# SP25: DATA - 228 Sec -12 Big Data Technologies and Applications Homework -3

Name - Manav Anandani SJSU ID: 018194917

# **Mandatory question:**

Did you find or come across solutions to similar problems by using Generative AI or other sources?

**Ans -** No, didn't use any LLMs to generate the answers for the assignments.

#### **Problem Definition:**

For this single exercise homework, you're tasked with developing a real-time credit card fraud detection system using Apache Kafka, Kafka Streams (KSQLDB), MongoDB, and Flink. This system will ingest simulated (or real – your choice) credit card transaction data, analyze it for suspicious patterns, and generate alerts in real time.

# Step 1:

# **Set up the Environment:**

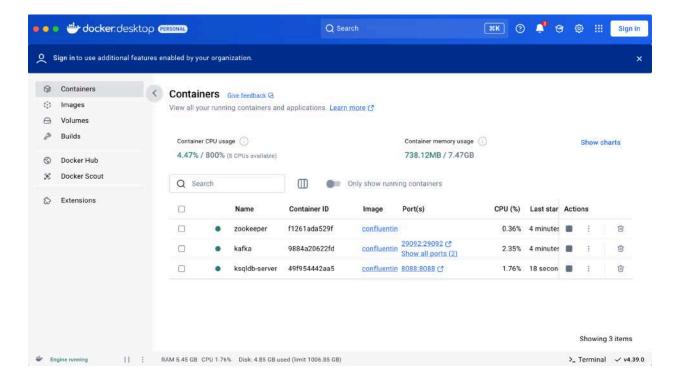
Here i manually stelled up the entire environment all the libraries and functions which were required by this assignment here i installed all the dependencies:

Used: pip install kafka-python confluent-kafka ksql kafka-streams faker pyflink pymongo

DATA-228 HW-3 018194917

```
spartan — -zsh — 98×30
(base) spartan@MLK-SCS-FVFHW4ULQ05N ~ % pip install kafka-python confluent-kafka ksql faker pyflin
k pymongo
Collecting kafka-python
  Using cached kafka_python-2.0.6-py2.py3-none-any.whl.metadata (9.0 kB)
Collecting confluent-kafka
  Using cached confluent_kafka-2.8.2-cp312-cp312-macosx_11_0_arm64.whl.metadata (22 kB)
Collecting ksal
  Using cached ksql-0.10.2.tar.gz (15 kB)
  Preparing metadata (setup.py) ... done
Requirement already satisfied: faker in /opt/anaconda3/lib/python3.12/site-packages (19.13.0)
Collecting pyflink
  Downloading pyflink-1.0-py3-none-any.whl.metadata (239 bytes)
Collecting pymongo
  Downloading pymongo-4.11.2-cp312-cp312-macosx_11_0_arm64.whl.metadata (22 kB)
Requirement already satisfied: requests in /opt/anaconda3/lib/python3.12/site-packages (from ksql)
Requirement already satisfied: six in /opt/anaconda3/lib/python3.12/site-packages (from ksql) (1.1
6.0)
Requirement already satisfied: urllib3 in /opt/anaconda3/lib/python3.12/site-packages (from ksql)
(2.2.3)
Collecting hyper (from ksql)
  Downloading hyper-0.7.0-py2.py3-none-any.whl.metadata (15 kB)
Requirement already satisfied: python-dateutil>=2.4 in /opt/anaconda3/lib/python3.12/site-packages
 (from faker) (2.9.0.post0)
Collecting dnspython<3.0.0,>=1.16.0 (from pymongo)
  Downloading dnspython-2.7.0-py3-none-any.whl.metadata (5.8 kB)
Collecting h2<3.0,>=2.4 (from hyper->ksql)
  Downloading h2-2.6.2-py2.py3-none-any.whl.metadata (27 kB)
Collecting hyperframe<4.0,>=3.2 (from hyper->ksql)
```

I personally used docker for setting up the entire tools which are required by the project here i created containers for kafka, zookeeper, ksqldb and kafka connect



Here we see that all the containers are up and running and we have installed all the required dependencies here which even include java and docker.

