# IoT Data Analytics with Lambda Architecture Dec 2018

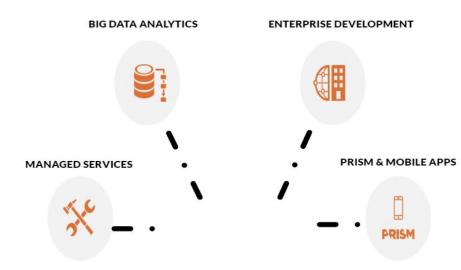




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# **Today's Presenters**

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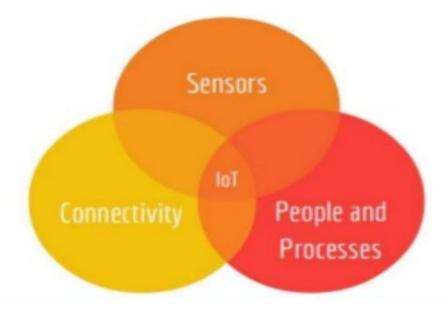
# Agenda

- Quick Recap of previous IoT Meetup
- What's new in this edition?
- Need of IoT Data Analytics and Introduction
- Lambda Architecture along with Use Cases
- Demo 1
- Tools and Frameworks
- Demo 2

# Quick Recap ... What is IoT

The Internet of Things (IoT) concept began as a simple idea of connected devices.

The Internet of Things (IoT) is the network of physical objects—devices, vehicles, buildings and other items embedded with electronics, software, sensors, and network connectivity—that enables these objects to collect and exchange data.





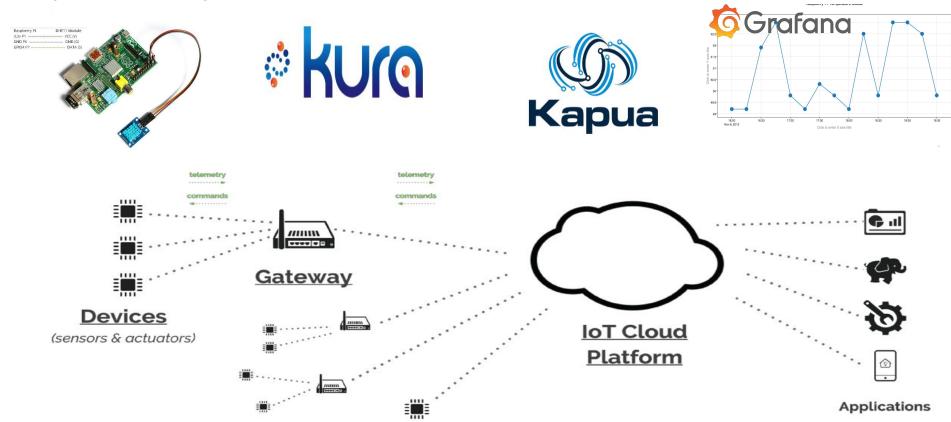
# Quick Recap ... What is IoT



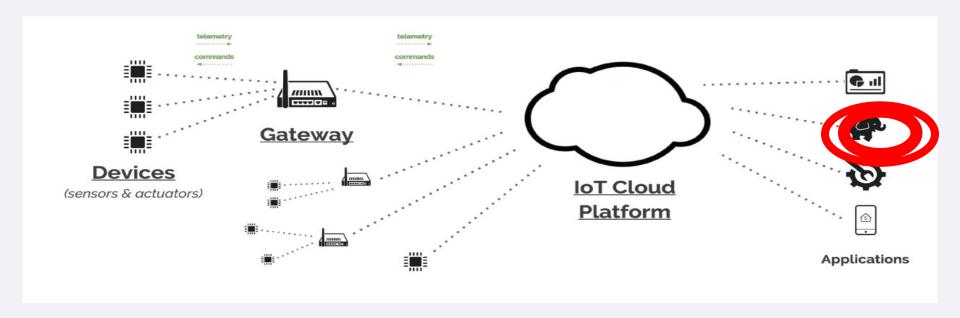
Courtesy: YouTube (https://www.youtube.com/watch?v=bCxZsexIUI0)

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# Quick Recap ... Architecture for our IoT Use Case



# What's new in this edition?





# **Need of IoT Data Analytics**

- **Analytics** rather than hardware novelties
- Insights & Predictions
- Variety of use cases
- Detect Patterns
- New business opportunities

