

# An Open-Source Streaming Machine Learning and Real-Time Analytics Architecture

Using an IoT example

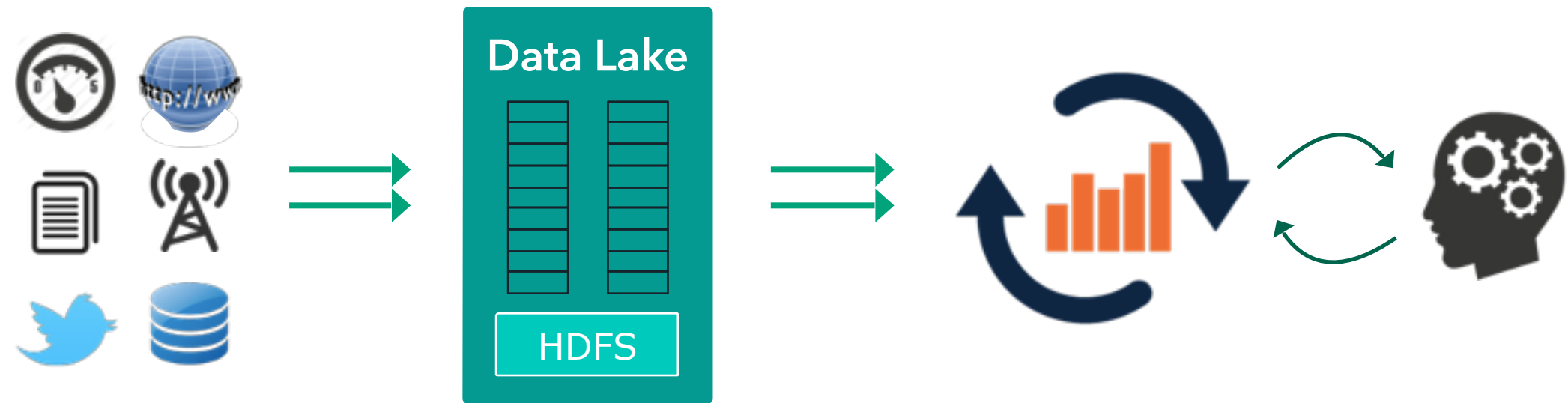


Fred Melo  
 @fredmelo\_br

William Markito  
 @william\_markito



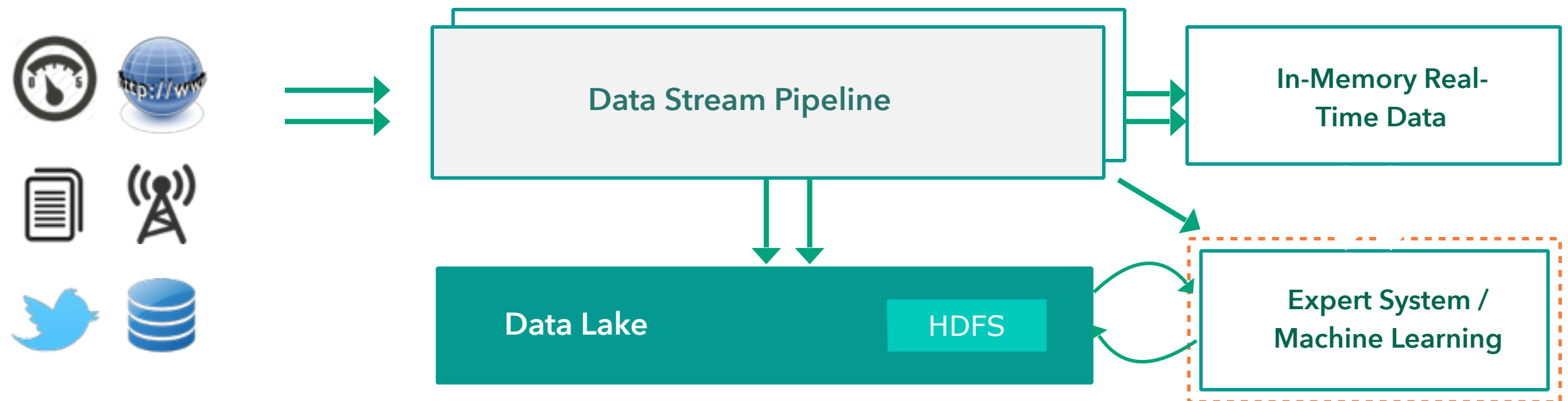
# Traditional Data Analytics - Limitations



*No real-time  
information  
ETL based  
Data-source specific*

*Hard to change  
Labor intensive  
Inefficient*

# Stream-based, Real-Time Closed-Loop Analytics

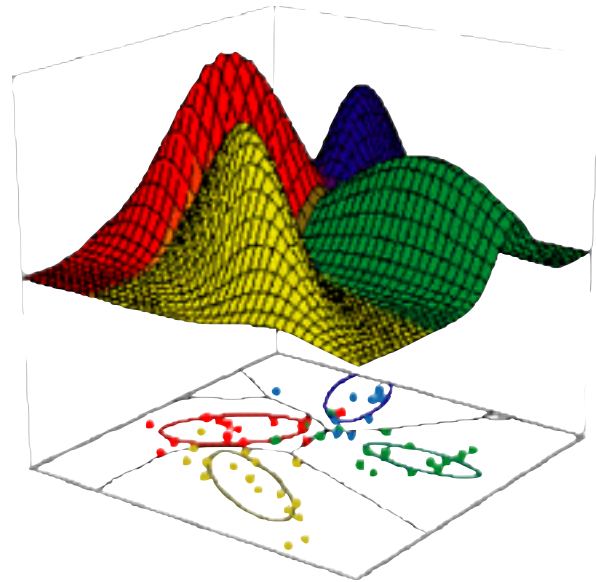


*Multiple Data  
Sources*

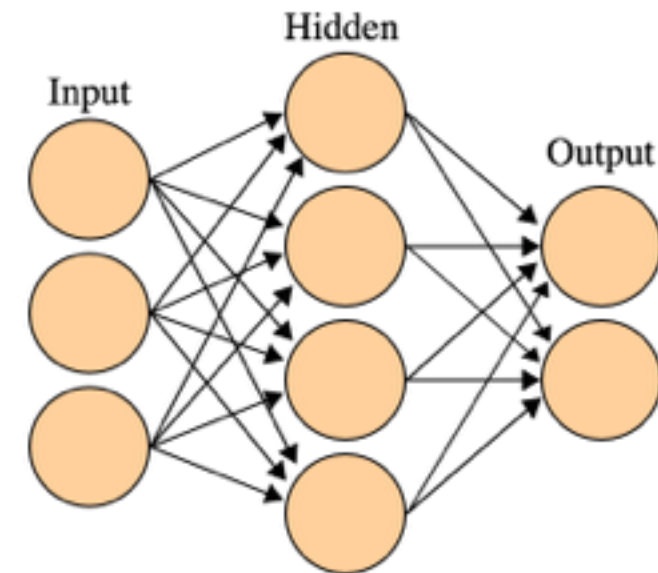
*Real-Time Processing  
Store Everything*

*Continuous Learning  
Continuous  
Improvement  
Continuous Adapting*

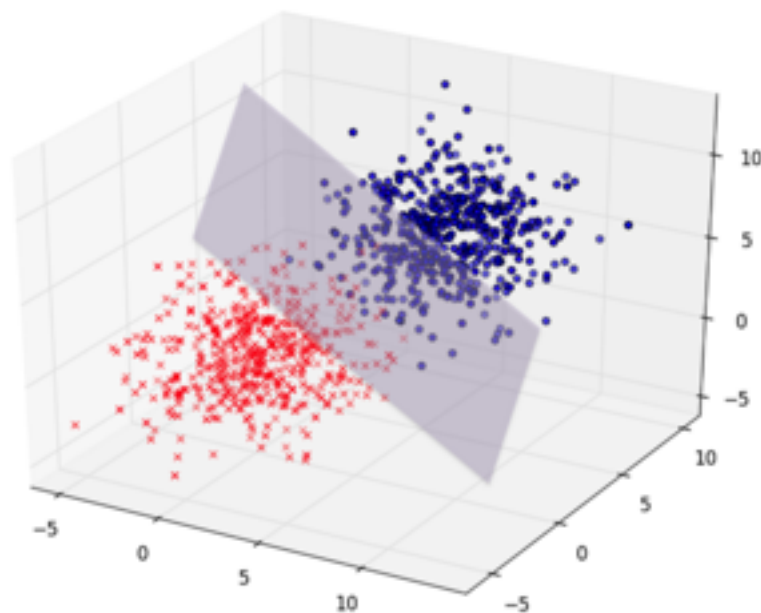
# Machine Learning and Smart Systems



Bayesian Methods



Neural Networks



Classifiers



Genetic Algorithms

# A Streaming Machine Learning for IoT Example

## Predictive Maintenance Scenario



Sensor Data

Real-Time

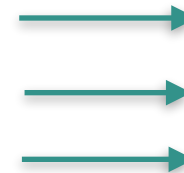


Evaluates **LIVE DATA**

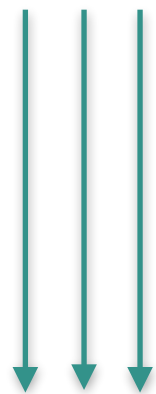
"According to historical trends, there's an 80% chance this equipment would fail in the next 12 hours"