# Creation of Kafka topics

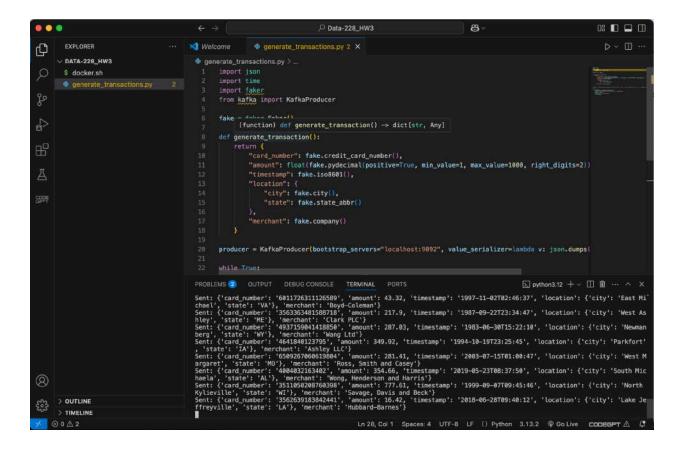
Here using the docker container for kafka we were able to create the kafka topic named **credit card transactions** 

Verification of the topics created

Now we will verify the created kafka topic which is **credit\_card\_transactions**:

# Step 2:

Generate Data Simulated using the Faker python package

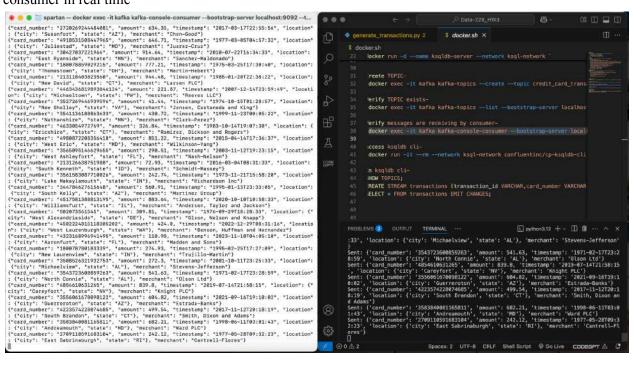


Step 3: Simulate transactions and send them to Kafka

Verification that the message are received by consumer in kafka

```
📵 🥮 📄 spartan — docker exec -it kafka kafka-console-consumer --bootstrap-server localhost:9092 --topic credit_card_trans...
(base) spartan@MLK-SCS-FVFHW4ULQ05N ~ % docker exec -it kafka kafka-console-consumer --bootstrap-server localhost:9092 -
topic credit_card_transactions -- from-beginning
{"card_number": "6011894551123607", "amount": 342.57, "timestamp": "1975-03-02T19:01:43", "location": {"city": "West Chri
stopher", "state": "AZ"}, "merchant": "Jones and Sons"}
"card_number": "4514391638447", "amount": 397.53, "timestamp": "2001-11-15T08:21:22", "location": {"city": "New Judyburg h", "state": "NM"}, "merchant": "Harvey Group"}
{"card_number": "38768369668239", "amount": 786.21, "timestamp": "2002-11-13T22:53:42", "location": {"city": "Nicoleville
     "state": "MO"}, "merchant": "Shannon Group"}
{"card_number": "565760482044", "amount": 926.88, "timestamp": "1990-04-14T00:32:25", "location": {"city": "New Darrenstad", "state": "RI"}, "merchant": "Smith, Brown and Haynes"}
"" state": "3508190301146924", "amount": 976.71, "timestamp": "2021-08-10T15:43:04", "location": {"city": "Palmerche ster", "state": "AZ"}, "merchant": "0lson and Sons"} 
{"card_number": "3580116946922974", "amount": 325.77, "timestamp": "2010-05-13T17:34:44", "location": {"city": "Lake Kristin", "state": "CA"}, "merchant": "Jackson-Lawrence"} 
{"card_number": "5398972959522284", "amount": 471.41, "timestamp": "1974-03-08T02:07:11", "location": {"city": "Jessicato"}
wn", "state": "VA"), "merchant": "Young-Cannon"} {"card_number": "4692160257275", "amount": 322.56, "timestamp": "1989-09-29T14:16:23", "location": {"city": "Port Elizabe th", "state": "CO"}, "merchant": "Dean-Sanchez"}
tn", "state": "CU"), "mercnant": "Dean-Sancnez"} 
{"card_number": "3589253873880841", "amount": 930.36, "timestamp": "2013-07-03T16:44:53", "location": {"city": "Lake Case yfurt", "state": "KY"}, "merchant": "Williams, Valdez and Torres"} 
{"card_number": "180028263494786", "amount": 199.97, "timestamp": "1988-12-23T03:50:24", "location": {"city": "New Charle ston", "state": "WI"}, "merchant": "Khan and Sons"} 
{"card_number": "6011688461666097", "amount": 445.12, "timestamp": "2017-12-17T17:55:41", "location": {"city": "Elizabeth furt", "state": "CO"}, "merchant": "Kennedy PLC"} 
{"card_number": "180084926118775", "amount": 244.42, "timestamp": "1973-10-14T06:03:14", "location": {"city": "Smithberg"
, "state": "PR"}, "merchant": "Cain, Ferguson and Griffin"}
{"card_number": "3584330188640600", "amount": 186.99, "timestamp": "1999-07-08T22:47:42", "location": {"city": "Kevinview
     "state": "NE"}, "merchant": "Burke and Sons"}
{"card_number": "4564083468634684", "amount": 984.25, "timestamp": "2022-09-09T07:07:25", "location": {"city": "East Tony
    "state": "NY"}, "merchant": "Rice Inc"}
{"card_number": "4367117201219", "amount": 570.45, "timestamp": "1984-05-29T03:42:06", "location": {"city": "New Kevinbur y", "state": "NJ"}, "merchant": "Thomas-Lee"}
```

