

A Internship Project Report on

Stock Market Real-Time Data Analysis Using Kafka BY

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MCA – II, SEM-IV

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Dr. Vishwanath Karad MIT World Peace University, Pune Pune - 411038

In Partial Fulfillment of the Degree of Master in Computer Application (M. C. A.)

Under the Guidance of

Name of Guide

Dr. Monika Gadre Academic Year 2023-24

CERTIFICATE

This is to certify that Mr. / Ms. VINAY WASUDEO YERNE has successfully completed his/her project work entitled "Stock Market Real-Time data Analysis Using Kafka" in partial fulfillment of MCA – II Semester-IV program for the year A.Y. 2023-2024. He /She have worked under our guidance and direction.

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Examiner 1 Examiner 2

Date: Place:

Company Offer Letter

HACKVEDA LIMITED



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Internship Selection Letter - AWS Internship

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Dear Candidate,

We are glad to have you onboard as an Intern for Hackveda (India) and Hackveda Limited, Canada.

Roles and Responsibilities

- Develop project source codes or Submit Tasks as per requirement specification
- Test working codes / task submissions, perform unit testing, automated testing and submit a test report / presentation
- Deploy working project on GitHub repositories or platforms recommended / provided by Hackveda Limited for live working demo or as per required specification
- Write standard software documentation for deployment and how to use the project
- Provide technical support to consumer in case of any errors on the existing software system

Financials

Probation period for Stipends: 2 months

Interns who may not perform well during their probation are allowed to continue their internships up-to 6 months or as desired without stipends. Evaluations for stipends will be done each month

Acknowledgment

I am grateful to <u>HACKVEDA</u> and my internship supervisor MS. <u>NEHA</u>, for providing me with the opportunity to completemy MCA internship at their organization. Their support and guidance helped me to understand the business world and gain valuable experience in my field.

From the moment I started my internship, **Mr. Devanshu Shukla sir**, provided me with clear direction and expectations, and was always available to answer my questions and provide valuable feedback. Their expertise and guidance helped me to understand the inner workings of the company and the industry, and allowed me to make the most of my internship.

Thankyou.

Student Name:-

Vinay Wasudeo Yerne

(TASK BASED)

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CHAPTER 1:- INTRODUCTION

Cloud computing is an on-demand delivery of compute power, database, storage, applications, and other IT resources via a cloud services platform via the internet with pay-as-you-go pricing. It offers rapidaccess to flexible and low-cost IT resources, allowing users to provision theright type and size of computing resources to power their ideas or operate their IT department. Cloud computing provides a simple way to access servers, storage, databases, and a broad set of application services over the internet. A cloud services platform like Amazon Web Services owns and maintains the network-connected hardware required for these application services, while user provision and use resources via a web application.

AWS service that allows developers to create conversational interfaces for applications using voice and text. It uses the same conversational engine as Amazon Alexa, allowing developers to create sophisticated, natural language chatbots in new and existing applications. It offers deep functionality and flexibility in natural language understanding and automatic speech recognition, enabling engaging user experiences and product categories.

Amazon Web Services (AWS) is the world's most comprehensive and broadly adopted cloud, offering over 200 fully featured services from data centers globally. Millions of customers—including the fastest-growing startups, largest enterprises, and leading government agencies—are using AWS to lower costs, become more agile, and innovate faster.

Setting up AWS Cloud for stock market data management and analysis involves creating a scalable and secure infrastructure on the Amazon Web Services (AWS) platform to store, process, and analyze vast amounts of financial data efficiently. This includes designing and deploying databases, implementing data ingestion pipelines, developing analytical models, and creating user-friendly interfaces for accessing insights and reports. Leveraging AWS services such as

Amazon S3 for storage, Amazon Redshift for data warehousing, AWS Glue for ETL (Extract, Transform, Load), and Amazon Sage Maker for machine learning, the system enables seamless data management, real-time analytics, and predictive modeling to support informed decision-making in the dynamic stock market environment.

Setting up AWS Cloud for stock market data management and analysis involves deploying a robust infrastructure capable of handling the storage, processing, and analysis of vast amounts of financial data. This infrastructure enables organizations to store historical and real-time market data securely, perform complex data analytics, and derive actionable insights to drive business strategies.

In this document, we will explore the process of setting up AWS Cloud for stock market data management and analysis, including the selection and configuration of AWS services, the design and implementation of data processing pipelines, and the development of analytical models to extract valuable insights from stock market data. Through this comprehensive approach, organizations can leverage the power of AWS Cloud to enhance their capabilities in managing and analyzing stock market data, ultimately driving better financial outcomes and staying ahead in today's competitive market environment.

AWS KAFKA:-

Amazon Managed Streaming for Apache Kafka (Amazon MSK) is a fully managed service that makes it easy for you to build and run applications that use Apache Kafka to process streaming data. Apache Kafka is an open-source platform for building real-time streaming data pipelines and applications.

