PROJECT

TOPIC : IoT Sensor Data Analytics Pipeline using Kafka, AWS Glue, DMS, Kinesis, S3, ClickHouse, Hive, and PySpark.

Use-Case : Vehicle Health Monitoring & Predictive Maintenance

Problem: Unexpected vehicle breakdowns cost businesses money and lead to downtime.

Solution: Build a Glue pipeline that processes RPM, engine load, coolant temp (cTemp), and DTC codes in real-time to:

- Detect early signs of engine trouble.
- Trigger alerts for overheating, frequent high RPMs, low battery, or high eLoad.

Pipeline Outcome:

- Alerting system on anomalies
- Maintenance recommendations
- Aggregated health reports in ClickHouse for dashboards

Made By,

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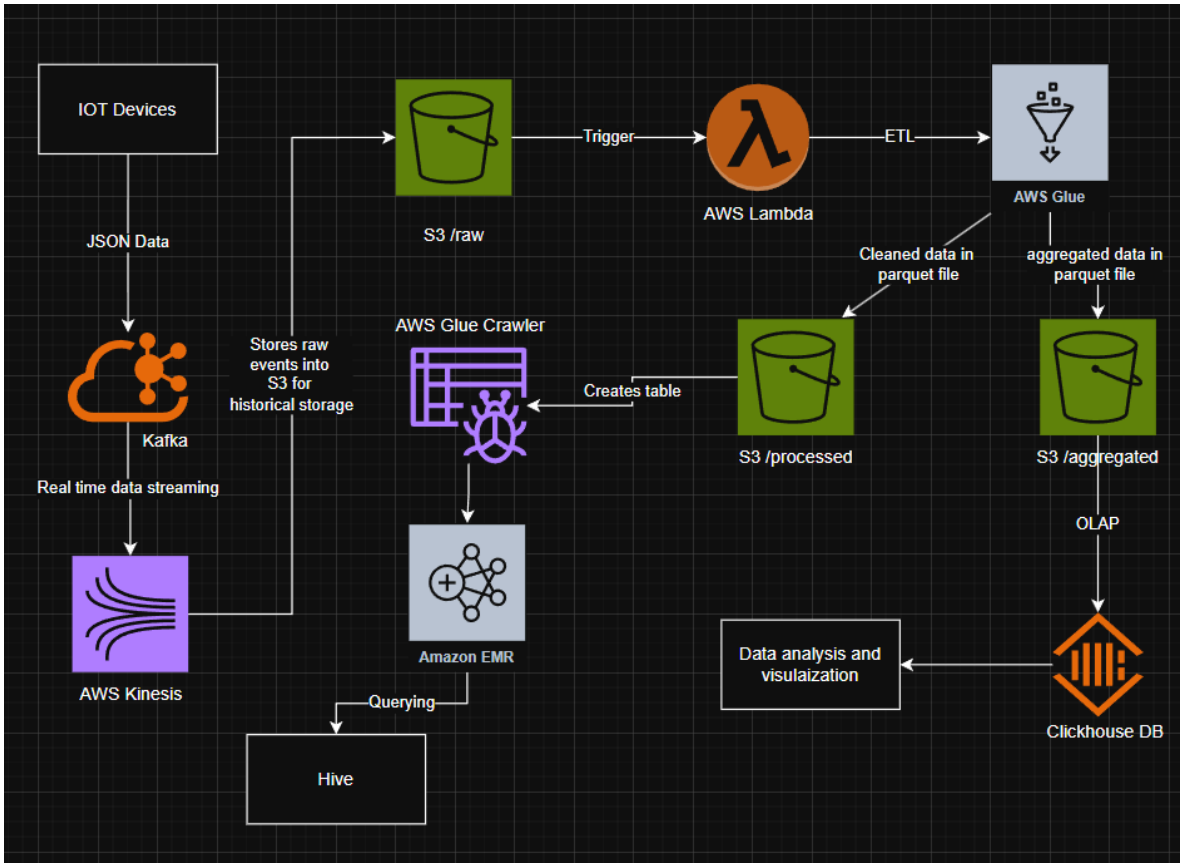
Architecture Diagram:



Dataset used:

This dataset contains vehicle telematics data collected in real-time from multiple passenger vehicles using IoT-enabled OBD-II and GPS modules. The data captures various engine parameters, diagnostic metrics, and driving behavior at one-second intervals, suitable for analysis of vehicle health, performance, and driver patterns.

1. Data Ingestion Layer Apache Kafka:

- Downloaded the dataset from kaggle.
- Converted the csv file into JSON using python script.

```
import csv
import json

input_csv = r"C:\Users\Leon\Downloads\BL-Project\vehicle.csv"
output_json = r"C:\Users\Leon\Downloads\BL-Project\output.json"

with open(input_csv, mode="r", encoding="utf-8") as csv_file, \
     open(output_json, mode="w", encoding="utf-8") as
json_file:

    reader = csv.DictReader(csv_file)

    for row in reader:
        json_line = json.dumps(row)
        json_file.write(json_line + "\n")

print("Success")
```

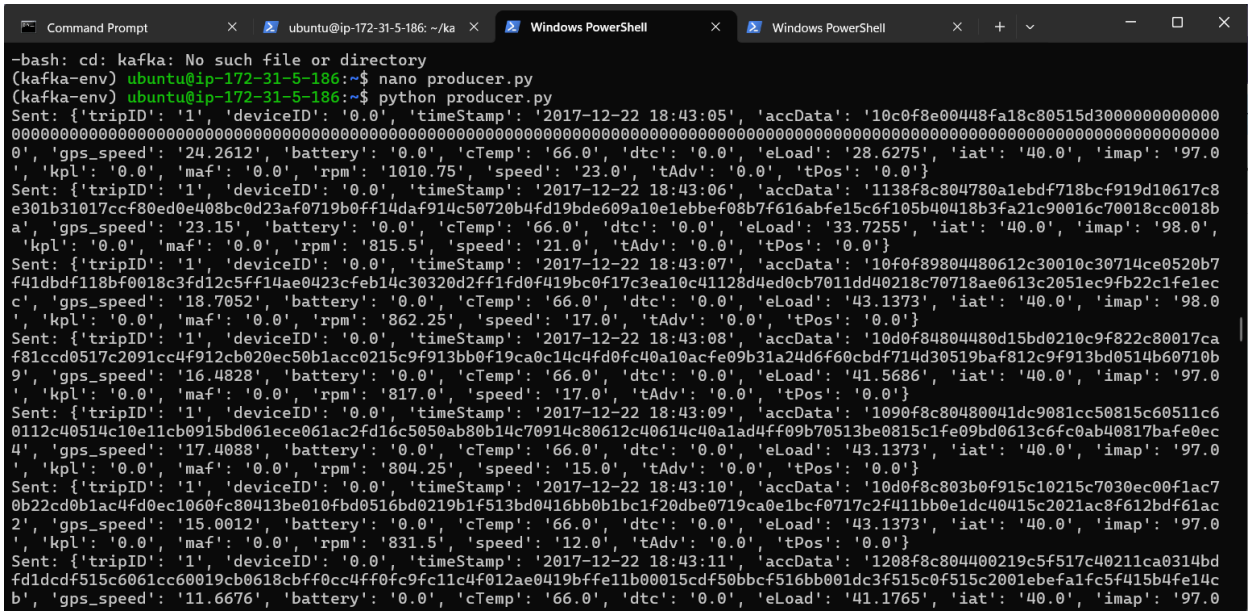
Send json file into ec2 through scp
scp -i /path/to/mykey.pem /path/to/data.json ec2-user@ec2-public-ip:/home/ec2-user/

- Downloaded and set up Kafka on EC2.
- Started zookeeper
  bin/zookeeper-server-start.sh config/zookeeper.properties
- Started server in another terminal
  bin/kafka-server-start.sh config/server.properties
  Create topic
  bin/kafka-topics.sh \
  --create \
  --topic sensor-data \
  --bootstrap-server localhost:9092 \
  --partitions 1 \
  --replication-factor 1

- Created topic sensor-data.
- Producer code

```
from kafka import KafkaProducer
import json
import time

producer = KafkaProducer(
    bootstrap_servers='localhost:9092',
    value_serializer=lambda v: json.dumps(v).encode('utf-8')
)

with open("output.json", "r") as file:
    for line in file:
        data = json.loads(line)
        producer.send("sensor-data", value=data)
        print("Sent:", data)
        time.sleep(0.1)

producer.flush()
producer.close()
```

Producer Sending the data:

