## Fast Cars, Big Data How Streaming Can Help Formula 1

Tugdual Grall @tgrall

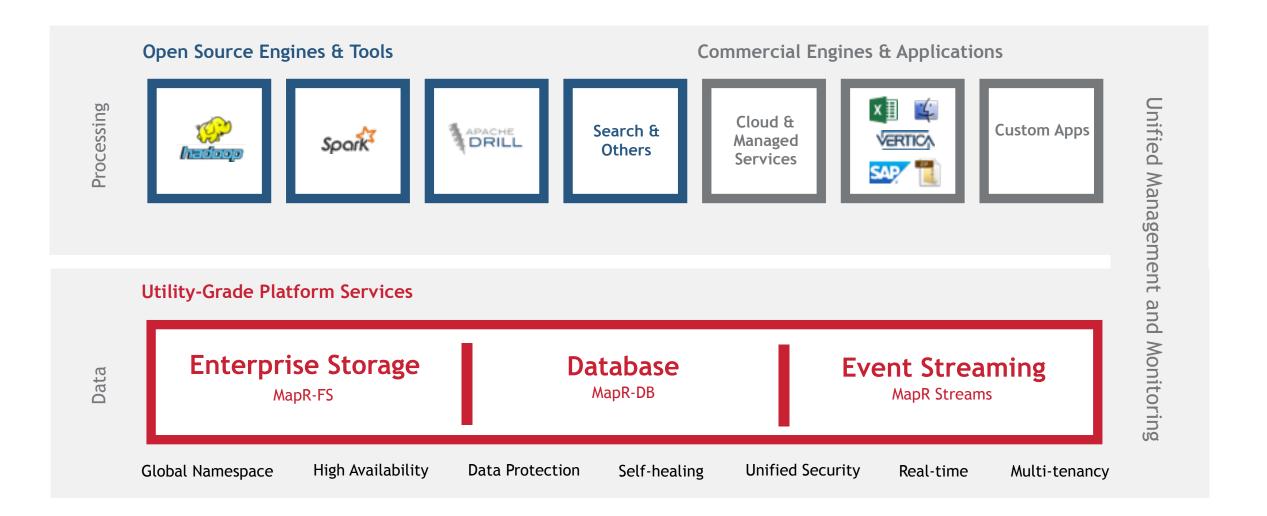
#### {"about" : "me"}

#### **Tugdual "Tug" Grall**

- MapR
  - Technical Evangelist
- MongoDB
  - Technical Evangelist
- Couchbase
  - Technical Evangelist
- eXo
  - CTO
- Oracle
  - Developer/Product Manager
  - Mainly Java/SOA
- Developer in consulting firms

- Web
  - @tgrall
  - http://tgrall.github.io
  - 💆 tgrall
- NantesJUG co-founder
- Pet Project :
  - http://www.resultri.com
- tug@mapr.com
- tugdual@gmail.com

## MapR Converged Data Platform



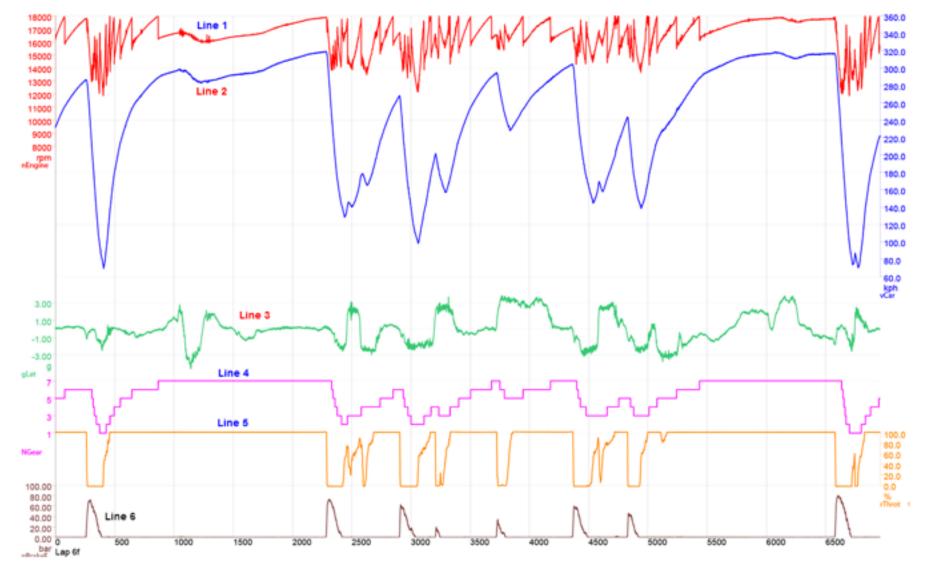
## Agenda

- What's the point of data in motorsports?
- Live demo
- Architecture
- What's next?