Smart System



Live data becomes historical over time



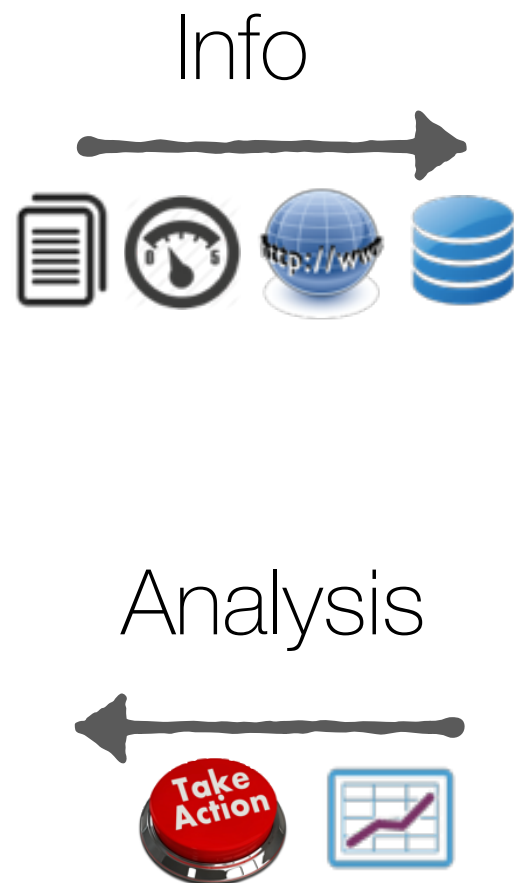
Historical



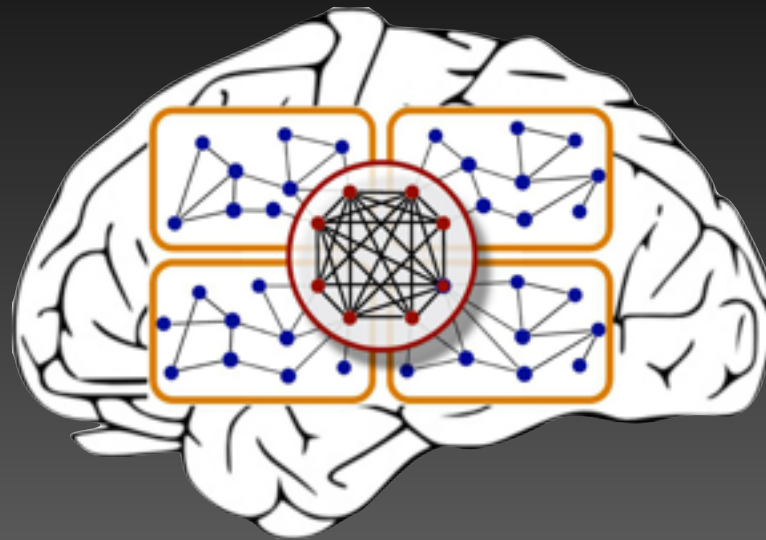
Learns with **HISTORICAL TRENDS**

"How were the temperature and vibration sensors reading when the latest failures happened? "

# Streaming Machine Learning



Machine Learning



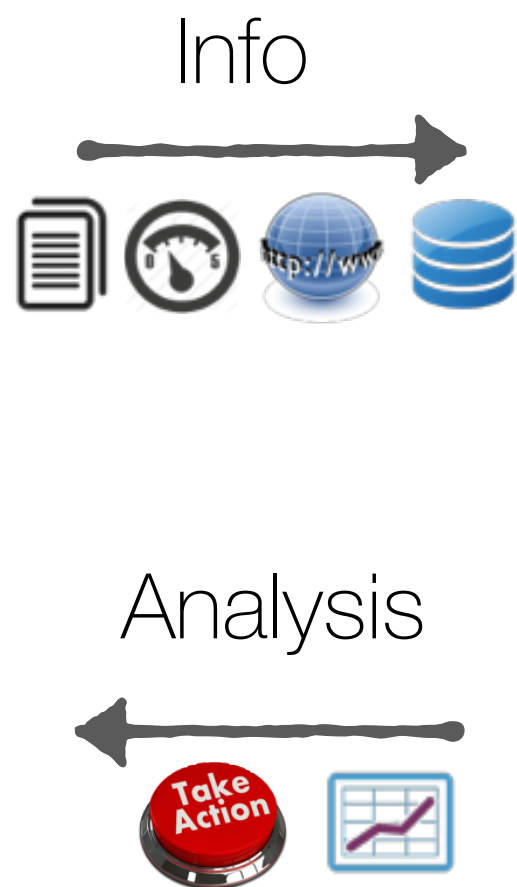
Look at past trends  
(for similar input)

Evaluate current input

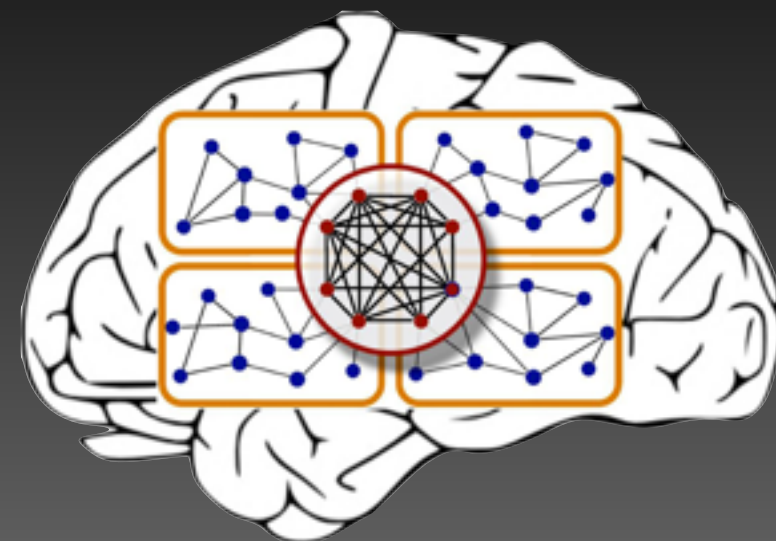
Score / Predict



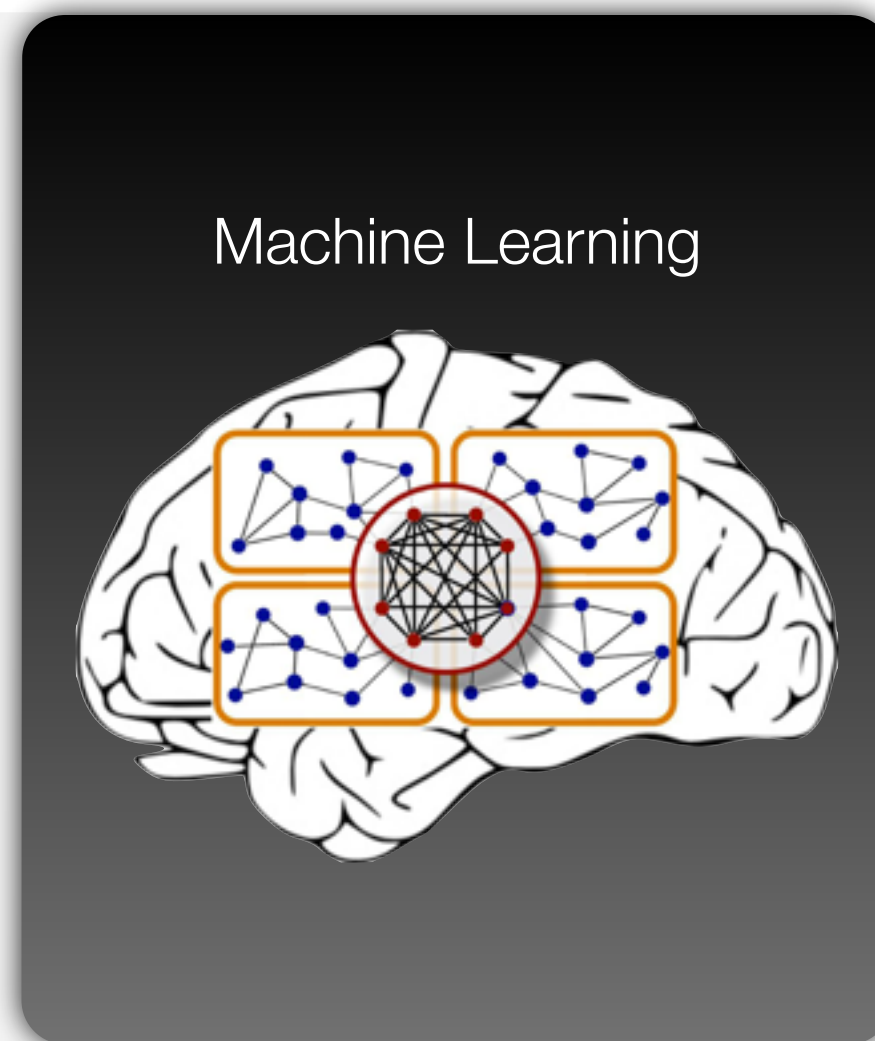
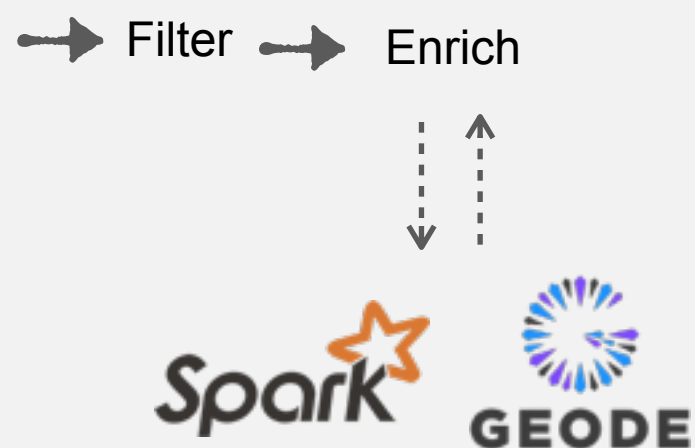
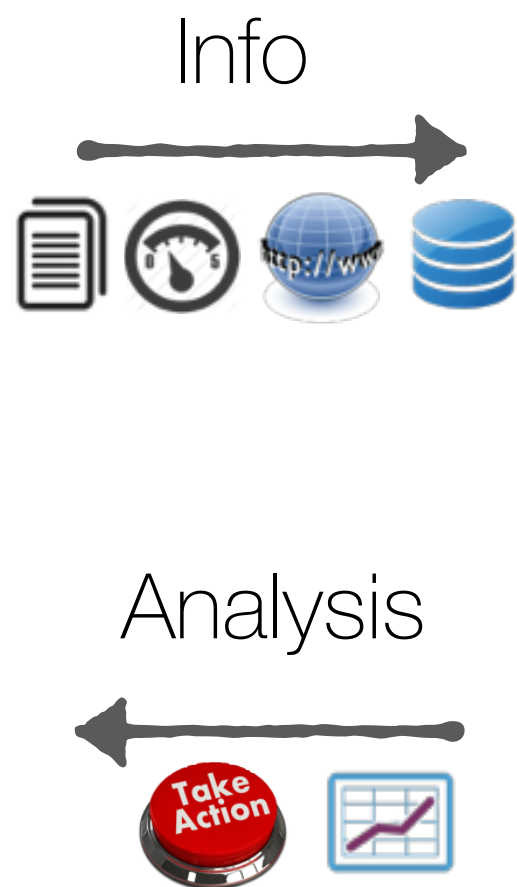
# Streaming Machine Learning