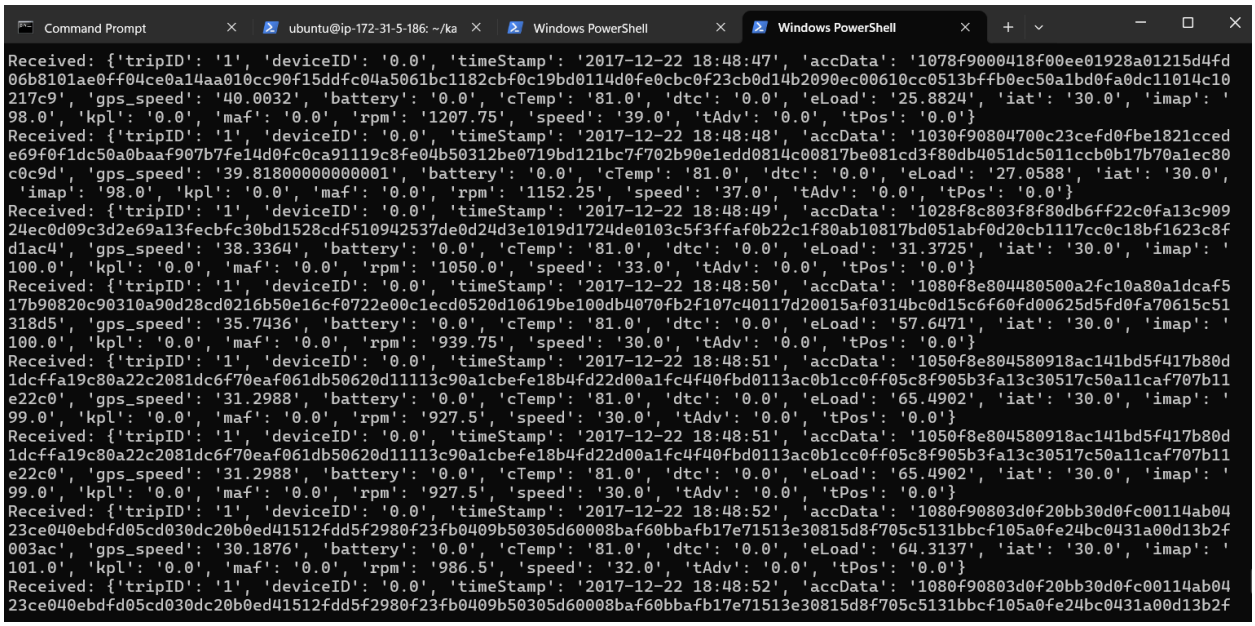


Testing if it is able to be received through consumer:

Consumer code:

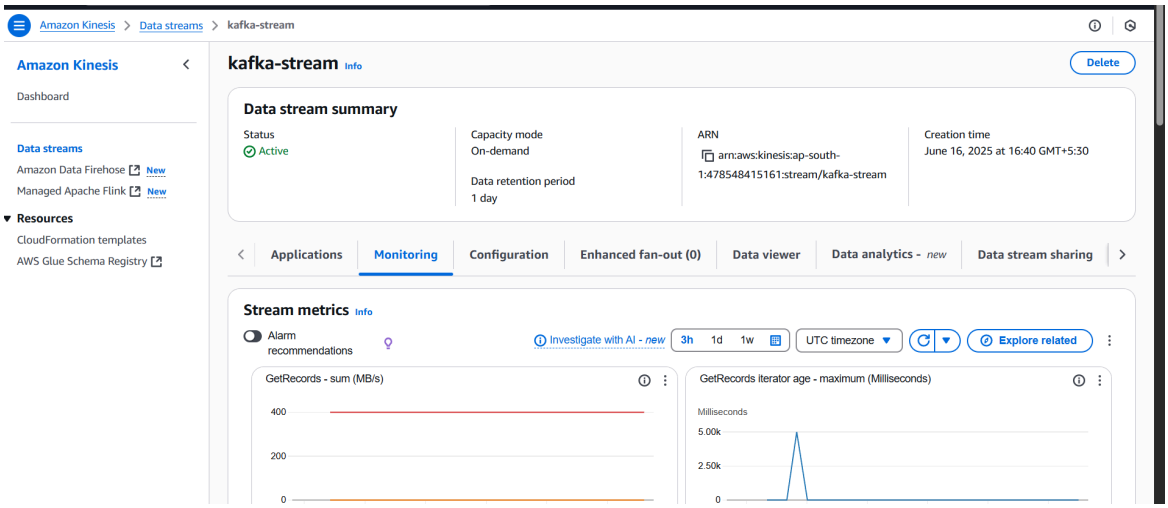
```
from kafka import KafkaConsumer
import json

consumer = KafkaConsumer(
    'sensor-data',
    bootstrap_servers='localhost:9092',
    auto_offset_reset='earliest',
    group_id='my-group',
    value_deserializer=lambda m: json.loads(m.decode('utf-8'))
)

print("Listening to messages...")
for message in consumer:
    print("Received:", message.value)
```

Output:



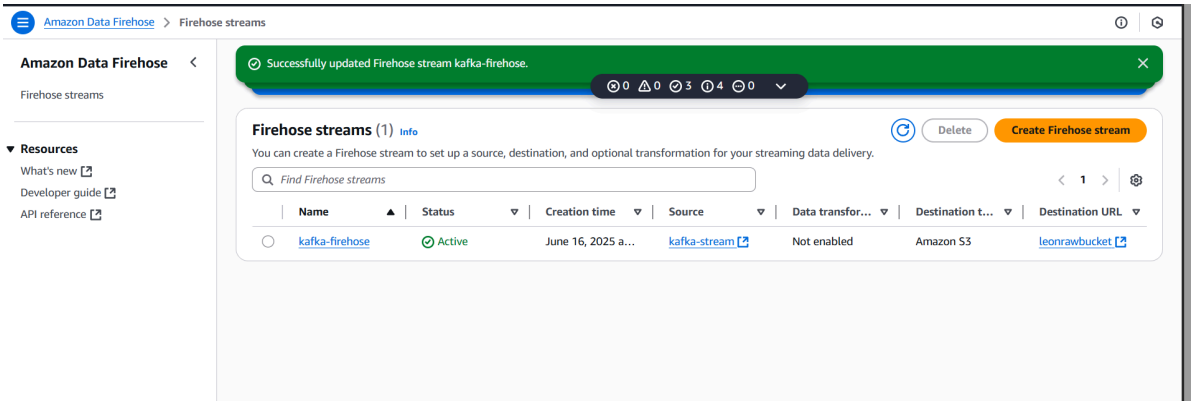
2. Real-time Streaming with Kinesis Amazon Kinesis Data Streams :

- Go to AWS Kinesis and create a kinesis stream called kafka-stream



- Create a kafka firehose with source as kafka-stream and destination as S3 bucket

- Firehose is used to take data from kinesis and store in in s3



- Type aws configure and enter all necessary details in ec2 where kafka is running.

- Clone the connector plugin
  git clone https://github.com/awslabs/kinesis-kafka-connector.git
  cd kinesis-kafka-connector
  ./gradlew shadowJar
  cp target/kinesis-kafka-connector.jar ~/kafka_2.13-3.7.0/libs/

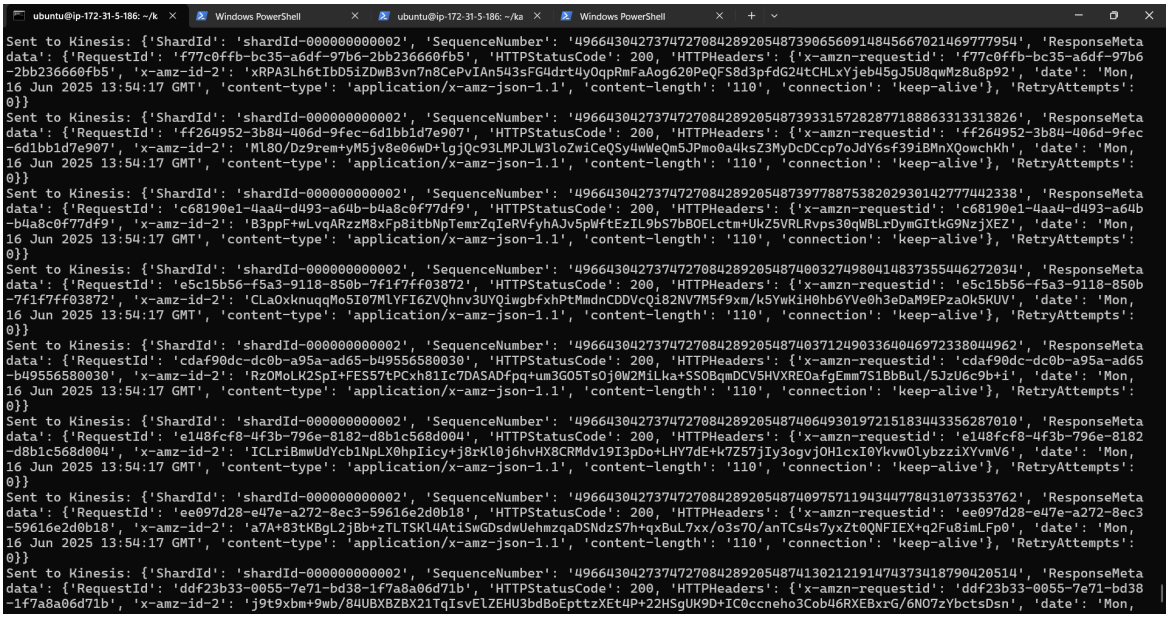
- In kinesis-sink-properties add
  name=kinesis-sink-connector
  connector.class=io.lenses.streamreactor.connect.kinesis.sink.KinesisSinkConnecto
  r
  tasks.max=1
  topics=sensor-data
  aws.kinesis.stream=kafka-stream
  aws.region=ap-south-1
  key.converter=org.apache.kafka.connect.storage.StringConverter
  value.converter=org.apache.kafka.connect.storage.StringConverter
  aws.access.key.id=ACCESS_KEY
  aws.secret.access.key=SECRET_KEY

- Start the connect worker
  bin/connect-standalone.sh config/connect-standalone.properties
kinesis-sink.properties

