



**GyG**

# Get your Grocery AG

Our Big Data Architecture | Competitive advantage in the 21<sup>st</sup> century

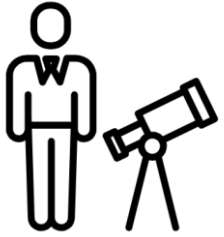
Niels Humbeck (CDE)

21.03.2021 Köln

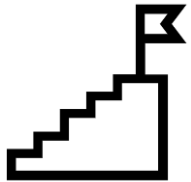
# Agenda

1. GyG - Our Vision & Mission
2. Our Big Data Ambitions | GAP analysis
3. Core requirements of big data processing systems
4. Stream versus Batch processing of big data
5. GyG - Our big data architecture
6. One step ahead – shaping our future at GyG

# GyG - Our Vision & Mission



**Save peoples time by delivering the most efficient supply solution for groceries in the needed quality all over the world.**



## **Our Big Data Ambitions:**

### Real time driven supply chain & sales

- Dynamic pricing
- Demand forecasting & optimization

### World leading customer experience

- Product recommendation engines
- Interactive voice/ chat bots
- Personalized offers & advertisement
- Digital Store Assistant

### Establishing multi-channel retailing


- Buy & collect
- Home delivery
- Localized product assortments

# Our Big Data Ambitions | GAP analysis

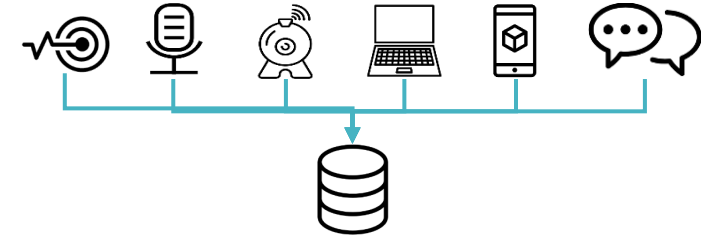
## Our Obsolete IT Infrastructure

## Our New IT Infrastructure

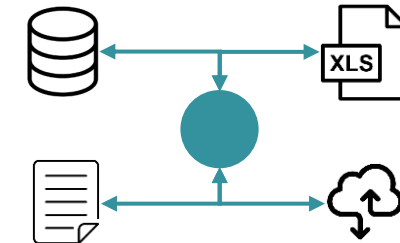
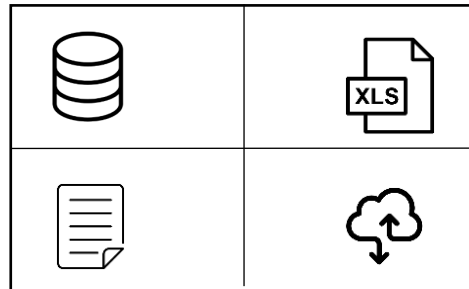
Multidimensional  
data ingestion



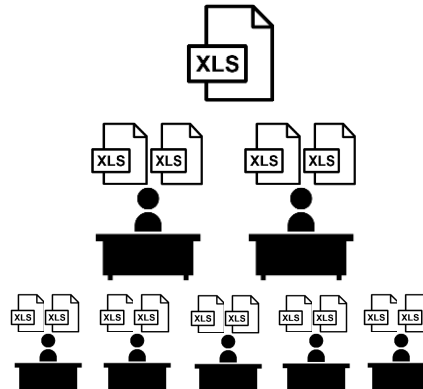
Costumer ID	Name	First Name	Age
50	Humbeck	Niels	29