Machine Learning

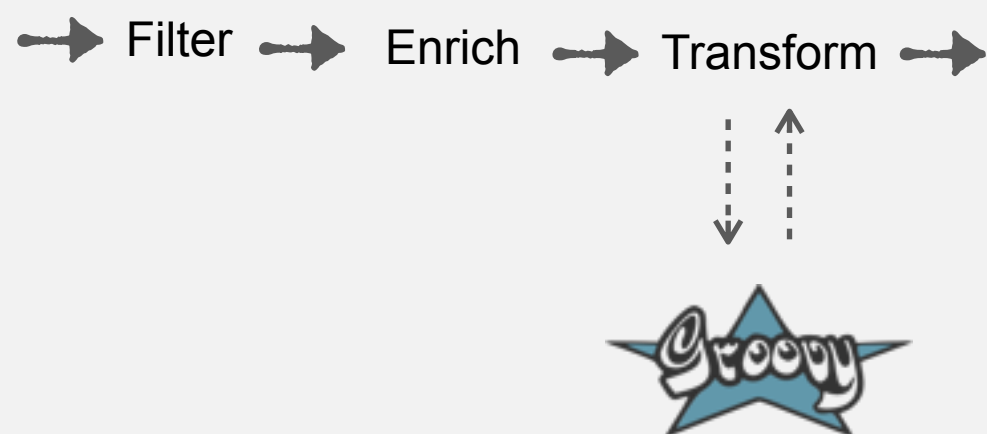
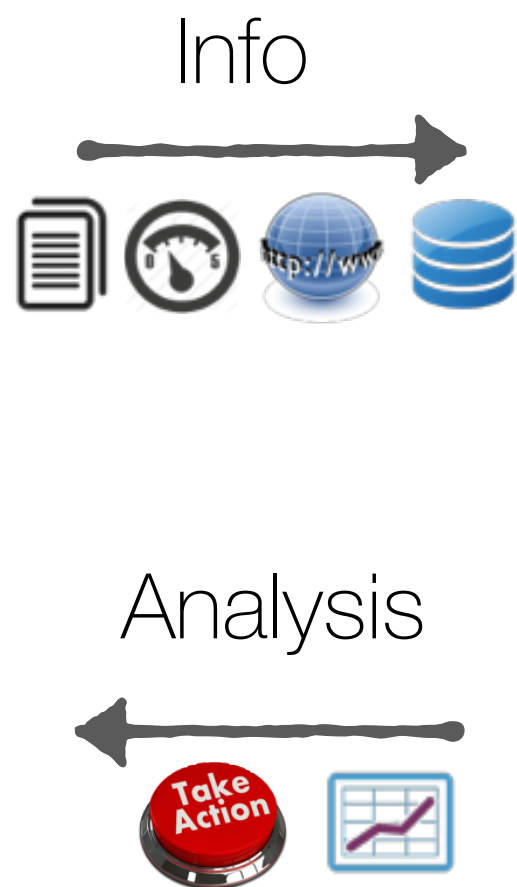