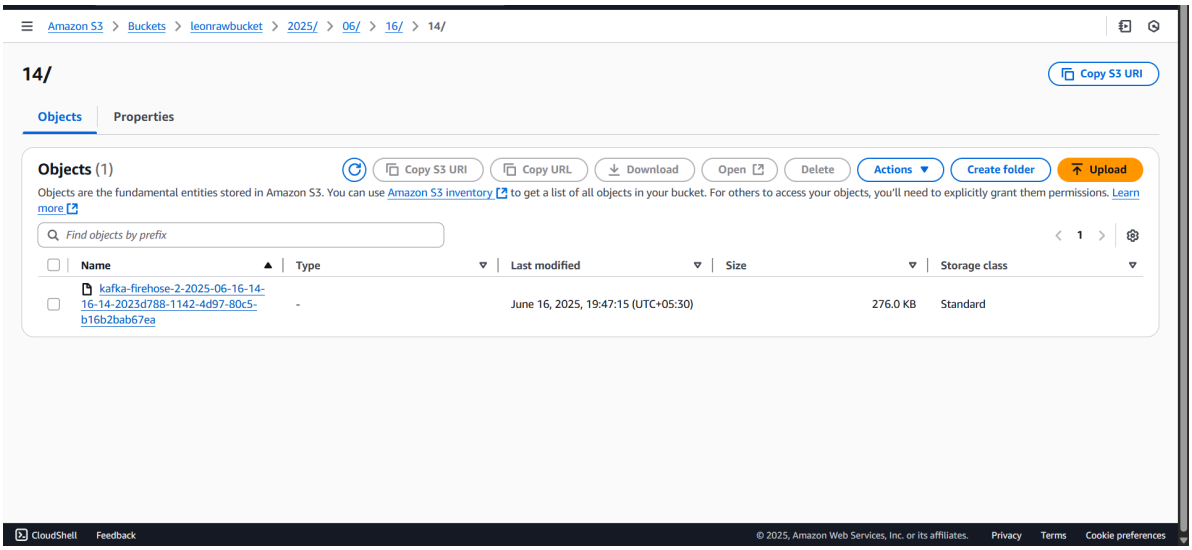
- Update consumer code to send data to kinesis stream

```
from kafka import KafkaConsumer
import boto3
import json

aws_region = 'ap-south-1'
kinesis_stream_name = 'kafka-stream'

kinesis_client = boto3.client('kinesis', region_name=aws_region)

consumer = KafkaConsumer(
   'sensor-data',
   bootstrap_servers=['localhost:9092'],
   auto_offset_reset='latest',
   enable_auto_commit=True,
   group_id='my-group',
   value_deserializer=lambda m: json.loads(m.decode('utf-8'))
)

for message in consumer:
   data = message.value

   payload = json.dumps(data).encode('utf-8')
   response = kinesis_client.put_record(
       StreamName=kinesis_stream_name,
       Data=payload,
       PartitionKey="partitionKey"
   )
   print(f"Sent to Kinesis: {response}")
```

- Run the zookeeper and then the producer as well as the consumer.

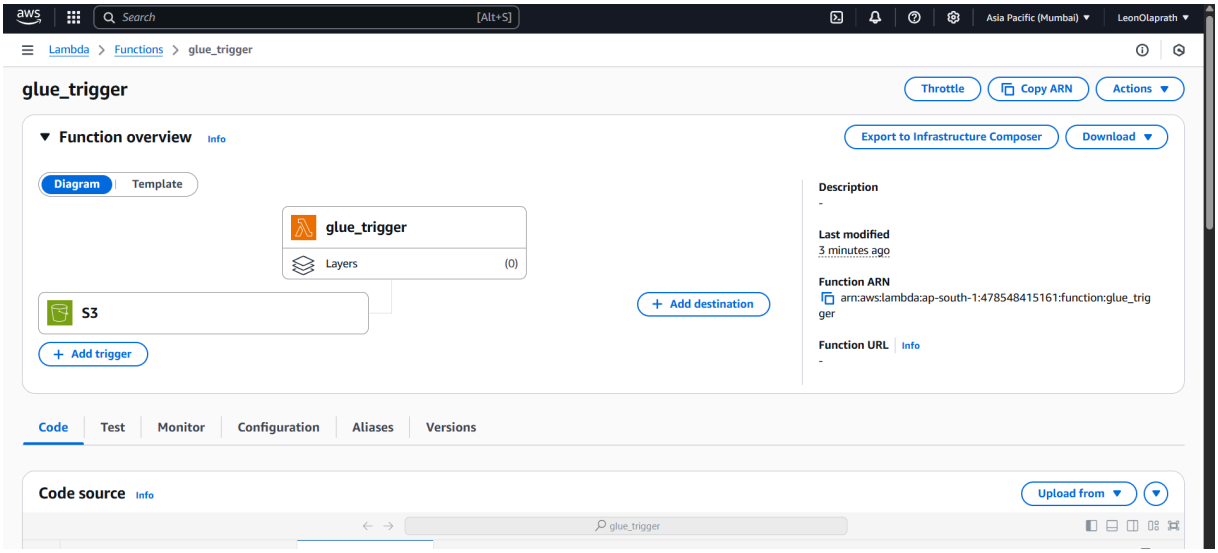


- Check S3 bucket if the files are present.



3. Processing with AWS Glue + PySpark AWS Glue Jobs

Create a lambda function to send the files into glue automatically immediately after it enters the s3 bucket from kinesis.



In the lambda IAM role include

```
{
  "Effect": "Allow",
  "Action": [
    "glue:StartJobRun"
  ],
  "Resource": "arn:aws:glue:region:account-id:job/glue-job-name"
}
```

```python
import boto3
import urllib.parse

glue = boto3.client('glue')

def lambda_handler(event, context):
    bucket = event['Records'][0]['s3']['bucket']['name']
    key = 
urllib.parse.unquote_plus(event['Records'][0]['s3']['object']['key'])

    print(f"New file in S3: s3://{bucket}/{key}")
```

```
    response = glue.start_job_run(
        JobName='Vehicle-stream',
        Arguments={
            '--source_s3_path': f's3://{bucket}/{key}'
        }
    )

    print(f"Glue job started: JobRunId = {response['JobRunId']}")
    return {
        'statusCode': 200,
        'body': f"Triggered Glue job for s3://{bucket}/{key}"
    }
```

Setup AWS- Glue:

- Go to ETL Jobs inside Glue
- Select Python and then choose Spark
- Then in script write the pyspark script to take data from s3 and then clean and send the processed data as parquet file back to s3 in processed and aggregated forms

```
from pyspark.context import SparkContext
from awsglue.context import GlueContext
from pyspark.sql.functions import (
    col, avg, max, count, window, to_timestamp
)
from pyspark.sql.types import StructType, StringType, DoubleType,
IntegerType

sc = SparkContext()
glueContext = GlueContext(sc)
spark = glueContext.spark_session

schema = StructType() \
    .add("tripID", StringType()) \
    .add("deviceID", StringType()) \
    .add("timeStamp", StringType()) \
    .add("accData", StringType()) \
    .add("gps_speed", DoubleType()) \
```

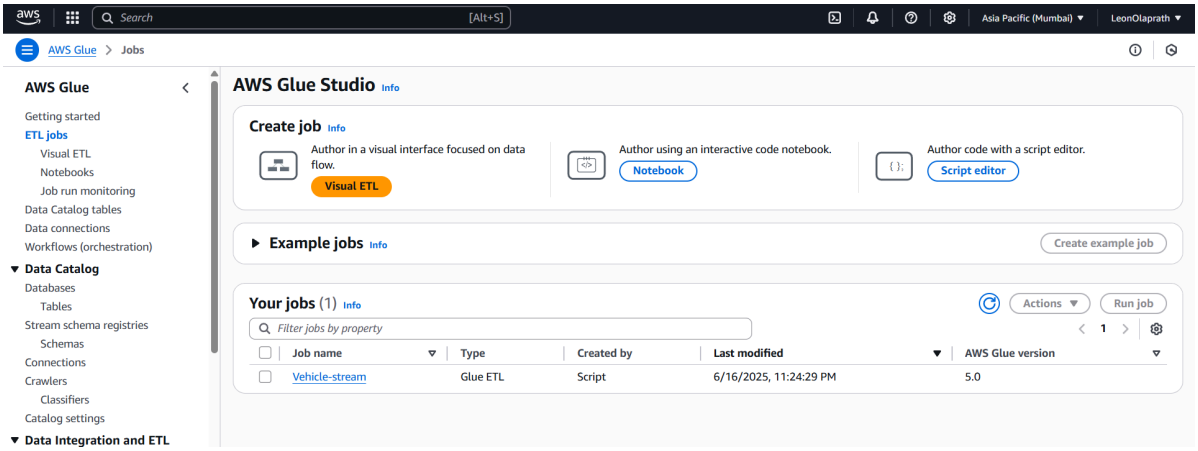
```
    .add("battery", DoubleType()) \
    .add("cTemp", DoubleType()) \
    .add("dtc", IntegerType()) \
    .add("eLoad", DoubleType()) \
    .add("iat", IntegerType()) \
    .add("imap", IntegerType()) \
    .add("kpl", DoubleType()) \
    .add("maf", DoubleType()) \
    .add("rpm", DoubleType()) \
    .add("speed", DoubleType()) \
    .add("tAdv", IntegerType()) \
    .add("tPos", IntegerType())


df = spark.readStream \
    .schema(schema) \
    .json("s3://leonrawbucket/2025/06/16/14/")

df_cleaned = df.fillna({
    "tripID": "unknown",
    "deviceID": "unknown",
    "timeStamp": "1970-01-01T00:00:00",
    "accData": "unknown",
    "gps_speed": 0.0,
    "battery": 0.0,
    "cTemp": 0.0,
    "dtc": 0,
    "eLoad": 0.0,
    "iat": 0,
    "imap": 0,
    "kpl": 0.0,
    "maf": 0.0,
    "rpm": 0.0,
    "speed": 0.0,
    "tAdv": 0,
    "tPos": 0
}).withColumn("event_time", to_timestamp("timeStamp",
"yyyy-MM-dd'T'HH:mm:ss"))

df_cleaned.writeStream \
```