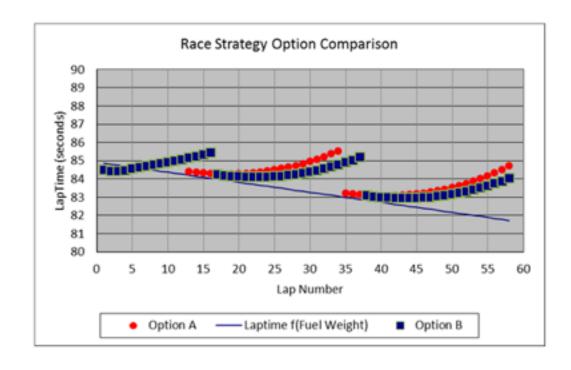
How data plays in F1 motorsports

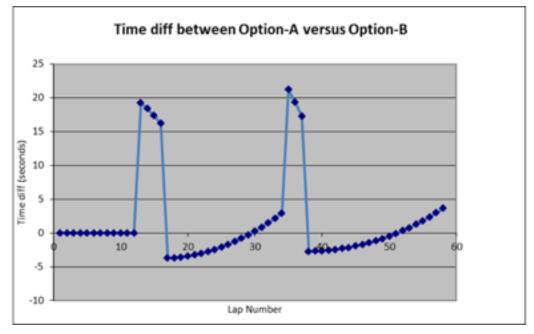


## Data in Motorsports



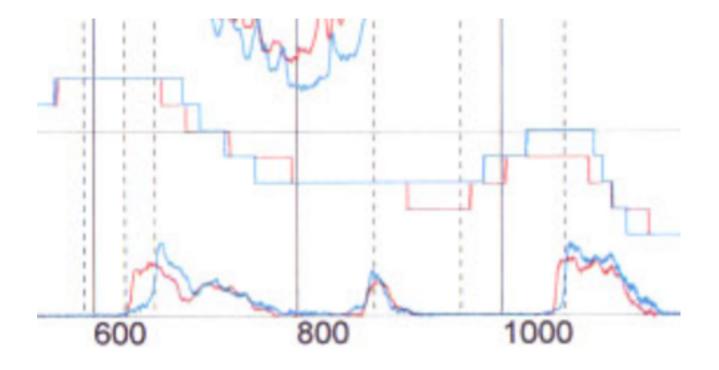
## Race Strategy





F1 Framework - http://f1framework.blogspot.be/2013/08/race-strategy-explained.html





Difference is due to later and sharper braking

- Up to 300 sensors per car
- Up to 2000 channels
- Sensor data are sent to the paddock in 2ms
- **1.5 billions** of data points for a race
- 5 billions for a full race weekend
- 5/6Gb of compressed data per car for 90mn

