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Department of Computer Science and Engineering B. Tech. CSE - 6th Semester Jan - May 2024

UE21CS343BB3 DATABASE TECHNOLOGIES (DBT)

PROJECT REPORT on

"CryptoStream: Real-time Cryptocurrency Data Analysis".

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1.Introduction:

CryptoStream a cutting-edge platform for real-time cryptocurrency data analysis. Leveraging Apache Spark, Kafka, and SQLite,

CryptoStream enables users to access live cryptocurrency data from a leading exchange via WebSocket connections. Through seamless integration with Spark's distributed computing capabilities, users can perform advanced data processing, transformations, and aggregations, gaining valuable insights into cryptocurrency price movements, trading volumes, and market trends.

With CryptoStream, users can stay ahead of the curve in the dynamic world of cryptocurrencies.

2.Installation of Software

Spark installation:

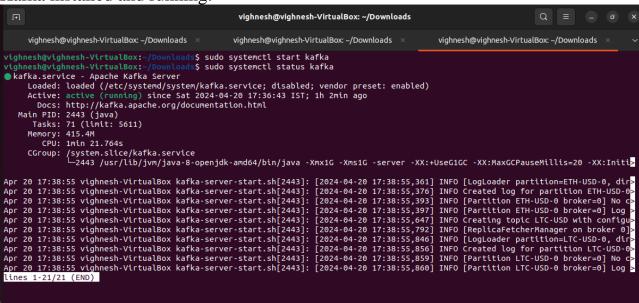
```
vighnesh@vighnesh-VirtualBox:-/Downloads$ spark-shell
24/04/20 18:30:13 MARN Utils: Your hostname, vighnesh-VirtualBox resolves to a loopback address: 127.0.1.1; using 10.0.2.15 instead (on interface enp0s3)
24/04/20 18:30:13 WARN Utils: Set SPARK_LOCAL_IP if you need to bind to another address
Setting default log level to "MARN".
To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
24/04/20 18:30:35 WARN NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applicable
Spark context Web UI available at http://10.0.2.15:4040
Spark context available as 'sc' (master = local[*], app id = local-1713618039059).
Spark session available as 'spark'.
Welcome to

Vight Spark context web UI available at http://olian.com/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/library/libr
```

spark running:

```
vighnesh@vighnesh-VirtualBox:~/Downloads × vighnesh@vighnesh-VirtualBox:~/Downloads × vighnesh@vighnesh-VirtualBox:~/Downloads / Spark/sbin/start-all.sh starting org.apache.spark.deploy.master.Master.logging to /home/vighnesh/Downloads/spark/logs/spark-vighnesh-org.apache.spark.deploy.master.Haster-1-vighnesh-VirtualBox.out localhost: starting org.apache.spark.deploy.worker.Worker, logging to /home/vighnesh/Downloads/spark/logs/spark-vighnesh-org.apache.spark.deploy.worker.logging to /home/vighnesh/Downloads/spark/logs/spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.apache.spark-vighnesh-org.
```

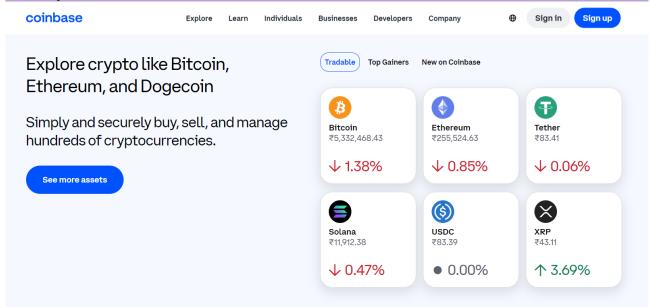
Kafka installed and running:



3.Input Data:

Source:

We are using the Coinbase Pro WebSocket API as the source for our input data.



The input data for our project consists of real-time ticker data for various cryptocurrencies, including Bitcoin (BTC), Ethereum (ETH), and Litecoin (LTC). This data is sourced from the Coinbase Pro WebSocket API, which provides a

continuous feed of market updates, including price changes, trading volume, best bid/ask prices, and other relevant metrics.