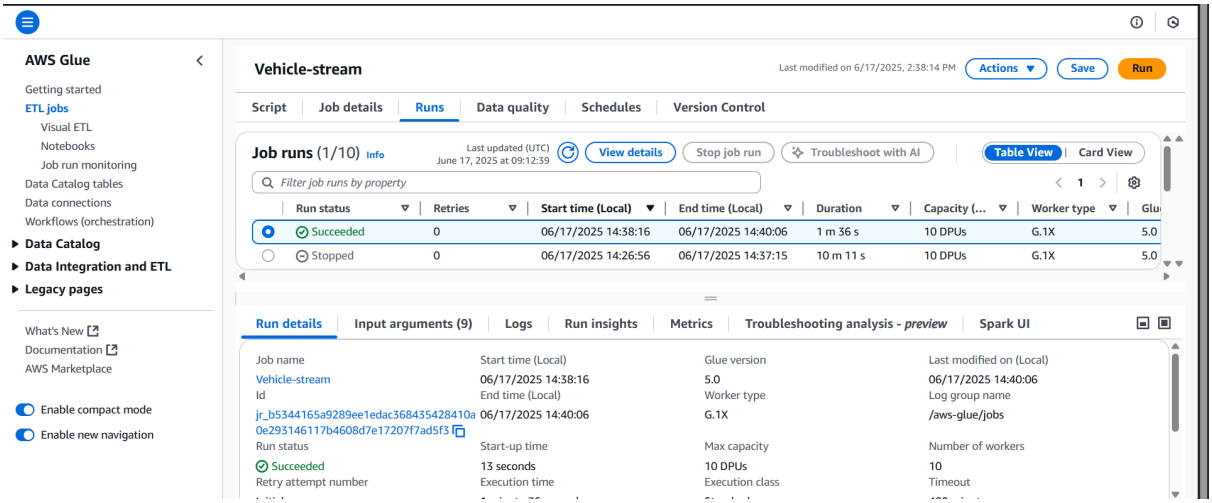
```
    .format("parquet") \
    .option("checkpointLocation",
"s3://leonprocessed/vehicleprocessed/_checkpoints/") \
    .option("path", "s3://leonprocessed/vehicleprocessed/") \
    .outputMode("append") \
    .start()

df_aggregated = df_cleaned.groupBy("deviceID").agg(
    avg("rpm").alias("avg_rpm"),
    avg("speed").alias("avg_speed"),
    avg("gps_speed").alias("avg_gps_speed"),
    avg("battery").alias("avg_battery"),
    avg("cTemp").alias("avg_coolant_temp"),
    avg("eLoad").alias("avg_engine_load"),
    avg("iat").alias("avg_intake_air_temp"),
    avg("imap").alias("avg_intake_pressure"),
    avg("kpl").alias("avg_kmpl"),
    avg("maf").alias("avg_mass_air_flow"),
    max("speed").alias("max_speed"),
    count("*").alias("record_count")
)

df_aggregated.writeStream \
    .format("parquet") \
    .option("checkpointLocation",
"s3://leonaggregated/vehiclehealth/_checkpoints/") \
    .option("path", "s3://leonaggregated/vehiclehealth/") \
    .outputMode("append") \
    .start() \
    .awaitTermination()
```

The processing done by this Glue script is :

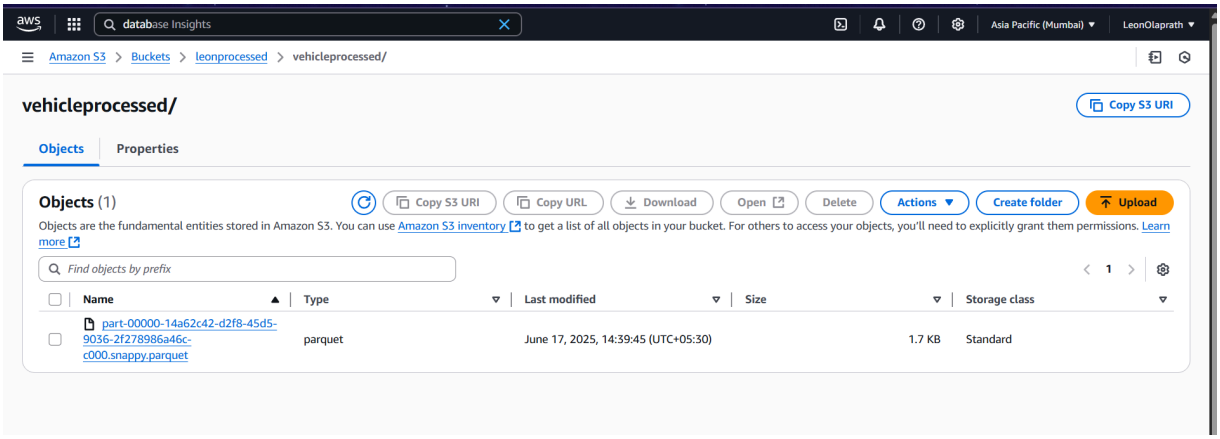
- Set up the Spark and Glue runtime environment.
- Defines schema for incoming JSON files.

- Reads streaming JSON telemetry data from the raw S3 path.
- Defines schema and fills missing rows with values
- Converts timestamp to timestamp type
- Sends cleaned data to s3
- Performs aggregation on the data and sends to s3

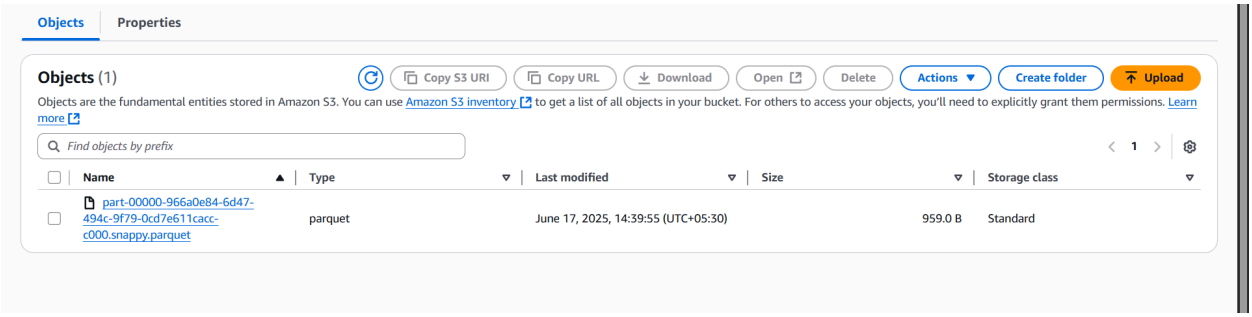




Cleaned Parquet file has been generated in s3



Aggregated data is stored in vehicleaggregated/

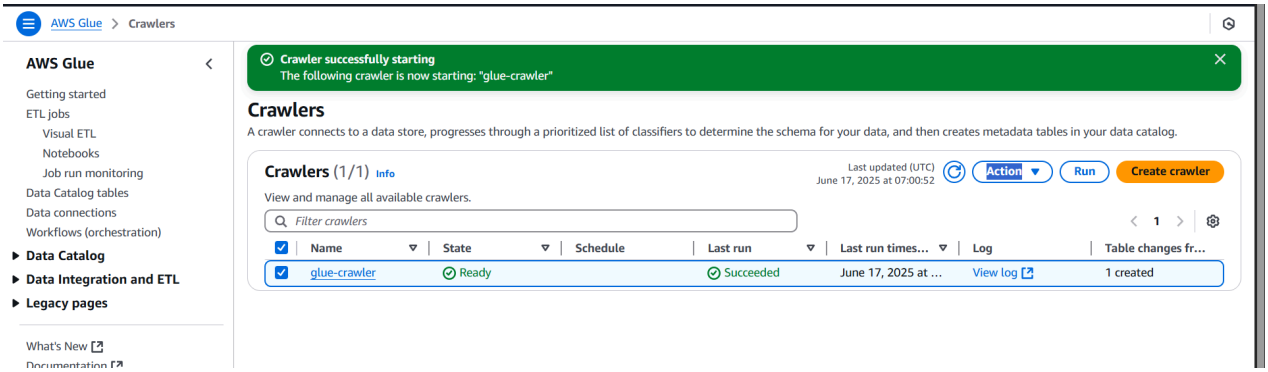


4. Data Storage Amazon S3

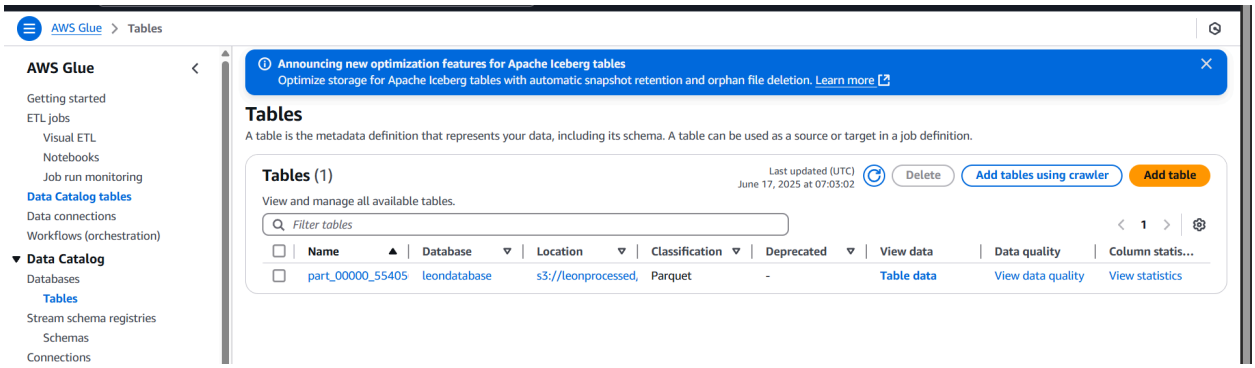
/raw : for storing the data from kinesis
/processed : for storing the processed data from glue
/aggregated : for aggregated data

5. Hive + Glue Catalog AWS Glue Crawler

- Create a Glue crawler.
- Choose the S3 bucket where the processed data is stored
- Give IAM role as per required.
- Create a database for storing the data.
- Run the crawler.