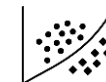
Connectivity &  
Latency



Increase analytic  
capabilities

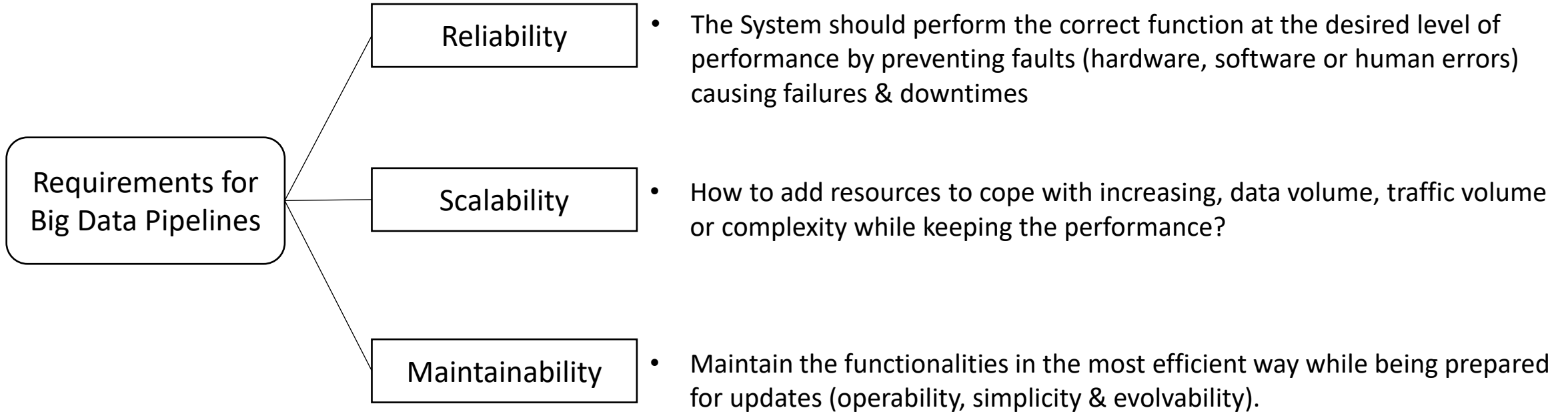


- Real Time analytics and visualization
- Automated reports
- Machine Learning



# Core requirements of big data processing systems

## Reliability, scalability and maintainability

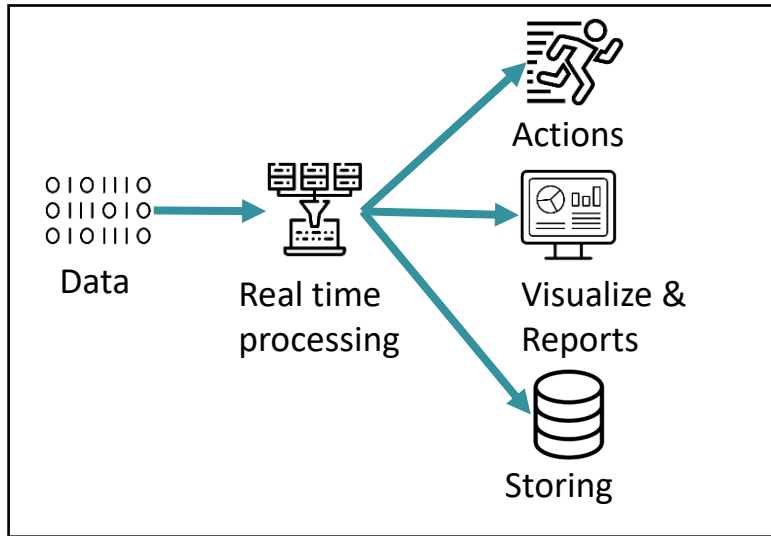


Focus on:

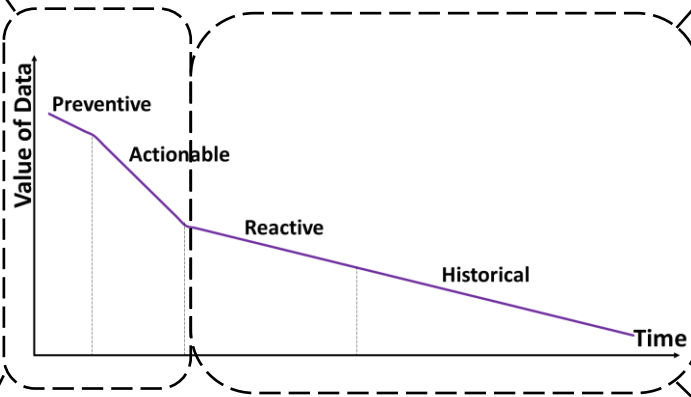
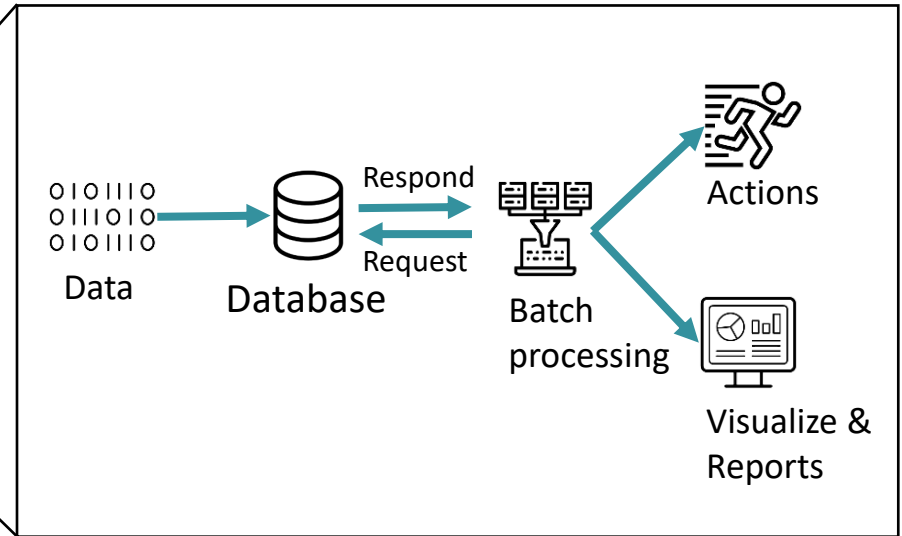
**Optimized for availability, high throughput and low latency**

# Stream versus Batch processing of big data

## Stream processing:



## Batch processing:



- ✚ • Low latency
- Up to date data
- ✖ • Expensive
- Higher complexity
- Normally less complex analysis