# **IoT Analytics Platform Built – Intro**

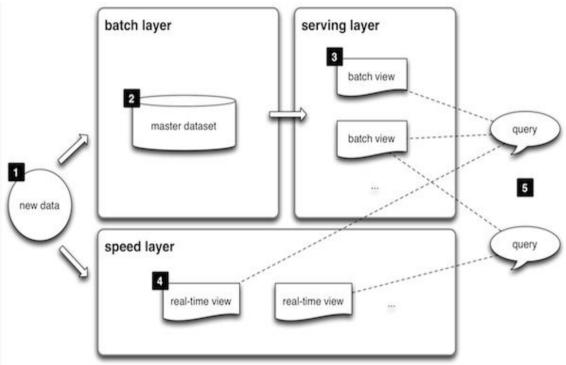
Choosing right Software Components

Handling Live Streaming and Historical data

#### Lambda Architecture

In order to accommodate the demand for real-time analytics, we need to design a system that can provide balance between the concept of "single version of truth" and "real-time analytics". Lambda Architecture is one such method. Lambda Architecture was invented by Nathan Marz to provide fault-tolerant and scalable data processing architecture.

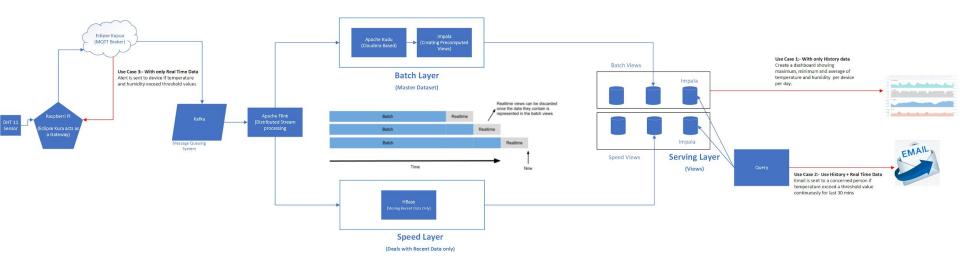
### Lambda Architecture



- All data entering the system is dispatched to both the batch layer and the speed layer for processing.
- The **batch layer** has two functions: (i) managing the master dataset (an immutable, append-only set of raw data), and (ii) to pre-compute the batch views.
- The serving layer indexes the batch views so that they can be queried in low-latency, ad-hoc way.
- The speed layer compensates for the high latency of updates to the serving layer and deals with recent data only.
- Any incoming query can be answered by merging results from batch views and real-time views.

Image source: Internet

# Design



# Components - What fits where and how?

#### Apache Kafka

- Kafka is a high-throughput, low-latency platform for handling real-time data feeds.
- An utility which reads data from Kapua and publish to a topic on Kafka.

#### **Apache Flink**

- It's a framework and distributed processing engine for **stateful computations** over unbounded and bounded data streams.
- Flink has been designed to run in all common cluster environments, perform computations at **in-memory** speed and at any scale.

#### Apache Kudu

- Kudu has tight integration with **Apache Impala**, allowing you to use Impala to insert, query, update, and delete data from tables using Impala's SQL syntax
- This is an alternative to using the Kudu APIs to build a custom Kudu application.

#### **Apache Hbase**

• It is mainly used for **ad hoc read** which is retrieved from most recent data entered into the system.



# **Use Cases**



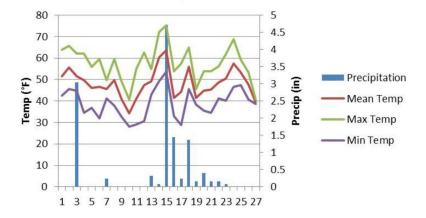


# Use Case 1 -

#### **Problem Statement: -**

'Show maximum, minimum and average temperature/humidity per device per day.'

**Data:-** History Data

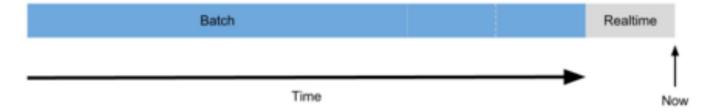


### Use Case 2 -

#### **Problem Statement: -**

'As per recent research in healthcare sector; an ill person exposed to temperature above 30 degree Celsius for half an hour then there are high chances of that person getting affected with health issues. Hence we need to continuously monitor the temperature for 30 mins and if it is above the set threshold then respective team is notified'

Data:- History + Real Time Data



# Use Case 3 -

#### **Problem Statement: -**

'If a temperature and humidity coming from sensor rise above threshold value for a particular device then raise an alarm'