Here are some key features and benefits of Amazon MSK:

- 1. Managed Service: Amazon MSK simplifies the setup, management, and monitoring of Apache Kafka clusters, allowing you to focus on building applications rather than managing infrastructure.
- 2. High Availability: Amazon MSK provides high availability by automatically replicating data across multiple availability zones within a region, ensuring durability and fault tolerance.
- 3. Scalability: You can easily scale your Apache Kafka clusters up or down based on your application's needs, without worrying about provisioning or managing additional resources.
- 4. Integration with AWS Services: Amazon MSK integrates seamlessly with other AWS services such as Amazon S3, Amazon CloudWatch, and AWS Identity and Access Management (IAM), enabling you to build end-to-end streaming data solutions.
- 5. Security: Amazon MSK offers built-in security features such as encryption at rest and in transit, authentication using AWS IAM, and network isolation through Amazon VPC, ensuring data confidentiality and integrity.
- 6. Monitoring and Management: Amazon MSK provides metrics and logs through Amazon CloudWatch, allowing you to monitor the performance and health of your Apache Kafka clusters in real-time.

1.1 Existing System:-

This section outlines the current system or processes that are in place for managing and analyzing stock market data. It may include details about any software tools, databases, or manual procedures currently used ionized software tools like Bloomberg Terminal or Reuters Eikon for real-time data and analysis. These tools are complemented by databases housing historical and real-time market data.

Tools like AWS Kafka or Terminal provide real-time and historical market data. Databases store vast data including prices, financials. Manual tasks involve validation and qualitative analysis. Data from exchanges, news, filings are integrated via APIs. Analysis includes fundamental, technical, and sentiment analysis.

1.1.1 Online Research: -

- Software Tools: Stock market data analysis often relies on specialized software tools such as Bloomberg Terminal, Reuters Eikon, or proprietary trading platforms. These tools provide real-time market data, historical data, charting, technical analysis tools, and sometimes algorithmic trading capabilities.
- Databases: Large volumes of stock market data are stored in databases for analysis. These databases may include historical price data, company financials, economic indicators, and other relevant information. Common database technologies used in this context include SQL databases like MySQL or PostgreSQL, as well as NoSQL databases like MongoDB for handling unstructured data.
- Manual Procedures: Despite the advancements in technology, some aspects of stock market analysis still require manual intervention. This could include tasks such as data validation, qualitative analysis of news or events impacting the market, or decision-making based on non-quantitative factors.
- Data Sources: The system likely integrates data from various sources such as stock exchanges, financial news outlets, regulatory filings, and third-party data providers. APIs (Application Programming Interfaces) are often used to automate data retrieval and ensure data accuracy.
- Analysis Techniques: Different analysis techniques are employed, including fundamental analysis (evaluating a company's financial health and prospects), technical analysis (examining price movements and trading volumes), and sentiment analysis (assessing market sentiment based on news articles, social media, etc.).

1.2 Need for System:-

- The current system for stock market data management and analysis may exhibit several shortcomings necessitating a new solution. Inefficiencies in data processing and analysis could lead to delays in decision-making, hindering traders and investors from capitalizing on timely market opportunities.
- Limitations in scalability may impede the system's ability to handle the ever-growing volumes of market data effectively, potentially leading to performance bottlenecks and decreased responsiveness. Furthermore, the existing system may lack integration with emerging data sources like social media, depriving users of valuable insights into market sentiment and trends.
- Additionally, there may be a need for additional features such as advanced analytics capabilities including machine learning algorithms for predictive modeling and risk assessment, which can provide a competitive edge in today's dynamic markets. Overall, the demand for a new system arises from the desire to overcome these inefficiencies, limitations, and to incorporate advanced features essential for robust and agile stock market data management and analysis.

1.2.1 Limitations of Conventional Methods:-

- Latency Issues: Despite Kafka's real-time processing capabilities, conventional methods may struggle to achieve low-latency data analysis due to inefficient data processing pipelines or resource constraints. This latency can impede timely decision-making in fast-paced market environments.
- Scalability Challenges: Conventional methods may face difficulties in effectively scaling data processing pipelines to handle the high volume of real-time data generated by stock markets. Scaling issues can lead to performance bottlenecks and decreased system responsiveness, especially during periods of high market activity.
- Complex Data Transformation: Efficiently transforming raw stock market data into a format suitable for analysis can be challenging for conventional methods. This complexity arises from the need for data normalization, aggregation, and enrichment processes, which may not be easily handled using traditional techniques.
- Limited Analytical Capabilities: Conventional methods may lack advanced analytical capabilities compared to modern data analytics platforms. This limitation restricts the types of analysis that can be performed on stock market data, potentially limiting insights and decision-making potential.
- Dependency on Legacy Systems: Conventional methods often rely on legacy systems and technologies that may not be optimized for real-time data analysis using Kafka. This reliance introduces compatibility issues and impedes the seamless integration of Kafka into existing data infrastructure.
- Risk of Data Loss: Inadequate fault tolerance mechanisms in conventional setups may pose a risk of data loss, particularly during system failures or network disruptions. This risk undermines data integrity and reliability, crucial factors in stock market data analysis.