### Use Cases:

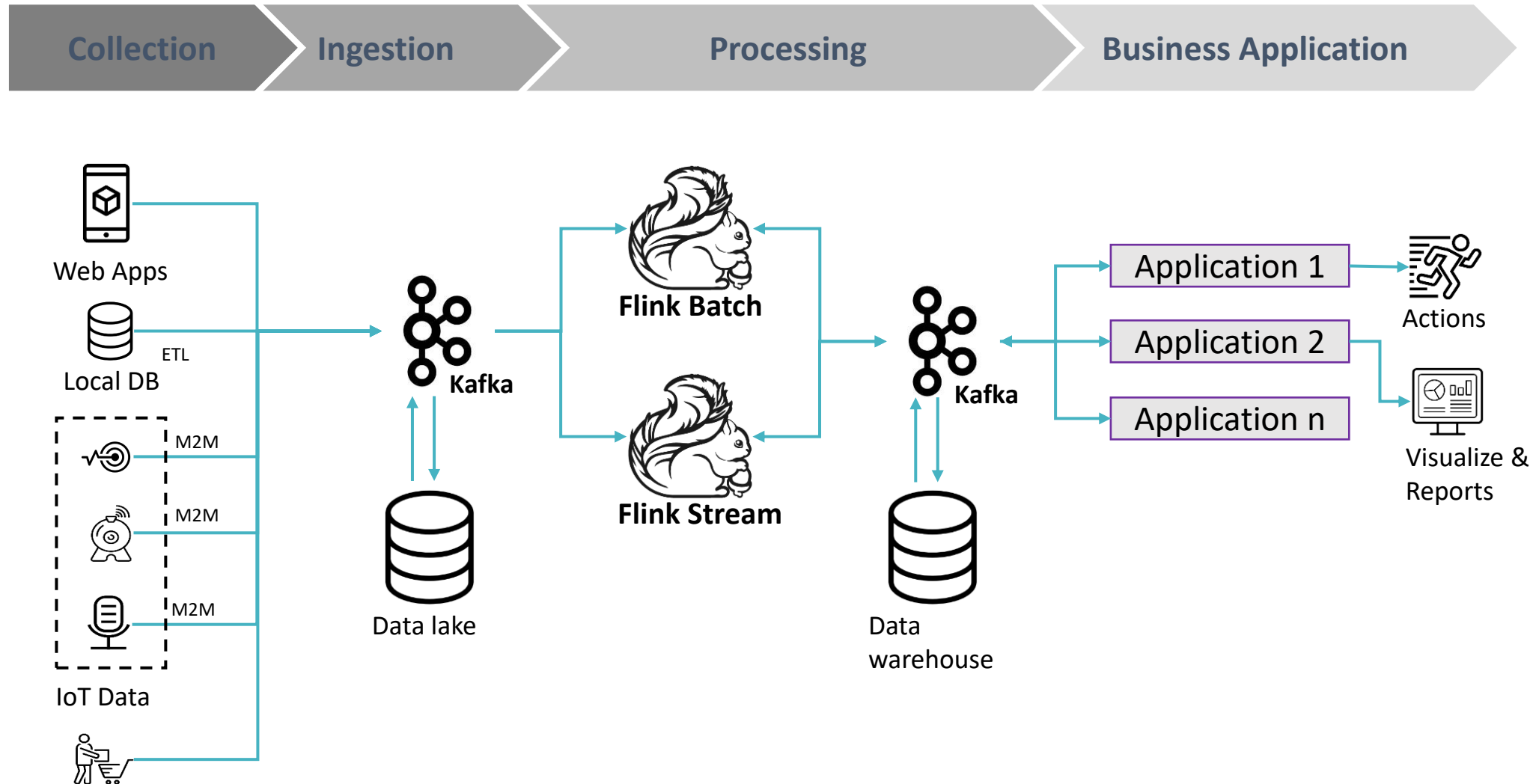
- Realtime inventory tracking system
- Recommendation system

- ✚ • Large batches of data
- Complex analytics & independency
- Efficient, cost effective
- ✖ • High latency