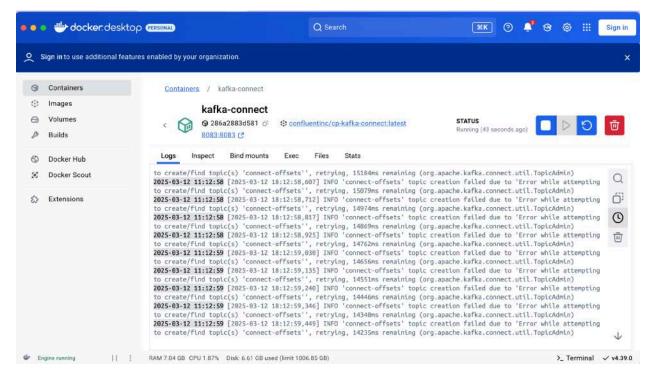
Now here we can see that messages are generated in real time from faker and are received by kafka consumer in real time



Step 4:
Real-time Fraud Detection with KSQLDB:

Starting and working with Kafka connect:

1. Configuring kafka connect using docker:



Here we used kafka connect using docker and here we see this is up and running and now and here we will use this to transfer the stream data from Kafka to ksqlDB

Setting up and installing of ksqlDB:

```
🌘 🌔 🌑 🚞 spartan — docker run -it --rm --network ksql-network confluentinc/cp-ksqldb-cli:latest http://ksqldb-server:8088 — 139×35
[(base) spartan@MLK-SCS-FVFHW4ULQ05N ~ % docker run -it --rm --network ksql-network confluentinc/cp-ksqldb-cli:latest http://ksqldb-server:8
Unable to find image 'confluentinc/cp-ksqldb-cli:latest' locally
latest: Pulling from confluentinc/cp-ksqldb-cli
409dc24a3db4: Pull complete
f37a8f90be5f: Pull complete
25177ac183d3: Pull complete
5acb731839c2: Pull complete
13c5de150e94: Pull complete
f44d477670c9: Pull complete
Digest: sha256:68a87a469e577c6b5c02f1a39a80d901f5af35a5027609ee77d525d010ef8fa7
Status: Downloaded newer image for confluentinc/cp-ksqldb-cli:latest
Debug Options:
                                         (_)
                               The Database purpose-built
                               for stream processing apps
Copyright 2017-2022 Confluent Inc.
CLI v7.9.0, Server v7.9.0 located at http://ksqldb-server:8088
Server Status: RUNNING
Having trouble? Type 'help' (case-insensitive) for a rundown of how things work!
ksql>
```