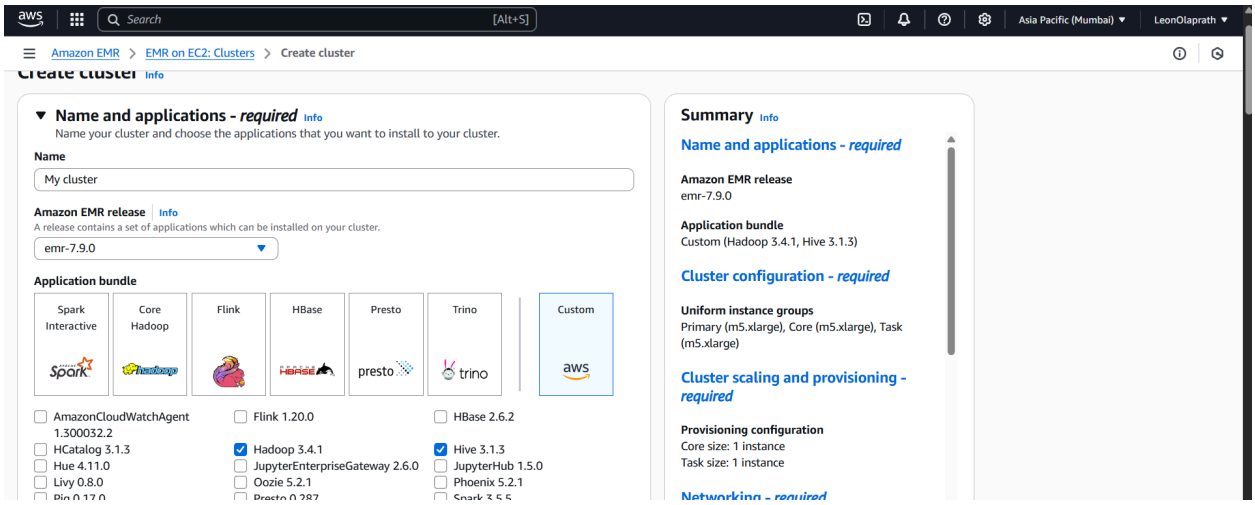


- Table will be created



Go to AWS EMR

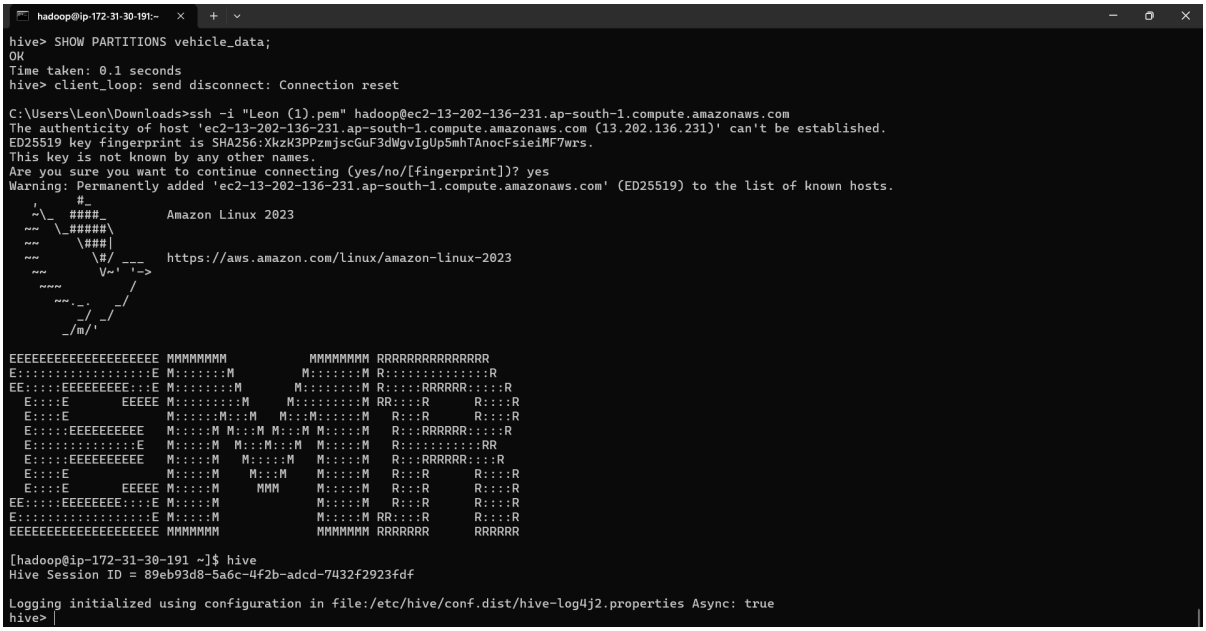
- Create cluster
- Choose Hive and hadoop in application bundle
- Create the required IAM roles
- Choose security groups and key pair



- In edit software settings add this so that it tells Hive to use the AWS Glue Data Catalog as the metastore instead of the default Hive metastore.

```
[
  {
    "classification": "hive-site",
    "properties": {
      "hive.metastore.client.factory.class":
"com.amazonaws.glue.catalog.metastore.AWSGlueDataCatalogHiveClient
Factory"
    }
  }
]
```

- In IAM permissions give glue full access
- After cluster is running connect into it through ssh
- Type hive
- Use database databasename
- Show tables ; will show the tables
- Query on the table created by the crawler



Now querying based on my use case i.e., focusing on vehicle health monitoring and predictive maintenance by analyzing historical telematics data such as RPM, engine load, coolant temperature, and battery. By identifying patterns like frequent overheating, high engine strain, or recurring faults, it enables the generation of maintenance recommendations and detection of long-term anomalies.

1. Detect High Engine Load

```
hive> SELECT tripID, deviceID, `timeStamp`, eLoad, 'High Engine Load Alert' AS alert
    > FROM vehicle_alerts
    > WHERE CAST(eLoad AS DOUBLE) > 80
    > LIMIT 20;
OK
tripid  deviceid        timestamp       eload   alert
1       0.0     2017-12-22 18:43:13     85.098  High Engine Load Alert
1       0.0     2017-12-22 18:45:21     83.9216 High Engine Load Alert
1       0.0     2017-12-22 18:45:39     83.1373 High Engine Load Alert
1       0.0     2017-12-22 18:49:27     83.9216 High Engine Load Alert
1       0.0     2017-12-22 18:49:28     83.9216 High Engine Load Alert
1       0.0     2017-12-22 18:49:29     83.9216 High Engine Load Alert
1       0.0     2017-12-22 18:50:48     86.6667 High Engine Load Alert
Time taken: 0.173 seconds, Fetched: 7 row(s)
hive>
```

2. Detect Low Battery

```
hive> SELECT tripID, deviceID, `timeStamp`, battery, 'Low Battery Alert' AS alert
    > FROM vehicle_alerts
    > WHERE CAST(battery AS DOUBLE) < 11.5
    > LIMIT 10;
OK
tripid  deviceid        timestamp       battery alert
2       0.0     2017-12-22 19:32:22     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:23     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:24     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:25     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:26     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:27     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:28     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:29     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:30     0.0     Low Battery Alert
2       0.0     2017-12-22 19:32:31     0.0     Low Battery Alert
Time taken: 0.147 seconds, Fetched: 10 row(s)
hive>
```

3. Detect High RPM

```
hive> SELECT tripID, deviceID, `timeStamp`, rpm, 'High RPM Alert' AS alert
    > FROM vehicle_alerts
    > WHERE CAST(rpm AS DOUBLE) > 2000
    > LIMIT 10;
OK
tripid  deviceid        timestamp       rpm     alert
2       0.0     2017-12-22 19:32:48     2413.0  High RPM Alert
2       0.0     2017-12-22 19:32:49     2115.25 High RPM Alert
2       0.0     2017-12-22 19:32:54     2049.75 High RPM Alert
1       0.0     2017-12-22 18:46:27     2222.25 High RPM Alert
1       0.0     2017-12-22 18:46:28     2159.5  High RPM Alert
Time taken: 0.118 seconds, Fetched: 5 row(s)
hive>
```

4. Detect overheating

```
hive> SELECT tripID, deviceID, `timeStamp`, cTemp, 'Overheating Alert' AS alert
    > FROM vehicle_alerts
    > WHERE CAST(cTemp AS DOUBLE) > 80
    > LIMIT 10;
OK
tripid  deviceid        timestamp       ctemp   alert
1       0.0     2017-12-22 18:46:53     81.0    Overheating Alert
1       0.0     2017-12-22 18:46:54     81.0    Overheating Alert
1       0.0     2017-12-22 18:46:55     81.0    Overheating Alert
1       0.0     2017-12-22 18:46:56     81.0    Overheating Alert
1       0.0     2017-12-22 18:46:57     81.0    Overheating Alert
1       0.0     2017-12-22 18:46:58     81.0    Overheating Alert
1       0.0     2017-12-22 18:46:59     81.0    Overheating Alert
1       0.0     2017-12-22 18:47:00     81.0    Overheating Alert
1       0.0     2017-12-22 18:47:01     81.0    Overheating Alert
1       0.0     2017-12-22 18:47:02     81.0    Overheating Alert
Time taken: 0.193 seconds, Fetched: 10 row(s)
hive>
```