# Streaming Machine Learning

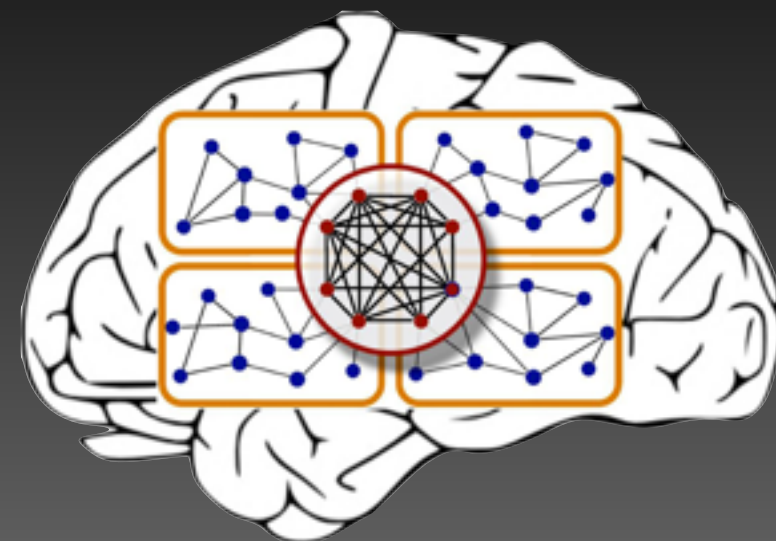




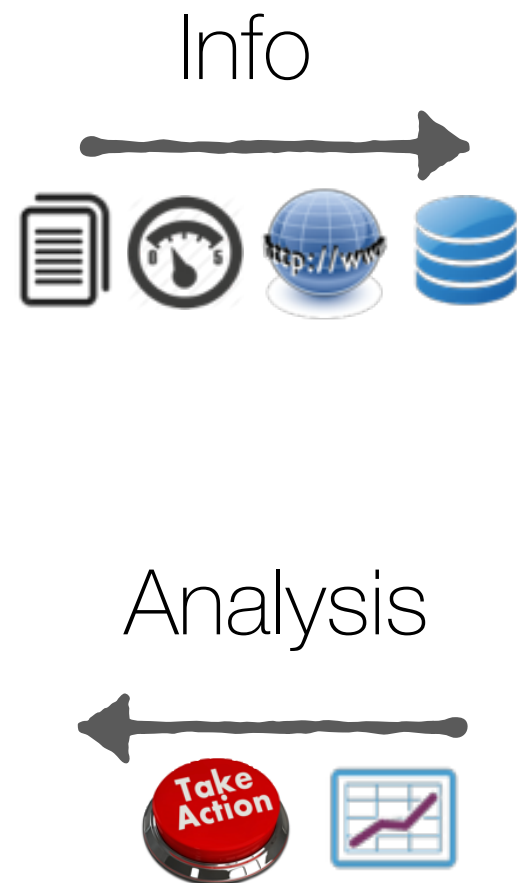
# Streaming Machine Learning



Machine Learning

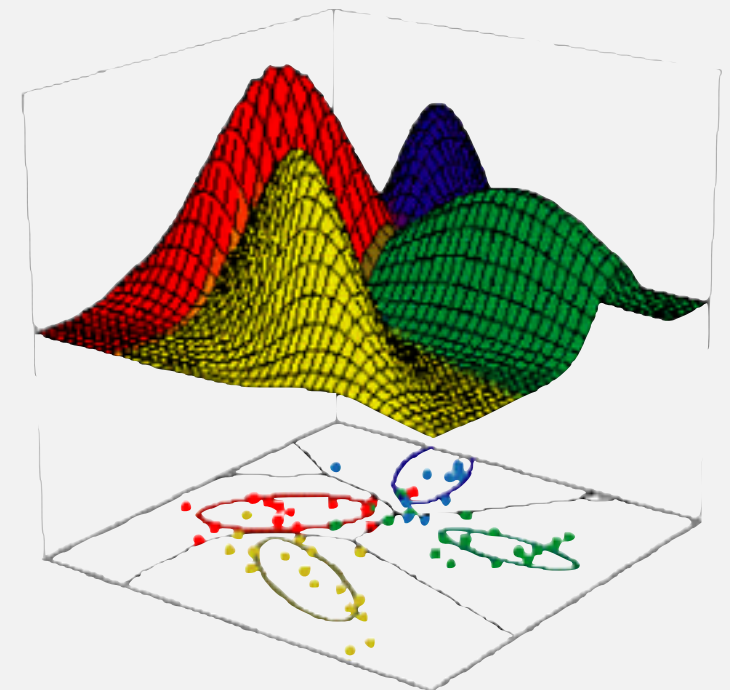


# Streaming Machine Learning

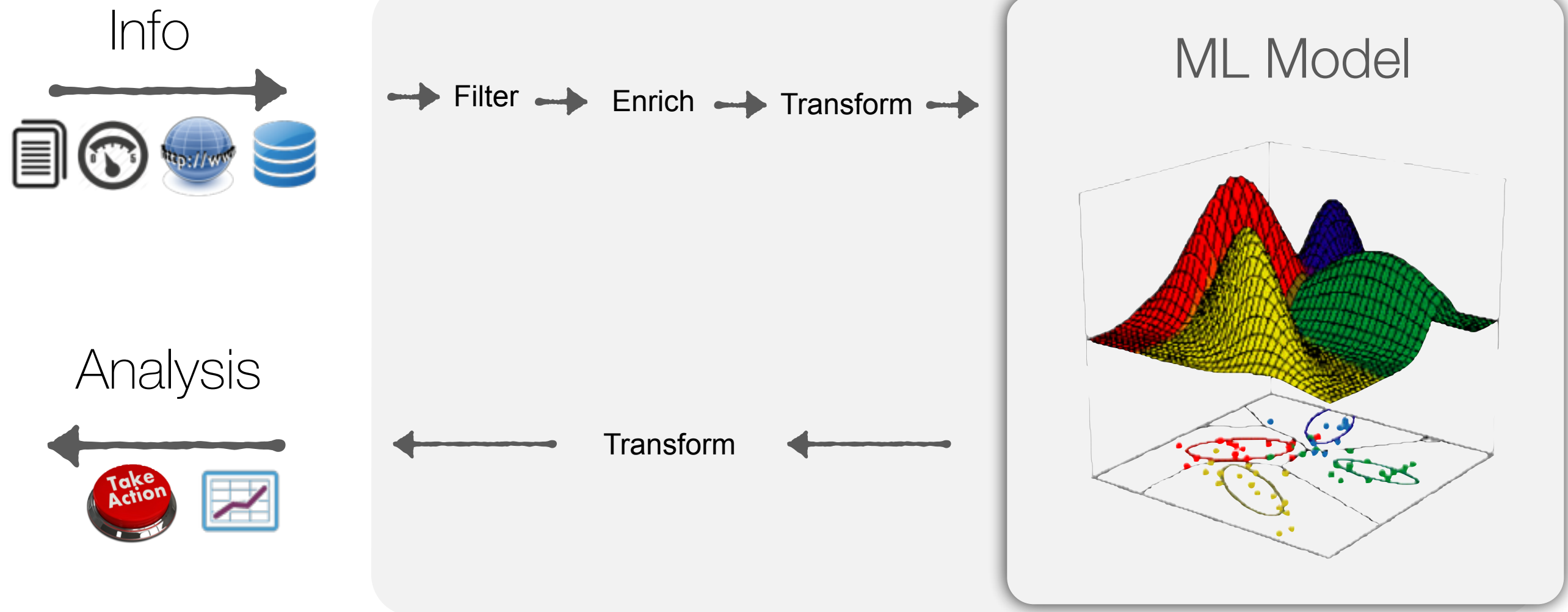


→ Filter → Enrich → Transform →

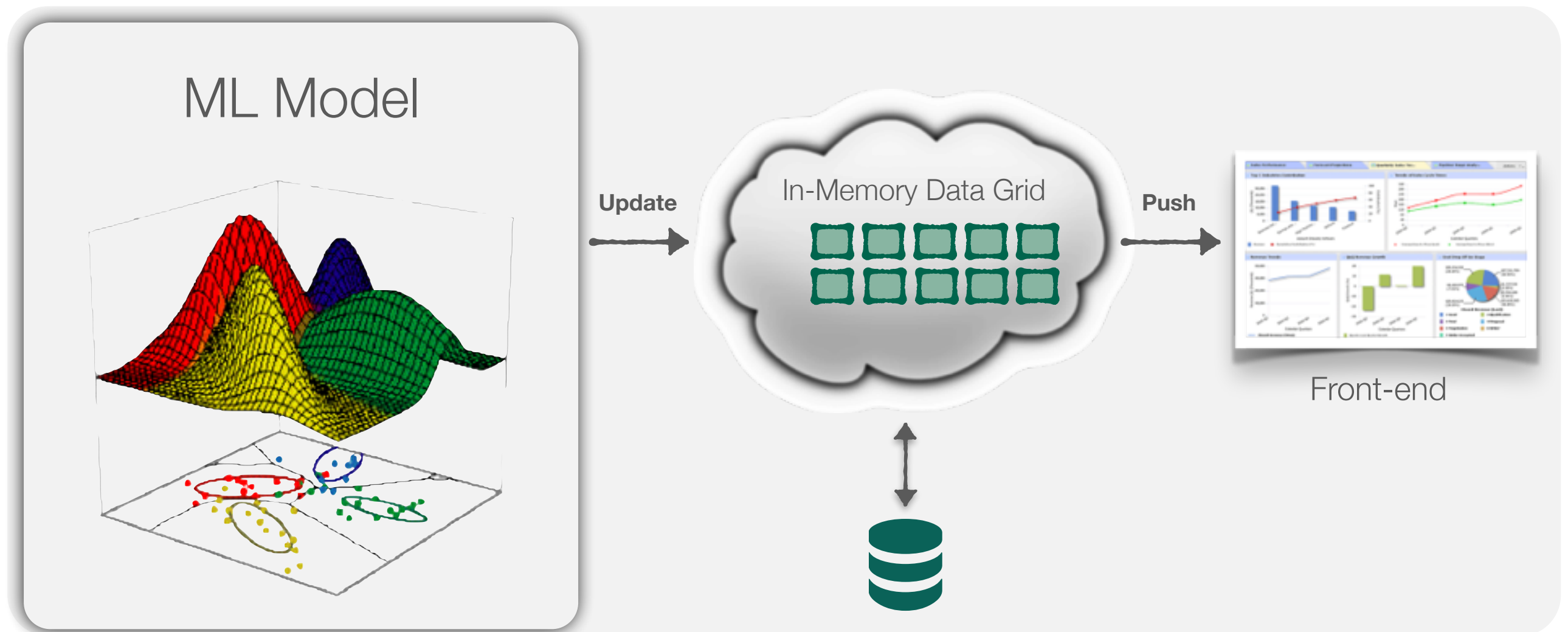
ML Model