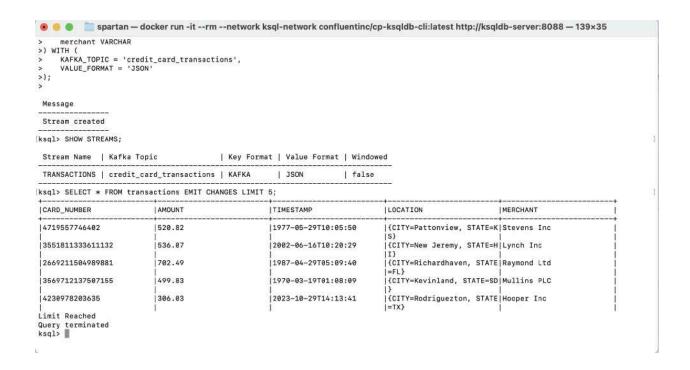
Now here we are inside the ksqlDB and here on we are going to perform various queries on this:

Create a ksqlDB stream to consume the transactions received from kafka



Here we created a ksql stream which will allow us to query the consumed data from kafka

 Now we will do the verification of the stream created and view the incoming data from kafka.



## **KSQL UDF:**

Here we defined the user defined function to classify the fraud detection with my own rules here:

#### **Rules for fraud detection:**

- 1. Any transactions exceeding over 500 will be detected as fraud
- 2. Transactions happening in an unusual locations are also flagged as fraud
  - Create a new kafka topic named as flagged transactions which would only contain the flagged transactions data in it:

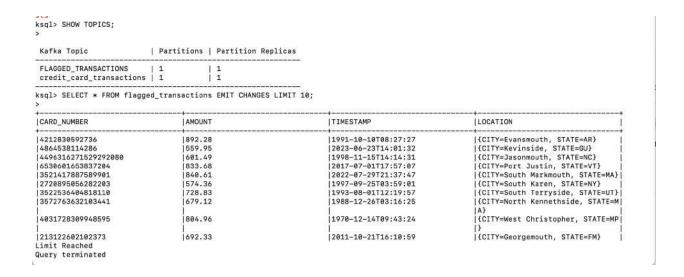
```
ksql> CREATE STREAM flagged_transactions AS
>SELECT card_number, amount, timestamp, location
>FROM transactions
|>WHERE amount > 500;

Message

Created query with ID CSAS_FLAGGED_TRANSACTIONS_1
|ksql> |
```

This shows that here we have successfully created a flagged\_transactions topic which will help us to identify the flagged transactions here.

 Verification of our UDF and the fraud detection system that we created it in a correct manner or not:



Now here we can see that the real time fraud data is being generated from our system in real time this table shows the verification of our fraud detection system in ksqlDB and we can perform various queries and analysis on the very same data.

• Create a new ksqldb stream to detect the fraud transactions

Creation and verification of streams:

ksql> Show streams;					
Stream Name	Kafka Topic	Key Format	Value Form	nat	Windowed
FLAG_TRANSACTIONS	flag_transactions	KAFKA	JSON	1	false
HIGH_RISK_MERCHANTS	HIGH_RISK_MERCHANTS	KAFKA	JSON	- 1	false
TRANSACTIONS	credit_card_transactions	KAFKA	JSON		false

Now i have created all the streams and all the topics which are required now we will perform our analysis on the real credit card transactions data using transactions stream and on flagged data using the flag\_transactions stream.