The Coinbase Pro WebSocket API offers a reliable and efficient way to access cryptocurrency market data in real time, making it an ideal data source for our project. By subscribing to specific channels and product IDs, we can tailor the data stream to our requirements and capture relevant information for analysis and visualization.

4.Streaming Mode Experiment

In the streaming mode experiment, we aimed to analyze real-time cryptocurrency market data using Apache Spark Streaming, Kafka, and SQLite. We set up a WebSocket connection to the Coinbase Pro API to receive live ticker data for Bitcoin (BTC), Ethereum (ETH), and Litecoin (LTC). This data was then ingested into Kafka topics using a Kafka producer. Meanwhile, Spark Streaming was utilized to consume the data from Kafka topics, perform various transformations and aggregations, and finally, store the processed data in a SQLite database.

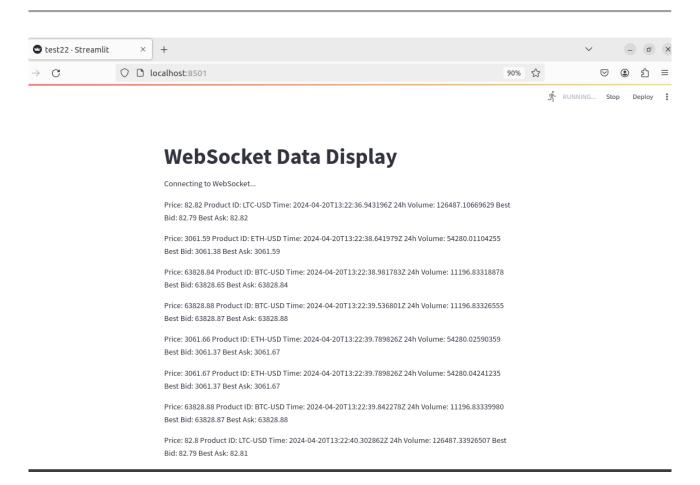
Streaming data using spark and kafka:

```
You can now view your Streamlit app in your browser.

Local URL: http://localhost:8501
Network URL: http://lo.0.2.15:8501

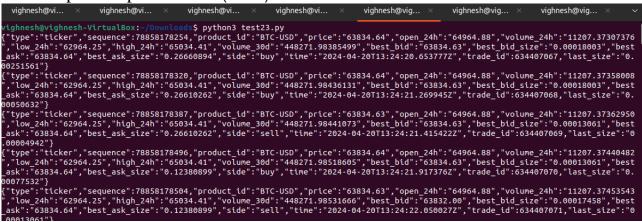
24/04/20 18:52:27 WARN Utils: Your hostname, vighnesh-VirtualBox resolves to a loopback address: 127.0.1.1; using 10.0.2.15 instead (on interface enp0s3)
24/04/20 18:52:27 WARN Utils: Set SPARK_LOCAL_IP if you need to bind to another address
Setting default log level to "WARN".
To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
24/04/20 18:52:30 WARN NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applicable
24/04/20 18:52:34 WARN Utils: Service 'SparkUI' could not bind on port 4040. Attempting port 4041.
The socket is open
```

1. **Data Ingestion Window:** This window represents the duration in which the WebSocket connection is open, and data is continuously received from the Coinbase Pro API.



Processing Window: Spark Streaming processes the incoming data in micro-batches, typically with a window size of a few seconds, to perform transformations and aggregations.