Data:- Real Time Data





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# **Technology Stack**

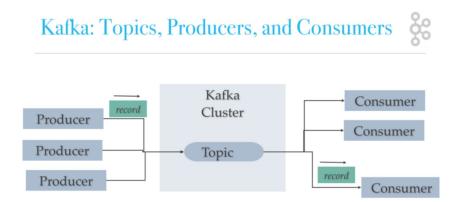


# **Apache Kafka**



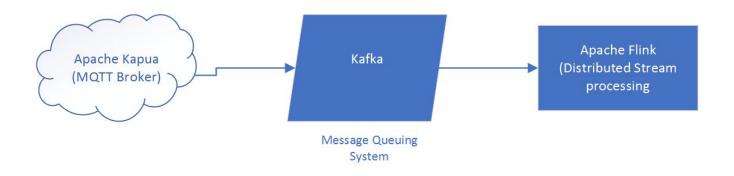


- Kafka is a publish-subscribe, distributed, message queuing system
- It allows users to publish messages to a certain topic; this is then distributed to the subscribers of the topic.
- All messages written to Kafka are persisted and replicated to peer brokers for fault tolerance
- Messages can stay for a configurable period of time (i.e., 7 days, 30 days, etc.)





### Role of Kafka in our design:



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# **Apache Flink**



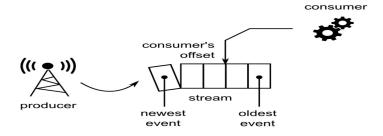


Apache Flink is an 'Open Source Stream Processing Framework'

A stream is a sequence of data elements made available over time

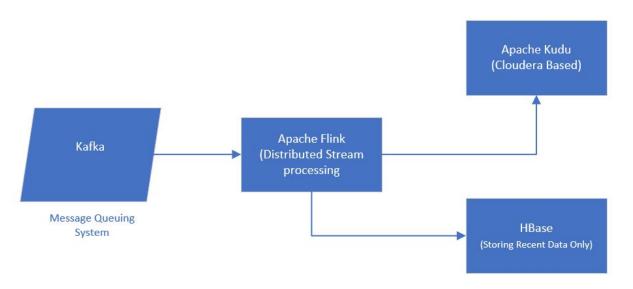
#### Streams can be :-

- Unbounded have a start but no defined end (streaming application, e.g. twitter feed)
- Bounded have a defined start and end (batch processing e.g. log files)





# Role of Flink in our design:



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#### Flink Features:

High Throughput and Low Latency :-

#### **Latency Bound Buffering:-**

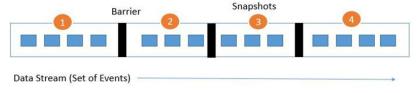
- Sending individual records over a network causes significant overhead.
- Buffering is the option.
- Disadvantage of buffering is that it adds latency because records are collected in a buffer instead of being immediately shipped.
- Flink ensures that each buffer is shipped after a certain period of time regardless of how much it is filled.



#### Flink Features Cont..

#### 2. Fault Tolerance :-

- Check pointing is Flink's backbone for providing consistent fault tolerance
- It keeps on taking consistent snapshots
- In case of any failure, Flink stops the executors and resets them and starts executing from the latest available checkpoint.



Flink Check Pointing & Barriers

- Stream barriers are core elements of Flink's snapshots.
- They are ingested into data streams without affecting the flow.
- Each barrier carries a unique ID. They group sets of records into a snapshot.
- Flink processes each record exactly once.



#### Flink Features Cont...

#### 3. Flexible streaming windows

- Flink supports data-driven windows
- Can design a window based on time, counts, or sessions.
- A window can also be customized which allows us to detect specific patterns in event streams.

#### 4. Stream and Batch in One Platform

- Flink provides APIs for both batch and stream data processing
- So once you set up the Flink environment, it can host stream and batch processing applications easily
- Considers batch processing as the special case of streaming.