### Use Cases:

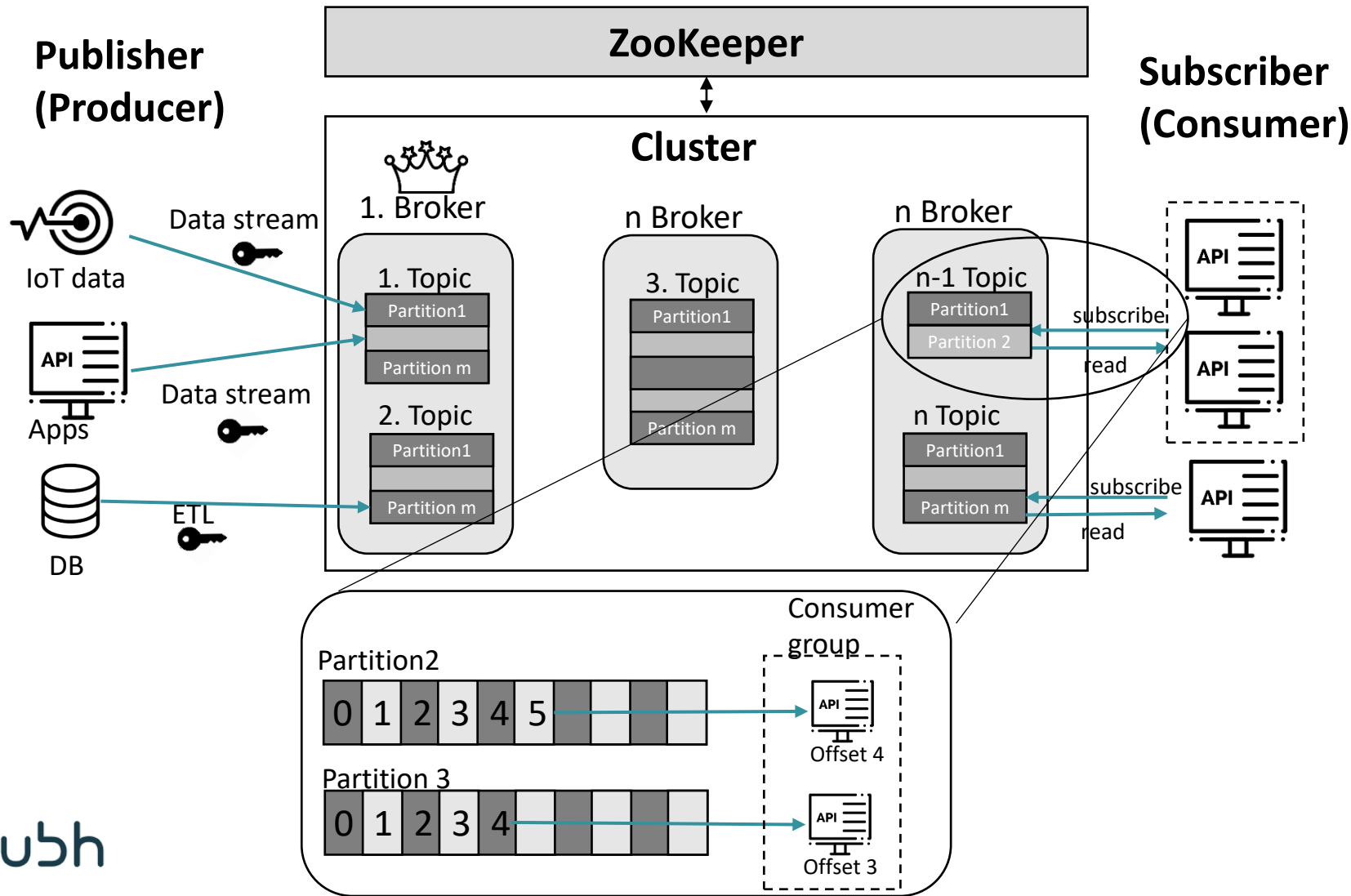
- Demand forecasting

# GyG - Our big data architecture



# GyG - Our big data architecture

## Kafka | Pub/Sub messaging system with distributed immutable commit log



### Reliability guaranties:

- Guaranties the order in a partition
- At-least once messages guarantee

### Scalability:

- Highly scalable (horizontal)

### Maintainability:

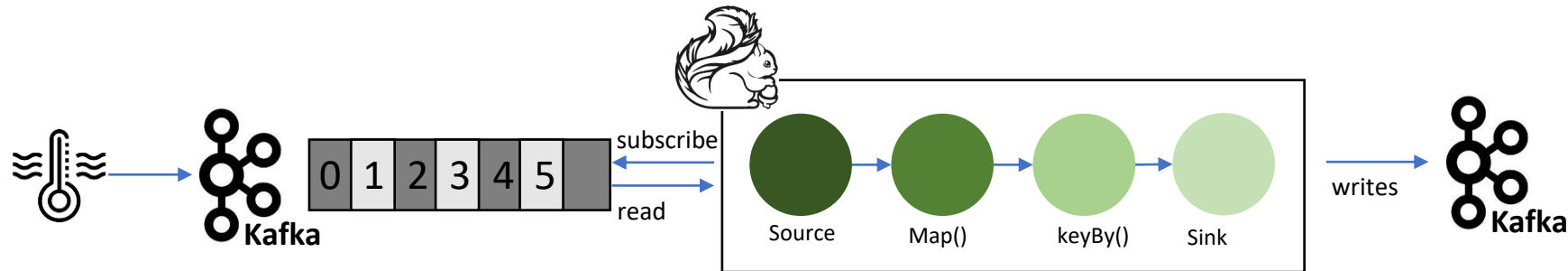
- Multiple producers/ consumers
- Reduces integration complexity
- Highly configurable + retention
- Fault tolerant