1.3 Operating Environment Hardware and Software

- •The proposed system requires hardware infrastructure capable of handling realtime data processing and analysis. This includes high-performance servers with sufficient CPU, memory, and storage resources to accommodate the volume of data generated by stock market feeds. Additionally, a distributed computing environment may be necessary for scalability and fault tolerance.
- •For the operating system, compatibility with both Linux and Windows environments is recommended to cater to diverse deployment scenarios. Linux distributions like Ubuntu, CentOS, or Red Hat Enterprise Linux are popular choices for server deployments due to their stability and performance benefits.
- •Regarding databases, support for both SQL and NoSQL databases is essential to accommodate various data storage and querying requirements. SQL databases such as PostgreSQL or MySQL are suitable for structured data storage, while NoSQL databases like Apache Cassandra or MongoDB can handle unstructured or semi-structured data efficiently.
- •Key software dependencies include Apache Kafka for real-time data ingestion and stream processing, along with stream processing frameworks like Apache Flink or Apache Spark Streaming for advanced analytics. Additionally, analytics tools such as Apache Hadoop or Apache Hive may be utilized for batch processing and data warehousing.
- •Containerization technologies like Docker and orchestration platforms like Kubernetes can facilitate deployment, management, and scalability of the system components in a distributed environment.

CHAPTER 2 -PROPOSED SYSTEM:-

2.1 Proposed System (Introduction of system):-

- The proposed system for stock market data management and analysis is a comprehensive platform designed to facilitate real-time processing, analysis, and decision-making in financial markets. It offers advanced features such as:
- Real-time Data Ingestion: Seamless integration with data sources to ingest real-time stock market data feeds from exchanges, news outlets, and other relevant sources.
- Scalable Architecture: Built on scalable infrastructure to handle high volumes of data efficiently, ensuring responsiveness during peak market activity.
- Advanced Analytics: Utilizes cutting-edge analytics techniques including machine learning and natural language processing for predictive modeling, sentiment analysis, and risk assessment.
- Flexible Data Storage: Supports both SQL and NoSQL databases for flexible data storage, enabling structured and unstructured data analysis.
- Stream Processing: Leverages stream processing frameworks like Apache Kafka and Apache Flink for real-time data processing, enabling instant insights and decision-making.
- Interactive Visualization: Provides interactive visualization tools for intuitive exploration and analysis of stock market trends, patterns, and anomalies.
- Robust Security: Implements robust security measures to protect sensitive financial data and ensure compliance with regulatory requirements.

2.2 Module specifications (Scope):-

- Data Ingestion: Module responsible for ingesting real-time stock market data from multiple sources such as exchanges, news feeds, and social media platforms.
- Data Storage: Module for storing and managing stock market data efficiently, supporting both SQL and NoSQL databases for structured and unstructured data.
- Data Processing: Module for processing raw data streams using stream processing frameworks like Apache Kafka and Apache Flink to enable real-time analytics.
- Analytics: Module incorporating advanced analytics techniques including machine learning, natural language processing, sentiment analysis, and statistical modeling for extracting insights from stock market data.
- Visualization: Module for interactive visualization of analyzed data, providing intuitive tools for exploring trends, patterns, and anomalies in stock market data.
- Alerting and Monitoring: Module for setting up alerts and monitoring systems to notify users of significant market events or deviations from predefined criteria.
- Security and Compliance: Module ensuring robust security measures to protect sensitive financial data and ensure compliance with regulatory requirements such as GDPR and FINRA.
- Integration and APIs: Module facilitating integration with external systems and providing APIs for seamless data exchange with third-party applications and services.

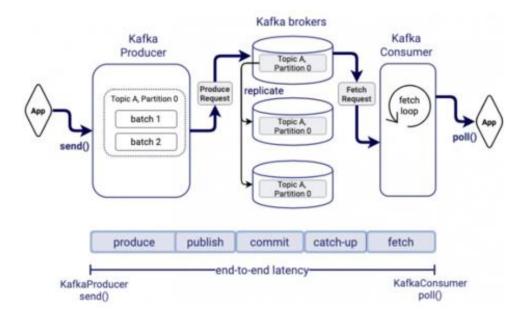
2.3 Objectives of System:-

The objectives of the proposed system are:-

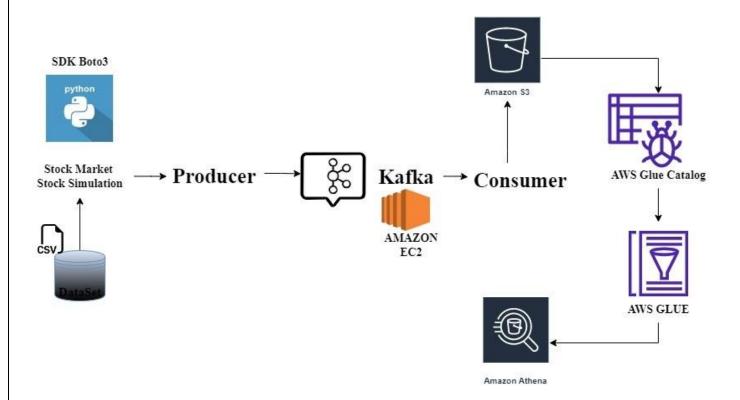
- Real-time Insights: Enable users to access real-time stock market data and gain actionable insights for informed decision-making.
- Scalability: Provide a scalable architecture capable of handling high volumes of data to support growing user demands.
- Advanced Analytics: Implement advanced analytics techniques such as machine learning and sentiment analysis to extract valuable insights from data.
- Efficiency: Streamline data processing and analysis workflows to improve efficiency and reduce latency in decision-making processes.
- Visualization: Offer intuitive visualization tools for interactive exploration of stock market trends, patterns, and anomalies.
- Security and Compliance: Ensure robust security measures and compliance with regulatory requirements to protect sensitive financial data.
- User-Friendly Interface: Design a user-friendly interface to enhance user experience and facilitate ease of use for traders, analysts, and investors.
- Integration: Facilitate integration with external systems and APIs to enable seamless data exchange and interoperability with other applications.