# Streaming Machine Learning

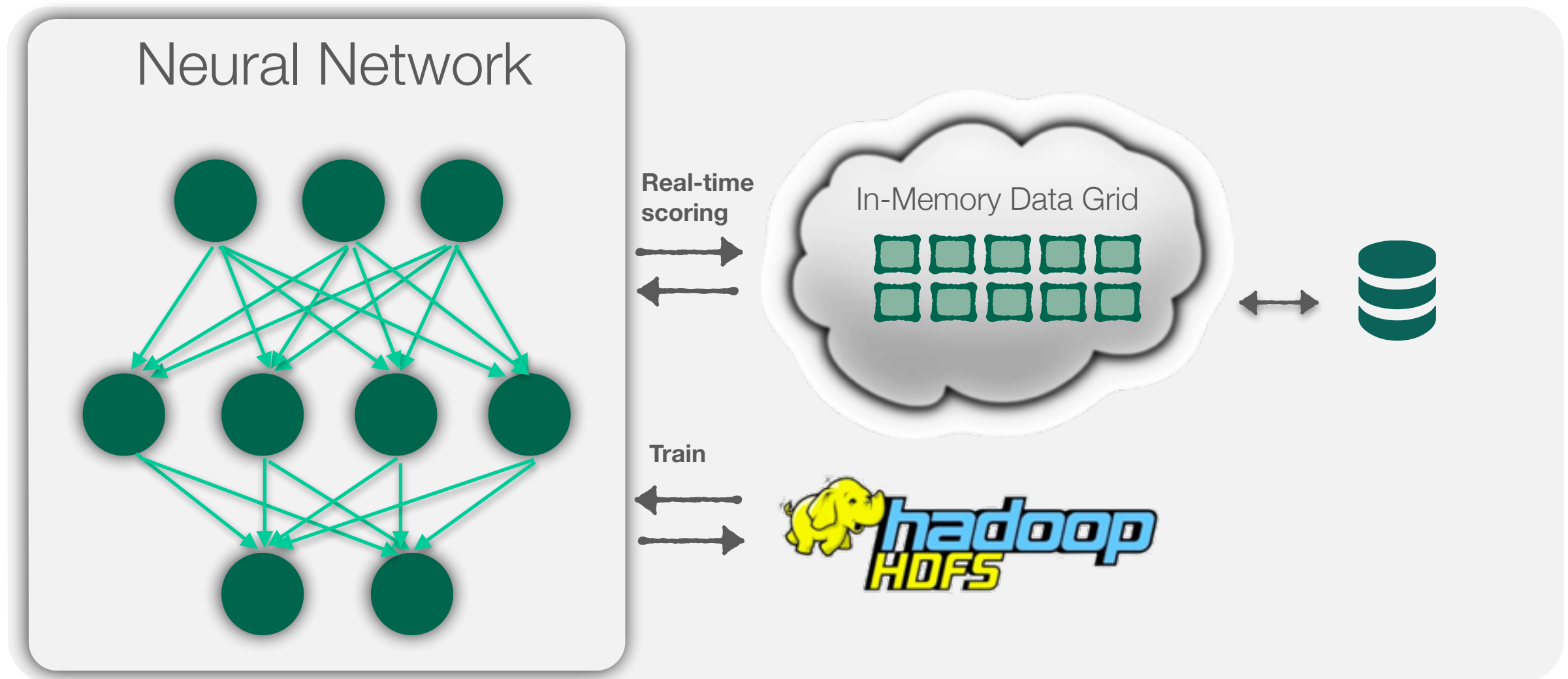


# Streaming Machine Learning

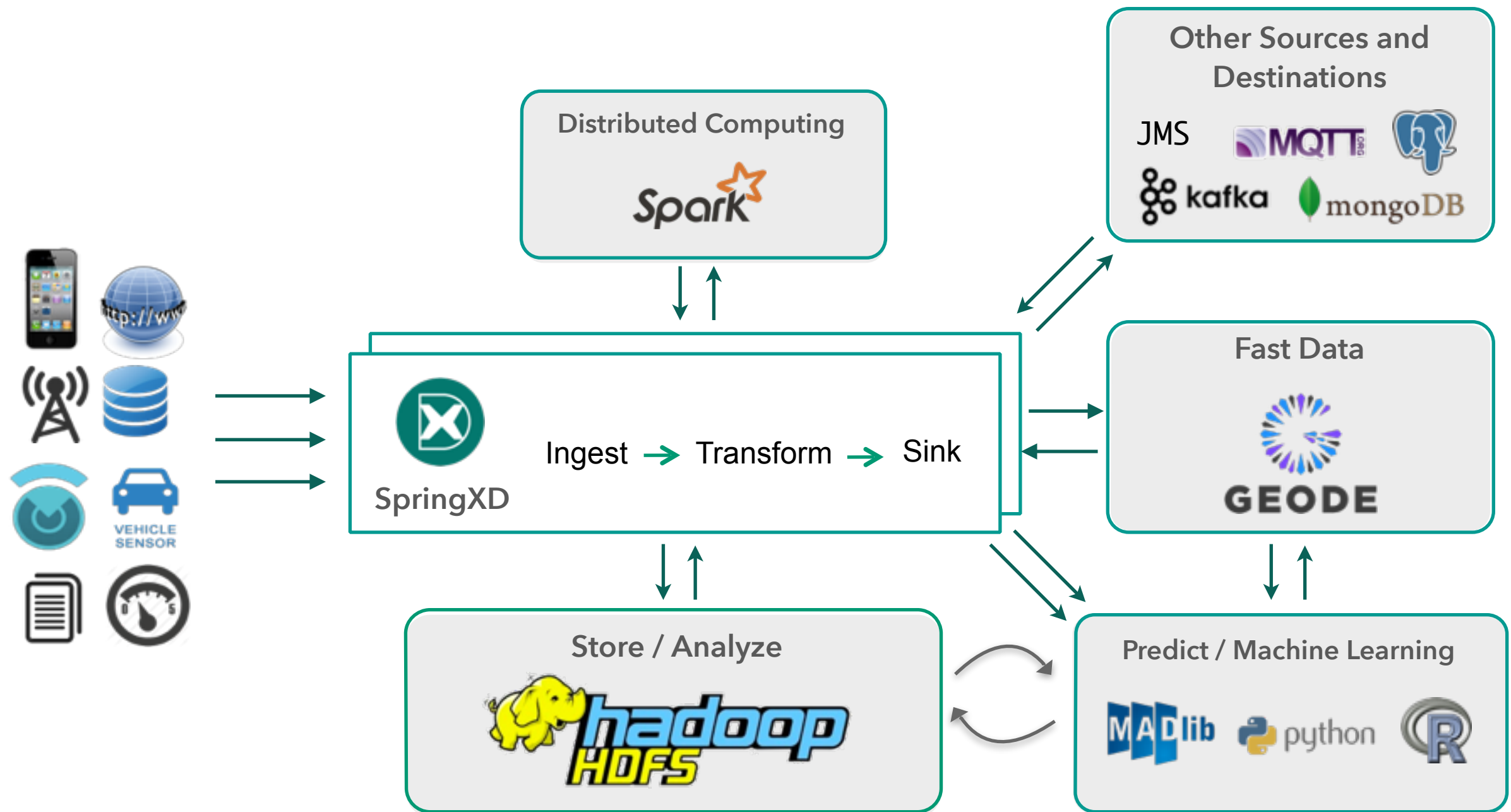


# Streaming Machine Learning

## Supervised Learning Example

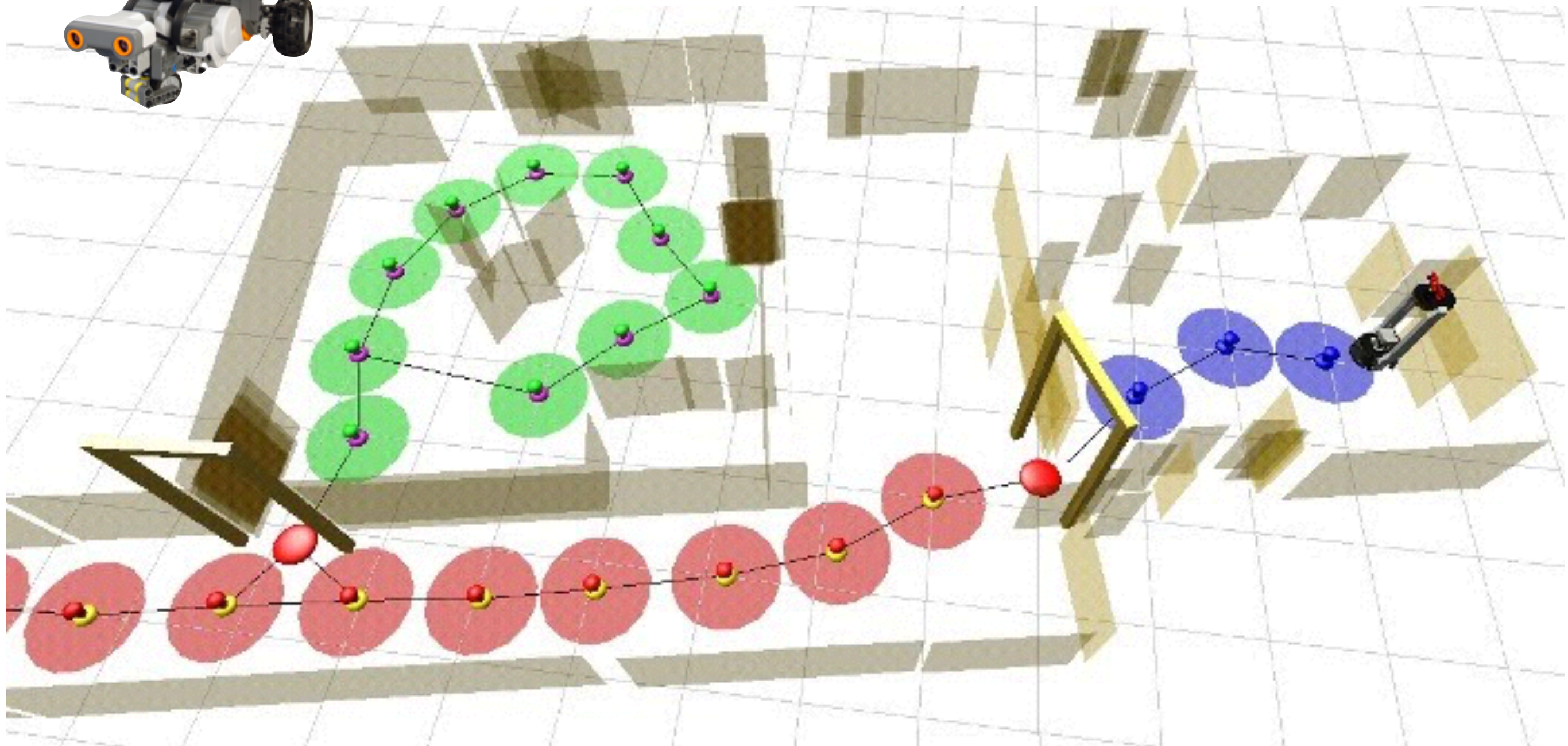


# A Streaming Machine Learning Reference Architecture





# Indoors Localization - Applied Example



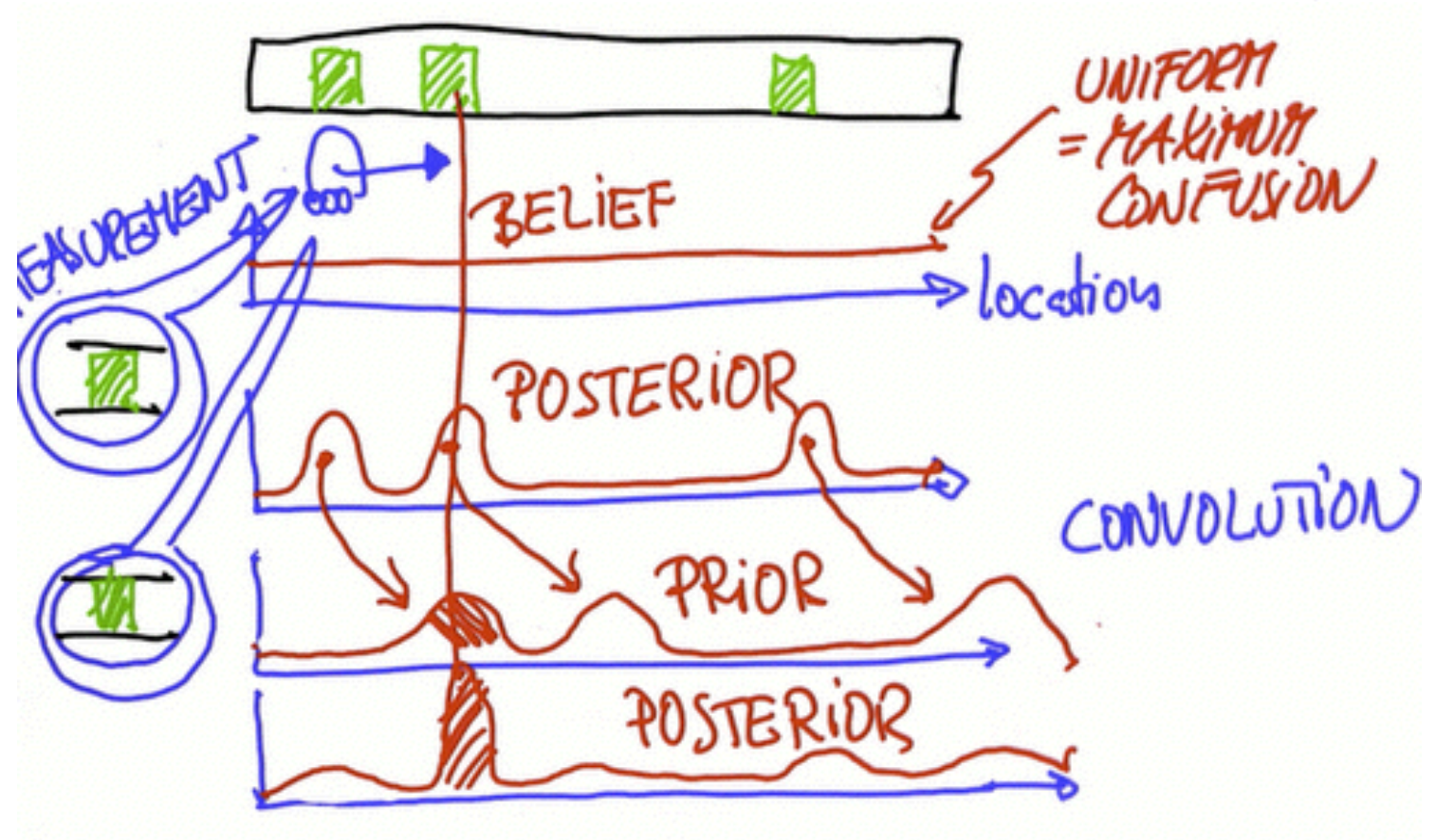
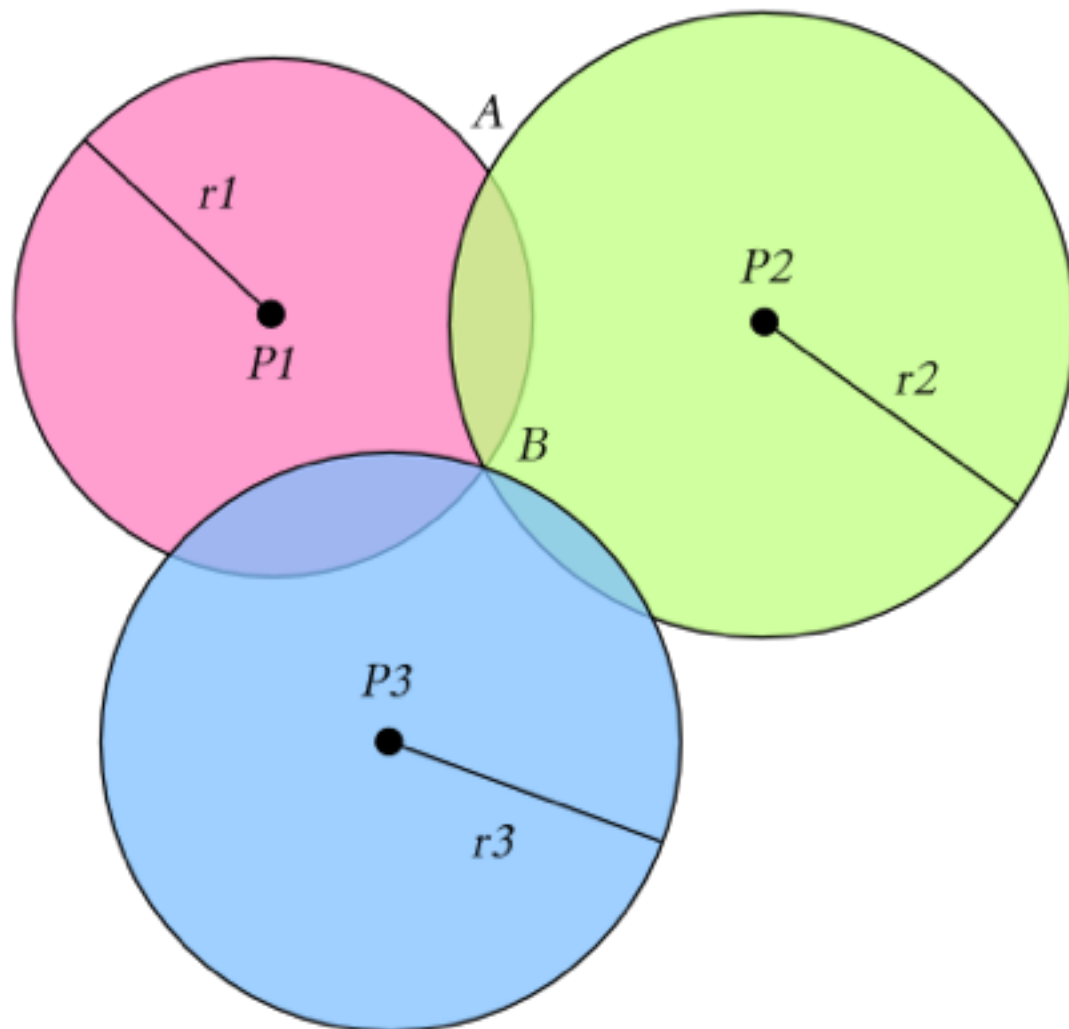


