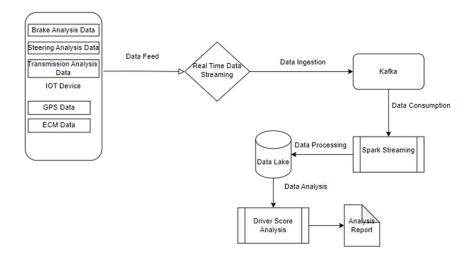
#### **Architecture**



### **Architecture Overview**

# 1. Data Acquisition & Ingestion:

• **IoT Devices:** Vehicles equipped with telematics devices continuously stream data, including GPS coordinates, speed, acceleration, braking patterns, engine status, etc.

# • Kafka for Data Ingestion:

- o **Producer:** Telematics devices act as producers, sending real-time data to Kafka topics.
- Topics: Data is categorized into different Kafka topics based on the data type (e.g., GPS, Speed, Acceleration).
- o **Consumer:** A Spark Streaming application consumes data from these Kafka topics in real time for further processing.

# 2. Data Transformation & Processing:

# • Spark Streaming:

 Data Cleansing: Spark processes the raw data to remove noise, handle missing values, and correct inaccuracies.

# Feature Engineering:

- Harsh Braking Detection: Analyzing deceleration patterns to detect instances of harsh braking.
- **Idle Engine Detection:** Identifying periods where the engine is on but the vehicle is stationary.
- Over Speeding Detection: Comparing speed data against predefined speed limits.
- Right Turn Detection: Calculating turn angles using GPS data to detect and analyze right turns.
- Threshold & Weightage Application: Configurable logic applies predefined thresholds and weightages to the features, preparing them for scoring.

## 3. Data Storage & Management:

- Scalable Storage Solutions:
  - o S3: Processed data is stored in as s3 files for both real-time access and long-term storage.
  - O Data Lake: A data lake architecture can be implemented to manage both raw and processed data, ensuring efficient data retrieval for analysis.
- Data Consistency: Delta Lake can be employed to maintain consistency across storage layers, enabling accurate and timely data retrieval.

## 4. Driver Score Calculation:

- Scoring Algorithm:
  - Implement a scoring algorithm in Spark that combines weighted factors into a single driver score. This score is calculated in near real-time as data is ingested and processed.
  - The algorithm is optimized to handle large volumes of streaming data, ensuring performance and scalability.

# • Customization & Flexibility:

• The scoring model allows for customization to accommodate different driving contexts or fleet-specific requirements, ensuring flexibility in score calculations.

#### **Deliverables**

1. **Architecture Diagram:** This diagram will visualize the entire pipeline, highlighting the flow from data ingestion to driver score calculation.

# 2. Running Code:

- o Kafka Producer: Simulates telematics devices sending data to Kafka.
- o Spark Streaming Job: Consumes data from Kafka, processes it, and calculates driver scores.
- o **Scoring Module:** Implements the driver scoring algorithm.
- Output: Driver scores are outputted to the console or terminal for verification.

### **Tech Stack**

• Data Ingestion: Apache Kafka

• **Data Processing:** Apache Spark (using Python)

• Data Storage: S3 and Delta Lake

• Scoring: Custom Spark-based algorithm