Kafka Topic	Partitions   Par	rtition Replicas		
FLAGGED_TRANSACTIONS   credit_card_transactions				
ksql> SELECT * FROM flagged_ >	transactions EMI	IT CHANGES LIMIT 10;		
CARD_NUMBER	AMOUNT		TIMESTAMP	LOCATION
4212830592736	1892.28		1991-10-10T08:27:27	{CITY=Evansmouth, STATE=AR}
4864538114286	1559.95		2023-06-23T14:01:32	{CITY=Kevinside, STATE=GU}
4496316271529292080	601.49		11998-11-15T14:14:31	{CITY=Jasonmouth, STATE=NC}
6530601653837204	833.68		2017-07-01T17:57:07	{CITY=Port Justin, STATE=VT}
3521417887589901	840.61		2022-07-29T21:37:47	{CITY=South Markmouth, STATE=MA}
2720895056282203	574.36		1997-09-25T03:59:01	{CITY=South Karen, STATE=NY}
3522536404818110	728.83		1993-08-01T12:19:57	{CITY=South Terryside, STATE=UT}
3572763632103441	679.12		1988-12-26T03:16:25	{CITY=North Kennethside, STATE=M  A}
4031728309948595	804.96		1970-12-14T09:43:24	{CITY=West Christopher, STATE=MP  }
213122602102373 Limit Reached Query terminated	692.33		2011-10-21T16:10:59	{CITY=Georgemouth, STATE=FM}

# Query and analysis on both streams and topics:

1. Flagging transactions with high risk merchants if transactions occur at fraud prone merchants

```
ksql> CREATE STREAM high_risk_merchants AS
>SELECT * FROM transactions

[>WHERE merchant IN ('ScamCorp', 'FraudTech', 'ShadyBiz');

Message

Created query with ID CSAS_HIGH_RISK_MERCHANTS_3

ksql>
```

2. Query using tumbling window

ksql> SELECT card\_number, COUNT(\*) AS txn\_count

>FROM transactions >WINDOW TUMBLING (SIZE 10 MINUTES) >GROUP BY card\_number |>EMIT CHANGES; | CARD\_NUMBER |TXN\_COUNT 180044116617390 |4432759808142823 |373557418074322 |1 |1 |1 |1 |1 |1 |1 |1 |1 |1 |1 |1 |1 3500857371720774 503802807415 30152148983623 4214403765160568667 |676306583987 |4013074380364572 4198263463950 3580756536538902 2709052543224092 13557339998772712 6011061451253385 30404529548156 36000024018507 3565266993554854 30329078031207 4950157915010 4532006619964782 |6011537616233388 ^CQuery terminated ksql>

3. Transactions with particular card number:

CARD_NUMBER	AMOUNT	TIMESTAMP	LOCATION	MERCHANT
 	355.75 	1985-07-25T17:37:23	{CITY=North Cassieborough  , STATE=OR}	Campbell LLC

4. Average transactions amount by account number in flag transactions

```
| Ksql > SELECT | CARD_NUMBER | AVG_TRANSACTION_AMOUNT | SEMIT CHANSACTION_AMOUNT | SEMIT CHANSACTIONS | SEMIT CHANSACTION | SEMIT CHANSACTION | SEMIT CHANSACTION | SEMIT CHANSACTION_AMOUNT | SEMIT CHANSACTION_
```

# 5. Counting transactions per location

## Step 5:

## **Store Data in MongoDB:**

For storing the data in mongo first of all:

- 1. We started the ming db server and check mongodb is correctly tunning on our system.
- 2. We created a file called **mongodb\_writer.py** which contains the logic fromm which the flagged data will be stored in mongodb by using connection from kafka and ksqldb