5. Total anomaly detection and maintenance recommendations

```
hive> SELECT
    >   tripID,
    >   deviceID,
    >   `timeStamp`,
    >   cTemp,
    >   rpm,
    >   battery,
    >   eLoad,
    >   CASE
    >     WHEN CAST(cTemp AS DOUBLE) > 100 THEN 'Overheating Alert - Check coolant system'
    >     WHEN CAST(rpm AS DOUBLE) > 3000 THEN 'High RPM Alert - Inspect engine speed'
    >     WHEN CAST(battery AS DOUBLE) < 11.5 THEN 'Low Battery Alert - Charge or replace battery'
    >     WHEN CAST(eLoad AS DOUBLE) > 80 THEN 'High Engine Load Alert - Check engine efficiency'
    >     ELSE 'No Alert'
    >   END AS alert_and_recommendation
    > FROM vehicle_alerts
    > WHERE
    >   CAST(cTemp AS DOUBLE) > 100
    >   OR CAST(rpm AS DOUBLE) > 3000
    >   OR CAST(battery AS DOUBLE) < 11.5
    >   OR CAST(eLoad AS DOUBLE) > 80
    > LIMIT 20;
OK
tripid  deviceid        timestamp       ctemp   rpm     battery eload   alert_and_recommendation
2       0.0     2017-12-22 19:32:22     72.0    802.25  0.0     41.9608 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:23     72.0    800.0   0.0     41.5686 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:24     72.0    800.0   0.0     42.7451 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:25     72.0    798.0   0.0     42.7451 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:26     72.0    800.0   0.0     42.3529 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:27     72.0    801.25  0.0     43.1373 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:28     72.0    802.25  0.0     43.1373 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:29     72.0    799.5   0.0     42.7451 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:30     73.0    799.5   0.0     43.1373 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:31     73.0    800.25  0.0     43.5294 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:32     73.0    804.5   0.0     43.1373 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:33     73.0    800.25  0.0     43.1373 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:34     73.0    802.75  0.0     43.1373 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:35     73.0    802.75  0.0     42.7451 Low Battery Alert - Charge or replace battery
2       0.0     2017-12-22 19:32:36     73.0    790.25  0.0     42.7451 Low Battery Alert - Charge or replace battery
```

6. ClickHouse: Fast OLAP store

- Using clickhouse inside docker in local machine
- Open cmd and enter docker exec -it clickhouse-server clickhouse-client
- Create table and give s3 as source which has the aggregated data as parquet file

```sql
CREATE TABLE IF NOT EXISTS vehicle_aggregated
(
    `deviceID` Int32,
    `avg_rpm` Float64,
    `avg_speed` Float64,
    `avg_gps_speed` Float64,
    `avg_battery` Float64,
    `avg_coolant_temp` Float64,
    `avg_engine_load` Float64,
    `avg_intake_air_temp` Float64,
    `avg_intake_pressure` Float64,
    `avg_kmpl` Float64,
    `avg_mass_air_flow` Float64,
    `max_speed` Int32,
    `record_count` UInt64
)
ENGINE = MergeTree
ORDER BY deviceID

Query id: d00f9455-29bb-4c84-a6ad-e7adffbc3225
```

```sql
74b5640f5218 :) INSERT INTO vehicle_aggregated
SELECT * FROM file('https://your-bucket.s3.amazonaws.com/path/.parquet', 'Parquet');
```

```sql
SELECT *
FROM vehicle_aggregated

Query id: f239e574-a37d-4bda-b70a-2c849c0ea614
```

	deviceID	avg_rpm	avg_speed	avg_gps_speed	avg_battery	avg_coolant_temp	avg_engine_load	avg_intake_air_temp	avg_intake_pressure	avg_kmpl	avg_mass_air_flow	max_speed	record_count
1.	0	1124.199242614373	29.384042300829584	29.998940102933176	0	62.56371768734379	34.581072765401764	30.247932918255298	92.99405574330713	0	0	149	290869
2.	1	1062.3124756809339	13.549416342412451	16.260199688716256	11.238449708171094	66.19075875486381	25.655451108949407	28.215175097276266	33.67451361867704	0	0	70	10280
3.	2	800.034629684164	15.266399175439888	15.666477449753554	0	52.98280939409556	27.03654983067048	29.56964588088051	74.77696385187366	0	0	74	27166
4.	3	988.7691394919575	31.236624498820156	32.397744826607536	0	63.776404486785076	39.84849244832528	29.39200546311384	92.58291018992435	6.080822174292237	11.438218200009246	120	147901
5.	4	867.7676619433198	15.409514170040486	14.62251473684208	10.95265192307701	65.17125506072874	25.357664868421065	26.70668016194332	35.79524291497976	0	0	53	9880
6.	5	1013.2892456948913	23.120822811370346	23.551899557525793	0	57.403859321048095	19.85580595536884	22.216477578095667	86.11972214655407	4.11714487068104	11.896503861523577	123	271654
7.	6	974.6370998072493	20.15391848056039	20.764462752104095	11.557844012975393	62.49085609515303	19.053833322363715	18.891260401485592	62.688355037374826	1.9507741243947057	22.312920408068912	86	21271
8.	7	811.9440432562498	16.328652137031717	16.78789361136014	10.807958883830972	68.39213936030427	39.13450381111581	30.125871140507964	14.806106438539201	0.4639017092927183	1.4542575970105338	139	180367
9.	8	1199.2944606413994	21.956846684251932	22.482383525382875	2.533915960942013	66.57874033967329	24.992164810958545	29.203271784904437	74.89976398722754	3.5555162316627658	14.880817714841621	134	43218
10.	9	1116.5035176712274	21.518670045492016	22.730690781214435	10.651195801039002	61.08042229501333	29.872156494933446	26.755832617295688	32.10297420955987	14.832412813875358	4.339884176367188	138	458542
11.	10	884.3427058652062	25.81783550038242	28.11780324094692	0	56.38984820850738	32.94942278918224	22.531620553359684	80.60097503837575	6.088200661346451	9.251108364041333	119	747868
12.	11	447.6944129146172	5.233446840711313	5.172800201790898	10.411048051456826	59.26699457686972	42.65533102535	30.163576743599446	5.673981586580905	0	0	51	7929
13.	12	929.6043588331278	18.859492538316925	20.722081317994196	0	62.9027188402649	28.474416697568298	25.73252626652082	83.18561760910659	3.735652721167567	10.745577966898836	135	773418
14.	14	185.07110091743118	0	0	3.0104454638124367	6.549439347604485	5.375069520897049	4.186544342507645	8.899082568807339	0	0	0	1962
15.	16	923.3071375522808	18.58422389868272	20.313356554738803	0	65.06609076339757	36.45126436461438	36.717101200015634	85.0545205800727	4.472215538443503	9.300017980687455	120	127915

15 rows in set. Elapsed: 0.002 sec.

```
SELECT
    deviceID,
    avg_coolant_temp,
    avg_engine_load,
    avg_kmpl,
    max_speed,
    record_count
FROM vehicle_aggregated

Query id: 152492db-f0b4-4eb5-bb11-2ee388280c77

    ┌─deviceID─┬──avg_coolant_temp─┬─────avg_engine_load─┬─────────────avg_kmpl─┬─max_speed─┬─record_count─┐
 1. │        0 │  62.56371768734379 │  34.581072765401764 │                    0 │       149 │       290869 │
 2. │        1 │  66.19075875486381 │  25.655451108949407 │                    0 │        70 │        10280 │
 3. │        2 │  52.98280939409556 │   27.036549830670 48 │                    0 │        74 │        27166 │
 4. │        3 │ 63.7764044867850 76 │   39.848492448325 28 │   6.080822174292237 │       120 │       147901 │
 5. │        4 │  65.17125506072874 │  25.357664868421065 │                    0 │        53 │         9880 │
 6. │        5 │ 57.403859321048095 │   19.855805955368 84 │   4.11714487068104 │       123 │       271654 │
 7. │        6 │  62.49085609515303 │ 19.053833322363715 │ 1.9507741243947057 │        86 │        21271 │
 8. │        7 │  68.39213936030427 │  39.13450381111581 │ 0.4639017092927183 │       139 │       180367 │
 9. │        8 │  66.57874033967329 │ 24.992164810958545 │ 3.5555162316627658 │       134 │        43218 │
10. │        9 │  61.08042229501333 │ 29.872156494933446 │ 14.832412813875358 │       138 │       458542 │
11. │       10 │  56.38984820850738 │  32.94942278918224 │  6.088200661346451 │       119 │       747868 │
12. │       11 │  59.26699457686972 │     42.65533102535 │                    0 │        51 │         7929 │
13. │       12 │   62.9027188402649 │ 28.474416697568298 │  3.735652721167567 │       135 │       773418 │
14. │       14 │  6.5494393476044 85 │  5.375069520897049 │                    0 │         0 │         1962 │
15. │       16 │  65.06609076339757 │  36.45126436461438 │  4.472215538443503 │       120 │       127915 │
    └──────────┴────────────────────┴────────────────────┴────────────────────┴───────────┴──────────────┘

15 rows in set. Elapsed: 0.002 sec.
```

```
SELECT
    deviceID,
    avg_rpm,
    avg_speed,
    avg_gps_speed,
    avg_battery
FROM vehicle_aggregated

Query id: f8fe1889-72d2-4f29-999d-0a20d69183a5

    ┌─deviceID─┬──────────avg_rpm─┬─────────avg_speed─┬──────avg_gps_speed─┬──────avg_battery─┐
 1. │        0 │  1124.199242614373 │  29.384042300829584 │  29.998940102933176 │                0 │
 2. │        1 │ 1062.3124756809339 │  13.549416342412451 │  16.260199688716256 │ 11.238449708171094 │
 3. │        2 │   800.0346296841 64 │  15.266399175439888 │  15.666477449753554 │                0 │
 4. │        3 │  988.7691394919575 │  31.236624498820156 │  32.397744826607536 │                0 │
 5. │        4 │  867.7676619433198 │  15.409514170040486 │   14.62251473684208 │  10.95265192307701 │
 6. │        5 │ 1013.2892456948913 │  23.120822811370346 │  23.551899557525793 │                0 │
 7. │        6 │  974.6370998072493 │   20.15391848056039 │  20.764462752104095 │ 11.557844012975393 │
 8. │        7 │  811.9440432562498 │  16.328652137031717 │   16.78789361136014 │ 10.807958883830972 │
 9. │        8 │ 1199.2944606413994 │  21.956846684251932 │  22.482383525382875 │ 2.5339159609422013 │
10. │        9 │ 1116.5035176712274 │  21.518670045492016 │  22.730690781214435 │ 10.651195801039002 │
11. │       10 │   884.3427058652062 │   25.81783550038242 │   28.11780324094692 │                0 │
12. │       11 │   447.6944129146172 │   5.233446840711313 │   5.172800201790898 │ 10.411048051456826 │
13. │       12 │   929.6043588331278 │  18.859492538316925 │  20.722081317994196 │                0 │
14. │       14 │ 185.07110091743118 │                   0 │                   0 │ 3.0104454638124367 │
15. │       16 │   923.3071375522808 │  18.58422389868272 │ 20.313356554738803 │                0 │
    └──────────┴────────────────────┴────────────────────┴────────────────────┴──────────────────┘

15 rows in set. Elapsed: 0.002 sec.
```