CHAPTER 3: ANALYSIS & DESIGN

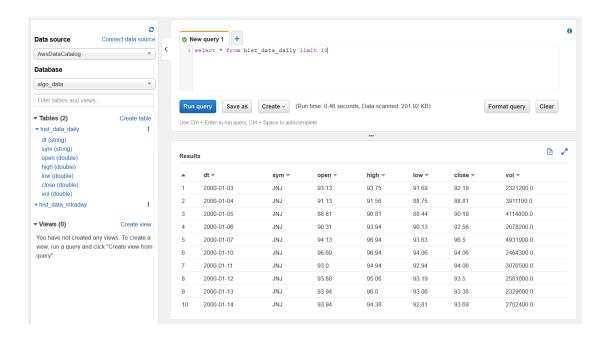
3.1 Process of Kafka Diagrams:-



3.2 Stock Market Data analysis Using Kafka Architecture.



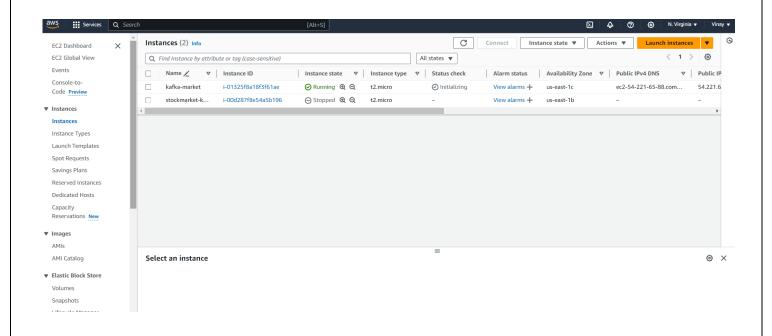
3.3 Database :-

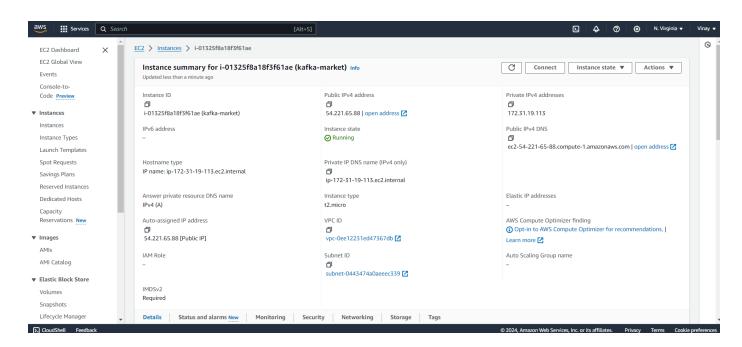


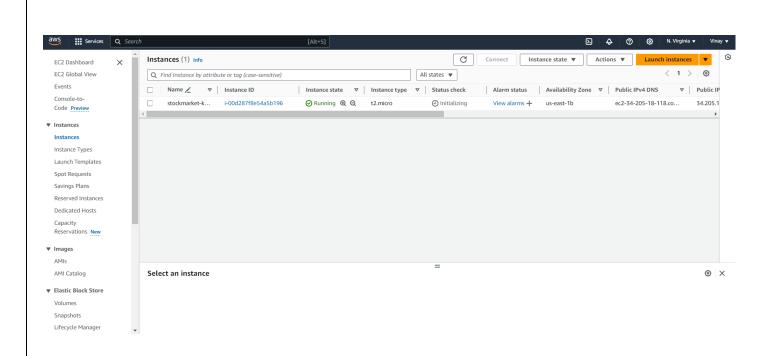


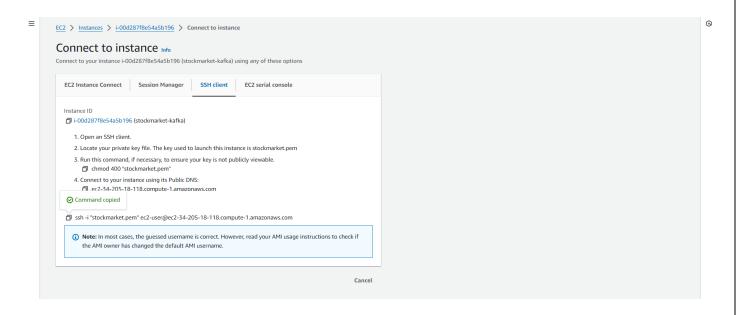
CHAPTER 4:- USER MANUAL

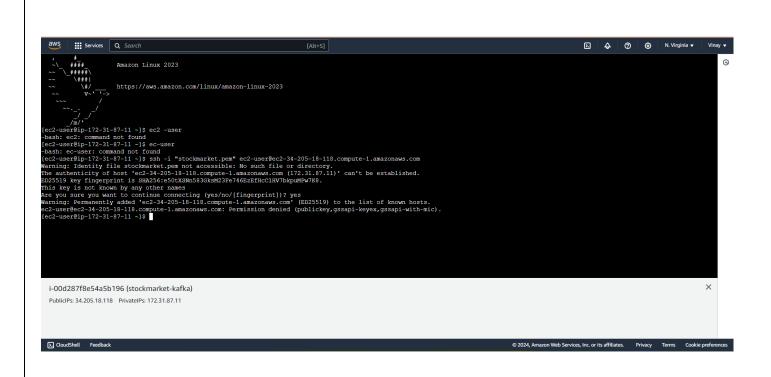
4.1 User Interface Screens (Input):-

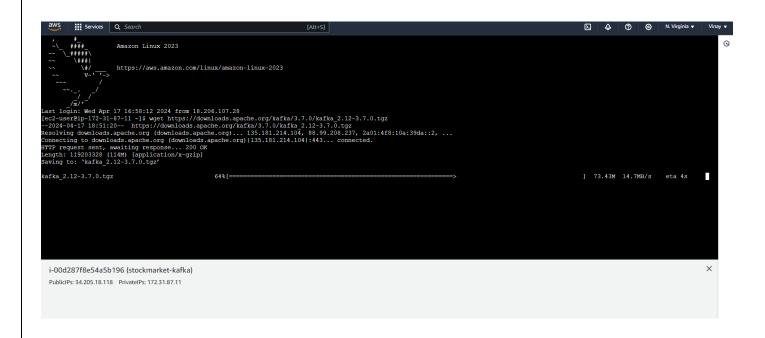


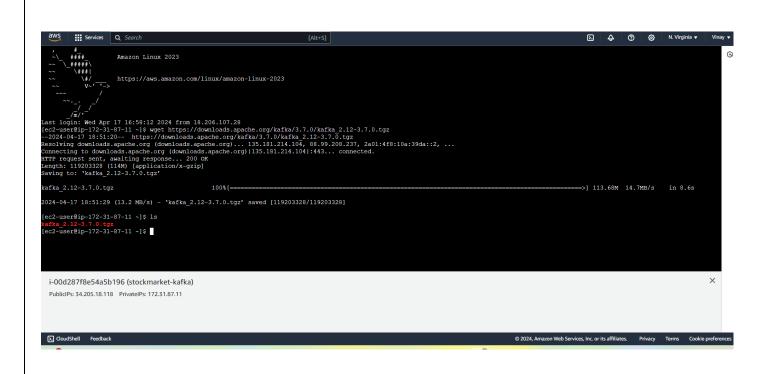


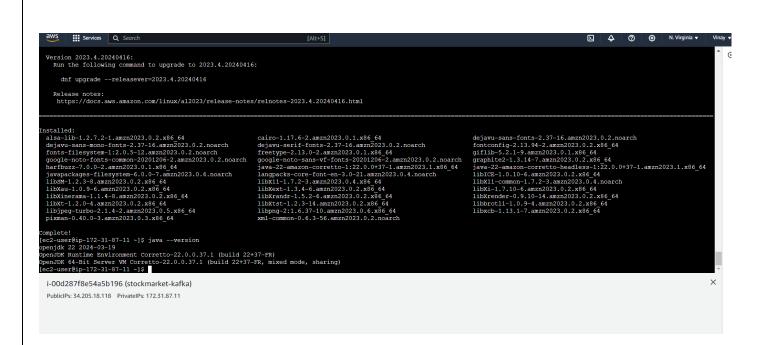


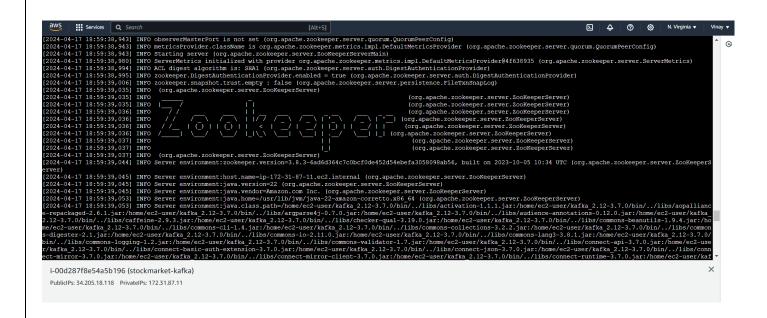


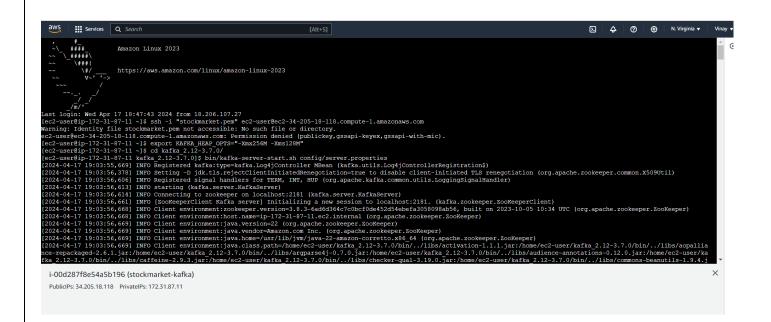


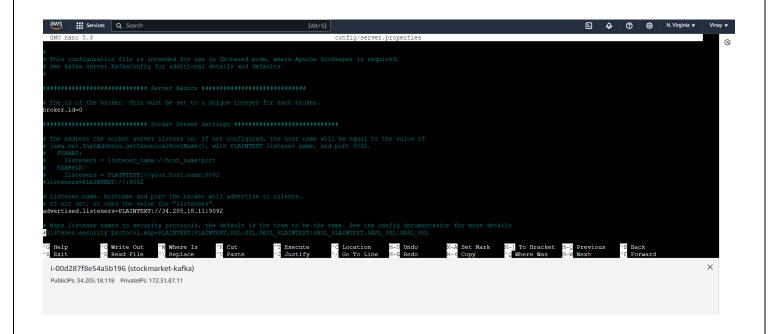




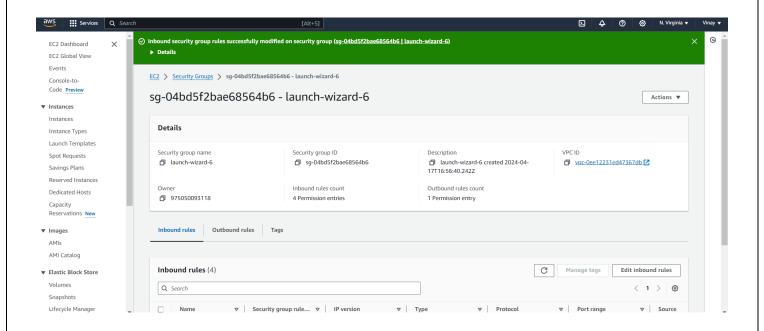


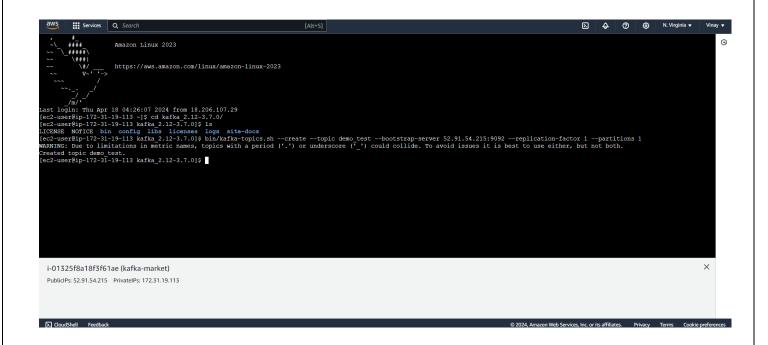


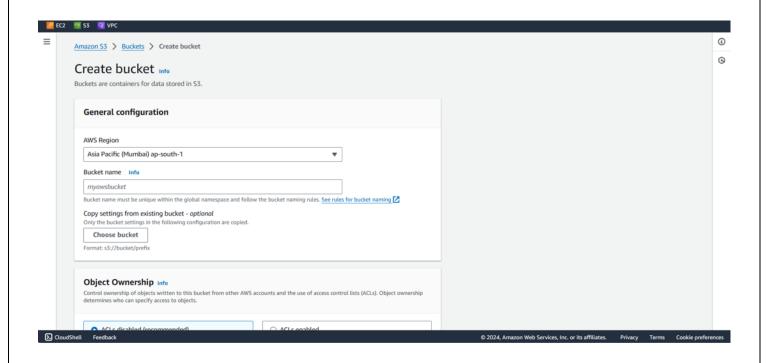


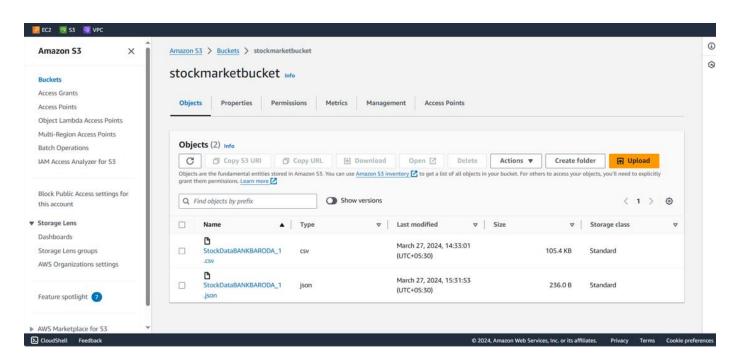


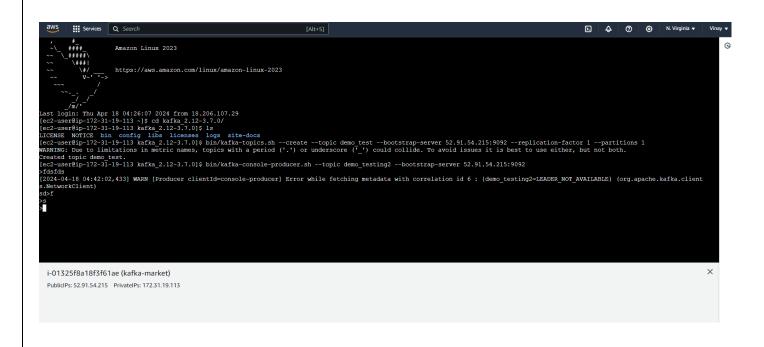
4.2 Output Screen:-

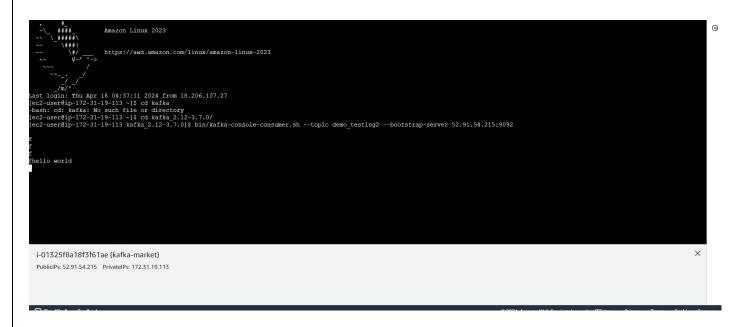






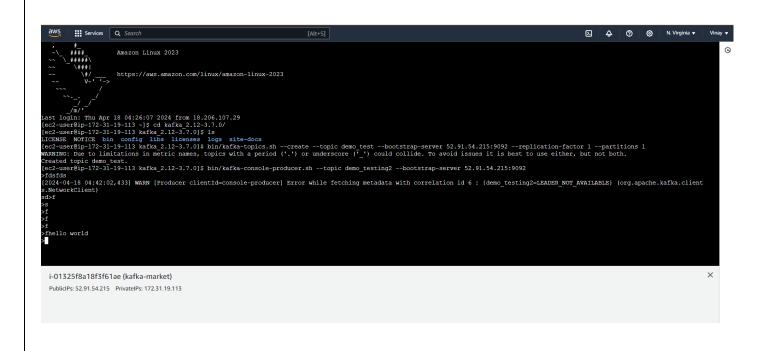




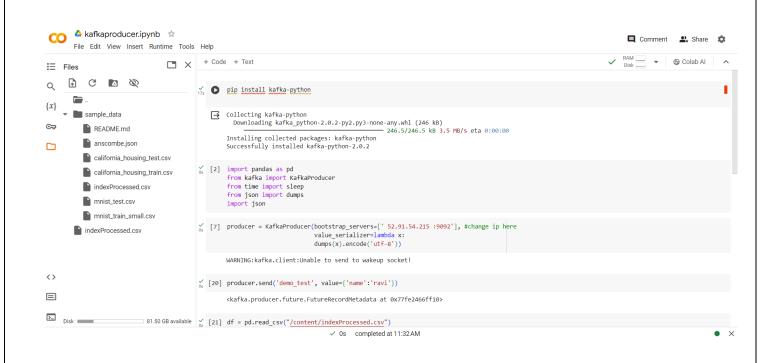


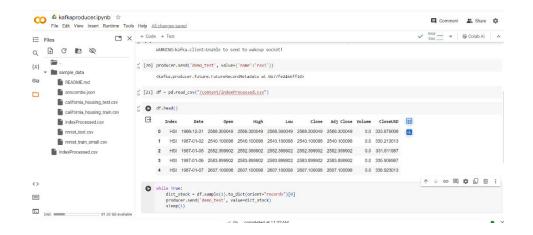


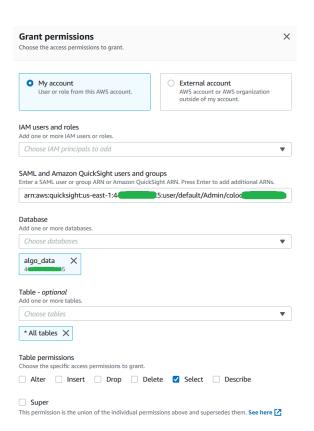