# Trilateration and its limitations

Noisy Data

Physical Barriers

Large Overlap Areas

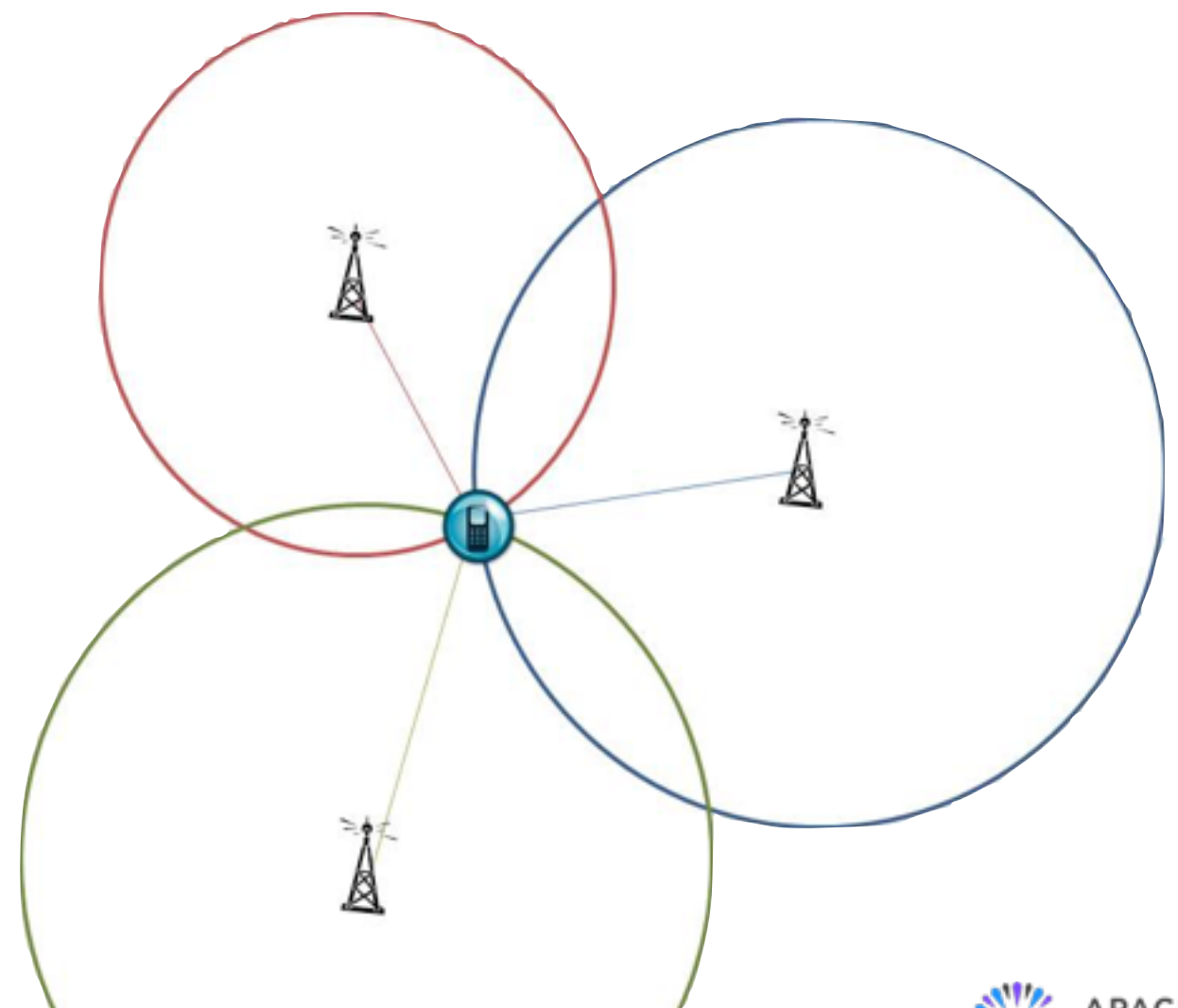
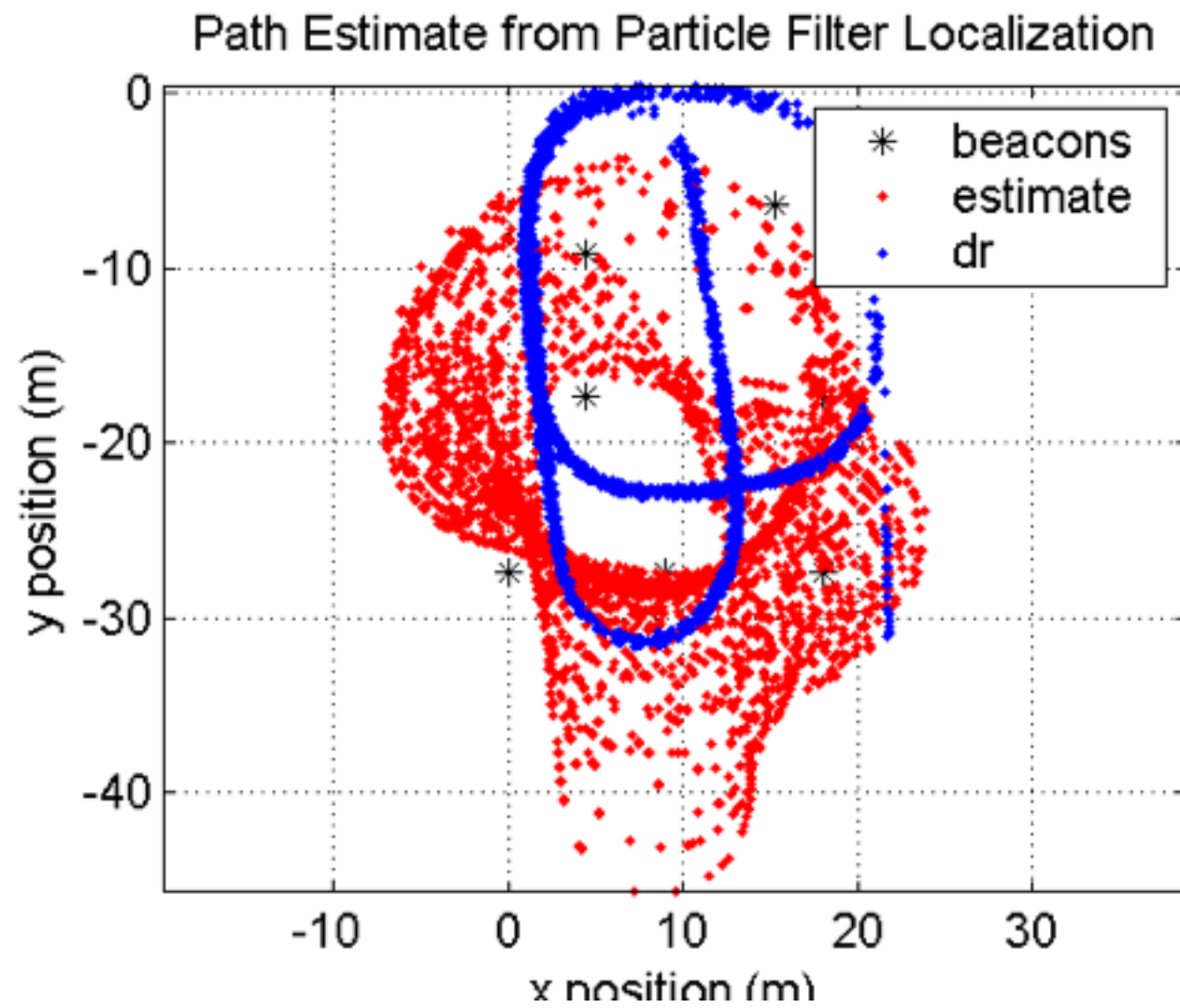