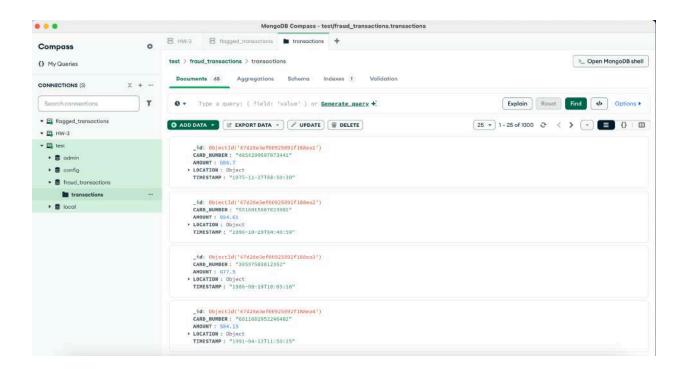
```
(base) spartan@MLK-SCS-FVFHW4ULQ05N Data-228_HW3 % python mongodb_writer.py
Listening for flagged transactions...

Ln 10, Col 10 Spaces: 4 UTF-8 LF {} Python 3.13.2 @ Go Live
```

```
generate_transactions.py 2 $ docker.sh
                                                                                                       mongodb_writer.py 2 ×
 mongodb_writer.py > ...
              from pymongo import MongoClient
            from kafka import KafkaConsumer
            client = MongoClient('localhost', 27017)
            db = client['fraud_transactions']
collection = db['transactions']
              consumer = KafkaConsumer(
                    'flag_transactions',
                      bootstrap_servers='localhost:9092',
                     auto_offset_reset='earliest',
                     value_deserializer=lambda v: json.loads(v.decode('utf-8'))
             print("Listening for flagged transactions...")
             for msg in consumer:
                 transaction = msg.value
                     print(f"Inserting transaction: {transaction}")
                    collection.insert_one(transaction)
 PROBLEMS 🐠 OUTPUT DEBUG CONSOLE TERMINAL PORTS
5-18-23108:23:04'}
Inserting transaction: {'CARD_NUMBER': '30532191504071', 'AMOUNT': 632.38, 'LOCATION': {'CITY': 'Julieshire', 'STATE': 'KY'}, 'TIMESTAMP': '2018-12-10T11:51:42'}
Inserting transaction: {'CARD_NUMBER': '3531301620809214', 'AMOUNT': 613.78, 'LOCATION': {'CITY': 'Matthewchester', 'STATE': 'RI'}, 'TIMESTAMP': '1993-01-09707:13:58'}
5-10-23T00:23:04'}
Inserting transaction: {'CARD_NUMBER': '348912036336458', 'AMOUNT': 501.56, 'LOCATION': {'CITY': 'Sonyashire', 'STATE': 'VI'}, 'TIMESTAMP': '1996—12-14708:49:57'}
Inserting transaction: {'CARD_NUMBER': '3587913350360631', 'AMOUNT': 632.36, 'LOCATION': {'CITY': 'Garciamouth', 'STATE': 'CO'}, 'TIMESTAMP': '20 14-09-10711:36:52'}
Inserting transaction: {'CARD_NUMBER': '3527547566161483', 'AMOUNT': 639.67, 'LOCATION': {'CITY': 'Lewischester', 'STATE': 'SC'}, 'TIMESTAMP': '1 979-04-05717:37:17'}
Inserting transaction: {'CARD_NUMBER': '375671984308119', 'AMOUNT': 923.66, 'LOCATION': {'CITY': 'Tracyview', 'STATE': 'AL'}, 'TIMESTAMP': '197-01-15705:09:48')
Inserting transaction: {'CARD_NUMBER': '343481851952539', 'AMOUNT': 648.33, 'LOCATION': {'CITY': 'South Moniqueburgh', 'STATE': 'RI'}, 'TIMESTAMP': '1974-10-28720:18:00'}
Inserting transaction: {'CARD_NUMBER': '340689159257186', 'AMOUNT': 623.46, 'LOCATION': {'CITY': 'Port Raymond', 'STATE': 'AK'}, 'TIMESTAMP': '197-01-10 COL10 Sarges: A LUTE-8 LE () Puthon 3.33 2 @ Golive
```

Now here we see that mongo db is listening and presenting the flagged transitions all in one:

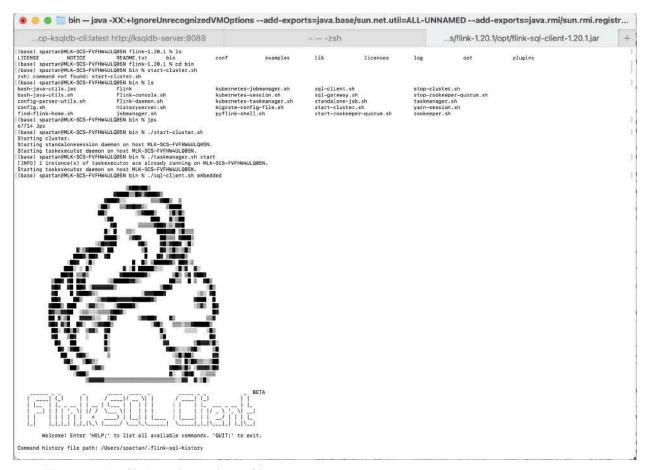
Now storing the data in mongodb



Step 6:

# **Real-time Analytics with Apache Flink:**

Installing apache Flink on the system

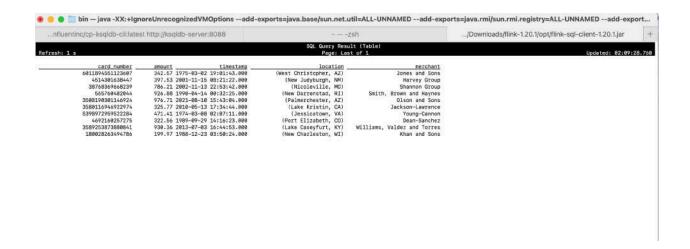


Installing apache flink and creation of jars:



Here the apache flink is running here and we have created a jar to run our system here:

# Data generation on apache flink:

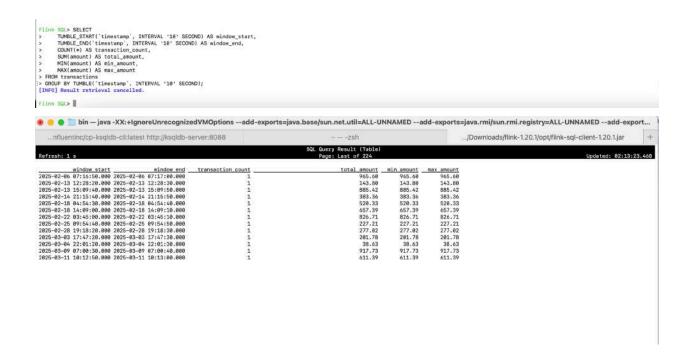


Here we see that fraud data is correctly generated in apache flink here in real time from ksqldb

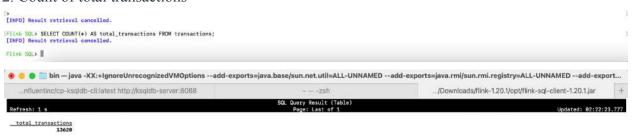
#### Table creations:

Performing and analysis on apache flink:

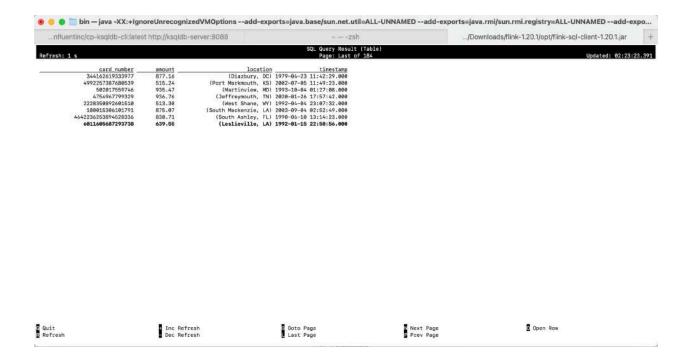
Querying on transactions data in flink:



## 2. Count of total transactions



## 3. Query where transactions are greater than 500



• Querying on the flagged data in flink

```
[Flink SQL) cfact TABLE FLAG_TRANSACTIONS;
[INFO] Execute statement succeeded.

Flink SQL> CRAD_NAMER_STRING.

AMGLAT DECIMAL(18, 2),

'IMESTAMP'STRING, — Change to STRING to avoid parsing issues

LOCATION ROW-CITY STRING, STATE STRING>

> WITH (

'connector' = 'kafka',

'topic' = 'flag_transactions',

'properties.bootstrap.servers' = 'localhost:9892',

'properties.group.id' = 'flink_consumer_group',

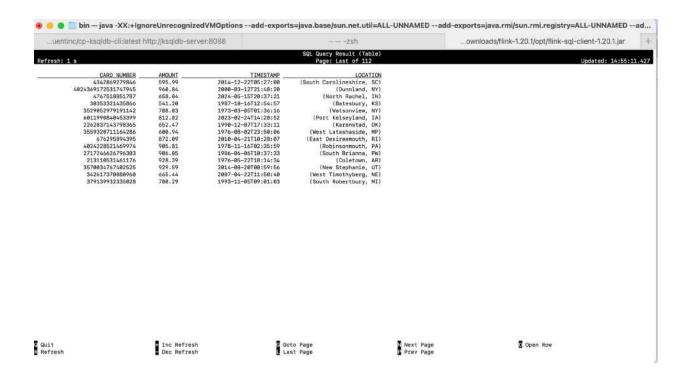
'scan.startup.mode' = 'earliest-offset',

'format' = 'json'

| Tipo | Execute statement succeeded.

Flink SQL> SELECT * FROM FLAG_TRANSACTIONS LIMIT 10;

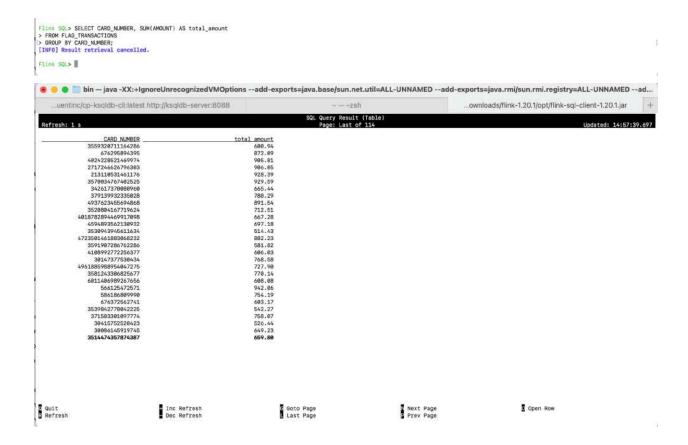
Flink SQL> SELECT = Transactions | Transactions
```



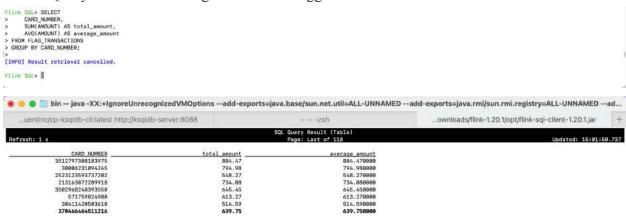
Here the flagged data from flag\_ransctions is visible in flink from ksqlDB

- Querying on the data
- 1. Group transactions based on amount ranges

2. Query to find amount by group by card number



3. Query to find the average amount of flagged transactions from a card



4. Query to group by flagged transactions by location

```
Flink SQL> SELECT LOCATION.CITY AS CITY, SUM(AMOUNT) AS TOTAL_FLAGGED_AMOUNT > FROM FLAG_TRANSACTIONS > GROUP BY LOCATION.CITY; |> [INFO] Result retrieval cancelled.
Flink SQL> ||
```



#### 7. Reflect and share:

## **Summary:**

Students gained practical knowledge of using Kafka alongside MongoDB and Flink for creating a live fraud detection solution through this assignment.

## **Challenges:**

**Kafka Integration:** Ensuring smooth communication between producer and consumer.

The system efficiently processes massive streams of current data.

**Data Synchronization:** Ensuring data integrity between Kafka and MongoDB.

# **Insights:**

Real-time analytics stands essential to prevent fraud from occurring.

The combination of Kafka and MongoDB and Flink provides powerful solutions although their integration process becomes complex.

Performance improvement in systems demands both data validation and asynchronous processing steps.

#### **Technical Difficulties:**

The application two systems achieved synchronization between Kafka offset management and MongoDB high-throughput performance.

## **Best Practices:**

Efficient error handling and data validation.

The system inserts data asynchronously into MongoDB to achieve better performance.

Blog link for more insights and detailed information:

https://github.com/manavanandani/Real time Fraud detection system

For more detailed information i have shared the link for my blog for reflect and share .