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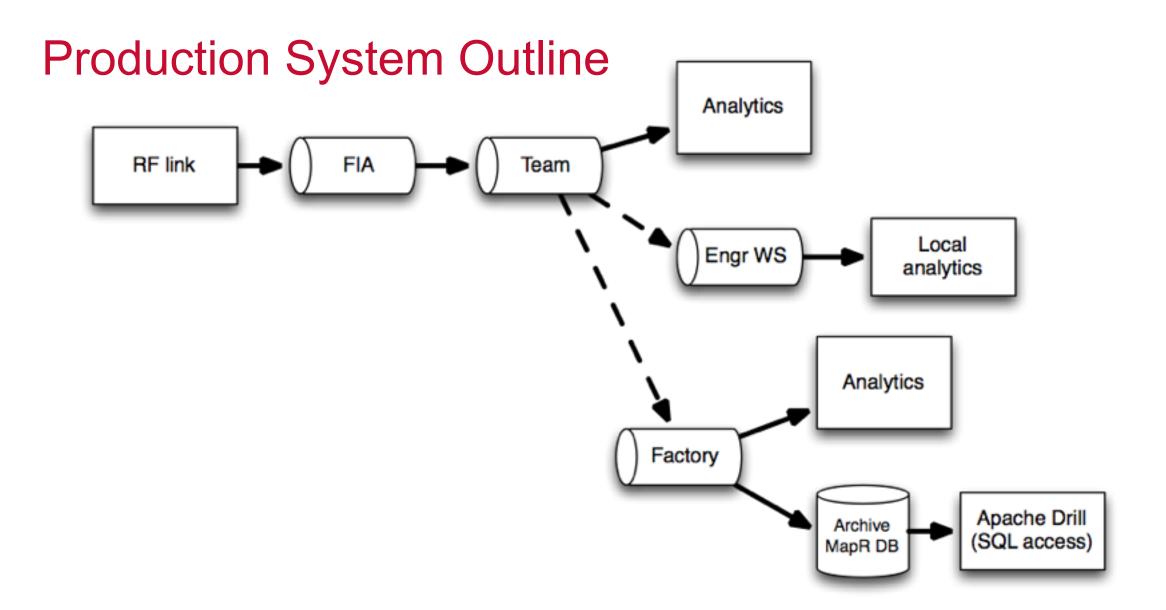
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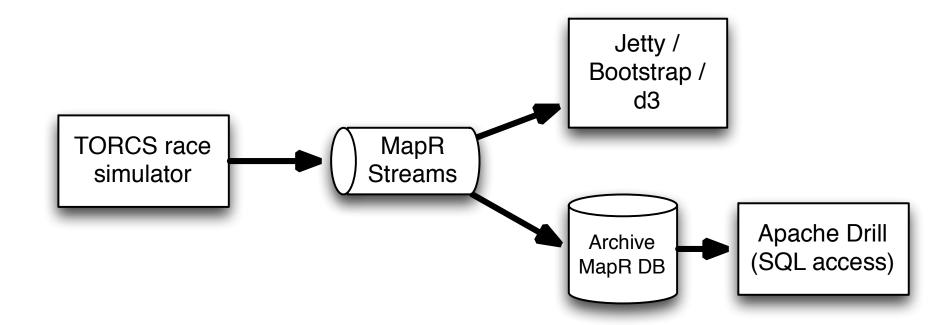
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US Grand Prix 2014: 243 Tb (race teams combined)

So how does that work? Especially for real-time data?