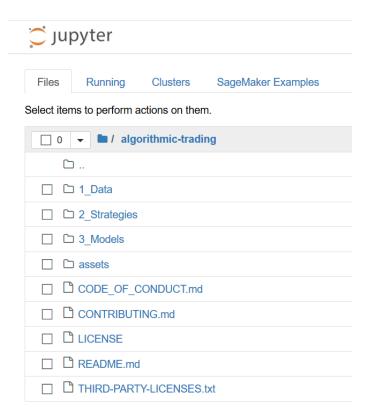


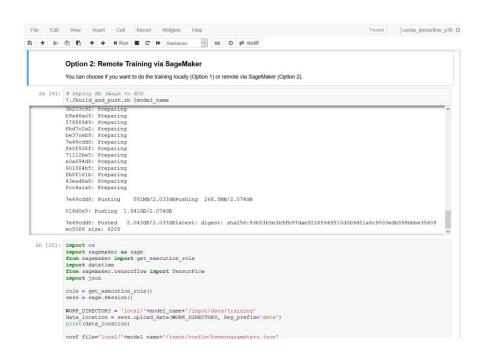




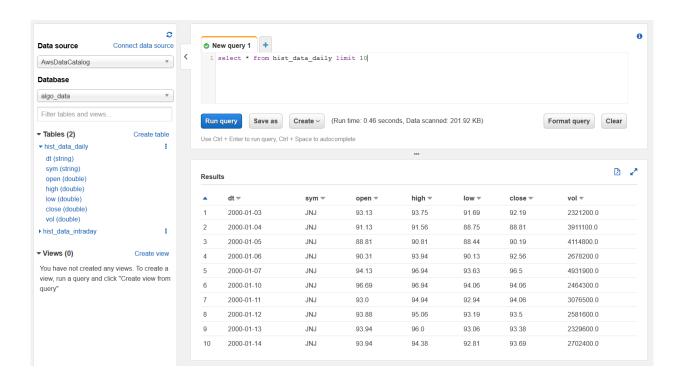








4.4 Test Procedures and cases:-





4.5 Sample Program code:-

Command Kafka:-

wget https://downloads.apache.org/kafka/3.3.1/kafka_2.12-3.3.1.tgz tar -xvf kafka_2.12-3.3.1.tgz

java -version sudo yum install java-1.8.0-openjdk java -version cd kafka 2.12-3.3.1

Start Zoo-keeper:

bin/zookeeper-server-start.sh config/zookeeper.properties

Open another window to start kafka But first ssh to to your ec2 machine as done above

Start Kafka-server:

Duplicate the session & enter in a new console -- export KAFKA_HEAP_OPTS="-Xmx256M -Xms128M" cd kafka_2.12-3.3.1 bin/kafka-server-start.sh config/server.properties

It is pointing to private server, change server.properties so that it can run in public IP

To do this, you can follow any of the 2 approaches shared belwo -- Do a "sudo nano config/server.properties" - change ADVERTISED_LISTENERS to public ip of the EC2 instance

| ~ | . 1 | . • |
|--------|-----|--------|
| Create | the | tonic |
| Create | uic | topic. |

Duplicate the session & enter in a new console --

