# Live processing with AWS (6 Weeks)

# **Steps to Implement the Workflow: [2 Week]**

## • Common Steps:

- o **Data Ingestion**:
  - Read data from the csv which reside inside bucket or local system.
  - Python script will write data into the kafka topic using producer.

## O Data Processing with Spark:

- Then you've to subscribe the topic and consume data using the spark structure streaming.
- Save the processed data into an output bucket in Parquet format with a filename pattern dd\_mm\_vyyy\_filename.

## **Output** Loading Data into SQL Database:

- Use AWS Glue (for AWS) to load the data from the output bucket to the RDS database.
- You've to use medallion architecture for this aws glue pipeline.

## **Detailed Steps for AWS:**

#### 1. Data Ingestion with Kafka:

 Create a Kafka producer in Python to read CSV files and send the data to a Kafka topic.

#### **Data Processing with Spark Structured Streaming:**

- o Create a Spark Structured Streaming job to subscribe to the Kafka topic.
- Process the data and write it to the Bronze layer in Amazon S3 in Parquet format (dd\_mm\_yyyy\_filename.parquet).
- Create subsequent Spark jobs to transform data from Bronze to Silver, and from Silver to Gold layers.

#### 2. Loading Data with AWS Glue:

- Create Glue jobs to load data from the Gold layer in Amazon S3 to Amazon RDS/Aurora.
- Schedule these jobs to run after the Spark jobs complete.

## Data Exploration and Machine Learning Modelling: [2 Week]

Parts below outline the steps for data exploration, implementation of machine learning (ML) modeling methods, and development of a user interface (UI) for querying data using natural language. The goal is to achieve the best fit for the model, to showcase findings using charts and dashboards along with a basic UI to facilitate in question—answer on the data.

## 2. Data Exploration

Utilising the data from the pipeline created above, include below parts in the data handling.

### 2.1 Data Preprocessing

- Handle missing values.
- Normalize and standardize data if necessary.
- Encode categorical variables.
- Split the data into training and testing sets using a 70-30 split ratio (0.3 parameter for train-test split).

### 2.2 Exploratory Data Analysis (EDA)

- Perform statistical analysis to understand data distribution.
- Visualize data using histograms, scatter plots, box plots, and correlation matrices.
- Identify any patterns, trends, and outliers in the data.

#### 3. Machine Learning Modelling

#### 3.1 Model Selection

- Experiment with various Machine learning and deep learning models appropriate according to data.
- Compare model performance using appropriate metrics based on data.

#### 3.2 Model Training and Validation

- Train models on the training set.
- Validate model performance on the test set.
- Perform hyperparameter tuning to optimize model performance.

#### 3.3 Model Evaluation

• Evaluate the models using appropriate metrics.

• Select the model with the best performance for deployment.

### 4. User Interface Development

### 4.1 LLM (large language model) Integration

- Implement a UI that allows users to ask questions about the data using natural language.
- Utilize LLM to understand and process user queries.
- Provide accurate and relevant answers based on the data.

#### 4.2 UI Design

- Develop a user-friendly interface for data querying.
- Ensure the UI is intuitive and easy to navigate.

#### 4.3 Features

- Allow users to input questions in natural language.
- Display answers in a clear and concise manner.
- Include options to visualize the queried data.

## 5. Findings and Visualization [1 Weeks]

#### **5.1 Findings**

- Summarize key findings from data exploration and model evaluation.
- Highlight important insights and patterns discovered during EDA.

#### 5.2 Visualization

- Create charts and dashboards (powerBI or Tableau) to visualize key findings.
- May use tools such as Matplotlib, Seaborn, and Plotly for creating visualizations along with other methods.
- Include visualizations such as(not limited to):
  - o Bar charts
  - Line graphs
  - o Heatmaps

- o Pie charts
- o Interactive dashboards

# **Diagram for AWS Setup:**

- o Amazon S3: Storage for raw and processed data.
- o Apache Spark: Read, process, and write data.
- o **AWS Glue**: Load data from Amazon S3 to SQL Database.
- o Amazon RDS/Aurora: Destination SQL database.
- Technical document

#### 6. Conclusion

This document outlines the steps to setup data pipeline, implement ML models, and develop a natural language-based querying UI. The objective is to find the best-fitting model and effectively communicate findings through visualizations and interactive dashboards.