### Flink DataStream API

```
DataStream<String> lines = env.addSource(
                                                                         Source
                                  new FlinkKafkaConsumer<>(...);
DataStream<Event> events = lines.map((line) -> parse(line));
                                                                         Transformation
DataStream<Statistics> stats = events
         .keyBy("id")
                                                                         Transformation
         .timeWindow (Time.seconds (10))
         .apply(new MyWindowAggregationFunction());
stats.addSink(new RollingSink(path));
                                                                         Sink
                                  Transformation
              Source
                                                                   Sink
                                     Operators
                                                                 Operator
             Operator
                                                keyBy()/
       Source
                            map()
                                                window()/
                                                                       Sink
                                                 apply()
                                     Stream
                                Streaming Dataflow
```

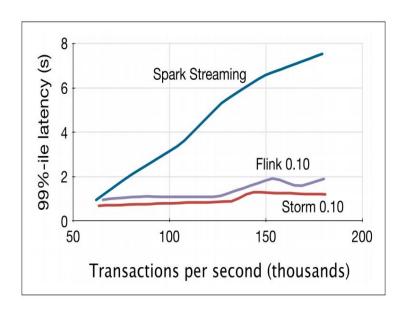
# Flink Vs Spark





Computation Model	Real time/ Event based	Near Real Time/ Micro batches
Abstraction	Batch Over Streaming	Streaming Over Batch
API	Dataset – Batch API DataStream – Streaming API	RDD – Batch API Dstream – Streaming API
Window Criteria	Record Based or User Defined	Time Based
Memory Management	Automatic	Configured
Fault Tolerance Guarantee	Exactly Once	

# Flink Performance: The Yahoo! Streaming Benchmark



#### Conclusion:-

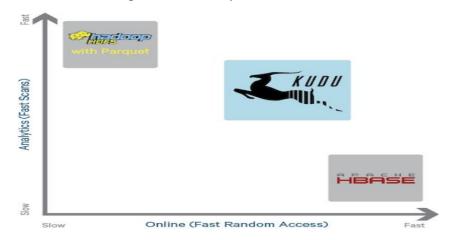
- Storm and Flink show low latencies at relatively high throughputs.
- Spark streaming supports high throughputs, but at a relatively higher latency.
- After Flink was rewritten with its own Window Function then it gave much more throughput that Storm.
- Flink runs with exactly once guarantees, Storm with at least once.

# **Apache Kudu**





- Kudu is an open source storage engine for structured data which supports -
  - efficient analytical access patterns
  - low-latency random access
- Kudu: Like HDFS and HBASE in one.
- Storage for Fast Analytics on Fast Data



#### Why Kudu

- Fast Columnar Data storage
- Fast Inserts/Updates
- Low-latency random access
- Tabular, SQL like schema
- As per TPC-H (Analytics benchmark), Kudu works well in querying data with low latency than HDFS\*

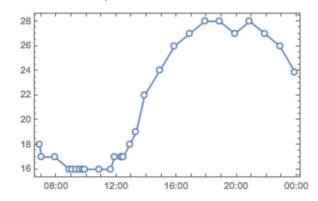
<sup>\*</sup>https://kudu.apache.org/kudu.pdf



#### **Kudu Use Case**

Kudu is best for use cases requiring a simultaneous combination of sequential and random reads and writes

Time Series
 Examples: Stream market data; fraud detection & prevention; network monitoring
 Workload: Insert, updates, scans, lookups





# Role of Kudu in our design:



# **Apache HBase**

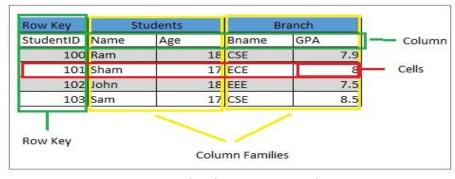




#### HBase is a NoSQL database used for real-time data streaming

#### **HBase Unique Features**

- A column family oriented database
- HBase is built for low latency operations
- Used extensively for random read and write operations
- HBase stores large amount of data in terms of tables



Sample Hbase Storage Schema

#### Why to Choose HBase?

e.g. A table for a popular web application may consist of billions of rows. If we want to search particular row from such huge amount of data, HBase is the ideal choice as query fetch time in less.



#### Where is HBase used?

#### **Telecom Industry**

**Problem Statement:** 

- Storing billions of CDR (Call detailed recording) log records generated by telecom domain
- Providing real-time access to CDR logs and billing information of customers considering cost effectiveness

#### Solution:

HBase is used to store billions of rows of call detailed records.

If 20TB of data is added per month to the existing RDBMS database, performance will deteriorate.

To handle a large amount of data in this use case, HBase is the best solution. HBase performs fast querying and display records.

That apart, HBase can be used -

- Whenever there is a need to write heavy applications.
- Performing online log analytics and to generate compliance reports.



### Role of HBase in our design:





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# What's Next....

Migration of this implementation to Google Cloud Platform (GCP)

Add Machine Learning component

IoT Devices and Communication Security

# Thank You!



### References

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