7. Dashboard using streamlit in python

- Install streamlit and clickhouse-connect in python
- Connect to clickhouse client and the required table
- Use plotly for visualization

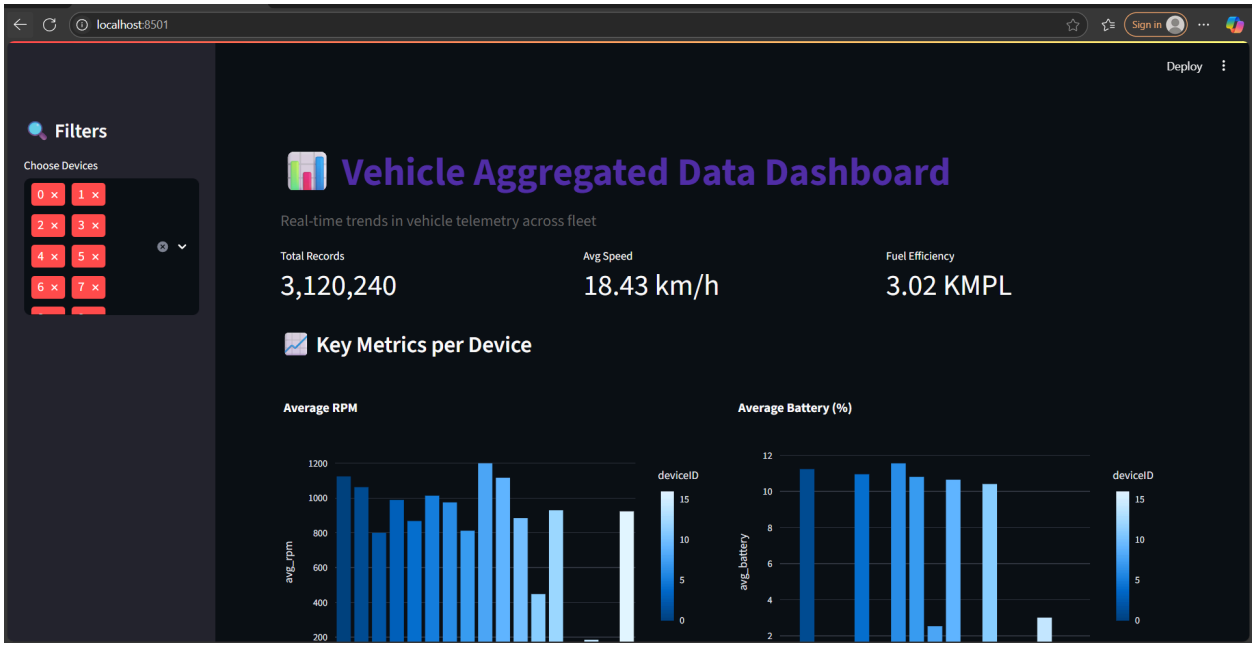
```
import streamlit as st
import clickhouse_connect
import pandas as pd
import plotly.express as px
import plotly.graph_objects as go


st.set_page_config(
    page_title="🚗 Vehicle Dashboard",
    layout="wide",
    initial_sidebar_state="expanded"
)


st.markdown("""
    <style>
        body {
            background-color: #f8f8fb;
            font-family: 'Segoe UI', sans-serif;
        }
        .css-1aumxhk, .css-ffhzg2 {
            background-color: #fff;
            border-radius: 12px;
            padding: 2rem;
            box-shadow: 0 2px 10px rgba(0,0,0,0.05);
        }
    </style>
""", unsafe_allow_html=True)



st.markdown("""
```

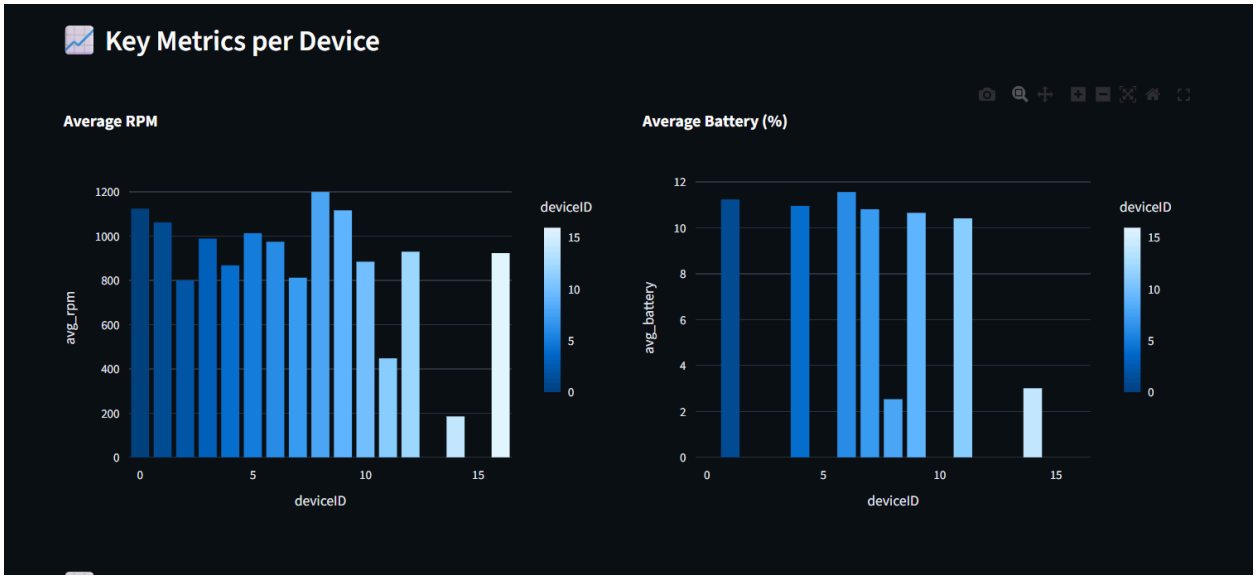
```
        <h1 style='font-size: 3em; color: #512da8;'>📊 Vehicle
Aggregated Data Dashboard</h1>
        <p style='font-size: 1.2em; color: #666;'>Real-time trends in
vehicle telemetry across fleet</p>
    """, unsafe_allow_html=True)




    st.sidebar.title("🔍 Filters")




    @st.cache_resource(ttl=600)
    def load_data():
        client = clickhouse_connect.get_client(
            host='localhost', port=8123, username='default',
password='mysecret', database='default')
        query = """
            SELECT * FROM vehicle_aggregated
            ORDER BY deviceID
        """
        return client.query_df(query)


    try:
        df = load_data()
    except Exception as e:
        st.error(f"❌ Error loading data: {e}")
        st.stop()


    devices = df['deviceID'].unique().tolist()
    selected_devices = st.sidebar.multiselect("Choose Devices",
devices, default=devices)
    filtered_df = df[df['deviceID'].isin(selected_devices)]


    col1, col2, col3 = st.columns(3)
```