Moving Targets

Innaccuracy

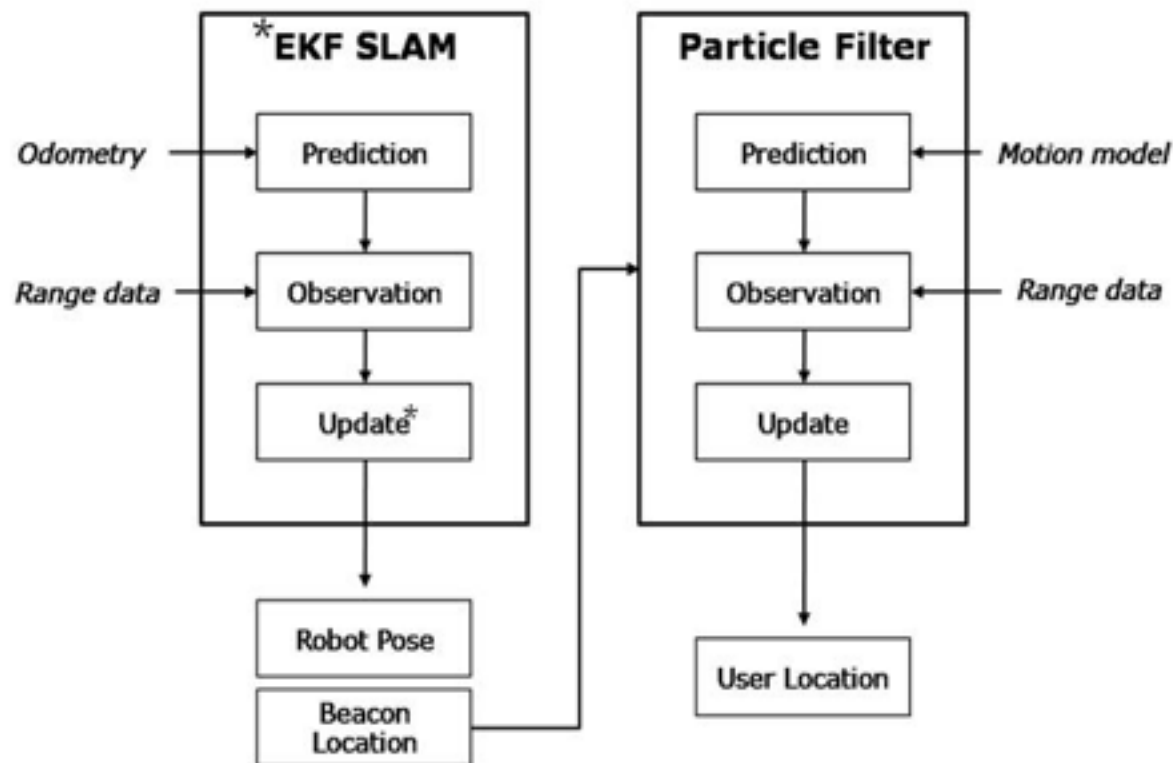
Large Overlap Areas

# Particle Filters - Calculating the optimum solution

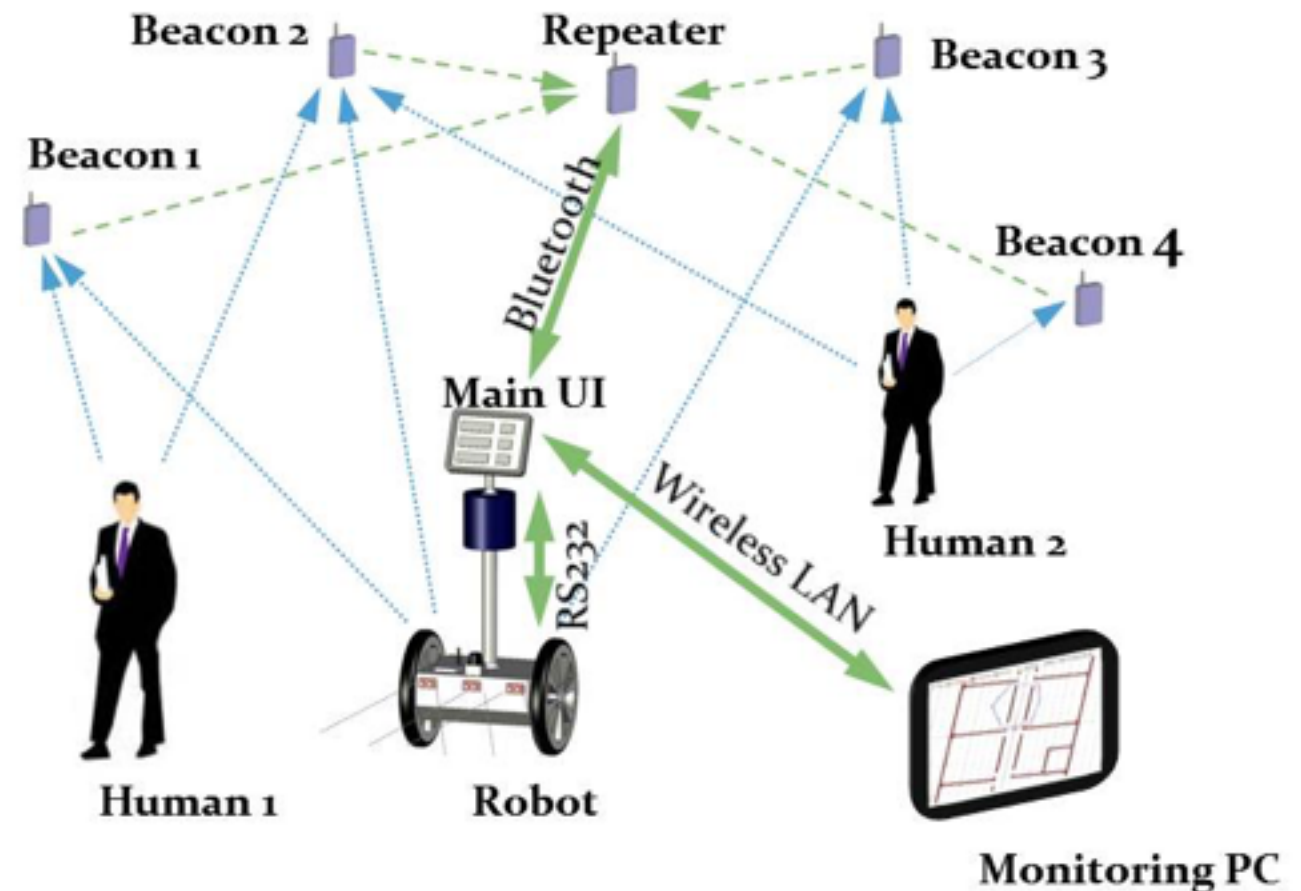


# Particle Filters - Calculating the optimum solution

## Localization Algorithm



## Prototype System



User localization based on the localization of robots and beacons

\* Extended Kalman Filter Simultaneous Localization And Mapping

Autonomous Navigation

# The Solution

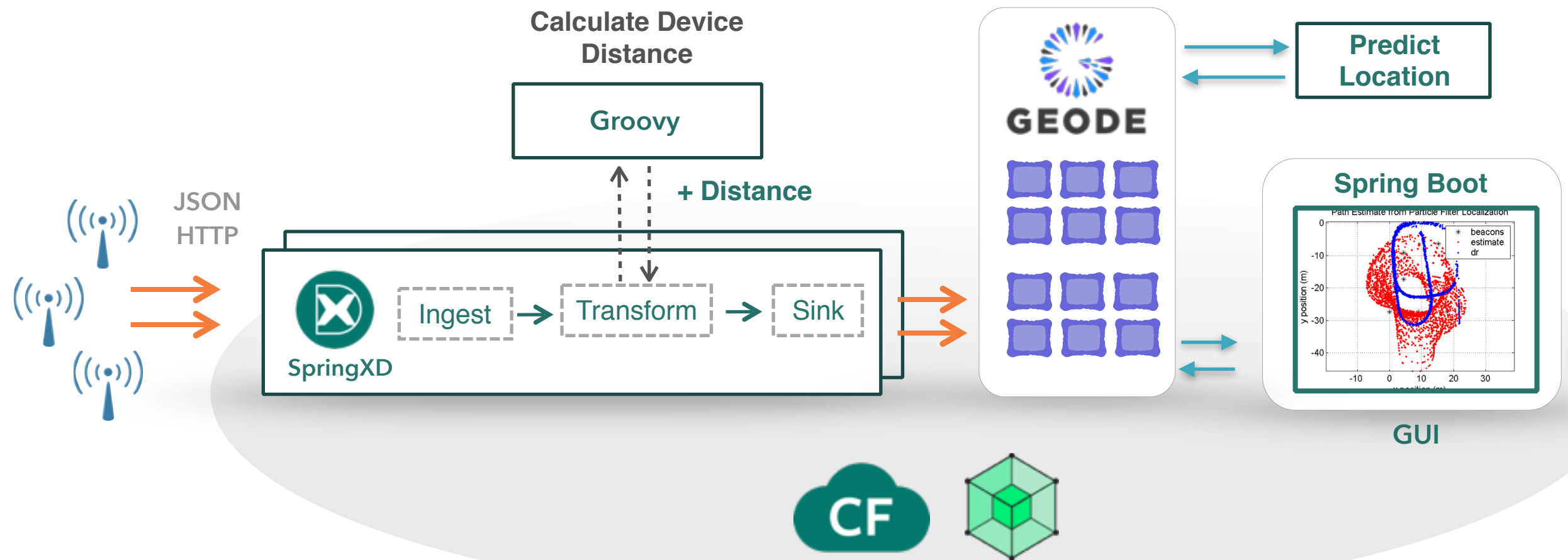
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1. Capture signal strength
2. Calculate distance from antenna
3. Trilaterate different sensors to predict location in real-time
4. Show on a map with live updates