- + ETL & Big Data ingestion
- High Throughput
- Fairly low latency
- Training
- Not for real low latency



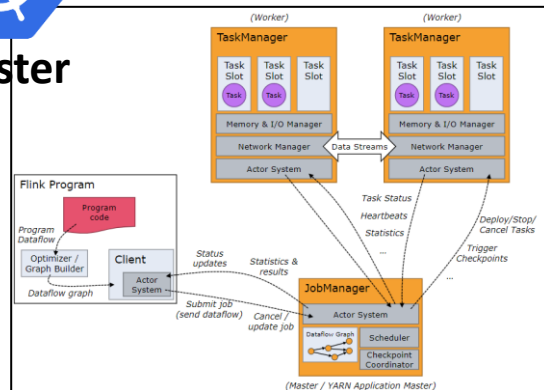
# GyG - Our big data architecture

Flink | Distributed processing engine for stateful computations over (un-) & bounded data streams



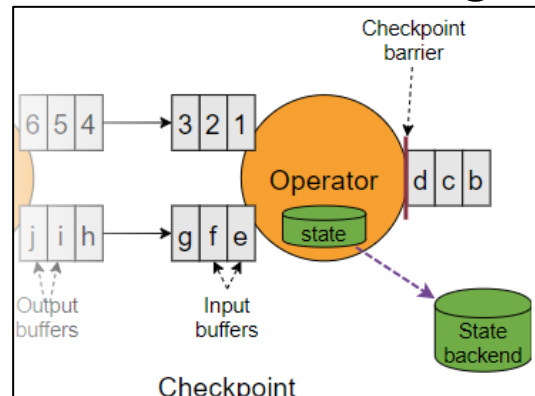
Cluster

## Distributed



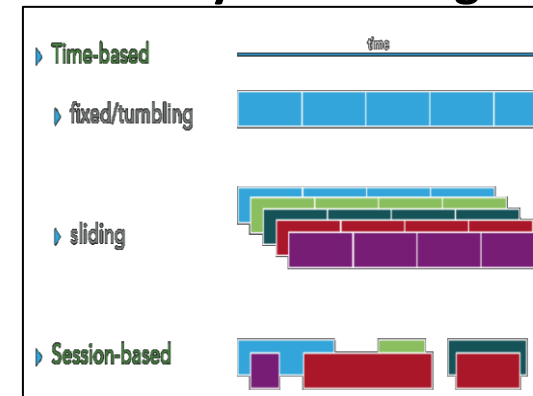
- Horizontal scaling
- High throughput

## Stateful Processing



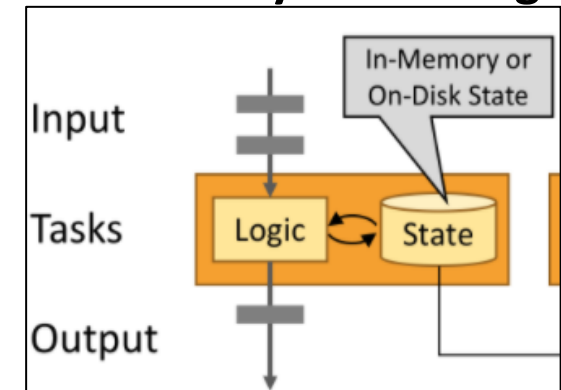
- Exactly once guarantee
- Fault tolerant