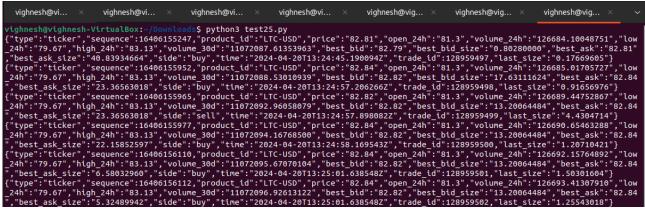
Kafka topic1 output Bitcoin (BTC)



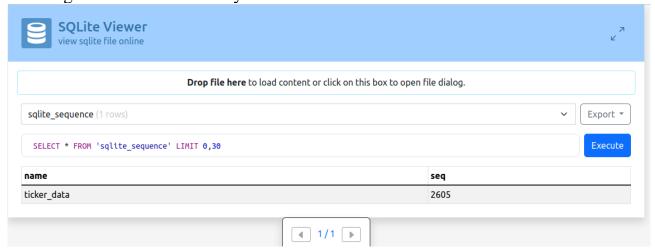
Kafka topic2 output Ethereum (ETH),

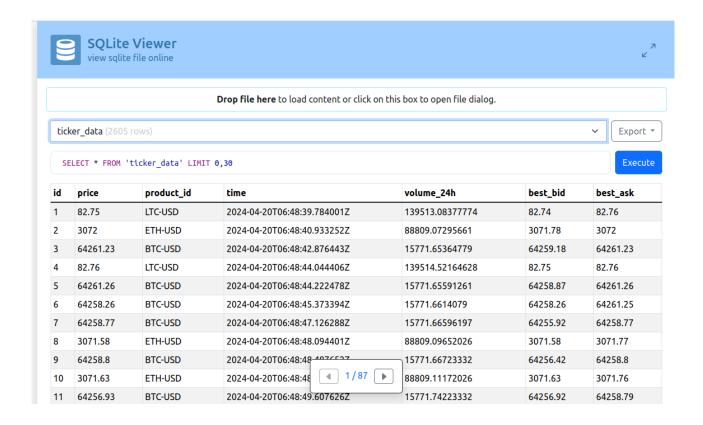
```
vighnesh@vi... × vighnesh@vi... × vighnesh@vi... × vighnesh@vig... vighnesh@vig... × vighnesh@vig... × vighnesh@vig
```

Kafka topic3 output Litecoin (LTC)



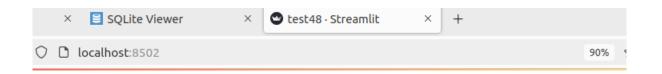
Storage Window: Once processed, the data is stored in the SQLite database, allowing for retrieval and analysis at a later time.





5.Batch Mode Experiment

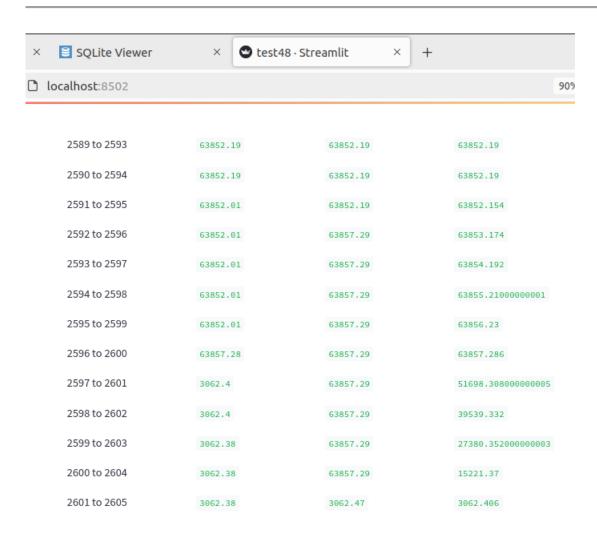
In the batch mode experiment, we conducted offline analysis of historical cryptocurrency market data using Apache Spark Batch Processing. We retrieved a large dataset containing historical price, volume, and other relevant metrics for Bitcoin (BTC), Ethereum (ETH), and Litecoin (LTC) from a reliable cryptocurrency exchange. The dataset covered a specific time period, allowing us to perform batch processing on a static snapshot of the data.



