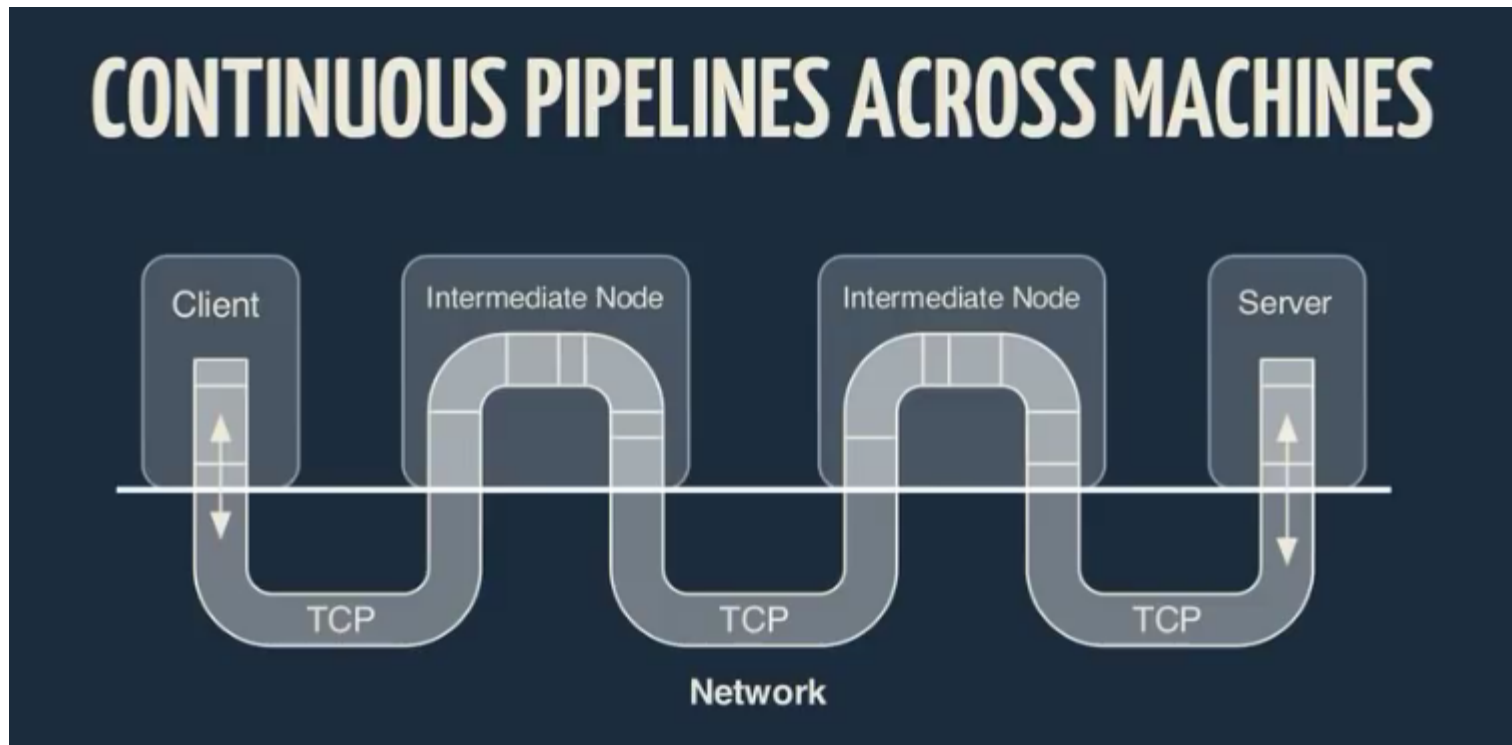
=> automatic back-pressure (avoid overflow)

## Kafka

=> consumers fully decoupled

=> keeps data for 1 week ! (Kinesis: 1 day)

# Avoid overflow in Akka: “tight seals”



source : <https://www.youtube.com/watch?v=o5PUDI4qi10> by @sirthias

# Avoid overflow in Kafka: “big lake”