## Timely Processing



- Enabling stream processing

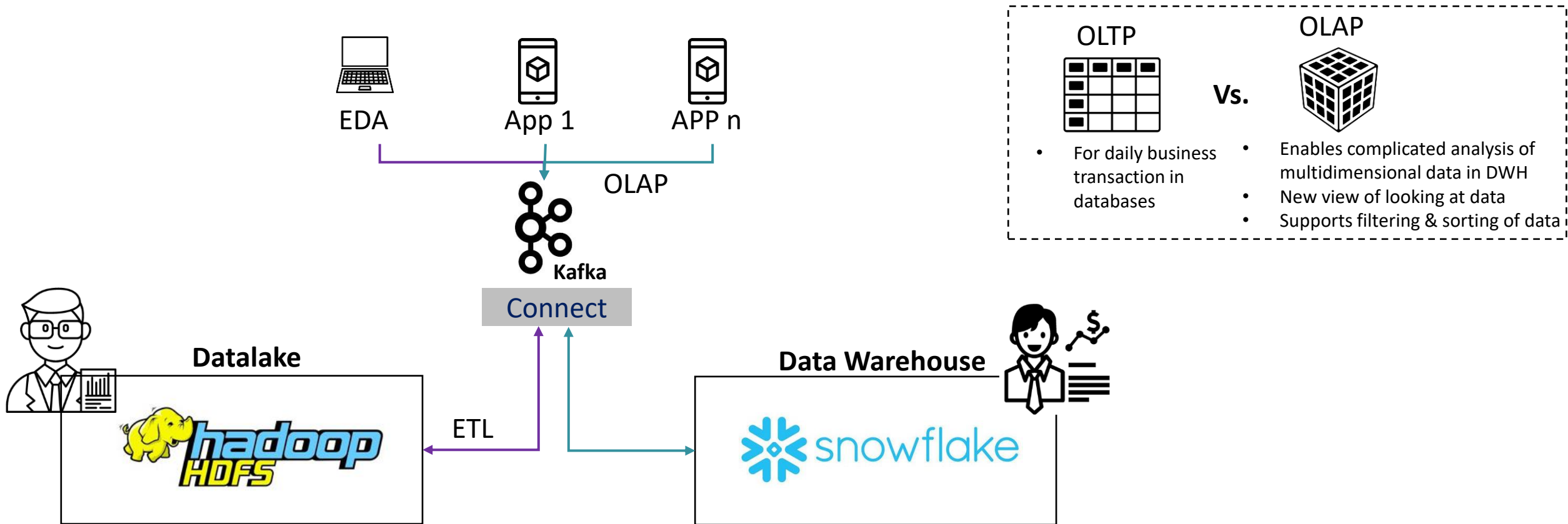
## In-Memory Processing



- Low latency

# GyG - Our big data architecture

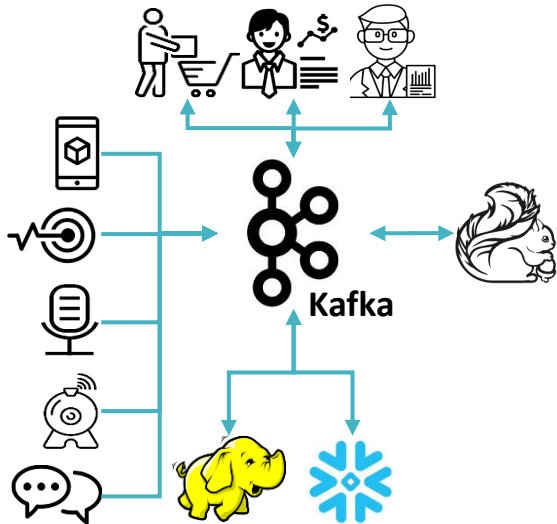
## Data storage & business application layer



# One step ahead – shaping our future at GyG

## State of the art Big Data infrastructure

- Reliable, maintainable and scalable real time big data system
  - Kafka as our data backbone enabling multi data ingestion & low latency connectivity
  - Real time analysis with Flink
  - Next generation cloud data warehouse



## Personalized shopping experience & advertisement

### Technology Trends:

- Facial recognition technology
- Social Media integration
  - Sentiment analysis (NLP)
  - Customer segmentation (ML)
- Interactive voice/ chat bots
- Block-Chain integration

### Product Developments:



#### Personal store assistant (App)

- Product recommendation engines
- Virtual try-on functionalities
- Product information



#### Personalized outdoor advertisement

- Product recommendation engines
- Personalized outdoor advertisement

### System Readiness:



- Highly configurable data ingestion of multiple sources
- Processing, analysis & ML of various data in real time
- Store & access multidimensional data fast in the cloud



Q&A

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