# Architecture Overview




Application Platform

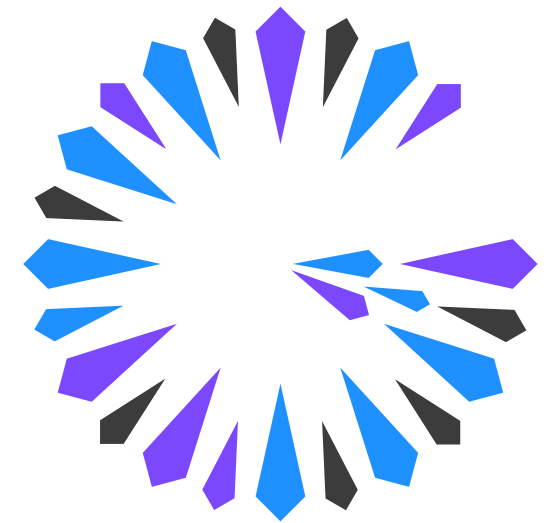




# Geode Basic Concepts

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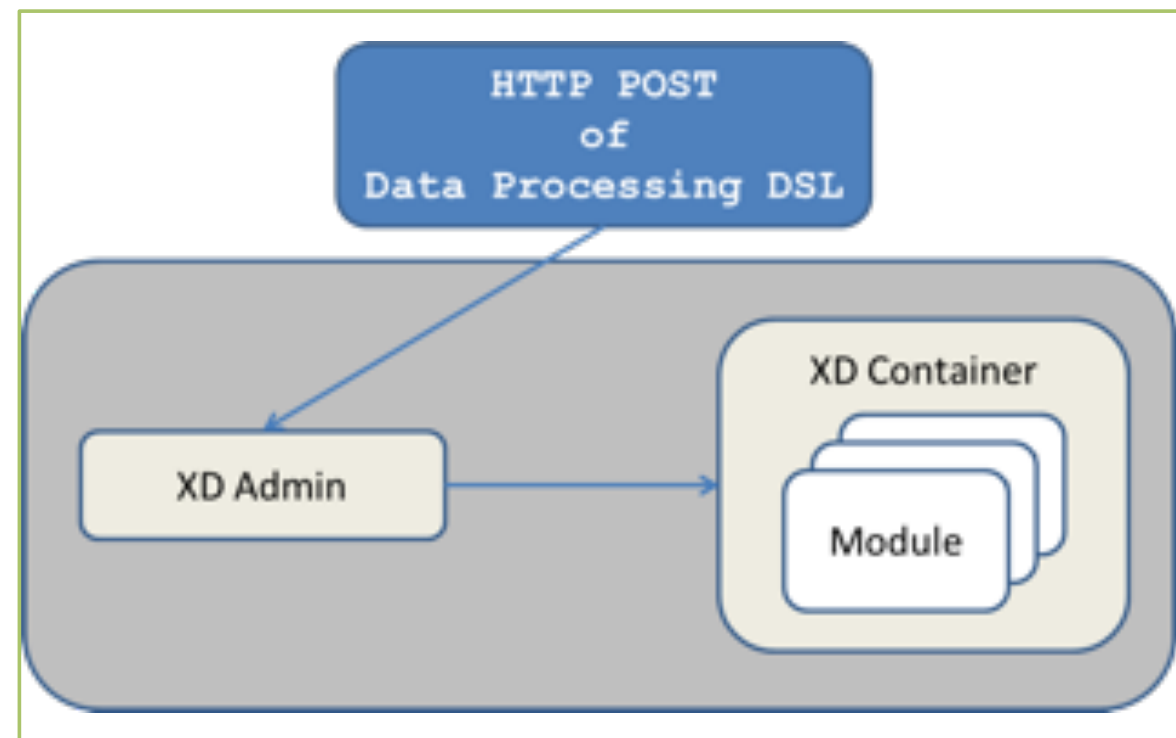
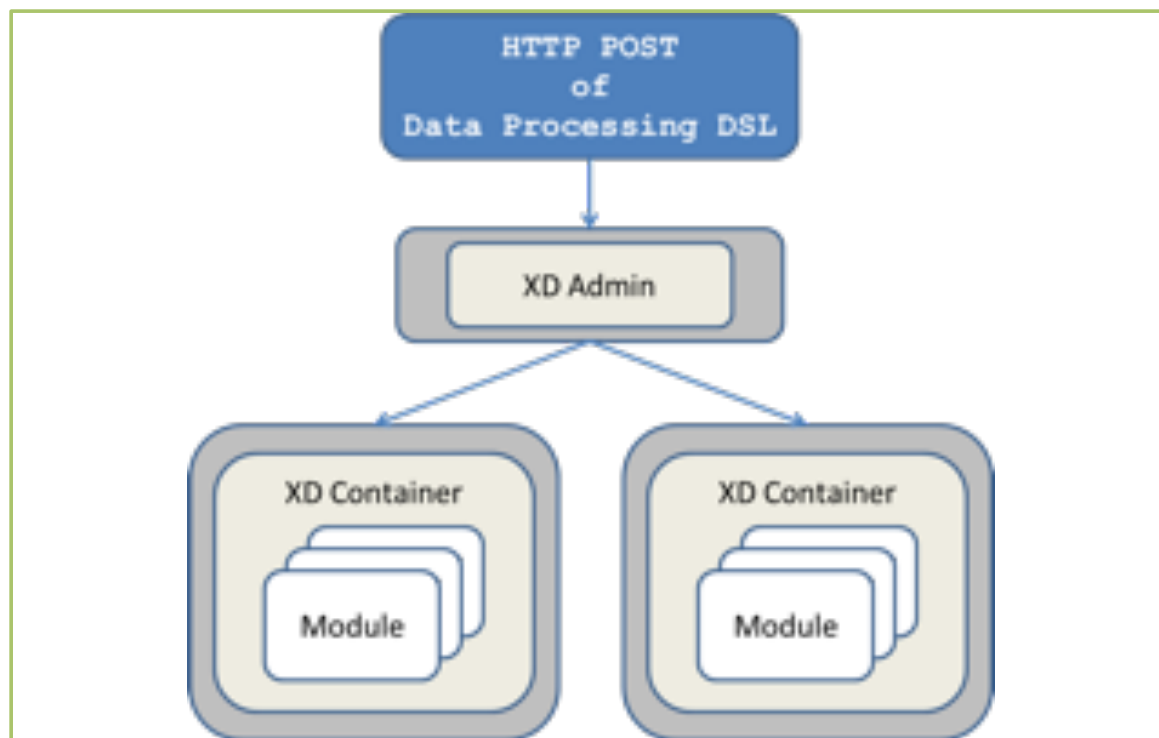
- Cache
  - Configurable through XML,  **spring**, Java
- Region
  - Distributed j.u.Map on steroids
  - Highly available, redundant
- Member
  - Locator, Server, Client
- Callbacks
  - Listener, Writer, AsyncEventListener, Parallel/Serial



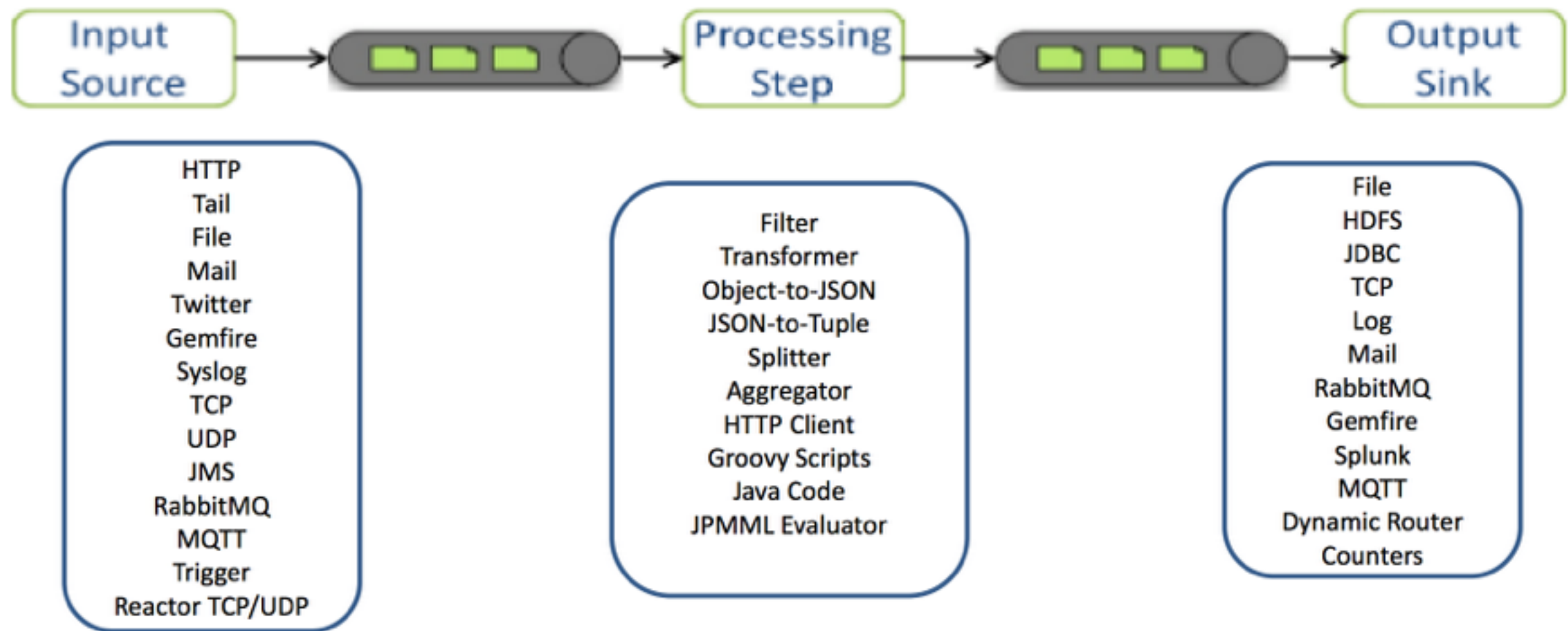
# Introduction to SpringXD

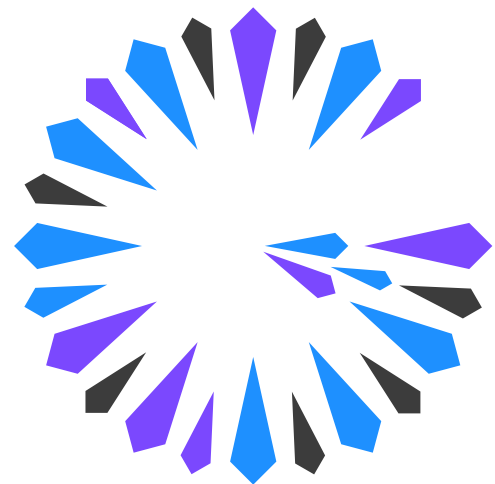


Runs as a distributed application or as a single node



A stream is composed from *modules*. Each module is deployed to a *container* and its channels are bound to the *transport*.





APACHE  
**GEODE**

Demo

# Why have we selected those projects

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- Iterative & Exploratory model
- Web based REPL
- Multiple Interpreters
  - Apache Geode
  - Apache Spark
  - Markdown
  - Flink
  - Python...



- Productivity
- Built-in connectors
- Cloud Agnostic
- Highly Scalable
- Easy to setup
- Streams without coding



- In-memory & Persistent
- Highly Consistent
- Extreme transaction processing
- Thousands of concurrent clients
- Reliable event model



Source code and detailed instructions available at:

<https://github.com/Pivotal-Open-Source-Hub/WifiAnalyticsIoT>

Tomorrow:

***Implementing a Highly Scalable In-Memory Stock  
Prediction System with Apache Geode (incubating),  
Spark MLlib and Spring XD***

*Room: Tohotom - 14:30, Sep 30*

*Fred Melo, Pivotal, William Markito, Pivotal*

**Follow us on GitHub!**

**Fred Melo**

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**William Markito**

 @william\_markito