## Simplified Demo System Outline



## TORCS for Cars, Physics and Drivers

The Open Racing Car Simulator http://torcs.sourceforge.net/

Car racing game Al & Research Platform

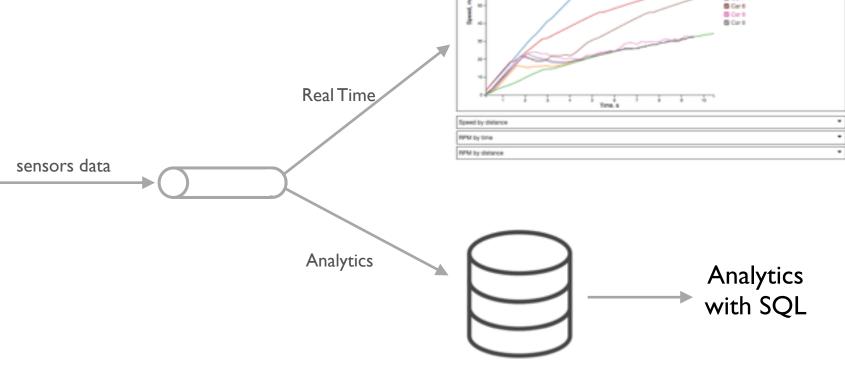




## Demo time!

# IoT: Racing Cars





Speed by time

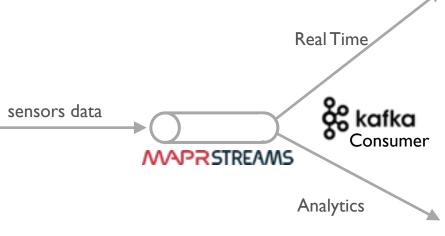
**Events Producers** 

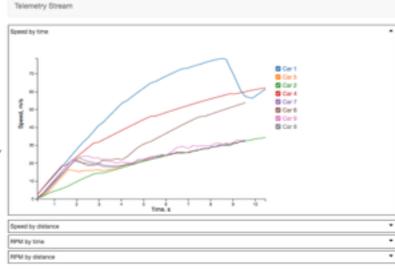
**Events Consumers** 

S Car 2 Car 4

## IoT: Racing Cars









& kafka Producer

**Events Producers** 

**Events Consumers** 

Where/How to store data ?

## **Big Datastore**



Distributed File System HDFS/MapR-FS



NoSQL Database HBase/MapR-DB



# Data Streaming

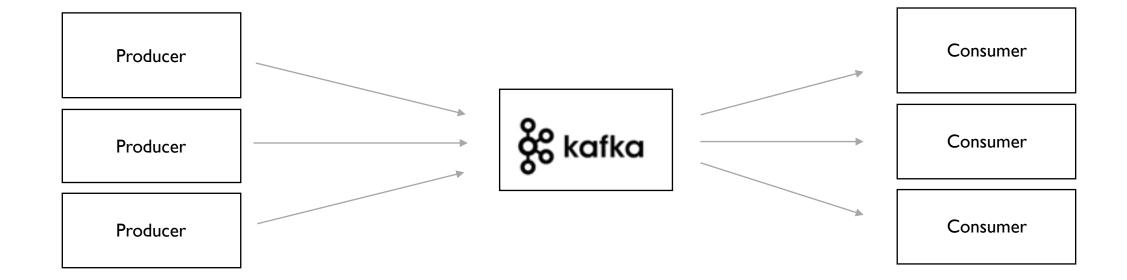
Data Streaming Moving millions of events per h mn s

#### What is Kafka?

- http://kafka.apache.org/
- Created at LinkedIn, open sourced in 2011
- Implemented in Scala / Java
- Distributed messaging system built to scale