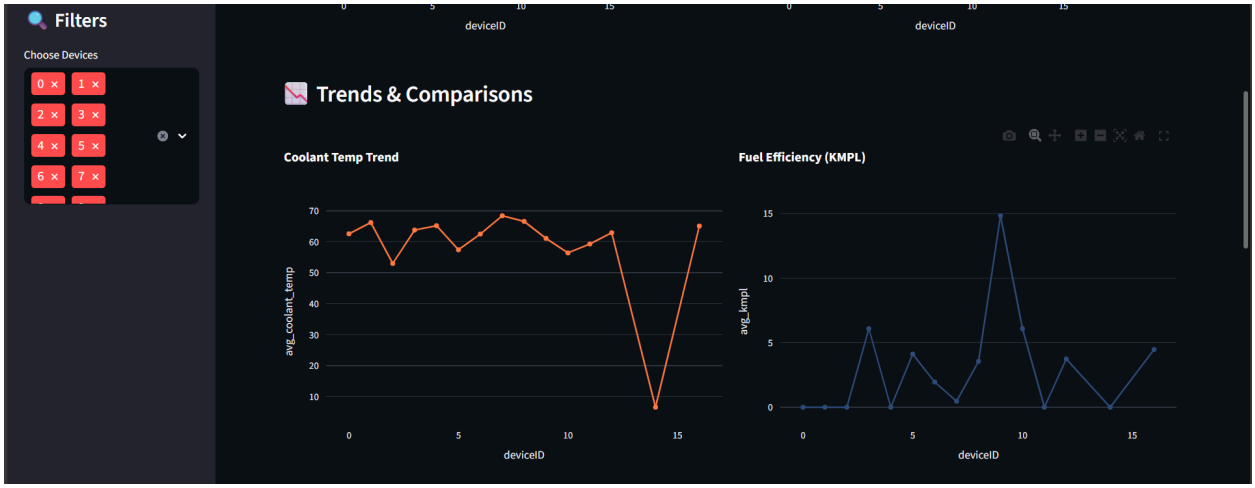
```
    col1.metric("Total Records",
f"{filtered_df['record_count'].sum():,}")
    col2.metric("Avg Speed", f"{filtered_df['avg_speed'].mean():.2f}
km/h")
    col3.metric("Fuel Efficiency",
f"{filtered_df['avg_kmpl'].mean():.2f} KMPL")


    st.markdown("### 📈 Key Metrics per Device")
    col4, col5 = st.columns(2)


    with col4:
        fig = px.bar(filtered_df, x='deviceID', y='avg_rpm',
color='deviceID',
                     color_discrete_sequence=['#ff6f61', '#6a1b9a'],
                     title="Average RPM")
        st.plotly_chart(fig, use_container_width=True)


    with col5:
        fig = px.bar(filtered_df, x='deviceID', y='avg_battery',
color='deviceID',
                     color_discrete_sequence=['#ffa600', '#a05195'],
                     title="Average Battery (%)")
        st.plotly_chart(fig, use_container_width=True)


    st.markdown("### 📉 Trends & Comparisons")
    col6, col7 = st.columns(2)


    with col6:
        fig = px.line(filtered_df, x='deviceID',
y='avg_coolant_temp', markers=True,
                     color_discrete_sequence=['#ff7c43'],
                     title="Coolant Temp Trend")
        st.plotly_chart(fig, use_container_width=True)
```

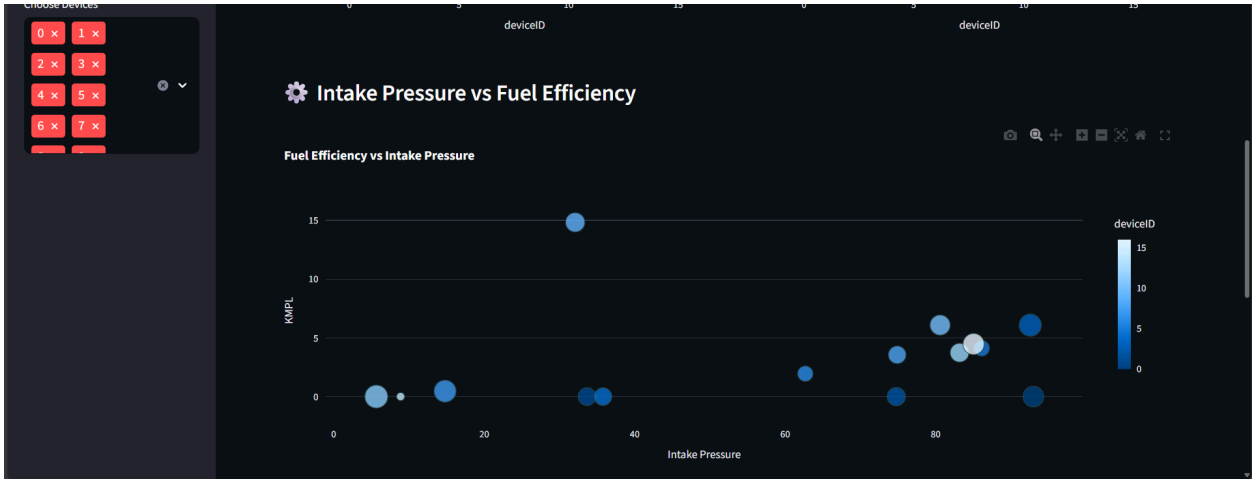
```
    with col7:
        fig = px.line(filtered_df, x='deviceID', y='avg_kmpl',
markers=True,
                      color_discrete_sequence=['#2f4b7c'],
                      title="Fuel Efficiency (KMPL)")
        st.plotly_chart(fig, use_container_width=True)


    st.markdown("### ⚙️ Intake Pressure vs Fuel Efficiency")
    fig = px.scatter(filtered_df, x='avg_intake_pressure',
y='avg_kmpl',
                     color='deviceID', size='avg_engine_load',

color_discrete_sequence=px.colors.qualitative.Prism,
                     title="Fuel Efficiency vs Intake Pressure",
                     labels={'avg_intake_pressure': 'Intake Pressure',
'avg_kmpl': 'KMPL'})
    fig.update_traces(marker=dict(opacity=0.8, line=dict(width=1,
color='DarkSlateGrey')))
    st.plotly_chart(fig, use_container_width=True)


    st.markdown("### 🧱 Engine Load Distribution")
    fig = px.box(filtered_df, x='deviceID', y='avg_engine_load',
color='deviceID',
                 color_discrete_sequence=px.colors.qualitative.Bold,
                 title="Engine Load Distribution per Device",
                 points="all")
    st.plotly_chart(fig, use_container_width=True)


    st.markdown("---")
    st.markdown("<p style='text-align:center;'>Designed by Leon
Jervis Olaprath</p>", unsafe_allow_html=True)
```

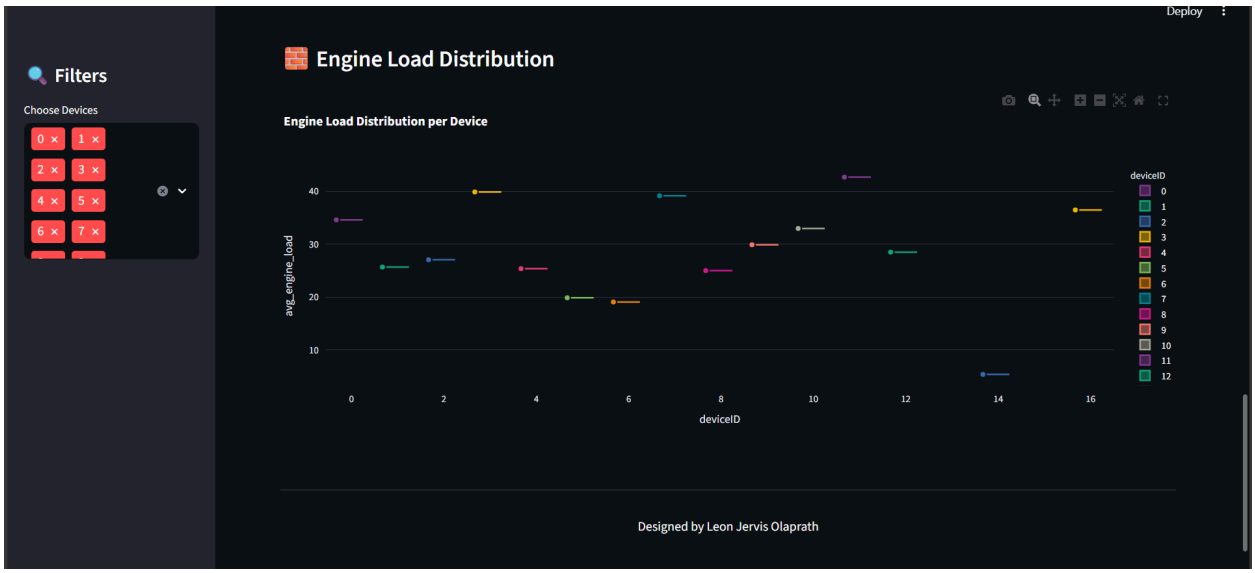
- Run streamlit using streamlit run file.py
- Make sure the clickhouse client is running
- The dashboard will be visible in localhost:8501


localhost:8501
Sign in
Deploy
Filters
Choose Devices
0 × 1 ×
2 × 3 ×
4 × 5 ×
6 × 7 ×
Vehicle Aggregated Data Dashboard
Real-time trends in vehicle telemetry across fleet
Total Records
3,120,240
Avg Speed
18.43 km/h
Fuel Efficiency
3.02 KMPL
Key Metrics per Device
Average RPM
avg_rpm
deviceID
Average Battery (%)
avg_battery
deviceID


Key Metrics per Device
Average RPM
avg_rpm
deviceID
Average Battery (%)
avg_battery
deviceID


Filters
Choose Devices
0 × 1 ×
2 × 3 ×
4 × 5 ×
6 × 7 ×
deviceID
deviceID
Trends & Comparisons
Coolant Temp Trend
avg_coolant_temp
70
60
50
40
30
20
10
0
5
10
15
deviceID
Fuel Efficiency (KMPL)
avg_kmpl
15
10
5
0
0
5
10
15
deviceID


0 × 1 ×
2 × 3 ×
4 × 5 ×
6 × 7 ×
deviceID
deviceID
Intake Pressure vs Fuel Efficiency
Fuel Efficiency vs Intake Pressure
KMPL
15
10
5
0
0
20
40
60
80
Intake Pressure
deviceID
15
10
5
0


Deploy
Filters
Choose Devices
0 × 1 ×
2 × 3 ×
4 × 5 ×
6 × 7 ×
Engine Load Distribution
Engine Load Distribution per Device
avg_engine_load
40
30
20
10
0
2
4
6
8
10
12
14
16
deviceID
deviceID
0
1
2
3
4
5
6
7
8
9
10
11
12
Designed by Leon Jervis Olaprath

REFERENCES

Kaggle :- https://www.kaggle.com/datasets/yunlevin/levin-vehicle-telematics