cd kafka_2.12-3.3.1

bin/kafka-topics.sh --create --topic demo_testing2 --bootstrap-server {Put the Public IP of your EC2 Instance:9092} --replication-factor 1 --partitions 1

Start Producer:

bin/kafka-console-producer.sh --topic demo_testing2 --bootstrap-server {Put the Public IP of your EC2 Instance:9092}

Start Consumer:

Duplicate the session & enter in a new console -- cd kafka_2.12-3.3.1

 $bin/kafka-console-consumer.sh --topic demo_testing2 --bootstrap-server \{Put \ the \ Public \ IP \ of \ your \ EC2 \ Instance:9092\}$

4.5.2Kafka producer:-

```
pip install kafka-python
                                                                                      In [ ]:
import pandas as pd
from kafka import KafkaProducer
from time import sleep
from json import dumps
import json
                                                                                      In []:
producer = KafkaProducer(bootstrap servers=[':9092'], #change ip here
                          value_serializer=lambda x:
                          dumps(x).encode('utf-8'))
                                                                                      In []:
producer.send('demo test', value={'surnasdasdame':'parasdasdmar'})
                                                                                      In []:
df = pd.read csv("data/indexProcessed.csv")
                                                                                      In []:
df.head()
                                                                                      In [ ]:
while True:
    dict_stock = df.sample(1).to_dict(orient="records")[0]
    producer.send('demo_test', value=dict_stock)
    sleep(1)
                                                                                      In []:
producer.flush() #clear data from kafka server
```

4.5.3 Kafka Consumer:-

```
from kafka import KafkaConsumer
from time import sleep
from json import dumps, loads
import json
from s3fs import S3FileSystem
                                                                                      In []:
consumer = KafkaConsumer(
    'demo test',
    bootstrap_servers=[':9092'], #add your IP here
    value deserializer=lambda x: loads(x.decode('utf-8')))
                                                                                      In [ ]:
# for c in consumer:
     print(c.value)
                                                                                      In []:
s3 = S3FileSystem()
                                                                                      In [ ]:
for count, i in enumerate(consumer):
    with s3.open("s3://kafka-stock-market-tutorial-youtube-
darshil/stock_market_{}.json".format(count), 'w') as file:
        json.dump(i.value, file)
                                                                                      In []:
```

4.6 Limitations and Bibliography

• Limitations:

- Integration Complexity: The system may face challenges in integrating with certain legacy systems or proprietary data sources, potentially limiting the scope of data analysis.
- Resource Constraints: Limitations in available resources, such as computing power or storage capacity, could affect the system's scalability and performance under heavy loads.
- Data Accuracy: Despite advanced analytics techniques, the accuracy of predictions and insights derived from stock market data may be influenced by factors such as data quality and market volatility.

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