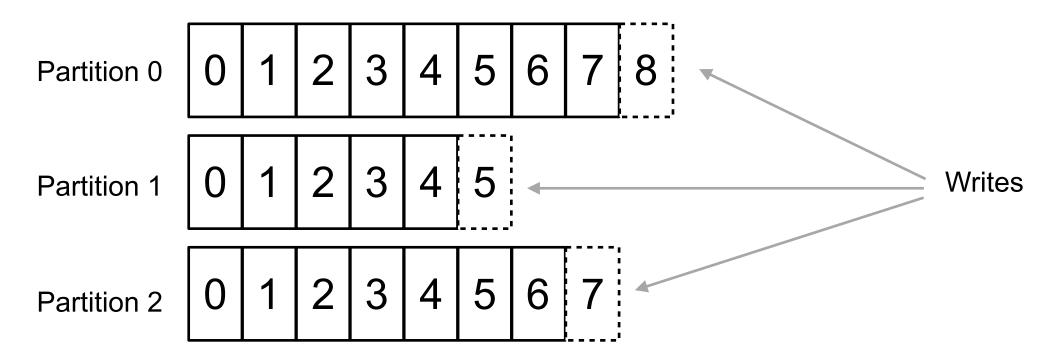


# Big Picture



## **Topics and Partitions**

Split topics into partitions for scalability

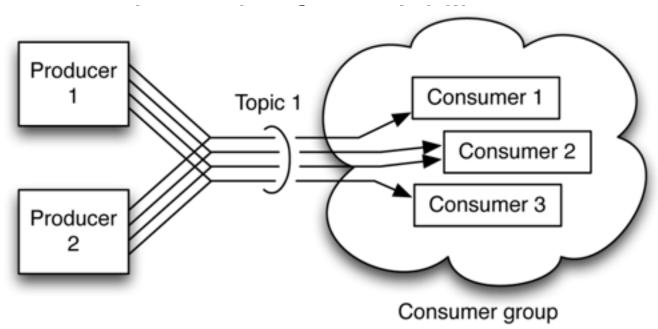


## Consumer Groups

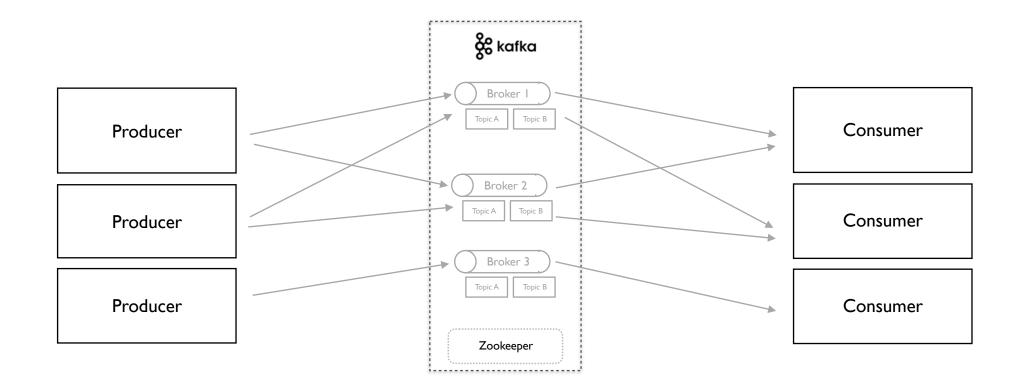
Single cor

Max 1 cor

Any numb



#### More real life Kafka ...



# **MAPRISTREAMS**

- Distributed messaging system built to scale
- Use Apache Kafka API 0.9.0
- No code change
- Does not use the same "broker" architecture
  - Log stored in MapR Storage (Scalable, Secured, Fast, Multi DC)
  - No Zookeeper

## Produce Messages

```
ProducerRecord<String, byte[]> rec = new ProducerRecord<>(
            "/apps/racing/stream:sensors_data",
            eventName,
            value.toString().getBytes());
producer.send(rec, (recordMetadata, e) -> {
if (e != null) { ... });
producer flush();
```



## Consume Messages

```
long pollTimeOut = 800;
while(true) {
    ConsumerRecords<String, String> records = consumer.poll(pollTimeOut);
    if (!records.isEmpty()) {
        Iterable<ConsumerRecord<String, String>> iterable = records::iterator;
        StreamSupport.stream(iterable.spliterator(), false).forEach((record) -> {
           // work with record object
            ... record.value();
        });
        consumer.commitAsync();
```



### What's next?

#### Sensor Data V1

- 3 main data points:
  - Speed (m/s)
  - RPM
  - Distance (m)
- Buffered

```
" id":"1.458141858E9/0.324",
"car" = "car1",
"timestamp":1458141858,
"racetime": 0.324,
"records":
         "sensors":{
            "Speed":3.588583,
            "Distance":2003.023071,
            "RPM":1896.575806
         "racetime": 0.324,
         "timestamp":1458141858
         "sensors":{
            "Speed":6.755624,
            "Distance":2004.084717,
            "RPM":1673.264526
         "racetime": 0.556,
         "timestamp":1458141858
      },
```

# Capture more data

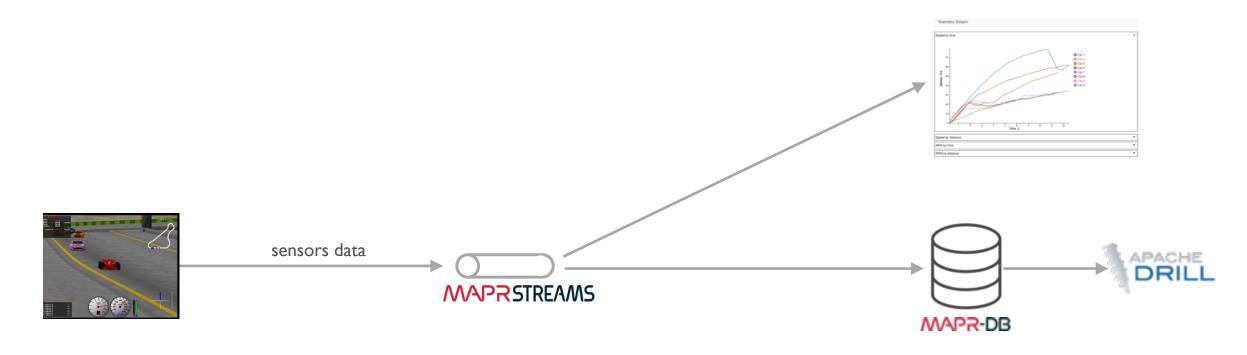
#### Sensor Data V2

- 3 main data points:
  - Speed (m/s)
  - RPM
  - Distance (m)
  - Throttle
  - Gear
  - •
- Buffered