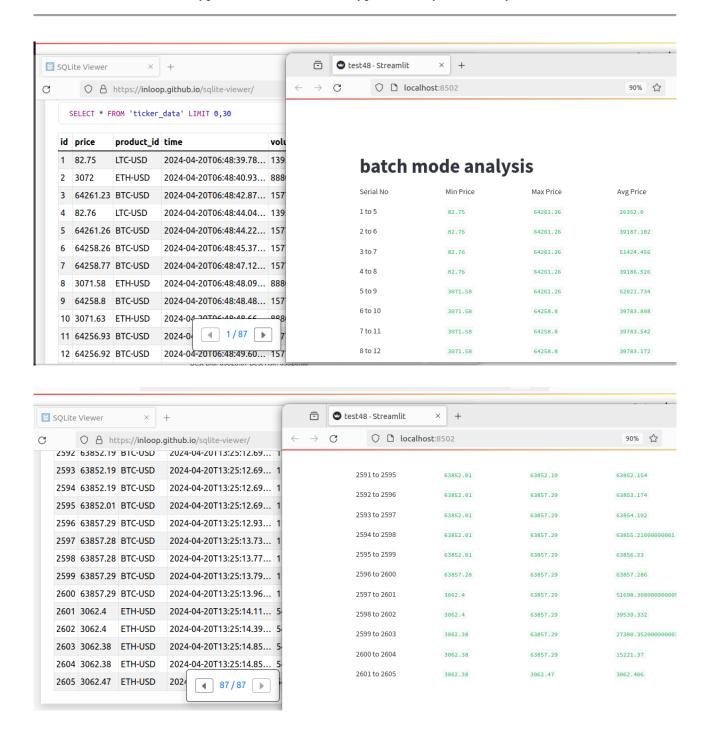
batch mode analysis

| Serial No | Min Price | Max Price | Avg Price |
|-----------|-----------|-----------|---------------------|
| 1 to 5 | 82.75 | 64261.26 | 26352.0 |
| 2 to 6 | 82.76 | 64261.26 | 39187.102 |
| 3 to 7 | 82.76 | 64261.26 | 51424.456 |
| 4 to 8 | 82.76 | 64261.26 | 39186.526 |
| 5 to 9 | 3071.58 | 64261.26 | 52021.734 |
| 6 to 10 | 3071.58 | 64258.8 | 39783.808 |
| 7 to 11 | 3071.58 | 64258.8 | 39783.542 |
| 8 to 12 | 3071.58 | 64258.8 | 39783.172 |
| 9 to 13 | 3071.63 | 64258.8 | 52020.2000000000004 |
| 10 to 14 | 3071.63 | 64256.93 | 52019.623999999996 |
| 11 to 15 | 64255.0 | 64256.93 | 64256.297999999995 |
| 12 to 16 | 64255.0 | 64256.92 | 64255.912 |
| 13 to 17 | 82.76 | 64256.72 | 51421.08 |

here we are using 5 window size and sliding window mechanism



6. Comparison of Streaming & Batch Modes



7. Conclusion:

1. Real-time vs. Offline Analysis:

- **Streaming Mode:** Provides real-time insights into cryptocurrency market dynamics, allowing for immediate response to changing market conditions and opportunities.
- **Batch Mode:** Conducts offline analysis on historical data snapshots, enabling in-depth exploration of long-term trends, patterns, and anomalies in the cryptocurrency market.

2. Data Freshness:

- **Streaming Mode:** Offers up-to-date information by processing data as it arrives, ensuring the freshest insights but may be subject to processing delays or latency.
- **Batch Mode:** Analyzes historical data in bulk, providing a comprehensive view of past market behavior but lacks immediacy in responding to current market events.

3. Resource Utilization:

- **Streaming Mode:** Requires continuous processing resources to handle incoming data streams in real-time, which may result in higher resource utilization and operational costs.
- **Batch Mode:** Utilizes resources intermittently for scheduled batch processing tasks, potentially resulting in lower overall resource consumption and cost-effectiveness.

8. References:

• Spark twitter streaming online course