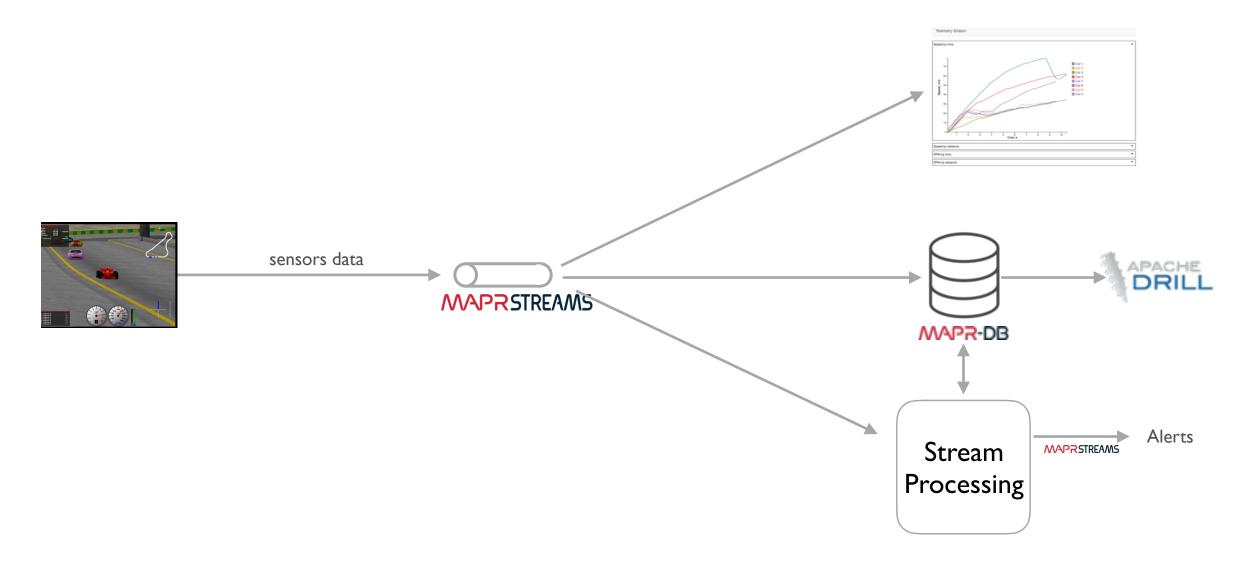
```
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            "gear" : 2
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```

### Add new services

### **New Data Service**



#### **New Data Service**



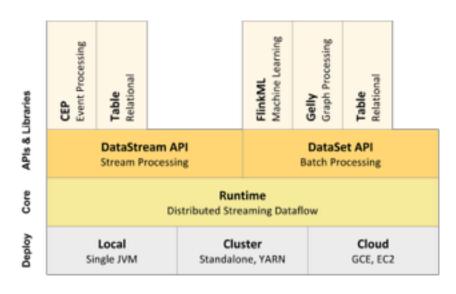


- Cluster Computing Platform
- Extends "MapReduce" with extensions
  - Streaming
  - Interactive Analytics
- Run in Memory
- http://spark.apache.org/





- Streaming Dataflow Engine
  - Datastream/Dataset APIs
  - CEP, Graph, ML
- Run in Memory
- https://flink.apache.org/



#### **DEMO**

Stream Processing with Flink

# Streaming Architecture & Formula 1

Stream data in real time

- Big Data Store to deal with the scale
  - NoSQL Database, Distributed File System
- Decouple the source from the consumer(s)
  - Dashboard, Analytics, Machine Learning
  - Add new use case....

# Streaming Architecture & Formula 1

Stream data in real time



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